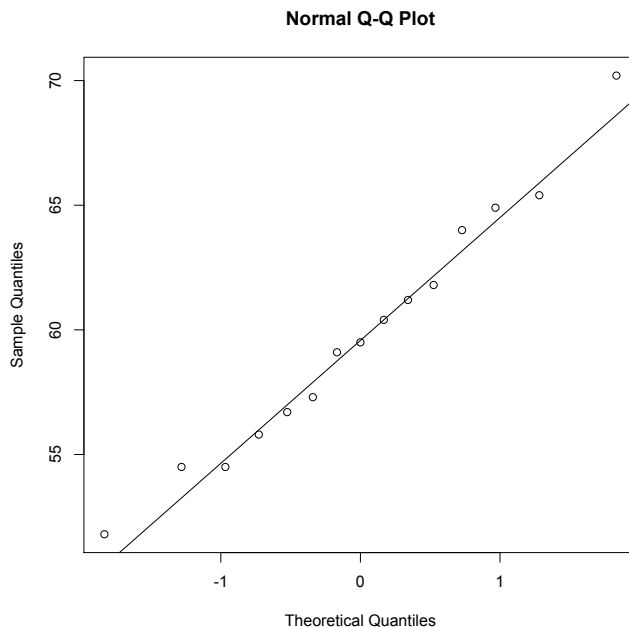


1. In a study of water absorption properties of cotton fabric, the following measurements were made of percentage of water pickup rates for 15 samples. Find a 95% upper confidence bound for μ the mean absorption percentage. What assumptions are needed on the data in order that your confidence interval be valid? Comment on how well the data satisfies the assumptions.

```
> x <- scan()
1: 51.8 59.5 59.1 61.8 61.2 55.8 57.3 64.9 65.4 54.5 54.5 60.4 64.0 70.2 56.7
16:
Read 15 items
> mean(x); sd(x)
[1] 59.80667
[1] 4.943317
> qqnorm(x); qqline(x)
```



As this is a small sample with σ unknown, we use the confidence interval for the mean coming from the t -distribution. We have $\nu = n - 1 = 14$ degrees of freedom and a confidence level of $\alpha = .05$, so the critical value for a one-sided interval is $t_{\alpha, \nu} = t_{.05, 14} = 1.761$ from Table A5. Thus the one-sided upper CI for μ is

$$\mu < \bar{X} + t_{\alpha, \nu} \frac{s}{\sqrt{n}} = 59.807 + 1.761 \cdot \frac{4.933}{\sqrt{15}} = \boxed{62.05}.$$

For the t -distribution based interval to be valid, we need to assume that the underlying distribution is approximately normal. Judging from the normal $P - P$ plot, the points line up nicely, so there is no indication that the data is not normal.

2. A 2002 study in *Journal of Environmental Engineering* looked at landfills containing demolition waste. It found that 33 of 42 such sites contained detectable levels of chromium. Find a 90% lower confidence interval on the probability that such a site contains detectable levels of chromium. How big should the sample size n be so that the lower bound is within .05 of \hat{p} ?

The estimator for p is $\hat{p} = \frac{X}{n} = \frac{33}{42} = \frac{11}{14} = 0.2619048$ where $n = 42$. $\hat{q} = \frac{9}{42}$ so that $n\hat{q} = 9$. Therefore the condition $n\hat{p} \geq 10$ and $n\hat{q} \geq 10$ fails and we must use the unrestricted score interval. We want the confidence level $\alpha = .10$ so for a one-sided interval the critical value is $z_\alpha = z_{0.10} = 1.282$ from Table A5. The one-sided lower confidence interval is

$$p > \frac{\hat{p} + \frac{z_\alpha^2}{2n} - z_\alpha \sqrt{\frac{\hat{p}\hat{q}}{n} + \frac{z_\alpha^2}{4n^2}}}{1 + \frac{z_\alpha^2}{n}} = \frac{\frac{11}{14} + \frac{(1.282)^2}{2 \cdot 42} - (1.282) \sqrt{\frac{\frac{11}{14} \cdot \frac{3}{14}}{42} + \frac{(1.282)^2}{4 \cdot 42^2}}}{1 + \frac{(1.282)^2}{42}} = \boxed{0.695}.$$

One could use Formula 7.12 with $w = .10 = 2 \cdot (.05)$, or its simplified approximation, but we expect that for small width, the sample size will be large and the traditional interval would be appropriate. Thus, we want the lower bound L within .05 of \hat{p} or $|L - \hat{p}| \leq .05$. The lower bound would be

$$p > L = \hat{p} - z_\alpha \sqrt{\frac{\hat{p}\hat{q}}{n}}$$

so

$$.05 \geq |L - \hat{p}| = z_\alpha \sqrt{\frac{\hat{p}\hat{q}}{n}}$$

or

$$n \geq \frac{z_\alpha^2 \hat{p}\hat{q}}{(.05)^2}.$$

This is implied if we take

$$n = \frac{.25 z_\alpha^2}{(.05)^2} = 164.4.$$

because $.25 \geq \hat{p}\hat{q}$ for any $\hat{p} = 1 - \hat{q}$. Thus a sample size of $\boxed{n = 165}$ would give the desired width.

3. Suppose that an urn contains three red balls, one white ball and two blue balls. If three balls are drawn at random without replacement, let X be the number of red balls and Y the number of white balls drawn. Then the joint probability mass function is given by $p(x, y)$. Let $Z = \max\{X, Y\}$ be the maximum of the two numbers. Find the pmf for the random variable Z . What is $E(Z)$? Are X and Y independent? Why? Find $\text{Cov}(X, Y)$.

$p(x, y)$		x				$p_Y(y)$
		0	1	2	3	
y	0	0	.15	.30	.05	.5
	1	.05	.30	.15	0	.5
$p_X(x)$.05	.45	.45	.05	

The pmf for $Z = \max(X, Y)$ is given for all possible values of $z \in \{0, 1, 2, 3\}$. $p_Z(0) = P(\{Z = 0\}) = P(\{(0, 0)\}) = p(0, 0) = 0$. $p_Z(1) = P(\{Z = 1\}) = P(\{(1, 0), (0, 1), (1, 1)\}) = p(1, 0) + p(0, 1) + p(1, 1) = .5$. $p_Z(2) = P(\{Z = 2\}) = P(\{(2, 0), (2, 1)\}) = p(2, 0) + p(2, 1) = .45$. $p_Z(3) = P(\{Z = 3\}) = P(\{(3, 0), (3, 1)\}) = p(3, 0) + p(3, 1) = .05$. So

z	0	1	2	3
$p_Z(z)$	0	.50	.45	.05

Thus

$$E(Z) = \sum_{z=0}^3 z p_Z(z) = 0 \cdot 0 + 1 \cdot (.50) + 2 \cdot (.45) + 3 \cdot (.05) = \boxed{1.55}.$$

Alternatively use the formula $E(Z) = \sum_{x=0}^3 \sum_{y=0}^1 \max(x, y) p(x, y)$.

The marginal probabilities are the row and column sums. X and Y are not independent because, e.g., $0 = p(0, 0) \neq p_X(0)p_Y(0) = (.05)(.5) = .025$.

We have

$$E(X) = \sum_{x=0}^3 x p_X(x) = 0 \cdot (.05) + 1 \cdot (.45) + 2 \cdot (.45) + 3 \cdot (.05) = 1.5,$$

$$E(Y) = \sum_{y=0}^1 y p_Y(y) = 0 \cdot (.5) + 1 \cdot (.5) = .5,$$

$$E(XY) = \sum_{x=0}^3 \sum_{y=0}^1 xy p(x, y) = 1 \cdot 1 \cdot (.30) + 2 \cdot 1 \cdot (.15) = .60.$$

Thus, using the short cut formula

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = .60 - (1.5)(.5) = \boxed{-.15}.$$

4. Suppose that the time in seconds a snowflake survives after landing on a sleeve is a random variable X with the probability density function with parameter $\theta > 0$ given by $f(x; \theta)$. Compute the expected value $E(X)$. Suppose that X, Y is a random sample of snowflake survival times taken from this distribution. Let α be a real number. Show that $\hat{\theta}_\alpha = \alpha X + (3 - \alpha)Y$ is an unbiased estimator for θ . Compute the standard error of $\hat{\theta}_\alpha$. Among the $\hat{\theta}_\alpha$, which one gives the best estimator for θ ? Why?

$$f(x; \theta) = \begin{cases} \frac{2}{\theta} - \frac{2}{\theta^2}x, & \text{if } 0 \leq x \leq \theta; \\ 0, & \text{otherwise.} \end{cases}$$

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx = \int_0^\theta x \left(\frac{2}{\theta} - \frac{2}{\theta^2}x \right) dx = \left[\frac{1}{\theta}x^2 - \frac{2}{3\theta^2}x^3 \right]_0^\theta = \boxed{\frac{\theta}{3}}.$$

Both X and Y are independent with the same distribution. Thus, the estimators $\hat{\theta}_\alpha$ are unbiased because

$$E(\hat{\theta}_\alpha) = E(\alpha X + (3 - \alpha)Y) = \alpha E(X) + (3 - \alpha)E(Y) = \frac{\alpha\theta}{3} + \frac{(3 - \alpha)\theta}{3} = \theta.$$

We have

$$E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_0^\theta x^2 \left(\frac{2}{\theta} - \frac{2}{\theta^2}x \right) dx = \left[\frac{2}{3\theta}x^3 - \frac{2}{4\theta^2}x^4 \right]_0^\theta = \boxed{\frac{\theta^2}{6}}$$

so

$$V(X) = E(X^2) - E(X)^2 = \frac{\theta^2}{6} - \frac{\theta^2}{9} = \frac{\theta^2}{18}.$$

Thus, since X and Y are independent,

$$V(\hat{\theta}_\alpha) = V(\alpha X + (3 - \alpha)Y) = \alpha^2 V(X) + (3 - \alpha)^2 V(Y) = [\alpha^2 + (3 - \alpha)^2] \frac{\theta^2}{18}.$$

Hence the standard error is

$$\sigma_{\theta_\alpha} = [\alpha^2 + (3 - \alpha)^2]^{\frac{1}{2}} \frac{\theta}{3\sqrt{2}}.$$

which is minimum when $\alpha = 1.5$. Thus the best of these estimators is the one with the smallest standard error, namely $\boxed{\hat{\theta}_{1.5}}$.

5. *Electronic books manufactured by Cedar City Readers have a mean lifetime of 6.75 years and a standard deviation of 0.90 years, while the Maeser Laser has a mean lifetime of 6.40 years and a standard deviation of 1.40 years. Let \bar{X} = the average lifetime for a random sample of 36 Cedar Readers and \bar{Y} = the average lifetime for a sample 49 Maeser Lasers. How is the random variable $\bar{X} - \bar{Y}$ distributed? What are its mean $\mu_{\bar{X}-\bar{Y}}$ and standard deviation $\sigma_{\bar{X}-\bar{Y}}$? Why? What is the probability that the average lifetime of a sample of 36 Cedar Readers will be at least one year more than the average lifetime of a sample of 49 Maeser Lasers?*

Because we assume that X_1, \dots, X_{36} are IID random variables from a distribution with $\mu_X = 6.75$ and $\sigma_X = 0.90$, and that $n_X = 36 > 30$, then the mean is approximately normally distributed by the rule of thumb for the Central Limit Theorem, $\bar{X} \sim N(\mu_X, \frac{\sigma_X}{\sqrt{n_X}})$.

Because Y_1, \dots, Y_{49} are IID random variables from a distribution with $\mu_Y = 6.40$ and $\sigma_Y = 1.40$, and that $n_Y = 49 > 30$, then the mean is also approximately normally distributed $\bar{Y} \sim N(\mu_Y, \frac{\sigma_Y}{\sqrt{n_Y}})$. Finally because the X_i 's are independent of the Y_j 's so \bar{X} is independent of \bar{Y} , we use the formula for linear combinations to find that

$$\begin{aligned}\mu_{\bar{X}-\bar{Y}} &= \mu_X - \mu_Y, \\ \sigma_{\bar{X}-\bar{Y}}^2 &= V(\bar{X} - \bar{Y}) = (1)^2V(\bar{X}) + (-1)^2V(\bar{Y}) = \frac{\sigma_X^2}{n_X} + \frac{\sigma_Y^2}{n_Y}.\end{aligned}$$

As the difference of two independent normal variables is also normal, we have

$$\bar{X} - \bar{Y} \sim N(\mu_{\bar{X}-\bar{Y}}, \sigma_{\bar{X}-\bar{Y}}).$$

Substituting values

$$\begin{aligned}\mu_{\bar{X}-\bar{Y}} &= \mu_X - \mu_Y = 6.75 - 6.40 = \boxed{.35}, \\ \sigma_{\bar{X}-\bar{Y}} &= \sqrt{\frac{\sigma_X^2}{n_X} + \frac{\sigma_Y^2}{n_Y}} = \sqrt{\frac{(0.90)^2}{36} + \frac{(1.40)^2}{49}} = \boxed{0.25}.\end{aligned}$$

Finally, standardizing

$$\begin{aligned}P(\bar{X} \geq \bar{Y} + 1) &= P(\bar{X} - \bar{Y} \geq 1) \\ &= P\left(Z = \frac{\bar{X} - \bar{Y} - \mu_{\bar{X}-\bar{Y}}}{\sigma_{\bar{X}-\bar{Y}}} \geq \frac{1 - .35}{.25} = 2.60\right) \\ &= 1 - P(Z < 2.60) = 1 - \Phi(2.60) = \Phi(-2.60) = \boxed{.0047}.\end{aligned}$$