

Modeling Integrated Public Health and Health Care Delivery System Response to Anthrax

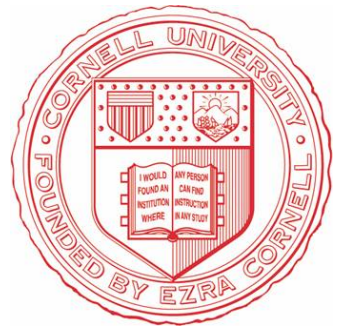
Nathaniel Hupert, MD, MPH

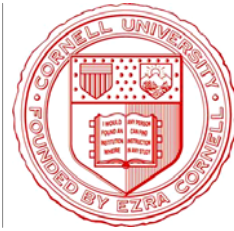
Departments of Pubic Health and Medicine
Weill Cornell Medical College
Cornell University

IOM Workshop, March 3 2008

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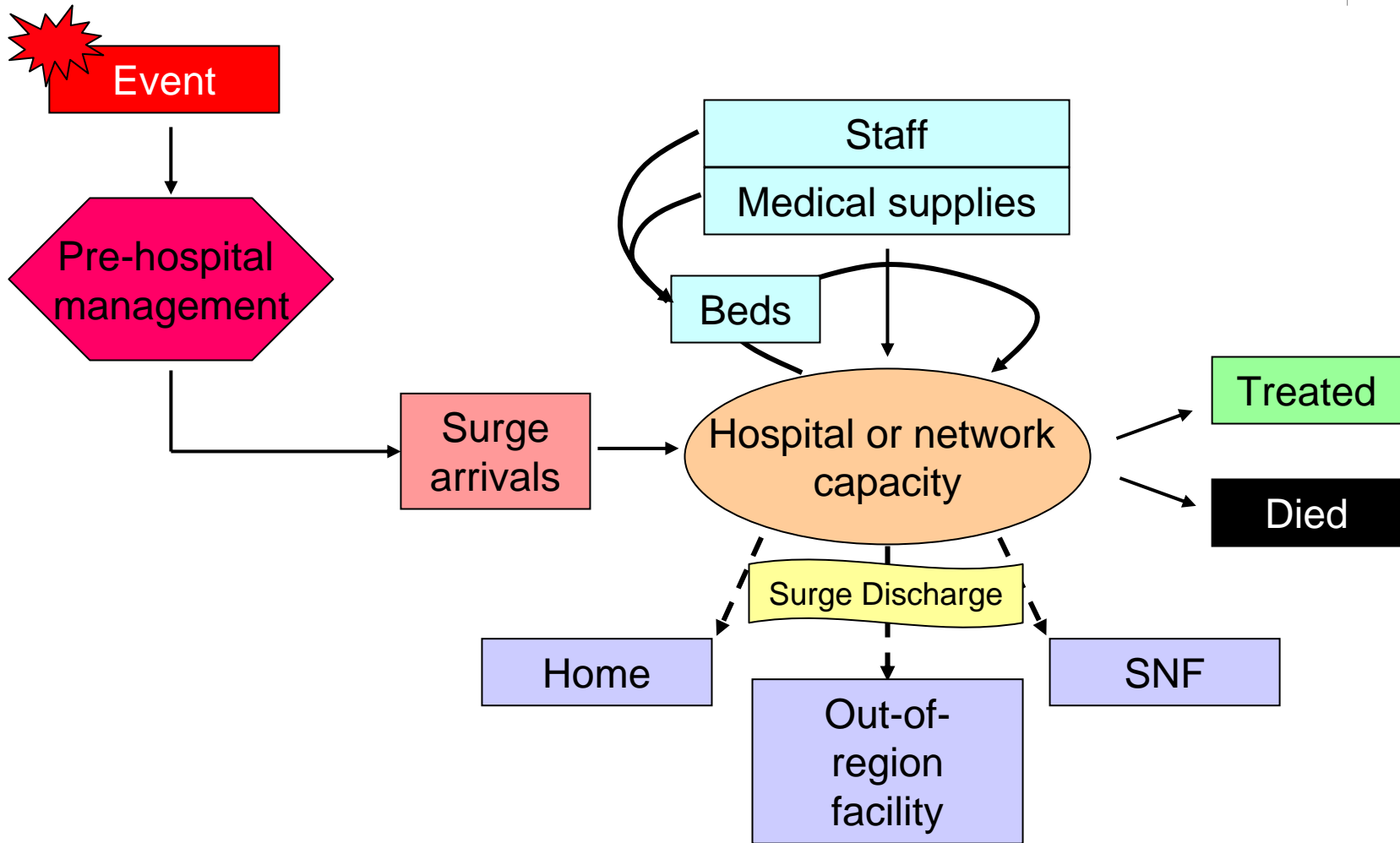
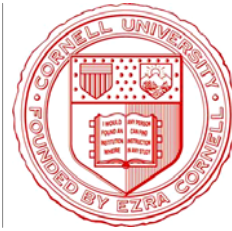


Goals of Mass Prophylaxis

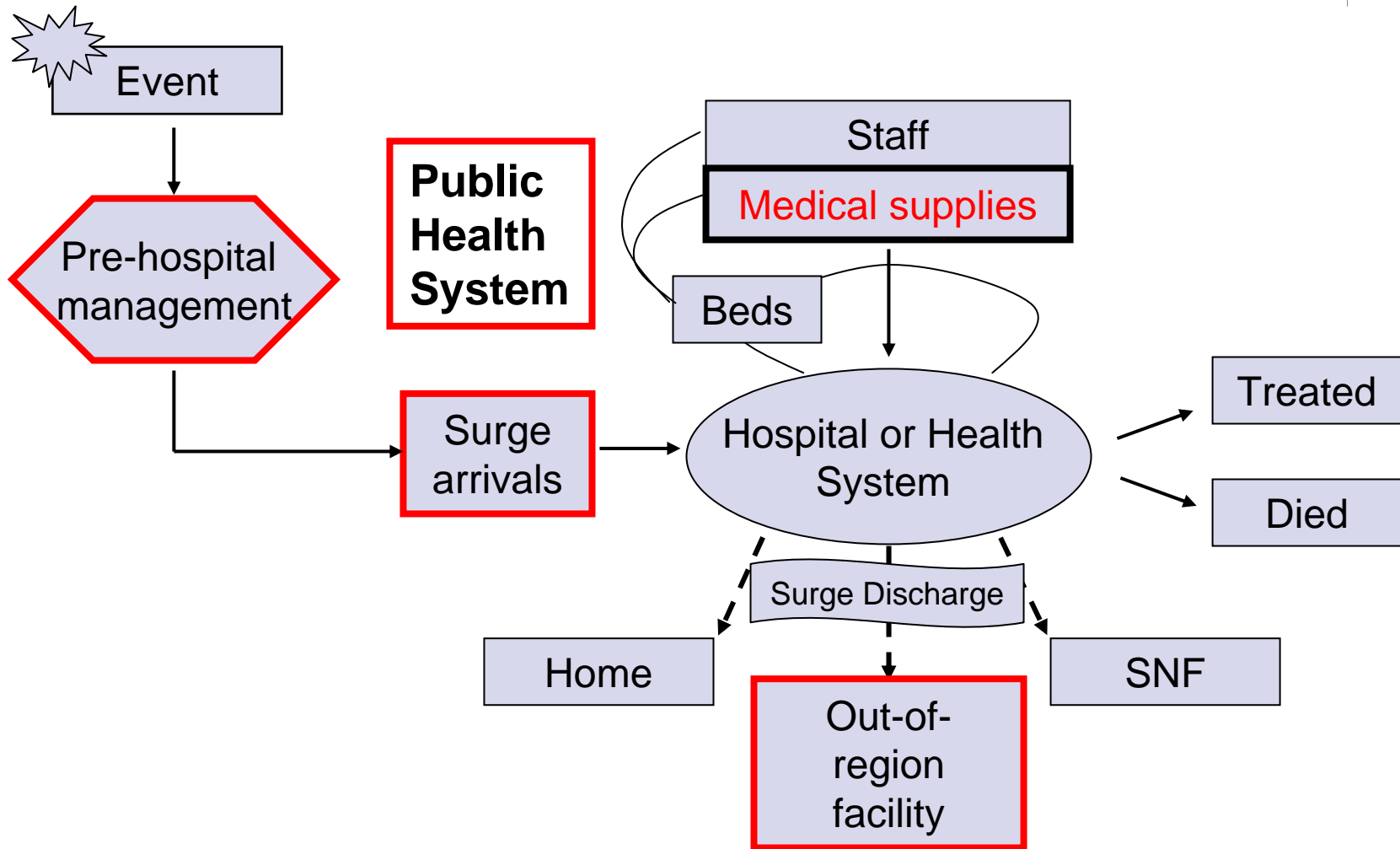
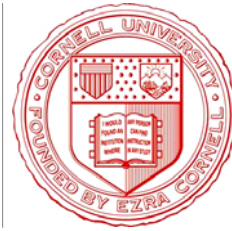
- 48 Hours! (the “when”)
- Maximize “Pills in People” (the “what”)
- Multiple secondary goals (the “why”):
 - Prevent illness
 - Social stability
 - **Prevent surge on medical treatment facilities**
 - Supply chain stability

What outcome metrics can define operational parameters for successful campaigns?

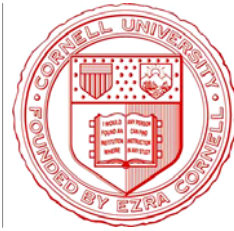
Modeling Schema



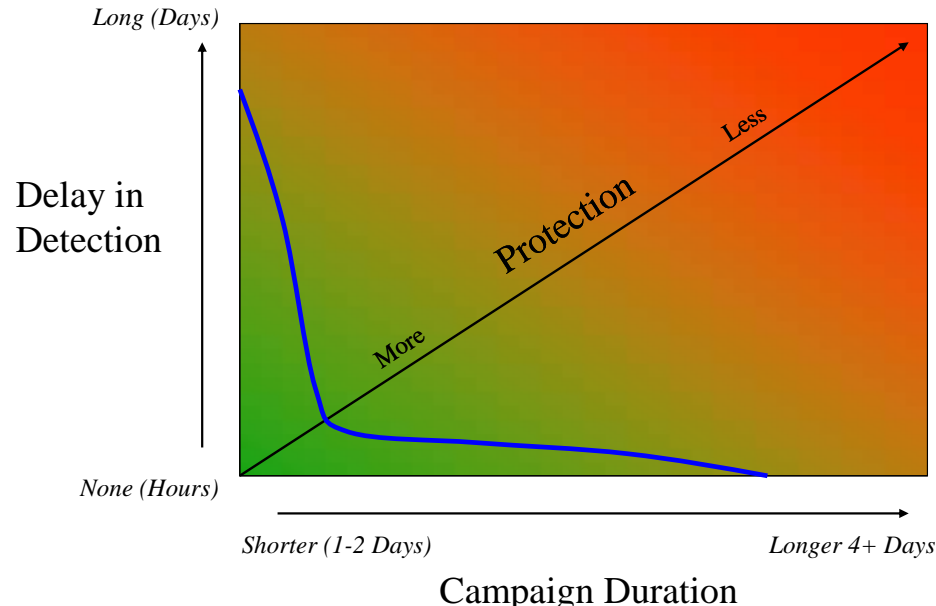
Modeling Schema



Weill Cornell Regional Hospital Caseload Calculator (AHRQ, 2003)



- State transition (compartment) model
- Excel-based, end-user oriented
- Contributed to 48h goal for CRI



Weill/Cornell Regional Hospital Caseload Calculator (RHCC) Model

Created by Christopher Neukermans, Jason Cuomo MPH, Mary Koshy MPA, and Nathaniel Hupert MD MPH

Under Contract 290-00-0013 from the DHHS Agency for Healthcare Research and Quality (AHRQ)

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Caseload Calculator Inputs

Caseload
Calculator
Results



1. Please select the disease and incidence curve from the list below.

2

1. Anthrax, USA-USSR Normal curve (mean 8.3, st. dev. 3.0)
2. Anthrax, (Brookmeyer) Log-normal curve (mean 2.398, st. dev. 0.713)
3. Anthrax, Log-normal curve (mean 6, st. dev. 2)
4. Anthrax, Normal curve (mean 6, st. dev. 2)
5. Anthrax, Gamma curve (alpha 9, beta 0.666)
6. Anthrax, Poisson distribution (mean 6)
7. Anthrax, Weibull curve (Brookmeyer, alpha 0.00785, beta 1.67)
8. Bubonic Plague, Log-normal curve (mean 4.3, st. dev. 1.8)
9. Bubonic Plague, Log-normal curve (mean 2.5, st. dev. 1.2)

2. What is your total population/community size?

2,000,000 people

3. How many people are exposed to the pathogen? (e.g. What is the attack size?)

50,000 people exposed

4. What percentage of exposed people develop symptoms if no treatment?

85%

5. How many days after the attack can the prophylaxis campaign begin? (i.e. 1 = day of attack)

2 day(s) after the attack

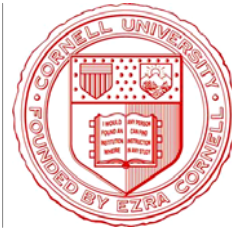
6. From the start of the campaign, how many days does it take to reach maximum prophylaxis capacity?

1 day(s) after the attack

7. What is the maximum number of people to whom you could deliver prophylactic antibiotics per day at peak operational capacity?

1,000,000 people per day which equals a 2 day campaign

Protection via Prophylaxis with Brookmeyer Anthrax Incidence Curve (Optimal Case)

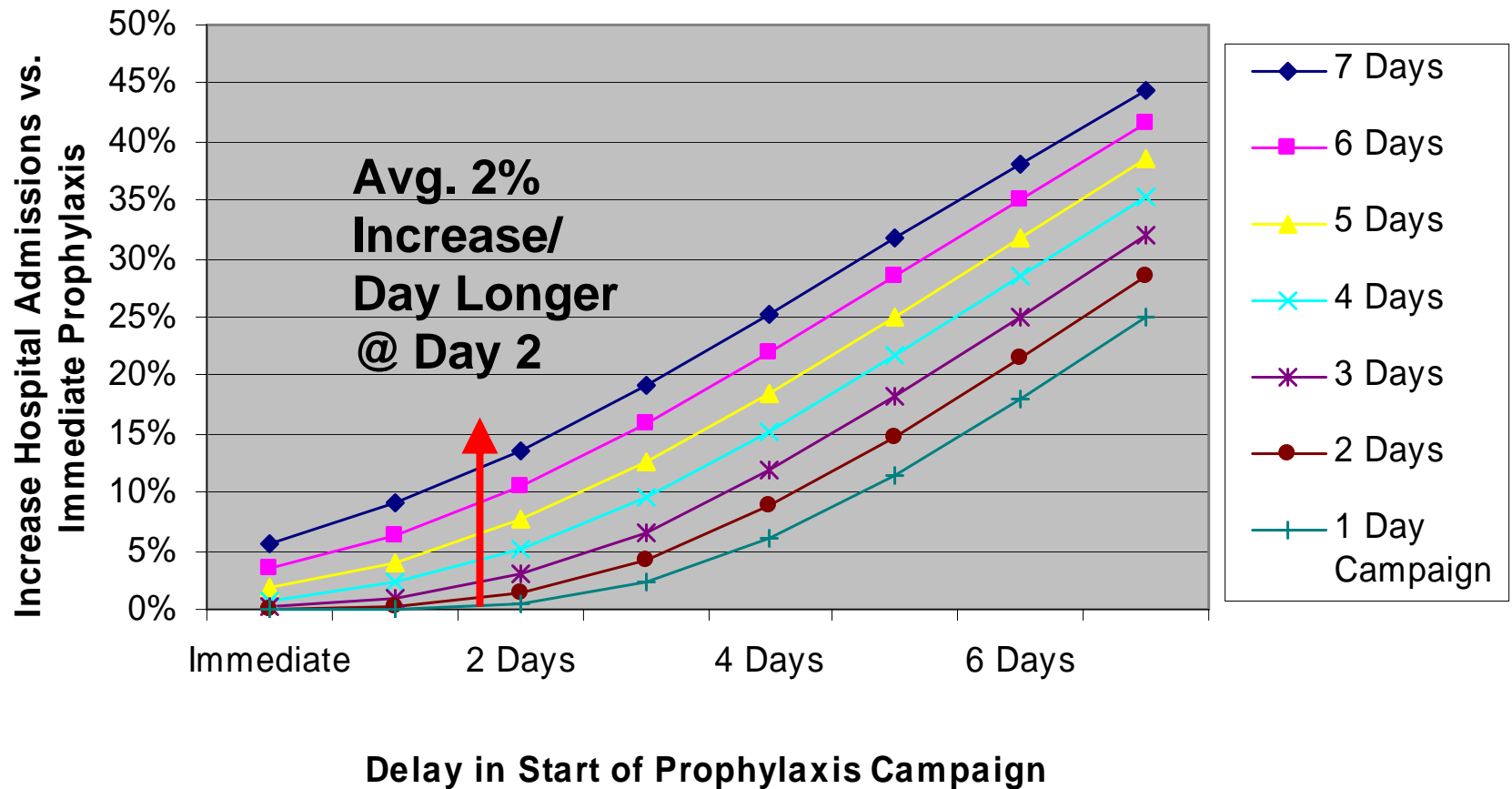


Campaign Duration Delay in Detection

	Immediate	1 Day	2 Days	3 Days	4 Days	5 Days	6 Days	7 Days
10 Days	84%	80%	74%	69%	63%	57%	51%	46%
9 Days	87%	83%	77%	72%	66%	60%	54%	48%
8 Days	90%	86%	81%	75%	69%	62%	56%	50%
7 Days	93%	89%	84%	78%	72%	65%	59%	53%
6 Days	95%	92%	87%	81%	75%	68%	62%	56%
5 Days	97%	94%	90%	85%	78%	72%	65%	59%
4 Days	99%	96%	93%	88%	82%	75%	68%	62%
3 Days	99%	98%	95%	91%	85%	78%	72%	65%
2 Days	100%	99%	97%	94%	88%	82%	75%	68%
1 Days	100%	100%	99%	96%	91%	85%	79%	72%

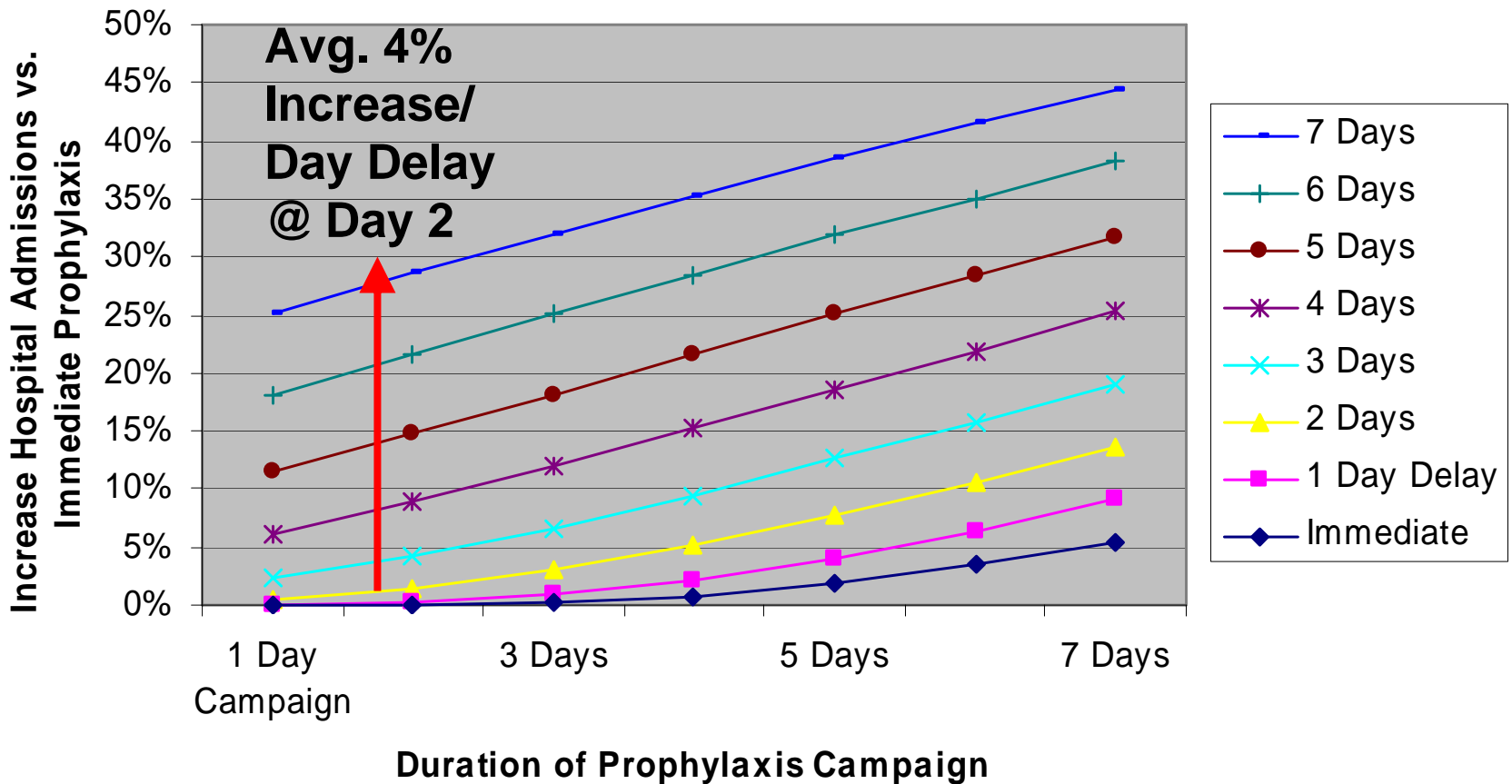
Weill Cornell Regional Hospital Caseload Calculator Output

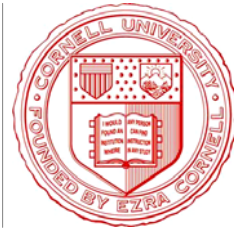
Increasing Hospital Surge with Slow Mass Prophylaxis: Impact of Duration of Campaign



Weill Cornell Regional Hospital Caseload Calculator Output

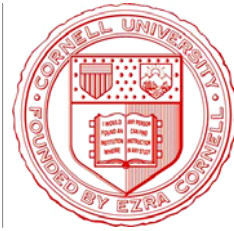
Increasing Hospital Surge with Slow Mass Prophylaxis: Impact of Delay to Initiation of Campaign





RHCC Operational Parameters

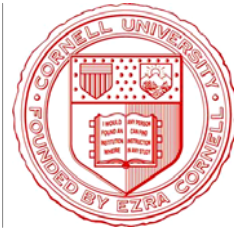
- For every day that an ideally-deployed anthrax mass prophylaxis campaign is shortened, there may be a 2-6% decrease in overall hospitalizations
- To accomplish this in the first week, POD efficacy must be increased by anywhere from 14% (7→6d) to 33% (3→2 days)
- *These are large increases in POD efficacy!*



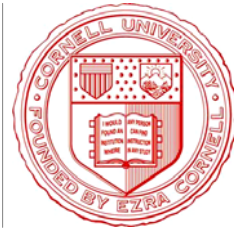
Determinants of POD Efficacy

1. **“Fill correctly”**
 - Inter-POD network load balancing
 - Intra-POD load balancing
2. **“Process efficiently”**
 - Staff availability and effectiveness
3. **“Sustain appropriately”**
 - Supply chain responsiveness and resiliency

Evaluation of Alternative POD Designs

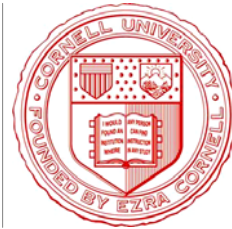


- Construct a stochastic simulation model that dynamically measures throughput capacity and patient flow times
- Experiment with alternative POD design strategies considering the effects of:
 - Timing of the demand for service over the planning horizon
 - Quantity of demand over time at each POD
 - Capacity (quantity and capabilities) of POD staff



1a. External Load Balancing

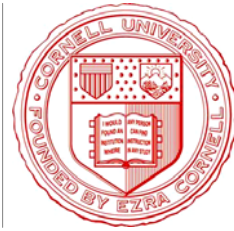
<i>POD</i>			<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>Per POD Per hour</i>
Balanced	Throughput	Unadjusted	5911(±13)	5901(±15)	5912(±12)	5918(±14)	493
		Adjusted	5993(±16)	5979(±15)	5946(±15)	5888(±19)	496
	Average number of people	Unadjusted	62.57(±7)	59.59(±7)	55.29(±7)	61.98(±8)	60
		Adjusted	32.75(±5)	32.50(±5)	32.25(±5)	32.01(±5)	33
Moderately Unbalanced	Throughput	Unadjusted	5991(±14)	6001(±18)	4793(±14)	3603(±12)	425
		Adjusted	5984(±15)	5981(±16)	5943(±15)	5903(±19)	496
	Average number waiting	Unadjusted	1206(±14)	605(±14)	3.17(±0.10)	0.89(±0.02)	454
		Adjusted	31.07(±3)	30.82(±3)	30.57(±3)	30.32(±3)	31
Significantly Unbalanced	Throughput	Unadjusted	6012(±18)	5993(±16)	4794(±13)	2389(±11)	400
		Adjusted	5983(±16)	5980(±16)	5951(±16)	5903(±19)	496
	Average number people	Unadjusted	1791(±16)	593(±13)	3.16(±0.1)	0.26(±0.01)	597
		Adjusted	28.05(±3)	27.80(±3)	27.54(3)	27.30(±3)	28



1a. External Load Balancing

POD			1	2	3	4	Per POD Per hour
Balanced	Throughput	Unadjusted	5911(±13)	5901(±15)	5912(±12)	5918(±14)	493
		Adjusted	5993(±16)	5970(±15)	5946(±15)	5888(±19)	496
	Average number of people	Unadjusted	62.57(±7)	59.59(±7)	55.29(±7)	61.98(±8)	60
		Adjusted	32.75(±5)	32.53(±5)	32.03(±5)	32.01(±5)	33
Moderately Unbalanced	Throughput	Unadjusted	5984(±15)	6001(±18)	4793(±14)	3603(±12)	425
		Adjusted	5984(±15)	5981(±16)	5943(±15)	5903(±19)	496
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Significantly Unbalanced	Throughput	Unadjusted	6012(±18)	5993(±16)	5911(±16)	5923(±14)	400
		Adjusted	5983(±15)	5970(±15)	5951(±16)	5903(±19)	496
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		Adjusted	28.05(±3)	27.80(±3)	27.54(3)	27.30(±3)	28

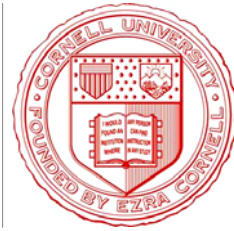
Role for Direct-to-Population Communications



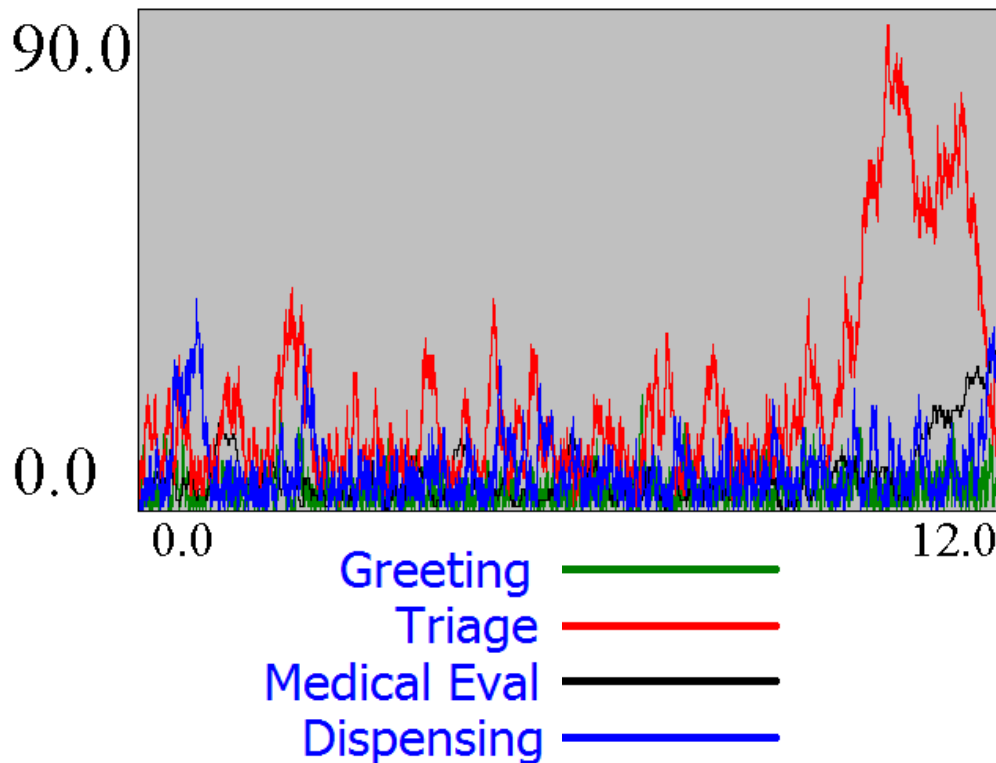
Traditional 500pph POD

- Classic station set-up
 - Greeting, Triage, Medical Evaluation, Dispensing
 - 5% G→ME, 5% T→ME
- Exponential processing times
 - G (20sec), T(1min), ME(5min), D(30sec)
- Optimized staffing using BERM
 - 4 Greeters, 9 Triage, 5 Med Eval, 5 Dispensers
- (G, T, D) considered cross-trainable
- 12-hour shift

1b. Internal Load Balancing: Baseline POD @500pph



Total Patients in POD

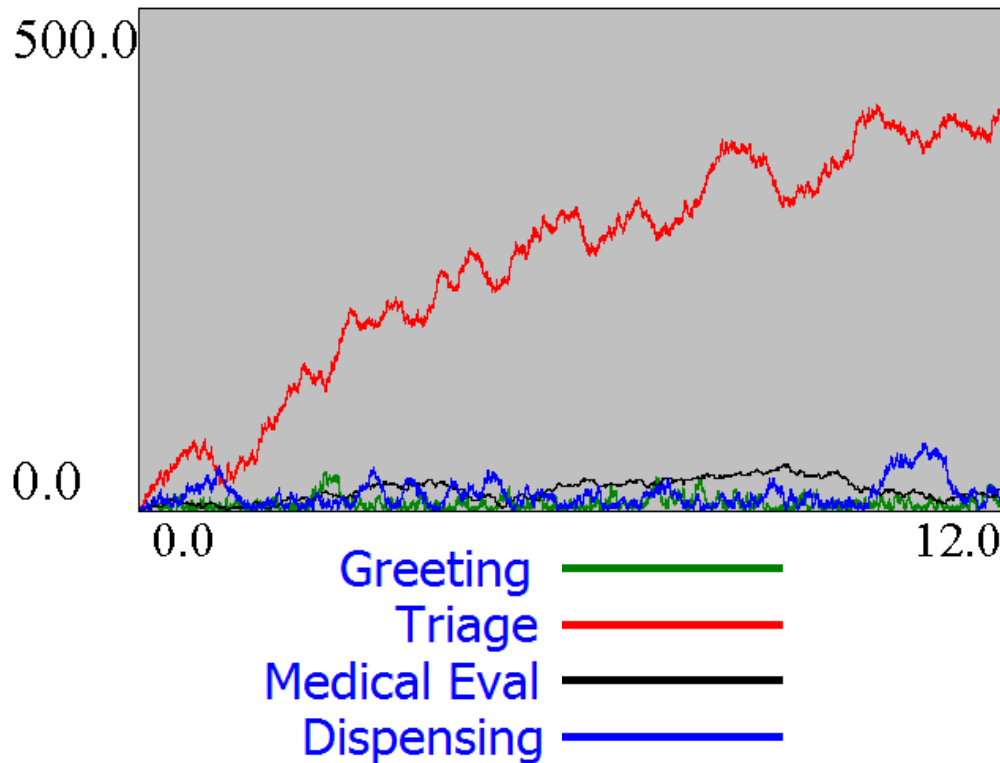


- Throughput
 - 497.3pph (+/- 1.3)
- Flow time
 - 3.5 minutes
- Average # in Queue
 - Greet 0.9 +/- 0.03
 - Triage 4.6 +/- 0.3
 - MedEval 2.3 +/- 0.3
 - Dispense 2.9 +/- 0.1

1b. Internal Load Balancing: Baseline POD @600pph

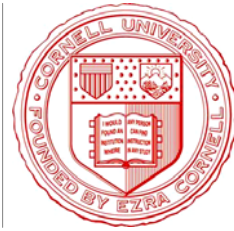


Total Patients in POD

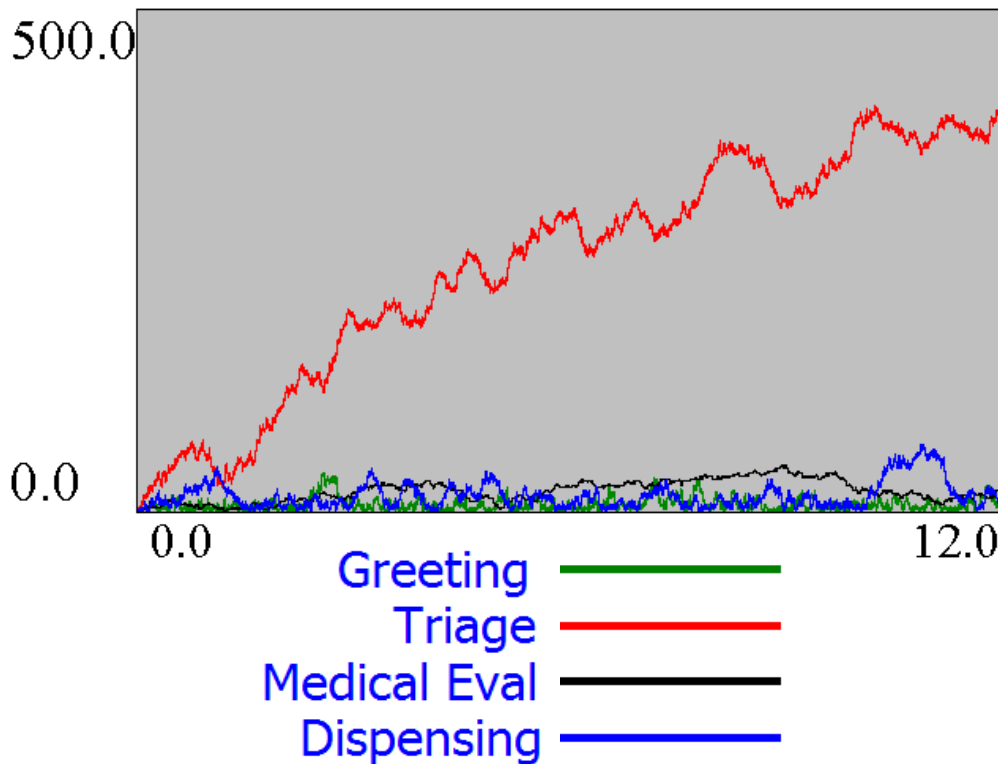


- Throughput
 - 563.8pph (+/- 1.3)
- Flow time
 - 24.3 min. (0→45 m)
- Average # in Queue
 - Greet 3.0 +/- 0.12
 - Triage 195.9 +/- 12.7
 - MedEval 8.9 +/- 1.3
 - Dispense 12.6 +/- 1.0

1b. Internal Load Balancing: Baseline POD @600pph



Total Patients in POD



Throughput

563.8pph (11/13)

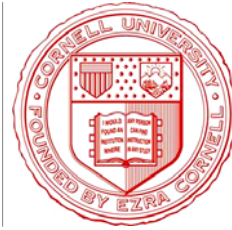
Flow time

24.0 min (0.7-45 min)

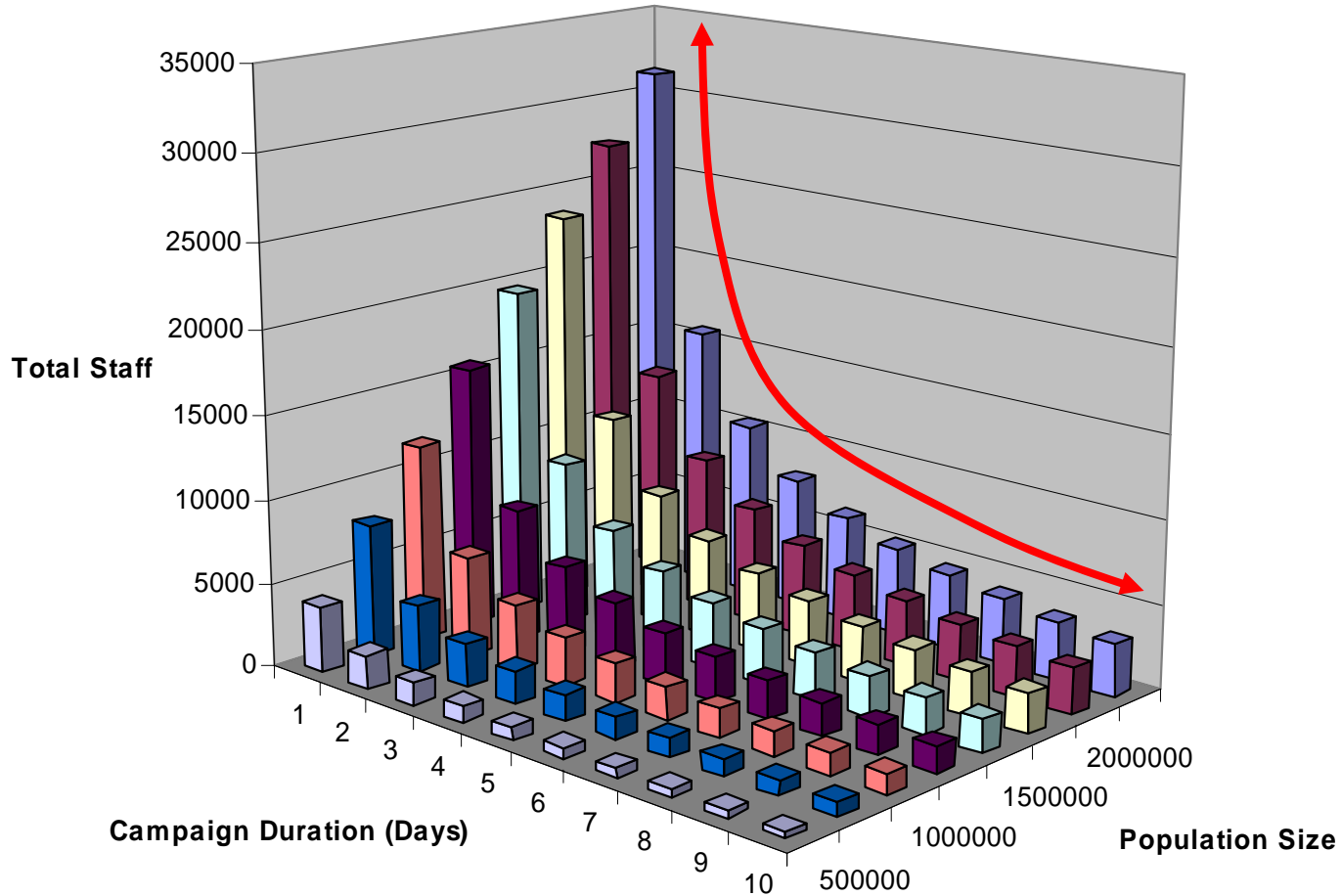
Average # in Queue

Check	3.0 +/- 0.12
Triage	9.7 +/- 12.7
MedEval	8.9 +/- 1.3
Dispense	12.6 +/- 1.0

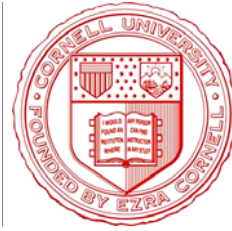
2. Time = Staff



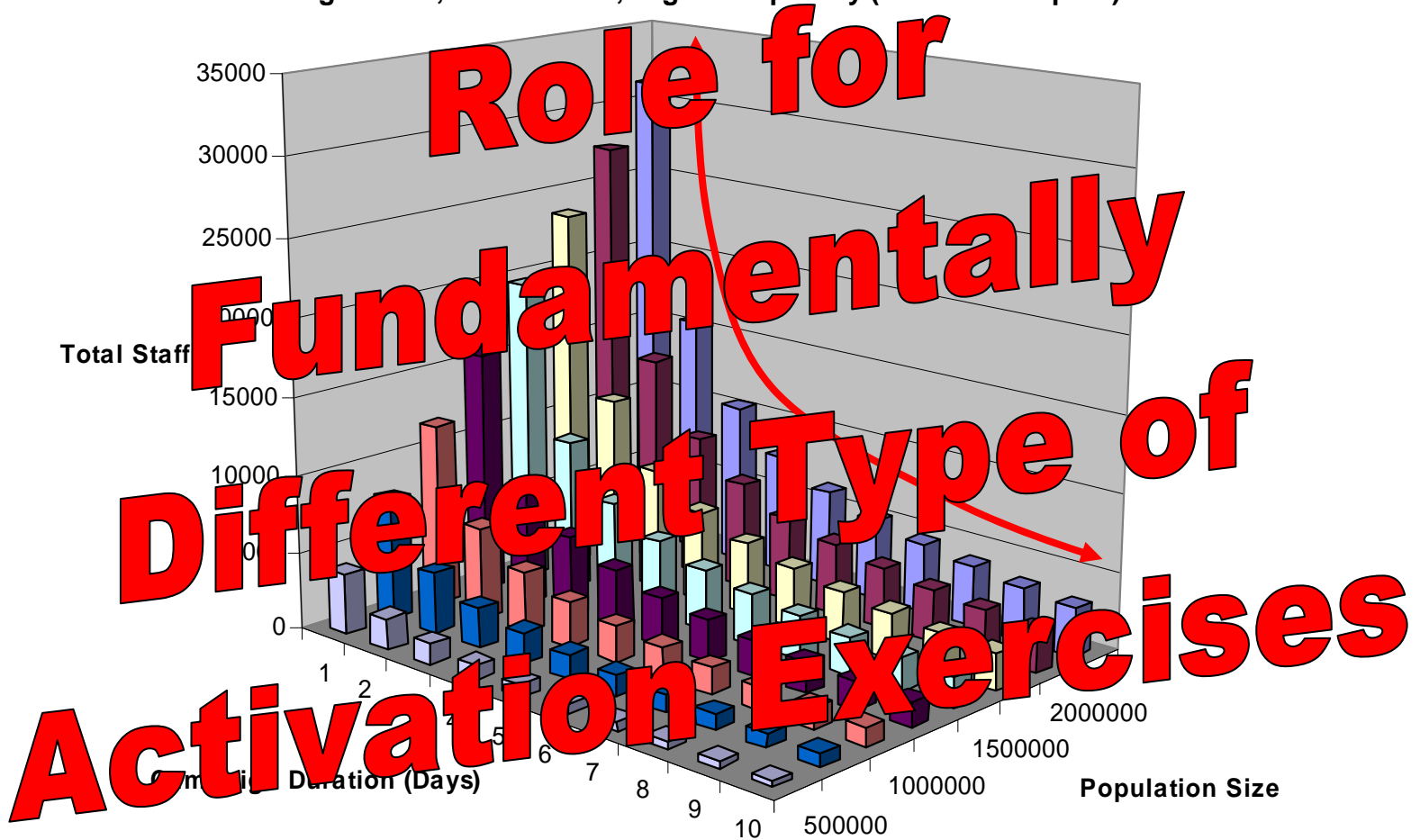
Staffing Needs, Athrax POD, High Complexity (BERM Floorplan)

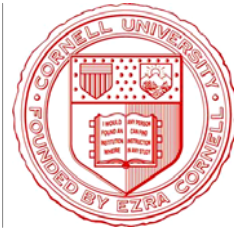


2. Time = Staff

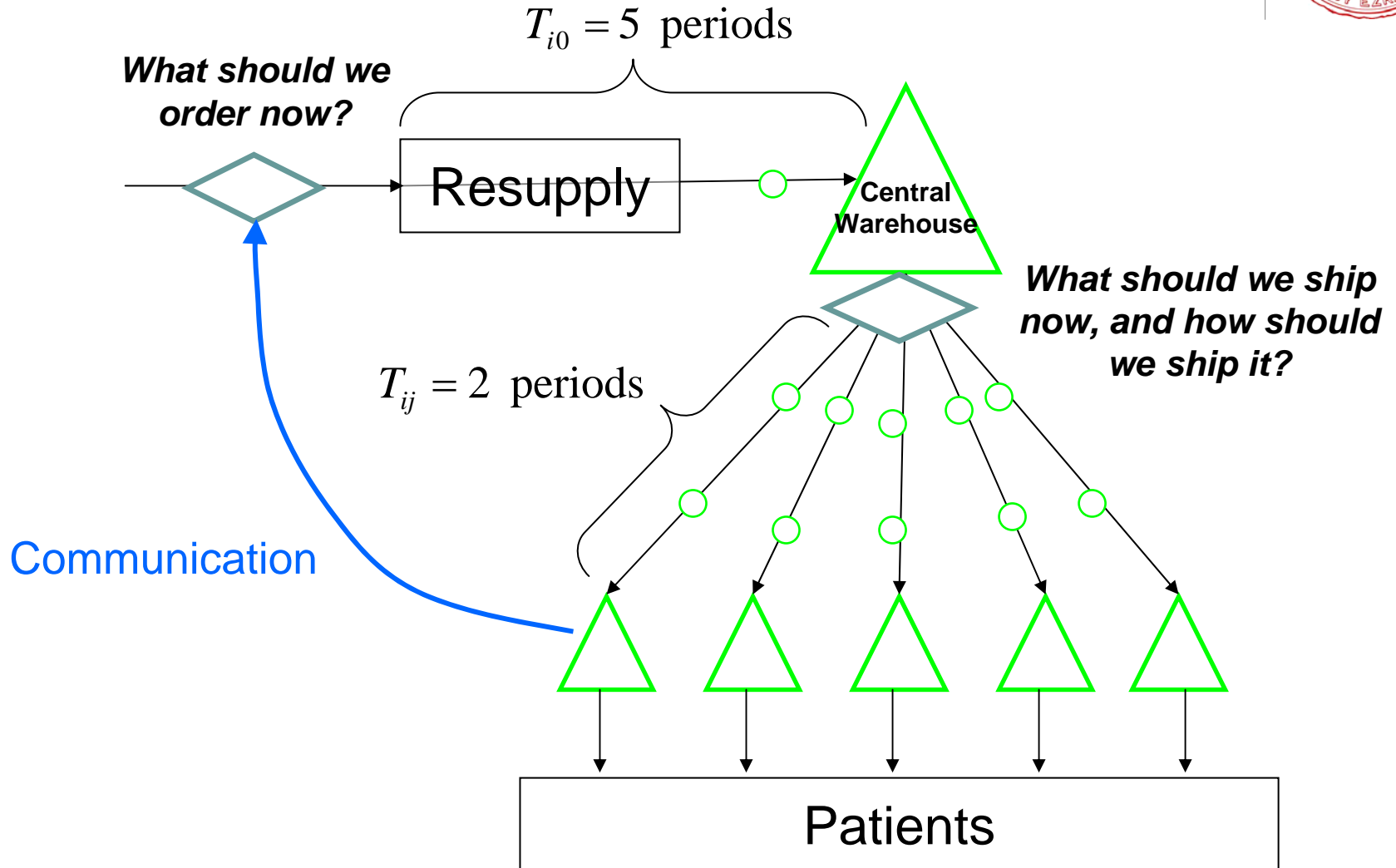


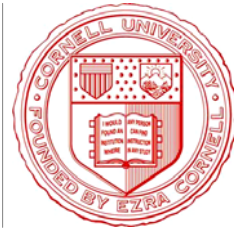
Staffing Needs, Athrax POD, High Complexity (BERM Floorplan)



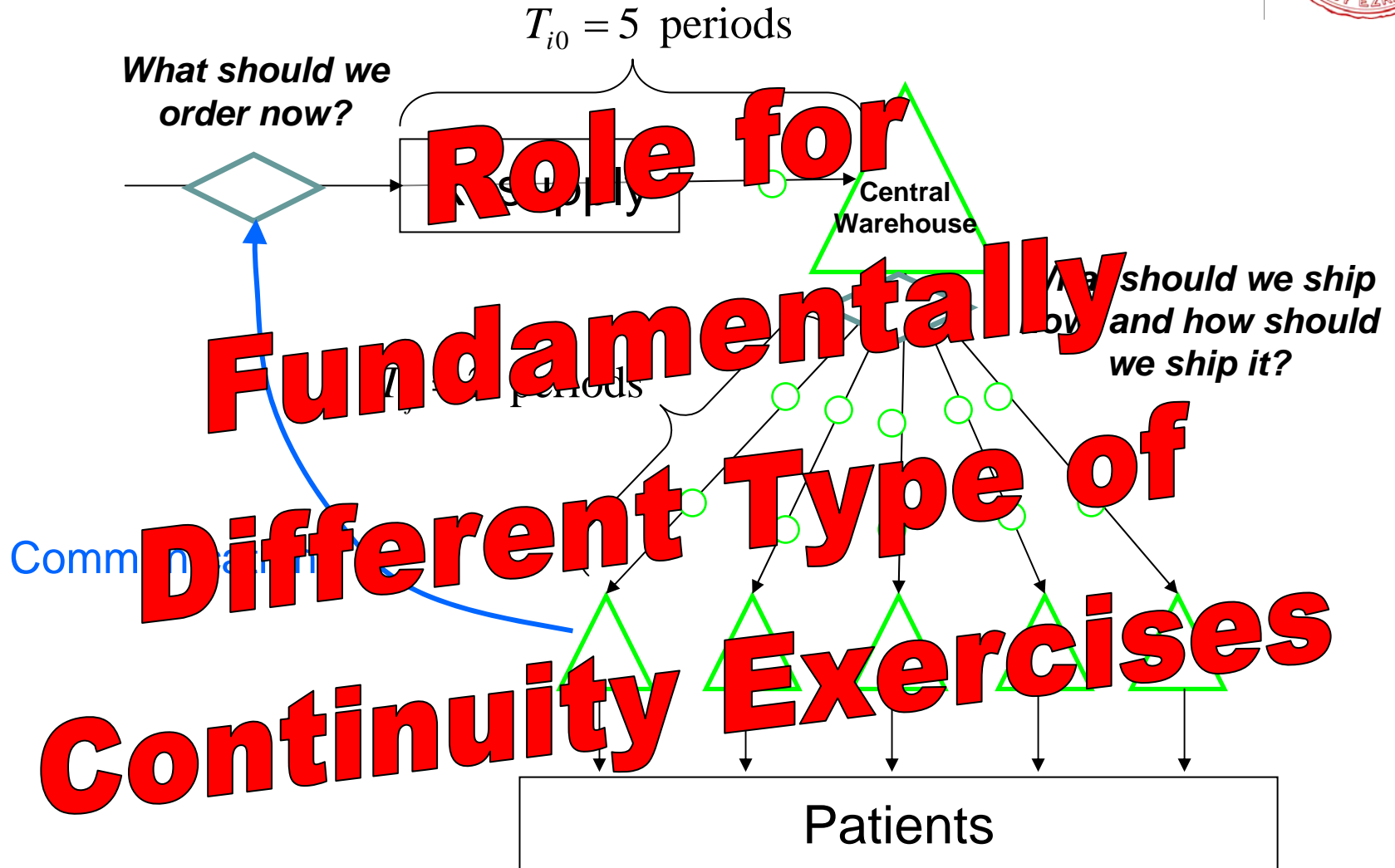


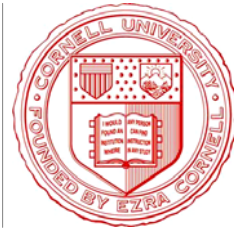
3. Inventory (Re-)Allocation





3. Inventory (Re-)Allocation

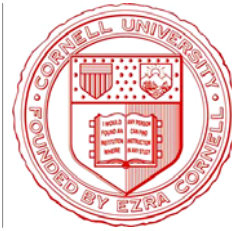




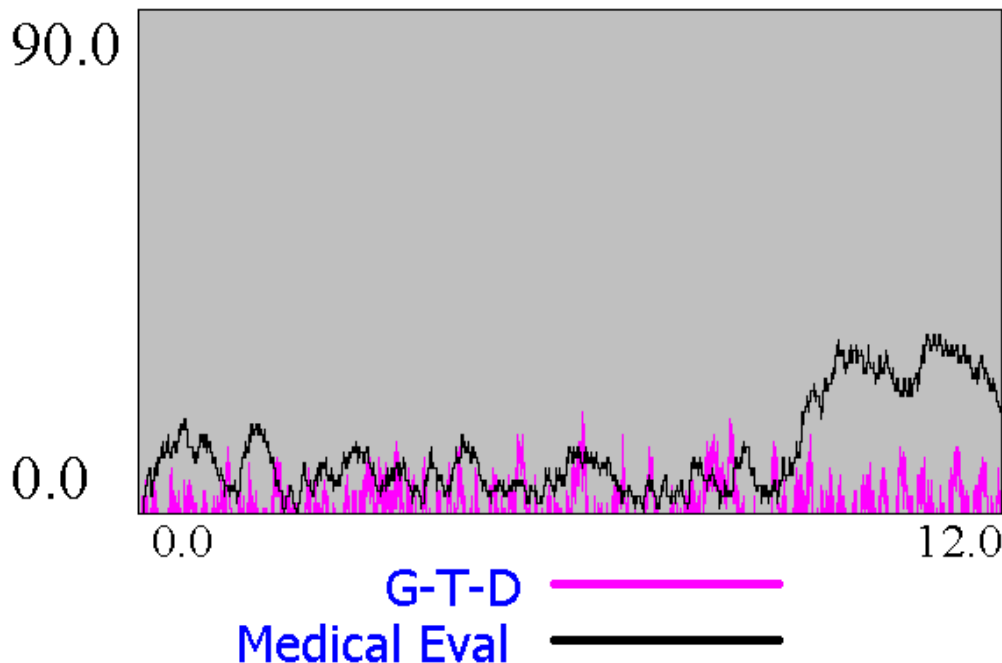
POD Alternatives

- “One-Stop Shop” in the POD
 - Eliminate the standard multi-station approach to internal POD design (goes further than
 - Extensive cross-training required
 - Drive-through may be variant of this
 - Certain stations may remain distinct due to different cadre of worker

One-Stop-Shop POD @600pph (Same total # staff)

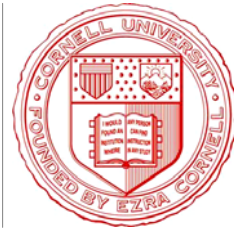


Total Patients in POD



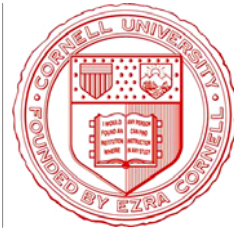
- Throughput
 - 597.3pph (+/- 1.4)
- Flow time
 - 3.2 minutes
- Average # in Queue

G-T-D	3.2 +/- 0.2
MedEval	7.4 +/- 0.9



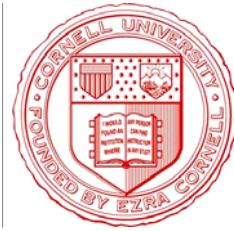
Other Alternatives

- Assortment of POD and non-POD approaches
 - Primary Care Provider Sites
 - Med Kit
 - Postal
 - Others to be discussed



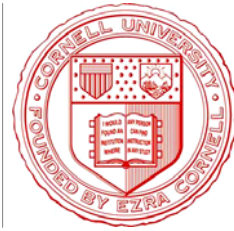
Modeling Realistic PODs

- Though the effects of uncertainty can be mitigated *to some extent* through collaboration and communication, fundamental uncertainty will always exist.
- Systems need to be designed so that they are robust in the presence of this uncertainty
- Models should capture and represent the implications of uncertainty on outcomes



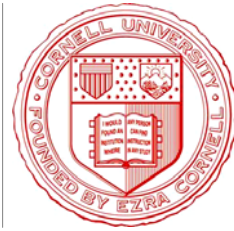
POD Planning Goal

POD systems needs to be *maximally effective and efficient* under conditions of *extreme uncertainty* relating to *care-seeking* and *resource availability*.



BERM +

- We are creating a collection of interacting models for **designing** POD systems that
 - Represent the dynamic nature of arriving care-seekers and calculate optimal time-varying POD resource requirements
 - Allocate arriving care-seekers optimally among PODs so as to minimize POD resource requirements and service times
 - Establish POD layout strategies that minimize care-seeker service times (maximize POD worker effectiveness)
- We are determining corresponding heuristics for **managing** real time POD operations
- Much of this work remains contingent on funding



Final Comment

Strategic, tactical, and operational decision making
with:

*awareness of the uncertain nature of the
operational environment, and*

*the interaction of that environment with
response mechanisms employed*

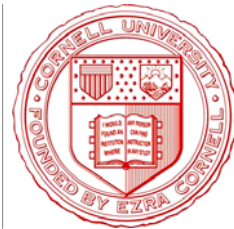
will lead to improved outcomes.



New web site for BERM and associated models

- www.simfluenza.org

Research Team & Collaborators



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 - Shirish Chinchalkar, PhD
 - Peter Jackson, PhD
 - Shane Henderson, PhD
 - Huseyin Topaloglu, PhD
- College of William and Mary
 - David Murray, PhD
- DHHS, Agency for Healthcare Research and Quality
 - Sally Phillips, RN, PhD
- DHHS, NIGMS
 - Irene Eckstrand, PhD
- DHHS, NIH/NIAID
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- New York-Presbyterian Hospital
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 - Nick Cagliuso, PhD
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- Cayuga Medical Center
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