

The background of the slide is an abstract network diagram. It features a complex web of thin, light gray lines connecting various nodes. The nodes are represented by circles of different sizes and colors, including dark blue, light blue, and gray. Some nodes are larger and more prominent, while others are smaller and less noticeable. The overall effect is a sense of interconnectedness and data flow.

## PROJECT 1

# CO CONCENTRATION LEVEL PREDICTION

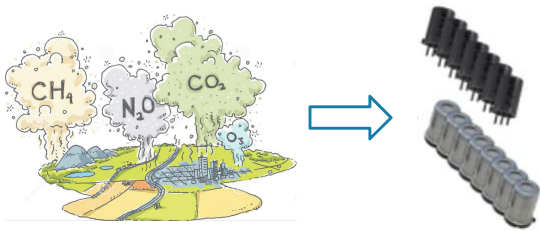
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8조 진형남, 박민수, 신명진, 장대근

# 목차

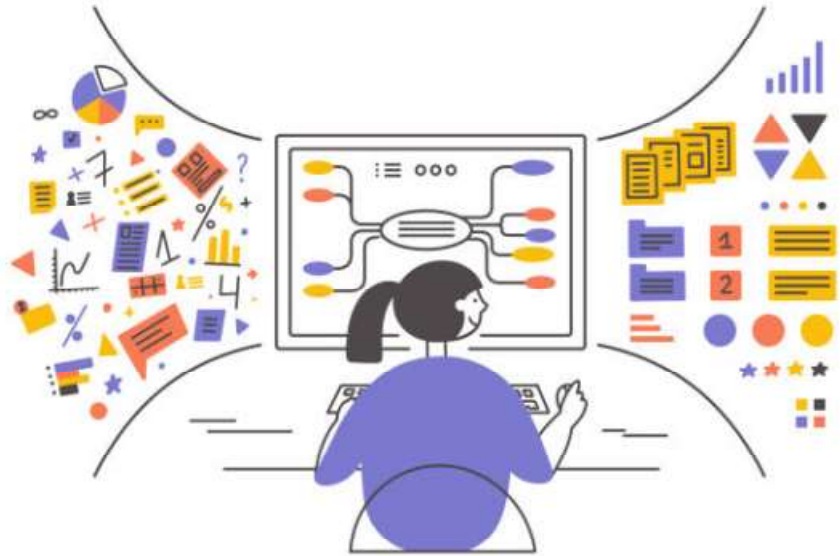
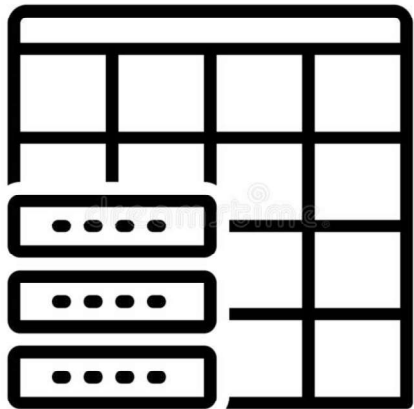
- Problem Statement
- Exploration
  - ✓ Domain Exploration (Literature Review)
  - ✓ Data Exploration
- Methods & Results
  - ✓ Overall Approach
  - ✓ Summary: Models & Results
- Conclusion

# Problem Statement



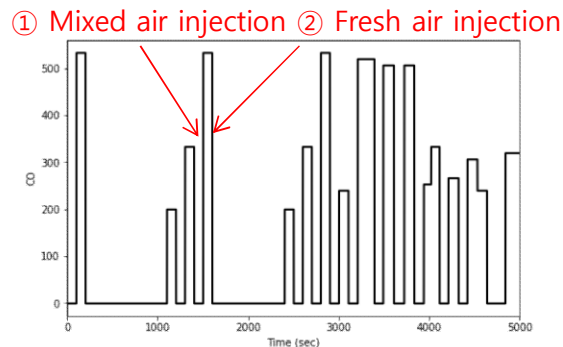
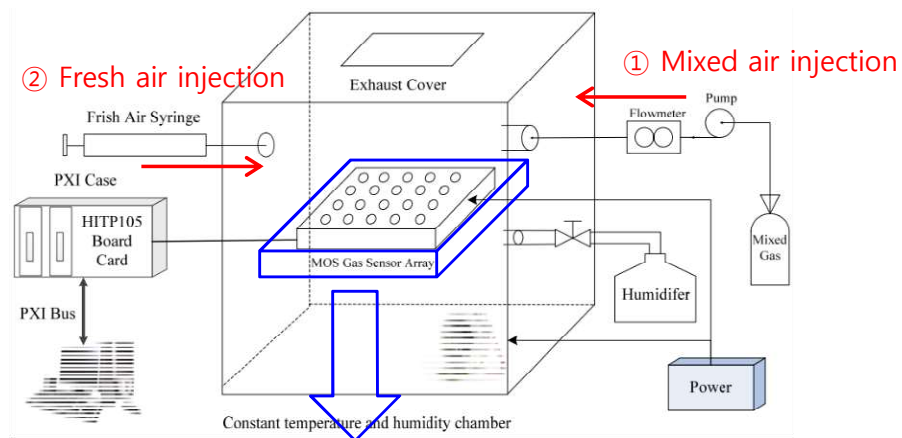
*Predict CO concentration levels  
from 16 gas sensor data*

Given data



# Domain Exploration (1/2) by Literature Review

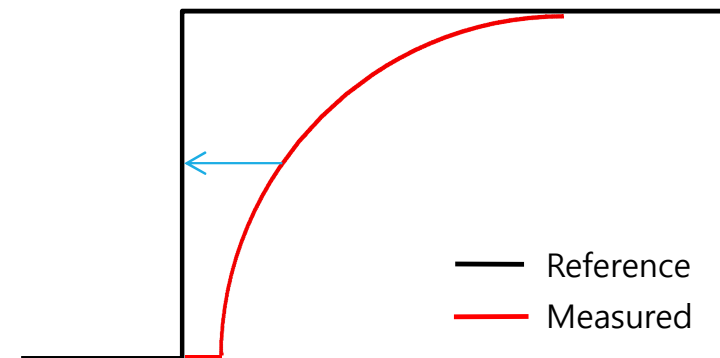
## ■ (General) In-Lab Mixed Gas Test Environment



## ■ Delayed sensor response due to:

- 1) Remaining gas in the chamber (before fully exhausted)
- 2) Sensor array structure
- 3) Sensor internal characteristics ( $\because$  analog lowpass filter)

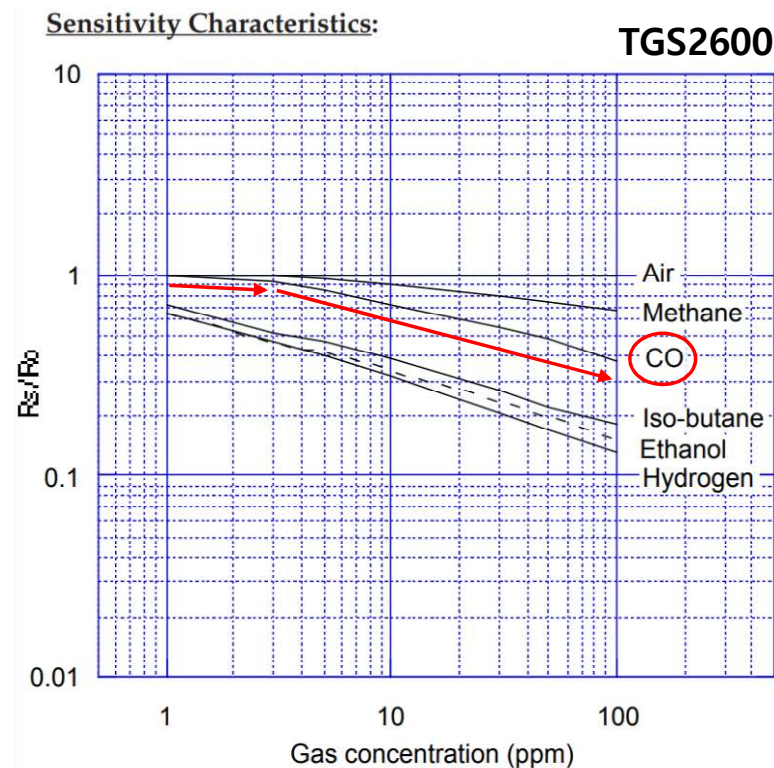
👉 **Need to align measured signals with reference CO**



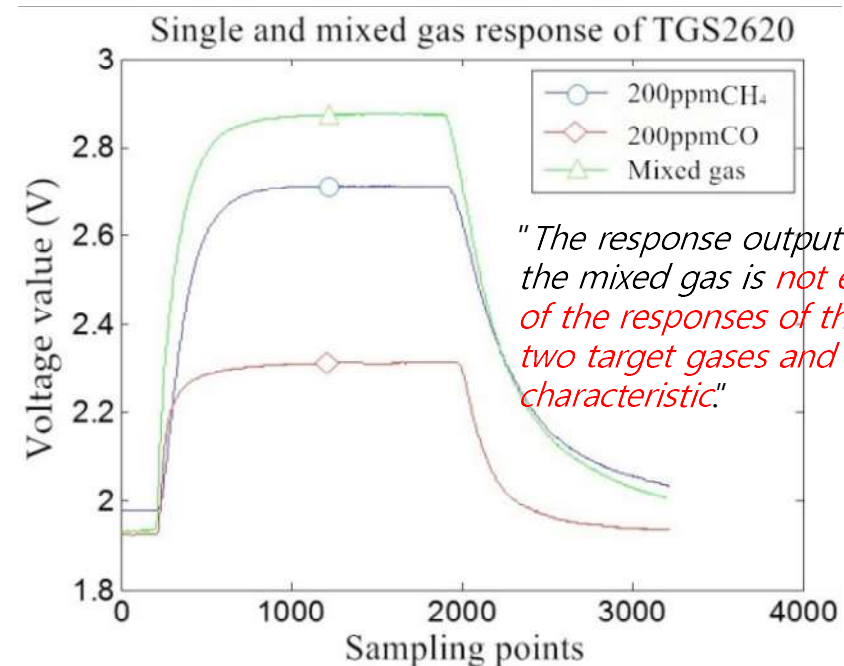
\*Ref.: Yonghui Xu, et al., "Research on a Mixed Gas Recognition and Concentration Detection Algorithm Based on a Metal Oxide Semiconductor Olfactory System Sensor Array," *Sensors*, 2018

# Domain Exploration (2/2) by Literature Review

- Nonlinearity in Cross-sensitivity characteristics & Mixed response



👉 **Model should represent these nonlinearities!**

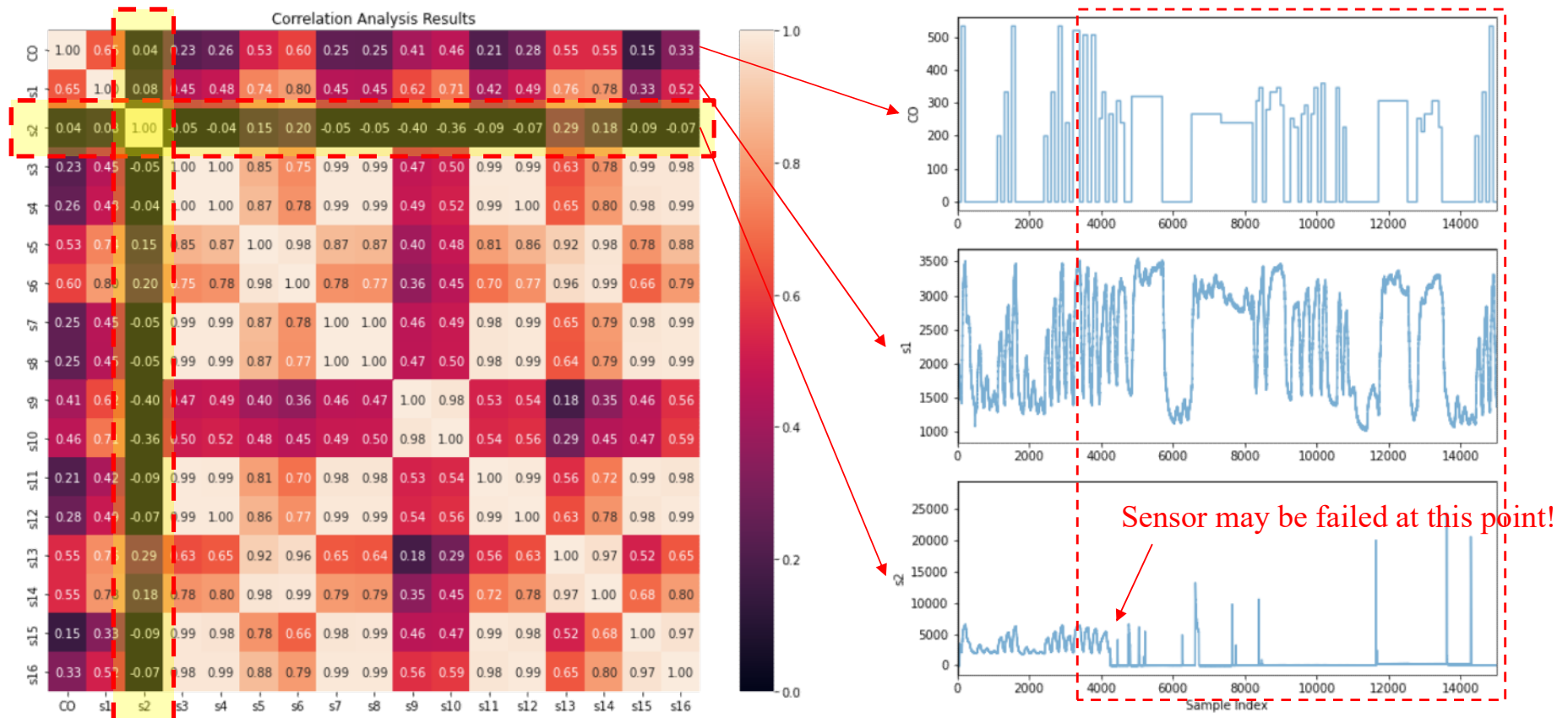


\*Ref.: Datasheets (Sensor specifications) of FIGARO TGS Series Sensors

\*Ref.: Yonghui Xu, et al., "Research on a Mixed Gas Recognition and Concentration Detection Algorithm Based on a Metal Oxide Semiconductor Olfactory System Sensor Array," *Sensors*, 2018

# Data Exploration (1/3)

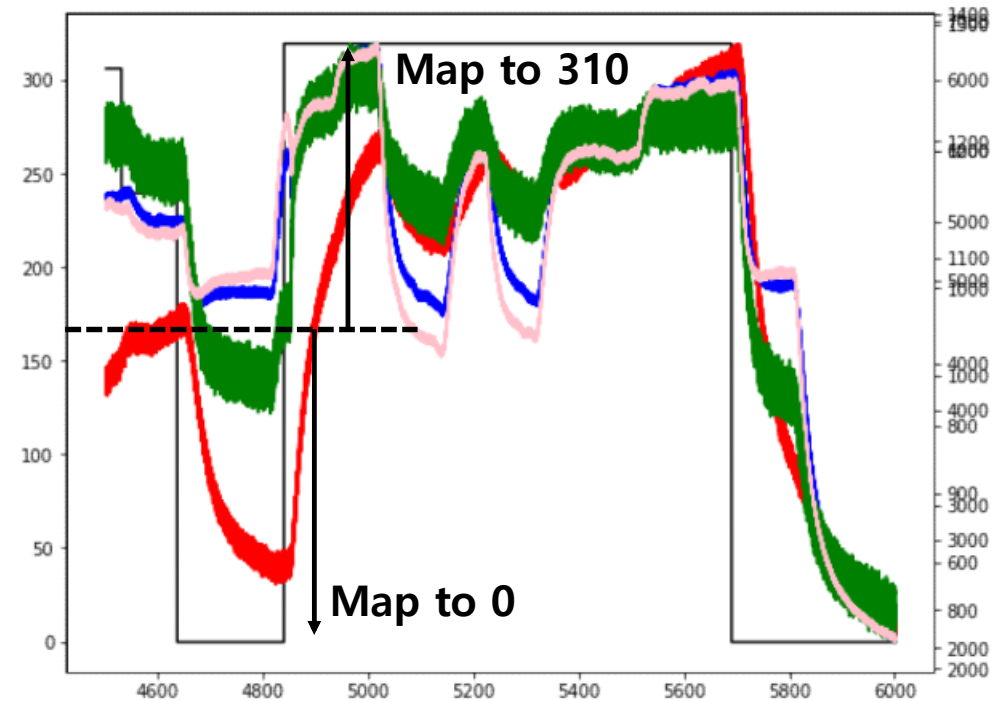
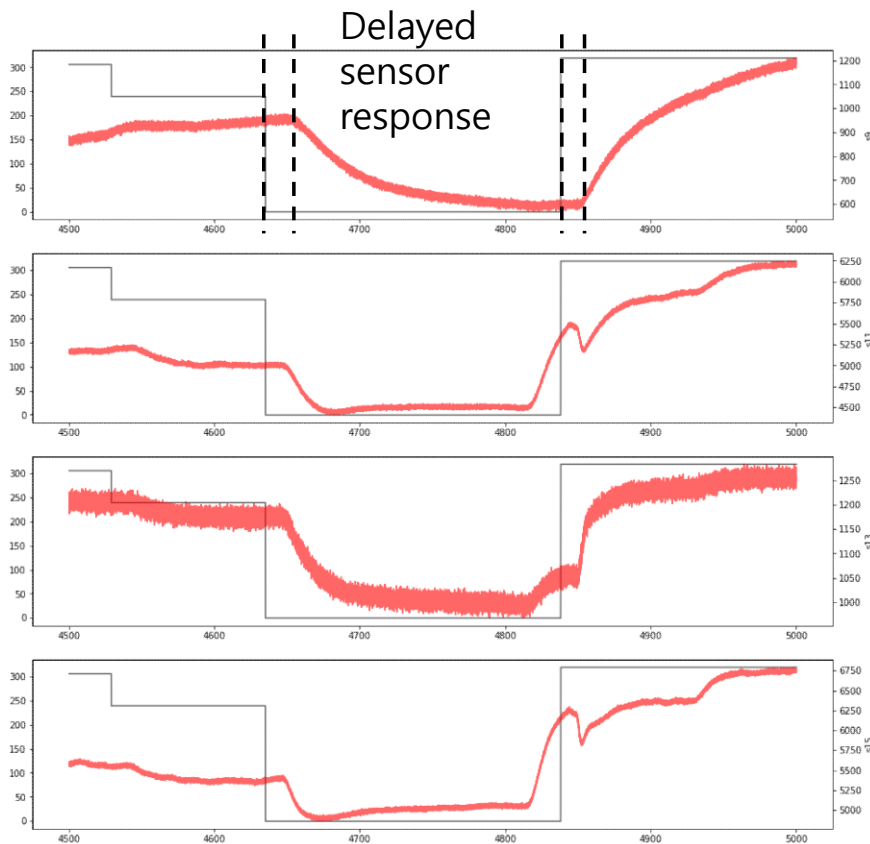
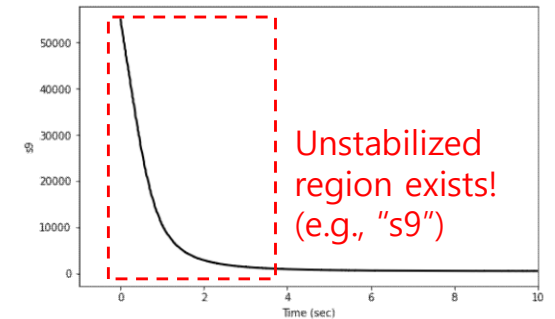
- Cross-Correlation Analysis: ① Sensor Failure @ "s2" → Exclusion?  
② Similar Sensors → Selection or Merge?





# Data Exploration (2/3)

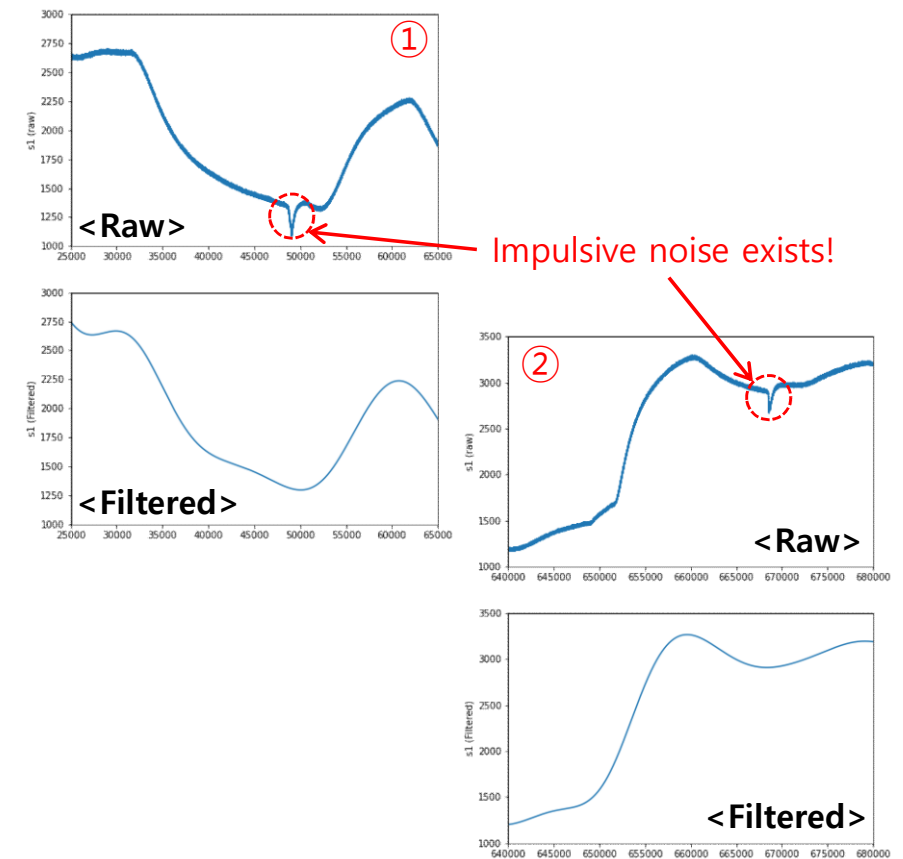
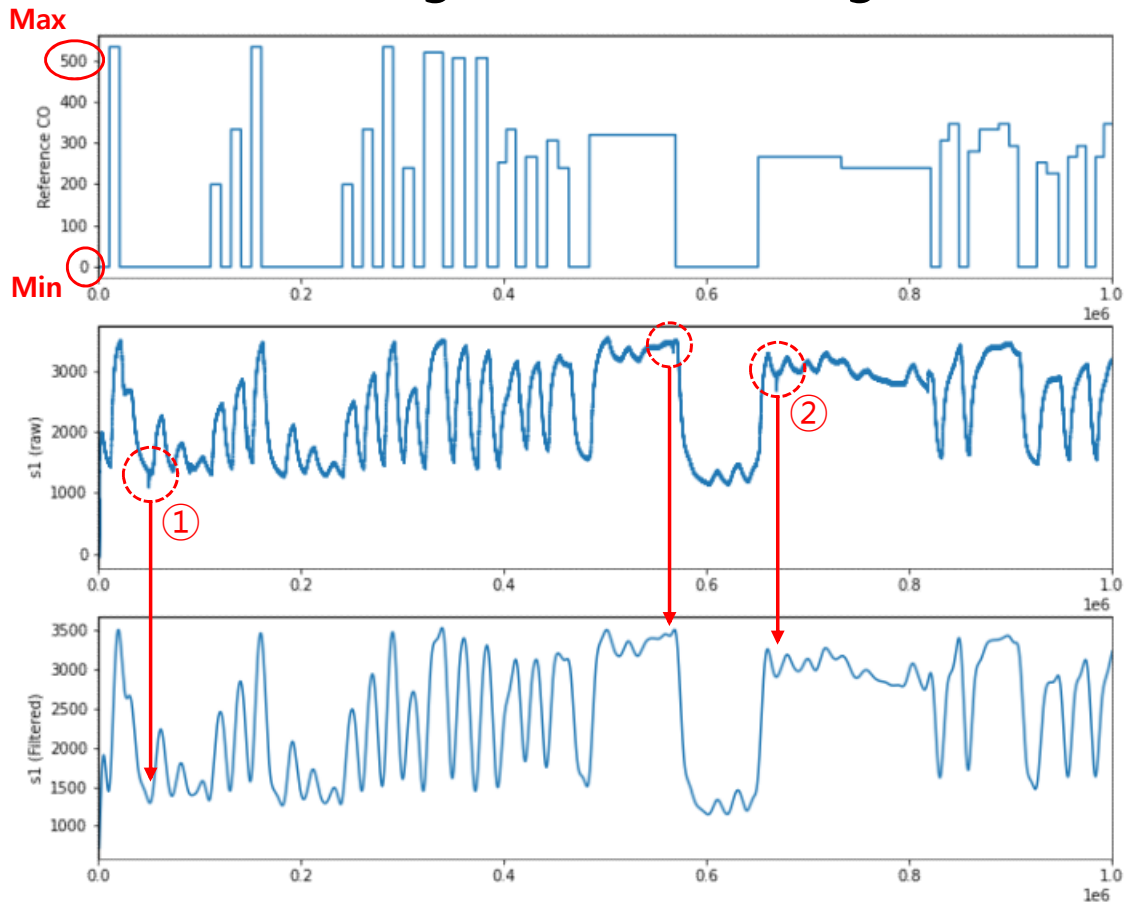
- Need of Signal Shifts → **2000 samples shifted!**



- \*Measured signal = Lowpass-filtered form of target signal
- \*\*Poor ability to track abrupt change in CO concentration level

# Data Exploration (3/3)

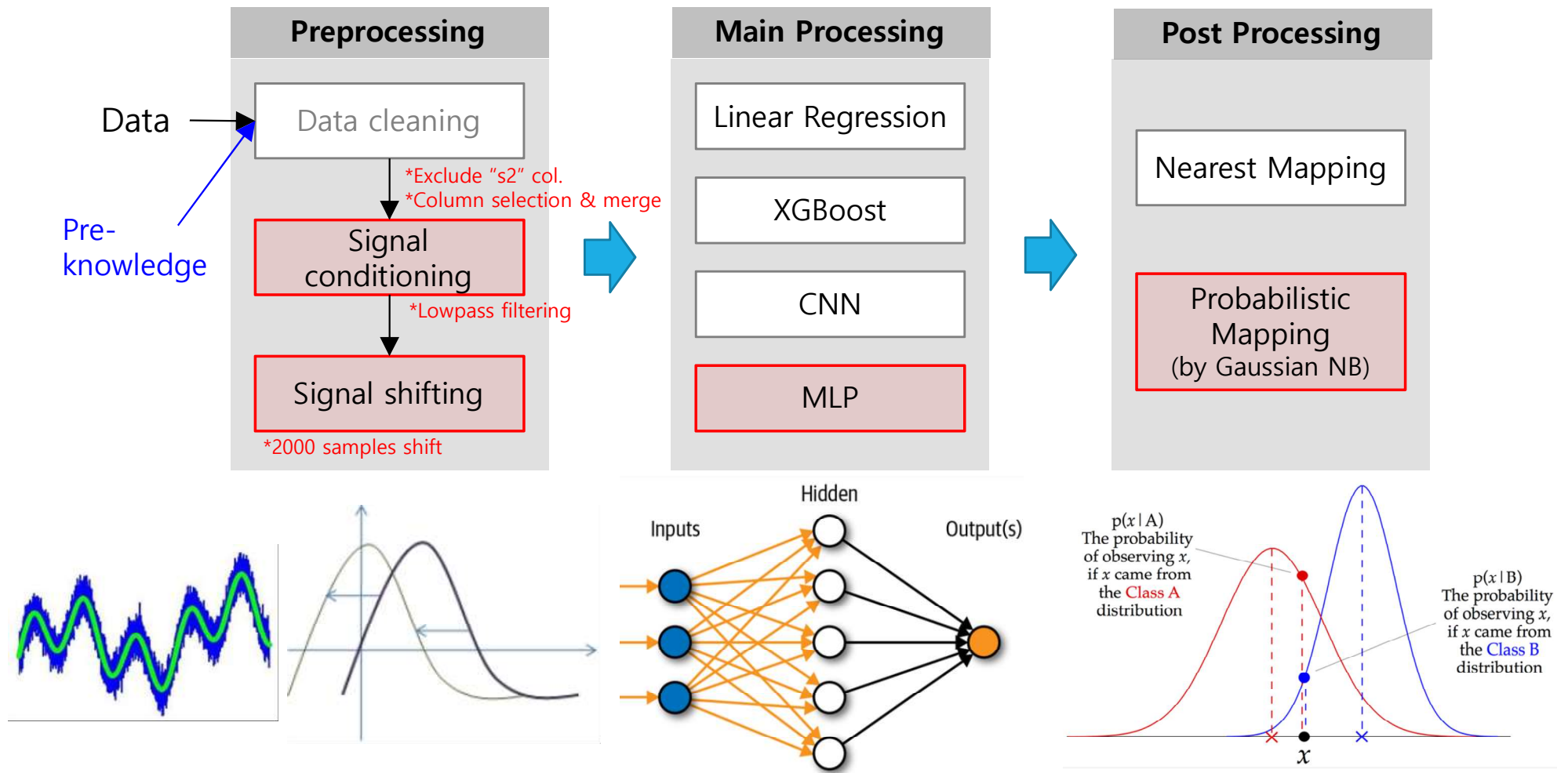
- Need of Signal Conditioning → **Zero-Phase Lowpass Filter Applied!**



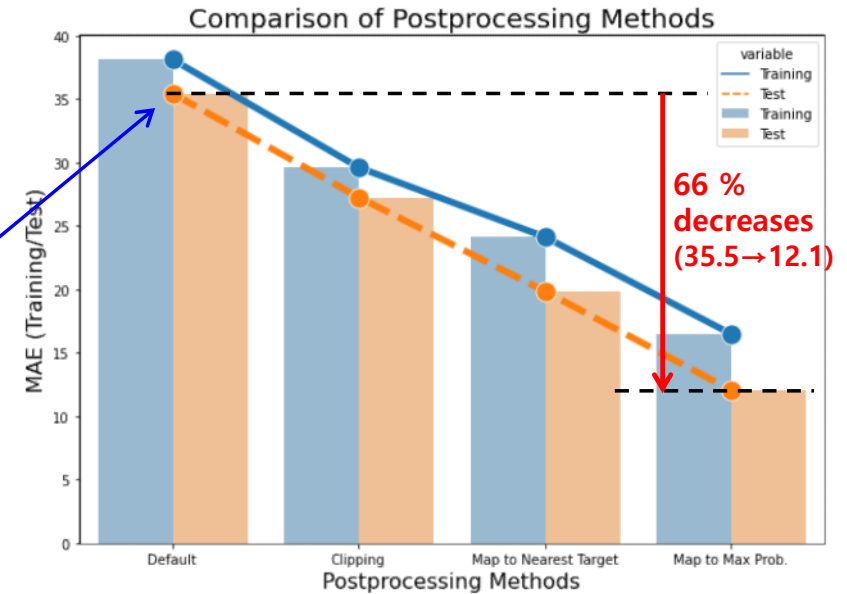
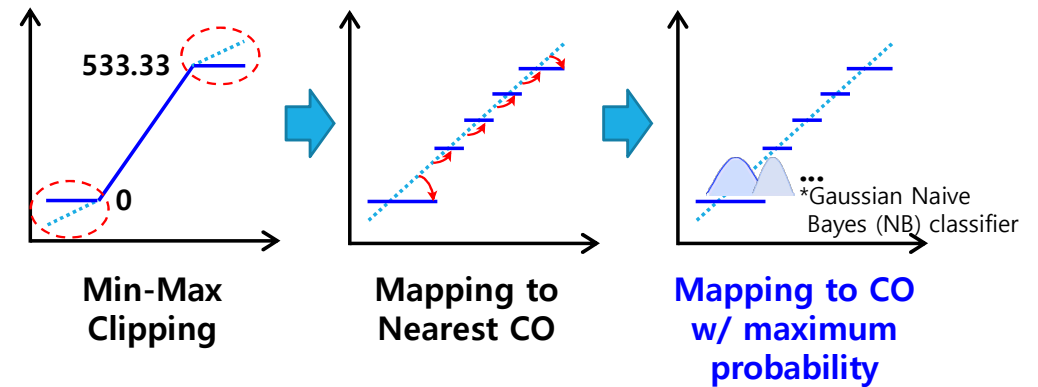
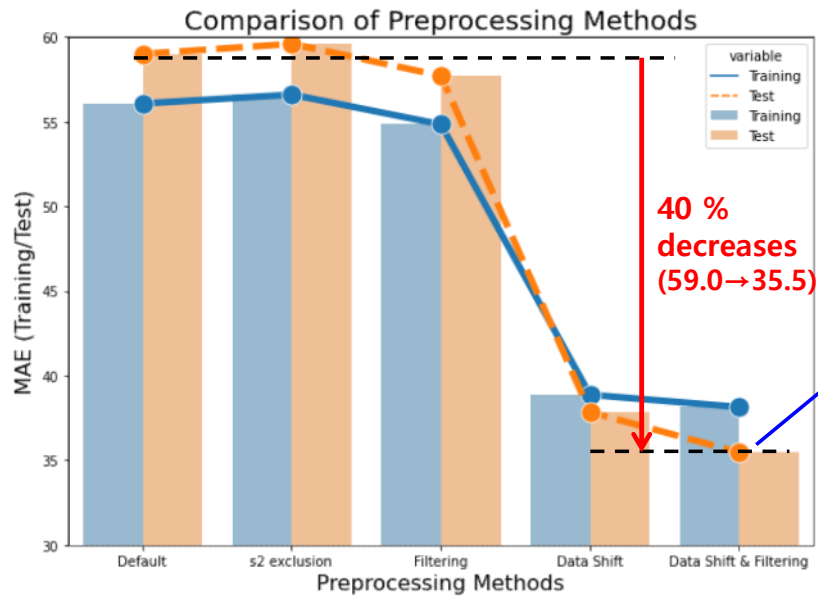
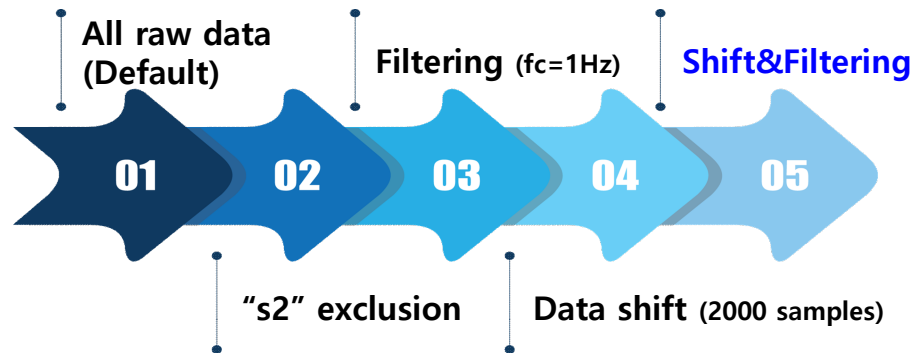


# Overall Approach

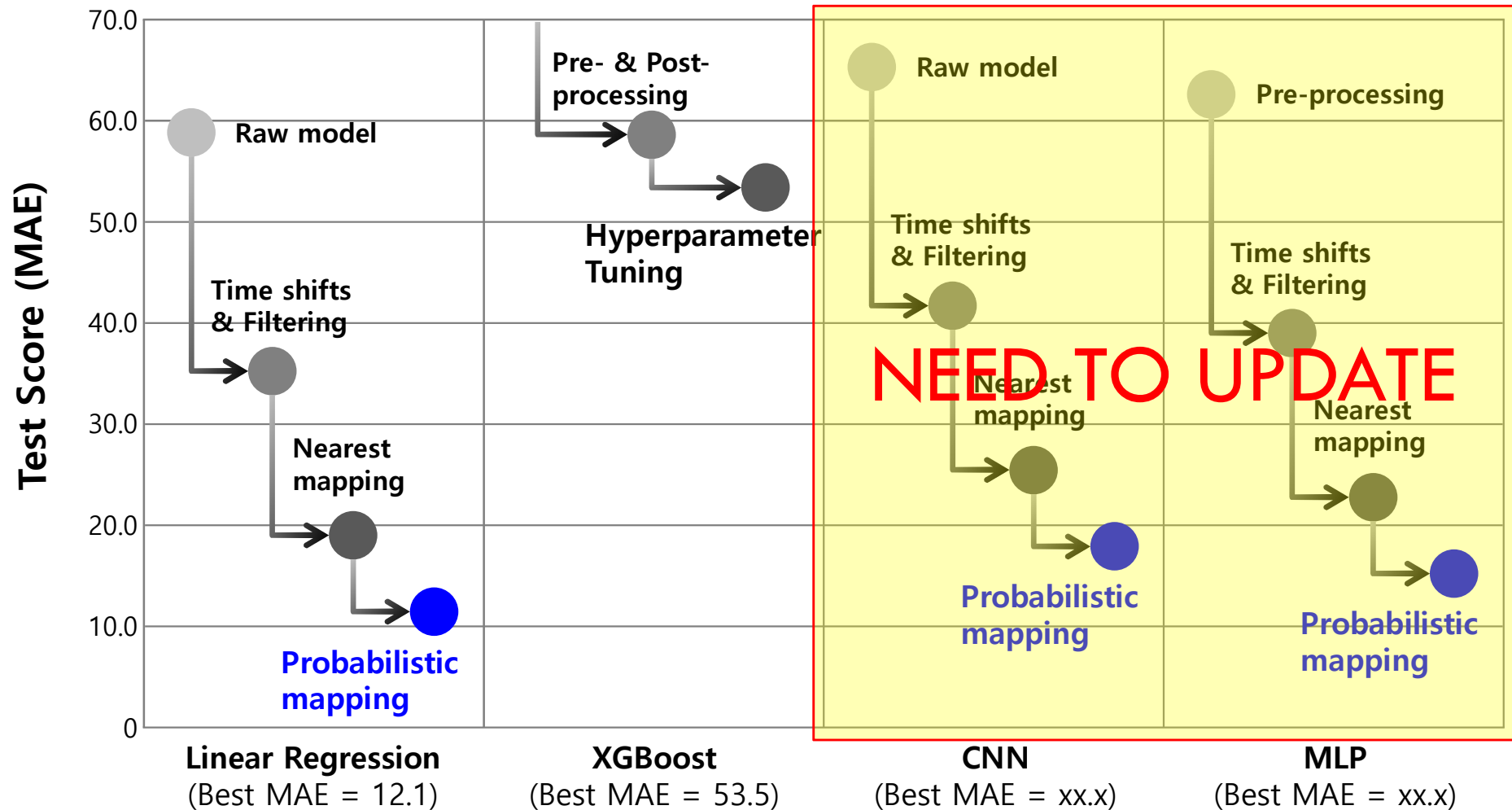
Best strategy



# Pre- & Post-processing (79.5 % MAE reduced w/ LR)



# Summary: (Combined) Models & Results



# Conclusion

## *“No Free Lunch”*

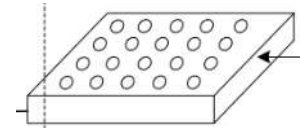
- **Linear regression**
  - ✓ Simple but powerful ( $\because$  highly correlated data)
  - ✓ Limited performance ( $\because$  only linear relationship)
- **XGBoost**
  - ✓ Good for classification (than regression tasks)
  - ✓ Poor at extrapolation (depends on training set)
- **CNN**
  - ✓ Good for data with spatial relationship
- **MLP - Achieved the best performance**
  - ✓ Good for time-series prediction (w/ tabular data)
  - ✓ But too many parameters ( $\because$  full connection)

\*Ref.: Towards Data Science, “Why XGBoost Can’t Solve All Your Problems”

\*Ref.: MachineLearningMastery, “When to use MLP, CNN, and RNN neural networks”

## *Further Works*

- **Different time shifts per sensors (columns)**



	s1	s11	s13	s15
Optimal Time Shift*	4700	2700	2300	2200

\*determined by the maximum cross-correlation

- **Use CNN w/ spatial representation** (e.g., spectrum)
- **Use RNN models - e.g., LSTM, Transformer**
- **Ensemble model**

