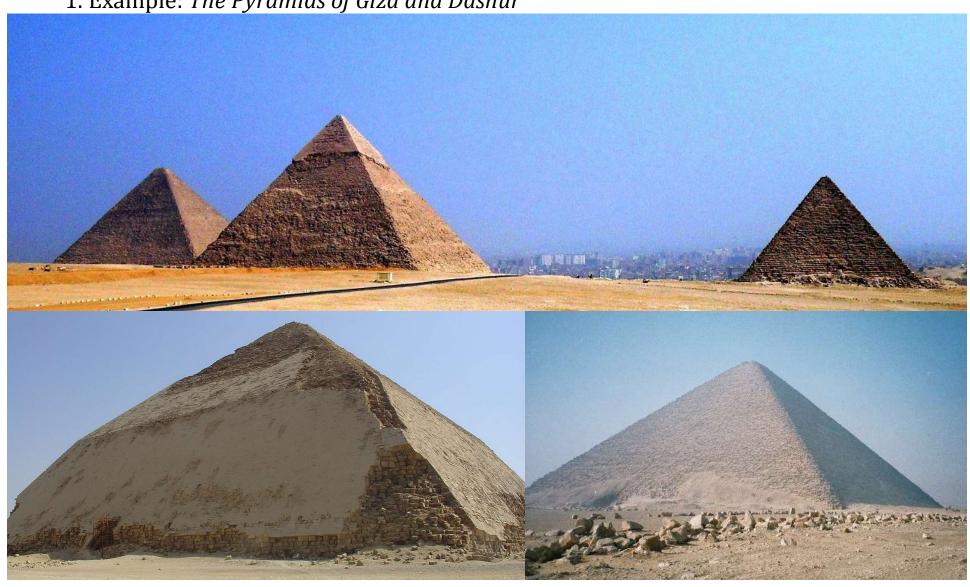
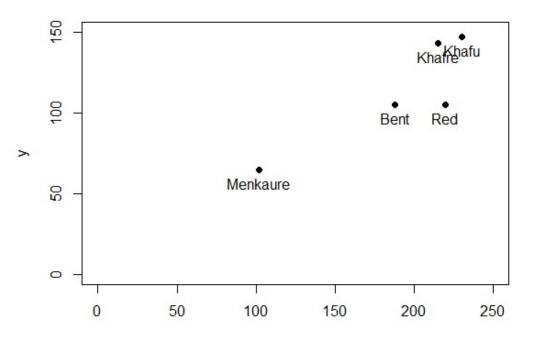
1. Example: The Pyramids of Giza and Dashur



1. Example: The Pyramids of Giza and Dashur

```
width <- c(230,220,215,188,102)
height <- c(147,105,143,105,65)
pyramids <- c("Khafu", "Red", "Khafre", "Bent", "Menkaure")
horiz <- c(0,250)
vert <- c(0,150)
plot(width, height, pch=16, xlim = horiz, ylim = vert)
text(width, height, labels=pyramids, pos=1, xpd=TRUE)
```



1. Example: The Pyramids of Giza and Dashur

i	X_i	y_i	$(x_i - \mu_x)$	$(x_i - \mu_x)^2$	$(y_i - \mu_y)$	$(y_i - \mu_y)^2$	$(x_i - \overline{x})(y_i - \overline{y})$
Khufu	230	147	39	1,521	34	1,156	1,326
Khafre	215	143	24	576	30	900	720
Menkaure	102	65	-89	7,921	-48	2,304	4,272
"Bent"	188	105	-3	9	-8	64	24
"Red"	220	105	29	841	-8	64	-232
Σ	955	565	$SST_{x} =$	10,868	$SST_y =$	4,488	6,110
Σ/n	$\mu_{x} = 191$	$\mu_{y} = 113$	$\sigma_{x}^{2} =$	2,173.6	$\sigma_y^2 =$	897.6	$\sigma_{xy}^2 = 1,222$

data = cbind.data.frame(pyramids,width,height)
covxy = cov(data[2:3], use = "pairwise.complete.obs"); covxy
bhat = covxy[1,2]/covxy[1,1]; bhat
ahat = mean(data\$height) - bhat*mean(data\$width); ahat

Slope:
$$\widehat{\beta_1} = \frac{cov(x,y)}{var(x)} = \frac{1222}{2173.6} = 0.562$$
 Intercept: $\widehat{\beta_0} = \overline{y} - \widehat{\beta_1}\overline{x} = 113 - (0.562) \cdot 191 = 5.62$

1. Example: The Pyramids of Giza and Dashur

R script: Lecture 1 Pyramids.R

Data frame: data

Explained variable: data\$height

Explanatory variable: data\$width

myModel = lm(data\$height ~ data\$width)
summary(myModel)

cor(myModel\$residuals, width) # check assumption of zero correlation...

sum(myModel\$residuals) # check assumption of zero mean residual

Question: "What determines the height of a pyramid?"

Answer: "The width of the base."

POLS 6481. Research Design and Quantitative Methods II

Lecture 1. Course Introduction

Reading: Daniel Pink, A Whole New Mind

2. Left Brain, Right Brain

Stimulus	Right-Hemisphere Injury	Left-Hemisphere Injury

POLS 6481. Research Design and Quantitative Methods II Lecture 1. Course Introduction Reading: Daniel Pink, *A Whole New Mind*

2. Left Brain, Right Brain

Left Hemisphere

Right vision field

Right motor skills

Details

Numbers

Verbal

Grammar, Vocab.

Text

Literal

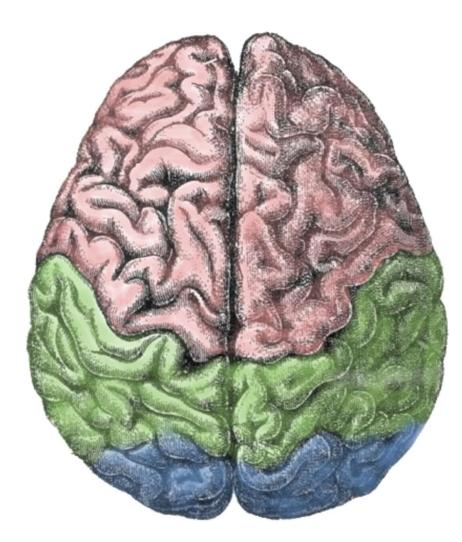
Sequential/Serial

Logical

Analytical

Cautious

Systematic



Right Hemisphere

Left vision field

Left motor skills

Holistic

Patterns

Nonverbal

Facial Expressions

Context (Emotions)

Figurative

Simultaneous

Intuitive

Imaginative

Impulsive

Empathetic

POLS 6481. Research Design and Quantitative Methods II Lecture 1. Course Introduction Reading: Daniel Pink, *A Whole New Mind*

2. Left Brain, Right Brain

However, there are some important caveats:

The idea that people use entirely (or predominantly) the right or left brain is a myth; you enlist both hemispheres all the time (as long as your corpus collosum is intact).

Names are stored in the left hemisphere while faces are stored in the right hemisphere.

This is an active field of neuroscience; almost any claim is controversial. For example, Simon Baron-Cohen (*The Essential Difference*) claims that there are gender differences along the empathizing-systemizing dimension — and that autism occurs at one extreme.

Nevertheless, this is a useful metaphor:

Daniel Pink (*A Whole New Mind*) argues that the Information Age prioritized left-brain aptitudes, but these are becoming less important in what he calls the Conceptual Age.

Iain McGilchrist (*The Master and His Emissary*) also makes the provocative claim that the evolution of Western culture increasingly has been dominated by the left-brain.

POLS 6481. Research Design and Quantitative Methods II Lecture 1. Course Introduction Reading: Daniel Pink, *A Whole New Mind*

2. Left Brain, Right Brain

My "challenge" is to teach left-brain material (equations, calculations, computations) without sacrificing the right-brain.

I will frequently emphasize the following:

Design (Appearance)

Story

Meaning

Connections between concepts

The 'big picture'

Readings: Ray Fair, Predicting Presidential Elections and Other Things, ch. 2

- 3. Regression in Seven (or More?) Steps
- 1. Begin with a theory
- 2. Collect data
- 3. Fit the data to a line
- 6. Interpret the results
- 7. Predict *y* given a set value of *x*
- 4. Test a hypothesis about the slope of the line
- 5. Think about pitfalls
- 8. Add more variables