

[SWCON253] Machine Learning – Lec.13

Model Evaluation

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References

- *Intro to Machine Learning* by Sebastian Raschka (<http://stat.wisc.edu/~raschka/teaching/stat451-fs2020/>)

Introduction

◆ Model Evaluation

- Estimates the **generalization performance**, i.e., the predictive performance of our model on future (**unseen**) **data**.
 1. **Model Test**: Want to **evaluate** the generalization performance of the **developed (final) model** before deploy.
 2. **Model Validation**: Want to **increase** the generalization performance **in the development stage** by tweaking the learning algorithm and **selecting the best** performing model from a given hypothesis space.

Training, Test, Validation Sets

◆ Training Set

- Use training set **to train** your model (**model fitting**)

◆ Test Set

- Use test set **only to evaluate** the performance of your final model (**model test**)
- **Caution: Do not use test set (or test result) to modify your model!**

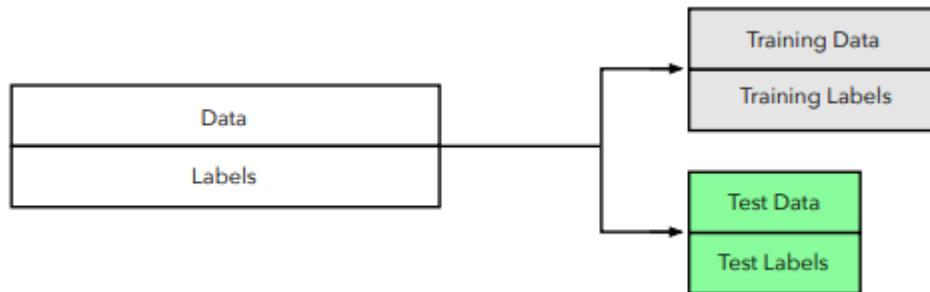
◆ Validation Set

- You can use validation set **to select** best model or hyper-parameters (**model selection**)
 - ★ e.g., in polynomial regression, select the order of the polynomial (i.e., model capacity)
 - ★ e.g., in L1 or L2 regularization, selecting the regularization-weight λ
 - ★ e.g., in SVM, selecting the slack-weight C and/or kernel parameters
 - ★ Note: "best" here means best performing on the validation set, implying the lowest generalization error

2-Way Holdout Method

◆ The Simplest Technique for **Testing** a Model

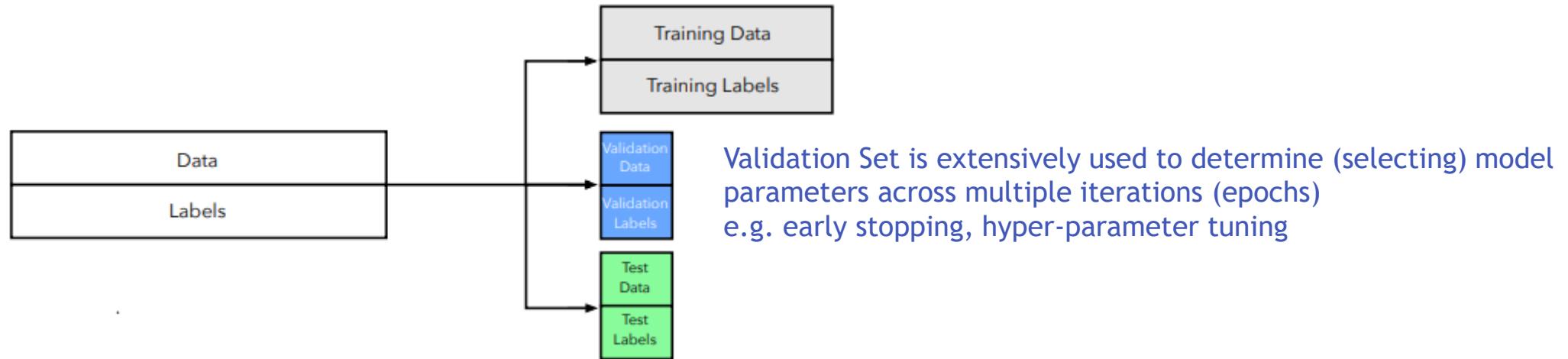
- ① *Divide* our available data into two subsets: a **training** and a **test** set.
 - ★ e.g., randomly assign 33% of the data to the test set, or assign 10% if the dataset is relatively large.
- ② *Train* your ML algorithm with the training set.
- ③ *Estimate* the generalization **performance** of the trained model with the **test** set.
- ④ *(Optional) Retrain* your ML algorithm with **all** the data, including both the training set and test set.
 - ★ As a rule of thumb, the model will have a better generalization performance if the algorithms uses more informative data.
 - ★ **Do not re-estimate the generalization performance with this retrained model!**



3-Way Holdout Method

◆ Holdout Method for **Selecting (Validating)** a Model

- ① *Divide* our available data into three subsets: a **training**, a **validation**, and a **test** set.
 - ★ e.g., assign 33% of the data to the test set, or assign 10% if the dataset is relatively large.
- ② *Train* your ML algorithm with the same training set but *with different hyperparameter settings*.
- ③ *Compare* the generalization **performance** of the trained models with the **validation** set and *Choose* the **hyperparameters** settings associated with the best performance.
- ④ *Estimate* the generalization **performance** of the trained model with the **test** set.
- ⑤ **(Optional)** *Retrain* your ML algorithm with **all** the data, i.e., including both the training set and test set.



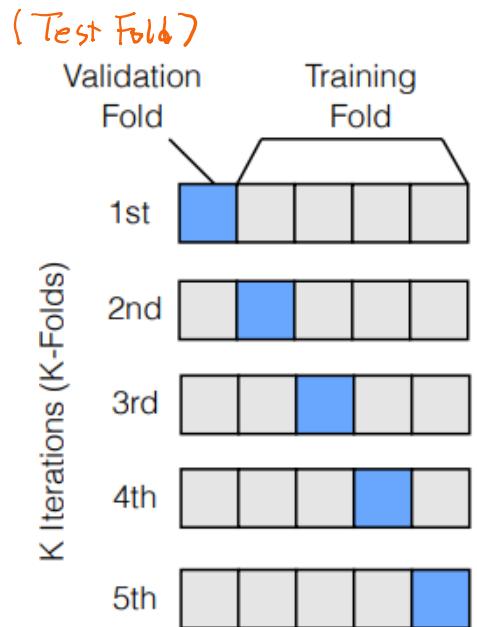
Cross-Validation

◆ Cross-Validation (CV)

- Cross-validation is very useful *when the available dataset size is small*.
- The main idea is that *each sample in our dataset* has the opportunity of being *tested*.
- Cf.) The term cross-validation is used loosely in literature, where the train/test holdout method is sometimes referred to as a cross-validation technique.

◆ k-Fold Cross-Validation

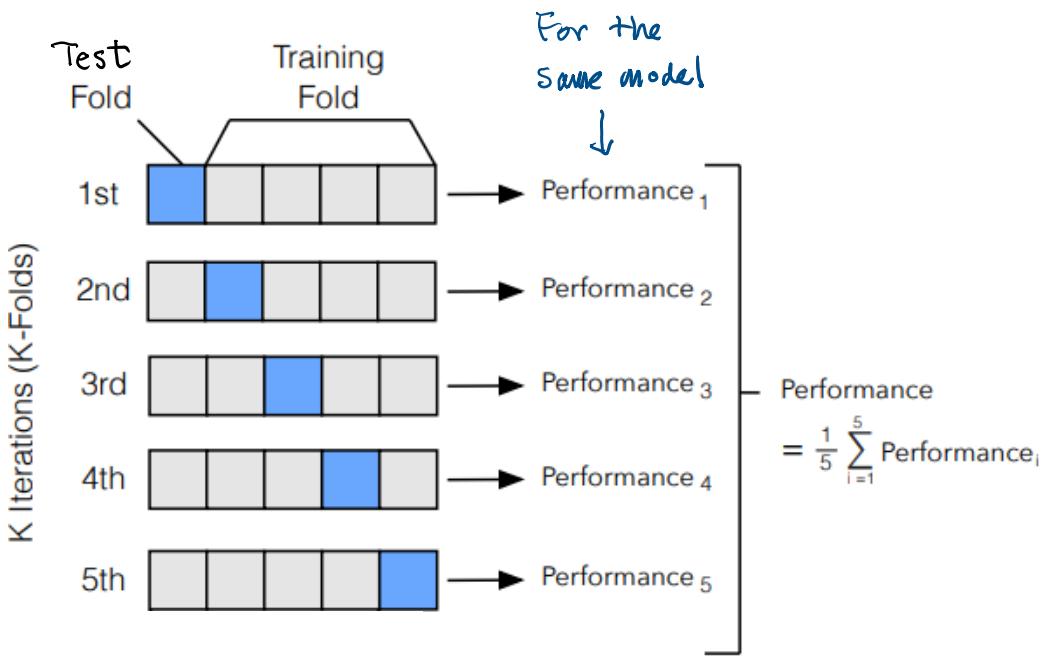
- *k-fold cross-validation* is a special case of cross-validation where we *iterate over a dataset set k times*.
 - ★ non-overlapping test (validation) folds; *utilizes all data for testing* (validation)
 - ★ overlapping training folds
- It is the most common technique for model evaluation & selection



k-Fold Cross-Validation

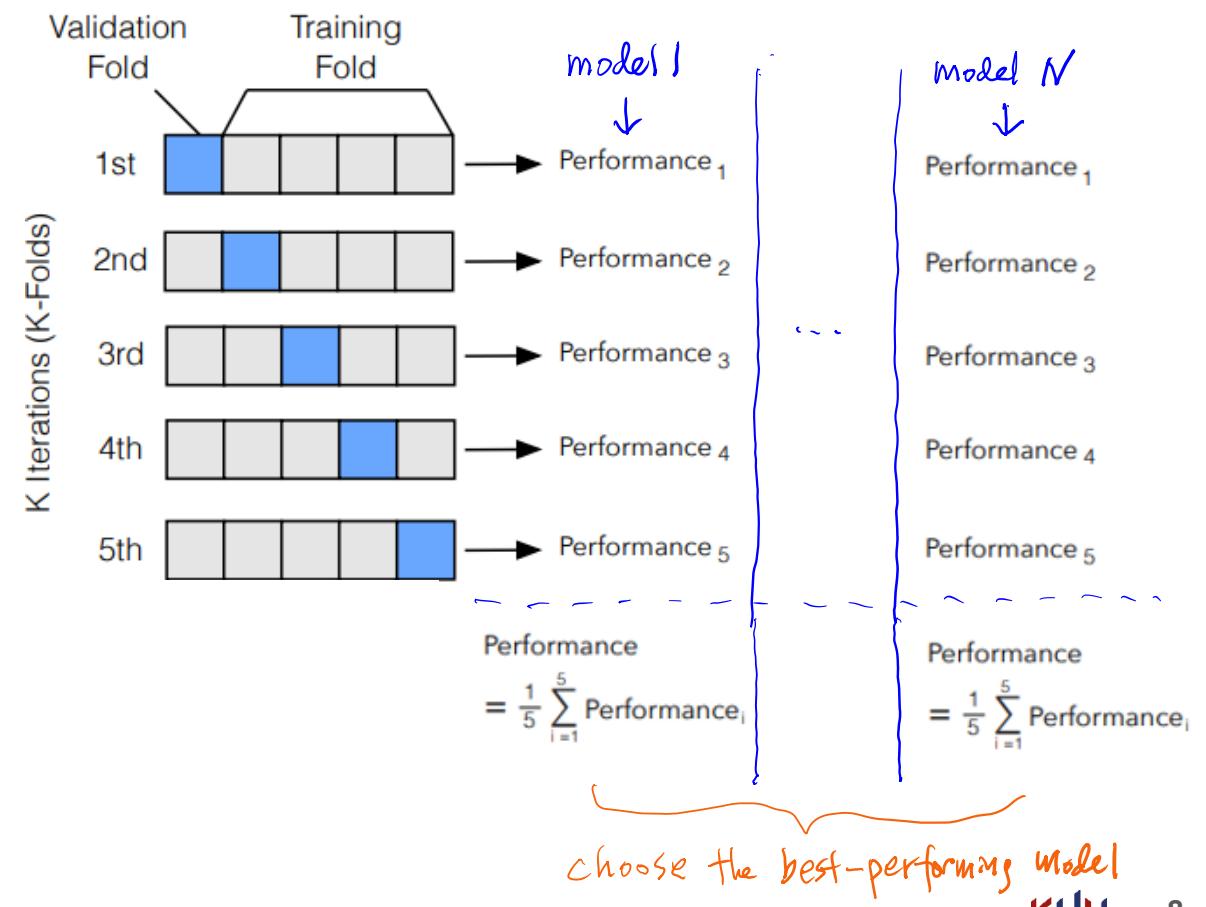
◆ k-Fold CV for Model **Testing**

- Evaluate the generalization performance of a given model (with *fixed hyperparameters*).



◆ k-Fold CV for Model (or Algorithm) **Selection**

- Compare the generalization performance of different models (with *different hyperparameters*).
- Choose the best performing model.



Q & A

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