

SGupta_HW03Question2

Question 2

Test the hypothesis that $\text{Salary} = 0$. Test the hypothesis that $\text{salary} = \text{ratio} = \text{expend} = 0$

```
library(faraway)
set.seed(1001)
data(sat)
head(sat)
```

	expend	ratio	salary	takers	verbal	math	total
Alabama	4.405	17.2	31.144	8	491	538	1029
Alaska	8.963	17.6	47.951	47	445	489	934
Arizona	4.778	19.3	32.175	27	448	496	944
Arkansas	4.459	17.1	28.934	6	482	523	1005
California	4.992	24.0	41.078	45	417	485	902
Colorado	5.443	18.4	34.571	29	462	518	980

```
model1 <- lm(total ~ expend + ratio + salary, data = sat)
summary(model1)
```

Call:

```
lm(formula = total ~ expend + ratio + salary, data = sat)
```

Residuals:

Min	1Q	Median	3Q	Max
-140.911	-46.740	-7.535	47.966	123.329

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1069.234	110.925	9.639	1.29e-12 ***
expend	16.469	22.050	0.747	0.4589
ratio	6.330	6.542	0.968	0.3383
salary	-8.823	4.697	-1.878	0.0667 .

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

Residual standard error: 68.65 on 46 degrees of freedom

Multiple R-squared: 0.2096, Adjusted R-squared: 0.1581

F-statistic: 4.066 on 3 and 46 DF, p-value: 0.01209

```
# Extract t-statistic and p-value for salary
summary(model1)$coefficients["salary", ]
```

	Estimate	Std. Error	t value	Pr(> t)
salary	-8.82263197	4.69679359	-1.87843723	0.06666771

The p value for salary predictor variable is 0.06 which is slightly higher than 0.05 significance level. Therefore we do not have enough evidence to reject null hypothesis. There is evidence that salary has no effect on SAT score.

Testing beta coefficients for expend, ratio and salary all are zero \rightarrow We can use the F-test to check if all predictors have any effect on the response variable. The F statistic value is 4.066 which is less than significant level therefore we reject the null hypothesis that all coefficients are zero. At least one of the predictor variable has an effect on the SAT score.

- expend: $p = 0.4589$ - not significant
- ratio: $p = 0.3383$ - not significant
- salary: $p = 0.0667$ marginal significance at the 10% level, but not at the 5% level

Now add takers to the model. Test the hypothesis that $\text{salary} = 0$.

```
data(sat)
model2 <- lm(total ~ expend + ratio + salary + takers, data = sat)
summary(model2)
```

```
Call:
lm(formula = total ~ expend + ratio + salary + takers, data = sat)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-90.531	-20.855	-1.746	15.979	66.571

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1045.9715	52.8698	19.784	< 2e-16 ***
expend	4.4626	10.5465	0.423	0.674
ratio	-3.6242	3.2154	-1.127	0.266
salary	1.6379	2.3872	0.686	0.496
takers	-2.9045	0.2313	-12.559	2.61e-16 ***

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

Residual standard error: 32.7 on 45 degrees of freedom

Multiple R-squared: 0.8246, Adjusted R-squared: 0.809

F-statistic: 52.88 on 4 and 45 DF, p-value: < 2.2e-16

```
summary(model2)$coefficients["salary", ]
```

	Estimate	Std. Error	t value	Pr(> t)
	1.6379172	2.3872480	0.6861110	0.4961632

Since the p-value (0.4962) is much greater than significance level, we fail to reject the null hypothesis. After accounting for the takers predictors in model2, salary does not appear to have a significant effect on the response.

```
anova(model1, model2)
```

Analysis of Variance Table

Model 1: total ~ expend + ratio + salary

Model 2: total ~ expend + ratio + salary + takers

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	46	216812				
2	45	48124	1	168688	157.74	2.607e-16 ***

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

- Model 1: total expend+ratio+salarytotal expend+ratio+salary
- Model 2:total expend+ratio+salary+takerstotal expend+ratio+salary+takers The RSS is reduced in model 2 and this is the extra variation by including the variable takers.The associated p value in model 2 is extremely small which indicates adding takers predictor variable significantly improved model.