

25 ROBOTICS

In which agents are endowed with physical effectors with which to do mischief.

25.1 INTRODUCTION

ROBOT

EFFECTOR

SENSOR

MANIPULATOR

MOBILE ROBOT

UGV

PLANETARY ROVER

UAV

Robots are physical agents that perform tasks by manipulating the physical world. To do so, they are equipped with **effectors** such as legs, wheels, joints, and grippers. Effectors have a single purpose: to assert physical forces on the environment.¹ Robots are also equipped with **sensors**, which allow them to perceive their environment. Present day robotics employs a diverse set of sensors, including cameras and lasers to measure the environment, and gyroscopes and accelerometers to measure the robot's own motion.

Most of today's robots fall into one of three primary categories. **Manipulators**, or robot arms (Figure 25.1(a)), are physically anchored to their workplace, for example in a factory assembly line or on the International Space Station. Manipulator motion usually involves a chain of controllable joints, enabling such robots to place their effectors in any position within the workplace. Manipulators are by far the most common type of industrial robots, with approximately one million units installed worldwide. Some mobile manipulators are used in hospitals to assist surgeons. Few car manufacturers could survive without robotic manipulators, and some manipulators have even been used to generate original artwork.

The second category is the **mobile robot**. Mobile robots move about their environment using wheels, legs, or similar mechanisms. They have been put to use delivering food in hospitals, moving containers at loading docks, and similar tasks. **Unmanned ground vehicles**, or UGVs, drive autonomously on streets, highways, and off-road. The **planetary rover** shown in Figure 25.2(b) explored Mars for a period of 3 months in 1997. Subsequent NASA robots include the twin Mars Exploration Rovers (one is depicted on the cover of this book), which landed in 2003 and were still operating six years later. Other types of mobile robots include **unmanned air vehicles** (UAVs), commonly used for surveillance, crop-spraying, and

¹ In Chapter 2 we talked about **actuators**, not effectors. Here we distinguish the effector (the physical device) from the actuator (the control line that communicates a command to the effector).

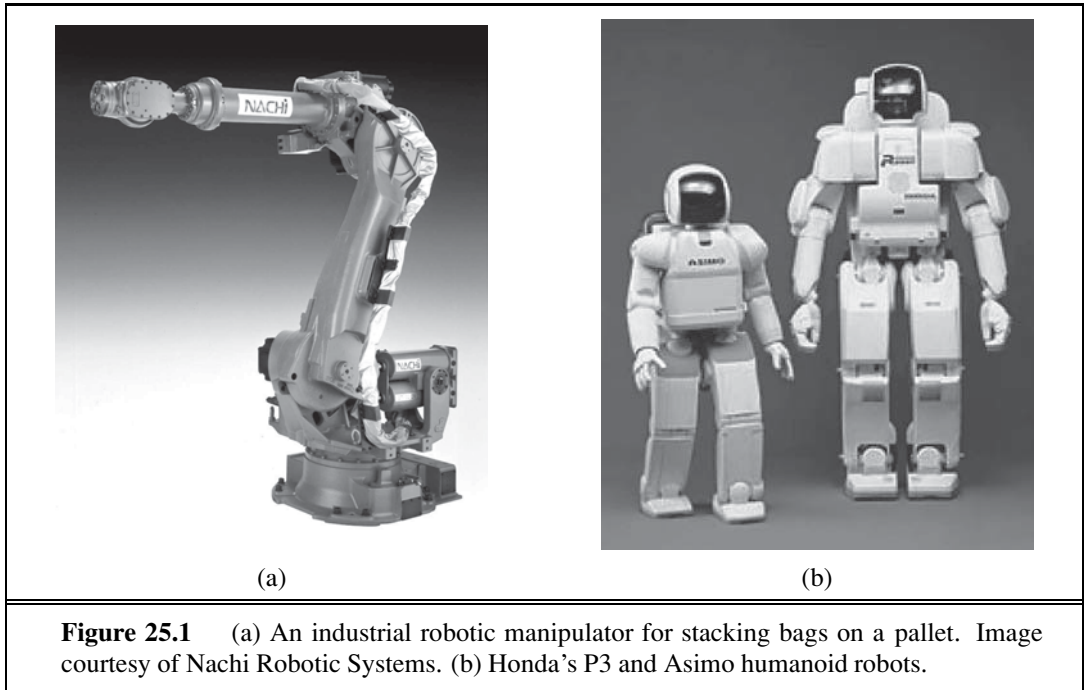


Figure 25.1 (a) An industrial robotic manipulator for stacking bags on a pallet. Image courtesy of Nachi Robotic Systems. (b) Honda's P3 and Asimo humanoid robots.

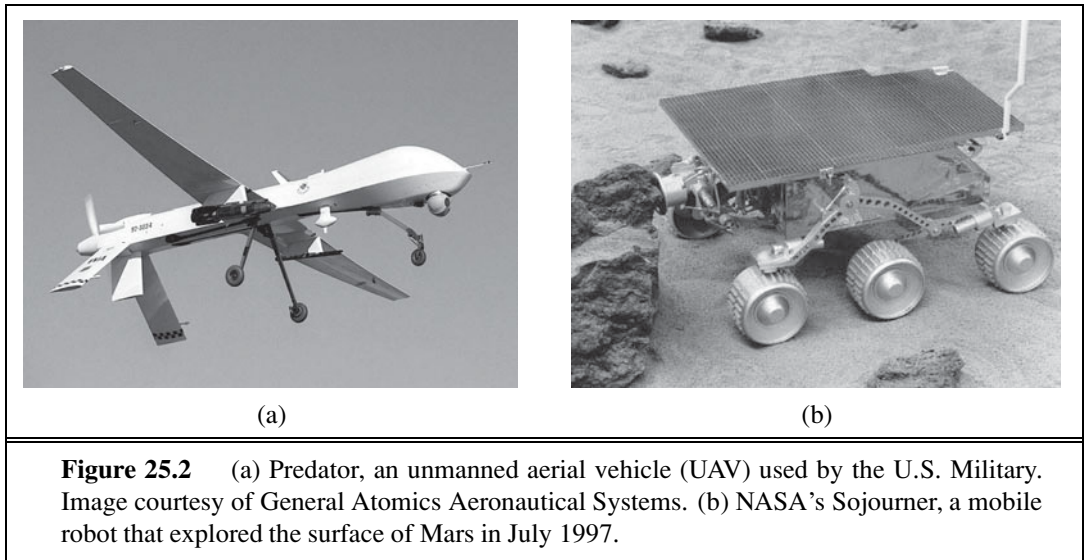


Figure 25.2 (a) Predator, an unmanned aerial vehicle (UAV) used by the U.S. Military. Image courtesy of General Atomics Aeronautical Systems. (b) NASA's Sojourner, a mobile robot that explored the surface of Mars in July 1997.

military operations. Figure 25.2(a) shows a UAV commonly used by the U.S. military. **Autonomous underwater vehicles (AUVs)** are used in deep sea exploration. Mobile robots deliver packages in the workplace and vacuum the floors at home.

The third type of robot combines mobility with manipulation, and is often called a **mobile manipulator**. **Humanoid robots** mimic the human torso. Figure 25.1(b) shows two early humanoid robots, both manufactured by Honda Corp. in Japan. Mobile manipulators

AUV

MOBILE
MANIPULATOR
HUMANOID ROBOT

can apply their effectors further afield than anchored manipulators can, but their task is made harder because they don't have the rigidity that the anchor provides.

The field of robotics also includes prosthetic devices (artificial limbs, ears, and eyes for humans), intelligent environments (such as an entire house that is equipped with sensors and effectors), and multibody systems, wherein robotic action is achieved through swarms of small cooperating robots.

Real robots must cope with environments that are partially observable, stochastic, dynamic, and continuous. Many robot environments are sequential and multiagent as well. Partial observability and stochasticity are the result of dealing with a large, complex world. Robot cameras cannot see around corners, and motion commands are subject to uncertainty due to gears slipping, friction, etc. Also, the real world stubbornly refuses to operate faster than real time. In a simulated environment, it is possible to use simple algorithms (such as the Q-learning algorithm described in Chapter 21) to learn in a few CPU hours from millions of trials. In a real environment, it might take years to run these trials. Furthermore, real crashes really hurt, unlike simulated ones. Practical robotic systems need to embody prior knowledge about the robot, its physical environment, and the tasks that the robot will perform so that the robot can learn quickly and perform safely.

Robotics brings together many of the concepts we have seen earlier in the book, including probabilistic state estimation, perception, planning, unsupervised learning, and reinforcement learning. For some of these concepts robotics serves as a challenging example application. For other concepts this chapter breaks new ground in introducing the continuous version of techniques that we previously saw only in the discrete case.

25.2 ROBOT HARDWARE

So far in this book, we have taken the agent architecture—sensors, effectors, and processors—as given, and we have concentrated on the agent program. The success of real robots depends at least as much on the design of sensors and effectors that are appropriate for the task.

25.2.1 Sensors

PASSIVE SENSOR

Sensors are the perceptual interface between robot and environment. **Passive sensors**, such as cameras, are true observers of the environment: they capture signals that are generated by other sources in the environment. **Active sensors**, such as sonar, send energy into the environment. They rely on the fact that this energy is reflected back to the sensor. Active sensors tend to provide more information than passive sensors, but at the expense of increased power consumption and with a danger of interference when multiple active sensors are used at the same time. Whether active or passive, sensors can be divided into three types, depending on whether they sense the environment, the robot's location, or the robot's internal configuration.

ACTIVE SENSOR

RANGE FINDER

SONAR SENSORS

Range finders are sensors that measure the distance to nearby objects. In the early days of robotics, robots were commonly equipped with **sonar sensors**. Sonar sensors emit directional sound waves, which are reflected by objects, with some of the sound making it

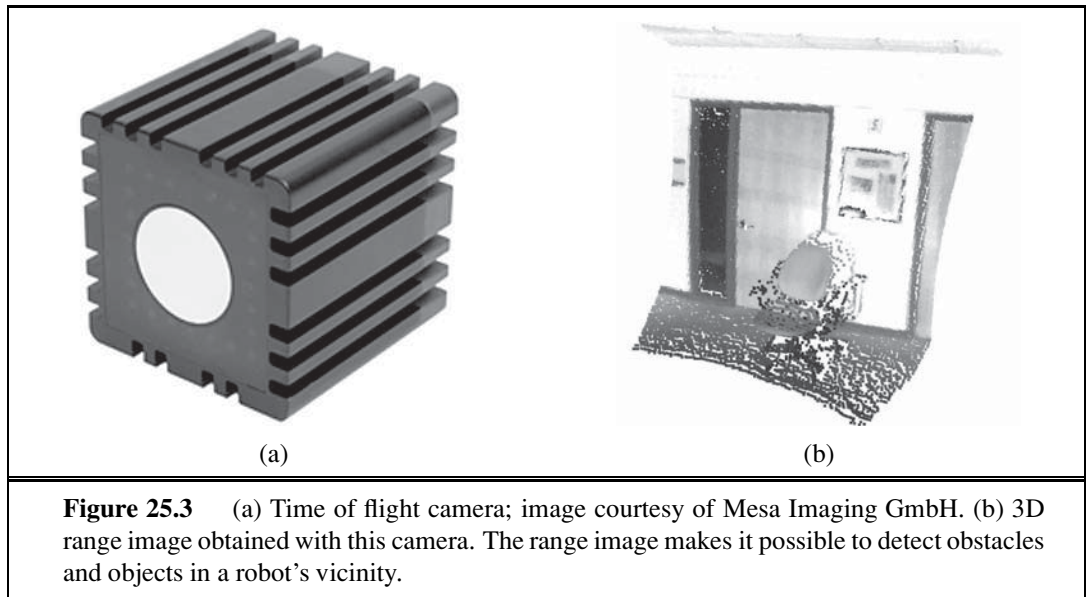


Figure 25.3 (a) Time of flight camera; image courtesy of Mesa Imaging GmbH. (b) 3D range image obtained with this camera. The range image makes it possible to detect obstacles and objects in a robot's vicinity.

back into the sensor. The time and intensity of the returning signal indicates the distance to nearby objects. Sonar is the technology of choice for autonomous underwater vehicles. **Stereo vision** (see Section 24.4.2) relies on multiple cameras to image the environment from slightly different viewpoints, analyzing the resulting parallax in these images to compute the range of surrounding objects. For mobile ground robots, sonar and stereo vision are now rarely used, because they are not reliably accurate.

Most ground robots are now equipped with optical range finders. Just like sonar sensors, optical range sensors emit active signals (light) and measure the time until a reflection of this signal arrives back at the sensor. Figure 25.3(a) shows a **time of flight camera**. This camera acquires range images like the one shown in Figure 25.3(b) at up to 60 frames per second. Other range sensors use laser beams and special 1-pixel cameras that can be directed using complex arrangements of mirrors or rotating elements. These sensors are called **scanning lidars** (short for *light detection and ranging*). Scanning lidars tend to provide longer ranges than time of flight cameras, and tend to perform better in bright daylight.

Other common range sensors include radar, which is often the sensor of choice for UAVs. Radar sensors can measure distances of multiple kilometers. On the other extreme end of range sensing are **tactile sensors** such as whiskers, bump panels, and touch-sensitive skin. These sensors measure range based on physical contact, and can be deployed only for sensing objects very close to the robot.

A second important class of sensors is **location sensors**. Most location sensors use range sensing as a primary component to determine location. Outdoors, the **Global Positioning System** (GPS) is the most common solution to the localization problem. GPS measures the distance to satellites that emit pulsed signals. At present, there are 31 satellites in orbit, transmitting signals on multiple frequencies. GPS receivers can recover the distance to these satellites by analyzing phase shifts. By triangulating signals from multiple satellites, GPS

STEREO VISION

TIME OF FLIGHT
CAMERA

SCANNING LIDARS

TACTILE SENSORS

LOCATION SENSORS

GLOBAL
POSITIONING
SYSTEM

DIFFERENTIAL GPS

receivers can determine their absolute location on Earth to within a few meters. **Differential GPS** involves a second ground receiver with known location, providing millimeter accuracy under ideal conditions. Unfortunately, GPS does not work indoors or underwater. Indoors, localization is often achieved by attaching beacons in the environment at known locations. Many indoor environments are full of wireless base stations, which can help robots localize through the analysis of the wireless signal. Underwater, active sonar beacons can provide a sense of location, using sound to inform AUVs of their relative distances to those beacons.

PROPRIOCEPTIVE SENSOR

The third important class is **proprioceptive sensors**, which inform the robot of its own motion. To measure the exact configuration of a robotic joint, motors are often equipped with **shaft decoders** that count the revolution of motors in small increments. On robot arms, shaft decoders can provide accurate information over any period of time. On mobile robots, shaft decoders that report wheel revolutions can be used for **odometry**—the measurement of distance traveled. Unfortunately, wheels tend to drift and slip, so odometry is accurate only over short distances. External forces, such as the current for AUVs and the wind for UAVs, increase positional uncertainty. **Inertial sensors**, such as gyroscopes, rely on the resistance of mass to the change of velocity. They can help reduce uncertainty.

SHAFT DECODER

ODOMETRY

INERTIAL SENSOR

FORCE SENSOR

TORQUE SENSOR

Other important aspects of robot state are measured by **force sensors** and **torque sensors**. These are indispensable when robots handle fragile objects or objects whose exact shape and location is unknown. Imagine a one-ton robotic manipulator screwing in a light bulb. It would be all too easy to apply too much force and break the bulb. Force sensors allow the robot to sense how hard it is gripping the bulb, and torque sensors allow it to sense how hard it is turning. Good sensors can measure forces in all three translational and three rotational directions. They do this at a frequency of several hundred times a second, so that a robot can quickly detect unexpected forces and correct its actions before it breaks a light bulb.

25.2.2 Effectors

DEGREE OF FREEDOM

Effectors are the means by which robots move and change the shape of their bodies. To understand the design of effectors, it will help to talk about motion and shape in the abstract, using the concept of a **degree of freedom** (DOF). We count one degree of freedom for each independent direction in which a robot, or one of its effectors, can move. For example, a rigid mobile robot such as an AUV has six degrees of freedom, three for its (x, y, z) location in space and three for its angular orientation, known as *yaw*, *roll*, and *pitch*. These six degrees define the **kinematic state**² or **pose** of the robot. The **dynamic state** of a robot includes these six plus an additional six dimensions for the rate of change of each kinematic dimension, that is, their velocities.

KINEMATIC STATE

POSE

DYNAMIC STATE

For nonrigid bodies, there are additional degrees of freedom within the robot itself. For example, the elbow of a human arm possesses two degree of freedom. It can flex the upper arm towards or away, and can rotate right or left. The wrist has three degrees of freedom. It can move up and down, side to side, and can also rotate. Robot joints also have one, two, or three degrees of freedom each. Six degrees of freedom are required to place an object, such as a hand, at a particular point in a particular orientation. The arm in Figure 25.4(a)

² “Kinematic” is from the Greek word for *motion*, as is “cinema.”

REVOLUTE JOINT
PRISMATIC JOINT

has exactly six degrees of freedom, created by five **revolute joints** that generate rotational motion and one **prismatic joint** that generates sliding motion. You can verify that the human arm as a whole has more than six degrees of freedom by a simple experiment: put your hand on the table and notice that you still have the freedom to rotate your elbow without changing the configuration of your hand. Manipulators that have extra degrees of freedom are easier to control than robots with only the minimum number of DOFs. Many industrial manipulators therefore have seven DOFs, not six.

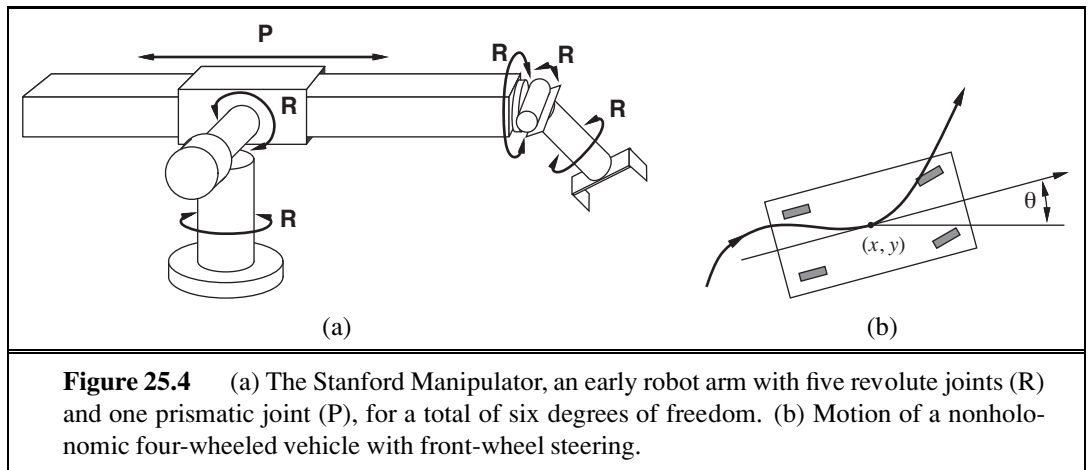


Figure 25.4 (a) The Stanford Manipulator, an early robot arm with five revolute joints (R) and one prismatic joint (P), for a total of six degrees of freedom. (b) Motion of a nonholonomic four-wheeled vehicle with front-wheel steering.

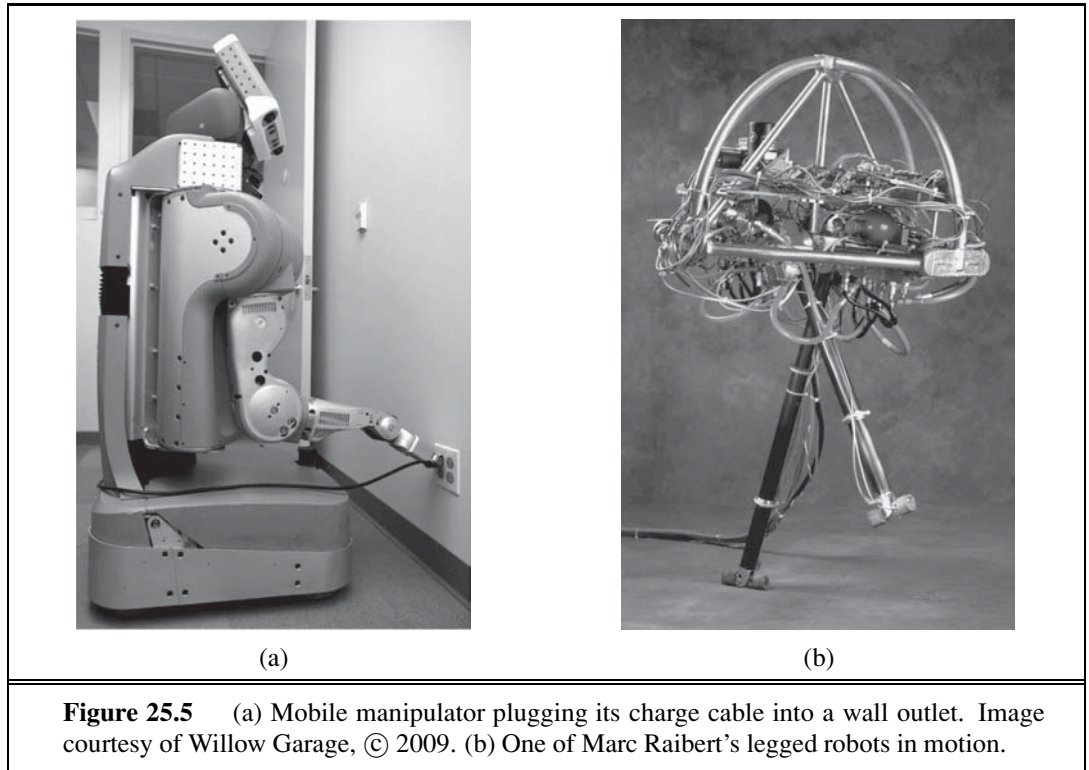
EFFECTIVE DOF
CONTROLLABLE DOF
NONHOLONOMIC

For mobile robots, the DOFs are not necessarily the same as the number of actuated elements. Consider, for example, your average car: it can move forward or backward, and it can turn, giving it two DOFs. In contrast, a car's kinematic configuration is three-dimensional: on an open flat surface, one can easily maneuver a car to any (x, y) point, in any orientation. (See Figure 25.4(b).) Thus, the car has three **effective degrees of freedom** but two **controllable degrees of freedom**. We say a robot is **nonholonomic** if it has more effective DOFs than controllable DOFs and **holonomic** if the two numbers are the same. Holonomic robots are easier to control—it would be much easier to park a car that could move sideways as well as forward and backward—but holonomic robots are also mechanically more complex. Most robot arms are holonomic, and most mobile robots are nonholonomic.

DIFFERENTIAL DRIVE
SYNCHRO DRIVE

Mobile robots have a range of mechanisms for locomotion, including wheels, tracks, and legs. **Differential drive** robots possess two independently actuated wheels (or tracks), one on each side, as on a military tank. If both wheels move at the same velocity, the robot moves on a straight line. If they move in opposite directions, the robot turns on the spot. An alternative is the **synchro drive**, in which each wheel can move and turn around its own axis. To avoid chaos, the wheels are tightly coordinated. When moving straight, for example, all wheels point in the same direction and move at the same speed. Both differential and synchro drives are nonholonomic. Some more expensive robots use holonomic drives, which have three or more wheels that can be oriented and moved independently.

Some mobile robots possess arms. Figure 25.5(a) displays a two-armed robot. This robot's arms use springs to compensate for gravity, and they provide minimal resistance to



external forces. Such a design minimizes the physical danger to people who might stumble into such a robot. This is a key consideration in deploying robots in domestic environments.

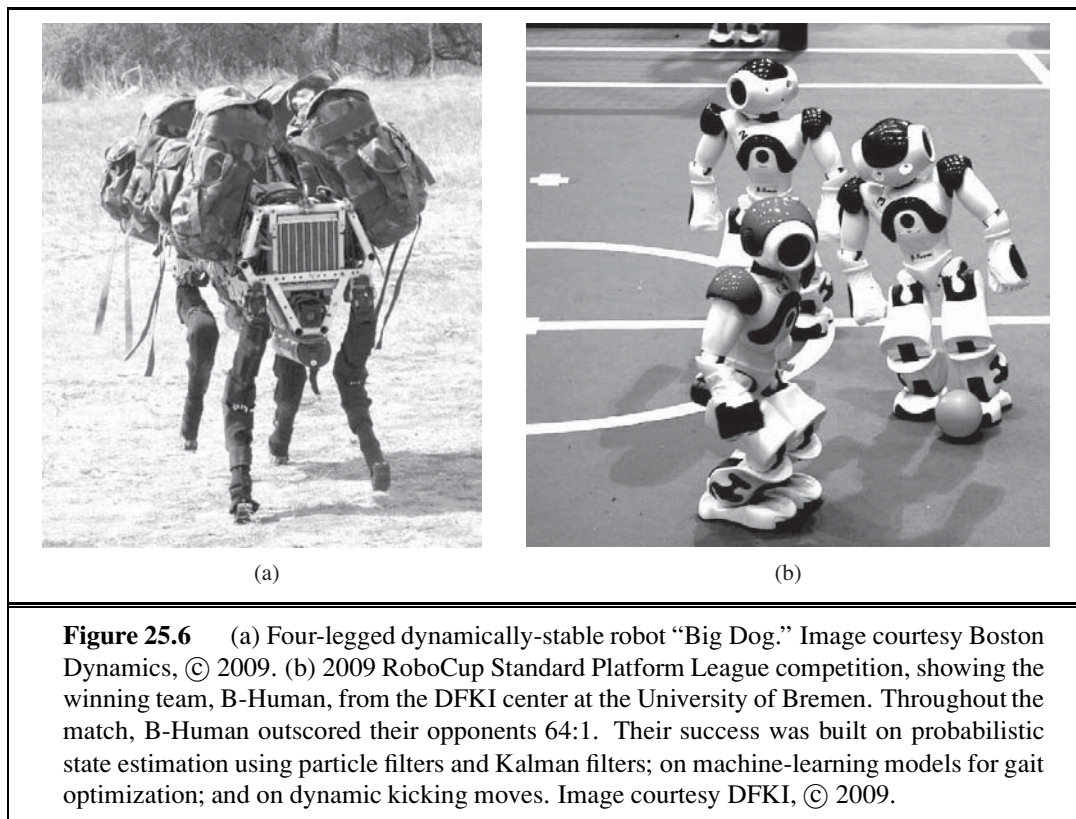
Legs, unlike wheels, can handle rough terrain. However, legs are notoriously slow on flat surfaces, and they are mechanically difficult to build. Robotics researchers have tried designs ranging from one leg up to dozens of legs. Legged robots have been made to walk, run, and even hop—as we see with the legged robot in Figure 25.5(b). This robot is **dynamically stable**, meaning that it can remain upright while hopping around. A robot that can remain upright without moving its legs is called **statically stable**. A robot is statically stable if its center of gravity is above the polygon spanned by its legs. The quadruped (four-legged) robot shown in Figure 25.6(a) may appear statically stable. However, it walks by lifting multiple legs at the same time, which renders it dynamically stable. The robot can walk on snow and ice, and it will not fall over even if you kick it (as demonstrated in videos available online). Two-legged robots such as those in Figure 25.6(b) are dynamically stable.

Other methods of movement are possible: air vehicles use propellers or turbines; underwater vehicles use propellers or thrusters, similar to those used on submarines. Robotic blimps rely on thermal effects to keep themselves aloft.

Sensors and effectors alone do not make a robot. A complete robot also needs a source of power to drive its effectors. The **electric motor** is the most popular mechanism for both manipulator actuation and locomotion, but **pneumatic actuation** using compressed gas and **hydraulic actuation** using pressurized fluids also have their application niches.

DYNAMICALLY
STABLE
STATICALLY STABLE

ELECTRIC MOTOR
PNEUMATIC
ACTUATION
HYDRAULIC
ACTUATION



25.3 ROBOTIC PERCEPTION

Perception is the process by which robots map sensor measurements into internal representations of the environment. Perception is difficult because sensors are noisy, and the environment is partially observable, unpredictable, and often dynamic. In other words, robots have all the problems of **state estimation** (or **filtering**) that we discussed in Section 15.2. As a rule of thumb, good internal representations for robots have three properties: they contain enough information for the robot to make good decisions, they are structured so that they can be updated efficiently, and they are natural in the sense that internal variables correspond to natural state variables in the physical world.

In Chapter 15, we saw that Kalman filters, HMMs, and dynamic Bayes nets can represent the transition and sensor models of a partially observable environment, and we described both exact and approximate algorithms for updating the **belief state**—the posterior probability distribution over the environment state variables. Several dynamic Bayes net models for this process were shown in Chapter 15. For robotics problems, we include the robot’s own past actions as observed variables in the model. Figure 25.7 shows the notation used in this chapter: \mathbf{X}_t is the state of the environment (including the robot) at time t , \mathbf{Z}_t is the observation received at time t , and A_t is the action taken after the observation is received.