

Modeling Demand for Medical Resources

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Flattening the Curve

Protect the Resources to Protect Us All

- China and Italy were the two hardest hit countries at the beginning of COVID-19
 - Both implemented early lockdowns
- Italy was running out of hospital beds and ventilators
 - Hospitals reported new arrivals every 5 minutes



By Johannes Kalliauer via Wikimedia Commons

Flattening the curve protects resources, but the allocation still has to work

COVID-19 in Maryland





Counties in Maryland by Incidence of COVID-19 Counties in Maryland by Number of Hospitals



APL,

Prior Public Health Optimization Work

- Resource Allocation
 - American Hospital Capacity and Projected Need for COVID-19 Care
 - Allocation of Intensive Care Unit Beds in Periods of High Demand
 - COVID-19 Capacity Analysis Tool for Hospitals and EDs
- Simulation
 - How Simulation Modelling Can Help Reduce the Impact of COVID-19
 - A Nonhomogeneous Agent-Based Simulation Approach to Modeling the Spread of Disease in a Pandemic Outbreak
 - Locally Informed Simulation to Predict Hospital Capacity Needs During the COVID-19 Pandemic



The Question

Can we prevent resource shortages from affecting our ability to treat patients?



Patient Allocation Model

- Goals
 - Maximize utilization of hospital beds
 - Minimize patient travel
 - Minimize displacement of future patients
- Decisions
 - What hospital do we send a patient to?
 - What penalty do we incur for potentially displacing patients?
- Implementation
 - Python and Pyomo
 - Solvable with CBC, CPLEX, GLPK

$$\min \sum_{j} d_{j}x_{j} + \sum_{j,t} p_{jt}$$
s.t.
$$\sum_{j} x_{j} = 1 \forall j$$

$$x_{j} \leq b_{j}x_{j} - o_{j}x_{j} - \sum_{t' \leq t} a_{jt'}x_{j} + \sum_{t' \leq t} d_{jt'}x_{j} + p_{jt} \forall j, t$$

$$x_{j} \in \{0,1\}$$

$$p_{jt} \geq 0$$

Simulating the Epidemic

- Hospital
 - Each hospital in Maryland with ICU beds
 - Count of ICU beds associated
- Population
 - Population taken at level of census tract
- Two hospital assignment policies:
 - Go to closest hospital
 - Allow LP to solve for "best" hospital

- Implementation
 - Discrete event simulation (DES)
 - Used Python and SimPy
- Hospitals are key resource in simulation
 - Time in ICU Weibull scaled by user parameter
- Patients generated at random intervals
 - Time interval exponential with user parameter
 - Tract selection weighted by population
- Metrics:
 - Distance Travelled
 - Number of Patients Displaced
 - Wait Time

Goal is to evaluate policies against each other

How OSG Helped

- Monte Carlo simulation
 - Requires many runs
 - All runs are independent of each other
 - Perfect HTC problem
 - OSG made for HTC
- 1600 runs for each policy made
 - Run may take multiple hours
 - Just need to capture metrics

- Python worked well
 - Used virtualenv per OSG documentation
 - All required packages added to the venv
 - Added compiled copy of glpk to venv
 - Felt shady but necessary for Pyomo
 - CSV files with patient stats written
 - Logfiles via spdlog generated
 - Results turned to OSG submit node
 - Processed stats in R
 - Logs provided info to answer questions

Distance Travelled



APL,





Positive Wait Times



Limitations and Future Directions

- Simulation Model
 - Fine tune parameters to generate patients more accurately
 - Working to acquire patient data to generate patients in simulation as actual COVID-19 epidemic
- Determine coefficients for the objective function
 - Distance to the nearest hospital is weighted the same as the penalty
 - We assume that the penalty is the same for each day
 - Adjust the coefficient function to more accurately represent the tradeoff between waiting and travelling
- Policies
 - Send patients to the closest hospital with an open bed
- Metrics
 - Bed Utilization



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