SIMULATION OF VIRAL INFECTION PROPAGATION THROUGH AIR TRAVEL

Ashok Srinivasan, University of West Florida

Collaborators

Sirish Namilae, Anuj Mubayi, Matthew Scotch, Robert Pahle, and C.D. Sudheer

We use Blue Waters to **analyze risk of infection spread due to movement of passengers during air travel**

VIPRA

VIRAL INFECTION PROPAGATION THROUGH AIR-TRAVEL

www.cs.fsu.edu/vipra

OUTLINE

- Introduction
- Modeling Passenger Movement
- Performance Optimizations
- Modeling Infection Spread
- Conclusions

INTRODUCTION

MOTIVATION

- Air travel is an important factor in infection spread
- There had been calls to ban flights from Ebola infected areas
  - This can have large human and economic impact
  - Fine-tuned policy prescriptions can be as effective
    - Reassures the public that action is being taken
    - Avoids negative human and economic impacts

PROJECT GOALS

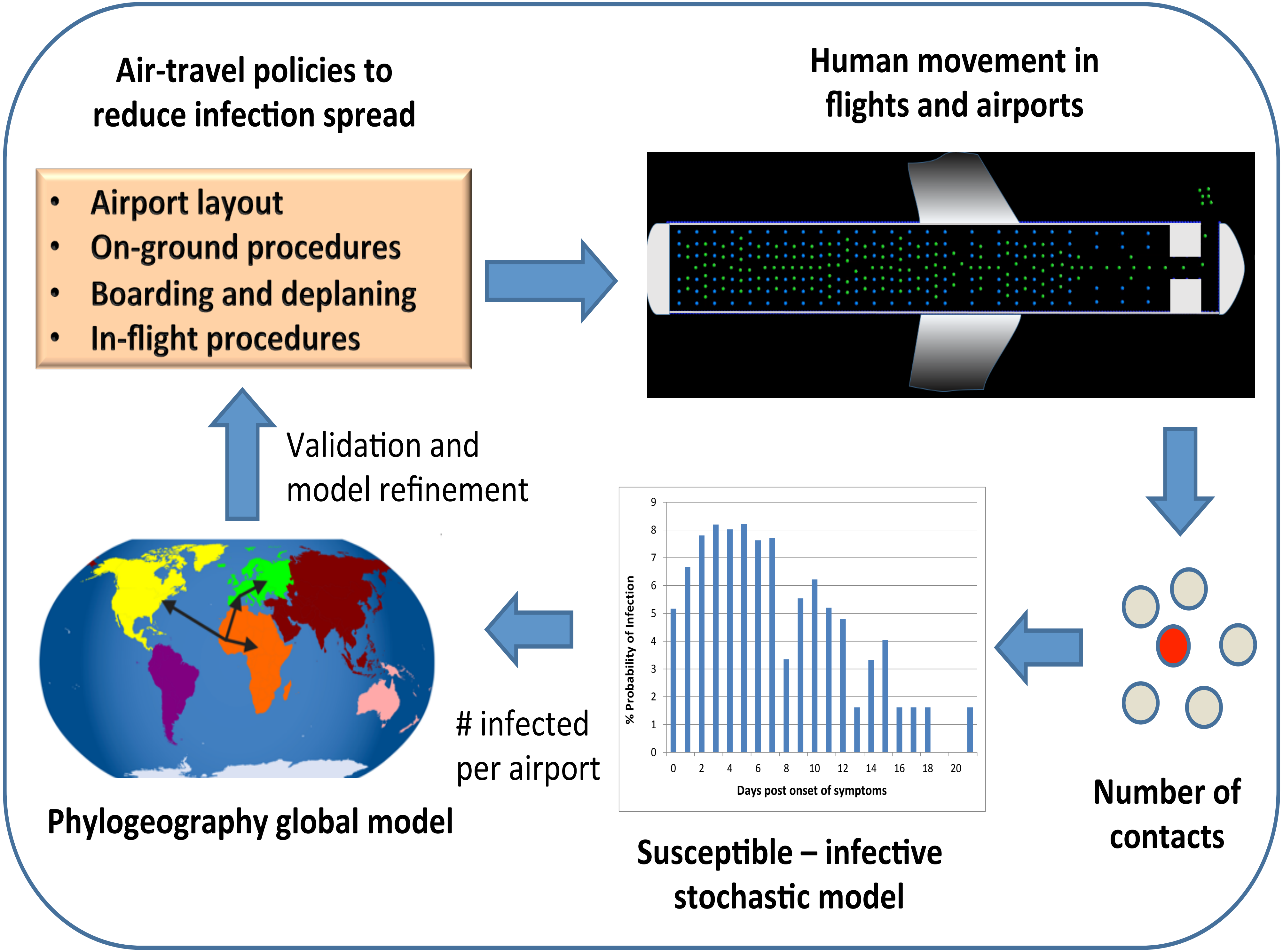
- Analyze the impact of different policies on spread of diseases through air-travel
  - Example: Different boarding procedures
- **Why it matters**
  - Provides insight to decision makers on policy or procedural choices that can reduce risk of infection spread without disrupting air travel

CURRENT MODELS

- Typically focused on scientific understanding, rather than policy analysis
  - Predictions are difficulty due to inherent uncertainties
- Usually at an aggregate level, which makes evaluation of impact of new policies difficult
- Example: Inaccurate predictions on Ebola
  - Predicted millions infected by early 2015 and hundreds of thousands dead

OUR MODELING APPROACH

- Use fine-scale model of human movement in planes to determine response to policies
- Parameterize sources of uncertainty
  - A parameter sweep over this space generates feasible scenarios
  - **Key challenge**
    - Large parameter space leads to high computational cost
  - **Why Blue Waters**
    - It provides the computational power to produce solutions in a national emergency



Air-travel policies to reduce infection spread
Airport layout
On-ground procedures
Boarding and deplaning
In-flight procedures
Human movement in flights and airports
Validation and model refinement
Phylogeography global model
infected per airport
% Probability of Infection
0
1
2
3
4
5
6
7
8
9
0
2
4
6
8
10
12
14
16
18
20
Days post onset of symptoms
Susceptible – infective stochastic model
Number of contacts

QUESTIONS ANSWERED

- Can simple policies reduce infection risk without causing major disruptions?
  - Change plane type
  - Change boarding and disembarkation procedures
  - Change airport layout and procedures
- **Broader impacts**

Come flu with me

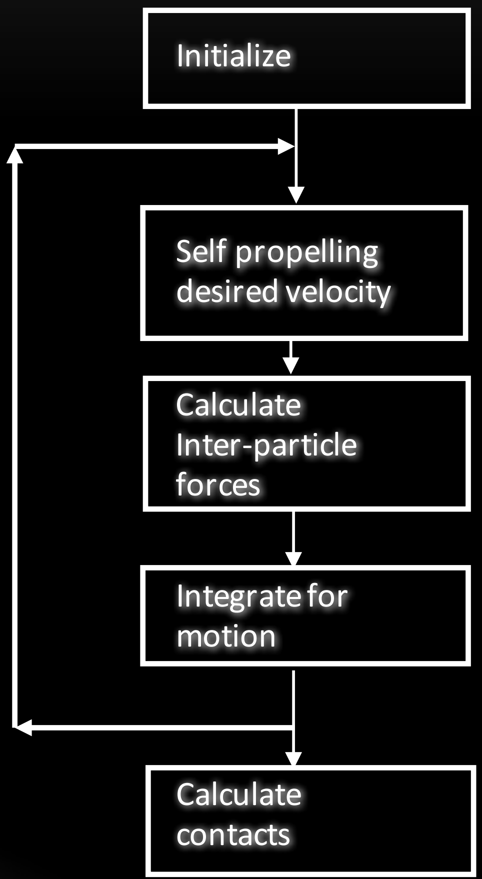
The way airlines board planes affects how easily bugs are spread among passengers

The way we board planes could actually be spreading diseases

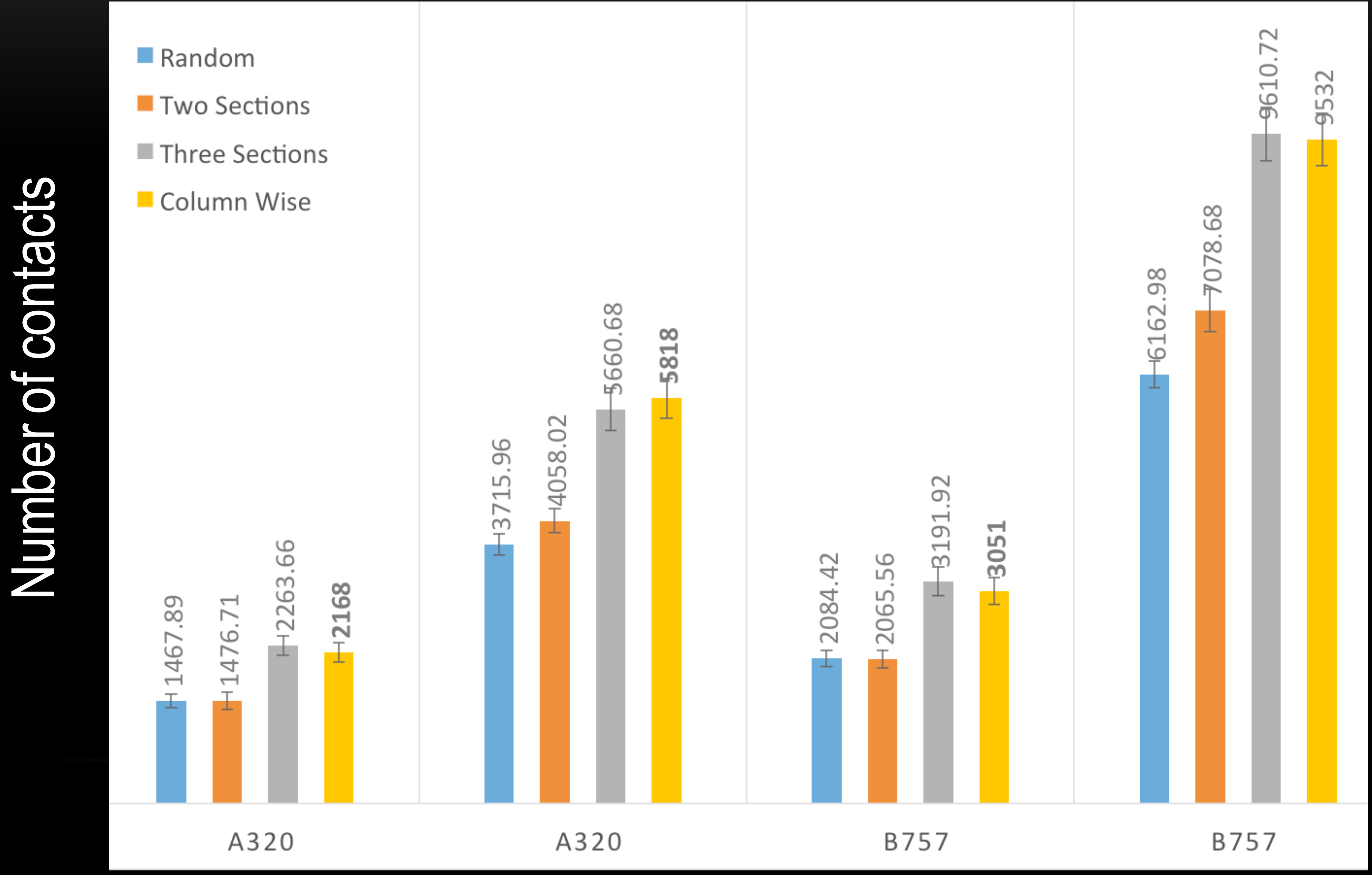
MODELING PASSENGER MOVEMENT

SELF PROPELLED ENTITY DYNAMICS MODEL

- Social dynamics is motivated by Molecular Dynamics, and treats entities as particles
  - Individuals experience self propulsion that induces them to move toward their desired goal
  - They experience repulsive forces from other persons and surfaces
- We add human behavioral characteristics to social dynamics
- Parameterize the **sources of uncertainty** and carry out a parameter sweep to identify their robustness under a variety of possible scenarios

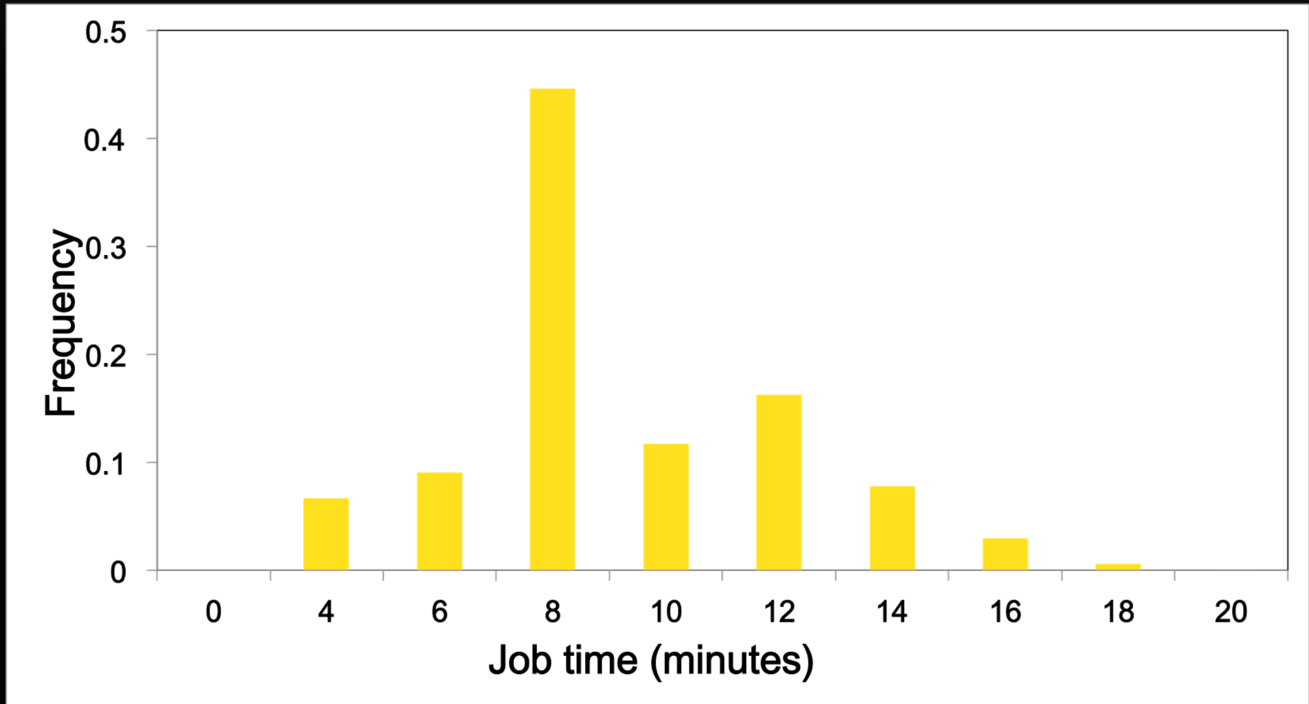


BOARDING STRATEGIES



PERFORMANCE OPTIMIZATIONS

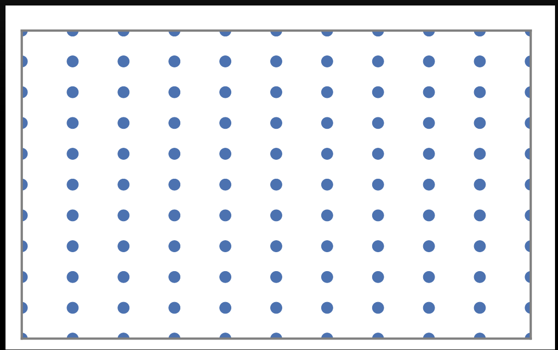
CONVENTIONAL OPTIMIZATIONS



Parallel parameter sweep with ~68K combinations

- **Blue Waters team** helped reduce parallel IO bottleneck, leading to a factor two performance gain

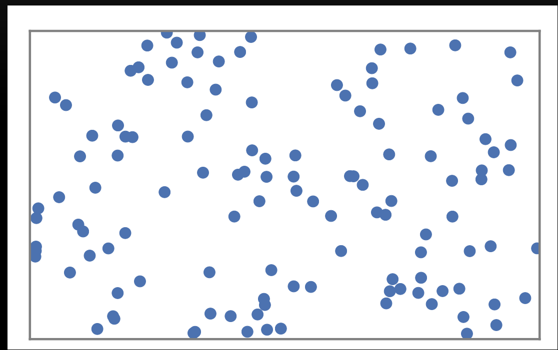
TYPES OF PARAMETER SWEEP



2D Lattice

Parameter space coverage: **inefficient**

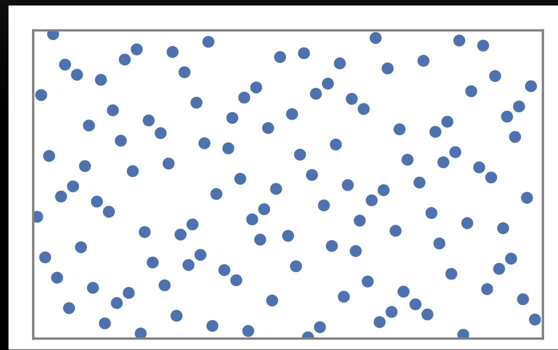
Convergence check: **inefficient**
Factor 2^d gap between convergence checks



2D Random

Parameter space coverage: **inefficient**

Convergence check: **efficient**
Factor 2 gap between convergence checks



2D LDS

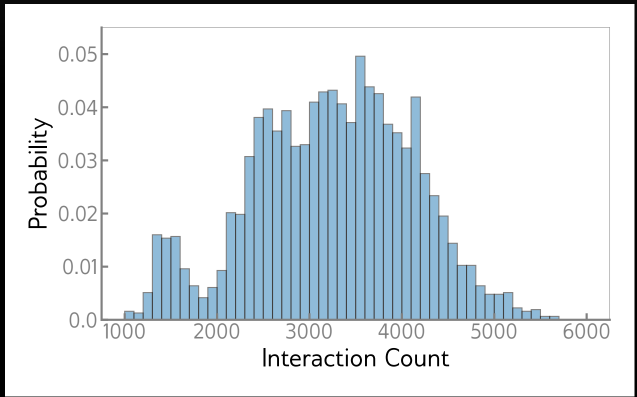
Parameter space coverage: **efficient**

Convergence check: **efficient**
Factor 2 gap between convergence checks

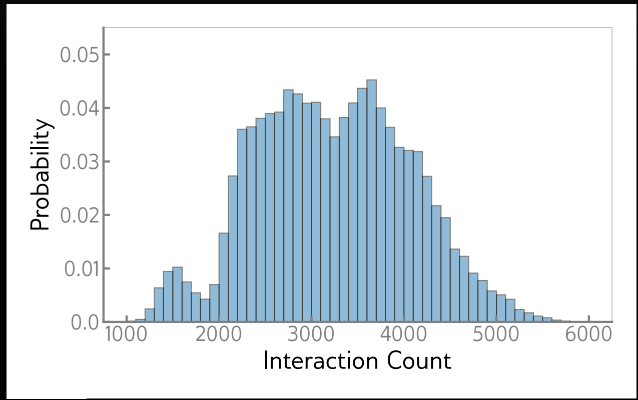
SPED model in this part of our study uses 5 parameters

- 5-D Lattice and 5-D Scrambled Halton Low Discrepancy Sequence (LDS) parameter sweeps used for infection spread analysis

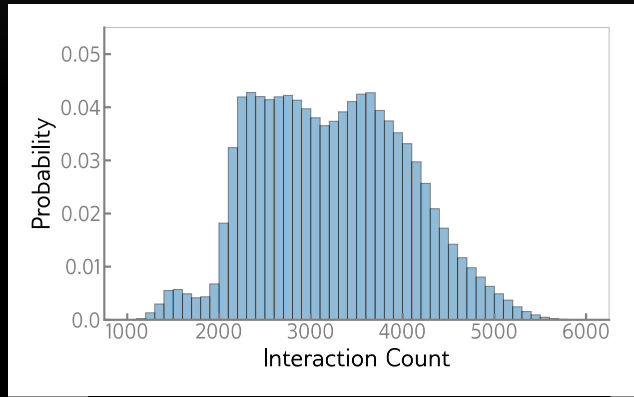
CONVERGENCE FOR LATTICE SWEEP



Histogram for subgrid of size 5^5

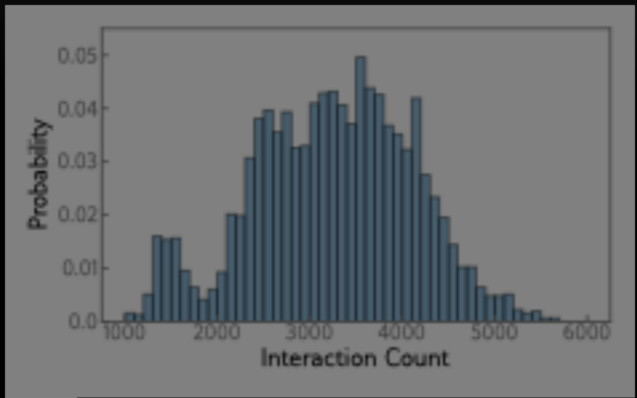


Histogram for subgrid of size 9^5

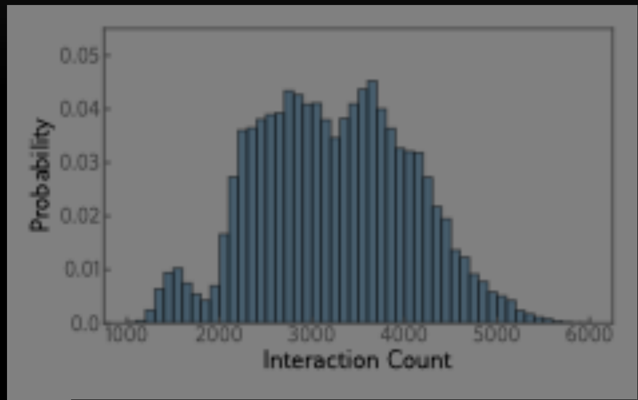


Histogram for 17^5 grid

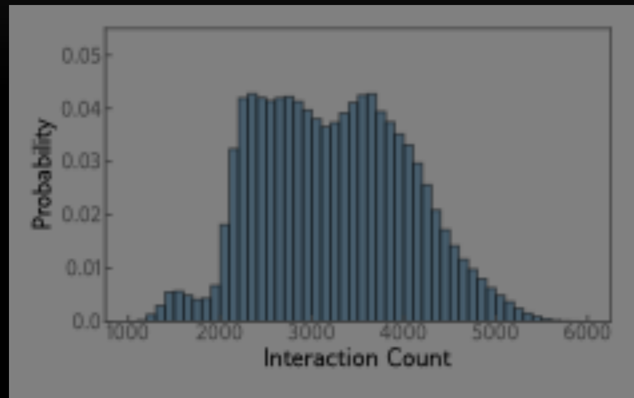
CONVERGENCE FOR LDS SWEEP



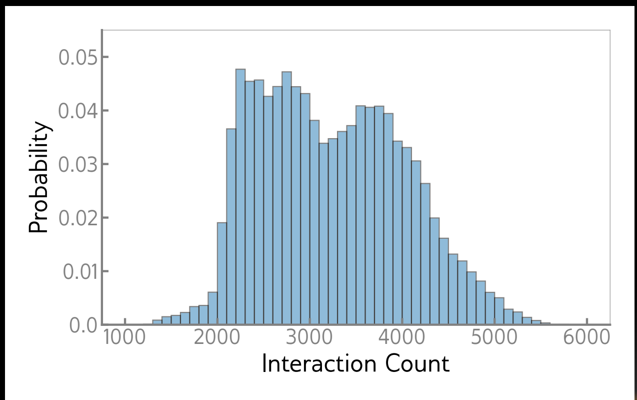
Histogram for subgrid of size 5^5



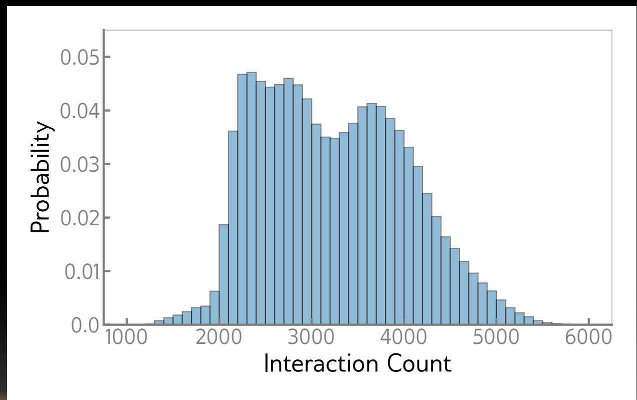
Histogram for subgrid of size 9^5



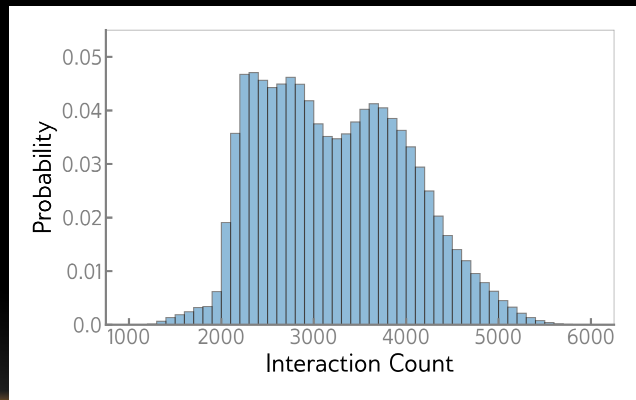
Histogram for 17^5 grid



Histogram for 32768 (2^{15}) points



Histogram for 262144 (2^{18}) points



Histogram for 17^5 points

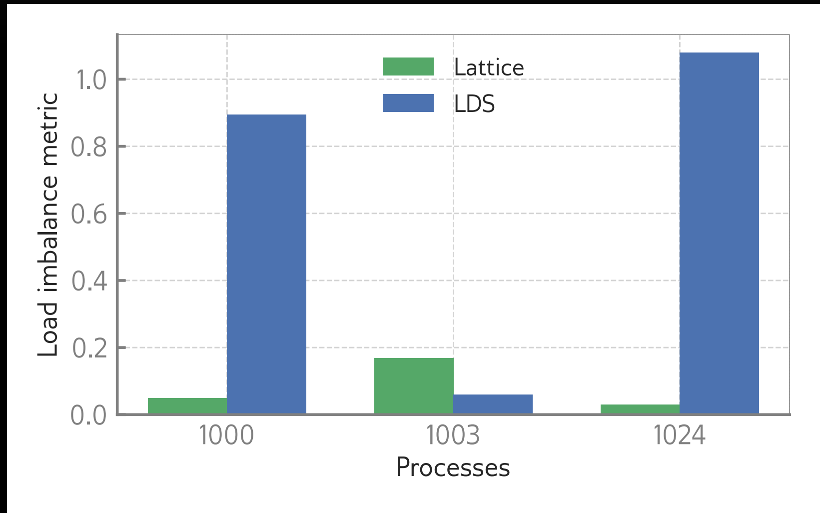
LOAD IMBALANCE IN LATTICE VS. LDS SWEEPS

Load imbalance across processes is defined as

$$|MaximumLoad - AverageLoad| / AverageLoad$$

0 when load is perfectly balanced

- Lattice sweep is well balanced
- LDS has a poor balance with 1000 and 1024 processes
- LDS performs better than Lattice for 1003 processes
    - 1003 is divisible by 17 (parameter values)

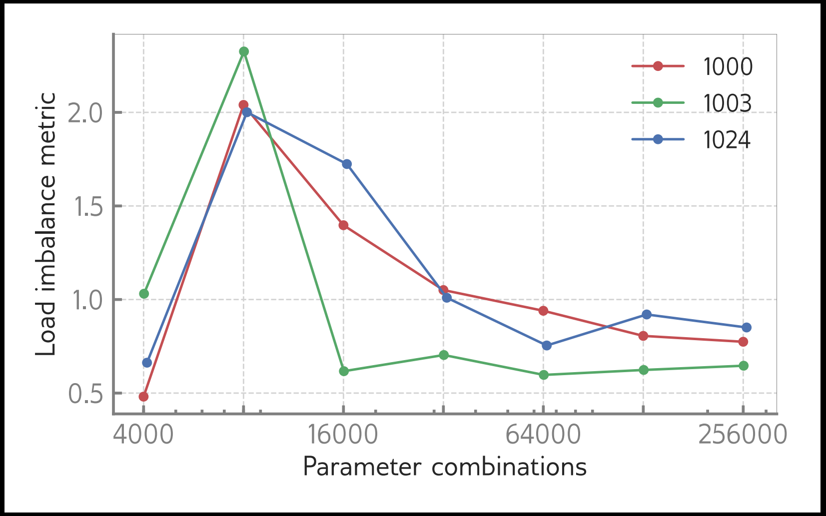


Load imbalance for Lattice and LDS sweep of the **entire data** set 17^5 (without convergence checks) using cyclic distribution

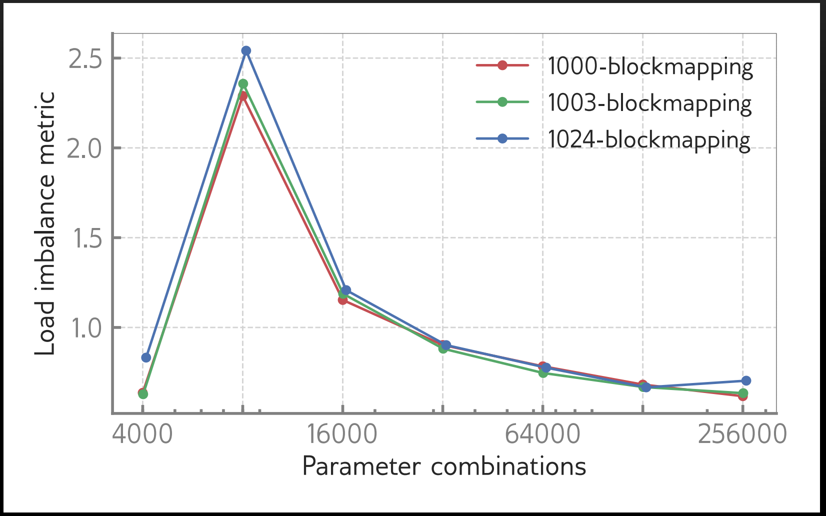
1000 and 1024 are products of primes used in the LDS

LOAD BALANCING LDS

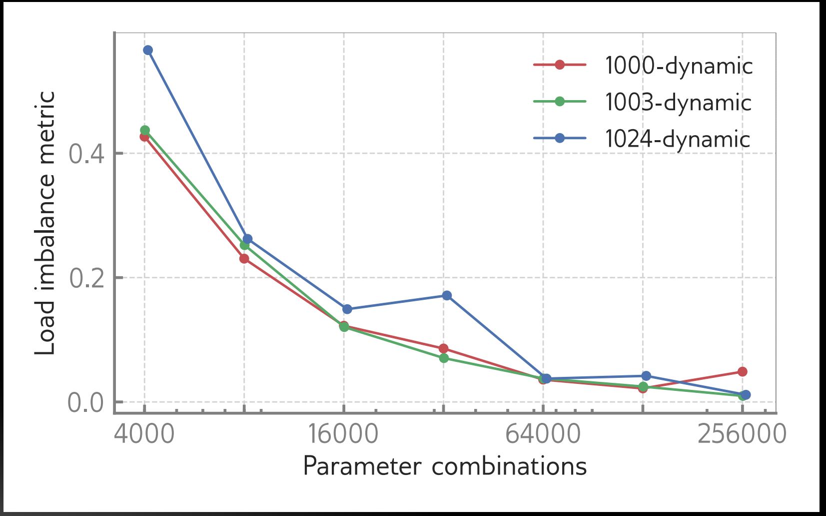
With convergence checks



Cyclic Distribution



Block Distribution

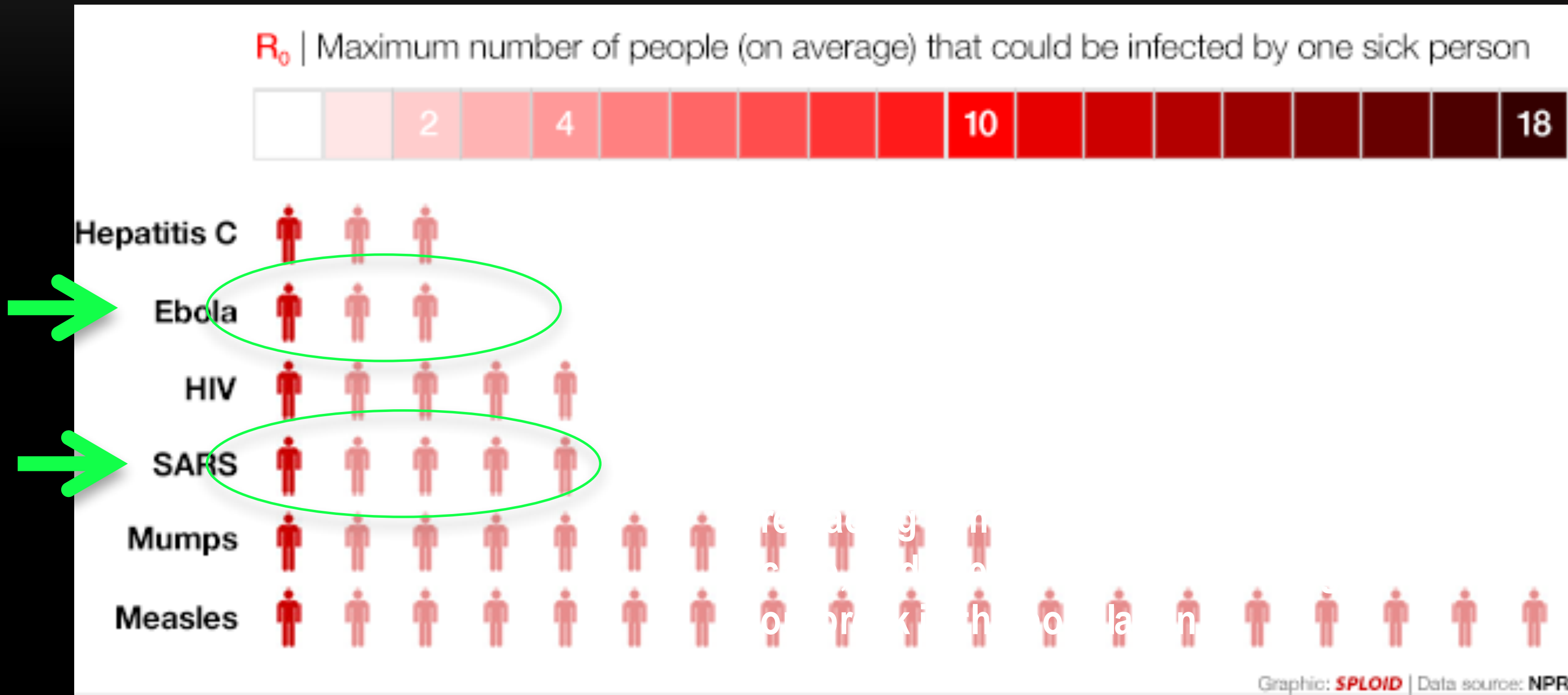


Dynamic Load balancing

- Cyclic: Load is not well balanced in the initial stages even with 1003 processes
- Block: Does not work well for small number of samples
- Dynamic: Master-worker based dynamic load balancing works best overall but is not scalable

MODELING INFECTION SPREAD

INFECTION TRANSMISSION

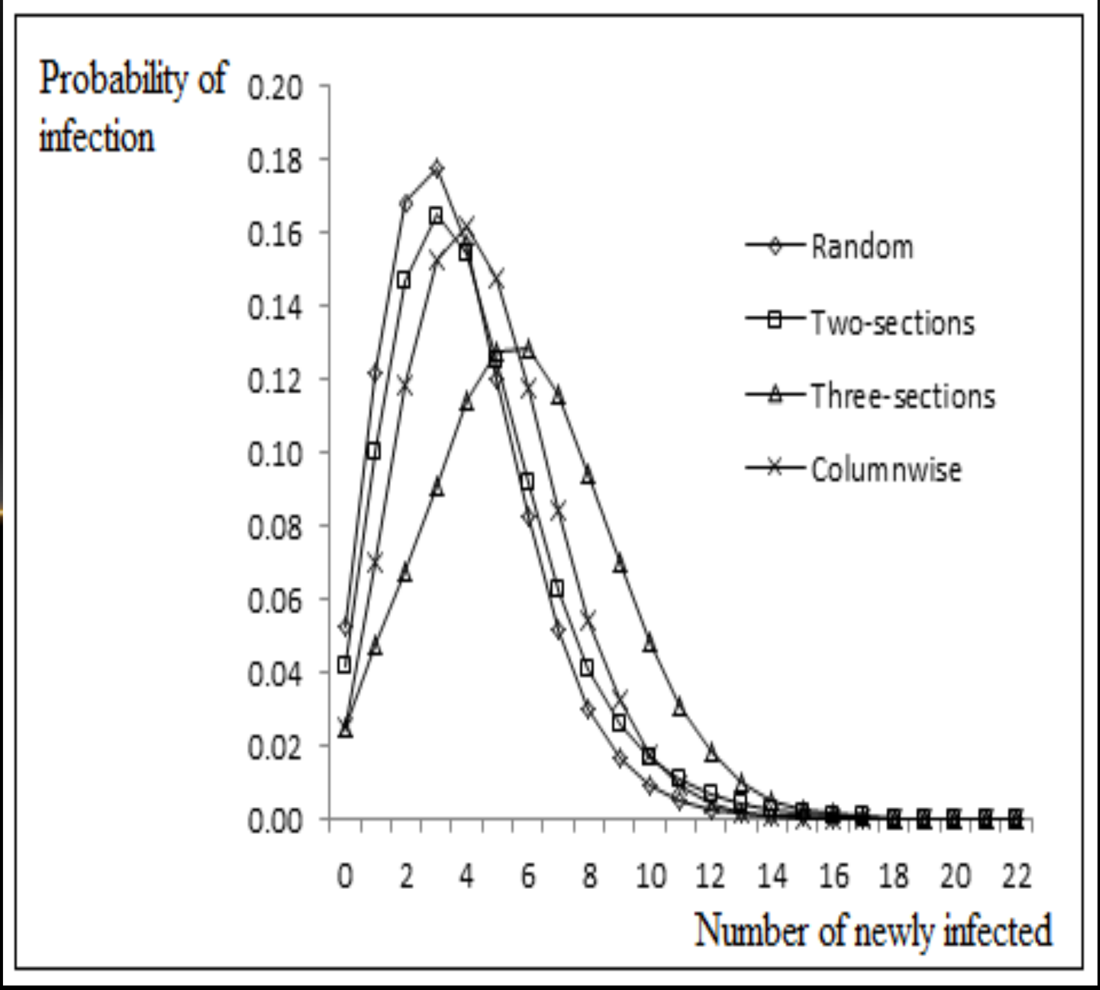


http://sploid.gizmodo.com/ebola-spreading-rate-compared-to-other-diseases-visuali-1642364575

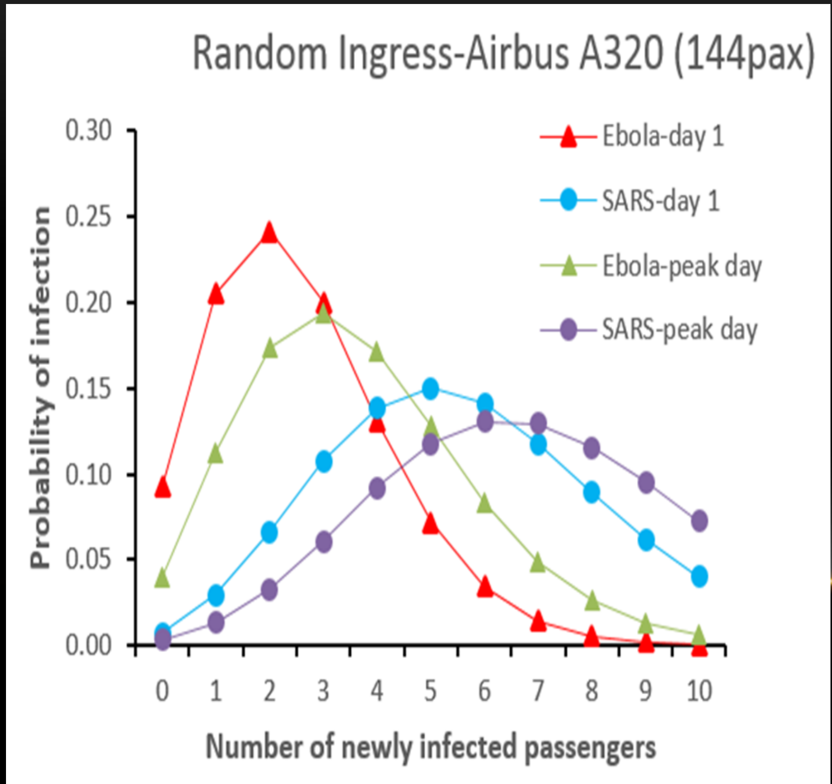
- Probability of infection transmission modeled as a function of distance to infected person, exposure time, and infectivity

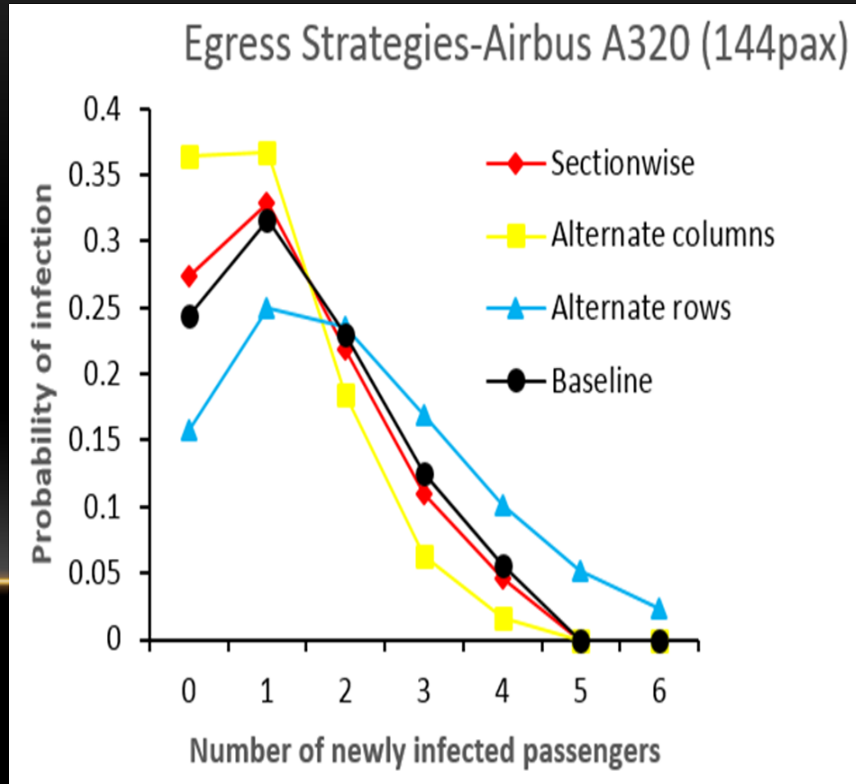
IMPACT OF BOARDING STRATEGIES

- Boarding Boeing 757-200
  - One passenger at the first day of infection
  - Infection probability = 0.06
  - Contact radius = 1.2 m
- Strategies that prevent clustering in the cabin reduce infection likelihood



LONG VS SHORT CONTACT RADIUS





- Infection contact radius
  - Ebola: 1.2 m
  - SARS: 2.1 m
- Model includes airport gates

CONCLUSIONS

COMPUTATIONAL OPTIMIZATIONS

https://www.cs.fsu.edu/vipra

- Parameter sweep with LDS is more efficient than with lattice
  - Better coverage of parameter sweep and faster convergence
  - It made feasible analysis that was not feasible earlier
- Load imbalance is a potential problem with LDS and is related to its number-theoretic properties
  - Identified techniques, that can lead to good load balancing under different applications scenarios

SUMMARY OF APPLICATION RESULTS

- Identified procedures that can lead to decrease in contacts
  - Random boarding leads to lower risk of infection spread
  - Boarding has a higher impact than deplaning
  - Smaller planes are better than larger ones
- Use of better procedures and smaller planes could have reduced Ebola risk by 87% without travel restrictions

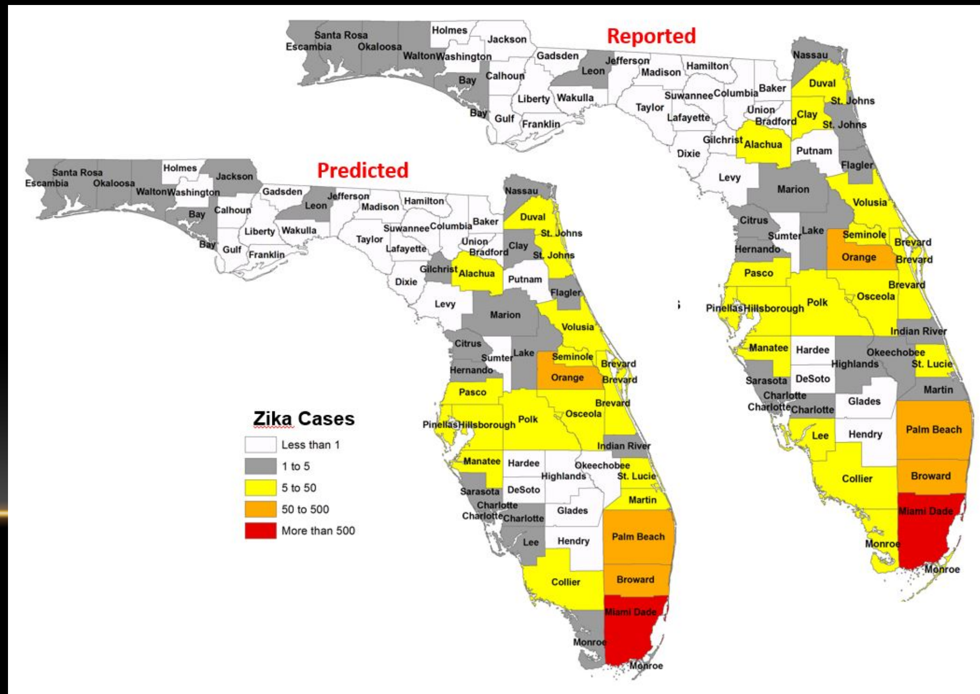
This material is based upon work supported by the National Science ACI under grants #1525061, #1524972, and #1525012 on **Simulation-Based Policy Analysis to Reduce Ebola Transmission Risk in Air Travel** and PRAC grant on **Petascale Simulation of Viral Infection Propagation through Air Travel.** Any opinion, findings, and conclusions or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of the National Science Foundation. We thank NCSA for providing use of the Blue Waters supercomputer.

FUTURE DIRECTIONS

- Extend this approach
  - Assimilate data into simulation model
  - Use domain adaptation to model related situations
  - Consider the consequences of air travel

Zika importation risk prediction

- Identify human mobility from social media data and link with metapopulation epidemic model
- Fine-grained results predict locations within Miami with granularity of the order of a square mile



www.cs.fsu.edu/vipra