# Using Python to Speed-up Applications with GPUs

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### Schedule

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#### Introduction

#### Why Python?

#### Pros

- A dynamic scripting language
- Rapid development
- Rich libraries NumPy, SciPy
- Great glue language to connect native libraries

#### Cons

- Global Interpretter Lock (GIL)
- Slow execution speed

#### Our Solution: Numba

- ► An opensource **JIT** compiler for **array-oriented programming** in CPython
- Turn numerical loops into fast native code
- Maximumize hardware utilization
- Just add a decorator to your compute intensive function

## Why Numba?

- Works with existing CPython
- ► Goal: Integration with scientific software stack
  - ► Numpy/SciPy/Blaze

#### Mandelbrot in Numba

```
from numba import autojit
Qautojit # <--- All we need to add
def mandel(x, y, max_iters):
    i = 0
    c = complex(x,y)
    z = 0.0j
    for i in range(max_iters):
        z = z*z + c
        if (z.real*z.real + z.imag*z.imag) >= 4:
            return i
    return 255
```

Over 170x speedup

## Need More Speed? Use NumbaPro "CUDA Python"

Our commerical product enables parallel compute on GPU.

```
from numbapro import cuda, uint8, f8, uint32
# use CUDA jit
@cuda.jit(argtypes=[uint8[:,:], f8, f8, f8, f8, uint32])
def mandel_kernel(image, min_x, max_x, min_y, max_y, iters
   height = image.shape[0]
   width = image.shape[1]
   pixel_size_x = (max_x - min_x) / width
   pixel_size_y = (max_y - min_y) / height
    # access thread indices
   x = cuda.threadIdx.x + cuda.blockIdx.x * cuda.blockDim
   y = cuda.threadIdx.y + cuda.blockIdx.y * cuda.blockDim
    # truncated ...
```

▶ **1255**x faster on K20

#### NumbaPro CU API

- Compute Unit (CU) API
- Similar to OpenCL
- Portable parallel programming for GPU, CPU
- Execute kernels asynchronously

## A Saxpy Example

```
def product(tid, A, B, Prod):
    Prod[tid] = A[tid] * B[tid]
def sum(tid, A, B, Sum):
    Sum[tid] = A[tid] + B[tid]
cu = CU('gpu') # or CU('cpu')
... # prepare data
cu.enqueue(product, ntid=dProd.size,
           args=(dA, dB, dProd))
cu.enqueue(sum, ntid=dSum.size,
           args=(dProd, dC, dSum))
... # do something while waiting?
cu.wait()
cu.close() # destroy the compute unit
```

## CUDA C Memory transfer

- cudaMalloc(devptr, size)
  - devptr: output device pointer
  - ▶ size: allocation size
- cudaFree(devptr)
- cudaMemcpy(dst, src, sz, flag)
  - dst: destination pointer
  - src: souce pointer
  - sz: transfer byte size
  - ► flag: cudaMemcpyHostToDevice, cudaMemcpyDeviceToHost

#### CUDA C Kernel Launch

```
dim3 griddim;
dim3 blockdim;
a_kernel<<<griddim, blockdim>>>(arg0, arg1, ..., argN);
```

## CUDA C Putting Everything Together

```
float host_ary[n];
// host -> device
float device_ary;
cudaMalloc(&device_ary, n * sizeof(float));
cudaMemcpy(device_ary, host_ary, cudaMemcpyHostToDevice);
// launch kernel
dim3 blockdim(256);
dim3 griddim(n / 256); // assume n is multiple of 256
inplace_square<<griddim, blockdim>>>(device_ary);
// device -> host
cudaMemcpy(host_ary, device_ary, cudaMemcpyDeviceToHost);
cudaFree(device_ary);
```

## Lab 1

Saxpy in "CUDA Python"

# Quick Lesson to CUDA Python

► Similar to CUDA-C

## Memory Transfer

- ▶ Host->Device
  - d\_ary = cuda.to\_device(ary)
- ► Host->Device (allocate only, no copy)
  - d\_ary = cuda.to\_device(ary, copy=False)
- ▶ Device->Host
  - d\_ary.to\_host()

#### Decorate kernel

cuda.autojit, cuda.jit

#### Kernel launch

- a\_kernel[griddim, blockddim](arg0, arg1)
- similar to C: a\_kernel << griddim, blockdim>>> (arg0, arg1)
- ▶ griddim: tuple of 1-2 ints
- ▶ blockdim: tuple of 1-3 ints

#### Inside the kernel

- threadIdx, blockIdx, blockDim -> cuda.threadIdx, cuda.blockIdx, cuda.blockDim
- i = cuda.threadIdx.x + cuda.blockIdx.x \*
  cuda.blockDim.x

#### Lab 2

A "Monty" Carlo Option Pricer in NumbaPro CU API

## Quick Lesson on NumbaPro CU API

- ► CU = Compute Unit
- ▶ A OpenCL-like API to heterogeneous parallel computing
- Instantiate a CU for 'gpu' or 'cpu'
  - cu = CU('gpu')
- Transfer data to the CU
  - d\_ary = cu.input(ary)
  - d\_ary = cu.output(ary)
  - d\_ary = cu.inout(ary)
  - d\_ary = cu.scratch(shape=arraylen, dtype=np.float32)
  - d\_ary = cu.scratch\_like(ary)

#### Continue...

- Enqueue kernels to the CU
  - cu.enqueue(kernel, ntid=number\_of\_threads, args=(arg0, arg1))
- ► The kernel runs asynchronously
- ▶ Wait for the kernel to complete
  - cu.wait()

## A Numpy Implementation

```
import numpy as np
from math import sqrt, exp
from timeit import default_timer as timer
def step(dt, prices, c0, c1, noises):
    return prices * np.exp(c0 * dt + c1 * noises)
def monte_carlo_pricer(paths, dt, interest, volatility):
    c0 = interest - 0.5 * volatility ** 2
    c1 = volatility * np.sqrt(dt)
    for j in xrange(1, paths.shape[1]):
        prices = paths[:, j - 1]
        noises = np.random.normal(0., 1., prices.size)
        paths[:, j] = step(dt, prices, c0, c1, noises)
```