

# Models and Likelihood

## Statistical Models

- Mathematical models
- Statistical models
  - Parameters
  - Likelihood

## First example: coin flipping

- Model a coin flipping experiment as a Bernoulli random variable with parameter  $p$ .
- Flip the coin 100 times and get 55 heads and 45 tails.

$$L = \binom{100}{55} p^{55} (1-p)^{45}$$

- **Maximum Likelihood** - forget the constant as it doesn't effect the result.

$$\frac{dL}{dp} = 55p^{54}(1-p)^{45} - 45p^{55}(1-p)^{44} = 0$$

yields

$$55(1-p) = 45p$$

or

$$p = 55/100 = .55$$

## Independent normally distributed errors

- Back to our temperature model. We assume that the errors in our measurements are normally distributed around zero. There is one parameter: the variance  $\sigma^2$  in our density function for a single measurement

$$p_\sigma(x) = \left( \frac{1}{\sigma\sqrt{2\pi}} \right) e^{-x^2/(2\sigma^2)}$$

- We make  $n$  independent measurements of temperature

$$x_1, \dots, x_n$$

What does this tell us about  $\sigma^2$ ? The likelihood for independent measurements is the density

$$L = \left( \frac{1}{\sigma\sqrt{2\pi}} \right)^n e^{-\|x\|^2/(2\sigma^2)}$$

- Maximize the density at this point. Use the *log-likelihood* as it is easier.

$$\log P(x) = -n \log \sigma - \frac{\|x\|^2}{2\sigma^2} + C$$

- Take the derivative and set it equal to zero.

- The *maximum likelihood estimate of the variance is the mean squared error!*

## Linear Regression and likelihood

- Model says that our  $N$  data points  $(x_i, y_i)$  arose from a process

$$y = mx + b + \epsilon$$

where  $\epsilon$  is a normally distributed error term with variance  $\sigma^2$ .

- How should we set  $m, b, \sigma$  to make the observed data most likely?
- The density function is

$$P(m, b, \sigma) = \left( \frac{1}{\sigma\sqrt{2\pi}} \right)^N e^{-\sum_{i=1}^N (y_i - mx_i - b)^2 / (2\sigma^2)}$$

- The *log likelihood* is

$$\log P = -N \log \sigma - \frac{1}{2\sigma^2} \sum_i (y_i - mx_i - b)^2$$

- The derivatives with respect to  $m$  and  $b$  give the least squares estimates.

- The derivative with respect to  $\sigma$  gives the best estimate when  $\sigma^2$  is the mean squared error.

- **Ordinary Least Squares is the maximum likelihood solution assuming independent normally distributed errors**