# **Transversal Machine Learning Concepts**

Baseline, Validation and Evaluation

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## Outline

- Baseline Model
- Splitting the Dataset
- 3 Learning Curves
- 4 Cross-Validation
- **5** Grid / Random Search
- **6** Classification Metrics

## Baseline Model

#### **Definition**

A **baseline model** is a simple reference point used to verify that a more complex model actually improves performance.

### Examples

- Classification always predict the majority class.
- Regression predict the mean or median of the target.

#### Why it matters

If your model does not beat the baseline, it has learned nothing useful.

# Train / Validation / Test Split

#### Principle

Divide the dataset to evaluate how well the model generalizes.

- Train fit the parameters.
- Validation tune hyperparameters.
- Test unbiased final check.

#### Python

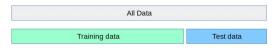


Figure: Sample a training set 80% while holding out 20% of the data for testing. (https://scikit-learn.org/stable/modules/cross<sub>v</sub>alidation)

# Learning Curves

#### Purpose

Show how training and validation scores evolve with the number of samples.

#### Reading the curves

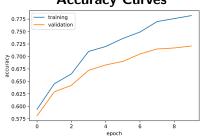
- ullet Large gap o overfitting.
- Both low  $\rightarrow$  underfitting.
- Converge high  $\rightarrow$  good fit.

## Learning Curves: Loss and Accuracy

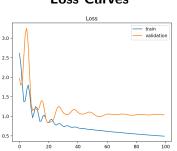
#### Visualization

Learning curves help us diagnose bias/variance trade-offs by observing how accuracy and loss evolve for both training and validation sets.

### **Accuracy Curves**



#### Loss Curves

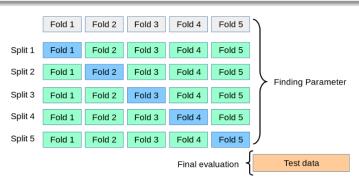


Train vs Validation Accuracy

Train vs Validation Loss

#### Idea

Split data into k folds; train on k-1, test on the remaining one, repeat k times.



k-fold CV, the training set is split into k smaller sets.

# Hyperparameter Tuning

#### Goal

Find the best combination of hyperparameters.

## **Approaches**

**Grid Search** – exhaustive combinations.

**Random Search** – random samples in parameter space.

### Python

## Main Metrics

- Accuracy = correct / total
- Precision = TP / (TP + FP)
- Recall = TP / (TP + FN)
- F1 = 2·(precision·recall)/(precision+recall)

#### Confusion Matrix

Summarizes TP, FP, TN, FN.

## **Practical Evaluation**

Python

# Summary

#### Checklist

- Build a baseline.
- Split properly.
- Plot learning curves.
- Validate with CV.
- Tune with Grid/Random Search.
- Evaluate with multiple metrics.