

# Toward Scientific Foundation Models for Aquatic Ecosystems

Abhilash Neog

PhD Computer Science, Virginia Tech

**H33D: Advancing Water Science Through Artificial Intelligence:  
Lessons, Strategies, and New Frontiers**

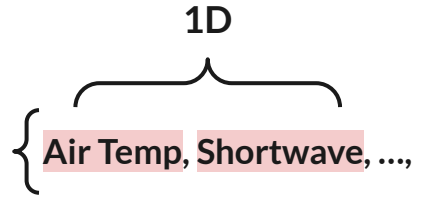
*AGU 2025, New Orleans, LA*

# Motivation

## Aquatic Ecosystem 1

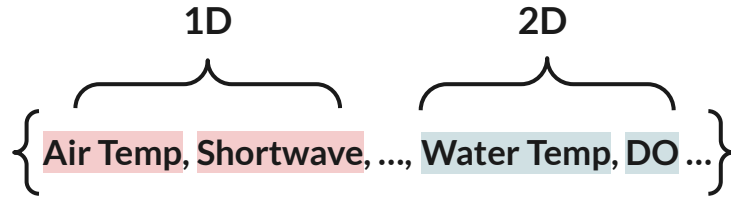
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## Aquatic Ecosystem 1



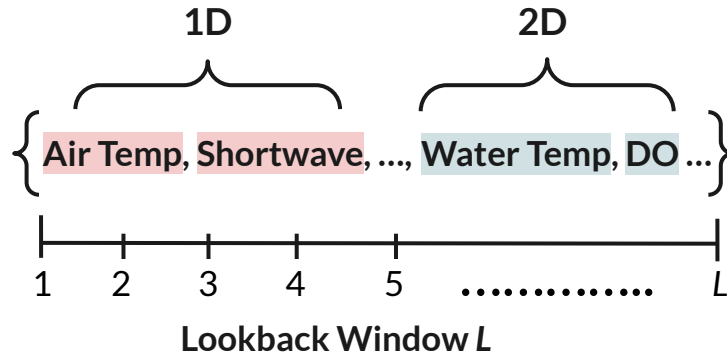
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## Aquatic Ecosystem 1



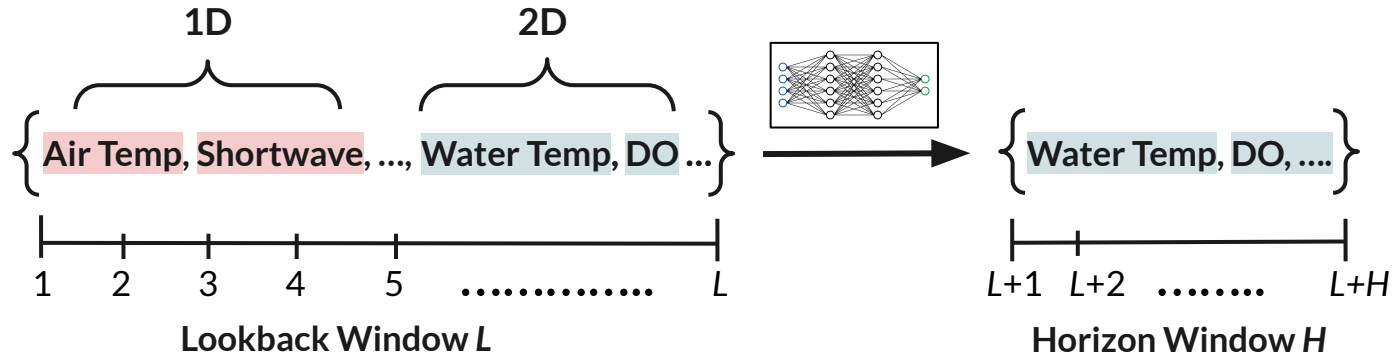
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## Aquatic Ecosystem 1



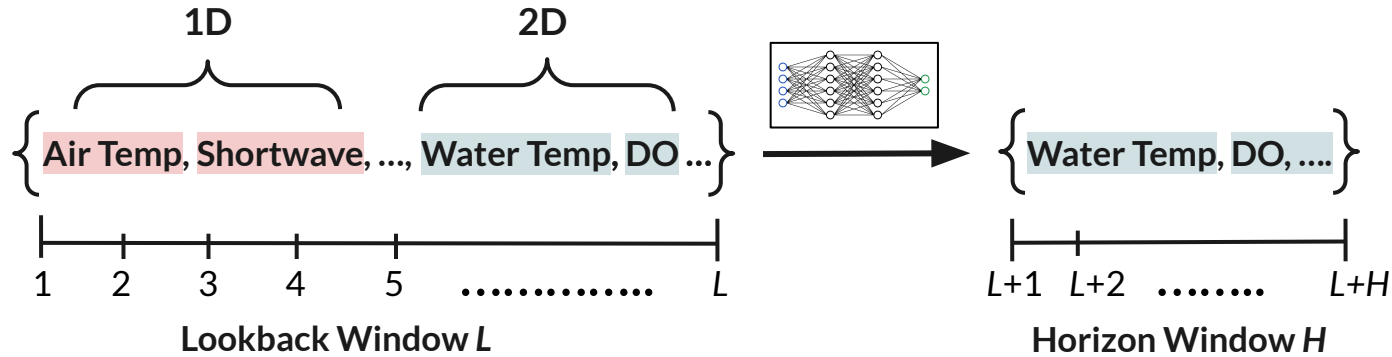
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## Aquatic Ecosystem 1



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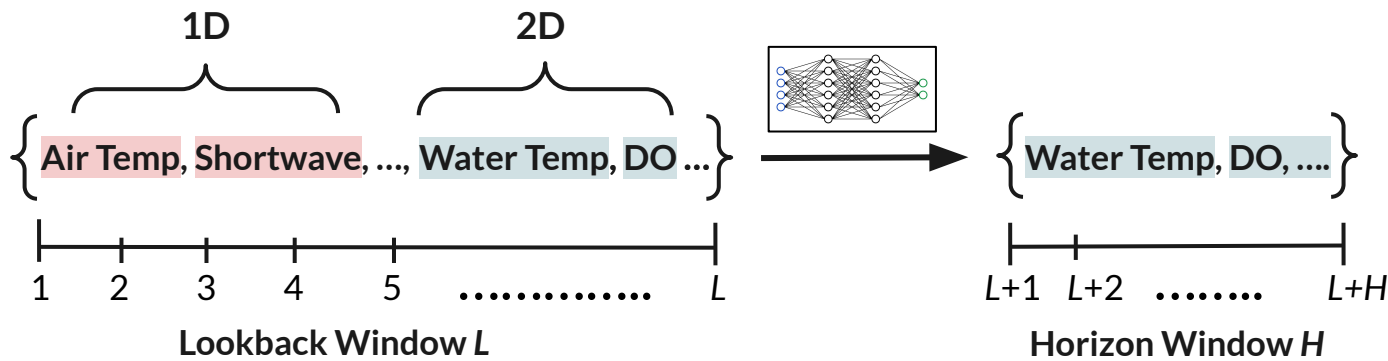
## Aquatic Ecosystem 1



⊗ Different subset of variables available in different ecosystems

# Motivation

## Aquatic Ecosystem 1

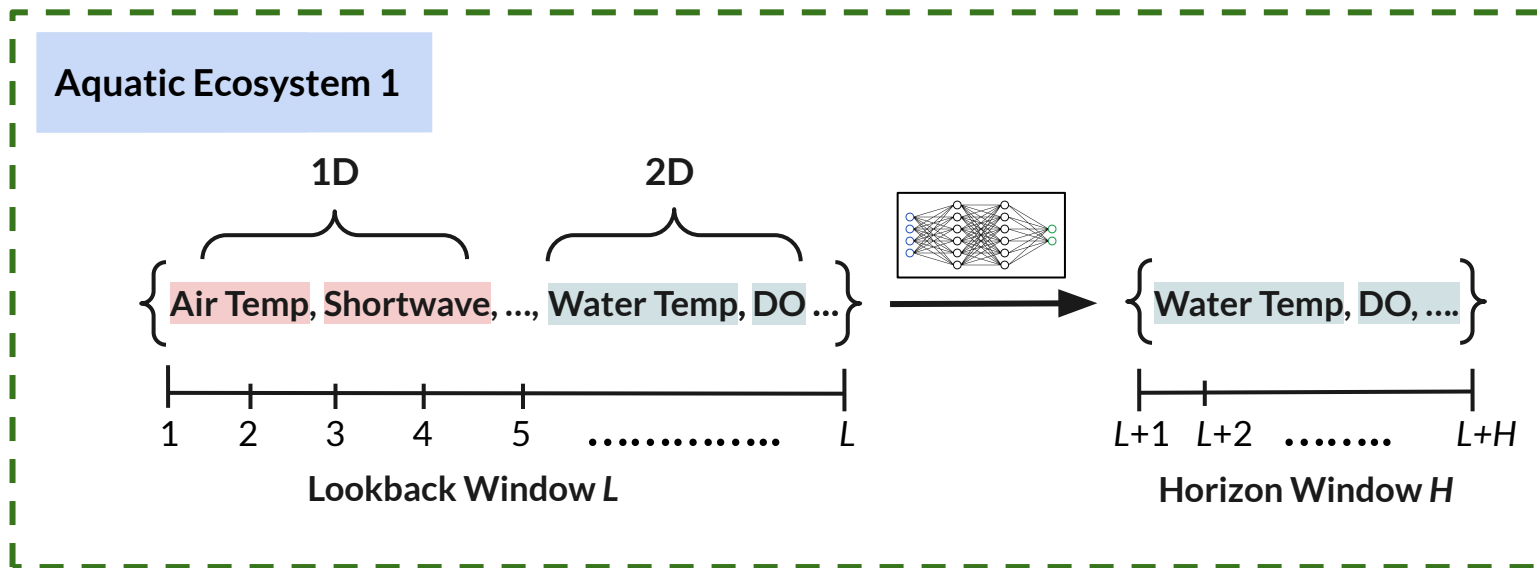


⊗ Different subset of variables available in different ecosystems

⊗ Large amounts of missing data (e.g., *Falling Creeks Reservoir, VA, has 50% missing data, 2017-04 to 2022-10*)



# Motivation



Transfer Learning



**Aquatic Ecosystem 2**  
*Well observed*

⊗ Different subset of variables available in different ecosystems

⊗ Large amounts of missing data (e.g., *Falling Creeks Reservoir, VA, has 70% missing data, 2017-04 to 2022-10*)

# Motivation

Aquatic Ecosystem 1

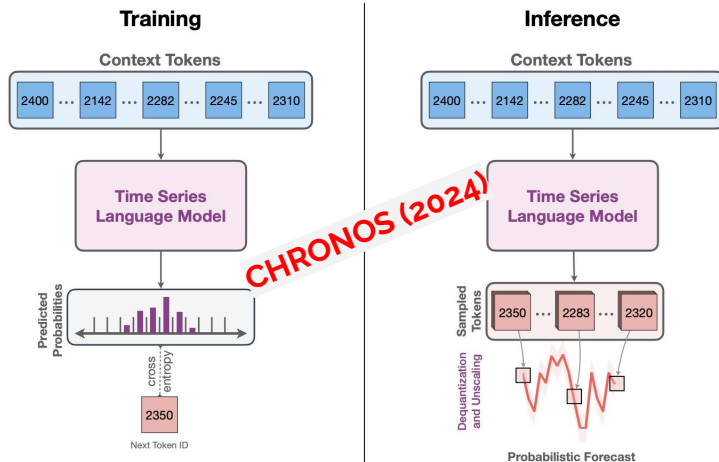
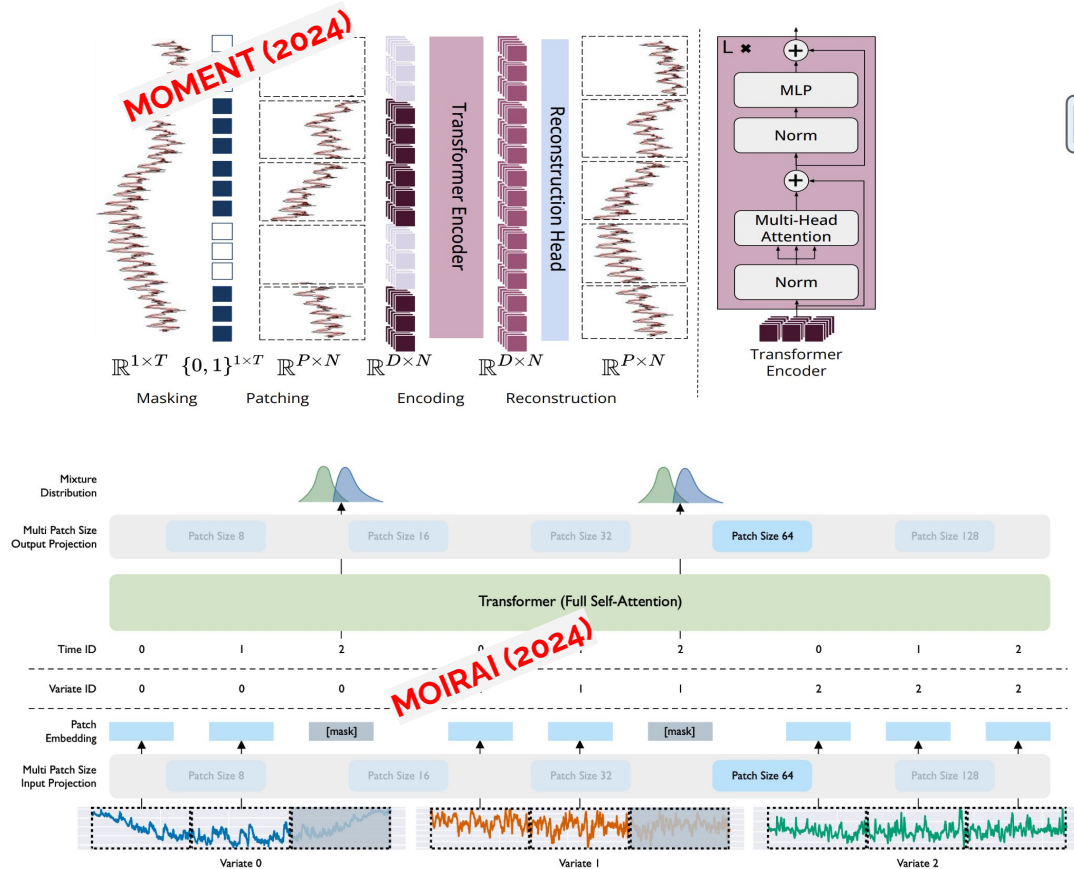
Can we build a Foundation Model that can generalize across different lake ecosystems with different variables and missing values?

⊗ Different subset of variables available in different ecosystems

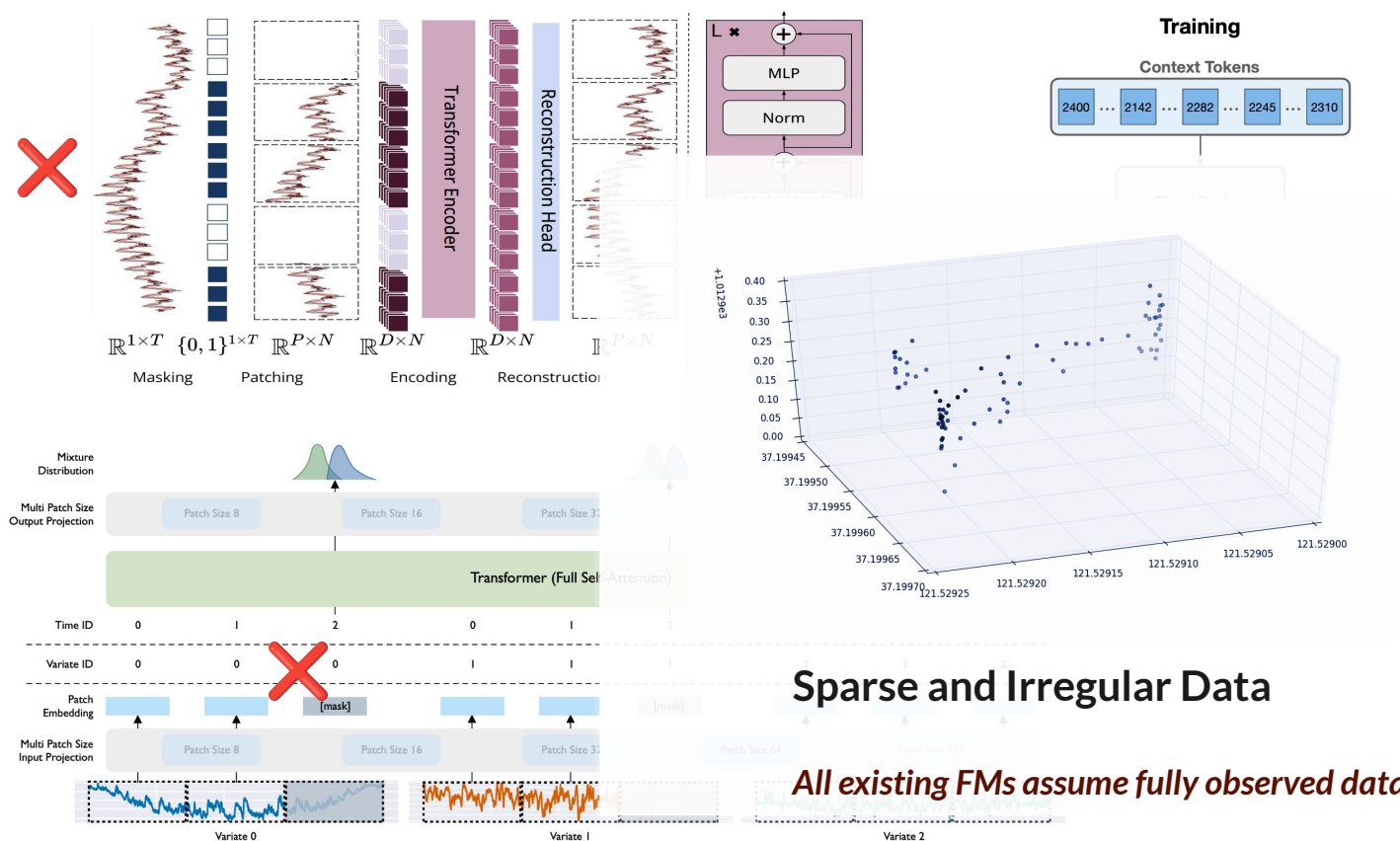
⊗ Large amounts of missing data (e.g., *Falling Creeks Reservoir, VA*, has 70% missing data, 2017-04 to 2022-10)

Aquatic Ecosystem 2  
Well observed

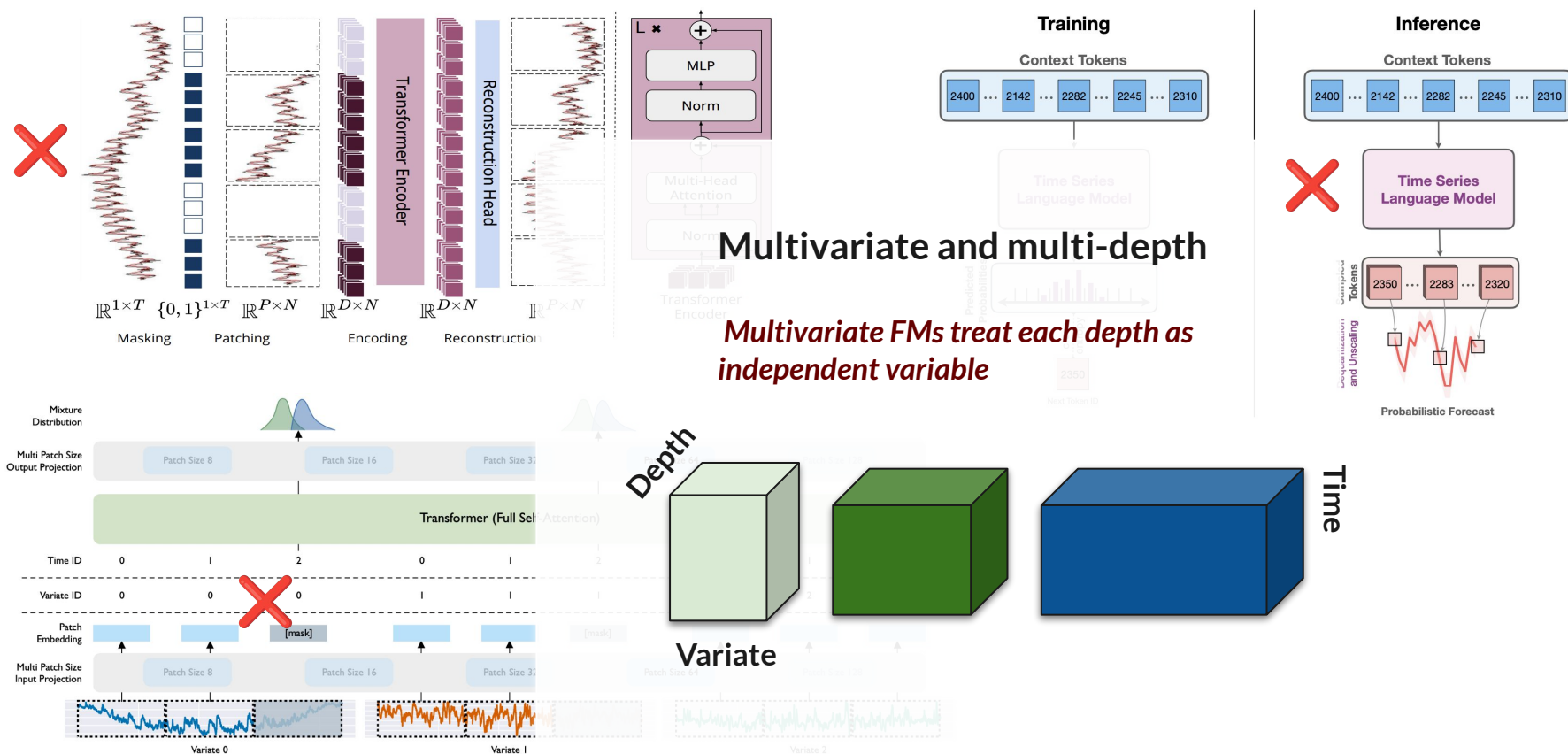
# Existing Foundation Models



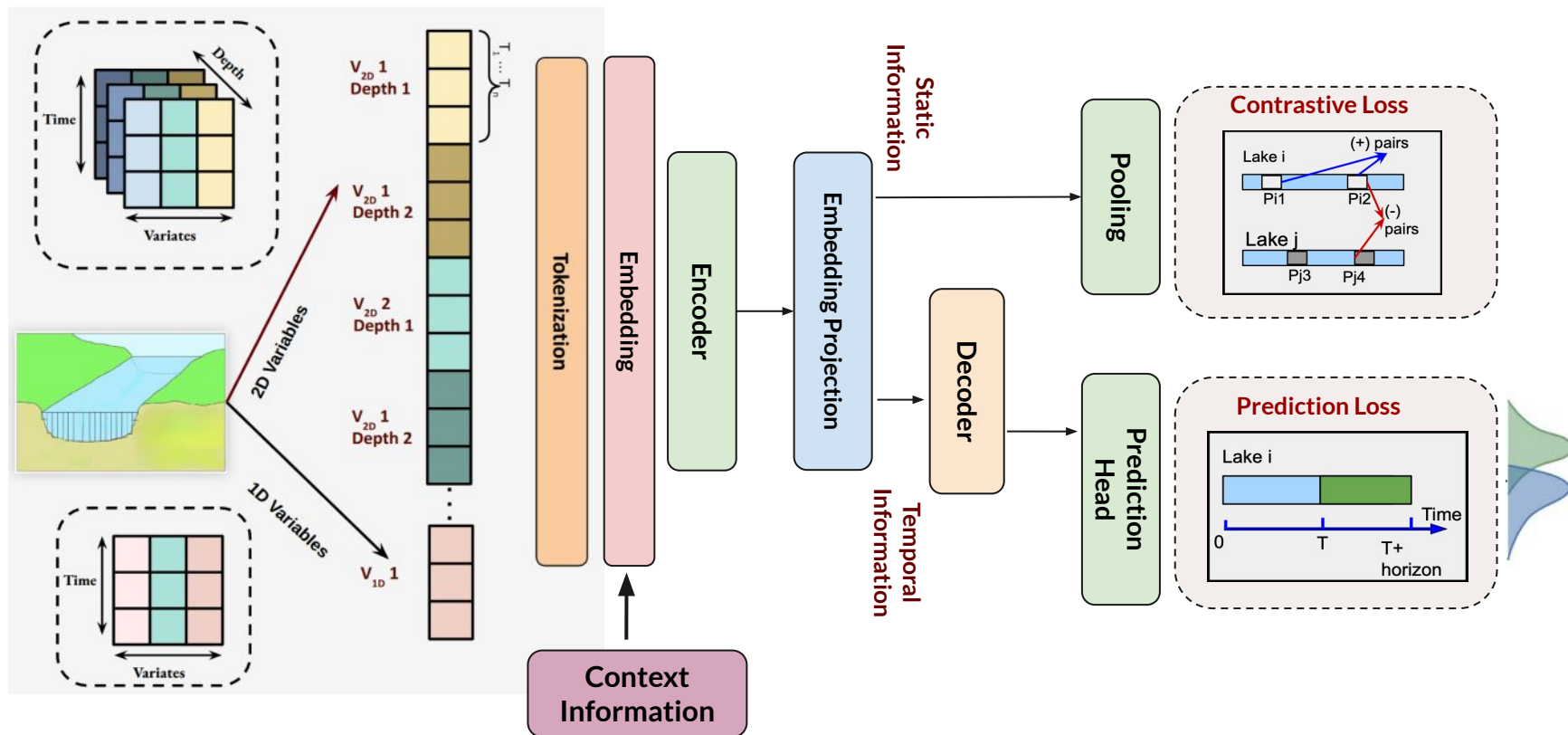
# Challenges with existing Foundation Models



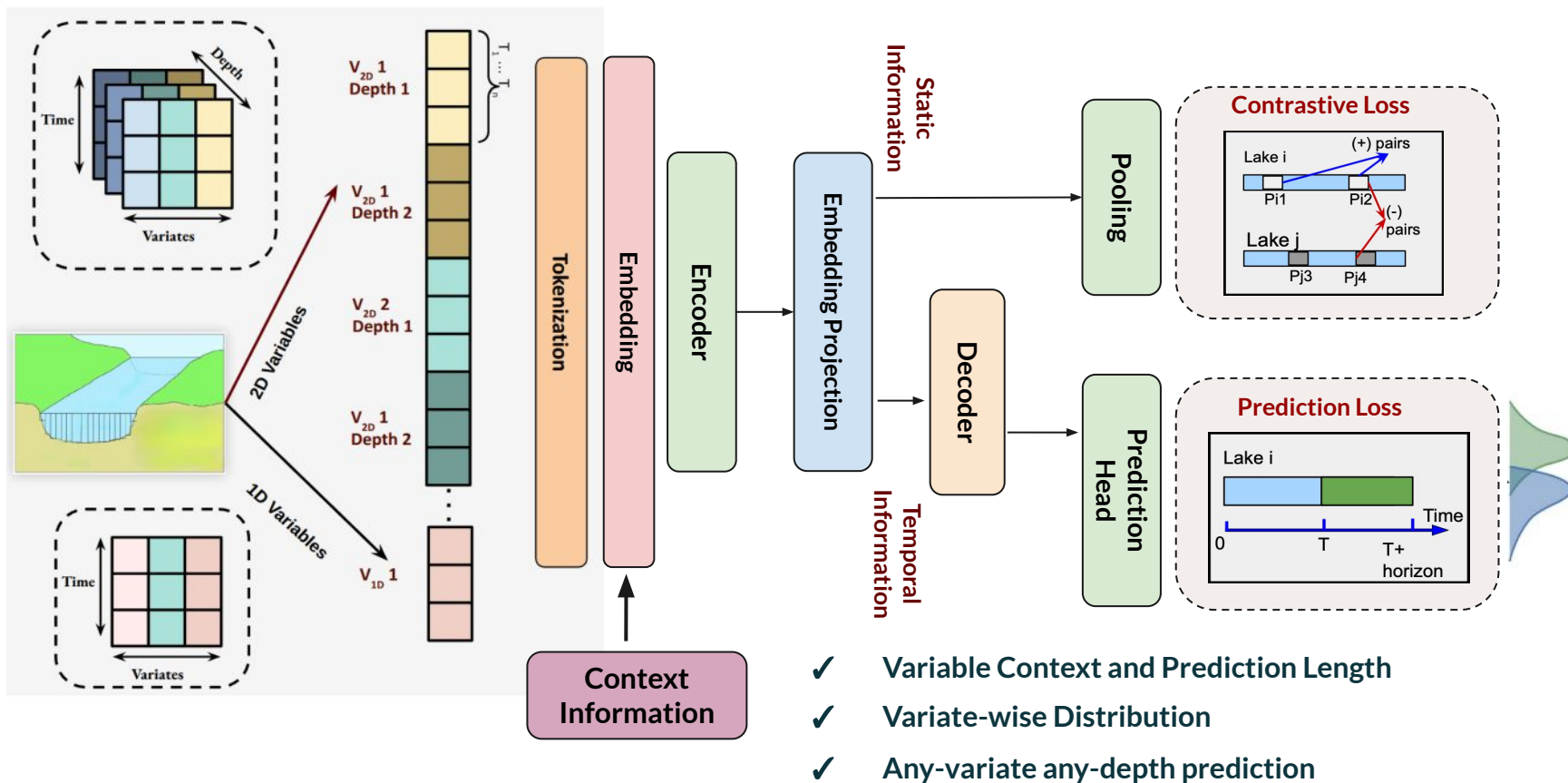
# Challenges with existing Foundation Models



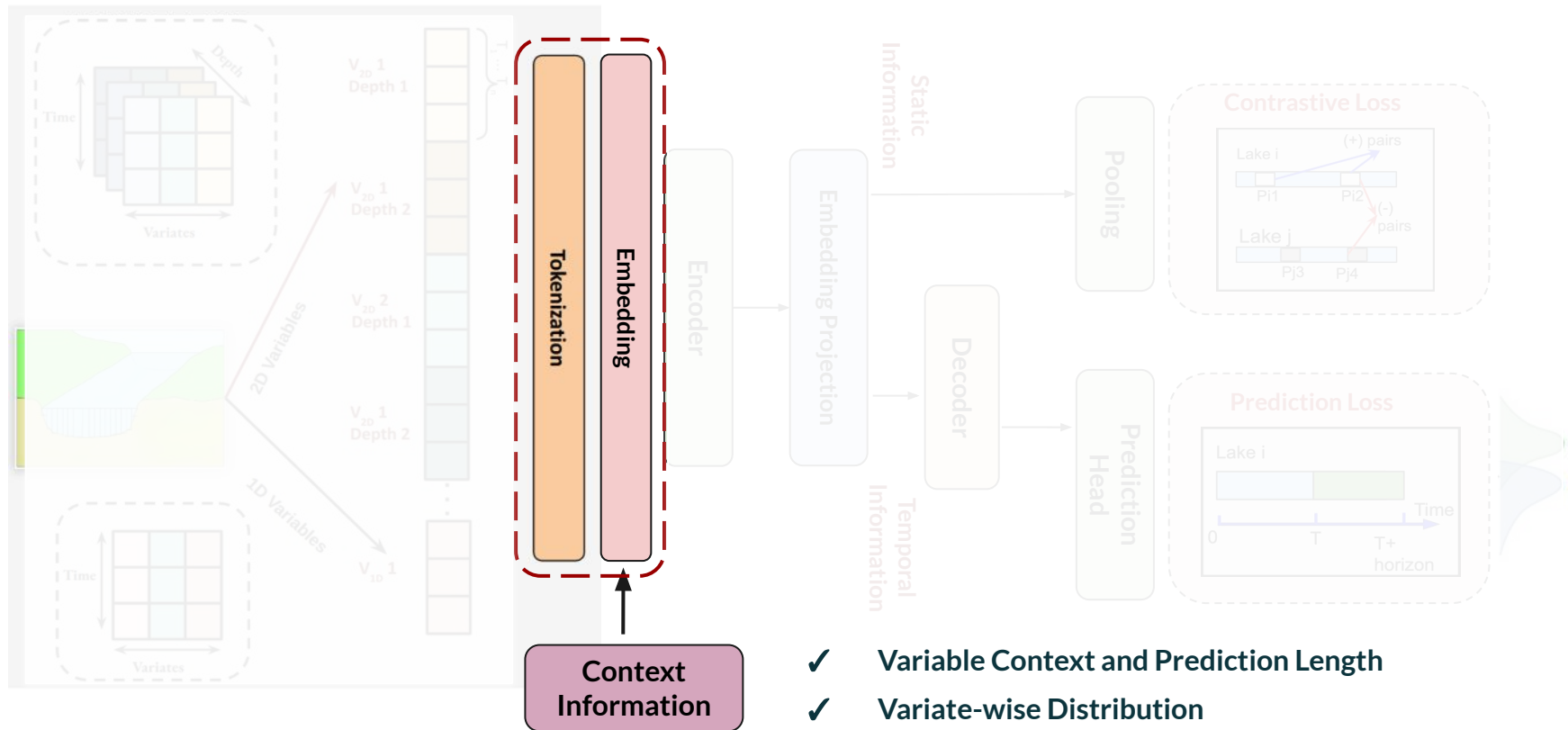
# Lake Foundation Model (LakeFM) - An Overview



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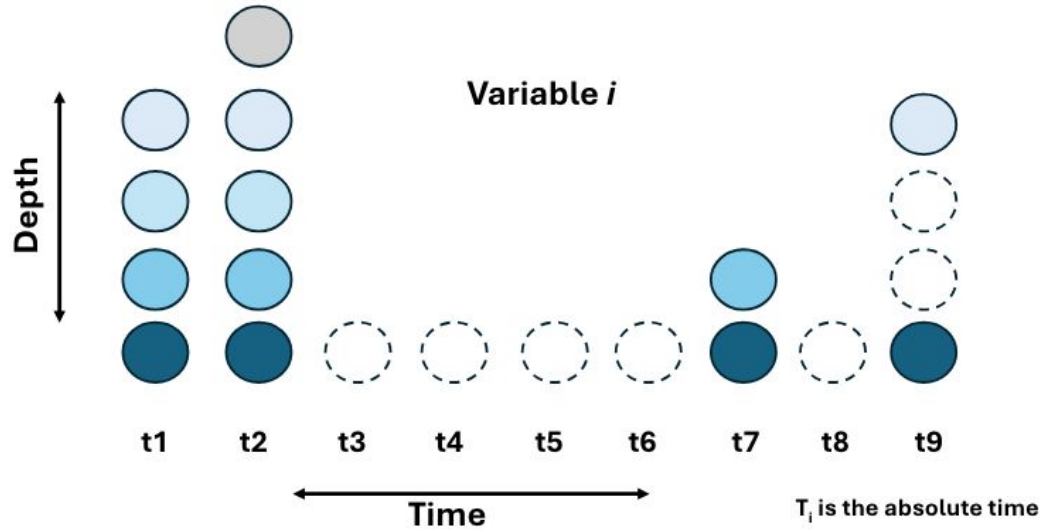
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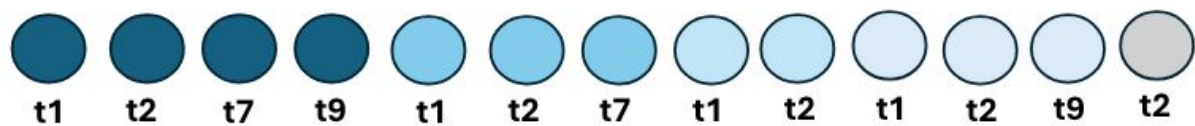
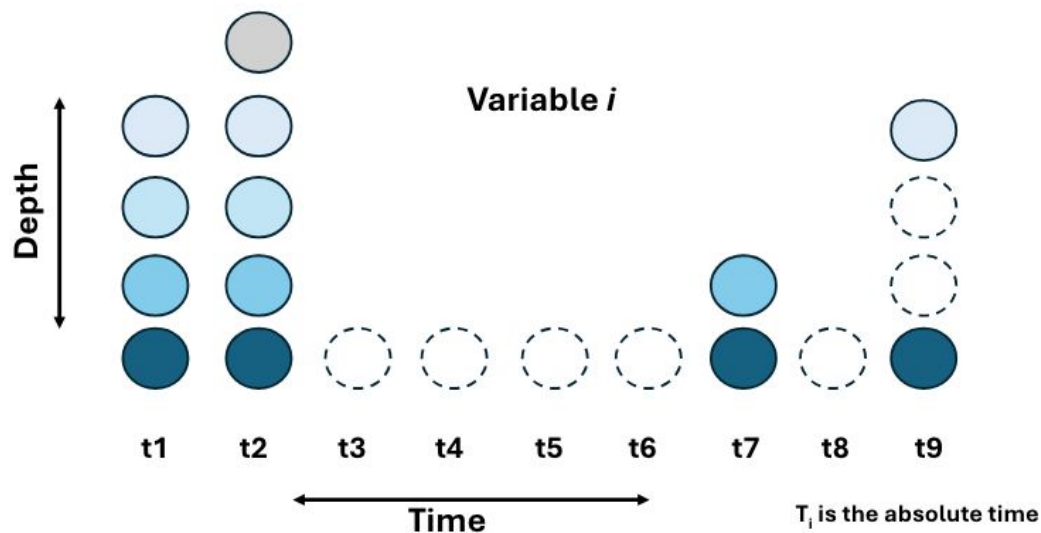
- ✓ Variable Context and Prediction Length
- ✓ Variate-wise Distribution
- ✓ Any-variate any-depth prediction



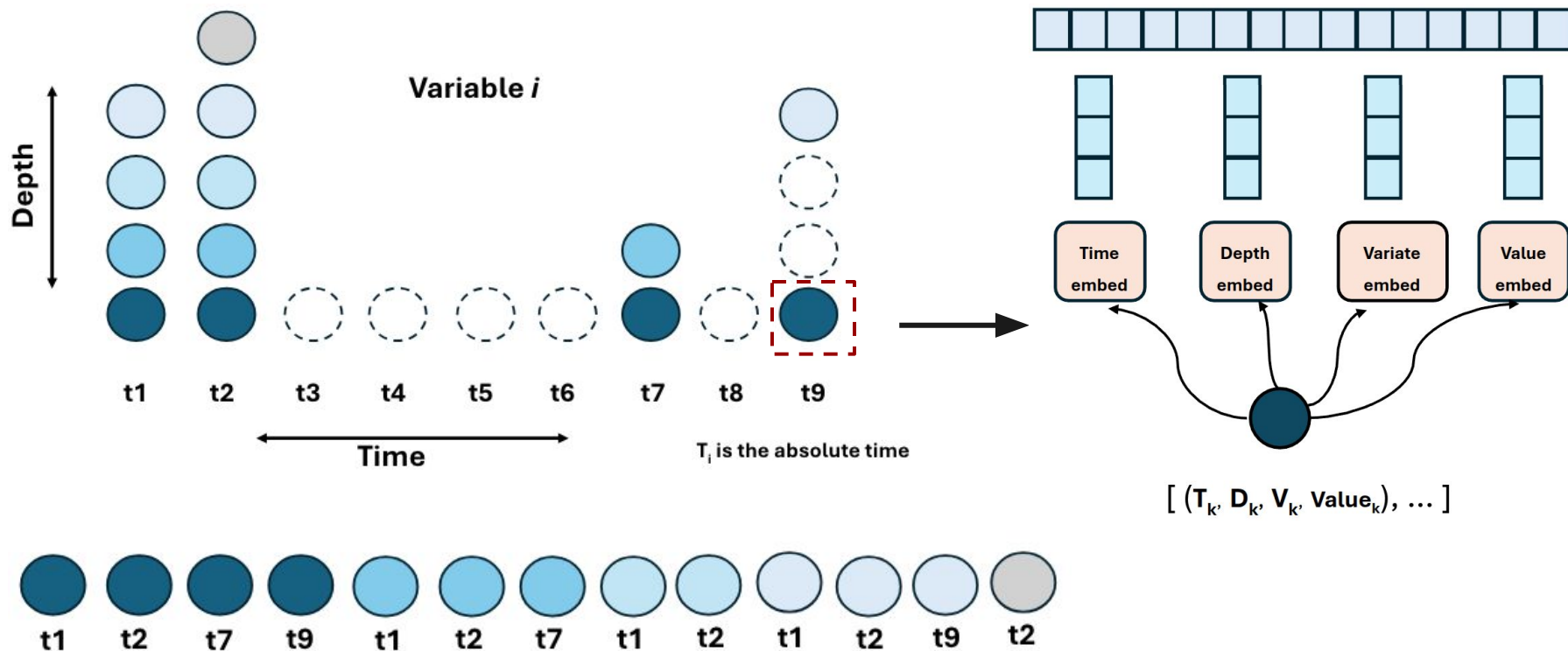
# Lake Foundation Model (LakeFM) - Tokenization & Embedding



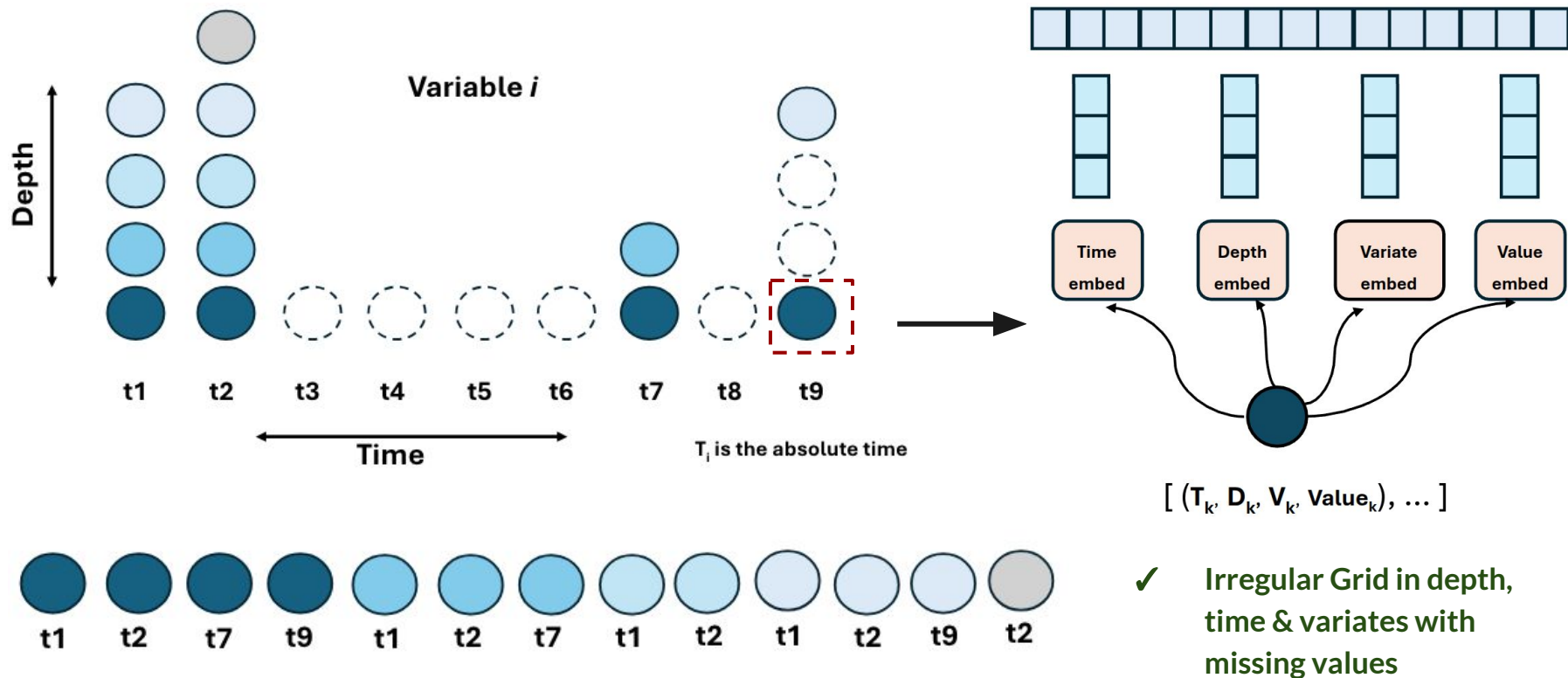
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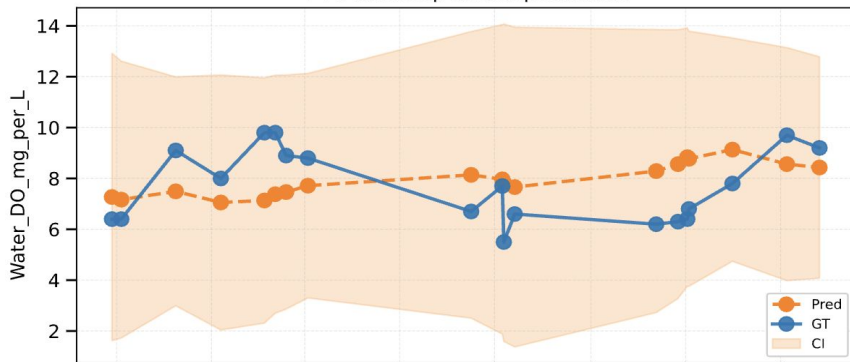
# Lake Foundation Model (LakeFM) - Tokenization & Embedding



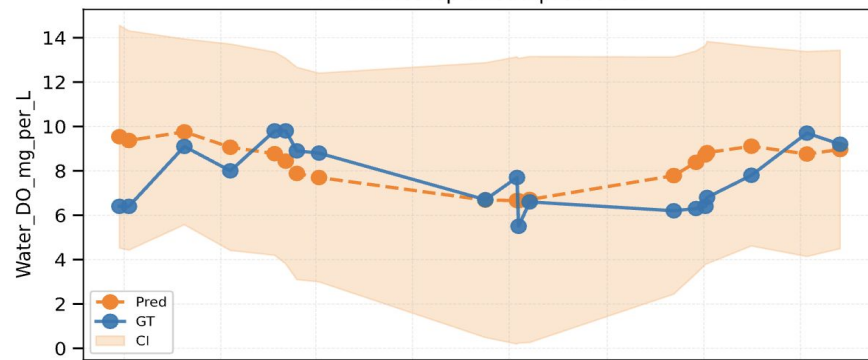
# Lake Foundation Model (LakeFM) - Findings (I) - Increasing Horizon Length

BARC @ depth=0.5m

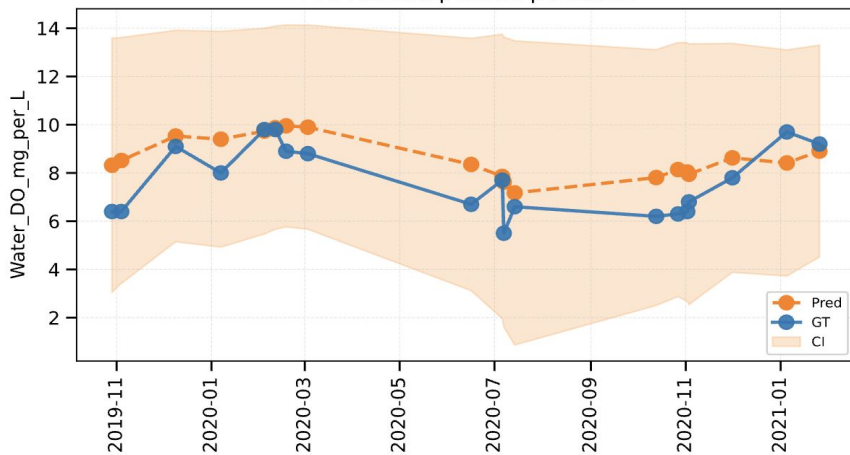
T+1 timestep ahead prediction



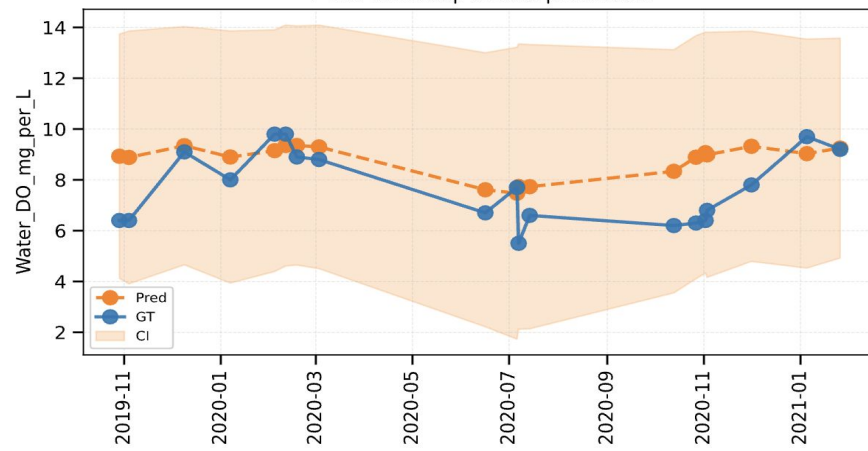
T+7 timestep ahead prediction



T+14 timestep ahead prediction



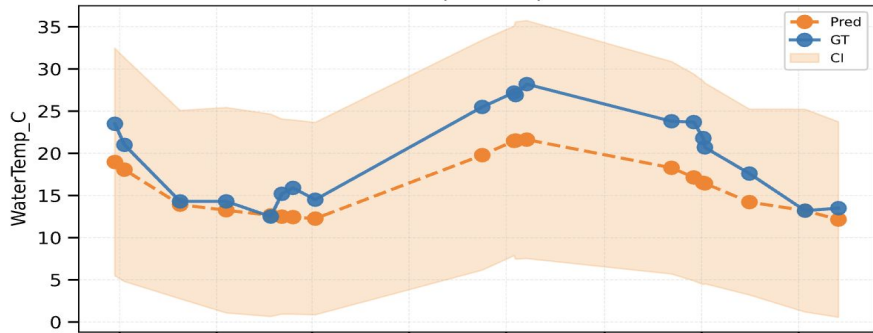
T+21 timestep ahead prediction



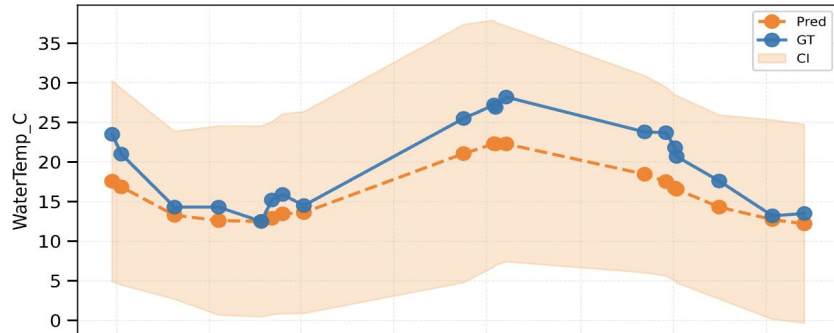
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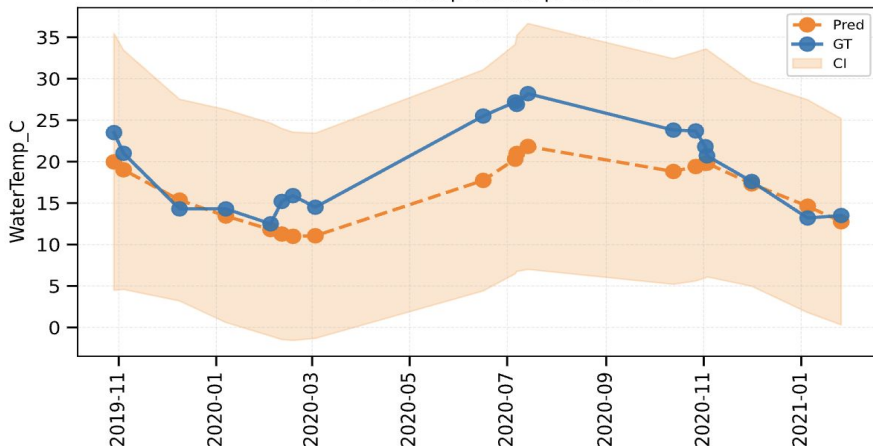
T+1 timestep ahead prediction



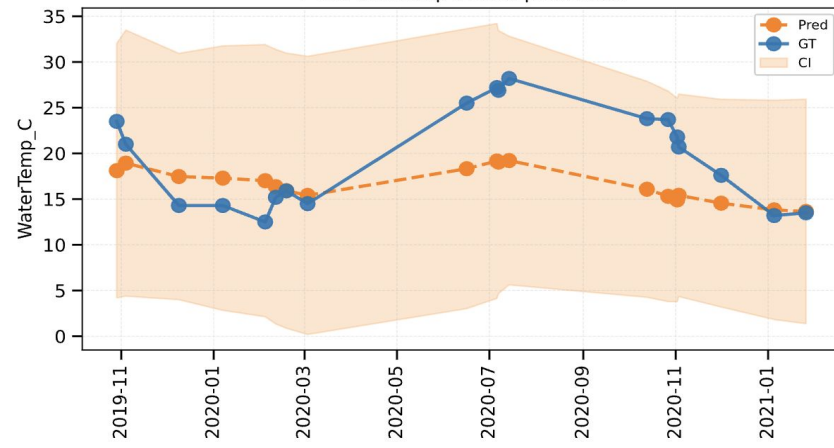
T+7 timestep ahead prediction



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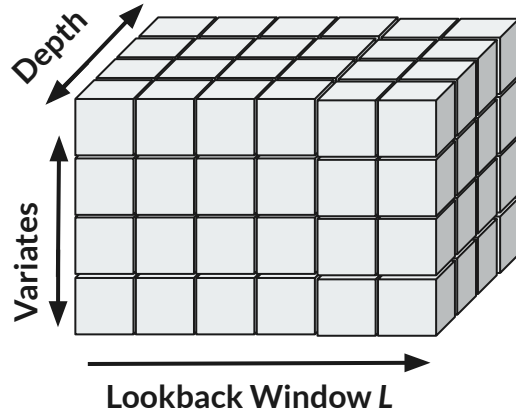


T+21 timestep ahead prediction

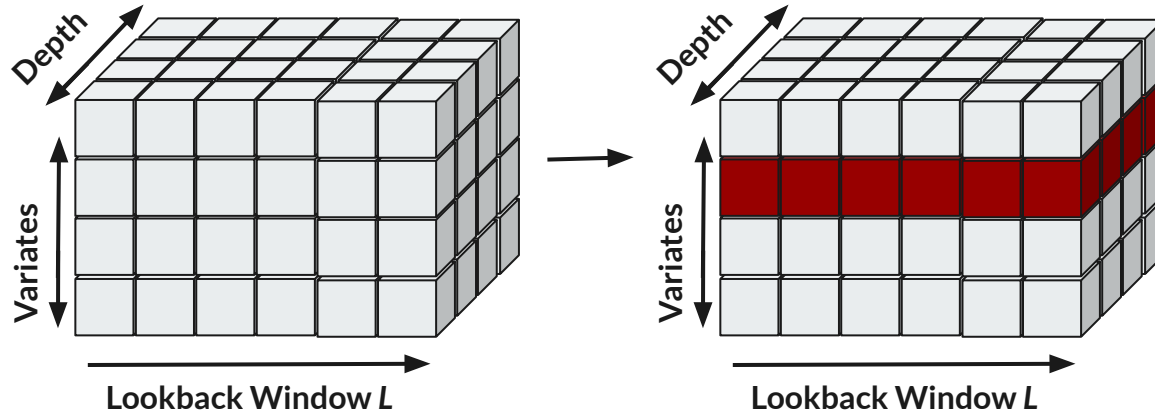


**Key Observation : Overall, LakeFM maintains stable performance across increasing horizon lengths**

# Lake Foundation Model (LakeFM) - Findings (II) - Incomplete Data (Variables)



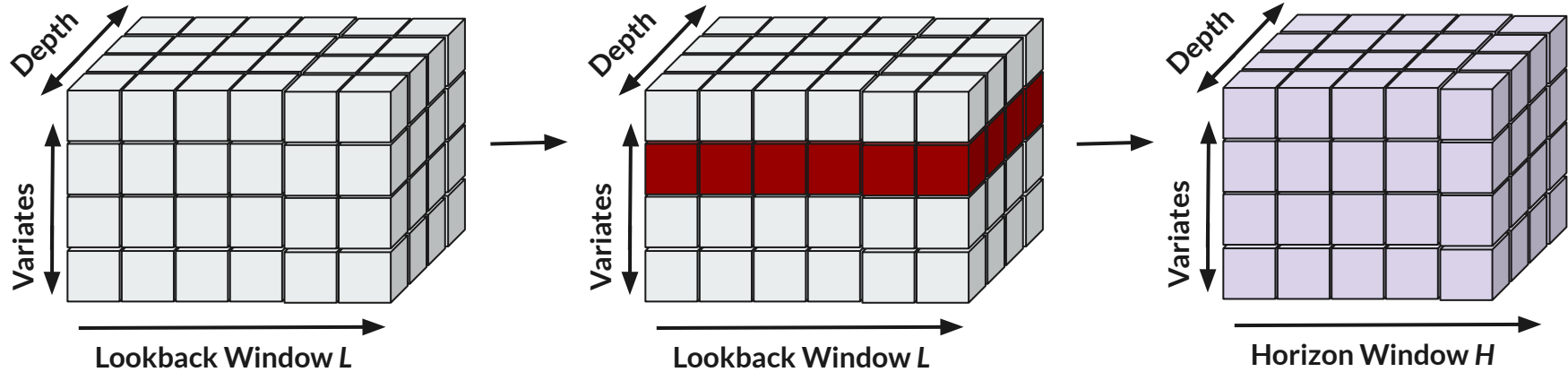
# Lake Foundation Model (LakeFM) - Findings (II) - Incomplete Data (Variables)



{ Air Temp, Shortwave, ..., ~~Water Temp~~, DO ... }



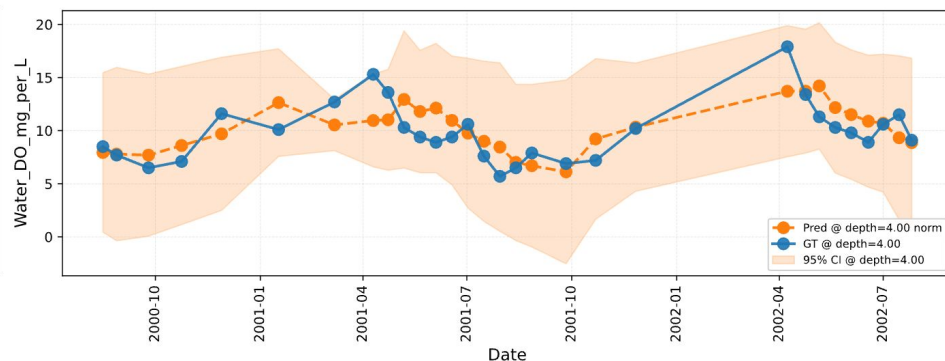
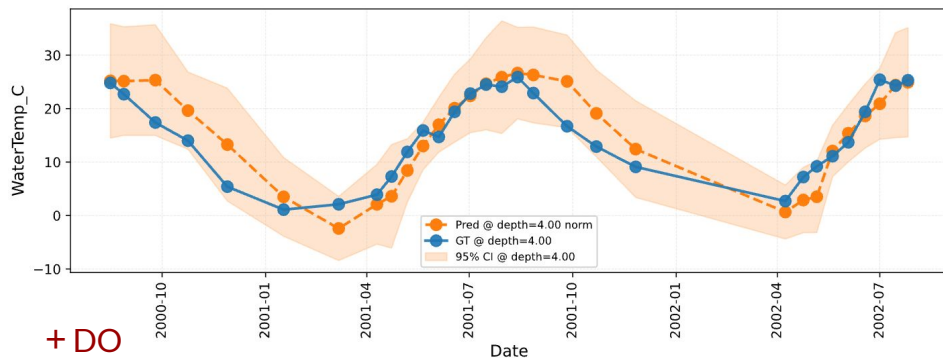
# Lake Foundation Model (LakeFM) - Findings (II) - Incomplete Data (Variables)



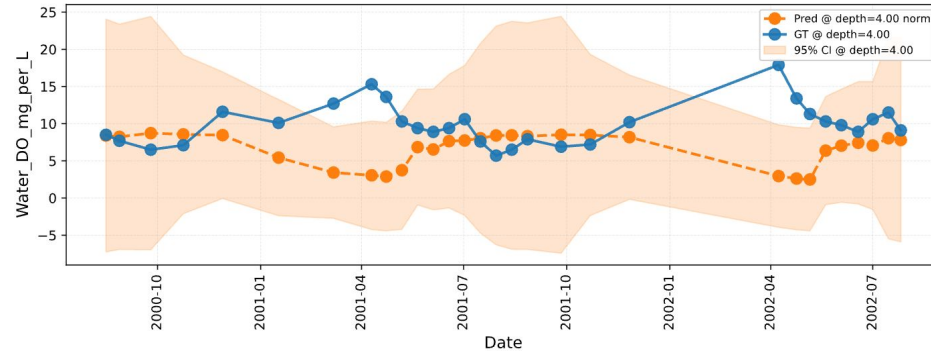
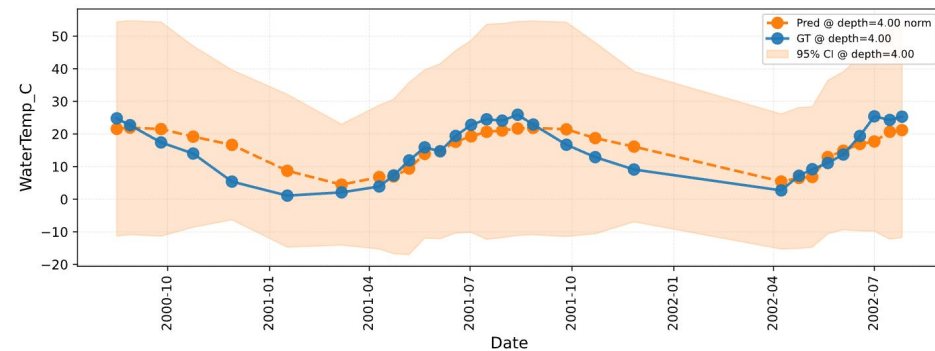
$\left\{ \text{Air Temp, Shortwave, ..., Water Temp, DO ...} \right\} \longrightarrow \left\{ \text{Air Temp, Shortwave, ..., Water Temp, DO ...} \right\}$

# Lake Foundation Model (LakeFM) - Findings (II) - Incomplete Data (Variables)

ME : 30 timesteps ahead forecast, at Depth 4.00m (shaded = 95% CI)



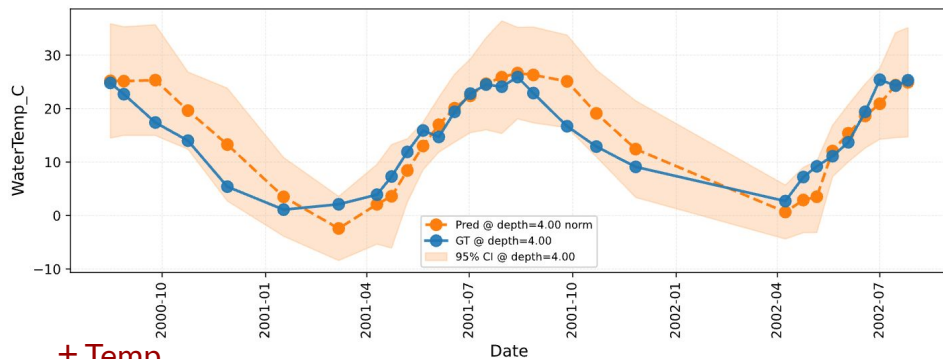
- DO



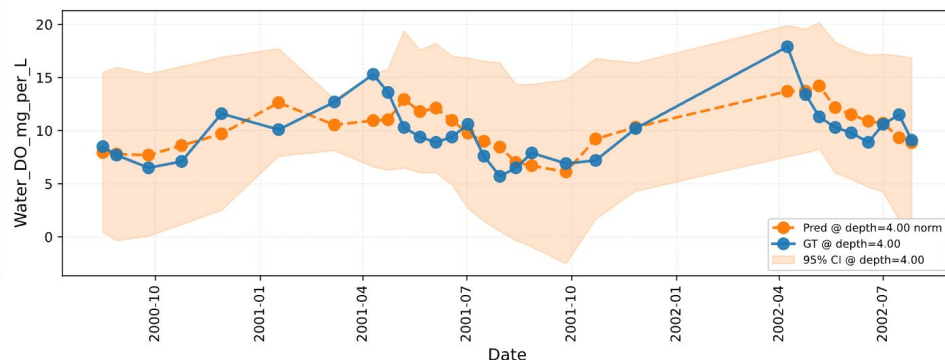
**Key Observation : Removing DO from inputs increases uncertainty in predictions of water temperature**

# Lake Foundation Model (LakeFM) - Findings (II) - Incomplete Data (Variables)

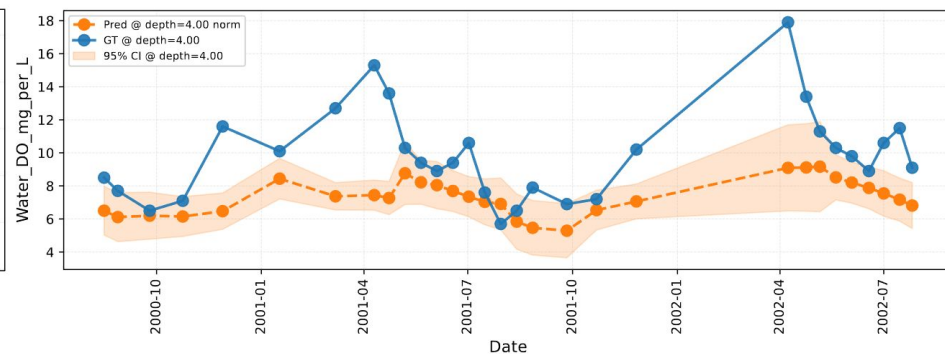
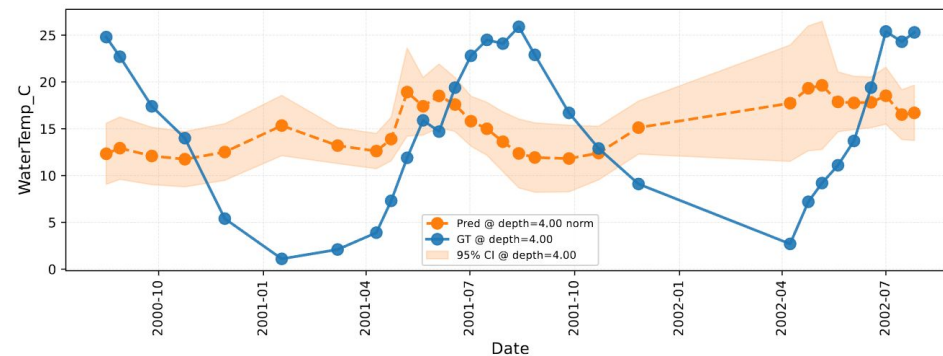
ME : 30 timesteps ahead forecast, at Depth 4.00m (shaded = 95% CI)



+ Temp

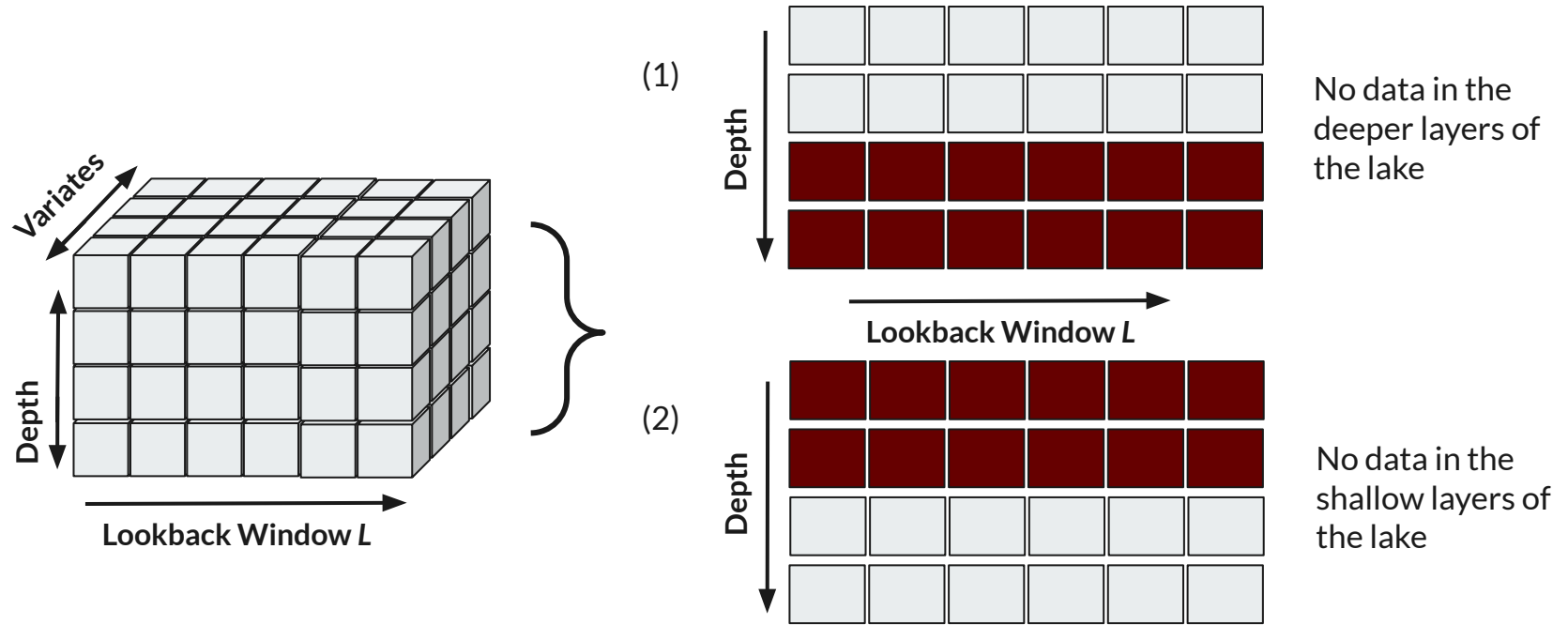


- Temp



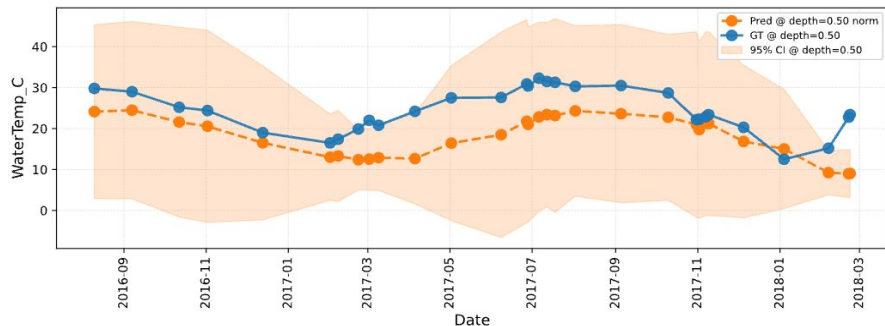
**Key Observation : Water temperature is a critical variable. Removing it degrades all predictions.**

# Lake Foundation Model (LakeFM) - Findings (III) - Incomplete Data (Depth)

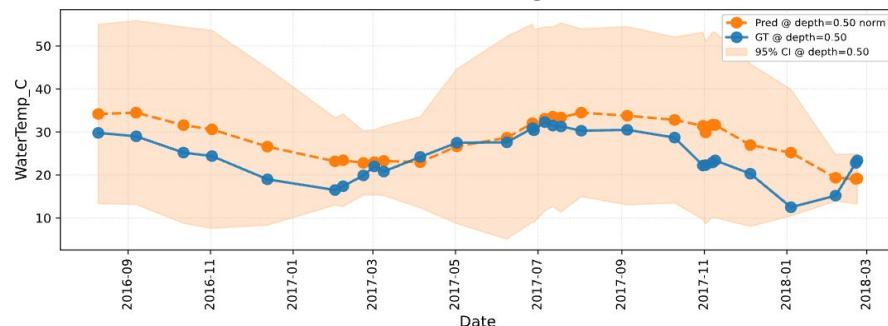


# Lake Foundation Model (LakeFM) - Findings (III) - Incomplete Data (Depth)

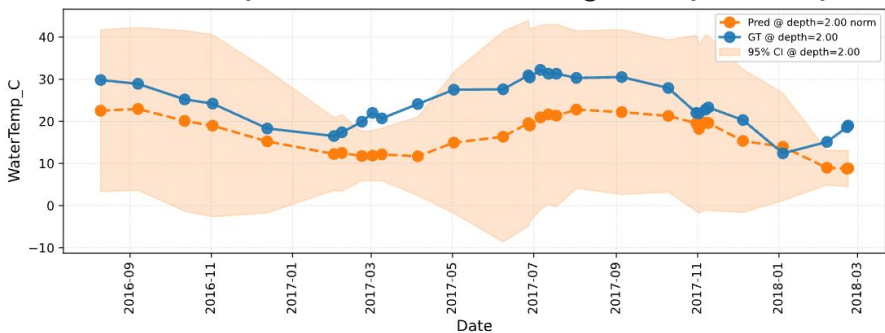
## Water Temp Predictions @ 0.5 m using full-depth history



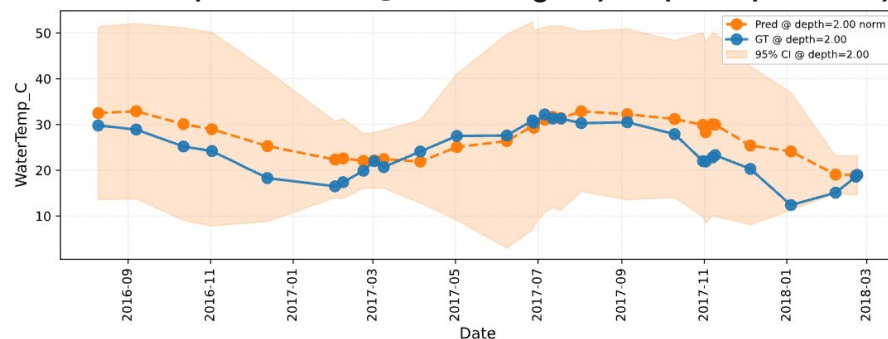
## Water Temp Predictions @ 0.5 m using only deeper-depth history



## Water Temp Predictions @ 2.0 m using full-depth history



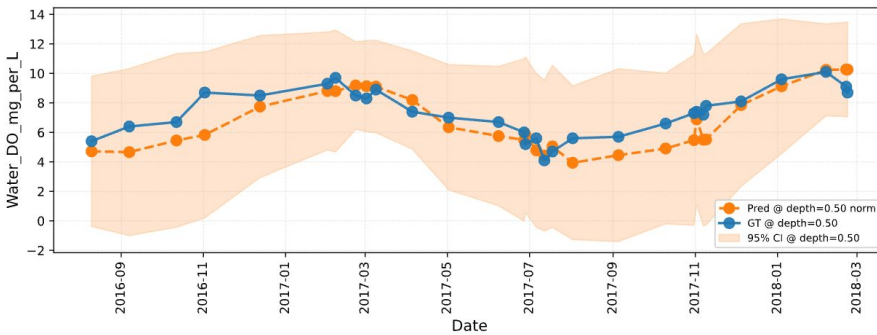
## Water Temp Predictions @ 2.0 m using only deeper-depth history



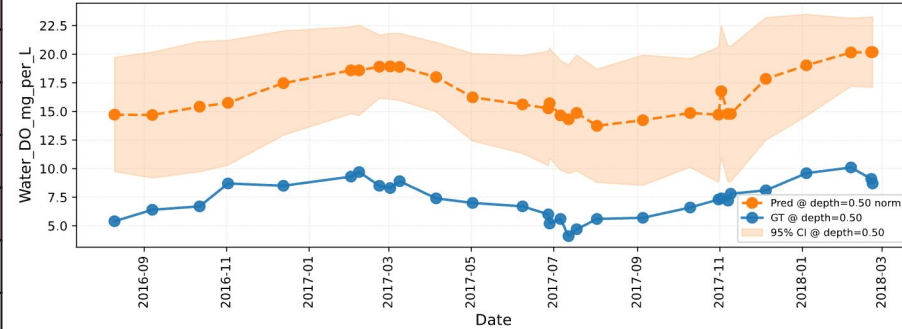
**Key Observation : Water temperature predictions remain stable even without shallow-layer variables**

# Lake Foundation Model (LakeFM) - Findings (III) - Incomplete Data (Depth)

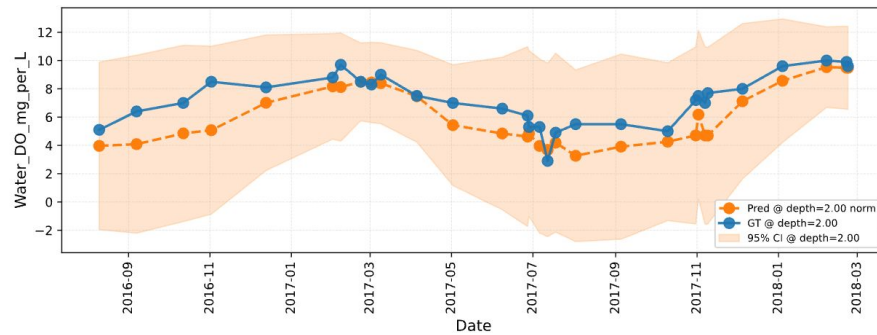
Water DO Predictions @ 0.5 m using full-depth history



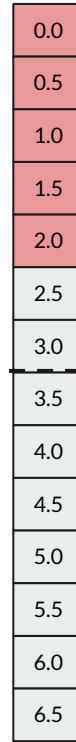
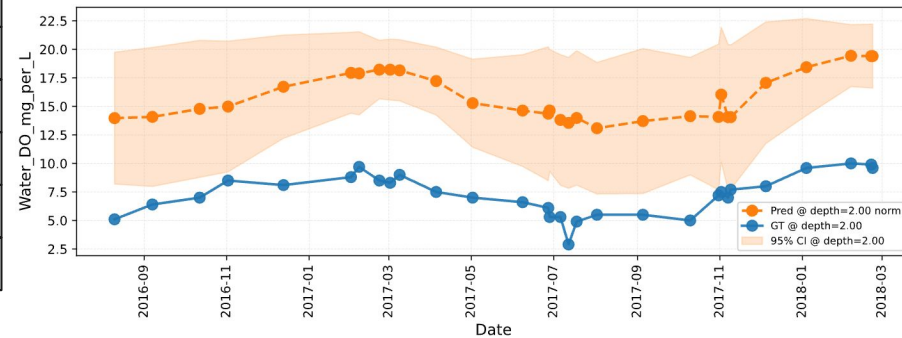
Water DO Predictions @ 0.5 m using only deeper-depth history



Water DO Predictions @ 2.0 m using full-depth history



Water DO Predictions @ 2.0 m using only deeper-depth history

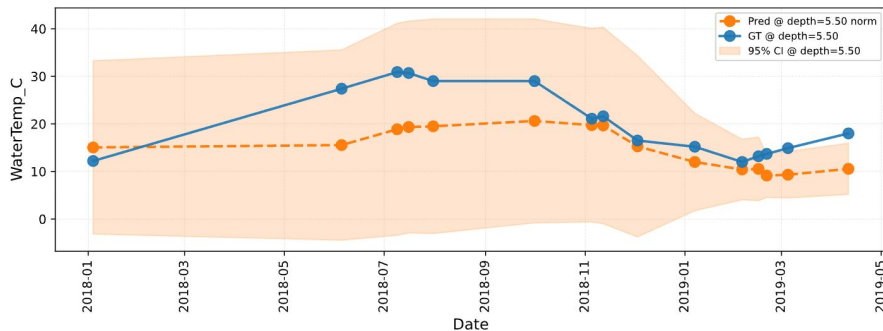


**Key Observation :** In contrast, DO predictions cannot rely on deeper-layer variables, indicating stronger vertical variability along the water column

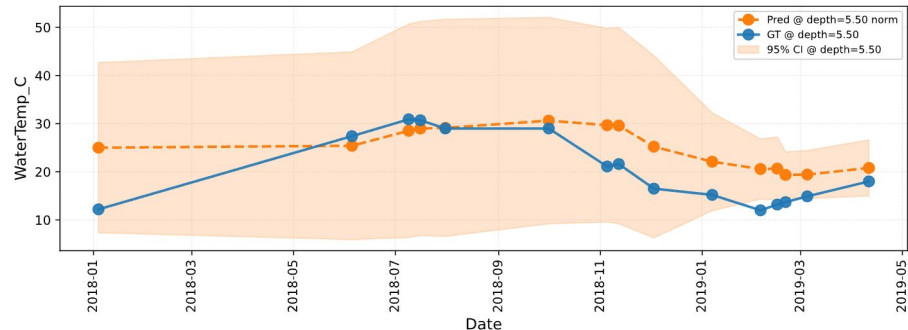


# Lake Foundation Model (LakeFM) - Findings (III) - Incomplete Data (Depth)

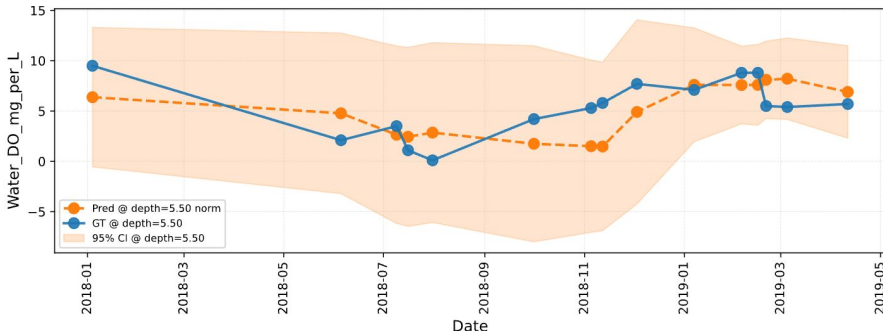
## Water Temp Predictions @ 5.5 m using full-depth history



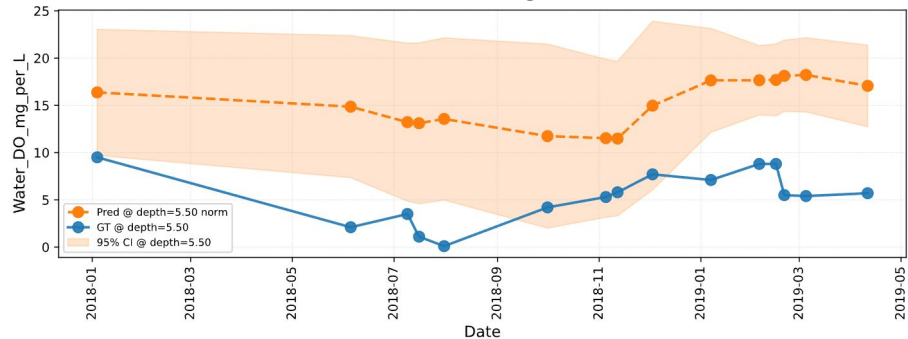
## Water Temp Predictions @ 5.5 m using only shallow-depth history



## Water DO Predictions @ 5.5 m using full-depth history



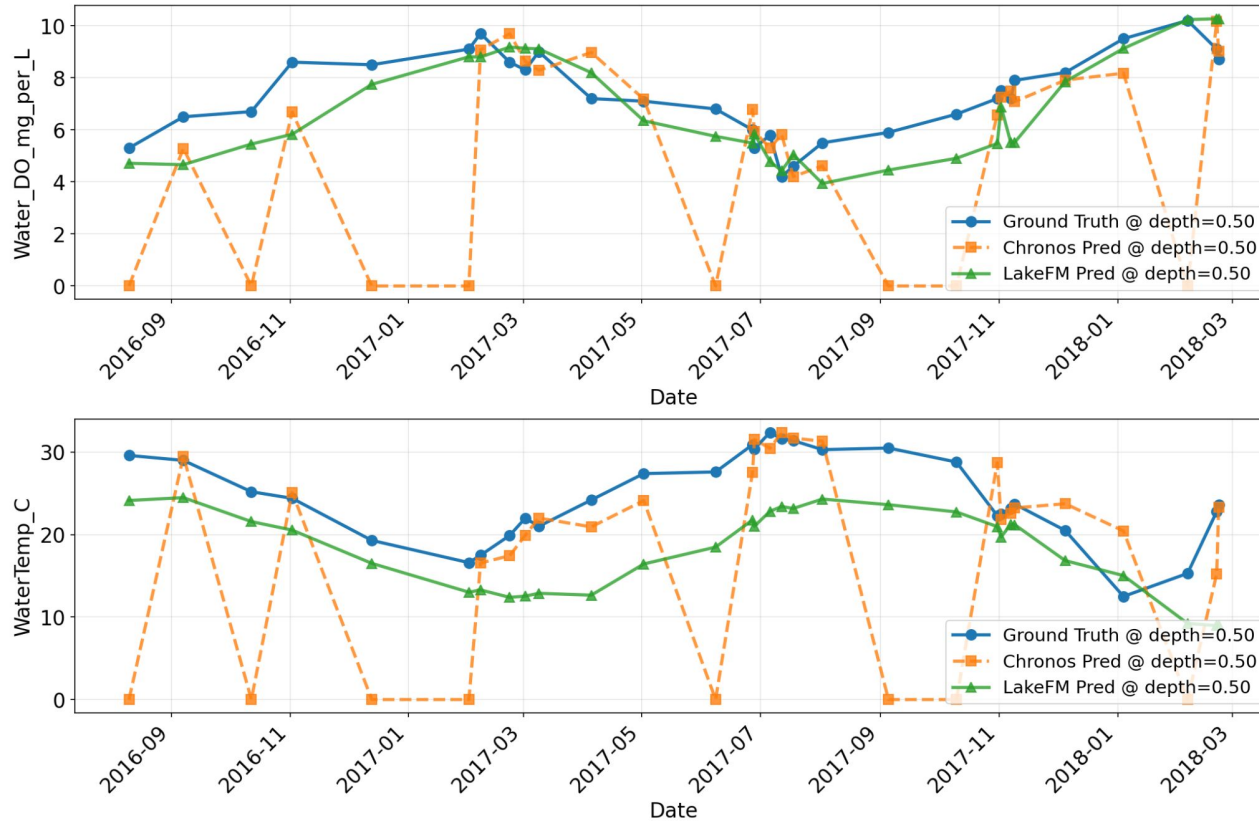
## Water DO Predictions @ 5.5 m using only shallow-depth history



**Key Observation :** In general, water temperature remains stable and is predictable using either shallow or deeper layers, while DO dynamics are tightly coupled to the local depth.

# Lake Foundation Model (LakeFM) - Findings (IV) - Performance Comparison

Comparing LakeFM predictions (on a horizon window of 30 timesteps) with Chronos Foundation Model on Lake BARC at Depth 0.5m



**Key Observation :**

**Chronos struggles with missing data, leading to context-dependent instability, whereas LakeFM remains stable under the same conditions**



# Lake Foundation Model (LakeFM) - Ongoing Work

## Lake representation analysis

- ❖ Visualization of learned lake embeddings
- ❖ Analyzing seasonal clustering patterns in lake representations
- ❖ Analyzing temporal trajectories of lake representations over time
- ❖ *Geographic structure emerging in embedding space*

## Variable representation analysis

- ❖ Visualization of learned variable embeddings
- ❖ Analyzing variable similarities inferred from embedding clusters
- ❖ Empirical verification of embedding-based variable similarities

**Thank you**