

2024 Fall Conference of the KWJS

Prediction of Welding Strength based on Friction Stir Welding (FSW) Tool Temperature and Machine Learning

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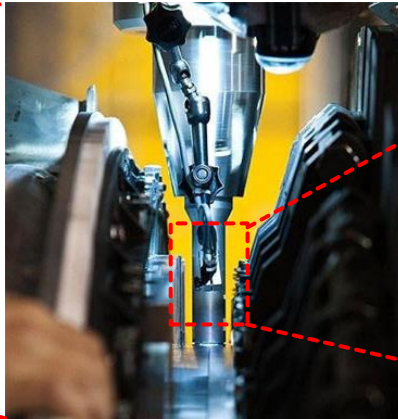
2024-10-18

Research Objective – Prediction of FSW Quality

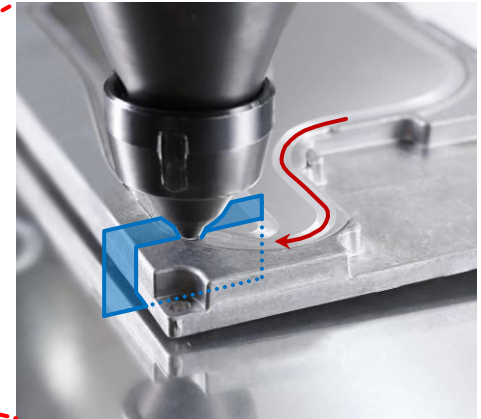
FSW Machine



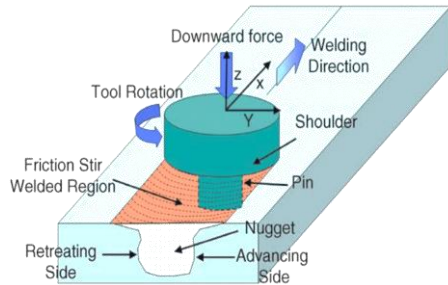
FSW Tool



FSW Joint

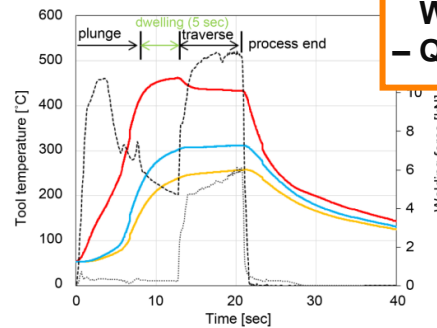


Process Parameters



- Tool rotation speed
- Feed rate
- Downward force

Tool Temperature

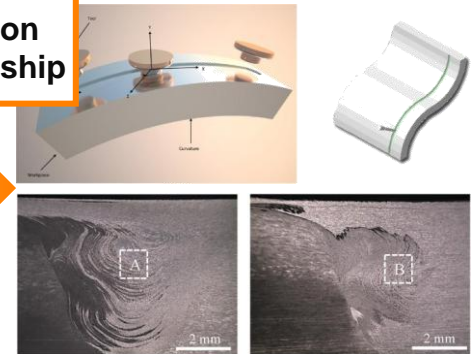


- Transient or stabilized temperature
- @ Probe, shoulder

Welding Condition – Quality Relationship



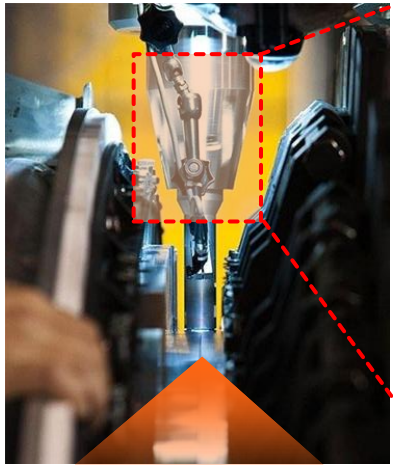
Welding Quality



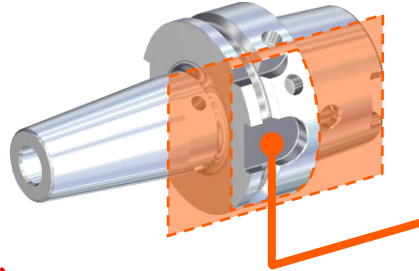
- Tensile strength
- Internal microstructure (pore size, fraction)

Approach

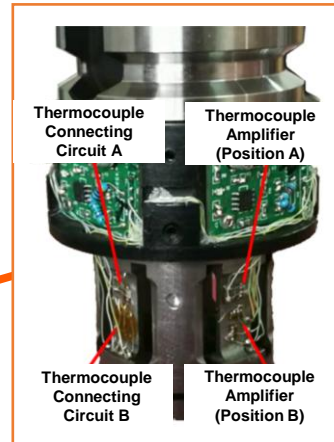
FSW Tool



Smart Tool Holder

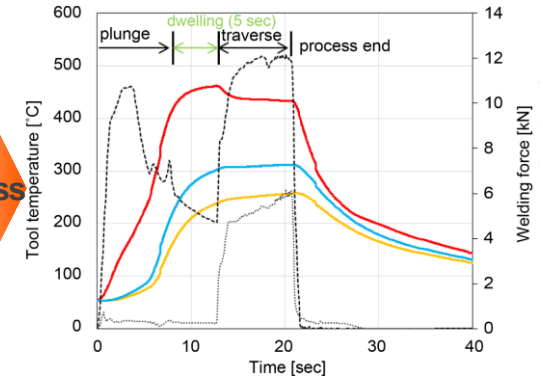


Tool temperature measurement



Wireless

Welding condition monitoring (Tool temperature)



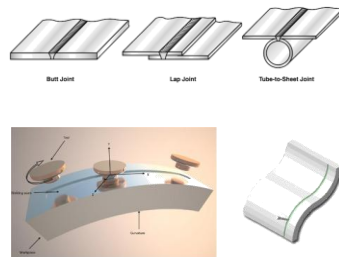
Welding cond.-Quality ML model



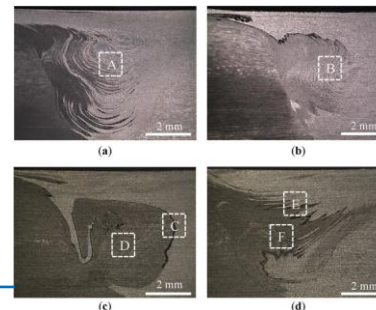
FSW Product



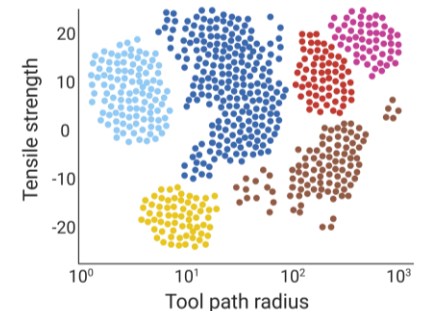
Welding geometry



Microstructures



Welding quality

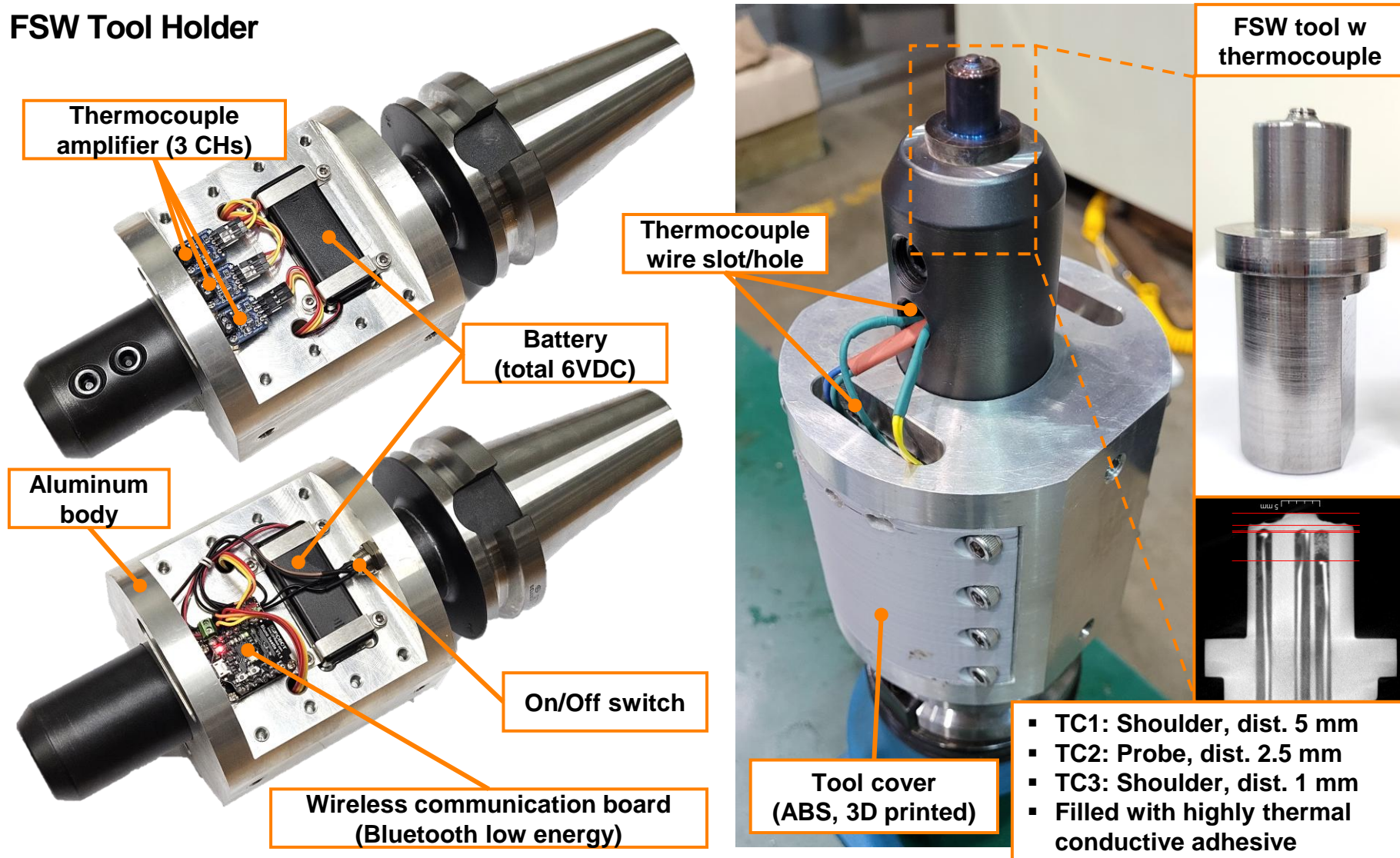


Physical Domain

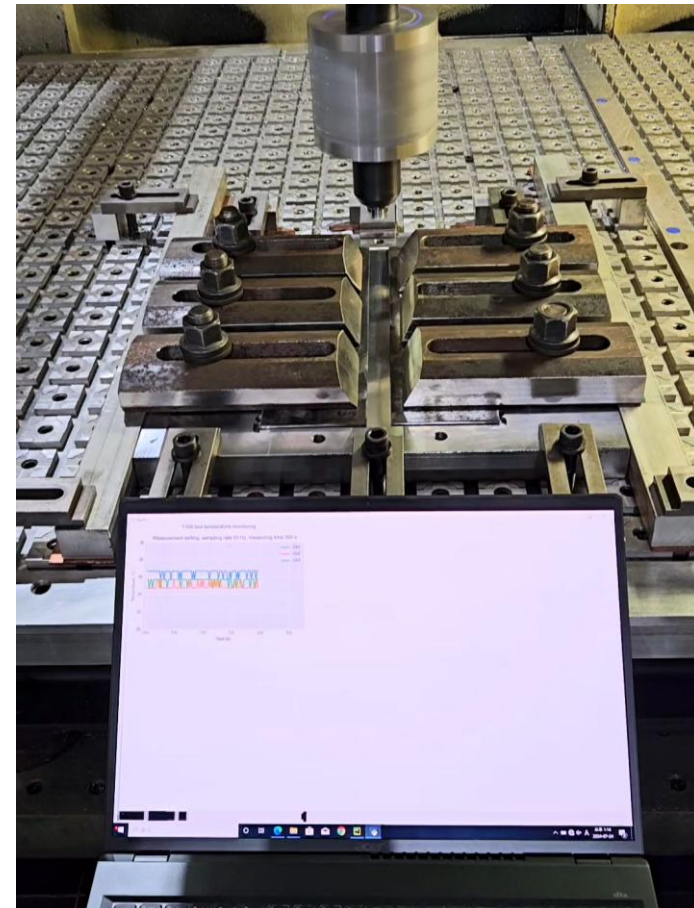
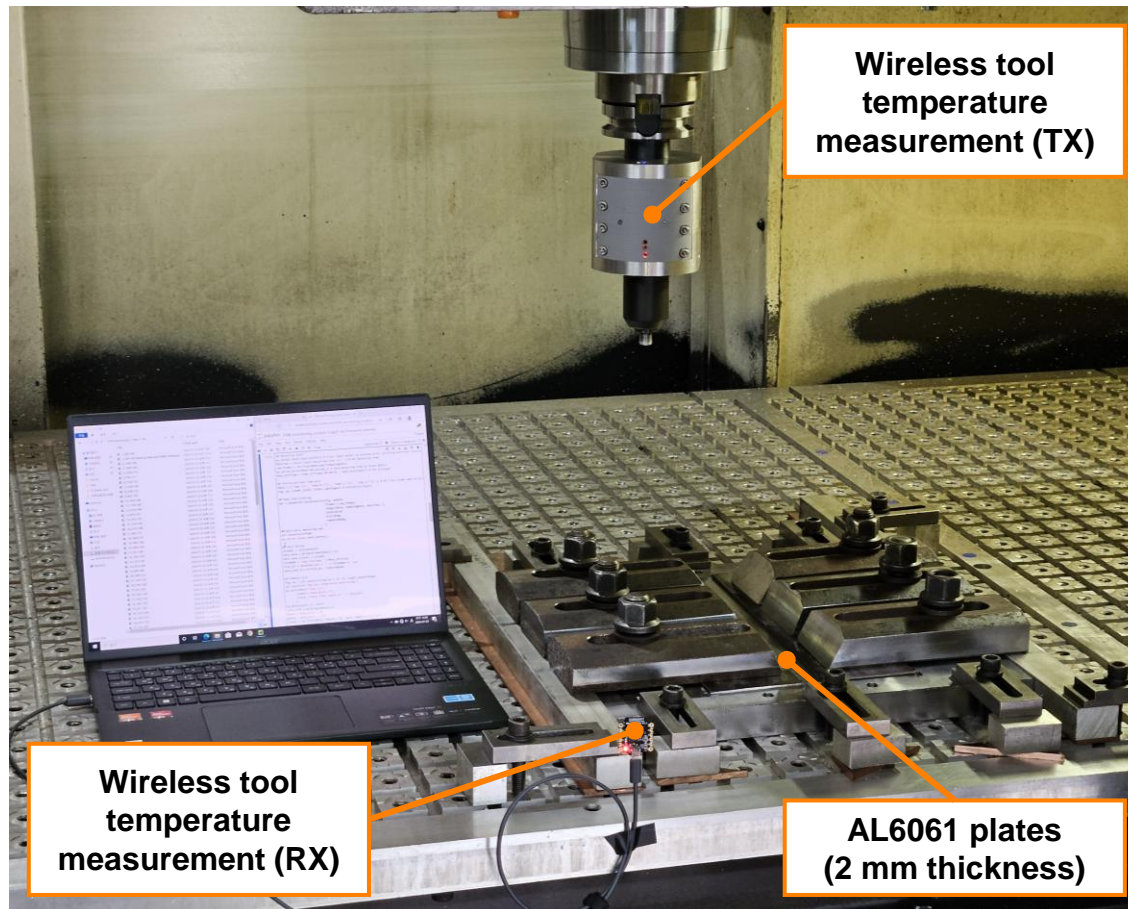
Digital Domain

Wireless FSW Tool Temperature Measurement

FSW Tool Holder



FSW Tool Temperature Measurement Setup

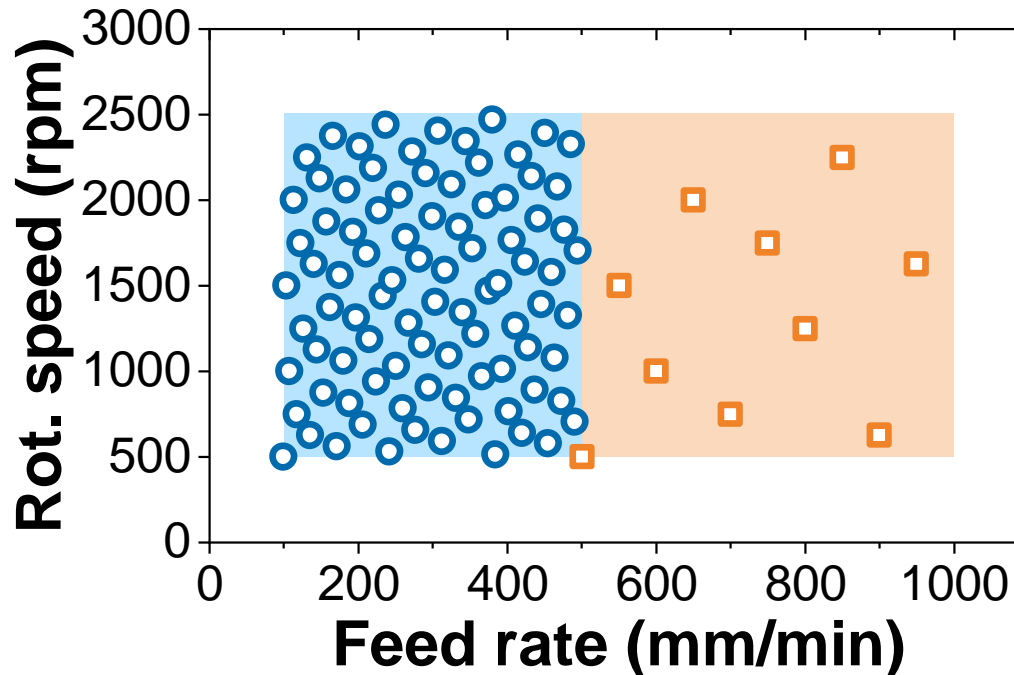


- Realtime tool temperature measurement
- Sampling rate 100 Hz/Ch, Operation time >4 hr
- No mechanical and electrical issue @ tool rotation speed ≤ 2500 rpm

FSW Temperature-Quality Data Preparation

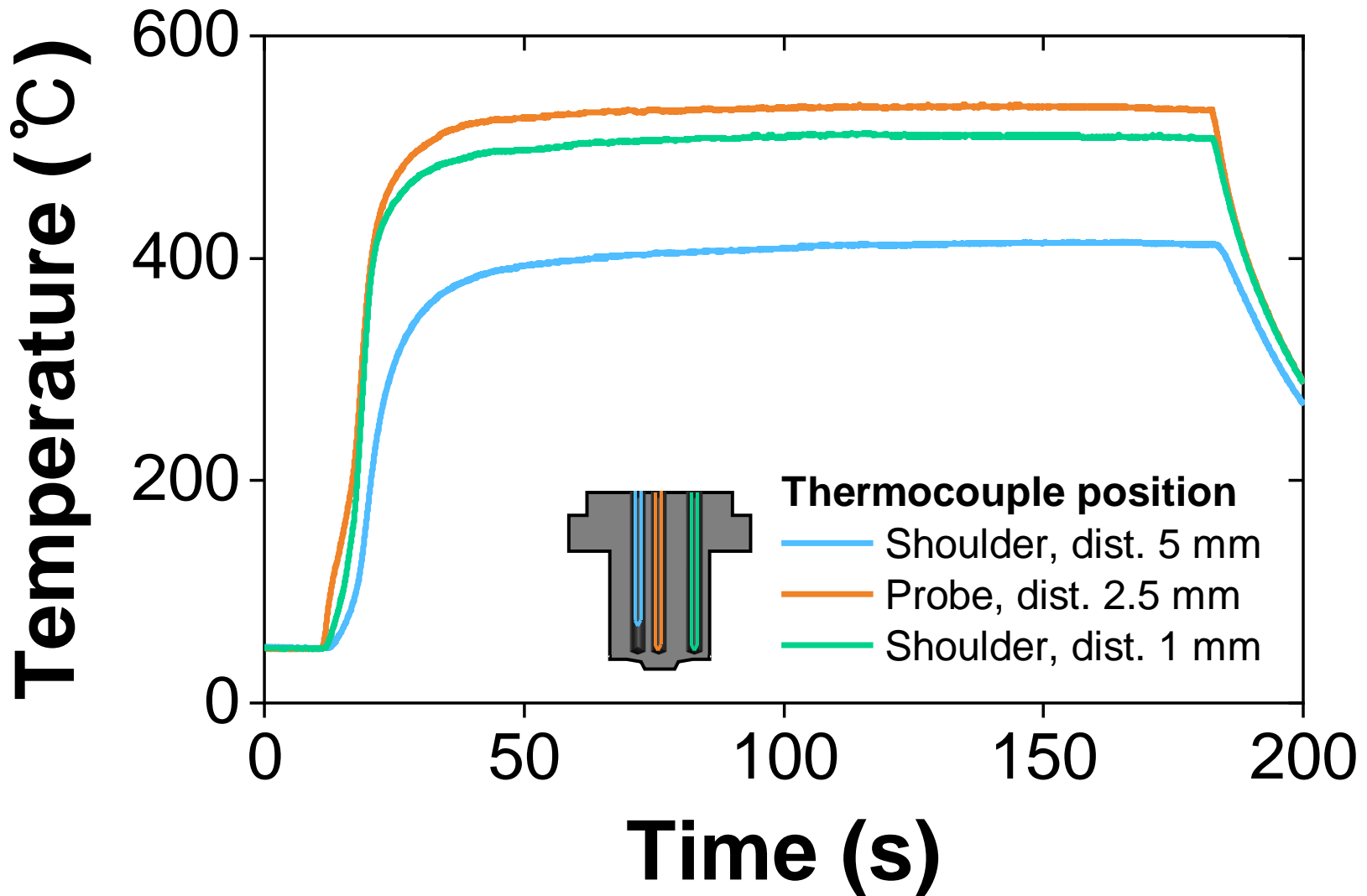
Design of Experiment (DoE)

- Process parameter: Tool rotation speed [500, 2500], Feed rate [100, 1000]
- DoE by *Hammersley Sequence Sampling (HSS), Total 100 samples
- Two HSS domain: Main / Extended
 - Main: 90 samples / Extended: 10 samples



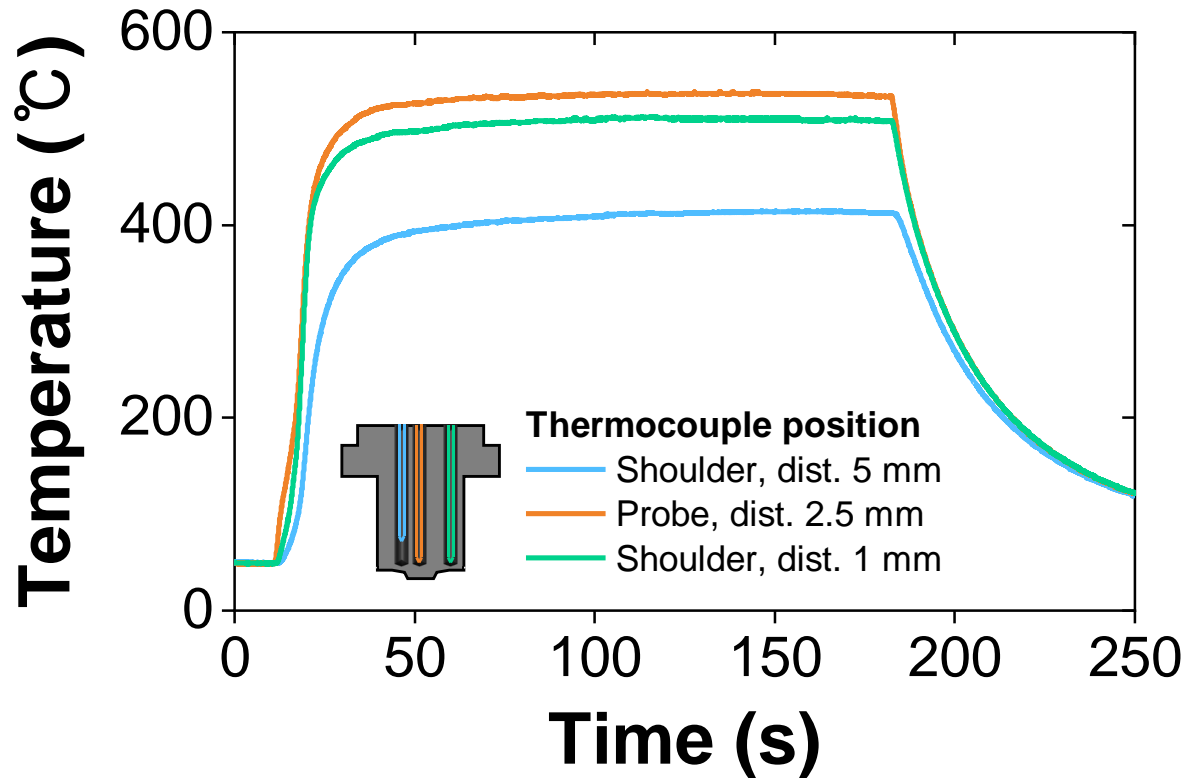
* Better DoE method for ML modeling than factorial or Latin hypercube sampling (LHS) due to minimized overlapping experimental conditions and low correlations between each experimental variable (Das S., Tesfamariam S., arXiv.2202.06416, 2022.)

Tool Temperature Profile at Different Positions

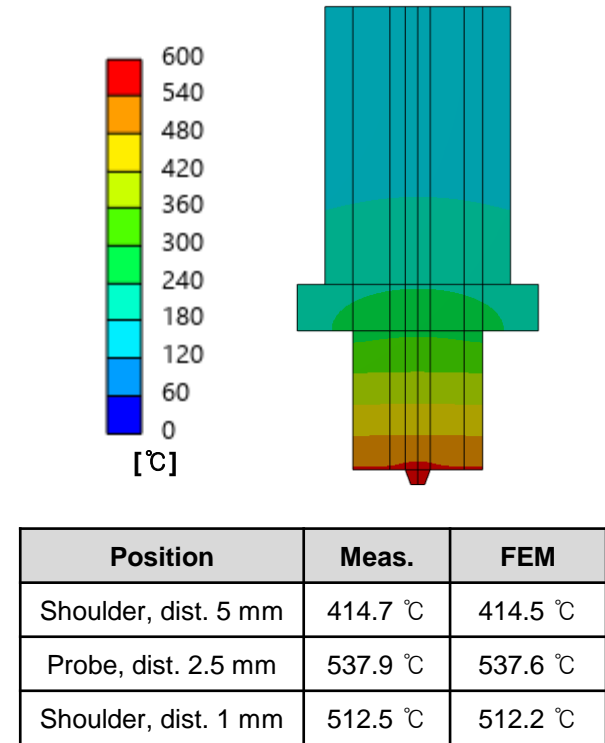


Tool Temperature Profile at Different Positions

Transient temperature profiles at different positions in FSW tool



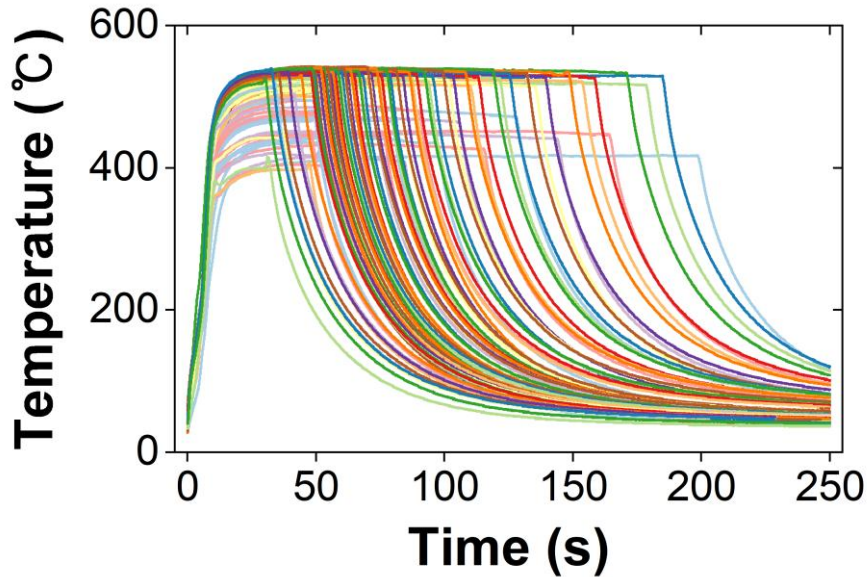
FEM thermal analysis



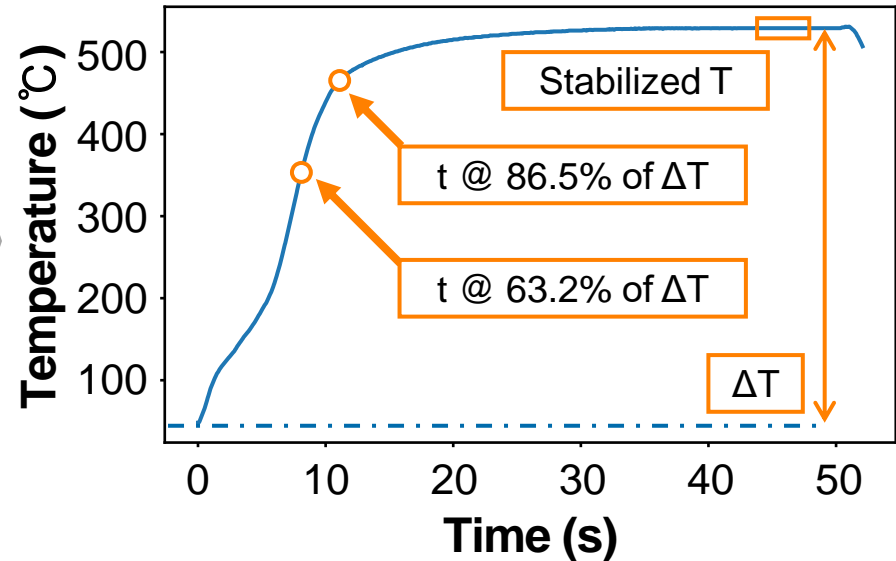
- Small difference to radial direction (measurement & FEM)
→ Only temperature measurement at probe is sufficient
(Probe, dist. 2,5 mm Vs. Shoulder, dist. 1 mm → Almost same depth)

FSW Temperature-Quality Data Preparation

Probe temperature at 100 different condition



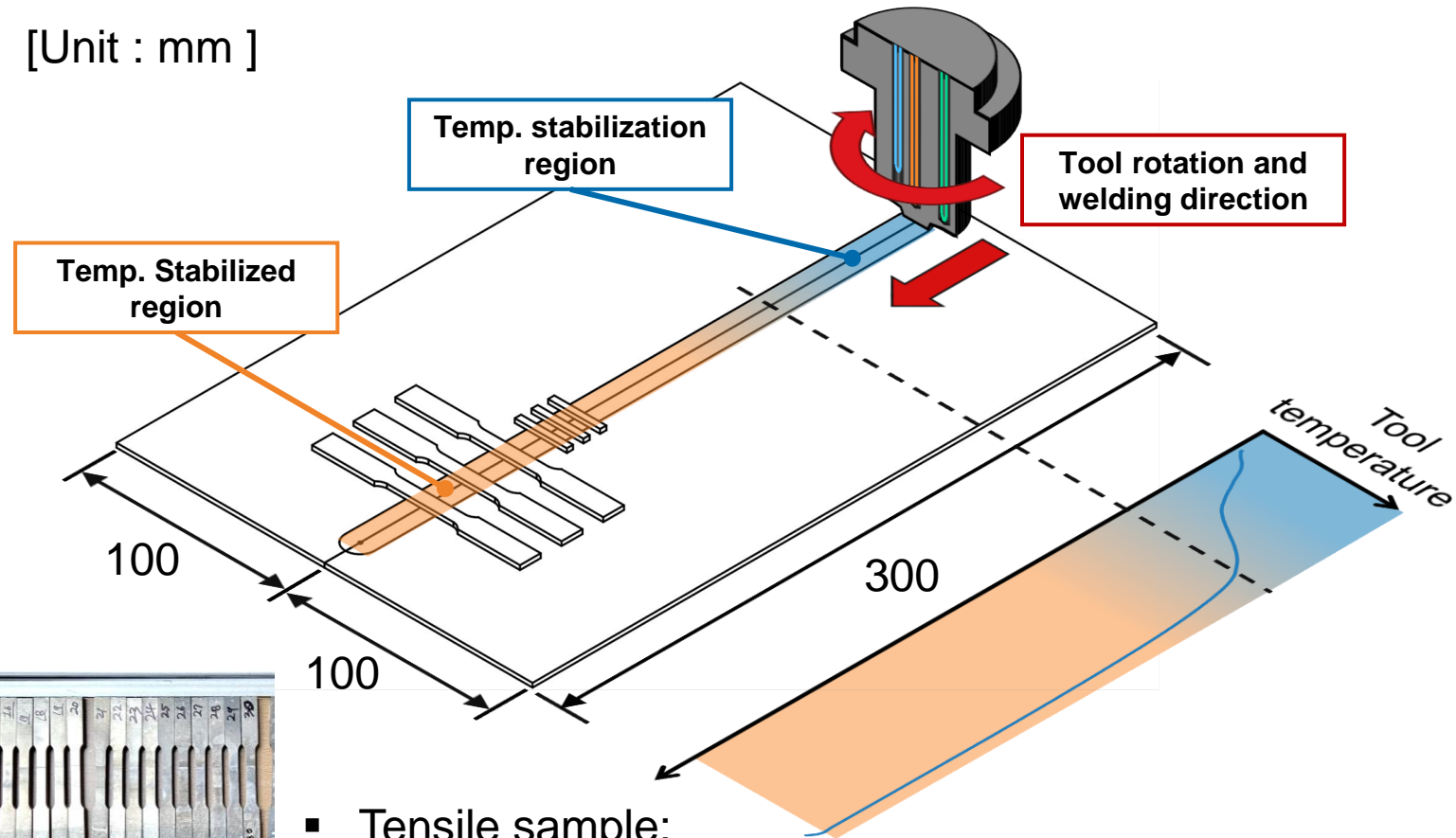
Feature extraction (e.g., 1578 rpm, 460 mm/min)



- Feature extraction by geometrical feature of the temperature profiles
- 1. Temperature after stabilization
- 2. Time at 63.2% of ΔT ($\sim 1 \tau$; Time constant)
- 3. Time at 86.5% of ΔT ($\sim 2 \tau$; Time constant)
- More features will be determined.

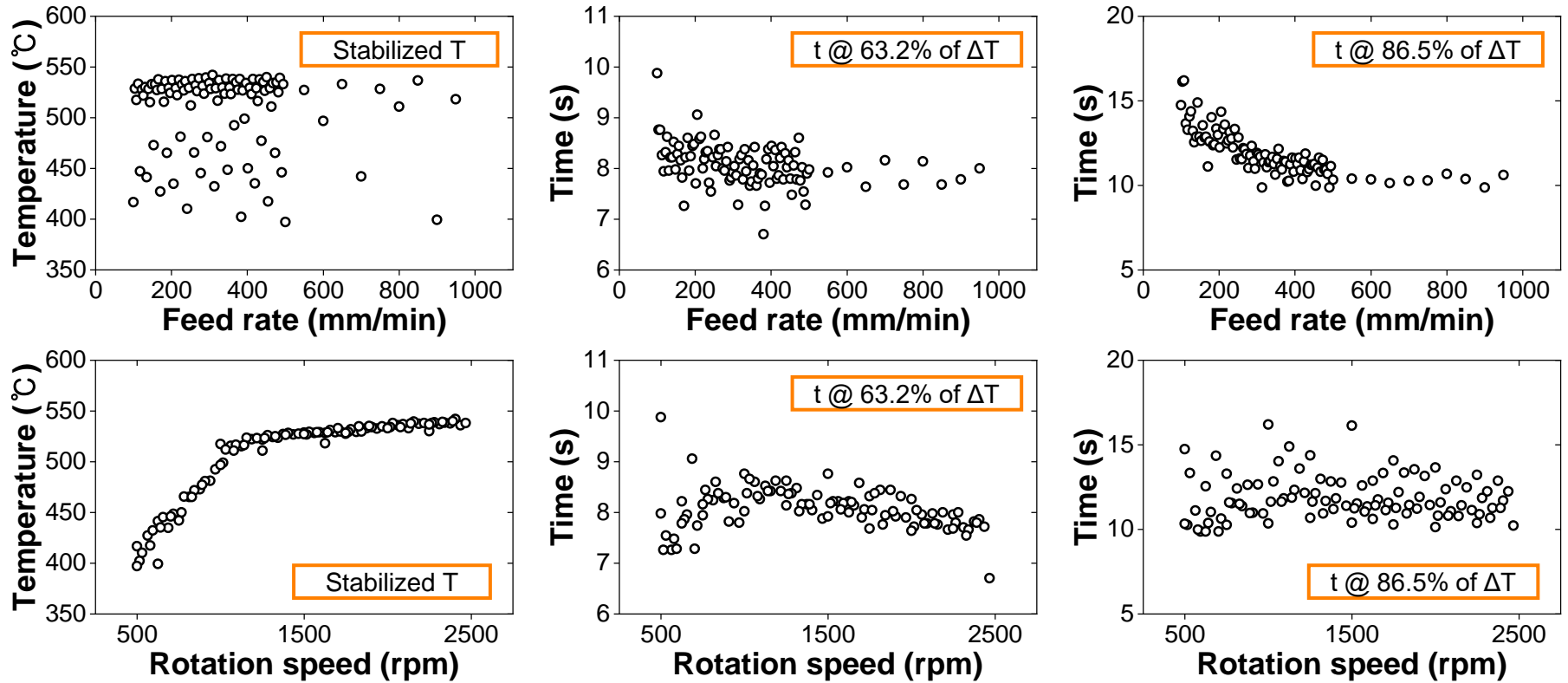
FSW Quality Measurement Sample Preparation

[Unit : mm]

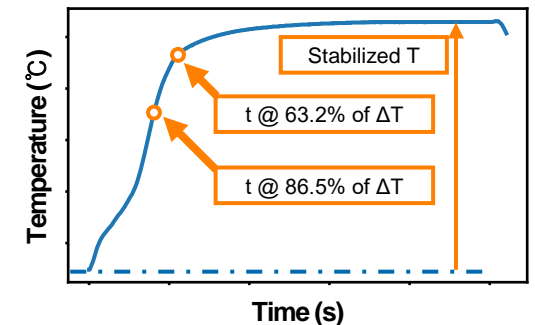


- Tensile sample:
ASTM E8/E8M-13a, 3 sample/condition
- Sample for microstructure analysis:
5x26x2 mm, 3 sample/condition
- All samples were from the temp stabilized region.

FSW Process Parameter – Feature Correlation

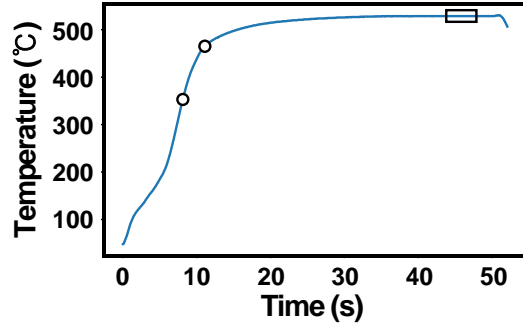


- Strong correlation between
 - a) Tool rotation speed – Stabilization temperature
 - b) Feed rate – Time at 63.2% of ΔT
 - c) Feed rate – Time at 86.5% of ΔT



Tool Temperature – Welding Strength ML Model

FSW tool temperature



Input Data Condition

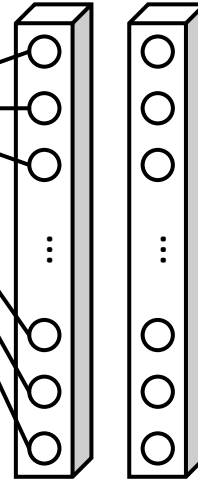
$t_{\Delta T 63.2\%}$

$t_{\Delta T 86.5\%}$

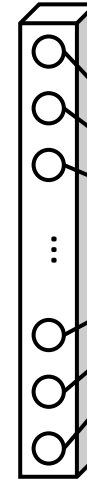
$T_{\text{Stabilized}}$



Fully connected hidden layers



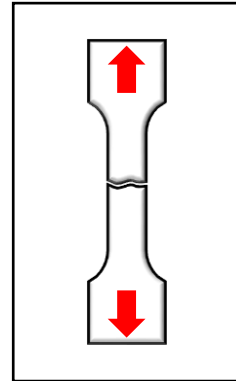
...



Single-task



Part quality



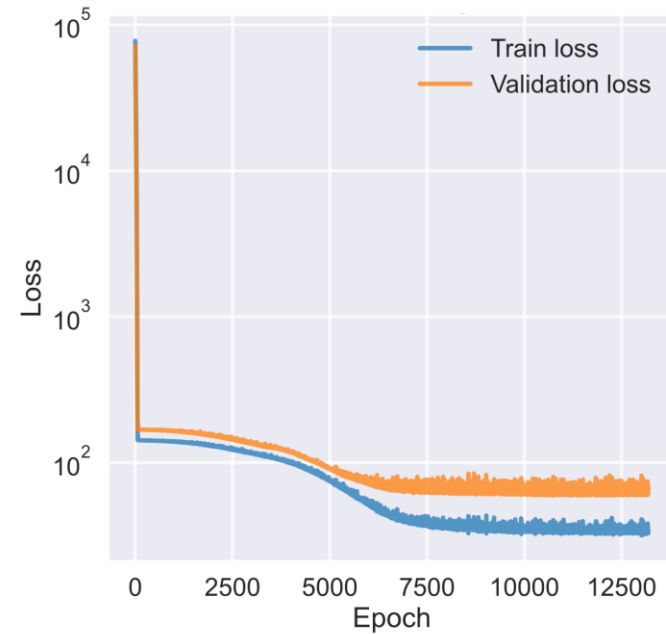
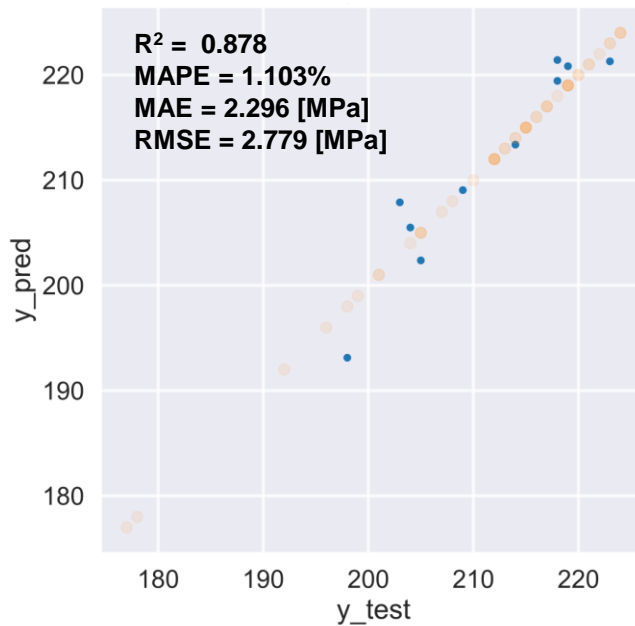
Tesile Strength

| Hyperparameter | Range | Optimized |
|--------------------------------|---|---------------|
| # of hidden layers | 1, 2, 3 | 3 |
| # of neuron per a hidden layer | 8, 16, 32 | 16 |
| Learning rate | 0.1, 0.01, 0.001 | 0.01 |
| Activation | ReLU, gelu, elu, swish, softplus | elu |
| Initializer | glorot_normal, glorot_uniform, he_normal, he_uniform, lecun_normal, lecun_uniform | Lecun_uniform |
| Optimizer | Adam, Adadelata, Adamax | Adamax |

* Hyperparameter optimization (HPO) by Hyperband algorithm with Tree-structured Parzen Estimator (TPE) and pruning.

* Early stopping and restoration of the best model were applied for main training after HPO

Prediction Performance of ML Model



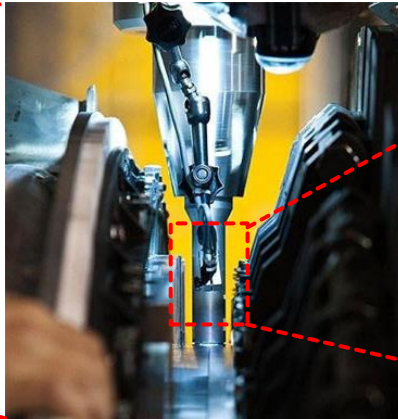
| Measured Tensile Strength (MPa) | Predicted Tensile Strength (MPa) | Error (%) | Measured Tensile Strength (MPa) | Predicted Tensile Strength (MPa) | Error (%) |
|---------------------------------|----------------------------------|-----------|---------------------------------|----------------------------------|-----------|
| 198 | 193.12 | 2.46 | 214 | 213.37 | 0.29 |
| 203 | 207.89 | 2.41 | 223 | 221.29 | 0.77 |
| 205 | 202.37 | 1.28 | 218 | 221.42 | 1.57 |
| 209 | 209.05 | 0.02 | 219 | 220.83 | 0.84 |
| 204 | 205.5 | 0.73 | 218 | 221.42 | 1.57 |

Research Objective – XAI Analysis

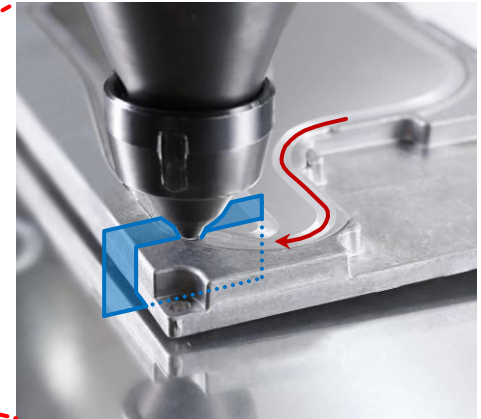
FSW Machine



FSW Tool

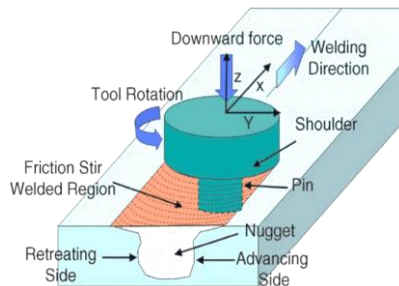


FSW Joint

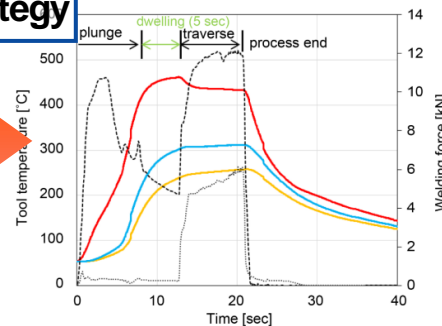


Process Parameters

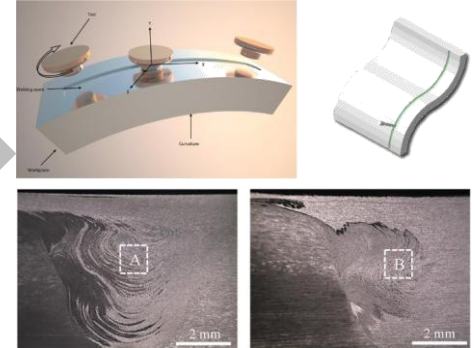
FSW process control strategy



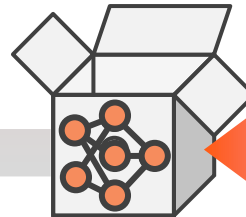
Tool Temperature



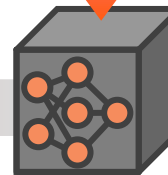
Welding Quality



- Influential features (Time, temperature)
→ Specific quality-targeted monitoring window, ML model improvement



XAI



Black box (Opaque characteristics)

Conclusions

- The developed FSW tool module was capable of measuring tool temperature to accurately predict the ultimate strength in FSW.
- Analysis of the measured tool temperature profiles presented that rotational speed had a greater influence on temperature than feed rate.
- The tool temperature profile features were used as input feature for the ANN to predict ultimate tensile strength.
- Overall prediction performance of the ANN model was $\geq 98\%$ accuracy (1-MAPE).

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KITECH-AU Joint International Cooperation Research
Weld Joint Analysis for Quality Prediction of Friction Stir Welding
(PI: Chanhoo Lee)

KITECH-Auburn Univ. Manufacturing Technology Innovation Center
(KAMTIC)



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