



## Why

We have a space  $X$  and a family of probability measures  $\mathcal{P}$  on this space. Assume  $x \in X$  drawn from a fixed, unknown measure  $P \in \mathcal{P}$ . Given  $x$ , how should we guess  $P$ ?

## Definition

A *probabilistic model* (or *statistical model*, *parametric statistical model*, *statistical experiment*) is a family of probability measures over the same measurable space  $(X, \mathcal{F})$ . Call the index set the *parameter set* or *set of parameters*. The set  $X$  is called the *sample space*. A *statistic* is any function on the sample space.

## Notation

Let  $(X, \mathcal{F})$  denote a measurable space. We usually denote the parameter by  $\Theta$ , and denote the family

$$\mathcal{P} = \{\mathbf{P}_\theta : \mathcal{F} \rightarrow [0, 1] \mid \mathbf{P}_\theta \text{ a measure}, \theta \in \Theta\}.$$

Often  $\Theta \subset \mathbf{R}^d$ .

## Example: coin flips

The usual model for  $n$  flips of a coin takes  $X = \{0, 1\}^n$ , the set of binary  $n$ -tuples. For  $\theta \in [0, 1]$ , a distribution  $p_\theta(x) = \theta^t(1-\theta)^{n-t}$  where  $t = t(x) = x_1 + \dots + x_n$  is defined on  $X$ . A probability measure  $\mathbf{P}_\theta$  is defined on  $\mathcal{P}(X)$  in the usual way. Thus, the probabilistic model is  $\{\mathbf{P}_\theta \mid \theta \in [0, 1]\}$ . Given  $x$ , we are asked to guess  $\theta$ .

## Decisions

A *decision procedure* (*estimator*, *statistical procedure*) is a measurable function  $A : \mathcal{X} \rightarrow \mathcal{A}$  where  $\mathcal{A}$  is a set, called the *actions* or *decisions*. Often  $\mathcal{A} = \Theta$ , in which case  $A(x)$  gives an *estimate* of  $\theta$ , which we denote  $\hat{\theta}(x)$ .

## Judging decisions

Given a *loss function*  $L : \mathcal{A} \times \Theta \rightarrow \bar{\mathbf{R}}$ , the *risk* of  $A$  is

$$R(A, \theta) = \mathbf{E}L(A(x), \theta).$$



