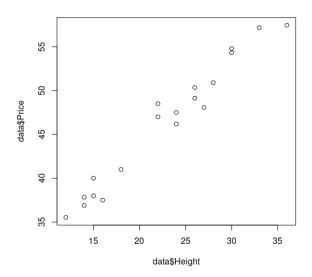
## EXAM 2

- 1. The following are the results of running an ANOVA on the three factors (temperature, fabric denier, and air pressure) at each of their respective three levels.
- 2. (a)
  - (b) The scatter plot indicates a positive, linear relatonship between height and price.



- (c) The equation of least squares is y = 0.98715x + 23.77215.
- (d) At a height of 27, the corresponding price is \$50.43. Given the summary below, the corresponding residual is the result of the least squares (50.4252) plus the median residual (-0.8796), which is 49.55.
- (e) In referring to the above figure, one will see that the adjusted R-squared is 0.96. So, 96% of the variation in sales price can be attributed to the approximate linear relationship between truss height and price.
- 3. (a) Below is the result of the following line of code, given the appropriate datset, data:

$$model \leftarrow lm(NO3\ CO, data = data);\ summary(model)$$

(b) The equation produced by the linear regression in R was

$$\hat{y} = 13.63x + 21.97$$

So, the prediction for x = 400 is 5473.97. Given that the mean of CO is 211.89, the following work shows that the prediction interval is (-139.5264, 563).

$$\hat{y} = mean(data\$CO); \ t = 0.627; \ error = 21.734; \ r^2 = deviance(model)$$

```
Call:
lm(formula = Price ~ Height, data = data)
Residuals:
             10 Median
                             30
   Min
                                    Max
-2.35522 -0.63584 -0.08796 0.92263 3.01053
          Estimate Std. Error t value Pr(>|t|)
(Intercept) 23.77215 1.11347 21.35 1.03e-13 ***
           Height
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.416 on 17 degrees of freedom
Multiple R-squared: 0.9631, Adjusted R-squared: 0.961
F-statistic: 444.1 on 1 and 17 DF, p-value: 1.271e-13
```

Figure 1: Summary of linear model.

Figure 2: Model of NO3 and CO data.

- (c) Without the largest value of CO, the fitted equation is  $\hat{y} = 15.21x + 21.70$ , yielding 6105.7 as the prediction: a number far outside the previous interval. So, the largest CO value has a substantial affect on the model.
- 4. Below I show that  $SSE = S_{yy} \hat{\beta}_1 S_{xy}$ .

$$SSE = \sum y_i^2 - \hat{\beta}_0 \sum y_i - \hat{\beta}_1 \sum x_i y_i$$

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

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

$$= \sum y_i^2 - \frac{1}{n} \sum y_i \sum y_i + \hat{\beta}_1 (\sum \bar{x} y_i - \sum x_i y_i)$$

$$= \sum y_i^2 - \frac{1}{n} (\sum y_i)^2 - \hat{\beta}_1 ((\sum x_i y_i) - \sum \bar{x} y_i)$$

$$= S_{yy} - \hat{\beta}_1 ((\sum x_i y_i) - \sum \bar{x} y_i)$$

$$= S_{yy} - \hat{\beta}_1 S_{xy}$$