

# Multicore Computing

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# OpenMP

- Open Multi-Processing
- An API that supports multi-platform shared-memory multiprocessing programming in C, C++, and Fortran

# OpenMP: Quick Start

```
for (int i = 0; i < N; ++i){  
    b[i] = a[i] + 1;  
}
```

# OpenMP: Quick Start

```
#pragma omp parallel for schedule(static) num_threads(8)
```

```
for (int i = 0; i < N; ++i){  
    b[i] = a[i] + 1;  
}
```

- g++ test.cc –fopenmp -o test -O2
- brew install libomp
- clang++ test.cc -o test -O2 -Xpreprocessor -fopenmp -I/usr/local/include  
-L/usr/local/lib -lomp

# OpenMP: Quick Start

```
int sum = 0;
```

```
for (int i = 0; i < N; ++i){  
    sum += a[i];  
}
```

# OpenMP: Quick Start

```
int sum = 0;  
#pragma omp parallel for schedule(static) default(shared)  
reduction(+:sum) num_threads(8)  
for (int i = 0; i < N; ++i){  
    sum += a[i];  
}
```

# OpenMP: Slow Start

- `#include<omp.h>`
- `void omp_set_num_threads(int num_threads)`
- `int omp_get_num_threads()`
- `int omp_get_thread_num()`
  
- `#pragma omp atomic (update/read/write/capture)`
- `#pragma omp critical`

# Gauss-Seidel Smoother

- Solving PDE
- Parallel
- Synchronization
- Lock
- Communication

# Matrix Multiplication (in Theory)

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{bmatrix}$$

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Arithmetic intensity: the ratio of the work to the memory traffic

# Strassen Algorithm

$$M_1 = (A_{11} + A_{22})(B_{11} + B_{22});$$

$$M_2 = (A_{21} + A_{22})B_{11};$$

$$M_3 = A_{11}(B_{12} - B_{22});$$

$$M_4 = A_{22}(B_{21} - B_{11});$$

$$M_5 = (A_{11} + A_{12})B_{22};$$

$$M_6 = (A_{21} - A_{11})(B_{11} + B_{12});$$

$$M_7 = (A_{12} - A_{22})(B_{21} + B_{22}),$$

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} M_1 + M_4 - M_5 + M_7 & M_3 + M_5 \\ M_2 + M_4 & M_1 - M_2 + M_3 + M_6 \end{bmatrix}$$

# HW 1: Logistic Regression

- Given matrix X and label Y, perform gradient descent of logistic regression
- 10 independent test cases. Each case weights 1 pt.
- The compilation is considered failed if it does not finish in **1 minute**.
- A test case is considered **incorrect** if it does not finish in **2 minutes**.
- The training accuracy must reach **60%**.
- The **summation** of the execution time across 10 cases will be used to rank **correct** solutions.
- Due: 09/13/2024 5:00 pm EDT

# Grading

• Homework	40%
• Reading	10%
• Project	50%
• $90\% \leq A \leq 100\%$	
• $80\% \leq B < 90\%$	
• $70\% \leq C < 80\%$	
• $60\% \leq D < 70\%$	
• $0\% \leq F < 60\%$	

- 5 pieces of homework.
- No late submissions.
- No 3<sup>rd</sup> party code
- Automatically tested: Please strictly follow the output format. An incorrect format is considered as a wrong answer.
- The best 4 scores among the 5 are counted in your final grade.
- The fastest correct solution in each homework gets 10% bonus score in the final grade.
- Other correct solutions that are no slower than 2X of the fastest one gets 5% bonus score in the final grade.

# Input Data

- First line contains 8 integers:  $N D x_0 x_1 A B C M$
- For  $i \geq 2$ 
  - $X[i] = (A * X[i - 1] + B * X[i - 2] + C) \% M$
- For all  $i$ 
  - $X[i] \neq M;$
- $N \leq 10^5$
- $D \leq 1600$

# Input Data

- First line contains 8 integers:  $N D x_0 x_1 A B C M$
  - For  $i \geq 2$ 
    - $X[i] = (A * X[i - 1] + B * X[i - 2] + C) \% M$
  - For all  $i$ 
    - $X[i] /= M;$
- Caution the potential overflow here!
- $N \leq 10^5$
  - $D \leq 1600$

# Output Format

- D lines
- Each line contains a floating number
  - The logistic regression parameters

# What Do We Need to Do?

- We are required to complete two scripts
- `compiler.sh`
  - it is executed once before the actual testing starts
- `run.sh`
  - it should takes two arguments, the first argument is the input file name, the second one is the file name that you should write your results into.

# Testing Environment

- ssh yourusername@granger.cs.rit.edu
  - Intel(R) Xeon(R) CPU E5-2650 v4 @ 2.20GHz
  - 48 threads in total (2 sockets, 12 cores per socket, 2 threads per core)
  - 251 GB memory  
  - Testing limit:
    - 8 threads

“Premature optimization is the root of all evil”

--- Sir Tony Hoare

# GPU Computing

# GPU (Graphics Processing Unit)

- Rendering
  - 3D surfaces
  - Textures
  - Lights
  - Views







VIA 9GAG.COM



# GPU Rendering

- Direct3D
- OpenGL
- Use primitives to render a frame

# GPU ~~Rendering~~ Computing (Before 2007)

- Direct3D
- OpenGL
- Use primitives to render a frame
- Make your 2D array as a frame and call render primitives

# GPU Computing (After 2007)

- CUDA (Compute Unified Device Architecture)
- C/C++, Fortran
- GPGPU (General-Purpose computing on Graphics Processing Units)
- OpenCL

# GPU Architecture

- 108 Streaming Multi-processor (SM)
- 40 GB High-Bandwidth Memory (HBM)
  - 1555 GB/sec
  - 6912 FP32 CUDA cores
- 432 Tensor Cores, TensorFloat-32(TF32) Dense Tensor (156 TFLOPs)
- 192KB \* 108 L1 Cache
- 40960 KB L2 Cache

# GPU Scheduling

- SIMT (Single Instruction Multiple Thread)
- Warp
- Dangerous to implement critical section (Pre Volta)
- Independent Thread Scheduling (After Volta)

# CUDA Programming

- Kernel
- Grid
- Block
- Thread
- Warp
- Host Memory
- Device Memory
  - Global Memory
  - Shared Memory

# CUDA Programming

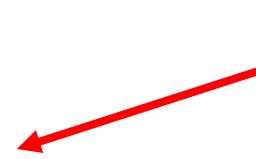
```
__global__
void saxpy(int n, float a, float *x, float *y){
    int i = blockIdx.x * blockDim.x + threadIdx.x;
    if (i < n)
        y[i] = a * x[i] + y[i];
}
```

# CUDA Programming

```
__global__
void saxpy(int n, float a, float *x, float *y){
    int i = blockIdx.x * blockDim.x + threadIdx.x;
    if (i < n)
        y[i] = a * x[i] + y[i];
}
saxpy<<<nB, nT>>>(n, a, x, y);
```

# CUDA Programming

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__global__
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```



device memory

# CUDA Programming

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    int i = blockIdx.x * blockDim.x + threadIdx.x;
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}
saxpy<<<nB, nT>>>(n, a, x, y);
```

The diagram illustrates the memory access pattern of the SAXPY kernel. It shows two parallel red arrows originating from the variable 'y' in the CUDA kernel code. These arrows point to the same memory location in the parameter list 'y' of the function call 'saxpy<<<nB, nT>>>(n, a, x, y);'. The text 'device memory' is written in red above the arrows, indicating that both points refer to memory located in the GPU's device memory space.

# CUDA Programming

```
__global__
void saxpy(int n, float a, float *x, float *y){
    int i = blockIdx.x * blockDim.x + threadIdx.x;
    if (i < n)
        y[i] = a * x[i] + y[i];
}
saxpy<<<nB, nT>>>(n, a, x, y);
```

device memory

float\* x;  
cudaMalloc(&x, n \* sizeof(float));

cudaError\_t cudaMalloc ( void\*\* devPtr, size\_t size )

# Host Memory vs. Device Memory

- `cudaMalloc`, `cudaFree`
- `cudaError_t cudaMemcpy( void* dst, const void* src, size_t count, cudaMemcpyKind kind )`
  - `cudaMemcpyHostToHost = 0`
    - Host -> Host
  - `cudaMemcpyHostToDevice = 1`
    - Host -> Device
  - `cudaMemcpyDeviceToHost = 2`
    - Device -> Host
  - `cudaMemcpyDeviceToDevice = 3`
    - Device -> Device
  - `cudaMemcpyDefault = 4`
    - Direction of the transfer is inferred from the pointer values. Requires unified virtual addressing

# CUDA Compilation

- nvcc a.cu -o a.out -O3 -Xptxas -O3
- cuda-gdb
  - -g -G (without optimizations)
  - info cuda threads
  - cuda thread 0
- cuda-memcheck
- nvprof
  - nvvp

# Scan

- Inclusive scan
- Exclusive scan
- Naïve scan
- Work-efficient scan