In our discussion of sequential decision problems, we have assumed that the transition and reward models are known. In many problems, however, these models are not known exactly, and the agent must learn to act through experience. By observing the outcomes of its actions in the form of state transitions and rewards, the agent is to choose actions that maximize its long-term accumulation of rewards. Solving such problems in which there is model uncertainty is the subject of the field of reinforcement learning and the focus of this part of the book. We will discuss several challenges in addressing model uncertainty. First, the agent must carefully balance exploration of the environment with the exploitation of that knowledge gained through experience. Second, rewards may be received long after the important decisions have been made, so credit for later rewards must be assigned to earlier decisions. Third, the agent must generalize from limited experience. We will review the theory and some of the key algorithms for addressing these challenges.