

Using Stochastic Programming to Improve Service Quality in an Outpatient Infusion Center

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Background

- Cancer Statistics
 - Second leading cause of death in the US
 - 1,638,910 new cases of cancer in 2012
- Outpatient Oncology
 - 29.2 million visits in the US with a primary diagnosis of cancer
 - 23 million adult patient visits for chemotherapy
- University of Michigan Comprehensive Cancer Center
 - 93,319 Outpatient Visits
 - 51,884 Infusion Treatments

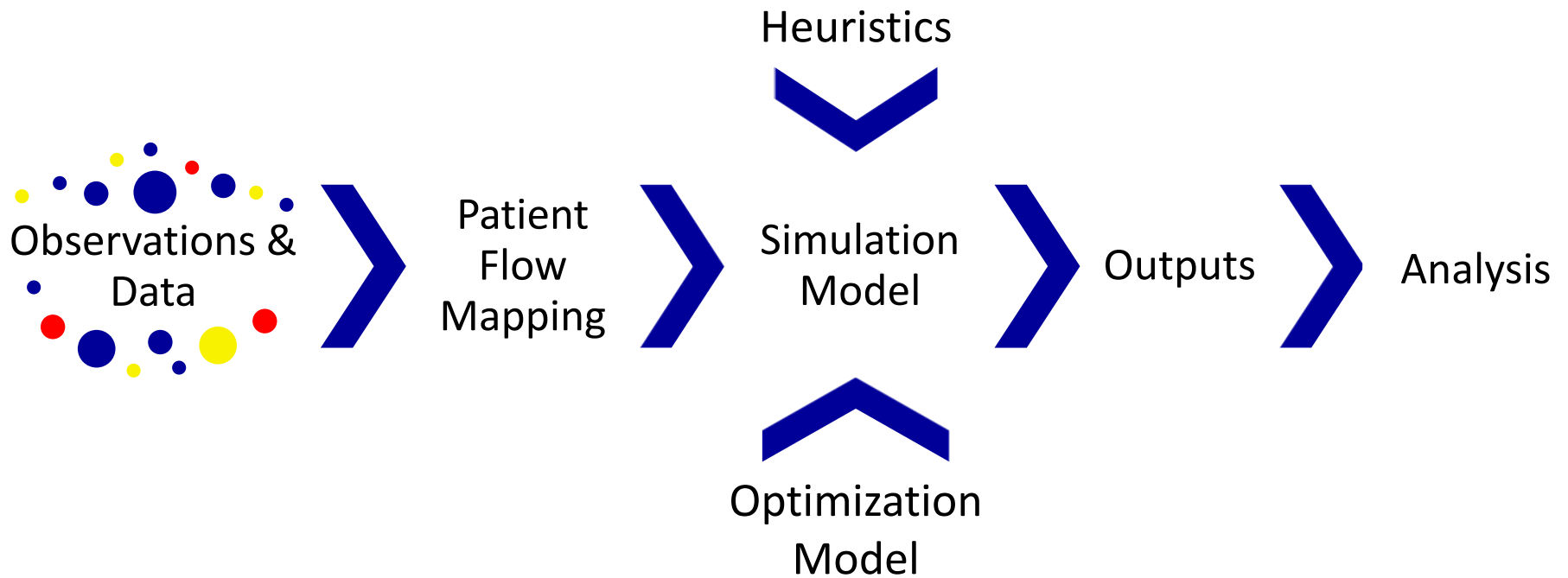


Project Goals

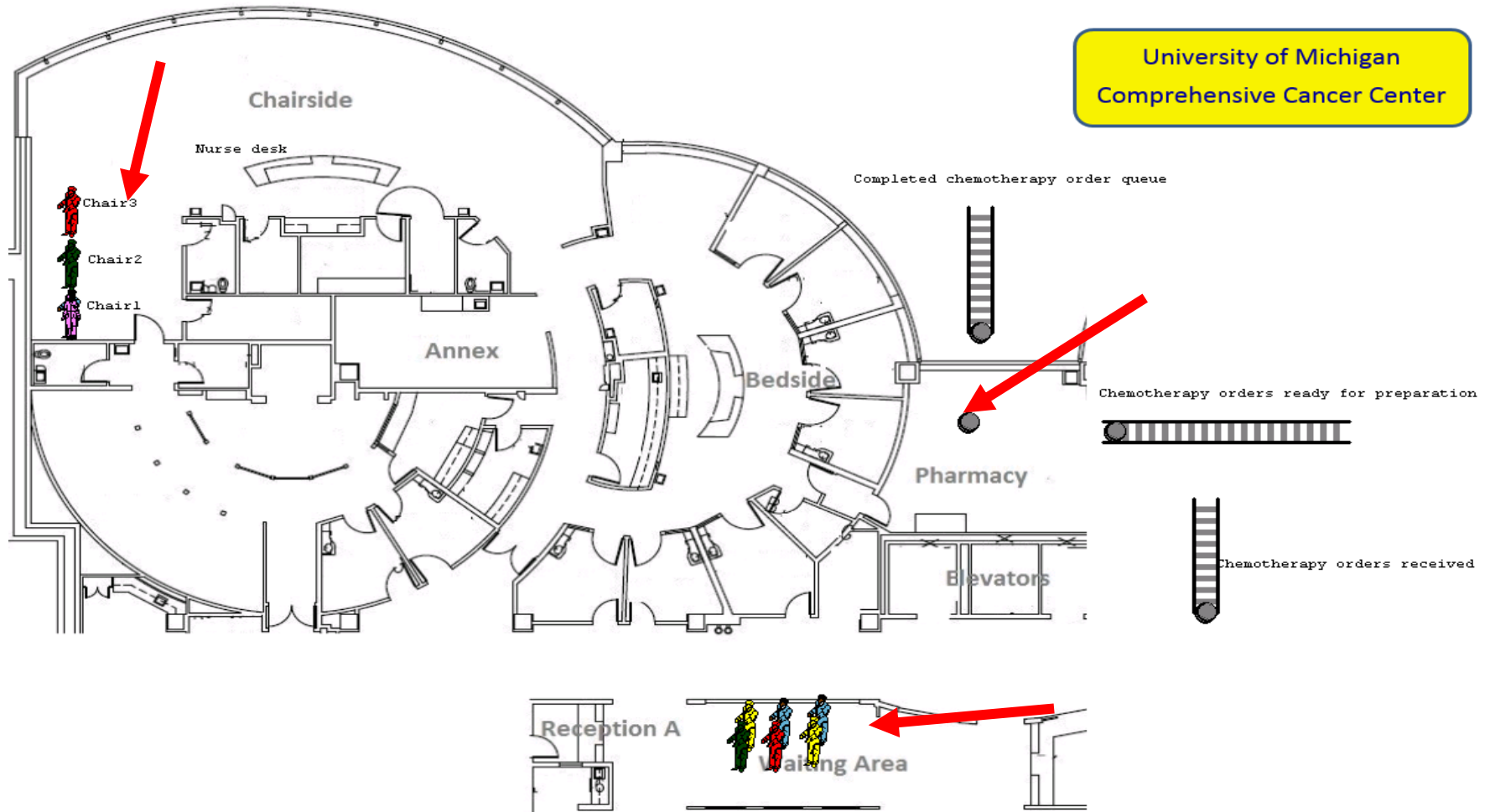
- Improve quality of cancer care delivery in the infusion center
 - Reduce patient waiting times
 - Reduce total length of day of operations
 - Improve patient and nurse safety



Approach



Computer Simulation Model



Computer Simulation Model

Inputs - Outputs

Inputs

Patient types

Nurse preparation time

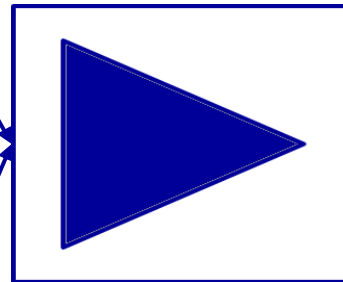
Nurse discharge time

Pharmacy preparation time

Patient appointment schedules

- Baseline
- LPT heuristic
- SPT heuristic
- Stochastic optimization model

Computer Simulation



Outputs

Average patient waiting times

Hours of operation

Chair utilization

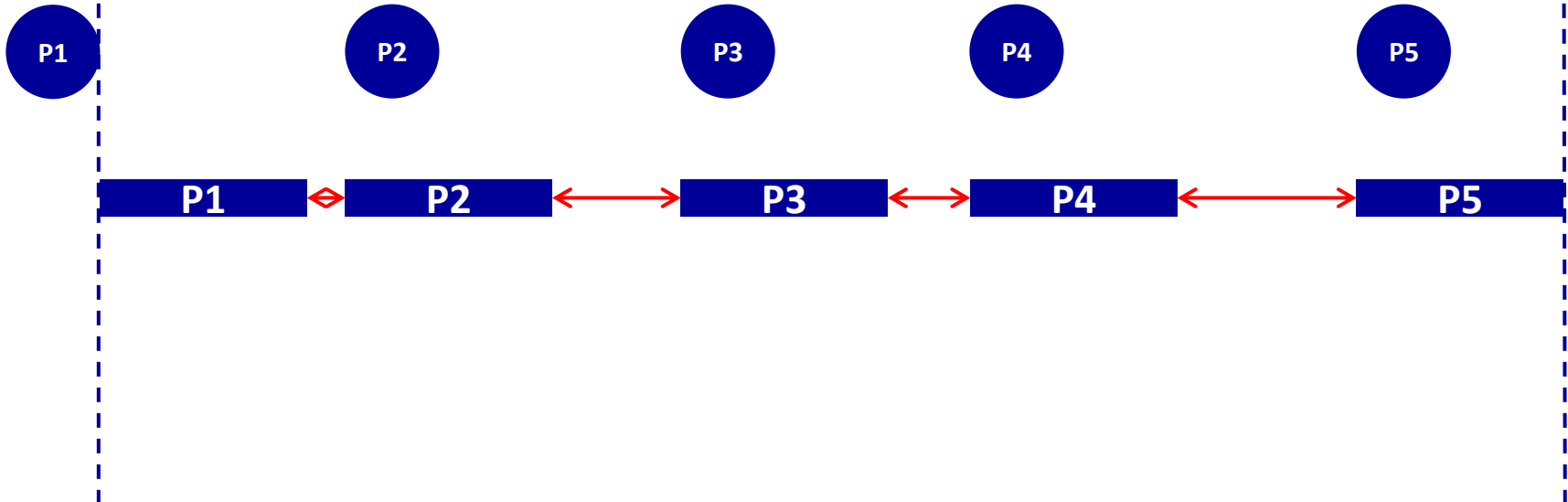
Average time in system



Patient wait times vs. Length of day

Start of Day

End of Day

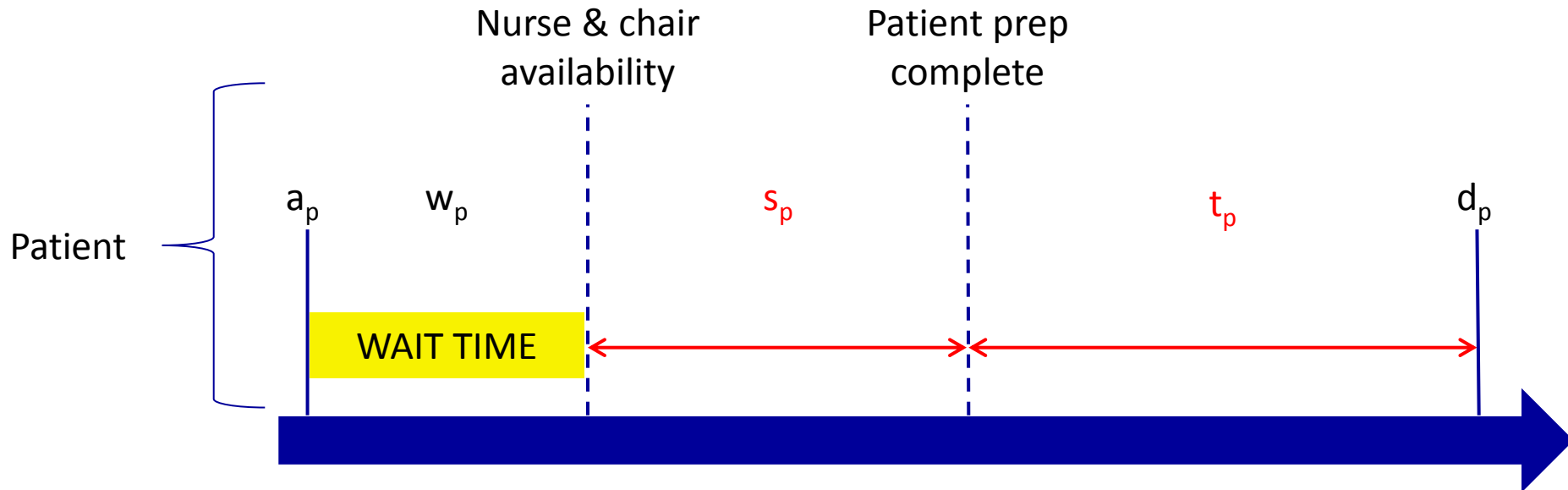


Stochastic Programming Model

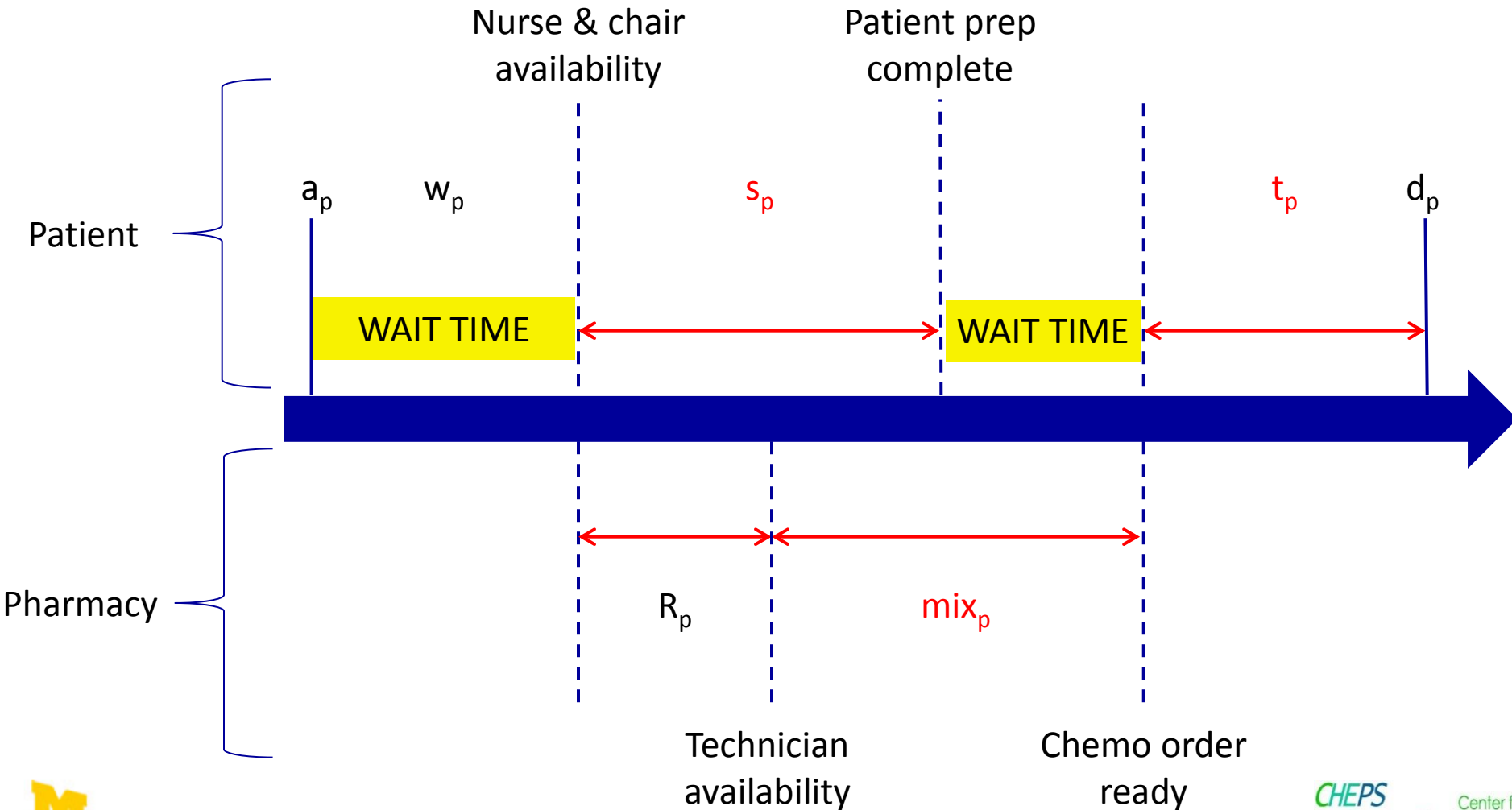
- Generates appointment schedule that reduces patient waiting times and total length of day of operations
- Decisions
 - Appointment times
 - Patient sequence
 - Patient-Chair assignment
- Stochasticity
 - Scenarios represent variability
 - Scenarios sample infusion, preparation, discharge, and drug mix times from distributions



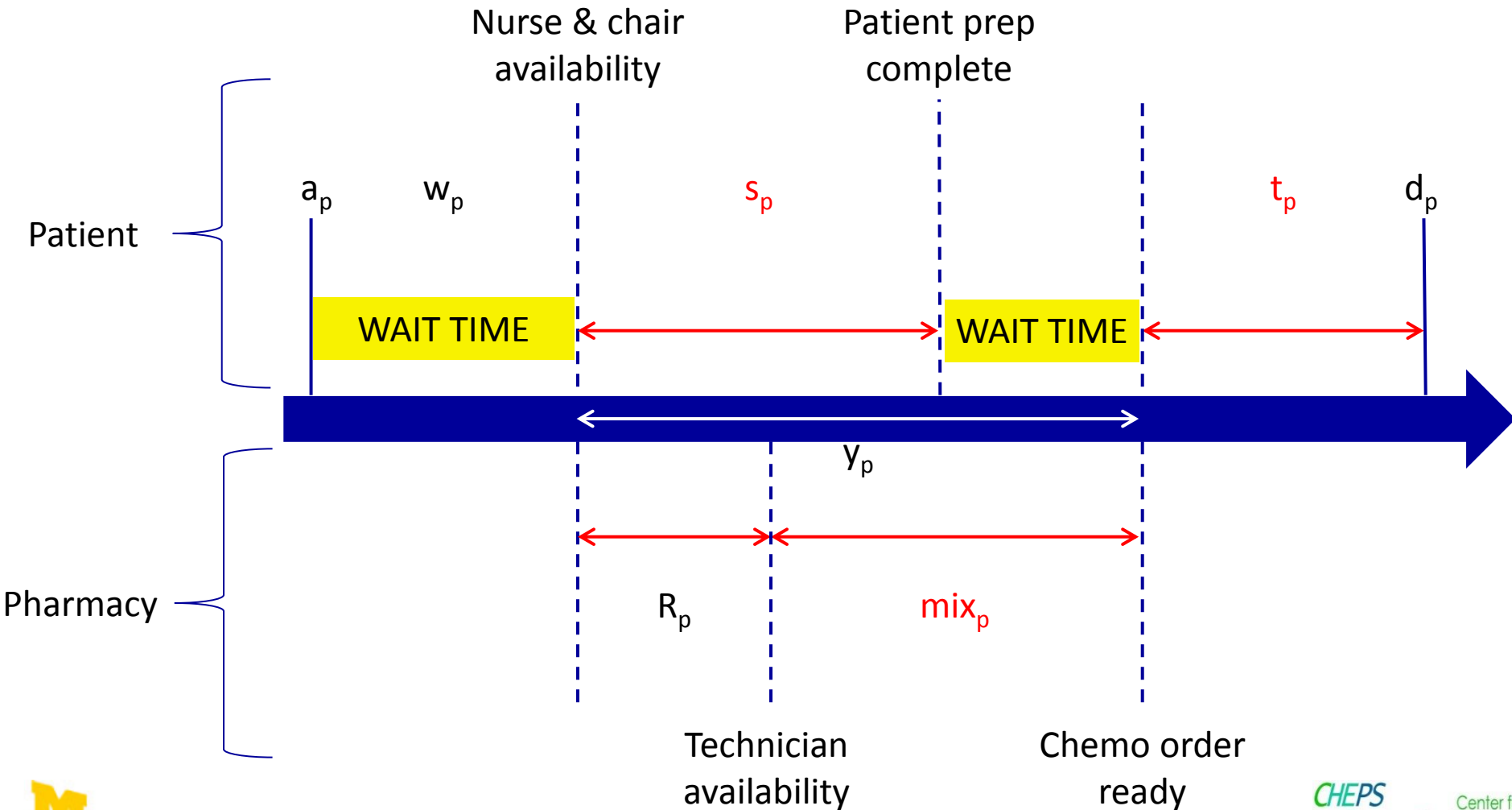
Stochastic Programming Model Variables & Parameters Explanation



Stochastic Programming Model Variables & Parameters Explanation



Stochastic Programming Model Variables & Parameters Explanation



Stochastic Programming Model

Sample Path Approximation

Decision Variables

- a_p : appointment time of patient p
- w_p^ω : waiting time of patient p in scenario ω
- R_p^ω : waiting time for available technician to mix patient's p chemo drug at the pharmacy in scenario ω
- y_p^ω : time patient p sits in the chair before infusion starts in scenario ω
- d_p^ω : exit time of patient p in scenario ω
- x_{pc}^ω : binary, 1 if patient p is assigned to chair c in scenario ω ; 0 otherwise
- $z_{p'p}$: binary, 1 if patient p' is scheduled before patient p ; 0 otherwise
- L^ω : end of day in scenario ω

Parameters

- s_p^ω : preparation time of patient p in scenario ω
- t_p^ω : infusion time plus discharge time of patient p in scenario ω
- mix_p^ω : chemo drug mix time for patient p in scenario ω
- m : number of scenarios
- λ : weight in the objective function
- M : large number



Stochastic Programming Model

Sample Path Approximation

$$\min \lambda \sum_{p \in P} \sum_{\omega \in \Omega} \frac{1}{m} [w_p^\omega + (y_p^\omega - s_p^\omega)] + (1 - \lambda) \sum_{\omega \in \Omega} \frac{1}{m} L^\omega$$

Objective Function: Trade-off between the total expected waiting time and the expected length of day



Stochastic Programming Model

Sample Path Approximation

$$\begin{aligned} \min \quad & \lambda \sum_{p \in P} \sum_{\omega \in \Omega} \frac{1}{m} [w_p^\omega + (y_p^\omega - s_p^\omega)] + (1 - \lambda) \sum_{\omega \in \Omega} \frac{1}{m} L^\omega \\ \text{subject to} \quad & \sum_{c \in C} x_{pc}^\omega = 1 \qquad \qquad \qquad \forall p \in P, \forall \omega \in \Omega \end{aligned}$$

Each patient should be assigned to exactly one infusion chair



Stochastic Programming Model

Sample Path Approximation

$$\begin{aligned} \min \quad & \lambda \sum_{p \in P} \sum_{\omega \in \Omega} \frac{1}{m} [w_p^\omega + (y_p^\omega - s_p^\omega)] + (1 - \lambda) \sum_{\omega \in \Omega} \frac{1}{m} L^\omega \\ \text{subject to} \quad & \sum_{c \in C} x_{pc}^\omega = 1 && \forall p \in P, \forall \omega \in \Omega \\ & a_p + w_p^\omega + y_p^\omega + t_p^\omega = d_p^\omega && \forall p \in P, \forall \omega \in \Omega \end{aligned}$$

Value of exit time of patient p in each scenario



Stochastic Programming Model

Sample Path Approximation

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 \min \quad & \lambda \sum_{p \in P} \sum_{\omega \in \Omega} \frac{1}{m} [w_p^\omega + (y_p^\omega - s_p^\omega)] + (1 - \lambda) \sum_{\omega \in \Omega} \frac{1}{m} L^\omega \\
 \text{subject to} \quad & \sum_{c \in C} x_{pc}^\omega = 1 && \forall p \in P, \forall \omega \in \Omega \\
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 & a_p + w_p^\omega + M(3 - x_{pc}^\omega - x_{p'c}^\omega - z_{p'p}) \geq d_{p'}^\omega && \forall c \in C, \forall p, p' \in P, \forall \omega \in \Omega
 \end{aligned}$$

Free chair constraint: A patient can sit in a chair only if all previously sequenced patients assigned to the same chair have been discharged



Stochastic Programming Model

Sample Path Approximation

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 \min \quad & \lambda \sum_{p \in P} \sum_{\omega \in \Omega} \frac{1}{m} [w_p^\omega + (y_p^\omega - s_p^\omega)] + (1 - \lambda) \sum_{\omega \in \Omega} \frac{1}{m} L^\omega \\
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 \end{aligned}$$

Available nurse constraint: A patient can sit in a chair if the nurse has finished preparing all previously sequenced patients



Stochastic Programming Model

Sample Path Approximation

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 \end{aligned}$$

FIFO rule: A patient that arrives before another patient has to sit in an infusion chair first



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Sample Path Approximation

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 & a_p + w_p^\omega + R_p^\omega + M(1 - z_{p'p}) \geq a_{p'} + w_{p'}^\omega + R_{p'}^\omega + mix_{p'}^\omega && \forall p, p' \in P, \forall \omega \in \Omega
 \end{aligned}$$

Available pharmacy technician constraint: A pharmacy technician can prepare only one chemotherapy order at a time



Stochastic Programming Model

Sample Path Approximation

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 & a_p + w_p^\omega + R_p^\omega + M(1 - z_{p'p}) \geq a_{p'} + w_{p'}^\omega + R_{p'}^\omega + mix_{p'}^\omega && \forall p, p' \in P, \forall \omega \in \Omega \\
 & L^\omega \geq d_p^\omega && \forall p \in P, \forall \omega \in \Omega
 \end{aligned}$$

All patients should be discharged to end the day



Stochastic Programming Model

Sample Path Approximation

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 & L^\omega \geq d_p^\omega && \forall p \in P, \forall \omega \in \Omega \\
 & (a_{p'} - a_p) + Mz_{p'p} \geq 0 && \forall p, p' \in P \\
 & z_{pp'} = 1 - z_{p'p} && \forall p, p' \in P
 \end{aligned}$$

Definition of variable $z_{p'p}$



Stochastic Programming Model

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 & y_p^\omega \geq s_p^\omega && \forall p \in P, \forall \omega \in \Omega \\
 & y_p^\omega \geq R_p^\omega + mix_p^\omega && \forall p \in P, \forall \omega \in \Omega
 \end{aligned}$$

Definition of variable y_p^ω



Stochastic Programming Model

Sample Path Approximation

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 & y_p^\omega \geq s_p^\omega && \forall p \in P, \forall \omega \in \Omega \\
 & y_p^\omega \geq R_p^\omega + mix_p^\omega && \forall p \in P, \forall \omega \in \Omega \\
 & x_{pc}^\omega \in \{0, 1\} && \forall c \in C, \forall p \in P, \forall \omega \in \Omega \\
 & z_{p'p} \in \{0, 1\} && \forall p, p' \in P \\
 & a_p \geq 0 && \forall p \in P \\
 & w_p^\omega, d_p^\omega, R_p^\omega \geq 0 && \forall p \in P, \forall \omega \in \Omega
 \end{aligned}$$

Binary and non-negativity constraints



Stochastic Programming Model Tractability

Optimized patient sequence
Optimized appointment times

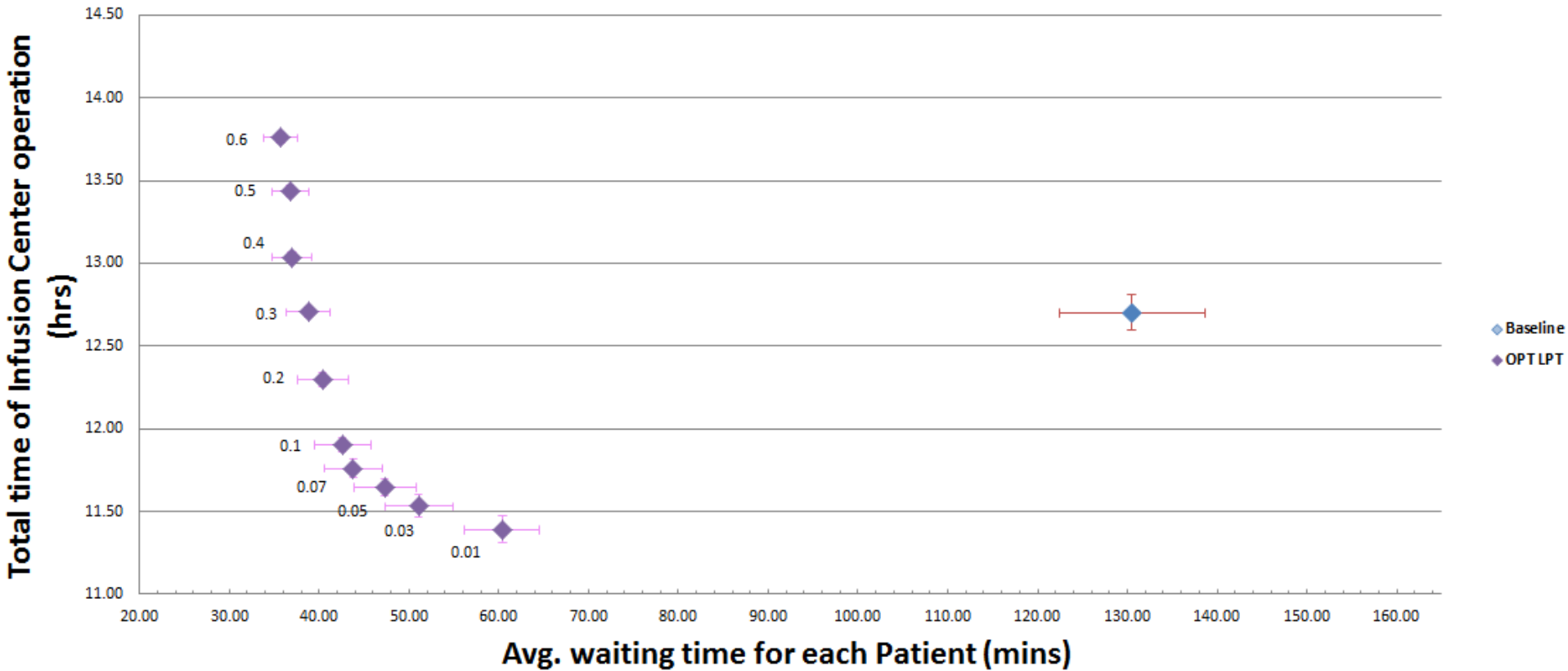


Fixed patient sequence
Optimized appointment times

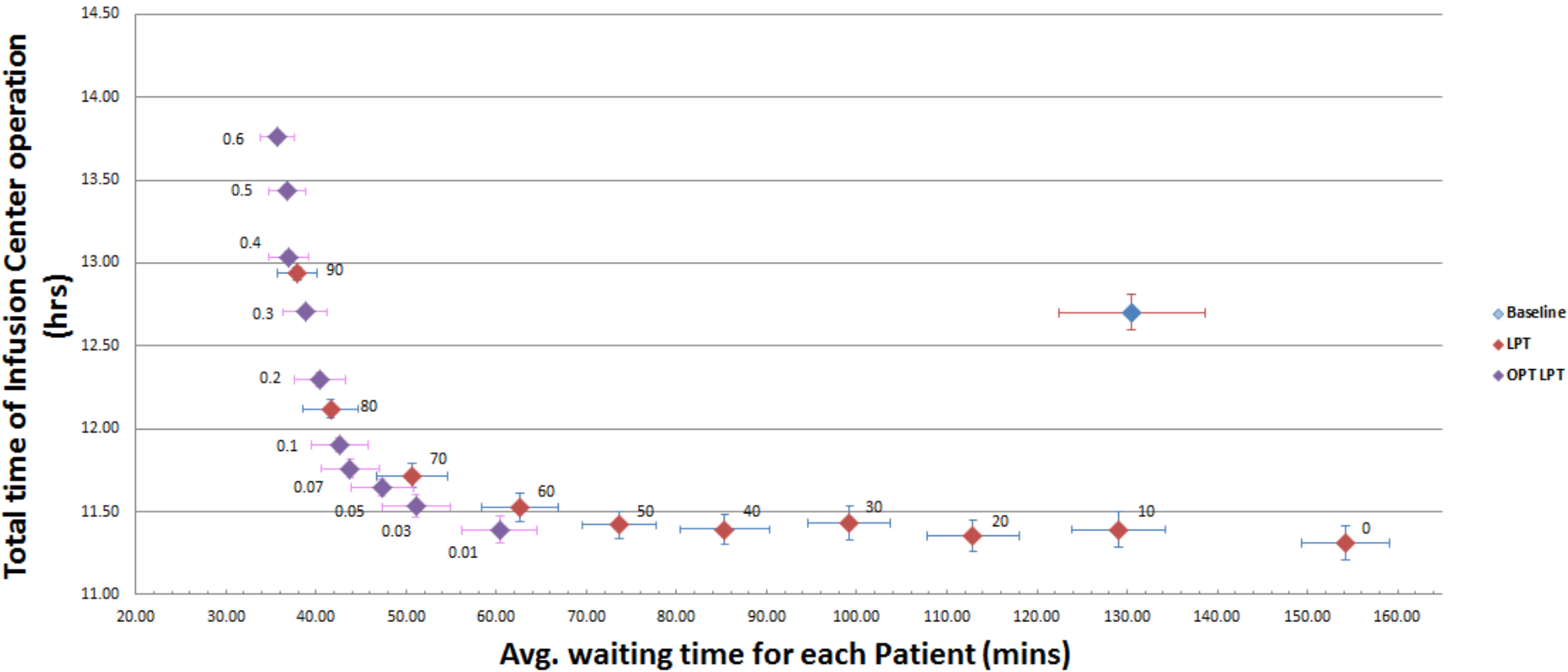
- Stochastic MIP model
 - Large number of constraints
 - Using big M in constraints results in weak relaxations
- Stochastic LP model assuming a fixed sequence of patients (e.g. LPT)
 - Eliminate $z_{p,p}$ and x_{pc}^{ω} decision variables



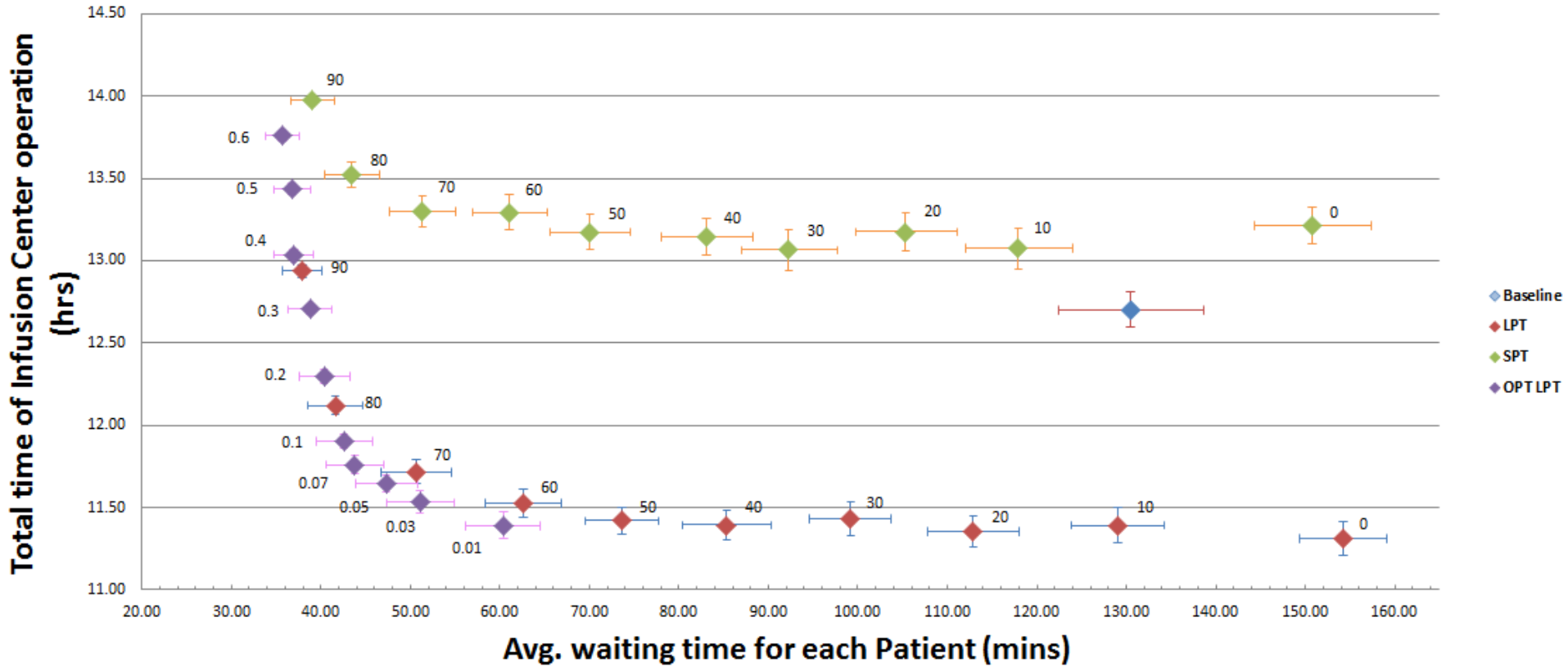
Results



Results



Results



Conclusions

- Key aspects of the infusion center scheduling process can be approximated as a two stage stochastic integer program
- Fixing sequencing decisions based on LPT results in an easy to solve continuous stochastic program that performs well compared to current practice and other heuristics
- The LPT sequence is near optimal when compared to solvable instances of the MIP model for a small number of scenarios



Future Steps

- Improve stochastic MIP model
- Development of a heuristic that can be easily implemented by schedulers
- Enhancing simulation model
 - Addition of oncology clinic



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