Elements of Formal Semantics
An Introduction to the Mathematical Theory of Meaning in Natural Language

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Open Access Materials: Chapters 1-3

Elements of Formal Semantics introduces some of the foundational concepts, principles and techniques in formal semantics of natural language. It is intended for mathematically-inclined readers who have some elementary background in set theory and linguistics. However, no expertise in logic, math, or theoretical linguistics is presupposed. By way of analyzing concrete English examples, the book brings central concepts and tools to the forefront, drawing attention to the beauty and value of the mathematical principles underlying linguistic meaning.

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See webpage below for further materials and information:

http://www.phil.uu.nl/~yoad/efs/main.html
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CHAPTER 1

INTRODUCTION

One of the most striking aspects of human language is the complexity of the meanings that it conveys. No other animal possesses a mode of expression that allows it to articulate intricate emotions, describe distant times and places, study molecules and galaxies, or discuss the production of sophisticated tools, weapons and cures. The complex meanings of natural language make it an efficient, general-purpose instrument of human thought and communication. But what are meanings? And how does language convey them?

To illustrate one aspect of the problem, let us consider a phrase in one of Bob Dylan’s famous love songs. The phrase opens the song’s refrain by describing a woman, whose identity is not disclosed. It goes like this:

\begin{equation}
\text{(1.1) sad-eyed lady of the lowlands, where the sad-eyed prophet says that no man comes}
\end{equation}

If we want to restate the meaning of this phrase in simpler terms, we can do it as follows:

\begin{equation}
\text{(1.2) There’s a lady. That lady has sad eyes. She is from the lowlands. Some prophet also has sad eyes. That prophet says “no man comes to the lowlands”.
}\end{equation}

Without doubt, this way of paraphrasing Dylan’s verse robs it of much of its poetic value. But at the same time it also highlights a remarkable property of meaning in natural language. When we hear a long expression like (1.1), we immediately draw from it all sorts of simple conclusions. This happens even in cases where we miss information that is important for understanding the “true meaning” of what is being said. Dylan’s song only gives vague clues about the identity of the lady. Yet upon hearing the refrain we unfailingly draw
from (1.1) the conclusions in (1.2). The converse is true as well: when Dylan invented his description of the sad-eyed lady, he must have implicitly assumed the statements in (1.2) as part of its meaning. This kind of back-and-forth reasoning occurs whenever we think and converse. When we hear, utter or think of an expression, we instinctively relate it to other phrases that we consider obvious conclusions. Drawing such trivial-looking inferences using our language is one of the abilities that characterize us as linguistic creatures. No other animal has this linguistic ability, and no current technology can accurately mimic it.

Our effortless manipulation of meaning is highly systematic, and relies on an ingrained ability to recognize structure in language. When we hear the phrase in (1.1), we mentally tack its words into short collocations like sad-eyed and the lowlands. Further, short expressions are tacked together into longer expressions such as sad-eyed lady from the lowlands. These syntactic dependencies between words and expressions lead to a complex hierarchical structure. In the case of (1.1), some main elements of this structure are represented below.

\[
(1.3) \text{[[ sad-eyed ] lady ] [ of[[ the lowlands ],[ where [[ the [[ sad-eyed ] prophet ]] [ says [ that [[ no man ] comes ]]]]]]]]
\]

The bracketed expressions in (1.3) represent constituents: sub-parts of the description in (1.1) that act as syntactic units – noun phrases, verb phrases, clauses etc. As the representation in (1.3) illustrates, constituents are often embedded within one another. For instance, the short sentence no man comes is embedded in the verb phrase says that no man comes, which is itself embedded within the sentence the sad-eyed prophet says that no man comes. In total, the expression in (1.1) has no fewer than seven levels of constituents that are embedded within each other. This complexity does not impair our ability to make sense of the description. Furthermore, it is part and parcel of our ability to understand it. In (1.1), the highly organized way in which the constituents are embedded makes it possible for us to immediately grasp Dylan’s complex vision as paraphrased in (1.2). In the case of complicated phrases like (1.1), it is clear that we would not be able to extract even the basic paraphrase in (1.2) if language did not support well-organized hierarchical structures. Furthermore, syntactic hierarchies help us to extract meaning from most other
linguistic expressions, including ones that are much more ordinary than Dylan’s verse.

The subfield of linguistics known as *formal semantics* studies how linguistic structure helps speakers to manipulate meaning. The word ‘formal’ stresses the centrality of linguistic forms in the enterprise. At the same time, the token ‘formal’ also expresses a motivation to account systematically for language meanings by using precise mathematical methods. Formal semanticists have benefited from the many breakthroughs in logic and computer science, two disciplines that constantly develop new artificial languages and address challenging questions about their meanings and forms. The dazzling achievements that logicians and computer scientists achieved in the twentieth century were based on a rich tradition of research in philosophy of language and the foundations of mathematics. It is only natural that in the 1960s, when semanticists started to systematically address questions about meaning and form in natural language, they turned to these neighboring disciplines in search of guiding principles. As a result, formal semantics relies on the mathematical foundations that were laid in major works on logic, philosophy of language and theoretical computer science.

The mathematical foundations of formal semantics give us precise tools for studying natural languages. Mathematical semantic models help us see what meanings are, and, more importantly, why they can be shared by different expressions. By examining meanings under the powerful microscope of mathematical theories, formal semantics has obtained effective methods for uncovering systematic regularities in the everyday use of language expressions.

The scientific value of this linguistic endeavor is further enhanced by recent developments in other branches of cognitive science that study natural language. In the emerging field of *cognitive neuroscience*, mathematical principles are becoming increasingly important for harnessing recent advances in brain imaging. As a leading cognitive neuroscientist puts it: “only mathematical theory can explain how the mental reduces to the neural. Neuroscience needs a series of bridging laws [...] that connect one domain to the other” (Dehaene 2014, p. 163). These laws are also needed in order to understand how the brain enables the semantic dexterity of language speakers. Mental semantic faculties are profitably described by mathematical laws. Recent works in natural language semantics have supported many of these
laws by statistically analyzing experimental data. As neuroscience brings more experimental data on the workings of the brain, it is becoming increasingly important to connect statistical generalizations about this data with models of our mental semantic abilities.

Similar procedures of mathematical theorizing are equally critical in current work in artificial intelligence. Recent advances in statistical machine learning make it possible to exploit formal semantic principles to enhance algorithms and computing technologies. In a recent state-of-the-art review, the authors describe this new direction, stating that “the distinction between logical and statistical approaches is rapidly disappearing with the development of models that can learn the conventional aspects of natural language meaning from corpora and databases” (Liang and Potts 2015, p. 356). In the new domain of computational semantics, mathematical and logical principles of formal semantics are increasingly employed together with statistical algorithms that deal with the parametrization of abstract semantic models by studying distributions of various linguistic phenomena in ordinary language.

Although these recent developments are not the focus of the current book, they do highlight new motivations for using precise principles and techniques in the study of natural language semantics. The achievements of formal semantics have formed a lively area of research, where new ideas, techniques, experimental results and computer systems appear every day. This book introduces you to some of the most important mathematical foundations of this field.

AIMS AND ORGANIZATION OF THIS BOOK

The two senses of the word ‘formal’ have a key role in this textbook. The book is a systematic introduction to the study of form and meaning in natural language. At the same time, it capitalizes on the precise mathematical principles and techniques that underlie their analysis. The aim is to help the reader acquire the tools that would allow her to do further semantic work, or engage in interdisciplinary research that relies on principles of formal semantics. Because of that, the book does not attempt to single out any of the current versions of formal semantic theory. Rather, it covers five topics that are of utmost importance to all of them.
Chapter 2 is a general overview of the major goals and techniques in formal semantics. It focuses on the principles of natural language semantics that support meaning relations as in (1) and (2). These semantic relations are called entailments. They are described by abstract mathematical models, and general principles of compositionality that connect forms with model-theoretical meanings.

Chapter 3 introduces semantic types as a means of systematizing the use of models. Typed meanings are derived from simpler ones by a uniform semantic operation of function application. A convenient notation of lambda-terms is introduced for describing semantic functions. This notation is illustrated for a couple of modification and coordination phenomena.

Chapter 4 uses the principles and tools of the two previous chapters for treating quantification. By focusing on the semantics of noun phrases that involve counting and other statements about quantities, Chapter 4 directly introduces one of the best-known parts of formal semantics: the theory of generalized quantifiers.

Chapter 5 extends the framework of the preceding chapters for treating meaning relations between expressions that appear a certain distance from each other. A principle of hypothetical reasoning is added to the system of Chapter 3. This principle works in duality with function application, and complements its operation. The two principles apply within a system of linguistic signs, which controls the interactions between forms and meanings.

Chapter 6 treats intensional expressions: expressions that refer to attitudes, beliefs or possibilities. Such expressions are treated in semantic models containing entities that represent possible worlds. Possible world semantics is introduced as a systematic generalization of the system developed in previous chapters.

Part of the material in Chapters 3 and 6 was covered by early textbooks on “Montague Grammar” (see further reading at the end of this chapter). Here, this material is introduced in a more general setting that takes recent findings into account and capitalizes on the mathematical architecture of type-theoretical grammars. Chapter 4 is unique in being a detailed textbook-level introduction to the central problem of quantification in natural language, which is fully based on the type-theoretical framework of Chapter 3. The treatment of long-distance dependencies in Chapter 5 is the first textbook-level
introduction of a general theoretical configuration known as Abstract Categorial Grammar.

At the end of each chapter there are exercises (see below) and references for suggested further reading. Further materials can be found through the website of Edinburgh University Press, at the following link:

edinburghuniversitypress.com/book/9780748640430

**ON THE EXERCISES IN THIS BOOK**

At the end of each chapter you will find some exercises, with model solutions to many of them. Acquiring the ability to solve these exercises constitutes an integral part of studying the material in this book. You will be referred to exercises at various points of the text, and further developments in the book often rely on the exercises in previous chapters. There are two kinds of exercise:

- Technical exercises, which should be solvable by using only the methods explained in the body of the textbook.
- More advanced exercises, which are specified at the beginning of each exercise section. Some of these advanced exercises introduce new notions that were not addressed in the text. These more “notional” advanced exercises are listed in **boldface** at the beginning of the exercises, and are especially recommended among the advanced exercises.

Upon finishing a chapter, and before moving on to the next chapter, it is advisable to make sure that you can correctly solve all of the technical exercises.

**WHO IS THIS BOOK FOR?**

The book is meant for any reader who is interested in human language and its mathematical modeling. For readers whose main interest is linguistic theory, the book serves as an introduction to some of the most useful tools and concepts in formal semantics, with numerous exercises to help grasp them. Readers who are mainly interested in mathematical models of language will find in the book an introduction to natural language semantics that emphasizes its empirical and methodological motivations.
INTRODUCTION

The book is especially suitable for the following audiences:

- general readers with the necessary mathematical background (see below)
- students and teachers of undergraduate linguistics courses on natural language semantics, which put sufficient emphasis on its set-theoretical background (see below)
- students and teachers of relevant undergraduate courses in artificial intelligence, computer science, cognitive science and philosophy
- researchers and advanced students in linguistics

PRESUPPOSED BACKGROUND

To be able to benefit from this book you should have some basic background in naive set theory. At the end of this chapter, you will find some suggestions for further reading, as well as some standard notation, exercises and solutions. By solving the exercises, you will be able to practice some basic set theory at the required level before you start reading. The book does not presuppose any prior knowledge in logic or theoretical linguistics. However, some general familiarity with these disciplines may be useful. Some suggestions for textbooks that introduce this background are given in the suggestions for further reading at the end of this chapter.

FOR THE INSTRUCTOR

The material in this book has been used for teaching undergraduate and graduate courses in linguistics, computer science and artificial intelligence programs. Different kinds of audiences may benefit from different complementary materials. For linguistics students, the most important additions should include more semantic and pragmatic theories of phenomena like anaphora, plurals, events, ellipsis, presupposition or implicature. In most linguistics programs, a short introduction to basic set-theoretical notions would be necessary in order to allow students to grasp the materials in this book (for materials see the further reading section below). For computer science and AI students, additional material on computational semantics may be useful, especially if it is accompanied by programming assignments. The type-theoretical semantics in this book is especially easy to adapt
for programming in strongly typed functional languages like Haskell. Some remarks about recommended literature are made at the end of the further reading section below.

FURTHER READING

Background material on linguistics, set theory and logic: For a general introduction to linguistics, see Fromkin et al. (2014). For a classical introduction to naive set theory, see Halmos (1960). Linguistics students may find the introduction in Partee et al. (1990, chs.1–3) more accessible. For a useful open-source introduction and exercises, see ST (2015). Two classical textbooks on logic are Suppes (1957); Barker-Plummer et al. (2011).


Other introductions to formal semantics: Chapters 3 and 6 overlap in some critical aspects with the early textbooks Dowty et al. (1981) and Gamut (1982), which introduced formal semantics as developed in Montague (1973). Zimmermann and Sternefeld (2013) is a friendly introduction to basic topics in formal semantics. For some of the topics covered in the present book, there are also more advanced textbooks that may be consulted. Carpenter (1997) and Jacobson (2014) are detailed introductions to compositional type-theoretical semantics. Jacobson’s book also contains an elaborate linguistic discussion. For introductions to formal semantics as it is often used in generative grammar, see Chierchia and McConnel-Ginet (1990); Heim and Kratzer (1997). For an introduction to formal semantics in the framework of Discourse Representation Theory, see Kamp and Reyle (1993). Readers who are interested in general perspectives on meaning besides formal semantics may consult Elbourne (2011); Saeed (1997).

For the instructor: On further important topics in formal semantics that are not covered in this textbook, see Chapter 7. For a textbook that uses the Haskell programming language to illustrate some of the core problems in formal semantics, see Van Eijck and Unger (2010).
INTRODUCTION

Concepts and notation from set theory

$x \in A$  \hspace{1em}  x \text{ is an element of the set } A = x \text{ is a member of } A

$x \notin A$  \hspace{1em}  x \text{ is not an element of } A

∅  \hspace{1em}  the empty set = the set that has no members

$A \subseteq B$  \hspace{1em}  the set $A$ is a subset of the set $B = B$ is a superset of $A = \text{ every element of } A \text{ is an element of } B$

$A \not\subseteq B$  \hspace{1em}  $A$ is not a subset of $B$

$\wp(A)$  \hspace{1em}  the powerset of $A = \text{ the set of all subsets of } A$. Example: $\wp(\{a, b\}) = \{\emptyset, \{a\}, \{b\}, \{a, b\}\}$

$A \cap B$  \hspace{1em}  the intersection of $A$ and $B = \text{ the set of elements that are in both } A \text{ and } B$

$A \cup B$  \hspace{1em}  the union of $A$ and $B = \text{ the set of elements that are in } A \text{ or } B \text{ (or both)}$

$A - B$  \hspace{1em}  the difference between $A$ and $B = \text{ the set of elements in } A \text{ that are not in } B$

$A^c$  \hspace{1em}  the complement of $A$ (in $E) = E - A$, where $E$ is a given superset of $A$

$|A|$  \hspace{1em}  the cardinality of $A = \text{ for finite sets: the number of elements in } A$

$\{x \in A : S\}$  \hspace{1em}  the set of elements in $A$ s.t. the statement $S$ holds

Example: $\{x \in \{a, b\} : x \in \{b, c\}\} = \{a, b\} \cap \{b, c\} = \{b\}$

$\{A \subseteq B : S\}$  \hspace{1em}  the set of subsets of $B$ s.t. the statement $S$ holds. Example: $\{A \subseteq \{a, b\} : |A| = 1\} = \{\{a\}, \{b\}\}$

$\langle x, y \rangle$  \hspace{1em}  an ordered pair of items $x$ and $y$

$A \times B$  \hspace{1em}  the cartesian product of $A$ and $B = \text{ the set of ordered pairs } \langle x, y \rangle \text{ s.t. } x \in A \text{ and } y \in B$

Example: $\{a, b\} \times \{1, 2\} = \{(a, 1), (a, 2), (b, 1), (b, 2)\}$

A binary relation between $A$ and $B$ is a subset of the cartesian product $A \times B$.

A function $f$ from $A$ to $B$ is a binary relation between $A$ and $B$ that satisfies: for every $x \in A$, there is a unique $y \in B$ s.t. $\langle x, y \rangle \in f$. If $f$ is a function where $\langle x, y \rangle \in f$, we say that $f$ maps $x$ to $y$, and write $f : x \mapsto y$ or $f(x) = y$.

Example: the binary relation $f = \{(a, 1), (b, 2)\}$ is a function from $\{a, b\}$ to $\{1, 2\}$, which is equivalently specified $[a \mapsto 1, b \mapsto 2]$ or by indicating that $f(a) = 1$ and $f(b) = 2$. 
\[ B^A \] is the set of functions from \( A \) to \( B \).

Example: \( \{1, 2\}^{\{a, b\}} \) is the functions from \( \{a, b\} \) to \( \{1, 2\} \)
\[ = \{(a\mapsto1, b\mapsto1), (a\mapsto1, b\mapsto2), (a\mapsto2, b\mapsto1), (a\mapsto2, b\mapsto2)\} \]

**EXERCISES**

1. Which of the following statements are true?
   - (i) \( a \in \{a, b\} \)
   - (ii) \( \{a\} \in \{a, b\} \)
   - (iii) \( \{a\} \subseteq \{a, b\} \)
   - (iv) \( \{a\} \subseteq \{a, b\} \)
   - (v) \( \{a\} \subseteq \{a\} \)
   - (vi) \( \{a\} \subseteq \{a, \{a\}\} \)
   - (vii) \( \{(a, b, c)\} \subseteq \wp(\{(a, b, c)\}) \)
   - (viii) \( \{(a, b, c)\} \in \wp(\{(a, b, c)\}) \)
   - (ix) \( \emptyset \in \{(a, \{b\}, \{c\}\} \)
   - (x) \( \emptyset \subseteq \{(a, \{b\}, \{c\}\} \)

2. Write down explicitly the following sets by enumerating their members, e.g. \( \wp(\{a\}) = \{\emptyset, \{a\}\} \).
   - (i) \( \wp(\{a, b\}) \)
   - (ii) \( \{a\} \cap \wp(\{a\}) \)
   - (iii) \( \{a\} \cap \wp(\{a, b\}) \)
   - (iv) \( \wp(\{a, b\}) \cap \wp(\{b, c\}) \)
   - (v) \( \wp(\{a\}) \cup \wp(\{\{b\}\}) \cup \wp(\{a, b\}) \)
   - (vi) \( \wp(\wp(\emptyset)) \)

3. Write down explicitly the following sets by enumerating their members.
   - (i) \( \{(a, b) \times \{c\}\} \)
   - (ii) \( \{(a, 1), (b, 2)\} \)
   - (iii) \( \{(a, 1), (a, 2)\} \)
   - (iv) \( \{(a, 1), (b, 1)\} \)
   - (v) \( \{(a, 1), (a, 2), (b, 1)\} \)

4. Which of the following binary relations are functions from \( \{a, b\} \) to \( \{1, 2\}\)?
   - (i) \( \{(a, 1), (b, 2)\} \)
   - (ii) \( \{(a, 1), (a, 2)\} \)
   - (iii) \( \{(a, 1), (b, 1)\} \)
   - (iv) \( \{(a, 1), (a, 2), (b, 1)\} \)

5. How many binary relations are there between \( \{a, b\} \) and \( \{1, 2\}\)? How many of them are functions?

6. Write down the functions in \( \{\text{no, yes}\}^{\{a, b, c\}} \). For each such function show a member of the powerset \( \wp(\{a, b, c\}) \) that intuitively corresponds to it.

7. Write down the functions in \( \{a, b, c\}^{\{\text{left, right}\}} \). For each such function show a member of the cartesian product \( \{a, b, c\} \times \{a, b, c\} \) that intuitively corresponds to it.

8. Write down explicitly the following sets of functions:
   - (i) \( \wp(\{a\})^{\wp(\{b\})} \)
   - (ii) \( \{1, 2\}^{\{a, b\} \times \{c\}} \)
   - (iii) \( \{(1, 2)^{\{c\}}\}^{\{a, b\}} \)

9. Consider the following function \( f \) in \( \{1, 2\}^{\{a, b\} \times \{c, d\}} \):
   \[ \langle a, c \rangle \mapsto 1, \langle a, d \rangle \mapsto 1, \langle b, c \rangle \mapsto 2, \langle b, d \rangle \mapsto 1 \]
   Write down the function \( g \) in \( \{(1, 2)^{\{c, d\}}\}^{\{a, b\}} \) that satisfies for every \( x \in \{a, b\} \), for every \( y \in \{c, d\} \):
   \( g(x)(y) = f(\langle x, y \rangle) \).
10. Write down explicitly the members of the following sets:
   (i) \( \{ f \in [a, b]^{[b, c]} : f(b) = b \} \)  
   (ii) \( \{ A \subseteq \{ a, b, c, d \} : |A| \geq 3 \} \)  
   (iii) \( \{ (x, y) \in [a, b, c] \times \{ b, c, d \} : x \neq y \} \)

SOLUTIONS TO EXERCISES

1. i, iii, v, vi, vii, x

2. (i) \( \emptyset, \{a\}, \{b\}, \{c\}, \{a, b\}, \{a, c\}, \{b, c\}, \{a, b, c\} \)  
   (ii) \( \emptyset \)  
   (iii) \( \{a\} \)  
   (iv) \( \emptyset, \{b\} \)  
   (v) \( \emptyset, \{a\}, \{b\} \)  
   (vi) \( \emptyset, \{\emptyset\} \)  

3. (i) \( \{a, c\} \)  
   (ii) \( \{\emptyset, \emptyset\}, \{\emptyset, \{a\}\}, \{\emptyset, \{b\}\}, \{\{\emptyset\}, \emptyset\}, \{\{\emptyset\}, \{\emptyset\}\}, \{\{\emptyset\}, \{a, b\}\} \)  
   (iii) \( \{(b, c), \{(a, c), (b, c)\}\} \)

4. ii, iv

5. 16; 4

6. \( [a \mapsto \text{no}, b \mapsto \text{no}, c \mapsto \text{no}] : \emptyset \)  
   \( [a \mapsto \text{yes}, b \mapsto \text{no}, c \mapsto \text{no}] : \{a\} \)  
   \( [a \mapsto \text{no}, b \mapsto \text{yes}, c \mapsto \no] : \{b\} \)  
   \( [a \mapsto \text{yes}, b \mapsto \text{yes}, c \mapsto \no] : \{a, b\} \)  
   \( [a \mapsto \text{yes}, b \mapsto \text{no}, c \mapsto \yes] : \{a, c\} \)  
   \( [a \mapsto \text{no}, b \mapsto \yes, c \mapsto \yes] : \{b, c\} \)  
   \( [a \mapsto \text{yes}, b \mapsto \yes, c \mapsto \yes] : \{a, b, c\} \)

7. \( \{\text{left} \mapsto a, \text{right} \mapsto a\} : \langle a, a \rangle \)  
   \( \{\text{left} \mapsto a, \text{right} \mapsto b\} : \langle a, b \rangle \)  
   \( \{\text{left} \mapsto a, \text{right} \mapsto c\} : \langle a, c \rangle \)  
   \( \{\text{left} \mapsto b, \text{right} \mapsto a\} : \langle b, a \rangle \)  
   \( \{\text{left} \mapsto b, \text{right} \mapsto b\} : \langle b, b \rangle \)  
   \( \{\text{left} \mapsto b, \text{right} \mapsto c\} : \langle b, c \rangle \)  
   \( \{\text{left} \mapsto c, \text{right} \mapsto a\} : \langle c, a \rangle \)  
   \( \{\text{left} \mapsto c, \text{right} \mapsto b\} : \langle c, b \rangle \)  
   \( \{\text{left} \mapsto c, \text{right} \mapsto c\} : \langle c, c \rangle \)

8. (i) \( [[\emptyset \mapsto \emptyset], [\emptyset \mapsto \emptyset], [\emptyset \mapsto \emptyset]] : [\emptyset \mapsto \emptyset], [\emptyset \mapsto \emptyset], [\emptyset \mapsto \emptyset] \)  
   (ii) \( [[\langle a, c \rangle \mapsto 1], \langle b, c \rangle \mapsto 1], \langle a, c \rangle \mapsto 1], \langle b, c \rangle \mapsto 2] \)  
   (iii) \( [[\langle a \mapsto c \mapsto 1\rangle, \emptyset \mapsto [c \mapsto 1], \emptyset \mapsto [c \mapsto 1]], \emptyset \mapsto [c \mapsto 2], \emptyset \mapsto [c \mapsto 2], \emptyset \mapsto [c \mapsto 2]] \)

9. \( [a \mapsto \{c \mapsto 1, d \mapsto 1\}, b \mapsto \{c \mapsto 2, d \mapsto 1\}] \)

10. (i) \( \{b \mapsto b, c \mapsto a\}, \{b \mapsto b, c \mapsto b\} \)  
   (ii) \( \{b, c, d\}, \{a, c, d\}, \{a, b, d\}, \{a, b, c, d\} \)  
   (iii) \( \{a, b\}, \{a, c\}, \{a, d\}, \{b, c\}, \{b, d\}, \{c, b\}, \{c, d\} \)
CHAPTER 2

MEANING AND FORM

This chapter introduces some of the key notions about the analysis of meaning in formal semantics. We focus on entailments: relations between premises and valid conclusions expressed as natural language sentences. Robust intuitions about entailment are distinguished from weaker types of reasoning with language. Speaker judgments on entailments are described using models: abstract mathematical structures, which emanate from semantic analyses of artificial logical languages. Model-theoretical objects are directly connected to syntactic structures by applying a general principle of compositionality. We see how this principle helps to analyze cases of structural ambiguity and to distinguish them from other cases of under-specification.

What do dictionaries mean when they tell us that semantics is “the study of meaning”? The concept that people intuitively refer to as “meaning” is an abstraction inspired by observing how we use language in everyday situations. However, we use language for many different purposes, and those various usages may inspire conceptions of meaning that are radically different from one another. We cannot reasonably expect a theory of meaning to cover everything that people do with their languages. A more tractable way of studying meaning is by discerning specific properties of language use that are amenable to scientific investigation. These aspects of language use, if stable across speakers and situations, will ultimately guide us toward a theory of language “meaning”.

ENTAILMENT

One of the most important usages of natural language is for everyday reasoning. For example, let us consider sentence (2.1):

(2.1) Tina is tall and thin.
From this sentence, any English speaker is able to draw the conclusion in (2.2) below:

(2.2) Tina is thin.

Thus, any speaker who considers sentence (2.1) to be true, will consider sentence (2.2) to be true as well. We say that sentence (2.1) entails (2.2), and denote it (2.1)⇒(2.2). Sentence (2.1) is called the premise, or antecedent, of the entailment. Sentence (2.2) is called the conclusion, or consequent.

The entailment from (2.1) to (2.2) exemplifies a relation that all English speakers will agree on. This consistency is remarkable, and all the more so since words like Tina, tall and thin are notoriously flexible in the way that they are used. For instance, you and I may have different criteria for characterizing people as being thin, and therefore disagree on whether Tina is thin or not. We may also disagree on the identity of Tina. You may think that Tina in sentences (2.1) and (2.2) is Tina Turner, while I may think that these sentences describe Tina Charles. However, we are unlikely to disagree on whether sentence (2.2) is a sound conclusion from (2.1).

We noted that when sentence (2.1) is judged to be true, so is sentence (2.2). However, the converse does not hold: (2.2) may be true while (2.1) is not – this is the case if Tina happens to be thin but not tall. Because of such situations, we conclude that sentence (2.2) does not entail (2.1). This is denoted (2.2)≰(2.1). Just as with positive judgments on entailment, rejections of entailment are also often uniform across speakers and circumstances of use. Therefore, we consider both positive and negative judgments on entailment as important empirical evidence for semantic theory.

When studying simple entailments, we often pretend that our language vocabulary is very small. Still, as soon as our vocabulary has some simple adjectives and proper names, we can easily find entailments and non-entailments by looking at their different combinations with words like and, or, is and not. For instance:

(2.3) a. Tina is tall, and Ms. Turner is not tall ⇒ Tina is not Ms. Turner.

b. Tina is tall, and Tina is not Ms. Turner ≮ Ms. Turner is not tall.
(2.4)  
a. Ms. Turner is tall, and Tina is Ms. Turner or Ms. Charles  
⇒ Tina is tall or Tina is Ms. Charles.
b. Ms. Turner is tall, and Tina is Ms. Turner or Ms. Charles  
⇒ Tina is tall.

The examples above may look unsurprising for anyone who is familiar with philosophical or mathematical logic. Indeed, similar entailments in natural language have inspired well-known logical formalisms like Propositional Logic and Predicate Logic. Readers may therefore wonder: don’t the entailments above demonstrate puzzles that were solved long ago by logicians? The answer is “yes and no”. Sure enough, these entailments can be translated to well-understood logical questions. However, logic does not traditionally focus on the details of the translation procedure from ordinary language to logical languages. This translation step is not “pure logic”: it also involves intricate questions about the sounds and the forms of human languages, and about the nature of semantic concepts in the human mind. Consequently, in modern cognitive science, the study of entailment in natural language is not the sanctuary of professional logicians. Entailment judgments bring to the forefront a variety of questions about language that are also of primary concern for linguists, computer scientists, psychologists and philosophers. For instance, let us consider the following entailments:

(2.5)  
a. Sue only drank half a glass of wine ⇒ Sue drank less than one glass of wine.
b. A dog entered the room ⇒ An animal entered the room.
c. John picked a blue card from the pack ⇒ John picked a card from the pack.

The entailments in (2.5) illustrate different aspects of language: measures and quantity in (2.5a); word meaning relations in (2.5b); adjective modification in (2.5c). These kinds of entailment are very common in natural language, but they were not systematically treated in classical logic. By studying the whole spectrum of entailments in ordinary language, formal semantics addresses various aspects of linguistic phenomena and their connections with human reasoning. Ideas from
logic are borrowed insofar as they are useful for analyzing natural language semantics. More specifically, later in this book we adopt concepts from type theory and higher-order logics that have proved especially well suited for studying entailment in natural language. As we shall see, incorporating these concepts allows formal semantics to develop important connections with theories about sentence structure and word meaning.

Among the phenomena of reasoning in language, entailment is especially central because of its remarkable stability. In other instances of reasoning with natural language, conclusions are not fully stable, since they may rely on implicit assumptions that emanate from context, world knowledge or probabilistic principles. These lead to meaning relations between sentences that are often fuzzier and less regular than entailments. Consider for instance the following two sentences:

(2.6) Tina is a bird.

(2.7) Tina can fly.

Sentence (2.7) is a likely conclusion from (2.6), and most speakers will not hesitate too much before drawing it. However, upon some reflection we can come up with many situations in which sentence (2.6) truthfully holds without supporting the conclusion in (2.7). Think of young birds, penguins, ostriches, or birds whose wings are broken. Thus, in many natural discourses speakers may accept (2.6) while explicitly denying (2.7):

(2.8) Tina is a bird, but she cannot fly, because ... (she is too young to fly, a penguin, an ostrich, etc.).

We classify the inferential relation between sentences like (2.6) and (2.7) as defeasible, or cancelable, reasoning. By contrast, entailments are classified as indefeasible reasoning: all of the assumptions that are needed in order to reach the conclusion of an entailment are explicitly stated in the premise. For instance, the entailment (2.1)⇒(2.2) cannot be easily canceled by adding further information to the premise.
A discourse like (2.9) below that tries to contradict sentence (2.2) after asserting (2.1) will normally be rejected as **incoherent**.

(2.9) #Tina is tall and thin, but she is not thin.

A speaker who wishes to support this incoherent line of reasoning would need to resort to self-contradictory or otherwise counter-communicative arguments like “because I am lying when saying that Tina is tall and thin”, or “because I am not using English in the ordinary way”. The incoherence of (2.9) is marked by ‘#’. Sentence (2.9) is intelligible but nonsensical: its communicative value is dubious. This sort of incoherent, contradictory sentence should be distinguished from **ungrammaticality**. The latter notion is reserved for strings of words that clearly do not belong to natural language, e.g. *is and tall Tina thin*. Such ungrammatical strings are standardly marked by ‘*’.

Taking stock, we adopt the following notion of entailment:

**Given an indefeasible relation between two natural language sentences** $S_1$ **and** $S_2$, **where speakers intuitively judge** $S_2$ **to be true whenever** $S_1$ **is true**, we say that $S_1$ **entails** $S_2$, **and denote it** $S_1 \Rightarrow S_2$.

Just as intuitive judgments about sentence grammaticality have become a cornerstone in syntactic theory, intuitions about entailments between sentences are central for natural language semantics. As in other linguistic domains, we aim to build our semantic theory on judgments that do not rely on training in linguistics, logic or other scholarly disciplines. Entailments that robustly appear in ordinary reasoning give us a handle on common semantic judgments about language.

Entailments between sentences allow us to define the related notion of **equivalence**. For instance, the sentence (2.1)=*$Tina is tall and thin$ and the sentence $S=*Tina is tall and Tina is thin$ are classified as equivalent, because they entail each other. We denote this equivalence $(2.1) \Leftrightarrow S$. For more examples of equivalent sentences see Exercise 4. Another classical semantic notion is **contradiction**, which was lightly touched upon in our discussion of sentence (2.9) above. See Exercise 7 for some elaboration on contradictions and their relations to entailment.
MODELS AND THE TRUTH-CONDITIONALITY CRITERION

With this background on entailments in natural language, let us now see how formal semantics accounts for them. As mentioned above, formal semantics relies on some central principles from traditional philosophical and mathematical logic. Most versions of formal semantics account for entailments using theoretical structures that are called models. Models are mathematical abstractions that we construct and use as descriptions of hypothetical situations. We call these situations “hypothetical” because they do not necessarily correspond to actual situations in the world. Some of our models may agree with how we look at the world, but some of them will also describe situations that are purely imaginary. For instance, the models that we use for analyzing the sentence Tina is thin will describe situations in which Tina is thin, as well as situations where she is not thin. If you know a woman called Tina and you think that she is thin, you will consider the first models as closer to reality than the others. However, this is irrelevant for our purposes. For the sake of our semantic analysis we consider all the models that we construct as hypothetical. As such, they are all equal.

In order to encode hypothetical situations in models, we let models link words to abstract mathematical objects. For instance, since we want our models to describe situations in relation to the word Tina, we let each model contain some or other abstract entity that is associated with this word. Similarly, when we analyze the words tall and thin, we also let our models associate these adjectives with abstract objects. In this chapter we let models link adjectives to sets of entities. Thus, in each model we include a set of entities that is associated with the word tall. These are the abstract entities in that model that are considered tall in the hypothetical situation that the model describes. Similarly, each model associates the adjective thin with the set of entities that are considered thin in the situation.

In addition to dealing with words, models also treat complex expressions: phrases and sentences that are made up of multiple words. For example, let us consider the complex phrase tall and thin. Just like we did in the case of simple adjective words, we let each model associate this phrase with a set of entities. These are the entities that are considered to be tall and thin in the hypothetical situation that the
model describes. Other words, phrases and sentences are associated with all sorts of abstract mathematical objects. The words, phrases and sentences that we treat are collectively referred to as expressions.

In each of our models, we associate abstract objects with all the expressions that we treat.

Summarizing, we state our general conception of models as follows:

A model is an abstract mathematical structure that we construct for describing hypothetical situations. Models are used for analyzing natural language expressions (words, phrases and sentences) by associating them with abstract objects.

Associating language expressions with abstract objects is part and parcel of a model definition. For instance, one of the models that we use, call it \( M \), may associate the word Tina with some abstract entity \( a \). In technical terms, we say that in the model \( M \), the word Tina denotes the entity \( a \). In all the models that we study in this chapter, the name Tina denotes some or other entity, and the adjective tall denotes some or other set of entities. Given a particular model \( M \), we refer to those denotations as \( \llbracket \text{Tina} \rrbracket^M \) and \( \llbracket \text{tall} \rrbracket^M \), respectively. Similarly, \( \llbracket \text{tall and thin} \rrbracket^M \) is the denotation of the phrase tall and thin in the model \( M \). In general, we adopt the following notational convention:

Let \( \exp \) be a language expression, and let \( M \) be a model. We write \( \llbracket \exp \rrbracket^M \) when referring to the denotation of \( \exp \) in the model \( M \).

To have a more concrete view on models and denotations, let us consider Figure 2.1. This figure describes two models, each of them containing three entities: \( a, b \) and \( c \). In model \( M_1 \), the name Tina denotes the entity \( a \), and the adjective thin denotes the set \( \{a, b\} \). In model \( M_2 \), the denotation of Tina is again the entity \( a \), but this time, the set denotation of thin is the set \( \{b, c\} \). We formally write it as follows:

\[
\llbracket \text{Tina} \rrbracket^{M_1} = a \quad \llbracket \text{thin} \rrbracket^{M_1} = \{a, b\} \\
\llbracket \text{Tina} \rrbracket^{M_2} = a \quad \llbracket \text{thin} \rrbracket^{M_2} = \{b, c\}
\]
Figure 2.1 Models map words and other expressions to abstract mathematical objects. $M_1$ and $M_2$ are models with an entity denotation of Tina and a set denotation of thin. The arrows designate the mappings from the words to their denotations, which are part of the model definition.

Figure 2.1 only illustrates the assignment of denotations to simple words. However, as mentioned above, models are used in order to assign denotations to all expressions that we analyze, including complex expressions that are made of multiple words. In particular, models specify denotations for sentences. There are various ideas about what kinds of abstract objects sentences should denote. In most of this book, we follow the traditional assumption that sentences denote the two abstract objects known as truth-values, which are referred to as ‘true’ and ‘false’. In more technical notation, we sometimes write ‘$\top$’ for ‘true’ and ‘$\bot$’ for ‘false’. Yet another convention, which is most convenient for our purposes, is to use the number 1 for ‘true’ and the number 0 for ‘false’.

Models assign truth-value denotations to sentences on the basis of the denotations they assign to words. For instance, the way we use models such as $M_1$ and $M_2$ in Figure 2.1 respects the intuition that the sentence Tina is thin is true in $M_1$ but false in $M_2$. Thus, we will make sure that $M_1$ and $M_2$ satisfy:

$$[[\text{Tina is thin}]]^{M_1} = 1 \quad [[\text{Tina is thin}]]^{M_2} = 0$$

As we move on further in this chapter, we see how this analysis is formally obtained.

The truth-value denotations that models assign to sentences are the basis for our account of entailment relations. Let us return to the entailment between the sentence Tina is tall and thin (=(2.1)) and the sentence Tina is thin (=(2.2)). When discussing this entailment, we
informally described our semantic judgment by saying that whenever sentence (2.1) is true, sentence (2.2) must be true as well. By contrast, we observed that the intuition does not hold in the other direction: when (2.2) is true, sentence (2.1) may be false. For this reason we intuitively concluded that sentence (2.2) does not entail (2.1). When analyzing (non-)entailment relations, we take into account these pre-theoretical intuitions about ‘truth’ and ‘falsity’. We analyze an entailment $S_1 \Rightarrow S_2$ by introducing the following requirement: if a model lets $S_1$ denote true, it also lets $S_2$ denote true. When truth-values are represented numerically, this requirement means that if a model lets $S_1$ denote the number 1, it also lets $S_2$ denote 1.

Specifically, in relation to the entailment (2.1)⇒(2.2), we require that for every model where sentence (2.1) denotes the value 1, sentence (2.2) denotes 1 as well. Another way to state this requirement is to say that in every model, the truth-value denotation of (2.1) is less than or equal to the denotation of (2.2). Let us see why this is indeed an equivalent requirement. First, consider models where the denotation of (2.1) is 1. In such models, we also want the denotation of (2.2) to be 1. Indeed, requiring $[[S_1]] \leq [[S_2]]$ boils down to requiring that $[[S_2]]$ is 1: this is the only truth-value for (2.2) that satisfies $1 \leq [[S_2]]$. Further, when we consider models where the denotation of (2.1) is 0, we see that such models trivially satisfy the requirement $0 \leq [[S_2]]$, independently of the denotation of (2.2).

To conclude: saying that $[[S_2]]$ is 1 in every model where $[[S_1]]$ is 1 amounts to saying that $[[S_1]] \leq [[S_2]]$ holds in every model. Accordingly, we translate our intuitive analysis of the entailment (2.1)⇒(2.2) to the formal requirement that $[[S_1]] \leq [[S_2]]$ holds in every model. More generally, our aim is to account for entailments using the ≤ relation between truth-values in models. This leads to a central requirement from formal semantic theory, which we call the **truth-conditionality criterion (TCC):**

\begin{quote}
A semantic theory $T$ satisfies the **truth-conditionality criterion** (TCC) for sentences $S_1$ and $S_2$ if the following two conditions are equivalent:

(I) Sentence $S_1$ intuitively entails sentence $S_2$.

(II) For all models $M$ in $T$: $[[S_1]]^M \leq [[S_2]]^M$.
\end{quote}
Figure 2.2 The TCC matches judgments on entailment with the $\leq$ relation. When the entailment $S_1 \Rightarrow S_2$ holds, the $\leq$ relation is required to hold between the truth-value denotations of $S_1$ and $S_2$ in all the models of the theory.

Table 2.1: Does $x \leq y$ hold?

<table>
<thead>
<tr>
<th>x</th>
<th>y=0</th>
<th>y=1</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>yes</td>
<td>yes</td>
</tr>
<tr>
<td>1</td>
<td>no</td>
<td>yes</td>
</tr>
</tbody>
</table>

Clause (I) of the TCC postulates an entailment between sentences $S_1$ and $S_2$. This is an empirical statement about the semantic intuitions of native speakers. By contrast, clause (II) is a statement about our theory’s treatment of sentences $S_1$ and $S_2$: the formal models we use indeed rely on intuitive notions of ‘truth’ and ‘falsity’, but they are purely theoretical. By imposing a connection between the empirical clause (I) and the theoretical clause (II), the TCC constitutes an adequacy criterion for semantic theory.

By way of recapitulation, Figure 2.2 illustrates how the TCC emulates the intuitive relation of entailment (‘$\Rightarrow$’) between sentences, by imposing the $\leq$ relation on sentence denotations in the models that our theory postulates. Table 2.1 summarizes how the requirement $x \leq y$ boils down to requiring that if $x$ is 1, then $y$ is 1 as well. Readers who are familiar with classical logic may observe the similarity between Table 2.1 and the truth table for implication. We will get back to this point in Chapter 5.
ELEMENTS OF FORMAL SEMANTICS

ARBITRARY AND CONSTANT DENOTATIONS

We have introduced the TCC as a general criterion for the empirical adequacy of formal semantics. Obviously, we want our theory to respect the TCC for as many intuitive (non-)entailments as possible. This will be our main goal throughout this book. Let us start our investigations by using the TCC to explain our favorite simple entailment: (2.1)⇒(2.2). We want to make sure that the models that we construct respect the TCC for this entailment. Thus, we need to define which models we have in our theory, and then check what truth-values each model derives for sentences (2.1) and (2.2). The models that we define fix the denotations of words like *Tina*, *tall* and *thin*. In technical jargon, words are also called *lexical items*. Accordingly, we will refer to the denotations of words as *lexical denotations*. Based on the lexical denotations that models assign to words, we will define the truth-values assigned to sentences containing them. To do that, let us explicitly state the assumptions that we have so far made about our models:

1. In every model $M$, in addition to the two truth-values 0 and 1, we have an arbitrary non-empty set $E^M$ of the entities in $M$. We refer to this set as the *domain of entities* in $M$. For instance, in models $M_1$ and $M_2$ of Figure 2.1, the entity domains $E^{M_1}$ and $E^{M_2}$ are the same: in both cases they are the set $\{a, b, c\}$.

2. In any model $M$, the proper name *Tina* denotes an arbitrary entity in the domain $E^M$ (cf. Figure 2.1).

3. In any model $M$, the adjectives *tall* and *thin* denote arbitrary sets of entities in $E^M$ (cf. Figure 2.1).

When the model is understood from the context, we often write $E$ for the domain of entities, suppressing the subscript $M$. We say that the domains of entities in the models we define are ‘arbitrary’ because we do not make any special assumptions about them: *any* non-empty set may qualify as a possible domain of entities in some model. Accordingly, we also treat the entity denotation of *Tina* as an arbitrary element of $E$. Whether this entity corresponds to a real-life entity like Tina Turner or Tina Charles is not our business here. We are not even insisting that the entity for *Tina* has ‘feminine’ properties, as might be suitable for a feminine English name. All we require is that in every model, the name *Tina* denotes *some* entity. In a similar fashion, we let
the adjectives \textit{tall} and \textit{thin} denote arbitrary sets in our models. This arbitrariness is of course an over-simplification, but it will do for our purposes in this book. Here we study the meanings of words only to the extent that they are relevant for the study of entailment. Of course, much more should be said on word meanings. This is the focus of research in \textit{lexical semantics}, which deals with many other important aspects of word meaning besides their contribution to entailments. For some readings on this rich domain, see some recommendations at the end of this chapter.

From now on we will often use words in \textbf{boldface} when referring to arbitrary denotations. For instance, by \textit{‘tina’} we refer to the element $[[\text{Tina}]]$ of $E$ that is denoted by the word \textit{Tina} in a given model. Similarly, \textit{‘tall’} and \textit{‘thin’} are shorthand for $[[\text{tall}]]$ and $[[\text{thin}]]$: the \textit{sets} of entities in $E$ denoted by the words \textit{tall} and \textit{thin}. Putting words in boldface in this way is a convenience that spares us the use of the double brackets $[[]]$. When we want to be more specific about the model $M$, we write $\text{tina}^M$ or $[[\text{Tina}]]^M$.

In our discussion of Figure 2.1, we noted that the sentence \textit{Tina is thin} is intuitively true in $M_1$ but false in $M_2$. We can now see how this intuition is respected by our precise definition of models. To achieve that, we make sure that the sentences \textit{Tina is thin} reflects a \textit{membership assertion}. We only allow the sentence to be true in models where the entity denoted by \textit{Tina} is a \textit{member} of the set denoted by the adjective. Therefore, we analyze the word \textit{is} as denoting a \textit{membership function}. This is the function sending every entity $x$ and set of entities $A$ to the truth-value 1 if $x$ is an element of $A$. If $x$ is not an element of $A$, the membership function sends $x$ and $A$ to the truth-value 0. When referring to the membership function that the word \textit{is} denotes, we use the notation ‘\text{'is}’. Formally, we define $\text{is}$ as the function that satisfies the following, for every entity $x$ in $E$ and every subset $A$ of $E$:

$$\text{is}(x, A) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$$

For example, let us reconsider the models $M_1$ and $M_2$ that we saw in Figure 2.1. With our new assumption on the denotation of the word \textit{is}, we now get:

In $M_1$: $[[\text{Tina is thin}]] = \text{is}(\text{tina}, \text{thin}) = \text{is}(a, \{a, b\}) = 1$ \textit{since} $a \in \{a, b\}$

In $M_2$: $[[\text{Tina is thin}]] = \text{is}(\text{tina}, \text{thin}) = \text{is}(a, \{b, c\}) = 0$ \textit{since} $a \notin \{b, c\}$
Thus, in $M_1$, the sentence $\text{Tina is thin}$ denotes 1, and in $M_2$ it denotes 0, as intuitively required. More generally, in (2.10) below we summarize the denotation that the sentence $\text{Tina is thin}$ is assigned in every model:

\[(2.10) \quad \llbracket \text{Tina is thin} \rrbracket^M = \text{is}(\text{tina}, \text{thin}) = \begin{cases} 1 & \text{if } \text{tina} \in \text{thin} \\ 0 & \text{if } \text{tina} \notin \text{thin} \end{cases} \]

When referring to denotations, we have made a difference between the font for the denotations $\text{tina}$, $\text{tall}$ and $\text{thin}$, and the font for the denotation $\text{is}$. There is a reason for this notational difference. As mentioned, the denotations of the words $\text{Tina}$, $\text{tall}$ and $\text{thin}$ are arbitrarily chosen by our models. We have presented no semantic ‘definition’ for the meaning of these words. Models are free to let the name $\text{Tina}$ denote any of their entities. Similarly, the adjectives $\text{tall}$ and $\text{thin}$ may denote any of set of entities. By contrast, the denotation of the word $\text{is}$ has a constant definition across models: in all models we define this denotation as the membership function. We will have more to say about this distinction between denotations in Chapter 3. In the meantime, let us summarize our notational conventions:

\[
\text{Let blik be a word in a language. When the denotation } \llbracket \text{blik} \rrbracket^M \text{ of blik is arbitrary, we mark it blik. When it has a constant definition across models we mark it BLIK.}
\]

**ANALYZING AN ENTAILMENT**

In order to analyze the entailment $(2.1) \Rightarrow (2.2)$, let us now also look at the denotation of the sentence $\text{Tina is tall and thin}$. Since we let the sentence $\text{Tina is thin}$ denote the truth-value of a membership assertion, it is only natural to analyze the sentence $\text{Tina is tall and thin}$ in a similar way. Thus, we want this sentence to be true if the entity $\text{tina}$ is a member of a set denoted by the conjunction $\text{tall and thin}$. But what should this set be? The same semantic intuitions that supported the entailment $(2.1) \Rightarrow (2.2)$ can guide us to the answer. Obviously, for Tina to be tall and thin, she has to be tall, and she also has to be thin. And vice versa: if Tina is tall, and if in addition she is also thin, there is no way to avoid the conclusion that she is tall and thin. Elementary as they are, these considerations suggest that if we are going to let the
conjunction \textit{tall and thin} denote a set, it had better be the \textit{intersection} of the two sets for the adjectives \textit{tall} and \textit{thin}. Formally, we write:

\[
[[\text{tall and thin}]]^M = [[\text{tall}]]^M \cap [[\text{thin}]]^M = \text{tall} \cap \text{thin}.
\]

Thus, we define the denotation of the word \textit{and} to be the \textit{intersection} function over $E$. This is the function $\text{AND}$ that satisfies the following, for all subsets $A$ and $B$ of $E$:

\[
\text{AND}(A, B) = A \cap B
\]

= the set of all members of $E$ that are both in $A$ and in $B$

Now there is also no doubt about the denotation of the sentence \textit{Tina is tall and thin}. Using the same kind of membership assertion that we used for the sentence \textit{Tina is thin}, we reach the following denotation for this sentence:

\begin{equation}
[[\text{Tina is tall and thin}]]^M = \text{IS}(\text{tina}, \text{AND}(\text{tall}, \text{thin}))
\end{equation}

\[
= \begin{cases} 
1 & \text{if } \text{tina} \in \text{tall} \cap \text{thin} \\
0 & \text{if } \text{tina} \notin \text{tall} \cap \text{thin}
\end{cases}
\]

In words: in every given model $M$, the sentence \textit{Tina is tall and thin} denotes the truth-value 1 if the entity \textit{tina} is in the intersection of the sets \textit{tall} and \textit{thin}; otherwise the sentence denotes 0.

We have now defined the truth-value denotations that the sentences \textit{Tina is tall and thin} and \textit{Tina is thin} have in every model. These are the truth-values specified in (2.11) and (2.10), respectively. Therefore, we can use the TCC in order to verify that our theory adequately describes the entailment between the two sentences. As a matter of set theory, the truth-value (2.10) must be 1 if the truth-value in (2.11) is 1: if the entity \textit{tina} is in the intersection \textit{tall} $\cap$ \textit{thin}, then, by definition of intersection and set membership, it is also in the set \textit{thin}. This set-theoretical consideration holds for all possible denotations \textit{tina}, \textit{tall} and \textit{thin}. Thus, it holds for all models. This means that our assignment of denotations to sentences (2.1) and (2.2) has been successful in meeting the TCC when accounting for the entailment between them.

At this point you may feel that the games we have been playing with entities, sets, functions and truth-values are just restating obvious intuitions. This is perfectly true. Indeed, there is reason to feel satisfied
about it. Semantic models provide us with a general and mathematically rigorous way of capturing common intuitions about entailment. A model is a small but precise description of a particular situation in which different sentences may be true or false. By specifying the denotations of the words Tina, tall and thin, a model describes, in an abstract way, who Tina is, and what the tall entities and thin entities are. As we have seen, and as we shall see in more detail throughout this book, models also take care of more “logical” denotations for words like and and is. This assignment of denotations to lexical items enables us to systematically assign denotations to complex expressions, including conjunctive phrases like tall and thin and sentences like Tina is tall and thin. If we are successful in assigning denotations to such complex expressions, we may be reasonably hopeful that our strategies will also be useful for much higher levels of hierarchical embedding (e.g. Dylan’s description of the sad-eyed lady on page 1). In fact, by defining truth and falsity in models for two simple sentences, we have been forced to dive rather deep into the meanings of conjunction, predication, adjectives and proper names, and the ways in which they combine with each other. As we shall see in the following chapters, much of our elementary set-theoretical maneuvering so far is valuable when tackling more advanced questions in formal semantics.

When looking at a class of models that is heterogenous enough, we can “see”, so to speak, whether one sentence must denote 1 when another sentence does. Let us get a feel of what is going on in the simple example we have been treating, by playing a little with some concrete models. Let us consider Table 2.2, which summarizes our assumptions so far and illustrates them concretely in the three models described in the rightmost columns. Each of these models has the set $E = \{a, b, c, d\}$ as its domain of entities. In model $M_1$, the word Tina denotes the entity a, and the word thin denotes the set of three entities \{a, b, c\}. Model $M_2$ assigns different denotations to these words: Tina denotes the entity b, and thin denotes the set \{b, c\}. In model $M_3$, the denotation of Tina remains the entity b, as in model $M_2$, while the adjective thin denotes a set of three entities: \{a, c, d\}. Accordingly, the truth-values in the three models for the sentence Tina is thin are 1, 1 and 0, respectively. Similarly, using the assumed denotations for tall, we can also verify that the truth-values in these three models for the sentence Tina is tall and thin are 0, 1 and 0, respectively. Satisfying
MEANING AND FORM

Table 2.2: Denotations for expressions in the entailment (2.1)⇒(2.2).

<table>
<thead>
<tr>
<th>Expression</th>
<th>Cat.</th>
<th>Type</th>
<th>Abstract denotation</th>
<th>Denotations in example models with $E = {a, b, c, d}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Tina</td>
<td>PN</td>
<td>entity</td>
<td>tina</td>
<td>$M_1 = {a}$, $M_2 = {b}$, $M_3 = {b}$</td>
</tr>
<tr>
<td>tall</td>
<td>A</td>
<td>set of entities</td>
<td>tall</td>
<td>$M_1 = {b, c}$, $M_2 = {b}$, $M_3 = {a, b, d}$</td>
</tr>
<tr>
<td>thin</td>
<td>A</td>
<td>set of entities</td>
<td>thin</td>
<td>$M_1 = {a, b, c}$, $M_2 = {b}$, $M_3 = {a, c, d}$</td>
</tr>
<tr>
<td>tall and thin</td>
<td>AP</td>
<td>set of entities</td>
<td>AND(tall, thin)</td>
<td>$M_1 = {b, c}$, $M_2 = {b}$, $M_3 = {a, d}$</td>
</tr>
<tr>
<td>Tina is thin</td>
<td>S</td>
<td>truth-value</td>
<td>IS(tina, thin)</td>
<td>$M_1 = 1$, $M_2 = 1$, $M_3 = 0$</td>
</tr>
<tr>
<td>Tina is tall and thin</td>
<td>S</td>
<td>truth-value</td>
<td>IS(tina, AND(tall, thin))</td>
<td>$M_1 = 0$, $M_2 = 1$, $M_3 = 0$</td>
</tr>
</tbody>
</table>

Categories: PN = proper name; A = adjective; AP = adjective phrase; S = sentence.

the TCC means that the latter value must be less than or equal to the former value, which is indeed the case.

Model $M_1$ in Table 2.2 shows that the TCC is also met for the non-entailment (2.2)⇒(2.1). This model makes sentence (2.2) true while making (2.1) false. This means that our theory respects the requirement in the TCC that, if an entailment is missing, then at least one model does not satisfy the $\leq$ relation between the truth-values of the two sentences in question. In formula, model $M_1$ satisfies $\Vert (2.2) \Vert^{M_1} \leq \Vert (2.1) \Vert^{M_1}$. Furthermore, model $M_1$ also respects our intuition of why an entailment is absent in this case. As pointed out above, if somebody tried to convince you that Tina must be tall and thin just because she happens to be thin, you might reasonably object by pointing out the possibility that Tina may not be tall. Model $M_1$ highlights this possibility.

**DIRECT COMPOSITIONALITY**

So far we have been paying little attention to sentence structure. However, as mentioned in the introduction, one of our main interests is how meanings of natural language expressions are related to the syntactic forms of these expressions. For instance, let us consider the following two sentences:

(2.12) a. All pianists are composers, and Tina is a pianist.
  b. All composers are pianists, and Tina is a pianist.
Sentences (2.12a) and (2.12b) contain the same words but in a different order. Consequently, the meanings of these two sentences are rather different. In particular, while (2.12a) entails the sentence *Tina is a composer*, sentence (2.12b) does not. The meaning of an expression is not a soup made by simply putting together the meanings of words. Rather, the order of the words in a complex expression and the hierarchical structures that they form affect its meaning in systematic ways. Since entailment relations between sentences reflect an aspect of their meanings, entailments are also sensitive to sentence structure. In the framework that we assume here, entailments are explained by appealing to model-theoretic denotations. Therefore, the question we are facing is: how are syntactic structures used when defining denotations of complex expressions? The general principle known as compositionality provides an answer to this question. According to this principle, the denotation of a complex expression is determined by the denotations of its immediate parts and the ways they combine with each other. For instance, in our analysis of the entailment (2.1)⇒(2.2), we treated the denotation of sentence (2.1) (*Tina is tall and thin*) as derived step by step from the denotations of its parts: the name *Tina*, the verb *is*, and the adjective phrase *tall and thin*. Figure 2.3 summarizes our compositional analysis.

Figure 2.3A shows the syntactic part-whole relations that we assume for the sentence. In this structure we group together the string of words *tall and thin* into one adjectival phrase, which we denote *AP*. More generally, tree diagrams as in Figure 2.3A represent the sentence’s constituents: the parts of the sentence that function as grammatical...
units. In this case, besides the sentence itself and the words it contains, the only syntactic constituent assumed is the adjectival phrase *tall and thin*. Figure 2.3B describes how denotations of constituents are derived from the denotations of their immediate parts. The denotation of the adjectival phrase *tall and thin* is determined by combining the denotations of its immediate parts: *tall*, *and* and *thin*. The denotation of the whole sentence is determined by the denotations of its parts: *tina*, *is*, and *and(* *tall, thin*)*. What we get as a result is the truth-value denotation in (2.11). The way in which this truth-value is derived is sanctioned by the compositionality principle on the basis of the structure in Figure 2.3A. Note that compositionality would not allow us to derive the truth-value in any other order. For instance, on the basis of the structure in Figure 2.3A, we would not be able to compositionally define the denotation of the whole sentence directly on the basis of the denotations of the adjectives *tall* and *thin*. These words are not among the sentence’s immediate parts. Therefore, according to the compositionality principle, they can only indirectly affect its denotation.

Summarizing, we have adopted the following general principle, and seen how we follow it in our analysis of the entailment (2.1)⇒(2.2).

**Compositionality**: The denotation of a complex expression is determined by the denotations of its immediate parts and the ways they combine with each other.

A word of clarification should be added here about the role of semantic formulas in our analysis. Consider for instance the formula IS( *tina, and(* *tall, thin*)*) that we derive in Figure 2.3B. This formula is not a representation of some abstract meaning, independent of the sentence structure. To the contrary, this formula is almost completely identical to the structure in Figure 2.3A, while adding only the necessary semantic details for describing how the denotation is derived for sentence (2.1). Most importantly, the formula specifies the function-argument relations between the denotations of the sentence constituents. Thus, the formula IS( *tina, and(* *tall, thin*)*) is simply the syntactic bracketing [Tina is [tall and thin]] imposed by the tree in Figure 2.3A, with two modifications: (i) symbols for words are replaced by symbols of their denotations; and (ii) symbols for denotations may
be shuffled around in order to follow the convention of putting function symbols to the left of their arguments. This respects the highly restrictive nature of compositional analysis: the process only requires a syntactic structure, denotations of lexical items, and a way to glue the denotations together semantically. The real “semantic action” is within these three components of the theory, not within the semantic formulas we use. This version of compositionality is sometimes referred to as direct compositionality. In this paradigm, denotations in the model are directly derived from syntactic structures, with no intermediate level of semantic or logical representation.

**STRUCTURAL AMBIGUITY**

Direct compositionality helps to clarify an important issue in linguistic theory: the phenomenon of structural ambiguity. Consider the following sentence:

(2.13) Tina is not tall and thin.

Let us consult our intuitions with respect to the following question: does (2.13) entail sentence (2.2) (=Tina is thin) or not? This is much harder to judge than in the case of the entailment (2.1)⇒(2.2). However, there is a common intuition that (2.13) entails (2.2), but only under particular usages. A speaker who wishes to convey the entailment (2.13)⇒(2.2) can do so by stressing the prosodic boundary after the word tall:

(2.14) Tina is not tall, and thin.

Without such an intonational pattern, a speaker can also use (2.13) felicitously for describing a situation where Tina is not thin. For instance, if we tell Sue that Tina is tall and thin, she may use (2.13) for denying the assertion, by saying something like:

(2.15) Come on, that isn’t true! Tina is not tall and thin: although she is indeed very tall, you couldn’t possibly think of her as thin!

In this reaction, the way in which Sue uses sentence (2.13) clearly indicates that she does not consider it to entail sentence (2.2).
Because the two usages of sentence (2.13) show differences between its “entailment potential”, we say that it is ambiguous. The “comma intonation” in (2.14) disambiguates the sentence. Another way of disambiguating sentence (2.13) is illustrated in (2.15), using the context of our conversation with Sue. In the former disambiguation, sentence (2.13) is used for entailing (2.2); in the latter disambiguation, the entailment from (2.13) to (2.2) is blocked. We refer to the two possible usages of sentence (2.13) as readings of the sentence. One reading entails (2.2), the other does not. Another way to describe the “two readings” intuition is to note that sentence (2.13) may be intuitively classified as both true and false in the same situation. Consider a situation where Tina is absolutely not thin. Context (2.15) highlights that sentence (2.13) may be used as true in this situation. By contrast, (2.14) highlights that the sentence also has the potential of being false in the same situation.

The striking thing about the ambiguity of sentence (2.13) is the ease with which it can be described when we assume the compositionality principle. Virtually all syntactic theories analyze sentence (2.13) as having two different syntactic structures, as illustrated in Figure 2.4. A simple phrase structure grammar that generates the structural ambiguity in Figure 2.4 is given in (2.16) below:

\[
\begin{align*}
\text{AP} & \rightarrow \text{tall, thin, ...} \\
\text{AP} & \rightarrow \text{AP and AP} \\
\text{AP} & \rightarrow \text{not AP}
\end{align*}
\]
In words: an adjective phrase (AP) can be a simple adjective, or a conjunction of two other APs, or a negation of another AP. These rules derive both structures in Figure 2.4. When a grammar generates more than one structure for an expression in this way, we say that it treats the expression as \textit{structurally ambiguous}. You may think that the syntactic ambiguity in Figure 2.4 is by itself already an elegant account of the semantic ambiguity in (2.13). However, there is a gap in this account: why does it follow that the structural ambiguity of sentence (2.13) also makes it \textit{semantically} ambiguous? Compositionality provides the missing link. When the two structures in Figure 2.4 are compositionally analyzed, we immediately see that the same model may assign them two different truth-values. Concretely, let us assume that the denotation of the negation word \textit{not} in (2.13) is the \textit{complement function}, i.e. the function \( \text{NOT} \) that maps any subset \( A \) of \( E \) to its complement set:

\[
\text{NOT}(A) = \overline{A} = E - A = \text{the set of all the members of } E \text{ that are not in } A
\]

Figure 2.5 uses the denotation \( \text{NOT} \) for illustrating the compositional analysis of the two structures in Figure 2.4. As Figure 2.5 shows, the compositional process works differently for each of the two structural analyses of sentence (2.13). For each of the denotations in Figures 2.5A and 2.5B to be 1, different requirements have to be satisfied. This is specified in (2.17a) and (2.17b) below:

\[
(2.17) \quad \begin{align*}
\text{a. } \text{IS}(\text{tina}, \text{AND(\text{NOT(tall)}, \text{thin}))}) &= 1 \\
\text{This holds if and only if (iff) } \text{tina} &\in \text{tall} \cap \text{thin}.
\end{align*}
\]

\[
\text{b. } \text{IS}(\text{tina}, \text{NOT(AND(tall, thin))})) &= 1 \\
\text{This holds if and only if } \text{tina} &\in \text{tall} \cap \overline{\text{thin}}.
\]

For the denotation in Figure 2.5A to be 1, the requirement in (2.17a) must hold. When this is the case, the entity \textit{tina} is in the set \textit{thin}, hence the truth-value assigned to the sentence \textit{Tina is thin} (= (2.2)) is also 1. Thus, our compositional analysis of the structure in Figure 2.4A captures the reading of sentence (2.13) that entails sentence (2.2). By contrast, the denotation (2.17b) that is derived in Figure 2.5B does not guarantee that the entity \textit{tina} is in the set \textit{thin}. This is because the entity \textit{tina} may be in the complement set \textit{tall} \cap \overline{\text{thin}} while being in the set \textit{thin}, as long as it is not in the set \textit{tall}. Specifically, consider a model...
Figure 2.5 Compositionality and ambiguity.

\[\text{A. } \ \text{IS(tina, AND(NOT(tall), thin))}\]

\[\text{tina IS AND(NOT(tall), thin)}\]

\[\text{NOT(tall) AND thin}\]

\[\text{NOT tall}\]

\[\text{B. } \ \text{IS(tina, NOT(AND(tall, thin)))}\]

\[\text{tina IS NOT(AND(tall, thin))}\]

\[\text{NOT AND(tall, thin)}\]

\[\text{tall AND thin}\]

Where the entities tina, mary and john are t, m and j, respectively. Suppose further that the model M assigns the following denotations to the adjectives thin and tall:

\[\text{thin} = \{t, j\} \quad \text{tall} = \{m, j\}\]

The model M represents a situation where Tina is thin but not tall, Mary is tall but not thin, and John is both thin and tall. In this model, the denotation in Figure 2.5B is the truth-value 1, but sentence (2.2) denotes the truth-value 0. This means that our compositional analysis of the structure in Figure 2.4B captures the reading of sentence (2.13) that does not entail sentence (2.2).

In compositional systems, the structure that we assume for a sentence strongly affects the entailment relations that our theory expects for it. When a sentence is assumed to be structurally ambiguous, a
compositional theory may assign different truth-values to its different structures. As a result, the theory may expect different entailment relations to hold for the different structures. Accordingly, when speakers are confronted with such a sentence, they are expected to experience what we informally call “semantic ambiguity”, i.e. some systematic hesitations regarding some of the sentence’s entailments. Structural ambiguity is used as the basis of our account of semantic ambiguity. Once we have acknowledged the possibility of ambiguity, we prefer to talk about the entailments that sentence structures show, and of the truth-values that are assigned to these structures. However, for the sake of convenience, we often say that sentences themselves have entailments and truth-values. This convention is harmless when the sentences in question are unambiguous, or when it is clear that we are talking about a particular reading.

Semanticists often distinguish the syntactic-semantic ambiguity of sentences like (2.13) from another type of under-specification, which is called vagueness. For instance, as we noted above, the sentence Tina is tall says little about Tina and her exact height. In some contexts, e.g. if Tina is known to be a Western female fashion model, the sentence may be used for indicating that Tina is above 1.70 meters. In other contexts, e.g. if Tina is known to be member of some community of relatively short people, the sentence may indicate that Tina is above 1.50 meters. However, we do not consider these two usages as evidence that the sentence must be assigned different structures with potentially different denotations. Rather, we say that the sentence Tina is tall is vague with respect to Tina’s height. Further specification of relevant heights is dealt with by augmenting our semantic treatment with a pragmatic theory. Pragmatic theories also consider the way in which sentences are used, and the effects of context on their use. Pragmatic theories aim as well to account for the way speakers resolve (or partly resolve) vagueness in their actual use of language. Classical versions of formal semantics did not aim to resolve vagueness, but current semantic theories often interact with pragmatics and describe the way context helps in resolving vagueness in actual language use.

Vagueness is very prominent in the way natural languages are used, and most sentences may be vague in one way or another. For instance, the sentence Sue is talking tells us nothing about Sue’s voice (loud or quiet, high or low, etc.), what Sue is talking about, who the addressee is, etc. However, upon hearing the sentence, we may often use the
context to infer such information. For instance, suppose that we are at a conference and know that Sue is one of the speakers. In such a context, we may draw some additional conclusions about the subject of Sue’s talk and the addressees. Hearers often use context in this way to extract more information from linguistic expressions, and speakers often rely on their hearers to do that. In distinction from entailment, such inferential processes which are based on contextual knowledge are defeasible. For instance, even when the context specifies a particular conference where Sue is a speaker, we may use the sentence *Sue is talking* to indicate that Sue is talking to a friend over the phone. What we saw in sentence (2.13) is quite different from the defeasible reasoning that helps speakers and hearers in their attempts to resolve vagueness of language utterances. The comma intonation in (2.14) illustrated that one phonological expression of sentence (2.13) indefeasibly entails sentence (2.2). This convinced us that both structures that we assigned to sentence (2.13) are semantically useful. The theoretical consideration was the key for our treatment of the sentence as semantically ambiguous, more than any “pure” linguistic intuition. Most decisions between ambiguity and vagueness involve similar theoretical considerations, rather than the direct judgments of a speaker’s linguistic intuitions.

**FURTHER READING**

**Introductory:** For methodological aspects of logical semantics, including truth-values, entailment and compositionality, see Gamut (1982, vol. 1). For more examples and discussion of structural ambiguity, see Zimmermann and Sternefeld (2013, ch. 3), and, in relation to vagueness, Kennedy (2011). For further discussion of compositionality, see Partee (1984). On defeasible reasoning, see Koons (2014). Levinson (1983) is an introductory textbook on pragmatics. On lexical semantics, see Cruse (1986); Murphy (2010). Meaning relations between words and concepts they refer to are extensively studied in the literature on categorization. See Laurence and Margolis (1999); Smith (1988); Taylor (1989) for introductions of these topics.

**Advanced:** The idea that sentences denote truth-values, and more generally, propositions (cf. Chapter 6), was proposed as central for
communication (Austin 1962; Searle 1969). The centrality of entailment and the model-theoretic TCC was also highlighted in semantic theories of non-indicative sentences, especially *interrogatives* (Groenendijk and Stokhof 1984, 2011). An alternative to the model-theoretic approach to entailment is *proof-theoretic semantics* (Schroeder-Heister, 2014). In its application to natural language, proof-theoretic approaches are sometimes referred to as *natural logic*. Some examples of work in this area are McAllester and Givan (1992); Sánchez (1991); Moss (2010). Defeasible reasoning in language is related to *common sense reasoning* in work in artificial intelligence (Brewka et al. 1997) and cognitive psychology (Stenning and van Lambalgen 2007; Adler and Rips 2008). For more on pragmatic theories, and specifically the notion of *implicature*, see Grice (1975); Geurts (2010); Chierchia et al. (2012). Much of this work pays close attention to the meaning and use of the word *or* (cf. the choice between ‘inclusive’ and ‘exclusive’ denotations in Exercise 6). Direct compositionality in contemporary semantics of natural language was first illustrated in Montague (1970a). For further work on compositionality, see Montague (1970b); Janssen (1983); Janssen with Partee (2011); Barker and Jacobson (2007); Pagin and Westerståhl (2010); Werning et al. (2012).

**EXERCISES (ADVANCED: 4, 5, 6, 7, 8, 9, 10)**

1. In the following pairs of sentences, make a judgment on whether there is an entailment between them, and if so, in which of the two possible directions. For directions in which there is no entailment, describe informally a situation that makes one sentence true and the other sentence false. For example, in the pair of sentences (2.1) and (2.2), we gave the judgment (2.1)⇒(2.2), and supported the non-entailment (2.2)⇏(2.1) by describing a situation in which Tina is thin but not tall.

   (i)  
   a. Tina got a B or a C.  
   b. Tina got a B.

   (ii)  
   a. Tina is neither tall nor thin.  
   b. Tina is not thin.

   (iii)  
   a. Mary arrived.  
   b. Someone arrived.

   (iv)  
   a. John saw fewer than four students.  
   b. John saw no students.

   (v)  
   a. The ball is in the room.  
   b. The box is in the room and the ball is in the box.

   (vi)  
   a. Hillary is not a blond girl.  
   b. Hillary is not a girl.
(vii)  a. Hillary is a blond girl.  b. Hillary is a girl.
(viii) a. Tina is a Danish flutist and a physicist.  b. Tina is a Danish physicist and a flutist.
(ix)  a. Tina is not tall but taller than Mary.  b. Mary is not tall.
(x)  a. Mary ran.  b. Mary ran quickly.
(xi)  a. I saw fewer than five horses that ran.  b. I saw fewer than five black horses that ran.
(xii) a. I saw fewer than five horses that ran.  b. I saw fewer than five animals that ran.
(xiii) a. Exactly five pianists in this room are French composers.  b. Exactly five composers in this room are French pianists.
(xiv) a. No tall politician is multilingual.  b. Every politician is monolingual.
(xv)  a. No politician is absent.  b. Every politician is present.
(xvi) a. At most three pacifists are vegetarians.  b. At most three vegetarians are pacifists.
(xvii) a. All but at most three pacifists are vegetarians.  b. At most three non-vegetarians are pacifists.

2. Each of the following sentences is standardly considered to be structurally ambiguous. For each sentence suggest two structures, and show an entailment that one structure intuitively supports and the other structure does not:
(i) I read that Dan published an article in the newspaper.
(ii) Sue is blond or tall and thin.
(iii) The policeman saw the man with the telescope.
(iv) Rich Americans and Russians like to spend money.
(v) Sue told some man that Dan liked the story.
(vi) Dan ate the lettuce wet.
(vii) Sue didn’t see a spot on the floor.

3. Table 2.2 shows different denotations for the expressions in sentences (2.1) and (2.2) in different models. We used these models and the truth-values they assign to the sentences to support our claim that the TCC explains the entailment (2.1)⇒(2.2), and the non-entailment (2.2)≠⇒(2.1).
The table on the right gives the expressions for the two analyses in Figure 2.5 of the sentence Tina is not tall and thin.
a. Add to this table the missing denotations of these expressions within the three models M1, M2 and M3.
b. Verify that the truth-values that you assigned to the two analyses in Figure 2.5 support the intuition that the analysis in Figure 2.4A entails sentence (2.2), whereas the analysis in Figure 2.4B does not entail (2.2).

<table>
<thead>
<tr>
<th>Expression</th>
<th>Denotations in example models with $E = {a, b, c, d}$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$M_1$</td>
</tr>
<tr>
<td>Tina</td>
<td>a</td>
</tr>
<tr>
<td>tall</td>
<td>{b, c}</td>
</tr>
<tr>
<td>thin</td>
<td>{a, b, c}</td>
</tr>
<tr>
<td>not tall</td>
<td></td>
</tr>
<tr>
<td>[not tall] and thin</td>
<td></td>
</tr>
<tr>
<td>Tina is [not tall] and thin</td>
<td></td>
</tr>
<tr>
<td>tall and thin</td>
<td></td>
</tr>
<tr>
<td>not [tall and thin]</td>
<td></td>
</tr>
<tr>
<td>Tina is [not [tall and thin]]</td>
<td></td>
</tr>
<tr>
<td>Tina is thin</td>
<td></td>
</tr>
</tbody>
</table>

4.a. Mark the pairs of sentences in Exercise 1 that you considered equivalent.

b. Give three more examples for pairs of sentences that you consider intuitively equivalent.

c. State the formal condition that a semantic theory that satisfies the TCC has to satisfy with respect to equivalent sentences.

5. Consider the ungrammaticality of the following strings of words.

(i) *Tina is both tall *Tina is both not tall *Tina is both tall or thin

To account for this ungrammaticality, let us assume that the word both only appears in adjective phrases of the structure both AP$_1$ and AP$_2$. Thus, a constituent both $X$ is only grammatical when $X$ is an and-conjunction of two adjectives or adjective phrases; hence the grammaticality of the string both tall and thin as opposed to the ungrammaticality of the strings in (i), where $X$ is tall, not tall and tall or thin, respectively. We assume further that the denotation of a constituent both AP$_1$ and AP$_2$ is the same as the denotation of the parallel constituent AP$_1$ and AP$_2$ as analyzed in this chapter.
a. With these syntactic and semantic assumptions, write down the denotations assigned to the following sentences in terms of the denotations **tina**, **tall**, **thin**, **IS** and **AND**.

(ii) Tina is both not tall and thin.  (iii) Tina is not both tall and thin.

b. Explain why the denotations you suggested for (ii) and (iii) account for the (non-)entailments \((ii) \Rightarrow (2.2)\) and \((iii) \not\Rightarrow (2.2)\).

c. Consider the equivalence between the following sentences:

(iv) Tina is both not tall and not thin.  (v) Tina is neither tall nor thin.

Suggest a proper denotation for the constituent **neither tall nor thin** in (v) in terms of the denotations **tall** and **thin** (standing for sets of entities). Explain how the denotation you suggest, together with our assumptions in items 5a and 5b above, explain the equivalence \((iv) \Leftrightarrow (v)\).

6. Consider the following sentence:

(i) Tina is \([\text{tall or thin}]\).

The **inclusive or exclusive** analyses for the coordinator **or** involve denotations that employ the **union** and **symmetric difference** functions, respectively – the functions defined as follows for all \(A, B \subseteq E\):

\[
\text{OR}_{\text{in}}(A, B) = A \cup B
\]

= the set of members of \(E\) that are in \(A\), in \(B\)
or in both \(A\) and \(B\)

\[
\text{OR}_{\text{ex}}(A, B) = (A - B) \cup (B - A) = (A \cup B) - (A \cap B)
\]

= the set of members of \(E\) that are in \(A\) or \(B\), but not both \(A\) and \(B\)

Consider the following sentential structures:

(ii) Tina is not \([\text{tall and thin}]\).

(iii) Tina is not \([\text{tall or thin}]\).

(iv) Tina is \([\text{not tall}]\) and \([\text{not thin}]\).

(v) Tina is \([\text{not tall}]\) or \([\text{not thin}]\).

a. Assuming that the word **or** denotes the function \(\text{OR}_{\text{in}}\), write down all the entailments that the TCC expects in (ii)–(v). Answer the same question, but now assuming that **or** denotes \(\text{OR}_{\text{ex}}\).
b. Which of the two entailment patterns in 6a better captures your linguistic intuitions about (ii)–(v)?

c. Under one of the two analyses of or, one of the structures (ii)–(v) is expected to be equivalent to (i). Which structure is it, and under which analysis? Support your answer by a set-theoretical equation.

7. A pair of sentences (or readings/structures) is said to be treated as **contradictory** if, whenever one of the sentences is taken to denote 1, the other denotes 0. For instance, under the analysis in this chapter, the sentences *Mary is tall* and *Mary is not tall* are contradictory.

a. Give more examples for contradictory pairs of sentences/structures under the assumptions of this chapter.

b. Consider the sentences *The bottle is empty* and *The bottle is full*. Suggest a theoretical assumption that would render these sentences contradictory.

c. Give an entailment that is accounted for by the same assumption.

d. Show that according to our account, the denotation of the sentence *Tina is tall and not tall* is 0 in any model. Such a sentence is classified as a **contradiction**. Show more examples for sentences that our account treats as contradictions.

e. Show that according to both our treatments of or in Exercise 6, the denotation of the sentence *Tina is tall or not tall* is 1 in any model. Such sentences are classified as **tautological**. Show more examples for sentences that our account treats as tautological.

f. Show that the TCC expects that any contradictory sentence entails any sentence in natural language, and that any tautology is entailed by any sentence in natural language. Does this expectation agree with your linguistic intuitions? If it does not, do you have ideas about how the problem can be solved?

8. We assume that entailments between sentences (or structures) have the following properties.

**Reflexivity:** Every sentence $S$ entails itself.

**Transitivity:** For all sentences $S_1$, $S_2$, $S_3$: if $S_1$ entails $S_2$ and $S_2$ entails $S_3$, then $S_1$ entails $S_3$.

Reflexivity and transitivity characterize entailments as a **preorder** relation on sentences/structures.
Consider the following entailments:
(i) Tina is tall, and Ms. Turner is neither tall nor thin ⇒ Tina is tall, and Ms. Turner is not tall.
(ii) Tina is tall, and Ms. Turner is neither tall nor thin ⇒ Tina is not Ms. Turner.
Show an entailment that illustrates transitivity together with entailments (i) and (ii).

9. Consider the following structurally ambiguous sentence (= (ii) from Exercise 2).
   (i) Tina is blond or tall and thin.
   a. For sentence (i), write down the denotations derived for the two structures using the inclusive denotation of or from Exercise 6, and the denotations Tina, blond, tall and thin.
   b. Give specific set denotations for the words blond, tall and thin that make one of these denotations true (1), while making the other denotation false (0).
   c. Using the both ... and construction from Exercise 4, find two unambiguous sentences, each of which is equivalent to one of the structural analyses you have given for sentence (i).
   d. Under an inclusive interpretation of or, which of the two sentences you found in 9c is expected to be equivalent to the following sentence?
      (ii) Tina is both blond and thin or both tall and thin.
   e. Write down the set-theoretical equation that supports this equivalence.
   f. Using our assumptions in this chapter, find a structurally ambiguous sentence whose two readings are analyzed as equivalent.

10. Consider the following entailment:
    (i) Tina has much money in her bank account, and Bill has one cent less than Tina in his bank account ⇒ Bill has much money in his bank account.
    a. We adopt the following assumption: Tina has \( m \) cents in her bank account, where \( m \) is some positive natural number. Further, we assume that entailment is transitive. Show that with these assumptions, you can use the entailment pattern in (i) to support an entailment with the following contradictory conclusion: Ms. X has much money in her bank account, and Ms. X has no money in her bank account.
b. The ability to rely on transitivity of entailments to support such absurd conclusions is known as the Sorites Paradox. Suggest a possible resolution of this paradox by modifying our assumptions in 10a and/or our assumption that entailment relations are transitive.

**SOLUTIONS TO SELECTED EXERCISES**

3. **Expression** | **Denotations in example models with**  
| | $E = \{a, b, c, d\}$  
| | $M_1$ | $M_2$ | $M_3$  
| Tina | a | b | b  
| tall | \{b, c\} | \{b, d\} | \{a, b, d\}  
| thin | \{a, b, c\} | \{b, c\} | \{a, c, d\}  
| not tall | \{a, d\} | \{a, c\} | \{c\}  
| [not tall] and thin | \{a\} | \{c\} | \{c\}  
| Tina is [[not tall] and thin] | 1 | 0 | 0  
| tall and thin | \{b, c\} | \{b\} | \{a, d\}  
| not [tall and thin] | \{a, d\} | \{a, c, d\} | \{b, c\}  
| Tina is [not [tall and thin]] | 1 | 0 | 1  
| Tina is thin | 1 | 1 | 0  

4.c. In any theory $T$ that satisfies the TCC, sentences $S_1$ and $S_2$ are equivalent if and only if for all models $M$ in $T$, $\llbracket S_1 \rrbracket^M = \llbracket S_2 \rrbracket^M$.

5.a–b. The truth-values and the accounts of the (non-)entailments are identical to the truth-values for the ambiguous sentence (2.13) and the corresponding (non-)entailment from (2.13) to (2.2).

c. $\llbracket \text{neither tall nor thin} \rrbracket = \overline{\text{tall}} \cap \overline{\text{thin}} = \text{AND}(\overline{\text{tall}}, \overline{\text{thin}}) = \llbracket \text{both not tall and not thin} \rrbracket$

6.a. $\text{OR}_{\text{in}}$: (iii)$\implies$(ii); (iv)$\implies$(ii); (ii)$\iff$(v); (iii)$\iff$(iv); (iii)$\implies$(v); (iv)$\implies$(v).

$\text{OR}_{\text{ex}}$: (iv)$\implies$(ii); (v)$\implies$(ii); (iv)$\implies$(iii).

c. (v); the $\text{OR}_{\text{ex}}$ analysis; $(\overline{A} - B) \cup (B - \overline{A}) = (B - A) \cup (A - B) = (A - B) \cup (B - A)$. 
7.a. Mary is neither tall nor thin, Mary is tall or thin; Mary is tall and not thin, Mary is thin; Mary is \{not tall\} or \{not thin\}, Mary is tall and thin.

b. The adjectives empty and full denote disjoint sets: \texttt{empty} \cap \texttt{full} = \emptyset.

c. The bottle is empty \Rightarrow \text{The bottle is not full}; \text{The bottle is full} \Rightarrow \text{The bottle is not empty}.

8. Tina is tall, and Ms. Turner is not tall \Rightarrow Tina is not Ms. Turner (=2.3a).

9.a. IS(tina, AND(ORin(blond, tall), thin))

b. blond = \{tina\}; tall = thin = \emptyset

c. Tina is both blond or tall and thin.

9. (A \cup B) \cap C = (A \cap C) \cup (B \cap C)

10.a. Consider the following general entailment scheme, based on (i):

\( (i') \) Ms. \( n \) has much money in Ms. \( n \)'s bank account and Ms. \( n + 1 \) has one cent less than Ms. \( n \) in Ms. \( n + 1 \)'s bank account \Rightarrow Ms. \( n + 1 \) has much money in Ms. \( n + 1 \)'s bank account.

We can deduce from \((i')\), by induction on the transitivity of entailment, that the following (unacceptable) entailment is intuitively valid:

Ms. 1 has much money in her bank account, and Ms. 1 has \( m \) cents in her bank account \Rightarrow Ms. \( m + 1 \) has much money in her bank account, and Ms. \( m + 1 \) has no cents in her bank account.
CHAPTER 3

TYPES AND MEANING COMPOSITION

This chapter introduces some of the elementary mathematical techniques in formal semantics. We systematize models by organizing denotations in domains of different types. This general type system allows models to describe sets, as well as relations and other operators with multiple arguments. Denotations of complex expressions are compositionally derived by a uniform semantic operation of function application. The resulting semantic framework is demonstrated by treating modified noun phrases (a tall man), reflexive pronouns (herself) and coordinations between different phrases. We avoid excess notation by defining denotations set-theoretically and introducing lambda-style shorthand when convenient.

This chapter systematically explains the way in which models allow linguistic expressions to denote abstract objects. This will give us a better insight into our theory of meaning and its relations with syntactic forms. The first step is to describe how denotations are organized in a model. Throughout Chapter 2, we used models freely to describe different mathematical objects. For sentences we used truth-values, for names like Tina we used entities, and for adjectives like tall we used sets of entities. In addition we used the membership operator for is, the intersection function for and, and the complement function for not. Using various mathematical objects in this manner was useful for expository purposes. However, in general it makes our compositionality principle hard to obtain. With each new mathematical notion we introduce, we need to see how it composes with other denotations. Too much mathematical freedom in the design of the denotations makes it hard to describe how they operate in different natural language expressions. The model structure that we introduce in this chapter helps us to make semantic distinctions between
language expressions within well-defined boundaries. In this way we gain a better understanding of denotations in general, and see more clearly how they interact with syntactic forms and with each other.

Some of the foundational themes in this chapter may seem intimidating at first glance. However, none of them is especially hard. To help you follow this chapter with ease, it is divided into four parts. Each of these parts covers a general topic that leads naturally to the topic of the next one. If you are a novice to the field, it is a good idea to solve the exercises referred to at the end of each part before reading on.

- Part 1 (‘Types and domains’) classifies denotations in models into different domains with different types. An important tool will be functions that characterize sets.
- Part 2 (‘Denotations at work’) elaborates on the composition of denotations and on how typed denotations in a compositional setting are translated to other set-theoretical concepts. An important tool here is functions that operate on other functions.
- Part 3 (‘Using lambda notation’) introduces a short notation for functions by using so-called ‘lambda-terms’. This helps us to define and use denotations in our analyses.
- Part 4 (‘Restricting denotations’) is about denotations that are systematically restricted by our models.

The formal system that is developed throughout this chapter is foundational to many works in formal semantics. For this reason, a technical summary of this chapter is included as an appendix to this book (page 239). This appendix gives readers a global overview of some of the most basic technical assumptions in formal semantics.

**PART 1: TYPES AND DOMAINS**

One of the main goals of this book is to systematically describe semantic distinctions between expressions as they are manifested in entailments. In Chapter 2, the basic difference was between entity denotations of names and truth-value denotations of sentences. Further, we used different functions as the denotations of the words *is*, *and* and *not*. Now we would like to analyze denotations of many more expressions. Therefore, it is high time to introduce some discipline into our semantic framework. In order to deal with denotations in a more
systematic way, we will formally specify the kind of mathematical objects that our models contain. In technical terms, such a specification is referred to as a *type system*.

**CHARACTERISTIC FUNCTIONS**

Many of the denotations in formal semantics are functions of different types. To illustrate a simple type of function in formal semantics, we start with a maximally simple sentence:

(3.1) Tina smiled.

Intuitively, the intransitive verb *smile* should denote a set of entities, just like the adjectives *tall* and *thin* in Chapter 2. We conceive of the denotation of the word *smiled* in (3.1) as the set of entities that smiled at a given moment in the past. For convenience, we often ignore the tense in our discussion, and refer to the verb *smiled* in (3.1) as being associated with “the set of smilers”. But now, how can sentence (3.1) denote a truth-value? Unlike the sentence *Tina is tall*, sentence (3.1) contains no word like *is* that may express the membership function. Thus, the set for *smiled* and the entity for *Tina* do not immediately give a truth-value in (3.1). To allow for their easy composition, we should change perspectives slightly. We still use sets of entities for describing denotations of intransitive verbs, but we do that indirectly using functions. For example, suppose that we want to describe a model with three entities: a, b and c. Suppose further that in the situation that the model describes, entities a and c smiled and entity b did not. In this case the set of smilers in the model, S, is the set \{a, c\}. Instead of defining the denotation of the verb *smile* to be the set S itself, we let it be a function that indirectly describes S. This function, which we denote \(\chi_S\), is a *function from entities to truth-values*. For each of the two elements in S, entities a and c, the function \(\chi_S\) returns the truth-value 1. For entity b, which is not in S, we let \(\chi_S\) return 0. Thus, \(\chi_S\) is the following function:

(3.2) \[\chi_S: a \mapsto 1 \quad b \mapsto 0 \quad c \mapsto 1\]

The function \(\chi_S\) is called the *characteristic function* of the set \{a, c\} over the set of entities \{a, b, c\}. In general, we define characteristic functions
as follows:

Let $A$ be a subset of $E$. A function $\chi_A$ from $E$ to the set $\{0, 1\}$ is called the characteristic function of $A$ in $E$ if it satisfies for every $x \in E$:

$$\chi_A(x) = \begin{cases} 
1 & \text{if } x \in A \\
0 & \text{if } x \notin A 
\end{cases}$$

For every element $x$ of $E$, the truth-value $\chi_A(x)$ indicates whether $x$ is in $A$ or not. Thus, $\chi_A$ uniquely describes a subset of $E$. The converse is also true: for every subset of $E$ there is a unique characteristic function. This means that sets of entities and their characteristic functions encode precisely the same information. For this reason we often interchangeably talk about subsets of $E$ or the functions that characterize them. Specifically, in Chapter 4, we will refer by ‘$f^*$’ to the set characterized by a function $f$. Further, we will see that functions can also be used for characterizing subsets of other domains besides $E$.

For the time being, for the sake of brevity, we use the general term ‘characteristic functions’ when referring exclusively to functions that characterize sets of entities in $E$. With this notion of characteristic functions, we can easily describe how the composition process in sentence (3.1) works. We assume that the denotation of the verb smile is an arbitrary characteristic function. This corresponds to our assumption that the verb smile can be associated with any set of entities. Suppose that the denotation of smile is the function $\chi_S$. In our analysis of sentence (3.1), this function applies to the entity denotation tina. In a formula, sentence (3.1) is analyzed as follows:

(3.3) $\text{smile(tina)}$

The expression in (3.3) describes the truth-value that the function smile assigns to the entity tina. For example, in the model we described above, the denotation smile is the function $\chi_S$ in (3.2). Suppose that in the same model, the denotation tina is the entity a. As a result, the denotation (3.3) of sentence (3.1) is $\chi_S(a)$, which equals 1. If tina is the entity b, the denotation (3.3) equals 0. This way, by letting the denotation of the verb smile characterize a set, we directly obtain a truth-value denotation for the sentence Tina smiled.
In our analysis of sentence (3.1), we have been using three denotations: an entity, a characteristic function and a truth-value. Each of these denotations has a different ‘nature’, which we distinguish by letting each of them come from a different domain. In Chapter 2, we already let every model $M$ contain a domain of entities $E^M$ and a domain of truth-values $\{0, 1\}$. Now it is time to also introduce domains for characteristic functions and other denotations. Each domain we introduce comes with a label that we call a type. We use the letter $e$ as the type for the domain $E^M$ in a given model $M$. Since we want to make an explicit connection between types and their respective domains, we also use the notation $'D^M_e'$ (the $e$ Domain in $M$) as an alternative name for $E^M$. As usual, when the model $M$ is clear from the context, we write ‘$D_e$’ rather than ‘$D^M_e$’. The letter $t$ is used as the type for the domain of truth-values. Accordingly, this domain is denoted ‘$D_t$’. Since we fix $D_t$ as the set $\{0, 1\}$ in all models, we do not mention the model when referring to this domain. In our example above, the name Tina takes its denotation from $D_e$, and the sentence Tina smiled takes its denotation from $D_t$. In short, we say that proper names are of type $e$ and sentences are of type $t$. We refer to the types $e$ and $t$ as the basic types of our type system. The domains for these types, $D_e$ and $D_t$, have been specified with no relation to other domains. For this reason, we refer to them as the basic domains in every model.

Now, we also want to define a type and a domain for characteristic functions like the denotation of smile. These are defined on the basis of the types $e$ and $t$, and the domains $D_e$ and $D_t$. Specifically, a characteristic function in a model $M$ is a function from the entities in $M$ to truth-values. Accordingly, we define the domain of characteristic functions as follows:

\[(3.4) \text{ The domain of characteristic functions in a model } M \text{ is the set of all the functions from } D^M_e \text{ to } D_t.\]

This domain is assigned the type ‘($et$)’. We often omit outermost parentheses, and refer to the same type as ‘$et$’. The corresponding domain is accordingly referred to as ‘$D^M_{et}$’, or simply ‘$D_{et}$’.

In set theory, there is a common way to refer to the set of all functions from a set $A$ to a set $B$. Formally, we use ‘$B^A$’ when referring to this set of functions. Thus, the definition of the domain $D_{et}$ in (3.4)
Table 3.1: Subsets of $D_e$ and their characteristic functions in $D_{et}$.

<table>
<thead>
<tr>
<th>Subset of $D_e$</th>
<th>Characteristic function in $D_{et}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\emptyset$</td>
<td>$f_1$: $a \mapsto 0$ $b \mapsto 0$ $c \mapsto 0$</td>
</tr>
<tr>
<td>${a}$</td>
<td>$f_2$: $a \mapsto 1$ $b \mapsto 0$ $c \mapsto 0$</td>
</tr>
<tr>
<td>${b}$</td>
<td>$f_3$: $a \mapsto 0$ $b \mapsto 1$ $c \mapsto 0$</td>
</tr>
<tr>
<td>${c}$</td>
<td>$f_4$: $a \mapsto 0$ $b \mapsto 0$ $c \mapsto 1$</td>
</tr>
<tr>
<td>${a, b}$</td>
<td>$f_5$: $a \mapsto 1$ $b \mapsto 1$ $c \mapsto 0$</td>
</tr>
<tr>
<td>${a, c}$</td>
<td>$f_6$: $a \mapsto 1$ $b \mapsto 0$ $c \mapsto 1$</td>
</tr>
<tr>
<td>${b, c}$</td>
<td>$f_7$: $a \mapsto 0$ $b \mapsto 1$ $c \mapsto 1$</td>
</tr>
<tr>
<td>${a, b, c}$</td>
<td>$f_8$: $a \mapsto 1$ $b \mapsto 1$ $c \mapsto 1$</td>
</tr>
</tbody>
</table>

can be formally written as follows:

$$D_{et} = D_t^{D_e}$$

Functions in the $D_{et}$ domain, as well as expressions of type $et$, are often referred to as one-place predicates over entities. Common alternative notations for the type of one-place predicates are $\langle e, t \rangle$ and $e \rightarrow t$. In this book we will stick to the shorter notation $et$.

All intransitive verbs like *smile*, *dance*, *run* etc. are assigned the type $et$. In a given model, each of these verbs may have a different denotation. For example, in the model we described above, the verb *smile* denotes the function $\chi_S$, which characterizes the set $S = \{a, c\}$. Other verbs like *dance* and *run* may be associated with different sets in the same model, and hence denote different characteristic functions. For this reason, the domain $D_{et}^M$ in a given model $M$ includes all the functions from $D_e^M$ to $D_t$. To see which functions these are in our example model, we first note that the domain $D_e$ has eight subsets in that model. These are: the empty set $\emptyset$; the singleton sets $\{a\}$, $\{b\}$ and $\{c\}$; the doubletons $\{a, b\}$, $\{a, c\}$ and $\{b, c\}$; and the whole set $D_e$, i.e. $\{a, b, c\}$. The domain $D_{et}$ includes the eight functions that characterize these sets, as shown in Table 3.1. In such models, where the entities are $a$, $b$ and $c$, intransitive verbs like *smile* and *run* must denote one of the eight functions in $D_{et}$. Specifically, the function $f_6$ in Table 3.1 is the same function $\chi_S$ that we assumed as the denotation of the verb *smile* in our example.
MANY TYPES, MANY DOMAINS

We have seen how to define the domain $D_{et}$ on the basis of $D_e$ and $D_t$. As we consider more expressions besides intransitive verbs, we will need more types and domains. The methods for defining them are similar to the definition of $D_{et}$. The construction of the type $et$ and the domain $D_{et}$ illustrates the general principle that we use for defining new types and domains. We have defined the type $(et)$ as the parenthesized concatenation of the basic types $e$ and $t$. The corresponding domain $D_{et}$ was defined as the set of functions from the domain $D_e$ to the domain $D_t$. We employ the same method for defining more new types and domains. Once two types $\tau$ and $\sigma$ and domains $D_\tau$ and $D_\sigma$ are defined, they are used for defining another type $(\tau \sigma)$ and a domain $D_{\tau \sigma}$, which consists of all the functions from $D_\tau$ to $D_\sigma$. More types and domains are defined in a similar way, with no upper limit on their complexity. Since the same method works for defining all types and domains from the basic ones, we refer to it as an inductive procedure. Specifically, types $e$, $t$ and $et$, and their respective domains, are used inductively for defining new types and domains. For instance, when using type $e$ twice, we get the type $ee$. The respective domain, $D_{ee}$, contains all the functions from $D_e$ to $D_e$. Further, combining the types $e$ and $et$, we get the type $e(et)$. The corresponding domain $D_{e(et)}$ is the set of functions from $D_e$ to $D_{et}$. As we will see below, this $e(et)$ domain is useful for denotations of transitive verbs, i.e. verbs that have both a subject and a direct object.

Definition 1 below formally summarizes our inductive method for specifying types:

**Definition 1.** The set of types over the basic types $e$ and $t$ is the smallest set $T$ that satisfies:

(i) \{e, t\} $\subseteq$ $T$

(ii) If $\tau$ and $\sigma$ are types in $T$ then $(\tau \sigma)$ is also a type in $T$.

This inductive definition specifies the set of types as an infinite set $T$, including, among others, the types given in Figure 3.1.

For every model $M$, we specify a domain for each of the types in the set $T$. Let us summarize how this is done. The domain $D_e^M$ is directly specified by the model. The domain $D_t$ is fixed as $\{0, 1\}$ for
all models. Domains for other types are inductively defined, as formally summarized in Definition 2 below:

**Definition 2.** For all types $\tau$ and $\sigma$ in $\mathcal{T}$, the domain $D_{\tau\sigma}$ of the type $(\tau\sigma)$ is the set $D^{D_\tau}_\sigma$ – the set of functions from $D_\tau$ to $D_\sigma$.

The induction in Definition 2, together with our stipulated basic domains $D_e$ and $D_t$, specify the domains for all the types derived from Definition 1. We have already discussed the domain $D_{et}$ of characteristic functions. Let us now consider in more detail the definition of the domain $D_{e(et)}$. When unfolding Definition 2, we see that it derives the following definition:

\begin{equation}
D_{e(et)} \text{ is the set of functions from } D_e \text{ to } D_{et} = \text{ the functions from entities to } D_{et} = \text{ the functions from entities to the functions from } D_e \text{ to } D_t = \text{ the functions from entities to the functions from entities to truth-values.}
\end{equation}

Thus, functions of type $e(et)$ return functions (of type $et$) as their result. This is a result of our inductive definitions. Our definitions above also make another situation possible: functions that take functions as their *arguments*. For instance, the type $(et)e$ describes functions that map $et$ functions to entities. Further, Definitions 1 and 2 also allow functions that take function arguments and map them to function results. Consider for instance the type $(et)(et)$. The corresponding domain, $D_{(et)(et)}$, contains the functions that map characteristic functions to characteristic functions. For instance, suppose that $F$ is a function in $D_{(et)(et)}$. This means that $F$ can receive any characteristic function $g$ in $D_{et}$ and return a characteristic function $h$ in $D_{et}$, possibly...
different from $g$. We describe this situation by writing $F(g) = h$. The functions $g$ and $h$ characterize sets of entities. Thus, we can view functions like $F$, of type $(et)(et)$, as mapping sets of entities to sets of entities. We already used one such function in Chapter 2, when we let the denotation of the word *not* map sets of entities to their complement. Functions from sets of entities to sets of entities, in their new guise as $(et)(et)$ functions, will often reappear in the rest of this book.

You are now advised to solve Exercises 1, 2 and 3 at the end of this chapter.

PART 2: DENOTATIONS AT WORK

Semantic types and their corresponding domains give us a powerful tool for analyzing natural language meanings: one that is empirically rich, and yet systematically constrained. Equipped with our expressive tool for describing denotations, it is high time to start using it for analyzing linguistic examples. In order to do that, we have to explain what principles allow denotations to combine with each other compositionally. One elementary principle lets functions apply to their arguments. As we will see, functions, and the rule of function application, allow us to encode many useful intuitions about meanings, using the technique known as *currying*. After introducing this technique of using functions, we will see how to develop systematic analyses by solving *type equations*. This will allow us to look back at what we did in Chapter 2, and systematically treat the copula *be* and predicate negation as part of our uniform type system.

FUNCTION APPLICATION

Types provide us with a record of the way denotations combine with each other. In our analysis of the simple example *Tina smiled* we saw how an $et$ function combines with an entity (type $e$) to derive a truth-value (type $t$). We write it as follows:

$$(et) + e = t.$$

The rule we used for combining denotations in the sentence *Tina smiled* is function application: we applied a function *smile* from entities to truth-values to an entity *tina*, and got a truth-value
smile(tina). In terms of denotations, we write it as follows:

\[
\text{smile}_{et} + \text{tina}_e = \text{smile(tina)} : t
\]

By the notation ‘\(\text{smile}_{et}\)’ we refer to the denotation of the verb \(\text{smile}\), and state that it is of type \(et\). Similarly for ‘\(\text{tina}_e\)’. An alternative notation is ‘\(\text{smile} : et\)’ and ‘\(\text{tina} : e\)’. This colon becomes more convenient when we wish to state the type \(t\) of the result \(\text{smile(tina)}\), and write ‘\(\text{smile(tina)} : t\)’. Following standard mathematical practice, we let the function \(\text{smile}\) appear to the left of its argument \(\text{tina}\). However, English verbs like \(\text{smile}\) normally follow the subject, as is the case in the sentence \(\text{Tina smiled}\). The workings of function application are not affected by this. So we also assume:

\[
e + (et) = t.
\]

Thus, when wishing to highlight the syntactic ordering, we also describe the composition of denotations in the sentence \(\text{Tina smiled}\) as follows:

\[
\text{tina}_e + \text{smile}_{et} = \text{smile(tina)} : t.
\]

In more general terms, our type-based rule of function application is given below:

---

**Function application with typed denotations:** Applying a function \(f\) of type \(\tau\sigma\) to an object \(x\) of type \(\tau\) gives an object \(f(x)\) of type \(\sigma\). In short:

<table>
<thead>
<tr>
<th>Types</th>
<th>((\tau\sigma) + \tau = \tau + (\tau\sigma) = \sigma)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Denotations</td>
<td>(f_{\tau\sigma} + x_\tau = x_\tau + f_{\tau\sigma} = f(x) : \sigma)</td>
</tr>
</tbody>
</table>

---

The equations that we gave above describe how types are combined with each other. For each type combination, there is a corresponding operation between denotations in the corresponding domains: function application. Such a system, which combines types and denotations, is called a type calculus. The type calculus above, which deals with function application, is known as the Ajdukiewicz Calculus (after K. Ajdukiewicz). In Chapter 5 we will return to type calculi, and extend their usages for other operations besides function application.
For now, let us see some more examples of the way we use Ajdukiewicz’s calculus:

\[ e + ee = e \quad \text{applying a function } g_{ee} \text{ to an entity } x_e \text{ gives an entity } g(x) \]

\[ e(et) + e = et \quad \text{applying a function } h_{e(et)} \text{ to an entity } x_e \text{ gives an et function } h(x) \]

\[ et + (et)(et) = et \quad \text{applying a function } F_{(et)(et)} \text{ to a function } k_{et} \text{ gives an et function } F(k) \]

These equations each contain two types and their combination using function application. However, for many pairs of types, function application cannot work. For instance, function application cannot combine a function \( f \) of type \( t(et) \) with a function \( g \) of type \( et \). The reason is twofold. First, the function \( f \) cannot apply to \( g \), since \( f \) takes truth-values as its argument, and \( g \) is not a truth-value. Second, the function \( g \) cannot apply to \( f \), since \( g \) takes entities as its argument, and \( f \) is not an entity. Such situations, where the type calculus does not produce any result, are referred to as a type mismatch.

**TRANSITIVE VERBS**

Now that we have seen how typed denotations are systematically combined with each other, let us consider the following sentence:

(3.6) Tina [praised Mary]

Sentences like (3.6), which contain both a subject and a direct object, are referred to as transitive sentences. In (3.6) we standardly assume that a transitive verb (praise) forms a constituent with the object noun phrase (Mary). This means that, in order to compositionally derive a truth-value for sentence (3.6), we first need to derive a denotation for the verb phrase praised Mary. To do that, we follow our treatment of intransitive verbs. In the same way that the denotation of the verb smiled characterizes the set of entities that smiled, we now want the denotation of the verb phrase praised Mary to characterize the set of entities that praised Mary. This is the function that sends every entity that praised Mary to 1, and any other entity to 0. How do we derive such an et function from the denotations of the words praised
Figure 3.2  Syntactic structure and semantic interpretation for Tina praised Mary.

and Mary? The key to doing so is to assume that the verb praise denotes a function of type $e(et)$. As we have seen in (3.5) above, functions of type $e(et)$ map entities to $et$ functions. Thus, we let the verb praise denote an $e(et)$ function $\text{praise}$. Applying this function to the entity $\text{mary}$, we get a denotation of type $et$ for the verb phrase praised Mary. This is formally written as follows:

$$ (3.7) \text{praise}_{e(et)} + \text{mary}_e = \text{praise(mary)} : et $$

Further, in the compositional analysis of the whole sentence (3.6), the $et$ function in (3.7) applies to the entity $\text{tina}$. What we get is the following truth-value:

$$ (3.8) \text{praise(mary)} + \text{tina}_e = (\text{praise(mary)})(\text{tina}) : t $$

In words: when the $et$ function $\text{praise(mary)}$ applies to the entity $\text{tina}$, the result is a truth-value. This truth-value is the denotation that our model assigns to the sentence (3.6). To increase readability, we often omit obvious types and parentheses, and write this truth-value as:

$$ (3.9) \text{praise(mary)}(\text{tina}) $$

To summarize the compositional process in our analysis of sentence (3.6), we repeat it in Figure 3.2 using tree notation.

Figure 3.2A is the tree notation of the structure we assumed in (3.6). Figure 3.2B is the same tree, but the nodes are now decorated with their
types and their denotations. We refer to tree diagrams like Figure 3.2B as *semantically interpreted structures*. In this interpreted structure, the nodes for the words *Tina, praised* and *Mary* are decorated by their lexical types and denotations. In addition, we have two nodes for the constituents assumed in the binary structure: the verb phrase *praised Mary* and the whole sentence. These nodes are decorated by their types and denotations, which are compositionally derived by function application.

Any \( e(\text{et}) \) function can combine with two entities, one entity at a time, returning a truth-value. Having seen how \( e(\text{et}) \) functions allow us to derive truth-values for transitive sentences like (3.6), we may still feel that functions that return functions as their result are excessively intricate when analyzing such simple sentences. Fortunately, there is an equivalent way of looking at \( e(\text{et}) \) denotations of transitive verbs, which better reflects their intuitive simplicity. Intuitively, denotations of verbs like *praised* can be viewed as *two-place relations* between entities, aka *binary relations*. Such relations are sets of *pairs* of entities. In our example, we may view the denotation of the verb *praised* as the set of pairs of entities \( \langle x, y \rangle \) that satisfy the condition \( x \text{ praised } y \). For instance, suppose that in our model, the entities \( t, j \) and \( m \) are the denotations of the respective names *Tina, John* and *Mary*. When the domain \( D_e \) is \{t, j, m\}, we may describe who praised who by the following binary relation \( U \):

\[
(3.10) \quad U = \{ \langle t, m \rangle, \langle m, t \rangle, \langle m, j \rangle, \langle m, m \rangle \}
\]

The relation \( U \) is useful for describing a situation with three people, where *Tina* only praised *Mary*, *John* praised no one, and *Mary* praised everybody, including herself. In this way, the relation \( U \) provides full answers to the following questions:

\[
(3.11) \begin{align*}
    &\text{a. \ Who praised Tina? \ Answer: only Mary.} \\
    &\text{b. \ Who praised John? \ Answer: only Mary.} \\
    &\text{c. \ Who praised Mary? \ Answer: only Tina and Mary herself.}
\end{align*}
\]

Conversely: anybody who gives the same answers as in (3.11) will have implicitly described the binary relation \( U \).

Now we can get back to \( e(\text{et}) \) functions, and observe that they give us the same information as binary relations like \( U \). In particular, the
e(et) denotation of praise also tells us, for each entity, which entities praised that entity. Thus, when our domain of entities $D_e$ is the set \{t, j, m\}, any e(et) function over this domain answers precisely the same questions as in (3.11). In particular, the same situation that the binary relation $U$ encodes is also described by the following $e(et)$ function in our model, which we call $\chi_U$:

\[
\chi_U : \begin{align*}
t & \mapsto [t \mapsto 0 \quad j \mapsto 0 \quad m \mapsto 1] \\
j & \mapsto [t \mapsto 0 \quad j \mapsto 0 \quad m \mapsto 1] \\
m & \mapsto [t \mapsto 1 \quad j \mapsto 0 \quad m \mapsto 1]
\end{align*}
\]

The function $\chi_U$ maps each of the three entities in $D_e$ to an et function. More specifically:

- $\chi_U$ maps the entity $t$ to the function characterizing the set \{m\}.
- $\chi_U$ maps the entity $j$ to the function characterizing the same set, \{m\}.
- $\chi_U$ maps the entity $m$ to the function characterizing the set \{t, m\}.

Note the parallelism between this specification of $\chi_U$ and the question–answer pairs in (3.11). When the denotation praise is $\chi_U$, our model describes the same situation that (3.11) describes in words, which is the same information described by the binary relation $U$. More generally, we conclude that $e(et)$ functions encode the same information as binary relations over entities. This is similar to how characteristic functions of type et encode the same information as sets of entities. In mathematical terms, we say that the domain of $e(et)$ functions is isomorphic to the set of binary relations over $D_e$. Because $e(et)$ functions take two entities before returning a truth-value, we sometimes also refer to them as two-place predicates.

CURRYING

There is a general lesson to be learned from our treatment of transitive sentences and $e(et)$ predicates. We have seen how a situation that is naturally described by a binary relation can equally be described by a function that returns functions. The general idea is useful in many other circumstances in formal semantics (as well as in computer science). A binary relation between entities is a set containing pairs of entities. We can characterize such a set by a function that takes pairs of
entities and returns “true” or “false”. Such two-place functions occur very often in mathematics. As another example, let us consider one of the most familiar two-place functions: number addition. This function takes two numbers, \( x \) and \( y \), and returns their sum, which we normally denote ‘\( x + y \)’. To highlight the fact that number addition is a two-place function, let us denote it using the function symbol \( \text{sum} \). Thus, we use the following notation:

\[
\text{sum}(x, y) = x + y
\]

Now, let us use the letter ‘\( n \)’ as the type for natural numbers. Using this type, we will now see how we can also encode addition as a function of type \( n(nn) \): a function from numbers to functions from numbers to numbers. Let us refer to this function as \( \text{ADD} \). To define \( \text{ADD} \), we need to define the result that it assigns to any given number. This result is an \( nn \) function: a function from numbers to numbers. Therefore, to define \( \text{ADD} \) we will now specify the \( nn \) function that \( \text{ADD} \) assigns to any number. For every number \( y \), we define:

\[
(3.13) \text{The } nn \text{ function } \text{ADD}(y) \text{ sends every number } x \text{ to the number } \text{sum}(x, y).
\]

As we expect from number addition, the function \( \text{ADD} \) takes two numbers and returns their sum. But it does it step by step: it first takes one number, \( y \), it returns a function \( \text{ADD}(y) \), and this function applies to another number \( x \) and returns the sum \( \text{sum}(x, y) \), or more simply: \( x + y \). As a result, for every two numbers \( x \) and \( y \), we get:

\[
\text{ADD}(y)(x) = \text{sum}(x, y) = x + y
\]

For example, let us consider how we calculate the sum of 1 and 5 using the function \( \text{ADD} \). We first give \( \text{ADD} \) the number 1 as an argument, and get the function \( \text{ADD}(1) \) as the result. This resulting function can take any number and return its sum with 1. Now, we choose to give the function \( \text{ADD}(1) \) the argument 5. Unsurprisingly, the result of calculating \( \text{ADD}(1)(5) \) is \( 5 + 1 \), or 6. Have we gained anything from reaching this obvious result in such a roundabout way? As strange as it may sound, we have! While calculating the sum of 5 and 1, we generated the function \( \text{ADD}(1) \). This is the \textit{successor} function: the function that sends every natural number to the number that
follows it. This function is of course useful for other purposes besides applying it to the number 5.

These different ways of looking at number addition are quite similar to what we saw in our syntactic-semantic analysis of sentence (3.6). In that analysis, we equivalently encoded situations either using binary relations like $U$ or using $e(\text{et})$ functions. Because we adopt the latter method, we got an intermediate result by applying the $e(\text{et})$ denotation of the verb \textit{praise} to the entity denotation of the object \textit{Mary}. This is the \textit{et} denotation of the verb phrase \textit{praised Mary}. Having such a denotation for the verb phrase is compositionally useful. As we will see later in this chapter, it gives us a natural treatment of conjunctive sentences like \textit{Tina smiled and praised Mary}, where the verb phrase \textit{praised Mary} does not combine directly with the subject \textit{Tina}.

The kind of maneuver we saw above will also be useful for treating many other phenomena besides transitive sentences. In its full generality, the idea is known as \textit{Currying} (after H. Curry) or, less commonly, as \textit{Schönfinkelization} (after M. Schönfinkel). In technical slang we often say that a one-place function like \textit{ADD} is a \textit{Curried} version of the two-place addition operator \textit{sum}. Conversely, we say that the addition function \textit{sum} is an \textit{uncurried} (or ‘deCurried’) version of \textit{ADD}. For more on Currying, see the suggested further reading at the end of this chapter.

**SOLVING TYPE EQUATIONS**

Using Currying, we now present a revised treatment of the copular sentences with adjectives from Chapter 2. Reconsider the following sentence:

(3.14) Tina [ is tall ]

In Chapter 2 we analyzed the verb \textit{is} in (3.14) as the membership function. This two-place function sends pairs, of entities and sets of entities, to truth-values. However, now we no longer have two-place functions and sets of entities in our models: we have replaced them by Curried functions and characteristic functions, respectively. Therefore, we need to revise our analysis of (3.14) along these lines. First, as for intransitive verbs, we let adjectives characterize sets of entities. Thus, in our analysis of sentence (3.14) we assume that the adjective \textit{tall} denotes an \textit{et} function. In many other languages
Besides English, this allows us to treat sentences of the form *Tina tall*. However, English requires the copula *be* to appear in such sentences. How should we now analyze the semantic role of the English copula? Let us first consider its type. Compositional interpretation of the structure in (3.14) means that the denotation of the constituent *is tall* has to be determined before we determine the denotation of the sentence. To be able to combine with the denotation of the subject *Tina*, the constituent *is tall* has to denote a function that applies to the entity *tina* and derives a truth-value. Therefore, the expression *is tall* should have the same type *et* as the adjective *tall*. We conclude that the denotation of *is* has to be a function of type (*et*)(*et*): a function from *et* functions to *et* functions. When such an (*et*)(*et*) denotation for the word *is* applies to an *et* function like *tall*, it gives a function of the same type, *et*, which we will use as the denotation for the constituent *is tall*.

What we have done above is a kind of puzzle solving. We assumed solutions to some parts of the puzzle: the types of the words *Tina* and *tall*, and the type *t* of the whole sentence. Using the sentence’s structure, we found a suitable type for the word *is* that allows function application to compositionally derive for the sentence a *t*-type denotation. The puzzle can be summarized as follows, with *X* and *Y* as the unknown types:

\[(3.15) [ \text{Tina}_e [ \text{is}_Y \text{tall}_{et} ]_X ]_t \]

In our solution, we found that *X* = *et* and *Y* = (*et*)(*et*). The solution process itself is described in Figure 3.3. This figure contains two type equations. One equation is:

\[\text{Eq. 1: } e + X = t \]

In words: which type(s) *X* combines with type *e* and derives type *t*?

By solving this equation, we see that the type for the constituent *is tall* must be *et*. This leads us to another equation in the compositional analysis of the sentence, which helps us to see how we can derive the type *et* for this constituent:

\[\text{Eq. 2: } Y + et = et \]

In words: which type(s) *Y* combines with type *et* and derives type *et*?
Solving this equation as well, we see that the type for the copula is must be \((et)(et)\).

**BACK TO COPULAS AND PREDICATE NEGATION**

Now, after determining the type of the copula is, we want its denotation to preserve the welcome results of our analysis in Chapter 2. To do that, we let the constituent is tall denote the same \(et\) function as the adjective tall. This is because, intuitively, we still want the sentence Tina is tall to be interpreted as a membership assertion, claiming that the entity tina is in the set that the function tall,\(et\) characterizes. Thus, we assume that the word is denotes the *identity function* for \(et\) functions: the function of type \((et)(et)\) that maps any \(et\) function to itself. Formally, we define the denotation is for the word is as the following \((et)(et)\) function:

\[
(3.16) \text{is} \text{ is the function sending every function } g \text{ of type } et \text{ to } g \text{ itself.}
\]
Our compositional, type-theoretical analysis is summarized in Figure 3.4.

Because the denotation of the copula *is* is now defined as the identity function, the truth-value that we get for sentence (3.14) can now be analyzed as follows in (3.17):

\[
\begin{align*}
\text{(3.17) a. } \text{IS(tall)} &= \text{tall} \quad \text{(by definition of IS in (3.16))} \\
\text{b. } \text{IS(tall)(tina)} &= \text{tall(tina)} \quad \text{(due to the equality in (3.17a))}
\end{align*}
\]

Due to this simple derivation, the truth-value that we derive in (3.17b) for sentence (3.14) is 1 if and only if the entity tina is in the set characterized by the function tall.

Functions of type (et)(et) are also useful for adjusting our account of predicate negation from Chapter 2. Let us reconsider, for example, the following negative sentence:

(3.18) Tina [ is [ not tall ]]

In Chapter 2 we let the negation word *not* denote the complement function, which sends every set of entities to its complement set. With our characteristic functions substituting sets, we now treat the word *not* as an (et)(et) function, of the same type as the copula *is*. Our definition of the denotation NOT in (3.19) below respects the status of the word *not* in (3.18) as a predicate negator:

(3.19) NOT is the (et)(et) function sending every et function g to the et function NOT(g) that satisfies the following, for every entity x (see next page):
In words: we define the function NOT, which takes a function \( g \) of type \( et \) and returns a function \( \text{NOT}(g) \) of the same type, separating into the following two cases:

- for any entity \( x \) such that \( g(x) = 0 \), we define \( \text{NOT}(g)(x) \) to be 1;
- for any entity \( x \) such that \( g(x) = 1 \), we define \( \text{NOT}(g)(x) \) to be 0.

In this way, the function \( \text{NOT}(g) \) reverses the value that \( g \) assigns to any entity. Because of that, \( \text{NOT}(g) \) is the \( et \) function characterizing the complement set of the set characterized by \( g \). Thus, by applying the function \( \text{NOT} \) to an \( et \) denotation \( \text{tall} \), we achieve the same analysis that we got in Chapter 2 by complementing the set characterized by \( \text{tall} \). More formally, we analyze the structure (3.18) as denoting the following truth-value:

\[
(3.20) \ (\text{IS}(et)(et)(\text{NOT}(et)(et)(\text{tall}_{et}))(\text{tina}_{e}))
\]

Because the function \( \text{IS} \) is the identity function, the truth-value in (3.20) is the same as:

\[
(\text{NOT}(et)(et)(\text{tall}_{et}))(\text{tina}_{e})
\]

By definition of the function \( \text{NOT} \), this truth-value is 1 if and only if the entity \( \text{tina} \) is in the complement of the set characterized by the \( et \) function \( \text{tall} \). Thus, the structure (3.18) is analyzed on a par with our analysis of the sentence in Chapter 2.

Let us now take stock of what we have done in Part 2. In this part we have aimed to maintain the informal analyses of Chapter 2 within a more structured type system. This system was fully defined by the two simple Definitions 1 and 2 in Part 1. As we have seen, these two definitions are highly expressive: they allowed models to mimic sets by using characteristic functions, and mimic two-place functions by using one-place Curried functions. Despite this expressiveness, so far we have done little to extend the empirical coverage of Chapter 2 besides adding a treatment of transitive verbs. However, by employing types as part of our theory we now have a rather powerful system that elucidates our notions of denotations and compositionality. This unified
framework will be employed throughout this book for dealing with new phenomena while relying on the same foundational principles.

You are now advised to solve Exercises 4, 5, 6 and 7 at the end of this chapter.

PART 3: USING LAMBDA NOTATION

Having functions of different types in our semantic framework gives us an extremely powerful tool. In order to use this tool efficiently, it is convenient to have a standard notation for the functions we use, and the way they apply to their arguments. In this part we study the common notation of lambda terms, and see how it is used within our semantic system.

DEFINING FUNCTIONS USING LAMBDA TERMS

Below we restate definition (3.16) of the denotation for the copula *is*:

(3.21) the function sending every function *g* of type *et* to *g* itself.

We may feel that (3.21) is an unnecessarily long and cumbersome way of defining the identity function. Indeed, in formal semantics we often use a more convenient notation, by employing lambda terms, or ‘λ-terms’. Let us illustrate it by rewriting definition (3.21) in our new notation:

- Instead of writing “the function sending every function *g* of type *et,*” we write “λ*get*.”
- Instead of “to *g* itself”, we write “*g*”.

After rewriting (3.21) in this way, we get the following formula:

(3.22) λ*get* · *g*

Since (3.22) is nothing but an alternative way of defining the function in (3.21), it gives us exactly the same information about it:

(i) The letter ‘λ’ tells us that it is a function.
(ii) The dot separates the specification of the function’s argument and the definition of the function’s result. Before the dot, writing ‘*get*’
introduces ‘g’ as an *ad hoc* name for the argument of the function. The type *et* in the subscript of *g* tells us that this argument can be any object in the domain *D*_*et*.

(iii) The re-occurrence of ‘g’ after the dot tells us that the function we define in (3.22) simply returns the value of its argument.

From (ii) and (iii) we immediately conclude that the function *λg* in (3.22) returns an object in the domain *D*_*et*. Hence this function is of type *(et)(et)*, as we wanted. Now, with our λ-term conventions, we are fully justified in saving space and writing our definition of the denotation for the copula *is* concisely, as in (3.23) below:

\[(3.23) \text{IS} = \lambda g_{\text{et}} \\cdot g\]

It is important to note that the letter ‘g’ has no special significance in definition (3.23). If we prefer, we may define the function IS equivalently, as *λhet.h*: “the function sending every function *h* of type *et* to *h* itself”. This would not change anything about our definition of the identity function, since it would still do the same thing: return the *et* function that it got as argument. When defining a function, it hardly matters if we decide to call the argument ‘g’, ‘h’, ‘x’, ‘y’ or any other name. Any name will do, as long as we use it consistently within the function definition.

With our new lambda notation, we adopt the following convention for writing function definitions:

| **Lambda notation:** When writing “*λx* τ.ϕ”, where τ is a type, we mean: |
| “the function sending every element *x* of the domain *D*_ τ to ϕ”. |

The expression ϕ within the lambda term *λx* τ.ϕ specifies the object that we want the function to return. In our definition of the identity function in (3.23), the expression ϕ was simply the argument *g* itself. However, in general, any mathematical expression ϕ that describes an object in one of our domains would be appropriate in such a λ-term. The type of the object that ϕ describes is the type of the value returned by the function. In (3.23), the type of the value *g* returned by the function is *et*, and hence the function is of type *(et)(et)*. Similarly,
we have:
\[ \lambda x_x \] the \( ee \) function sending every entity \( x \) to \( x \) itself
\[ \lambda f_{e.t.} tina_e \] the \( (e)t \) function sending every \( e \) function \( f \) to the entity \( tina \)
\[ \lambda h_{e.t.} h(tina_e) \] the \( (e)t \) function sending every \( e \) function \( h \) to the truth-value \( h(tina) \) that \( h \) assigns to the entity \( tina \)

More generally, when \( \varphi \) describes an object of type \( \sigma \), the whole \( \lambda \)-term \( \lambda x_{\tau} . \varphi \) describes a function of type \( \tau \sigma \): from objects in \( D_\tau \) to objects in \( D_\sigma \).

**FUNCTION APPLICATION WITH LAMBDA TERMS**

Now we can see how \( \lambda \)-terms are used in the semantic analysis. Based on definition (3.23), we can rewrite the equation \( \text{IS}(\text{tall}_{e.t}) = \text{tall} \) in (3.17a) as follows:

(3.24)  
\begin{align*}
\text{a. } & \text{IS}(\text{tall}_{e.t}) \\
& = (\lambda g_{e.t.} g)(\text{tall}) \\
& = \text{tall}
\end{align*}

The move from (3.24a) to (3.24b) is according to the definition of the function \( \text{IS} \). The move from (3.24b) to (3.24c) involves applying a function to its argument. As a result, we replace the abstract description in (3.24b) “what the identity function returns when it gets the argument \( \text{tall} \)” by writing more simply and concretely “\( \text{tall} \)”. Function application with \( \lambda \)-terms always involves this sort of concretization. Suppose that we have a function \( f \) described by the instruction:

“for every \( x \) of type \( \tau \) do such and such to \( x \) and return the result”.

Suppose further that we apply \( f \) to an argument \( a \) of the right type. What we get as the value \( f(a) \) is whatever “doing such and such” does to \( a \).

In our lambda notation, the expression \( \varphi \) in a \( \lambda \)-term \( \lambda x . \varphi \) describes what the function does with its argument. Within the identity function \( \lambda g_{e.t.} g \) in (3.24), the expression \( \varphi \) is the argument \( g \) itself. However, \( \varphi \) may encode a more complex operation on the function argument. For instance, let us again consider operations on numbers. Consider a
function \textsc{doubler} that maps any number to its multiplication by 2:

\textsc{doubler} is the function from numbers to numbers sending every number \( x \) to \( 2 \cdot x \).

In lambda format, this definition is written as follows, where \( n \) is again the type of numbers:

\begin{equation}
\textsc{doubler}_{nn} = \lambda x_{nn}. 2 \cdot x
\end{equation}

Having a name like \textsc{doubler} for this function is convenient when we commonly want to double numbers. However, when using \( \lambda \)-terms we can also avoid giving this function a name, and apply the term that corresponds to the function definition directly to the function’s argument, as we did in (3.24b) above. Suppose that we apply the function \textsc{doubler} to the number 17. We get the following term:

\begin{equation}
(\lambda x_{nn}. 2 \cdot x)(17)
\end{equation}

This corresponds to the following verbal description:

“the result of applying the function sending every number \( x \) to \( 2 \cdot x \), to the number 17”.

Of course, the result is the value of the arithmetic expression \( 2 \cdot 17 \). We get this expression by substituting ‘17’ for ‘\( x \)’ in the result definition \( 2 \cdot x \), as it is defined in the \( \lambda \)-term \( \lambda x_{nn}. 2 \cdot x \). In sum, we get:

\begin{equation}
(\lambda x_{nn}. 2 \cdot x)(17) = 2 \cdot 17
\end{equation}

In general, we describe this kind of simplification as follows:

\begin{quote}
\textbf{Function application with lambda terms:} The result \((\lambda x_{\tau}. \varphi)(a_{\tau})\) of applying a function described by a lambda term \( \lambda x_{\tau}. \varphi \) to an argument \( a_{\tau} \) is equal to the value of the expression \( \varphi \), with all occurrences of \( x \) replaced by \( a \).
\end{quote}

Consider how this convention works in (3.27). The expression ‘\( \varphi \)’ within the lambda term \( \lambda x_{nn}. 2 \cdot x \) is the expression \( 2 \cdot x \). The argument ‘\( a \)’ is the number 17. Substituting 17 for \( x \) in \( 2 \cdot x \), we get the result \( 2 \cdot 17 \). Something similar happens when we use a \( \lambda \)-term for defining
the function \textit{ADD} that we introduced in (3.13):

\[ \text{ADD} = \lambda y_n.\lambda x_n. y + x \]

As in (3.13), this \(\lambda\)-term defines the function \textit{ADD} by specifying that it sends every number \(y\) to the function \(\lambda x_n. y + x\): the function sending every number \(x\) to \(y + x\). When describing the result of applying the function \textit{ADD} to the values 1 and 5, we get the following simplifications:

(3.28) \((\lambda y_n.\lambda x_n. y + x)(1))(5) = (\lambda x_n.1 + x)(5) = 1 + 5\)

The simplification in (3.28) involves two steps. First, when the function \textit{ADD} takes the argument 1, this value is substituted for \(‘y’\). The result is the function \(\lambda x_n.1 + x\). This function applies to 5. When 5 is substituted for \(‘x’\), we get the result \(1 + 5\).

In (3.24), (3.27) and (3.28) above we use \(\lambda\)-terms for function application. In all of these cases, when a function \(\lambda x.\varphi\) applies to an argument \(a\), the result \((\lambda x.\varphi)(a)\) was displayed in a simplified notation, by substituting \(a\) for \(x\)’s occurrences in \(\varphi\). This substitution technique is based on common mathematical intuitions about the workings of functions. However, as usual, we should be careful when formalizing intuitions. There are some cases where naively using substitution as described above fails to derive the correct results. When using \(\lambda\)-terms for actual calculations, we need a more general rule for simplifying \(\lambda\)-terms under function application. This rule is known as \textbf{beta-reduction}, and it constitutes part of the rule system for computing \(\lambda\)-term equivalences, known as the \textbf{lambda calculus}. For our purposes in this book we will not need the full lambda calculus. Rather, the informal notational conventions above for writing and simplifying \(\lambda\)-terms will be sufficient. The reader is assured that none of our examples involve the formal complications that motivated the more intricate rule of beta-reduction in the lambda calculus. For more on this point, see the further reading at the end of this chapter.

\textbf{REFLEXIVE PRONOUNS}

In order to get a better feel for the value of lambda notation, let us now see how it works in a more complicated example: the case of \textit{reflexive pronouns} like \textit{herself} or \textit{himself}. Consider the following sentence:
(3.29) Tina [praised herself]

Intuitively, this sentence should be treated as having the following truth-value:

(3.30) praise_{et}(tina)(tina)

In words, (3.30) is the truth-value that results from applying the denotation of praise twice to the entity denotation of Tina. What denotation of the pronoun herself can guarantee that this truth-value is derived for sentence (3.29)? One obvious way to derive (3.30) from (3.29) is to let the pronoun herself denote the entity tina. This would be a correct treatment of sentence (3.29), but it definitely could not work as a general account of reflexive pronouns. To see why, let us consider the sentence Mary praised herself. Here, we must guarantee that the entity for Mary is given twice to the function praise. More generally, for any entity $x$ denoted by the subject, we must guarantee that that same entity is given twice as an argument to the function praise. This “sameness” cannot be described by treating the pronoun herself as denoting an entity of type $e$. For instance, if we analyzed herself as denoting the entity tina, the unwelcome result would be that the sentence Mary praised herself would be analyzed as having the same truth-value as Mary praised Tina. Similar problems would appear for any analysis of the pronoun herself as an entity-denoting noun phrase.

Let us see how we solve the problem. First, following our previous analyses, we let the verb phrase praised herself in (3.29) denote a function of type $et$. In this way we can treat it on a par with verb phrases like praised Mary. Thus, we want to solve the following type equation for the verb phrase praised herself, where $X$ is the type for herself:

$$e(et) + X = et$$

Letting $X$ be $e$ was fine type-theoretically, but it did not allow us to respect the semantic properties of reflexive pronouns. Fortunately, there is another solution: $X = (e(et))(et)$. Thus, we will let the pronoun herself denote a function that takes the $e(et)$ denotation praise as an argument, and returns a function of type $et$: the denotation of the phrase praised herself. These type-theoretical ideas about the analysis of sentence (3.29) are summarized in Figure 3.5.
The analysis in Figure 3.5 does not yet define the \((e(et))(et)\) function that the pronoun \(\text{HERSELF}\) should denote. To define it correctly, we rely on our intuition that sentence (3.29) has to have the same truth-value as (3.30). Thus, what we want to achieve is the following identity:

\[
(3.31) \quad [[\text{praised herself}]](\text{tina}) = \text{praise}(\text{tina})(\text{tina})
\]

In words: the one-place predicate for \(\text{praised herself}\) holds of the entity \(\text{tina}\) if and only if the two-place predicate \(\text{praise}\) holds of the pair \(\langle \text{tina}, \text{tina} \rangle\).

The denotation \([[\text{praised herself}]]\) in (3.31) is obtained by applying the denotation of \(\text{herself}\) to the function \(\text{praise}\). Thus, by spelling out the denotation \([[\text{praised herself}]]\), we restate (3.31) as the following identity:

\[
(3.32) \quad (\text{HERSELF}_{(e(et))(et)}(\text{praise}_{e(et)}))(\text{tina})
\]

\[
= [[\text{praised herself}]](\text{tina})
\]

\[
= \text{praise}(\text{tina})(\text{tina})
\]

In words: when the function \(\text{HERSELF}\) applies to the denotation \(\text{praise}\), the result is the \(et\) function \([[\text{praised herself}]]\) in (3.31) above. As we saw, this resulting \(et\) function sends the entity \(\text{tina}\) to 1 if and only if the two-place predicate \(\text{praise}\) holds of the pair \(\langle \text{tina}, \text{tina} \rangle\).
Since the denotations of the verb praise and the name Tina are arbitrary, we want the function HERSELF to satisfy the identity in (3.32) for all possible $e(\text{et})$ and $e$ denotations. We therefore conclude:

(3.33) For all functions $R$ of type $e(\text{et})$, for all entities $x$:
$$\text{HERSELF}(e(\text{et}))(R))(x) = R(x)(x)$$

This generalization defines the result of the function HERSELF for every argument of type $e(\text{et})$. Thus, we get:

(3.34) HERSELF($e(\text{et})$) is
the function sending every function $R$ of type $e(\text{et})$ to the $e$ function sending every entity $x$ to $R(x)(x)$.

This definition is rather long. Let us save space and use a $\lambda$-term for rewriting it:

(3.35) HERSELF($e(\text{et})$)
$$= \lambda R_{e(\text{et})}. \text{the } e \text{ function sending every entity } x \text{ to } R(x)(x)$$

But behold: we can save more space! The $e$ function that HERSELF returns can also be written as a lambda term: $\lambda x. R(x)(x)$. Here is what we get when we use it in (3.35):

(3.36) HERSELF($e(\text{et})$)
$$= \lambda R_{e(\text{et})}. (\lambda x. R(x)(x))$$

Since we don’t like unnecessary parentheses, we will write instead:

(3.37) $\lambda R_{e(\text{et})}. \lambda x. R(x)(x)$

And this is as much as we get by abbreviating our definition for the function HERSELF.

We can now write our analysis of the truth-value of sentence (3.29) using $\lambda$-terms alone, with some notes for clarification, but without any
description of functions in natural language:

(3.38) a. \((\text{HERSELF}(e(et))(\text{praise}_{e(et)}))(\text{tina})\)

\(\triangleright\) compositional analysis (Figure 3.5)

b. \(= ((\lambda R_{e(et)} \lambda x_e. R(x)(x))(\text{praise}))(\text{tina})\)

\(\triangleright\) definition (3.37) of HERSELF

c. \(= (\lambda x_e. \text{praise}(x)(x))(\text{tina})\)

\(\triangleright\) application to praise

d. \(= \text{praise}(\text{tina})(\text{tina})\)

\(\triangleright\) application to tina

Our compositional analysis assigns the verb phrase *praised herself* the denotation \(\text{HERSELF}(\text{praise})\), of type \(et\). This is the same type we assigned to the intransitive verb *smiled* and to the verb phrase *praised Mary*. As we see in (3.38b–c), with our definition of the function HERSELF, the denotation we get for *praised herself* is the \(et\) function \(\lambda x_e. \text{praise}(x)(x)\), which characterizes the set of self-praisers, as intuitively required by sentence (3.29). By letting all verb phrases denote \(et\) functions we obtain pleasing uniformity in our system. This uniformity will prove useful later on in this chapter, when we analyze verb phrase conjunctions like *smiled and praised herself*, as we will do in (3.53)–(3.54) below, and in Exercise 12c.

You are now advised to solve Exercises 8, 9, 10 and 11 at the end of this chapter.

PART 4: RESTRICTING DENOTATIONS

In Chapter 2 we introduced the distinction between constant and arbitrary denotations. This distinction is also instrumental with our new type system. For words like *Tina, smile, tall* and *praise*, we assume arbitrary denotations of the type we assign. By contrast, when analyzing functional words like *is, not* and *herself*, we focus on one denotation of the relevant type. In this part we study more examples where denotations are restricted in this way. This will lead us to some general conclusions about the relations between formal semantics and the specification of lexical meanings.

PROPOSITIONAL NEGATION AND PREDICATE NEGATION

Let us now get back to the negation in sentence (3.18) (*Tina is not tall*). We analyzed this sentence using definition (3.19), which describes
the set complement operation using characteristic functions. Using \(\lambda\)-terms we restate this definition as follows:

\[
(3.39) \ \text{NOT} = \lambda_{et} \lambda_x e. \begin{cases} 
1 & \text{if } g(x) = 0 \\
0 & \text{if } g(x) = 1 
\end{cases}
\]

In words: the function \(\text{NOT}\) sends every \(et\) function \(g\) to the function that sends every entity \(x\) to 1 in case \(g(x) = 0\), and to 0 in case \(g(x) = 1\). Driven by our wish to save space, we can shorten up the “piece-wise” definition in (3.39) by using propositional negation: the function from truth-values to truth-values that sends 0 to 1, and 1 to 0. We denote this \(tt\) function ‘\(\sim\)’. For its formal definition we can use subtraction, as stated in (3.40) below:

\[
(3.40) \ \sim = \lambda_x. 1 - x
\]

Using this definition of propositional negation, we can rewrite definition (3.39) above more concisely as:

\[
(3.41) \ \text{NOT} = \lambda_{et} \lambda_x e. \sim(g(x))
\]

The three definitions (3.19), (3.39) and (3.41) are equivalent, and they all boil down to the analysis of the word \(not\) as set complementation. However, since we now work with characteristic functions, our use of propositional negation in (3.41) has some presentational advantages. Given the structure in (3.18) (page 62), we analyze the sentence \(Tina\ is\ not\ tall\) as in (3.42) below.

\[
(3.42) \ \begin{align*}
\text{a. } & (\text{IS}(et)(et)(\text{NOT}(et)(et)(\text{tall}))) (\text{tina}) \\
& \triangleright \text{compositional analysis of structure (3.18)} \\
\text{b. } & = ((\lambda_{et} g. (\text{NOT}(\text{tall})))) (\text{tina}) \\
& \triangleright \text{definition of IS as identity function} \\
\text{c. } & = (\text{NOT}(\text{tall})) (\text{tina}) \\
& \triangleright \text{application to NOT(tall)} \\
\text{d. } & = ((\lambda_{et} x. \sim(g(x)))(\text{tall}))(\text{tina}) \\
& \triangleright \text{definition (3.41) of NOT} \\
\text{e. } & = ((\lambda_x x. \sim(\text{tall}(x))))(\text{tina}) \\
& \triangleright \text{application to tall} \\
\text{f. } & = \sim(\text{tall}(\text{tina})) \triangleright \text{application to tina}
\end{align*}
\]
In (3.42) we see two different representations of the same truth-value that our model assigns to sentence (3.18). In (3.42a–c) it is easy to see that the truth-value assigned to the sentence is 1 if the denotation tina is in the complement of the set characterized by the et denotation tall. This was the view that we adopted in Chapter 2. It is also very much in line with our compositional analysis of the sentence. After simplifying the representation by employing propositional negation, (3.42d–f) makes it easier to see that the truth-value that we assign to (3.18) is 1 if the function tall sends tina to 0. In simple sentences such as (3.18) these two perspectives are equivalent. However, the general question of how we should use propositional negation when analyzing the semantics of negative sentences is more complex. Later on in this chapter, and in Exercise 12e below, we briefly touch upon this problem. See also the further reading at the end of this chapter.

PROPOSITIONAL CONJUNCTION AND PREDICATE CONJUNCTION

Another useful propositional operator is propositional conjunction. In Chapter 2 we focused on predicate conjunction of adjectives and simple entailments as in (3.43) below:

(3.43) Tina is tall and thin ⇒ Tina is thin.

We also briefly noted the use of the conjunction and between sentences, and the very similar entailments that it often leads to. This is illustrated again in (3.44) below.

(3.44) Tina is tall and Tina is thin ⇒ Tina is thin.

For the sentential conjunction in (3.44) we assume the following binary structure:

(3.45) \[ Tina \text{[ is tall]} \]\text{[ and] }[ Tina \text{[ is thin]}]

In (3.45) the conjunction word and combines with the right-hand sentence and forms the constituent and Tina is thin. This expression combines with the left-hand sentential conjunct, of type t, and together they form a sentence of the same type. Thus, the type we assign to the constituent and Tina is thin in (3.45) is a function of type
tt, from truth-values to truth-values. This function is derived by combining the conjunction word \textit{and} with a sentence of type \( t \), and hence we conclude that the word \textit{and} in (3.45) is of type \( t(tt) \): a function from truth-values to functions from truth-values to truth-values. We define this denotation as the Curried version of the classical \textit{propositional conjunction} operator \( \land \). Propositional conjunction is a two-place function that maps two truth-values into a single truth-value. In numeric terms, it amounts to \textit{multiplication} between truth-values: the binary function that maps a pair of truth-values to 1 if both of them are 1, and to 0 otherwise. Formally:

\[(3.46) \text{ For any two truth-values } x \text{ and } y: \text{ the truth-value } x \land y \text{ is } x \cdot y, \text{ the multiplication of } x \text{ by } y.\]

In (3.47) below, we define our Curried \( t(tt) \) version of propositional conjunction using lambda notation. In order to distinguish this \( t(tt) \) denotation of \textit{and} from other denotations of this word, we refer to this \( t(tt) \) function as ‘\textit{AND}':

\[(3.47) \text{ AND}' = \lambda x_t. \lambda y_t. y \land x \]

In words: the denotation of sentential \textit{and} maps any truth-value \( x \) to the function mapping any truth-value \( y \) to the multiplication \( y \land x \) of \( x \) and \( y \). For example, for structure (3.45), we get:

\[(3.48) \text{ AND}'(\llbracket \text{Tina is thin}\rrbracket)(\llbracket \text{Tina is tall}\rrbracket) = \llbracket \text{Tina is tall}\rrbracket \land \llbracket \text{Tina is thin}\rrbracket \]

The denotation of the second conjunct in (3.45) is used as the first argument of the function \textit{AND}'\( t \). When \textit{AND}'\( t \) sends its arguments to the \( \land \) operator, we let it reverse their order, so that the first argument of \textit{AND}'\( t \) is the second argument of \( \land \). This is innocuous, since \( y \land x \) is the same as \( x \land y \). The reason we reverse the order is merely aesthetic: it is more pleasing to the eye to see a ‘\( y \land x \)’ notation when \( x \) is the denotation of the right-hand conjunct in the sentence.

For a fuller analysis of structure (3.45), see Figure 3.6. The semantic interpretation uses the functions \textit{AND}'\( t \) for \textit{and}, and the identity function \textit{IS} for \textit{is}.

In (3.49) below we use our definition of the denotations \textit{AND}'\( t \) and \textit{IS} to analyze the truth-value derived in Figure 3.6.
Figure 3.6  Syntactic structure and semantic interpretation for Tina is tall and Tina is thin.

(3.49) a. \((\text{AND}^t((\text{IS}\,\text{thin})(\text{tina})))((\text{IS}\,\text{tall})(\text{tina})) : t\)  
    \(\triangleright\) compositional analysis in Figure 3.6

b. \(= (\text{AND}^t((\text{IS}\,\text{tall})(\text{tina})))((\text{IS}\,\text{tall})(\text{tina}))\)  
    \(\triangleright\) applying IS (identity function)

c. \(= ((\lambda x. \lambda y. y \land x)((\text{thin}(\text{tina})))((\text{tall}(\text{tina})))\)  
    \(\triangleright\) definition of \text{AND}^t

d. \(= ((\lambda y. \lambda y. y \land (\text{thin}(\text{tina}))))((\text{tall}(\text{tina})))\)  
    \(\triangleright\) application to \text{thin(tina)}

e. \(= \text{tall(tina)} \land \text{thin(tina)}\)  
    \(\triangleright\) application to \text{tall(tina)}
As in our example of number addition, we see that lambda notation allows us to use Curried functions while reverting to traditional notation when convenient. The Curried notation in (3.49a–b) is convenient when considering the truth-value derivation from syntactic forms as in Figure 3.6. The standard notation in (3.49e) is convenient if we are interested in the relations between our results and classical logical analyses. However, the two notations are equivalent in terms of the truth-values they represent.

As noted in Chapter 2, the connections between our compositional treatment of conjunction and classical analyses become less straightforward when we consider predicate conjunction. In Chapter 2 we analyzed the conjunction between adjectives in the sentence *Tina is tall and thin* using set intersection. Given our current emphasis on the use of Curried functions, it is convenient to analyze the same sentence analogously to (3.45), using only binary structures. Thus, we now assume the structure in (3.50) below:

(3.50) Tina [ is [ tall [ and thin ]]]

To mimic our intersective analysis of conjunction, here we let the word *and* denote a function of type \((et)((et)(et))\). When we read it in English, this is the type that describes *functions from characteristic functions to functions from characteristic functions to characteristic functions*. You may say gee whiz, but when replacing “characteristic functions” by “sets”, we see that it is just our usual Currying practice. The functions we have just mentioned correspond to functions that map every pair of sets to a set. The intersection operator is such a function: it sends every pair of sets to their intersection. Thus, type \((et)((et)(et))\) is the proper type for defining an operator on \(et\) functions that mimics set intersection.

When we define the denotation of predicate conjunction in (3.50) as a function of type \((et)((et)(et))\), it is convenient, as we did in the case of predicate negation, to use the corresponding propositional operator. Below we give our denotation of predicate conjunction, denoted ‘\(^{\text{AND}}_{et}\)’, using the operator \(\wedge\) between truth-values:

(3.51) \(^{\text{AND}}_{et} = \lambda f_{et}.\lambda g_{et}.\lambda x_{et}.g(x) \wedge f(x)\)
In words: the denotation of the predicate conjunction *and* maps any *et* function $f$ to the function mapping any *et* function $g$ to the function mapping any entity $x$ to $g(x) \land f(x)$. This definition has the following simple property: when two *et* functions $h_1$ and $h_2$ characterize sets of entities $A_1$ and $A_2$, respectively, the result of applying the function $\text{AND}^{et}$ to $h_1$ and $h_2$ is the function characterizing the intersection of $A_1$ and $A_2$ (see Exercise 12). Thus, treating predicate conjunction using $(et)((et)(et))$ functions encodes the same analysis of (3.50) as in Chapter 2, where we used set intersection.

To see in more detail how definition (3.51) works, consider the analysis of structure (3.50) in (3.52) below:

\[ (3.52) \]

d. \[ = (\text{tall}(x) \land \text{thin}(x))(\text{tina}) \] \> application to \text{tall}

e. \[ = (\text{tall}(\text{tina}) \land \text{thin}(\text{tina})) \] \> application to \text{tina}

The truth-value that we end up deriving for sentence (3.50) (*Tina is tall and thin*) is the same as the one we got for (3.45) (*Tina is tall and Tina is thin*). Thus, we have captured the intuitive semantic equivalence between these sentences. However, note that simplifications as in (3.49) and (3.52) are only one way of explaining this equivalence, and sometimes they can obscure insights about denotations of constituents. Specifically, in (3.52f) we no longer see that our compositional analysis of predicate conjunction is equivalent to the set intersection of Chapter 2. The direct compositional analysis in (3.49a) highlights this fact more clearly. More generally, we analyze the equivalence between (3.50) and (3.45) as following from elementary considerations about the relationships between set intersection
(in (3.50)) and multiplication of truth-values (in (3.45)). In formula, let \( \chi_A \) and \( \chi_B \) be *et* functions that characterize the sets of entities \( A \) and \( B \), respectively. For every entity \( x \), we have:

\[
x \in A \cap B \text{ if and only if } \chi_A(x) \cdot \chi_B(x) = 1
\]

For instance, the entity *tina* is in the intersection of the sets characterized by the *et* functions *tall* and *thin* if and only if both functions send *tina* to 1. Thus, we can see that we account for the equivalence already in the interpreted structures (3.52a) and (3.49a), without any simplification of \( \lambda \)-terms. This point is of some historical interest: unlike early analyses of conjunction in the 1960s, our compositional semantics directly interprets the structures of both sentences (3.50) and (3.45). Neither of these structures is assumed to be “deeper” or “more basic” than the other.

Our treatment of adjective conjunction can now be directly used for other predicates, especially verb phrases as in the examples below:

(3.53) Tina smiled and danced. Tina smiled and praised Mary. Tina praised Mary and smiled. Tina praised Mary and thanked John.

(3.54) Tina smiled and praised herself. Tina thanked Mary and praised herself.

These sentences with verb phrase conjunctions also demonstrate an equivalence with sentential conjunction. Consider for instance the following equivalence:

\[
\text{Tina thanked Mary and praised herself } \leftrightarrow \text{ Tina thanked Mary and Tina praised herself}
\]

Such equivalences are immediately explained by our analysis of propositional conjunction and predicate conjunction. However, it should be noted that the equivalence scheme is not valid in general: it does not necessarily hold with more complex subjects. For instance, the sentence *someone is smiling and someone is dancing* does not entail *someone is smiling and dancing*: the existence of people doing two different things does not mean that someone is doing both. We will return to this point in Chapter 4.

As with our treatment of negation, the lambda notation exposes the semantic relation between a propositional operator (\( \land \)) and a
set-theoretical operator (intersection), where the latter is presented by means of characteristic functions. However, in contrast to our treatment of negation, where propositional negation was not directly used in our compositional analysis, now it becomes clear that both propositional conjunction and set intersection are useful as denotations of conjunction: the first has a straightforward use with sentential conjunctions, the latter with predicate conjunction. To conclude, our type system requires the use of two denotations for conjunction: \( \text{AND} \) for sentential conjunction and \( \text{AND}^* \) for predicate conjunction. Although the two functions are of different types, namely \( t(tt) \) and \( (et)(et)(et) \), they are logically related. In fact, the parallelism we have observed only reflects a small part of the semantic relations between and conjunctions of different categories in natural language. Similar relations exist between disjunctive coordinations with or in different categories, and, to a lesser extent, between negation in different categories. These relations are often analyzed as revealing Boolean structures in natural language semantics. Relevant details can be found in the further reading at the end of this chapter.

INTERSECTIVE ADJECTIVES AND SUBSECTIVE ADJECTIVES

Let us move on to another example where set intersection and propositional conjunction play a major role: the different usages of adjectives. So far we have only considered adjectives in sentences like Tina is tall, where they appear in predicative positions, following the copula is. However, English also allows adjectives to precede nouns, as in the following sentences:

(3.55) Tina is a tall woman; the tall engineer visited us; I met five tall astronomers.

Occurrences of adjectives before the noun as in (3.55) are often referred to as ‘attributive’, or modificational. In many cases, we find strong semantic relations between these modificational occurrences and the predicative use. Consider for instance the following equivalences:

(3.56) a. Tina is a Chinese pianist ↔ Tina is Chinese and Tina is a pianist.
b. My doctor wears no *white* shirts ⇔ No shirts that my
doctor wears are *white*.
c. Dan saw six *carnivorous* animals ⇔ Six animals that Dan
saw are *carnivorous*.

In each of these examples, we see an equivalence between a sentence
with a modificational adjective, and another sentence with the same
adjective in a predicative position. When analyzing such equivalences,
we will concentrate on (3.56a) as a representative example. Before
further analyzing the sentence *Tina is a Chinese pianist* in (3.56a), let
us first consider the following simpler sentence:

(3.57) Tina [ is [ a pianist ]]

Given the constituency that we assume in (3.57), we analyze the noun
*pianist* as denoting a function pianist of type et. This is similar to
our treatment of intransitive verbs and predicative adjectives. The
indefinite article *a* is analyzed, similarly to the copula *is*, as denoting
the identity function of type \((et)(et)\). Formally:

\[(3.58) A(et)(et) = 1s = \lambda g_{et}.g\]

With these assumptions, structure (3.57) is analyzed as denoting the
following truth-value.

\[(3.59) (1s(A(pianist)))(tina) = pianist(tina)\]

This truth-value is 1 when the entity *tina* is in the set characterized
by the function pianist, and 0 otherwise. As for the modificational
construction in (3.56a), we now assume the following structure:

(3.60) Tina [ is [ a [ Chinese pianist ]]]

We already know how to analyze sentences with predicative adjectives
like *Tina is Chinese*, where we let the adjective denote an et function.
It might be tempting to let the modificational occurrence of the
adjective *Chinese* in (3.60) denote the same et function. However,
with such an analysis, we would have a problem when treating the
constituent *Chinese pianist* in (3.60). If both the adjective and the
noun are assigned the type et, function application would not be
able to combine their denotations. Our solution to this problem is to assume that there are two different denotations of the adjective Chinese. One denotation is the arbitrary et function chinese that we use for the predicative use, e.g. in Tina is Chinese. The other denotation is used for modificational occurrences as in (3.60). The modificational denotation is a function of type (et)(et), from characteristic functions to characteristic functions. This will allow the adjective to combine with nouns like pianist, as in (3.60). To highlight its use, we refer to this function as ’chinese\text{mod}'. Now, since there is a semantic connection between the two usages of the adjective Chinese, we define the (et)(et) denotation chinese\text{mod} on the basis of the et function chinese. Specifically, we associate chinese\text{mod} with the function mapping any set A to the intersection of A with the Chinese entities. In this way, the expression Chinese pianist is associated with the set of pianists who are also Chinese. In lambda notation we define the function chinese\text{mod} as follows:

(3.61) $\text{chinese}^{\text{mod}}_{(et)(et)} = \lambda f_{et}. \lambda x_e. \text{chinese}(x) \land f(x)$

The function chinese\text{mod} maps any et function f to the function that maps an entity x to 1 if and only if x is in both sets that are characterized by the et functions f and chinese. Treating the constituent Chinese pianist using this denotation, we get the following analysis of sentence (3.60):

(3.62) $(\text{is}(\lambda (\text{chinese}^{\text{mod}}(\text{pianist}))(\text{tina})))$  
▷ compositional analysis of (3.60)  
$= (\text{chinese}^{\text{mod}}(\text{pianist}))(\text{tina})$  
▷ applying is and $\land$ (identity functions)  
$= ((\lambda f_{et}. \lambda x_e. \text{chinese}(x) \land f(x))(\text{pianist}))(\text{tina})$  
▷ definition (3.61) of chinese\text{mod}  
$= (\lambda x_e. \text{chinese}(x) \land \text{pianist}(x))(\text{tina})$  
▷ application to pianist  
$= \text{chinese}(\text{tina}) \land \text{pianist}(\text{tina})$  
▷ application to tina

The analysis in (3.62) immediately accounts for the equivalence we observed in (3.56a). Similar analyses of the modificational usage of
adjectives also account for the equivalences in (3.56b) and (3.56c) with the adjectives white and carnivorous. Because their analysis as modifiers involves intersecting the noun denotation with the predicative adjective’s denotation, adjectives like Chinese, white and carnivorous are referred to as intersective.

Intersective adjectives also support equivalences like the one between (3.63a) and (3.63b) below:

(3.63)  
   a. Tina is a Chinese pianist and a biologist.  
   b. Tina is a Chinese biologist and a pianist.

We can now easily account for such equivalences as well. Our analysis of the pre-nominal usage of the adjective Chinese is based on set intersection. The occurrences of Chinese in (3.63a) and (3.63b) are interpreted by intersecting the set C of Chinese entities with another set: the set P of pianists, and the set B of biologists, respectively. With these notations, let us consider the following set-theoretical equality:

(3.64)  \((C \cap P) \cap B = (C \cap B) \cap P\)

In words: the intersection of the set B with the intersection of C and P is the same as the intersection of P with the intersection of C and B. Having observed this equality, we see that the equivalence in (3.63) follows directly from our analysis of intersective adjectives.

It is important to note that, although intersective interpretations are pretty common, adjectives may also show non-intersective behavior in their modificational use. Unlike what we saw with the adjective Chinese, there are other adjectives that do not support the equivalence pattern in (3.63). Consider for instance the adjective skillful in (3.65) below:

(3.65)  
   a. Tina is a skillful pianist and a biologist.  
   b. Tina is a skillful biologist and a pianist.

Sentence (3.65a) can be true if Tina is competent as a pianist but amateurish as a biologist. Thus, (3.65a) does not entail (3.65b). For a similar reason, (3.65b) does not entail (3.65a). This lack of equivalence shows that we cannot analyze constructions like skillful pianist or skillful biologist by intersecting the set of pianists/biologists with
some set of “skillful entities”. Such an analysis, together with the set-theoretical equality we saw in (3.64), would lead us to incorrectly expect equivalence in (3.65). Intuitively, we say that the adjective skillful shows a non-intersective behavior in (3.65). How can we describe this usage?

To solve this problem, we assume that the basic denotation of the adjective skillful is of type \((et)(et)\). This immediately allows us to construct models where one of the sentences in (3.65a–b) denotes 1 and the other denotes 0. The reason is that, when the \((et)(et)\) function associated with skillful is arbitrary, we no longer assume that the denotations of skillful pianist and skillful biologist are formed by intersection. Specifically, let us look at a model with the following denotations:

\[
\begin{align*}
\text{pianist}_{et} & : \text{characterizes the singleton set } \{tina\} \\
\text{biologist}_{et} & : \text{characterizes the set } \{tina, mary\} \\
[[\text{skillful}]]_{(et)(et)}(\text{pianist}) & : \text{characterizes the singleton set } \{tina\} \\
[[\text{skillful}]]_{(et)(et)}(\text{biologist}) & : \text{characterizes the singleton set } \{mary\}
\end{align*}
\]

In words: the only pianist is Tina, the only biologists are Tina and Mary, the only skillful pianist is Tina, and the only skillful biologist is Mary.

In this model, we use the \((et)(et)\) denotation for skillful, and not an \(et\) function denotation. Thus, we consider Tina skillful relative to the denotation of pianist, but not relative to the denotation of biologist. Because our analysis does not use any set of “skillful entities”, the fact that Tina is a biologist in the model does not entail that she is considered a skillful biologist. Thus, sentence (3.65a) denotes 1 whereas (3.65b) denotes 0. This agrees with our intuitions about the lack of entailment between these sentences.

But now we have to treat another property of the adjective skillful. Let us consider the following entailment:

\[(3.66) \text{ Tina is a skillful pianist } \Rightarrow \text{ Tina is a pianist.}\]

This entailment reflects the obvious intuition that every skillful pianist is a pianist. Adjectives like skillful that show this behavior are often
called *subsective adjectives*. An alternative name for these adjectives is ‘restrictive’ or ‘restricting’. The judgment about the validity of (3.66) must lead us to think a little more about the \((et)(et)\) denotation of the adjective *skillful*. This denotation cannot be allowed to be any \((et)(et)\) function. If such arbitrariness were allowed, models might let the denotation of *skillful pianist* be any set of entities they contain, not necessarily a subset of the set of pianists. For instance, a function from sets to set can send the set \(\{a, b\}\) to the set \(\{c, d\}\). To avoid such situations, we require that the denotation of *skillful* sends any set \(P\) to a *subset* of \(P\). More formally, we define this general restriction as follows:

\[(3.67)\] For any model \(M\), the denotation of *skillful* in \(M\) is an \((et)(et)\) function \(\text{skillful}^{\text{mod}}\) that satisfies the following: for every \(et\) function \(f\) in \(M\), the set characterized by \(\text{skillful}^{\text{mod}}(f)\) is a subset of the set characterized by \(f\).

Another possible way to state the restriction, which is elegant though harder to grasp, is to define \(\text{skillful}^{\text{mod}}\) so that it satisfies (3.67) by relying on another, arbitrary, function of type \((et)(et)\). When we denote this function \(\text{skillful}^{\text{arb}}\), the definition of \(\text{skillful}^{\text{mod}}\) reads as follows:

\[(3.68)\] \(\text{skillful}^{\text{mod}} = \lambda_{f et}. \lambda_{x e}. (\text{skillful}^{\text{arb}}(f))(x) \land f(x)\)

In words: the function \(\text{skillful}^{\text{mod}}\) sends every \(et\) function \(f\) to the characteristic function of the (arbitrary) set characterized by \(\text{skillful}^{\text{arb}}(f)\), intersected with the set characterized by \(f\). If the restriction (3.67) is all that we want our models to specify about the denotation of the adjective *skillful*, then it is equivalent to the definition in (3.68). The \((et)(et)\) function \(\text{skillful}^{\text{arb}}\) is free to vary from one model to another like the other arbitrary denotations we have assumed. The superscript \(arb\) is a reminder that we assume that the function \(\text{skillful}^{\text{arb}}\) is arbitrary, although it is not the denotation of the word *skillful*.

With the denotation in (3.68), the sentence *Tina is a skillful pianist* is analyzed as follows:
The analysis in (3.69) immediately accounts for the entailment in (3.66). At the same time, it also accounts for the lack of entailment in the opposite direction: when Tina is a pianist, it does not follow that she is a skillful pianist. This is easily explained, since when Tina is a pianist the truth-value derived in (3.69) may still be 0. This happens in models where the entity Tina is not in the set characterized by skillful\text{arb}(\text{pianist}).

We should note that intersective adjectives also show entailments as in (3.66). This is directly accounted for in our analysis. In other words, our treatment of intersective adjectives correctly expects them to be a sub-class of the adjectives we classified as subsective. Below we summarize the concepts of intersective and subsective adjective, and intersective and subsective adjective functions. For convenience, we look at functions from sets of entities to sets of entities.

The following entailments define an adjective \( A \) as being intersective/subsective, where \( X \) is a proper name and \( N \) is a common noun:

\( A \) is intersective \(-\) \( X \) is a \( A \) \( N \) \( \Leftrightarrow \) \( X \) is \( A \) and \( X \) is a \( N \)

e.g. Dan is a Dutch man
\( \Leftrightarrow \) Dan is Dutch and Dan is a man

\( A \) is subsective \(-\) \( X \) is a \( A \) \( N \) \( \Rightarrow \) \( X \) is \( N \)

e.g. Dan is a skillful pianist \( \Rightarrow \) Dan is a pianist

For a function \( F \) from \( \wp(E) \) to \( \wp(E) \) we define:

\( F \) is intersective \(-\) There is a set \( A \), s.t. for every set \( B \):
\( F(B) = A \cap B \).

\( F \) is subsective \(-\) For every set \( B \): \( F(B) \subseteq B \).

By treating the denotations of adjectives as intersective and subsective functions of type \((et)(et)\), we have been able to analyze the behavior of intersective and subsective adjectives.
The examples above show some cases of intersective adjectives like Chinese, as well as one example of a non-intersective, subsective adjective, skillful. Classifying adjectives into intersective and non-intersective is a highly complex problem, both empirically and theoretically. To see one example of the difficulty, let us reconsider our favorite adjectives tall and thin. So far, we have assumed that these adjectives denote arbitrary et functions. Therefore, in their modifi
cational usages, e.g. in tall woman and thin woman, we may like to use the intersective analysis. However, this would be questionable: tall children are not necessarily tall; thin hippos are not necessarily thin. On the other hand, treating tall and thin as subsective adjectives may also lead to complications. For instance, when saying that Tina is a child and she is tall, our assertion that Tina is a child affects our understanding of the adjective tall in much the same way as it does in the sentence Tina is a tall child. Following this kind of observation, many researchers propose that adjectives like tall and thin should be treated as intersective, while paying more attention to the way they are affected by the context of the sentence. This and other questions about the semantics of adjectives constitute a large body of current research. For some of these problems, see the further reading at the end of this chapter.

The semantic concepts of subsective and intersective functions are useful for other categories besides adjectives. Consider for instance the following entailments:

(3.70)  
a. Tina [smiled [charmingly]] ⇒ Tina smiled.
   b. Tina [ran [with John]] ⇒ Tina ran.

(3.71)  
a. Tina [is [a [pianist [from Rome]]]]
   ⇔ Tina is a pianist and Tina is from Rome.
   b. Tina [is [a [pianist [who [praised herself]]]]]
   ⇔ Tina is a pianist and Tina praised herself.

In sentences (3.70a–b), we see that the adverbial modifiers charmingly and with John give rise to ‘subsective’ entailments. Furthermore, the adnominal prepositional modifier from Rome in (3.71a) and the relative clause who praised herself in (3.71b) show ‘intersective’ equivalences. A simple analysis of the adverb charmingly may assign it a subsective denotation of type (et)(et). Similarly, the prepositional phrases with John and from Rome can be analyzed as subsective, or
even intersective, \((et)(et)\) functions. The relative clause \textit{who praised herself} can also be analyzed as an intersective \((et)(et)\) function. By solving Exercise 13, you may analyze these constructions in more detail.

SUMMARY: LEXICAL DENOTATIONS

By way of recapitulation, let us take stock of the variety of lexical denotations that we have so far assumed. First, for some words like \textit{is} and \textit{not} we assigned denotations on the basis of a \textit{definition}. The denotations of these words are fully specified by our analysis, with little freedom left for models to change them. These denotations are subdivided into two classes:

1. Denotations like \textit{is} and \textit{herself}: functions that are exclusively defined by means of their workings on other functions, without further definitions or assumptions. Such denotations are functions that can be expressed as ‘pure’ \(\lambda\)-terms, and they are also referred to as \textit{combinators}.
2. Denotations like \textit{not}, \textit{and} and \textit{and} \((et)\): functions that are defined by means of some additional concepts, e.g. the functions of propositional negation and propositional conjunction. Because these constant denotations rely on truth-values, they are often referred to as \textit{logical}.

Most words whose denotations are combinatorially or logically defined belong in the class that linguists call \textit{function words} or \textit{functional words}. These words are contrasted with \textit{content words}, which are the bulk of the lexicon in all natural languages. We have seen two kinds of denotations for content words:

3. \textit{Arbitrary} denotations, for which no restrictions hold in our models besides those following from their types. We gave such denotations to proper names (\textit{Tina}), common nouns (\textit{pianist}), verbs (\textit{smile}, \textit{praise}) and predicative usages of adjectives.
4. Denotations that are logically or combinatorially defined on the basis of other, arbitrary denotations. This is how we accounted for modificational adjectives, when deriving their denotations from arbitrary denotations.
Table 3.2: Lexical denotations and their restrictions.

<table>
<thead>
<tr>
<th>Denotation</th>
<th>Type</th>
<th>Restrictions</th>
<th>Category</th>
</tr>
</thead>
<tbody>
<tr>
<td>tina</td>
<td>e</td>
<td>-</td>
<td>proper name</td>
</tr>
<tr>
<td>smile</td>
<td>et</td>
<td>-</td>
<td>intransitive verb</td>
</tr>
<tr>
<td>praise</td>
<td>e(et)</td>
<td>-</td>
<td>transitive verb</td>
</tr>
<tr>
<td>pianist</td>
<td>et</td>
<td>-</td>
<td>common noun</td>
</tr>
<tr>
<td>chinese</td>
<td>et</td>
<td>-</td>
<td>predicative adjective</td>
</tr>
<tr>
<td>chinese^mod</td>
<td>(et)(et)</td>
<td>intersective: λf_εt, λx_ε.</td>
<td>modificational adjective</td>
</tr>
<tr>
<td></td>
<td></td>
<td>chinese(x) ∧ f(x)</td>
<td></td>
</tr>
<tr>
<td>skillful^mod</td>
<td>(et)(et)</td>
<td>subsective: λf_εt, λx_ε.</td>
<td>modificational adjective</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(skillful^{\text{ab}}(f))(x) ∧ f(x)</td>
<td></td>
</tr>
<tr>
<td>IS</td>
<td>(et)(et)</td>
<td>combinator: λg_εt, g</td>
<td>copula (auxiliary verb)</td>
</tr>
<tr>
<td>A</td>
<td>(et)(et)</td>
<td>combinator: λg_εt, g</td>
<td>indefinite article</td>
</tr>
<tr>
<td>HERSELF</td>
<td>(e(et))(et)</td>
<td>combinator: λR_{εt}, λx_ε, R(x)(x)</td>
<td>reflexive pronoun</td>
</tr>
<tr>
<td>NOT</td>
<td>(et)(et)</td>
<td>logical: λg_εt, λx_ε, ¬(g(x))</td>
<td>predicate negation</td>
</tr>
<tr>
<td>AND^t</td>
<td>(et)(et)</td>
<td>logical: λx_ε, λy_ε, y ∧ x</td>
<td>sentential conjunction</td>
</tr>
<tr>
<td>AND^et</td>
<td>(et)((et)(et))</td>
<td>logical: λf_εt, λg_εt, λx_ε, g(x) ∧ f(x)</td>
<td>predicate conjunction</td>
</tr>
</tbody>
</table>

The lexical denotations that we assumed, together with their restrictions, are summarized in Table 3.2.

This summary of our restrictions on lexical denotations only scratches the surface of a vast topic: the organization of lexical meanings. Let us briefly mention some of the questions that we have left untreated. Restrictions on lexical meanings that affect entailments do not only involve restricting the possible denotations of single entries. There are also many strong semantic relations between different lexical entries. Consider for instance the following examples:

(3.72) a. Tina danced ⇒ Tina moved.
     b. John is a bachelor ⇒ John is a man and John is not married.

The entailments in (3.72) illustrate that theories of entailment should also constrain the relations between lexical denotations. Entailment (3.72a) can be explained if the set associated with the verb dance is contained in the set for move; (3.72b) is explained when the set for bachelor is contained in the intersection between the set of men and the complement set of the married entities. Our theory should include an architecture that allows encoding such lexical restrictions on denotations more systematically than we have attempted to do here.
The role of syntactic theory in analyzing lexical meanings is another issue that should be further emphasized. For instance, consider the constituent structure Tina \([\text{is [not thin]}]\) that we assumed for sentences with predicate negation. This structure made us adopt the \((et)(et)\) type for the word \textit{not}. Now, suppose that for syntactic reasons we used the structure \textit{not [Tina [is thin]]}. A treatment of the word \textit{not} as propositional negation of type \(tt\) would then be forthcoming. We conclude that the choice between the types \((et)(et)\) and \(tt\) hinges heavily on theoretical syntactic questions about the structure of negation.

Such puzzles are highly challenging for theories about the relations between formal semantics and other parts of grammar, especially lexical semantics and syntactic theory. They constitute a fascinating and very active area of research in linguistics. Some references for works in this area can be found in the further reading below.

You are now advised to solve Exercises 12 and 13 at the end of this chapter.

FURTHER READING


\textbf{Advanced}: For more on type theory and higher-order logics see Thompson (1991); Kamareddine et al. (2004). The Ajdukiewicz Calculus is from Ajdukiewicz (1935). The original idea of Currying appeared in Schönfinkel (1924). Solving type equations is part of the more general problem of \textit{type inference} in programming languages (Gunter 1992), especially in relation to \textit{functional programming} (Hutton 2007; Van Eijck and Unger 2010). For an early semantic treatment of reflexive pronouns see Keenan (1989). For more on variable-free semantics, see Jacobson (1999); Keenan (2007); Hendriks (1993); Steedman (1997); Szabolcsi (1987). On the \(\lambda\)-calculus see Barendregt et al. (2013), and, in relation to combinators, Hindley

EXERCISES (ADVANCED: 7, 9, 10, 11, 12, 13)

1. For $D_e = \{u, v, w, m\}$:
   a. Give the functions in $D_{et}$ that characterize: (i) the set $\{u, w\}$; (ii) the empty set; (iii) the complement set of $\{u, w\}$, i.e. $\overline{\{u, w\}}$, or $D_e - \{u, w\}$.
   b. Give the set that is characterized by the function that sends every element of $D_e$ to 1.

2. a. Give the types for the following English descriptions (cf. (3.5)):
   (i) functions from functions from entities to entities to functions from entities to truth-values;
   (ii) functions from functions from entities to truth-values to entities;
   (iii) functions from functions from entities to truth-values to functions from truth-values to entities;
   (iv) functions from entities to functions from truth-values to functions from entities to entities;
   (v) functions from functions from entities to truth-values to functions from entities to functions from entities to entities;
   (vi) functions from entities to functions from functions from entities to truth-values to functions from truth-values to entities;
   (vii) functions from functions from functions from truth-values to truth-values to functions from truth-values to entities to functions from entities to functions from entities to entities.
   b. Give English descriptions (cf. (3.5)) for the following types:
      $(et)t$, $t(te)$, $(tt)e$, $(e(et))t$, $e((et)t)$, $(e(et))(e(tt))$.

3. a. Give the functions in $D_{tt}$.
   b. For $D_e = \{u, v, w, m\}$, give the functions in $D_{te}$ and $D_{et}$.
   c. For $D_e = \{l, n\}$, give the functions in $D_{(ee)t}$. 
4. For each of the following pairs of types say if our function application rule holds. If that’s the case, write the resulting type:

\[ e + e(et), \quad (et)t + et, \]
\[ tt + (ee)(tt), \quad et + e(tt), \]
\[ (e(et))(et) + ee, \quad e(et) + (e(et))(et), \]
\[ (e(et))(et) + e, \quad (e((et)t))(tt) + e((et)t), \]
\[ e((et)t)(tt) + e, \quad (ee)(et) + e, \]
\[ e((et)(et)) + e(\), \quad (et)(et) + ((et)(et))(e(e(et))). \]

5. a. In a model with \( D_e = \{ t, j, m \} \), suppose that Tina, John and Mary praised Mary, that Tina and Mary praised John, and that nobody else praised anybody else. What should the denotation of the verb praise be in this model?

b. Consider sentences of the form \( John \ [[read Mary] Moby Dick] \). We let the ditransitive verb \( read \) be of type \( e(e(et)) \). Assume that John read Mary Moby Dick, Tina read Mary Lolita, and nobody else read anything else to anybody else. In such a model, give the \( et \) denotation of the expression \( read Mary Moby Dick \) and the \( e(et) \) denotation of the expression \( read Mary \).

6. a. Solve the following type equations:

\[ tt + X = t(tt), \quad Y + e = ee, \quad Z + t = et, \quad (et)t + M = e((et)t). \]

b. Each of the following type equations has two solutions. Find them:

\[ t(tt) + X = tt, \quad Y + ee = e, \quad Z + et = t, \quad e((et)t) + M = (et)t. \]

c. Complete the general conclusions from your answers to 6a and 6b:

(i) Equations of the form \( X + y = z \) always have the solution \( X = _. \)

(ii) Equations of the form \( X + yz = z \) always have the solutions \( X = _. \) and \( X = _. \)

7. a. The sentence structures below introduce “typing puzzles” similar to (3.15). Solve these puzzles and find appropriate types for the underlined words.

(i) \([\text{Mary}_e \ [\text{walked}_{et} \text{quickly}_{et}]_Y ]_t; \)

(ii) \([\text{Mary}_e \ [\text{walked}_{et} \text{in}_X \text{Utrecht}_{et}]_Z ]_Y ]_t; \)

(iii) \([ [ \text{the}_X \text{pianist}_{et} ]_t e \ [\text{smiled}_{et}]_t; \)

(iv) \([ [ \text{the}_X [\text{skillful}_Y \text{pianist}_{et}]_et ]_e e \text{smiled}_{et} ]_t; \)
(v) \[[ \text{Walk}_{et} \text{ing}_X ]e \ [ \text{is}_{(et)(et)} \text{fun}_{et} \ ]_{et} \]
(vi) \[[ \text{the}_X [ \text{man}_{et} [ \text{who}_Y \text{walked}_{et} ]_Z ]_M ]_e \text{smiled}_{et} \]
(vii) \[[ \text{if}_X [ \text{you}_e \text{smile}_{et} ]_Y [ \text{you}_e \text{win}_{et} ]_t \]
(viii) \[[ I_e [ \text{love}_{e(et)} \text{it}_e ]_Y [ \text{when}_X [ \text{you}_e \text{smile}_{et} ]_t ]_Z ]_M ]

b. Now consider the following puzzle: \[[ \text{no}_X \text{man}_{et} ]_Y \text{smiled}_{et} \].
Give two solutions for \( Y \). For each solution give the corresponding solution for \( X \).

c. Based on your answers to 7a and 7b, find at least one solution for \( X, Y \) and \( Z \) in the following puzzle:
\[[ \text{There}_X [ \text{is}_{(et)(et)} [ \text{trouble}_{et} [ \text{in}_Y \text{Paradise}_{e} ]_Z ] ]_t \].
Can you find any more solutions?

8. a. Give English descriptions (cf. (3.34)) for the following \( \lambda \)-terms:
   (i) \( \lambda f_e t. \lambda y_e t. f(\lambda z_e t. y) \)
   (ii) \( \lambda x_e t. \lambda f_e t. f(x) \)
   (iii) \( \lambda f_e t. \lambda g_{et} t. \lambda x_e t. g(f(x)) \)
   (iv) \( \lambda f_{ee} t. \lambda g_{ee} t. \lambda x_e t. g(\lambda y_e t. f(x)) \)

b. Give \( \lambda \)-terms for the following English descriptions:
   (i) the function sending every function \( f_{ee} \) to the function sending every entity \( x \) to the result of applying \( f \) to \( f(x) \);
   (ii) the function sending every function of type \( (ee)e \) to its value on the identity function of type \( ee \);
   (iii) the function sending every function \( R \) of type \( e(et) \) to its inverse, i.e. the function \( R^{-1} \) of type \( e(et) \) that satisfies for every two entities \( x \) and \( y \): \( (R^{-1}(x))(y) = (R(y))(x) \).

9. a. Simplify the following \( \lambda \)-terms as much as possible using function application:
   (i) \( (\lambda x_e t. \lambda f_e t. f(x))(\text{tina}_e) ) (\text{smile}_{et}) \)
   (ii) \( (\lambda f_e t. \lambda g_{et} t. \lambda x_e t. g(f(x)))(\text{smile}_{et}) ) (\text{y}_e t. y) \)
   (iii) \( (((\lambda g_{ee} t. \lambda x_e t. g(y)(x))(\text{praise}_{e(et)}))(\text{tina}_e))(\text{mary}_e) \)

b. We analyze sentences like Mary is Tina by letting the word \text{is} denote an \( e(et) \) function \( F \), which is different from the identity function of type \( (et)(et) \). Define \( F \) as a \( \lambda \)-term, basing your definition on the equality formula \( x = y \), which has the truth-value \( 1 \) iff \( x \) and \( y \) are equal. Write the \( \lambda \)-term for the structure \text{Lewis Carroll} \[ \text{is} [ \text{C. L. Dodgson} ] \], and then simplify it as much as possible.
c. We analyze the sentence \([\text{Tina [praised [her mother]]}]\) using the following denotations:
\[
\text{HER} = \lambda_{ee}.\lambda_{e(e)}.\lambda_{x_e}.g(f(x))(x) \quad \text{and} \quad \text{mother}_{ee} = \text{the function sending each entity } x \text{ to } x' \text{'s mother.}
\]
Assuming these denotations, give the \(\lambda\)-term we derive for the sentence. Then simplify the \(\lambda\)-term you got as much as possible using function application.

10. a. Give English descriptions for the \(\lambda\)-terms:
   (i) \(\lambda_y.\text{(DOUBLER)}(y)\) (see (3.25))
   (ii) \(\lambda_y.\text{(}\lambda_x.x)(y)\)

b. Write simpler descriptions for the functions you described in 10a.

c. Complete the following conclusion:
   for any function \(f_{\tau\sigma}\), the function _____ equals \(f\).
   In the \(\lambda\)-calculus, the simplification rule that this conclusion supports is known as eta-reduction.

d. Simplify the following \(\lambda\)-term as much as possible using function application and eta-reductions:
   \[
   ((\lambda_{e(e)}.\lambda_{e(e)}.\lambda_{x_e}.\lambda_{y_e}.g(f(x))(y))(\lambda_{u_e}.\lambda_{z_e}.\text{praise}(z)(u)))(\lambda_{w_1}.w)
   \]

11. Consider the equivalence between the active sentence \(\text{Mary [praised Tina]}\) and its passive form \([\text{Tina [was praised by]} \text{Mary}]\).
   In this structure, we unrealistically assume that the string \(\text{was praised by}\) is a constituent. Express the denotation of this constituent in terms of the denotation \(\text{praise}\), in a way that captures the equivalence between the active and passive sentences.

12. a. Simplify the following \(\lambda\)-term as much as possible using function application and the definition of \(\text{AND}^t\):
   \[
   (\lambda_{e(e)}.\lambda_{x_e}.\lambda_{y_e}.(\text{AND}^t(((f(x))(y))((f(y))(x))))((\text{praise}_{e(e)}))(\lambda_{u_1}.u)
   \]
   Describe in words the \(e(e)\) function that you got.

b. (i) Give the two binary structures for the sentence \(\text{Tina is not tall and thin}\).
   (ii) For each of these structures, give the semantic interpretation derived using the lexical denotations \(\text{NOT}_{e(e)}\) and \(\text{AND}^t\).
   (iii) Simplify the resulting \(\lambda\)-terms as much as possible using function application and the definitions of \(\text{NOT}_{e(e)}\) and \(\text{AND}^t\).

c. (i) Give binary structures for the sentences in (3.53) and (3.54).
   (ii) For each of these structures, give the semantic interpretation derived using the lexical denotations \(\text{AND}^t\) and \(\text{HERSELF}\).
(iii) Simplify the resulting $\lambda$-terms as much as possible using function application and the definitions of $\text{AND}^{et}$ and $\text{HERSELF}$.

d. For every function $f$ of type $et$, we use the notation $f^*$ to denote the set of entities $\{x \in D_e : f(x) = 1\}$ that $f$ characterizes. For instance, for the function $\chi_S$ in (3.2) we denote: $\chi_S^* = S = \{a, c\}$. Show that for all functions $h_1, h_2$ of type $et$, the following holds:

$$(\text{AND}^{et}(h_2)(h_1))^* = h_1^* \cap h_2^*.$$ 

In words: the function $\text{AND}^{et}(h_2)(h_1)$ characterizes the intersection of the sets that are characterized by $h_1$ and $h_2$.

e. Assume that the sentence $\text{Tina is not tall}$ has the structure $\text{not} [\text{Tina is tall}]$. What should the type and denotation of $\text{not}$ be under this analysis? How would you account for the ambiguity of $\text{Tina is not tall and thin}$? Explain all your structural assumptions.

13. a. Account for the entailment (3.70a) with the adverb $\text{charmingly}$: describe the restriction on the adverb’s denotation by completing the following sentence:

the function $\text{charmingly}$ of type ____ maps any set $A$ characterized by ____ to ____.

b. Account for the same entailment by postulating a $\lambda$-term for $\text{charmingly}$ in terms of an arbitrary function $\text{charmingly}^{arb}$. Simplify the $\lambda$-terms for the two sentences in (3.70a), and explain why the $\leq$ relation must hold between them in every model.

c. (i) Repeat your analysis in 13a, but now for the entailment (3.70b) and the preposition $\text{with}$. Complete the following sentence:

the function $\text{with}$ of type ____ maps any entity (e.g. for $\text{John}$), to a function mapping any characteristic function $\chi_A$ (e.g. for $\text{ran}$) to ____.

(ii) Repeat your analysis in b, but now for (3.70b). Postulate a $\lambda$-term for $\text{with}$ in terms of an arbitrary function $\text{with}^{arb}$. Simplify the $\lambda$-terms you get for the sentences in (3.70b).

d. Account for the equivalence (3.71a) by defining the denotation $\text{from}$ (of which type?) on the basis of an arbitrary function $\text{from}^{arb}$ of type $e(et)$. Show that after simplifications, the truth-values you get for the two sentences in (3.71a) are the same.
e. Account for the entailments (i) *Tina is very tall* ⇒ *Tina is tall*, and (ii) *Tina is a [[very tall] student]* ⇒ *Tina is a tall student*: postulate proper restrictions on the denotation of the word *very* in (i) and (ii), of types *(et)(et)* and *(et)(et)((et)(et))* respectively.

f. Account for the equivalence in (3.71b) by postulating a proper type and a proper restriction on the denotation of the word *who*. Give it a λ-term. Show that after simplifications, the truth-values you get for the two sentences in (3.71b) are the same.

**SOLUTIONS TO SELECTED EXERCISES**

1. a. \([u\mapsto 1, v\mapsto 0, w\mapsto 1, m\mapsto 0]; [u\mapsto 0, v\mapsto 0, w\mapsto 0, m\mapsto 0]; [u\mapsto 0, v\mapsto 1, w\mapsto 0, m\mapsto 1] \). b. \([u, v, w, m]\), i.e. the whole domain \(D_e\).

2. a. (i) *(ee)(et)*, (ii) *(et)e*, (iii) *(et)(te)*, (iv) *(ee)(et)*, (v) *(et)(ee)*, (vi) *(e(et))(te)*, (vii) *(tt)(te))(e(ee))*.

3. a. \(D_{et} = \{[0\mapsto 0, 1\mapsto 0], [0\mapsto 0, 1\mapsto 1], [0\mapsto 1, 1\mapsto 0], [0\mapsto 1, 1\mapsto 1]\}\).

b. \(D_{et} =\{
[u\mapsto 0, v\mapsto 0, w\mapsto 0, m\mapsto 0], [u\mapsto 0, v\mapsto 0, w\mapsto 0, m\mapsto 1],
[u\mapsto 0, v\mapsto 0, w\mapsto 1, m\mapsto 0], [u\mapsto 0, v\mapsto 0, w\mapsto 1, m\mapsto 1],
[u\mapsto 0, v\mapsto 1, w\mapsto 0, m\mapsto 0], [u\mapsto 0, v\mapsto 1, w\mapsto 0, m\mapsto 1],
[u\mapsto 0, v\mapsto 1, w\mapsto 1, m\mapsto 0], [u\mapsto 0, v\mapsto 1, w\mapsto 1, m\mapsto 1],
[u\mapsto 1, v\mapsto 0, w\mapsto 0, m\mapsto 0], [u\mapsto 1, v\mapsto 0, w\mapsto 0, m\mapsto 1],
[u\mapsto 1, v\mapsto 0, w\mapsto 1, m\mapsto 0], [u\mapsto 1, v\mapsto 0, w\mapsto 1, m\mapsto 1],
[u\mapsto 1, v\mapsto 1, w\mapsto 0, m\mapsto 0], [u\mapsto 1, v\mapsto 1, w\mapsto 0, m\mapsto 1],
[u\mapsto 1, v\mapsto 1, w\mapsto 1, m\mapsto 0], [u\mapsto 1, v\mapsto 1, w\mapsto 1, m\mapsto 1]\}\). The solution for \(D_{et}\) is along similar lines.

c. For \(D_e = \{l, n\}\), there are four functions in \(D_{ee} : [l\mapsto l, n\mapsto l], [l\mapsto l, n\mapsto l]\) and \([l\mapsto n, n\mapsto l]\) and \([l\mapsto n, n\mapsto n]\). If we substitute these four functions for \(u, v, w\) and \(m\) in the answer to 3b, we get the sixteen functions in \(D_{(ee)}\).

4. *et, t, –, –, et, –, tt, –, –, ee(et)*.

5. a. *praise* = \(t\mapsto [t\mapsto 0 j\mapsto 0 m\mapsto 0], j\mapsto [t\mapsto 1 j\mapsto 0 m\mapsto 1], m\mapsto [t\mapsto 1 j\mapsto 1 m\mapsto 1]\)
8. a. (i) the function sending every function \(f\) of type (et)t to the function sending every truth-value \(y\) to the result of applying \(f\) to the constant et function sending every entity to \(y\).
   (ii) the function \(I\) sending every entity \(x\) to the function sending every et function \(f\) to the truth-value result of applying \(f\) to \(x\) (\(I\) sends every \(x\) to the characteristic function of the set of functions characterizing subsets of \(D_x\) containing \(x\)).
   (iii) the function \(C\) sending every et function \(f\) to the function sending every tt function \(g\) to the function from entities \(x\) to the result of applying \(g\) to \(f(x)\) (\(C\) returns the function composition of \(g_{tt}\) on \(f_{et}\)).
   (iv) the function sending every ee function \(f\) to the function sending every (ee)t function \(g\) to the function from entities \(x\) to the result of applying \(g\) to the constant function sending every entity to \(f(x)\).

b. (i) \(\lambda f_{ee}.\lambda x_{ee}. f(f(x))\);
   (ii) \(\lambda f_{(ee)e}. f(\lambda x_{(ee)}. x)\);
   (iii) \(\lambda R_{(e)(et)}. \lambda x_{(et)}. \lambda y_{(et)}. (R(y))(x)\).

9. a. (i) \texttt{smile(tina)}; (ii) \(\lambda x_{c}. \texttt{smile}(x)\); (iii) \(\texttt{praise(mary)})(\texttt{tina})\)
b. \(F = \lambda y_{c}. \lambda x_{c}. x = y; (F(\texttt{cld})) (\texttt{l} c) = (\texttt{l} c = \texttt{cld})\).
c. \(((\text{HER(mother)})(\text{praise}_{\text{et}}))(\text{tina}_e) = \\
(\text{praise(mother(tina))})(\text{tina})\)

10. a. (i) the function sending every number \(y\) to \(\text{DOUBLER}(y)\)
    (ii) the function sending every entity \(y\) to the result that the function sending every entity \(x\) to \(x\) returns for \(y\)

b. (i) \(\text{DOUBLER}\)
    (ii) the function sending every entity \(y\) to \(y\), a.k.a. the function sending every entity \(x\) to \(x\)

c. the function \(\lambda x. f(x) = f\)

d. \(\text{praise}\).

11. \(\text{[was praised by]} = \lambda x_e. \lambda y_e. \text{praise}(y)(x)\)

12. a. \(\lambda x_e. \lambda y_e. (\text{praise}(y))(x) \land (\text{praise}(x))(y)\) – the function sending \(x\) to the function sending \(y\) to 1 iff \(x\) and \(y\) praised each other.

b. (i) Tina \([\text{is [not [tall [and thin]]]}]\); Tina \([\text{is [[not tall] [and thin]]]}\]
    (ii) \(\text{IS(NOT(AND_et(thin)))(tall)})(\text{tina})\);
    (iii) \(\sim (\text{tall(tina)} \land \text{thin(tina)}); (\sim(\text{tall(tina)})) \land \text{thin(tina)}\).

d. Suppose \(h_1^* = A_1, \ h_2^* = A_2\). By def. of \(\text{AND}_{et}\):
    \((\text{AND}_{et}(h_2)(h_1))^* = (\lambda x_e. h_1(x) \land h_2(x))^*,\) i.e. the set \(A = \{x \in E : (h_1(x) \land h_2(x)) = 1\}\). By def. of \(\land\), for every \(y \in E: y \in A\) iff \(h_1(y) = 1\) and \(h_2(y) = 1\), i.e. \(y \in A_1 \land A_2\), i.e. \(y \in A_1 \cap A_2\).

c. \text{with}\) of type \(e((\text{et})(\text{et}))\) maps any entity \(x\) (e.g. for \(\text{John}\)), to a function mapping any characteristic function \(\chi_A\) (e.g. for \(\text{ran}\)) to a function characterizing subsets of \(A\);
    \text{with} = \lambda x_e. \lambda f_{et}. \lambda y_e. ((\text{with}_{\text{arb}}_{e((\text{et})(\text{et}))})(x))(f)(y) \land f(y).

d. \text{from}_{e((\text{et})(\text{et}))} = \lambda x_e. \lambda f_{et}. \lambda y_e. ((\text{from}_{\text{arb}}_{e((\text{et})(\text{et}))})(x))(y) \land f(y).

f. \text{WHO}\) is assigned type \((\text{et})(\text{et})\), which leads to the term:
    \((\text{IS(A(WHO(HERSELF(praise)(pianist)))))(tina)}\). With the assumption \(\text{WHO=AND}_{et}\), this term is simplified to:
    \(\text{pianist(tina)} \land \text{praise(tina)(tina)}\). We get the same term when simplifying the term for the sentence \(\text{Tina is a pianist and Tina praised herself}\.\)


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