

PLSC 503 – Spring 2022

“Stupid Regression Tricks” + Multivariate Regression

February 2, 2022

Stupid Regression Tricks

Africa (2001) Data

```
> temp<-getURL("https://raw.githubusercontent.com/PrisonRodeo/PLSC503-2022-git/master/Data/africa2001.csv")
> africa<-read.csv(text=temp, header=TRUE)
> summary(africa)
```

ccode	cabbr	country	population	popthou
Min. :404	AGO : 1	Angola : 1	Min. : 470000	Min. : 470
1st Qu.:452	BDI : 1	Benin : 1	1st Qu.: 3446000	1st Qu.: 3446
Median :510	BEN : 1	Botswana : 1	Median : 9662000	Median : 9662
Mean :510	BWA : 1	Burundi : 1	Mean : 17388558	Mean : 17390
3rd Qu.:556	CAF : 1	Cameroon : 1	3rd Qu.: 19150000	3rd Qu.: 19189
Max. :651	CIV : 1	Central African Republic: 1	Max. :117000000	Max. :116929
	(Other):37	(Other) :37		

popden	polity	gdppppd	tradegdp	war	adrate
Min. :0.0022	Min. : -9.000	Min. : 0.500	Min. : 4.03	Min. :0.000	Min. : 0.10
1st Qu.:0.0134	1st Qu.: -4.500	1st Qu.: 0.855	1st Qu.: 7.64	1st Qu.:0.000	1st Qu.: 2.70
Median :0.0357	Median : 0.000	Median : 1.200	Median : 13.56	Median :0.000	Median : 6.00
Mean :0.0643	Mean : 0.512	Mean : 2.159	Mean : 30.49	Mean :0.116	Mean : 9.37
3rd Qu.:0.0683	3rd Qu.: 5.500	3rd Qu.: 2.040	3rd Qu.: 30.01	3rd Qu.:0.000	3rd Qu.:12.90
Max. :0.5740	Max. :10.000	Max. :10.800	Max. :272.69	Max. :1.000	Max. :38.80

healthexp	subsaharan	muslperc	literacy	internalwar	intensity
Min. :2.00	Not Sub-Saharan: 6	Min. : 0.0	Min. :17.0	Min. :0.000	Min. :0.000
1st Qu.:3.45	Sub-Saharan :37	1st Qu.: 10.0	1st Qu.:43.0	1st Qu.:0.000	1st Qu.:0.000
Median :4.40		Median : 20.0	Median :61.0	Median :0.000	Median :0.000
Mean :4.60		Mean : 36.0	Mean :60.1	Mean :0.302	Mean :0.581
3rd Qu.:5.80		3rd Qu.: 55.5	3rd Qu.:78.5	3rd Qu.:1.000	3rd Qu.:1.000
Max. :8.60		Max. :100.0	Max. :89.0	Max. :1.000	Max. :3.000

A Very Simple Regression

```
> fit<-with(africa, lm(adrates~muslperc))  
> summary(fit)
```

Call:

```
lm(formula = adrates ~ muslperc)
```

Residuals:

Min	1Q	Median	3Q	Max
-13.828	-5.206	0.279	2.022	23.521

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	15.2787	1.8322	8.34	0.00000000023 ***
muslperc	-0.1644	0.0369	-4.45	0.00006390853 ***

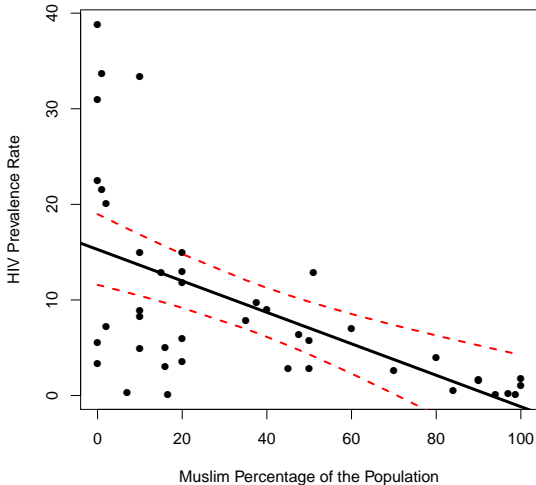
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 8.28 on 41 degrees of freedom

Multiple R-squared: 0.326, Adjusted R-squared: 0.31

F-statistic: 19.8 on 1 and 41 DF, p-value: 0.0000639

Scatterplot of HIV/AIDS Rates on Muslim Population Percentage, Africa 2001



Adding a Constant to X

```
> africa$muslplusten<-africa$muslperc+10
> fit2<-with(africa, lm(adrate~muslplusten,data=africa))
> summary(fit2)
```

Call:

```
lm(formula = adrate ~ muslplusten, data = africa)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-13.828	-5.206	0.279	2.022	23.521

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	16.9232	2.1152	8.00	0.00000000066 ***
muslplusten	-0.1644	0.0369	-4.45	0.00006390853 ***

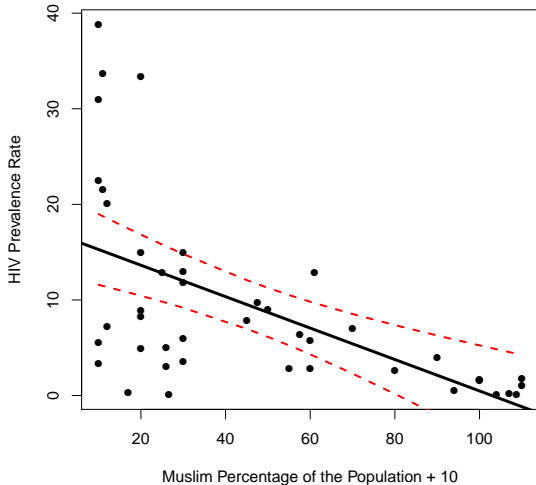
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 8.28 on 41 degrees of freedom

Multiple R-squared: 0.326, Adjusted R-squared: 0.31

F-statistic: 19.8 on 1 and 41 DF, p-value: 0.0000639

Scatterplot of HIV/AIDS Rates on Rescaled Muslim Population Percentage



Multiplying Y by a Constant

```
> africa$screwrate<-africa$adrate*(-314)
> fit3<-with(africa, lm(screwrate~muslperc))
> summary(fit3)
```

Call:

```
lm(formula = screwrate ~ muslperc)
```

Residuals:

Min	1Q	Median	3Q	Max
-7386	-635	-88	1635	4342

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-4797.5	575.3	-8.34	0.00000000023 ***
muslperc	51.6	11.6	4.45	0.00006390853 ***

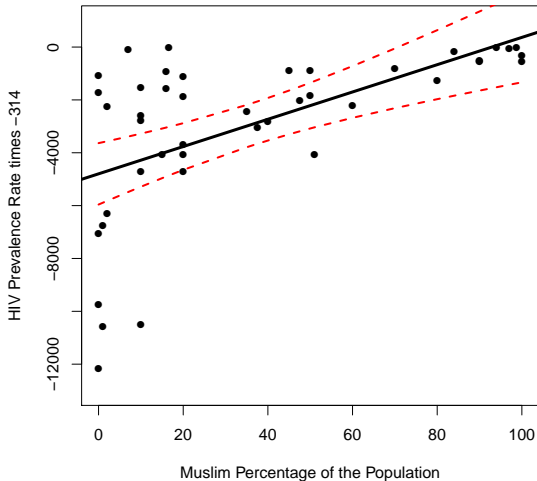
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 2600 on 41 degrees of freedom

Multiple R-squared: 0.326, Adjusted R-squared: 0.31

F-statistic: 19.8 on 1 and 41 DF, p-value: 0.0000639

Scatterplot of Rescaled HIV/AIDS Rates on Muslim Population Percentage



Reversing the scales of X and Y

```
> africa$nonmuslimpct <- 100 - africa$muslperc  
> africa$noninfected <- 100 - africa$adrate  
> fit4<-lm(noninfected~nonmuslimpct,data=africa)  
> summary(fit4)
```

Call:

```
lm(formula = noninfected ~ nonmuslimpct, data = africa)
```

Residuals:

Min	1Q	Median	3Q	Max
-23.521	-2.022	-0.279	5.206	13.828

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	101.1660	2.6808	37.74	< 2e-16 ***
nonmuslimpct	-0.1644	0.0369	-4.45	0.000064 ***

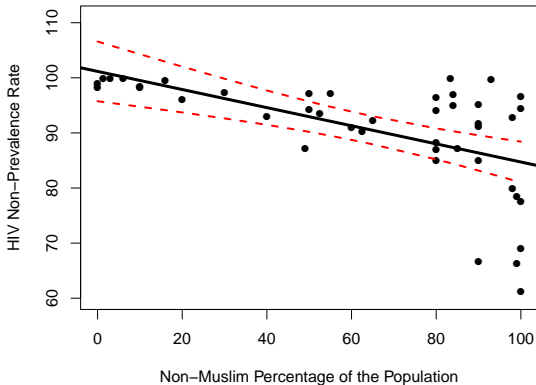
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 8.28 on 41 degrees of freedom

Multiple R-squared: 0.326, Adjusted R-squared: 0.31

F-statistic: 19.8 on 1 and 41 DF, p-value: 0.0000639

Scatterplot of HIV/AIDS Non-Infection Rates on Non-Muslim Population Percentage



Linear Transformations

- Adding (subtracting) a positive constant to X shifts the X -axis to the left (right).
- Adding (subtracting) a positive constant to Y shifts the Y -axis downwards (upwards).
- Multiplying X (Y) times a positive constant greater than 1.0 stretches the X (Y) axis.
- Multiplying X (Y) times a positive constant less than 1.0 shrinks the X (Y) axis.
- Multiplying X (Y) times a negative constant inverts the X (Y) axis, and stretches / shrinks it as above.

Use: “Centering” a Variable

```
> africa$muslcenter<-africa$muslperc - mean(africa$muslperc, na.rm=TRUE)
> fit5<-lm(adrate~muslcenter,data=africa)
> summary(fit5)
```

Call:

```
lm(formula = adrate ~ muslcenter, data = africa)
```

Residuals:

Min	1Q	Median	3Q	Max
-13.828	-5.206	0.279	2.022	23.521

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	9.3651	1.2622	7.42	0.0000000042 ***
muslcenter	-0.1644	0.0369	-4.45	0.0000639085 ***

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

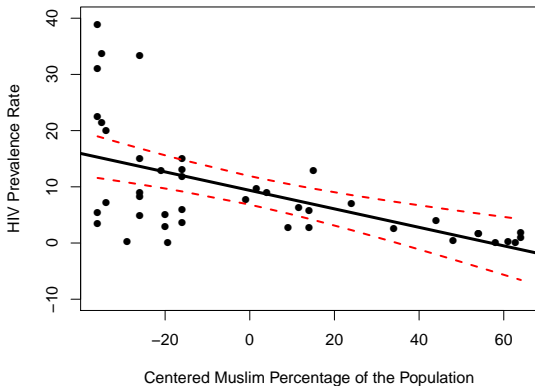
Residual standard error: 8.28 on 41 degrees of freedom

Multiple R-squared: 0.326, Adjusted R-squared: 0.31

F-statistic: 19.8 on 1 and 41 DF, p-value: 0.0000639

```
> mean(africa$adrate)
[1] 9.365116
```

Scatterplot of HIV/AIDS Infection Rates on (Centered) Muslim Population Percentage



Use: Rescaling X for Interpretability

```
> fit6<-lm(adrate~population,data=africa)
> summary(fit6)
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	10.5883163475	1.9140361989	5.53	0.000002 ***
population	-0.0000000703	0.0000000671	-1.05	0.3

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

Residual standard error: 9.95 on 41 degrees of freedom

Multiple R-squared: 0.0261, Adjusted R-squared: 0.00234

F-statistic: 1.1 on 1 and 41 DF, p-value: 0.301

```
> africa$popmil<-africa$population / 1000000
> fit7<-lm(adrate~popmil,data=africa)
> summary(fit7)
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	10.5883	1.9140	5.53	0.000002 ***
popmil	-0.0703	0.0671	-1.05	0.3

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

Residual standard error: 9.95 on 41 degrees of freedom

Multiple R-squared: 0.0261, Adjusted R-squared: 0.00234

F-statistic: 1.1 on 1 and 41 DF, p-value: 0.301

Dichotomous Xs: Bivariate Regression \equiv *t*-test

```
> fit8<-lm(adrate~subsaharan,data=africa)
> summary(fit8)
```

Residuals:

Min	1Q	Median	3Q	Max
-10.58	-6.23	-1.78	2.22	28.12

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.27	3.88	0.33	0.75
subsaharanSub-Saharan	9.41	4.19	2.25	0.03 *

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

Residual standard error: 9.51 on 41 degrees of freedom

Multiple R-squared: 0.11, Adjusted R-squared: 0.088

F-statistic: 5.05 on 1 and 41 DF, p-value: 0.03

```
> with(africa,
+       t.test(adrate~subsaharan, var.equal=TRUE))
```

Two Sample t-test

data: adrate by subsaharan

t = -2.2, df = 41, p-value = 0.03

alternative hypothesis: true difference in means is not equal to 0

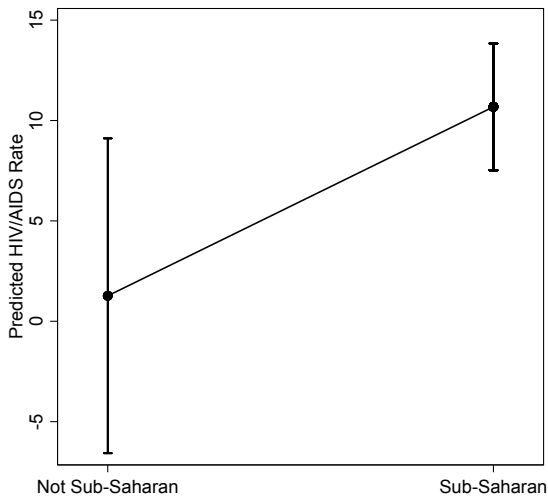
95 percent confidence interval:

-17.8659 -0.9576

sample estimates:

mean in group Not Sub-Saharan	mean in group Sub-Saharan
1.267	10.678

Expected Values of HIV/AIDS Infection Rates in Saharan and Sub-Saharan Africa



The results:

```
> fit<-lm(adrate~muslperc, data=africa)
```

```
> summary.lm(fit)
```

Call:

```
lm(formula = adrate ~ muslperc, data = africa)
```

Residuals:

Min	1Q	Median	3Q	Max
-13.828	-5.206	0.279	2.022	23.521

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	15.2787	1.8322	8.34	0.00000000023 ***
muslperc	-0.1644	0.0369	-4.45	0.00006390853 ***

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

Residual standard error: 8.28 on 41 degrees of freedom

Multiple R-squared: 0.326, Adjusted R-squared: 0.31

F-statistic: 19.8 on 1 and 41 DF, p-value: 0.0000639

The table:

Table: OLS Regression Model of HIV/AIDS Rates in Africa, 2001

Variables	Model I
(Constant)	15.28 (1.83)
Muslim Percentage of the Population	-0.164* (0.037)
Adjusted R^2	0.31

Note: $N = 43$. Cell entries are coefficient estimates; numbers in parentheses are estimated standard errors. Asterisks indicate $p < .05$ (one-tailed). See text for details.

Another Table (using default-y stargazer)

Table: OLS Regression Model of HIV/AIDS Rates in Africa, 2001

	Model I
(Constant)	15.28*** (1.83)
Muslim Percentage of the Population	-0.16*** (0.04)
Observations	43
R ²	0.33
Adjusted R ²	0.31
Residual Std. Error	8.28 (df = 41)
F Statistic	19.83*** (df = 1; 41)

Note:

*p<0.1; **p<0.05; ***p<0.01

Some Guidelines (“Rules”?)

Tables:

- *Use column headings descriptively.*
- *Use multiple rows / columns rather than multiple tables.*
- *Learn about significant digits, and don't report more than 4-5 of them.*
- *Use a figure to replace a table when you can.*
- *Be aware of norms about *s.*

Figures:

- *Report the scale of axes, and label them.*
- *Use as much “space” as you need, but no more.*
- *Use color sparingly.*

“Multivariate” Regression

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{u}$$

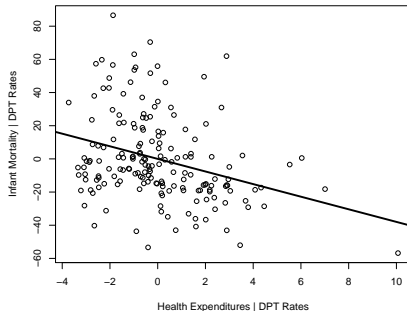
$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \dots + \beta_K X_{Ki} + u_i$$

$$\begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_N \end{bmatrix} = \begin{bmatrix} 1 & X_{11} & X_{21} & \cdots & X_{K1} \\ 1 & X_{12} & X_{22} & \cdots & X_{K2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & X_{1N} & X_{2N} & \cdots & X_{KN} \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_K \end{bmatrix} + \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_N \end{bmatrix}.$$

Diversion: “Added Variable Plots”

- Regress Y on X_1 and save the residuals \hat{u}_i ,
- Regress X_2 on X_1 and save the residuals (call these \hat{e}_i),
- Plot \hat{u}_i (conventionally on the y -axis) vs. \hat{e}_i (conventionally on the x -axis).

Example: Infant Mortality and Health Expenditures Given DPT Immunization Rates



Residuals:

$$\mathbf{u} = \mathbf{Y} - \mathbf{X}\beta$$

The inner product of \mathbf{u} :

$$\begin{aligned} \mathbf{u}'\mathbf{u} &= \begin{bmatrix} u_1 & u_2 & \cdots & u_N \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_N \end{bmatrix} \\ &= u_1^2 + u_2^2 + \dots + u_N^2 \\ &= \sum_{i=1}^N u_i^2 \end{aligned}$$

$$\begin{aligned}\mathbf{u}'\mathbf{u} &= (\mathbf{Y} - \mathbf{X}\beta)'(\mathbf{Y} - \mathbf{X}\beta) \\ &= \mathbf{Y}'\mathbf{Y} - 2\beta'\mathbf{X}'\mathbf{Y}' + \beta'\mathbf{X}'\mathbf{X}\beta\end{aligned}$$

Now get:

$$\frac{\partial \mathbf{u}'\mathbf{u}}{\partial \beta} = -2\mathbf{X}'\mathbf{Y} + 2\mathbf{X}'\mathbf{X}\beta$$

Solve:

$$\begin{aligned}-2\mathbf{X}'\mathbf{Y} + 2\mathbf{X}'\mathbf{X}\beta &= 0 \\ -\mathbf{X}'\mathbf{Y} + \mathbf{X}'\mathbf{X}\beta &= 0 \\ \mathbf{X}'\mathbf{X}\beta &= \mathbf{X}'\mathbf{Y} \\ (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{X}\beta &= (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y} \\ \beta &= (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}\end{aligned}$$

“Do not compute the least squares estimates using $(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}$!”

– Weisberg (p. 61)

Most software uses:

$$\mathbf{X} = \mathbf{QR}$$

where \mathbf{Q} is orthogonal ($\mathbf{Q}'\mathbf{Q} = \mathbf{I}$) and \mathbf{R} is upper-triangular.

Why??? See e.g. [here](#), or section 3.19, [here](#)

1. Expectation-Zero Disturbances

$$E(\mathbf{u}) = \mathbf{0}$$

2. Homoscedasticity / No Error Correlation

$$\begin{aligned} \mathbf{u}\mathbf{u}' &= \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_N \end{bmatrix} \begin{bmatrix} u_1 & u_2 & \cdots & u_N \end{bmatrix} \\ &= \begin{bmatrix} u_1^2 & u_1 u_2 & \cdots & u_1 u_N \\ u_2 u_1 & u_2^2 & \cdots & u_2 u_N \\ \vdots & \vdots & \ddots & \vdots \\ u_N u_1 & u_N u_2 & \cdots & u_N^2 \end{bmatrix} \end{aligned}$$

Expectation must be:

$$E(\mathbf{u}\mathbf{u}') = \begin{bmatrix} \sigma^2 & 0 & \cdots & 0 \\ 0 & \sigma^2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \sigma^2 \end{bmatrix} = \sigma^2 \mathbf{I}_{N \times N}$$

3. “Fixed” \mathbf{X} ...

- No *measurement error* in the \mathbf{X} s, and
- $\text{Cov}(\mathbf{X}, \mathbf{u}) = \mathbf{0}$.

4. \mathbf{X} is full column rank.

Means:

- no exact linear relationship among \mathbf{X} , and
- $K < N$.

5. Normal Disturbances

$$\mathbf{u} \sim \mathbf{N}(\mathbf{0}, \sigma^2 \mathbf{I})$$

Unbiasedness:

$$\mathbf{Y} = \mathbf{X}\beta + \mathbf{u}$$

Substitute OLS $\hat{\beta}$:

$$\begin{aligned}\hat{\beta} &= (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'(\mathbf{X}\beta + \mathbf{u}) \\ &= (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{X}\beta + (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u} \\ &= \beta + (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u}\end{aligned}$$

and so:

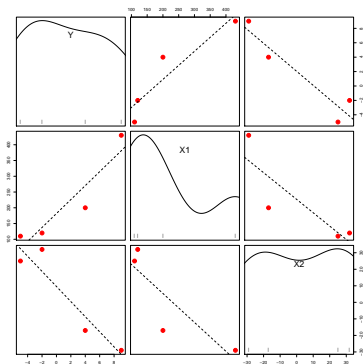
$$\hat{\beta} - \beta = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u}.$$

By $\text{Cov}(\mathbf{X}, \mathbf{u}) = \mathbf{0}$, we have $E(\hat{\beta}) = \beta$.

A Toy Example

$$\mathbf{Y} = \begin{bmatrix} 4 \\ -2 \\ 9 \\ -5 \end{bmatrix}$$

$$\mathbf{X} = \begin{bmatrix} 1 & 200 & -17 \\ 1 & 120 & 32 \\ 1 & 430 & -29 \\ 1 & 110 & 25 \end{bmatrix}$$



Example, continued

So:

$$\mathbf{X}'\mathbf{X} = \begin{bmatrix} 4 & 860 & 11 \\ 860 & 251400 & -9280 \\ 11 & -9280 & 2779 \end{bmatrix}$$

$$(\mathbf{X}'\mathbf{X})^{-1} = \begin{bmatrix} 3.2453 & -0.0132 & -0.05694 \\ -0.0132 & 0.000058 & 0.0002468 \\ -0.0569 & 0.000247 & 0.001409 \end{bmatrix}$$

$$\mathbf{X}'\mathbf{Y} = \begin{bmatrix} 6 \\ 3880 \\ 518 \end{bmatrix}$$

So:

$$\begin{aligned} \hat{\beta} &= \begin{bmatrix} 3.2453 & -0.0132 & -0.05694 \\ -0.0132 & 0.000058 & 0.0002468 \\ -0.0569 & 0.000247 & 0.001409 \end{bmatrix} \begin{bmatrix} 6 \\ 3880 \\ 518 \end{bmatrix} \\ &= \begin{bmatrix} -2.264 \\ 0.0190 \\ -0.1141 \end{bmatrix} \end{aligned}$$

Minimal Example: Correlation

```
Y<-c(4,-2,9,-5)
X1<-c(200,120,430,110)
X2<-c(-17,32,-29,25)
data<-cbind(Y,X1,X2)
```

```
cor(data)
```

	Y	X1	X2
Y	1.0000	0.9285	-0.9425
X1	0.9285	1.0000	-0.8613
X2	-0.9425	-0.8613	1.0000

→ Regression

```
fit<-lm(Y~X1+X2)
summary(fit)
```

```
Call:
lm(formula = Y ~ X1 + X2)
```

Residuals:

1	2	3	4
0.531	1.639	-0.201	-1.970

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-2.2643	4.7284	-0.48	0.72
X1	0.0190	0.0200	0.95	0.52
X2	-0.1141	0.0985	-1.16	0.45

Residual standard error: 2.62 on 1 degrees of freedom

Multiple R-Squared: 0.941, Adjusted R-squared: 0.823

F-statistic: 7.99 on 2 and 1 DF, p-value: 0.243

- Pick some $\mathbf{H}_A : \boldsymbol{\beta} = \boldsymbol{\beta}_A$
- Estimate $\hat{\boldsymbol{\beta}}$
- Determine distribution of $\hat{\boldsymbol{\beta}}$ under \mathbf{H}_A
- Form a *test statistic* $\hat{\mathbf{S}} = h(\boldsymbol{\beta}, \hat{\boldsymbol{\beta}})$
- Assess $\Pr(\hat{\mathbf{S}}|\mathbf{H}_A)$

The Importance of $\mathbf{V}(\hat{\beta})$

$$\begin{aligned}\mathbf{V}(\hat{\beta}) &= E[\hat{\beta} - E(\hat{\beta})]^2 \\ &= E\{[\hat{\beta} - E(\hat{\beta})][\hat{\beta} - E(\hat{\beta})]'\}\end{aligned}$$

Rewrite:

$$\begin{aligned}\mathbf{V}(\hat{\beta}) &= E(\hat{\beta} - \beta)(\hat{\beta} - \beta)' \\ &= E\{[(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u}][(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u}]'\} \\ &= E[(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u}\mathbf{u}'\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}]\end{aligned}$$

The Importance of $\mathbf{V}(\hat{\beta})$

Taking expectations:

$$\begin{aligned}\mathbf{V}(\hat{\beta}) &= (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{E}(\mathbf{u}\mathbf{u}')\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1} \\ &= (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\sigma^2\mathbf{I}\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1} \\ &= \sigma^2(\mathbf{X}'\mathbf{X})^{-1}\end{aligned}$$

Estimating $\mathbf{V}(\hat{\beta})$

Empirical estimate:

$$\hat{\sigma}^2 = \frac{\hat{\mathbf{u}}' \hat{\mathbf{u}}}{N - K}$$

Yields:

$$\widehat{\mathbf{V}(\hat{\beta})} = \hat{\sigma}^2 (\mathbf{X}' \mathbf{X})^{-1}$$

Single Coefficient Hypothesis Tests

We know that:

$$\hat{\beta} \sim \mathcal{N}[\beta, \sigma^2(\mathbf{X}'\mathbf{X})^{-1}].$$

In practice, using $\hat{\sigma}^2$ means

$$\hat{\beta} - \beta \sim t_{N-K}$$

Procedure:

- Choose a value of β_k that you want to test (say, $\beta_k = 0$),
- Calculate the t -statistic for the coefficient associated with X_k , which is:

$$\hat{t}_k = \frac{\hat{\beta}_k - \beta_k}{\sqrt{\mathbf{V}(\hat{\beta}_k)}}$$

- Compare \hat{t}_k to a t distribution with $N - K$ degrees of freedom.

Multivariate Hypothesis Testing

E.g.: $H_0 : \beta_1 = \beta_2 = \dots = \beta_K = 0$

or: $H_0 : \beta_3 = \beta_6 = 0$

Generally: *Linear restrictions*:

$$\underset{q \times k}{\mathbf{R}} \underset{k \times 1}{\boldsymbol{\beta}} = \underset{q \times 1}{\mathbf{r}}$$

E.g.:

$$\beta_2 = -2 \iff (0 \ 1 \ 0) \begin{pmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \end{pmatrix} = -2$$

Recall:

$$\text{TSS} = \text{MSS} + \text{RSS}$$

Consider:

$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \beta_3 X_{3i} + \beta_4 X_{4i} + u_{Ui}$$

and the restriction:

$$H_a : \beta_2 = \beta_4 = 0.$$

Restricted model:

$$\begin{aligned} Y_i &= \beta_0 + \beta_1 X_{1i} + 0X_{2i} + \beta_3 X_{3i} + 0X_{4i} + u_i \\ &= \beta_0 + \beta_1 X_{1i} + \beta_3 X_{3i} + u_{Ri} \end{aligned}$$

F-tests: Sums of Squared Residuals

“Unrestricted”:

$$RSS_U \equiv \hat{\mathbf{u}}_U' \hat{\mathbf{u}}_U = \sum_{i=1}^N \hat{u}_{Ui}^2$$

“Restricted”:

$$RSS_R \equiv \hat{\mathbf{u}}_R' \hat{\mathbf{u}}_R = \sum_{i=1}^N \hat{u}_{Ri}^2$$

F-statistic:

$$\begin{aligned}\mathbf{F} &= \frac{(\text{RSS}_R - \text{RSS}_U)/q}{\text{RSS}_U/(N - K)} \\ &= \frac{(R_U^2 - R_R^2)/q}{(1 - R_U^2)/(N - K)}\end{aligned}$$

Testing:

$$\mathbf{F} \sim F_{q, N-K}$$

Consider:

$$\begin{aligned}H_b : \quad & \beta_1 + \beta_4 = 1 \\ & \beta_1 = 1 - \beta_4\end{aligned}$$

Implies:

$$\begin{aligned}Y_i &= \beta_0 + (1 - \beta_4)X_{1i} + \beta_2X_{2i} + \beta_3X_{3i} + \beta_4X_{4i} + u_{R'i} \\ &= \beta_0 + X_{1i} - \beta_4X_{1i} + \beta_2X_{2i} + \beta_3X_{3i} + \beta_4X_{4i} + u_{R'i} \\ &= \beta_0 + X_{1i} + \beta_2X_{2i} + \beta_3X_{3i} + \beta_4(X_{4i} - X_{1i}) + u_{R'i}\end{aligned}$$

implying restricted model:

$$Y_i - X_{1i} = \beta_0 + \beta_2X_{2i} + \beta_3X_{3i} + \beta_4(X_{4i} - X_{1i}) + u_{R'i}$$

$$F = \frac{(\hat{\beta}_q - \beta_q^H)' \hat{\mathbf{V}}_q^{-1} (\hat{\beta}_q - \beta_q^H)}{q \hat{\sigma}^2}$$

Implies:

$$\Pr \left[\frac{(\hat{\beta}_q - \beta_q^H)' \hat{\mathbf{V}}_q^{-1} (\hat{\beta}_q - \beta_q^H)}{q \hat{\sigma}^2} \leq F_{q, N-K} \right] = 1 - \alpha. \quad (1)$$

→ “confidence region” of all points satisfying:

$$(\hat{\beta}_q - \beta_q^H)' \hat{\mathbf{V}}_q^{-1} (\hat{\beta}_q - \beta_q^H) \leq q \hat{\sigma}^2 F_{q, N-K}.$$

Prediction:

$$\hat{Y}_j = \mathbf{X}_j \hat{\beta}$$

Variance:

$$\widehat{\mathbf{V}(\hat{Y}_j)} = \hat{\sigma}^2 [1 + \mathbf{X}_j (\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}_j']$$

Standard error:

$$\widehat{\text{s.e.}(\hat{Y}_j)} = \sqrt{\hat{\sigma}^2 [1 + \mathbf{X}_j (\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}_j']}$$

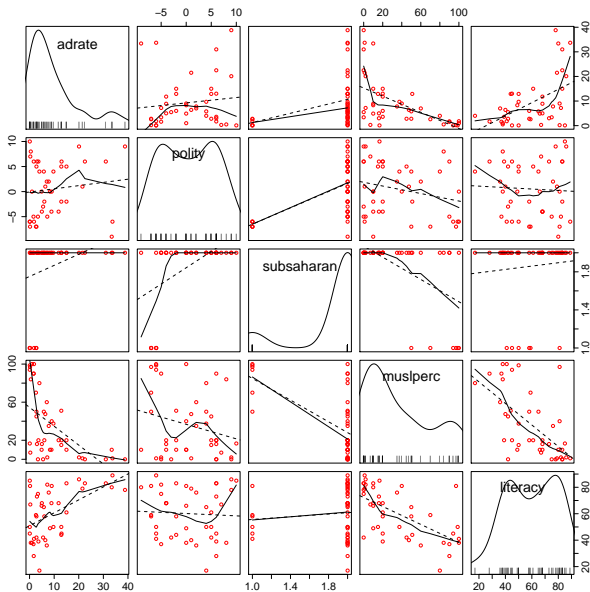
Example: Africa Data

```
> library(RCurl)
> temp<-getURL("https://raw.githubusercontent.com/PrisonRodeo/PLSC503-2022-git/master/Data/africa2001.csv")
> Data<-read.csv(text=temp, header=TRUE)
> Data<-with(Data, data.frame(adrates, polity,
+   subsaharan=as.numeric(as.factor(subsaharan))-1,
+   muslperc, literacy))

> summary(Data)
      adrate      polity      subsaharan      muslperc      literacy
Min.   : 0.10  Min.   :-9.000  Min.   :0.000  Min.   : 0.0  Min.   :17.0
1st Qu.: 2.70  1st Qu.: -4.500  1st Qu.:1.000  1st Qu.: 10.0  1st Qu.:43.0
Median : 6.00  Median : 0.000  Median :1.000  Median : 20.0  Median :61.0
Mean   : 9.37  Mean   : 0.512  Mean   :0.861  Mean   : 36.0  Mean   :60.1
3rd Qu.:12.90  3rd Qu.: 5.500  3rd Qu.:1.000  3rd Qu.: 55.5  3rd Qu.:78.5
Max.   :38.80  Max.   :10.000  Max.   :1.000  Max.   :100.0  Max.   :89.0

> cor(Data)
      adrate      polity      subsaharan      muslperc      literacy
adrates    1.0000  0.11794    0.33129   -0.5709  0.51489
polity     0.1179  1.00000    0.52820   -0.2392 -0.05079
subsaharan 0.3313  0.52820    1.00000   -0.5773  0.09473
muslperc   -0.5709 -0.23917   -0.57725    1.0000 -0.61960
literacy    0.5149 -0.05079    0.09473   -0.6196 1.00000
```


Africa Data



A Regression

```
> model<-lm(adrate~polity+subsaharan+muslperc+literacy,data=Data)
> summary(model)
```

Call:

```
lm(formula = adrate ~ polity + subsaharan + muslperc + literacy,
    data = Data)
```

Residuals:

Min	1Q	Median	3Q	Max
-15.4681	-4.3947	-0.5251	3.4246	22.9358

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-4.39843	14.94744	-0.294	0.7702
polity	-0.01390	0.27969	-0.050	0.9606
subsaharan	3.72969	5.43093	0.687	0.4964
muslperc	-0.08689	0.06282	-1.383	0.1747
literacy	0.16575	0.09433	1.757	0.0869 .

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

Residual standard error: 8.264 on 38 degrees of freedom

Multiple R-squared: 0.3771, Adjusted R-squared: 0.3115

F-statistic: 5.751 on 4 and 38 DF, p-value: 0.001013

Variance-Covariance Matrix of $\hat{\beta}$

```
> options(digits=4)
> vcov(model)
```

	(Intercept)	polity	subsaharan	muslperc	literacy
(Intercept)	223.4259	1.088030	-72.2628	-0.771309	-1.002421
polity	1.0880	0.078229	-0.6642	-0.000293	0.001968
subsaharan	-72.2628	-0.664212	29.4950	0.206067	0.171765
muslperc	-0.7713	-0.000293	0.2061	0.003946	0.004098
literacy	-1.0024	0.001968	0.1718	0.004098	0.008898

Test $H_0 : \beta_{\text{polity}} = \beta_{\text{subsaharan}} = 0$:

```
> library(lmtest)
> modelsmall<-lm(adrate~muslperc+literacy,data=Data)
> waldtest(model,modelsmall)
```

Wald test

Model 1: adrate ~ polity + subsaharan + muslperc + literacy

Model 2: adrate ~ muslperc + literacy

	Res.Df	Df	F	Pr(>F)
1	38			
2	40	-2	0.27	0.76

Test $H_0 : \beta_{\text{muslperc}} = 0.1$:

```
> library(car)
> linearHypothesis(model,"muslperc=0.1")
```

Linear hypothesis test

Hypothesis:
muslperc = 0.1

Model 1: restricted model

Model 2: adrate ~ polity + subsaharan + muslperc + literacy

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	39	3200				
2	38	2595	1	605	8.85	0.0051 **

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

Test $H_0 : \beta_{\text{literacy}} = \beta_{\text{muslperc}}$:

```
> linearHypothesis(model,"literacy=muslperc")
```

Linear hypothesis test

Hypothesis:

- muslperc + literacy = 0

Model 1: restricted model

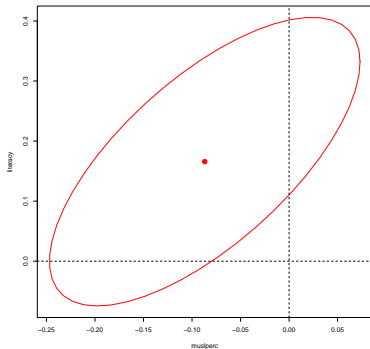
Model 2: adrate ~ polity + subsaharan + muslperc + literacy

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	39	3534				
2	38	2595	1	938	13.7	0.00067 ***

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

Confidence Regions / Ellipses

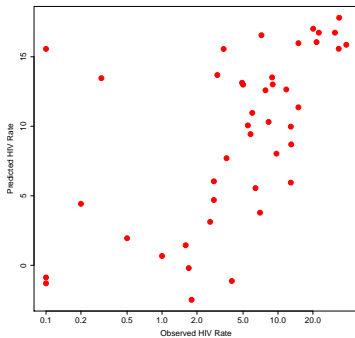
```
> confidenceEllipse(model=model,which.coef=c(4,5),  
                    xlab="Muslim Percentage",ylab="Literacy")  
> abline(h=0,v=0,lty=2)
```



Predicted Values

```
> hats<-fitted(model)
> # Or, alternatively:
> fitted<-predict(model,se.fit=TRUE, interval=c("confidence"))
> scatterplot(model$fitted~adrate,log="x",smooth=FALSE,boxplots=FALSE,
  reg.line=FALSE,xlab="Observed HIV Rate",ylab="Predicted HIV Rate",
  pch=16,cex=2)
```

Predicted and Actual HIV/AIDS Rates (X-Axis Logged)



An Even More Useful Plot

```
> library(plotrix)
> plotCI(Data$adrate,model$fitted,uiw=(1.96*(fitted$se.fit)),
         log="x",xlab="Observed HIV Rate",ylab="Predicted HIV Rate")
> lines(lowess(Data$adrate>Data$adrate),lwd=2)
```

Predicted and Actual HIV/AIDS Rates, with 95% C.I.s

