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Full file at

<https://answersun.com/download/instructors-solutions-manual-probability-and-statistics-for-engineers-and-scienti>

Chapter 1

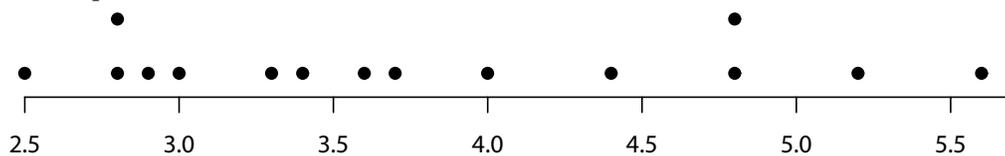
Introduction to Statistics and Data Analysis

1.1 (a) 15.

(b) $\bar{x} = \frac{1}{15}(3.4 + 2.5 + 4.8 + \dots + 4.8) = 3.787$.

(c) Sample median is the 8th value, after the data is sorted from smallest to largest: 3.6.

(d) A dot plot is shown below.



(e) After trimming total 40% of the data (20% highest and 20% lowest), the data becomes:

2.9 3.0 3.3 3.4 3.6
3.7 4.0 4.4 4.8

So, the trimmed mean is

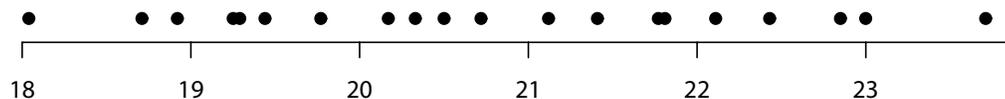
$$\bar{x}_{tr20} = \frac{1}{9}(2.9 + 3.0 + \dots + 4.8) = 3.678.$$

(f) They are about the same.

1.2 (a) Mean=20.7675 and Median=20.610.

(b) $\bar{x}_{tr10} = 20.743$.

(c) A dot plot is shown below.



(d) No. They are all close to each other.

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$$4.67 \quad E[g(X, Y)] = E(X/Y^3 + X^2Y) = E(X/Y^3) + E(X^2Y).$$

$$E(X/Y^3) = \int_1^2 \int_0^1 \frac{2x(x+2y)}{7y^3} dx dy = \frac{2}{7} \int_1^2 \left(\frac{1}{3y^3} + \frac{1}{y^2} \right) dy = \frac{15}{84};$$

$$E(X^2Y) = \int_1^2 \int_0^1 \frac{2x^2y(x+2y)}{7} dx dy = \frac{2}{7} \int_1^2 y \left(\frac{1}{4} + \frac{2y}{3} \right) dy = \frac{139}{252}.$$

$$\text{Hence, } E[g(X, Y)] = \frac{15}{84} + \frac{139}{252} = \frac{46}{63}.$$

$$4.68 \quad P = I^2R \text{ with } R = 50, \mu_I = E(I) = 15 \text{ and } \sigma_I^2 = \text{Var}(I) = 0.03.$$

$E(P) = E(I^2R) = 50E(I^2) = 50[\text{Var}(I) + \mu_I^2] = 50(0.03 + 15^2) = 11251.5$. If we use the approximation formula, with $g(I) = I^2$, $g'(I) = 2I$ and $g''(I) = 2$, we obtain,

$$E(P) \approx 50 \left[g(\mu_I) + 2 \frac{\sigma_I^2}{2} \right] = 50(15^2 + 0.03) = 11251.5.$$

Since $\text{Var}[g(I)] \approx \left[\frac{\partial g(i)}{\partial i} \right]_{i=\mu_I}^2 \sigma_I^2$, we obtain

$$\text{Var}(P) = 50^2 \text{Var}(I^2) = 50^2 (2\mu_I)^2 \sigma_I^2 = 50^2 (30)^2 (0.03) = 67500.$$

$$4.69 \quad \text{For } 0 < a < 1, \text{ since } g(a) = \sum_{x=0}^{\infty} a^x = \frac{1}{1-a}, \quad g'(a) = \sum_{x=1}^{\infty} x a^{x-1} = \frac{1}{(1-a)^2} \text{ and}$$

$$g''(a) = \sum_{x=2}^{\infty} x(x-1)a^{x-2} = \frac{2}{(1-a)^3}.$$

$$(a) \quad E(X) = (3/4) \sum_{x=1}^{\infty} x(1/4)^x = (3/4)(1/4) \sum_{x=1}^{\infty} x(1/4)^{x-1} = (3/16)[1/(1-1/4)^2]$$

$$= 1/3, \text{ and } E(Y) = E(X) = 1/3.$$

$$E(X^2) - E(X) = E[X(X-1)] = (3/4) \sum_{x=2}^{\infty} x(x-1)(1/4)^x$$

$$= (3/4)(1/4)^2 \sum_{x=2}^{\infty} x(x-1)(1/4)^{x-2} = (3/4^3)[2/(1-1/4)^3] = 2/9.$$

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

$$2/9 + 1/3 - (1/3)^2 = 4/9, \text{ and } \text{Var}(Y) = 4/9.$$

$$(b) \quad E(Z) = E(X) + E(Y) = (1/3) + (1/3) = 2/3, \text{ and}$$

$\text{Var}(Z) = \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) = (4/9) + (4/9) = 8/9$, since X and Y are independent (from Exercise 3.79).

$$4.70 \quad (a) \quad g(x) = \frac{3}{2} \int_0^1 (x^2 + y^2) dy = \frac{1}{2}(3x^2 + 1) \text{ for } 0 < x < 1 \text{ and}$$

$$h(y) = \frac{1}{2}(3y^2 + 1) \text{ for } 0 < y < 1.$$

Since $f(x, y) \neq g(x)h(y)$, X and Y are not independent.

$$(b) \quad E(X+Y) = E(X) + E(Y) = 2E(X) = \int_0^1 x(3x^2 + 1) dx = 3/4 + 1/2 = 5/4.$$

$$E(XY) = \frac{3}{2} \int_0^1 \int_0^1 xy(x^2 + y^2) dx dy = \frac{3}{2} \int_0^1 y \left(\frac{1}{4} + \frac{y^2}{2} \right) dy$$

$$= \frac{3}{2} \left[\left(\frac{1}{4} \right) \left(\frac{1}{2} \right) + \left(\frac{1}{2} \right) \left(\frac{1}{4} \right) \right] = \frac{3}{8}.$$

$$(c) \quad \text{Var}(X) = E(X^2) - [E(X)]^2 = \frac{1}{2} \int_0^1 x^2(3x^2 + 1) dx - \left(\frac{5}{8} \right)^2 = \frac{7}{15} - \frac{25}{64} = \frac{73}{960}, \text{ and}$$

$$\text{Var}(Y) = \frac{73}{960}. \text{ Also, } \text{Cov}(X, Y) = E(XY) - E(X)E(Y) = \frac{3}{8} - \left(\frac{5}{8} \right)^2 = -\frac{1}{64}.$$

$$(d) \quad \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y) = 2 \frac{73}{960} - 2 \frac{1}{64} = \frac{29}{240}.$$

$$4.71 \quad (a) \quad E(Y) = \int_0^{\infty} ye^{-y/4} dy = 4.$$

- 9.9 $n = 20$, $\bar{x} = 11.3$, $s = 2.45$, and $t_{0.025} = 2.093$ with 19 degrees of freedom. A 95% confidence interval for the population mean is

$$11.3 - (2.093)(2.45/\sqrt{20}) < \mu < 11.3 + (2.093)(2.45/\sqrt{20}),$$

or $10.15 < \mu < 12.45$.

- 9.10 $n = 12$, $\bar{x} = 79.3$, $s = 7.8$, and $t_{0.025} = 2.201$ with 11 degrees of freedom. A 95% confidence interval for the population mean is

$$79.3 - (2.201)(7.8/\sqrt{12}) < \mu < 79.3 + (2.201)(7.8/\sqrt{12}),$$

or $74.34 < \mu < 84.26$.

- 9.11 $n = 9$, $\bar{x} = 1.0056$, $s = 0.0245$, and $t_{0.005} = 3.355$ with 8 degrees of freedom. A 99% confidence interval for the population mean is

$$1.0056 - (3.355)(0.0245/3) < \mu < 1.0056 + (3.355)(0.0245/3),$$

or $0.978 < \mu < 1.033$.

- 9.12 $n = 10$, $\bar{x} = 230$, $s = 15$, and $t_{0.005} = 3.25$ with 9 degrees of freedom. A 99% confidence interval for the population mean is

$$230 - (3.25)(15/\sqrt{10}) < \mu < 230 + (3.25)(15/\sqrt{10}),$$

or $214.58 < \mu < 245.42$.

- 9.13 $n = 12$, $\bar{x} = 48.50$, $s = 1.5$, and $t_{0.05} = 1.796$ with 11 degrees of freedom. A 90% confidence interval for the population mean is

$$48.50 - (1.796)(1.5/\sqrt{12}) < \mu < 48.50 + (1.796)(1.5/\sqrt{12}),$$

or $47.722 < \mu < 49.278$.

- 9.14 $n = 15$, $\bar{x} = 3.7867$, $s = 0.9709$, $1 - \alpha = 95\%$, and $t_{0.025} = 2.145$ with 14 degrees of freedom. So, by calculating $3.7867 \pm (2.145)(0.9709)\sqrt{1 + 1/15}$ we obtain $(1.6358, 5.9376)$ which is a 95% prediction interval the drying times of next paint.

- 9.15 $n = 100$, $\bar{x} = 23,500$, $s = 3,900$, $1 - \alpha = 0.99$, and $t_{0.005} \approx 2.66$ with 60 degrees of freedom (use table from the book) or $t_{0.005} = 2.626$ if 100 degrees of freedom is used. The prediction interval of next automobile will be driven in Virginia (using 2.66) is $23,500 \pm (2.66)(3,900)\sqrt{1 + 1/100}$ which yields $13,075 < \mu < 33,925$ kilometers.

- 9.16 $n = 12$, $\bar{x} = 79.3$, $s = 7.8$, and $t_{0.025} = 2.201$ with 11 degrees of freedom. A 95% prediction interval for a future observation is

$$79.3 \pm (2.201)(7.8)\sqrt{1 + 1/12} = 79.3 \pm 17.87,$$

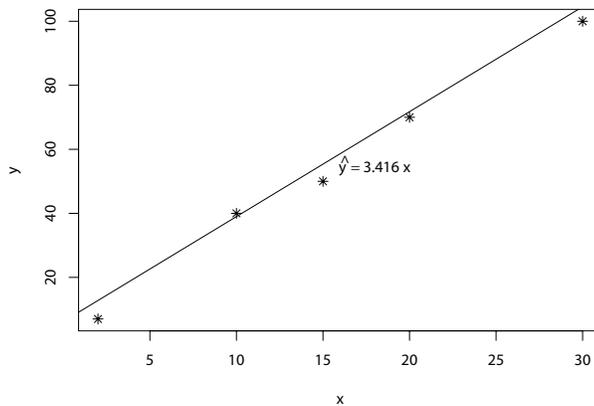
which yields $(61.43, 97.17)$.

Taking derivative of the above with respect to b_1 and setting the derivative to zero, we have $-2 \sum_{i=1}^n x_i(y_i - b_1 x_i) = 0$, which implies $b_1 = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n x_i^2}$.

$$(b) \sigma_{B_1}^2 = \frac{\text{Var}\left(\sum_{i=1}^n x_i Y_i\right)}{\left(\sum_{i=1}^n x_i^2\right)^2} = \frac{\sum_{i=1}^n x_i^2 \sigma_{Y_i}^2}{\left(\sum_{i=1}^n x_i^2\right)^2} = \frac{\sigma^2}{\sum_{i=1}^n x_i^2}, \text{ since } Y_i \text{'s are independent.}$$

$$(c) E(B_1) = \frac{E\left(\sum_{i=1}^n x_i Y_i\right)}{\sum_{i=1}^n x_i^2} = \frac{\sum_{i=1}^n x_i (\beta_1 x_i)}{\sum_{i=1}^n x_i^2} = \beta_1.$$

11.29 (a) The scatter plot of the data is shown next.



$$(b) \sum_{i=1}^n x_i^2 = 1629 \text{ and } \sum_{i=1}^n x_i y_i = 5564. \text{ Hence } b_1 = \frac{5564}{1629} = 3.4156. \text{ So, } \hat{y} = 3.4156x.$$

(c) See (a).

(d) Since there is only one regression coefficient, β_1 , to be estimated, the degrees of freedom in estimating σ^2 is $n - 1$. So,

$$\hat{\sigma}^2 = s^2 = \frac{SSE}{n-1} = \frac{\sum_{i=1}^n (y_i - b_1 x_i)^2}{n-1}.$$

$$(e) \text{Var}(\hat{y}_i) = \text{Var}(B_1 x_i) = x_i^2 \text{Var}(B_1) = \frac{x_i^2 \sigma^2}{\sum_{i=1}^n x_i^2}.$$

(f) The plot is shown next.

14.2 The hypotheses of the three parts are,

(a) for the main effects brands,

$$H'_0 : \alpha_1 = \alpha_2 = \alpha_3 = 0,$$

$$H'_1 : \text{At least one of the } \alpha_i\text{'s is not zero;}$$

(b) for the main effects times,

$$H''_0 : \beta_1 = \beta_2 = \beta_3 = 0,$$

$$H''_1 : \text{At least one of the } \beta_i\text{'s is not zero;}$$

(c) and for the interactions,

$$H'''_0 : (\alpha\beta)_{11} = (\alpha\beta)_{12} = \cdots = (\alpha\beta)_{33} = 0,$$

$$H'''_1 : \text{At least one of the } (\alpha\beta)_{ij}\text{'s is not zero.}$$

$$\alpha = 0.05.$$

Critical regions: (a) $f_1 > 3.35$; (b) $f_2 > 3.35$; and (c) $f_3 > 2.73$.

Computations: From the computer printout we have

Source of Variation	Sum of Squares	Degrees of Freedom	Mean Square	Computed f
Brands	32.7517	2	16.3758	1.74
Times	227.2117	2	113.6058	12.04
Interaction	17.3217	4	4.3304	0.46
Error	254.7025	27	9.4334	
Total	531.9875	35		

Decision: (a) Do not reject H'_0 ; (b) Reject H''_0 ; (c) Do not reject H'''_0 .

14.3 The hypotheses of the three parts are,

(a) for the main effects environments,

$$H'_0 : \alpha_1 = \alpha_2 = 0, \text{ (no differences in the environment)}$$

$$H'_1 : \text{At least one of the } \alpha_i\text{'s is not zero;}$$

(b) for the main effects strains,

$$H''_0 : \beta_1 = \beta_2 = \beta_3 = 0, \text{ (no differences in the strains)}$$

$$H''_1 : \text{At least one of the } \beta_i\text{'s is not zero;}$$

(c) and for the interactions,

$$H'''_0 : (\alpha\beta)_{11} = (\alpha\beta)_{12} = \cdots = (\alpha\beta)_{23} = 0, \text{ (environments and strains do not interact)}$$

$$H'''_1 : \text{At least one of the } (\alpha\beta)_{ij}\text{'s is not zero.}$$

$$\alpha = 0.01.$$

Critical regions: (a) $f_1 > 7.29$; (b) $f_2 > 5.16$; and (c) $f_3 > 5.16$.

Computations: From the computer printout we have