Leveraging Electronic Data in "Real-Time"

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Outline

- Background
- 2. What is real-time?
- 3. Tracking the essentials "Air Traffic Control"
- 4. Identifying patterns "Weather Forecasting"
- 5. Some final thoughts







Background

- Learning Health Systems Researcher
 - NICHD K23: "A Learning Health Systems Approach to Precision Sedation and Analgesia for Critically III Children"
 - NINDS R01: "A Biodigital Rapid Alert for Identifying Neuromorbidity in Critically III Children"
- Operations Informatics
 - UPMC Department of Critical Care Medicine
 - >40 Adult and Pediatric ICUs
 - Cerner x2, Epic x2, Allscripts, Medtech
- Embedded Clinical Trials
 - EHR to support traditional EDC
 - Workflow integration







What is "Real-Time"?

On-demand data or information with essentially no delay



- Discern explorer
- Tap into a recovery server
- Analytic instance
- ETL data for specific use cases



- Clarity reporting DB
- Cognitive computing platform
- Slicer Dicer



- Multiple data sources
- 'Real-time' ETL at a cadence of the slowest refresh













Tracking the essentials -"Air Traffic Control"









Tracking the essentials -"Air Traffic Control"

System Capacity – Can I accept this patient?

Patient Transfers and Triage - Where do I put this patient?

Tracking Staffing/Personnel - Who is going to care for this patient?

Supply Chain Management - Do we have what we need to provide care?

Strategic Stockpile Management - How long can we last?

'Real-time' Patient Identification and Enrollment - Where is the patient?







METHODOLOGY

Open Access

Implementation of the Randomized Embedded Multifactorial Adaptive Platform for COVID-19 (REMAP-COVID) trial in a US health system—lessons learned and recommendations



The UPMC REMAP-COVID Group, on behalf of the REMAP-CAP Investigators¹







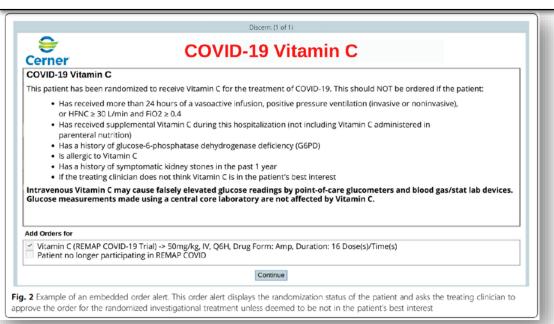
Advantages 1, 2 and 3

1) Patient Identification

UP	MC COVID-19 Intake Form	FIN: 123456789					
Symptoms							
What symptoms were present prior to or at the time of presentation? (Check all that apply)							
	Abdominal Pain	Fever (Temperature >=38°C / 100°F) or subjective fever					
	Body Aches / Myalgias	Headache					
	Chills	Malaise / Fatigue / Lethargy					
	Confusion	Nausea or Vomiting					
	Cough	Shortne					
	Diarrhea (>/= 3 loose/looser than normal stools/24 hr period)	Sputum None 2) Trial Data					
How many days since the onset of symptoms? (The number recognized by the patient) Collection							
Anticipated Treatment							
Please read the following to the patient or their representative and record their response:							
"There are few known treatments for COVID-19. Would the patient like to hear about potential additional therapies and studies?" She or he will only be contacted if potentially eligible.							
○Yes	○ No ○ Ur	hknown ease only click if the patient or representative is unable to respond. An unknown response will lay potential additional therapies and studies.)					
requests p		bedded into a patient's electronic health record, solicits basic clinical information, and if representative if s/he is interested in potential additional therapies for COVID-19. The REMAP-COVID trial at UPMC					

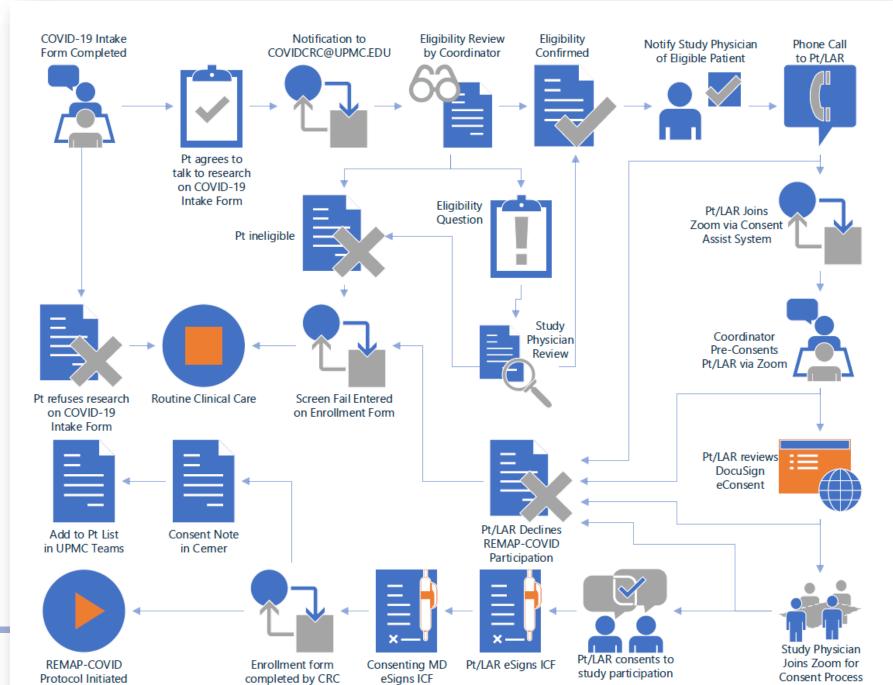
Trials

3) Deploy Interventions







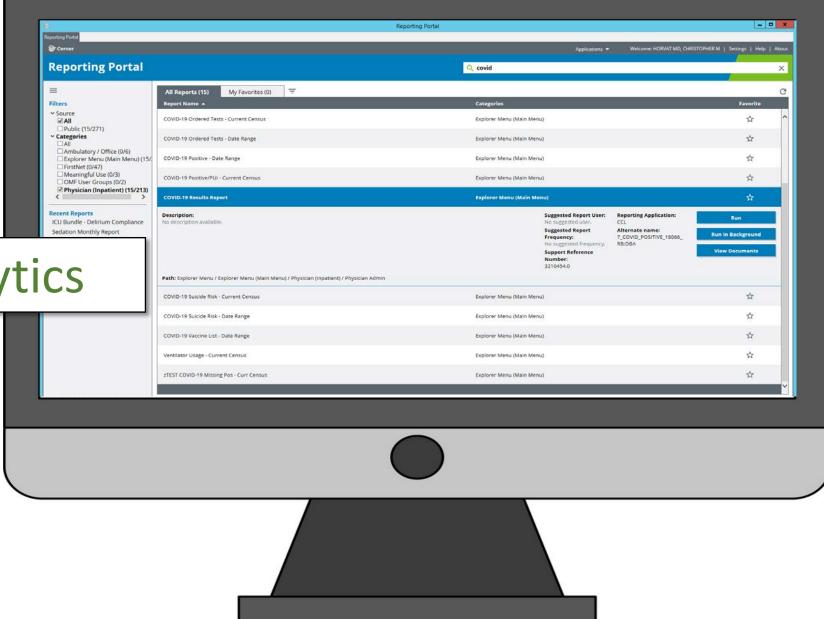






Advantage 4

4) EHR-based Analytics

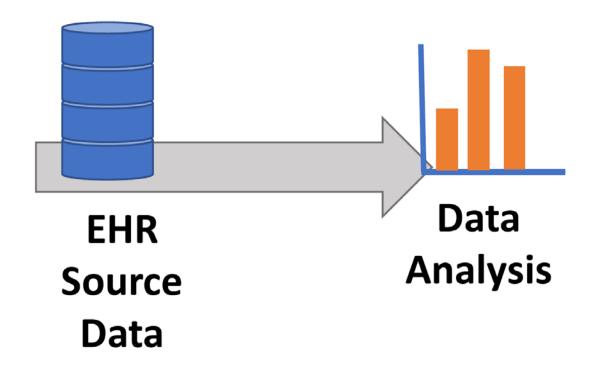






Advantage 5

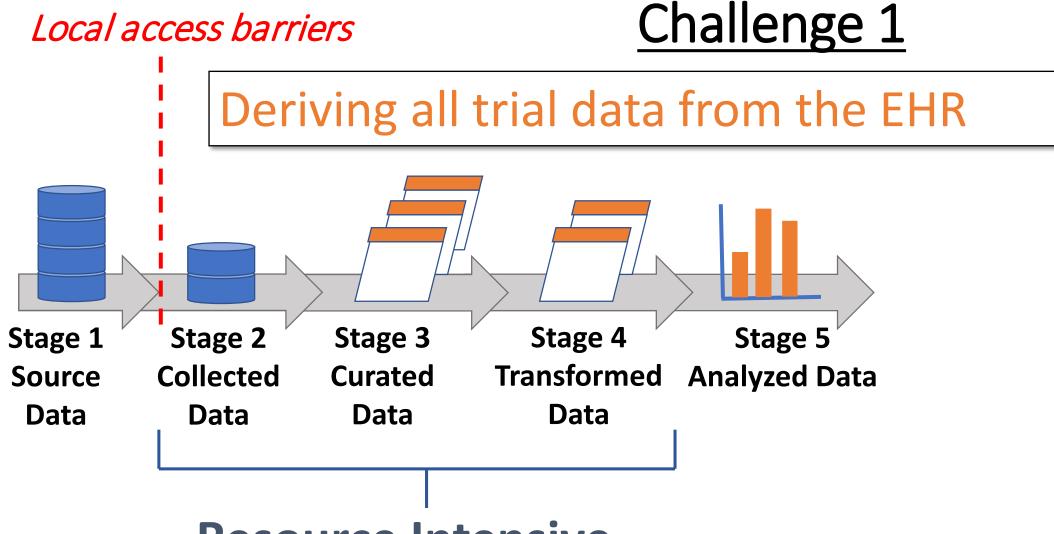
5) Deriving all trial data from the EHR



















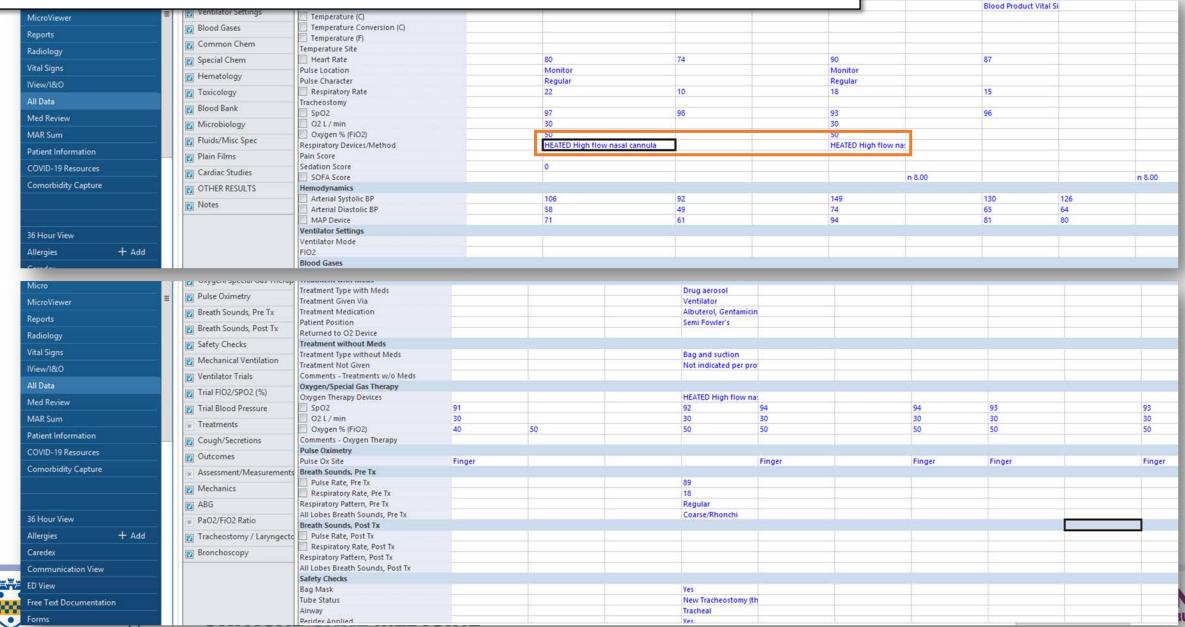
Deriving all trial data from the EHR

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MicroViewer	■ Hemodynamics	Temperature Site			dder	Bladder		
	☑ Ventilator Settings	Heart Rate		73	9211	78		
Reports	Blood Gases	Pulse Location		Mo	nitor	Monitor		
Radiology		Pulse Character		Re	gular	Regular		
10.12.13.10.000	Other ICU Labs	Respiratory Rate		8		11		
ital Signs	Common Chem	Ventilator Rate				10		
/iew/I&O		Total Respiratory Rate				10	_	
116.1	Special Chem	Respiratory Rate #2		10		10	_	
II Data	Hematology	SpO2 02 L / min		97		96	-	
Med Review		Oxygen % (FiO2)		40		40	_	
MAR Sum	Court (Respiratory Devices/Method			dotracheal Tube, Ve	Endotracheal Tub	e Ve	
	Immunodeficiency	Sedation Score			octocitical rabe, v.	3	-,	
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OVID-19 Resources		Vital Signs Comments				ECMO circuit set a	at 37	
STANDAMENTAL STANDAMENT	Blood Bank	Hemodynamics						
Comorbidity Capture	Microbiology	Arterial Systolic BP		105		116		
	Plain Films	Arterial Diastolic BP		59		64		
 ;		MAP Device		70		77		
	CT CT	Ventilator Settings						
6 Hour View	Cardiac Studies	Ventilator Mode				A/C		
A SECTION OF THE PROPERTY OF T	(Mary)	FiO2 - Ventilator				40	40	
llergies + Add	OTHER RESULTS	Tidal Volume - Set				450		
aredex	Notes Notes	Pressure Support Ventilation Peak Pressure				26	-	
		Endotube Placement				28	-	
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D View		FIO2				Center reedi		
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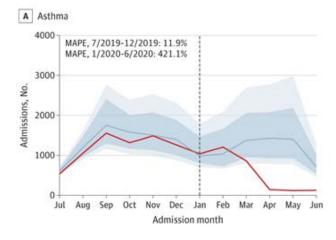


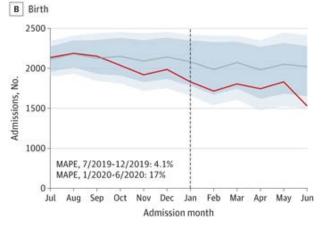
Deriving all trial data from the EHR

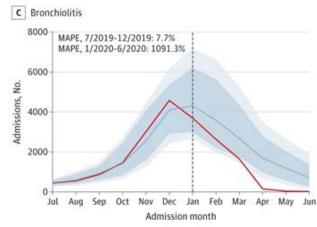


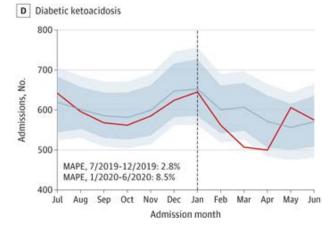
Identifying Patterns – "Weather Forecasting"

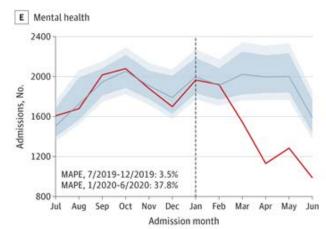


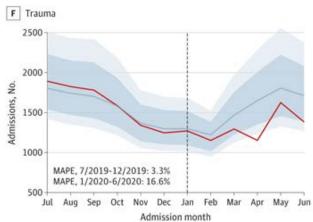






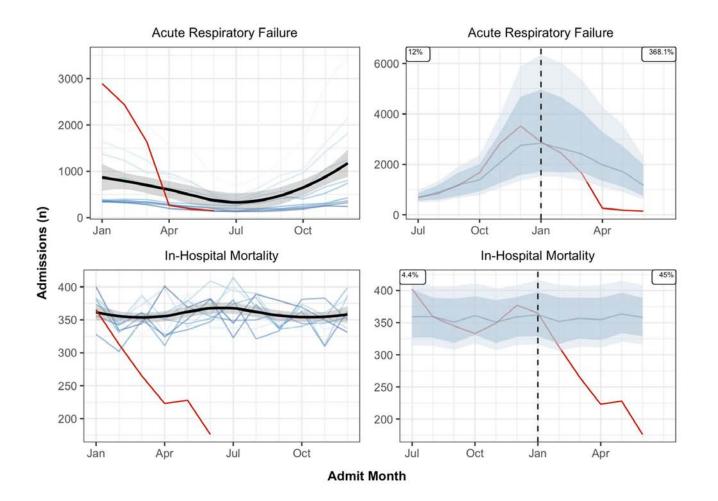


















"...we're all just gambling with probabilities." - Henry Fonda, 12 Angry Men







The origin of "evidence-based medicine"

"The evidence presented shows that physicians do not manage uncertainty very well, that many physicians make major errors in probabilistic reasoning, and that these errors threaten the quality of medical care."

- David Eddy, MD PhD

https://www.cambridge.org/core/books/judgment-under-uncertainty/problems-and-opportunities/661E12D1ECD669EDB5B410407A4BB570

Table 1. Accuracy of mammography in diagnosing benign and malignant lesions

Results of X ray	Malignant lesion (cancer)	Benign lesion (no cancer)
Positive	.792	.096
Negative	.208	.704

Source: The numbers are from Snyder (1966).

Bayes' formula can be applied to assess the probability. This formula

$$P(\text{ca}|\text{pos}) = \frac{P(\text{pos}|\text{ca}) P(\text{ca})}{P(\text{pos}|\text{ca}) P(\text{ca}) + P(\text{pos}|\text{benign}) P(\text{benign})}$$

where

P(ca|pos) is the probability that the patient has cancer, given that she has a positive X-ray report (the posterior probability)

P(pos|ca) is the probability that, if the patient has cancer, the radiologist will correctly diagnose it (the true-positive rate, or sensitivity)

P(ca) is the probability that the patient has cancer (prior probabili-

P(benign) is the prior probability that the patient has benign disease [P(benign) - 1 - P(ca)]

P(pos benign) is the probability that, if the patient has a benign lesion, the radiologist will incorrectly diagnose it as cancer (the false-positive rate)

Table 1 summarizes the numbers given by Snyder. The entries in the cells are the appropriate probabilities (e.g., P(pos | ca) = .792).

Using 1% as the physician's estimate of the prior probability that the mass is malignant and taking into account the new information provided by the test, we obtain

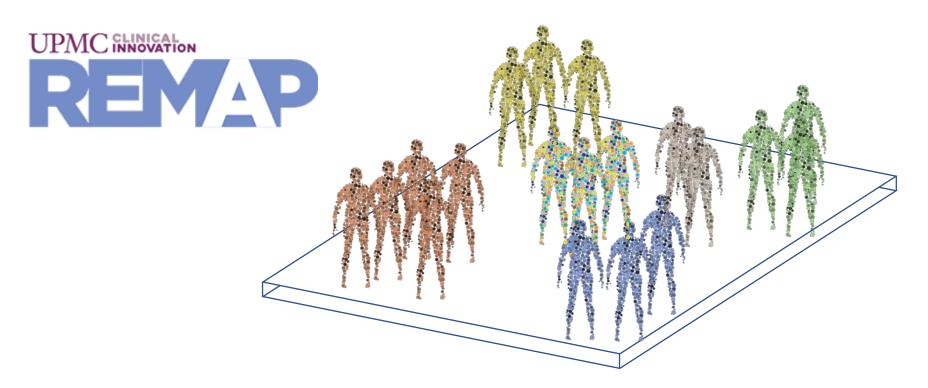
$$P(\text{ca} \mid \text{pos}) = \frac{(0.792)(0.01)}{(0.792)(0.01) + (0.096)(0.99)} = 0.077$$

Thus, the physician should estimate that there is approximately an 8% chance that the patient has cancer.









$$\begin{split} \log\left(\frac{\pi}{1-\pi}\right) &= \sum_{R=1}^{R} \nu_R + \sum_{k=1}^{K} \sum_{s=1}^{S} \alpha_{s,g_k} + \sum_{age=1}^{AGE} \lambda_{age} + \sum_{T=1}^{T} \theta_T + \sum_{d=1}^{D} \sum_{j=1}^{Ja} \beta_{d_j} \\ &+ \sum_{k=1}^{K} \sum_{d=1}^{D} \sum_{j=1}^{Ja} I(g_k = 2) \gamma_{kd_j} + \sum_{d=1}^{D} \sum_{j=1}^{Ja} \sum_{d'=d+1}^{D} \sum_{j'=1}^{J'_{d'}} \delta_{d_j d'_{j'}} \end{split}$$

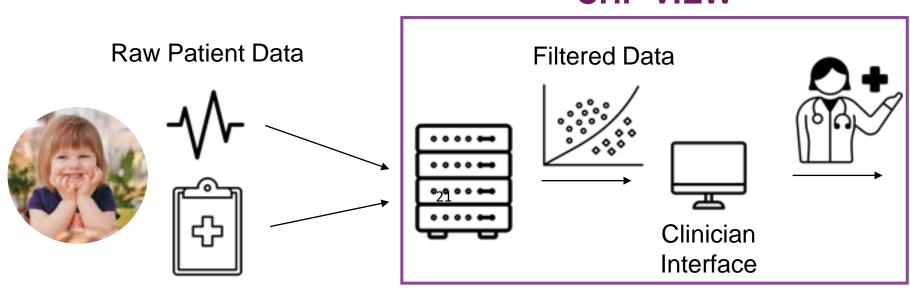
Unified Statistical Analysis Plan





Intelligence Augmentation (IA)

CHP VIEW





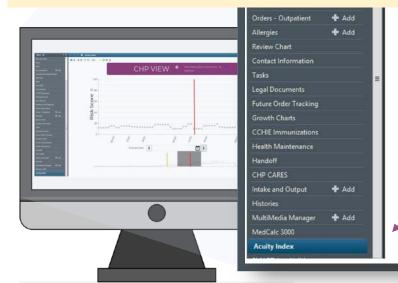




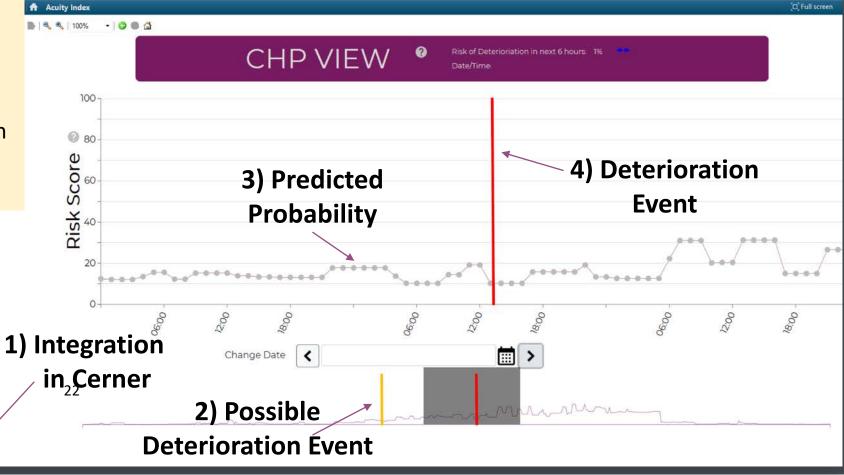


Probability of **deterioration in the next 6 hours** defined by any one of the following:

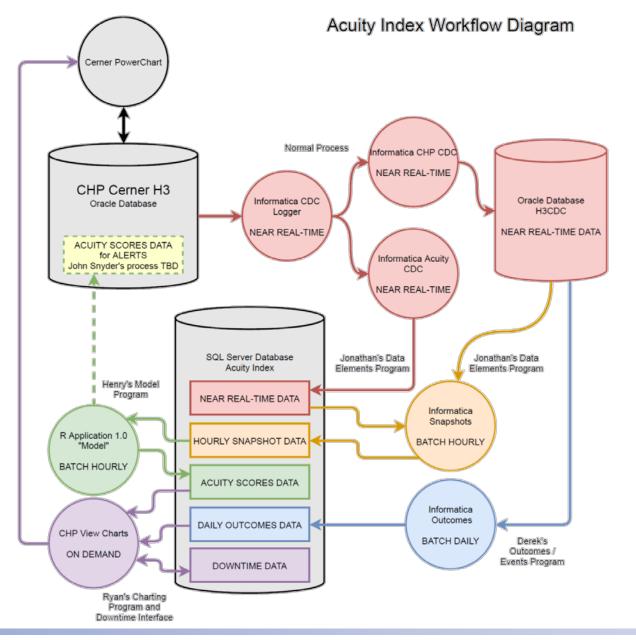
- CPR
- Condition A or C
- Multiple Blood Transfusions
- Emergent Transfer to the ICU
 - PPV or vasoactive gtt within
 12 hours
- ECMO



Human Centered Design



















Summary

- 'Real-time' can refer to a range of cadences and depends on technical infrastructure
- Tracking the essentials is a major undertaking
- Advanced analytics can augment our current capabilities
- There is no such thing as 'automation' this all requires substantial investment











