Concurrent & Parallel Programming in Python

Kai Anter, Caspar Sachsenmaier

04.07.2023

Structure

- Python Overview
- Libraries
 - \circ threading
 - asyncio
 - multiprocessing
 - concurrent.futures
- Outlook

Overview: Python

- Interpreted, dynamically typed language
- Implementations:
 - CPython (reference implementation)
 - MicroPython
 - Stackless Python (\rightarrow see other Lightning Talk)

```
o ...
```

```
def my_funcion():
    print("Hello World")

if __name__ == "__main__":
    my_function()
```

CPython

- Written in C
- Current version: 3.11.4
- User scripts are read \rightarrow compiled to bytecode \rightarrow executed
- Uses a single thread to^[1]:
 - Run the user's program and
 - the memory garbage collector
- Has the infamous Global Interpreter Lock

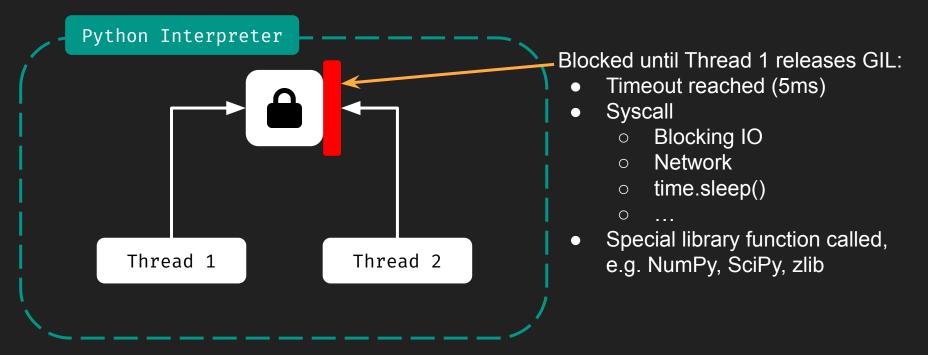
The Global Interpreter Lock (GIL)

- Basically a mutex
 - \rightarrow only 1 Python thread can run bytecode at the same time
- Ensures exclusive access to interpreter internals for current thread^[2]
- Mostly not a problem for performance^[3]
 - Exception: CPU-heavy workloads implemented in Python
 - Larger issue: blocking IO operations

 \rightarrow (In general,) A single Python interpreter can run code concurrently, but not in parallel

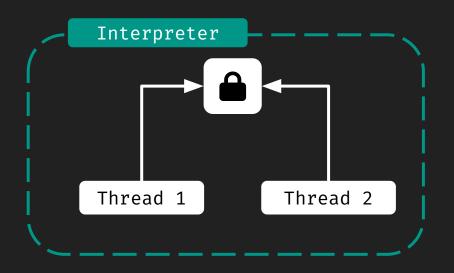
[2] Beazley (2010) [3] Reitz & Schlusser (2016)

The Global Interpreter Lock (GIL)



Library: Threading

- Thread based, similar to Java Threads
- Creates new OS level threads (not user level)
- Limited parallelism due to GIL
- Use Cases:
 - Running blocking IO operations
 - Background (daemon) services



threading: Simple Example

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t.start()

from threading import Thread
import time

```
def my_func(line: str):
    time.sleep(5)
    print(f"Output: {line}")
```

pass

t = Thread(target=my_func, args=("test",))

non-blocking,
wait for finish
with t.join()

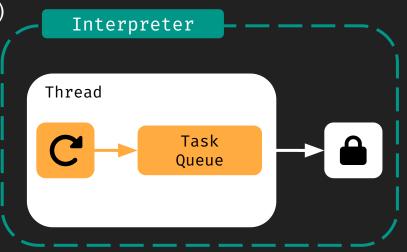
```
# Alternative: Create Subclass of Thread
class MyThread(Thread):
    def run():
        # ..
```

pass daemon=True for background service

args as tuple or array

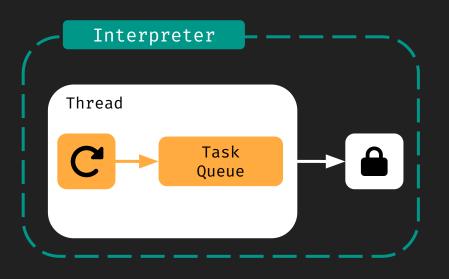
Library: asyncio

- Based on Coroutines + Event Loop
- Introduces new syntax: def async / await (Similar to JS Promises)
- Use cases:
 - Asynchronous IO (without creating new OS threads per blocking operation)



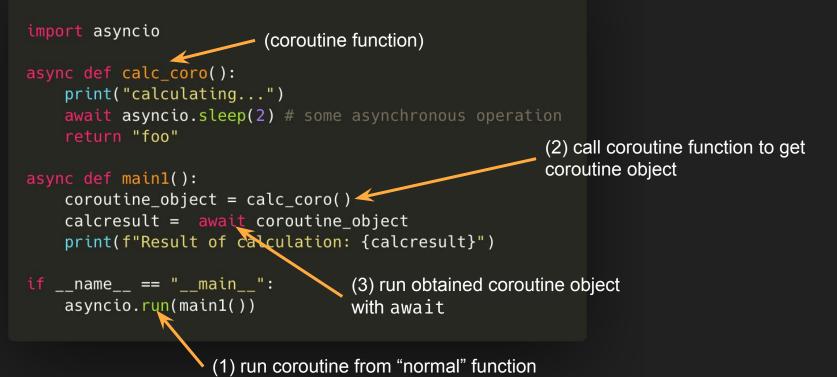
Library: asyncio

- Use async def to define a coroutine function \rightarrow return a coroutine object
- Three ways to run coroutine objects:
 - o asyncio.run(...)
 - await *awaitable_obj*
 - asyncio.create_task(...)
- await can be used on *awaitables*^[4]:
 - Coroutine objects
 - Tasks



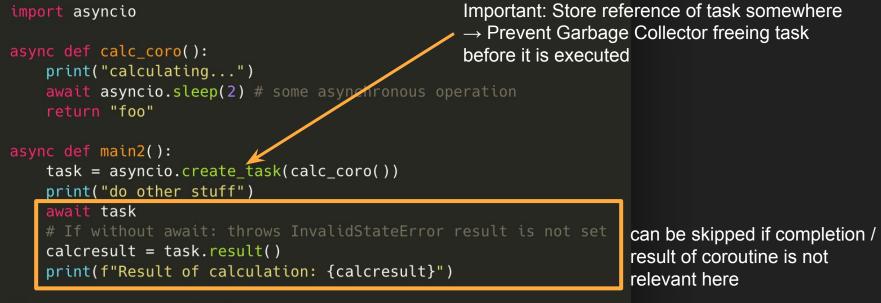
asyncio: Run Coroutine with await



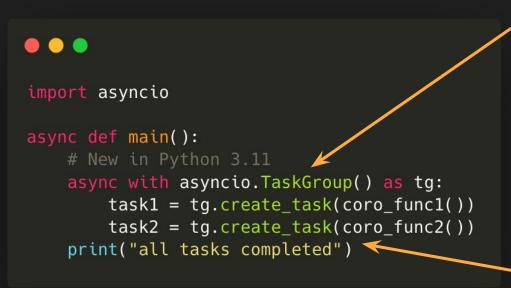


asyncio: Run Coroutine as Task

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asyncio: Run Coroutines with TaskGroups



Alternative to storing Task reference: asyncio.TaskGroup

blocks until all Tasks in TaskGroup are finished

From: Python asyncio documentation

Library: multiprocessing

- Uses processes instead of threads
- Creates new subprocesses of the Python interpreter
 - spawn (Unix & Windows)
 - fork (Unix only)
 - forkserver (some Unix platforms)
- Bypassing GIL by using one for every process
- Use Cases:
 - CPU intensive programs without IO operations



multiprocessing: Communication between processes

- Processes can communicate by Queues or Pipes
- Queues are process safe
- Pipes can get corrupted by multiple processes accessing the same end at the same time → only one sending & one receiving process per pipe

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```
from multiprocessing import Process, Queue

def f(q):
    q.put(['python', 'is', 'gr' + str(8)])

if __name__ == '__main__':
    q = Queue()
    p = Process(target=f, args=(q,))
    p.start()
    print(q.get())  # prints "['python', 'is', 'gr8']"
    p.join()
```

•••

from multiprocessing import Process, Pipe

```
def f(conn):
    conn.send(['python', 'is', 'gr' + str(8)])
    conn.close()
```

```
if __name__ == '__main__':
    parent_conn, child_conn = Pipe()
    p = Process(target=f, args=(child_conn,))
    p.start()
    print(parent_conn.recv())  # prints "['python', 'is', 'gr8']"
    p.join()
```

From: Python multiprocessing documentation

multiprocessing: Shared memory

- use multiprocessing.sharedctypes
- multiprocessing.manager can be used to hold Python objects
- A manager can also be shared on different computers (forkserver)

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```
from multiprocessing import Process, Manager

def f(d, l):
    d['python'] = 'good'
    d['java'] = 'not so much'
    d[1] = '@'

if __name__ == '__main__':
    with Manager() as manager:
        d = manager.dict()
        l = manager.list(range(10))
        p = Process(target=f, args=(d, l))
        p.start()
        p.join()
        print(d)
        print(l)
```

From: Python multiprocessing documentation

multiprocessing: Process Pools

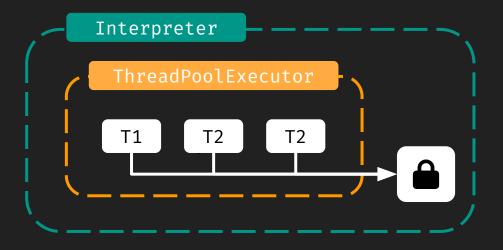
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```
import multiprocessing
def worker_function(x, y):
    return x * y
def main():
    x_{values} = (1, 2, 3)
    y_values = (9, -5, -9)
    with multiprocessing.Pool() as pool:
        results = pool.starmap(worker_function, zip(x_values, y_values))
        print(results)
if __name__ == '__main__':
    main()
```

Like map, but returns iterator

Library: concurrent.futures

- ThreadPoolExecutor or ProcessPoolExecutor
- Use Threads for I/O intensive programs, Processes for CPU intensive ones
- Beware of deadlocks by waiting on another future



concurrent.futures: ProcessPoolExecutor

•••

import concurrent.futures

```
def worker_function(x, y):
    return x * y
```

```
def main():
    x_values = (1, 2, 3)
```

 $y_values = (1, 2, 3)$ $y_values = (9, -5, -9)$

```
with concurrent.futures.ProcessPoolExecutor() as executor:
    results = list(executor.map(worker_function, x_values, y_values))
    print(results)
```

```
if __name__ == '__main__':
    main()
```

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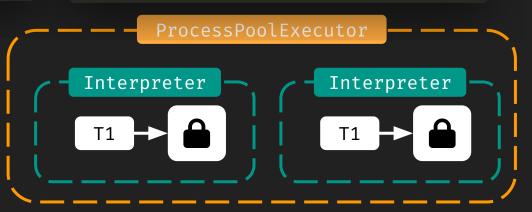
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import multiprocessing
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```
def worker_function(x, y):
    return x * y
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def main(): x_values = (1, 2, 3) y_values = (9, -5, -9)

```
with multiprocessing.Pool() as pool:
    results = pool.starmap(worker_function, zip(x_values, y_values))
    print(results)
```

```
if __name__ == '__main__':
    main()
```



Outlook: A Per-Interpreter GIL

- PEP 684 A Per-Interpreter GIL
 - \circ Currently: Interpreters in the same process share GIL \rightarrow sharing of global states...
 - Proposal: Stop sharing GIL between Interpreters
 - Proposal was accepted, feature will be released with 3.12

Outlook: Making the Global Interpreter Lock Optional

- PEP 703 Making the Global Interpreter Lock Optional in CPython
 - Currently: No parallelism possible in threads because of the GIL
 - Proposal: Making it possible to disable the GIL
 - Proposal just a Draft

References

- Reitz, K., & Schlusser, T. (2016). The hitchhiker's guide to python. O'Reilly Media.
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- <u>https://docs.python.org/3/library/multiprocessing.html</u> (Retrieved: 30.06.2023)
- <u>https://docs.python.org/3/library/concurrent.futures.html (Retrieved: 30.06.2023)</u>