

## Tutorial 6

2. For the given data, we can fit a best fit line but we can see by plotting that error is high.

Hence we try fitting the data using a polynomial. If we take polynomial of degree 2, the data is fitted much better.

Hence we can consider variables  $1, x, x^2$  and  $y = \beta_0 + \beta_1 x + \beta_2 x^2$ .

The model is still "Linear" in these transformed variable and thus basis expansion is done.

3. a) Since we have  $x_1, \dots, x_n$  independent random variables and all are Bernoulli, for an  $x_i$ ,

$$f(x_i; p) = p^{x_i} (1-p)^{1-x_i}$$

and

$$\text{Likelihood } L(p) = \prod_{i=1}^n f(x_i; p) \quad \left[ \begin{array}{l} \text{as all } x_i \text{ are} \\ \text{independent} \end{array} \right]$$

$$= p^{x_1} (1-p)^{1-x_1} \dots p^{x_n} (1-p)^{1-x_n}$$

$$= p^{\sum_{i=1}^n x_i} (1-p)^{n - \sum_{i=1}^n x_i}$$

Log likelihood,

$$\log L(p) = (\log p) \sum_{i=1}^n x_i + (\log(1-p)) (n - \sum_{i=1}^n x_i)$$

Diff wrt  $p$  and equate to 0,

$$\frac{\sum x_i}{p} - \frac{(n - \sum x_i)}{1-p} = 0$$

$$\Rightarrow (1-p) \sum x_i = p(n - \sum x_i)$$

$$\Rightarrow \sum x_i - np = 0$$

$$\Rightarrow \boxed{\begin{array}{c} \hat{p}_{MCE} = \frac{\sum_{i=1}^n x_i}{n} \end{array}}$$



4. We know, 
$$P(\theta|S) = \frac{P(S|\theta) P(\theta)}{\int P(S|\theta) P(\theta) d\theta}$$

Here, since we are using head or tail,  
 $P(S|\theta)$  is from a Bernoulli dist.

Also since we did 20 trials and got Heads  
 $\frac{1}{4} \times 20 = 5$  times. We got tails 15 times.

The prior dist is Beta (#heads, #tails)  
 $= \text{Beta}(5, 15)$

$$\therefore P(\theta|S) = (\text{const}) \cdot \theta^y (1-\theta)^{n-y} \theta^4 (1-\theta)^{14}$$

$$[ \int P(S|\theta) P(\theta) d\theta \text{ is const} ]$$

$$= \theta^{y+4} (1-\theta)^{n-y+14} \Rightarrow \text{Beta}(y+5, n-y+15)$$

$$\therefore \text{Likelihood} = \prod_{i=1}^n \theta^{x_i+4} (1-\theta)^{n-x_i+14}$$

$$\text{Log likelihood} = \sum_{i=1}^n \left( (x_i+4) \log \theta + (n-x_i+14) \log(1-\theta) \right)$$

diff w.r.t  $\theta$  and equate to 0,

$$\Rightarrow \frac{\sum x_i + 4n}{\theta} - \frac{(n^2 + 14n - \sum x_i)}{1-\theta} = 0$$

$$\Rightarrow \sum x_i + 4n - \theta(n^2 + 14n - \sum x_i) = 0$$

$$\Rightarrow \sum x_i + 4n = \theta(n^2 + 14n - \sum x_i + \sum x_i + 4n)$$

$$\Rightarrow \boxed{\theta_{\text{MLE}} = \frac{\sum_{i=1}^n x_i + 4n}{n^2 + 18n}}$$