

MATH 308 Assignment 18:

Exercises 8.5

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$p = P(X \geq 20) = 1 - \text{ppois}(19, \lambda = 15) \approx 12.5\%$. This implies that more than 10% of the months will have a birth rate as extreme as the one observed, which means that we cannot reject the null ($\lambda = 15$) even at the 5% significance level.

lead to patients wasting money on a useless drug.

A Type II error would be failing to conclude that the drug is effective, when in fact it is. This would lead to a waste of the research money put into the drug, since the pharmaceutical company would then not be able to sell it.

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a)

Running the test gives a p -value of 10^{-4} . This lets us reject the null hypothesis (that the difference in means is zero) at the 1% significance level.

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b)

The random assignment of seedlings to plots tends to remove confounding effects. Thus, the result hints at a causal relationship between presence of competition and diameter change.

Under the null, $\bar{X} \sim \mathcal{N}(25, 16/30)$. Thus $Z = \frac{\bar{X} - 25}{4/\sqrt{30}}$, where Z is a standard normal random variable. Thus, the null is rejected if $Z \geq z_{1-\alpha} \implies \frac{\bar{X} - 25}{4/\sqrt{30}} \geq z_{0.95} \implies \bar{X} \geq 25 + \frac{4z_{0.95}}{\sqrt{30}} \equiv C$.

But $\bar{X} \sim \mathcal{N}(27, 16/\sqrt{30}) \implies Z = \frac{\bar{X} - 27}{4/\sqrt{30}}$. So,

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Running the chi-squared test gives a p -value of 0.012, allowing us to reject the null at the 5% significance level.

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A Type I error would be incorrectly concluding that the drug is effective when in fact it is not. This would

$$\begin{aligned} 1 - \beta &= P(\bar{X} \geq C | \mu = 27) \\ &= P\left(\frac{4Z}{\sqrt{30}} + 27 \geq C\right) \\ &= P\left(Z \geq \frac{\sqrt{30}(C - 27)}{4}\right) \equiv P(Z \geq C_1) \\ &= 1 - P(Z < C_1) \approx 86.3\% \end{aligned}$$

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The null is rejected if $\bar{X} \geq \mu_0 + \frac{z_{1-\alpha}\sigma}{\sqrt{n}} \equiv C$, where $\mu_0 = 1, \alpha = 0.01$ and $\sigma = 0.3$. So,

$$\begin{aligned}
1 - \beta &= P(\bar{X} \geq C | \mu = \mu_1) \\
&= P\left(\frac{Z\sigma}{\sqrt{n}} + \mu_1 \geq C\right) \\
&= P\left(Z \geq \frac{\sqrt{n}(C - \mu_1)}{\sigma}\right) \equiv P(Z \geq C_1) \\
&= 1 - P(Z < C_1) \\
&\implies \beta = P(Z < C_1) \\
&\implies z_\beta = C_1 = \frac{\sqrt{n}(C - \mu_1)}{\sigma} \\
\implies \frac{z_\beta\sigma}{\sqrt{n}} + \mu_1 &= C = \frac{z_{1-\alpha}\sigma}{\sqrt{n}} + \mu_0 \\
\implies n &= \left(\frac{\sigma(z_\beta - z_{1-\alpha})}{\mu_0 - \mu_1}\right)^2 = 8
\end{aligned}$$

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$$n = \left(\frac{\sigma(z_\beta - z_{1-\alpha})}{\mu_0 - \mu_1}\right)^2 = 1$$

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$$f(x) = \lambda e^{-\lambda x} \implies F(x) = \int_0^x \lambda e^{-\lambda t} dt = 1 - e^{-\lambda x}$$

a)

$$\begin{aligned}
F_{X,\min}(x) &= \prod_{i=1}^{15} P(X_i \leq x) \\
&= (1 - F(x))^{15} \\
&= e^{-15\lambda x} \\
\implies \alpha &= P(X_{\min} \geq 1 | \lambda = 1/5) \\
&= 1 - F_{X,\min}(1) |_{\lambda=1/5} \\
&= 1 - e^{-3} \approx 95\%
\end{aligned}$$

b)

$$\begin{aligned}
1 - \beta &= P(X_{\min} \geq 1 | \lambda = 1/25) \\
&= 1 - F_{X,\min}(1) |_{\lambda=1/25} \\
&= 1 - e^{-3/5} \approx 45\%
\end{aligned}$$

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The most powerful test simply always rejects the null. Thus, the critical region is $X_1, X_2 \in [0, n]$.

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Under the null, the likelihood is maximised at $\mu = \mu_0$ and $\sigma^2 = \Sigma(X_i - \mu_0)^2/n \equiv \hat{\sigma}^2$. Thus, the likelihood under the null is

$$L_{H,0} = \left(\frac{1}{\hat{\sigma}\sqrt{2\pi}}\right)^n \exp\left(\frac{-\Sigma(X_i - \mu_0)^2}{2\hat{\sigma}^2}\right) = \frac{e^{-n/2}}{(\hat{\sigma}\sqrt{2\pi})^n}$$

To maximise the likelihood, we simply use $\mu = \bar{X}$ and $\sigma = \sqrt{\Sigma(X_i - \bar{X})^2/n}$, giving

$$L = \left(\frac{1}{S\sqrt{2\pi}}\right)^n \exp\left(\frac{-\Sigma(X_i - \bar{X})^2}{2S^2}\right) = \frac{e^{-n/2}}{(S\sqrt{2\pi})^n}$$

The likelihood ratio is therefore $(S/\hat{\sigma})^n$. So, we reject if

$$\begin{aligned}
\left(\frac{\Sigma(X_i - \bar{X})^2}{\Sigma(X_i - \mu_0)^2}\right)^{n/2} &\leq c \implies \frac{\Sigma(X_i - \bar{X})^2}{\Sigma(X_i - \mu_0)^2} \leq c_1 \\
\implies c_1 &\leq \frac{\Sigma(X_i - \mu_0)^2}{\Sigma(X_i - \bar{X})^2} = \frac{\Sigma(X_i - \bar{X} + \bar{X} - \mu_0)^2}{\Sigma(X_i - \bar{X})^2} \\
&\leq 1 + \frac{n(\bar{X} - \mu_0)^2 - 2(\bar{X} - \mu_0)\Sigma(X_i - \bar{X})}{\Sigma(X_i - \bar{X})^2} \\
&\leq 1 + \frac{n}{\Sigma(X_i - \bar{X})^2}(\bar{X} - \mu_0)^2 \\
\implies c_2 &\leq \sqrt{\frac{n}{\Sigma(X_i - \bar{X})^2}}(\bar{X} - \mu_0) = \frac{\bar{X} - \mu_0}{S/\sqrt{n}}
\end{aligned}$$

which is a t -statistic. Thus, the test is a one-sided t -test. \square

MATH 308 Assignment 19: Exercises 9.7

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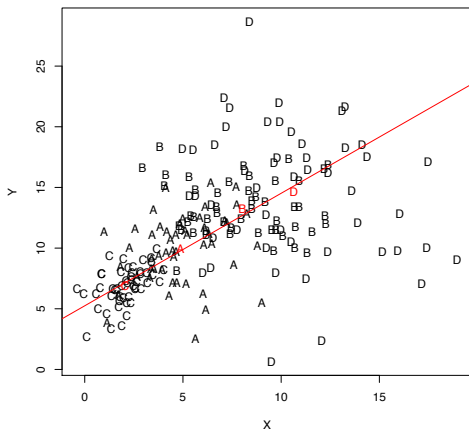
a)

$$\rho \approx 0.50$$

b)

	A	B	C	D
\bar{X}	4.88	8.03	2.06	10.64
\bar{Y}	9.91	13.24	6.89	14.68

c)



$\rho = 0.99$, which is much higher than the result from (a).

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$$\begin{aligned} g(a, b) &= \Sigma(y_i - \hat{y}_i)^2 \\ \Rightarrow \frac{\partial g}{\partial a} &= \Sigma \left(2(y_i - \hat{y}_i) \frac{\partial(y_i - (a + bx_i))}{\partial a} \right) \\ &\Rightarrow 0 = -2\Sigma(y_i - \hat{y}_i) \\ &\Rightarrow \Sigma(y_i - \hat{y}_i) = 0 \quad \square \end{aligned}$$

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a)

$$\begin{aligned} r &= \frac{ss_{xy}}{\sqrt{ss_x ss_y}}, b = \frac{ss_{xy}}{ss_x} \\ \Sigma(\hat{y}_i - \bar{y})^2 &= \Sigma(a + bx_i - (a + b\bar{x}))^2 \\ &= b^2 \Sigma(x_i - \bar{x})^2 = b^2 ss_x = \frac{ss_{xy}^2}{ss_x} \\ &= r^2 ss_y \\ \Rightarrow s_{\hat{y}} &= \sqrt{\frac{\Sigma(\hat{y}_i - \bar{y})^2}{n-1}} = r \sqrt{\frac{ss_y}{n-1}} \\ &= r \sqrt{\frac{\Sigma(y_i - \bar{y})^2}{n-1}} = r s_y \quad \square \end{aligned}$$

b)

$$\begin{aligned}
 \bar{e} &= \Sigma(y_i - \hat{y}_i)/n = 0 \\
 \Rightarrow s_e &= \sqrt{\frac{\Sigma(y_i - \hat{y}_i)^2}{n-1}} \\
 \Rightarrow (n-1)s_e^2 &= \Sigma(y_i - \hat{y}_i)^2 \\
 &= \Sigma(y_i - \bar{y} + \bar{y} - \hat{y}_i)^2 \\
 &= \Sigma(y_i - \bar{y})^2 + \Sigma(\bar{y} - \hat{y}_i)^2 \\
 &\quad - 2\Sigma(y_i - \bar{y})(\hat{y}_i - \bar{y}) \\
 &= ss_y + r^2 ss_y - 2\Sigma(y_i \hat{y}_i - y_i \bar{y} - \hat{y}_i \bar{y} + \bar{y}^2) \\
 &= ss_y(1 + r^2) - 2\Sigma(\hat{y}_i(y_i - \bar{y})) \\
 \Rightarrow s_e &= \sqrt{\frac{(1 + r^2)ss_y}{n-1}} = \sqrt{1 + r^2} s_y \quad \square
 \end{aligned}$$

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a)

$$\begin{aligned}
 b &= r \frac{s_w}{s_h} \approx 1.6a = \bar{w} - b\bar{h} \approx 20.1 \\
 \Rightarrow \hat{w} &= a + bh \approx 20.1 + 1.6h
 \end{aligned}$$

b)

$$\hat{w}(5 \text{ ft}) = \hat{w}(60 \text{ in}) = 116.5 \text{ pounds}$$

c)

$r^2 \approx 0.56$, which means that the model explains 56% of the variability of the weights of the individuals.

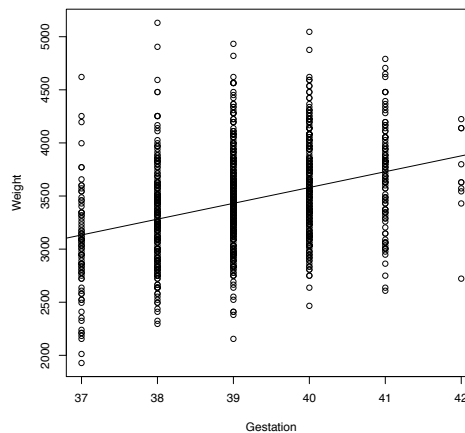
17

a)

Correlation is ≈ 0.35

b)

$$y = -2380 + 149x$$

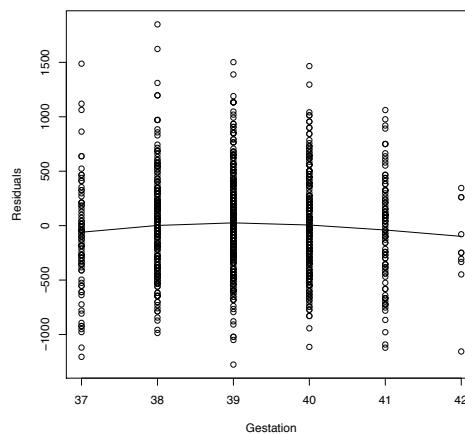


c)

The slope is interpreted as ‘a week increase in gestation period is correlated with a 149g increase in weight’.

The $r^2 \approx 0.12$ is interpreted as ‘the linear model explains 12% of the variation in weights’.

d)



The residual plot shows a curve peaking at week 39. The linear model is therefore inappropriate.

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a)

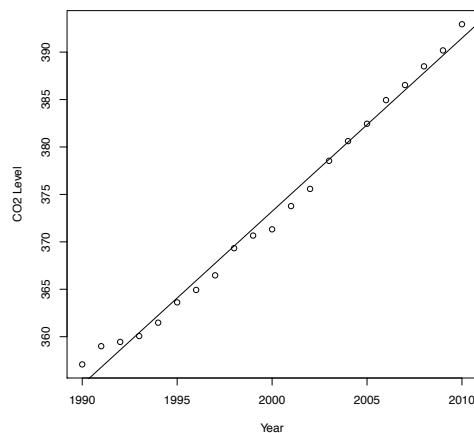
457.4 g

b)

$$\begin{aligned}
 1 - \alpha &= 0.95 \implies 1 - \alpha/2 = 0.975 \\
 \implies t &\equiv t_{n-2, 1-\alpha/2} = t_{1007, 0.975} \approx 1.96 \\
 \therefore 95\% \text{ CI: } \hat{\beta} \pm \frac{tS}{\sqrt{SS_x}} &= 149 \pm 24.8 \\
 &= (124.2, 173.8)
 \end{aligned}$$

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a)

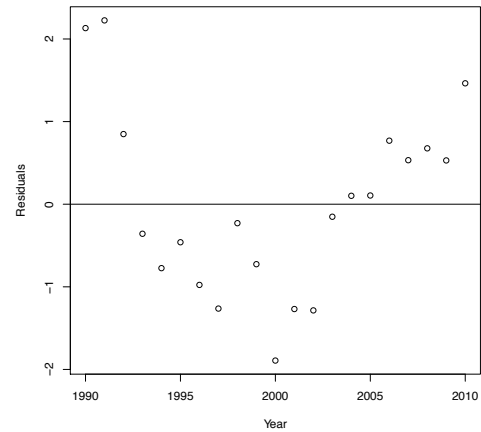


The data is strongly linear, with a correlation of over 0.99.

b)

$$y = -3279.593 + 1.826x$$

c)



At first, it appears that there is a pattern to the residuals, decreasing until year 2000 and then increasing again. However, looking at the scale of the y -axis we see that the largest residual is only 2, but the CO_2 Levels are of order 350.

Thus, the apparent pattern can be put down to random error and ignored, making a linear model appropriate.

MATH 308 Assignment 20: Power Function

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1

The null is rejected if $\bar{X} > \mu_0 + \frac{z_{1-\alpha}}{\sqrt{n}} \equiv C$. But,
 $\bar{X} \sim \mathcal{N}(\mu, 1^2/n) \implies \bar{X} = \mu + \frac{Z}{\sqrt{n}}$

$$\begin{aligned} \therefore \beta(\mu) &= P(\bar{X} > C) \\ &= 1 - P(\bar{X} \leq C) \\ &= 1 - P\left(\mu + \frac{Z}{\sqrt{n}} \leq C\right) \\ &= 1 - P(Z \leq (C - \mu)\sqrt{n}) \\ &= 1 - \Phi((C - \mu)\sqrt{n}) \\ &= 1 - \Phi(z_{1-\alpha} + (\mu_0 - \mu)\sqrt{n}) \end{aligned}$$

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$$\begin{aligned} 0.8 &= 1 - \Phi(z_{0.95} + (0 - 1)\sqrt{n}) \\ &\implies z_{0.2} = z_{0.95} - \sqrt{n} \\ &\implies n = (z_{0.95} - z_{0.2})^2 \approx 7 \end{aligned}$$

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Setting $\alpha = 0.05$, $\mu_0 = 0$ and $n = 100$, we get

$$\beta(\mu) = 1 - \Phi(z_{0.95} - 10\mu)$$

