

Figure 1: img

(a) As  $\theta$  increased, the height of the curve decreased.

(b)

$$L(\theta) = \prod_{i=1}^{n} \frac{\theta+1}{2} X_i^2 (1 - X_i)^{\theta}$$
$$l(\theta) = \ln \left( \prod_{i=1}^{n} \frac{\theta+1}{2} X_i^2 (1 - X_i)^{\theta} \right)$$
$$l(\theta) = \sum_{i=1}^{n} \ln \left( \frac{\theta+1}{2} X_i^2 (1 - X_i)^{\theta} \right)$$

$$l(\theta) = \sum_{i=1}^{n} \left( \ln\left(\frac{\theta+1}{2}\right) + \ln\left(X_{i}^{2}\right) + \ln\left((1-X_{i})^{\theta}\right) \right)$$

$$l(\theta) = \sum_{i=1}^{n} \left( \ln\left(\frac{\theta+1}{2}\right) + 2\ln\left(X_{i}\right) + \theta\ln\left(1-X_{i}\right) \right)$$

$$l(\theta) = n \ln\left(\frac{\theta+1}{2}\right) + 2\sum_{i=1}^{n} \ln\left(X_{i}\right) + \theta\sum_{i=1}^{n} \ln\left(1-X_{i}\right)$$

$$\frac{\delta l(\theta)}{\delta \theta} = \frac{n}{\theta+1} + \sum_{i=1}^{n} \ln(1-X_{i}) = 0$$

$$\hat{\theta}_{MLE} = \frac{-n}{\sum_{i=1}^{n} \ln(1-X_{i})} - 1$$

$$\hat{\eta}_{MLE} = \sqrt{\hat{\theta}_{MLE} + 1}$$

 $\mathbf{2}$ 

(c)

(a) As  $\xi$  increased, the height of the curve decreased.

(b)
$$L(\xi) = \prod_{i=1}^{n} \frac{1}{\sqrt{2\pi}\xi} exp\left(-\frac{1}{2}\left(\frac{X_{i}-2}{\xi}\right)^{2}\right)$$

$$l(\xi) = \ln\left(\prod_{i=1}^{n} \frac{1}{\sqrt{2\pi}\xi} exp\left(-\frac{1}{2}\left(\frac{X_{i}-2}{\xi}\right)^{2}\right)\right)$$

$$l(\xi) = \sum_{i=1}^{n} \left(\ln\left(\frac{1}{\sqrt{2\pi}\xi}\right) - \frac{1}{2}\left(\frac{X_{i}-2}{\xi}\right)^{2}\right)$$

$$l(\xi) = -n \ln\left(\sqrt{2\pi}\xi\right) - \frac{1}{2}\sum_{i=1}^{n} \left(\frac{X_{i}-2}{\xi}\right)^{2}$$

$$l(\xi) = -n \ln\sqrt{2\pi} - n \ln\xi - \frac{1}{2}\sum_{i=1}^{n} \left(\frac{X_{i}-2}{\xi}\right)^{2}$$

$$\frac{\delta l(\xi)}{\delta \xi} = -\frac{n}{\xi} + \sum_{i=1}^{n} \frac{(X_{i}-2)^{2}}{\xi^{3}} = 0$$

$$\frac{1}{\xi^{3}} \sum_{i=1}^{n} (X_{i}-2)^{2} = \frac{n}{\xi}$$

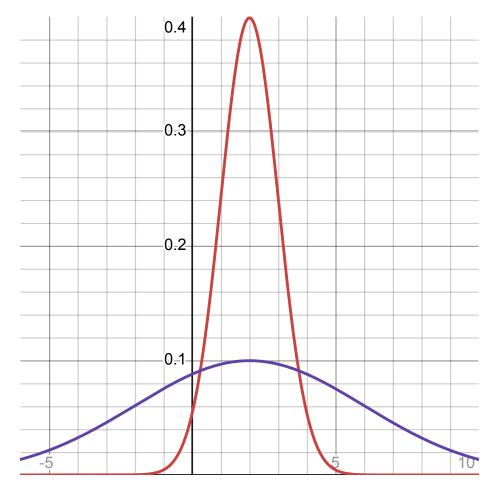


Figure 2: img

$$\sum_{i=1}^{n} (X_i - 2)^2 = n\xi^2$$

$$\hat{\xi}_{MLE} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (X_i - 2)^2}$$

$$\hat{\xi}_{MLE} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (X_i - 2)^2}$$

(c) 
$$\hat{\alpha}_{MLE} = log(\hat{\xi}_{MLE})$$

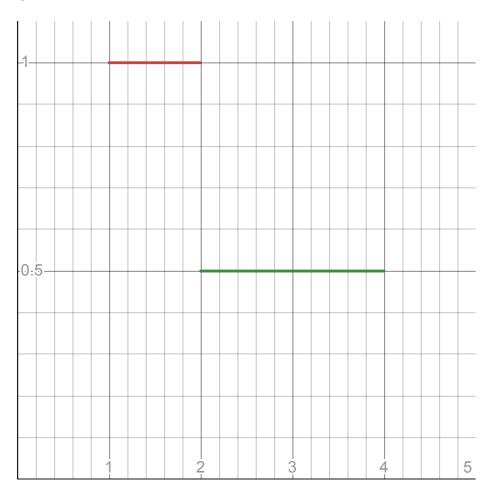


Figure 3: img

(a) As  $\beta$  increases, the line is lower but is also longer.

(b) 
$$L(\beta) = \begin{cases} \left(\frac{1}{\beta}\right)^n & \text{if } \beta < \min(X_i) \text{ and } \max(X_i) < 2\beta \\ 0 & \text{otherwise} \end{cases}$$
 
$$\frac{\max(X_i)}{2} < \beta < \min(X_i)$$

We know that  $L(\beta)$  is monotonic. To maximize  $L(\beta)$ , we want to minimize  $\beta$ .

$$\hat{\beta}_{MLE} = \begin{cases} \frac{max(X_i)}{2} & \text{if } \frac{max(X_i)}{2} < min(X_i) \\ DNE & \text{otherwise} \end{cases}$$

4

(a) As  $\theta$  increases, the line is higher and also shorter.

(b) 
$$L(\theta) = \begin{cases} \frac{1}{1-\theta} & \text{if } \theta < min(X_i) \text{ and } max(X_i) < 1 \\ 0 & \text{otherwise} \end{cases}$$

To maximize  $L(\theta)$ , we want  $\theta$  to be as close to 1 as possible.

$$\hat{\theta}_{MLE} = \begin{cases} min(X_i) & \text{if } max(X_i) < 1\\ DNE & \text{otherwise} \end{cases}$$

5

(a) As  $\sigma$  increases, the graph starts more towards the right and is higher.

(b) 
$$F(x) = \int_{-\infty}^{x} \frac{3\sigma^{3}}{z^{4}} dz$$

$$F(x) = \int_{\sigma}^{x} \frac{3\sigma^{3}}{z^{4}} dz$$

$$F(x) = 3\sigma^{3} \int_{\sigma}^{x} \frac{1}{z^{4}} dz$$

$$F(x) = \left[3\sigma^{3} * \frac{-1}{3z^{3}}\right]_{\sigma}^{x}$$

$$F(x) = \left[\frac{-\sigma^{3}}{z^{3}}\right]_{\sigma}^{x}$$

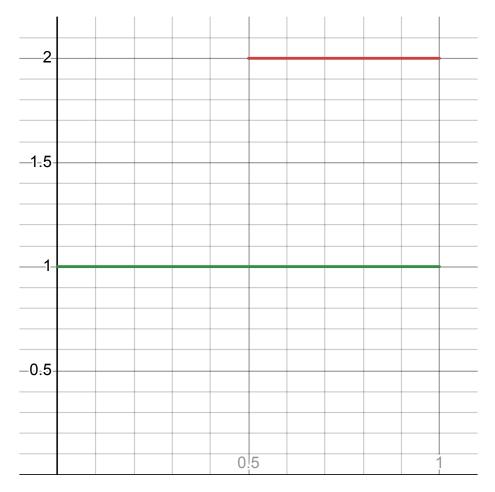


Figure 4: img

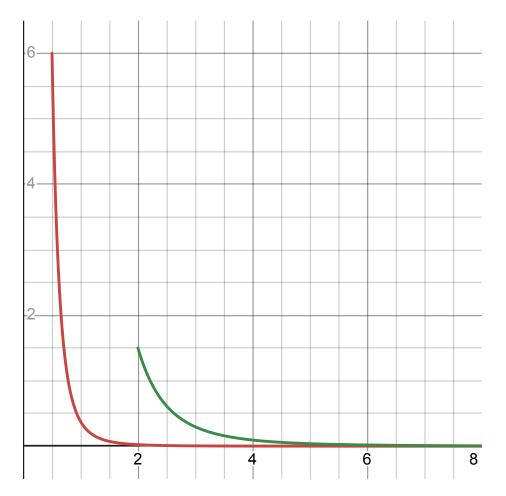


Figure 5: img

$$F(x) = \frac{-\sigma^3}{x^3} + 1$$
 
$$F(x) = \begin{cases} 1 - \frac{\sigma^3}{x^3} & \text{if } x > \sigma \\ 0 & \text{otherwise} \end{cases}$$

(c)
$$E[X_1] = \int_{\sigma}^{\infty} x \, \frac{3\sigma^3}{x^4} \, dx = \int_{\sigma}^{\infty} \frac{3\sigma^3}{x^3} \, dx$$

$$E[X_1] = 3\sigma^3 \int_{\sigma}^{\infty} \frac{1}{x^3} \, dx$$

$$E[X_1] = 3\sigma^3 \left[\frac{-1}{2x^2}\right]_{\sigma}^{\infty}$$

$$E[X_1] = \frac{3\sigma^3}{2\sigma^2}$$

$$E[X_1] = \frac{3\sigma}{2}$$

(d) 
$$L(\sigma) = \begin{cases} \prod_{i=1}^n \frac{3\sigma^3}{X_i^4} & \text{if } min(X_i) > \sigma \\ 0 & \text{otherwise} \end{cases}$$

To maximize  $L(\sigma)$ , we want to maximize  $\sigma$ .

$$\hat{\sigma}_{MLE} = min(X_i)$$

$$\pi_2 = 4\pi_1$$

$$\pi_1 + \pi_2 + \pi_3 = 1$$

$$(\hat{\pi}_1)_{MLE} = ?$$

$$\pi_3 = 1 - \pi_2 - \pi_1$$

$$\pi_3 = 1 - 4\pi_1 - \pi_1$$

$$\pi_3 = 1 - 5\pi_1$$

$$\begin{split} L(\pi) &= \pi_1^{Y_1} \pi_2^{Y_2} \pi_3^{Y_3} \\ L(\pi) &= \pi_1^{Y_1} (4\pi_1)^{Y_2} (1 - 5\pi_1)^{Y_3} \\ l(\pi) &= log(\pi_1^{Y_1}) + log((4\pi_1)^{Y_2}) + log((1 - 5\pi_1)^{Y_3}) \end{split}$$

$$\begin{split} l(\pi) &= Y_1 log(\pi_1) + Y_2 log(4\pi_1) + Y_3 log(1 - 5\pi_1) \\ l(\pi) &= Y_1 log(\pi_1) + Y_2 log(4) + Y_2 log(\pi_1) + Y_3 log(1 - 5\pi_1) \\ l(\pi) &= (Y_1 + Y_2) log(\pi_1) + Y_3 log(1 - 5\pi_1) + Y_2 log(4) \\ &\frac{\delta l(\pi)}{\delta \pi} = \frac{Y_1 + Y_2}{\pi_1} - \frac{5Y_3}{1 - 5\pi_1} = 0 \\ &\frac{Y_1 + Y_2}{\pi_1} = \frac{5Y_3}{1 - 5\pi_1} \\ &(Y_1 + Y_2)(1 - 5\pi_1) = 5Y_3\pi_1 \\ &Y_1 + Y_2 - 5\pi_1 Y_1 - 5\pi_1 Y_2 = 5Y_3\pi_1 \\ &Y_1 + Y_2 = \pi_1 * 5(Y_1 + Y_2 + Y_3) \\ &(\hat{\pi}_1)_{MLE} = \frac{Y_1 + Y_2}{5(Y_1 + Y_2 + Y_3)} = \frac{Y_1 + Y_2}{5n} \end{split}$$

$$\pi_2 = 2\pi_1$$

$$\pi_4 = 3\pi_1$$

$$\pi_1 + \pi_2 + \pi_3 + \pi_4 = 1$$

$$L(\pi_1) = \pi_1^{Y_1} \pi_2^{Y_2} \pi_3^{Y_3} \pi_4^{Y_4}$$

(a)

$$(\hat{\pi}_1)_{MLE} = ?$$

$$\pi_3 = 1 - \pi_4 - \pi_2 - \pi_1$$

$$\pi_3 = 1 - 3\pi_1 - 2\pi_1 - \pi_1$$

$$\pi_3 = 1 - 6\pi_1$$

$$L(\pi_1) = \pi_1^{Y_1} (2\pi_1)^{Y_2} (1 - 6\pi_1)^{Y_3} (3\pi_1)^{Y_4}$$

$$l(\pi_1) = Y_1 log(\pi_1) + Y_2 log(2\pi_1) + Y_3 log(1 - 6\pi_1) + Y_4 log(3\pi_1)$$

$$l(\pi_1) = (Y_1 + Y_2 + Y_4) log(\pi_1) + Y_2 log(2) + Y_3 log(1 - 6\pi_1) + Y_4 log(3)$$

$$\frac{\delta l(\pi_1)}{\delta \pi_1} = \frac{Y_1 + Y_2 + Y_4}{\pi_1} - \frac{6Y_3}{1 - 6\pi_1} = 0$$

$$\frac{Y_1 + Y_2 + Y_4}{\pi_1} = \frac{6Y_3}{1 - 6\pi_1}$$

$$(Y_1 + Y_2 + Y_4)(1 - 6\pi_1) = 6Y_3\pi_1$$

$$Y_1 + Y_2 + Y_4 = 6n\pi_1$$

$$(\hat{\pi}_1)_{MLE} = \frac{Y_1 + Y_2 + Y_4}{6n}$$

(b) 
$$(\hat{\pi}_3)_{MLE} = ?$$

$$\pi_1 = \frac{1 - \pi_3}{6}$$

$$\pi_2 = \frac{1 - \pi_3}{3}$$

$$\pi_4 = \frac{1 - \pi_3}{2}$$

$$\begin{split} L(\pi_3) &= \left(\frac{1-\pi_3}{6}\right)^{Y_1} \left(\frac{1-\pi_3}{3}\right)^{Y_2} \pi_3^{Y_3} \left(\frac{1-\pi_3}{2}\right)^{Y_4} \\ l(\pi_3) &= Y_1 \, \log\left(\frac{1-\pi_3}{6}\right) + Y_2 \, \log\left(\frac{1-\pi_3}{3}\right) + Y_3 \, \log(\pi_3) + Y_4 \, \log\left(\frac{1-\pi_3}{2}\right) \\ l(\pi_3) &= (Y_1 + Y_2 + Y_4) \, \log\left(1-\pi_3\right) + Y_3 \, \log(\pi_3) - Y_1 \, \log(6) - Y_2 \, \log(3) - Y_4 \, \log(2) \\ \frac{\delta l(\pi_3)}{\delta \pi_3} &= \frac{-(Y_1 + Y_2 + Y_4)}{1-\pi_3} + \frac{Y_3}{\pi_3} = 0 \\ \frac{Y_3}{\pi_3} &= \frac{Y_1 + Y_2 + Y_4}{1-\pi_3} \\ Y_3(1-\pi_3) &= (Y_1 + Y_2 + Y_4)\pi_3 \\ Y_3 &= (Y_1 + Y_2 + Y_3 + Y_4)\pi_3 \\ (\hat{\pi}_3)_{MLE} &= \frac{Y_3}{\pi} \end{split}$$

$$L(\theta) = \begin{cases} \frac{1}{\theta^{27}} & \text{if } \max(X_i) \le \theta \\ 0 & \text{otherwise} \end{cases}$$

$$max(X_i) = max(9.2, 3) = 9.2$$

To maximize  $L(\theta)$ , we want to minimize  $\theta$ .

$$\hat{\theta}_{MLE} = 9.2$$