

ESCAPE

Epidemic Simulator Compiler and
Programming Environment

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The ESCAPE Framework

- ESCAPE: Epidemic Simulator Compiler and Programming Environment
- A novel framework for creating agent-based epidemics simulators
- A meta-simulator for creating epidemic simulators on contact networks
- Supports multiple interacting contagions
- Aim 1: Make agent-based simulations easier to write
- Aim 2: Make agent-based simulations faster to run

ESL: Epidemic Simulation Language

- A domain specific language for describing epidemic simulations
- Syntax inspired by Python/Julia
- Comes with Editor/IDE support
 - Provides custom Language Server
 - Currently tested on NeoVim and Jupyter Lab
 - Easy to port to VSCode, Emacs, ..
- Compiles down to C++ with OpenMP / CUDA*

```
contagion c1
  state type c1_state_t

  transition
    E -> Ipresymp, p = p_E_Ipresymp, dwell = fixed3
    E -> Iasymp, p = p_E_Iasymp, dwell = normal5
    Ipresymp -> Isymp, dwell = fixed2
    Isymp -> R, dwell = isymp_r_dwell
    Iasymp -> R, dwell = normal5
  end

  transmission
    Ipresymp => S -> E
    Isymp => S -> E
    Iasymp => S -> E
  end

  susceptibility fn c1_susceptibility
  infectivity fn c1_infectivity
  transmissibility fn c1_transmissibility
  enabled fn c1_enabled
end

def p_E_Ipresymp(n: node) -> float:
  return 0.65
end

def p_E_Iasymp(n: node) -> float:
  return 0.35
end

def c1_susceptibility(v: node) -> float:
  if v.c1.state == S:
    return 1.0
  else:
    return 0.0
  end
end

def c1_infectivity(v: node) -> float:
  if v.c1.state == Isymp or v.c1.state == Iasymp:
    return 1.0
  else:
    return 0.0
  end
end

def c1_transmissibility(e: edge) -> float:
  return transmissibility_scale * e.duration / 86400
end

def c1_enabled(e: edge) -> bool:
  if (e.is_non_home_edge and
      (is_isolating(e.source_node) or is_isolating(e.target_node))):
    return False
  elif e.is_school_edge and not is_school_day():
    return False
  else:
    return True
  end
end
```

Preliminary performance results

- Test case: SIR model for whole USA
 - 330M agents
 - 12B contacts per day
 - 365 days
- Test cluster: Rivanna @ UVA
 - CPU: Intel Xeon 2 x 20 cores
 - MEMORY: 384 GB
- Digital similar:
 - 48 US States + DC
- ESCAPE init time: 55s
- ESCAPE time per tick: 3.2s

Population	Metric	EpiHiper	ESCAPE
USA	Runtime	28m 20s	11m 40s
	# Nodes	15	1
	Speedup	1	36.43

Preliminary performance results

- Test case: In-school NPIs [1]
 - Hybrid Learning: schools open Monday, Tuesday, and Wednesday for in-person learning; Thursday and Friday remote learning.
 - In-school testing: Daily Antigen Test on in-person school days (80% sensitivity); Weekly Antigen Test on Mondays (95% sensitivity)
 - Voluntary home isolation: Stay at home for 14 days after testing positive
 - Seeding: 5 infections every day for 10 days
- Test cluster: Anvil @ Perdue
 - CPU: AMD EPYC 2x64 cores
 - MEMORY: 256 GB
- Digital similars: CA, VA

Population	Metric	EpiHiper	ESCAPE
Virginia	Runtime	2m 4s	19.7s
	# Nodes	1	1
	Speedup	1	6.29
California	Runtime	6m 12s	1m 47s
	# Nodes	4	1
	Speedup	1	13.90

[1]: <https://epihiper.readthedocs.io/en/latest/examples/examples.html>



What makes the **ESCAPE** simulations faster?



Design: Compiled vs
interpreted code



Data structure design: data
oriented vs. object oriented

Compiled vs Interpreted code

Fewer instructions to execute

- How to get susceptibility for a node?
- EpiHiper approach:
 - build up data structure
 - query at runtime
- ESCAPE approach
 - Convert to code
 - Call at runtime

```
CModel {  
    ...  
    vector<CHealthState> mStates;  
    map<string, CHealthState *> mId2State;  
    ...  
}
```

```
float_type _c1__susceptibility(node_index_type _v) {  
    if (NODE_TABLE->_c1_state[_v] == _S) {  
        return 1.0;  
    }  
    else {  
        return 0.0;  
    }  
}
```

Compiled vs Interpreted code

Compile time optimizations

```
transition
  E -> Ipresymp, p = p_E_Ipresymp, dwell = fixed3
  E -> Iasymp, p = p_E_Iasymp, dwell = normal5
  Ipresymp -> Isymp, dwell = fixed2
  Isymp -> R, dwell = isymp_r_dwell
  Iasymp -> R, dwell = normal5
end
```

State transitions

- Only state "E" has multiple exit states,
- This known at compile time
- No need to go through/generate code paths that test/check if state has multiple exit states

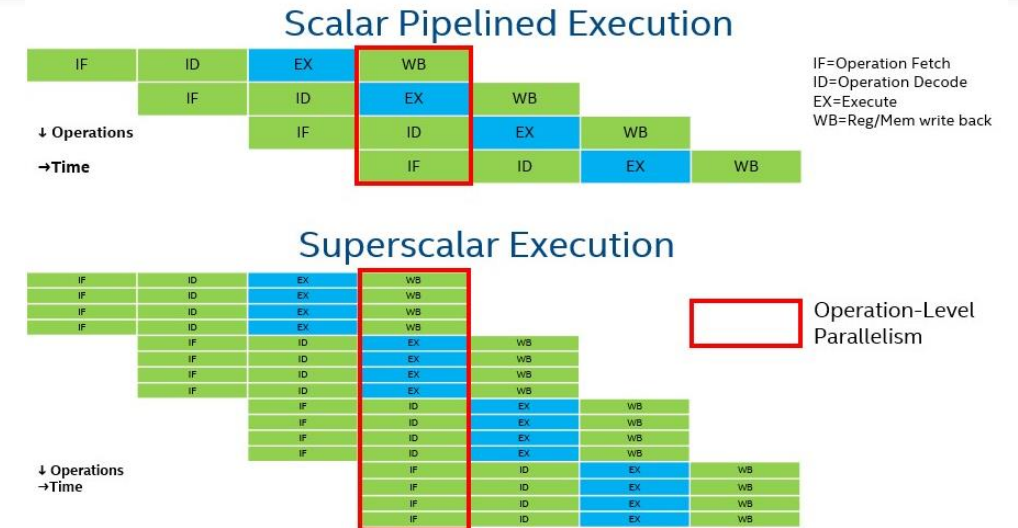
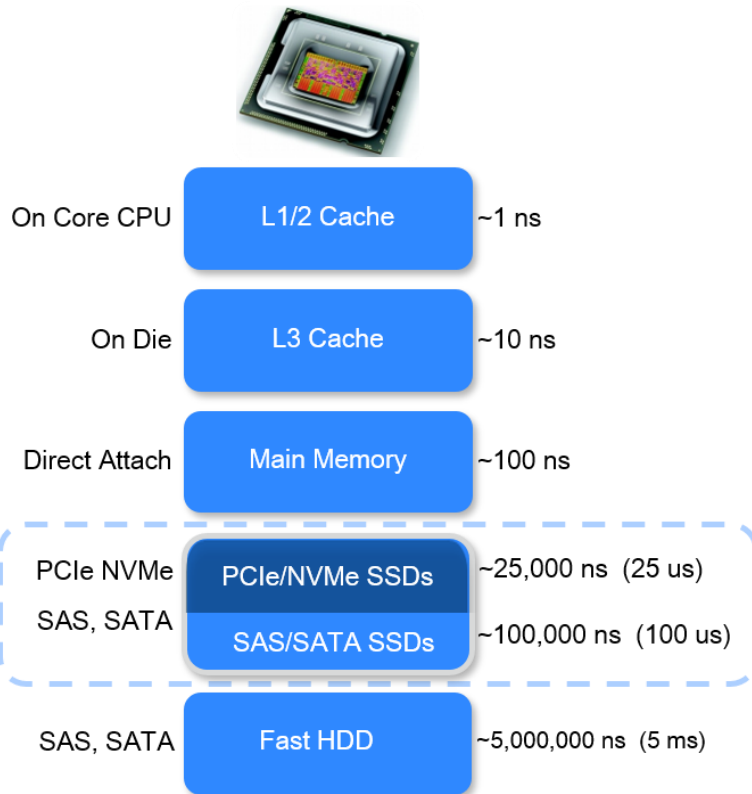
```
switch (state) {
case _Ipresymp:
  next_state = _Isymp;
  dwell_time = _fixed2();
  break;
case _Isymp:
  next_state = _R;
  dwell_time = _isymp_r_dwell();
  break;
case _Iasymp:
  next_state = _R;
  dwell_time = _normal5();
  break;
case _E:
  {
    float_type cum_probs[2] = {0};
    cum_probs[0] = _p_E_Ipresymp(v);
    cum_probs[1] = _p_E_Iasymp(v);
    cum_probs[1] += cum_probs[1 - 1];

    auto p = uniform01(*THREAD_RND_STATE);
    p *= cum_probs[2 - 1];

    if (p <= cum_probs[0]) {
      next_state = _Ipresymp;
      dwell_time = _fixed3();
      break;
    }
    if (p <= cum_probs[1]) {
      next_state = _Iasymp;
      dwell_time = _normal5();
      break;
    }
  }
}
```

Generated C++ code

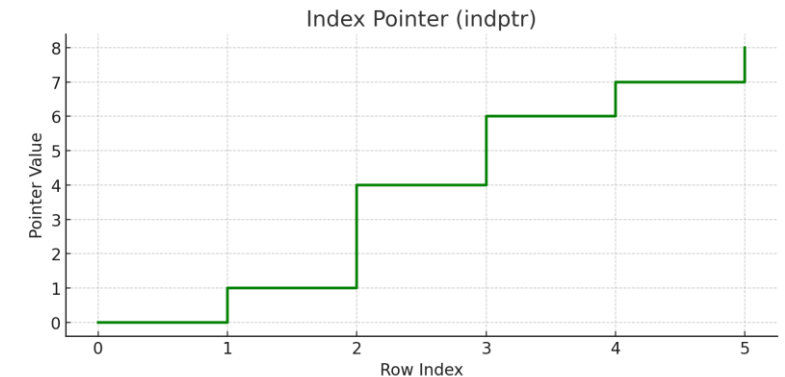
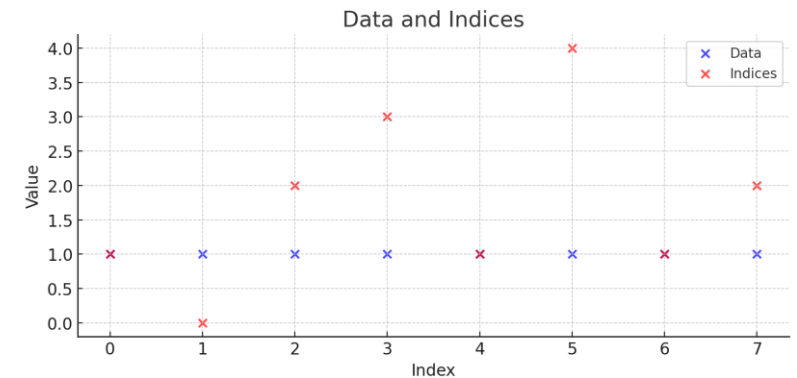
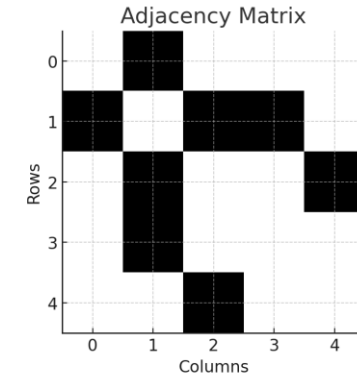
Data oriented vs. object-oriented design: Background



- Typical numbers
 - 1 CPU instruction: 0.25-0.5ns
 - Branch misprediction: 5ns
 - L1 cache reference: 0.5-1ns
 - L2 cache reference: 4ns
 - Main memory reference: 100ns

Data oriented vs. object-oriented design: Graph structure memory layout

- Graph data structure
 - Sparse incidence matrix
 - (<node index>, <incoming edge index>)
- Node attributes table
 - Column oriented (struct of arrays)
 - Sorted by Node ID
- Edge attributes table
 - Column oriented (struct of arrays)
 - Sorted by (target node ID, source node ID)

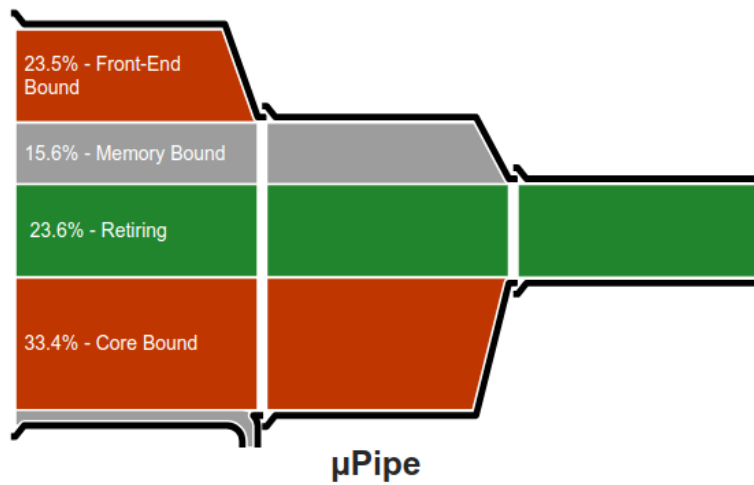


VTune: Microarchitecture exploration**

Test case: VA Montgomery; Test system: PB's Laptop; Test Scenario: Example 1 (C)

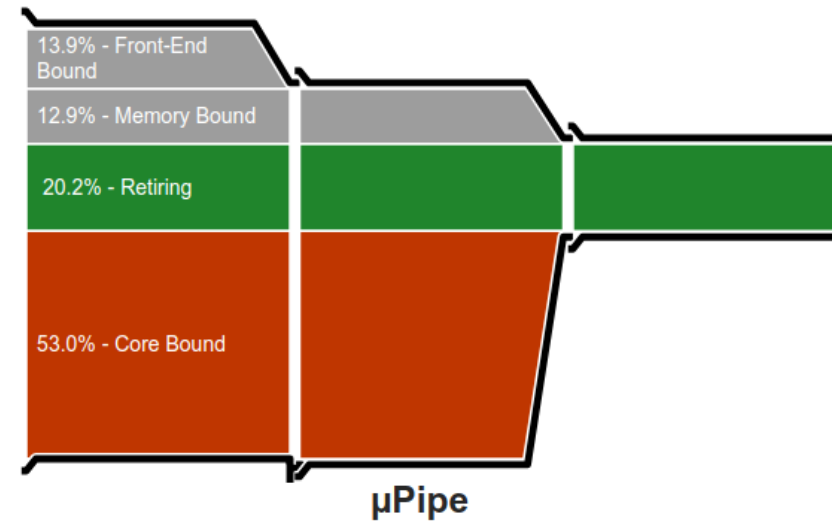
EpiHiper

- Larger frontend bound section
 - CPU waiting for instructions
 - Branch mispredictions
 - Function pointer jumps
- Slightly larger memory bound section
 - CPU waiting for data



ESCAPE

- Smaller frontend bound section
- Slightly smaller memory bound section
- Larger core bound section
 - Limited by CPU power



ESL: A DSL for writing ESCAPE simulations

- Syntax inspired by Python/Julia
- Domain specific constructs
 - Node/edge table
 - Contagion
 - Transitions/Transmissions
 - Dwell time distributions
 - Transition probability functions
 - Node/edge sets
- Serial programming constructs
 - Scalar variables: local/global (int/float/bool/...)
 - Custom enumerated types
 - Conditionals: if/elif/else
 - Loops: while/for*
 - Expressions / functions
- Parallel constructs
 - Select, Apply and Reduce

ESL: A DSL for writing ESCAPE simulations

- ESL compiles down to C++/CUDA*
- Compiler written in Python
 - Current parser uses TreeSitter
 - Supports GLR grammar
 - Grammar written in Javascript
 - Grammar compiles down to C
- 3 phase compiler
 - Source -> Parse tree
 - Parse tree -> AST
 - AST -> Generated code
- Editor support
 - Syntax highlighting
 - TreeSitter: Vim / Emacs
 - Lezer: Jupyter
 - Custom Language Server
 - Currently only error checks
 - Code completion
 - Future: code formatting
 - Vim, Emacs, VScode, Jupyter

Limitations and Future Work

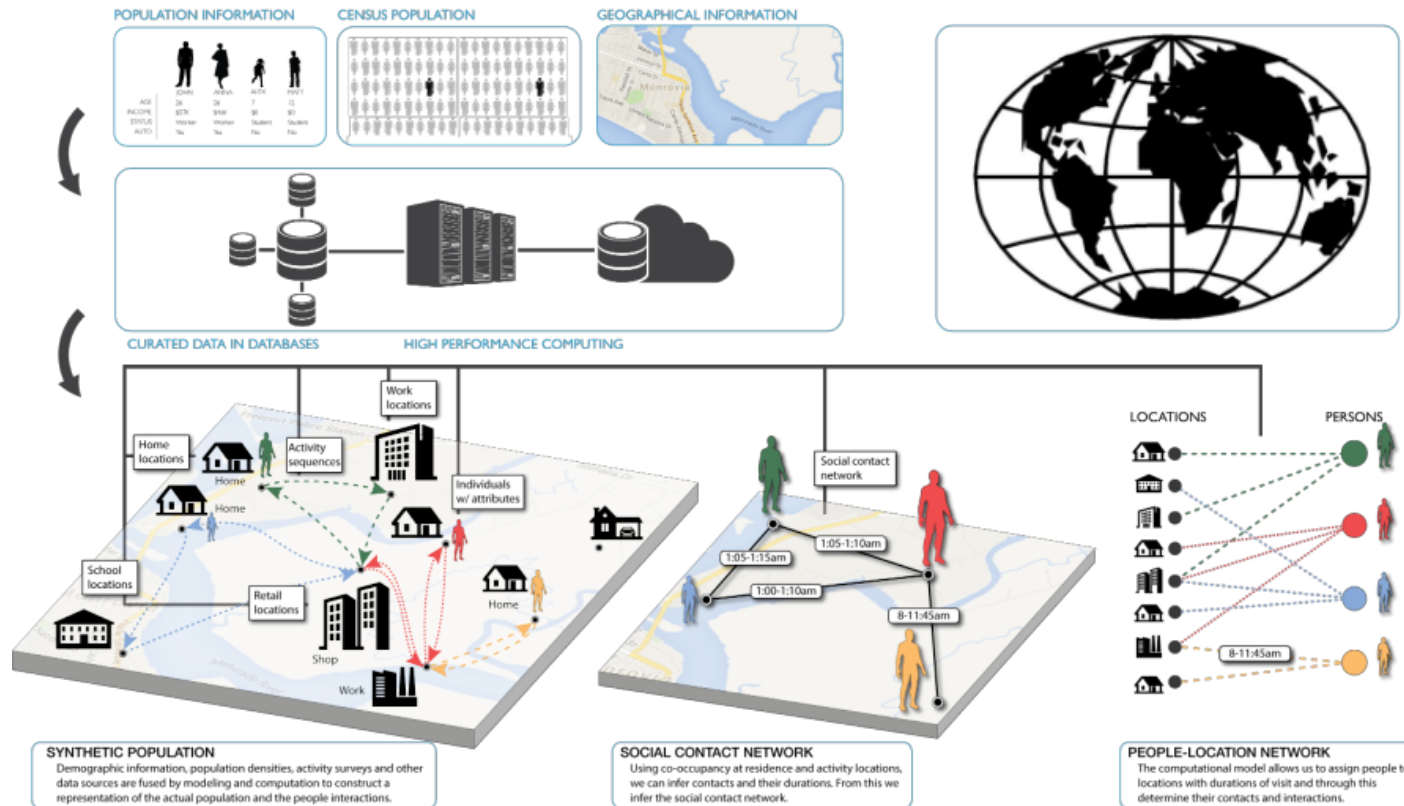
- Can work with only a finite number of contagions
- Supports only a single (possibly time-varying) contact network
- Doesn't natively support vector borne diseases
- CUDA backend is work in progress
- A distributed memory (MPI) backend implementation

`github.com/NSSAC/escape-abm`

`parantapa@virginia.edu`

Thank you!

Digital Similar of Populations



- ESCAPE simulators require a population to run contagions on
- The "contact network" network needs to be provided to it as an input.
- The Biocomplexity Institute maintains detailed digital similars of USA and many other countries.