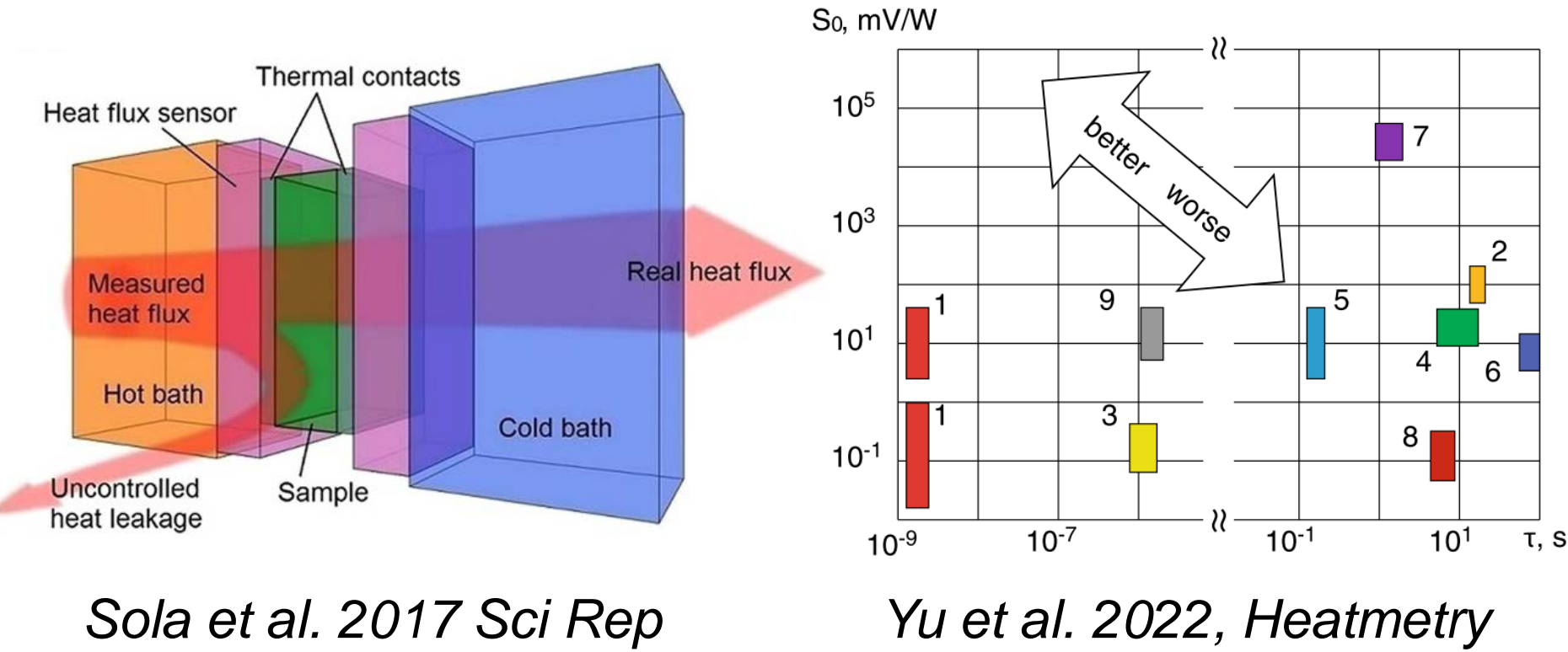


## Background

### Heat Flux Measurement in Power Systems

#### Surface-mounted sensors

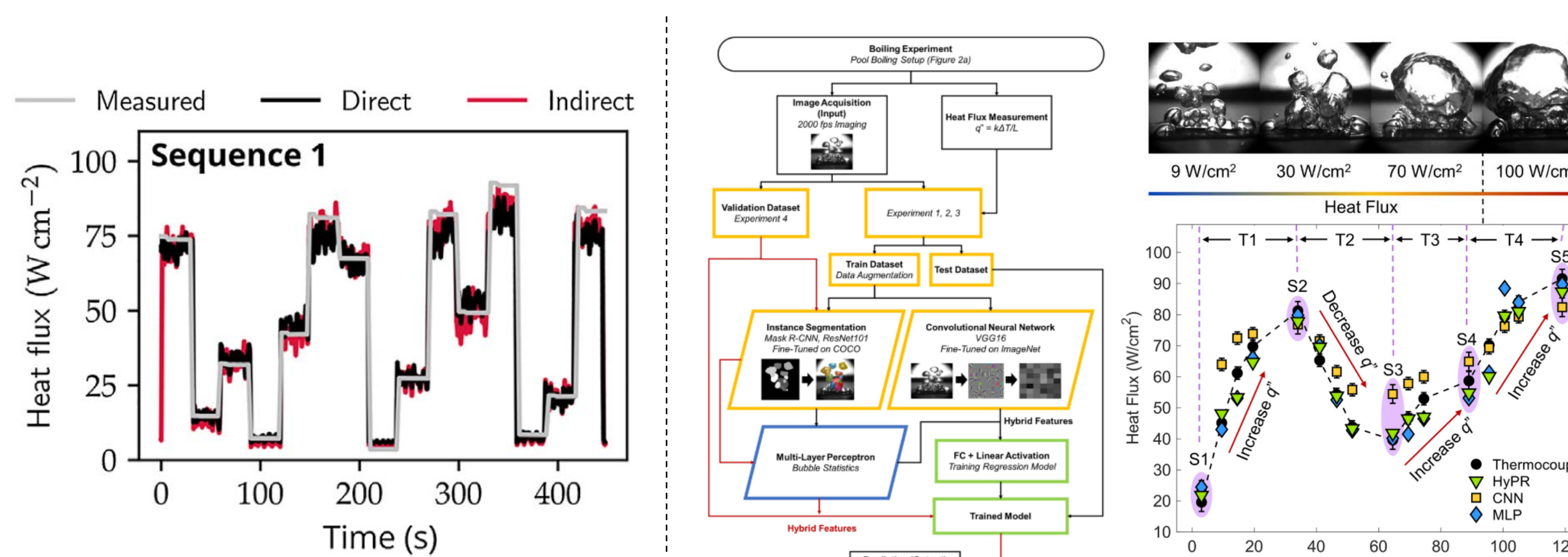
- Transverse thermoelectric effect method
- Temperature gradient method
- Joule heating effect method



Sola et al. 2017 Sci Rep

Yu et al. 2022, Heatmetry

### Non-Intrusive Sensing and Machine Learning

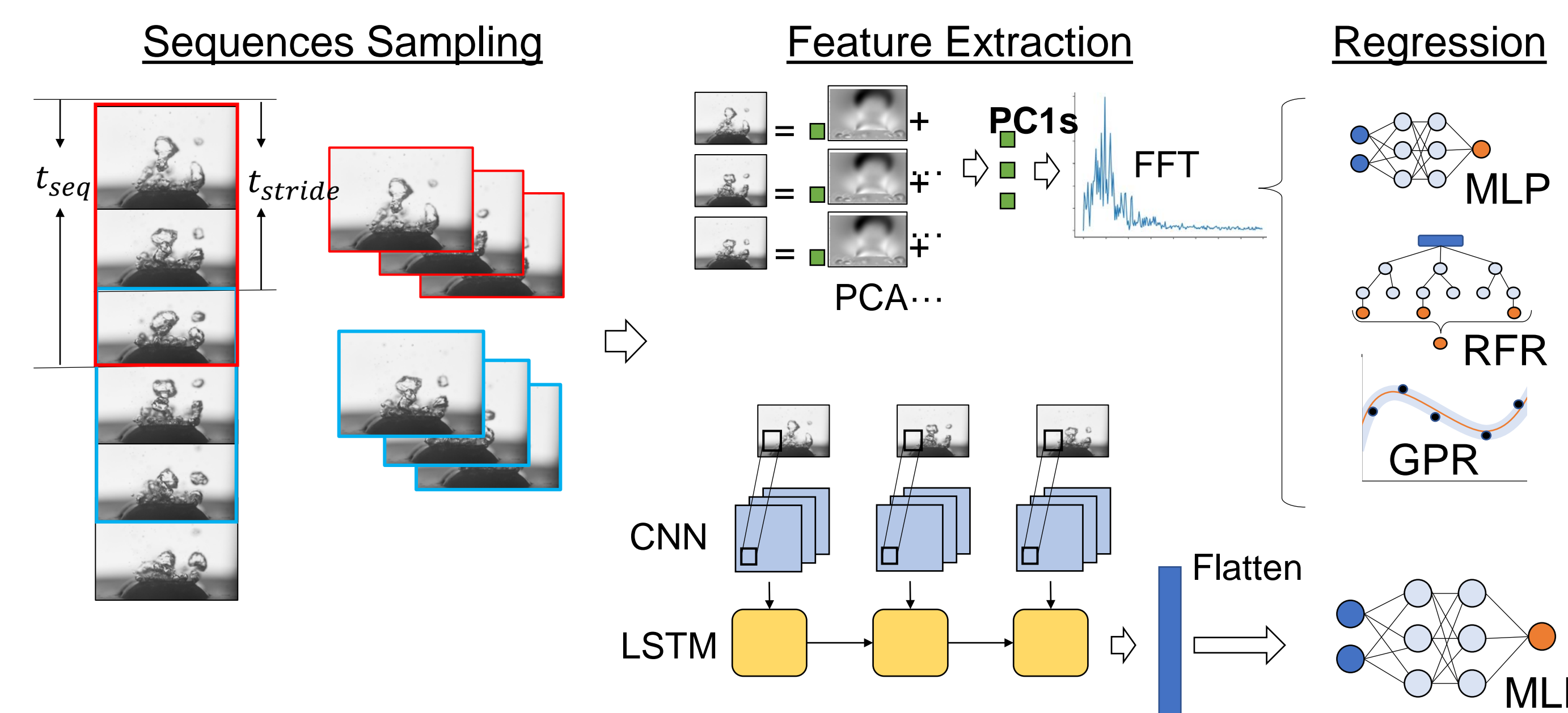


Hobold et al. 2019 Int J Heat Mass Transf

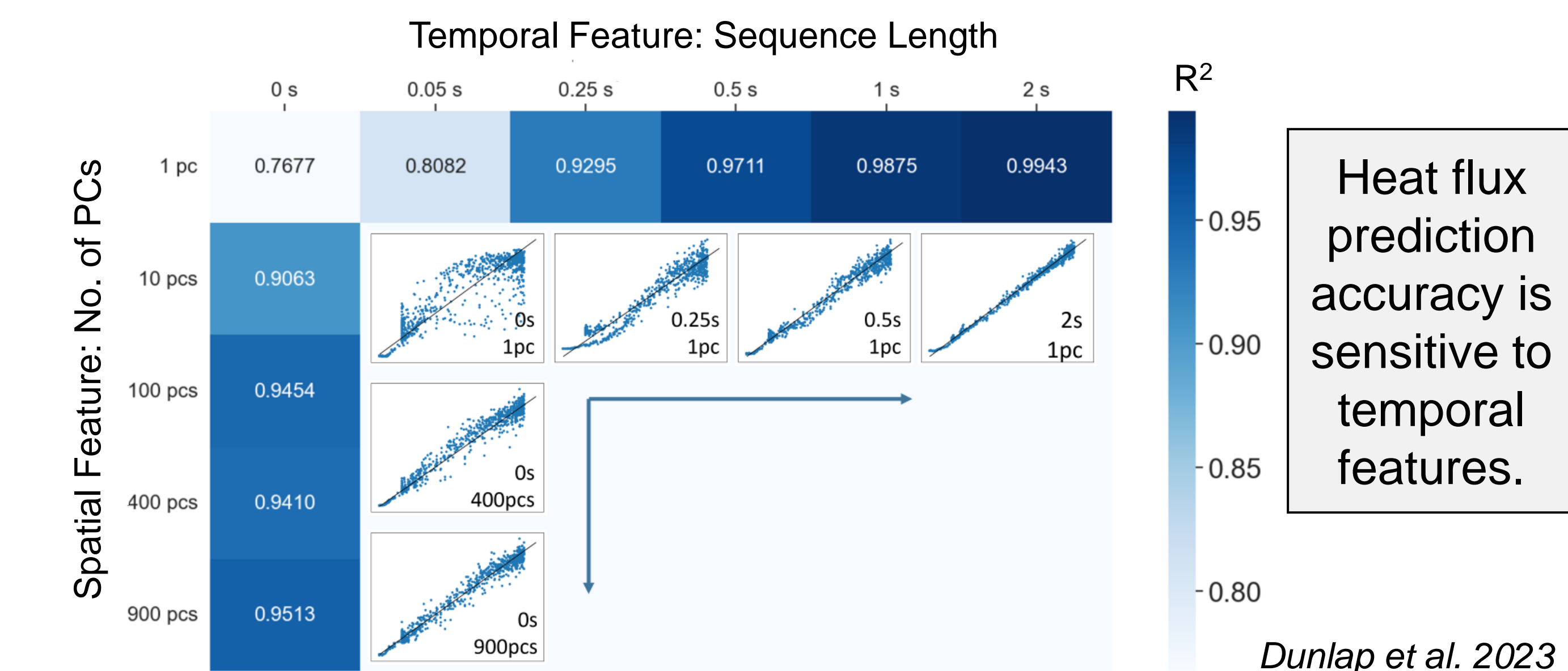
Suh et al. 2021 Sci Rep

## Significance of Temporal Features

### Image Sequence-Based Heat Flux Quantification

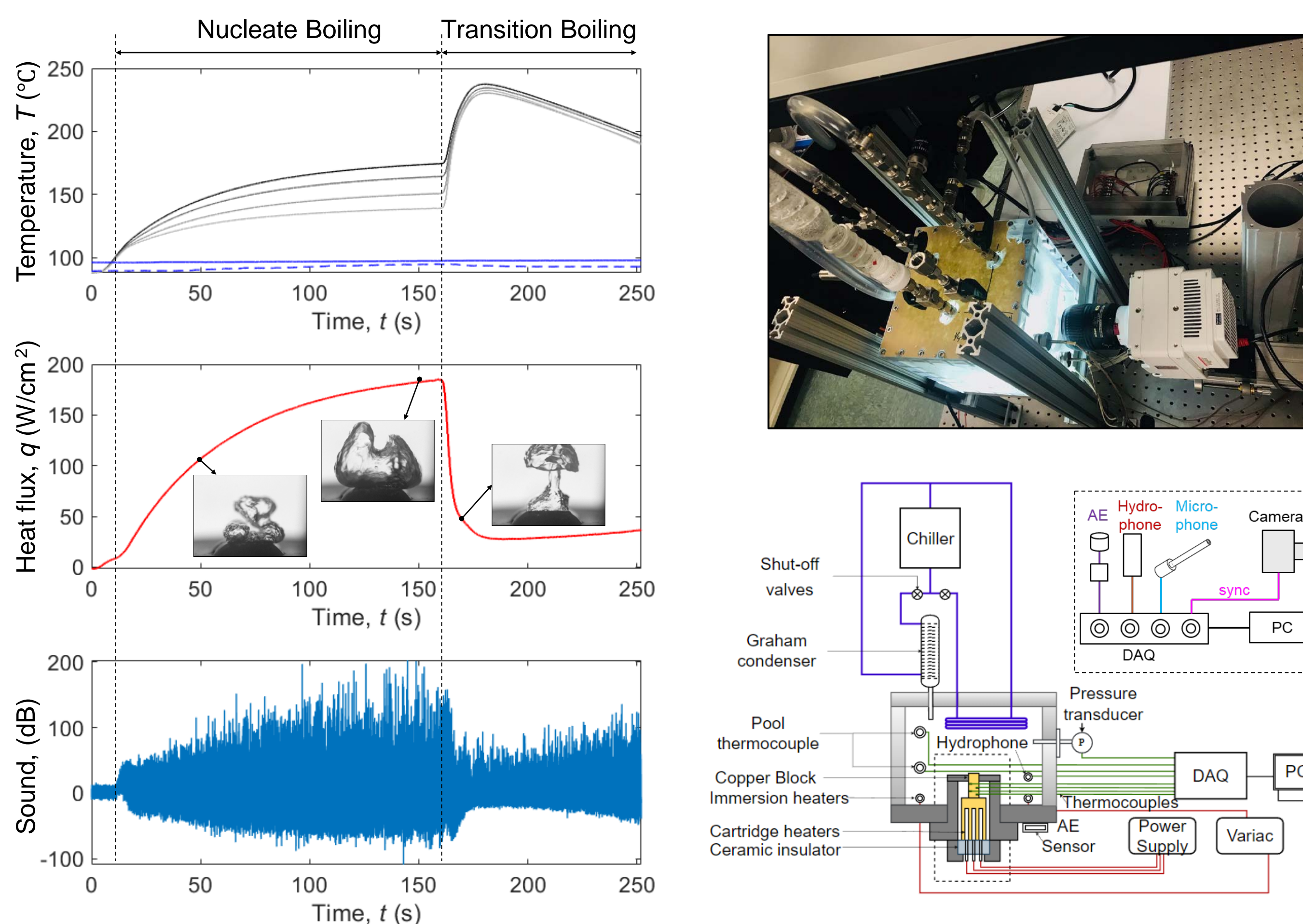


### Effect of Temporal and Spatial Features on Model Accuracy

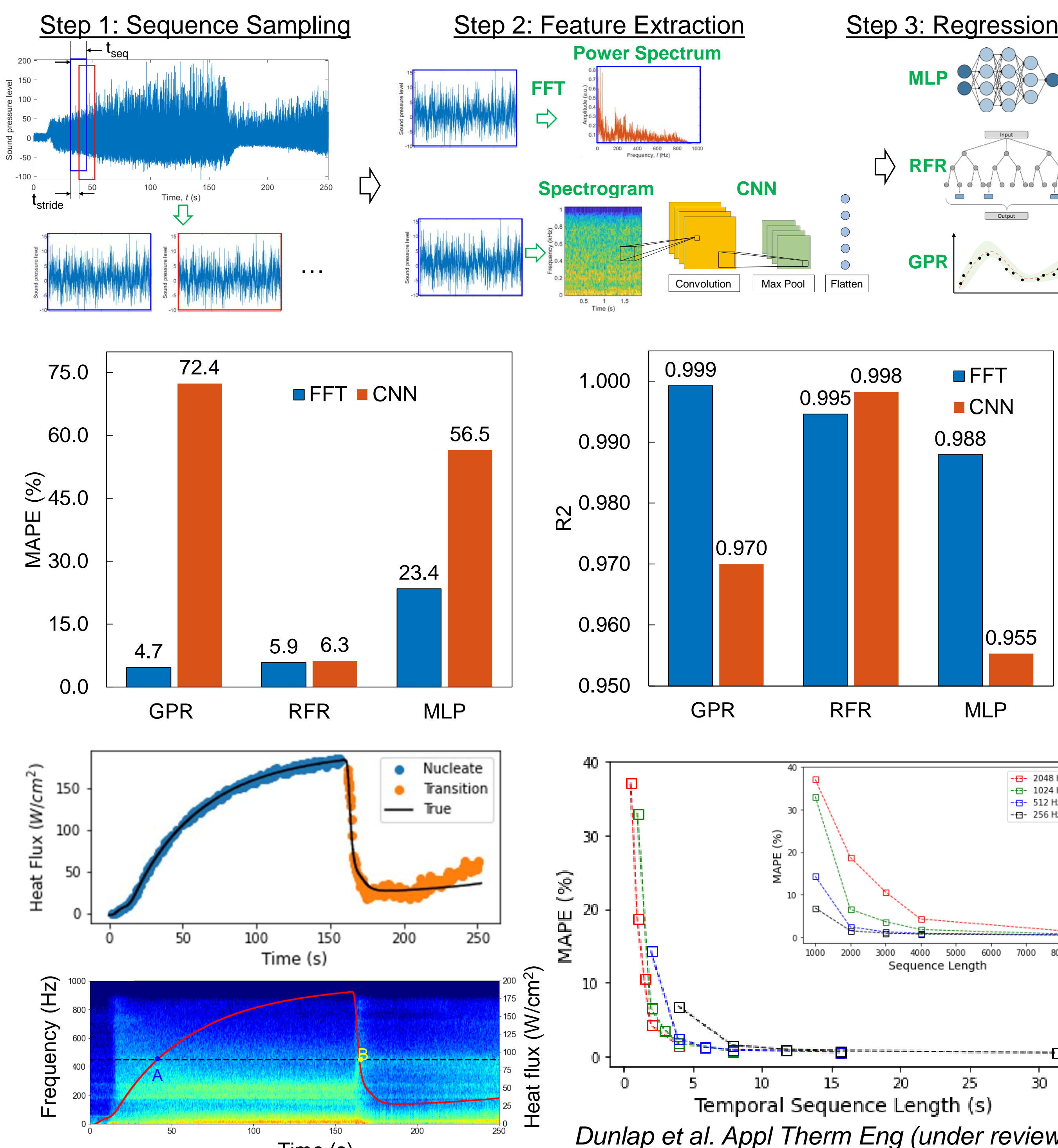


## Acoustic-Base Heat Flux Quantification

### Multimodal Signals from Two-Phase Heat Transfer



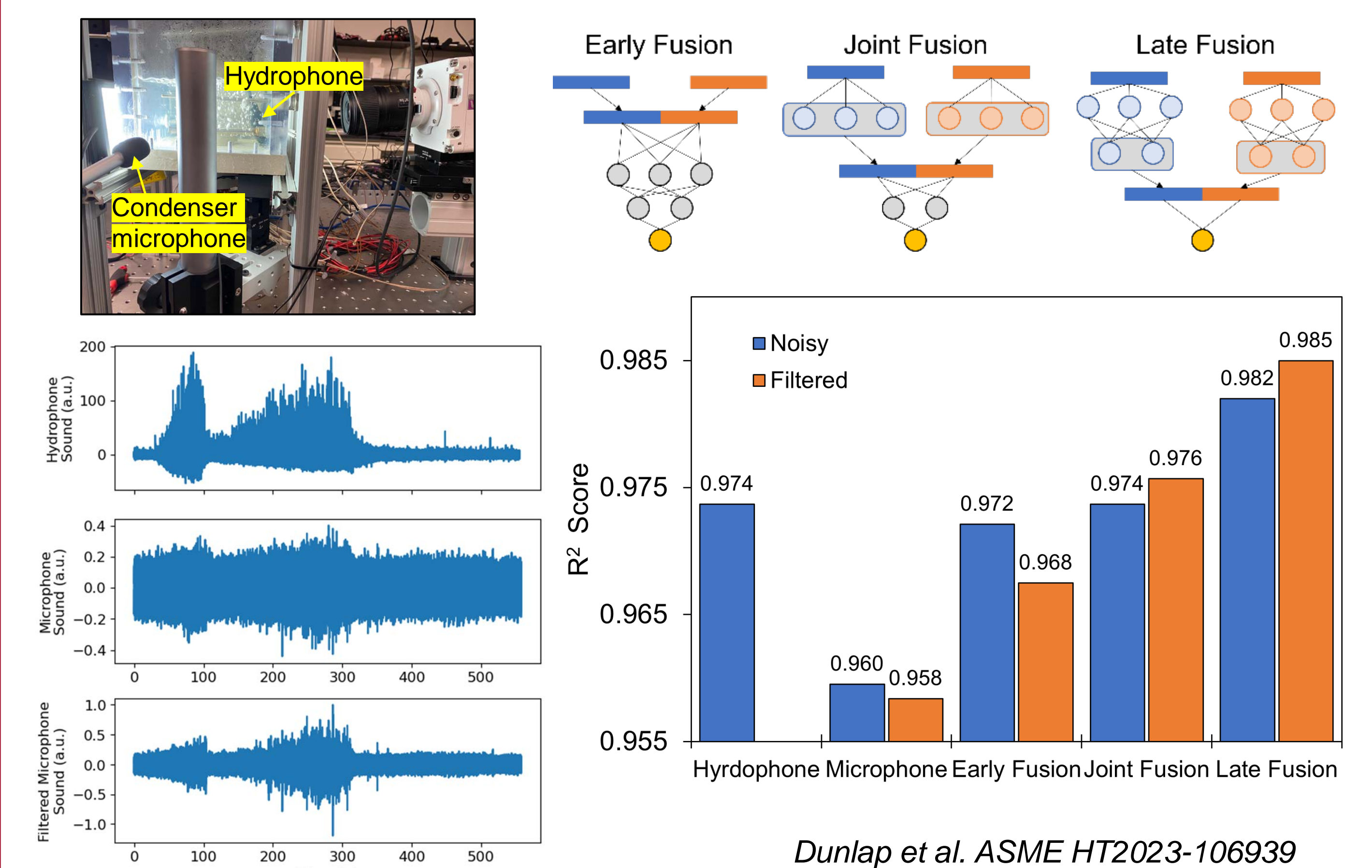
### Acoustic Sequence Sampling and Regression



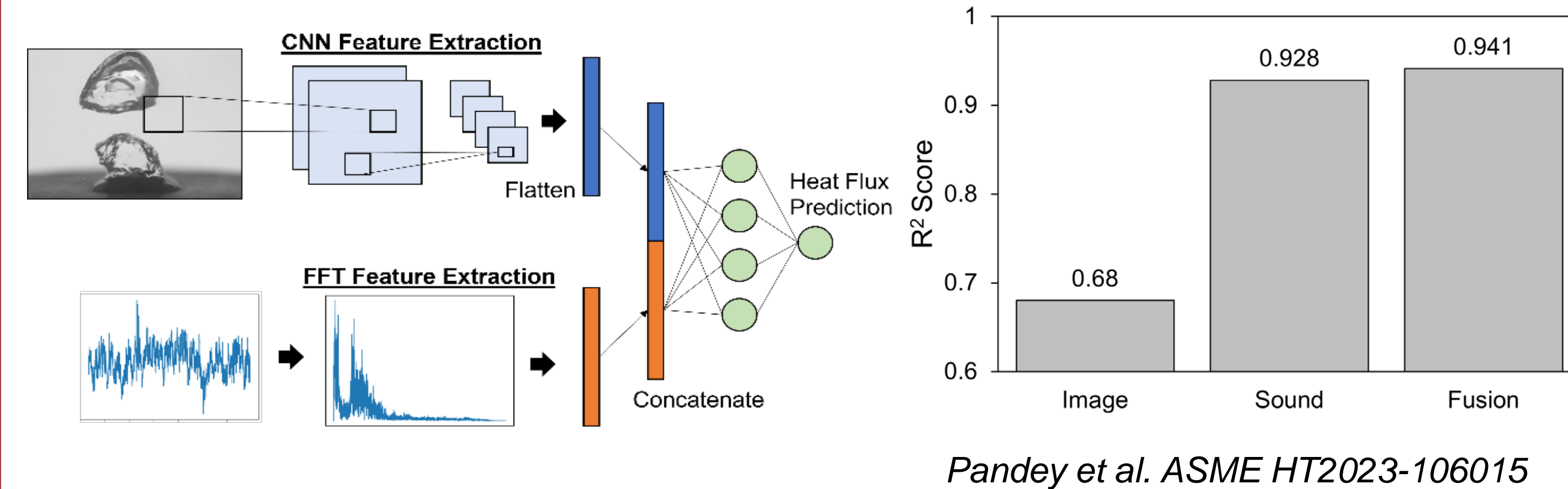
Temporal sequence length dictates prediction accuracy for rate > 256 Hz; acoustic features from 200 - 500 Hz are critical to heat flux prediction.

## Multimodal Sensor Fusion

### Acoustic Sensor Fusion – Hydrophone and Microphone



### Optical-Acoustic Sensor Fusion



Multimodal sensor fusion improves the accuracy of heat flux predictions.

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