

## ANSWER KEY - PROBLEM SET II

12.20)

$$a - S_{xy} = 1.0785$$

$$S_{xx} = 1.1155$$

$$\hat{\beta}_1 = \frac{S_{xy}}{S_{xx}} = \frac{1.0785}{1.1155} = .9668$$

$$\begin{aligned}\hat{\beta}_0 &= \bar{y} - \hat{\beta}_1 \bar{x} = 0.80 - .9668(0.455) \\ &= 0.365\end{aligned}$$

$$\begin{aligned}b.) \text{Lichen } N &= 0.365 + .9668(\text{NO}_3 \text{ deposition}) \\ &= 0.365 + .9668(.5) \\ &= 0.8485\end{aligned}$$

$$c.) \hat{\sigma} = \sqrt{\frac{SSE}{n-2}} = \sqrt{\frac{.4106}{11}} = .1932$$

$$\begin{aligned}12.25) n\hat{\beta}_0 + (\sum x_i) \hat{\beta}_1 &= \sum y_i \\ n(\bar{y} - \hat{\beta}_1 \bar{x}) + (\sum x_i) \hat{\beta}_1 &= \sum y_i\end{aligned}$$

$$\hat{\beta}_1 = \frac{\sum x_i y_i - \frac{(\sum x_i)(\sum y_i)}{n}}{\sum x_i^2 - \frac{(\sum x_i)^2}{n}}$$

$$\sum x_i \left[ \bar{y} - \hat{\beta}_1 \bar{x} \right] + \sum x_i^2 \hat{\beta}_1 = \sum x_i y_i$$

$$\sum x_i \bar{y} + \hat{\beta}_1 (\sum x_i)^2 + \sum x_i^2 \hat{\beta}_1 = \sum x_i y_i$$

$$\frac{\sum x_i \sum y_i}{n} + \left[ \frac{\sum x_i y_i - \frac{(\sum x_i)(\sum y_i)}{n}}{\sum x_i^2 - \frac{(\sum x_i)^2}{n}} \right] \left[ \frac{\sum x_i^2}{n} - \frac{(\sum x_i)^2}{n} \right]$$

$$+ \frac{\sum x_i \sum y_i}{n} + \sum x_i y_i - \frac{\sum x_i \sum y_i}{n} = \sum x_i y_i$$

28.) a. - See Attached R printout  
 While the two lines have the same slope, the second one is shifted to the left by the value of  $\bar{x}$ .

b. - See attached R printout  
 From the output, we see that  $B_0^*$  is equivalent to  $\bar{y}$  and that  $B_1^*$  is equivalent to  $B_1$ .

29.) See R printout  
 Linear regression would be most effective for data set #3 because it has the highest  $r^2$  and  $s$  values.

$$81.) a. - S_x^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} \quad S_y^2 = \frac{\sum (y_i - \bar{y})^2}{n-1}$$

$$\frac{S_y^2}{S_x^2} = \frac{\sum (y_i - \bar{y})^2}{\sum (x_i - \bar{x})^2}$$

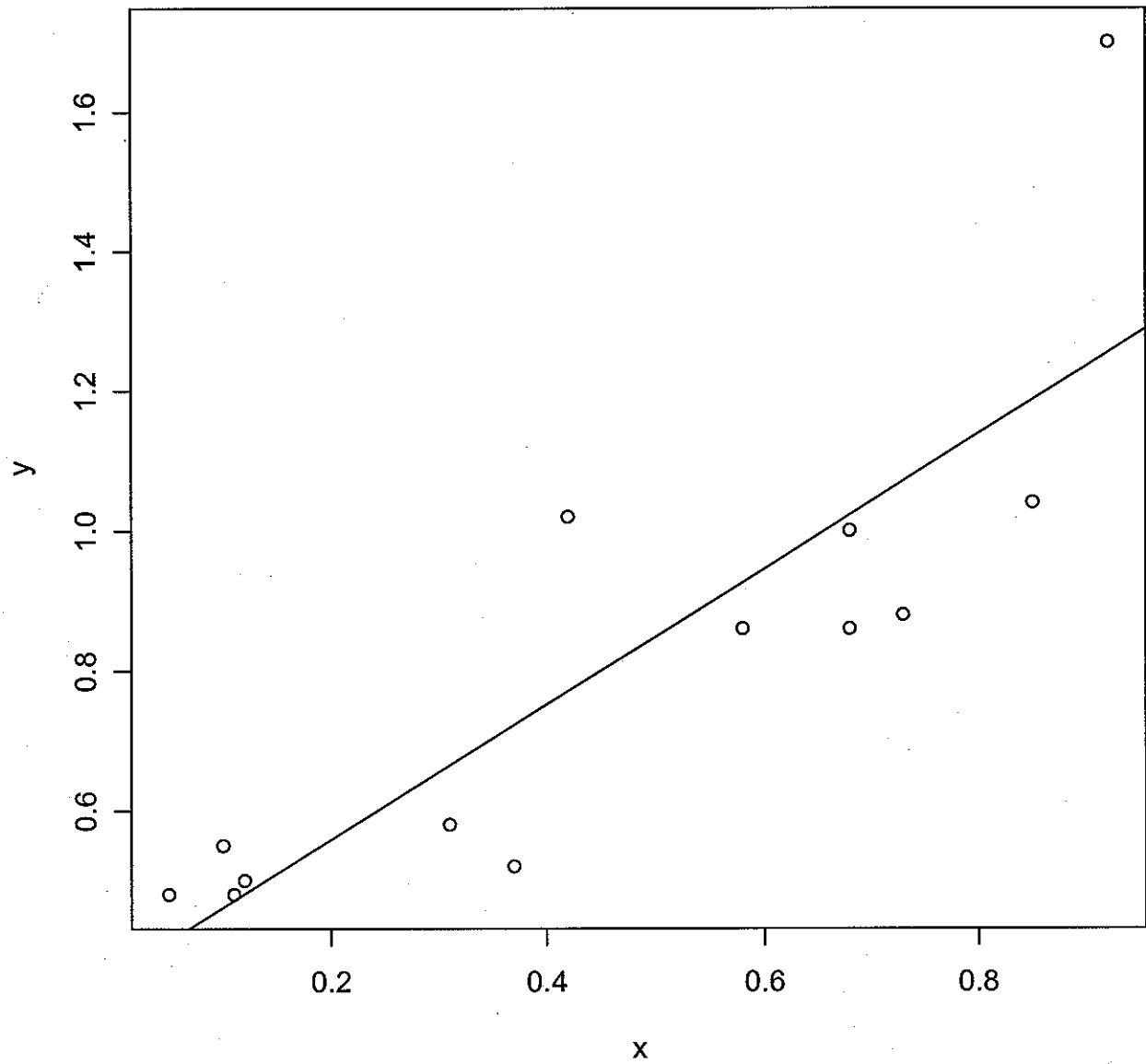
$$\frac{S_y}{S_x} = \frac{\sum (y_i - \bar{y})}{\sum (x_i - \bar{x})} \Rightarrow \frac{S_y}{S_x} = \frac{\sum (y_i - \bar{y})}{\sum (x_i - \bar{x})}$$

$$y = \hat{B}_0 + \hat{B}_1 x = (\bar{y} - \hat{B}_1 \bar{x}) + \hat{B}_1 x = \bar{y} + \hat{B}_1 (x - \bar{x})$$

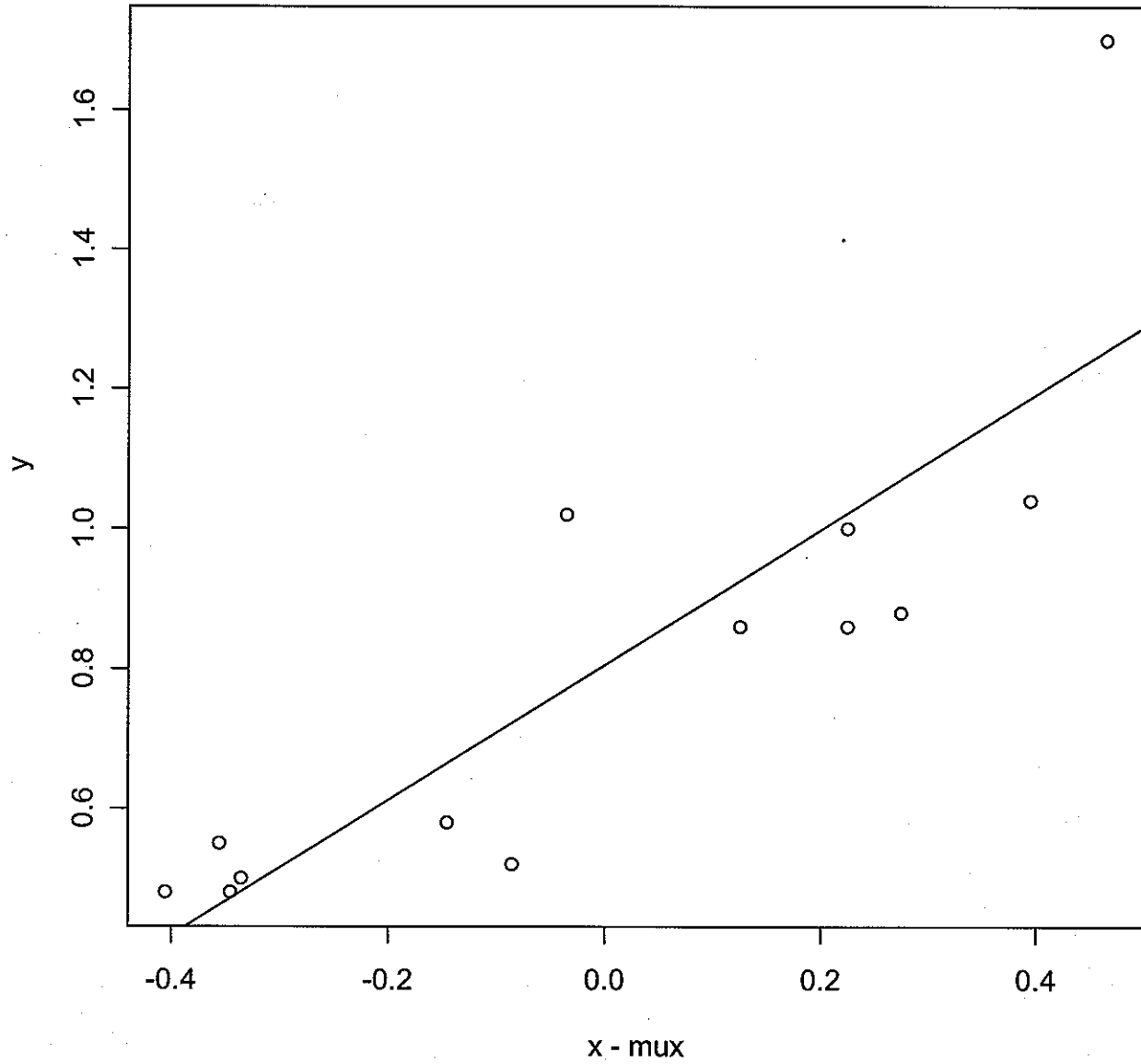
$$\hat{B}_1 = \frac{S_{xy}}{S_{xx}} = \frac{S_{xy}}{\sqrt{S_{xx}} \sqrt{S_{yy}}} \cdot \frac{S_y}{S_x} = r \cdot \frac{S_y}{S_x}$$

$$y = \bar{y} + r \frac{S_y}{S_x} (x - \bar{x}) \quad \checkmark$$

Problem 12.28(a)



Problem 12.28(a)



# Problem 12.28(b)

```
> z = x-mux
> abline(lm(y~z))
> a = lm(y~z)
> summary(a)
```

```
Call:
lm(formula = y ~ z)
```

```
Residuals:
    Min     1Q   Median     3Q      Max
-0.20283 -0.14691 -0.02255  0.06655  0.44541
```

```
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.80538   0.05358  15.031 1.12e-08 ***
z            0.96683   0.18292   5.286 0.000258 ***
```

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.1932 on 11 degrees of freedom
Multiple R-Squared: 0.7175, Adjusted R-squared: 0.6918
F-statistic: 27.94 on 1 and 11 DF, p-value: 0.0002581
```

```
> b = lm(y~x)
> summary(b)
```

```
Call:
lm(formula = y ~ x)
```

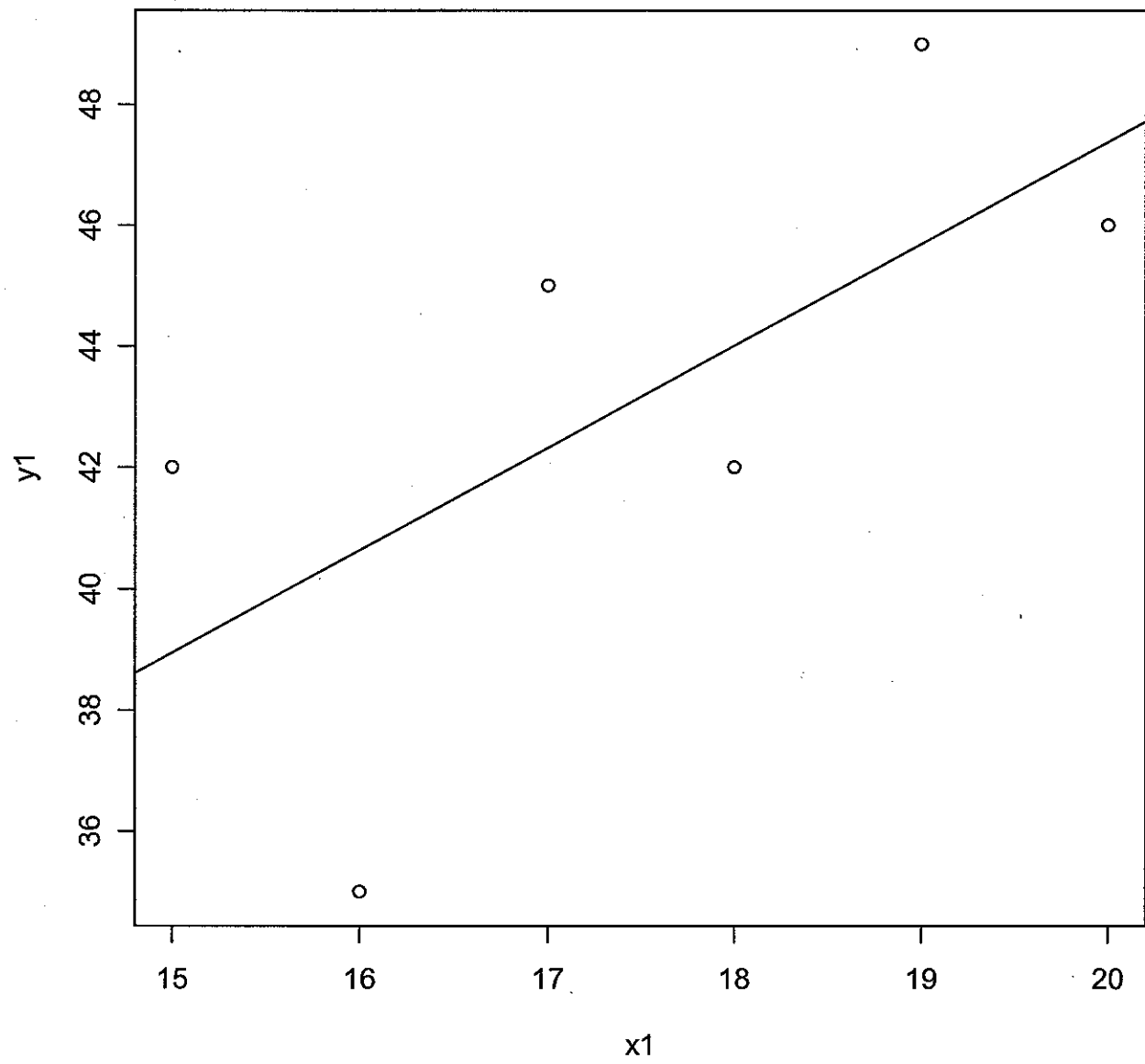
```
Residuals:
    Min     1Q   Median     3Q      Max
-0.20283 -0.14691 -0.02255  0.06655  0.44541
```

```
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.36510   0.09904   3.686 0.003586 **
x            0.96683   0.18292   5.286 0.000258 ***
```

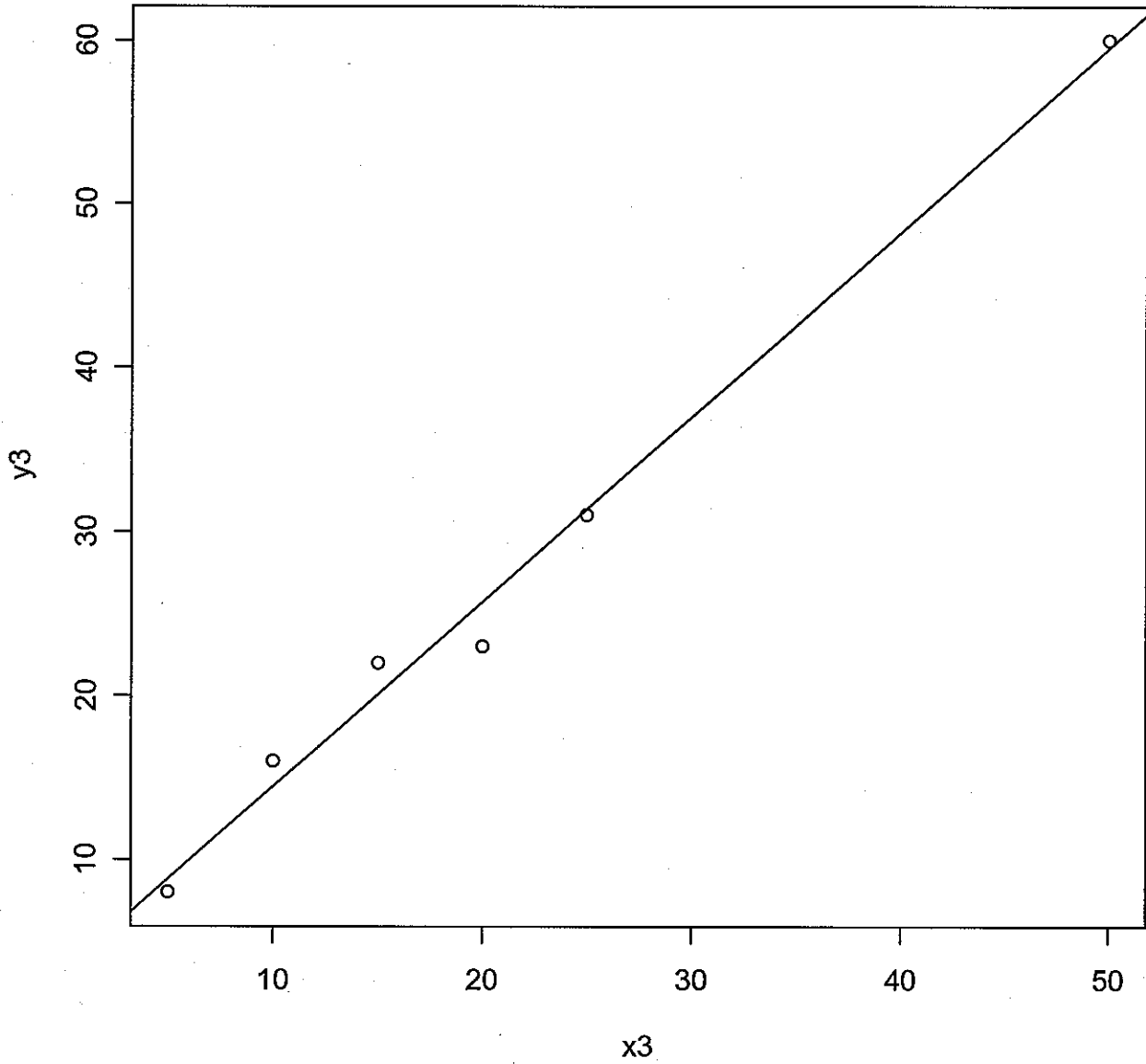
```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.1932 on 11 degrees of freedom
Multiple R-Squared: 0.7175, Adjusted R-squared: 0.6918
F-statistic: 27.94 on 1 and 11 DF, p-value: 0.0002581
```

Problem 12.29



# Problem 1229



# Problem 12.29

## Data Set 1

Call:  
lm(formula = y1 ~ x1)

Residuals:  
1 2 3 4 5 6  
3.048 -5.638 2.676 -2.010 3.305 -1.381

Coefficients:  
Estimate Std. Error t value Pr(>|t|)  
(Intercept) 13.6667 16.9572 0.806 0.465  
x1 1.6857 0.9644 1.748 0.155

Residual standard error: 4.034 on 4 degrees of freedom  
Multiple R-Squared: 0.433, Adjusted R-squared: 0.2913  
F-statistic: 3.055 on 1 and 4 DF, p-value: 0.1554

## Data Set 2

Call:  
lm(formula = y2 ~ x2)

Residuals:  
1 2 3 4 5 6  
-2.58033 2.70820 3.99672 -5.71475 1.57377 0.01639

Coefficients:  
Estimate Std. Error t value Pr(>|t|)  
(Intercept) 7.8689 2.8760 2.736 0.0521 .  
x2 2.1423 0.1132 18.930 4.59e-05 \*\*\*

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.034 on 4 degrees of freedom  
Multiple R-Squared: 0.989, Adjusted R-squared: 0.9862  
F-statistic: 358.4 on 1 and 4 DF, p-value: 4.587e-05

## Data Set 3

Call:  
lm(formula = y3 ~ x3)

Residuals:  
1 2 3 4 5 6  
-0.8295 1.5377 1.9049 -2.7279 -0.3607 0.4754