Deep NN in acoustic emission classification and hysteresis analysis

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Experiment 1

4 sensors

- Continuous acoustic emission
- ▶ 5 levels of sharpness
- ▶ 5 separate holes
- Shorter sections of signal sampled



Figure: Drilling setup.

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Experiment 1

- Dimensions of final dataset: (12500, 4, 6000)
- Train:Validation:Test split ratio = 7:2:1



Figure: Example of AE signal for new and very blunt drill.

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Experiment 2

- 4 sensors
- Continuous AE
- ▶ 8 load weight levels
- Constant spinning speed
- Shorter samples
- Dataset of size (16000,4,5000)



Figure: Ball bearing loading setup.

NN architectures - 1. Conv2Net

- Based on LeNet-5
- 2 convolutional layers with filters of size 501 a 81
- ReLU activation
- Use of max-pooling



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NN architectures - 2. ResNet-16

- Based on ResNets (2015)
- 4 residual blocks

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- Each block contains 4 convolutional layers
- BatchNorm a ReLU activation
- max-pooling between blocks

$$\begin{pmatrix} l_{i} \\ N_{i} \end{pmatrix} \stackrel{d}{=} \sim \begin{array}{c} l_{i} \\ N_{i} \\ N_{i} \end{pmatrix} \stackrel{d}{=} \begin{pmatrix} l_{i} \\ N_{i} \end{pmatrix} \stackrel{d}{=}$$



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NN architectures - 3. InceptionTime

- 3 IncTime blocks with a residual connection
- Adjusted filter lengths 41, 101, 301
- Flattening using avg-pooling
- 9 IncTime blocks in total



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NN architectures - 4. Pooled InceptionTime

- Adjustment of IncTime block: avg-pooling to lower the length of processed signals
- Increasing number of channels, decreasing length of filters



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NN architectures - 5. TSiT

- Transformer based on ViT (tsai)
- Signal divided into 1D patches (4x1x40)
- Classification head only on [class] token



Vision Transformer (ViT)

Results AE - Experiment 1

Table: Classification accuracy on data from 1. experiment (Drill)

	Network type				
	Conv2Net	ResNet-16	InceptionTime	Pooled IncT	TSiT
Mean accuracy	80,4%	81,5%	83,2%	86,4%	56,2%
Std dev	1,96	1,57	4,73	2,15	1,11



Figure: Confusion matrix - Pooled Inception Time.

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Results AE - Experiment 2

Table: Classification accuracy on data from 2. experiment (Bearings).



Figure: Confusion matrix - Pooled Inception Time.

Results AE - Experiment 2, regression approach

 Pooled Inception Time network adjusted for continuous prediction of load weight (MAE = 1,05kg)



Figure: Boxplots of load weight predictions.

Results AE - Length of sampled signals



Figure: Classification accuracy using Pooled Inception Time depending on the length of cut-out signals.

- Hysteresis memory efect in a dynamic system
- Modeled using Preisach-Mayergoyzova model
- We assume a mixture distribution on PM space



Figure: Synthetic PM space with Guyer1 dominant distribution.

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- Synthetic dataset 8 types of distributions
- ▶ PM space + fixed load profile \rightarrow hysteresis curve
- Goals:
 - Classification of hyst. curves by dominant distribution
 - Prediction of mixing parameters λ_i



Figure: Normed load profile and its corresponding hysteresis curve originating in PM space with dominant Laplace distribution.

- Inception Time with filters of length {10, 20, 40}
- Accuracy of classification by dominant distribution: 69,3%
- In regression setting, predicting λ : MAE = 0,075



Figure: Final MSE and MAE for individual predicted coefficients.

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Figure: Confusion matrix - classification using Inception Time.

Thanks for attention