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Stat318 HW10

4)

a)

```
> baseball.data= read.delim("clipboard", header = T)
> attach(baseball.data)
> baseball.mod= lm(BA44 ~ BA43)
> summary(baseball.mod)
```

Call:

```
lm(formula = BA44 ~ BA43)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.052177	-0.017221	0.006301	0.027441	0.039027

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.10368	0.06636	1.562	0.1332
BA43	0.67517	0.25128	2.687	0.0138 *

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

Residual standard error: 0.02968 on 21 degrees of freedom

Multiple R-squared: 0.2558, Adjusted R-squared: 0.2204

F-statistic: 7.22 on 1 and 21 DF, p-value: 0.0138

b)

```
> anova(baseball.mod)
```

Analysis of Variance Table

Response: BA44

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
BA43	1	0.0063595	0.0063595	7.2196	0.0138 *
Residuals	21	0.0184984	0.0008809		

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

c)

Variability explained by the line is the SSR= .0063595

d)

Variability not explained by the line is the SSE=.0184984

e)

The residual standard error provides an estimate for  $\sigma^2$ , as the RSE is equivalent to the square root of the MSE, which is just  $\sigma$ , so squaring the RSE gives us  $\sigma^2$ .  $RSE=.02968$ ,  $.02968^2 = \underline{7.279204E-4 = \sigma^2}$

f)

$R^2 = .2558$ . This low of an  $R^2$  value tells us the line is a bad fit.

Interpretation: 25.58% of the variability in 1944 batting averages is explained by the line with 1943 batting averages.