

## Learning Objectives

- Introductions
- Mayo Clinic Strategy for Implementing Al in Nursing Practice
- Outcomes from Generative Al Adoption
   Case Study: In Basket Augmented
   Reply Technology
- The Future of AI for Nursing at Mayo Clinic Case Study: Nursing Virtual Assistant
- Summary & Close

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### **Our Team**



Cheristi Cognetta-Rieke, DNP, RN

Chief Nursing Officer – Enterprise Transformation

Dr. Cognetta-Rieke is an Assistant Professor of Nursing at the Mayo Clinic College of Medicine and Science and serves as the chief nursing officer for Mayo Clinic leading the enterprise Department of Nursing transformation work.



Angie Griffin

Director – Clinical Systems Strategy and Innovation

Angie has a rich background in EHR implementation and optimization, specifically in achieving EHR success in outpatient specialties. Currently, Angie supports Clinical Systems by articulating strategy, championing innovation – including Generative AI, automation, and ambient documentation – and demonstrating value to support Mayo Clinic's mission to Cure, Connect, and Transform.

# Mayo Clinic Multidisciplinary Academic, Practice, Education & Research Center – AZ, FL, MN, WI, IA, International Epic version – November 2023



Rochester Minnesota



Jacksonville Florida



Phoenix & Scottsdale

Arizona



Health System
Minnesota • Iowa • Wisconsin



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Hospitals

**20** 

Surgeries

141,000

Organ transplants

1,858

**Outpatient visits** 

14 mil

Virtual visits

15-20%

Staff

76,000

## Disclosure of relevant financial relationship(s) with industry

Nothing to disclose

## References to off-label usage(s) of pharmaceuticals or instruments

Nothing to disclose

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## Strategy for Implementing AI in Nursing Practice

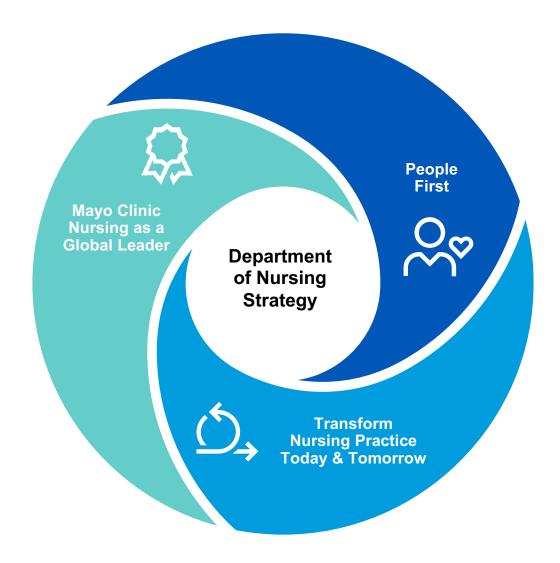
**Advancing Nursing Practice with Al** 

- Front-line nursing engagement in initial feedback/development and rollout strategy
- Nursing leadership engaged in project approval and oversight
- Deliver tools that allow nursing staff to focus on patient care
- Human in the loop: Gen AI can summarize, write drafts, and/or suggest care, but nurses determine what to use and what goes in the record
- Leverage teamwork from multiple teams across Mayo Clinic working to transform clinical practice

## **Mayo Clinic Nursing**

The definitive source of unparalleled nursing knowledge, trusted expertise and innovative patient care.

24,000 Nursing staff



**Generative AI at Mayo Clinic** 

### Vision

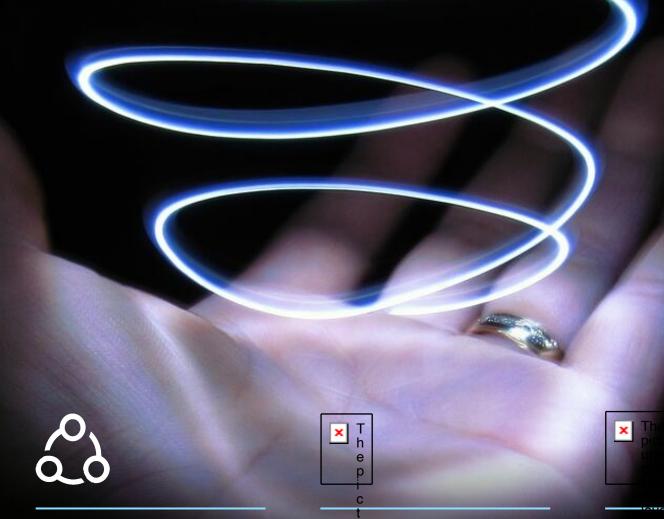
A reimagined healthcare system where clinicians, researchers, educators, administrators and allied health professionals all practice at the top of their license, ensuring the most optimal care for patients.

## **Purpose**

To seek out, develop and integrate into Mayo Clinic the most impactful generative Al solutions to reduce administrative burden, improve the clinical practice and transform healthcare.



# Workstreams Supporting Generative Al





## Supporting Staff-Generated Innovation

Helping progress novel ideas from mayo clinic employees

## Foundation **Model**

'Mayo Al doc' for the world



Implement solutions to generate optimal workflows

#### **Skunkworks**

Leverage infrastructure for LLM development to foster collaboration and innovation

## Plummer Chart (Epic) Subcommittee

Ease, Elevate, Evolve,

Mayo Clinic is exploring generative Al use cases that can help our clinicians and staff. As an early adoption partner with Epic, we will provide ongoing rapid feedback to speed the delivery of use cases from concept to release and maturation.

The goal is to **leverage the technology side of healthcare** in innovative ways that increase
productivity and efficiency, reduce administrative
burden, and allow our clinicians to **focus more on the human side of healthcare: our patients**.

## Learning Objectives

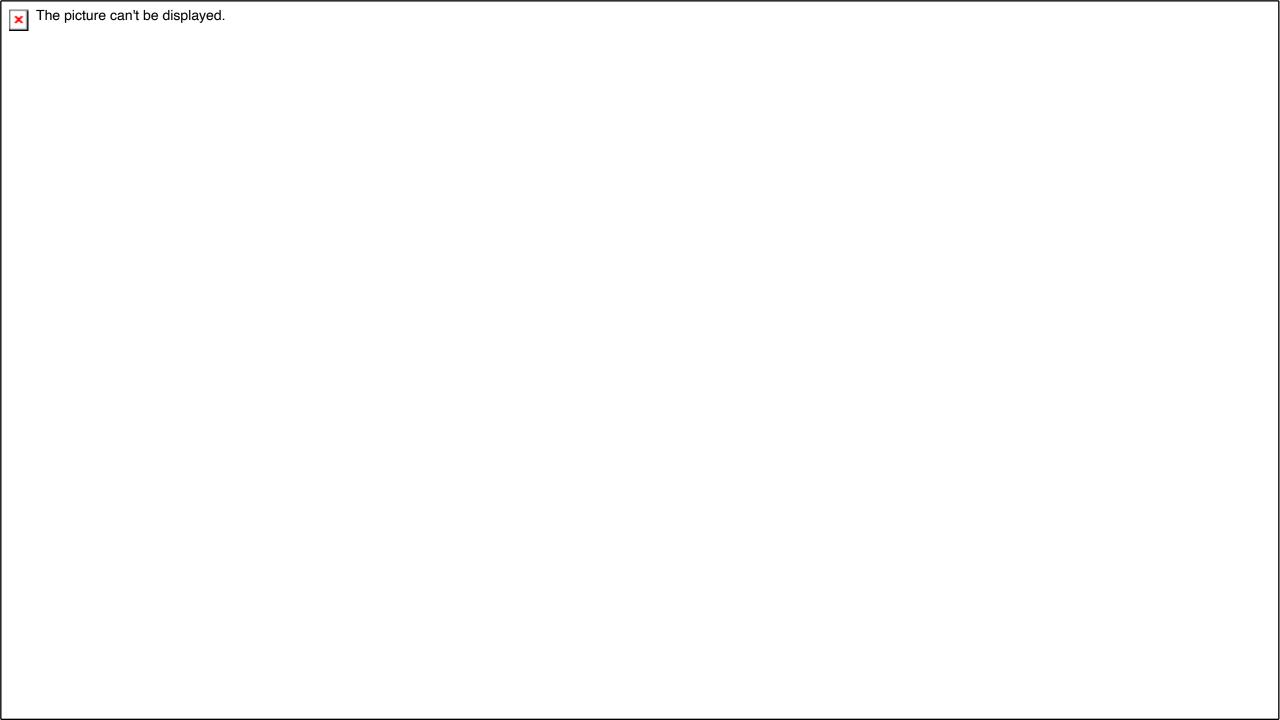
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## **Outcomes from Generative Al Adoption**

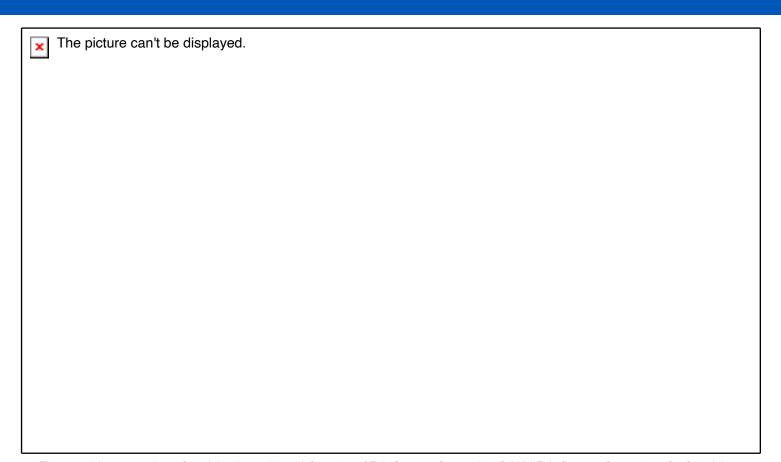
**Advancing Nursing Practice with Al** 

### Case Study: In Basket Augmented Reply Technology

- High-quality, empathetic messages to patients
- Reported reduced cognitive burden & time savings
- Metrics
  - Adoption
  - Appropriate Use
  - ROI



## In Basket Augmented Response Technology



What is it? Generative Al drafts replies to patient portal messages, which staff review and edit, if necessary, before sending.

## **Background**

**Pilot Details:** The feature was rolled out to 30 nurses and 10 providers July – Dec 2023

#### **Pilot Observations**

- No observed hallucinations
- More valuable for nursing than providers as the prompt explicitly avoids medical decision making
- Nursing feedback shows value in the tool; drafts prevent "writer's block" and improve empathy
- Overall, high draft usage (30-50% of messages started from draft); utility varies by individual and specialty; need more data to fully understand this

#### **Pilot Outcomes**

Limited timing studies (n=25) indicate that on average, nursing users spent 30 seconds less per message when using the Al-generated draft, showing potential for time savings with a larger roll-out

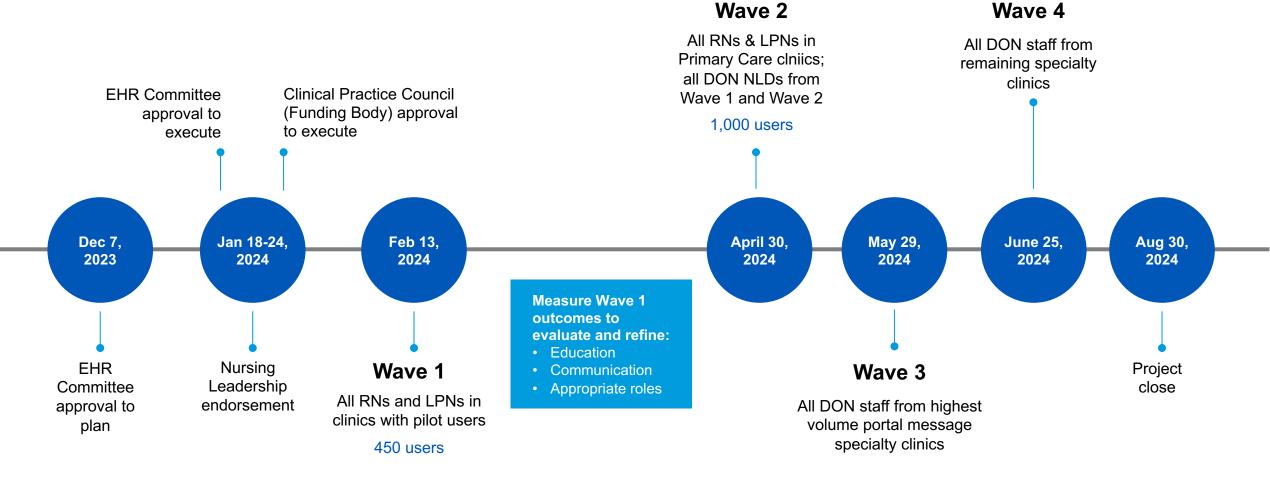
#### **Rollout Project**

Approval to Execute secured on Feb 1

### **Rollout Plan**

- Focus on nurses (RN, LPN); limited value for providers
- Prioritize specialties with the highest volumes of Patient Advice Portal Messages
  - 1. Primary Care
  - Med Specialities, OBGYN, Oncology
  - 3. Remaining specialties and Peds
- Following Wave 1, expanded to include non-licensed delegates under the Department of Nursing
- Exploring expansion to administrative non-licensed delegate (administrative assistants, scheduling staff)

## **Project Timeline**



## **Training Strategy**

#### Wave 1

**Materials:** Providing online training modules (4 mins) and Quick Reference Guides managers to distribute to their staff

**Strategy:** Distribute to managers, managers share with staff

#### **Outcome:**

Wave 1 – Week 1 Opt-In: 53%

Wave 1 – Week 1 Draft Usage: 40%

#### Wave 2

Materials: No change

**Strategy:** Offer "Go-Live Open House" sessions for Nurse Mangers, Supervisors, and NES

Ask nurse leaders to share training materials with staff

#### **Outcome:**

Wave 2 – Week 1 Opt-In: 53%

Wave 2 – Week 1 Draft Usage: 51%

## Implications for Nursing Practice

#### The generative draft reply:

- Will enhance the nurse's workflow
- Will reduce cognitive burden
- Will provide a starting point for replies to patient messages
- Will allow for easy edits including discarding the generated draft
- Will include empathetic and professional text
- Will NOT replace clinical judgement
- Will NOT change scope of practice
- Will NOT remove nurses' responsibility for message content

## **Outcomes**

## **Current Usage Data**

April 28th – May 5th, 2024

Draft Usage Data			
Viewed drafts used	<b>42%</b> 1,728 of 4,153		

User Adoption Data			
Eligible users who have opted In	<b>60%</b> 914 of 1,536		
Opted in users who used a draft	<b>46%</b> 425 out of 914		

ROI Data				
Total minutes saved	tbd			
Avg % of sent messages written by AI	87%	1		

Appropriate Use Data			
Unchanged drafts	<b>29.6%</b> 506 out of 1,708	1	
Avg % of draft used per message	89%	1	

<sup>\*</sup>In a blinded qualitative assessment, our team has shown that messages started with the AI draft are empathetic, address the patient need, and are appropriate length and reading level.

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## Future Al for Nursing at Mayo Clinic

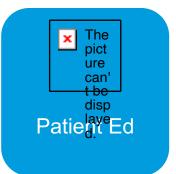
**Advancing Nursing Practice with Al** 

- Nursing Virtual Assistant (case study)
- Epic Draft Shift Summary Note
- Ambient nursing documentation
- Generative AI Patient Summaries

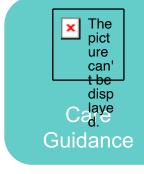
## **Nursing Virtual Assistant**

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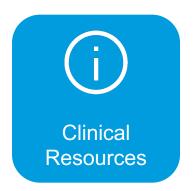


Nursing Virtual Assistant









## **Purpose and Problems**

We aim to develop a **Nursing Virtual Assistant (NVA)** that will **guide individualized nursing care**, including identification of missed care

## Regulatory

Care planning is not personalized which is out of alignment with Joint Commission requirements

#### **Support**

We have many new, float, and traveling nurses who need support with planning care

### **Usability**

Care planning documentation lacks ease of use and direct value to patients and nursing workflow

#### Resources

Care

Nurses may not have time or know how to quickly find relevant care guidance resources

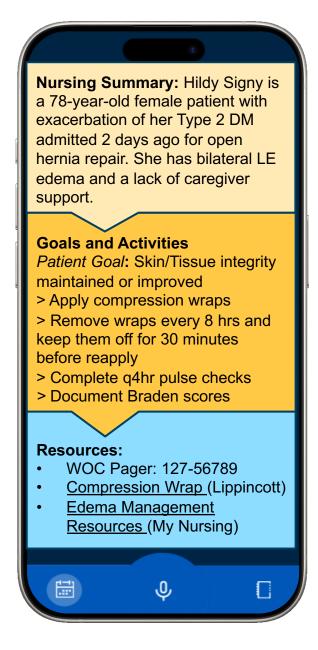
## **Target MVP**

### **Platform: Mayo Clinic Chatbot**

Patient Summary: nursing-centric patient summary designed for start of shift or transfer of care

Goals and Activities: Patient & care goals with supporting interventions. Amplification of "care signals" that may indicate risk, missed care, or priority activities which impact patient outcomes.

**Resources:** Source of recommendations and options for getting more information



### **Workflow**

#### **START**

Nurse is assigned to patients in Epic

Nurse enters patient's chart and selects NVA icon in Storyboard or Rover

Nurse selects patient within NVA

Assigned patients auto-populate to NVA dropdown list

Nurse uses clinical judgement to identify appropriate care inventions

Nurse views unit specific algorithms and resources

Nurse reviews auto-populated output

Patient data and internal resources are consumed and incorporated into output

Playground (PLY) – Hyperspace (Central) – Mayo Clinic – RST RODO

| Policy | Patient Lookup | ED Track Board | Procedural /
| Patient Lookup | ED Track Board | Procedural /
| Applecrisp, Denny-IPR... ×

Summary	Charte		
Summary	Charte		
Overview	Active Country		
Applecrisp	Male, 17 v.o., 1/31/2007	No Travel/Exposure Section	No Travel/Exposure Section
No Travel/Exposure Section	No Travel/Exposu		

Nurse returns to patient's chart and facilities tasks, orders, and documentation as appropriate

Nurse personalizes Care
Plan details and adds
patient specific
interventions

Nurse continues with cares and accesses NVA as needed throughout shift



**END** 

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### 3. The Future of AI for Nursing at Mayo Clinic

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