

# What is Operations Research!?

(...or what i think about when i think about work)

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CURIE academy, July 2021

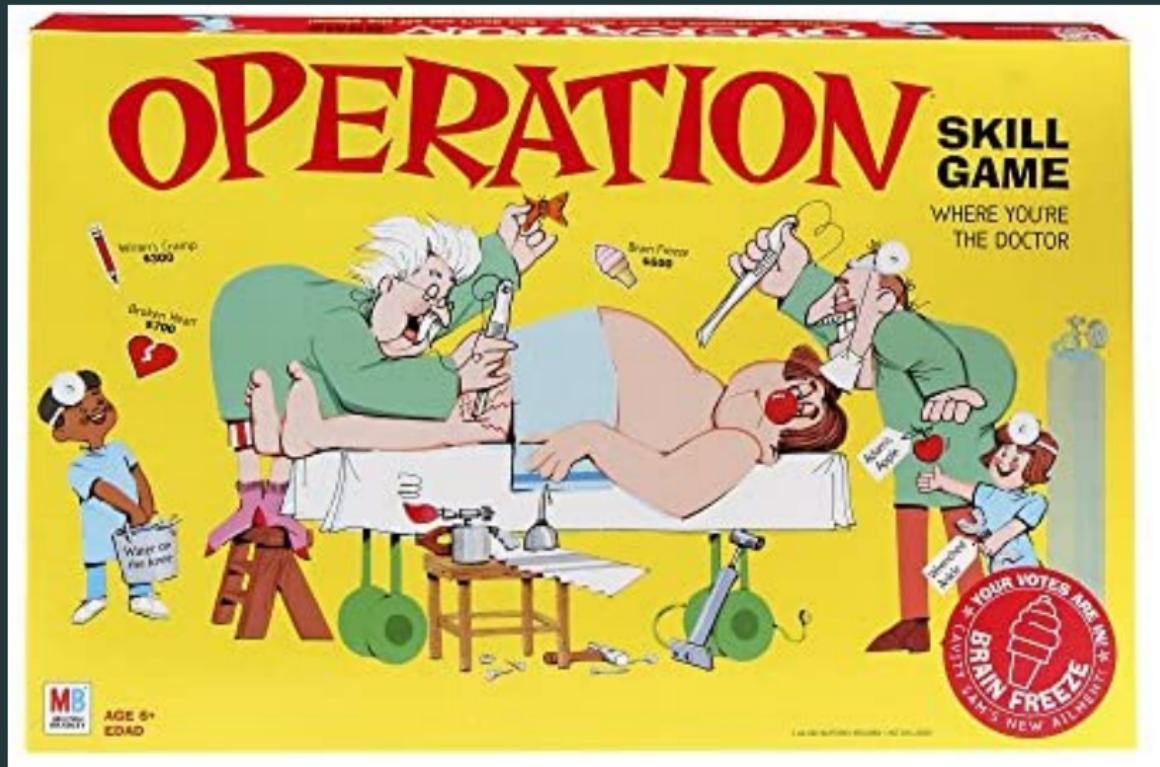
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- Aerospace engineering = stuff which **fly through air/in space**
- Civil engineering = **civilian** stuff (buildings, transportation, etc.)

by that logic...



## If not that, then what is Operations Research?

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**learn from data**

to

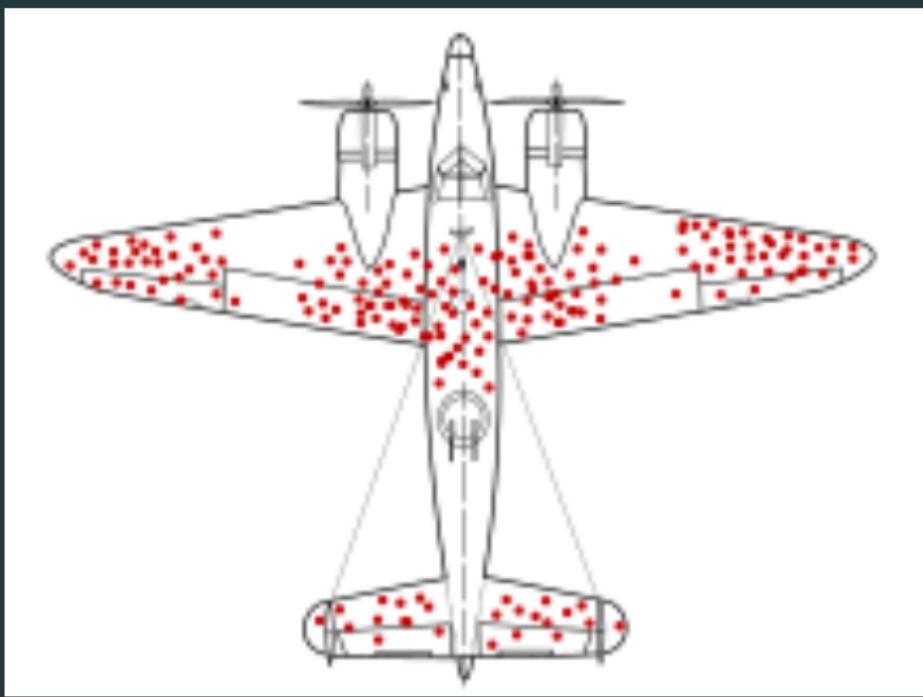
**build models**

for

**making better decisions**

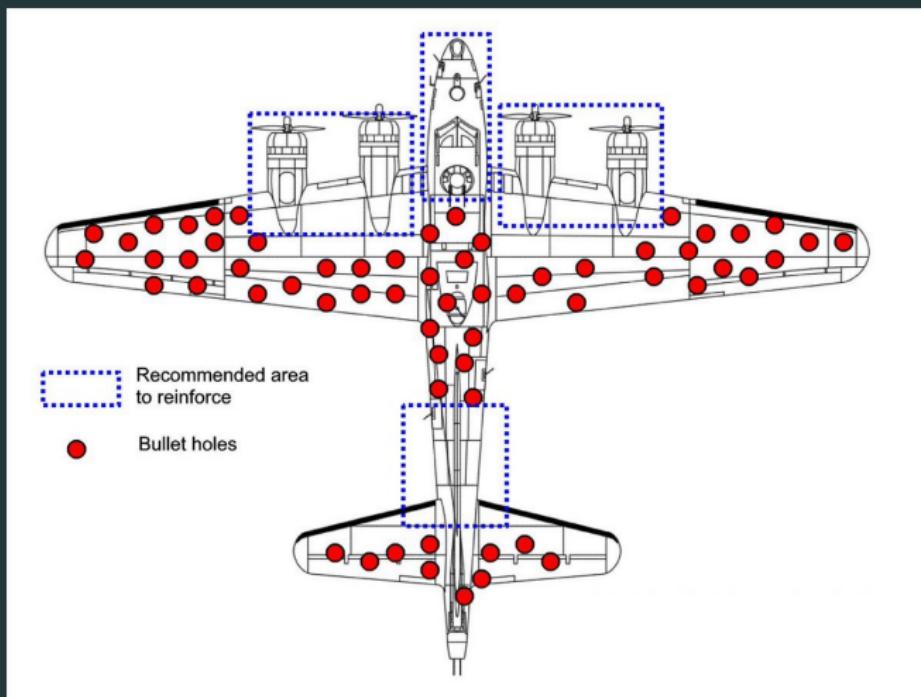
## An example (and some history)

the name comes from the original application in **military** operations



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## The OR mantra

learn from data      Statistics,  
ML, data mining, etc

to

build models      Simulation, applied  
Probability game theory, microecon,  
operations management

for

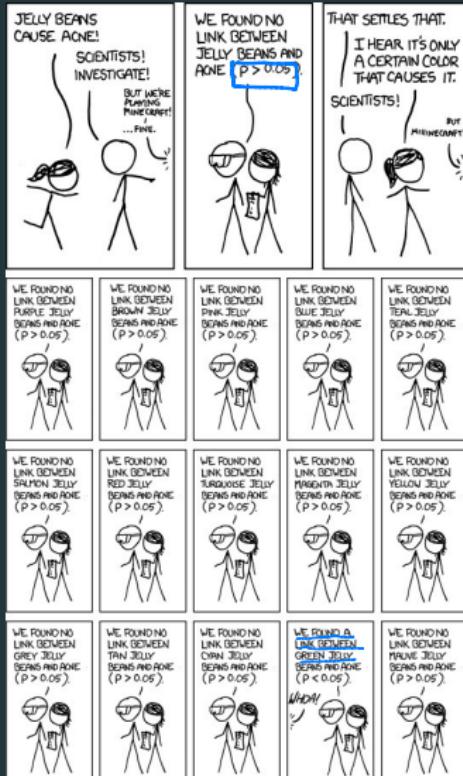
making better decisions

Optimization

- linear / integer programming
- 'stochastic' optim

**What does an Operations Researcher think  
about?**

# Understanding randomness and sampling



What are the fundamental misunderstandings here?

# Risk Analysis: Choosing Projects



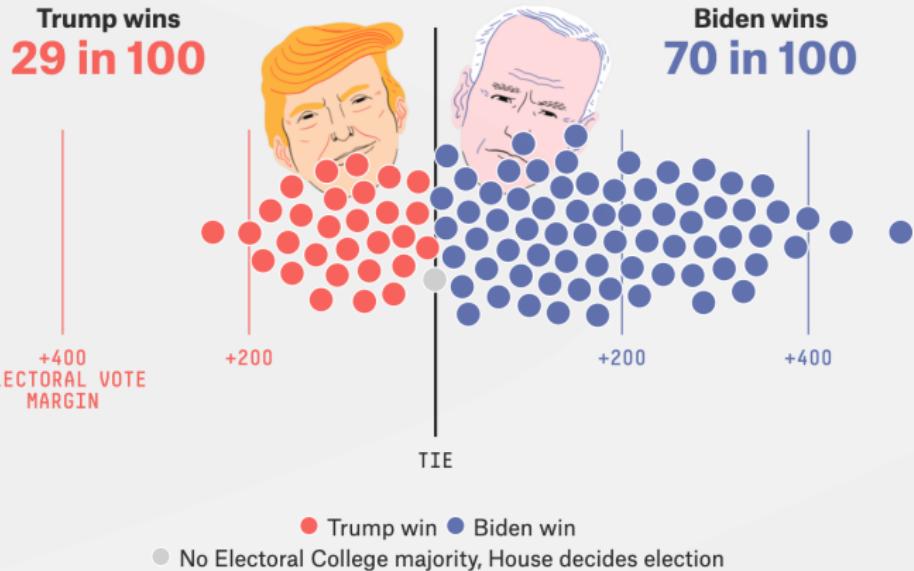
What are the fundamental misunderstandings here?

The project has a 70% chance of success. Why not try?

Why not increase the chance of a success by doing 10?

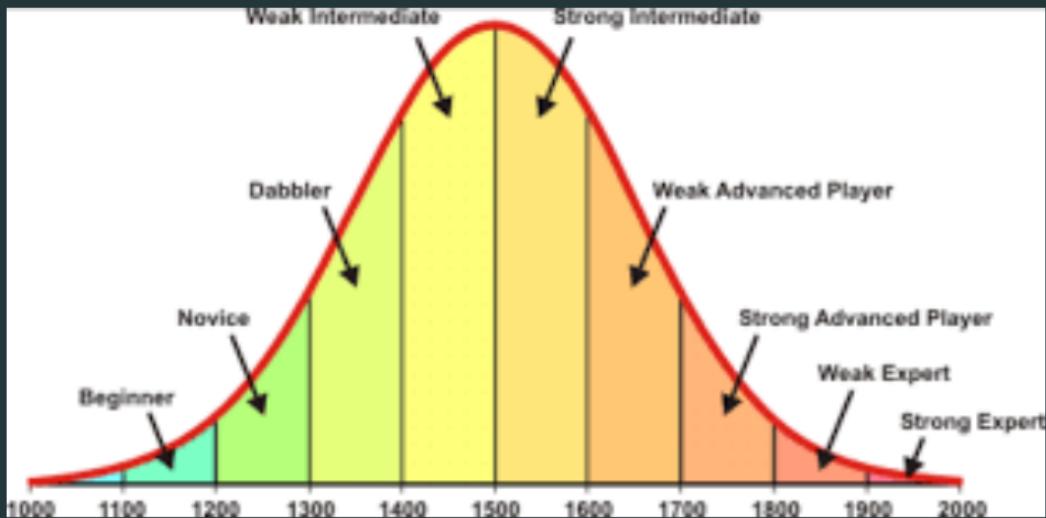
# Prediction

We simulate the election 40,000 times to see who wins most often. The sample of 100 outcomes below gives you a good idea of the range of scenarios our model thinks is possible.



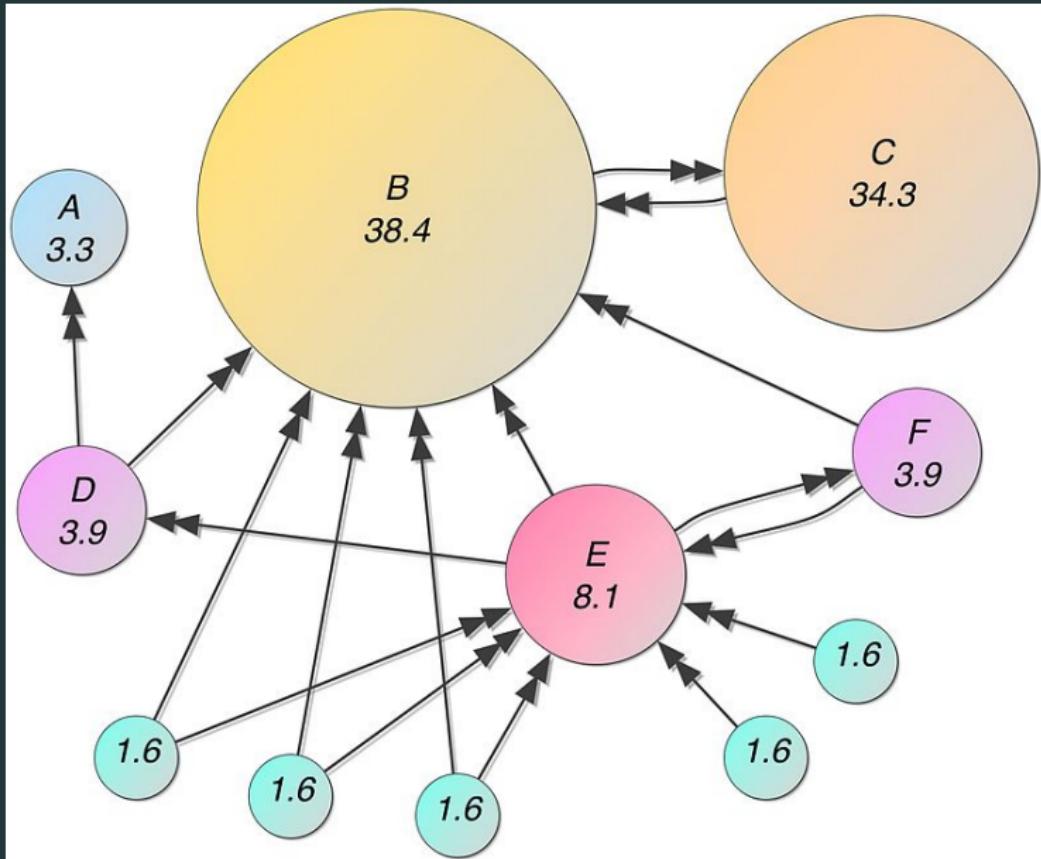
# Quantifying Quality

## ELO Rankings

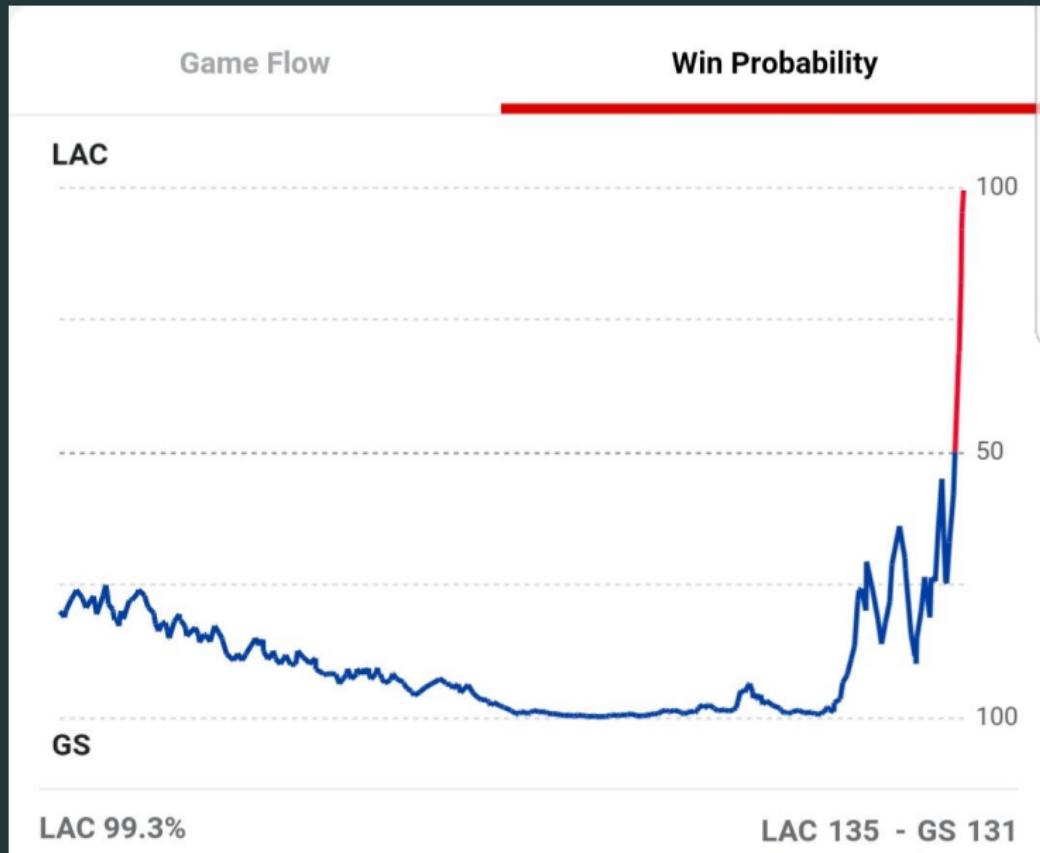


# Quantifying Utility

'PageRank'



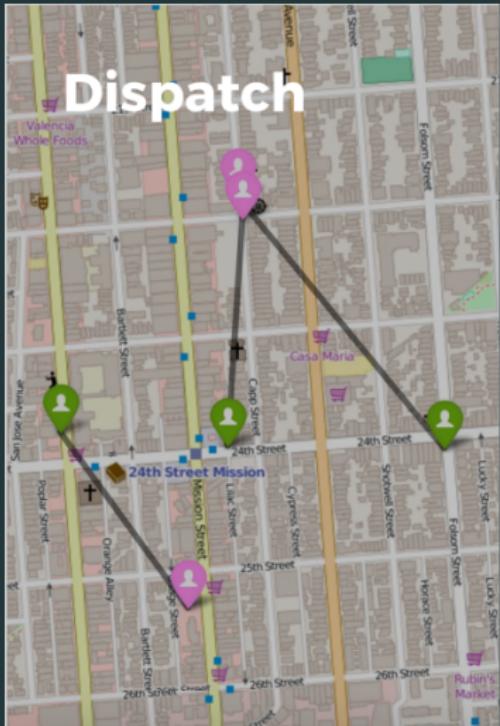
# Surprise!



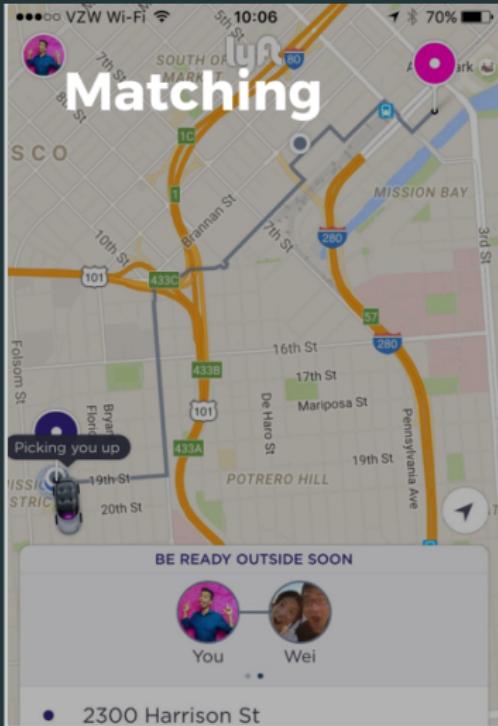
# Designing complex systems



# research in ridesharing: logistics



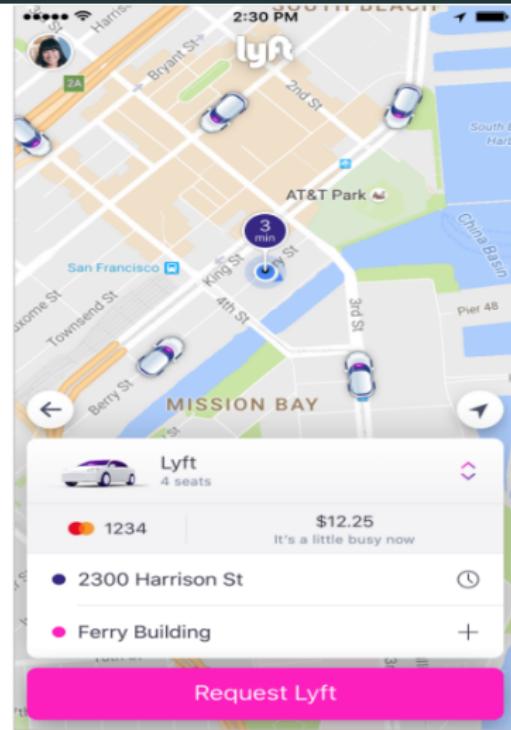
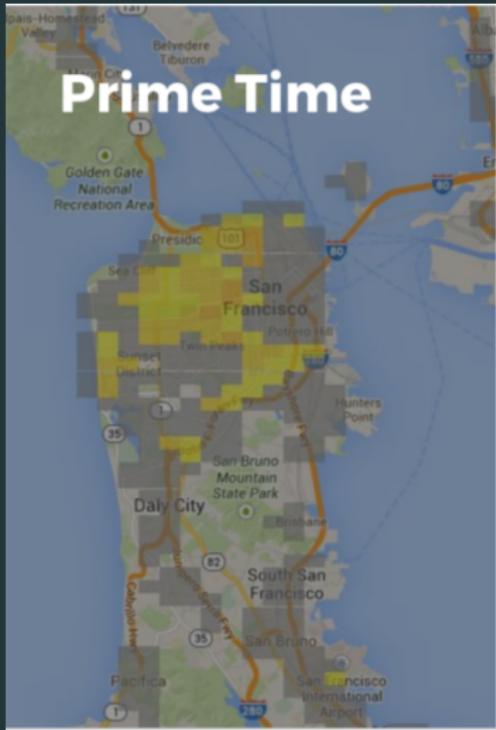
Dispatch



Matching

credit: lyft research science

# research in ridesharing: market design



credit: lyft research science

# Counterfactual Analysis

## COVID-19 Mathematical Modeling for Cornell's Fall Semester

PhD Students: J. Massey Cashore, Ning Duan, Alyf Janmohamed, Jiayue Wan, Yujia Zhang

Faculty: Shane Henderson, David Shmoys, Peter Frazier\*

June 15, 2020

### Executive Summary:

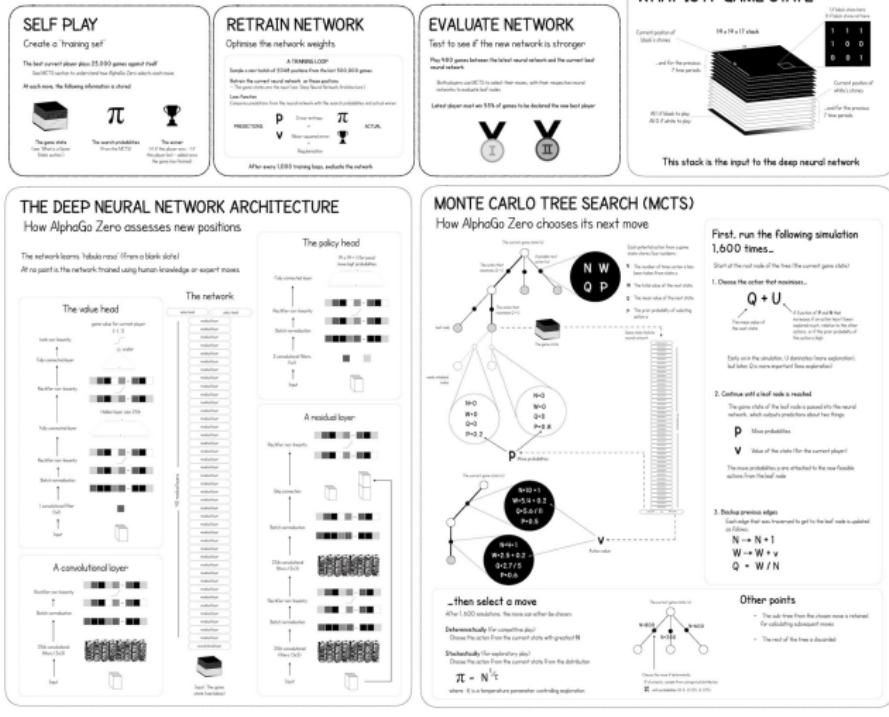
- Initial modeling results suggest that a combination of contact tracing, asymptomatic surveillance, and low initial prevalence (supported through testing students prior to, and upon, returning to campus) can achieve meaningful control over outbreaks on Cornell's Ithaca campus in the fall semester if asymptomatic surveillance is sufficiently frequent and if we have sufficient quarantine capacity. This would dovetail with a complementary effort at Cornell to reduce transmissions through housing policy, class organization, and regulations on social gatherings.
- We use our model to predict outcomes for a full return of students, faculty and staff in the fall semester over a 16 week time period, with cases imported from returning students and from Tompkins county, counterbalanced by aggressive asymptomatic surveillance where every member of the campus community is tested every 5 days. The course of the epidemic is random and we directly model that randomness. Accordingly, our model produces a range of potential futures. In the median random potential future, under our nominal set of parameters, 3.6% of the campus population (1254 people) become infected, and 0.047% of the campus population (16 people) require hospitalization. The 90% quantile rises to 4.02% infected and 0.051% requiring hospitalization. Of the 1254 infections in the median outcome, 570 are due to direct outside infections and ensuing additional infections prior to isolation, while 31 (0.09% of the campus population) are infected before arrival to campus but missed in the test-on-return protocol. There are an additional set of people infected before arrival, found through test-on-

...and beyond: Optimization, control and RL

## Reinforcement Learning

ALPHAGO ZERO CHEAT SHEET

The training pipeline for AlphaGo Zero consists of three stages, executed in parallel:



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