

EXERCISE 1

① A histogram is a suitable representation for this variable. Boxplot is also good.

Mean and median are similar (7.34 and 7.42 respectively).

The dispersion can be described by the standard deviation of 1.498.

The distribution is pretty symmetric, as $Q_1 = 6.358$ and $Q_3 = 8.418$ have a similar distance from the median, the whiskers in the boxplot have similar length. We only note few lower outliers.

② quantile (Sleep Quality, 0.95) = 9.668

quantile (Sleep Quality, 0.90) = 9.26 ③

③ $Q_1 - 1.5(Q_3 - Q_1) = 3.267$ is the threshold

There are 3 outliers: 3.198778, 2.390676, 3.200246

④ We investigate through a scatterplot and calculating the sample correlation $r = 0.3884$

Plausibly there a mild positive linear association

⑤ Percentage of subjects who suffer from sleep disorder "insomnia" or "other" among normal BMI: $(19+7)/(200+9+7) = 7.4\%$

among overweight BMI: $(64+65)/(64+65+19) = 87\%$

Clearly higher among overweight BMI!

EXERCISE 2

① $W \sim N(250, 100^2)$

$p_{\text{norm}}(0, 250, 100) = 0.0062 = p$

② $q_{\text{norm}}(0.005, 250, 100) = -7.5829$ $q_{\text{norm}}(0.001, 250, 100) = -19.023$

\wedge loss of 7.5829

\wedge loss of -19.023

③ X := number of investment resulting in a financial loss out of $n=100$

$X \sim \text{Binomial}(n, p)$

$X \sim \text{Poisson}(np)$ approximately

i) $1 - p_{\text{binom}}(3, n, p) = 0.0036$

$1 - p_{\text{poiss}}(3, np) = 0.0038$

both good

④

$1 - p_{\text{binom}}(2, n, p) = 0.248$

$1 - p_{\text{poiss}}(2, np) = 0.252$

both good

ii) $\sum W_i \sim N(250, 100)$

$p_{\text{norm}}(0, 250, 10) \approx 0$

$\sum W_i \sim N(250, 100)$

$p_{\text{norm}}(0, 250, 10) \approx 0$

$$\textcircled{B} P(T-20 > 40) \leq \dots \leq \frac{100}{40^2} = 0.0625$$

EXERCISE 3

$$\textcircled{a} P(T > 60) = P(T-15 > 45) \leq P(|T-15| > 45) \leq \frac{\text{Var}(T)}{45^2} = \frac{100}{45^2} = 4.94\%$$

$$\textcircled{b} P(T \geq C) \leq \dots \leq \frac{\text{Var}(T)}{(C-15)^2} = 0.01$$

$$C = \left(\frac{100}{0.01}\right)^{1/2} + 15 = 115$$

$$\textcircled{c} C = \left(\frac{100}{0.01}\right)^{1/2} + 20 = 120$$

$$\textcircled{2} P(XY=0) = p \rightarrow \text{FALSE}$$

$$E[XY] = p \rightarrow \text{TRUE}$$

\textcircled{d} As $X \sim \text{Bernoulli}(p)$ $Y \sim \text{Exp}(1)$, see notes:

$$E[X] = p \quad \text{Var}(X) = p(1-p) \quad E[Y] = 1 \quad \text{Var}(Y) = 1$$

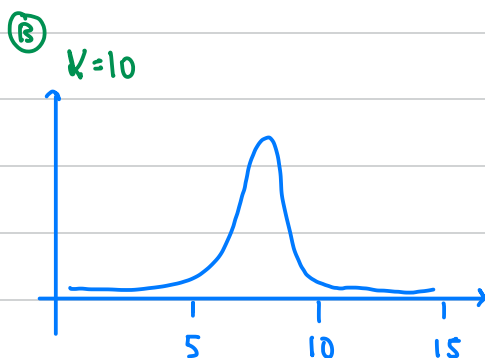
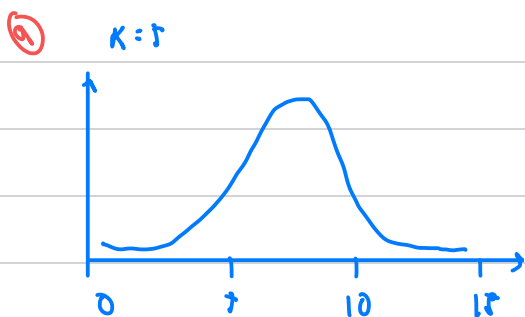
$$\text{Var}(XY) = E[X^2 Y^2] - E^2[XY]$$

$$= E[X^2] E[Y^2] - E^2[X] E^2[Y]$$

$$= (E[X] + \text{Var}(X)) (E[Y] + E^2[Y]) - E^2[X] E^2[Y]$$

$$= (p(1-p) + p^2)(2) - p^2 = 2p - p^2$$

EXERCISE 4



By trying different values of K (or studying briefly the function):

- as K increases the mode slightly increases towards $\lambda=8$
- as K increases the distribution becomes more concentrated around the mode.

(b)
$$l(K) = \prod_{i=1}^n P(x_i; K, \lambda) = \prod_{i=1}^n \frac{K}{\lambda} \left(\frac{x_i}{\lambda}\right)^{K-1} e^{-(x_i/\lambda)^K}$$

$$L(K) = -\log(l(K)) = -\sum \left[\log K - \log \lambda + (K-1) \log x_i - (K-1) \log \lambda - \left(\frac{x_i}{\lambda}\right)^K \right]$$

(c)
$$L'(K) = -\sum \left[\frac{1}{K} + \log x_i - \log \lambda + \left(\frac{x_i}{\lambda}\right)^K \log \left(\frac{x_i}{\lambda}\right) \right]$$

to obtain the MLE \hat{K} the condition is $L'(K)=0$

(d) Using R we evaluate: (B)

$$L(8) = 521.4 \quad L(8.5) = 517.4$$

$$L(9) = 515.9 \quad L(9.5) = 515.6$$

$$L(10) = 516.6 \quad L(10.5) = 518.9$$

These are preferable,

lowest value of L

hence highest likelihood!

② Through grid search in R , evaluate $L(k)$ for e.g

$$k = 8, 8.01, 8.02, \dots, 10.99, 11$$

to find $\hat{k} = 9.36$

③ Compute $\bar{x} = \text{mean}(\text{Sleep} \# \text{ Sleep Duration}) = 7.13$

Again we use grid search in R , for values of k and setting $\lambda = \frac{\bar{x}}{\Gamma(1 + \frac{1}{k})}$ every time to satisfy the constraint.

Obtain $\hat{k} = 10.15$ $\hat{\lambda} = 7.49$

EXERCISE 5

(a) i) $H_0: p \leq 0.35 = p_0$

$H_1: p > 0.35$

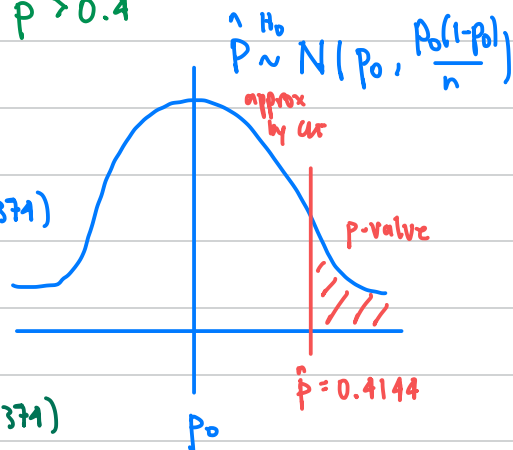
ii) $\hat{p} = \frac{77+78}{374} = 0.4144$

p-value = $1 - \text{pnorm}(\hat{p}, 0.35, \sqrt{0.35 \cdot 0.65 / 374})$
 $= 0.0015$ reject H_0 (iv)

(B)

$H_0: p \leq 0.4 = p_0$

$H_1: p > 0.4$



(B)

p-value = $1 - \text{pnorm}(\hat{p}, 0.4, \sqrt{0.4 \cdot 0.6 / 374})$
 $= 0.2843$ do not reject H_0 (iv)

(iii) the p-value is the probability to obtain a \hat{p} as extreme or more extreme than $\hat{p} = 0.4144$ if the null hypothesis H_0 is true.

(b) $H_0: \mu_D = \mu_N$
 $H_1: \mu_D < \mu_N$

$\bar{x}_D = 6.97$ $\bar{x}_N = 7.06$

$s_D^2 = 0.746$ $s_N^2 = 0.98$

$n_D = 71$ $n_N = 73$

p-value = $\text{pnorm}(\bar{x}_D - \bar{x}_N, 0, \sqrt{\frac{s_D^2}{n_D} + \frac{s_N^2}{n_N}}) = 0.2747$

↓
do not reject H_0 !

