

Application of F-tests

Signif. codes - how significant a predictor is (more stars - more signif.)

6.1 - Data and model

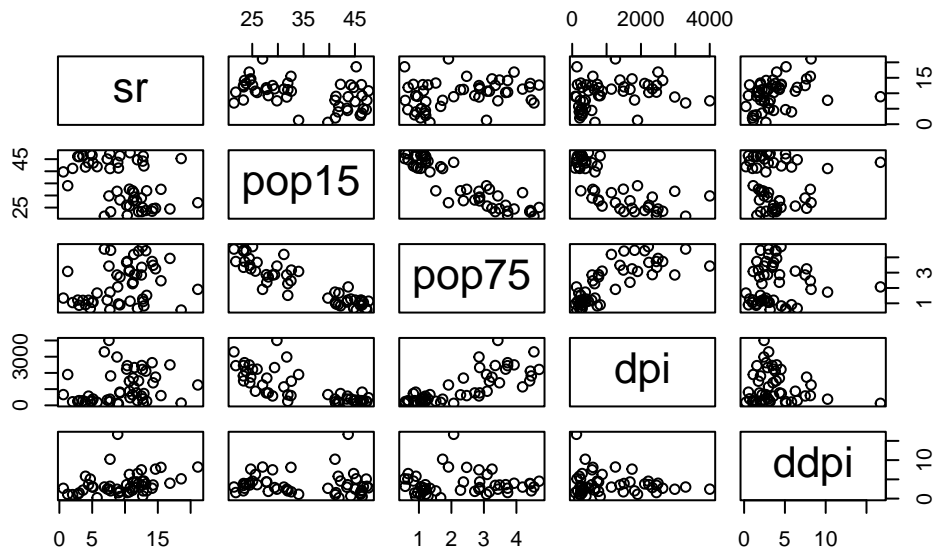
```
library(faraway)
data(savings)
attach(savings)
head(savings)
```

	sr	pop15	pop75	dpi	ddpi
Australia	11.43	29.35	2.87	2329.68	2.87
Austria	12.07	23.32	4.41	1507.99	3.93
Belgium	13.17	23.80	4.43	2108.47	3.82
Bolivia	5.75	41.89	1.67	189.13	0.22
Brazil	12.88	42.19	0.83	728.47	4.56
Canada	8.79	31.72	2.85	2982.88	2.43

```
nrow(savings)
```

```
[1] 50
```

```
pairs(savings)
```



```
fit = lm(sr ~ pop15 + pop75 + dpi + ddp, data = savings)
summary(fit)
```

Call:

```
lm(formula = sr ~ pop15 + pop75 + dpi + ddp, data = savings)
```

Residuals:

Min	1Q	Median	3Q	Max
-8.2422	-2.6857	-0.2488	2.4280	9.7509

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	28.5660865	7.3545161	3.884	0.000334 ***
pop15	-0.4611931	0.1446422	-3.189	0.002603 **
pop75	-1.6914977	1.0835989	-1.561	0.125530
dpi	-0.0003369	0.0009311	-0.362	0.719173
ddp	0.4096949	0.1961971	2.088	0.042471 *

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

Residual standard error: 3.803 on 45 degrees of freedom

Multiple R-squared: 0.3385, Adjusted R-squared: 0.2797

F-statistic: 5.756 on 4 and 45 DF, p-value: 0.0007904

6.2 - Testing all the predictors

```
fit0 = lm(sr ~ 1, data = savings)

X = model.matrix(fit)
n = nrow(X)
p = ncol(X)

RSS = deviance(fit)
RSS0 = deviance(fit0)

F = ((RSS0 - RSS) / (p - 1)) / (RSS / (n - p))
F
```

```
[1] 5.755681
```

```
quantile = qf(0.95, p - 1, n - p)
quantile
```

```
[1] 2.578739
```

```
p_val = 1 - pf(F, p - 1, n - p)
p_val
```

```
[1] 0.0007903779
```

6.3 - F test for single predictor

```
fit1 = lm(sr ~ pop75 + dpi + ddpi)

p1 = 4
RSS1 = deviance(fit1)

F = ((RSS1 - RSS) / (p - p1)) / (RSS / (n - p))
F
```

```
[1] 10.16659
```

```
p_val = 1 - pf(F, p - p1, n - p)
p_val
```

```
[1] 0.002603019
```

t - test

```
X = model.matrix(fit)
beta = coefficients(fit)
c = c(0, 1, 0, 0, 0)

SEpsi = sigma(fit) * sqrt(t(c) %*% solve(t(X) %*% X) %*% c)

t = (t(c) %*% beta) / SEpsi
t
```

```
      [,1]
[1,] -3.18851
```

```
p_val = 2 * (1 - pt(abs(t), n - p))
p_val
```

```
      [,1]
[1,] 0.002603019
```

anova

```
anova(fit1, fit)
```

Analysis of Variance Table

Model 1: sr ~ pop75 + dpi + ddpi

Model 2: sr ~ pop15 + pop75 + dpi + ddpi

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	46	797.72				
2	45	650.71	1	147.01	10.167	0.002603 **

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

6.4 - excluding pop75 and ddpi

```
fit2 = lm(sr ~ pop15 + dpi, data = savings)
anova(fit2, fit)
```

Analysis of Variance Table

Model 1: sr ~ pop15 + dpi

Model 2: sr ~ pop15 + pop75 + dpi + ddpi

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	47	744.12				
2	45	650.71	2	93.411	3.2299	0.04889 *

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

6.5 - same coefficients

```
fit3 = lm(sr ~ I(pop15 + pop75) + dpi + ddpi, data = savings)
anova(fit3, fit)
```

Analysis of Variance Table

Model 1: sr ~ I(pop15 + pop75) + dpi + ddpi

Model 2: sr ~ pop15 + pop75 + dpi + ddpi

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	46	673.63				
2	45	650.71	1	22.915	1.5847	0.2146

6.6 - testing concrete value of coefficient

```
fit4 = lm(sr ~ pop15 + pop75 + dpi + offset(0.5 * ddpi), data = savings)
anova(fit4, fit)
```

Analysis of Variance Table

Model 1: $sr \sim pop15 + pop75 + dpi + offset(0.5 * ddpi)$

Model 2: $sr \sim pop15 + pop75 + dpi + ddpi$

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	46	653.78				
2	45	650.71	1	3.0635	0.2119	0.6475

```
X = model.matrix(fit)
beta = coefficients(fit)
c = c(0, 0, 0, 0, 1)
psi = 0.5
SEpsi = sigma(fit) * sqrt(t(c) %*% solve(t(X) %*% X) %*% c)

t = (t(c) %*% beta - psi) / SEpsi
t
```

```
      [,1]
[1,] -0.4602772
```

```
t ^ 2
```

```
      [,1]
[1,] 0.2118551
```

```
p_val = 2 * (1 - pt(abs(t), n - p))
p_val
```

```
      [,1]
[1,] 0.6475337
```