

## 19.2.1 Example problem: Restaurant waiting

We will describe a sample supervised learning problem in detail: the problem of deciding whether to wait for a table at a restaurant. This problem will be used throughout the chapter to demonstrate different model classes. For this problem the output,  $y$ , is a Boolean variable that we will call *WillWait*; it is true for examples where we do wait for a table. The input,  $x$ , is a vector of ten attribute values, each of which has discrete values:

1. **ALTERNATE**: whether there is a suitable alternative restaurant nearby.
2. **BAR**: whether the restaurant has a comfortable bar area to wait in.
3. **FRI/SAT**: true on Fridays and Saturdays.
4. **HUNGRY**: whether we are hungry right now.
5. **PATRONS**: how many people are in the restaurant (values are *None*, *Some*, and *Full*).
6. **PRICE**: the restaurant's price range (\$, \$\$, \$\$\$).
7. **RAINING**: whether it is raining outside.
8. **RESERVATION**: whether we made a reservation.
9. **TYPE**: the kind of restaurant (French, Italian, Thai, or burger).
10. **WAITESTIMATE**: host's wait estimate:  
0 – 10, 10 – 30, 30 – 60, or >60 minutes.

A set of 12 examples, taken from the experience of one of us (SR), is shown in [Figure 19.2](#). Note how skimpy these data are: there are  $2^6 \times 3^2 \times 4^2 = 9,216$  possible combinations of values for the input attributes, but we are given the correct output for only 12 of them; each of the other 9,204 could be either true or false; we don't know. This is the essence of induction: we need to make our best guess at these missing 9,204 output values, given only the evidence of the 12 examples.

Example	Input Attributes										Output
	<i>Alt</i>	<i>Bar</i>	<i>Fri</i>	<i>Hun</i>	<i>Pat</i>	<i>Price</i>	<i>Rain</i>	<i>Res</i>	<i>Type</i>	<i>Est</i>	<i>WillWait</i>
$x_1$	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>Yes</i>	<i>Some</i>	\$\$\$	<i>No</i>	<i>Yes</i>	<i>French</i>	0–10	$y_1 = \text{Yes}$
$x_2$	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>Yes</i>	<i>Full</i>	\$	<i>No</i>	<i>No</i>	<i>Thai</i>	30–60	$y_2 = \text{No}$
$x_3$	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>Some</i>	\$	<i>No</i>	<i>No</i>	<i>Burger</i>	0–10	$y_3 = \text{Yes}$
$x_4$	<i>Yes</i>	<i>No</i>	<i>Yes</i>	<i>Yes</i>	<i>Full</i>	\$	<i>Yes</i>	<i>No</i>	<i>Thai</i>	10–30	$y_4 = \text{Yes}$
$x_5$	<i>Yes</i>	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>Full</i>	\$\$\$	<i>No</i>	<i>Yes</i>	<i>French</i>	>60	$y_5 = \text{No}$
$x_6$	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>Yes</i>	<i>Some</i>	\$\$	<i>Yes</i>	<i>Yes</i>	<i>Italian</i>	0–10	$y_6 = \text{Yes}$
$x_7$	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>None</i>	\$	<i>Yes</i>	<i>No</i>	<i>Burger</i>	0–10	$y_7 = \text{No}$
$x_8$	<i>No</i>	<i>No</i>	<i>No</i>	<i>Yes</i>	<i>Some</i>	\$\$	<i>Yes</i>	<i>Yes</i>	<i>Thai</i>	0–10	$y_8 = \text{Yes}$
$x_9$	<i>No</i>	<i>Yes</i>	<i>Yes</i>	<i>No</i>	<i>Full</i>	\$	<i>Yes</i>	<i>No</i>	<i>Burger</i>	>60	$y_9 = \text{No}$
$x_{10}$	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Full</i>	\$\$\$	<i>No</i>	<i>Yes</i>	<i>Italian</i>	10–30	$y_{10} = \text{No}$
$x_{11}$	<i>No</i>	<i>No</i>	<i>No</i>	<i>No</i>	<i>None</i>	\$	<i>No</i>	<i>No</i>	<i>Thai</i>	0–10	$y_{11} = \text{No}$
$x_{12}$	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Full</i>	\$	<i>No</i>	<i>No</i>	<i>Burger</i>	30–60	$y_{12} = \text{Yes}$

**Figure 19.2** Examples for the restaurant domain.