The following is the output we get when we use the preceding command:

How it works...

We create the process and then monitor its lifetime by the is_alive() method. Then, we finish it with a call to terminate():

p.terminate()

Finally, we verify the status code when the process is finished, and read the attribute of the ExitCode process. The possible values of ExitCode are, as follows:

- == 0: This means that no error was produced
- > 0: This means that the process had an error and exited that code
- < 0: This means that the process was killed with a signal of -1 * ExitCode</p>

For our example, the output value of the ExitCode code is equal to -15. The negative value -15 indicates that the child was terminated by an interrupt signal identified by the number 15.

How to use a process in a subclass

To implement a custom subclass and process, we must:

- ▶ Define a new subclass of the Process class
- Override the _init__(self [,args]) method to add additional arguments
- Override the run(self [, args]) method to implement what Process should when it is started

Once you have created the new Process subclass, you can create an instance of it and then start by invoking the start() method, which will in turn call the run() method.

We will rewrite the first example in this manner:

```
#Using a process in a subclass Chapter 3: Process Based #Parallelism
import multiprocessing

class MyProcess(multiprocessing.Process):
    def run(self):
        print ('called run method in process: %s' %self.name)
        return

if __name__ == '__main__':
    jobs = []
    for i in range(5):
        p = MyProcess ()
        jobs.append(p)
        p.start()
        p.join()
```

To run the script from the Command Prompt, type the following command:

```
python subclass_process.py
```

The result of the preceding command is as follows:

C:\Python CookBook\Chapter 3 - Process Based Parallelism\Example Codes
Chapter 3>python subclass_process.py

```
called run method in process: MyProcess-1 called run method in process: MyProcess-2 called run method in process: MyProcess-3 called run method in process: MyProcess-4 called run method in process: MyProcess-5
```

How it works...

Each Process subclass could be represented by a class that extends the Process class and overrides its run() method. This method is the starting point of Process:

```
class MyProcess (multiprocessing.Process):
    def run(self):
        print ('called run method in process: %s' %self.name)
        return
```

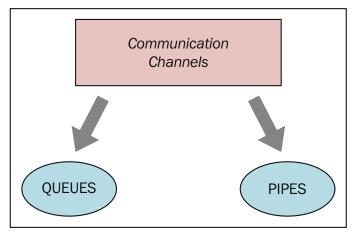
In the main program, we create several objects of the type MyProcess(). The execution of the thread begins when the start() method is called:

```
p = MyProcess()
p.start()
```

The join() command just handles the termination of processes.

How to exchange objects between processes

The development of parallel applications has the need for the exchange of data between processes. The multiprocessing library has two communication channels with which it can manage the exchange of objects: queues and pipes.



Communication channels in the multiprocessing module

Using queue to exchange objects

As explained before, it is possible for us to share data with the queue data structure.

A queue returns a process shared queue, is thread and process safe, and any serializable object (Python serializes an object using the pickable module) can be exchanged through it.

In the following example, we show you how to use a queue for a producer-consumer problem. The producer class creates the item and queues and then, the consumer class provides the facility to remove the inserted item:

```
import multiprocessing
import random
import time
class producer(multiprocessing.Process):
    def __init__(self, queue):
        multiprocessing.Process. init (self)
        self.queue = queue
    def run(self) :
        for i in range(10):
            item = random.randint(0, 256)
            self.queue.put(item)
            print ("Process Producer : item %d appended to queue %s"\
                   % (item,self.name))
            time.sleep(1)
            print ("The size of queue is %s"\
                   % self.queue.qsize())
class consumer(multiprocessing.Process):
    def __init__(self, queue):
        multiprocessing.Process.__init__(self)
        self.queue = queue
    def run(self):
        while True:
            if (self.queue.empty()):
                print("the queue is empty")
                break
            else :
                time.sleep(2)
                item = self.queue.get()
                print ('Process Consumer : item %d popped from by %s \n'\
                       % (item, self.name))
                time.sleep(1)
```

if __name__ == '__main__':

```
queue = multiprocessing.Queue()
          process_producer = producer(queue)
          process consumer = consumer(queue)
          process producer.start()
          process consumer.start()
          process producer.join()
          process_consumer.join()
This is the output that we get after the execution:
C:\Python CookBook\Chapter 3 - Process Based Parallelism\Example Codes
Chapter 3>python using_queue.py
Process Producer: item 69 appended to queue producer-1
The size of queue is 1
Process Producer: item 168 appended to queue producer-1
The size of queue is 2
Process Consumer : item 69 popped from by consumer-2
Process Producer: item 235 appended to queue producer-1
The size of queue is 2
Process Producer : item 152 appended to queue producer-1
The size of queue is 3
Process Producer: item 213 appended to queue producer-1
Process Consumer: item 168 popped from by consumer-2
The size of queue is 3
Process Producer: item 35 appended to gueue producer-1
The size of queue is 4
Process Producer : item 218 appended to queue producer-1
The size of queue is 5
Process Producer: item 175 appended to queue producer-1
Process Consumer: item 235 popped from by consumer-2
The size of queue is 5
Process Producer : item 140 appended to queue producer-1
The size of queue is 6
Process Producer: item 241 appended to queue producer-1
The size of queue is 7
```

Process Consumer : item 152 popped from by consumer-2 Process Consumer : item 213 popped from by consumer-2

```
Process Consumer: item 35 popped from by consumer-2
Process Consumer: item 218 popped from by consumer-2
Process Consumer: item 175 popped from by consumer-2
Process Consumer: item 140 popped from by consumer-2
Process Consumer: item 241 popped from by consumer-2
the queue is empty
```

How it works...

The multiprocessing class has its Queue object instantiated in the main program:

Then, we create the two processes, producer and consumer, with the Queue object as an attribute:

```
process_producer = producer(queue)
process consumer = consumer(queue)
```

The process producer is responsible for entering 10 items in the queue using its put() method:

```
for i in range(10):
    item = random.randint(0, 256)
    self.queue.put(item)
```

The process consumer has the task of removing the items from the queue (using the get method) and verifying that the queue is not empty. If this happens, the flow inside the while loop ends with a break statement:

There's more...

A queue has the JoinaleQueue subclass. It has the following two additional methods:

- ▶ task_done(): This indicates that a task is complete, for example, after the get() method is used to fetch items from the queue. So, it must be used only by queue consumers.
- join(): This blocks the processes until all the items in the queue have been achieved and processed.

Using pipes to exchange objects

The second communication channel is the pipe data structure.

A pipe does the following:

- ▶ Returns a pair of connection objects connected by a pipe
- ▶ In this, every object has send/receive methods to communicate between processes

How to do it...

Here is a simple example with pipes. We have one process pipe the gives out numbers from 0 to 9 and another process that takes the numbers and squares them:

```
import multiprocessing
def create_items(pipe):
    output_pipe, _ = pipe
    for item in range(10):
        output_pipe.send(item)
    output pipe.close()
def multiply_items(pipe_1, pipe_2):
    close, input pipe = pipe 1
    close.close()
    output_pipe, _ = pipe_2
    try:
        while True:
            item = input_pipe.recv()
            output pipe.send(item * item)
    except EOFError:
        output_pipe.close()
```

```
if __name__== '__main__':
#First process pipe with numbers from 0 to 9
   pipe_1 = multiprocessing.Pipe(True)
   process_pipe_1 = \
                   multiprocessing.Process\
                   (target=create items, args=(pipe 1,))
   process pipe 1.start()
#second pipe,
   pipe 2 = multiprocessing.Pipe(True)
   process_pipe_2 = \
                   multiprocessing.Process\
                   (target=multiply items, args=(pipe 1, pipe 2,))
   process_pipe_2.start()
   pipe 1[0].close()
   pipe_2[0].close()
   try:
       while True:
            print (pipe_2[1].recv())
   except EOFError:
       print("End")
```

The output obtained is as follows:

How it works...

Let's remember that the pipe() function returns a pair of connection objects connected by a two way pipe. In the example, out_pipe contains the numbers from 0 to 9, generated by the target function create items():

```
def create_items(pipe):
   output_pipe, _ = pipe
   for item in range(10):
        output_pipe.send(item)
   output pipe.close()
```

In the second process, we have two pipes: the input pipe and final output pipe that contains the results:

These are finally printed as:

How to synchronize processes

Multiple processes can work together to perform a given task. Usually, they share data. It is important that the access to shared data by various processes does not produce inconsistent data. Processes that cooperate by sharing data must therefore act in an orderly manner in order to access that data. Synchronization primitives are quite similar to those encountered for the library and threading.

They are as follows:

- ▶ **Lock**: This object can be in one of the states: locked and unlocked. A lock object has two methods, acquire() and release(), to manage the access to a shared resource.
- ▶ Event: This realizes simple communication between processes, one process signals an event and the other processes wait for it. An Event object has two methods, set() and clear(), to manage its own internal flag.
- ► **Condition**: This object is used to synchronize parts of a workflow, in sequential or parallel processes. It has two basic methods, wait() is used to wait for a condition and notify_all() is used to communicate the condition that was applied.

- Semaphore: This is used to share a common resource, for example, to support a fixed number of simultaneous connections.
- ▶ **RLock**: This defines the recursive lock object. The methods and functionality for RLock are the same as the Threading module.
- ▶ **Barrier**: This divides a program into phases as it requires all of the processes to reach it before any of them proceeds. Code that is executed after a barrier cannot be concurrent with the code executed before the barrier.

The example here shows the use of barrier() to synchronize two processes. We have four processes, wherein process1 and process2 are managed by a barrier statement, while process3 and process4 have no synchronizations directives:

```
import multiprocessing
from multiprocessing import Barrier, Lock, Process
from time import time
from datetime import datetime
def test_with_barrier(synchronizer, serializer):
    name = multiprocessing.current process().name
    synchronizer.wait()
    now = time()
    with serializer:
        print("process %s ----> %s" \
              % (name, datetime.fromtimestamp(now)))
def test without barrier():
    name = multiprocessing.current process().name
    now = time()
    print("process %s ----> %s" \
          %(name ,datetime.fromtimestamp(now)))
if __name__ == '__main__':
    synchronizer = Barrier(2)
    serializer = Lock()
    Process(name='p1 - test with barrier'\
            ,target=test with barrier, \
            args=(synchronizer,serializer)).start()
    Process(name='p2 - test_with_barrier'\
            ,target=test with barrier, \
            args=(synchronizer, serializer)).start()
    Process(name='p3 - test_without_barrier'\
```

```
,target=test_without_barrier).start()
Process(name='p4 - test_without_barrier'\
    ,target=test_without_barrier).start()
```

By running the script, we can see that process1 and process2 print out the same timestamps:

C:\Python CookBook\Chapter 3 - Process Based Parallelism\Example Codes
Chapter 3>python process_barrier.py

```
process p1 - test_with_barrier ----> 2015-05-09 11:11:33.291229
process p2 - test_with_barrier ----> 2015-05-09 11:11:33.291229
process p3 - test_without_barrier ----> 2015-05-09 11:11:33.310230
process p4 - test without barrier ----> 2015-05-09 11:11:33.333231
```

How it works...

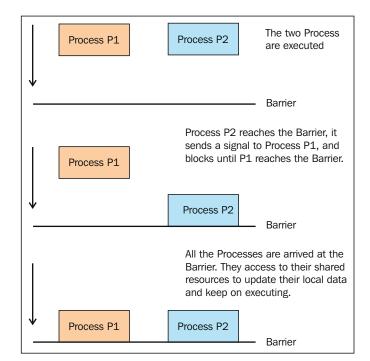
In the main program, we created four processes; however, we also need a barrier and lock primitive. The parameter 2 in the barrier statement stands for the total number of process that are to be managed:

The test with barrier function executes the barrier's wait () method:

```
def test_with_barrier(synchronizer, serializer):
   name = multiprocessing.current_process().name
   synchronizer.wait()
```

When the two processes have called the wait() method, they are released simultaneously:

```
now = time()
with serializer:
    print("process %s ----> %s" %(name \
        ,datetime.fromtimestamp(now)))
```



The following figure shows you how a barrier works with the two processes:

Process management with a barrier

How to manage a state between processes

Python multiprocessing provides a manager to coordinate shared information between all its users. A manager object controls a server process that holds Python objects and allows other processes to manipulate them.

A manager has the following properties:

- ▶ It controls the server process that manages a shared object
- It makes sure the shared object gets updated in all processes when anyone modifies it

Let's see an example of how to share a state between processes:

- 1. First, the program creates a manager list, shares it between *n* number of taskWorkers, and every worker updates an index.
- 2. After all workers finish, the new list is printed to stdout:

The output is as follows:

```
C:\Python CookBook\Chapter 3 - Process Based Parallelism\Example Codes
Chapter 3>python manager.py
```

```
key = 0 value = 0
key = 2 value = 4
key = 6 value = 12
key = 4 value = 8
key = 8 value = 16
key = 7 value = 14
key = 3 value = 6
key = 1 value = 2
key = 5 value = 10
key = 9 value = 18
Results: {0: 0, 1: 2, 2: 4, 3: 6, 4: 8, 5: 10, 6: 12, 7: 14, 8: 16, 9: 18}
```

How it works...

We declare the manager with the following statement:

```
mgr = multiprocessing.Manager()
```

In the next statement, a data structure of the type ${\tt dictionary}$ is created:

```
dictionary = mgr.dict()
```

Then, the multiprocess is launched:

Here, the target function taskworker adds an item to the data structure dictionary:

```
def taskWorker(dictionary, key, item):
    dictionary[key] = value
```

Finally, we get the output and all the dictionaries are printed out:

How to use a process pool

The multiprocessing library provides the Pool class for simple parallel processing tasks. The Pool class has the following methods:

- ▶ apply(): It blocks until the result is ready.
- apply_async(): This is a variant of the apply() method, which returns a result object. It is an asynchronous operation that will not lock the main thread until all the child classes are executed.
- map(): This is the parallel equivalent of the map() built-in function. It blocks until the result is ready, this method chops the iterable data in a number of chunks that submits to the process pool as separate tasks.

map_async(): This is a variant of the map() method, which returns a result object. If a callback is specified, then it should be callable, which accepts a single argument. When the result becomes ready, a callback is applied to it (unless the call failed). A callback should be completed immediately; otherwise, the thread that handles the results will get blocked.

How to do it...

This example shows you how to implement a process pool to perform a parallel application. We create a pool of four processes and then we use the pool's map method to perform a simple calculation:

```
def function_square(data):
    result = data*data
    return result

if __name__ == '__main__':
    inputs = list(range(100))
    pool = multiprocessing.Pool(processes=4)
    pool_outputs = pool.map(function_square, inputs)
    pool.close()
    pool.join()
    print ('Pool :', pool_outputs)
```

This is the result that we get after completing the calculation:

C:\Python CookBook\Chapter 3 - Process Based Parallelism\Example Codes
Chapter 3>\python process pool.py

```
Pool : [0, 1, 4, 9, 16, 25, 36, 49, 64, 81, 100, 121, 144, 169, 196, 225, 256, 289, 324, 361, 400, 441, 484, 529, 576, 625, 676, 729, 784, 841, 900, 961, 1024, 1089, 1156, 1225, 1296, 1369, 1444, 1521, 1600, 1681, 1764, 1849, 1936, 2025, 2116, 2209, 2304, 2401, 2500, 2601, 2704, 2809, 2916, 3025, 3136, 3249, 3364, 3481, 3600, 3721, 3844, 3969, 4096, 4225, 4356, 4489, 4624, 4761, 4900, 5041, 5184, 5329, 5476, 5625, 5776, 5929, 6084, 6241, 6400, 6561, 6724, 6889, 7056, 7225, 7396, 7569, 7744, 7921, 8100, 8281, 8464, 8649, 8836, 9025, 9216, 9409, 9604, 9801]
```