# Models and Likelihood

### Statistical Models

- $\bullet$  Mathematical models
- ullet Statistical models
  - Parameters
  - Likelihood

### First example: coin flipping

- Model a coin flipping experiment as a Bernoulli random variable with parameter p.
- $\bullet\,$  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

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

ullet We make n independent measurements of temperature

$$x_1, \ldots, x_n$$

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

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

 $\bullet$  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