

Questions based on lecture 1

- (1) (1 pt.) Which claim is true?
- (a) The test data for a machine learning model is usually assumed to be drawn from the same known distribution D that generated the training data
 - (b) Consistent hypothesis correctly classifies all the training samples
 - (c) Version space is the space of all the hypotheses of the hypothesis class

- (2) (1 pt.) Which claim is true?
- (a) If a ROC curve shows a diagonal line $x = y$, this means that the classifier performs perfectly
 - (b) To obtain the ROC curve, the decisions of a binary classifier (classes -1 and 1) are analysed with various different thresholds θ , with decision function f as

$$f(x) = \begin{cases} -1 & \text{if } P(y = 1|x) < \theta \\ 1 & \text{otherwise} \end{cases}$$

- (c) A random classifier plotted in a ROC curve has AUC=1.
- (3) (1 pt.) You are working in a bank, building a machine learning system whose goal is to raise an alert for fraudulent-looking transactions. It is very important that no fraudulent transactions (=positive class) get past this model, but if some legitimate transactions (=negative class) happens to be flagged as fraudulent it would be quickly cleared up and no big deal.
- If the precision of the system is 99%, how would you assess the machine learning model?
- (a) This is a good model and performs very well for the task.
 - (b) You are unsure if the model is good or not: this metric does not give you enough information to make an informed assessment.
 - (c) The model is not performing well.
- (4) (1 pt.) Continuing with the setting of the previous question, if instead you obtain recall of 99% (without knowing the precision), how would you assess the model then?
- (a) This is a good model and performs very well for the task.
 - (b) You are unsure if the model is good or not: this metric does not tell you enough information to make an informed assessment.
 - (c) The model is not performing well.

- (5) (1 pt.) [*Programming exercise*] Consider the diabetes dataset loaded in the provided python code. The data is split to training and test sets: assume the test set is not available during the training phase.
- Fit a linear regression model to the data without considering bias (intercept). What is the root mean squared error on the test data?
- (a) 55.52
 - (b) 167.69
 - (c) 28120.65

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import numpy as np
from sklearn.datasets import load_diabetes
from sklearn.metrics import mean_squared_error

# load the data
X, y = load_diabetes(return_X_y=True)
print(X.shape, y.shape)

# division into training and testing
np.random.seed(0)
order = np.random.permutation(len(y))
tst = np.sort(order[:200])
tr = np.sort(order[200:])

Xtr = X[tr, :]
Xtst = X[tst, :]
Ytr = y[tr]
Ytst = y[tst]

# assume that the test data is not known during the training stage
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