



Lessons Learned Implementing Machine Learning in Workflows

ROADMAP



Background



Embed in Workflows



Measure & Monitor

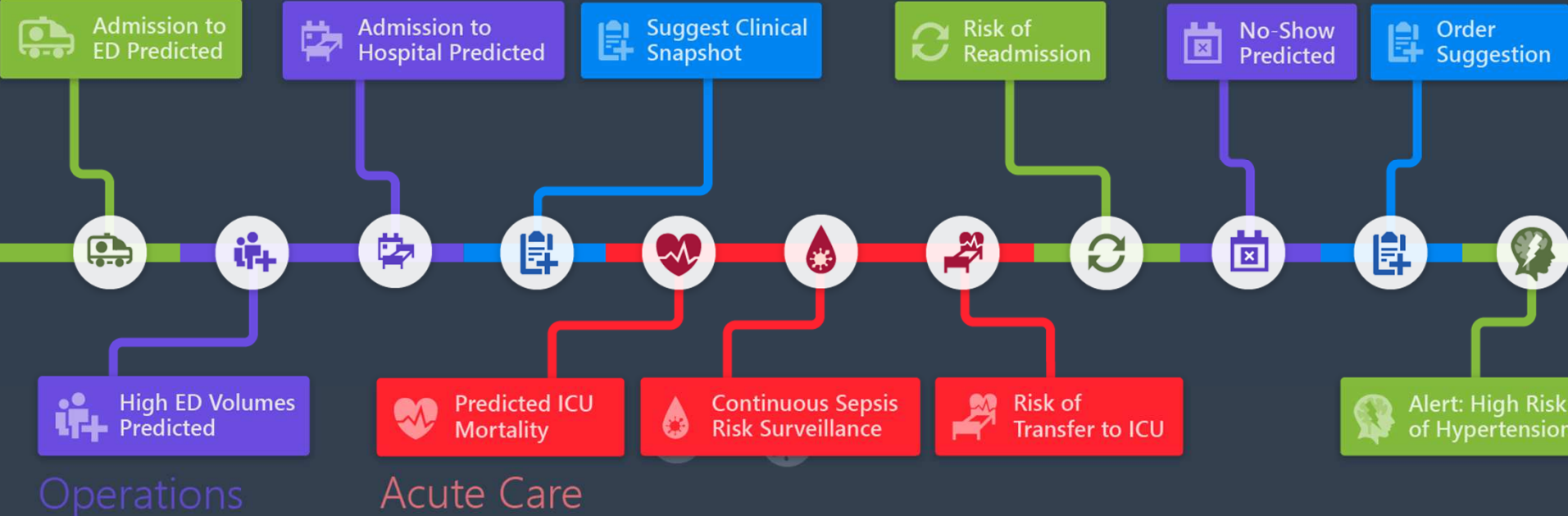


Future Applications

ACROSS THE CONTINUUM

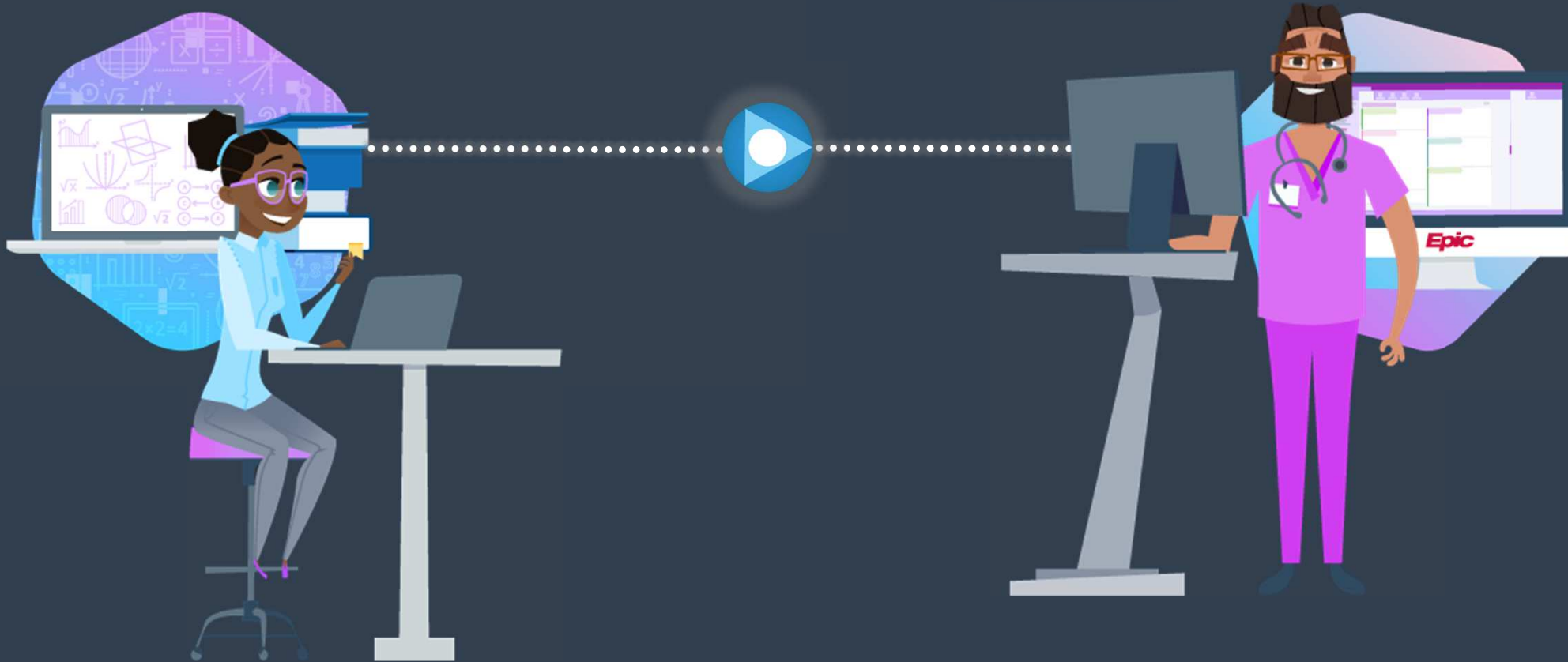
Pop Health

Precision Charting



Rule #1: Don't be afraid to launch a product without machine learning.

LAST MILE CHALLENGE



EPIC MACHINE LEARNING

160

organizations with
models in production



17

machine learning
models





Better Math:
Machine Learning
Algorithms

More Data:
Weather, Local Events,
Census

+

Adapt:
To local populations

Enhanced Decision-making.



Embed in Workflows

DESIGNING A WORKFLOW



Knowledge Driven

MDs & RNs

Domain knowledge
Workflow focus

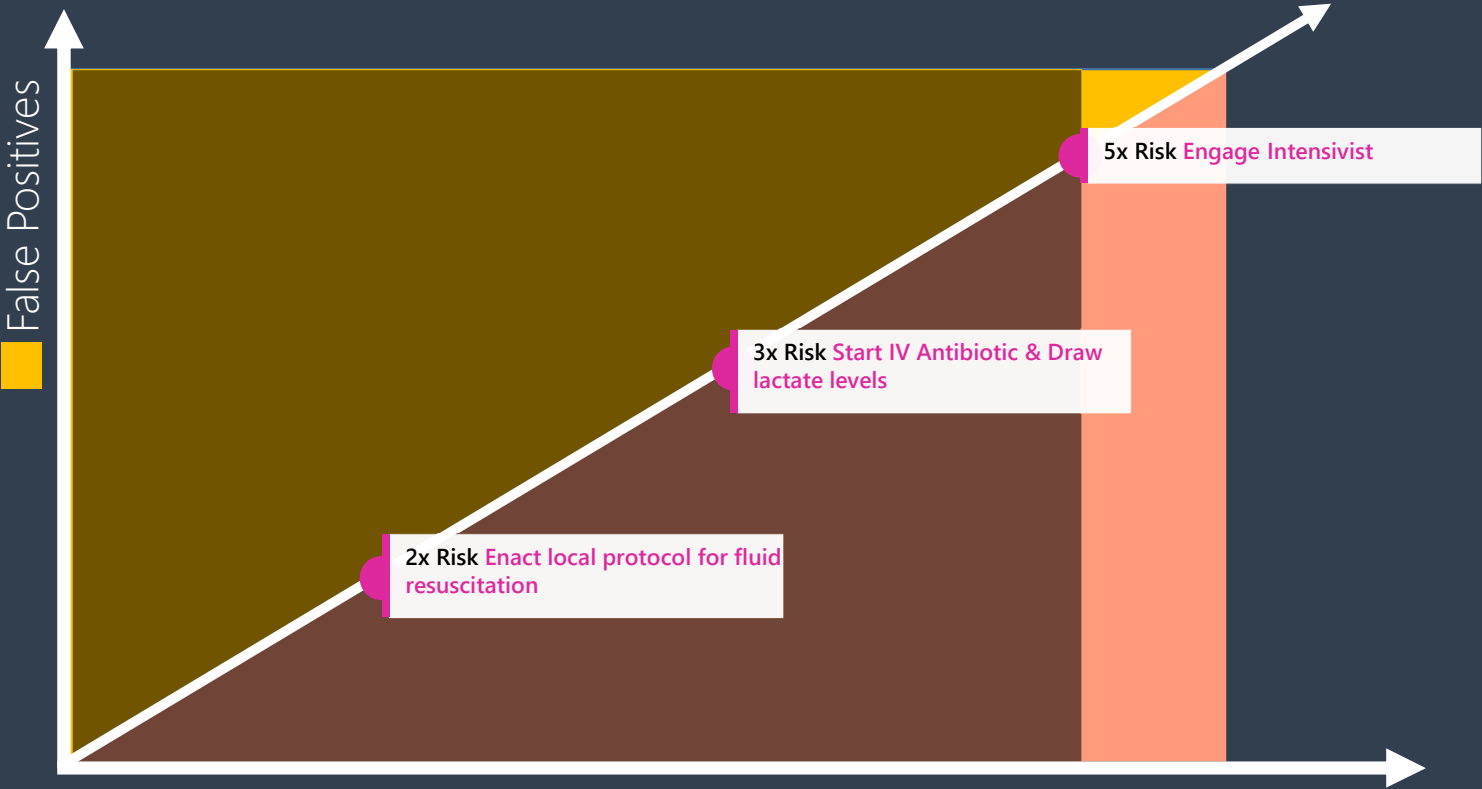


Data Driven

Data Scientists

Statistical approach
Data reveals insight

INTERVENTION SLIDING SCALE



DEFINING INTERPRETABILITY

Explainable AI (XAI)

Understand results of the model

PRESENTING INSIGHTS



Measure & Monitor

DEFINING MODEL ACCURACY

Risk of Hospital Admission or ED Visit Model Validation

Choose a cutoff threshold. A patient with a risk score above the cutoff threshold is predicted to have a hospital admission or ED visit in the next year. The default medium risk threshold in the system is 0.2.

Cutoff Threshold:

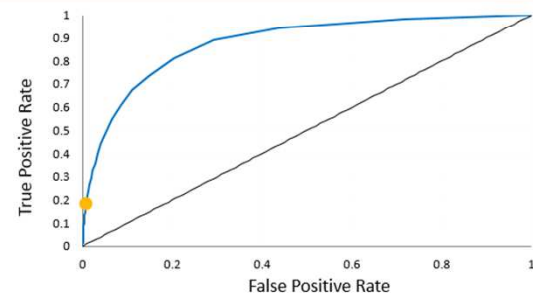
Performance Measures

Accuracy: The number of patients correctly identified divided by the total number of patients.	0.910
True Positive Rate: The number of patients who were in the hospital or ED in the last year with a risk greater than the threshold / total number of patients who were in the hospital or ED in the last year. Higher is better.	0.183
False Positive Rate: The number of patients who were <i>not</i> in the hospital or ED in the last year with a risk greater than the threshold / total number of patients who were <i>not</i> in the hospital or ED in the last year. Lower is better.	0.008
Positive Predictive Value: The number of patients who were in the hospital or ED in the last year with a risk greater than the threshold / total number of patients with a risk greater than the threshold. Higher is better.	0.732
Negative Predictive Value: The number of patients who were <i>not</i> in the hospital or ED in the last year with a risk lower than the threshold / total number of patients with a risk lower than the threshold. Higher is better.	0.914
C - Statistic: Also known as AUC, this is the area under the ROC curve. A C-statistic of 0.5 is no better than chance, while a C-statistic of 1.0 represents perfect accuracy.	0.880

Confusion Matrix



ROC Curve



A receiver operating characteristic curve, or ROC curve, is a graphical plot that illustrates the performance of a binary classifier system as its discrimination threshold is varied. The curve is created by plotting the true positive rate against the false positive rate at various threshold settings. This indicates how the model's ability to discriminate between positive and negative cases compares to pure chance (pure chance is depicted as the black line).

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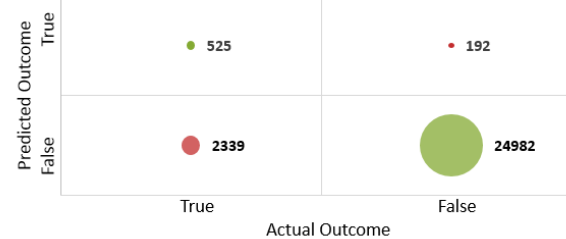
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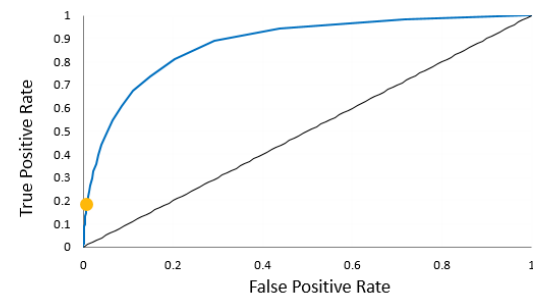
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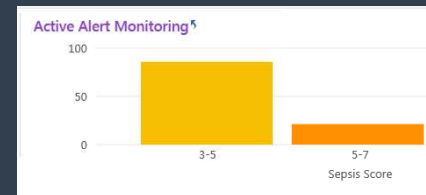
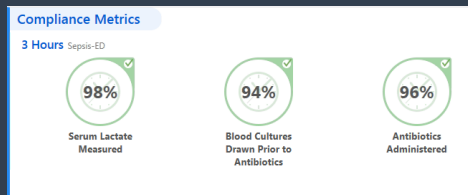
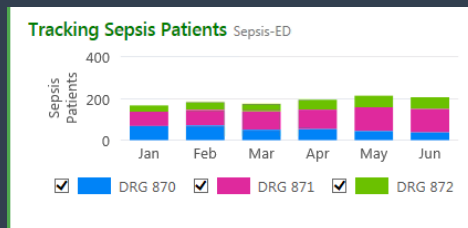


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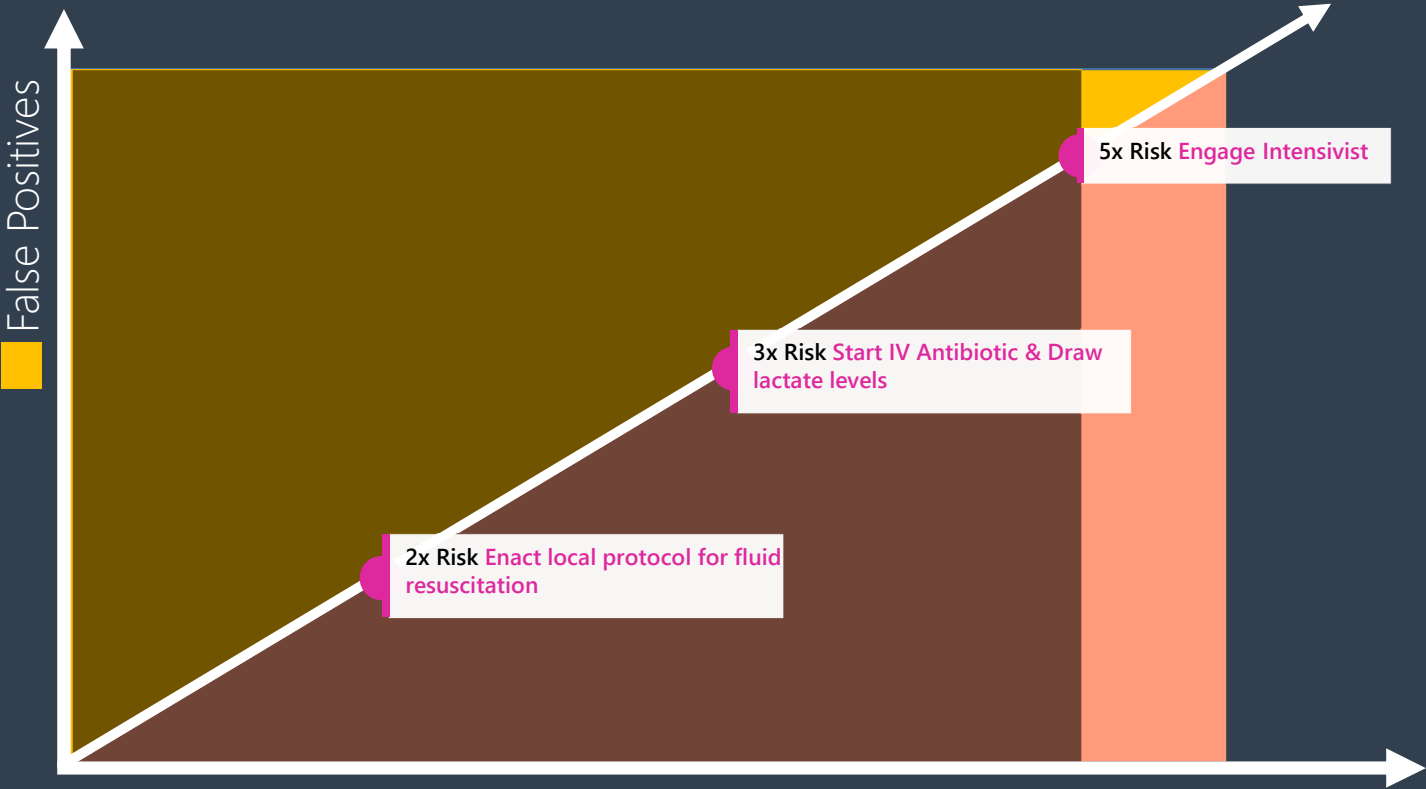


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MONITORING OUTCOMES



OPTIMIZING THRESHOLDS





Future Applications

EXPANDING MACHINE LEARNING



Resource Maximization
Streamline operations



Helping Physicians
real-time guidance and recommendations

HELPING PHYSICIANS



Helping Physicians

real-time guidance and recommendations



**Shopper
One**

**Shopper
Two**

HELPING PHYSICIANS



Helping Physicians

real-time guidance and recommendations

ONGOING MONITORING



CONCLUSIONS



Design a workflow to maximize benefit to end users



Establish a strategy for optimizing thresholds & monitoring impact



Be thoughtful in ongoing model management capabilities