Expanding Horizons: Developing and Implementing AI Models for Delirium Prediction in Critical Care

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## Disclosures

None



#### Learning objectives

- Understand the principles behind AI model development for delirium prediction.
- Identify challenges like transparency, fairness, and bias in AI.
- Explore the lifecycle of an AI algorithm and its clinical implementation.
- Envision the future of AI in delirium care, including multimodal approaches.
- Apply insights to improve AI development and integration in healthcare.

Detected Imagine an ICU where delirium risk is detected early, interventions are timely and personalized, and outcomes improve for every delirious patient. This future is possible and relies on tools that seamlessly integrate into clinical workflows.

DELIRUM RISK D EANS

RATIEIT CARE

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I'll walk you through how we approached building a delirium prediction model that aligns with this vision, prioritizing usability and interpretation.

### Model Building



#### Delirium prediction to use as a screening tool

- 16546 patients
- Continuous prediction using sliding observation windows

JOURNAL ARTICLE

#### Delirium prediction in the ICU: designing a screening tool for preventive interventions 3

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# Model Building -Outcome

- Typically diagnosed using CAM-ICU
- We analyzed *timing* and *frequency* to plan prediction



# Model Building - Features



L.00	Table 1.     Variables included in the prediction models
).75	Demographic data
).50	Age, gender, height, weight
	Vital signs
).25	Oxygen saturation (SpO <sub>2</sub> ), heart rate (HR), temperature
	Other measurements
0.00	Sofa, sofa without GCS, Ventilation
	Laboratory measurements
-0.25	White blood cell count (WBC), sodium (Na), blood urea nitrogen (BUN), glucose, hemoglobin, platelets, potassium, chloride, bicarbonate, creatinine
	Medications as continuous drips
	Dopamine, epinephrine, norepinephrine, phenylephrine (all calculated as norepinephrine equivalent)

# Model Building -Missingness

Variable	eICU (%)	MIMIC-III (%)
Serum Bicarbonate	2.18	14.58
BUN	0.54	9.30
Chloride	0.53	13.65
Serum creatinine	0.61	14 53
Glucose	0.34	5 71
Heart Bate	2 00	0.05
Homoglahin	0.59	10.21
	0.39	0.12
Oxygen Saturation	2.05	0.13
Platelets	1.65	16.04
Serum Potassium	0.51	13.49
Total norepinephrine dose	0.00	0.00
Serum Sodium	0.51	13.57
SOFA	0.00	0.00
SOFA without GCS	0.00	0.00
Temperature	0.09	4.08
Ventilation	0.00	0.00
WBC	0.65	15.52

## Model Building – Observation and Prediction Windows



#### Model Building - Optimization



	Not optimized	Optimized
Precision	0.38	0.27
Recall	0.86	0.94

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Model Building – Challenges Ahead

**Fairness Metrics** 

Transparency

Explainability  $\rightarrow$  Causality

Real time performance

Deal with discrepancies

#### Implementation: Research to Practice

- Standards, best practices, and operational tools
- Quality assurance
- Transparency and accountability
- Risk management for AI models in health care



Reporting of delirium prediction algorithms



#### Future of AI models for Delirium



"The best way to predict the future is to create it." – Peter Drucker

"Together, we can shape the future of delirium care through innovation, collaboration, and action."

# What do we do next?



#### Think about your role



Standardize AI building and implementation



Advocacy for Al integration, data sharing



Multidisciplinary teams and settings to address challenges



#### Questions?

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