

QUANTITATIVE ANALYSIS

BIVARIATE REGRESSION

AGENDA

1. Follow-up
2. Tables and Plots
3. The Failings of Simple Models
4. Bivariate Regression Theory
5. Bivariate Regression in Stata

1 FOLLOW-UP

2 TABLES AND PLOTS

TABLES

- ▶ Table format will depend on medium - print vs. projection

	mpg	weight	foreign
mpg	1.00	-	-
weight	-.807***	1.00	-
foreign	.393***	-.593***	1.00

* - $p < .05$; ** - $p < .01$; *** - $p < .001$

TABLES

- Tables for projection should have less detail and emphasize readability

	mpg	weight	foreign
mpg	1.00	-	-
weight	-.807***	1.00	-
foreign	.393***	-.593***	1.00

* - $p < .05$; ** - $p < .01$; *** - $p < .001$

TABLES

- ▶ Print tables use minimum horizontal lines
- ▶ Tables are included at end of paper with references left in text where they should be placed: <<<<< TABLE 2 ABOUT HERE >>>>>

TABLE 2 - Pearson's r Correlation Results

	miles per gallon	weight (lbs.)	foreign
miles per gallon	1.00		
weight (lbs.)	-.807***	1.00	
foreign	.393***	-.593***	1.00

Notes: * - $p < .05$; ** - $p < .01$; *** - $p < .001$

2. TABLES AND PLOTS

TABLES

- Use your word processor of choice to create a table and edit it to match requested format for journal submissions

	A	B	C	D
1	TABLE 2 - Pearson's r Correlation Results			
2		miles per gallon	weight (lbs.)	foreign
3	miles per gallon	1.00		
4	weight (lbs.)	-.807***	1.00	
5	foreign	.393***	-.593***	1.00
6	Notes: * - $p < .05$; ** - $p < .01$; *** - $p < .001$			

PLOTS

- ▶ Plot format will depend on medium - print vs. projection
- ▶ Plots for projection can (and often should!) use color with fonts sized appropriately for slides
- ▶ Many journals have a specific standard for how plots are designed, and typically required black and white or grayscale plots
- ▶ Fonts for print plots need to be sized appropriately for printing

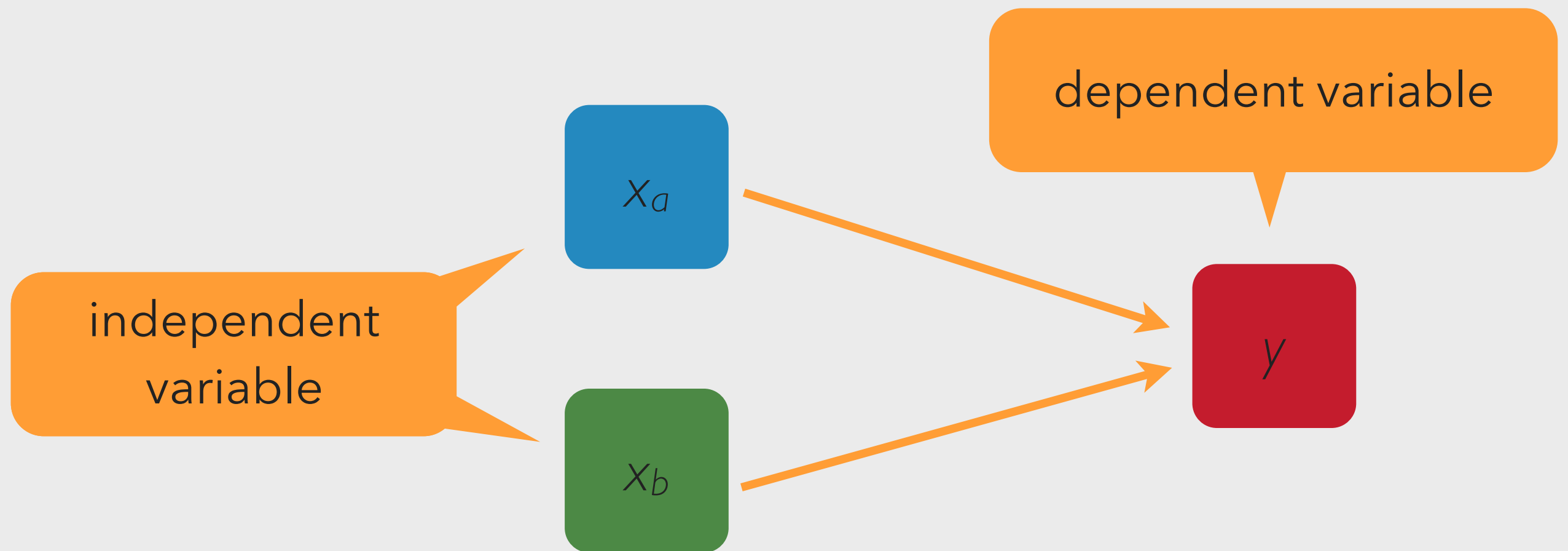
PRESENTATIONS

► See:

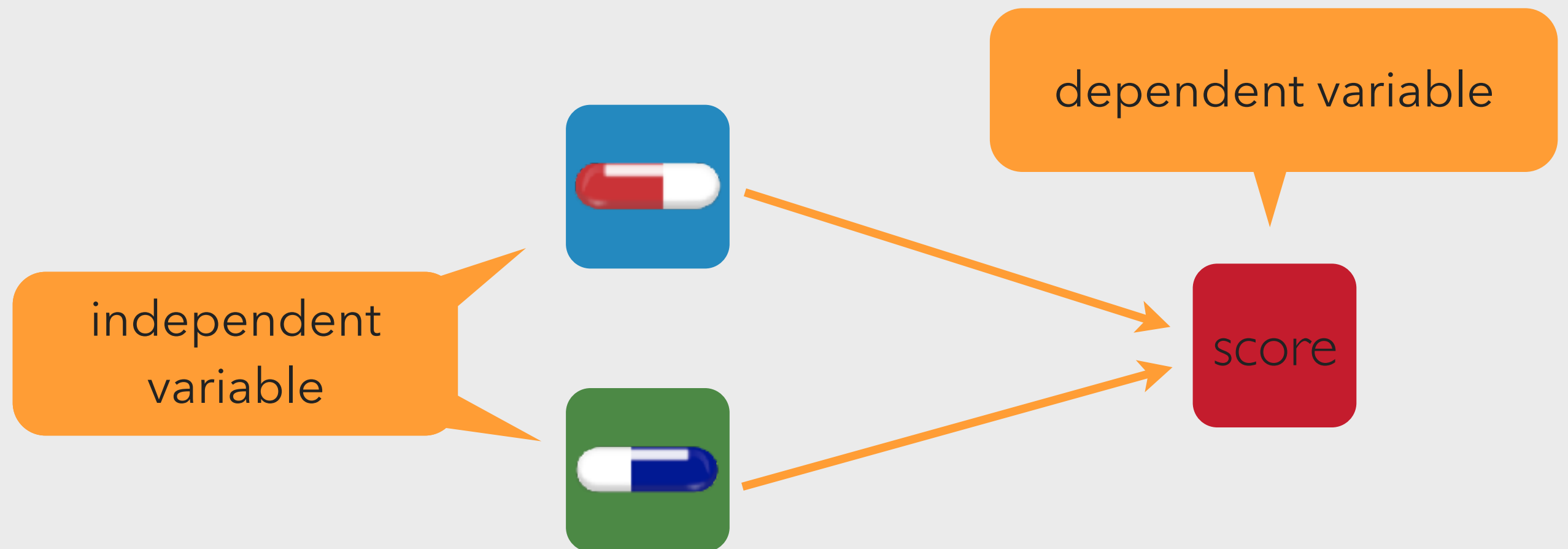
<https://github.com/chris-prener/Presentations>

3 THE FAILINGS OF SIMPLE MODELS

DIFFERENCE OF MEANS

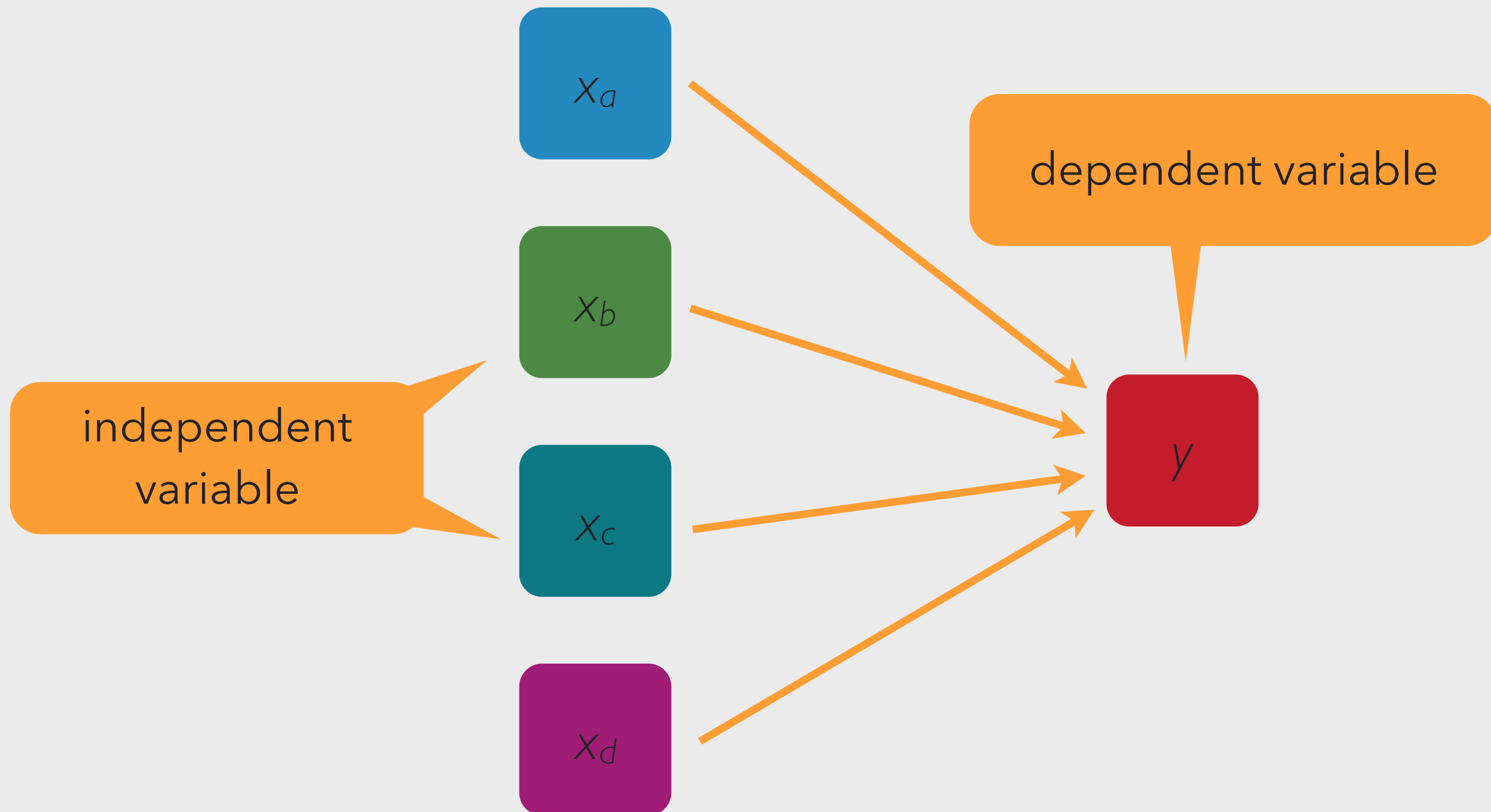


DIFFERENCE OF MEANS



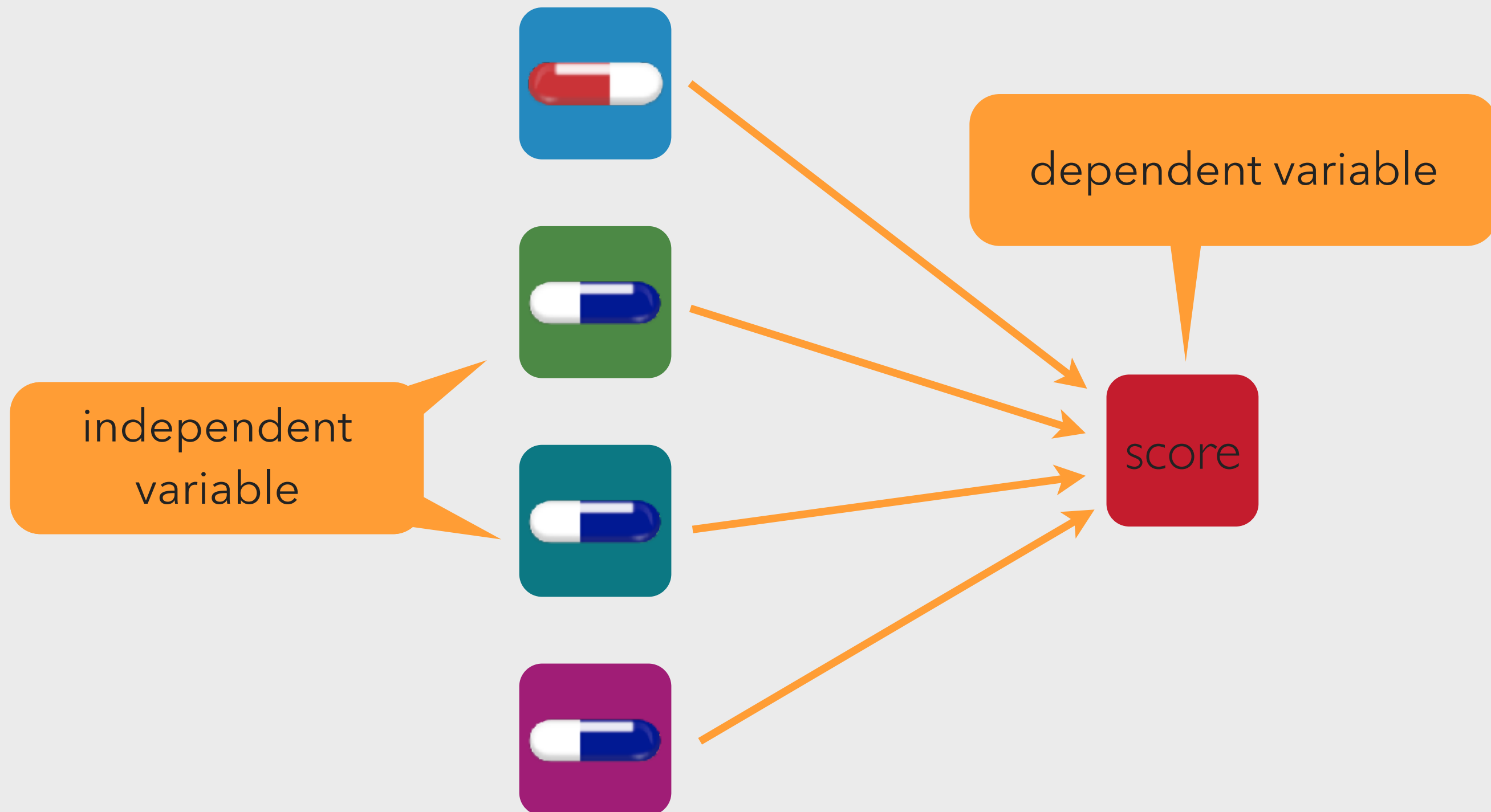
3. THE FAILINGS OF SIMPLE MODELS

ANOVA



3. THE FAILINGS OF SIMPLE MODELS

ANOVA



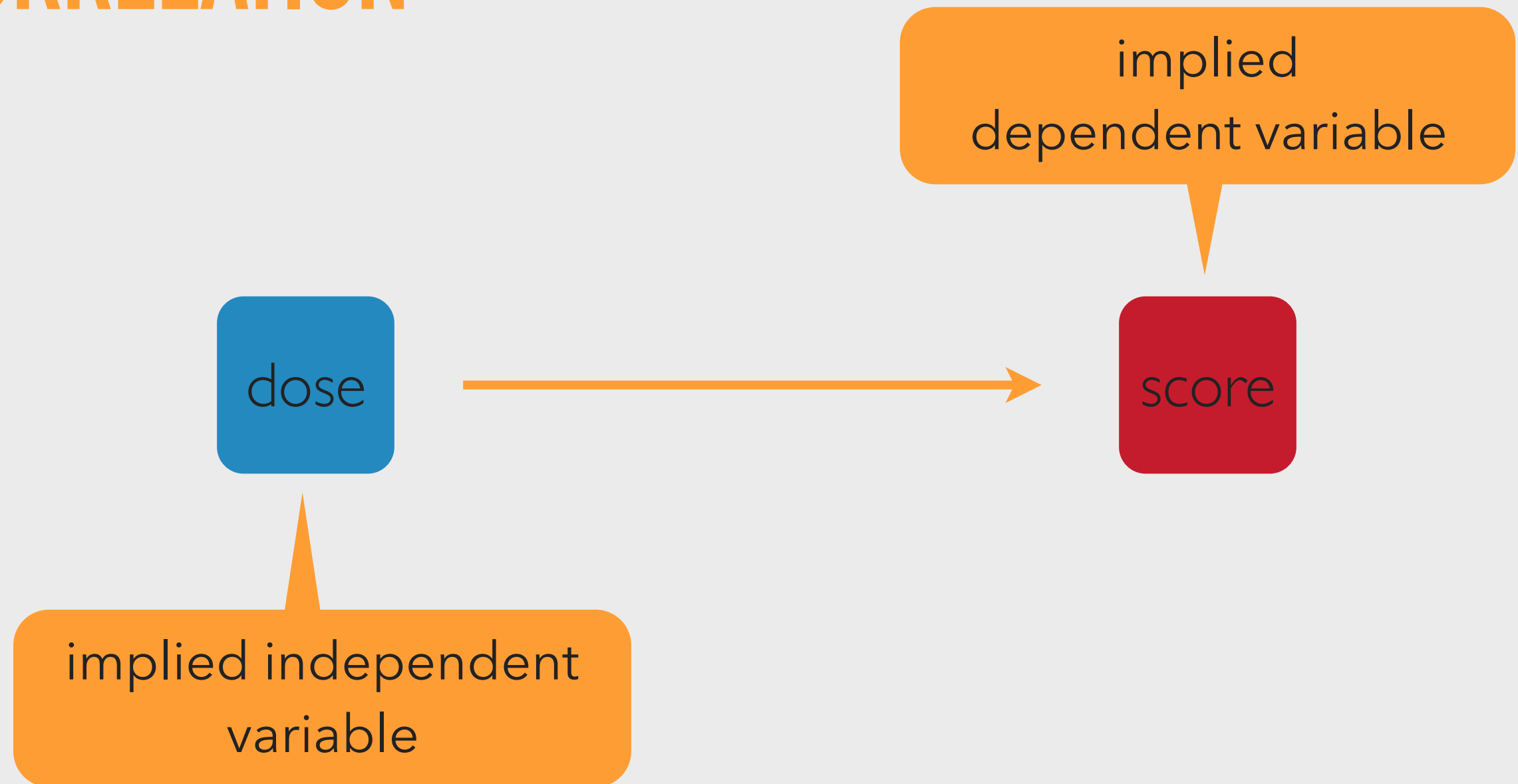
3. THE FAILINGS OF SIMPLE MODELS

CORRELATION



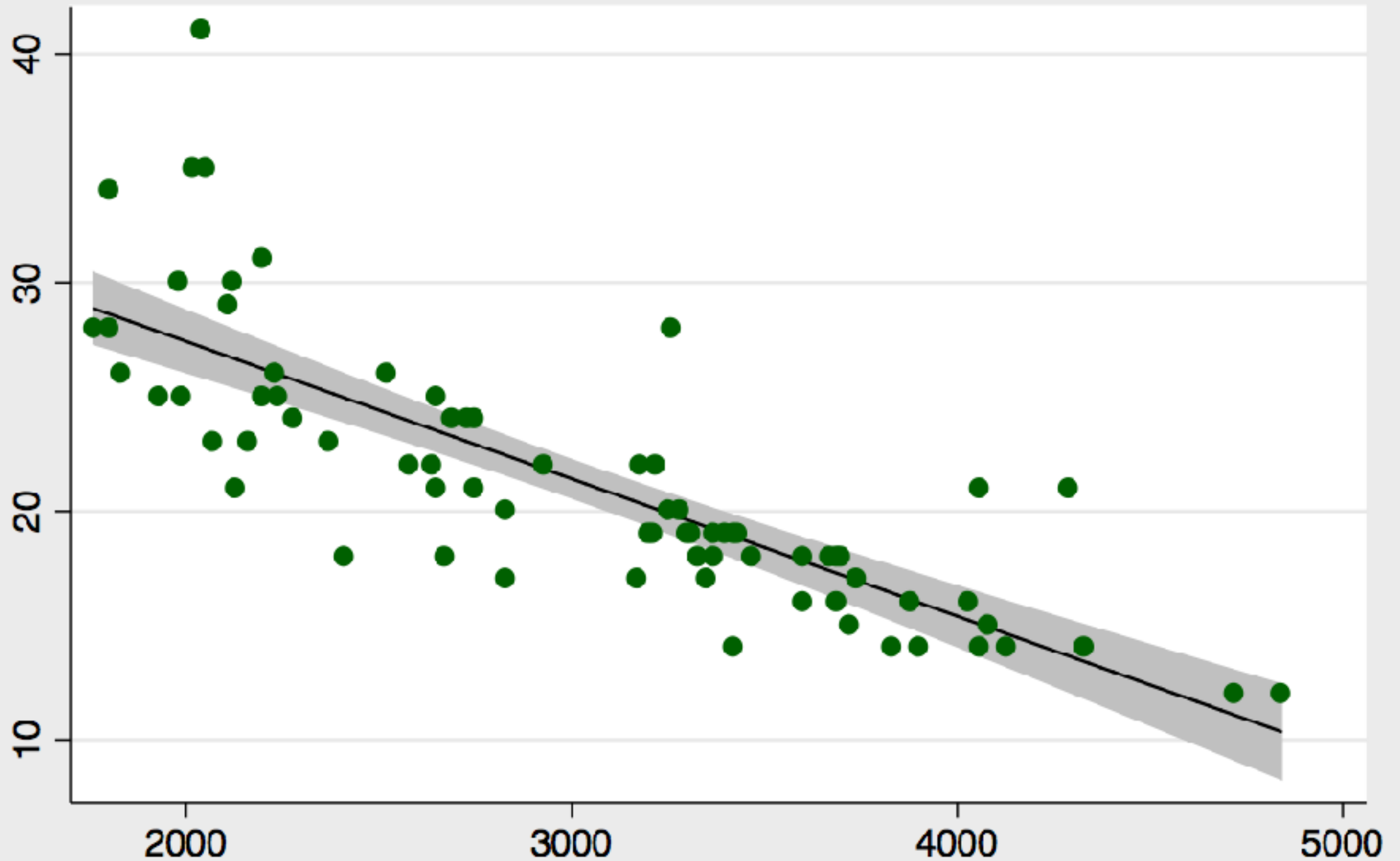
3. THE FAILINGS OF SIMPLE MODELS

CORRELATION



3. THE FAILINGS OF SIMPLE MODELS

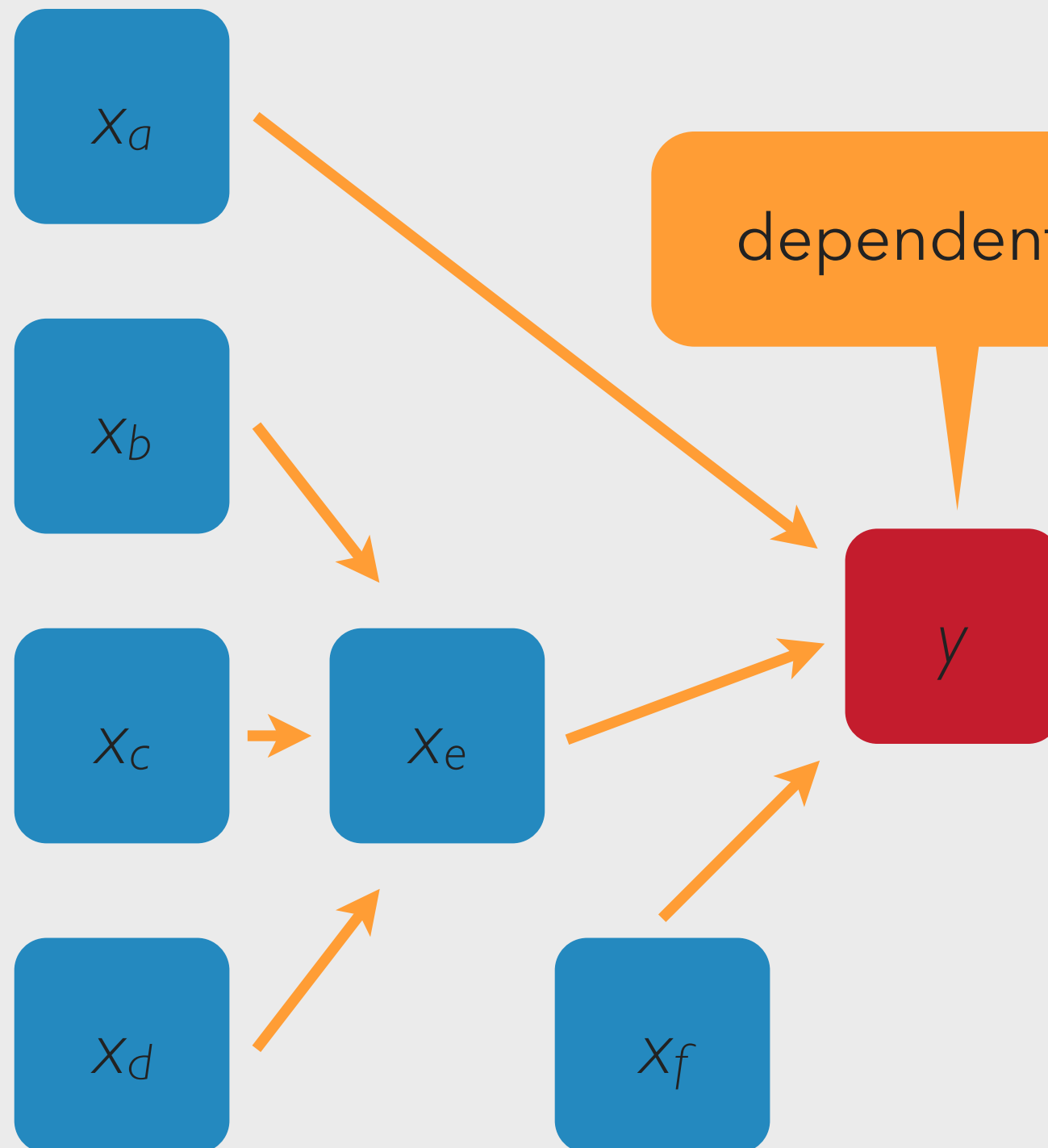
CORRELATION



3. THE FAILINGS OF SIMPLE MODELS

THE “REAL” WORLD

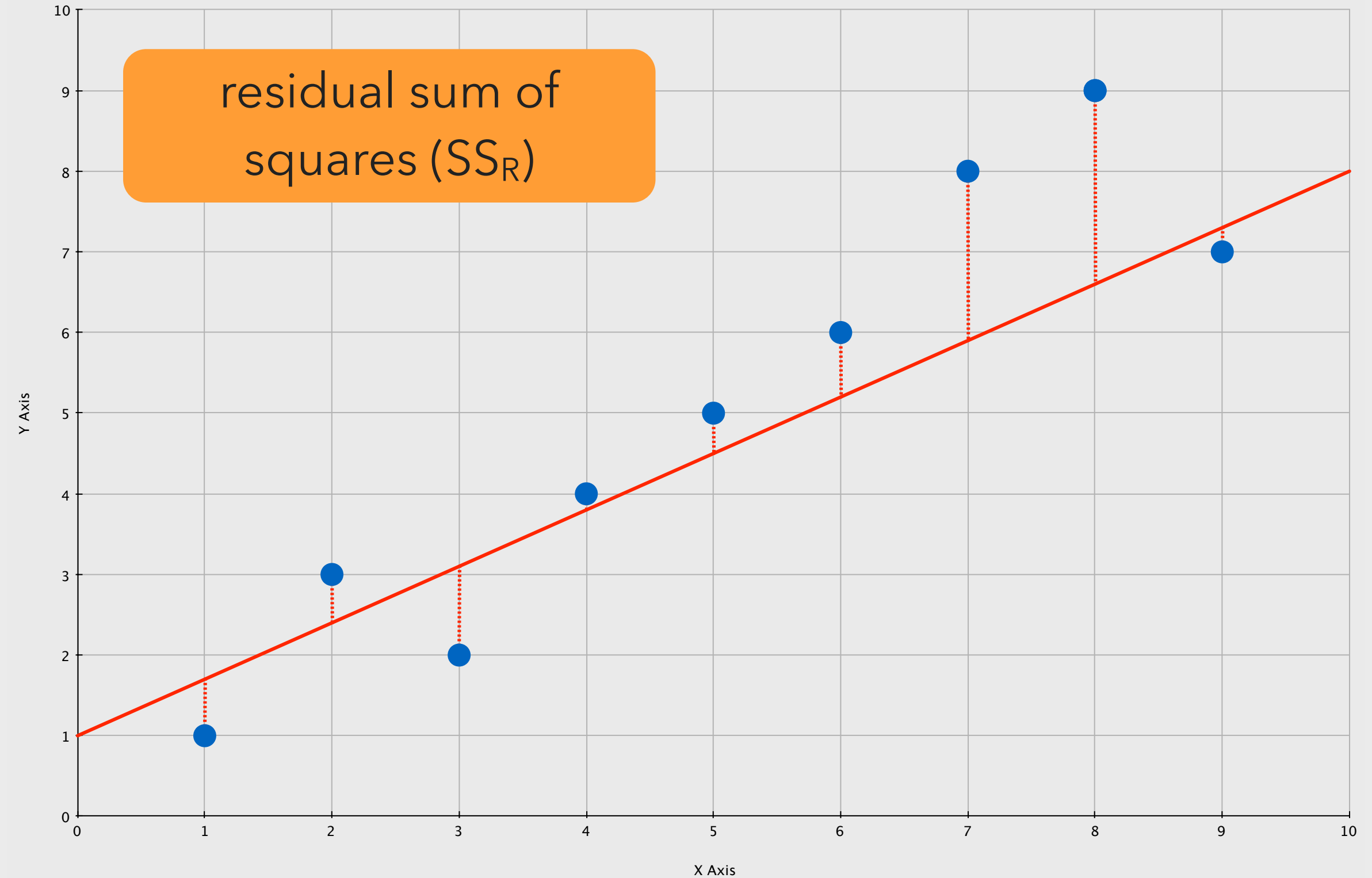
independent
variable



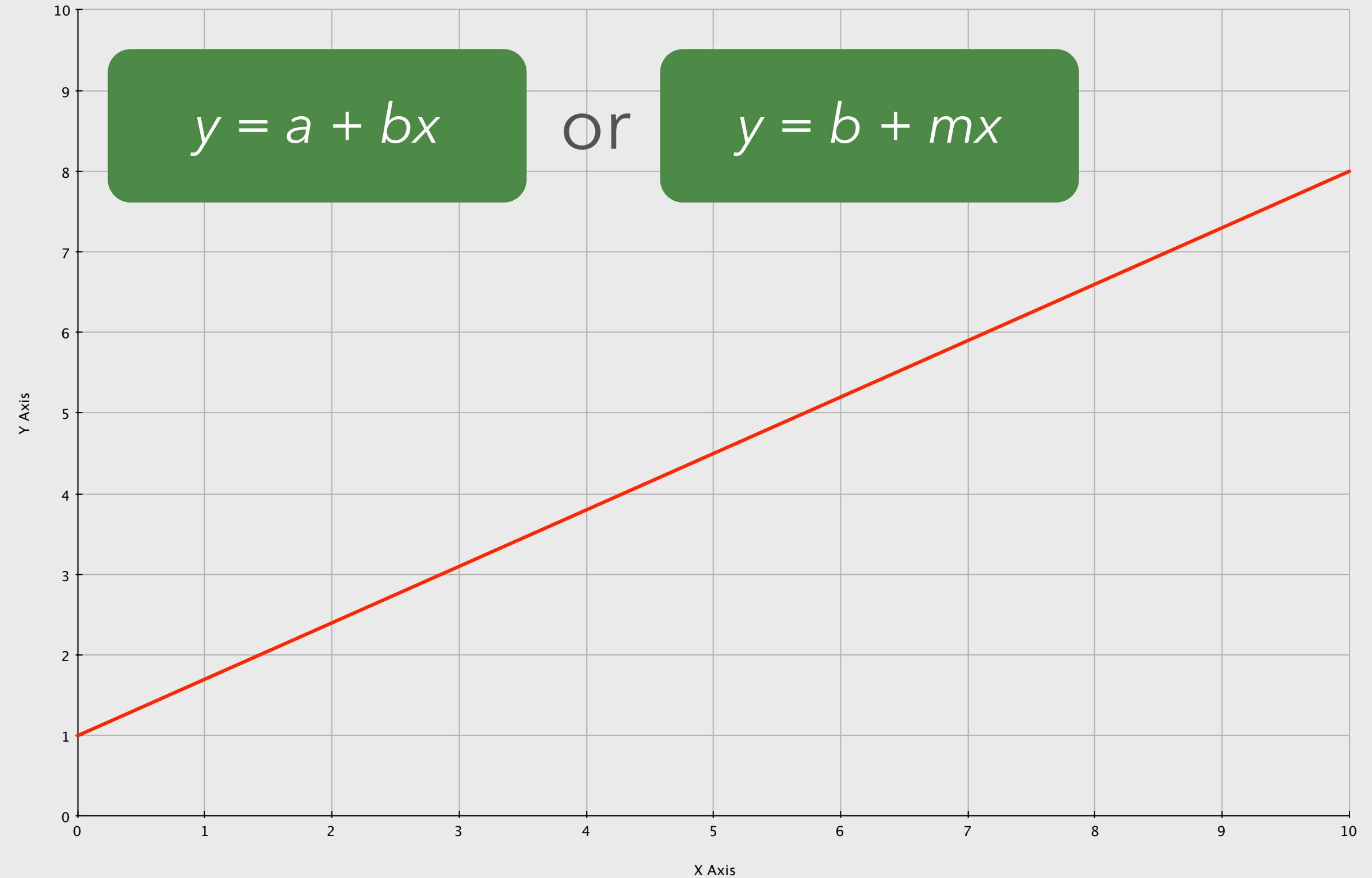
dependent variable

4 BIVARIATE REGRESSION THEORY

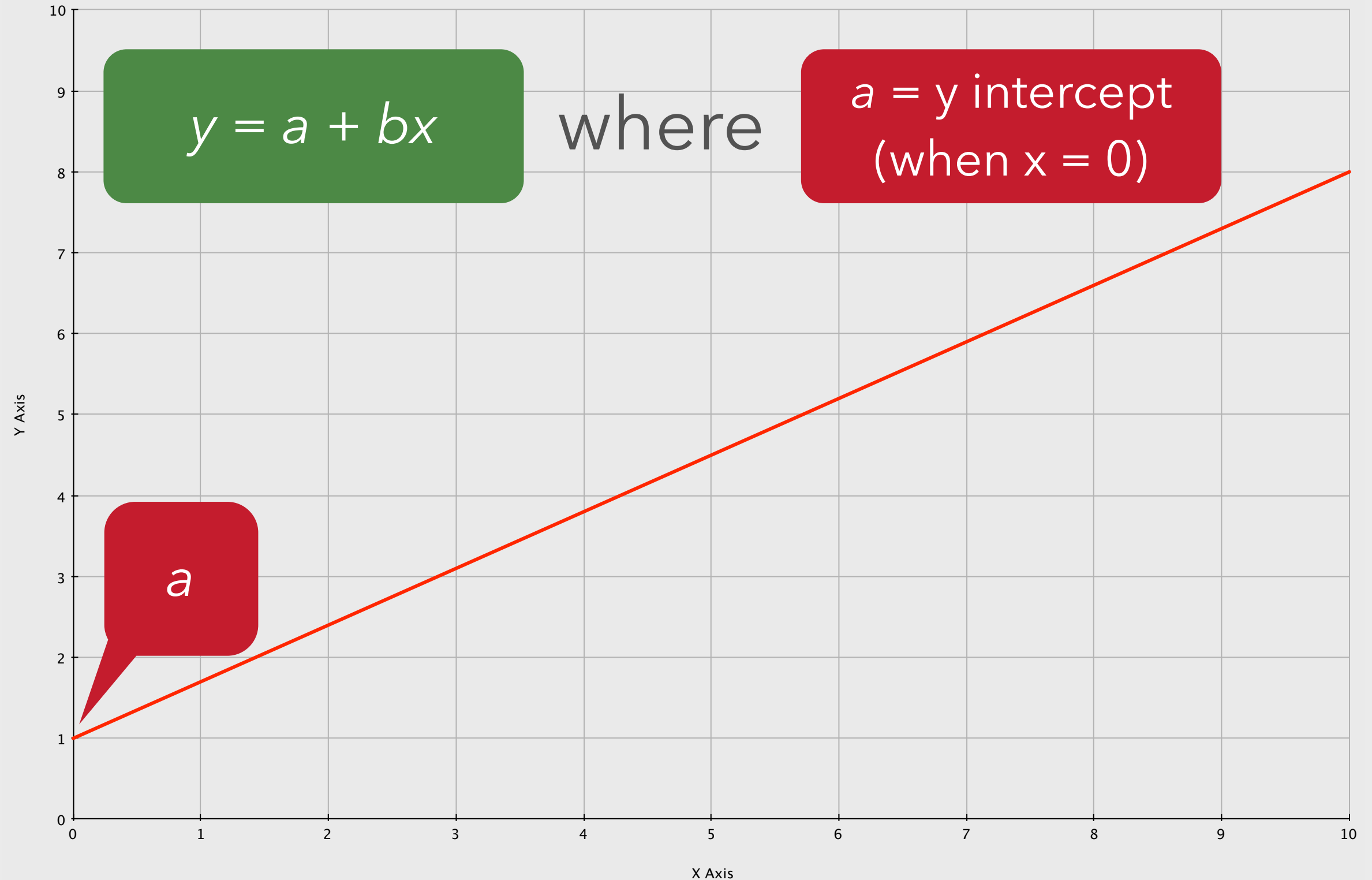
THE GOAL OF OLS REGRESSION



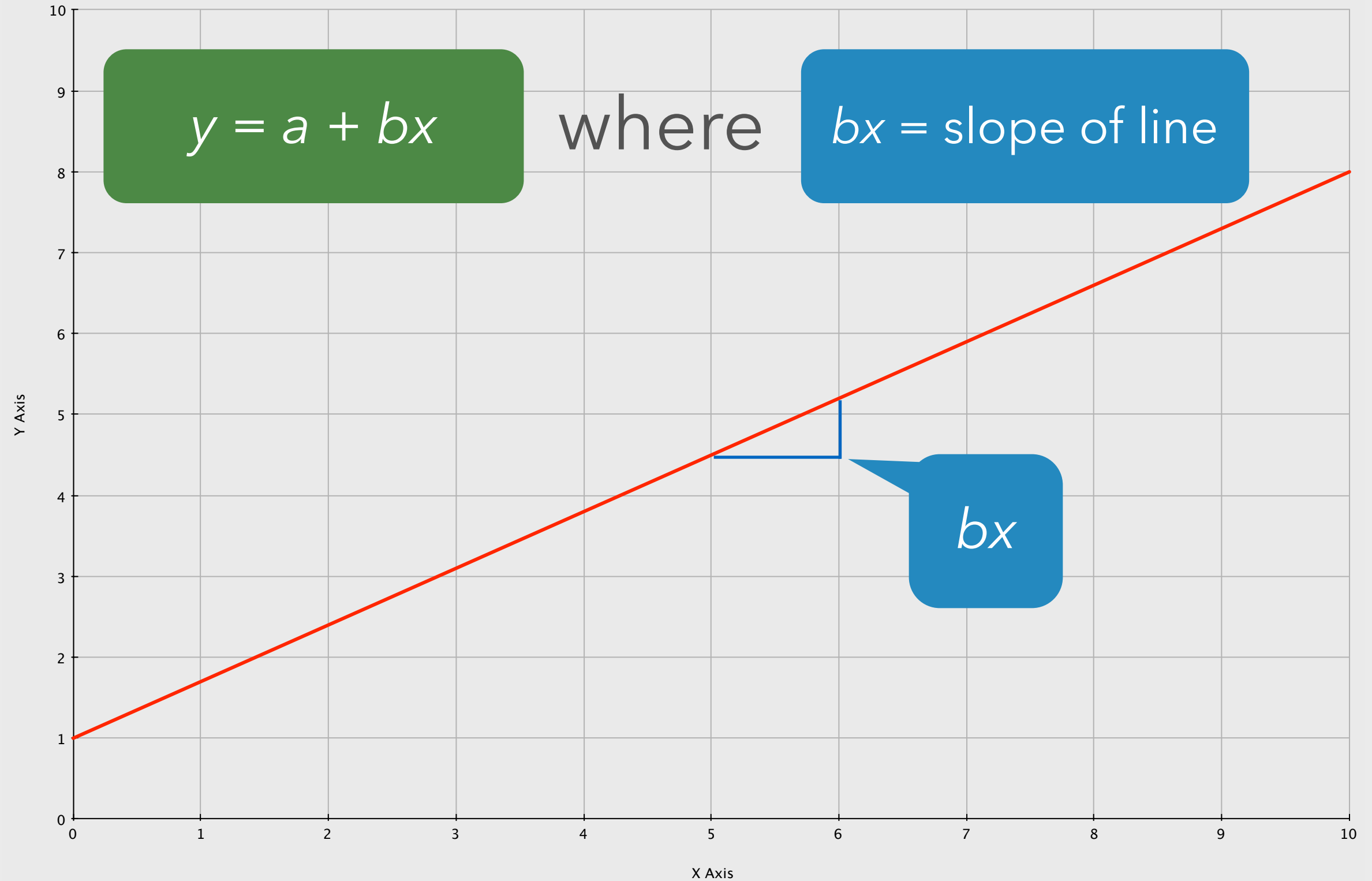
OLS REGRESSION BASICS



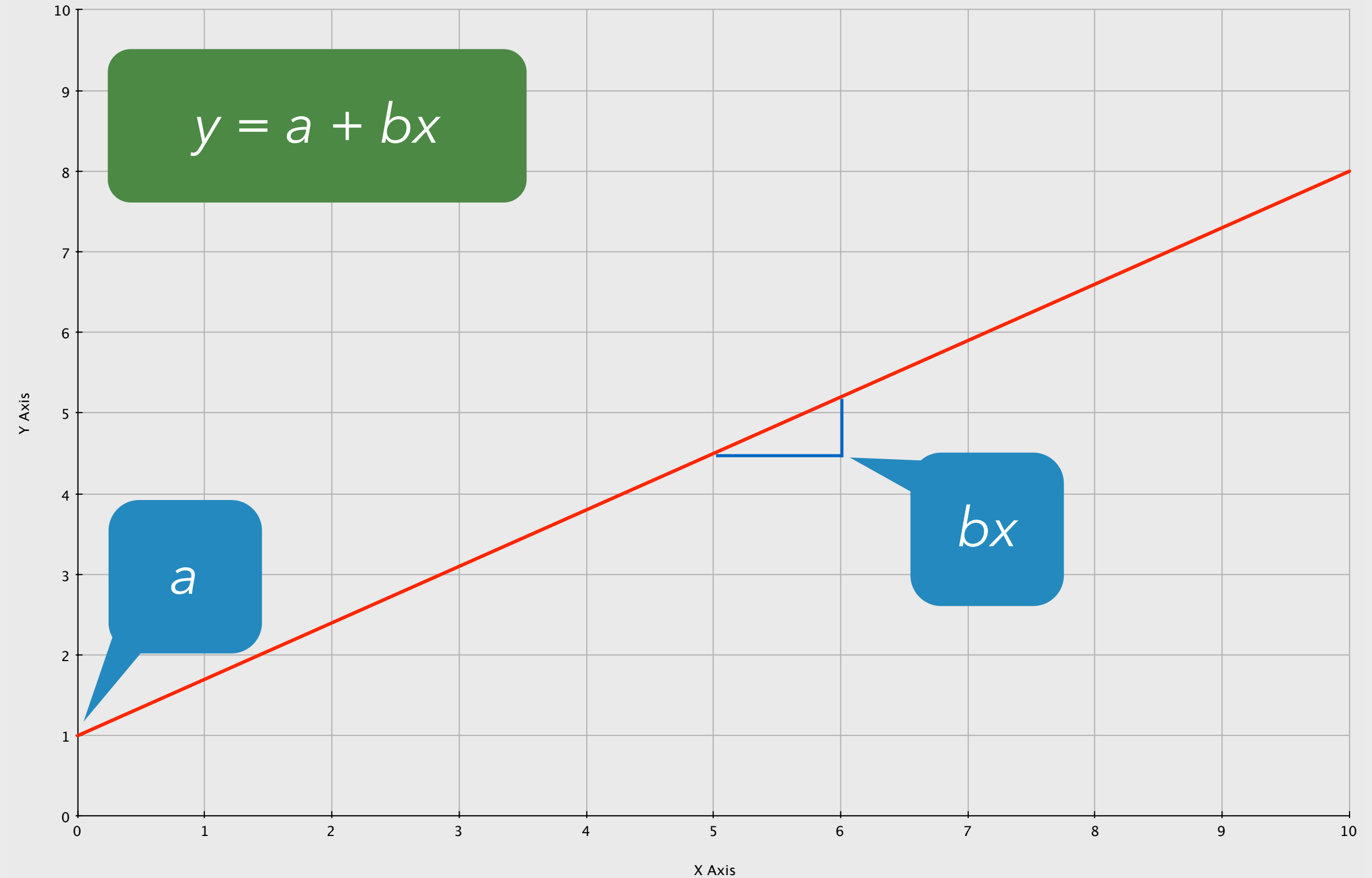
OLS REGRESSION BASICS



OLS REGRESSION BASICS



OLS REGRESSION BASICS



OLS REGRESSION BASICS

$$y = a + bx$$

y is the
dependent
variable (DV)
in regression
analysis

$$y = b_0 + b_i x_i + \varepsilon$$

b_0 is called
the 'constant'
rather than
the 'intercept'

OLS REGRESSION BASICS

subscript used
because the
slope of y is
dependent on
multiple
factors

$$y = a + bx$$

$$y = b_0 + b_i x_i + \varepsilon$$

ε is included
because we
are estimating
the line, there
may be
unexplained
variation in y

MODEL BUILDING

y = dependent variable
 b_0 = constant
 x_i = independent variable i
 b_i = beta value of IV i

DV = height
IV = gender
(where 0 = male
& 1 = female)

$$y = b_0 + b_i x_i + \varepsilon$$

$$y_{height} = b_0 + b_1 x_{female} + \varepsilon$$

MODEL BUILDING



$$y_{height} = b_0 + \beta_1 x_{male} + \varepsilon$$

constant is "male"

MODEL BUILDING

$$y = b_0 + b_i x_i + \varepsilon$$

y = dependent variable

b_0 = constant

x_i = independent variable i

b_i = beta value of IV i

DV = height

IV = school year

(where 0 = pre-school,

1 = elementary,

& 2 = middle school)

$$y_{height} = b_0 + b_1 x_{elementary} + b_2 x_{middle} + \varepsilon$$

MODEL BUILDING

$$y = b_0 + b_i x_i + \varepsilon$$

y = dependent variable

b_0 = constant

x_i = independent variable i

b_i = beta value of IV i

DV = occupational prestige


IV = race/ethnicity

(where 0 = white,

1 = African American,
& 2 = Hispanic)

$$y_{prestige} = b_0 + b_1 x_{black} + b_2 x_{hispanic} + \varepsilon$$

SOLVING REGRESSION EQUATIONS


$$\hat{y}_{prestige} = 40 + (-5)x_{black} + (-7)x_{hispanic} + \varepsilon$$

what is the mean occupational
prestige for African American respondents?

$$y_{prestige} = b_0 + b_1x_{black} + b_2x_{hispanic} + \varepsilon$$

SOLVING REGRESSION EQUATIONS

$$y_{prestige} = b_0 + b_1 x_{black} + b_2 x_{hispanic} + \varepsilon$$



$$\hat{y}_{prestige} = 40 + (-5)(1) + (-7)(0) + \varepsilon$$

what is the mean occupational
prestige for African American respondents?

SOLVING REGRESSION EQUATIONS

$$y_{prestige} = b_0 + b_1 x_{black} + b_2 x_{hispanic} + \varepsilon$$



$$\hat{y}_{prestige} = 40 + (-5) + 0 + \varepsilon$$

what is the mean occupational
prestige for African American respondents?

SOLVING REGRESSION EQUATIONS

$$y_{prestige} = b_0 + b_1 x_{black} + b_2 x_{hispanic} + \varepsilon$$



$$\hat{y}_{prestige} = 35$$

what is the mean occupational
prestige for African American respondents?

SOLVING REGRESSION EQUATIONS

$$y_{prestige} = b_0 + b_1 x_{black} + b_2 x_{hispanic} + \varepsilon$$



$$\hat{y}_{prestige} = 40 + (-5)(1) + (-7)(0) + \varepsilon$$

what is the mean occupational
prestige for Hispanic respondents?

SOLVING REGRESSION EQUATIONS

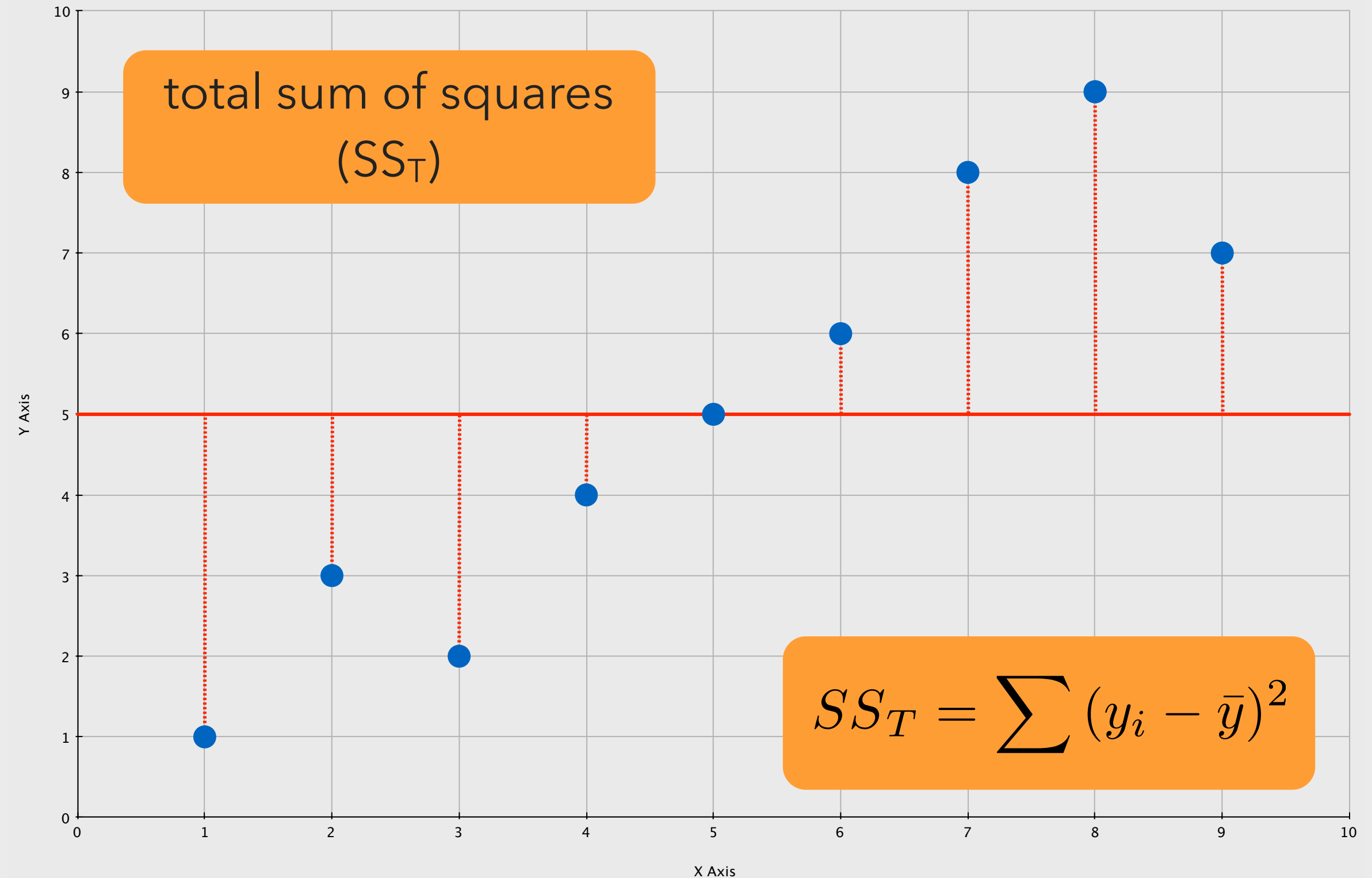
$$y_{prestige} = b_0 + b_1 x_{black} + b_2 x_{hispanic} + \varepsilon$$



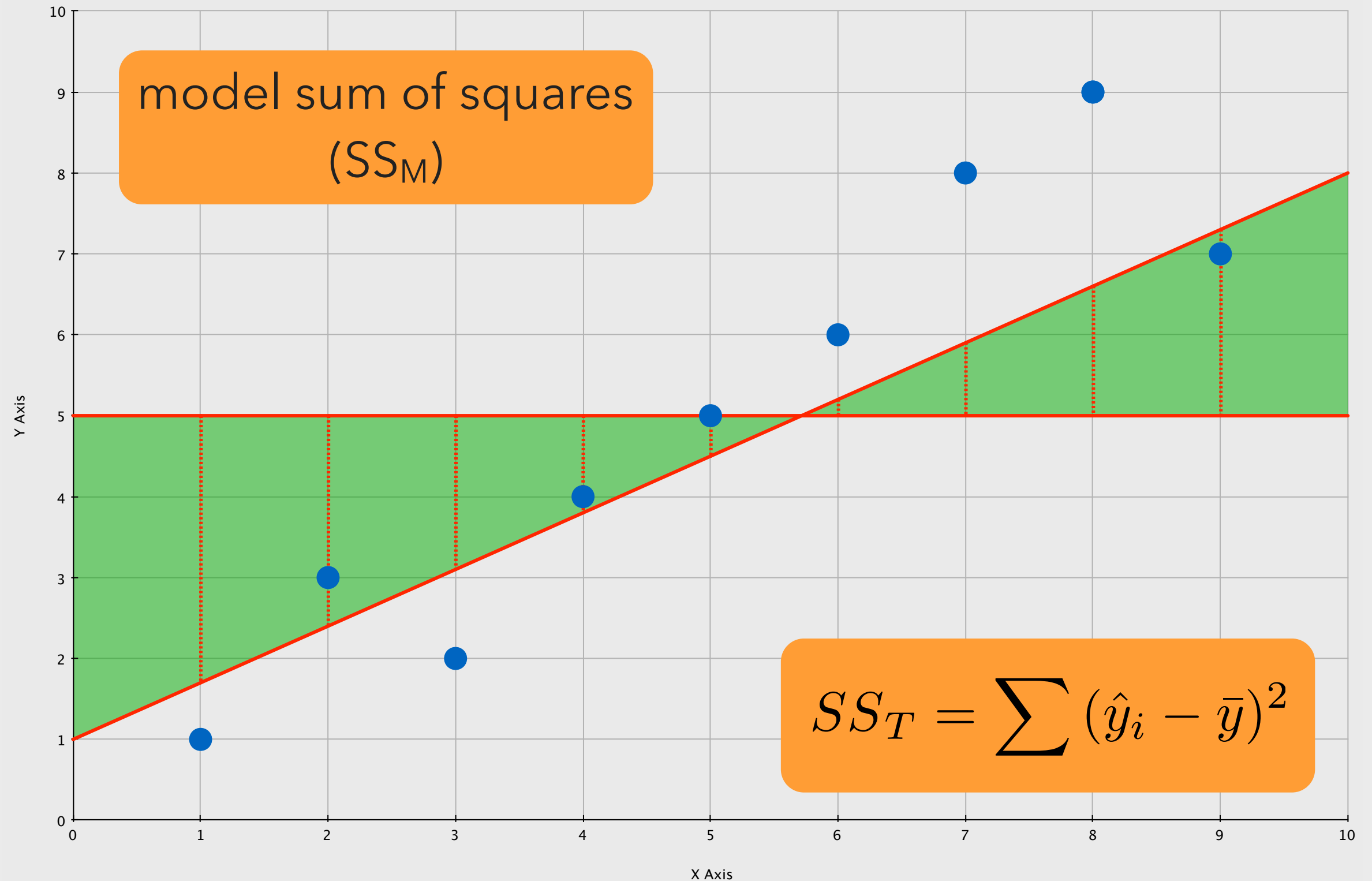
$$\hat{y}_{prestige} = 33$$

what is the mean occupational
prestige for Hispanic respondents?

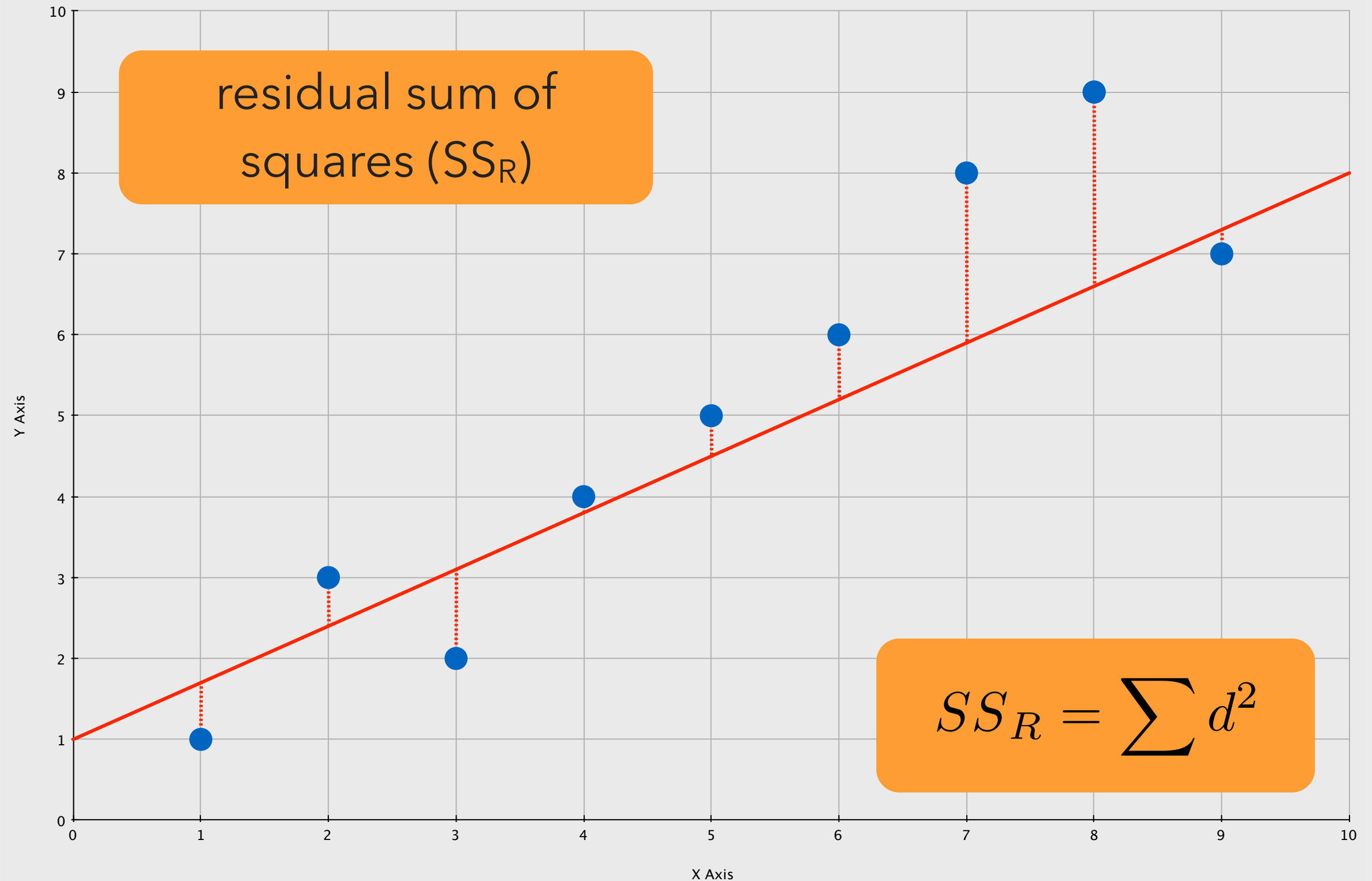
THE MECHANICS OF OLS REGRESSION



THE MECHANICS OF OLS REGRESSION



THE MECHANICS OF OLS REGRESSION



EXPLAINED VARIATION

the amount of
variation in y that our
model explains,
known as r^2

$$r^2 = \frac{SS_M}{SS_T}$$

ASSUMPTIONS

- ▶ DV must be continuous
- ▶ IV may be:
 - ▶ Dummy
 - ▶ Ordinal
 - ▶ Continuous
- ▶ IV's must have a variance > 0
- ▶ Relationship is linear
- ▶ DV should be normal
- ▶ No significant outliers

There are also assumptions about *multicollinearity* and *autocorrelation*, but we'll address these next week.

5 BIVARIATE REGRESSION IN STATA

ESTIMATING REGRESSION MODELS

```
▪ regress mpg weight, robust
```

Number of obs	=	74
F(1, 72)	=	105.83
Prob > F	=	0.0000
R-squared	=	0.6515
Root MSE	=	3.4389

		Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
mpg							
weight		-.0060087	.0005841	-10.29	0.000	-.007173	-.0048443
_cons		39.44028	1.98832	19.84	0.000	35.47664	43.40393

RELATIONSHIP WITH BIVARIATE TESTS

```
. pwcorr mpg weight
```

	mpg	weight
mpg	1.0000	
weight	-0.8072	1.0000

```
. display -0.8072^2  
-.65157184
```

```
. regress mpg weight, robust
```

Linear regression

Number of obs	=	74
F(1, 72)	=	105.83
Prob > F	=	0.0000
R-squared	=	0.6515
Root MSE	=	3.4389

	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
mpg						
weight	-.0060087	.0005841	-10.29	0.000	-.007173	-.0048443
_cons	39.44028	1.98832	19.84	0.000	35.47664	43.40393

```
. ttest mpg, by(foreign)
```

Two-sample t test with equal variances

Group	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
Domestic	52	19.82692	.657777	4.743297	18.50638	21.14747
Foreign	22	24.77273	1.40951	6.611187	21.84149	27.70396
combined	74	21.2973	.6725511	5.785503	19.9569	22.63769
diff		-4.945804	1.362162		-7.661225	-2.230384

diff = mean(Domestic) - mean(Foreign) t = -3.6308
Ho: diff = 0 degrees of freedom = 72

Ha: diff < 0
Pr(T < t) = 0.0003

Ha: diff != 0
Pr(|T| > |t|) = 0.0005

Ha: diff > 0
Pr(T > t) = 0.9997

```
. regress mpg foreign, robust
```

Linear regression

Number of obs = 74
F(1, 72) = 10.26
Prob > F = 0.0020
R-squared = 0.1548
Root MSE = 5.3558

mpg	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
foreign	4.945804	1.544418	3.20	0.002	1.867062	8.024546
_cons	19.82692	.660407	30.02	0.000	18.51043	21.14342

5. BIVARIATE REGRESSION IN STATA

INTERPRETATIONS: MODEL FIT

```
. regress mpg weight
```

Source	SS	df	MS	Number of obs	=	74
				F(1, 72)	=	134.62
Model	1591.9902	1	1591.9902	Prob > F	=	0.0000
Residual	851.469256	72	11.8259619	R-squared	=	0.6515
				Adj R-squared	=	0.6467
Total	2443.45946	73	33.4720474	Root MSE	=	3.4389

mpg	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
weight	-.0060087	.0005179	-11.60	0.000	-.0070411	-.0049763
_cons	39.44028	1.614003	24.44	0.000	36.22283	42.65774

The results of the ANOVA test ($f = 134.62$, $p < .001$) indicate that the model is a good fit for the data. Overall, the model explains 64.67% of the variation in miles per gallon.

5. BIVARIATE REGRESSION IN STATA

INTERPRETATIONS: MODEL FIT

```
. regress mpg weight, robust
```

```
Linear regression               Number of obs   =           74
                               F(1, 72)         =        105.83
                               Prob > F          =         0.0000
                               R-squared         =         0.6515
                               Root MSE      =         3.4389
```

		Robust					
mpg	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]		
weight	-.0060087	.0005841	-10.29	0.000	-.007173	-.0048443	
_cons	39.44028	1.98832	19.84	0.000	35.47664	43.40393	

The results of the ANOVA test ($f = 105.83$, $p < .001$) indicate that the model is a good fit for the data. Overall, the model explains 64.67% of the variation in miles per gallon.

5. BIVARIATE REGRESSION IN STATA

INTERPRETATIONS: CONTINUOUS PREDICTORS

```
. regress mpg weight, robust
```

```
Linear regression               Number of obs   =           74
                                F(1, 72)         =        105.83
                                Prob > F          =         0.0000
                                R-squared          =         0.6515
                                Root MSE       =         3.4389
```

		Robust				
mpg	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
weight	-.0060087	.0005841	-10.29	0.000	-.007173	-.0048443
_cons	39.44028	1.98832	19.84	0.000	35.47664	43.40393

The model finds that a unit increase in weight is associated with a -0.006 ($p < .001$) miles per gallon decrease. Heavier cars are, on average, less fuel efficient than lighter cars.

5. BIVARIATE REGRESSION IN STATA

INTERPRETATIONS: BINARY PREDICTORS

```
. regress mpg foreign, robust
```

```
Linear regression              Number of obs      =           74
                              F(1, 72)            =          10.26
                              Prob > F             =          0.0020
                              R-squared            =          0.1548
                              Root MSE         =          5.3558
```

		Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
mpg							
foreign		4.945804	1.544418	3.20	0.002	1.867062	8.024546
_cons		19.82692	.660407	30.02	0.000	18.51043	21.14342

The model finds that foreign cars are associated with a 4.946 ($p < .001$) miles per gallon increase. Foreign cars are, on average, more fuel efficient than domestic cars..

DOCUMENT DETAILS

Document produced by [Christopher Prener, Ph.D](#) for the Saint Louis University course SOC 5050: QUANTITATIVE ANALYSIS - APPLIED INFERENTIAL STATISTICS. See the [course wiki](#) and the repository [README.md](#) file for additional details.



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