

# ML Section 1.4: Evaluation of Supervised Learning

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## 1 Hypothesis Evaluation

General questions:

- How can one estimate the performance of a learned hypothesis on future data?
- How good is the estimate?
- Comparative performance evaluations.

**Formally:** Given a hypothesis  $h$  and a data sample containing  $n$  examples drawn at random according to the distribution  $D$ , what is the best estimate of the accuracy of  $h$  over future instances drawn from the same distribution? What is the probable error in this accuracy estimate?

## 2 Evaluation Problems

- Limited samples of data may be misleading (e.g. prime numbers and data set =  $\{3,5,7\}$  leads to hypothesis of odd numbers).

- Observed accuracy on training data is often too optimistic (e.g. due to overfitting).
- Solution: use independent test examples.
- Problem: estimate may still depend on the specific makeup of the set of training/test examples.

### 3 Preliminary Definitions:

$f$  The target customisation function to be learned ( $f: \text{Examples} \rightarrow \text{Categories}$ ).

$h$  The hypothesis learned ( $h: \text{Examples} \rightarrow \text{Categories}$ ).

$S$  Data sample of size  $n$ .

$D$  Probability distribution over all data points.

#### Sample Error

$$error_s(h) = \frac{1}{2} \sum_{x \in S} \delta(f(x), h(x)) \quad (1)$$

Where:

$$\delta(y, z) = 1 \text{ if } y \neq z, \text{ and } 0 \text{ otherwise.} \quad (2)$$

#### True Error

$$error_D(h) = Pr_{x \in D}[f(x) \neq h(x)] \quad (3)$$

### 4 Confidence Intervals