

15.777 Healthcare Lab: Introduction to Healthcare Delivery in the United States Professor Jónas Jónasson | Fall 2020



Optimization Modeling for Allergy Testing

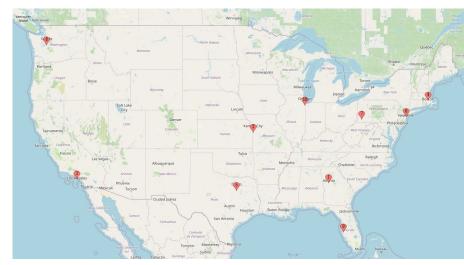
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Background and Project Motivation

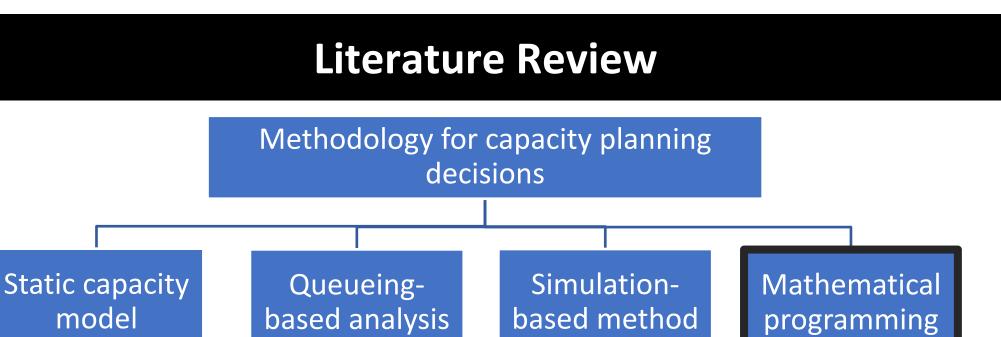
- Quest Diagnostics has 22 labs across the US and processes ~25 million allergy tests per year
 - 400+ unique allergens
- Existing equipment is approaching end of useful life, and Quest needs to purchase new testing equipment for all sites
- Operational costs such as reagent costs play a significant role over the lifespan of the machines





Objective

- Develop and implement an optimization model to determine the number of testing machines to purchase at each lab
- Build operational strategy and menu offering to address infrequently tested allergens



Methodology

- Descriptive Analytics: Quest provided several data sets, which included (i) historical daily testing volume by allergen type, (ii) cost data for each lab (labor, transportation, etc.), and (iii) instrument throughput. Through analysis of testing volume data, we discovered that roughly 60% of site-test combinations were responsible for 99% of the volume which satisfied minimum 3-month site-level volume thresholds to avoid reagent waste. This observation significantly lowered the complexity of our optimization model design
- **Optimization Model High Volume Tests:** Our model, shown on the right, sought to answer two questions: (i) how many and where testing machines should be purchased, and (ii) which origin labs should process all their own test volume, and which should divert to another lab. With the mathematical model built, we then used the Gurobi modeling program to code and solve our optimization problem
- Heuristic Design Low Volume Tests: The mathematical model served as a useful instrument to design operations for most of the allergen testing volume, but it fails to optimally allocate for low volume tests, where pooling at centers of excellence becomes part of design considerations. For each of these test types, we calculated the requisite volumes needed at each lab such that the cost of wasted reagents remained below the shipping cost to other labs and built decision rules based on projected volume

Definitions & Decision Variables

$$\begin{split} M_{j} &= number \ of \ Machines \ Purchased \ at \ Lab \ j \\ X_{ijk} &= number \ of tests \ that \ travel \ from \ lab \ i \ to \ lab \ j \ during \ month \ k \\ T_{ij} &= Travel \ cost \ from \ lab \ i \ to \ lab \ j \ where \ T_{ij} &= 0 \ \forall \ i = j \\ L_{j} &= Aggregate \ variable \ labor \ cost \ per \ test \ at \ lab \ j \\ D_{ik} &= Demand \ at \ lab \ i \ during \ month \ k \\ \beta &= 1 \ Month \ ImmunoCap \ throughput \\ \alpha_{j} &= slack \ capacity \ at \ lab \ j \ needed \ for \ given \ service \ level \ (ex: 90\%) \\ &\forall \ i, j \in \{1, ..., 10\} \ and \ k \in \{1, ..., 120\} \end{split}$$

Objective Function

Subject

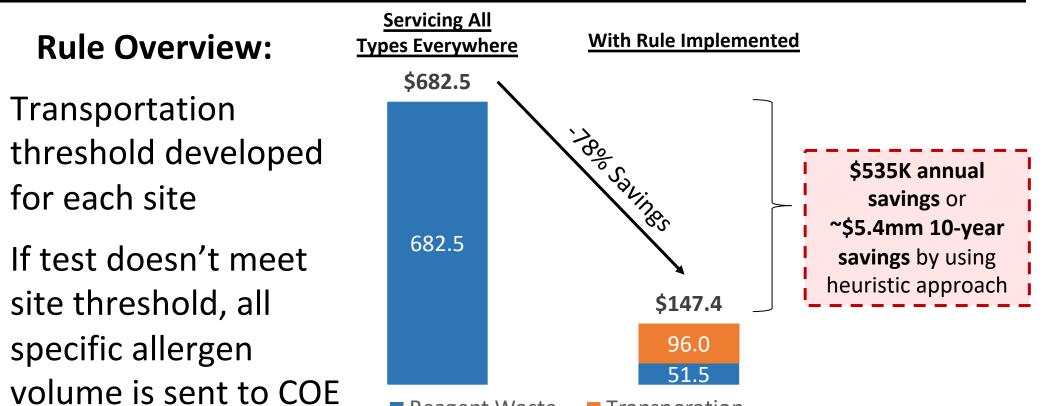
 \bullet

$$Min \quad \sum_{k}^{120} \sum_{j}^{10} \sum_{i}^{10} X_{ijk} * T_{ij} + \sum_{j}^{10} \left(\left(\sum_{k}^{120} \sum_{i}^{10} X_{ijk} \right) * L_{j} \right) + \$250,000 * \sum_{j}^{10} M_{j}$$

$$\underline{To}$$

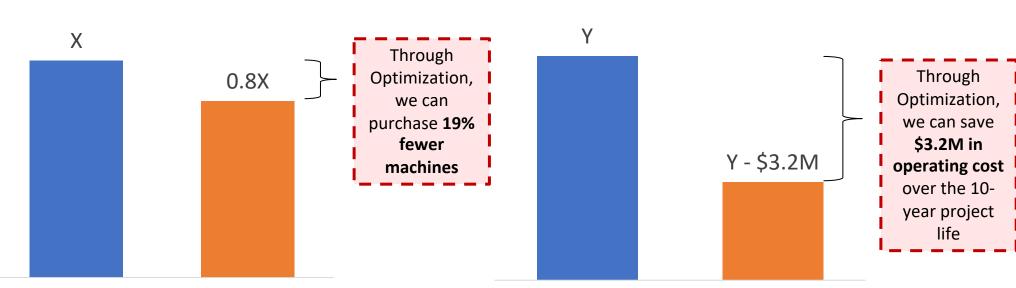
$$\begin{split} &\sum_{j}^{10} X_{ijk} = D_{ik} \ \forall \ i,k \quad (Forces \ all \ demand \ at \ Lab \ i \ to \ be \ sent \ to \ a \ lab \ j \ for \ testing) \\ &\beta \ast M_j \geq \sum_{i}^{10} \alpha_j X_{ijk} \ \forall \ j,k \quad \begin{pmatrix} Throughput \ must \ meet \ testing \ needs \ at \ lab \ j \ in \ month \ k, \\ with \ slack \ capacity \ of \ alpha \\ &Where, \ \alpha_j = \frac{\mu_j + 1.282 \ast \sigma_j}{\mu_j} \ (capture \ 90th \ percentile \ demand \ variability) \\ &M_j, X_{ijk,} \geq 0; \ M_j, X_{ijk,} \in Int \ (Non - negativity \ and \ integer \ constraints) \end{split}$$

Heuristic Results for Low-Volume Tests



Optimization Results

Machine Purchases



10-year Operating Cost

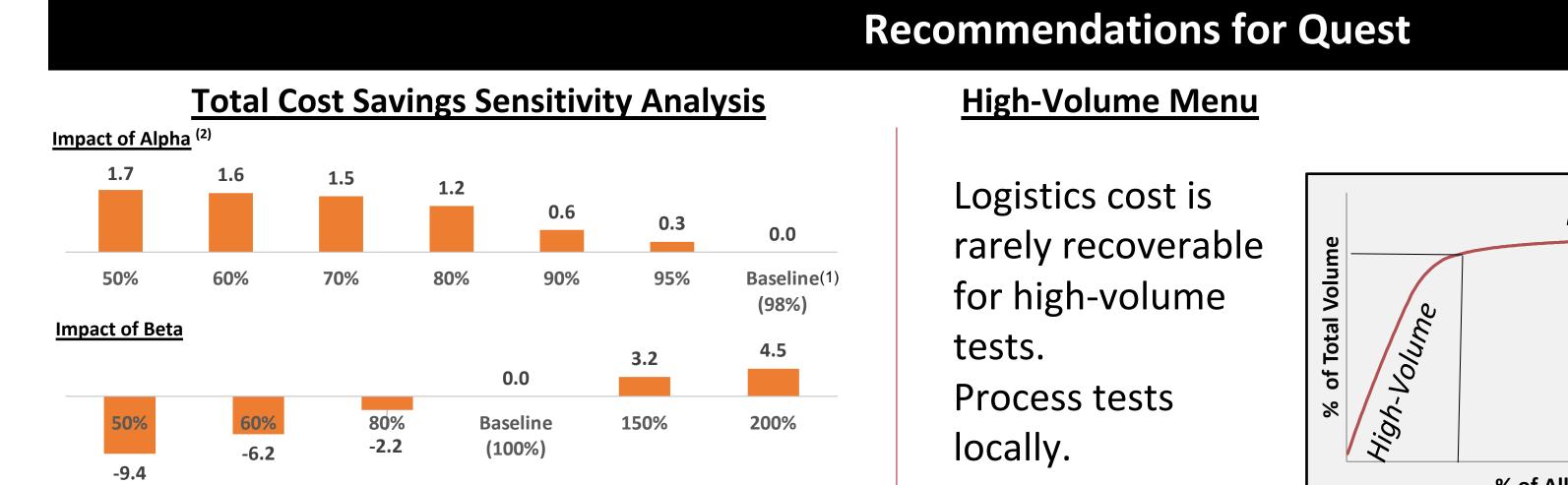
No Inter-lab Transit Optimal (98% α)

volume is sent to COE

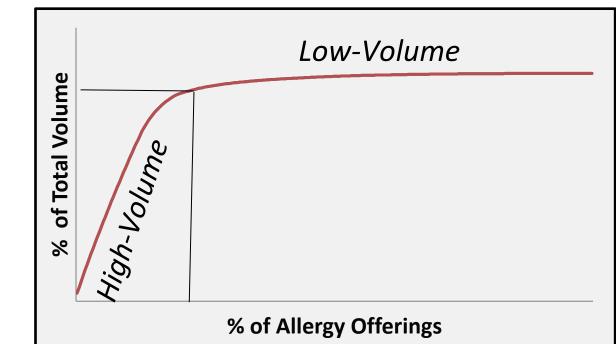
Transporation Reagent Waste

Spencer Moss

MBA 2021



(1) Baseline case is defined as the scenario where alpha is 98%, beta (throughput) assumes instruments running 12 hours/day, 30 days/month; (2) Alpha represents service level, determined by historical distributions of demand



Jordan Leising

Supply Chain 2021

Low-Volume Menu

Build a center of excellence in lowest cost lab. Simple heuristic can lower scrap cost by 78%

The Team Key Findings Asset Total Menu Allocation Design Savings 3.2MM + 5.4MM =\$8.6MM

Stella Chen

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