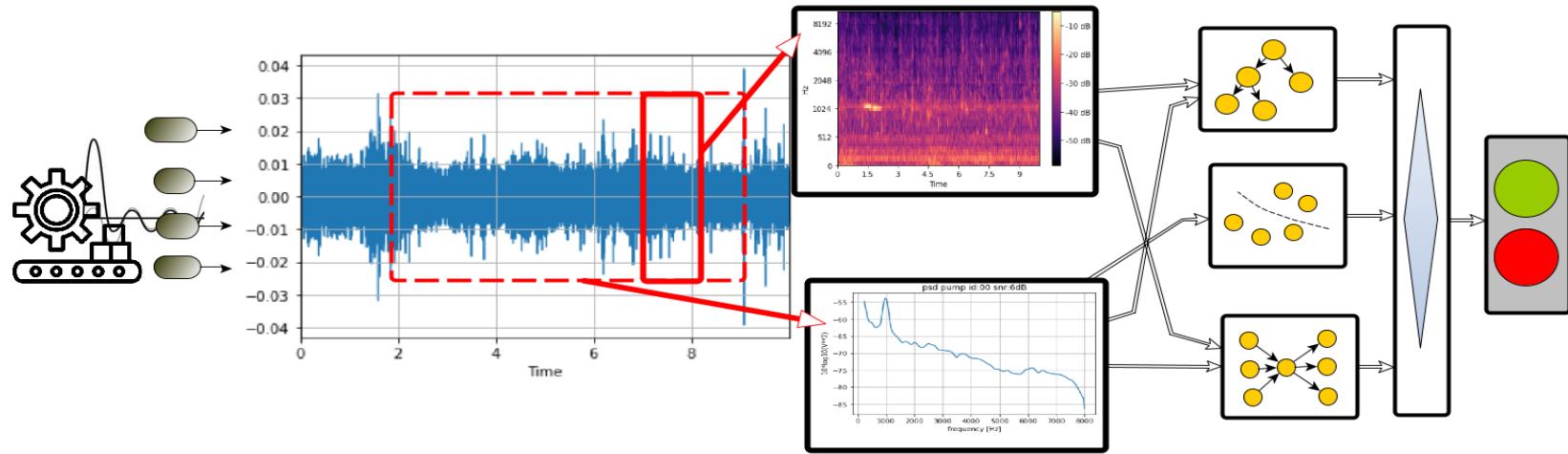


# **Graded Exercise 2:**

# **Anomaly Detection on Acoustic Data**

**Teaching Assistant: Chenghao Xu**  
**Dr. Zhan Ma**

**Date: Oct. 3rd, 2024**



Reference: Can a Machine Hear If a Machine Is Broken?

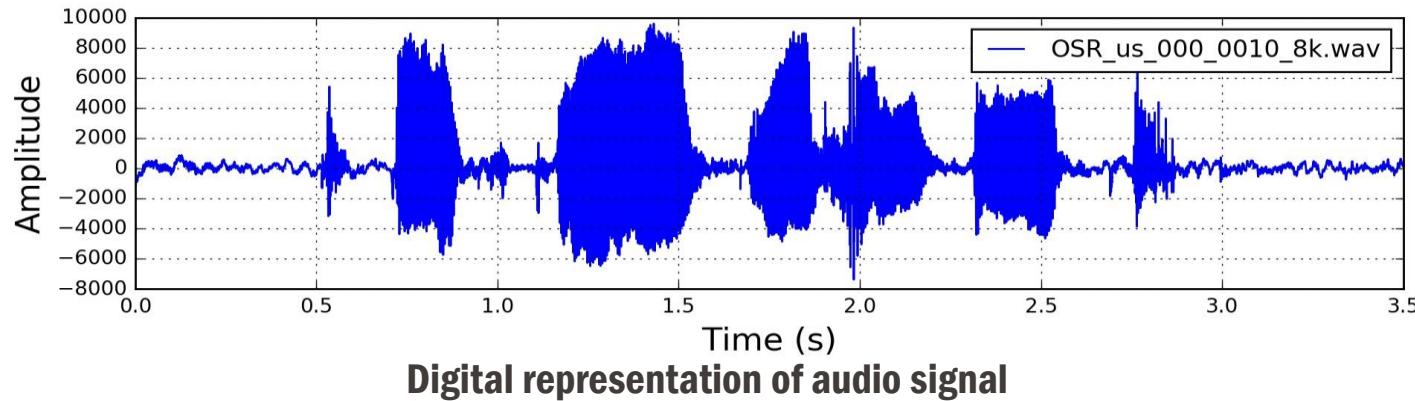
## (Step 0: Signal Preprocessing)

### Step 1: Feature Extraction

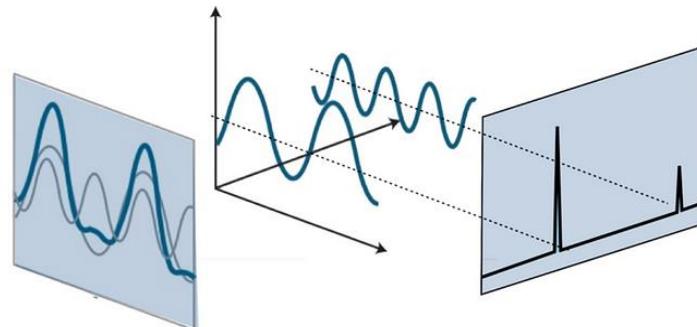
### Step 2: Model Training

### Step 3: Decision-making / Evaluation

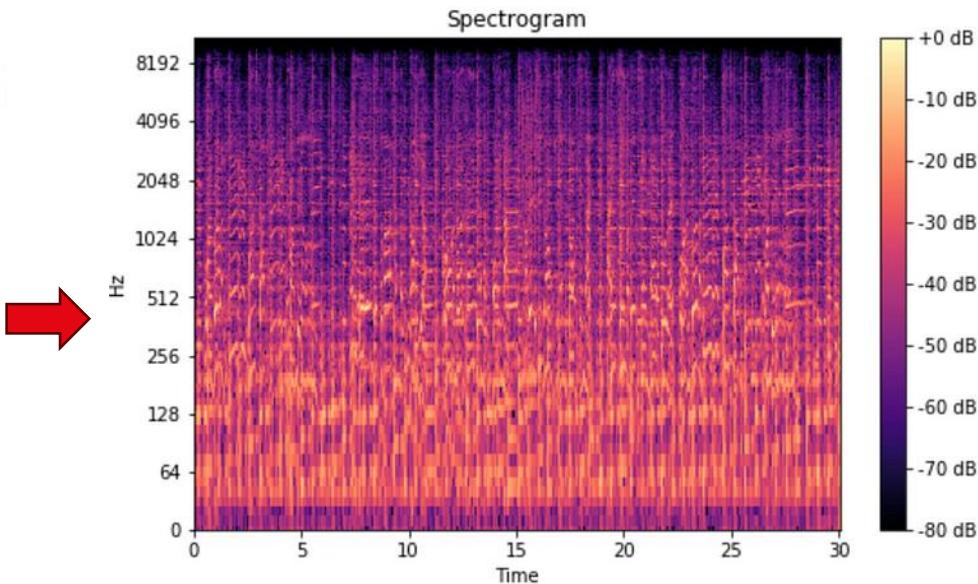
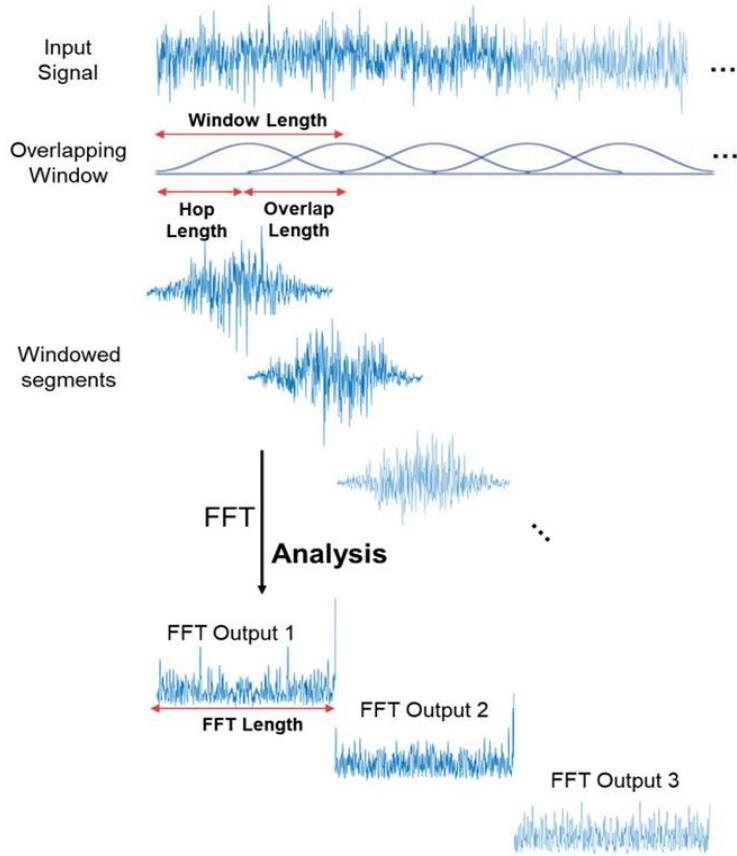
# I – Mel Spectrogram



*Fourier Transform*

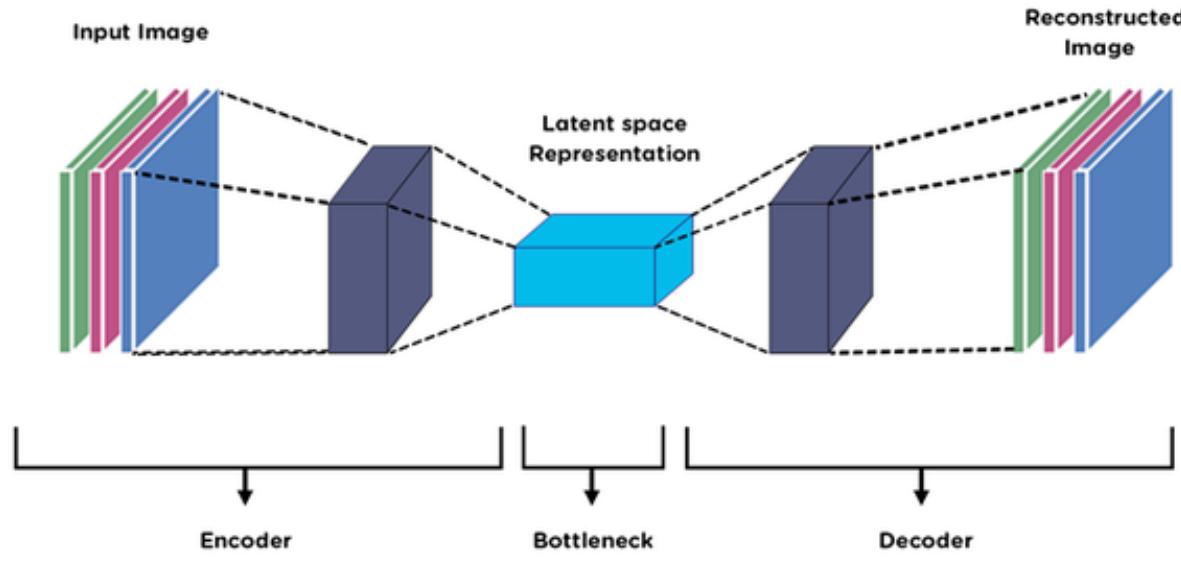


- How to obtain spectrum on non-periodic signals?



Mel Scale: Human better at detecting differences in lower frequencies than higher frequencies.

# II – AutoEncoder

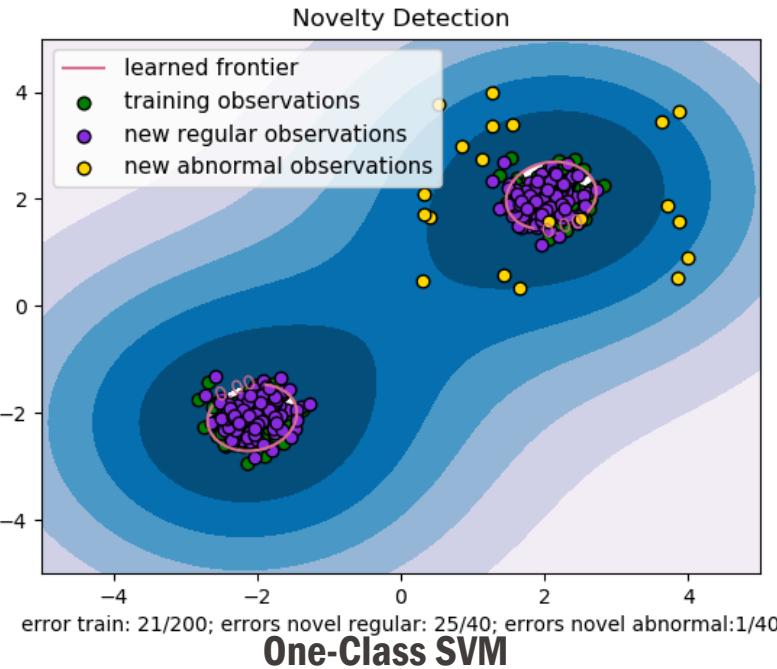


### Autoencoder Architecture

An autoencoder is a type of neural network architecture designed to efficiently compress (encode) input data down to its essential features, then reconstruct (decode) the original input from this compressed representation.

# III – One-Class SVM & Isolation Forest

- Anomaly Detection is widely applied in industrial application or imbalanced data distribution
- One-class SVM: find a hypersphere to separate normal samples and minimize its volume.
- Isolation Forest: find the sample point most likely to be separated (low density, and far from the cluster with high density)




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**Algorithm 1 :  $iForest(X, t, \psi)$**

**Inputs:**  $X$  - input data,  $t$  - number of trees,  $\psi$  - sub-sampling size

**Output:** a set of  $t$   $iTrees$

- 1: **Initialize Forest**
- 2: set height limit  $l = \text{ceiling}(\log_2 \psi)$
- 3: **for**  $i = 1$  to  $t$  **do**
- 4:      $X' \leftarrow \text{sample}(X, \psi)$
- 5:      $Forest \leftarrow Forest \cup iTree(X', 0, l)$
- 6: **end for**
- 7: **return**  $Forest$

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**Isolation Forest**

# IV – Exercise Introduction