

Systems Engineering Methods for Improving Patient Access to Care

Brian Denton

North Carolina State University

Edward P. Fitts Department of Industrial and Systems Engineering

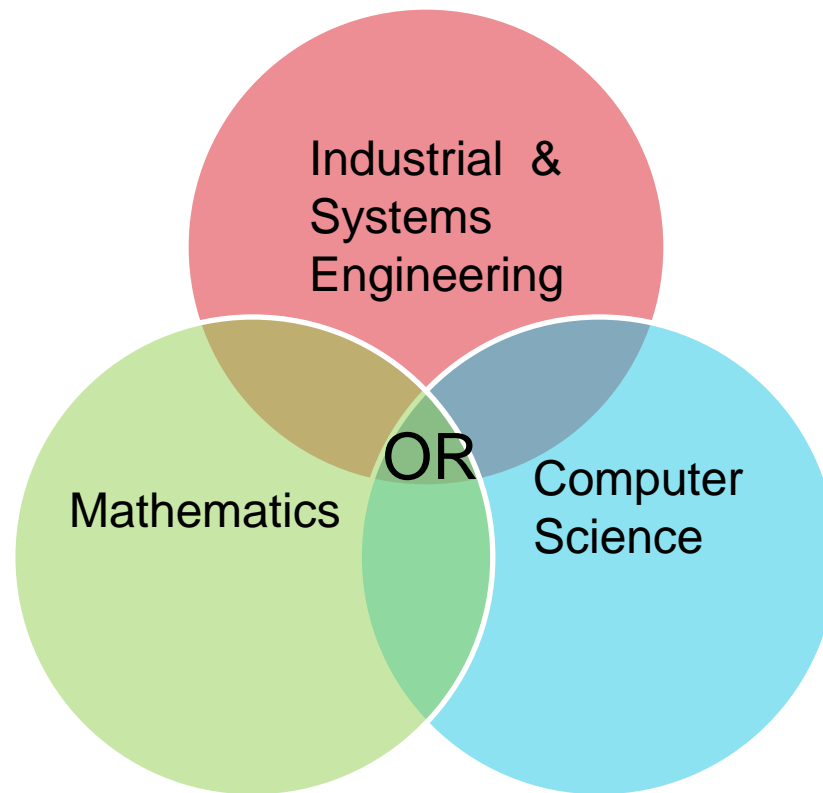
June 10, 2008



Summary

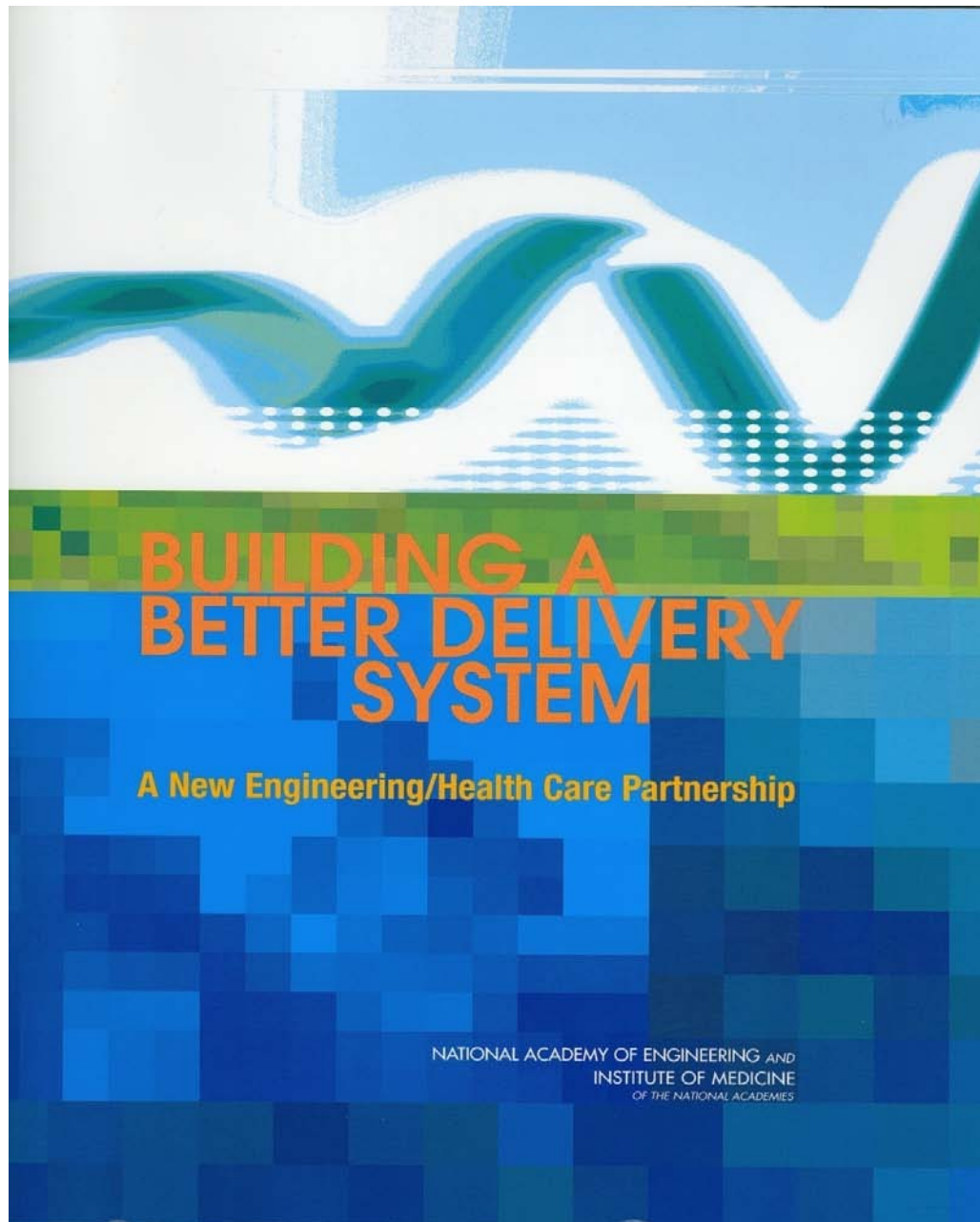
- Opportunities for systems engineering methods to improve health care delivery
- Two examples:
 - (1) Modeling patient flow through a family medicine clinic
 - (2) An optimization based decision support system for scheduling thoracic surgery

Operations Research



A multidisciplinary field that focuses on analytical methods for optimization of complex systems





Analysis of Patient Flow Through UNC Family Medicine Clinic

North Carolina State University
Edward P. Fitts Department of Industrial and Systems Engineering
Senior Design Project

Matthew Diering
Kylie Goodell
Drew Johnson

Sponsored by:
The University of North Carolina
Family Practice

Advised by:
Mr. Kevin Tate, UNC Family Practice
Dr. Brian Denton, North Carolina State University
Dr. Javad Taheri, North Carolina State University

April 23, 2008



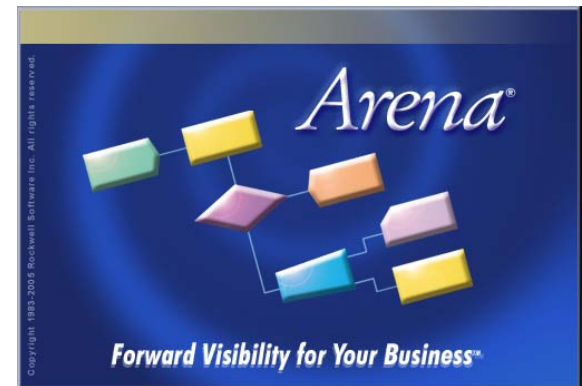
Project Objectives

- Develop a discrete event simulation model of patient flow through UNC Family Medicine Clinic
- Provide Recommendations to Clinic
 - Identify Bottlenecks
 - Increasing Patient Throughput
 - Decreasing Patient Waiting Time



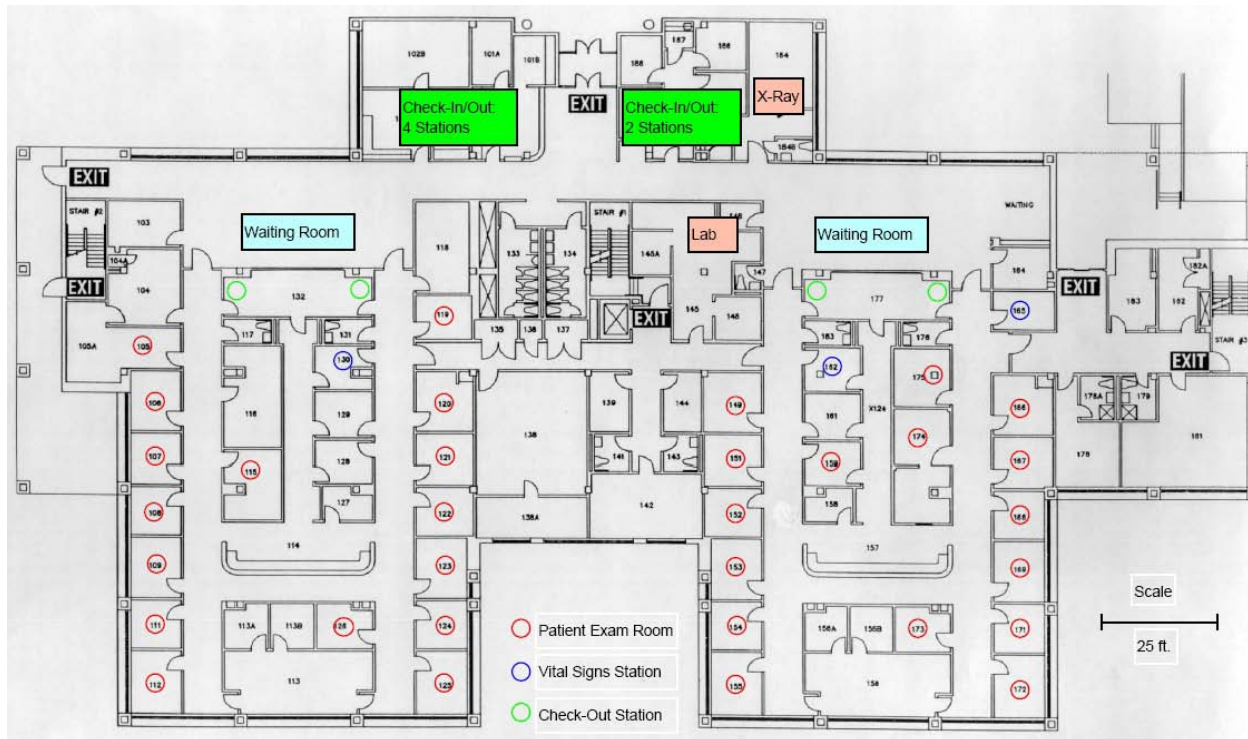
Discrete Event Simulation

- Why Discrete Event Simulation?
 - Cost Effective
 - Time Effective
 - Evaluate Multiple Alternatives
 - Evaluate the effects of uncertainty by averaging across many scenarios (days)



Creating a Conceptual Model

- On Site Meetings, Clinic Tour
- Track Process with Conference Calls



Patient Flow and Resource Constraints

- Process Mapping:
 - Sequence of activities for typical patient flow
 - Alternative sequences and collections
 - Unpredictable events
- Potential Resource Constraints:

Exam Rooms	Vitals Rooms
Lab	Checkin/out
Physicians	Nurses
Preceptors	



Creating a Data Model

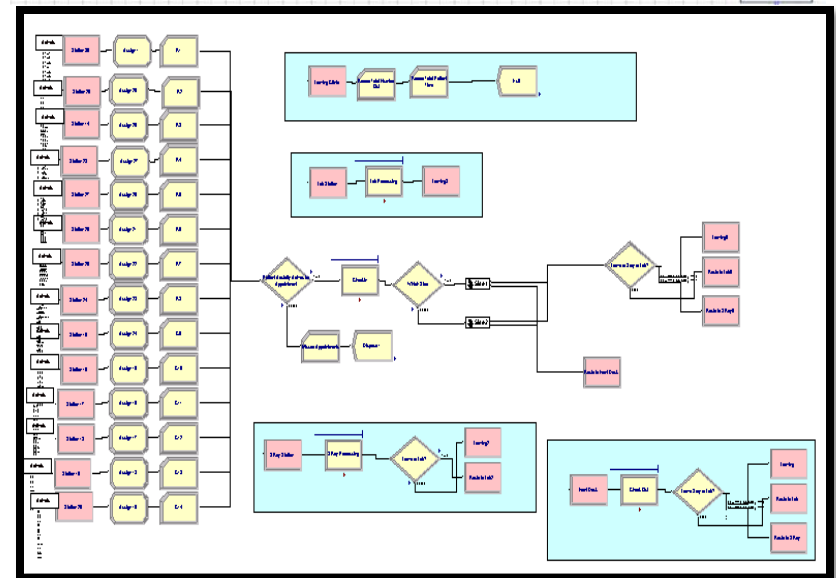
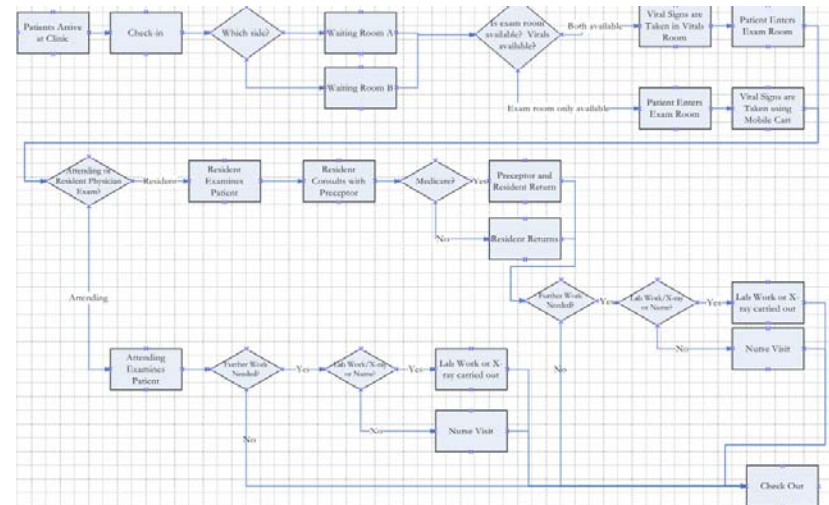
- Large amounts of historical data required
 - Sampled activity times (e.g. patient with physician, patient with nurse, lab time, checkin/out times)
 - Probabilities
 - Arrival Schedule

	A	V	W	X	Y	Z	AA	AB	AC	AD	AE	AF	AG
	Date	Early for Appt	Late for Appt	Time From Appt Time	Time from Arrival	Arrival-Check In	Check In Time (mins)*	Time in Waiting Room	Waiting Room Time (Mins)	Length of Vitals	Vital sign time (Mins)	Exam Room	Waiting in Exam room (mins)
2	9/28	0:00		1:13	1:13	0:07	7	0:02	2	0:03	3	0:02	2
3	9/28	0:02		1:11	1:13	0:01	1	0:02	2	0:04	4	0:08	8
4	9/28	0:15		1:22	1:22	0:06	6	0:18	18	0:08	8	0:00	0.5
5	9/28	0:09		0:40	0:49	0:03	3	0:13	13	0:03	3	0:15	15
6	9/28	0:17		0:52	1:09	0:03	3	0:02	2	0:07	7	0:05	5
7	9/28		0:06	0:46	0:46	0:01	1	0:05	5	0:08	8	0:05	5
8	9/28		0:08	0:17	0:17	0:04	4	0:07	7	0:00		0:01	1
9	9/28		0:23	0:18	0:18	0:02	2	0:03	3	0:02	2	0:00	0
10	9/29	0:02		0:51	0:53	0:02	2	0:02	2	0:03	3	0:02	2
11	9/29	0:19		0:36	0:55	0:03	3	0:06	6	0:16	16	0:06	6
12	9/29		0:03	1:07	1:07	0:03	3	0:00	0	0:03	3	0:02	2
13	9/29	0:16		1:05	1:21	0:06	6	0:33	33	0:03	3	0:02	2
14	9/29	0:05		0:50	0:55	0:03	3	0:12	12	0:02	2	0:02	2
15	9/29		0:05	0:26	0:26	0:00	0.5	0:04	4	0:02	2	0:06	6
16	9/29	0:10		0:52	1:02	0:00	0.5	0:01	1	0:04	4	0:10	10
17	9/29	0:19		1:27	1:46	0:05	5	0:14	14	0:05	5	0:36	36
18	9/29	0:10		1:30	1:40	0:02	2	0:12	12	0:03	3	0:42	42
19	10/3	0:02		0:47	0:49	0:03	3	0:04	4	0:04	4	0:26	26
20	10/3	0:32		0:12	0:44	0:03	3	0:05	5	0:03	3	0:07	7
21	10/3	0:27		0:50	1:17	0:12	12	0:25	25	0:03	3	0:12	12
22	10/3	0:05		1:04	1:09	0:04	4	0:17	17	0:07	7	0:13	13
23	10/3	0:08		0:37	0:45	0:07	7	0:01	1	0:07	7	0:22	22
24	10/3	0:02		1:12	1:14	0:04	4	0:02	2	0:03	3	0:07	7
25	10/3	0:00		0:20	0:20	0:02	2	0:03	3	0:02	2	0:03	3
26	10/3	0:07		0:30	0:37	0:03	3	0:04	4	0:04	4	0:02	2
27	10/3	0:05		0:42	0:47	0:04	4	0:02	2	0:04	4	0:06	6
28	10/3	0:09		0:40	0:49	0:04	4	0:10	10	0:05	5	0:05	5
29	10/5	0:13		1:12	1:25	0:08	8	0:07	7	0:04	4	0:24	24
30	10/5	0:00		0:42	0:42	0:02	2	0:09	9	0:04	4	0:16	16
31	10/5	0:15		0:15	0:30	0:03	3	0:00	0	0:04	4	0:08	8
32	10/5		0:05	0:47	0:47	0:01	1	0:06	6	0:05	5	0:08	8
33		0:09	0:09	0:48	0:56	0:03	0:07		0:04			0:09	
34													
35													
36													

* FOR ALL 0 TIMES, ASSUMED TIME WAS .5 MINUTES

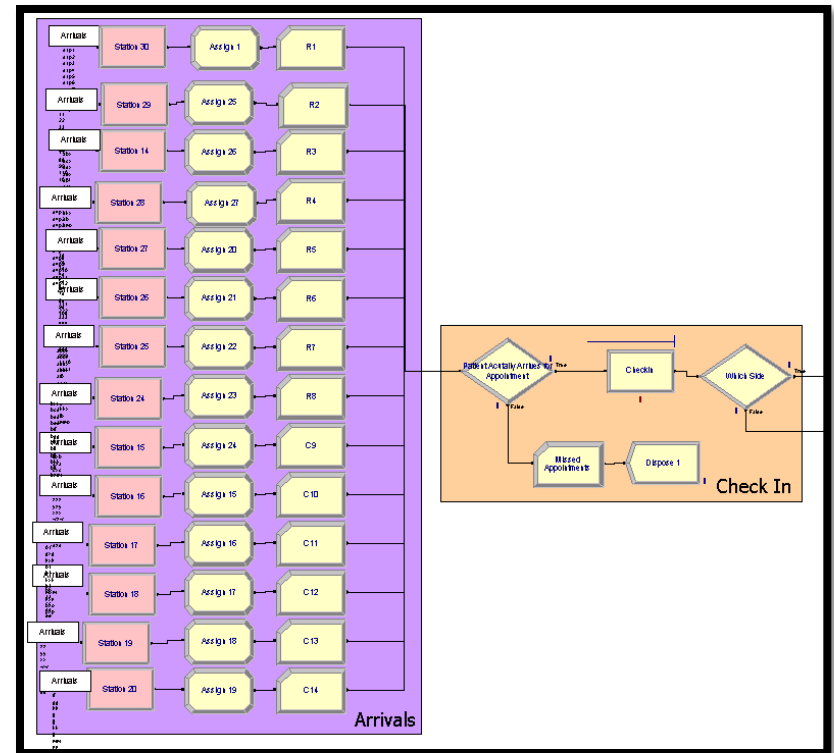
Model Construction

- The base model was constructed in modules
 - Arrivals and Check In
 - Waiting Area and Vital Signs
 - Exam Rooms and Examinations
 - Radiology, Laboratory, and Checkout
- The Model is divided into two sides with two halls per side

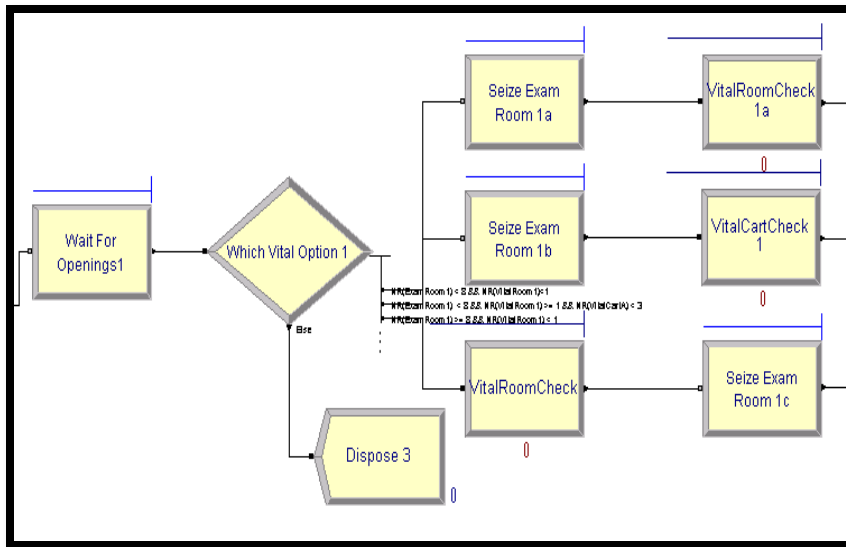


Arrivals and Check In

- Arrival of patients is determined by scheduling templates
- Each patient is assigned a specific physician and a number of other attributes based on a discrete distribution



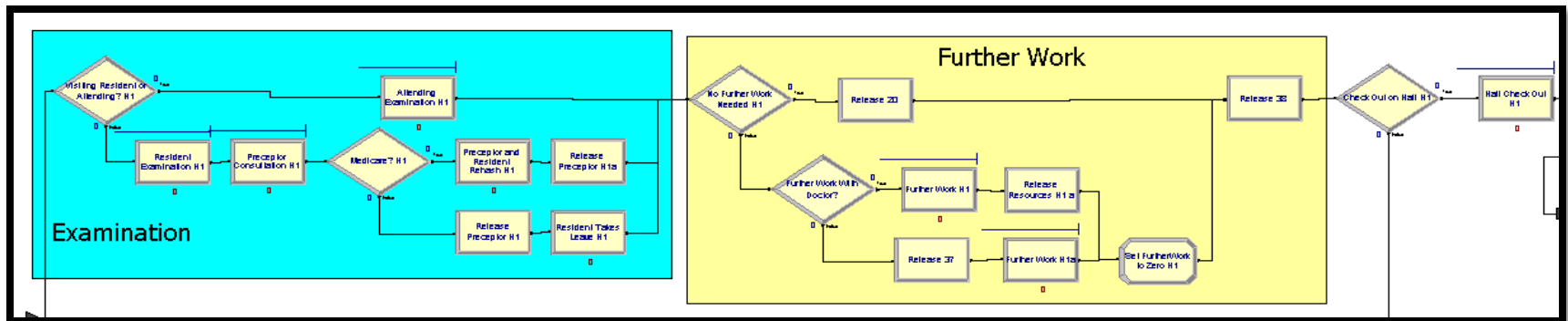
Resource Constraints: Waiting Area and Vital Signs



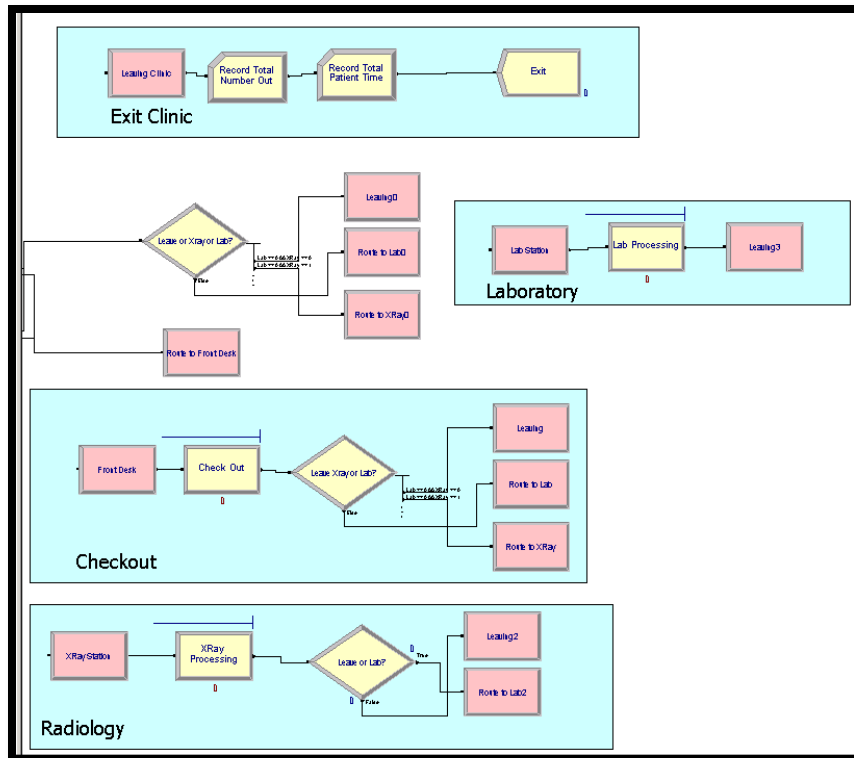
- Patients wait for one of two conditions to become true:
 - Vital room is available
 - Exam room and vital cart are available

Resource Constraints: Exam Rooms and Examination

- Patients wait for their physician to become available
- Resident procedure vs. Attending procedure
- Further Work
 - Only Nurse
 - Nurse and Physician
- Hall Checkout



Resource Constraints: Radiology, Laboratory, and Checkout



- After leaving the exam room, each patient has a probability of needing an X-ray and/or laboratory work
- If not already done, patient checks out then leaves the clinic

Staffing

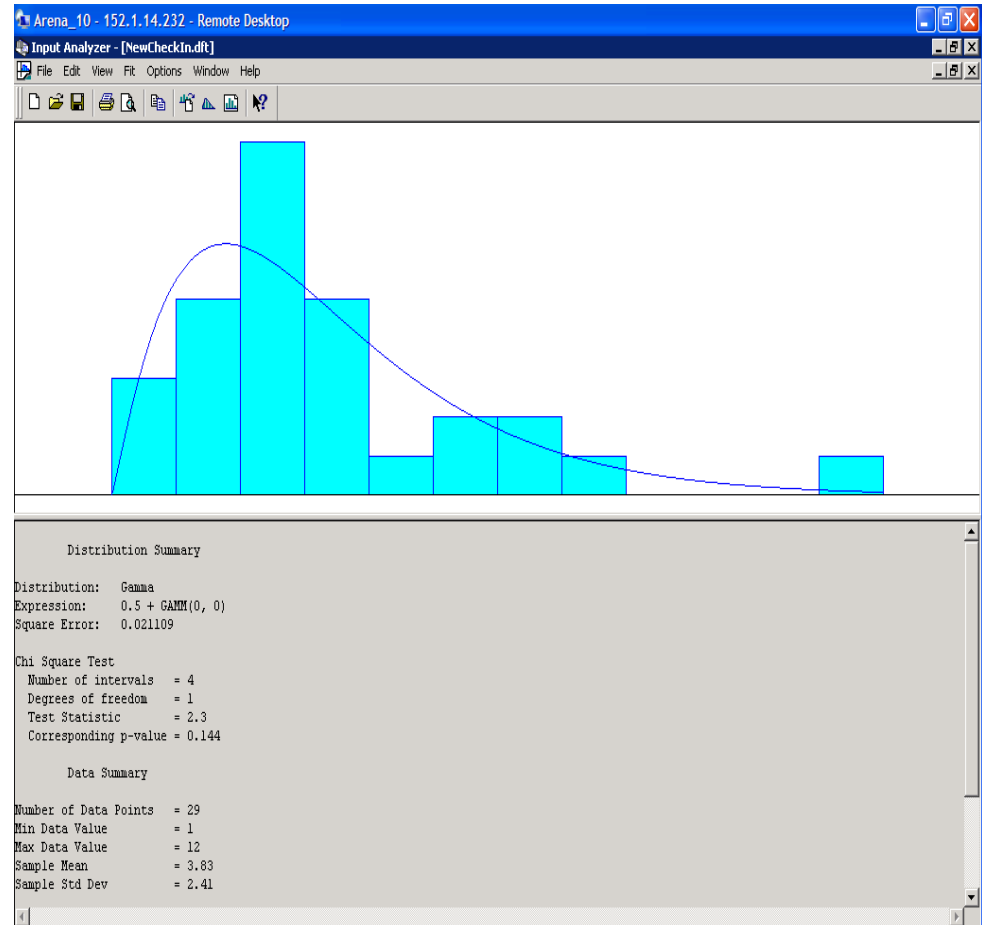
- Staffing levels were determined through communication with Operations Manager Kevin Tate
 - Given the total number of resources
 - Determined a logical distribution of resources
- The base model uses the typical staffing levels

	Number of Resources		Number of Resources
Attending Doctors		Exam Rooms	
Hall 1	2	Hall 1	8
Hall 2	2	Hall 2	8
Hall 3	2	Hall 3	8
Hall 4	2	Hall 4	8
Residents		Hall Check Out	
Hall 1	2	Hall 1	1
Hall 2	1	Hall 2	1
Hall 3	2	Hall 3	1
Hall 4	1	Hall 4	1
Nurses		Preceptors	
Hall 1	4	Side 1	1
Hall 2	3	Side 2	1
Hall 3	4	Vital Carts	
Hall 4	3	Side 1	3
Vital Rooms		Side 2	3
Hall 1	1	Front Desk	6
Hall 2	1	Radiology	1
Hall 3	1	Laboratory	1
Hall 4	1		



Process Time Distributions

- Time distributions for all process activities were determined using Arena Input Analyzer
 - Part of the Arena Software package
 - Allows fitting a distribution to a set of data



Validation of Base Model

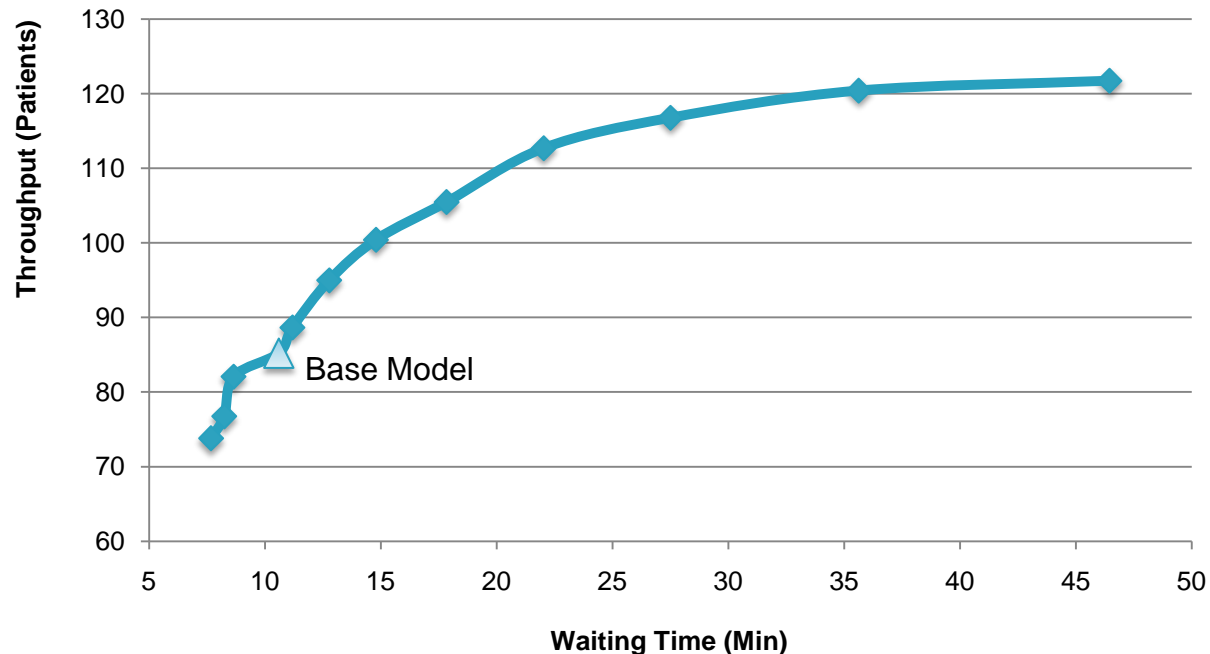
Base Model Results		
	Per Shift	Per Day
Patients	89.4	178.8
Throughput	85.2	170.4
Patients Left in Clinic	4.2	8.4
Cycle Time (min)	51.82	51.82
Wait Time (min)	10.6	10.6
Value Added Time (min)	37.55	37.55

- A comparison of our values of the primary performance measures show that our model is valid.
- Both our base model and the clinic have patients left in the system at the end of a shift.

Performance Measures

We utilized two primary performance measures to validate our model:

- **Throughput:** number of patients who pass through clinic during a shift
- **Cycle Time:** how long patients are in the clinic
 - Waiting Time: time spent waiting for a resource to become available
 - Value Added Time: the amount of time spent with a resource



Results of Experiments

Experiment 1: Adding Two Physicians

- Base Model – 14 physicians per shift
 - 3 physicians on hallways 1 & 3
 - 4 physicians on hallways 2 & 4
- Proposed scenario: Adding two physicians
 - Adding one physicians to hallway 1
 - Adding one physicians to hallway 3
 - Result: Each hall has 4 physicians



Results of Experiments

Results:

	Base Model	Adding Two Doctors
Throughput (patients)	85.2	99.02
Waiting Time (minutes)	10.6071	10.5814

– Resources accommodated the higher volume

Recommendation: hire two additional physicians



Results of Experiments

Experiment 2: No hall checkout rooms

	Base Model	No Hall Checkout Rooms
Throughput (patients)	85.2	85.63
Waiting Time (minutes)	10.6071	10.4408

Recommendation: Convert the hall checkout rooms for other uses

Results of Experiments

Experiment 3: Altering number of exam rooms per hall

	Base Model	7 Exam Rooms	9 Exam Rooms
Throughput (patients)	85.2	85.36	85.26
Waiting Time (minutes)	10.6071	10.5466	10.6287

Recommendation: Do not add additional exam rooms



Results of Experiments

Experiment 4: Staggered Shifts

- Base model – 2 shifts
 - 8 AM – 12 PM; 1 PM – 5 PM (approximate)
 - 14 physicians each shift
- Adding a third shift
 - Shift 1: 8 AM – 12 PM, 9 physicians
 - Shift 2: 10 AM – 2 PM, 10 physicians
 - Shift 3: 12 PM – 4 PM, 9 physicians
- Desired result: help smooth scheduling



Results of Experiments

Results:

	Base Model	Adding a Third Shift
Daily throughput (patients)	170.4	164
Waiting Time (minutes)	10.6071	10.5814

Recommendation: Do not stagger shifts

Summary

- Some things that seemed like obvious ways to improve the system were not
- The model building process generated many more ideas for experiments
- Evaluation of new ideas is now “cheap”

A Surgery Scheduling Decision Support System

North Carolina State University
Edward P. Fitts Department of Industrial and Systems Engineering
Senior Design Project

Zach Adams
Kristin Moore
Ashley Wampler

Sponsored by:
Mayo Clinic
Thoracic Suergery

Advised by:
Todd Huschka, Mayo Clinic
Hari Balasubramanian, Mayo Clinic
Dr. Brian Denton, North Carolina State University

April 23, 2008



Project Goal

- Develop a decision support system for automated surgery scheduling
- Design a heuristic for computing near optimal surgery schedules given OR and scheduling constraints
- Enable decision makers to manually change surgery schedules
- Provide advanced performance measures for the analysis of surgery schedules

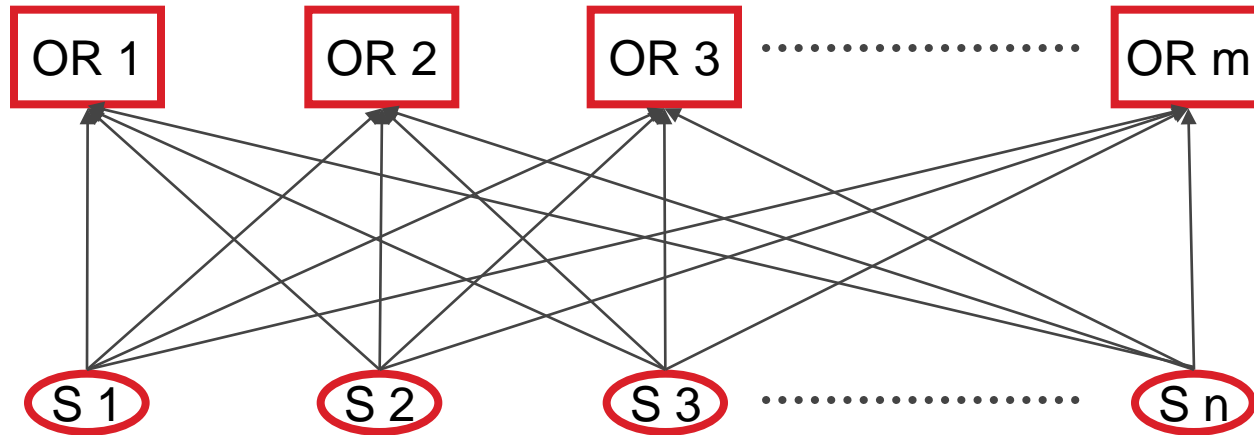


Problem Description

- Thoracic Unit Surgery Scheduling
 - 4 primary surgeons, multiple residents
 - 22 different procedures
 - Average of 1,500 patients per year
 - Typically 6 operating rooms
- Current method
 - Manual
 - No formal heuristic used



Multi-Operating Room Scheduling



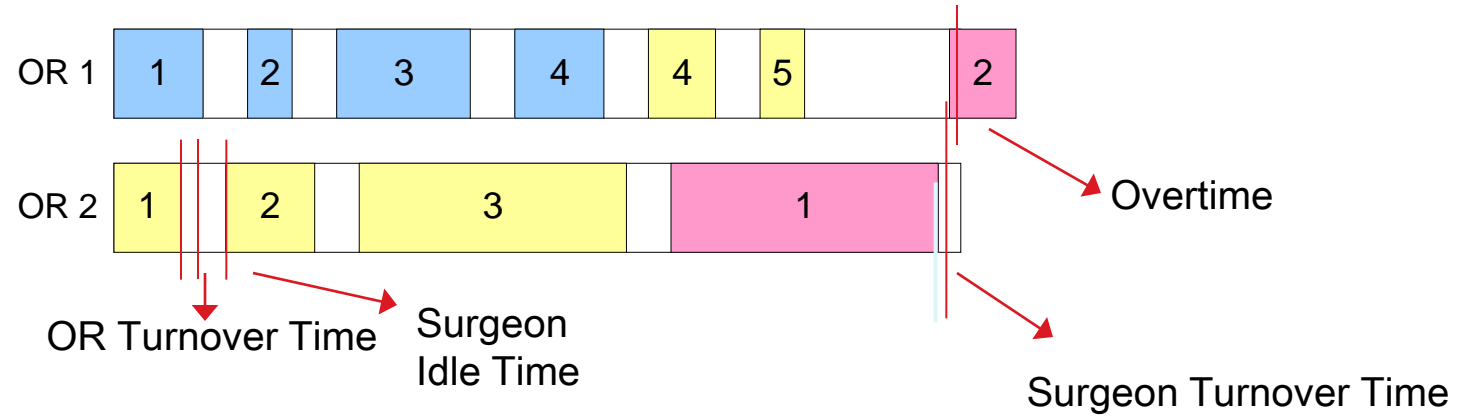
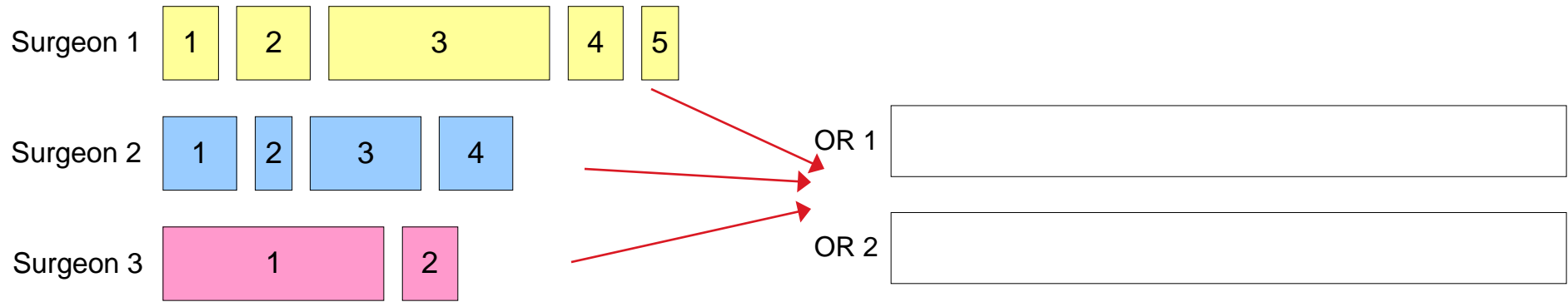
Decisions:

- How many ORs to open?
- Where to schedule each surgery?

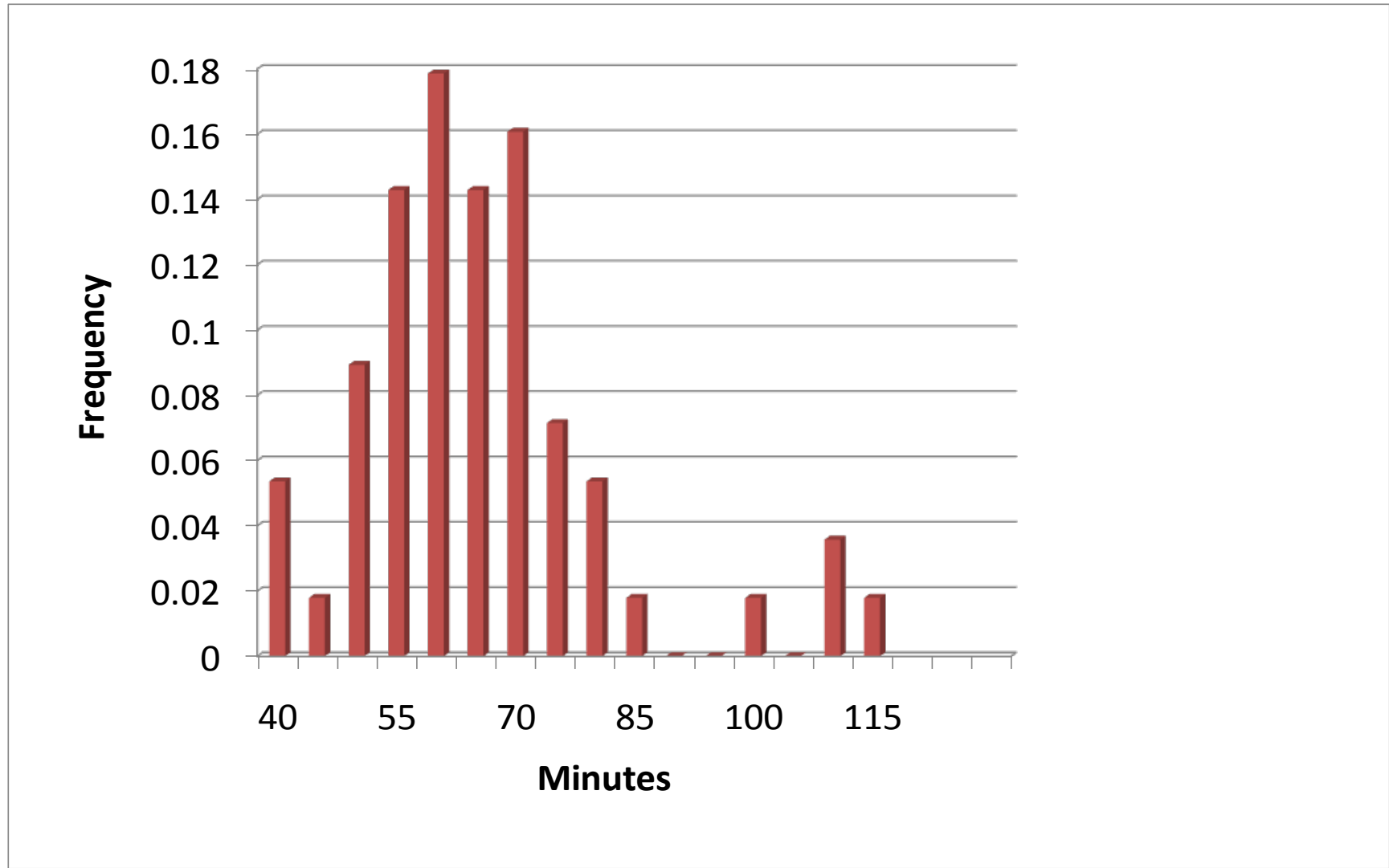
Performance Measures:

- Cost of operating rooms opened
- Overtime costs for operating rooms

Resource Constraints



Surgery Duration Variability



Simulation Model

- A Discrete event simulation model
 - Inputs:
 - Automatically generated or user selected surgery schedule
 - User defined parameters (# of ORs, # of surgeries, type of surgeries, cost of OR, cost of overtime)
 - Outputs:
 - Average overtime
 - Probability of overtime in each OR
 - Total cost (cost of ORs + cost of overtime)



Data Sampling

- 22 Different Types of surgeries
 - Various combinations
 - Up to five surgeries per patient
- Over 3 million combinations
- “Scenarios” generated by statistical sampling from historical data

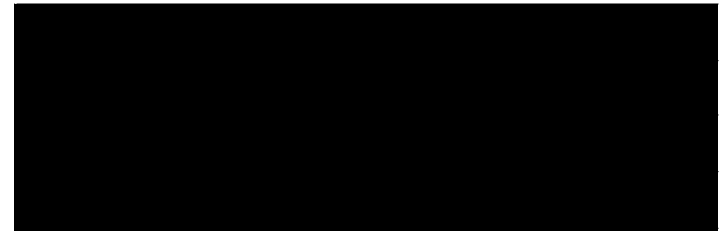
Automated Scheduling Algorithm

- Compute mean duration for each patient
- Group by resident surgeon
- Select the lower bound number of OR's needed
- Schedule resident with longest duration to least loaded OR
 - Repeat until all residents have been scheduled
- Compute over time statistics
- View optimized schedule



Algorithm Example

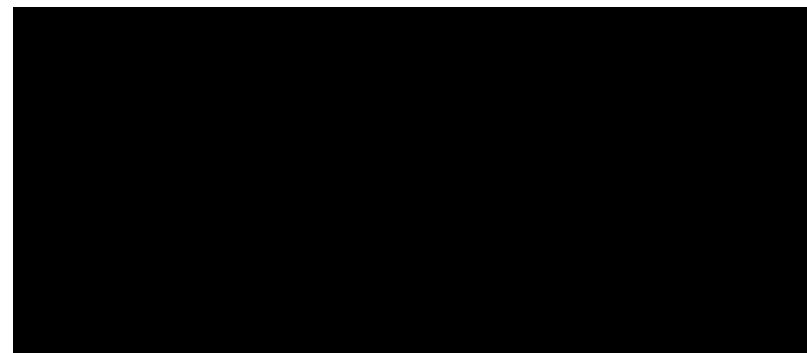
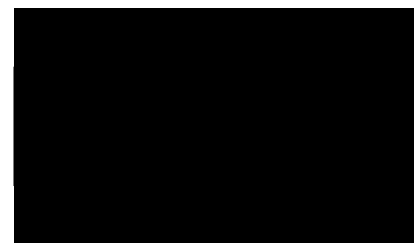
	Sample Surgery Time (mins)	Resident
Patient 1	100	1
Patient 2	120	2
Patient 3	300	3
Patient 4	150	5
Patient 5	110	6
Patient 6	75	1
Total Time	855	



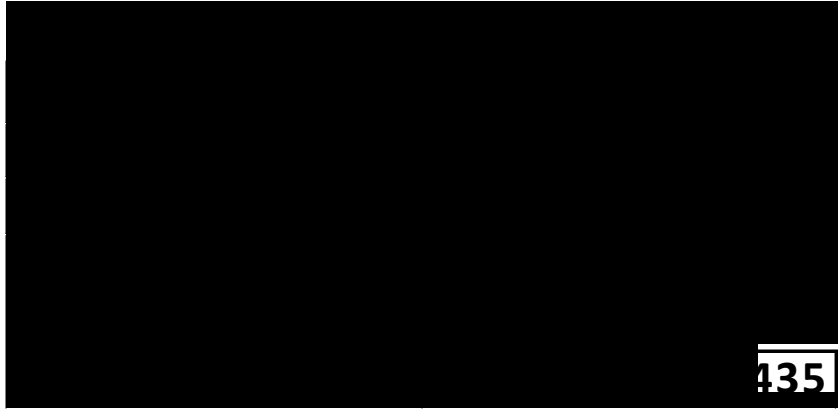
$$\text{Minutes per day} = 9 \text{ hours} * \frac{60 \text{ min}}{\text{hr}} = 540 \text{ minutes}$$

$$\text{Number of ORs} = \frac{\text{Surgery Minutes}}{\text{Minutes per day}} = \frac{855 \text{ min}}{540 \text{ min}} = 1.583 = 2$$

Algorithm Cont.



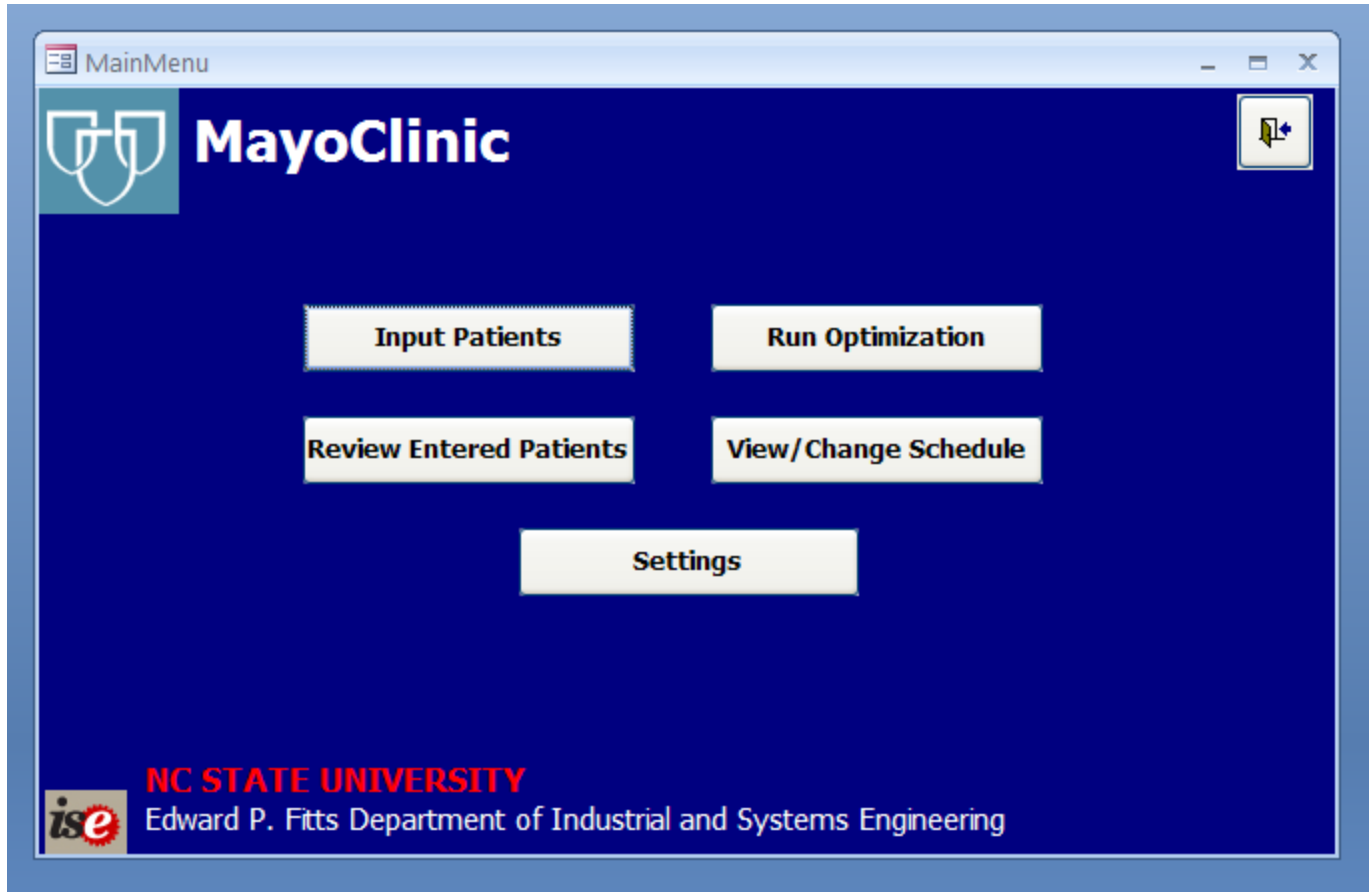
Algorithm Cont.



Algorithm Cont.

- Allowable overtime is set by user
- If any OR's time exceeds the max allowable time in minutes:
 - (1) increase the number of open OR's by one
 - (2) recalculate the new schedule
- Return the best (minimum cost) schedule found

Demo



DSS Scheduling vs. Current Scheduling

	Actual Rooms Opened	Actual Base Cost	DSS Rooms Opened	DSS Cost
1/3/2005	4	\$ 17,748	3	\$ 15,165
1/4/2005	6	\$ 26,622	5	\$ 24,997.00
1/5/2005	4	\$ 17,748	3	\$ 14,322.00
1/6/2005	3	\$ 13,311	4	\$ 18,946.00
1/7/2005	4	\$ 17,748	4	\$ 18,207.00
1/26/2005	4	\$ 17,748	3	\$ 15,349.00
11/13/2006	5	\$ 22,185	4	\$ 20,415.00
11/14/2006	4	\$ 17,748	2	\$ 9,148.00
11/15/2006	6	\$ 26,622	4	\$ 18,430.00
Total Cost		\$ 177,480		\$ 154,979
Average Savings/Day		\$ 2,500		
Average Savings/Year		\$ 912,540.56		



Summary

- There are many opportunities to use systems engineering methods to improve health care delivery
- Environments that are “data rich” are well positioned to use systems engineering
- A small number of (the right) resources can go a long way!