

THE CARE-AI STUDY CREATING ACCOUNTABLE AND RESPONSIBLE ETHICS FOR ARTIFICIAL INTELLIGENCE IN HEALTHCARE

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Equity in Health Systems Lab

OBJECTIVES

- Identify and understand the challenges for professionalism introduced by the use of AI technologies, including algorithmic bias, data privacy, and ethical use of AI-generated insights.
- 2. Apply the CARE-Al principles to scenarios specifically designed for the use of Al in healthcare.



ACKNOWLEDGEMENTS

This work builds off of previous work in professionalism and social media created by:

Rachel Ellaway, University of Calgary Janet Corral, University of Ottawa and University of Calgary; Friend of the Wilson Centre, University of Toronto

They have approved the use of Professionalism Principles for Social Media to be used in this work and related work, and have participated in its creation.

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WHY DO WE NEED A PROFESSIONALISM FRAMEWORK?

Equitable Care

Transparency

Trust

Enduring Behaviors

Informed Decision Making

IDENTIFIED GAP: PROFESSIONALISM FRAMEWORK IS NEEDED



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Article Text

Clinicians and Al use: where is the professional guidance? 8

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HOW SHOULD WE USE AI RESPONSIBLY AND PROFESSIONALLY IN HEALTH CARE, EDUCATION, AND CLINICAL PRACTICE?



DEVELOPING CARE-AI

Build off Social Media Principles (2011) and iteratively develop with inclusive, diverse and expert voices:



CARE-AI FRAMEWORK

Download to Your Device Keep Open To Refer To During Scenarios

OVERVIEW OF CARE-AI PRINCIPLES

- **1.** Responsible Al is a professional duty.
- 2. Responsible Al use should complement, not replace, human judgment.
- **3.** Think before you say, do, or interact with an AI system present.
- **4.** Use AI with honesty and integrity.
- 5. Work within the law.
- **6.** Ensure ethical information stewardship.
- 7. Advocate for mechanisms to reduce bias in Al tools.
- **8.** Digital security awareness is part of your professional responsibility and your professional reputation.
- **9.** Consider the long-term implications of your digital activities.
- **10.** Build and maintain literacy in Al.



KEY CONSIDERATIONS FOR CLINICIANS CONSIDERING ADOPTING AN ML MODEL INTO PRACTICE AT THEIR INSTITUTION



<u>Teaching old tools new tricks—preparing emergency medicine for the impact</u> of machine learning-based risk prediction models

Vinyas Harish, Keerat Grewal, ... Venkatesh Thiruganasambandamoorthy in <u>Canadian Journal of</u> Emergency Medicine

18 March 2023

Consideration	Recomm
1. Is the model needed for the outcome of interest?	Is there a
	Is the pro the use of
2. Was the model robustly developed?	What pre predictor
	Are the re specific a
	Has the n mortality
	Checklist users app
3. Is there sufficient technical capacity available to support potential deployment?	Consult e and how electronic
	If the mo
4. Is model performance robust in the target patient population?	Conduct a these pre
	The lengt need mor
	Use resul updated ^c
	Define a c measurin
5. Does the model appear to be trustworthy (i.e., in terms of bias, interpretability/explainability)?	Conduct a socioecor
	Audit the or sensib
	Audit mo
6. Have all relevant stakeholders considered how the model will impact clinical workflows?	Create an
	Ensure th interface
7. Is there a means of continually monitoring model performance once it has been deployed?	Keep in n
	Create au drops bel

nendation

a perceived need among frontline clinicians to use a prediction model in this context?

oblem sufficiently common, challenging, and does it have enough practice variation to warrant of an ML model?^a

edictors were used to develop the model? Are they clinically plausible or sensible?^b Are these ors difficult to capture? (e.g., family history of sudden death)

reported performance metrics appropriate? Keep in mind the choice of performance metric is task and requires clinical insight

model demonstrated potential to improve meaningful clinical outcomes? (e.g., morbidity, :y, length-of-stay)

ts such as TRIPOD-AI, STARD-AI, DECIDE-AI, and SPIRIT/CONSORT-AI can help prospective endpraise studies of ML models

existing organizational information technology and data governance groups regarding feasibility the model would be integrated within existing information technology systems (e.g., the hic medical record, cloud computing systems, etc.)

odel is commercially available, determine what supports the vendor can provide through the

t a silent trial at the site of potential deployment whereby the model is generating predictions, but redictions are not being used to impact patient care^c

gth of the silent trial is dependent on the prevalence of the condition of interest (e.g., rarer diseases pre time) and duration needed for outcomes to accrue (e.g., 30-day mortality)

Its from silent trial to determine if the model needs to be retrained or if predictors need to be I^c

cut-off for determining poor model performance, this is highly task specific and may involve ng the baseline performance of clinicians in that task^c

a bias assessment in key population subgroups (e.g., by age, sex, race if that data are available, ponomic status, etc.)^c

e predictors used in the model to determine which are most important, are they clinically plausible ble?^{b,c} Are there any signs that may suggest overfitting?

odel false positive and negatives, would clinicians make similar errors on these examples?^c

n interdisciplinary working group to explore how to navigate concerns and potential barriers

hat team members have expertise in overlooked areas, such as human factors and user e/experience design

mind that dataset shifts can cause deterioration in model performance once a model is deployed

utomated alerts and a protocol to revert to baseline operating procedures if model performance elow a clinically significant threshold^c

CARE-AI FACILITATOR AND PARTICIPANT WORKBOOKS

GUIDING WITH CARE-AI

CARE-AI SERIES Creating Accountable & Responsible Ethics for Artificial Intelligence

> A Practical Framework for Facilitating AI Conversations in Health Professions Education

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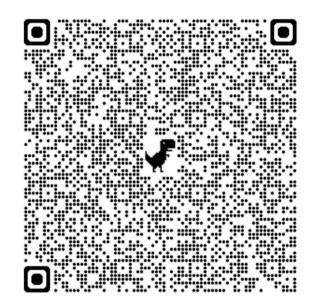






NAVIGATING WITH CARE-AI

A Participant Workbook for **Real-World AI Scenarios in Health Professions Education**



Scan here

GET INVOLVED: CARE-AI STUDY

- CARE-AI Study: Survey Link https://redcap.link/CAREAI1
 - Scan the QR Code to participate in the Delphi-based CARE-AI survey.
 - Your Feedback Matters Survey Deadline: April 20
- REGISTER: AI in Healthcare Think Shop
 - Virtual Event: June 9th, 1–4 pm ET.
 - Don't miss this opportunity to engage in key discussions!

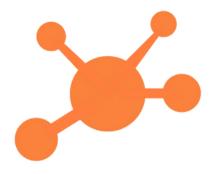




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in Equity in Health Systems Lab















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Let's co-create impactful change

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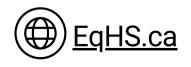






THANK YOU

FOR YOUR PARTICIPATION







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