

**Emergency Department Wait Time Reduction Commission** 

Meeting #2



#### Agenda

- Complexity Score Presentation– Jim Scheulen and Heather Blondy
- Hospital & Regional Factors Associated with ED LOS– Geoff Dougherty
- Prioritization Discussion–All
- Legislative Updates–Jon Kromm
- Subcommittee Updates–Tina Simmons
- Next Steps–Tina Simmons







# The impact of the patient population on ED operations: Patient Complexity and Throughput

James Scheulen & Heather Blonsky AAAEM Benchmark Committee Vizient, Inc



Benchmarking in Emergency Medicine Complexity of Building a Cohort for ED Operations

> "Our patients are sicker..." "Our patients need more..."



# Introductions





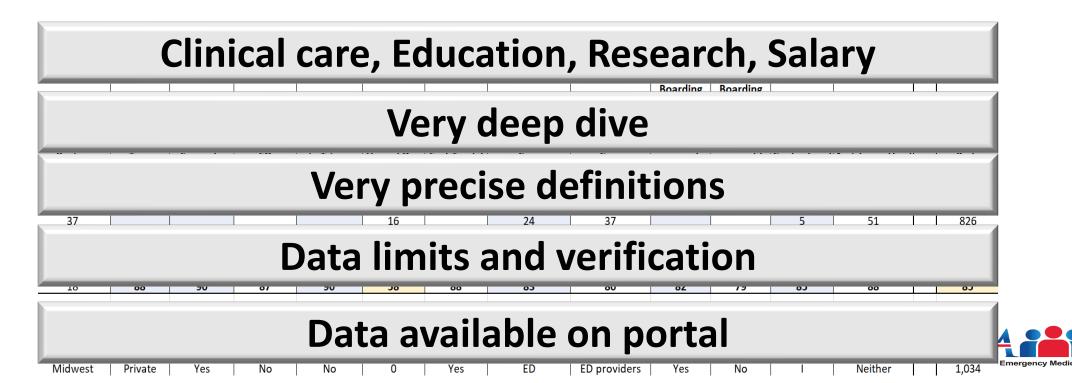


#### **Benchmark Committee: 20 EM Administrators and Physician Leaders**





### Comparing the activity or operations of one emergency department with others for the purpose of quality or process improvement



DEMOGRAPHICS HOSPITAL ED TREATMENT SPACES ED DISPOSITIONS % FMS Hosps **ED PATIENT POPULATION** ESI / TRIAGE LEVELS **ED-BASED OBSERVATION** ED THROU **BOARDING BREAKDOWNS** UTILIZATION AND TURNAROUND \_\_\_\_ NURSING AND NURSING SUPPORT % Actual vs Budgeted Budgeted Patients Actual Patients Patients Patients per RN -Actual RN % Actual vs Nursing Nursing Nursing per RN per RN per RN -Behavioral Hours Budgeted Budgeted Support Support Support ED Case Weekday Weekend ED Social Weekday Main ED Fast Track Obs. Unit Health Worked **RN Hours RN Hours** Hours Hours Hours Managers? **CM Hours** CM Hours Workers? SW Hours 2 1 2 2 52 52 4.2% 21 13 18.3% 8 0 8 4 4 4 4 142,004 153,586 91.6% 51,239 59,010 85.7% Yes 12 10 Yes 15 4 5 4 4 180,484 191,464 100.0% 75,505 78,869 99.7% 85% 16 16 91% 24 4 6 5 5 255,101 239,201 109.5% 159,697 160,303 117.9% 24 22 24 7 10 6 6 6,909,229 6,292,922 149.0% 2,797,979 2,362,722 820.8% No 36 36 No 32 4 5 4 4 439,210 516,134 98.6% 232,786 187,264 121.1% 15% 17 15 9% 19 4 5 4 4 180,484 191.464 100.0% 75,505 78.869 99.7% 16 16 24 Emergency Medicine 61 49 26 35 35 35 35 34 34 34 82 51 48 82 60

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Peer Group Close							Help for Peer	Groups
se Saved Peer Group								
Scheulen Peer ED Compare Group(G	Gai Col로)	Primary academic ONLY - edite Staffed beds: 500-1400 Licensed Beds: 500-1500 Trauma Level 1 only (added HU Peds volume no more than 10% Annual Pt volume 55-80K Total treatment spaces 40-110 Acute treatment spaces 25-110	P back)		R	35 EDs		
- OR hoose Year for Basis of Peer Group	5	Type of Survey						
alculations	•	Primary Academic	•	Store Peer Gro	up	Delete Peer Group		
Hospital Environment Filters	ED Environ	ment Filters Clinical Coverage Filt	ers Pa	atient Population Fil	ers ED Name Fil	ters Custom Filters		
Hospital Environment Filters Checkbox includes unanswered res (40004) Staffed Hospital Bee	esponses	ment Filters Clinical Coverage Filt	ers Pa	atient Population Fil	ters ED Name Fil	ters Custom Filters		

## The Academic ED

Fiscal Year 2023	Median		
Hospital Beds	604		
Licensed ED Beds	57		
Total Bed Hours	536,560		
% Bed Hours to MAIN	69%		
ED Treat & D/C	38,248		
ED Admissions	14,803		
Hospital Observation	2,854		
Total Visits	63,591		
Hospitalized Rate (Calc)	27.8%		
Unique visits	66.4%		

**63,591 Visits** Range: 28,011 – 144,710

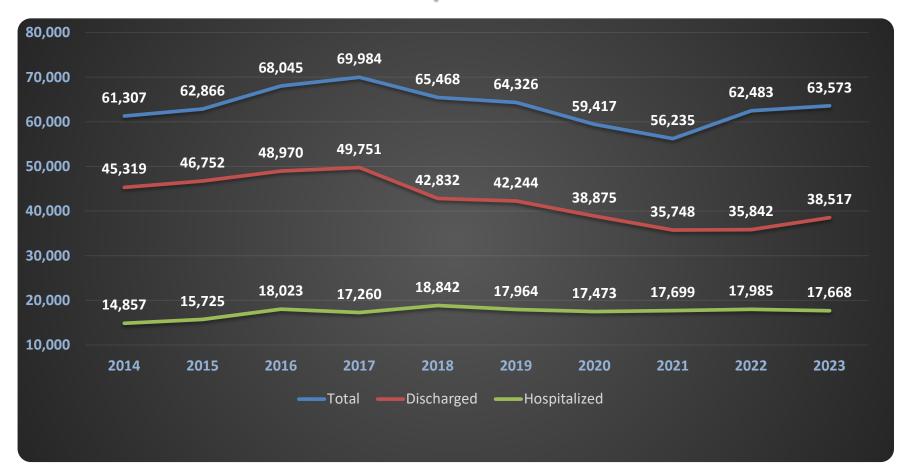


<sup>40,878</sup> Unique visits



# Patient Volume Trend: Median

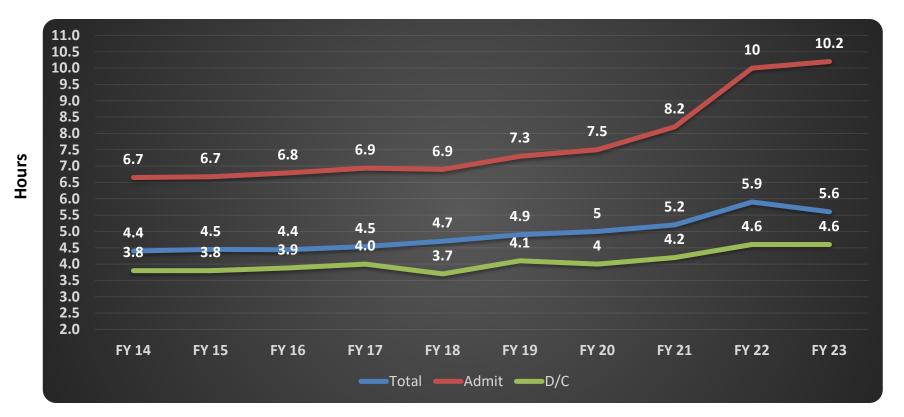
#### Patient Volume Trend—All Responders FY 23





## LOS Trend: Median LOS

#### **Median Emergency Department Length of Stay**

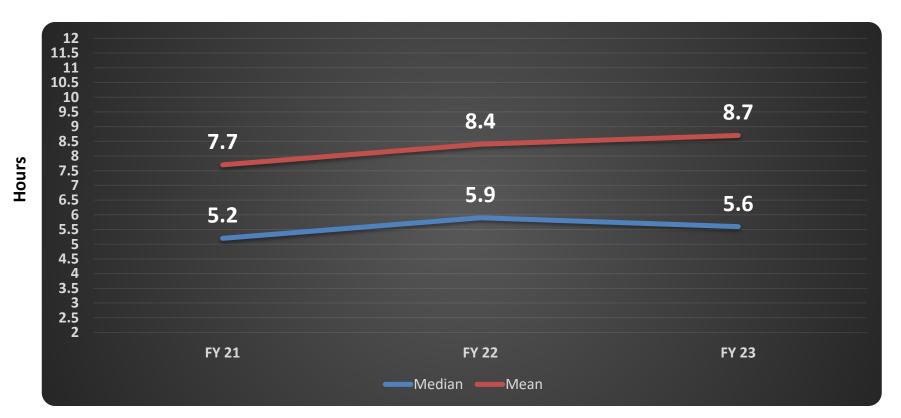


Median times represented here Mean times are longer Distribution with a long right tail



### LOS Trend: Median vs Mean LOS

#### Total ED LOS: Mean vs Median



Mean times represent what staff and patients experience Data distribution has a long right tail



# Sub-cycle Time

### DISCHARGED PATIENTS

- Arrival to Provider: 1.0 hr
- Provider to Decision: 3.9 hr
- *Decision to Depart:* 1.0 hr

### **ADMITTED PATIENTS**

- Arrival to Provider: 1.0 hr
- Provider to Decision: 5.5 hr
- Decision to Depart: 8.0 hr



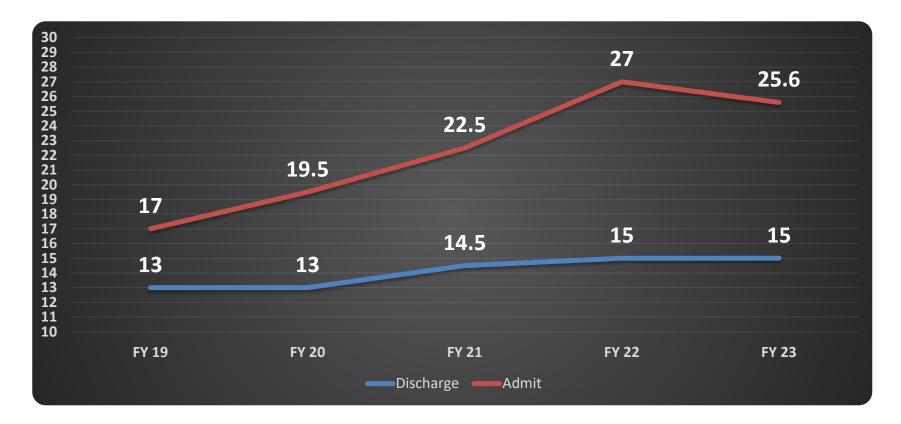
# Ancillary Resource Utilization

FY 19	FY 20	FY 21	FY 22	FY 23			
Itilization 25.5% Mean = 2.3 hours vs 2.8 hours 31.5%							
2.7%	2.8%	2.8%	2.9%	3.2%			
Mean CT Process time has increased by 39%							
Mean MR process time has increased by 169%							
CT and MR Utilization account for 65,000 hours of process time							
If patients are in beds, we now dedicate AN ADDITIONAL 2 beds entirely to CT/MR wait: 7 beds entirely dedicated to process wait time							
4.0 hr	Mean = 5	<mark>.0 hours vs</mark>	11.3 hrs	5.5 hr			
	25.5% 2.7% CT Process tin R process tin ation accoun we now ded beds entirely	25.5% Mean = 2. 2.7% 2.8% T Process time has increased R process time has increased ation account for 65,000 we now dedicate AN Au beds entirely dedicated	25.5% Mean = 2.3 hours vs 2 2.7% 2.8% 2.8% T Process time has increased by 39 R process time has increased by 16 ation account for 65,000 hours of p we now dedicate AN ADDITIONAL beds entirely dedicated to process	25.5%Mean = 2.3 hours vs 2.8 hours2.7%2.8%2.8%2.7%2.8%2.9%CT Process time has increased by 39%IR process time has increased by 169%ation account for 65,000 hours of process timewe now dedicate AN ADDITIONAL 2 beas entitledbeds entirely dedicated to process wait time			



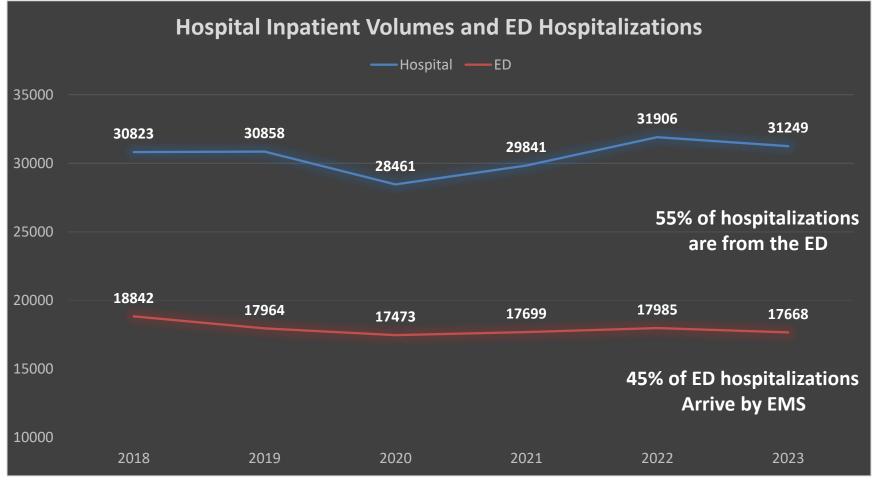
# LOS Behavioral Health

#### **Behavioral Health Patients = 5.7% of Arrivals or 3500 patients**



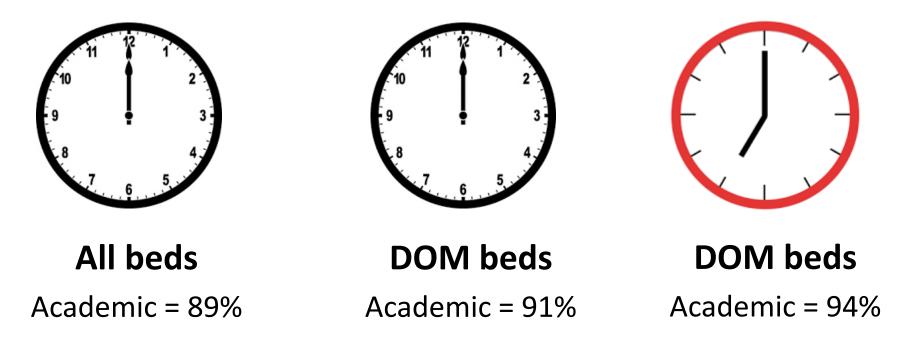


# **Hospitalization Data**





# Inpatient Occupancy

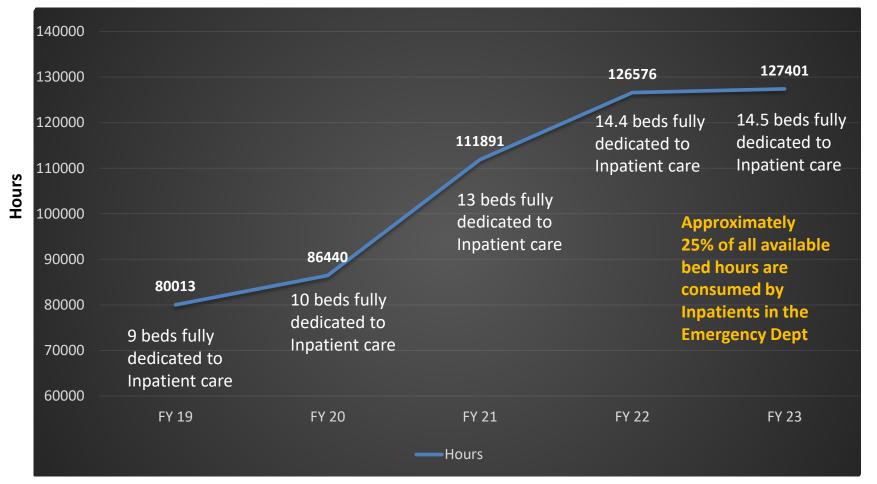


Approximately 80% of all patients in DOM come from the ED Approximately 67% of all ED hospitalizations go to DOM



## **Boarding Time**

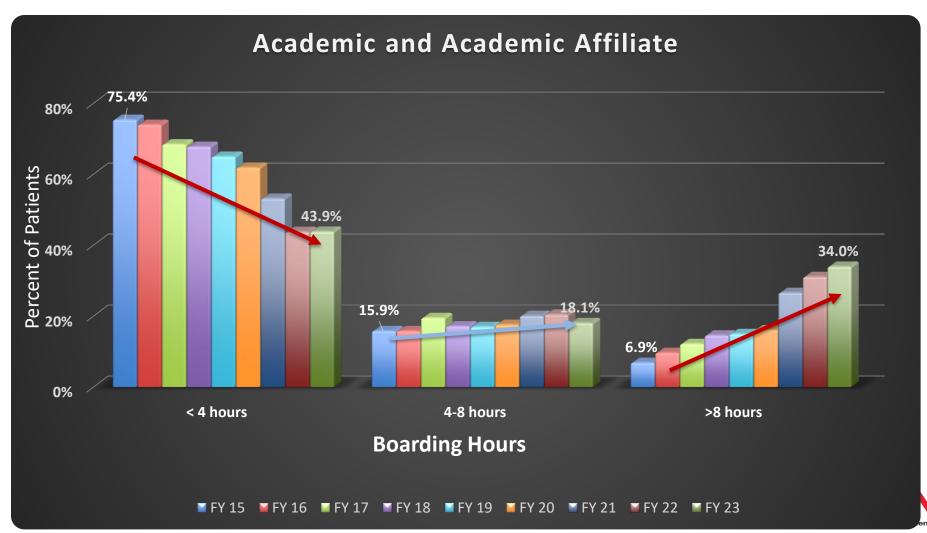
#### **Total Boarding Hours**



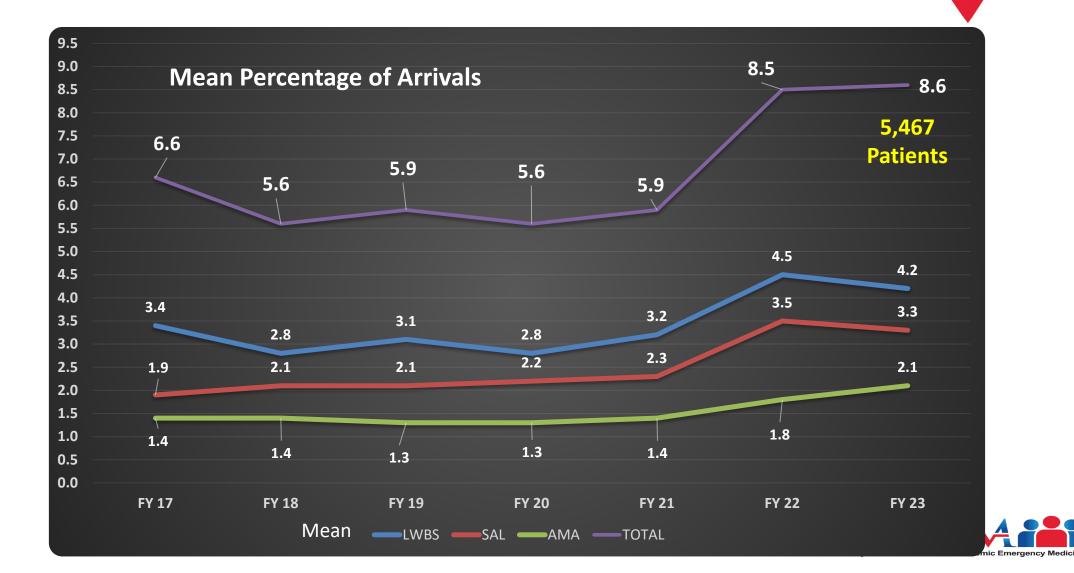


# **Boarding Distribution**

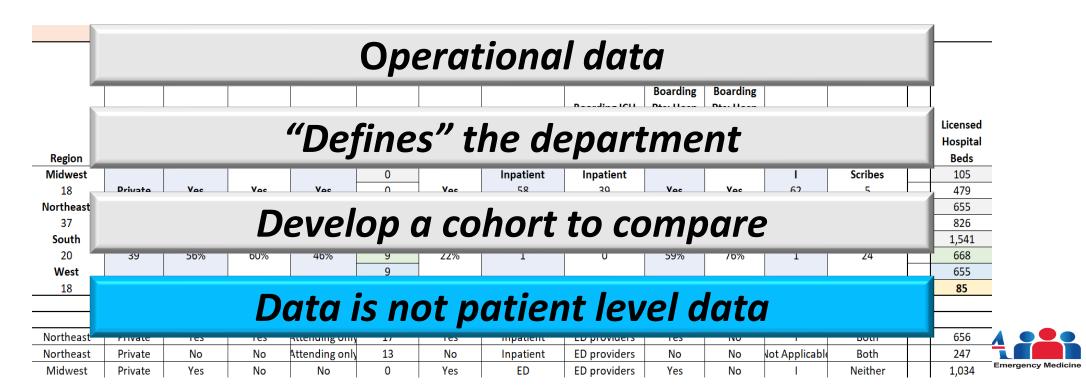
#### FY 2015 vs FY 2023



# Left Before Treatment Complete



### Comparing the activity or operations of one emergency department with others for the purpose of quality or process improvement



# Defining your department

# Developing the right cohort Understanding resource needs

### **Operational Variables**

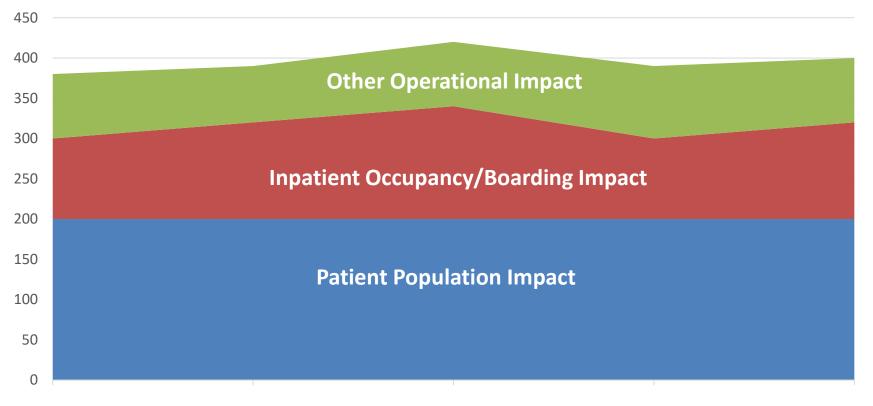
#### **Patient Population**

- Visit volume
- Teaching vs Community
- Hospitalization rate
- EMS arrivals

- Patient history
- Presenting complaint
- Co-morbidities
- Social needs



# Throughput Impact Layers



**Impact of Patient Population vs Operations/Boarding** 

■ Population ■ Boarding ■ Other





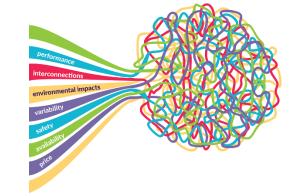
How does the composition or complexity of the patient population impact operations



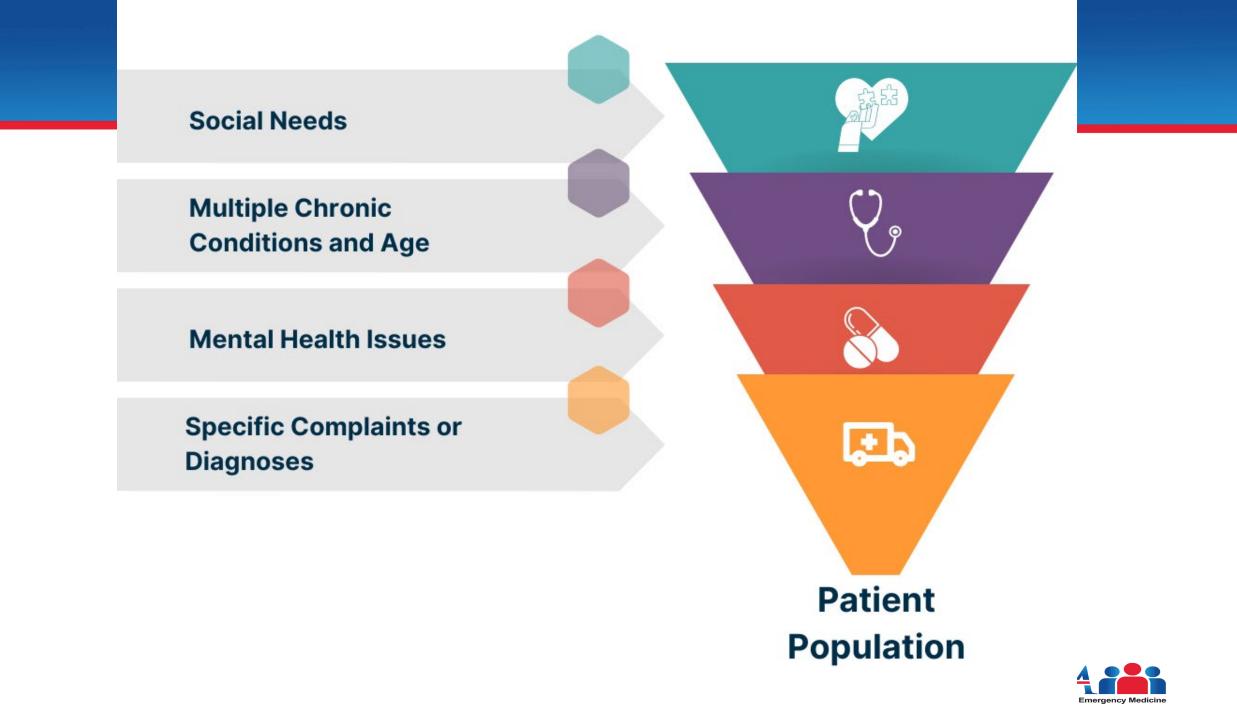
# Patient Population Definition

Understanding the *composition and complexity of the patient population* in each emergency department as a way to better understand the *resources required* to care for that patient population.

- *Time* as a proxy for resource demands
- Patient level data
- On any given day, what do we face?
  - Interactions between patient variables









- A way to describe a patient population
- A way to consider multiple patient based variables
- A way to compare among ourselves
- A way to compare ourselves to ourselves over time

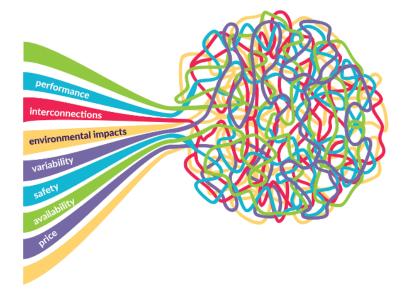


# Acuity versus Complexity

#### <u>Acuity</u>

Severity of illness Priority setting Implies **SPEED** is required





### **Complexity**

Multiple care needs Personal, social and clinical needs together Implies **TIME** is required



# **Previous Efforts**

# OPERATIONAL METRICS CASE MIX INDEX FOR ED ADMISSIONS COMPLEXITY INDEX DEVELOPMENT



## Inpatient Case-Mix Index

- Inpatient Case Mix Index: Hospitalized from ED
  - Resource based index
  - Indicates acuity/complexity but impacted by high cost treatments

	CMI w/o HAC
Primary Academic	1.94
ED Admissions	1.80
Non-ED	2.07
Community	1.45
ED Admissions	1.45
Non-ED	1.49

**AMC > Community Hospital** 

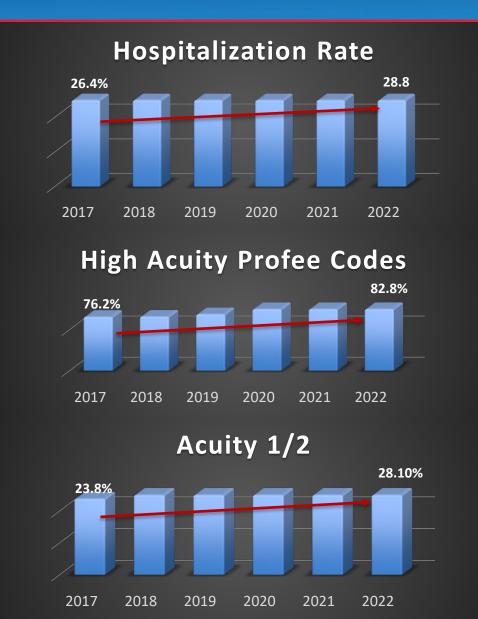
AMC ED > Community ED

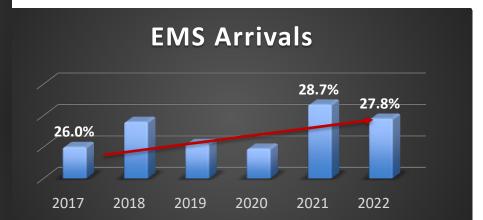
**Non-ED > ED Hospitalizations** 

**Community ED = Non-ED** 

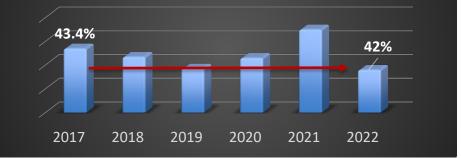


### Complexity Metrics: As a group





#### **EMS Admissions**



AAAEAA

### **Operations Based Complexity Index**

Complexity Index: Data Preprocessing and Methodological Comparison

AAAEM/AACEM Benchmark Committee and Roundtable Analytics, Inc.

February 28, 2020

#### Produce an Index Score and Rank for each Academic Center

- Number of Arrivals
- Ratio of % ESI-1/2 to % ESI-4/5
- % ED Arrivals Hospitalized
- % Arrivals by EMS
- % EMS Arrivals Hospitalized
- % Profee 4/5/CC

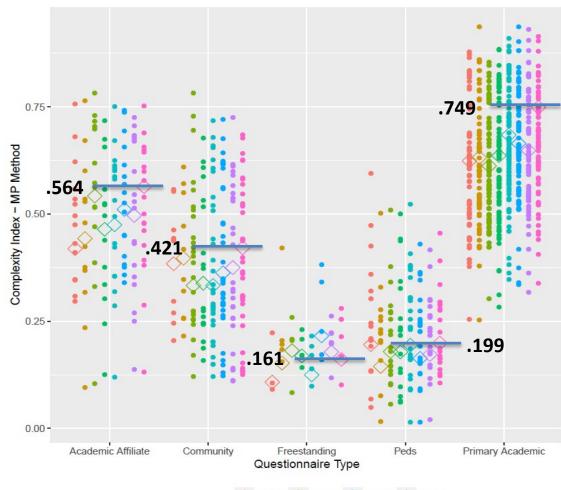
4 Versions of Complexity Index Principal Component Analysis Blended Versions



### **Operations Based Complexity Index**

### Fiscal Year 2021

- Hospitalization rate
- EMS Arrivals
- EMS Admissions
- High Acuity Codes
- Acuity 1&2 vs 3&4



2014 📀 2016 📀 2018

2015 🕢 2017 🔷 2019 🔶 2021

Fiscal Year

2020



### **Operations Based Complexity Index**

#### **Complexity Index Ranking**

University	🝷 ED Type 💽	Complexity Index 💌	Overall Ran 🕶 I	ED Type R 🔽
University of Massachusetts / Baystate	Primary Academic	0.942708333	1	1
University of Florida, Gainesville	Primary Academic	0.902083333	2	2
The Ohio State University	Primary Academic	0.890625	3	3
Harvard University / Beth Israel Deaconess	Primary Academic	0.88125	4	4
Medical College of Wisconsin	Primary Academic	0.877083333	5	5
University of Kansas School of Medicine	Primary Academic	0.864583333	6	6
Vanderbilt University	Primary Academic	0.855208333	7	7
Virginia Tech University	Primary Academic	0.845833333	8	8
Harvard Medical School	Primary Academic	0.840625	9	9
University of Texas Health Sciences - Houston	Primary Academic	0.817708333	10	10
Washington University @ St. Louis	Primary Academic	0.796875	11	11
University of Texas, Southwestern	Academic Affiliate	0.788541667	12	1
Yale University	Primary Academic	0.7875	13	12
Loma Linda University	Primary Academic	0.780208333	14	13
Penn State University	Primary Academic	0.778125	15	14
University of Michigan	Primary Academic	0.777083333	16	15
University of Rochester	Primary Academic	0.777083333	17	16
Texas A&M University	Primary Academic	0.776041667	18	17
Duke University	Primary Academic	0.772916667	19	18



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# Patient Based Complexity

- Collaboration with Vizient
  - Membership PI Organization
  - Most AMCs (95% of our members)
  - Clinical Data Base from members
- Patient Based Complexity Measure
  - Patient level data
  - Encounter specific metrics: Hospital Coding
    - Demographics (Age)
    - Presentations
    - Diagnoses
    - Co-morbidities
    - Social needs
  - Impact on Throughput





Heather Blonsky





# Patient Based Complexity

Quantify differences in the complexity of cases or definition of the patient population seen in different EDs or one ED over time

- Provide context to understanding variables impacting throughput
  - Patient Clinical Data
  - Social Needs
  - Variability (Operations)



#### Hypothesis:

An emergency department that sees patients with more clinical needs and patients with more social needs will have longer throughput times.



## Patient Based Complexity Model

#### **Creating the Model**

Initial Data Set: 4 Hospitals Vizient Clinical Data Base 280 patient level variables

Small sample size for model

Provided throughput data points 2 years of daily patient level data Principal Component Analysis Streamlined variables

#### On this day in the ED

- Age and co-morbidities
- Current diagnoses
  - Psychosis
  - Alcohol and/or drugs/depression
  - Trauma
- Complex history
- PCP desert and 7 day returns
- Patients from neighborhoods with high social needs (transportation domain)



# Patient Based Complexity Model

- Expanded the number and type of hospitals
  - 10 Health Systems
    - JHHS, UC Health, Northwestern, Mass General Brigham, UMass, Michigan, Cincinnati, OSU, Jefferson, U Virginia
  - 27 Hospitals
    - 11 Academic Medical Centers
    - 7 Large Community Hospitals (Affiliates)
    - 7 Small Community Hospitals
    - 2 Critical Access Hospitals



– 52 Hospitals



# Summary Patient Based Model

## **Population most impacting operations:**

- More patients
- Higher proportion of patients with chronic effects of ETOH
- Higher proportion of patients with mental health issues (Psychosis)
- Higher proportion of elderly and/or complex patients
  - More than 4 Elixhauser comorbidities
- Higher proportion of patients with oncology Dx
- Higher proportion of patients from neighborhoods with high social needs
  - Transportation challenges
  - Access to health care/PCP desert



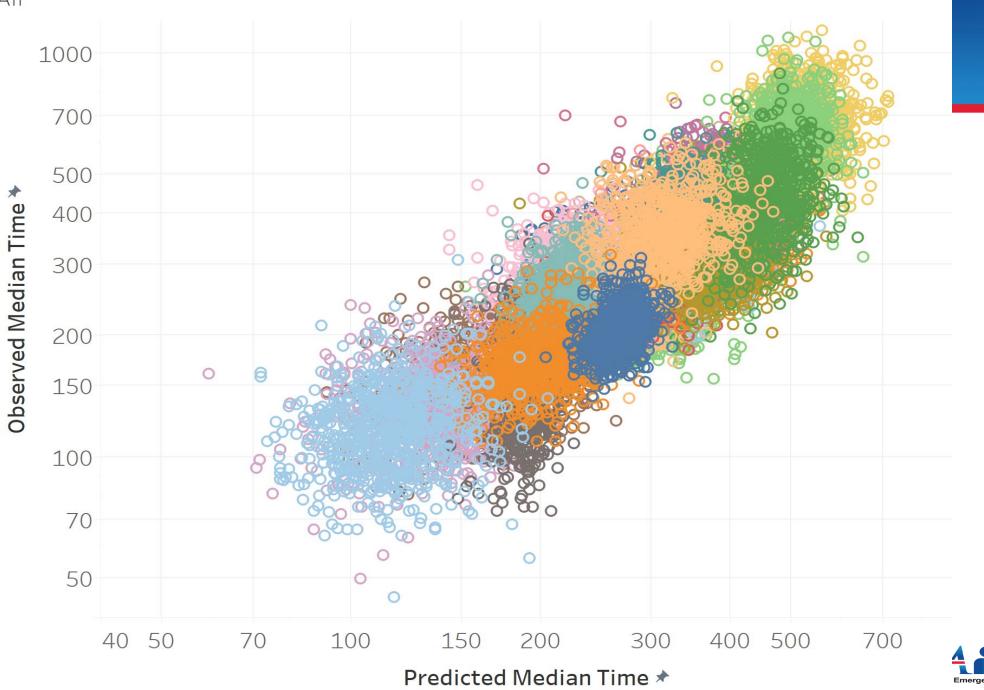
## Patient Based Complexity Model

#### Variables provide a good fit (r<sup>2</sup> = 0.71)

- More patients
- More patients with chronic ETOH
- More patients with psychosis
- More elderly and/or with comorbidities
- More patients with oncology Dx
- Patients from neighborhoods with high social needs/PCP desert
- Reduced time = more patients with:
  - Current drug or alcohol overdose
  - History of 7-day returns to the ED
  - Severe trauma

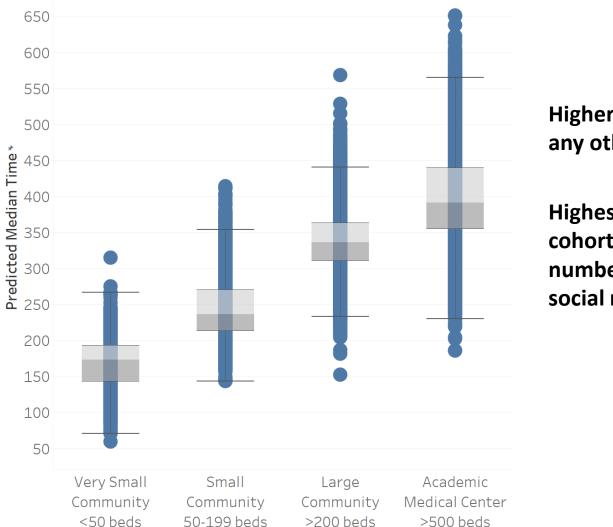
		Parameter	Estimates			
Variable	DF	Parameter Estimate		t Value	Pr >  t	Variance Inflation
Intercept	1	1.72420	0.00844	204.39	<.0001	0
logcountEncounters	1	0.26441	0.00332	79.66	<.0001	1.80271
elderlycomplex_pct	1	0.54292	0.03768	14.41	<.0001	3.61055
elderlyorcomplex_pct	1	0.13184	0.01351	9.76	<.0001	1.98338
depression_pct	1	0.30474	0.02607	11.69	<.0001	2.98850
psychosis_pct	1	0.73521	0.05640	13.04	<.0001	2.17136
drugs_pct	1	-0.65709	0.01535	-42.81	<.0001	2.01709
alcohol_chronic_pct	1	1.38677	0.09917	13.98	<.0001	1.08316
hf_pct	1	0.42824	0.04181	10.24	<.0001	2.38631
ami_pct	1	1.39015	0.12972	10.72	<.0001	1.05410
oncology_pct	1	0.53311	0.02819	18.91	<.0001	2.10574
stroke_pct	1	0.49120	0.09667	5.08	<.0001	1.38436
covid_pct	1	0.17618	0.01594	11.06	<.0001	1.02545
trauma_pct	1	0.56129	0.02515	22.31	<.0001	2.20454
trauma_severe_pct	1	-0.77730	0.02538	-30.62	<.0001	2.58417
vvi_pct	1	0.23934	0.00489	48.99	<.0001	2.41054
transportation_pct	1	0.14704	0.00366	40.22	<.0001	1.62815
access_pct	1	1.39098	0.08132	17.10	<.0001	1.11890
pcp_pct	1	0.06822	0.00504	13.54	<.0001	1.58394
ed7day_pct	1	-0.15454	0.01771	-8.72	<.0001	1.61210





All

## Patient Based Complexity Model

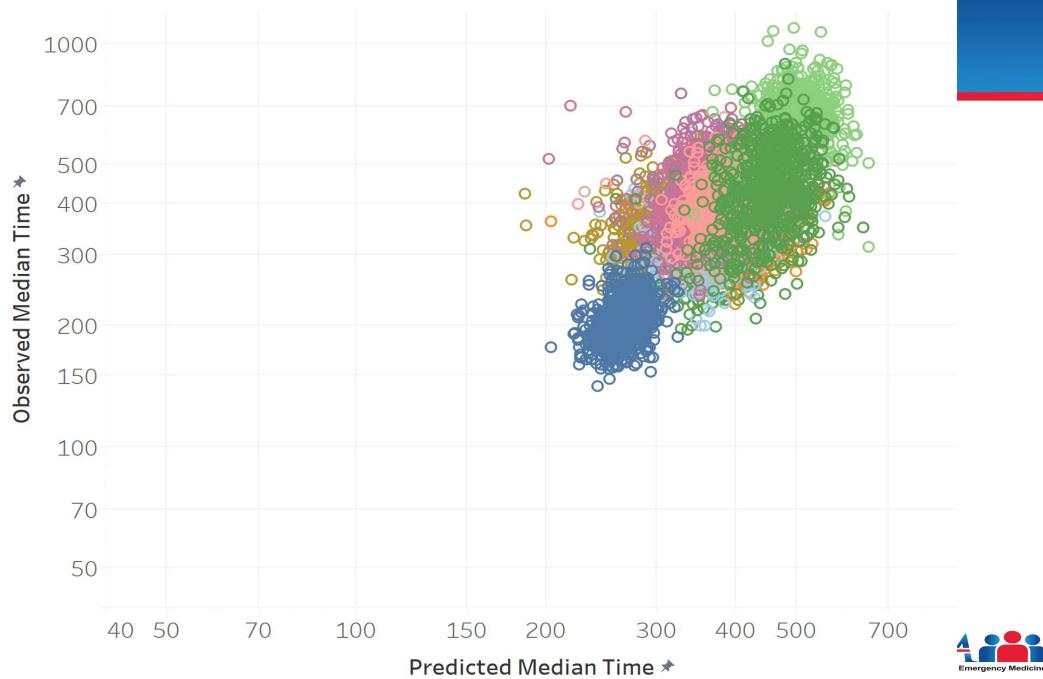


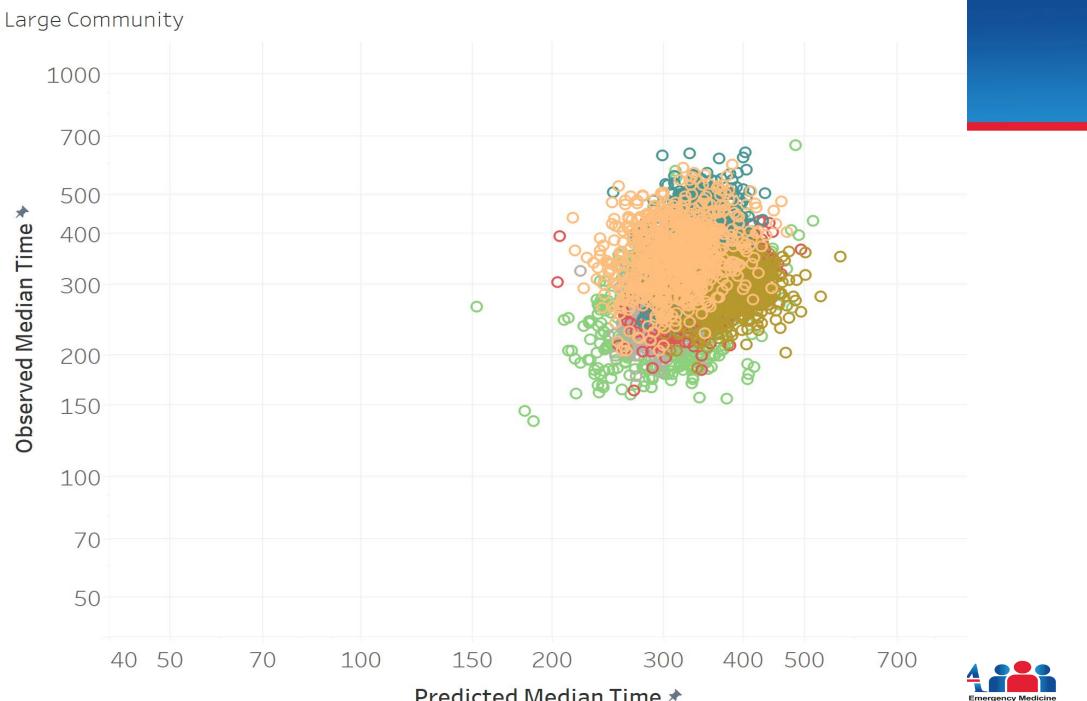
Higher complexity for AMCs than for any other cohort

Highest complexity within any one cohort tends to include higher numbers of patients with increased social needs

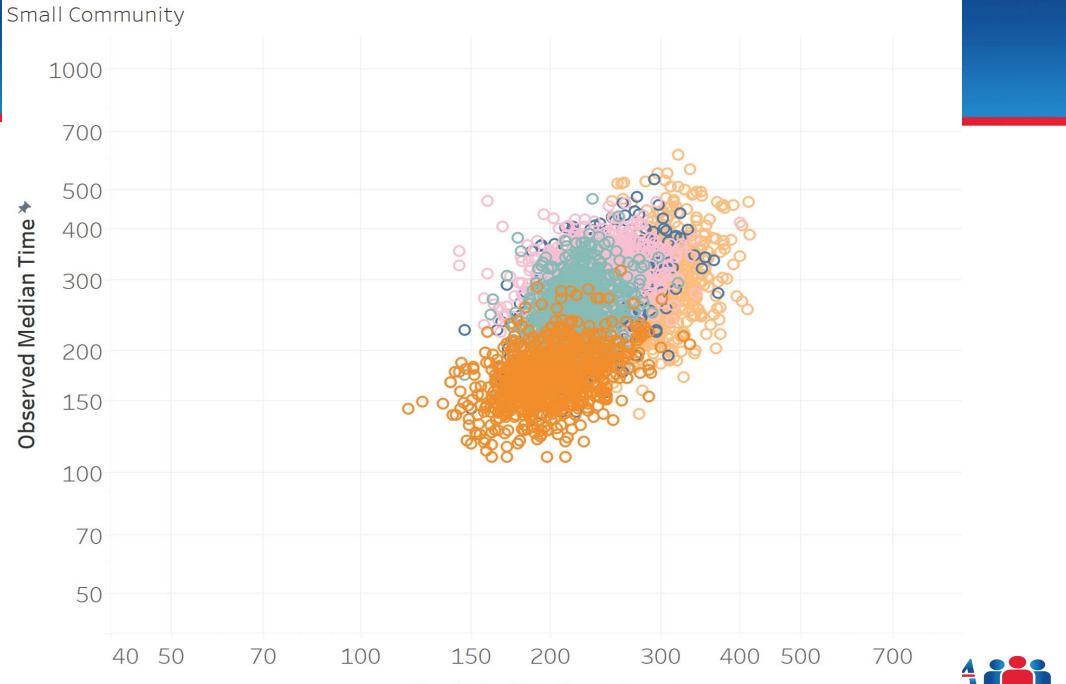


Academic Medical Center



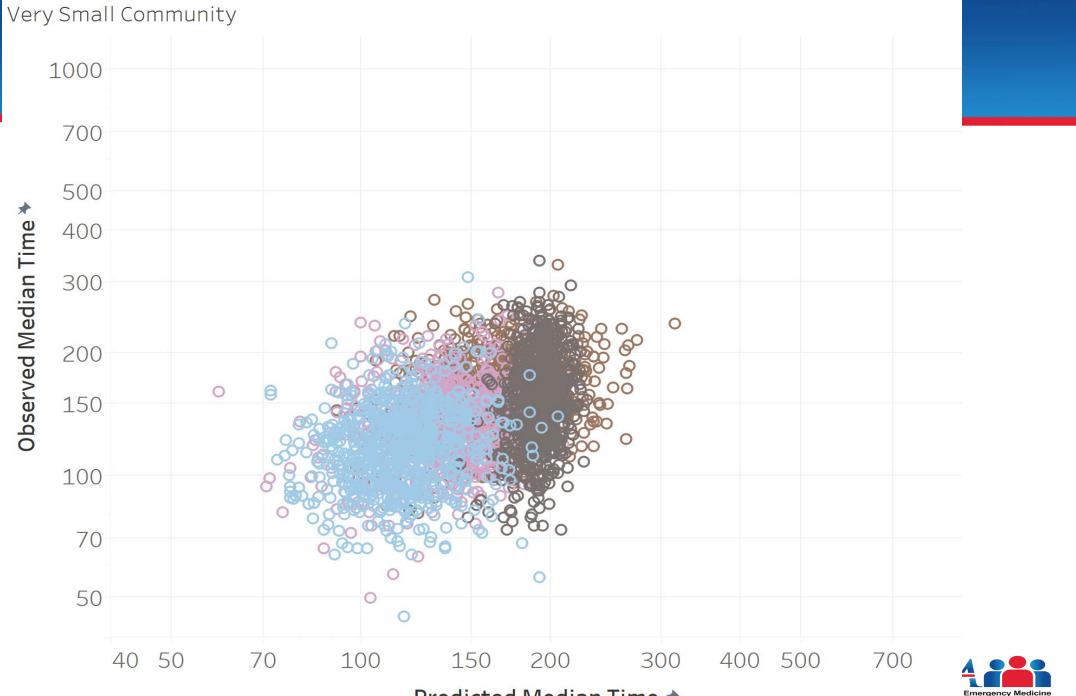


Predicted Median Time \*

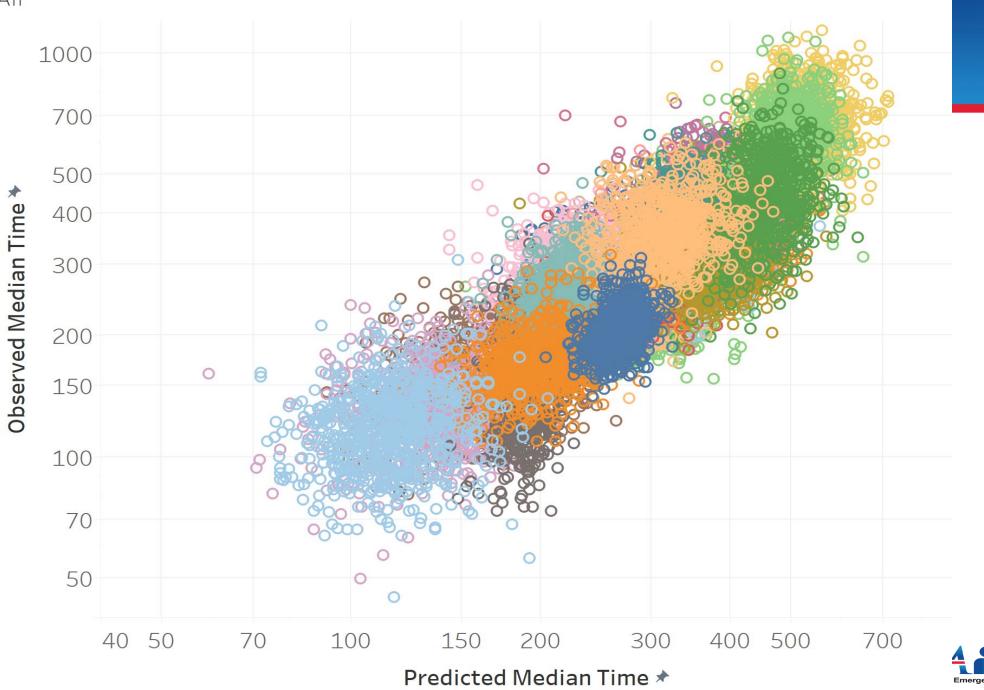


Predicted Median Time \*

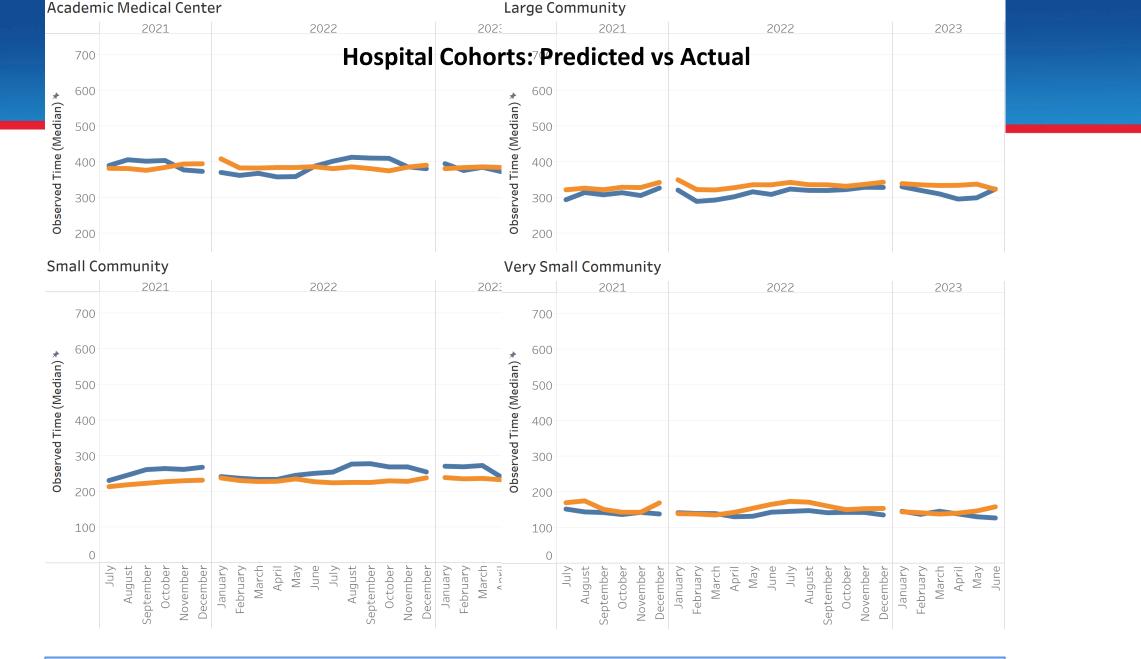
Emergency Medicine



Predicted Median Time 🖈

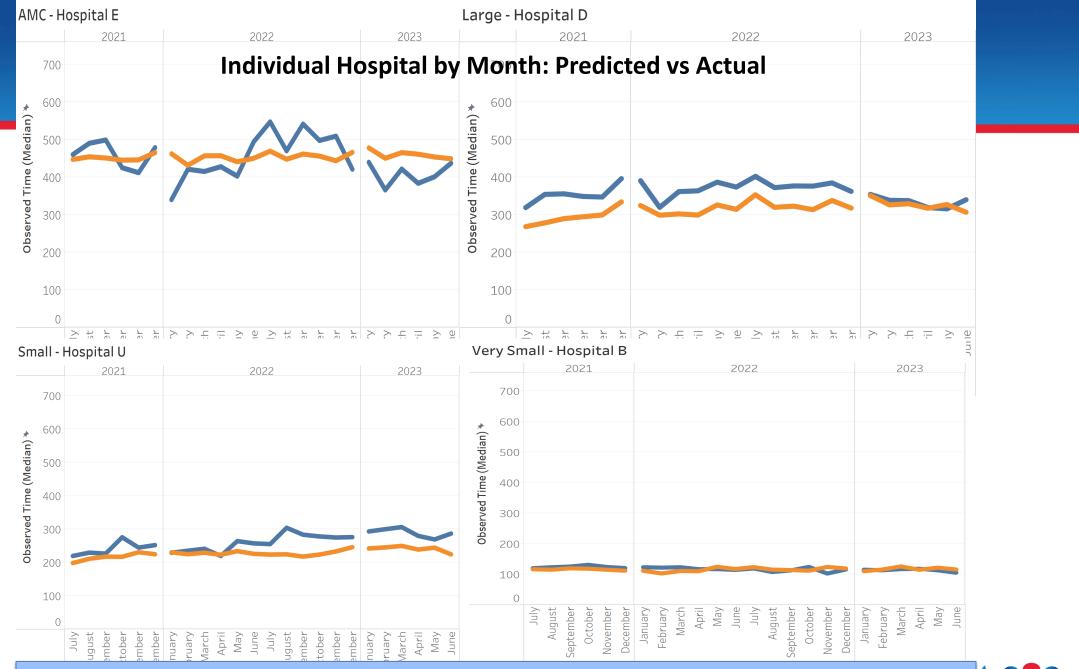


All



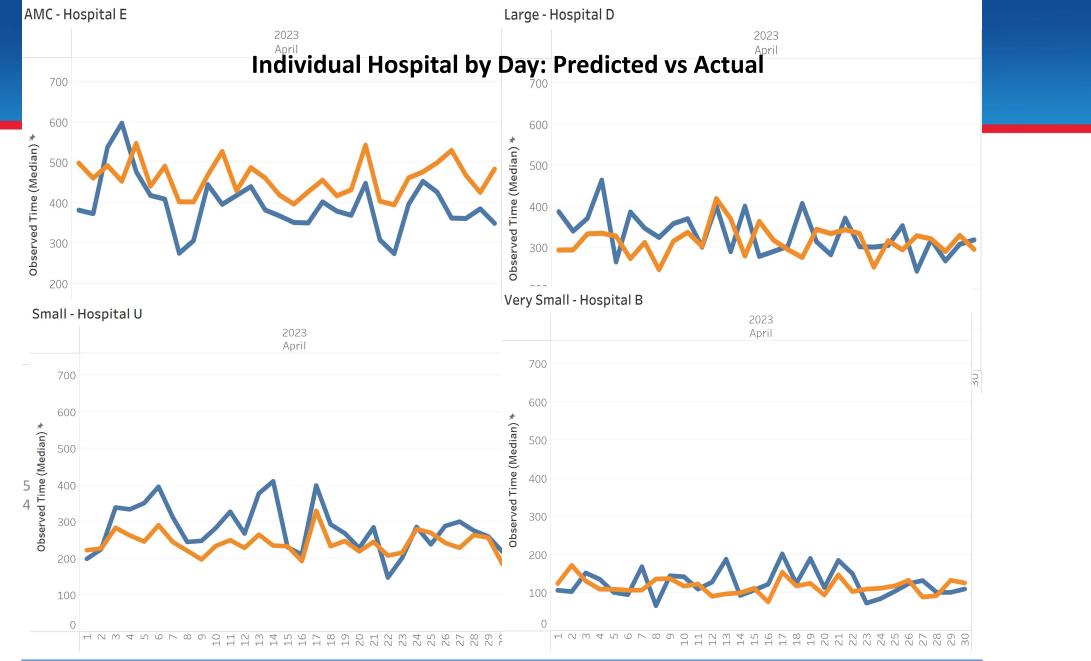






Low variability at the hospital level (Monthly)





Emergency Medicine

Expected variability at the hospital level BY DAY



Emergency Departments that care for patients with *more clinical and social needs* can expect *longer throughput times* than those who care for a population with fewer clinical and social needs.





# Patient Based Complexity Model

#### Use Case:

- Build a cohort
- Demonstrate comparison
- Observed over Expected

#### **Next Steps:**

- Finalize model
- Add impact of boarding
- Data access/Rollout



# Implications





# **THANK YOU**



# Hospital & Regional Factors Affecting ED LOS





### Motivating Questions for Today's Presentation

- How does Maryland performance on ED Length of Stay (LOS) compare to nation prior to and during the TCOC Model?
- What is the relative contribution of regional and hospital-specific determinants of ED Length of Stay on a national level?
- What kinds of improvements in ED Length of Stay can we expect from specific interventions on these determinants?
- What policies/programs are suggested by these analytic results?



## **Statistical Modeling Approach**

- We modeled
  - Hospital Referral Region (N=306)
  - Individual Hospital (N=3019)
- The model assesses the degree to which each determinant is associated with added ED Length of Stay
  - e.g.,: "A change of one year in median population age is associated with an increase of 10 minutes ED Length of Stay"
- The model also provides guidance on what proportion of variation in ED Length of Stay is driven by HRR and hospitalspecific factors
- Finally, we evaluated factors underlying one particular determinant of ED Length of Stay: inpatient occupancy rate



#### **Data Sources**

#### **Hospital Referral Region**

- <u>US Census</u>: Population size, age, density
- <u>CDC</u>: Social Vulnerability Index
- AHA Survey: IP Beds per capita
- <u>CMS</u>: PCPs and SNFS per capita
- <u>Dartmouth Atlas</u>: Primary care access and surgical volume for Medicare population

<u>2019 AHA Survey:</u> ED visits, IP visits, services provided, teaching status, hospital staffing, IP occupancy

CMS Hospital Compare

• 2019 ED1 and OP18



## **Additional Details**

- Presentation focuses on ED1, but takeaways for OP18 and composite measure are similar
- Data that would have been helpful but weren't obtainable
  - ED-specific hospital staffing and resources (AHA mostly provides hospital-wide numbers)
  - Patient acuity
- Because of data limitations and policy interest, this work focuses on impact of factors related to ED inflow and output, rather than movement of patients within the ED
- Our model does not account for interactions and non-linear relationships. More on this later
- Data are mostly self-reported and cross-sectional

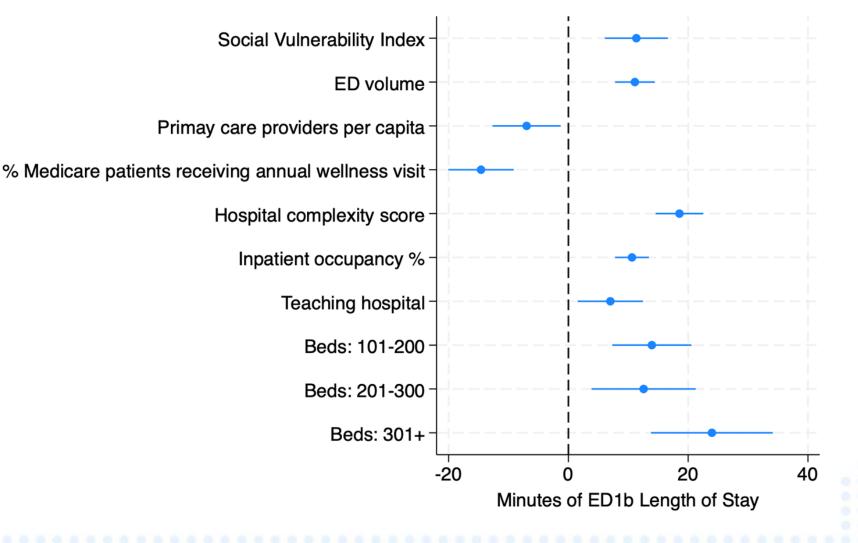


## Summary of Analytic Findings

- Differences between Hospital Referral Regions account for 37% of variation in Median Time from ED Arrival to ED Departure for Admitted ED Patients (ED1b)
- Differences between hospitals within Hospital Referral Regions account for 63% variation in ED1b performance
  - This indicates that hospital factors (e.g. staffing, bed management, organizational structure) are likely driving ED performance
  - HRR/regional factors (IP Beds per capita, SNF beds) are less important
- Primary care access is an important and modifiable determinant of ED length of stay
- Addressing social determinants may also improve ED length of stay performance
- Structural hospital factors (Bed size, complexity, teaching status, ED size) that are not as easily modifiable have a large effect on ED performance)



#### Relative Strength of Association with ED Length of Stay

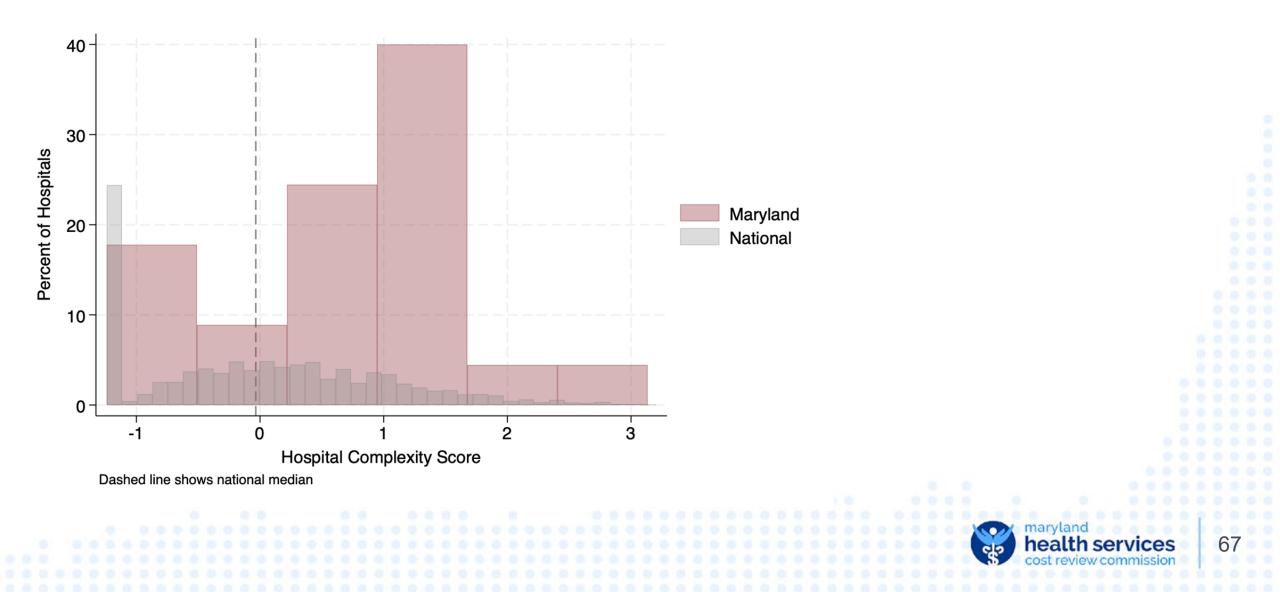


Comparative ED Length of Stay effect size of all statistically significant variables in national model

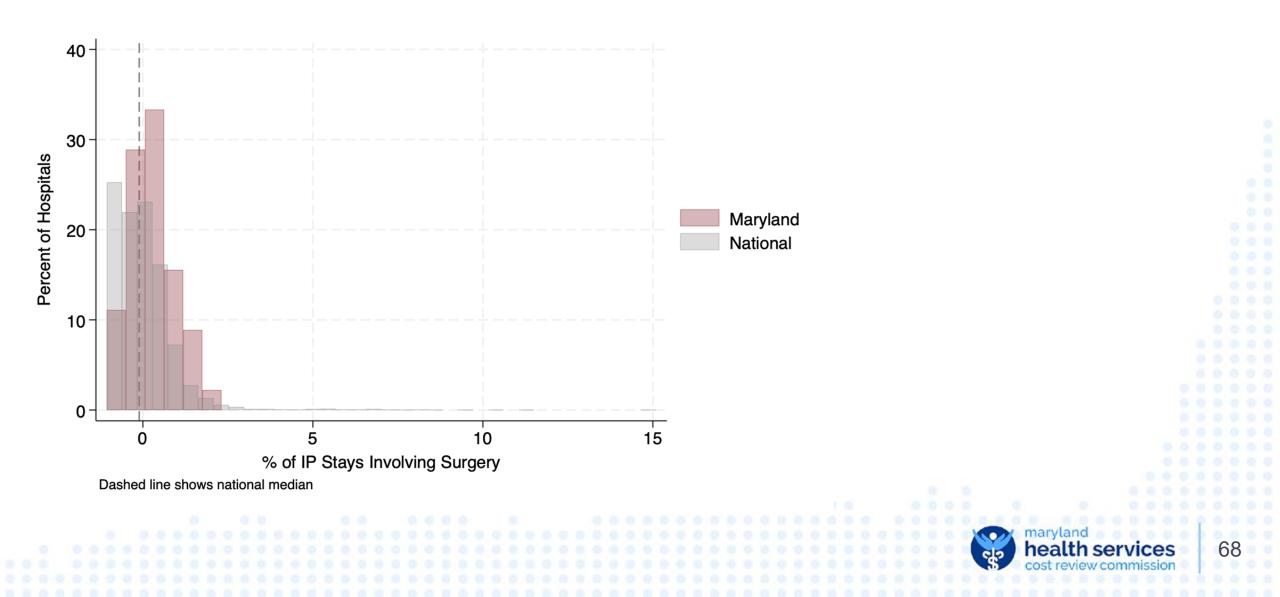
Model accounts for 67% of variation in ED1b performance across hospitals



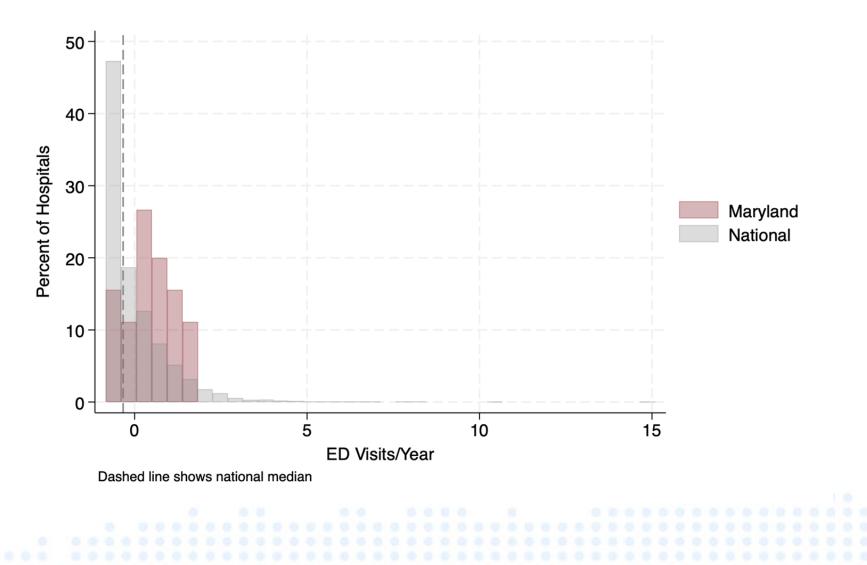
#### MD Hospitals Are More Complex Than Others



#### MD Hospitals Have High Surgical Volume



#### MD Hospitals Have Larger ED's

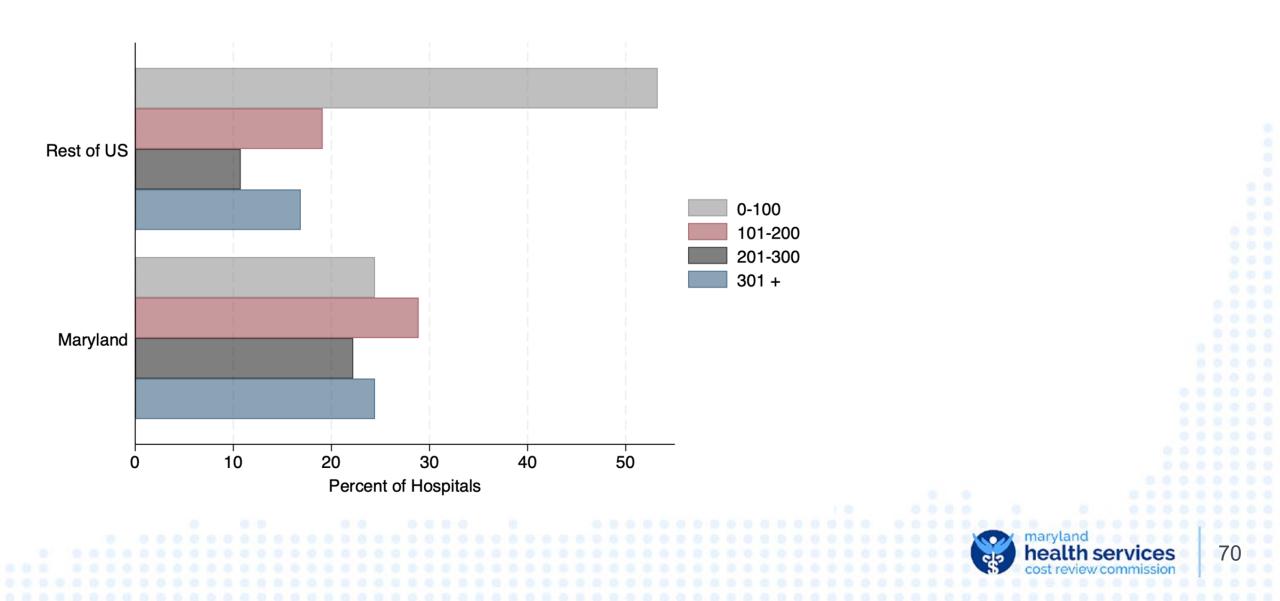


 Note than while ED volume is higher in MD, volume per capita is lower

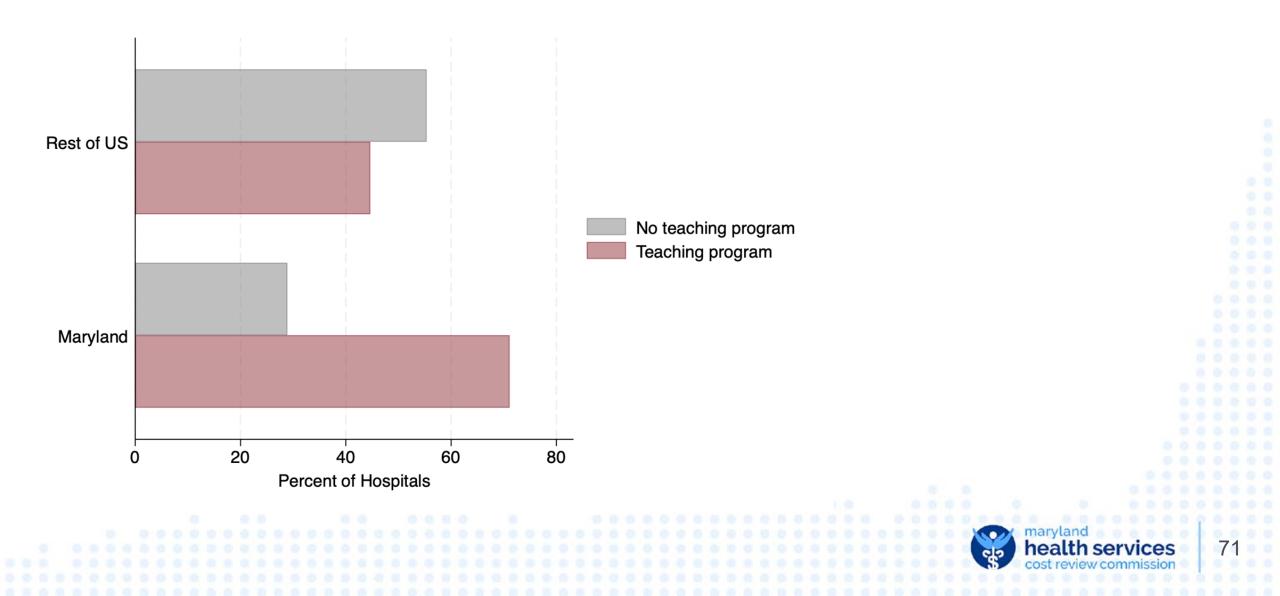
 Maryland has fewer ED's than elsewhere, but they see more volume



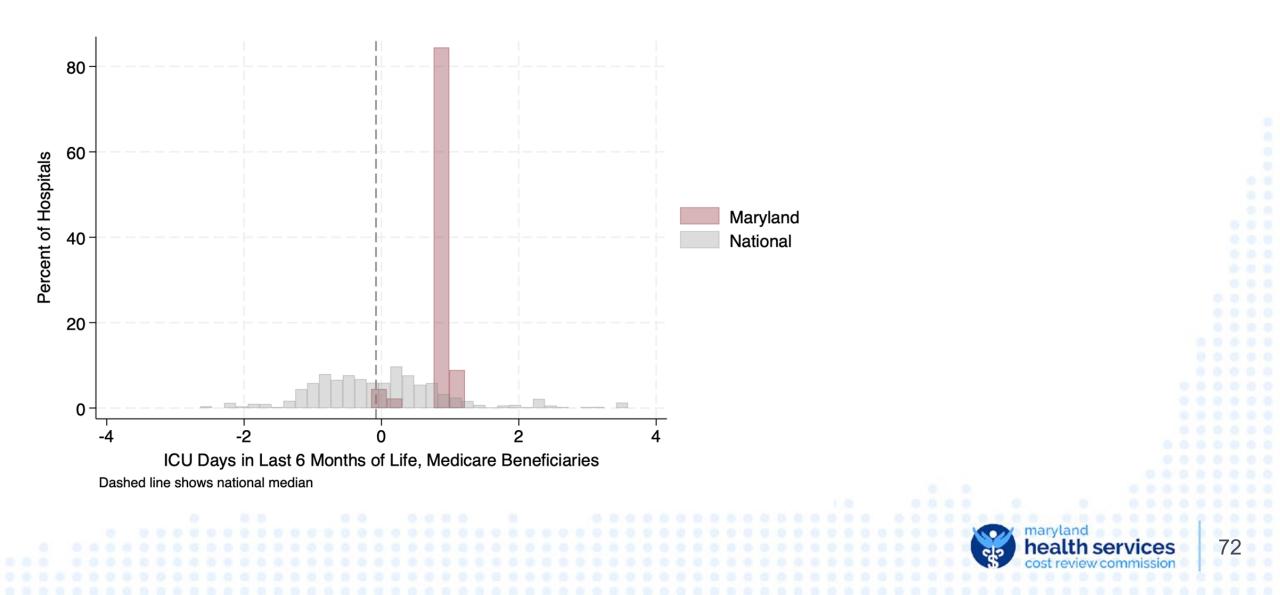
#### MD Hospitals Are Larger Than Those Elsewhere



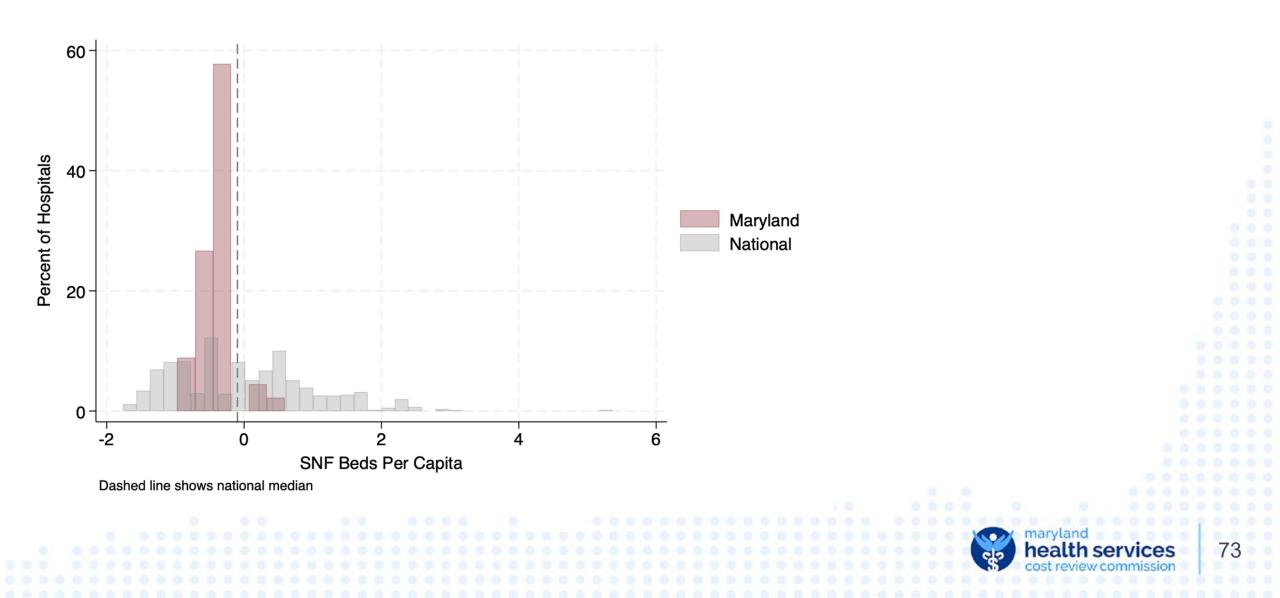
#### MD Hospitals Are More Likely to Have Teaching Programs



#### MD Hospitals Provide More End of Life ICU



#### MD Hospital Regions Have Fewer SNF Beds



## Performance of MD Hospitals vs. Nation

- Maryland hospitals are larger, more complex, and more likely to be teaching facilities. All of these factors are associated with longer ED Length of Stay
- This is a blessing and a curse. Larger, higher-volume and more complex hospitals typically provide better outcomes in terms of risk-adjusted mortality, readmission and inpatient length of stay
- After accounting for structural differences, Maryland hospitals are not doing as poorly as reported
  - However, some big, complicated hospitals nationally still perform well in ED Length of Stay (See Appendix B), so Maryland has significant room for improvement
- Can we provide both excellent IP results and better streamlined ED experience by finding ways to make big hospitals feel more like small ones (or high performing hospitals elsewhere in the nation that are big and complicated)?

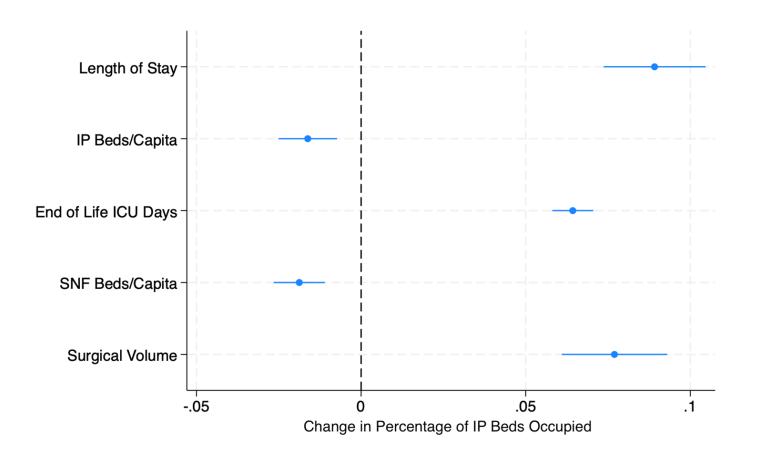


### What About Occupancy?

- Hospital occupancy is an important determinant of ED Length of Stay, and a complex topic in its own right
- We evaluated the independent association of multiple variables with inpatient occupancy
  - Staffed IP beds per capita
  - Length of Stay
  - End of Life Care
  - SNF beds per capita
  - Surgical volume
- Occupancy = AHA IP bed days / (365\* IP beds staffed EOY)



### Relative Strength of Association with IP Occupancy



- Surgical volume, LOS, end of life ICU days, and SNF availability are significant determinants of occupancy
- MD differs from the nation unfavorably on all measures
- Increasing MD staffed IP beds to national average would change occupancy by 0.5%
- MD staffed beds per capita (exclusive of beds in nearby regions, e.g., DC) are lower than national average due to reduced demand under TCOC model



# What Does Analysis Tell Us About Policy/Program Directions?

- Policies addressing primary care may result in improved ED Length of Stay
  - Reimbursement Enhancements: Maryland Primary Care Program (MDPCP)
  - Investments in additional primary care supply
- Policies addressing social determinants may also result in improved ED Length of Stay
- Policies addressing IP occupancy may result in improved ED Length of Stay
  - Improved hospice access
  - Improved SNF access
  - Planning elective surgery and medical admissions to avoid constraining ED admissions
- Increasing inpatient bed capacity is not likely to be a viable and sustainable solution to ED Length of Stay in Maryland
  - Stacking more beds in institutions that have structural impediments to low ED throughput may worsen the problem
  - Expanding IP capacity would likely be a costly, long-range solution that has negative implications for TCOC model performance
- Other interventions discussed above may provide similar or better outcomes with limited cost and downside



# **Testing Interventions**



## **Developing a Testing Platform**

- Because conventional statistical modeling (i.e., regression) does not account for nonlinearities, bidirectional causation, etc., it does not always provide a clear picture of the impact of future interventions
- Simulation exercises are a standard way to address this blind spot
  - If SimCity had an emergency room ...
- Long history of this type of work in operations science and hospital performance literature
- Most straightforward modeling approach divides hospital areas of interest into buckets or "stocks", and moves patients between them with flow rates



### **Simulation Process Overview**

- Identify the setting for the model's base case
- Identify variables that are to be reflected in model
- Obtain real-world values for these variables
- Build the base case model using Models are be populated with real-world data and tested to ensure they reproduce real-world conditions prior to testing hypotheses



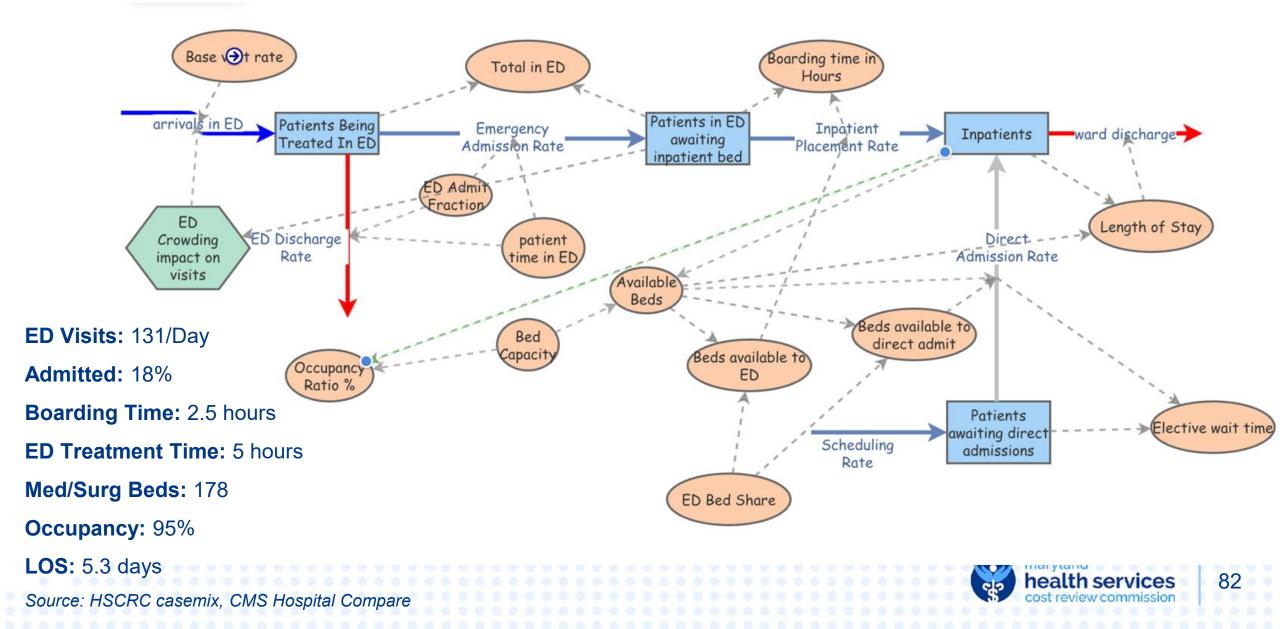
### **Our Model**

- Hospital: Suburban Baltimore community hospital with ED LOS, bed count and ED volume close to the state average in 2019
- Key stocks
  - Patients being treated in ED
  - Patients admitted and awaiting a bed
  - Patients on inpatient service, patients awaiting direct/elective admission
- Key flows
  - ED visit volume
  - Admission rate from ED and direct admit
  - IP discharge rate (linked to LOS)
- System Dynamics
  - Bottlenecks
  - Thresholds
  - See Lane et al. (2000) 'Looking in the wrong place for healthcare improvements: A system dynamics study of an accident and emergency department', The Journal of the Operational Research Society.



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### **Baseline Model**



### **Interventions Tested**

- Reduce ED volume by 5%, reflecting modest cut in volume from multi-visit patients (more on this later)
- Reduce LOS by 5%, reflecting modest increase in SNF/behavioral beds
- Reduce daily elective/direct admit volume by 1 patient/day

### How to measure impact of interventions?

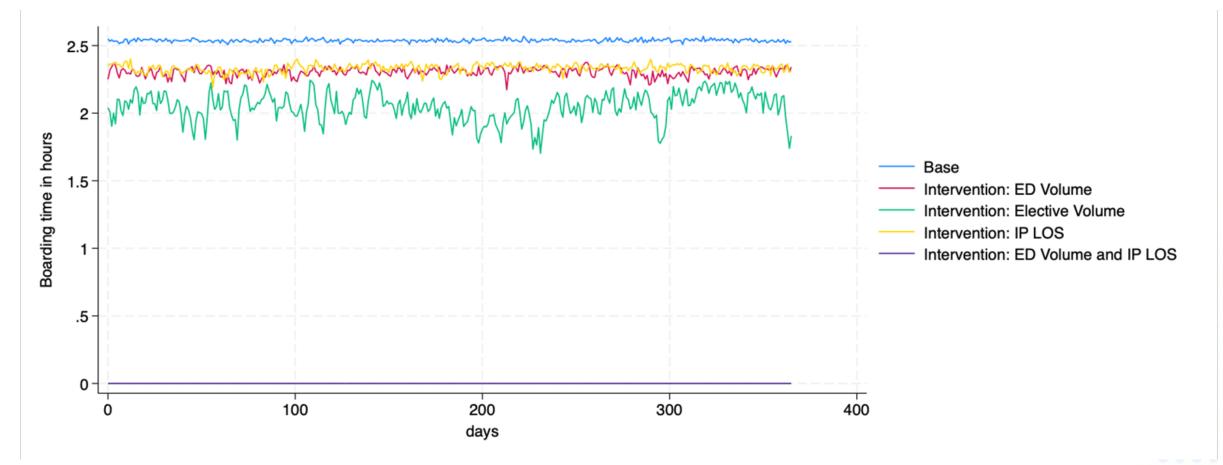
- Boarding time
- Elective admit wait time
- Total # of patients in ED

### How to interpret results?

- Model provides evidence of plausible effect of system changes
- Best viewed as qualitative/directional, rather than as precise estimates



## Small Interventions Yielded Large Improvements in Performance



Base case boarding time is ~2.5 hours. Small changes in ED volume, elective volume and LOS result in modest improvement. Combined intervention eliminates boarding time.



## Key takeaways

- Findings are consistent with our understanding of complex systems chaotic systems can be tamed with seemingly small, but carefully selected changes
  - By contrast, other changes, such as reducing patient treatment times in ED, may have unexpected consequences
- Interventions that are cheaper and/or quicker than adding physical beds may significantly improve patient experience and outcomes
- There are a wide variety of programs and policies that could achieve results similar to those shown here
- Simulation results are consistent with those from regression models
  - Hospital-level interventions can be effective
  - Reducing IP occupancy through better SNF/behavioral/hospice access and reducing ED volume through hospice and care management are important areas for further exploration



### Caveats

- There are likely many ways to build a model that uses real-world data and reproduces conditions on the ground
- There's no guarantee that our interventions, tested within a different but equally plausible model, would yield the same results
- However, our results are consistent with those from regression modeling, and also with principles developed over decades of research into complex systems theory
- The model does not address some important dynamics
  - Hour-by-hour fluctuation in ED arrival and IP departure volume
  - Specific actions to reduce LOS or ED volume

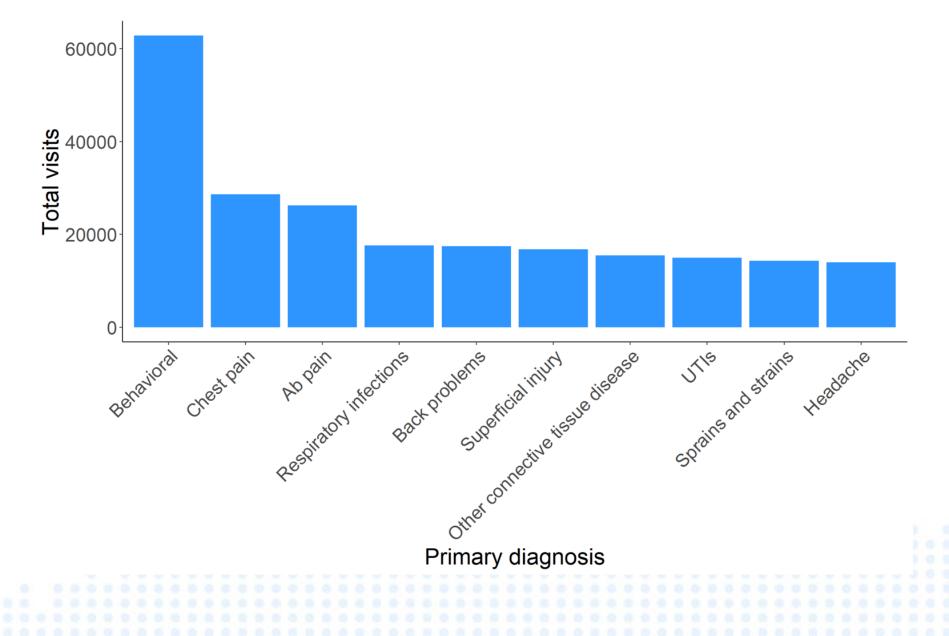




# **HSCRC** Opportunities



### The Multi-Visit Patient Opportunity



- MVP: Patient w/ >=4 ED visits in year
- Accounted for 29% of 2019 ED visits
- 18% were admitted
- Of outpatient visits by MVPs, 62% are for low-acuity principal diagnoses
- Wide variation in MVP ED visit and admission rates between hospitals



# **IP LOS Opportunity**

- IP LOS has increased since the confounding effects of the COVID-19 pandemic
- However, after accounting for acuity change and mix change (e.g., shift of surgeries to Outpatient and Ambulatory Surgery Centers), there still appears to be a statewide IP LOS increase of 4.26% from 2019 to 2022
  - Unadjusted LOS increased by ~16%
  - Variance of 4.26% between risk and mix adjusted LOS statistics suggests operational inefficiency opportunity
- If investments were made to make Maryland's risk adjusted LOS equivalent to 2019 experience, staffed bed capacity would increase by 246 beds
  - Would effectively add a new hospital; Average licensed bed size is 220 in FY 2024

Type of Bed Day	2022 Bed Days	Added Days (Current Bed Days X 4.26%)	Bed days if Risk Adjusted LOS Increase was Eliminated	
		,		
Medical Surgical Acute	1,665,578	70,876	1,594,702	
Medical Surgical Intensive Care	245,980	10,467	235,513	
Oncology	45,511	1,937	43,574	
Definite Observation	60,499	2,574	57,925	
Shock Trauma	34,391	1,463	32,928	
Pedatrics Acute	34,002	1,447	32,555	
Pediatric Intensive Care	17,831	759	17,072	
Burn Care	1,755	75	1,680	
Coronary Care	5,070	216	4,854	
Total	2,110,617	89,813	2,020,804	
Bed Count (Total/365)	5,783	246	5,536	



# Summary of ED-related Policy Recommendations

### Recommendations for ongoing measurement and engagement

- EDDIE Continue to steward rapid cycle improvement in ED performance
- Other Efforts Coordinated with Maryland Hospital Association

### Recommendations for payment policy

- Quality-Based Reimbursement (QBR) policy Staff proposal provides new incentive for improvement on CMS ED-1 measure
- Multi-Visit Patient policy Financial reward for reduction in percentage of ED visits accounted for by patients with 4 or more visits per year
- Workgroup to monitor impact of policies on ED performance, propose payment policy changes and provide periodic reporting to General Assembly
  - Potentially establish a stand-alone pay-for-performance program weighted at 1% of inpatient revenue that incents improvements in ED LOS root causes and continued improvement in EDDIE.



# **Goal:** Leave the meeting with 3 top priorities to align and focus efforts of the subgroups.

\*See next slide for key priority suggestions based on stakeholder feedback.





# **Priorities Discussion**



# **Key Priorities**

- Key Priority Identified: Hospital Throughput & ED Boarding
- Staff recommend focusing on the following key drivers impacting hospital throughput & ED boarding:
  - Optimize capacity across the continuum of care (ambulatory, acute, post-acute, and community resources)
    - Utilize Access Study Analysis to prioritize regional capacity needs
    - Distribution of inpatient beds
  - Care transitions within the hospital that impact throughput (best practice subgroup focused on these efforts)
    - Incentivize health systems to make operational changes that reduce Inpatient Length of Stay, reduce ED boarding and improve overall hospital throughput
  - Care transitions to post-acute levels of care, inclusive of skilled nursing, palliative care, and home health
    - Improve discharge processes and address transitions delays to post-acute care





# 2025 Legislative Session





# Subcommittee Updates



### **Commission Subcomittees**

### Access to Non-Hospital Care

- Integrate and optimize best practices and data analytics for advanced primary care, specialty care, home health, post-acute care, and ancillary services in an effort to reduce avoidable ED and hospital utilization and improve care transition workflows throughout the continuum of care.
- Meetings every six to eight weeks.

### **Data Subcommittee**

- Identify different data sources across healthcare platforms to include ambulatory, acute care, postacute care, and third-party data.
- Meetings every six to eight weeks.

### ED Hospital "Throughput" Incentives

- Develop a set of hospital best practices and scoring criteria to improve overall hospital throughput and reduce ED length of stay, advise on revenue at-risk and scaled financial incentives, and provide input on data collection and auditing.
- Meetings every four weeks.

#### Hospital Capacity, Operations & Staffing

- Subgroup will convene in April 2025.
- Planned focus of the subgroup is to assess access and capacity across the State, collaborate with commercial payers, Medicare, and Medicaid, and optimize workforce development opportunities.
- Meetings every four to six weeks.



### Subcommittee Updates

- Access to Non-Hospital Care
  - Top priorities identified are advanced primary care and post-acute (discharge barriers to postacute and post-acute capacity).
  - Consider engagement with PointClickCare for focused discussions on post-acute care transitions
     and capacity opportunities
  - Next Meeting: February 6<sup>th</sup>
- ED Hospital "Throughput" Best Practices
  - Best Practices Policy Draft presented to HSCRC Commission on 1/8. Final policy will be presented in March. *Discussed on next slide.*
- Data Subcommittee
  - 1<sup>st</sup> meeting scheduled for February 5<sup>th</sup>
- Hospital, Capacity, Operations & Staffing
  - Plans to convene April 2025



## ED Hospital Throughput Best Practices Subgroup Activities

<ul> <li>April</li> <li>Examined CMS ED LOS of MD vs. Nation Performance</li> <li>Evaluated ED 1 measures, and decided ED 1b stratification (non- Psych patients)</li> <li>Introduced Incentive structures and risk adjustment for occupancy and discharge disposition</li> </ul>		<ul> <li>Discussed the nomenclature and not other</li> <li>Explored the observation s in the measur</li> <li>Conducted E improvement</li> </ul>	<ul> <li>July</li> <li>Discussed that the measure nomenclature focuses on the ED and not other stakeholders</li> <li>Explored the impact of observation status being included in the measure</li> <li>Conducted ED LOS improvement literature review</li> <li>Explore improvement modeling scenarios</li> </ul>		<ul> <li>September</li> <li>Discussed alignment with Maryland ED Wait Time Reduction Commission</li> <li>Identified six ED best practices from data collection and members ranked the top recommendations</li> <li>Established a sample measure scoring example based on a best practice</li> </ul>		October • Attended the AHRQ Webinar on ED Boarding to inform work • Finalized 4 of 6 ED best practice interventions	
		ier system that ociated KPIs	• Finalized all 6 interventions	<b>cember</b> 6 ED best practice tted for review on	<ul> <li>Policy prese Commission</li> <li>Policy commission</li> <li>2/19 and prepolicy to HS</li> </ul>	nent period 1/8 to epare to present final CRC in March and measure at maryland	services	

### **Best Practices Subcommittee Update**

Draft Recommendation for RY 2027 (CY 2025 Performance Period)



- Building upon the ongoing work of staff and key stakeholders, refine the specifications developed by the Best Practice subgroup on a set of up to six Hospital Best Practices that are designed to improve emergency department (ED) and hospital throughput and reduce ED length of stay (LOS).
  - For each best practice identified, develop three weighted tiers with corresponding measures that reflect the fidelity and intensity of each best practice.
- 2. Require hospitals to select two Best Practices to implement and report data on for RY 2027
  - Failure to implement and report data to the Commission by October 2025 will result in a 0.1 percent penalty on all-payer, inpatient revenue to be assessed in January 2026.
- 3. We propose that subsequent rate years will have 0.25 percent inpatient hospital revenue at risk tied to performance on these best practice metrics but intend to evaluate the impact of the best practices and make a final recommendation for subsequent rate years after the Year 1 Best Practice program impact is assessed.



### **Final Six Best Practices Selected**

Each hospital will select 2 interventions from the 6 interventions below:

- Interdisciplinary Rounds
- Bed capacity Alert Process
- Standard Daily/Shift Huddles
- Expedited Care Bucket (inclusive of expediting team, rapid medical evaluation team, rapid medical evaluation unit and patient observation management)
- Patient Flow Throughput PI Council
- Establishing Clinical Pathways



### **Benefits of Best Practices Proposal**

- Increased focus on ED & Hospital Throughput
- Significant collaboration within and across hospitals
- Foundation for Quality Improvement Partnership



### **Best Practices Next Steps**

- Continue development of measure definition, tiers, and targets with hospital groups
- Comment period through 2/19
- Final policy presented to HSCRC Commission at March Commission meeting



### **Next Steps**

- Next Meeting: March 26, 2025
- Please visit the <u>ED Wait Time Reduction Commission Webpage</u> for all materials.





# Appendix





- MHA Report: <u>https://mhaonline.org/caring-for-communities/quality-</u> safety/hospital-throughput/general-assembly-hospital-throughput-workgroup/
- <u>The ED Capacity Crisis: Hard Truths and Real Solutions from NYC's</u> <u>Mount Sinai (February 4th, 2025 | 12:00 PM - 1:00 PM CT)</u>

