

# **A Statistical Framework for Machine Learning**

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A statistical framework for machine learning is presented.  
The following topics are described.

**(a)** Description of the Phenomenon

**(b)** Development of the Model

Model 1.

Model 2.

### *A. Description of the Phenomenon.*

Let the input space of a system be specified by  $X$ .

Let the output space of a system be specified by  $Y$ .

Let  $X \times Y \triangleq Z$ .

There is an unknown distribution  $\mathcal{D}$  over  $Z$ .

The learner collects  $m \in \mathbb{P}$  samples which are drawn from the distribution  $\mathcal{D}$ , where  $\mathbb{P}$  is the set of positive integers. The sample set is:

$$\{(x_i, y_i) \mid 1 \leq i \leq m\}$$

## *B. Development of the Model:*

Different models for the evaluation of the learning model are developed.

*Model 1:* Let  $\mathcal{F}$  be a collection of models, where for each  $f \in \mathcal{F}$ , we have  $f : X \rightarrow Y$  predicts  $y$  for a given  $x$ .

Using  $m \in \mathbb{P}$  observations, the goal is to select a model  $f_m \in \mathcal{F}$  which predicts *well*. One way to define the word ‘well’ is as follows. Let

$$err(f) = P_{Z \sim \mathcal{D}}(f(x) \neq y) = \text{error probability of } f$$

The model  $f$  is said to have learned the phenomenon if

$$f^* = \arg \min_{f \in \mathcal{F}} err(f)$$

$$|err(f) - err(f^*)| \leq \varepsilon$$

for a specified tolerance level  $\varepsilon > 0$ .

*Model 2:* A loss function  $\ell(f(x), y)$  is used to measure the discrepancy between the predicted output  $h(x)$  and the true output  $y$ .

1. The *empirical risk* is:

$$R_{emp}(f) = \frac{1}{m} \sum_{i=1}^m \ell(f(x_i), y_i)$$

2. The *theoretical risk* is:

$$R_{theor}(f) = \mathcal{E}(\ell(f(X), Y))$$

where  $\mathcal{E}(\cdot)$  is the expectation operator.

*Examples of loss function  $\ell(\cdot, \cdot)$ :*

1. Classification, 0-1 loss:  $\ell(f(x), y) = 1[f(x) \neq y]$ .
2. Regression, square loss:  $\ell(f(x), y) = (y - f(x))^2$ .