

EE 503

Homework 10 Solution

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Problem 1. Suppose the random variables X_i , $i = 1, 2, \dots, n$ are uncorrelated and have the same mean μ and variance σ^2 . Define the sample mean \bar{X} as

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

and the sample variance \bar{V} as

$$\bar{V} = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2.$$

Show

- $E(\bar{X}) = \mu.$
- $Var(\bar{X}) = \sigma^2/n.$
- $E(\bar{V}) = \sigma^2.$

Solution:

- Since $\{X_i\}_{i=1}^n$ is a sequence of iid random variables with mean $\mathbf{E}[X_i] = \mu$ and variance $\text{var}(X_i) = \sigma^2$, $\forall i \in \{1, 2, \dots, n\}$, the expectation of the sample mean \bar{X} is

$$\mathbf{E}[\bar{X}] = \frac{1}{n} \sum_{i=1}^n \mathbf{E}[X_i] = \frac{1}{n} \cdot n\mu = \mu,$$

b. while the variance of the sample mean \bar{X} is

$$\begin{aligned}
 \text{var}(\bar{X}) &= \mathbf{E}[(\bar{X} - \mu)^2] \\
 &= \mathbf{E}\left[\left(\frac{1}{n} \sum_{i=1}^n (X_i - \mu)\right)^2\right] \\
 &= \frac{1}{n^2} \mathbf{E}\left[\left(\sum_{i=1}^n (X_i - \mu)\right)^2\right] \\
 &= \frac{1}{n^2} (n\sigma^2 + n(n-1) \cdot 0), \text{ only } n \text{ out of } n^2 \text{ terms are nonzero} \\
 &= \frac{\sigma^2}{n}.
 \end{aligned}$$

Note that $n(n-1)$ terms in the expectation are zeros due to independence.

c. The expectation of the sample variance \bar{V} is

$$\begin{aligned}
 \mathbf{E}[\bar{V}] &= \mathbf{E}\left[\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2\right] \\
 &= \frac{1}{n-1} \mathbf{E}\left[\sum_{i=1}^n ((X_i - \mu) - (\bar{X} - \mu))^2\right] \\
 &= \frac{1}{n-1} \sum_{i=1}^n \mathbf{E}[(X_i - \mu)^2 - 2(X_i - \mu)(\bar{X} - \mu) + (\bar{X} - \mu)^2] \\
 &= \frac{1}{n-1} \sum_{i=1}^n \left(\sigma^2 - 2\mathbf{E}[(X_i - \mu)(\bar{X} - \mu)] + \frac{\sigma^2}{n}\right).
 \end{aligned}$$

Notice that

$$\begin{aligned}
 \mathbf{E}[(X_i - \mu)(\bar{X} - \mu)] &= \mathbf{E}\left[(X_i - \mu) \frac{1}{n} \sum_{j=1}^n (X_j - \mu)\right] \\
 &= \frac{1}{n} \sigma^2,
 \end{aligned}$$

since $\mathbf{E} \left[(X_i - \mu) \frac{1}{n} \sum_{j=1, j \neq i}^n (X_j - \mu) \right] = 0$. Finally,

$$\mathbf{E}[\bar{V}] = \frac{1}{n-1} \sum_{k=1}^n \left(\sigma^2 - \frac{2\sigma^2}{n} + \frac{\sigma^2}{n} \right) = \sigma^2.$$

Problem 2. Suppose the random variable X is normally distributed with mean 2 and variance 9. Find

- $P(X < 5)$.
- $P(X > -1)$.
- $P(-1 < X < 5)$.
- $P(X < 10)$.

Solution:

Let $\Phi(z)$ be the standard normal cdf, i.e. $Z \sim \mathcal{N}(0, 1)$ and

$$\Phi(z) = F_Z(z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt.$$

- $P[X < 5] = \Phi\left(\frac{5-2}{3}\right) = \Phi(1) = 0.8413$.
- $P[X > -1] = 1 - \Phi\left(\frac{-1-2}{3}\right) = 1 - \Phi(-1) = 0.8413$.
- $P[-1 < X < 5] = \Phi(1) - \Phi(-1) = 0.6827$.
- $P[X < 10] = \Phi\left(\frac{10-2}{3}\right) = 0.9962$.

Problem 3. Let X be binomially distributed with $n = 50$ and $p = \frac{1}{3}$. Find

- $P(X = 17)$.
- $P(X \leq 20)$.

Solution:

- $P[X = 17] = \binom{50}{17} p^{17} (1-p)^{33} = 0.1178$.

b. $P[X \leq 20] = \sum_{k=0}^{20} \binom{50}{k} p^k (1-p)^{50-k} = 0.8741$.

Problem 4. We can think of X in Problem 3 as the sum of 50 Bernoulli random variables with success probability $p = \frac{1}{3}$. So let us use the CLT to approximate our answers in Problem 3. Using the CLT find

a. $P(X = 17)$.

b. $P(X \leq 20)$.

Remember to use continuity correction as we discussed in our class example.

Solution:

Using CLT with $\frac{S-\mu_S}{\sigma_S} \sim \mathcal{N}(0, 1)$, $\mu_S = np = \frac{50}{3}$ and $\sigma_S = \sqrt{np(1-p)} = \frac{10}{3}$, we have

a. $P[X = 17] \approx P\left[\frac{16.5-\mu_S}{\sigma_S} \leq Z \leq \frac{17.5-\mu_S}{\sigma_S}\right] = 0.1186$,

b. $P[X \leq 20] \approx P\left[Z \leq \frac{20.5-\mu_S}{\sigma_S}\right] = 0.8749$.