Examples

The multiprocessing module

```
from __future__ import print_function
import multiprocessing
def countdown(count):
    while count > 0:
        print("Count value", count)
        count -= 1
    return
if __name__ == "__main__":
    p1 = multiprocessing.Process(target=countdown, args=(10,))
    p1.start()
    p2 = multiprocessing.Process(target=countdown, args=(20,))
    p2.start()
    p1.join()
    p2.join()
```

Here, each function is executed in a new process. Since a new instance of Python VM is running the code, there is no GIL and you get parallelism running on multiple cores.

The Process.start method launches this new process and run the function passed in the target argument with the arguments args. The Process join method waits for the end of the execution of processes p1 and p2.

The new processes are launched differently depending on the version of python and the plateform on which the code is running e.g.:

- Windows uses spawn to create the new process.
- With unix systems and version earlier than 3.3, the processes are created using a fork. Note that this method does not respect the POSIX usage of fork and thus leads to unexpected behaviors, especially when interacting with other multiprocessing libraries.
- With unix system and version 3.4+, you can choose to start the new processes with either fork, forkserver or spawn using multiprocessing set start method at the beginning of your program. forkserver and spawn methods are slower than forking but avoid some unexpected behaviors.

POSIX fork usage :

After a fork in a multithreaded program, the child can safely call only async-signal-safe functions until such time as it calls execve. (see)

Using fork, a new process will be launched with the exact same state for all the current mutex but only the MainThread will be launched. This is unsafe as it could lead to race conditions e.g.:

• If you use a Lock in MainThread and pass it to an other thread which is suppose to lock it at some point. If the fork occures simultaneously, the new process will start with a locked lock which will never be released as the second thread does not exist in this new process.

Actually, this kind of behavior should not occured in pure python as multiprocessing handles it properly but if you are interacting with other library, this kind of behavior can occures, leading to crash of your system (for instance with numpy/accelerated on macOS).

The threading module

```
from __future__ import print_function
import threading
def counter(count):
    while count > 0:
        print("Count value", count)
        count -= 1
t1 = threading.Thread(target=countdown,args=(10,))
t1.start()
t2 = threading.Thread(target=countdown,args=(20,))
t2.start()
```

In certain implementations of Hytnon such as CHytnon, true parallelism is not achieved using threads because of using what is known as the GIL, or **G** lobal **I** nterpreter **L** ock.

Here is an excellent overview of Python concurrency:

Python concurrency by David Beazley (YouTube)

Passing data between multiprocessing processes

Because data is sensitive when dealt with between two threads (think concurrent read and concurrent write can conflict with one another, causing race conditions), a set of unique objects were made in order to facilitate the passing of data back and forth between threads. Any truly atomic operation can be used between threads, but it is always safe to stick with Queue.

```
import multiprocessing
import queue
my_Queue=multiprocessing.Queue()
#Creates a queue with an undefined maximum size
#this can be dangerous as the queue becomes increasingly large
#it will take a long time to copy data to/from each read/write thread
```

Most people will suggest that when using queue, to always place the queue data in a try: except: block instead of using empty. However, for applications where it does not matter if you skip a scan cycle (data can be placed in the queue while it is flipping states from queue.Empty==True to queue.Empty==False) it is usually better to place read and write access in what I call an lftry block, because an 'if' statement is technically more performant than catching the exception.

```
import multiprocessing
import queue
'''Import necessary Python standard libraries, multiprocessing for classes and queue for the
def Queue_Iftry_Get(get_queue, default=None, use_default=False, func=None, use_func=False):
     ''This global method for the Iftry block is provided for it's reuse and
standard functionality, the if also saves on performance as opposed to catching
the exception, which is expencive.
       It also allows the user to specify a function for the outgoing data to use,
 and a default value to return if the function cannot return the value from the queue'''
       if get_queue.empty():
           if use default:
               return default
       else:
            try:
               value = get_queue.get_nowait()
            except queue.Empty:
               if use default:
                   return default
               if use func:
                   return func(value)
               else:
                   return value
   def Queue_Iftry_Put(put_queue, value):
          'This global method for the Iftry block is provided because of its reuse
and
standard functionality, the If also saves on performance as opposed to catching
the exception, which is expensive.
       Return True if placing value in the queue was successful. Otherwise, false'''
        if put_queue.full():
           return False
       else:
               put_queue.put_nowait(value)
           except queue.Full:
                return False
```

Syntax

Parameters

Remarks

The Python developers made sure that the API between threading and multiprocessing is similar so that switching

between the two variants is easier for programmers.