# The Surg-e-Screener HF Tool: Improving Heart Failure Recognition using Electronic Health Records

#### e-Learning Module

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### The Problem



Treatments for heart failure (HF) proven to extend & improve quality of life are <u>limited by clinicians' ability to diagnose the disease</u> in early stages.





### The Problem

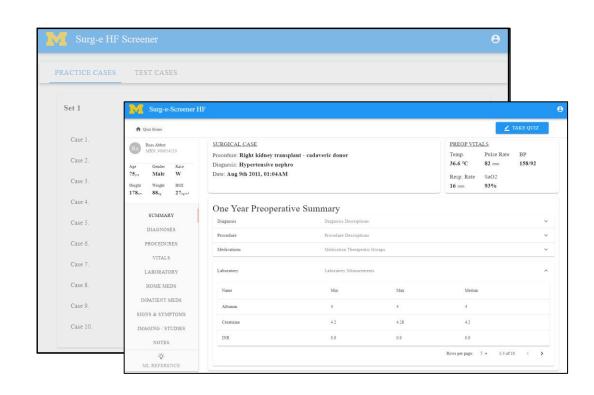


Unrecognized or untreated heart failure is the single greatest risk factor for <u>cardiovascular complications and mortality</u> after major surgery.





# The Solution: Surg-e-Screener HF



#### **Surg-e-Screener HF:**

Automated screening & educational tool

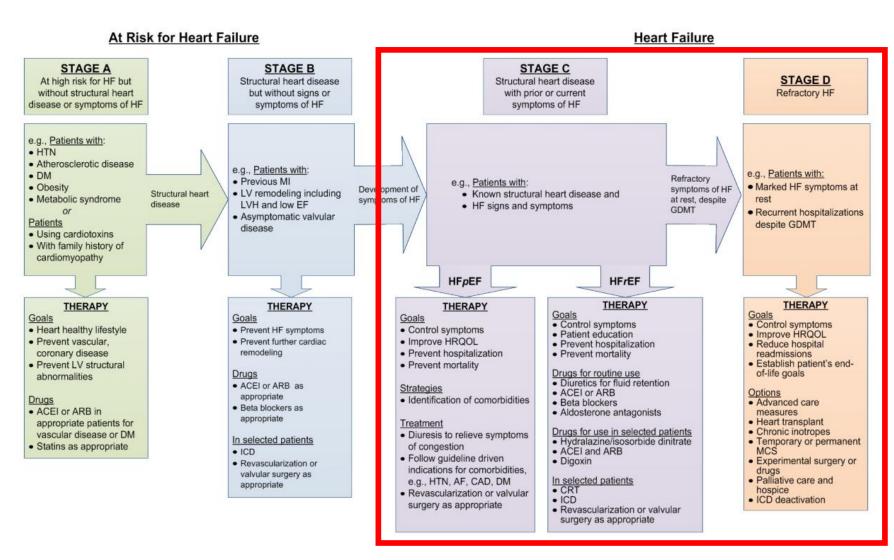
#### **Education & Training –**

- It is a quiz-based webapp to improve HF recognition with surgical cases in EHR
- It provides a report after completing quizzes to compare your clinical judgment to a pane of HF experts
- It is designed to augment clinicians' ability using EHR data and machine learning (ML) reference





#### **Chronic Heart Failure Definition for this Tool**



For purposes of this tool, **Chronic HF** defined as:

• Chronic:

Signs +/- symptoms or underlying pathophysiology persistent for  $\geq 3$  months (with onset defined as the *start* of this  $\geq 3$  month period)

• HF:

ACCF/AHA Stage C or D (*prior* or current symptoms + structural heart disease)



Yancy CW, Jessup M, Bozkurt B, et al. 2013 ACCF/AHA guideline for the management of heart failure: a report of the American College of Cardiology Foundation/American Heart Association Task Force on Practice Guidelines. *Journal of the American College of Cardiology*. 2013;62(16):e147-239.



# Purpose of this study

To understand how EHR-based <u>heart failure (HF) recognition</u> can be improved as aided by the Surg-e-Screener HF tool.





### **Case Review Process**

#### 1. Case Reviews



Assume you are a preop physician reviewing the patient's medical history prior to surgery in order to identify factors pertaining to presence/absence of HF

- Previous HF diagnoses (simplest way to detect, but occasionally inaccurate)
- Signs & symptoms, labs, studies/imaging, medications, clinical notes.

2. <u>Pre-test</u>: HF Recognition Quiz



For the first 10 cases, you <u>use EHR data</u> to decide if a patient had HF (chronic ACC/AHA Stage C or D) before the start of the surgery.

3. ML Reference



After completing the quiz, you have access to the *Machine Learning (ML) reference,* ML performance indicators and a list of risk factors ranked by an algorithm.

4. <u>Post-test</u>: HF Recognition Quiz



For the next 10 new cases, you <u>use EHR data</u> **AND** <u>the ML reference</u> to decide if the patient had HF (chronic ACC/AHA Stage C or D) before the start of the surgery.

5. Expert Review Comparison



After completing 20 surgical cases, you can access a result report, comparing your HF decision to pre-determined answers and a short case summary by HF experts.

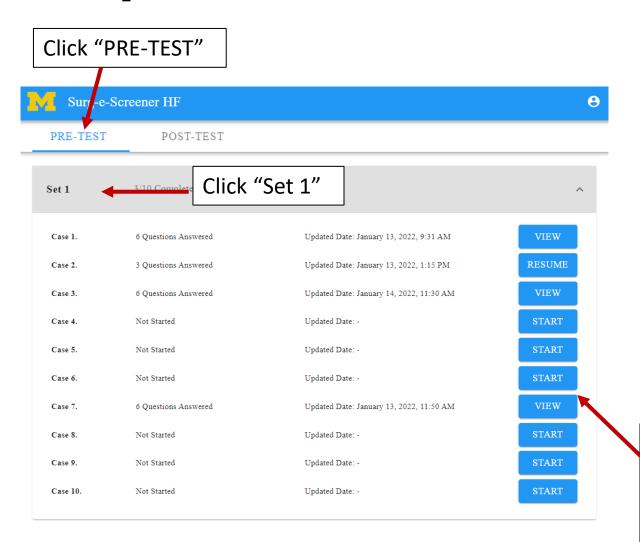
### How to use the tool:

Step-by-step Guide





### **Step 1. Case Reviews**



#### **Check your assigned cases**

- 10 surgical cases are assigned to *Pre-Test* and
   10 surgical cases are in *Post-Test*.
- Click Pre-Test to review 10 surgical cases.
- Click the *Start* button to begin your case review. The button indicates your status.

#### **Status of your progress:**

**START** No questions completed.

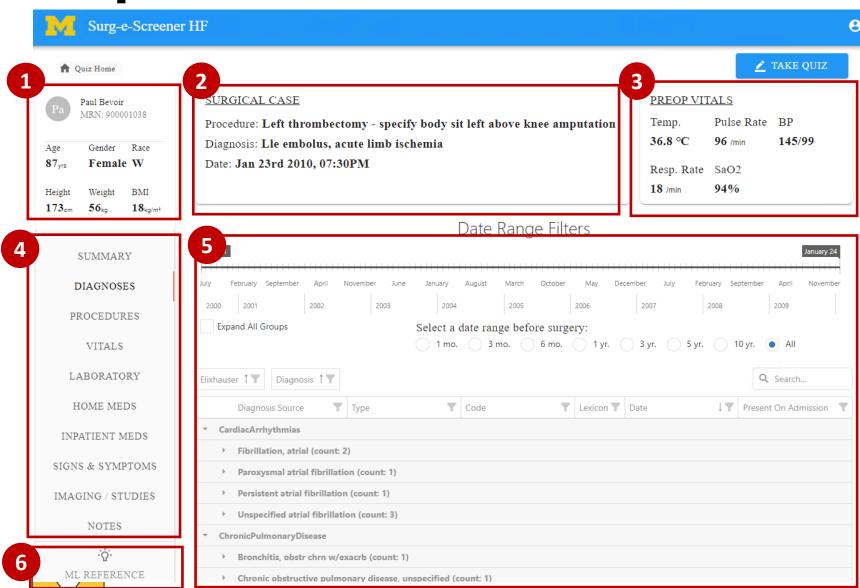
**RESUME** Partially answered questions, not yet completed.

**VIEW** Completed questions. You can *only* view the case.





# Step 1. Case Reviews: Dashboard

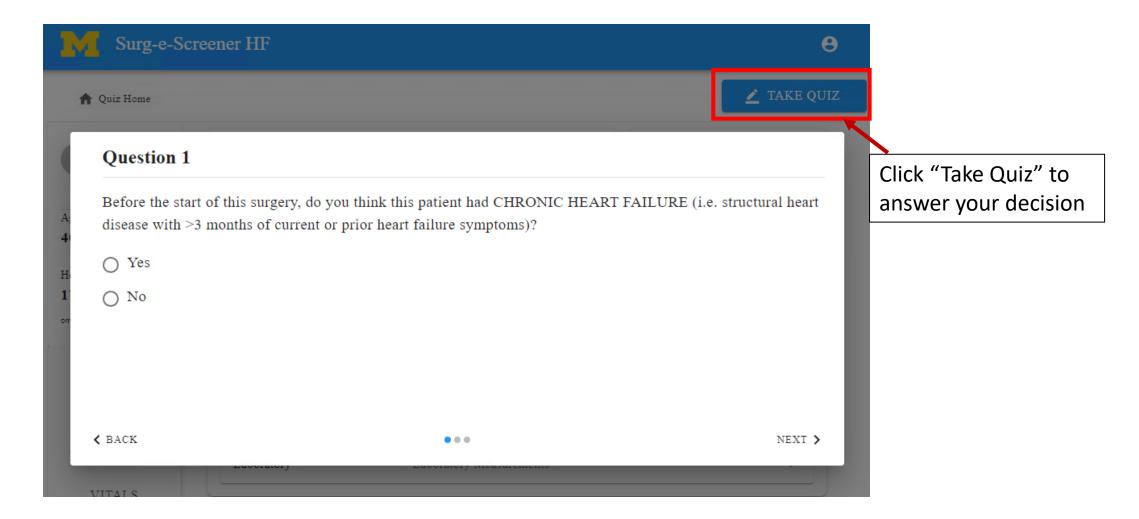


#### **Key components**

- 1. Demographics
- 2. Surgical Case
- 3. Preoperative Vitals
- 4. Subject Domains
- 5. Date Filter & EHR data
- 6. ML reference



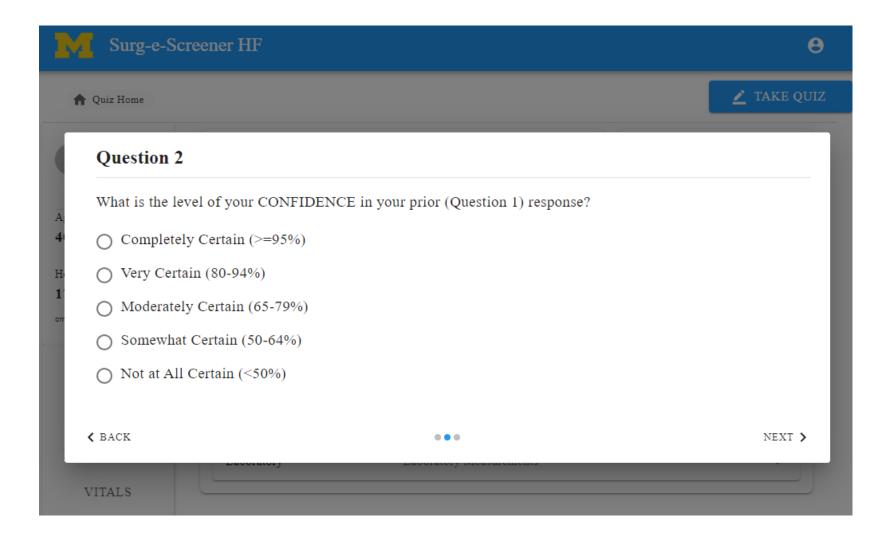
# Step 2. Pre-test: HF Recognition Quiz







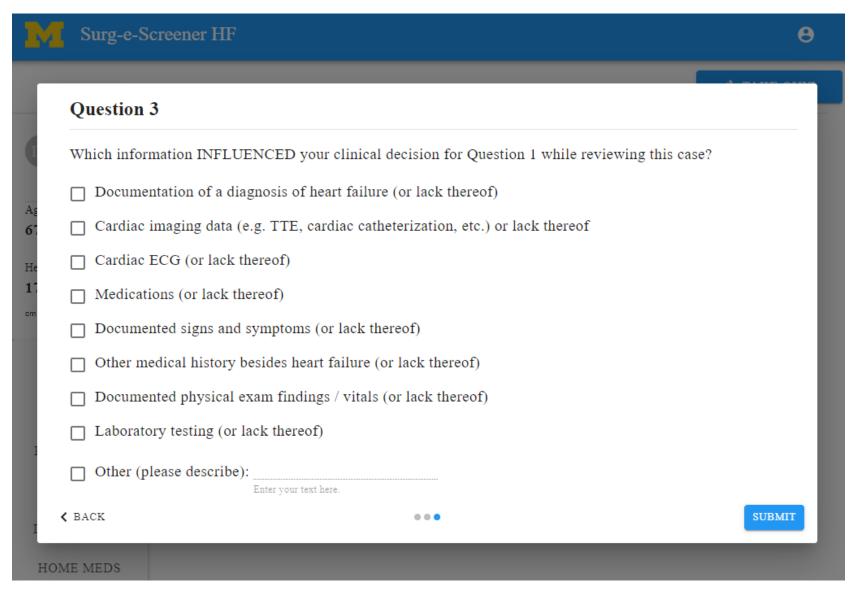
# Step 2. Pre-test: HF Recognition Quiz







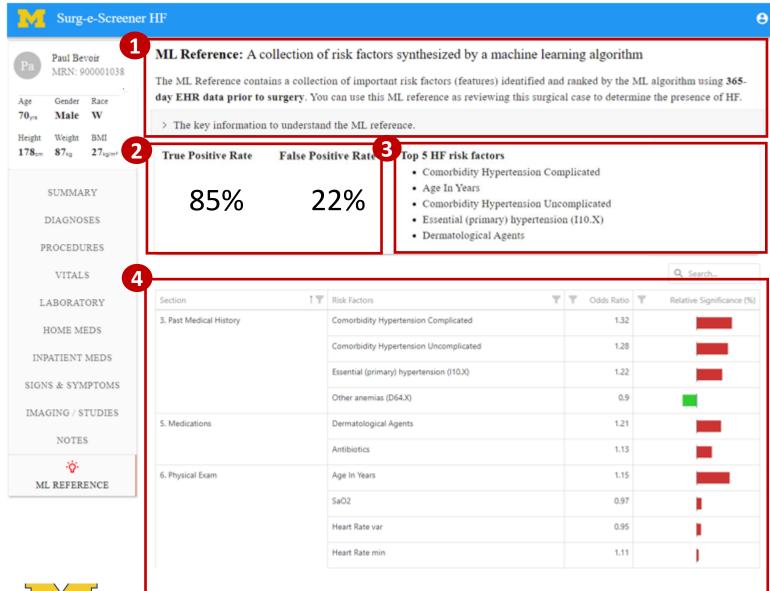
# Step 2. Pre-test: HF Recognition Quiz







# Step 3. Machine Learning (ML) Reference



#### **Key components**

- 1. The ML Reference description
- 2. TPR and FPR
- 3. Top 5 HF risk factors
- 4. Risk factors calculated and ranked by ML algorithms



# Using the Machine Learning (ML) Reference

### **Key HF features are identified and ranked in each category**

#### **Odds Ratio**

• **OR = 1:** No Effect

OR > 1: Higher Odds to HF

OR < 1: Lower Odds to HF</li>

#### **Risk Factor Relative Significance (%)**

- Positive assoc. with HF in red, meaning a patient is more likely to have chronic HF
- Negative assoc. with HF in green, meaning a patient is less likely to have chronic HF

Category	Feature Name	Odds Ratio	Relative Significance (%)
	Acute Pain*	1.12	0.76
	angina*	1.21	1.30
	associated symptoms*	1.15	0.95
	back pain*	1.13	0.85
	calf pain*	0.91	-0.64
	cough*	0.94	-0.40
	Dizziness*	1.05	0.34
	dry skin*	1.18	1.17
	dyspnea on exertion*	1.21	1.33
Signs and Sumptoms	eye pain*	1.20	1.27
Signs and Symptoms	fatigue*	1.23	1.44
	Heartburn*	1.17	1.09
	nervousness*	0.88	-0.89
	photophobia*	1.16	1.04
	rashes*	1.16	1.03
	rest pain*	0.92	-0.56
	stridor*	1.16	1.04
	tarry stools*	0.90	-0.75
	urinary symptoms*	1.18	1.12
	wheezing*	1.43	2.44
Past HF History	Cardiomyopathy	1.45	2.55
	Heart Failure	1.85	4.25
	Disorders of fluid, electrolyte, and acid-base balan	1.14	0.91
Past Medical History	Other and unspecified anemias (285.X)	1.11	0.69
	Cardiac dysrhythmias (427.X)	1.26	1.61
	apnea*	0.93	-0.52
	Comorbidity Chronic Pulmonary Disease	1.13	0.84
	Comorbidity Diabetes Complicated	1.28	1.73
	Comorbidity Hypertension Complicated	1.32	1.92
	Comorbidity Hypertension Uncomplicated	1.28	1.71
	Comorbidity Peripheral Vascular Disorders	1.43	2.45
	Comorbidity Pulmonary Circulation Disorders	1.24	1.49
	Comorbidity Valvular Disease	1.28	1.72
	Coronary Artery Disease	1.40	2.31
	Other anemias (D64.X)	0.90	-0.73
	Other disorders of fluid, electrolyte and acid-base l	0.95	-0.36
	Nicotine dependence (F17.X)	1.12	0.80
	Other anxiety disorders (F41.X)	0.84	-1.16
	Essential (primary) hypertension (I10.X)	1.22	1.38

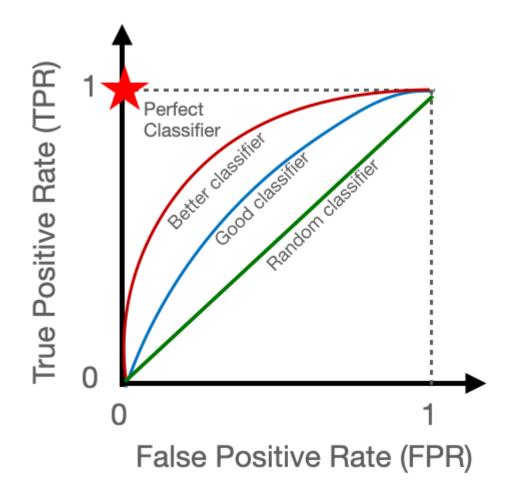
Past Medical History	Gastro-esophageal reflux disease (K21.X)	0.85	-1.10
	Abnormalities of breathing (R06.X)	1.31	1.87
	Abnormal findings on diagnostic imaging of lung (R	0.89	-0.78
	Abnormal results of function studies (R94.X)	0.92	-0.55
	Smoking Classification - Former Smoker	0.91	-0.67
	Personal history of malignant neoplasm (Z85.X)	0.84	-1.22
Past Surgical History	Presence of cardiac and vascular implants and gra	1.77	3.92
Medications	ANTIBIOTICS	1.13	0.83
	ANTIVIRALS	0.88	-0.91
	AUTONOMIC_DRUGS	1.11	0.72
wedications	MUSCLE_RELAXANTS	0.81	-1.44
	SEDATIVE_HYPNOTICS	1.14	0.89
	SKIN_PREPS	1.21	1.33
	Age In Years	1.15	0.99
	Baseline Blood Pressure Diastolic	1.23	1.43
	Baseline Blood Pressure Systolic	0.83	-1.28
	BMI max	1.23	1.41
	BP Systolic mean	0.94	-0.39
	Heart Rate max	1.08	0.54
	Heart Rate min	1.11	0.69
Physical Exam	Heart Rate var	0.95	-0.35
Filysical Exam	Physical Exam Resp Rate	1.10	0.64
	Physical Exam Temperature	0.92	-0.55
	pitting edema*	1.28	1.72
	Race African American	1.16	1.04
	Respiratory Rate max	0.91	-0.64
	SaO2	0.97	-0.20
	SPO2 min	0.84	-1.18
	Temperature max	0.98	-0.11
Test - Lab	Albumin max	0.89	-0.84
	Creatinine var	1.23	1.41
	HbA1c last is below 5.7	0.85	-1.12
	Hematocrit var	0.89	-0.83
	No INR Test	0.80	-1.53
	INR variance	0.92	-0.60
	PlateletCount last is b/w 150-449.99 K/uL	0.90	-0.71
	PlateletCount var	0.72	-2.22
	Potassium last is b/w 3.5-4.99 mmol/L	1.12	0.77
	Troponin var	0.97	-0.23
	WBC var	0.92	-0.61
Test - Image	LVEF min is below 50	2.04	4.92
rest - illiage	No LVEF Test	0.57	-3.84
Test - ECG	Atrioventricular and left bundle-branch block	1.24	1.45





### **ML Quick Review – Evaluation Metrics**

Area Under the Receiver Operating characteristic Curve (AUROC or AUC)



AUROC Values	Test Quality
1.00	Perfect test; 100% accurate
0.90	Better, Excellent
0.80	Good, Acceptable
0.70	Poor
0.50	No better than coin flip

**TPR** = Sensitivity

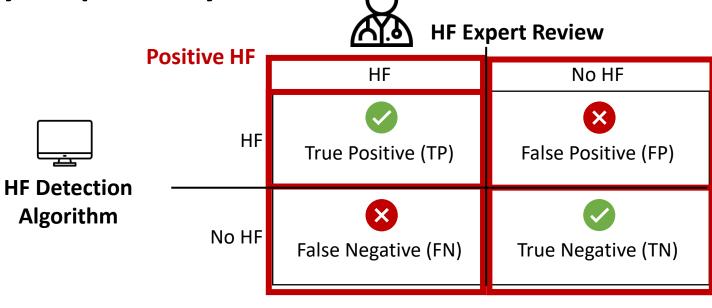
FPR = 1 - Specificity





### **ML Quick Review – Evaluation Metrics**





**TPR** = Sensitivity

**Negative HF** 

 $\mathbf{FPR} = 1 - \text{Specificity}$ 

**Sensitivity** = 
$$\frac{TP}{TP + FN}$$

The ability of HF algorithm to correctly detect patients with HF.

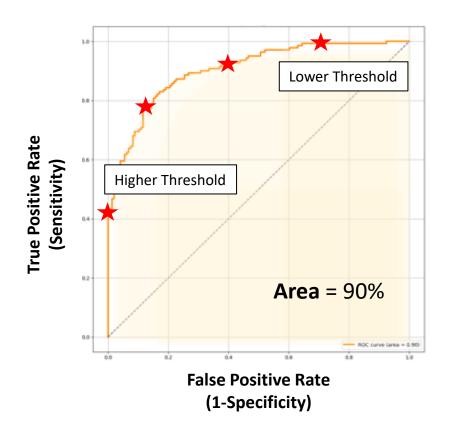
The ability of HF algorithm correctly detect patients without HF.

Specificity =





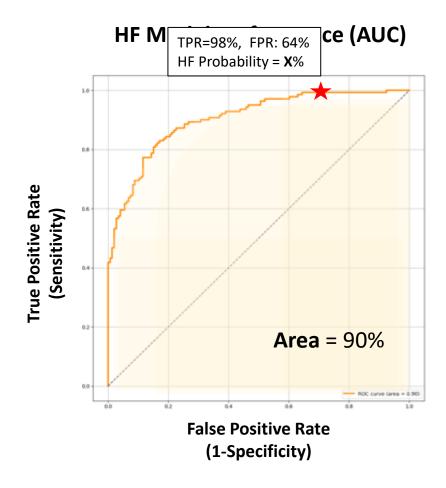
#### **HF Model Performance (AUC)**



- AUC is a curve that plots TPR against FPR at various thresholds.
- Threshold is a cut-off probability to discriminate HF from No HF, a probability between 0 – 1 estimated by the HF model.
- Lower vs. High threshold
- In this study, we show TPR and FPR when <u>threshold</u> is set to the HF probability, estimated by the algorithm.







#### Ex1. TPR=98% and FPR=64%

The model was able to *correctly* detect 98% of patients with HF, but *incorrectly* detect 64% of patients without HF.

#### **Key takeaways**

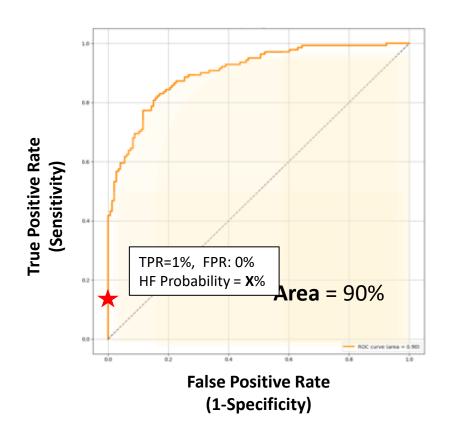
- High true positives (98%), but also high false positives (64%)
- This is due to the **low threshold** (e.g., 12%), HF probability higher than 12% is HF.

Do you think this patient had HF or No HF?





#### **HF Model Performance (AUC)**



#### Ex2. TPR=1% and FPR=0%

The model was able to *correctly* detect 1% of patients with HF, but *incorrectly* detect 0% of patients without HF.

#### **Key takeaways**

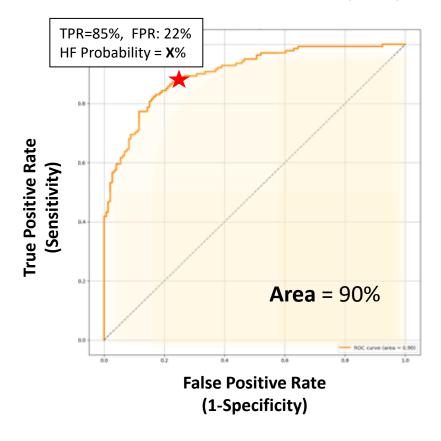
- Low true positives (1%), but also low false positives (0%)
- This is due to the **high threshold** (e.g., 96%), HF probability higher than 96% is HF.

Do you think this patient had HF or No HF?





#### **HF Model Performance (AUC)**



#### Ex3. TPR=85% and FPR=22%

The model was able to *correctly* detect 85% of patients with HF, but *incorrectly* detect 22% of patients without HF.

#### **Key takeaways**

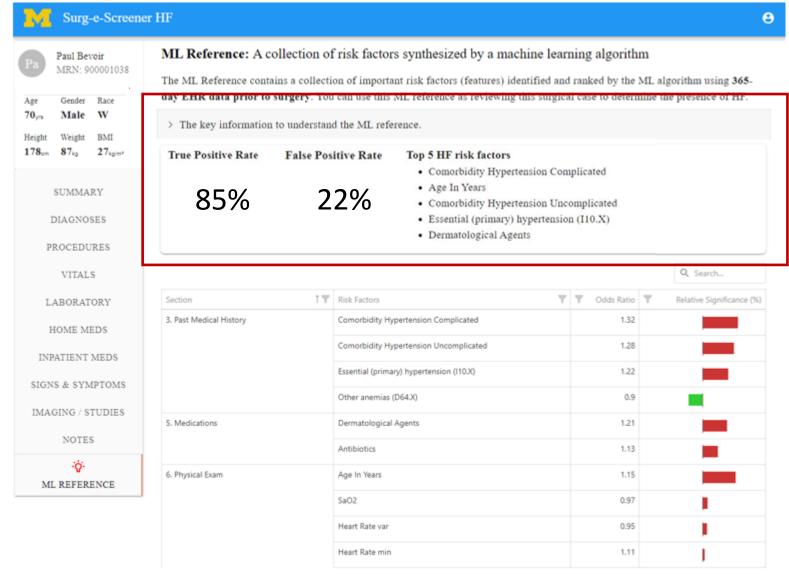
- Relatively high true positives (85%) and relatively low false positives (22%)
- The **threshold** is in gray zone (e.g., 45%) where HF and No HF overlap the most. HF probability higher than 45% is HF.

Do you think this patient had HF or No HF?





# Step 3. Machine Learning (ML) Reference



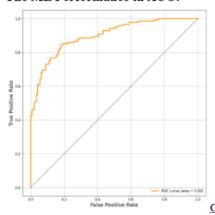




# Step 3. Machine Learning (ML) Reference

v The key information to understand the ML reference.

#### The ML Performance in AUC:



#### Abbreviations:

- LVEF: Left Ventricular Ejection Fraction
- INR: International Normalized Ratio of prothrombin times
- WBC: White Blood Cell
- SaO2: Oxygen saturation of the arterial blood
- · SpO2: Oxygen saturation from a pulse oximeter
- var: variance of the measurements (e.g., Creatinine var)
- min: minimum value (e.g., Heart Rate min)
- max: maximum value (e.g., Heart Rate max)
- last: last (latest) value before surgery (e.g., HbA1c last)

#### Note:

- An asterisk next to the risk factor (e.g., wheezing\*) describes that the risk factor was "mentioned" in clinical notes, which does not indicate either "positive" or "negative". You can confirm positive/negative by searching the risk factor from Signs and Symptoms or Notes.
- Risk factors or entire section(s) will **not be displayed**if they are older than one year or not documented in EHR.

#### **Key components**

- 1. The ML Performance in AUC
- 2. Abbreviations
- 3. Notes about risk factors





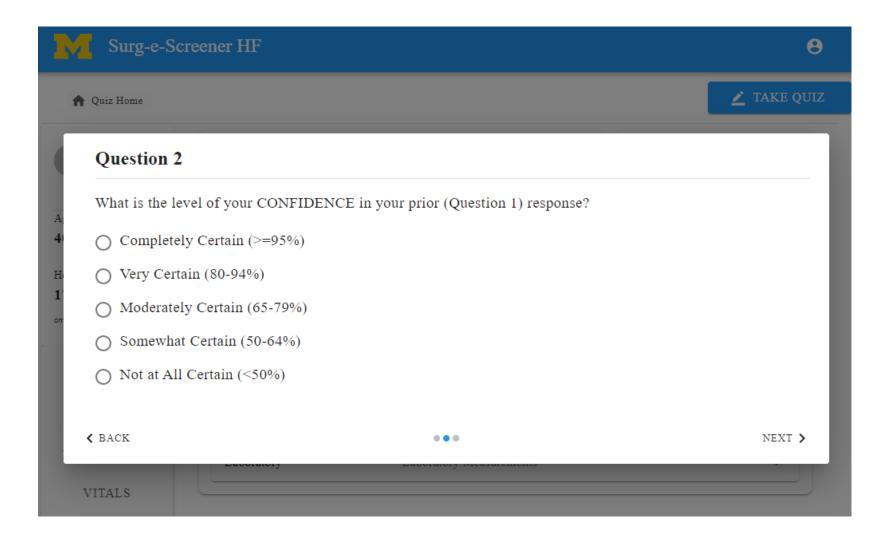
# Step 4. Post-test: HF Recognition Quiz







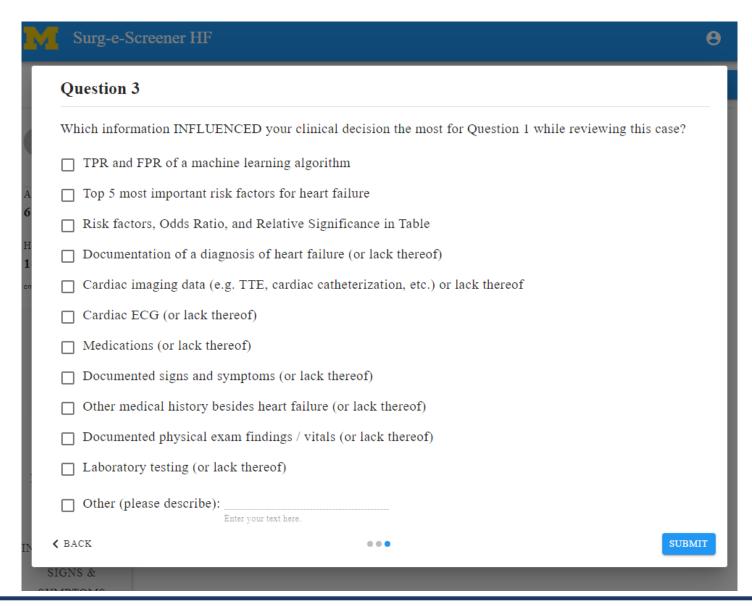
# Step 4. Post-test: HF Recognition Quiz







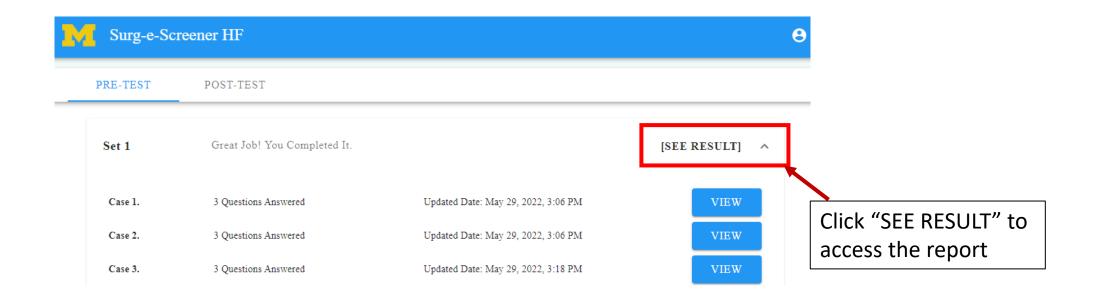
# Step 4. Post-test: HF Recognition Quiz







# Step 5. Result Report Access



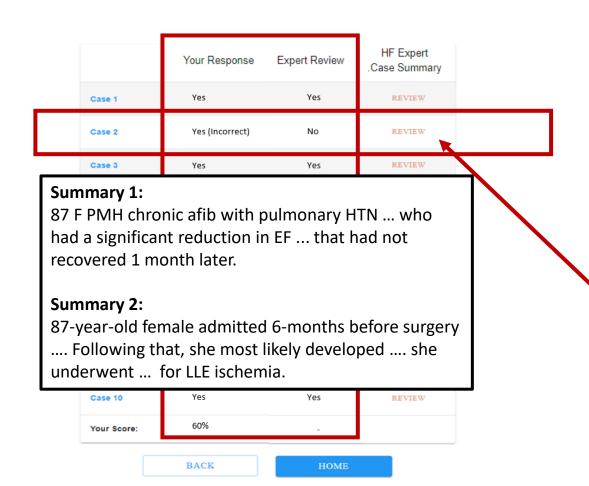
#### **Key components**

 SEE RESULT will be available after reviewing 20 surgical cases, 10 case review results in PRE-TEST and 10 case review results in POST-TEST.





# Step 5. Expert Review Comparison



#### Review your result report

- Compare your HF recognition with the pre-determined answers by HF experts in each case.
- Re-examine cases when you are misaligned with the answers.
- Review a short description of the surgical cases summarized by the HF experts.





# Summary of e-Learning Module

After watching this e-Learning module, you should be able to:

- Understand the process and activities required to complete the HF recognition study
- Interpret evaluation metrics such as AUC, True Positive Rate (Sensitivity), and False Positive Rate (1-Specificity)
- 3. Describe how **TPR and FPR change** as the threshold increases or decreases

The short quiz will ask you mainly about Point 2 and Point 3.





# Demo



# **Final Step**

• No further actions after completing the test cases. We appreciate your participation.

- Reimbursement:
  - Follow-up if any case reviews remain incomplete
  - Pending follow-up / completion of all cases, you will receive reimbursement of \$50 as a check delivered to your mailing address
- Email Hyeon Joo (thejoo@med.umich.edu) if any questions



