

1. A bottling machine can be regulated so that it discharges an average of μ ounces per bottle. It has been observed that the amount of fill dispensed by the machine is normally distributed with $\sigma = 1$ ounce. A sample of $n = 9$ filled bottles is randomly selected from the output of the machine on a given day (all bottled with the same machine setting) and the ounces of fill are measured for each.

- (a) Find the probability that the sample mean will be within .3 ounces of the true mean for the chosen machine setting? (Use a probability table).

Solution: Let Y_1, \dots, Y_9 denote the ounces of fill to be observed. We know that the random variables Y_i , $1 \leq i \leq 9$ are all normally distributed with mean μ and variance $\sigma^2 = 1$. According to corollary 5.6.2, the sample mean \bar{Y} is normally distributed with mean μ and variance $1/9$. We want to find

$$\begin{aligned} P(|\bar{Y} - \mu| \leq 0.3) &= P(-0.3 \leq \bar{Y} - \mu \leq 0.3) \\ &= P\left(\frac{-0.3}{1/\sqrt{9}} \leq \frac{\bar{Y} - \mu}{1/\sqrt{9}} \leq \frac{0.3}{1/\sqrt{9}}\right) \\ &= P(-0.9 \leq Z \leq 0.9) \text{ where } Z \sim \mathcal{N}(0, 1) \\ &\approx 0.6318 \text{ (from a probability table.)} \end{aligned}$$

- (b) How many observations should be included in the sample if we wish \bar{Y} to be within 0.3 ounces of μ with probability 0.95? (Use a probability table.)

Solution: Now we want

$$P(|\bar{Y} - \mu| \leq 0.3) = P(-3 \leq \bar{Y} - \mu \leq 0.3) = 0.95.$$

Dividing each term of the inequality by $\sigma_{\bar{Y}} = 1/\sqrt{n}$, we have

$$P(-0.3\sqrt{n} \leq Z \leq 0.3\sqrt{n}) = 0.95, \text{ where } Z \sim \mathcal{N}(0, 1).$$

By using a probability table, we find

$$P(-1.96 \leq Z \leq 1.96) = 0.95.$$

It must follow that $0.3\sqrt{n} = 1.96$ or, equivalently, $n = \left(\frac{1.96}{0.3}\right)^2 \approx 42.68$. Practically, we should include 42 observations.

2. Suppose that a particle is located at the origin of the 3D-space at $t = 0$. Denote (X, Y, Z) the coordinates of the particle at any time $t > 0$. The random variables X , Y and Z are assumed to be i.i.d. and, at any time $t > 0$, each of them has the normal distribution with mean 0 and variance $\sigma^2 t$. Find the probability that, at $t = 2$, the particle will lie within a sphere whose center is at the origin and whose radius is 4σ .

Solution: At $t = 2$, the variables X , Y and Z are independent from the normal distribution $\mathcal{N}(0, 2\sigma^2)$ hence the variables $\frac{X}{\sqrt{2}\sigma}$, $\frac{Y}{\sqrt{2}\sigma}$ and $\frac{Z}{\sqrt{2}\sigma}$ are independent from the standard normal distribution and the variable $\frac{X^2 + Y^2 + Z^2}{2\sigma^2}$ follows a χ^2 -distribution with 3 degrees of freedom. Thus

$$\begin{aligned} P(\text{At } t = 2, \text{ particle lies within a sphere centered at origin with radius is } 4\sigma) &= P(X^2 + Y^2 + Z^2 \leq 16\sigma^2) \\ &= P\left(\frac{X^2 + Y^2 + Z^2}{2\sigma^2} \leq 8\right) \\ &= \Phi_{\chi^2(3)}(8) \\ &\approx 0.954. \end{aligned}$$

3. Suppose that X_1, \dots, X_n form a random sample from the normal distribution with mean μ and variance σ^2 . Assume that the sample size is $n = 16$. Determine the values of

(a) $P\left(\frac{\sigma^2}{2} \leq \frac{1}{n} \sum_{i=1}^n (X_i - \mu)^2 \leq 2\sigma^2\right)$

Solution: Notice that, for all $1 \leq i \leq 16$, $\frac{(X_i - \mu)^2}{\sigma^2} \sim \mathcal{N}(0, 1)$ so $\sum_{i=1}^n \frac{(X_i - \mu)^2}{\sigma^2} \sim \chi^2(16)$. Hence, when $n = 16$,

$$\begin{aligned} P\left(\frac{\sigma^2}{2} \leq \frac{1}{n} \sum_{i=1}^n (X_i - \mu)^2 \leq 2\sigma^2\right) &= P\left(\frac{n}{2} \leq \sum_{i=1}^n \frac{(X_i - \mu)^2}{\sigma^2} \leq 2n\right) \\ &= \Phi_{\chi^2(16)}(32) - \Phi_{\chi^2(16)}(8) \\ &\approx 0.939 \end{aligned}$$

(b) $P\left(\frac{\sigma^2}{2} \leq \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n)^2 \leq 2\sigma^2\right)$.

Solution: By theorem 8.3.1, $\sum_{i=1}^n \frac{(X_i - \bar{X}_n)^2}{\sigma^2} \sim \chi^2(n - 1)$. Hence, when $n = 16$,

$$\begin{aligned} P\left(\frac{\sigma^2}{2} \leq \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n)^2 \leq 2\sigma^2\right) &= P\left(\frac{n}{2} \leq \sum_{i=1}^n \frac{(X_i - \bar{X}_n)^2}{\sigma^2} \leq 2n\right) \\ &= \Phi_{\chi^2(15)}(32) - \Phi_{\chi^2(15)}(8) \\ &\approx 0.918 \end{aligned}$$

4. The tensile strength for a type of wire is normally distributed with unknown mean μ and unknown variance σ^2 . Six pieces of wire are randomly selected from a large roll; Y_i , the tensile strength for portion i , is measured for $1 \leq i \leq 6$. The population mean μ and variance σ^2 can be estimated by \bar{Y} and S^2 , respectively. Because $\sigma_{\bar{Y}}^2 = \frac{\sigma^2}{n}$ (corollary 5.6.2), it follows that $\sigma_{\bar{Y}}^2$ can be estimated by $\frac{S^2}{n}$. Find the approximate probability that \bar{Y} will be within $\frac{2S}{\sqrt{n}}$ of the true population mean μ .

Solution: We want

$$P\left(-\frac{2S}{\sqrt{n}} \leq \bar{Y} - \mu \leq \frac{2S}{\sqrt{n}}\right) = P\left(-2 \leq \sqrt{n} \frac{\bar{Y} - \mu}{S} \leq 2\right) = P(-2 \leq T \leq 2)$$

where T has a t -distribution with, in this case, $n - 1$ df. From a probability table, we can see that

$$P(-2.015 \leq T \leq 2.015) = 0.90.$$

Therefore, the probability that \bar{Y} is within 2 estimated standard deviation of μ is slightly less than 0.9.

5. Suppose that Y_1, Y_2 and Y_3 form a random sample from an exponential distribution with parameter θ . Consider the estimators $\hat{\theta}_1 = Y_1$, $\hat{\theta}_2 = \frac{1}{2}(Y_1 + Y_2)$ and $\hat{\theta}_3 = \bar{Y}$.

- (a) Show that $\hat{\theta}_1, \hat{\theta}_2$ and $\hat{\theta}_3$ are unbiased estimators for $\frac{1}{\theta}$.

Solution: $E(\hat{\theta}_1) = E(Y_1) = 1/\theta$, $E(\hat{\theta}_2) = \frac{1}{2}(E(Y_1) + E(Y_2)) = 2/2\theta = 1/\theta$ and $E(\hat{\theta}_3) = \frac{1}{3}(E(Y_1) + E(Y_2) + E(Y_3)) = 3/3\theta = 1/\theta$ so $\hat{\theta}_1, \hat{\theta}_2$ and $\hat{\theta}_3$ are unbiased estimators for $\frac{1}{\theta}$.

- (b) Which of $\hat{\theta}_1, \hat{\theta}_2$ and $\hat{\theta}_3$ has the least variance?

Solution: $\text{Var}(\hat{\theta}_1) = \text{Var}(Y_1) = 1/\theta^2$, $\text{Var}(\hat{\theta}_2) = (1/4) \cdot (\text{Var}(Y_1) + \text{Var}(Y_2)) = 1/2\theta^2$ and $\text{Var}(\hat{\theta}_3) = (1/3) \cdot (\text{Var}(Y_1) + \text{Var}(Y_2) + \text{Var}(Y_3)) = 1/3\theta^2$ so $\hat{\theta}_3$ has the smallest variance, that is the smallest MSE and, therefore, is the best estimator.

6. Suppose that Y_1, Y_2 and Y_3 form a random sample from a uniform distribution on the interval $(\theta, \theta + 1)$.

- (a) Calculate the bias for \bar{Y} as an estimator of θ .

Solution: $\text{Bias}(\bar{Y}) = E(\bar{Y}) - \theta = (2\theta + 1)/2 - \theta = 1/2$.

- (b) Calculate the bias for $\bar{Y} - \frac{1}{2}$ as an estimator of θ .

Solution: $\text{Bias}(\bar{Y} - 1/2) = E(\bar{Y}) + 1/2 - \theta = 0$.

- (c) Calculate the MSE for \bar{Y} as an estimator of θ .

Solution: $\text{MSE}(\bar{Y}) = \text{Var}(\bar{Y}) + (\text{Bias}(\bar{Y}))^2 = \frac{1}{9} \cdot \frac{3}{12} + \frac{1}{4} = \frac{5}{18}$.