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Parallelisation with Python

Practical Course on Parallel Computing



Python offers two ways of parallelization:

- Multi-Threading:

Sub-tasks on several CPUs share the same memory Requires proper memory synchronisation

- Multi-Processing:

Each process has own memory, processes run completely independent from each other

More stable, but additional overhead in in memory consumption



- Included modules in Python:
 - thread (deprecated, not supported any more in Python3)
 - threading
- The threading module:
 - Provides the following functions:
 - threading.activeCount() Returns the number of thread objects that are active.
 - threading.currentThread() Returns information of the thread from where it is called.
 - threading.enumerate() Returns a list of all thread objects that are currently active.
- Threads can be initialized and started with

```
thread = threading.Thread(target=f, args=(var1, var2,))
thread.start()
(executes function f(var1,var2) in a thread)
```



Example ThreadTest_0.py:

```
#!/usr/bin/python
import threading
def myfunction(a,b):
    print a*b
# Create new threads
thread1 = threading.Thread(target=myfunction, args=(2,3))
thread2 = threading.Thread(target=myfunction, args=(4,6))
# Start the threads
thread1.start()
thread2.start()
```



Everything that you want to execute in a parallel thread can also be written in a Thread **class:**

Methods of the class:

- ____init___ (self [,args]) Initialization
- run() The method contains the code to be ran in parallel.
- start() The method starts a thread by calling the run method.
- join(timeout=None) Waits for threads to terminate.
- isAlive() The method checks whether a thread is still executing.
- getName() The method returns the name of a thread.
- setName() The method sets the name of a thread.
- **To use threading, you have to write your own sub-class of the** Thread **dass**:
 - Define new sub-class
 - Override the __init__() and run() methods



```
Example ThreadTest_1.py:
#!/usr/bin/python
```

```
import threading
import time
class myThread (threading.Thread):
    def __init__(self, threadID, name, counter):
        threading.Thread.__init__(self)
        self.threadID = threadID
        self.name = name
        self.name = name
        self.counter = counter
    def run(self):
        print "Starting " + self.name
        print_time(self.name, 5, self.counter)
        print "Exiting " + self.name
```

. . .



...

```
def print time(threadName, counter, delay):
   while counter:
      time.sleep(delay)
      print "%s: %s" % (threadName, time.ctime(time.time()))
      counter -= 1
# Create new threads
thread1 = myThread(1, "Thread-1", 1)
thread2 = myThread(2, "Thread-2", 2)
# Start new Threads
thread1.start()
thread2.start()
print "Exiting Main Thread"
```



Output:

- Starting Thread-1
- Starting Thread-2
- Exiting Main Thread

Thread-1:	Wed	May	8	16:37:06	2019		
Thread-1:	Wed	May	8	16:37:07	2019		
Thread-2:	Wed	May	8	16:37:07	2019		
Thread-1:	Wed	May	8	16:37:08	2019		
Thread-1:	Wed	May	8	16:37:09	2019		
Thread-2:	Wed	May	8	16:37:09	2019		
Thread-1:	Wed	May	8	16:37:10	2019		
Exiting Thread-1							
Thread-2:	Wed	May	8	16:37:11	2019		
Thread-2:	Wed	May	8	16:37:13	2019		
Thread-2:	Wed	May	8	16:37:15	2019		
Exiting Thread-2							



Add the following lines to your code after you started the threads:

for t in threading.enumerate(): print t

Output:

<_MainThread(MainThread, started 139672845326144)>
<myThread(Thread-1, started 139672823293696)>
<myThread(Thread-2, started 139672814900992)>

There will always be a MainThread



Synchronizing Threads

- □ Synchronization needed for threads that depend on each other
 → Requires communication between threads
- Easiest way for communication: the threading.Event() object:
 Provided functions:
 - set() and clear(): set an internal flag to true or false
 - isSet(): checks if the event has been set by the set() method
 - wait(): blocks further processing until another thread calls the set()
 method; can be used with wait(t) to time-out after t second



Synchronizing: Example

```
Example ThreadTest_2.py:
```

#!/usr/bin/python

```
import threading
import time
```

```
class myThread (threading.Thread):
    def __init__(self, threadID, name, e):
        threading.Thread.__init__(self)
        self.threadID = threadID
        self.name = name
        self.e = e
```

def run(self):

```
if(self.threadID==1):
    print "Start blocking"
    time.sleep(10)
    self.e.set()
    print "End blocking"
```

```
if(self.threadID==2):
    print "Start waiting process"
    while not self.e.isSet():
        event_is_set = self.e.wait(2)
        print "event set: ", event_is_set
        if event_is_set:
            print "processing event"
        else:
            print "doing other things"
e = threading.Event()
```

```
# Create new threads
thread1 = myThread(1, "blocking", e)
thread2 = myThread(2, "non-blocking", e)
```

```
# Start new Threads
thread1.start()
thread2.start()
```

```
#Wait for threads to finish
thread1.join()
thread2.join()
```

```
print "Exiting"
```



Synchronizing: Example

Output:				
Start	block	ing	waiting	process
Start				
event	set:	Fals	se	
doing	other	thir	ngs	
event	set:	Fal	lse	
doing	other	thir	ngs	
event	set:	Fal	lse	
doing	other	thir	ngs	
event	set:	Fal	lse	
doing	other	thir	ngs	
event	set:	Fal	lse	
End bl	locking	9		
event	set:	True	9	
proces	ssing			
Exiti	ing e	event		



Locking is alternative method to synchronize threads

The threading.Lock() object:

- acquire (): change the state to locked
- release(): unlock the state

If one thread calls acquire() for a lock object, all other threads calling acquire() have to wait until the first thread calls release().



Modify previous example (see ThreadTest_3.py):

-Add definition of Lock object to main part of the program:

```
myLock = threading.Lock()
```

- Modify run () function of first example:

```
def run(self):
    print "Starting " + self.name
    # Get lock to synchronize threads
    myLock.acquire()
    print_time(self.name, self.counter, 3)
    # Free lock to release next thread
    myLock.release()
```



Output:

Starting Thread-1									
Starting Thread-2									
Thread-1:	Fri	May	10	17:18:50	2019				
Thread-1:	Fri	May	10	17:18:51	2019				
Thread-1:	Fri	May	10	17:18:52	2019				
Thread-2:	Fri	May	10	17:18:54	2019				
Thread-2:	Fri	May	10	17:18:56	2019				
Thread-2:	Fri	May	10	17:18:58	2019				

 \rightarrow Thread 2 has to wait for thread 1 to finish



^I Multi-core processing configuration is very similar to threading

The multiprocessing **class**:

- Similar methods as the threading class, but:
 - Thread → Process
 - threading → multiprocessing
- Example (MPTest_0.py):

```
#!/usr/bin/python
```

```
import multiprocessing
```

```
def myfunction(a,b):
    print a*b
```

```
# Create new processs
```

```
process1 = multiprocessing.Process(target=myfunction, args=(2,3))
process2 = multiprocessing.Process(target=myfunction, args=(4,6))
```

```
# Start the processs
process1.start()
process2.start()
```



In multi-threading, each thread accesses the same memory
 In multi-processing, memory is completely separated

```
Example (MPTest_1.py):
```

#!/usr/bin/python

import multiprocessing
import threading
import random

```
result =[]
def myfunction():
    result append(random randimt(1,1)
```

```
result.append(random.randint(1,10))
```

Now run myfunction () in multiprocessing and threading mode



Difference to Threading

Create new processes
process1 = multiprocessing.Process(target=myfunction)
process2 = multiprocessing.Process(target=myfunction)

Start the processes
process1.start()
process2.start()

```
#Wait for processes to finish
process1.join()
process2.join()
```

```
print "Multi-processing result: " , result
```

result=[]

```
# Create new threads
thread1 = threading.Thread(target=myfunction)
thread2 = threading.Thread(target=myfunction)
```

```
# Start the threads
thread1.start()
thread2.start()
```

```
#Wait for threads to finish
thread1.join()
thread2.join()
```

```
print "Multi-threading result: " , result
```

Output:

Multi-processing result: []

Multi-threading result: [9, 2]



- Multi-processing can be done by creating a Pool:
 - How to create a pool:

```
Import multiprocessing as mp
```

```
#create a pool with all available CPUs
```

```
pool= mp.Pool(mp.cpu_count())
```

- The Pool class provides the following functions:
 - apply
 - map
 - starmap
 - close(): stops the pool cluster

apply and map are similar to the default functions apply and map, but arguments are executed in parallel on the pool.



Calling apply executes a single function on the pool, see MPTest_2.py:

```
#!/usr/bin/python
import multiprocessing as mp
def myfunction(a, b):
  return a*b
pool = mp.Pool(mp.cpu count())
results = [ pool.apply(myfunction, args=(a,2)) for a in range(1,100) ]
pool.close()
print results
   [2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30,
  32, 34, 36, 38, 40, 42, 44, 46, 48, 50, 52, 54, 56, 58,
  60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 82, 84, 86,
  88, 90, 92, 94, 96, 98, 100, 102, 104, 106, 108 ...
```

/



map takes one iterable as argument and executes processes for each iterable
object, see MPTest_3.py:

```
#!/usr/bin/python
```

```
import multiprocessing as mp
```

```
def myfunction(a):
    return a*2
```

```
pool = mp.Pool(mp.cpu count())
```

```
results = pool.map(myfunction, [a for a in range(1,100)])
```

```
pool.close()
```

print results

[2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38, 40, 42, 44, 46, 48, 50, 52, 54, 56, 58, 60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 82, 84, 86, 88, 90, 92, 94, 96, 98, 100, 102, 104, 106, 108 ...



starmap (available only since Python 3.3) also takes one iterable as argument, but each iterable object can be iterable again, see MPTest_4.py:

```
#!/usr/bin/python3
```

```
import multiprocessing as mp
```

```
def myfunction(a,b):
    return a*b
```

```
pool = mp.Pool(mp.cpu_count())
```

results = pool.starmap(myfunction, [(a,2) for a in range(1,100)])

pool.close()

```
print(results)
```

[2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38, 40, 42, 44, 46, 48, 50, 52, 54, 56, 58, 60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 82, 84, 86, 88, 90, 92, 94, 96, 98, 100, 102, 104, 106, 108 ...



Asynchronous Execution

- Asynchronous execution is often faster then the default synchronized way
- Output of individual jobs is not ordered
- Need specific functions to retrieve the unordered output
- In python:
 - Pool.apply_async
 - Pool.map_async
 - Methods provide just list of output objects, not the output of the parallel calculation

Output can be collected via:

- Call the pool. ApplyResult.get()
- Define callback routine as argument of apply_async or map_async that is called after all jobs finished



Asynchronous: Example 1

```
Example with get (), see MPTest_5.py:
#!/usr/bin/python
import multiprocessing as mp
def myfunction(a, b):
   return a*b
pool = mp.Pool(mp.cpu count())
results_objects = [pool.apply_async(myfunction, args=(a,2)) for a in range(1,100)]
results = [r.get() for r in results_objects]
pool.close()
print results
```



```
Example with callback function, see
#!/usr/bin/python
import multiprocessing as mp
results = []
def myfunction(a, b):
   return a*b
# Define callback function to collect the output:
def collect result(result):
    global results
    results.append(result)
pool = mp.Pool(mp.cpu count())
for a in range(1,100):
   pool.apply async(myfunction, args=(a,2), callback=collect result)
pool.close()
print results
```