

# Mini-project overview

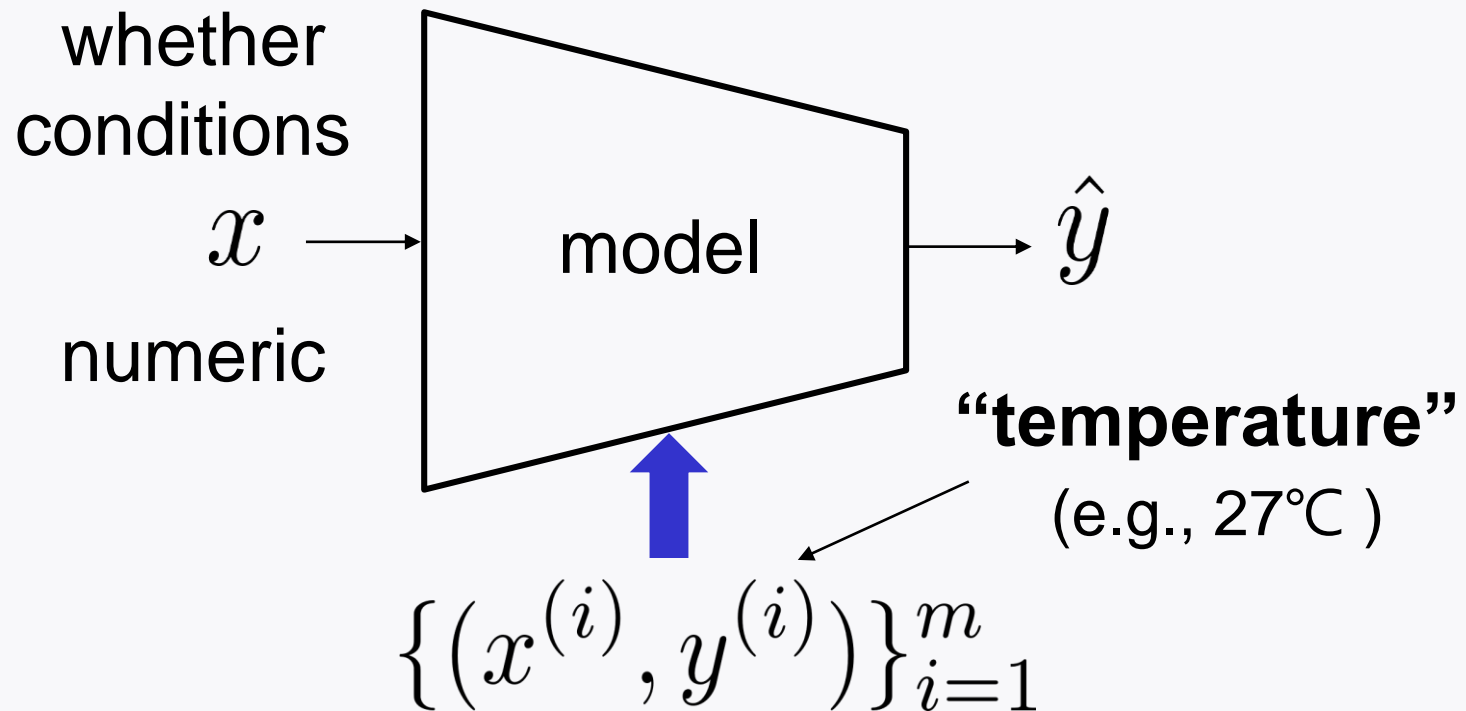
## Lecture 17

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# Mini-project #2

# Task: Weather prediction



# Jena climate dataset

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## Data:

weather features  $\in \mathbf{R}^{14}$

Example: pressure, humidity, wind speed, ...

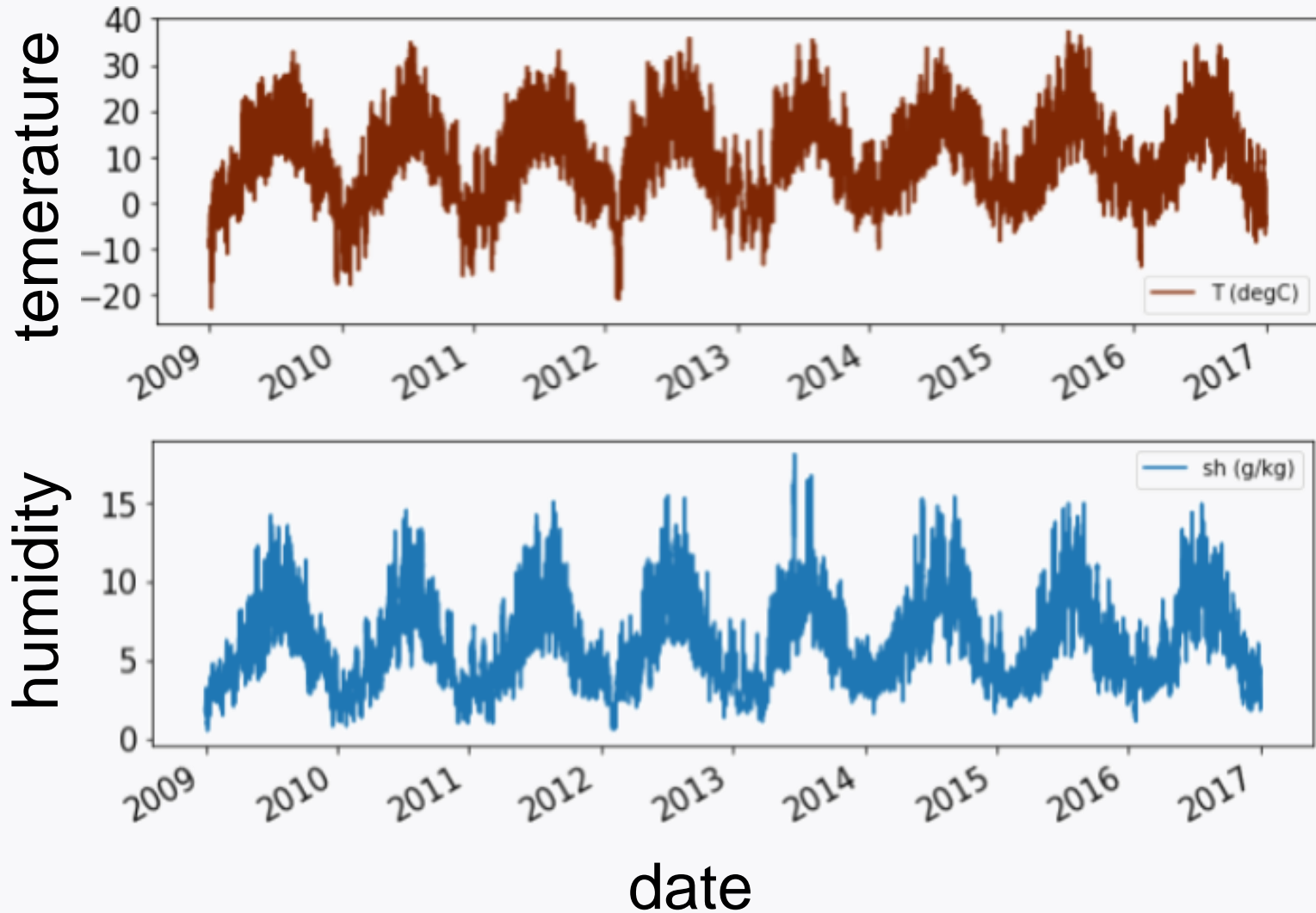
**Label:** celcius temperature  $\in \mathbf{R}$

Collected in Jena from 2009 to 2016 and measured every 10 minutes:

420,551 samples

Load data in **csv** file using pandas.

# Visualization



# Data split

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Note: **Time series** data

Split dataset into **train/val/test** sets with:

**7:2:1** (in *chronological* order)

To this end, will use `train_test_split`

```
from sklearn.model_selection import train_test_split
```

# Model selection

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Will try two models:

**DNN and RNN**

# Performance measure

## Root-mean-square error (RMSE):

$$\sqrt{\frac{1}{m_{\text{test}}} \sum_{i=1}^{m_{\text{test}}} \|y^{(i)} - \hat{y}^{(i)}\|^2}$$

## Normalized RMSE:

RMSE

$$\sqrt{\frac{1}{m_{\text{test}}} \sum_{i=1}^{m_{\text{test}}} \|y^{(i)} - \mu\|^2} \leftarrow \sigma_{\text{test}}$$



# Target performance

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**1. DNN:** NRMSE  $\sim 0.09$

**2. RNN:** NRMSE  $\sim 0.08$

# Advanced techniques we will apply

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Regularization, early stopping

Weight initialization

Learning rate decaying

Hyperparameter search:

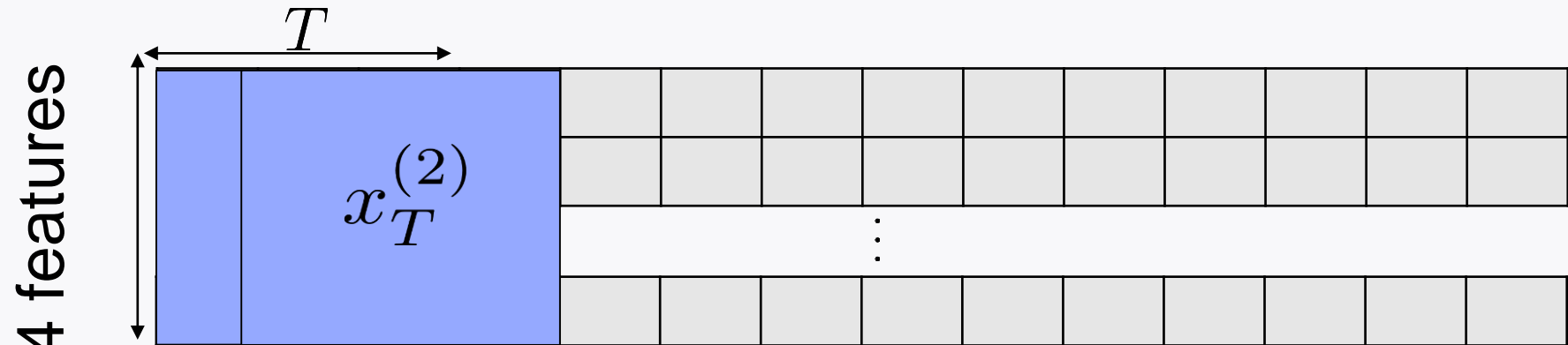
$T$  (window size), learning rate, ...

# Hyperparameter: time window $T$

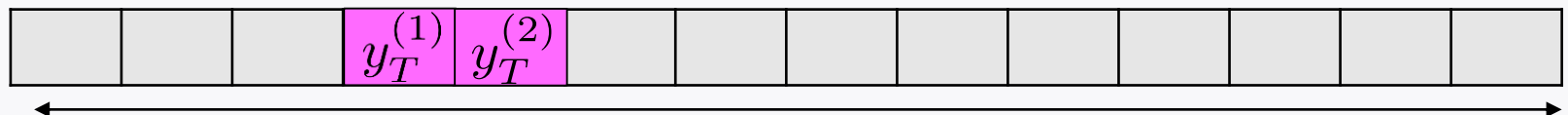
Will generate time series dataset according to  $T$ :

$$\{x_T^{(i)}, y_T^{(i)}\}_{i=1}^{m_T}$$

data:



label:



# of examples

# Saving

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1. “loss” curves
2. log files
3. parameters of trained models