



# Weill Cornell Medicine

## AI System for Disease Detection and Risk Prediction from Longitudinal Multi-Modal Medical Data

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## Background & Unmet Need

- Much of biomedical research and healthcare is based on characterizing and responding to relevant changes in data that have been collected over time, such as imaging, laboratory results, pathology, free-text clinical notes, and wearable signals
- Many artificial intelligence (AI)-based diagnostic tools rely primarily on single-timepoint data—such as a single imaging scan, a one-off lab reading, or both—when assessing a patient's health status
- This can result in suboptimal early detection or less accurate predictions of disease course, since meaningful patterns often emerge over multiple patient visits
- Classical risk models for disease prediction rely on relatively few variables and do not exploit the richness of longitudinal image data or other repeated clinical information
- **Unmet Need:** Methods for integrating multimodal longitudinal data to support clinical decisions

## Technology Overview

- **The Technology:** A time-aware AI architecture which integrates multimodal longitudinal data to provide improved clinical decision support
- The system accommodates virtually any data type including imaging data, lab values, pathology, clinical notes, genomic data, and wearable signals
- The system learns temporal relationships across a patient's entire clinical history to perform a variety of tasks (e.g., disease classification, regression, multi-label tasks, survival/risk estimation)
- **PoC Data:** The inventors developed a longitudinal model for mammogram risk prediction with superior predictive value (AUC 0.92 at 1-year follow up vs. 0.9 and 0.83 for comparator models)
- An additional model for AD progression in normal vs. MCI patients based on multimodal inputs (cognitive assessments, MRI, genetic, and clinical data) with an AUC curve improved by up to 15% over state-of-the-art models

## Inventors:

Mert Sabuncu  
Batuhan Karaman

## Patents:

Provisional Filed

## Publications:

[Karaman et al. Big Data Mining and Analytics](#). 2026.  
[Khawaled. MIDL 2026 Poster](#). 2026. (poster)  
[Kim et al. Proc Natl Acad Sci USA](#). 2025.  
[Nguyen et al. MELBA](#). 2024.  
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## Cornell Reference:

D-11493



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AD: Alzheimer's Disease; AI: Artificial Intelligence; MCI: Mild Cognitive Impairment

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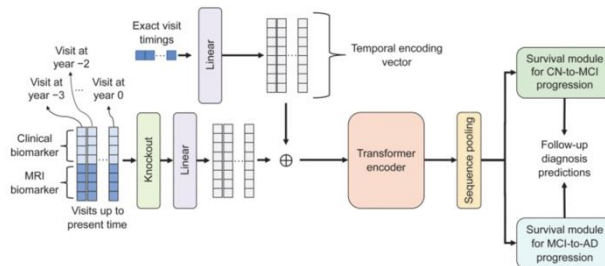
## Technology Applications

- Clinical decision support (e.g., diagnosis, detection, risk stratification, therapy selection, monitoring)
- Population health management (e.g., patient outreach and screening, proactive interventions, healthcare policy and budget planning)
- Research and clinical trials (e.g., longitudinal outcome analysis, biomarker discovery, adaptive trial design)

## Technology Advantages

- The architecture is flexible and can incorporate virtually any data modality or set of data modalities
- The architecture design enables robust handling missing or asynchronous data without severely degrading performance
- The system is generalizable; it scales to large datasets and can be applied to numerous clinical domains (e.g., oncology, neurology, and cardiology)

## Supporting Data / Figures



| Model                          | Longitudinal data modality | Sequential network architecture | Three-year MCI-to-AD AUROC (%) |
|--------------------------------|----------------------------|---------------------------------|--------------------------------|
| Cui et al. <sup>[1]</sup>      | MRI                        | RNN                             | 73.03                          |
| VGG-TSwinformer <sup>[6]</sup> | MRI                        | Transformer                     | 81.53                          |
| LongForMAD <sup>[12]</sup>     | MRI                        | Transformer                     | 78.65                          |
| LongForMAD <sup>[12]</sup>     | MRI + COGN                 | Transformer                     | 92.03                          |
| Li and Liu <sup>[23]</sup>     | MRI                        | —                               | 74.60                          |
| BMNet <sup>[26]</sup>          | MRI                        | —                               | 79.26                          |
| LSM-AD (ours)                  | MRI                        | Transformer                     | 81.33                          |
| LSM-AD (ours)                  | MRI + COGN                 | Transformer                     | 92.54                          |

**Figure 1:** The LSM-AD model integrates clinical and MRI biomarkers over a series of patient visits to predict control vs. MCI-to-AD progression and outperforms prior state-of-the-art longitudinal models.

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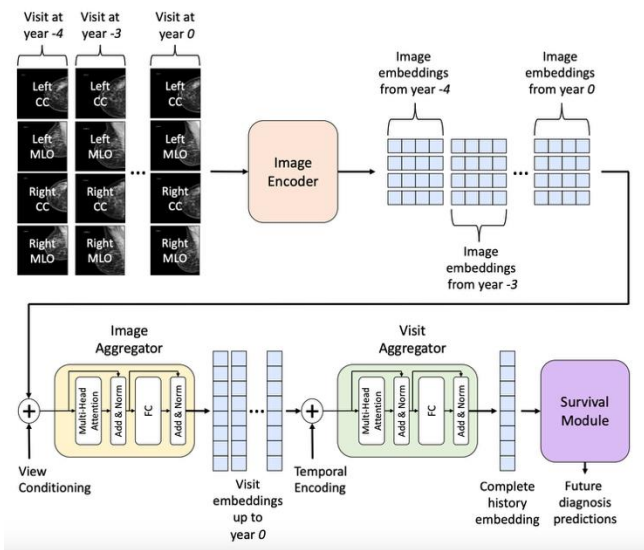


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## Supporting Data / Figures



**Figure 2:** The LoMaR model integrates time-series mammography data over time to predict breast cancer risk and outperforms single visit-state-of-the-art prediction models.

| Model        | History duration | C-index             | Follow-up year ROCAUC |                     |                     |                     |                     |
|--------------|------------------|---------------------|-----------------------|---------------------|---------------------|---------------------|---------------------|
|              |                  |                     | 1-year                | 2-year              | 3-year              | 4-year              | 5-year              |
| Image-Only   | 0                | 0.75<br>(0.73–0.77) | 0.83<br>(0.81–0.86)   | 0.79<br>(0.77–0.81) | 0.75<br>(0.73–0.77) | 0.73<br>(0.71–0.75) | 0.71<br>(0.69–0.73) |
| DL [18]      | 0                | 0.81<br>(0.79–0.82) | 0.90<br>(0.89–0.92)   | 0.86<br>(0.84–0.88) | 0.82<br>(0.80–0.84) | 0.80<br>(0.79–0.82) | 0.78<br>(0.76–0.80) |
| Mirai [20]   | 0                | 0.81<br>(0.78–0.83) | 0.90<br>(0.90–0.94)   | 0.86<br>(0.83–0.86) | 0.82<br>(0.83–0.85) | 0.80<br>(0.81–0.83) | 0.78<br>(0.80–0.82) |
| LoMaR (Ours) | 0                | 0.81<br>(0.78–0.83) | 0.92<br>(0.90–0.94)   | 0.84<br>(0.83–0.86) | 0.84<br>(0.83–0.85) | 0.82<br>(0.81–0.83) | 0.81<br>(0.80–0.82) |
| LoMaR (Ours) | 1*               | 0.82<br>(0.78–0.83) | 0.92<br>(0.90–0.94)   | 0.85<br>(0.83–0.86) | 0.84<br>(0.83–0.85) | 0.82<br>(0.81–0.83) | 0.81<br>(0.80–0.82) |
| LoMaR (Ours) | 2*               | 0.82<br>(0.78–0.83) | 0.92<br>(0.90–0.94)   | 0.84<br>(0.83–0.86) | 0.84<br>(0.83–0.85) | 0.82<br>(0.81–0.83) | 0.83<br>(0.82–0.84) |
| LoMaR (Ours) | 3*               | 0.82<br>(0.78–0.83) | 0.92<br>(0.90–0.94)   | 0.84<br>(0.83–0.86) | 0.84<br>(0.82–0.85) | 0.83<br>(0.82–0.84) | 0.85<br>(0.84–0.86) |
| LoMaR (Ours) | 4*               | 0.82<br>(0.78–0.83) | 0.92<br>(0.90–0.93)   | 0.84<br>(0.83–0.86) | 0.84<br>(0.83–0.86) | 0.85<br>(0.84–0.86) | 0.86<br>(0.85–0.87) |
| LoMaR (Ours) | 4†               | 0.82<br>(0.78–0.83) | 0.92<br>(0.90–0.93)   | 0.84<br>(0.83–0.86) | 0.84<br>(0.83–0.86) | 0.84<br>(0.84–0.86) | 0.84<br>(0.85–0.87) |

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