Modeling Basics / Intro to R

READING: 0.1, 0.2

EXERCISES: ALL CHAPTER 0

ASSIGNED: HW 1

PRODUCER: DR. MARIO

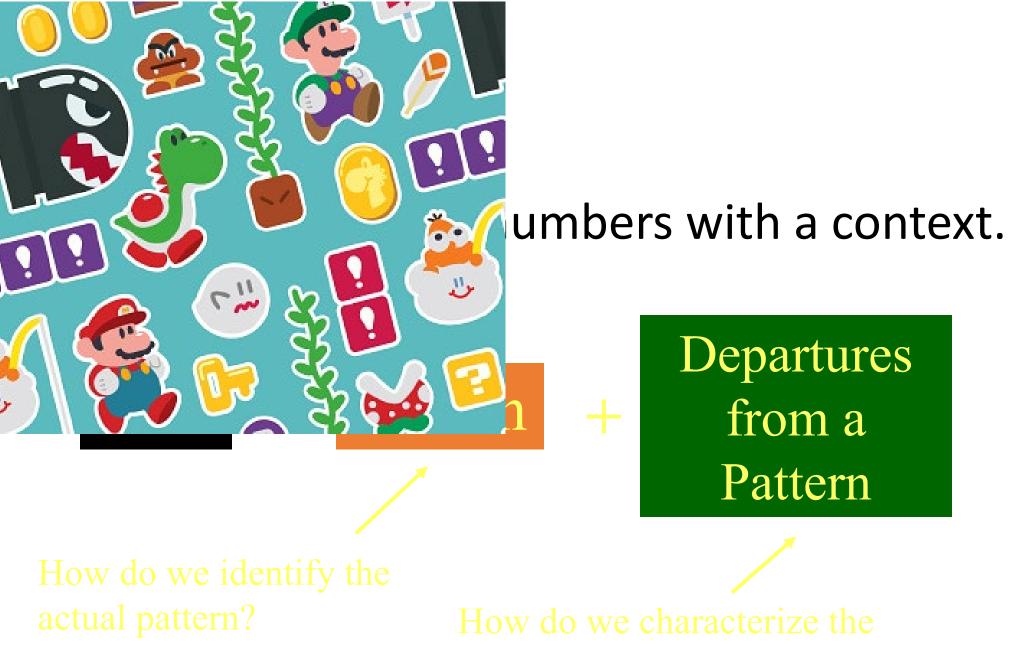


Course Website / Syllabus

- Access Course Website Through Canvas
- Cover Syllabus
 - Office Hours
 - Grading and Curving
 - Attendance: UNC Check-In App
 - Homework
 - Quizzes
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Statistical Modeling

Find a model for a relationship between a response variable (Y) and one (or more) predictor/explanatory variables $(X_1, X_2, ..., X_k)$.

Purposes for Statistical Modeling

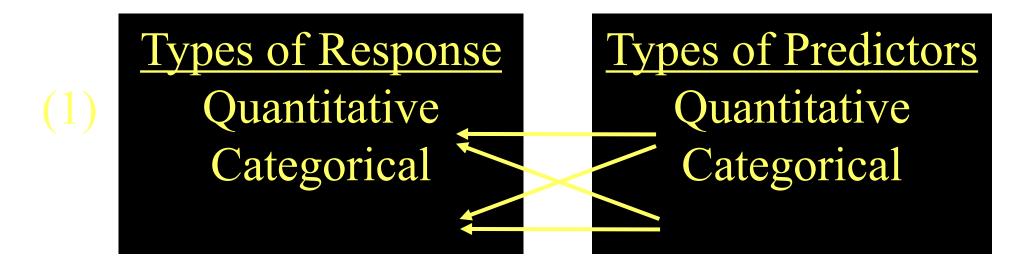
- 1. Making predictions
- 2. Understanding relationships
- 3. Assessing differences

Types of Variables

```
Quantitative: expressible as
numbers for which arithmetic
makes sense
Categorical: divides sample points
into groups
```

Binary = categorical with just two groups

Two Main Themes of STOR 455



(2) Allow for models with *multiple* predictors.

Building a Statistical Model: Four Step Process

- 1. CHOOSE Pick a form for the model
- 2. FIT Estimate any parameters
- 3. ASSESS Is the model adequate?
- Could it be simpler? Are conditions met?
- 4. USE Answer the question of interest

General form of a model:

$$Y = f(X) + \varepsilon$$
Random Error

"Expected" Y for some combination of predictors

Data

Model

Error

Example: Lego Prices

Question:

How can we predict the price of a Lego set?

Predictor variables: Start with none.

Example: Constant Model

$$Y = c + \varepsilon$$
 where c is an (unknown) constant

Terminology:

The constant c is a parameter of this model. We use data to provide a sample estimate of c.

How should we estimate *c* from data?

Predicted Value for Response

Get an for Y using the predictors and the model with estimated parameters.

Notation: The predicted y is denoted \hat{y}

For the constant model: $\hat{y} = \hat{c}$

Examples: (sample mean)

(sample median)

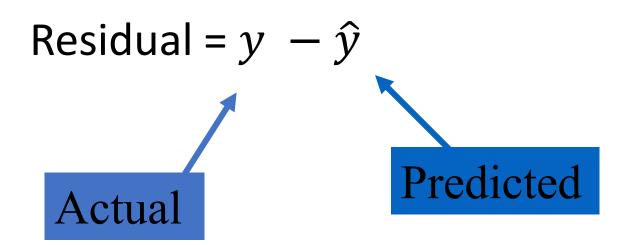
Questions

(1) Which estimator (mean or median) is better?

(2) Is either model any good?

Residuals

Using the predicted value for each sample case the residual is



Criteria to Minimize Residuals?

Sum of residuals:

Sum of absolute deviations:

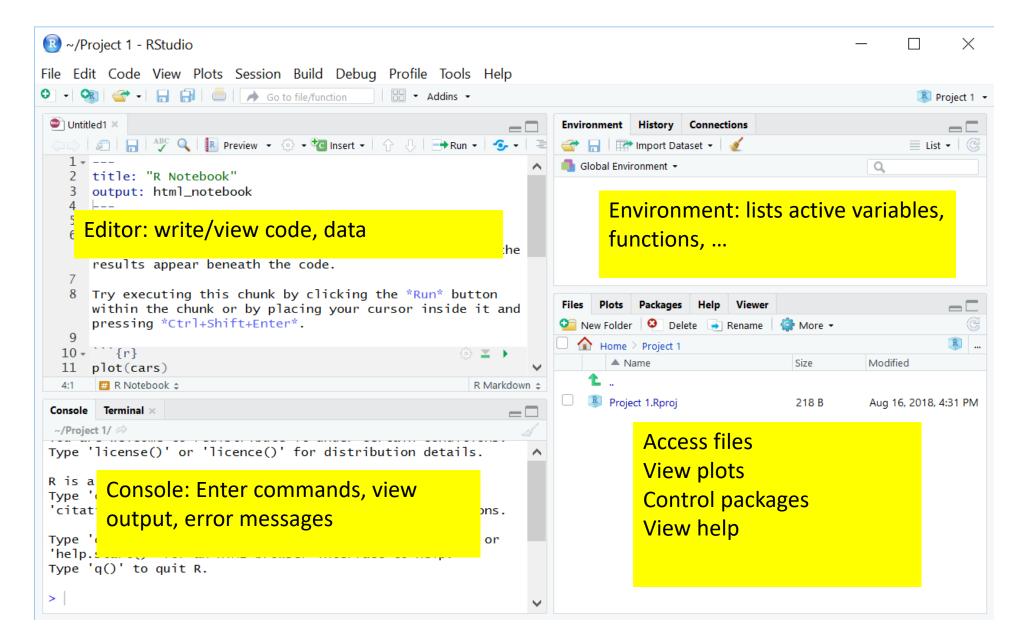
Sum of squared errors:

Technology

We need software to automate computations...

R – a free, widely used, open source statistics package Rstudio – an interface for R

RStudio



- Load the *Lego* data into R
- Summarize the Amazon_Price variable
 - Numerical: mean and median
 - Graphical: histogram, boxplot
- Compute and evaluate residuals

Load the *lego* data into R

```
```{r}
loads a package needed to use the read csv() function
install package before first using it for the first time
library(readr)
loads the lego dataframe into the environment from GitHub
lego <- read csv("https://raw.githubusercontent.com/JA-McLean/STOR455/master/data/lego.csv")</pre>
Alternative way to load dataframe (remove # to use)
lego.csv must be saved in the same folder as this notebook!
#lego <- read csv("lego.csv")</pre>
Shows the variables and first 6 cases (by default)
head(lego)
```

Summarize the *Amazon\_Price* variable - Numerical: mean and median

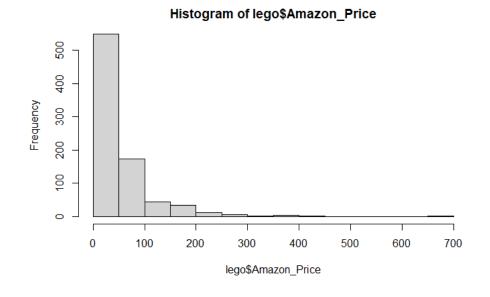
```
dataframe$variable_name
mean(lego$Amazon_Price, na.rm = TRUE)
median(lego$Amazon_Price, na.rm = TRUE)

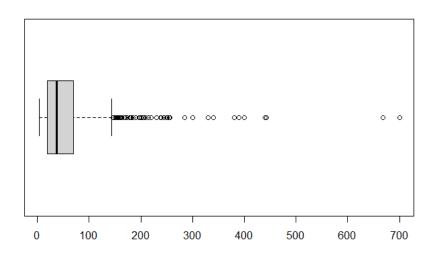
[1] 57.8232
[1] 37.325
```

Summarize the Amazon\_Price variable - Graphical: histogram, boxplot

hist(lego\$Amazon\_Price)

boxplot(lego\$Amazon Price, horizontal = TRUE)





#### Compute and evaluate residuals

```
removes NA Amazon Prices
lego rm AP na = subset(lego, is.na(Amazon Price) == FALSE)
Assignment operators in R: = vs. <-
xbar = mean(lego rm AP na$Amazon Price)
m = median(lego rm AP na$Amazon Price)
residxbar = lego rm AP na$Amazon Price - xbar
residm = lego rm AP na$Amazon Price - m
sum(residxbar^2)
sum(residm^2)
```

### Can we use a predictor to improve the model?

See Chapter 1

Two-sample t-test for a difference in means (comparing only two themes)

#### Homework 0

```
Homework 0.Rmd ×

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 *C - | ↑ ↓ | → Run - | • - | = | A
 2 title: 'STOR 455 Homework #0'
 3 output:
 pdf_document: default
 html_notebook: default
 Welcome to STOR 455! This assignment is to show that you:
 10
 have R/RStudio installed and ready to use for the course.
 11
 12 ×
 are able to knit files to PDF.
 13
 are able to submit assignments through Gradescope.
 14 ×
 15
 16
 17 Fill in your name and pronouns in the chunk below. Then knit the file to PDF and
 submit the PDF to Gradescope. If you see LaTeX errors when knitting directly to
 PDF, try knitting to a word file and then converting that word file to a PDF.
 18
 19 - ```{r}
 ⊹ 🛂 🕨
 21 Name = 'My preferred name is: *delete this text and enter your prefered name*.'
 Pronouns = 'My pronouns are: *delete this text and enter your pronouns*. '
 24
 25 -
 26
 27 • ```{r}
 ∰ ≚ ▶
 28 Name
 29 Pronouns
 30 - ` ` `
 31
 32
```