

①

$$T_1 = \sum_{i=1}^n X_i$$

Joint pdf is given by:

$$f(\underline{y}_n; \theta) = \begin{cases} \prod_{i=1}^n \theta^{y_i} (1-\theta)^{1-y_i} & \text{if } y_i = 0, 1 \\ 0 & \text{if otherwise} \end{cases}$$

Let $K(\underline{y}_n)$ be a function such that,

$$K(\underline{y}_n) = \begin{cases} 1 & \text{if all } y_i \in \{0, 1\} \\ 0 & \text{if any } y_i \notin \{0, 1\} \end{cases}$$

$$\Rightarrow f(\underline{y}_n; \theta) = \prod_{i=1}^n \theta^{y_i} (1-\theta)^{1-y_i} K(\underline{y}_n)$$

$$= \left(\frac{\theta}{1-\theta} \right)^{\sum_{i=1}^n y_i} (1-\theta)^N K(\underline{y}_n)$$

$$= \theta^{\sum y_i} (1-\theta)^{N - \sum y_i} K(\underline{y}_n)$$

$\theta^{\sum y_i} (1-\theta)^{N - \sum y_i}$ is a function of y_i

only through $T_1 = \sum_{i=1}^n y_i$

consider $\theta^{\sum y_i} (1-\theta)^{N - \sum y_i} = \phi(T_1; \theta)$

$$\Rightarrow \phi(T_1; \theta) = \theta^{T_1} (1-\theta)^{N - T_1}$$

$$\Rightarrow f(\underline{y}_N, \theta) = \phi(T_1, \theta) K(\underline{y}_N)$$

By the Neyman-Fisher factorisation theorem,
 T_1 is a sufficient statistic for θ .

$$\text{Now, } E(T) = E\left(\sum_{i=1}^N Y_i\right) = \sum_{i=1}^N E(Y_i)$$

$$\text{For } Y_i \in \{0, 1\} \quad f(y_i, \theta) = \begin{cases} \theta & y = 1 \\ 1 - \theta & y = 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\Rightarrow E(Y_i) = \theta \times 1 + (1 - \theta) \times 0 \\ = \theta$$

$$\therefore E(T) = NE(Y_i) \\ = N\theta$$

Since T is a complete sufficient statistic for θ , we just need to do bias correction to make it MVUE.

$$\Rightarrow T_{\text{new}} = \frac{1}{N} T_1 \quad (\Rightarrow E(T_{\text{new}}) = \frac{1}{N} E(T_1)) \\ \Rightarrow T_{\text{new}} \text{ is unbiased} \quad \left(\frac{1}{N} \times \theta N = \theta \right)$$

\therefore The MVUE for θ is

$$\frac{1}{N} \sum_{i=1}^N Y_i$$