

$$X_i \stackrel{i.i.d.}{\sim} \text{Beta}(\alpha, \alpha) \quad f(x|\alpha) = \frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} x^{\alpha-1} (1-x)^{\alpha-1}$$

A. LIKELIHOOD FUNCTION

$$x \in (0,1), \alpha > -1$$

$$\text{Want } L(\alpha | \underline{x}) = f(x|\alpha)$$

$$X_i \text{ i.i.d. } \longrightarrow = \frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} x_i^{\alpha-1} (1-x_i)^{\alpha-1}$$

$$= \left(\frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} \right)^m \left(\prod_{i=1}^m x_i (1-x_i) \right)^{\alpha-1}$$

B. $T(X)$ SUFFICIENT

$$T(X) = \frac{1}{m} \sum_{i=1}^m \log(x_i (1-x_i)) < 0$$

$$\Rightarrow m T(X) = \sum_{i=1}^m \log(x_i (1-x_i)) = \log \prod_{i=1}^m (x_i (1-x_i))$$

$$\Rightarrow e^{m T(X)} = \prod_{i=1}^m x_i (1-x_i)$$

Likelihood above can be rewritten as

$$L(\alpha | \underline{x}) = \left(\frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} \right)^m \left(e^{m T(X)} \right)^{\alpha-1}$$

Therefore Set $g(T(X) | \alpha) = \left(\frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} \right)^m e^{m T(X)}$
 $h(\underline{x}) = 1$
Depends on \underline{x} only through $T(\underline{x})$

$$L(\alpha | \underline{x}) = g(T(X) | \alpha) h(\underline{x}) \quad \text{independent of } \alpha$$

By factorisation theorem $T(X)$ is sufficient

C GENERALISED LIKELIHOOD RATIO

$$W(\underline{x}) = \frac{\sup_{\alpha \in \Theta_1} L(\alpha | \underline{x})}{\sup_{\alpha \in \Theta_0} L(\alpha | \underline{x})} \quad \Theta_1 = \{1, 2\}$$
$$\Theta_0 = \{1\}$$

Note: the notation for this question simplifies when we consider $L(\alpha | \underline{x}) = g(\tau(\underline{x}) | \alpha) h(\underline{x})$ from part b.

$$L(\alpha | \underline{x}) = \frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} e^{m\tau(\underline{x})(\alpha-1)}$$

$$L(1 | \underline{x}) = 1 \quad L\left(\frac{1}{2} | \underline{x}\right) = \left(\frac{\Gamma(1)}{\Gamma(\frac{1}{2})^2}\right)^m e^{-\frac{m}{2}\tau(\underline{x})}$$

$$\text{Note } \Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}, \Gamma(m) = (m-1)!$$
$$= \left(\frac{1}{\pi} e^{-\frac{\tau(\underline{x})}{2}}\right)^m$$

$$L(2 | \underline{x}) = \left(\frac{\Gamma(4)}{\Gamma(2)^2}\right)^m e^{m\tau(\underline{x})}$$
$$= (6e^{\tau(\underline{x})})^m$$

$$\Rightarrow W(\underline{x}) = \max \left\{ \left(\frac{e^{-\frac{\tau(\underline{x})}{2}}}{\pi}\right)^m, (6e^{\tau(\underline{x})})^m \right\}$$

$$D. \quad \mathbb{E}_{X \sim 1} T(X) \quad \& \quad \text{Var}_{X \sim 1} T(X)$$

Note: $f(x|X=1) = 1$

$$\Rightarrow \mathbb{E}_{X \sim 1} [T(X)] = \mathbb{E} \left[\frac{1}{m} \sum \log(X_i(1-X_i)) \right]$$

$$\stackrel{IID}{=} \mathbb{E} \left[\log(X(1-X)) \right]$$

$$= \mathbb{E} \left[\log(X) + \log(1-X) \right]$$

$$= \int_0^1 (\log(x) + \log(1-x)) \cdot 1 \, dx$$

$$= \text{Wolfram} = -2$$

$$\text{Var}_{X \sim 1} (T(X)) = \mathbb{E} (T(X)^2) - \mathbb{E} (T(X))^2$$

$$= \mathbb{E} \left[\frac{1}{m^2} \sum \log^2(X_i(1-X_i)) \right] - 4$$

$$= \frac{1}{m^2} \sum_i \mathbb{E} \left[\log(X_i(1-X_i)) \log(X_i(1-X_i)) \right]$$

when $i=j \Rightarrow \mathbb{E} \log^2(X(1-X))$

$i \neq j \Rightarrow 4$

$$= \frac{1}{m^2} \left[m \mathbb{E} \left[\log^2(X(1-X)) \right] + m(m-1) 4 \right]$$

$$= \text{Wolfram} = 8 - \frac{\pi^2}{3}$$

$$\text{Var}_{X \sim 1} (T(X)) = \frac{1}{m} \left(8 - \frac{\pi^2}{3} + (m-1) 4 \right) - 4$$

$$= \frac{1}{m} \left(8 - \frac{\pi^2}{3} + (m-1) 4 - m 4 \right)$$

$$= \frac{1}{m} \left(4 - \frac{\pi^2}{3} \right)$$

E P-value corresponding to W

Note: Because H_1 is not simply $\alpha \neq 1$ p-values based on $T(x)$ alone does not test our hypotheses. Therefore $W(x)$ must be used to derive the p-value

$$P_W(\underline{\alpha}) = P_{\alpha=1}(W(x) \geq W(\underline{\alpha})) \quad \text{where } W(\underline{\alpha}) = y$$

is calculated from
some realisation \underline{x}

$$\Rightarrow P_{\alpha=1}(W(x) \geq y)$$

Consider $W(x) \geq y$ $y > 0$

$$\Rightarrow \left(\frac{e^{-\frac{T(x)}{2}}}{\pi} \right)^m \geq y \quad \text{or} \quad (b e^{T(x)})^m \geq y$$

Interested in $y > 1$

$$m \left(-\frac{T(x)}{2} - \log \pi \right) \geq \log y \quad \text{or} \quad m(\log b + T(x)) \geq \log y$$

$$-\frac{T(x)}{2} - \log \pi \geq \frac{\log(y)}{m} \quad \text{or} \quad \log b + T(x) \geq \frac{\log y}{m}$$

$$\text{Set } w = \frac{\log(y)}{m}$$

$$\Rightarrow -T(x) \geq 2(w + \log \pi)$$

$$T(x) \leq -2(w + \log \pi) \quad \text{or} \quad T(x) \geq w - \log b$$

$$\Rightarrow P(T(x) \leq -2(w + \log \pi) \text{ or } T(x) \geq w - \log b)$$

$$\text{Note } P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

Note $-2(w + \log \pi) \leq w - \log b$ when $y > 1$

$$\therefore \mathbb{P}(w - \log b \leq T(X) \leq -2(w + \log \pi)) = 0$$

$$\Rightarrow P_W(W(X)) = \mathbb{P}(T(X) \leq -2(w + \log \pi)) + \mathbb{P}(T(X) \geq w - \log b)$$

$$\text{Note } g(Y) = \frac{Y - E_{\pi} T(X)}{\sqrt{\text{Var}_{\pi} T(X)}}$$

$$g(T(X)) = Z \sim \mathcal{N}(0, 1)$$

$$\Rightarrow P_W(W(X)) = \mathbb{P}(Z \leq g(-2(w + \log \pi))) + \mathbb{P}(Z > g(w - \log b))$$

$$F(z) = \mathbb{P}(Z \leq z)$$

$$P_W(W(X)) = F(g(-2(w + \log \pi))) + 1 - F(g(w - \log b))$$

F. CONSTRUCT HYPOTHESIS TEST of POWER $\alpha^* \geq 0.05$

State decision Rule

$$\phi(x) = \begin{cases} 1, & P_W(x) < \alpha^* \\ 0, & \text{otherwise} \end{cases}$$

with $P_W(x)$ defined as above

Decision Rule describes Hypothesis test with power α^* under $H_0: \alpha^* = 1$

9 PROB OF ACCEPTING H_0 WHEN NOT TRUE

○ $\#P_{\alpha=\frac{1}{2}}(P_W(W(z)) > \alpha^*)$ and $P_{\alpha=2}(P_W(W(z)) > \alpha^*)$

Only going to answer for #

$P_{\alpha=\frac{1}{2}}(P_W(W(z)) \geq \alpha^*)$

$P_W(W(z)) \geq \alpha^*$

$\Rightarrow W(z) \geq P_W^{-1}(\alpha^*)$

○ $\Rightarrow T(z) \leq -2 \left(\frac{\log(P_W^{-1}(\alpha^*))}{m} + \log \pi \right)$

or $T(z) \geq \frac{\log(P_W^{-1}(\alpha^*))}{m} - \log 6$

Define $h(Y) = \frac{Y - E_{\alpha=\frac{1}{2}} T(z)}{\sqrt{\text{Var}_{\alpha=\frac{1}{2}} T(z)}} = \frac{Y + 2.74}{\sqrt{\frac{3.3042}{m}}}$

○ Note: $f(z | \alpha = \frac{1}{2}) = \begin{cases} \frac{(\alpha(1-\alpha))^{-\frac{1}{2}}}{\pi}, & z \in (0,1) \\ 0, & \text{otherwise} \end{cases}$

Find $E_{\alpha=\frac{1}{2}}(\cdot)$ & $\text{Var}_{\alpha=\frac{1}{2}}(\cdot)$ as before

$\Rightarrow P(P_W(W(z)) \geq \alpha^*)$

$= P(Z \leq h(\lambda(z))) + P(Z \geq h(\delta(z)))$