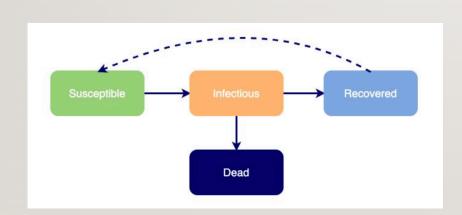
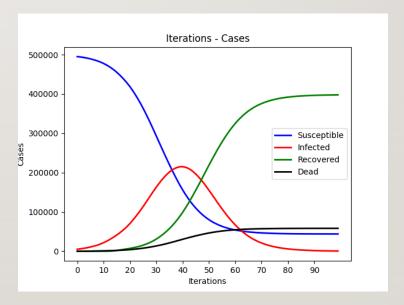
## Parallelize Massively Transmission Simulations of COVID-19 Diseases

## Preface & Motivation

## SIR Model (Susceptiable, Infectious, Recovered)

- Epidemiological Model
- Compute # of the infected in a closed population





## Challenges & Limitation

```
For iterations:
    For node1 in allNodes:
        CheckWhetherInfectious(node1)
    For node2 in allNodes:
        CheckWhetherSusceptible(node2)
        // Euclidean Distance
        distance = CalculateDistance(node1, node2)
        if (distance < infectious_radius) {
            prob = CalculateInfectiousRate(node1, node2)
        }
        UpdateInfectedNode()

For node in allNodes:
        move(node)</pre>
```

```
F(x) = a * exp(-b * x) \quad (0 <= x <= r) \ a,b: infectious\_param \quad r: infectious\_radius
```

#### Computation-Intensive Workload

Computation overhead dreasitically increases when # of node increase

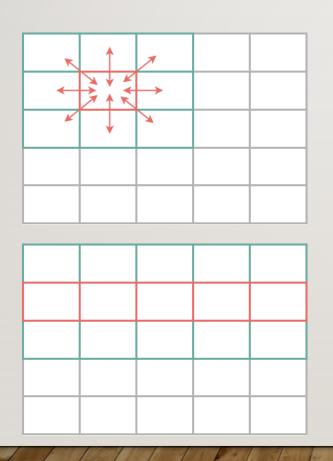
### Iteration Dependency

Hard to parallize between phases and load balancing

### Randomly Distribution of Nodes

No common data split method to parallelize

## Blocked-Version Parallelism with OpenMP



#### **Blocked Version**

- Split map into blocks
- Make width/height of block = infectious radius
   => Only need to consider nearby 9 blocks for each block
- Only parallelize within iteration
- One thread maps to one blocks
- Lock Awareness? Synchronization?

#### **Row-Regioned Version**

- Split map into rows
- One thread maps to one rows

#### Optimization

## Blocked-Version Parallelism with OpenMP

#### Data Computation Parallelism

Use OpenMP Thread library
Collapse 2 layers for-loop in blocked version
Prevent synchronization with "nowait"

# Lock-Free with Backward State Assignment

Reduce the lock contention & fully parallelize

### **Both Spatial & Temporal Locality**

Reduce the mem-copy time Increase the cache hit

### Utilize row-based split method

Reduce # of times that thread context switches

### CUDA VERSION IMPLEMENTATION

#### Distance & Infection Rate Calculation

Pair-Wise Calculation of Distance & Infection Rate

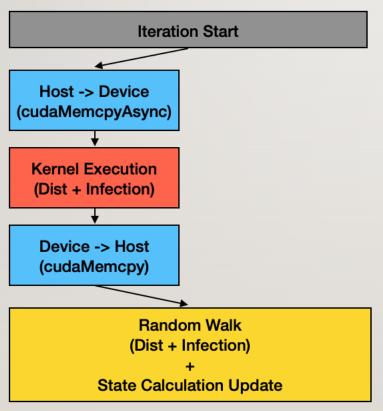
### **Memory Concern**

Tradeoff Between Memory and Speed

### GPU Related Optimization

Specific Technique about Cuda Implementation

## Blocked-Version on GPU



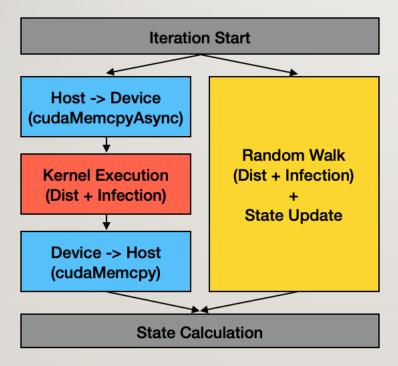
#### Pro:

- Row Region Version Reduced Caculation Workload
- Minimize Memory Space Utilization

#### Con:

- Complex Data Structure
- Moving Logic Add More Burden On CPU Calculation
- Map Data Structure Should be Locked to Avoid Race Condition
- Can't Overlap Computation Between CPU & GPU

## Adjacency Matrix on GPU



#### Pro:

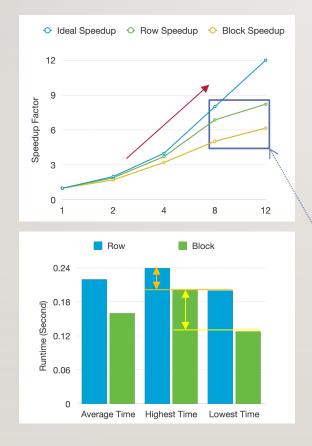
- Calculation Overlapping Between CPU & GPU
- OpenMP Integration in CPU Part
- Less Branch Diverge on Kernel Execution
- Coalesced Memory Access Pattern

#### Con:

- Memory-Consuming Can Only Simulate Roughly 35000
   People on GTX1660 Super (6G)
- Unnecessary Calculation Each Pair of Node Will be Calculate Twice

#### Experiment

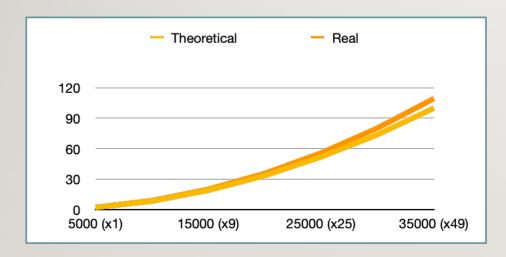
## OpenMP Parallelism for SIR Model



- Larger data block > smaller data block when data randomly distributes
  - Greater load balancing when data block is bigger
  - But when data is too intensive, the straggler thread would emerge
  - Needs to find balancing way to get optimal load balancing result
- Gradually converge when # threads > 8

#### Experiment

## CUDA Profiling & Scalability



#### Time Complexity:

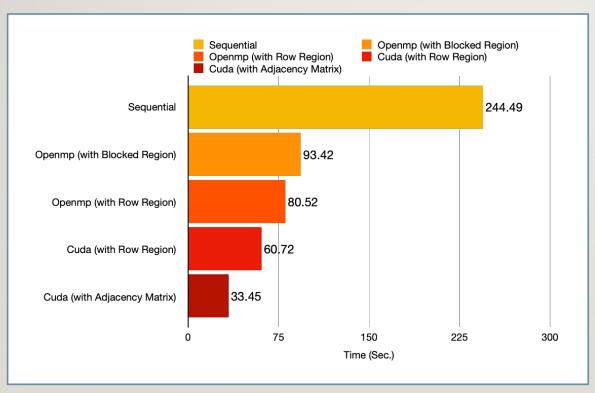
- O(V^2)
- Some Overhead:
  - Memory Copy Between Device & Host
  - Data Structure Implementation
  - Code Structure

### **Profiling Result**

- Global Load Throughput: 107.63GB/s
- Achieved Occupancy: 0.775715
- SM Efficiency: 99.98%
- Global Store Throughput: 344.41GB/s

#### Experiment

## Cuda + OpenMP Performance



#### **Overall Performance:**

- Node Count: 25000
- GTX 1660 SUPER (1408 cuda cores)
- AMD Ryzen 7 3700X 8-Core Processor (16 Thread)
- clock\_gettime with MONOTONIC

## Reference

- How simulation modelling can help reduce the impact of COVID-19[J]. Journal of Simulation. Currie C S M, Fowler J W, Kotiadis K, et al. 2020: 1-15.
- A Time-dependent SIR model for COVID-19 with Undetectable Infected Persons. Yi-Cheng Chen, Ping-En Lu, Cheng-Shang Chang, Tzu-Hsuan Liu. 2020
- Mathematical models of SIR disease spread with combined non-sexual and sexual transmission routes.
   Joel C.Miller. 2017
- Simulations for epidemiology and public health education[M]//Operational Research for Emergency Planning in Healthcare: Volume 2. Huang CY, Tsai YS, Wen TH. 2016
- A simulation model of the epidemiology of urban dengue fever: literature analysis, model development, preliminary validation, and samples of simulation results[J]. Focks D A, Daniels E, Haile D G, et al. The American journal of tropical medicine and hygiene, 1995, 53(5): 489-506.
- Directed-graph epidemiological models of computer viruses[M]//Computation: the micro and the macro view. Kephart J O, White S R. 1992: 71-102.