Toward an ethical framework for Artificial Intelligence in Biomedical and Behavioral Research: *Transparency for Data and Model Reuse* 

## Workshop Goals and Expectations https://www.scgcorp.com/ethicalframework2024/Default

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## Background

- There are multiple examples of **unintended consequences from AI models in healthcare** and biomedicine. The root cause could be data, metadata, models, context of use, ... combinations or all of these... Remediating impacts is hard.
- Increased attention to model testing and assurance. What tests and metrics are important? What potential biases or consequences are relevant?
- Federally funded, open science, is characterized by reuse of data and models.
- NIH researchers using and developing AI want to advance ethical AI but lack guidance.

## Co-chairs and **Breakout Leads**



Ansu Chatterjee, **U** Maryland







Vinnie Liu, Kaiser Permanente



Courtney Lyles, **UC Davis** 



Shazia Siddique, UPenn



Eric Stahlberg, Frederick National Lab



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## Goals



Begin to develop transparency guidelines for NIH awardees using, developing, or contributing to AI



Identify tools and capability gaps.



Look to the future: Identify trends in AI and transparency

## Goals



## Approach to Transparency



## Approach to Transparency



## Goal: Guidelines for NIH Awardees

NIH Awardees could be PI's sharing an AI model, data repository owners, Clinical investigator collecting data...



## Goal: Identify Tools and Capability Gaps

What new capabilities are needed for information to flow?



## Goal: Look to the Future

## What does transparency look like in 3-5 years?









Session 4: Future Trends Readout Plenary

## Session 1: Use Cases

GOAL: Considering your breakout theme, identify relevant stakeholders and their information needs / concerns / decisions



#### <u>Tips:</u>

Consider the full data and AI development cycle Stakeholders could include

- Researchers collecting data
- End users
- Model developers
- Model reusers...

# Session 2: Achieving Transparency and Mapping Capability Gaps

GOAL: Considering your breakout theme, Map stakeholder needs to information sources; and identify capability gaps



#### <u>Tips:</u>

- Consider the full data and AI development cycle
- Articulate a desired state →
   Identify capability gaps

## Session 3: Developing Guidance

## GOAL: Develop draft guidance/best practices for \*actions\* taken by NIH awardees

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#### <u>Tips:</u>

Consider actions of NIH awardees. Examples could be:

- Share data
- Publish an AI model
- Tune a foundation model
- •••

## Plenary Discussion: Synthesizing Guidance

GOAL: Review and Synthesize Draft Guidance from Breakouts in Preparation for an RFI

#### **Questions:**

Have we covered all relevant actions for NIH awardees? Are there other categories of guidance that need to be developed?

For each action, have we captured best practices and considerations? What additional guidance may be needed?

What tools or capabilities are needed to facilitate adoption of these guidelines?

## Session 4: Future Trends

### GOAL: Identify trends in AI transparency



#### <u>Tips:</u>

 What factors are driving change? E.g. policy, awareness, technology, cultural changes?

New Breakout Groups

 What are risks and opportunities for ethical AI?



- Our approach to transparency is based on information and decision needs of stakeholders
- Transparency vs Moral Judgment
  - We want to develop systems of tools and practices that allow people to make informed, responsible, and ethical
    decisions
- Bias vs Causality vs Prejudice
- Breakout themes and instructions are starting points not constraints
  - Think about meta questions
- Think about the full data and model development cycle. Think about the full data ecosystem.
  - There are many different kinds of NIH awardee.
- Consider both health and biomedical research applications.
- One goal of the breakouts is to have all voices heard

## **Breakout Themes**



**1.Synthetic Data:** This breakout focuses on synthetic data —how they are generated, how they might be used, and how they could have both positive and negative impacts on human health. We will discuss specific considerations for the need for and challenges related to synthetic data, including realism, bias, degradation, ethical concerns, and generalizability.

**2.Data Sharing for General Reuse:** Responsible reuse of shared data for AI requires technical, operational, ethical, privacy, and regulatory considerations to assess whether the data are fit for purpose.

**3.Multimodal Data:** This breakout will discuss using multimodal data within artificial intelligence (AI) model development, validation, and translation for clinical implementation (e.g., combining structured data, such as diagnoses, with unstructure d data, such as text or images).

**4.Foundational Models:** This breakout series will explore key concepts behind these models, as well as the implications when creating and using these models in multiple settings involving the clinician, patient, researcher, developer, and community as a whole.

**5.Proxy Variables:** In this breakout session, we will examine the use of proxy variables in algorithms. Proxy variables are confounders and therefore are used (intentionally or unintentionally) in place of another variable that has a true causal relationship with the outcome.

## Breakout Themes and Room Numbers

#### Sessions 1-3



## Breakout Themes and Room Numbers

#### Sessions 4



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