

# Elementary Statistical Inference:

## Binomial distribution;

Let  $X = \begin{cases} 1 & \text{success,} \\ 0 & \text{otherwise} \end{cases}$   $X \sim \text{Bin}(n, \theta), \theta \in [0, 1]$

Likelihood estimate based on a sample of  $X$ :

$$L(\theta) = \prod_{i=1}^n p(x_i; \theta) = \theta^{\sum_{i=1}^n x_i} (1-\theta)^{n - \sum_{i=1}^n x_i}$$

$$\text{logs: } l(\theta) = \sum_{i=1}^n x_i \log(\theta) + \left(n - \sum_{i=1}^n x_i\right) \log(1-\theta)$$

$$\frac{\partial l(\theta)}{\partial \theta} = \frac{\sum_{i=1}^n x_i}{\theta} - \frac{n - \sum_{i=1}^n x_i}{1-\theta}$$

solving for  $\theta$ , we get  $\hat{\theta}_{mb} = n^{-1} \sum_{i=1}^n x_i = \bar{x}$ ;  
proportion of successes in "n" indep. Bernoulli trials.

eg. In a study, we find that 18 out of 79 patients are at risk of a disease "A". Likelihood estimator:

$$\hat{\theta}_{mb} = \frac{18}{79} \approx \underline{\underline{0.228}}$$

Reference: Introduction to mathematical statistics  
Hogg; McKean; Craig, 2019