

Fundamentals of Machine Learning for Predictive Data Analytics

Chapter 8: Evaluation Sections 8.1, 8.2, 8.3

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- 1 **Big Idea**
- 2 **Fundamentals**
- 3 **Standard Approach: Measuring Misclassification Rate on a Hold-out Test Set**
- 4 **Summary**

- The most important part of the design of an evaluation experiment for a predictive model is ensuring that the data used to evaluate the model is not the same as the data used to train the model.

- The purpose of evaluation is threefold:
 - 1 to determine which model is the most suitable for a task
 - 2 to estimate how the model will perform
 - 3 to convince users that the model will meet their needs

Standard Approach: Measuring Misclassification Rate on a Hold-out Test Set

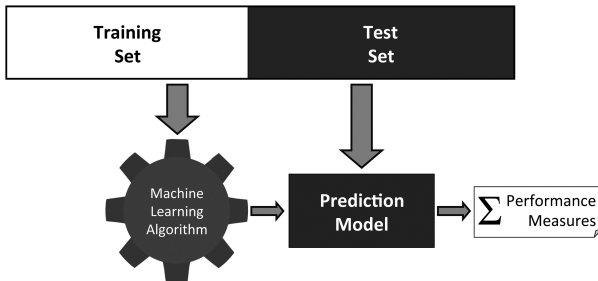


Figure: The process of building and evaluating a model using a **hold-out test set**.

Table: A sample test set with model predictions.

ID	Target	Pred.	Outcome	ID	Target	Pred.	Outcome
1	spam	ham	FN	11	ham	ham	TN
2	spam	ham	FN	12	spam	ham	FN
3	ham	ham	TN	13	ham	ham	TN
4	spam	spam	TP	14	ham	ham	TN
5	ham	ham	TN	15	ham	ham	TN
6	spam	spam	TP	16	ham	ham	TN
7	ham	ham	TN	17	ham	spam	FP
8	spam	spam	TP	18	spam	spam	TP
9	spam	spam	TP	19	ham	ham	TN
10	spam	spam	TP	20	ham	spam	FP

$$\text{misclassification rate} = \frac{\text{number incorrect predictions}}{\text{total predictions}} \quad (1)$$

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$$\text{misclassification rate} = \frac{(2 + 3)}{(6 + 9 + 2 + 3)} = 0.25$$

- For binary prediction problems there are 4 possible outcomes:
 - 1 True Positive (TP)
 - 2 True Negative (TN)
 - 3 False Positive (FP)
 - 4 False Negative (FN)

Table: The structure of a confusion matrix.

		Prediction	
		positive	negative
Target	positive	<i>TP</i>	<i>FN</i>
	negative	<i>FP</i>	<i>TN</i>

Table: A confusion matrix for the set of predictions shown in Table 1 [7].

		Prediction	
		'spam'	'ham'
Target	'spam'	6	3
	'ham'	2	9

$$\text{misclassification accuracy} = \frac{(FP + FN)}{(TP + TN + FP + FN)} \quad (2)$$

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$$\text{classification accuracy} = \frac{(TP + TN)}{(TP + TN + FP + FN)} \quad (3)$$

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$$\text{classification accuracy} = \frac{(6 + 9)}{(6 + 9 + 2 + 3)} = 0.75$$

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