

# Building Electronic Tools To Enhance and Reinforce CArdiovascular REcommendations for Heart Failure (BETTER CARE-HF):

A Pragmatic Cluster-Randomized Trial to Improve Heart Failure Care

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NIH Pragmatic Clinical Trials Collaboratory
Grand Rounds
June 27th 2025

#### **Funding and Disclosures**

- No relevant disclosures
- Funding:
  - NIH/NHLBI (K23HL171636-01; 1R01HL155149-01)
  - AHA (23CDA1042602)
  - NYU CTSI Pilot Program (NIH/NCATS UL1TR001445)
  - NYU Chairman's Circle Research Award
  - Prior support from NIH/NHLBI (2T32HL098129-12)
  - We thank Allen Thorpe for funding the NYU Langone Learning Health System program and NYU Langone Health for providing in-kind contributions



### **Educational Objectives/Outline**

- Describe gaps in evidence-based care for patients with heart failure
- Review the BETTER CARE-HF Study for the development of an effective electronic health record tool to improve prescribing
  - Development and qualitative pilot-testing
  - Cluster-randomized trial
  - Follow up studies
  - Current ongoing pragmatic, multi-center trial



### Heart failure is a major public health issue



Over 6 million
Americans



Over \$30 billion in expenditures



A leading cause of hospitalization



High mortality (40-60% at 5 yrs)



## Mineralocorticoid antagonists (MRA) are life-saving, but under-prescribed in HFrEF

- Approximately 65-75% of patients eligible for MRA are not prescribed this life-saving medication.
- Closing this treatment gap could save **OVER** 20,000 lives per year in the United States.

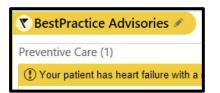


## Electronic health record (EHR) tools are low-cost, scalable, and can improve prescribing

- When developed for other medications, EHR tools have shown modest effectiveness (4.4 percentage points in a metanalysis).
- However, there is wide variability in EHR tool development and design.
- The optimal delivery and timing of EHR tools is **Unknown**.



#### Two EHR tools: Alerts and Messages

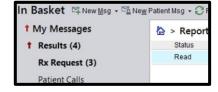


- Single patient at a time
- During clinical encounter
- Could disturb workflow
- Could cause "alert fatigue"









- Multiple patients at once
- Seen between encounters
- Does not disturb workflow
- Could cause "burnout"





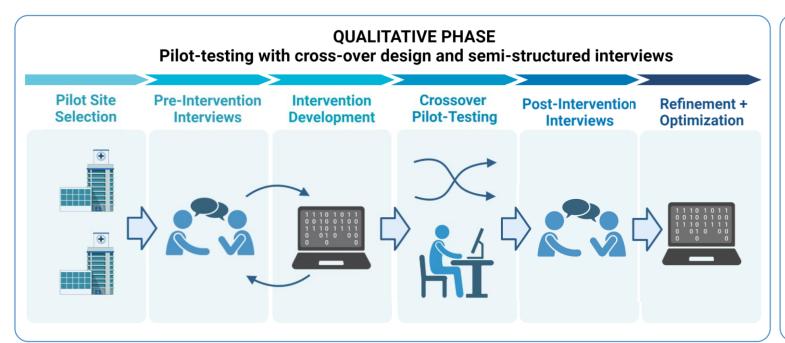


### **BETTER CARE-HF Hypotheses**

- Among patients with HFrEF who are evaluated by a cardiologist in the outpatient setting, an alert or a message will improve prescribing of MRA as compared to usual care.
- An alert will also be more effective than a message at improving prescribing of MRA.



### Mixed Methods Study Design





### Qualitative Phase: Theories and Frameworks Inspiring EHR Tool Development



#### Cognitive Load and Nudge Theories



#### Cognitive Load Theory

- "Split attention effect"
- "Transient information effect"



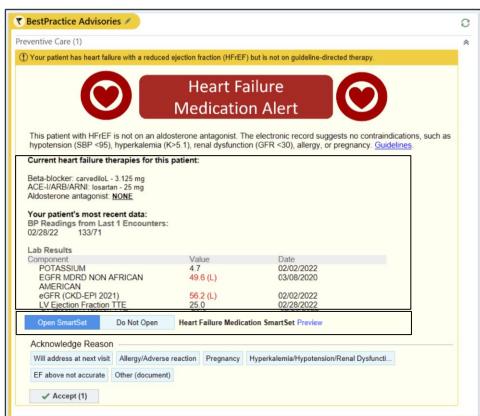
#### **Nudge Theory**

- "Default option"
- "Positioning"
- "Social influence" (illusion of being monitored)



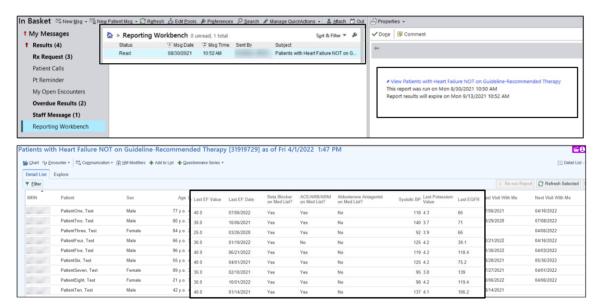
### Applying theories to the EHR alert

- Positioning:
   First alert visualized
   at the top of the chart
- Split attention effect: Multiple components of patient's data all in one place
- Default option:
   Pre-selected order set



### Applying theories to the EHR message

- Transient information effect:
   Information stays in EHR inbox
   until deleted
- Social influence: There is an illusion of being monitored (although we did not actually monitor!)
- Split attention effect:
   Multiple components of patient's data all in one place





## Guiding Framework: The 5 Rights of Clinical Decision Support



### Participants for pilot-testing and interviews

11 cardiologists participated in pilot-testing and qualitative interviews



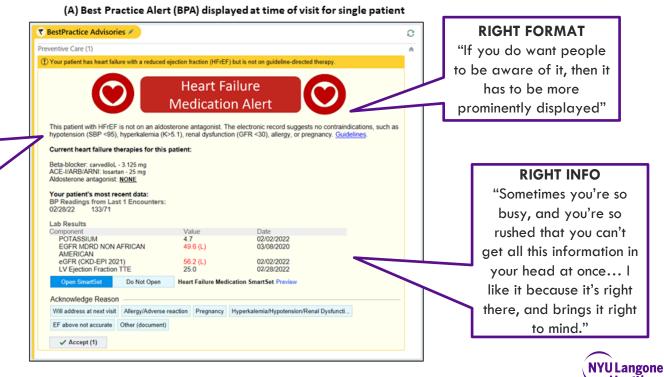
	<u> </u>						
Cha	racteristic	Number (%) or mean (std)					
Age							
•	31-50 years	7 (64)					
•	51-70 years	4 (36)					
Gen	der						
•	Female	3 (27)					
•	Male	8 (73)					
Yea	rs in Practice						
•	0-5	2 (18)					
•	11-20	3 (27)					
•	>20	3 (27)					
Subs	specialty Training						
•	None/General	6 (55)					
•	Electrophysiology	3 (27)					
•	Interventional	1 (9)					
•	Advanced imaging	1 (9)					
Estir	nated percent of patients with HFrEF						
•	0-25%	9 (81)					
•	26-50%	2 (18)					
Clini	c sessions (half-day) per week	6.4 (2.6)					
Patie	ents seen per clinic session	14.5 (2.8)					
Carc	liologist is primary person making decision for MRA prescription	11 (100)					



## Pilot-testing and user interviews lead to intervention improvements

#### **RIGHT INFO**

"I'd like having a little bit more information about the contraindications... the criteria... [for example] not just hyperkalemia but hyperkalemia greater than 5.2"



## Pilot-testing and user interviews lead to intervention improvements

01/14/2021

#### **RIGHT PESON**

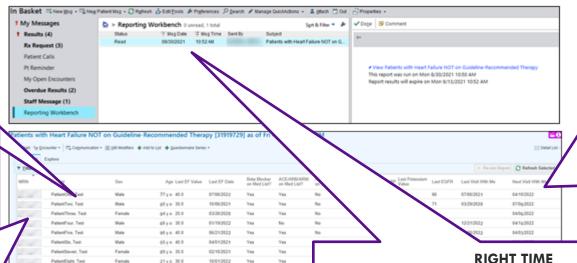
"One of my patients is in the list with TTR amyloid. Amyloid patients should be dropped"

#### **RIGHT FORMAT**

"I basically just ignored them because there were so many patients on the list... the inbasket message was very good, but I think it's probably too much... it was a huge list."

PatientTen, Test

(B) In-Basket Message sent between visits with list of patients



#### **RIGHT INFO**

"I think you need the date when the patient's coming in... somebody's coming on this date, and then I'll look at it, but otherwise, I'm not going to try to figure out when this patient's coming in."

"Less frequent is better. Because if I get it every two weeks, I'm going to get enough where I just start ignoring it."



## The importance of guiding theories, frameworks, and pilot-testing

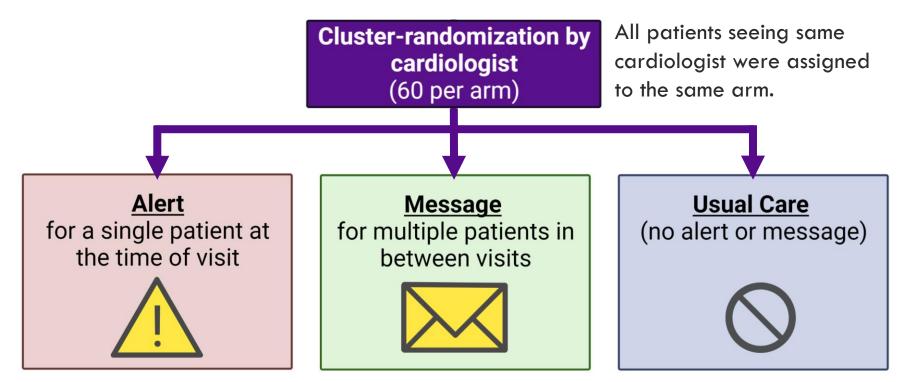
- We made several modifications to the EHR alerts and messages in response to pilot-testing.
- Guiding frameworks and pilot-testing are critical to designing an electronic intervention.



#### Pragmatic, Cluster-Randomized Trial



#### We conducted a trial to test the EHR tools



The study was conducted across over 60 practices



#### Patient Inclusion and Exclusion Criteria

#### **Inclusion Criteria**

- Age ≥ 18
- Seen in outpatient cardiology practice
- Most recent EF ≤ 40%
- Not already prescribed MRA therapy

### Excluded if MRA Contraindicated

- Hypotension (SBP < 90 mm Hg)</li>
- Hyperkalemia (most recent K > 5.0 mmol/L)
- Kidney disease (GFR ≤ 30 mL/min/1.73m²
- Documented MRA allergy or intolerance

#### Additional Exclusion Criteria

- Ventricular assist device
- Cardiac amyloid
- Hospice

Selected Group of Patients Reduced "Alert Fatique"



### **Study Setting**





#### **Outcomes**

#### Primary outcome

New MRA prescription during study period

#### Secondary outcome

Prescription of new BB, ACEI, ARB, or ARNI



#### **Statistical Considerations**

#### Sample size

• In order to detect at least a 10 percentage point difference between each two-way comparison with 80% power and two-tailed alpha = 0.05, with Bonferroni adjustment, we required 1,503 patients, which we estimated would require a 6 month study period.

#### Pre-specified, intention-to-treat analysis

Generalized linear mixed effects model with binomial distribution, log link function, and random intercept by cardiologist to account for clustering at the provider level.



#### Patient Characteristics (N=2,211)

Median age: 73 years Sex: 71% male

Race: 69% White

Ethnicity: 11% Hispanic

Insurance: 69% Medicare

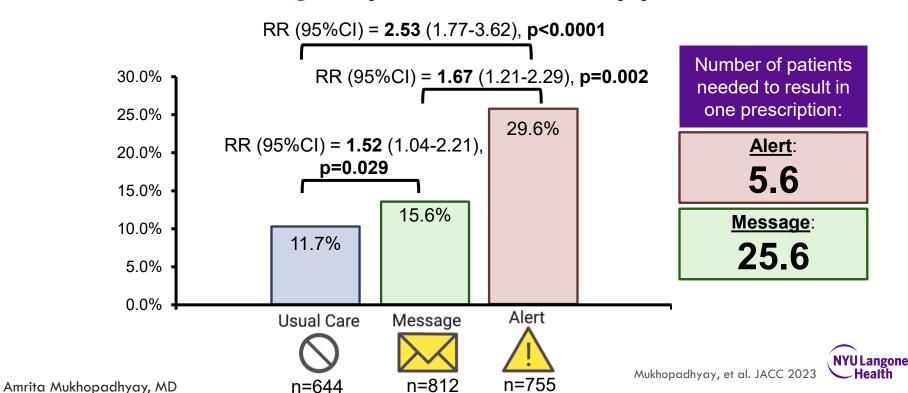
Background BB therapy: 80%

Background ACEI/ARB/ARNI: 74%

Seen by General or HF: 75%

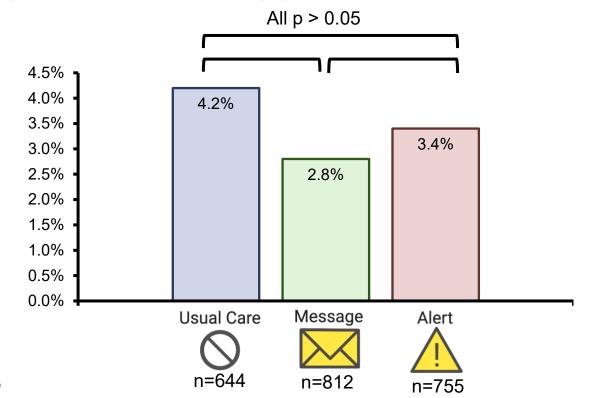


## Primary Outcome Percent of MRA-eligible patients with newly prescribed MRA



## **Secondary Outcome**

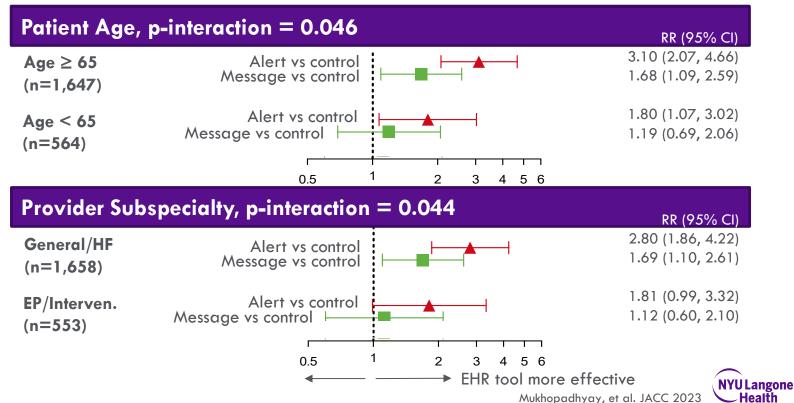
Percent of patients with newly prescribed BB, ACEI, ARB, or ARNI





## Pre-Specified Subgroup Analysis

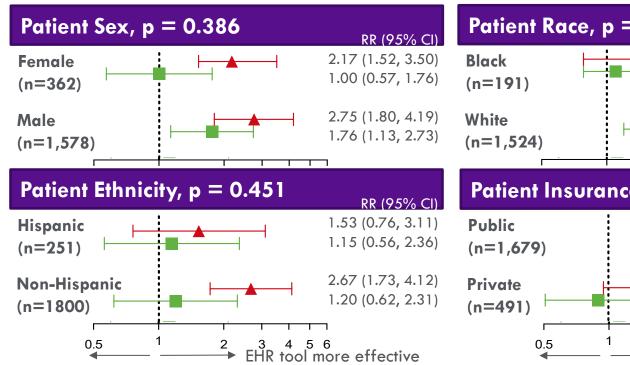
Alert vs Usual Care
Message vs Usual Care

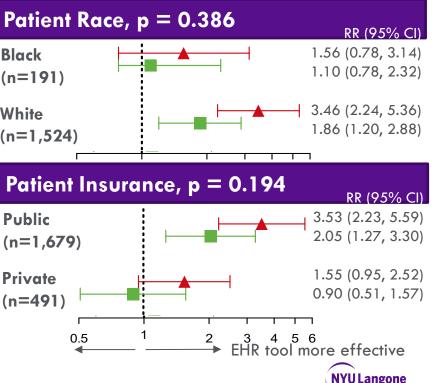


## **Pre-Specified Subgroup Analysis**

Alert vs Usual Care
Message vs Usual Care

Health





Mukhopadhyay, et al. JACC 2023

#### Limitations

- Generalizability
- Targeted to cardiologists
- Specific and selective for MRA
- Sustainability and interaction with other decision support tools is unknown



### Well-designed EHR tools could save lives!

- An automated, EHR-embedded, tailored, and selective alert delivered at the time of visit more than doubled prescribing of MRA as compared to usual care.
- If the amount of improvement in MRA prescription seen in our study was achieved nationally, we could save an estimated >3,500 lives per year (and even more hospitalizations)!
- But is this generalizable across physicians and practice settings?



## Despite EHR tool effectiveness, busy physicians may still be hesitant

- Too many EHR tools can cause fatigue/burnout.
- Concerns about WORK oad and time costs may hinder uptake in busy practice settings.
- Conversely, EHR tools that Save time and reduce cognitive load may be more beneficial in busy practices.







## Does cardiologist workload modify the effectiveness of our EHR tools?

#### Cardiologist Visit Volume Subgroups

- Total number of visits seen during the study period by the cardiologist
  - High-volume = upper tertile
  - Non-high volume = lower two tertiles

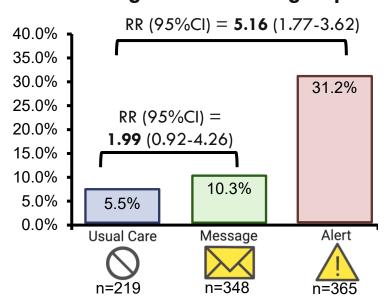
#### Sensitivity Analysis

Average number of visits seen per half-day session by the cardiologist

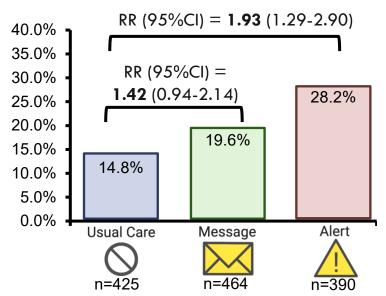


## Primary Outcome Stratified by Visit Volume Percent of MRA-eligible patients with newly prescribed MRA

#### High visit volume group



#### Non-high visit volume group





#### Subgroup Analysis for Primary Outcome Percent of MRA-eligible patients with newly prescribed MRA

Subgroup	RR (95% CI)	p-interaction
Alert vs Usual Care - High-volume group - Non-high volume group	5.16 (2.57, 10.38) 1.93 (1.29, 2.90)	0.02
Message vs Usual Care - High-volume group - Non-high volume group	1.99 (0.92, 4.26) 1.42 (0.94, 2.14)	0.46
	0.5	→ EHR tool more effective



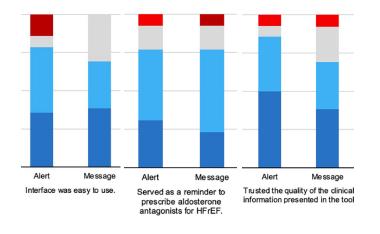
## Sensitivity Analysis for Primary Outcome (Visit volume defined as average # visits per half-day session)

Subgroup	RR (95% CI)									p-interaction
Alert vs Usual Care										
- High-volume group	4.54 (2.34, 8.79)			F-1		_				0.04
- Non-high volume group	1.99 (1.33, 2.97)		١	•	—				·	
Message vs Usual Care										
- High-volume group	1.61 (0.80, 3.25)		<b>⊢</b>	•						0.99
- Non-high volume group	1.60 (1.06, 2.43)		<b>⊢</b>	-						
3 3 1	, , , , ,		i							_
		0.5	1	2	3	4	5	6	7 8 9 1	0
		•		<b>→</b> E	HR too	l moi	re e	ffe	ctive	

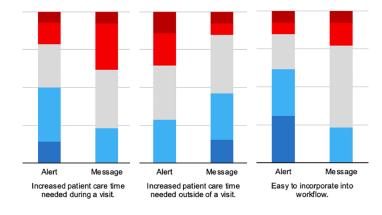


## Post-Trial Survey (N= 27 Cardiologists)

 Overall, cardiologist perceptions were favorable towards both EHR tools.



 With some notable differences when asked about workflow.







#### Will our alert work at other institutions?

 We are conducting multi-center trial to assess the effectiveness of the alert across 3 high-volume health systems around the country

Just because it worked here...



May or may not mean it will work here...



We are looking for more institutions to join us! (please reach out if interested)



#### Current Progress, Resources, and Support

With Epic support, we have built tools for standard alert implementation across sites

Our team has also built tools for standard, de-identified data extraction across sites

NYU Grossman School of Medicine is currently serving as the single IRB for this study

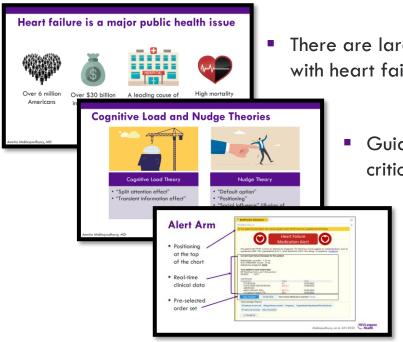
Our team provides technical support, and will conduct all statistical analysis for this work

Currently, one site is live and two others will go live in the coming weeks

We want to partner with you! Amrita.Mukhopadhyay@nyulangone.org



### **Learning Points**



 There are large gaps in evidence-based care for patients with heart failure.

Guiding theories, frameworks, and pilot-testing are critical for successful development of EHR tools.

 Well designed EHR tools can improve prescribing of life-saving therapies.



#### Thank you!

#### Co-Investigators:

Dr. Saul Blecker (Primary Mentor, Population Health)

Dr. Stuart Katz (Cardiology, Heart Failure)

Dr. Harmony Reynolds (Cardiology)

Dr. Leora Horwitz (Population Health)

#### Organizational Leadership:

Dr. Lawrence Phillips (Outpatient Cardiology)

Dr. Arielle Nagler (Outpatient System Integration)

Dr. Eduardo Iturrate (Research Informatics)

Dr. Glenn Fishman (Chief of Cardiology)

Dr. Paul Testa (Chief Information Officer)

Dr. Judith Hochman (NYU CTSI)

#### MCIT Leadership and Developers:

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## We thank NYULH administrative leadership, physicians, and patients

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Administrative Support:

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Cassidy Fitchett

Zaina Laaroussi-Tribek

#### Funding Sources for BETTER CARE-HF:

NYU CTSI Pilot Program (NIH/NCATS UL1TR001445)

NIH/NHLBI (2T32HL098129-12)

NYU Langone Rapid RCT Lab/Learning Health System





