

# Road Network Simulation using FLAME GPU

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# Overview

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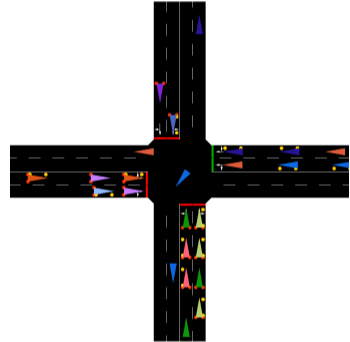
# Introduction

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# Road Network Simulation

Need for improved road simulation systems [4, 8]

- Increasing number of vehicles globally
- Poor utilisation of existing infrastructure
- Relatively cheap
- Decision making



An example of traffic microsimulation (SUMO)

# Why Agent Based Simulation on the GPU

## Why ABS / Microsimulation?

- More natural method of description
- Allow emergence of more complex behaviour
- Good for modelling congested networks

## Why GPGPU?

- *Not* embarrassingly parallel but it is *well suited* for GPGPU computing
- Aspects are SIMD (*Same Instruction Many Data*) in nature
- Has been demonstrated as GPGPU suitable [7, 9]
- Speed-up allows for increased complexity / scale

- Demonstrate performance of road network simulation using FLAME GPU
- Evaluate performance scalability using an artificial road network.
  - Scale population size
  - Scale population and environment
- Demonstrate interactive visualisation using instancing

# Gipps' Car Following Model

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# Car Following

- Key vehicle behaviour
- Drive at desired speed without colliding into other vehicles
- Considering factors such as *reaction time*, *vehicle limitations*, *neighbouring vehicles*  
...
- Many car following models exist
  - Safety-distance models
  - Psycho-physical models



# Gipps' Car Following Model

Gipps' Car Following Model defined in 1981 by Peter Gipps

- Safety Distance Model
- Considers driver & vehicle characteristics
- Only considers the preceding vehicle
- One of the most commonly used models

# Aims - Gipps' Car Following Model

“The model should mimic the behaviour of real traffic” [3]

“parameters which correspond to obvious characteristics of drivers and vehicles” [3]

“should be well behaved when the interval between successive recalculations of speed and position is the same as the reaction time” [3]

# Notation - Gipps' Car Following Model

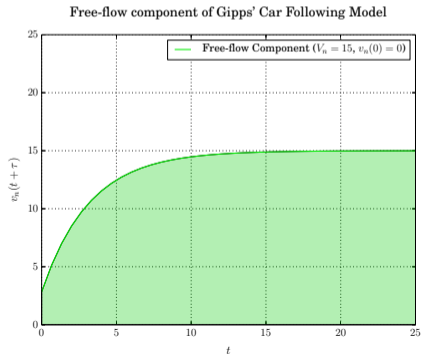
$a_n$	the maximum acceleration of vehicle $n$
$b_n$	the most severe braking that the vehicle $n$ will undertake
$s_n$	the effective size of vehicle $n$ , including a margin
$V_n$	the target speed of vehicle $n$
$x_n(t)$	the location of the front of vehicle $n$ at time $t$
$v_n(t)$	the speed of vehicle $n$ at time $t$
$\tau$	constant reaction time for all vehicles
$\hat{b}$	estimate of leading vehicles most severe braking

Notation for variables used by Gipps' car following model

# Equations - Gipps' Car Following Model

## Free-flow Conditions

$$v_n(t + \tau) \leq v_n(t) + 2.5a_n\tau(1 - v_n(t)/V_n)(0.025 + v_n(t)/V_n)^{\frac{1}{2}}$$

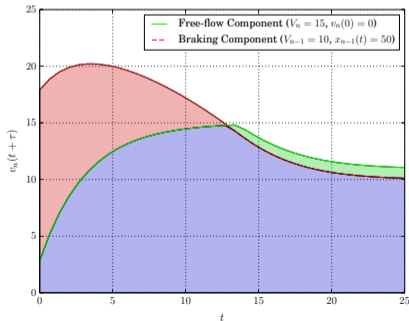


# Equations - Gipps' Car Following Model

## Congested Conditions (Braking)

$$v_n(t + \tau) \leq b_n \tau + \sqrt{b_n^2 \tau^2 - b_n(2[x_{n-1}(t) - s_{n-1} - x_n(t)] - v_n(t)\tau - v_{n-1}(t)^2/\hat{b})}$$

Free-flow and Braking components of Gipps' Car Following Model



# Limitations - Gipps' Car Following Model

- Time-step should be set to reaction time  $\tau$
- Assumes drivers:
  - Drive in a safe manner
  - Can make accurate observations

# Implementation

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# Artificial Road Network

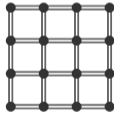
- Scales consistently unlike real world networks
- Single lane uniform grid
- Grid made of  $N$  rows and columns
- 2 sections of road between each adjacent junction
- $N^2$  junctions and  $4N(N - 1)$  one-way roads



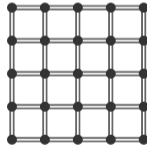
$N = 2$



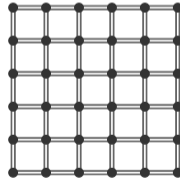
$N = 3$



$N = 4$



$N = 5$

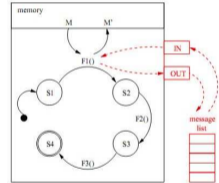


$N = 6$



FLAME GPU is a “template based simulation environment” for agent based simulation on Graphics Processing Unit (GPU) architecture [6]

- Agents are represented as X-Machines
- Agents can communicate via globally accessible message lists
- Messages are crucial for interaction
- Message lists can be partitioned to “ensure the most optimal cycling of messages”[6]



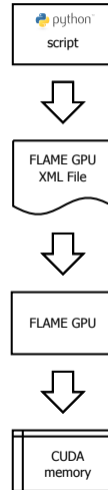
There are currently 3 defined message partitioning schemes

- **Non-partitioned messaging**
  - All to All
- **Discrete partitioned messages**
  - 2D non-mobile agents only (i.e. Cellular Automata)
- **Spatially partitioned messages**
  - Continuous space
  - Requires *radius* and *environment bounds*

Aims to reduce the size of message lists

# Implementing Gipps' Car Following Model using FLAME GPU

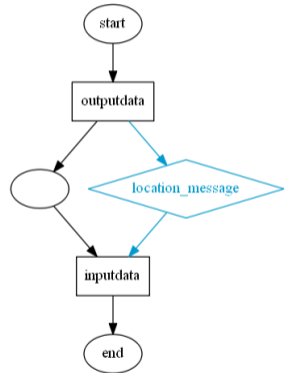
- Each vehicle represented by an agent
- Initial values generated with *python* script and stored in a *FLAME GPU XML file*
- Road network stored in *CUDA constant memory*
  - Does not change
  - Agents interact with same network
  - *CUDA Read-Only Data Cache* could allow larger road networks (> 64kB of memory)



# Implementing Gipps' Car Following Model using FLAME GPU

For each step in the simulation

- Agents output their observable properties (`outputdata`)
- Agents iterate through their message lists for the lead vehicle (`inputdata`)
  - Gipps' car following model is applied using the lead vehicle information
  - *Forward Euler* used to calculate location and velocity
  - New roads randomly assigned at junctions



# Experiments & Results

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# Experiments, Model Parameters, Hardware

## Experiments

	Grid Size	Agent Count	Road Length
Fixed Grid	$N = 16$	256 to 262144	10000m
Scaled Grid	$N = 2$ to $N = 24$	512 to 141312	1000m

(64 vehicles per 1000m)

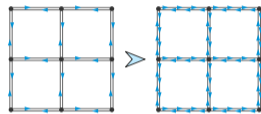
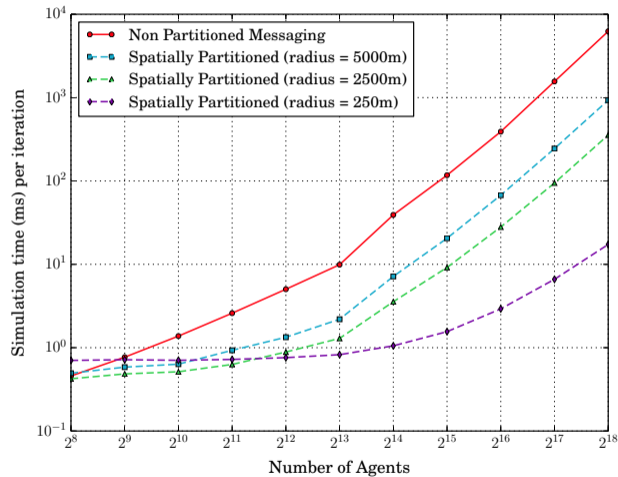
## Model Parameters proposed by Gipps

$a_n$	sampled from the normal distribution $N(1.7, 0.3^2) \text{ m/sec}^2$
$b_n$	$-2.0a_n$
$s_n$	sampled from the normal distribution $N(6.5, 0.3^2) \text{ m}$
$V_n$	sampled from the normal distribution $N(20.0, 3.2^2) \text{ m/sec}$
$\tau$	2/3 seconds
$\hat{b}$	the minimum of $-3.0$ and $(b_n - 3.0)/2 \text{ m/sec}^2$

## Hardware/Software

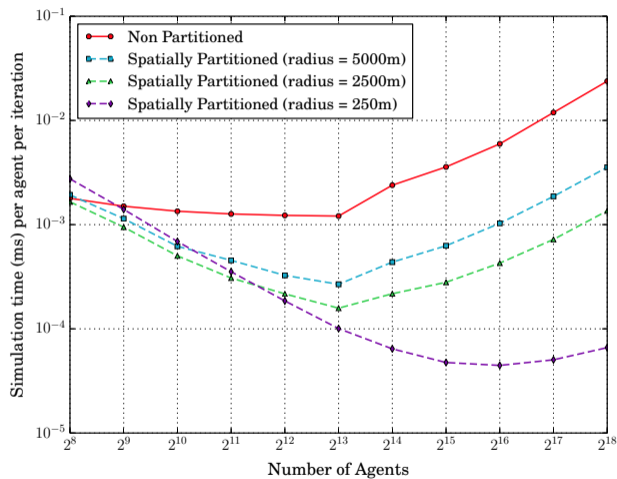
- FLAME GPU 1.4 for CUDA 7.0
- Intel Core i7 4770K
- NVIDIA Tesla K20c

# Fixed Grid Network



- Spatially partitioned messaging outperforms non-partitioned messaging
- Smaller radii outperforms larger radii beyond overhead
- Distinct gradient change at  $2^{13}$  agents

# Fixed Grid Network - Per Agent

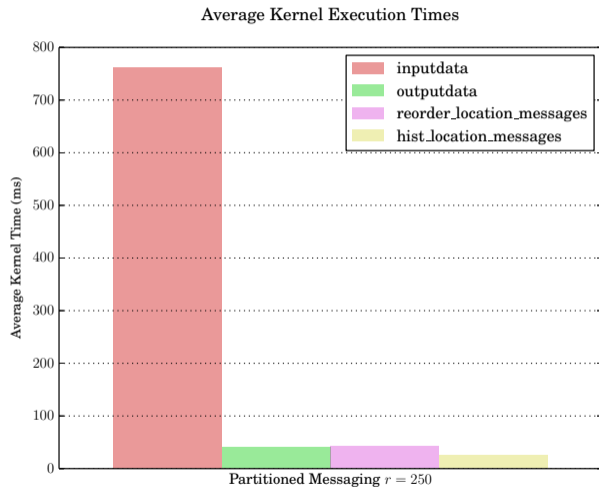


- Distinct gradient change at  $2^{13}$  agents - hardware utilisation vs larger message lists
- Non-partitioned outperformed by partitioned messaging
- $r = 250$  scales much better per agent
- Maximum message count

Non-partitioned	262144
Partitioned $r = 5000$	19662
Partitioned $r = 2500$	9720
Partitioned $r = 250$	309

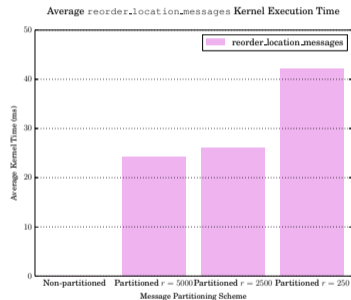
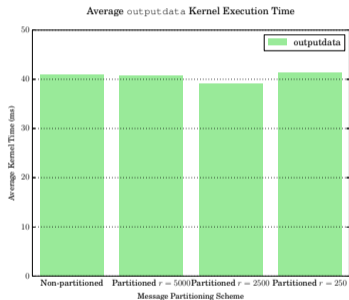
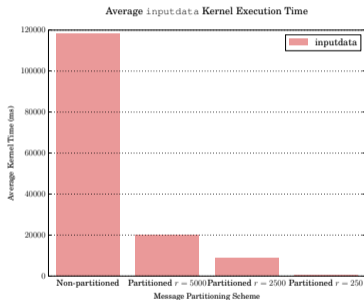


# Fixed Grid Network - Kernel Profiling



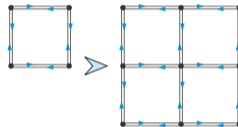
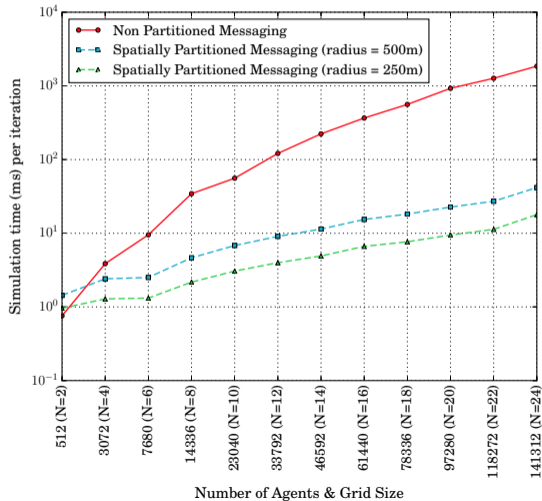
- Kernel times averaged over 10 iterations
- Some Kernels omitted
- 32768 Agents
- Spatial Partitioned messaging with  $r = 250$
- `inputdata` kernel is dominant

# Fixed Grid Network - Kernel Profiling



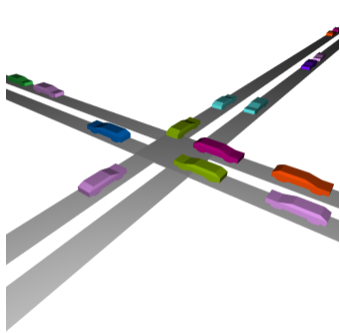
# Scaled Grid Network

Average iteration execution time for increasing Grid Size  $N$  with a fixed vehicle density of 64 agents per 1000m

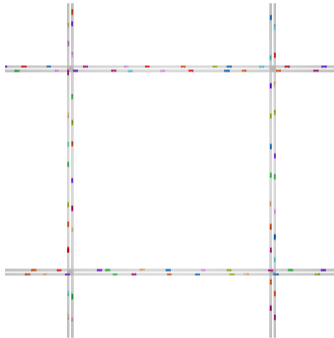


- As scale increases performance decreases
- Spatially partitioned messaging outperforms non-partitioned beyond overhead
- Spatial partitioning scales better
- Up to 103x performance increase for spatial partitioning than non-partitioned

# Interactive Visualisation



Nearby



Overview

- Cross platform C++, OpenGL & libSDL<sup>[2]</sup>
- OpenGL Interop<sup>[5]</sup> & instanced rendering<sup>[1]</sup> used to avoid unnecessary host-device memory transfers
- $N = 8$ , length 1000m, 8192 vehicles & 1000 iterations
- NVIDIA GeForce GTX 660

Console	15079ms
Visualisation	16291ms
Increase	1.08x

## Conclusions & Future Work

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# Conclusions

- Two experiments carried out, demonstrating suitability of FLAME GPU for road network simulation
- Scaling behaviour has been investigated
- Performance difference between messaging communication schemes highlighted

# Future Work

- Message partitioning techniques for network based communication
- Support wider range of road networks
- Non-uniform vehicle distribution
- Increased accessibility through visualisation of aggregate data on the GPU
- Increased variation of vehicles using procedural instancing

Thank You

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flamegpu.com



# References I

- [1] OpenGL SDK glDrawArraysInstanced manpage. <https://www.opengl.org/sdk/docs/man/html/glDrawArraysInstanced.xhtml>
- [2] Simple DirectMedia Layer (libSDL). <https://www.libsdl.org/>
- [3] Gipps, P.G.: A behavioural car-following model for computer simulation. *Transportation Research Part B: Methodological* 15(2), 105–111 (1981)
- [4] Neffendorf, H., Fletcher, G., North, R., Worsley, T., Bradley, R.: *Modelling for intelligent mobility* (Feb 2015)
- [5] Nvidia, C.: *Cuda c programming guide*. [http://docs.nvidia.com/cuda/pdf/CUDA\\_C\\_Programming\\_Guide.pdf](http://docs.nvidia.com/cuda/pdf/CUDA_C_Programming_Guide.pdf) (Mar 2015), last accessed 2015-03-30

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- [6] Richmond, P.: Flame gpu technical report and user guide. Tech. rep., technical report CS-11-03. Technical report, University of Sheffield, Department of Computer Science (2011)
- [7] Strippgen, D., Nagel, K.: Multi-agent traffic simulation with cuda. In: High Performance Computing & Simulation, 2009. HPCS'09. International Conference on. pp. 106–114. IEEE (2009)
- [8] UK Department for Transport: Quarterly Road Traffic Estimates: Great Britain Quarter 4 (October - December) 2014 (Feb 2015)
- [9] Wang, K., Shen, Z.: A gpu based trafficparallel simulation module of artificial transportation systems. In: Service Operations and Logistics, and Informatics (SOLI), 2012 IEEE International Conference on. pp. 160–165. IEEE (2012)