

Simulating Access and Patient Flow for a Cardiovascular ICU

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RESEARCH MOTIVATION

What is the aortic dissection (AD) patient experience?

> Ensure adequate capacity for all transfer requests to the Cardiovascular Center (CVC) at Michigan Medicine (MM)

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INITIAL RESEARCH QUESTION





INITIAL RESEARCH QUESTION

It's a lot more complicated than that.



THE BIGGER PICTURE



= All Other Cardiac Patient Types

PATIENT FLOW IN CARDIOVASCULAR SURGERY



TRANSFER REQUESTS

Preliminary analysis conducted by the CVC staff showed that the most common reason for patient deferral when requesting transfer to Michigan Medicine is attributed to unavailable ICU beds.



APPROACH TO IMPROVING ICU UTILIZATION



SIMULATION FRAMEWORK

FIXED INPUTS

- Bed Count per Unit
- Time Horizon
- Number of Replications

VARIABLE INPUTS

- Patient Arrival Rate
- Length of Stay in ICU and Step Down (SDn) units
- Bounce Back Probability

SIMULATION FRAMEWORK



SIMULATION FRAMEWORK

METRICS

Number of			
Patient Arrivals			
Accepted Patients			min
Denied Patients			
ICU	Step Down (SDn)		
 Patient LOS Unnecessary days in an ICU bed (SDn status) Bed Utilization 	 Patient LOS Unnecessary days in a SDn bed (ICU status) Bed Utilization 	med	max

ANALYSES

SDn Variation

 Change number of shared SDn beds

Bounce Back Rate

 Change the rate of bounce back incrementally 2

BASE CASE PARAMETERS

- 1 Patient Type
- Arrival Rate = 0.33 patient/hr
- Time Horizon = 1 Year
- Replications = 1,000

- Bernoulli trial for transfer and discharge from respective units
 - P_{ICU Transfer} = 0.25



ANALYSIS 1: SDN VARIATION

Allocated Step Down Beds	28	32	36	40
Annual Patient Arrival	2879	2875	2878	2877
Patients Denied	8.1%	5.8%	4.8%	4.5%
ICU Average LOS ICU Status	3.45 days	3.44 days	3.44 days	3.44 days
ICU Average LOS SDn Status	0.34 days	0.14 days	0.04 days	0.01 days
SDn Average LOS	3.70 days	3.78 days	3.82 days	3.83 days
SDn Bed Utilization	94.23%	88.98%	82.06%	74.73 %

- Time Horizon = 1 Year
- Replications = 1,000

- 36 ICU Beds
- 16 Dedicated SDn Beds

ANALYSIS 2: BOUNCE BACK

Bounce Back Rate Increments

Bounce Back Rate	0%	5%	10%	15%
Annual Patient Arrival	2875	2871	2873	2872
Patients Denied	5.8%	13%	22%	31%
ICU Average LOS ICU Status	3.44 days	3.85 days	4.09 days	4.39 days
ICU Average LOS SDn Status	0.14 days	0.31 days	0.68 days	1.15 days
SDn Average LOS	3.78 days	4.19 days	4.43 days	4.45 days
SDn Average LOS ICU Status	0 days	0.15 days	0.45 days	0.76 days
ICU Bed Utilization	78.88%	85.95%	89.53%	92.02%

- Time Horizon = 1 Year
- Replications = 1,000

- 36 ICU Beds
- 32 SDn Beds

ANALYSES TAKEAWAYS

Analysis 1: SDn Variation

- The unnecessary ICU bed days decreases as SDn beds are added to a certain point
- Trade-offs will be necessary

Analysis 2: Bounce Back

- Small rates of bounce back impact utilization and flow
- Patient information would allow us to more accurately predict bounce back rates

FUTURE RESEARCH

- Expanding the tool
 - Adding more patient types
 - Adding patient predictors of bounce back
- Conducting Analysis
 - More data!
 - Explore smoothing elective surgery

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Questions?



LITERATURE REVIEW

First Author	Reference	Year	Objective/Purpose
Levin, S.	[5]	2011	To test policies to reduce patient's length of stay (LOS) and increase patient throughput.
Marmor, Y.	[6]	2013	To predict minimum bed needs to achieve the high patient service level demanded for the cardiovascular ICU.
Levin, S.	[7]	2015	To estimate patients' wait time while integrating the effect of the transition process (i.e. wait time for a bed to become available) with queuing using embedded regression models.
Kolker, A.	[8]	2009	To establish a quantitative link between the daily load leveling of elective surgeries (i.e. elective schedule smoothing) and ICU diversion of multiple ICU units including cardio ICU.

INTRODUCTION | PROBLEM STATEMENT | LITERATURE REVIEW | SIMULATION | ANALYSIS | FUTURE RESEARCH

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