

001 - Introduction to Inferential Statistics

EPIB 607 - FALL 2020

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slides compiled on September 2, 2020



Objectives for this course

1. Visualize/Analyze/Interpret data using statistical methods with a computer

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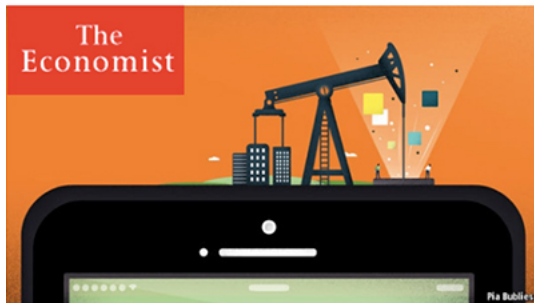
1. Visualize/Analyze/Interpret data using statistical methods with a computer
2. Gather data into analysis ready format
3. Learn regression
4. Understand the statistical results in a scientific paper
5. Learn the tools for creating reproducible analyses

Data is the new oil¹

Fuel of the future

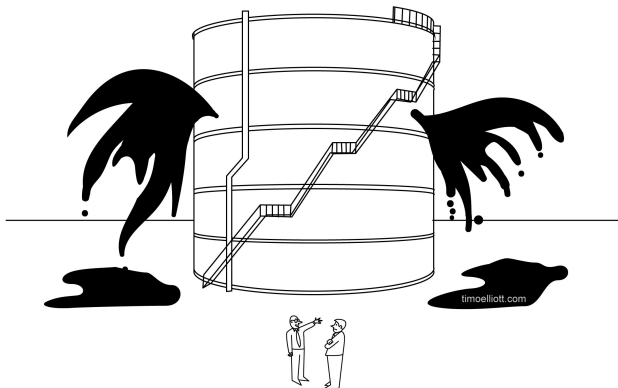
Data is giving rise to a new economy

How is it shaping up?



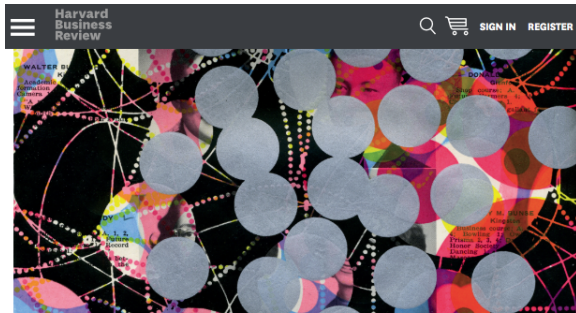
¹<https://www.economist.com/briefing/2017/05/06/data-is-giving-rise-to-a-new-economy>

Danger²



“Data is the new oil? Absolutely—toxic if mishandled!...”

²<https://timoelliott.com/blog/2018/03/data-is-the-new-oil-yes-toxic-if-mishandled.html>



ARTWORK: TAMAR COHEN, ANDREW J BUBOLTZ, 2011, SILK SCREEN ON A PAGE FROM A HIGH SCHOOL YEARBOOK, 8.5" X 12"

DATA

Data Scientist: The Sexiest Job of the 21st Century

by **Thomas H. Davenport** and **D.J. Patil**

FROM THE OCTOBER 2012 ISSUE

³<https://hbr.org/2012/10/data-scientist-the-sexiest-job-of-the-21st-century>

Why R ?

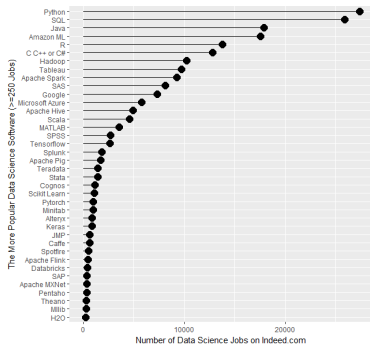


Figure: Data as of May 2019

<http://r4stats.com/articles/popularity/>

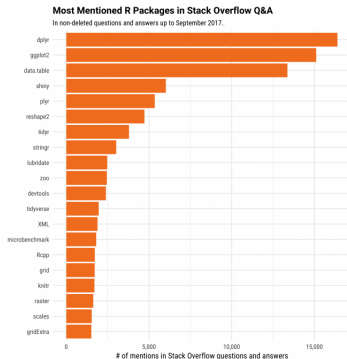


Figure: Popular R packages

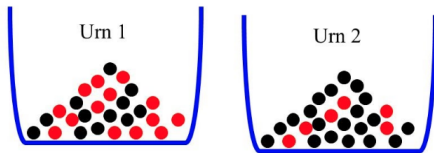
<https://stackoverflow.blog/2017/10/10/impressive-growth-r/>

First day in a statistics course

Example:

We have two urns. Urn 1 contains 14 red balls and 12 black balls. Urn 2 contains 6 red balls and 20 black balls.

An Urn is selected at random and a ball is selected from that urn.



If the ball turns out to be red what is the probability that it came from the first urn?

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
1	5.1	3.5	1.4	0.2	setosa
2	4.9	3.0	1.4	0.2	setosa
3	4.7	3.2	1.3	0.2	setosa
4	4.6	3.1	1.5	0.2	setosa
5	5.0	3.6	1.4	0.2	setosa
6	5.4	3.9	1.7	0.4	setosa
7	4.6	3.4	1.4	0.3	setosa
8	5.0	3.4	1.5	0.2	setosa
9	4.4	2.9	1.4	0.2	setosa
10	4.9	3.1	1.5	0.1	setosa
11	5.4	3.7	1.5	0.2	setosa
12	4.8	3.4	1.6	0.2	setosa
13	4.8	3.0	1.4	0.1	setosa
14	4.3	3.0	1.1	0.1	setosa
15	5.8	4.0	1.2	0.2	setosa

Second day in a statistics course



z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	.5000	.5040	.5080	.5120	.5160	.5199	.5239	.5279	.5319	.5359
0.1	.5398	.5438	.5478	.5517	.5557	.5596	.5636	.5675	.5714	.5753
0.2	.5793	.5832	.5871	.5910	.5948	.5987	.6026	.6064	.6103	.6141
0.3	.6179	.6217	.6255	.6293	.6331	.6368	.6406	.6443	.6480	.6517
0.4	.6554	.6591	.6628	.6664	.6700	.6736	.6772	.6808	.6844	.6879
0.5	.6915	.6950	.6985	.7019	.7054	.7088	.7123	.7157	.7190	.7224
0.6	.7257	.7291	.7324	.7357	.7389	.7422	.7454	.7486	.7517	.7549
0.7	.7580	.7611	.7642	.7673	.7704	.7734	.7764	.7794	.7823	.7852
0.8	.7881	.7910	.7939	.7967	.7995	.8023	.8051	.8078	.8106	.8133
0.9	.8159	.8186	.8212	.8238	.8264	.8289	.8315	.8340	.8365	.8389
1.0	.8413	.8438	.8461	.8485	.8508	.8531	.8554	.8577	.8599	.8621
1.1	.8643	.8665	.8686	.8708	.8729	.8749	.8770	.8790	.8810	.8830
1.2	.8849	.8869	.8888	.8907	.8925	.8944	.8962	.8980	.8997	.9015
1.3	.9032	.9049	.9066	.9082	.9099	.9115	.9131	.9147	.9162	.9177
1.4	.9192	.9207	.9222	.9236	.9251	.9265	.9279	.9292	.9306	.9319
1.5	.9332	.9345	.9357	.9370	.9382	.9394	.9406	.9418	.9429	.9441
1.6	.9452	.9463	.9474	.9484	.9495	.9505	.9515	.9525	.9535	.9545
1.7	.9554	.9564	.9573	.9582	.9591	.9599	.9608	.9616	.9625	.9633
1.8	.9641	.9649	.9656	.9664	.9671	.9678	.9686	.9693	.9699	.9706
1.9	.9713	.9719	.9726	.9732	.9738	.9744	.9750	.9756	.9761	.9767
2.0	.9772	.9778	.9783	.9788	.9793	.9798	.9803	.9808	.9812	.9817
2.1	.9821	.9826	.9830	.9834	.9838	.9842	.9846	.9850	.9854	.9857
2.2	.9861	.9864	.9868	.9871	.9875	.9878	.9881	.9884	.9887	.9890
2.3	.9893	.9896	.9898	.9901	.9904	.9906	.9909	.9911	.9913	.9916
2.4	.9918	.9920	.9922	.9925	.9927	.9929	.9931	.9932	.9934	.9936
2.5	.9938	.9940	.9941	.9943	.9945	.9946	.9948	.9949	.9951	.9952
2.6	.9953	.9955	.9956	.9957	.9959	.9960	.9961	.9962	.9963	.9964
2.7	.9965	.9966	.9967	.9968	.9969	.9970	.9971	.9972	.9973	.9974
2.8	.9974	.9975	.9976	.9977	.9977	.9978	.9979	.9979	.9980	.9981
2.9	.9981	.9982	.9982	.9983	.9984	.9984	.9985	.9985	.9986	.9986
3.0	.9987	.9987	.9987	.9988	.9988	.9989	.9989	.9989	.9990	.9990
3.1	.9990	.9991	.9991	.9991	.9992	.9992	.9992	.9992	.9993	.9993
3.2	.9993	.9993	.9994	.9994	.9994	.9994	.9994	.9995	.9995	.9995
3.3	.9995	.9995	.9995	.9996	.9996	.9996	.9996	.9996	.9996	.9997
3.4	.9997	.9997	.9997	.9997	.9997	.9997	.9997	.9997	.9997	.9998
3.5	.9998	.9998	.9998	.9998	.9998	.9998	.9998	.9998	.9998	.9998
3.6	.9998	.9998	.9999	.9999	.9999	.9999	.9999	.9999	.9999	.9999

Tidy data

- Each variable forms a column.
- Each observation forms a row.
- Each type of observational units forms a table
- Tidy data is ready for regression routines and plotting

country	year	cases	population
Afghanistan	1999	181	197071
Afghanistan	2000	2666	205360
Brazil	1999	3737	172362
Brazil	2000	8488	1744898
China	1999	21258	1272272
China	2000	21766	1280583

variables

country	year	cases	population
Afghanistan	1999	181	197071
Afghanistan	2000	2666	205360
Brazil	1999	3737	172362
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observations

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Brazil	1999	3737	172362
Brazil	2000	8488	1744898
China	1999	21258	1272272
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values

Example: Does a full moon affect behaviour?

- Many people believe that the moon influences the actions of some individuals.
- A study of dementia patients in nursing homes recorded various types of disruptive behaviors every day for 12 weeks.
- Days were classified as moon days if they were in a 3-day period centered at the day of the full moon.
- For each patient, the average number of disruptive behaviors was computed for moon days and for all otherdays.

patient	moon_days	other_days
1	3.33	0.27
2	3.67	0.59
3	2.67	0.32
4	3.33	0.19
5	3.33	1.26
6	3.67	0.11
7	4.67	0.30

Is it tidy?

patient	moon_days	other_days
1	3.33	0.27
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3	2.67	0.32

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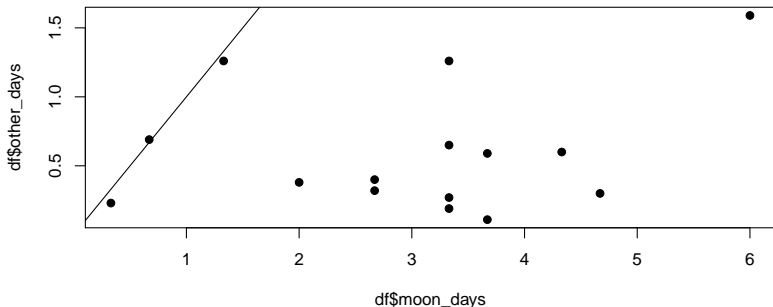
Question: Can I plot the data?

Is it tidy?

patient	moon_days	other_days
1	3.33	0.27
2	3.67	0.59
3	2.67	0.32

Question: Can I plot the data?

```
plot(df$moon_days, df$other_days, pch = 19)  
abline(a=0,b=1)
```



Is it tidy?

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Question: Can I fit a meaningful regression model directly to the variables in the data?

Is it tidy?

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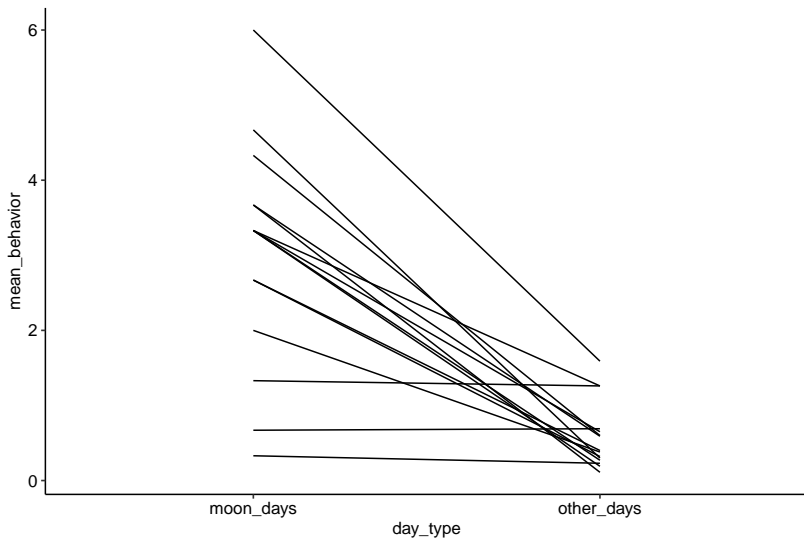
```
## Call: lm(formula = moon_days ~ other_days, data = df)
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    2.56      0.66      3.9   0.002
## other_days     0.79      0.91      0.9   0.402
##
## Residual standard error: 1.5 on 13 degrees of freedom
## Multiple R-squared:  0.055, Adjusted R-squared: -0.018
## F-statistic: 0.75 on 1 and 13 DF,  p-value: 0.4
```

Is it tidy?

patient	day_type	mean_behavior
1	moon_days	3.33
1	other_days	0.27
2	moon_days	3.67
2	other_days	0.59
3	moon_days	2.67
3	other_days	0.32
4	moon_days	3.33
4	other_days	0.19
5	moon_days	3.33
5	other_days	1.26

Plotting with tidy data

```
ggplot(data = df_tidy, mapping = aes(x = day_type, y = mean_behavior, group = patient)) + geom_line()
```



Regression with tidy data

```
fit <- lme4::lmer(mean_behavior ~ day_type + (1|patient), data = df_tidy)
summary(fit)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: mean_behavior ~ day_type + (1 | patient)
## Data: df_tidy
##
## REML criterion at convergence: 90.3
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.27236 -0.30142 -0.04023  0.48540  2.44753
##
## Random effects:
## Groups Name Variance Std.Dev.
## patient (Intercept) 0.1563  0.3954
## Residual 1.0659  1.0324
## Number of obs: 30, groups: patient, 15
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)    3.0220    0.2854  10.587
## day_typeother_days -2.4327    0.3770  -6.453
##
## Correlation of Fixed Effects:
##              (Intr)
## dy_typtthr_d -0.660
```

Not tidy vs. tidy data

patient	moon_days	other_days
1	3.33	0.27
2	3.67	0.59
3	2.67	0.32
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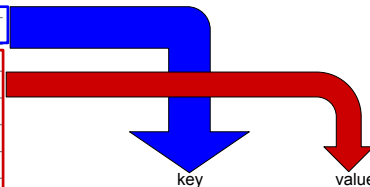
Not tidy

tidy

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3	moon_days	2.67
3	other_days	0.32
4	moon_days	3.33
4	other_days	0.19
5	moon_days	3.33
5	other_days	1.26

tidyr::pivot_longer()

patient	moon_days	other_days
1	3.33	0.27
2	3.67	0.59
3	2.67	0.32
4	3.33	0.19
5	3.33	1.26



patient	key	value
1	moon_days	3.33
1	other_days	0.27
2	moon_days	3.67
2	other_days	0.59
3	moon_days	2.67
3	other_days	0.32
4	moon_days	3.33
4	other_days	0.19
5	moon_days	3.33
5	other_days	1.26

```
tidyr::pivot_longer(data = df, cols = -patient, names_to = "day_type", values_to = "mean_behavior")
```


Traditional stats textbook

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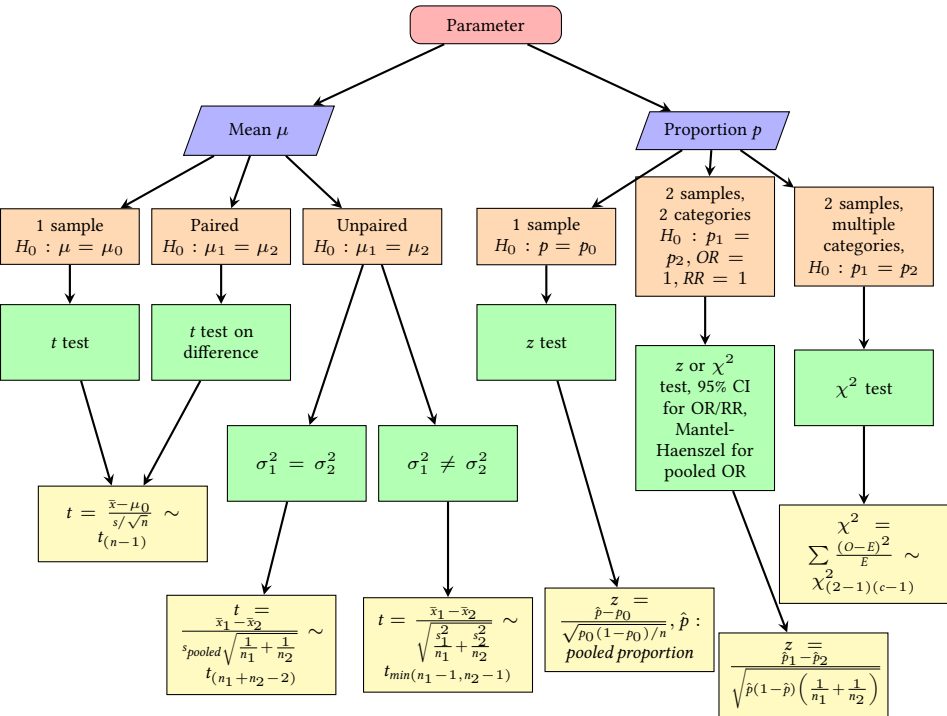
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Statistical concepts

RESULTS The total populations were 462 445 in the Iowa border counties and 272 385 in the Illinois border counties. Population density was higher in the Iowa counties (114.2 people per square mile) than in the Illinois counties (78.2 people per square mile). Trends of cumulative COVID-19 cases per 10 000 residents for the Iowa and Illinois border counties were comparable before the Illinois stay-at-home order, which went into effect at 5:00 PM on March 21 (March 15 to March 21: 0.024 per 10 000 residents vs 0.026 per 10 000 residents). After that, cases increased more quickly in Iowa and more slowly in Illinois. Within 10, 20, and 30 days after the enactment of the stay-at-home order in Illinois, the difference in cases was -0.51 per 10 000 residents (SE, 0.09; 95% CI, -0.69 to -0.32 ; $P < .001$), -1.15 per 10 000 residents (SE, 0.49; 95% CI, -2.12 to -0.18 ; $P = .02$), and -4.71 per 10 000 residents (SE, 1.99; 95% CI, -8.64 to -0.78 ; $P = .02$), respectively. The estimates indicate excess cases in the border Iowa counties by as many as 217 cases after 1 month without a stay-at-home order. This estimate of excess cases represents 30.4% of the 716 total cases in those Iowa counties by that date. Sensitivity analyses addressing differences in timing of closing schools and nonessential businesses and differences in county population density and poverty rates between the 2 states supported these findings.

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⁴<https://jamanetwork.com/journals/jamanetworkopen/fullarticle/2766229>

Statistical concepts

Table 1. Difference-in-Differences Estimates of COVID-19 Cases Comparing Border Counties in Iowa With Those in Illinois Before and After the Stay-at-Home Order Was Issued in Illinois^a

Period	Difference in COVID-19 cases per 10 000 residents ^b	Heteroskedasticity robust SE (95% CI) ^c	P value	Excess cases in Iowa border counties	Excess cases as proportion of total cases, %
3/22-3/26	-0.14	0.04 (-0.23 to -0.06)	.001	6	32.4
3/27-3/31	-0.51	0.09 (-0.69 to -0.32)	<.001	24	38.0
4/01-4/05	-0.41	0.17 (-0.74 to -0.07)	.02	19	15.2
4/06-4/10	-1.15	0.49 (-2.12 to -0.18)	.02	53	17.8
4/11-4/15	-3.35	1.19 (-5.70 to -0.99)	.006	154	30.0
4/16-4/20	-4.71	1.99 (-8.64 to -0.78)	.02	217	30.4

Abbreviation: COVID-19, coronavirus disease 2019.

^a The regression model was estimated separately for each of 5-day period. The regression was estimated using least squares weighted by the 2019 county population. The regression adjusted for county and day fixed effects. The number of county-day observations was 180 for each regression.

^b This indicates the estimated difference-in-differences association of a stay-at-home order with COVID-19 cases in a given period relative to March 15 to March 21 (ie, the period before the stay-at-home order in Illinois was enacted).

^c Heteroskedasticity robust SEs were estimated because homoscedasticity is rejected for all post-period regressions.

5

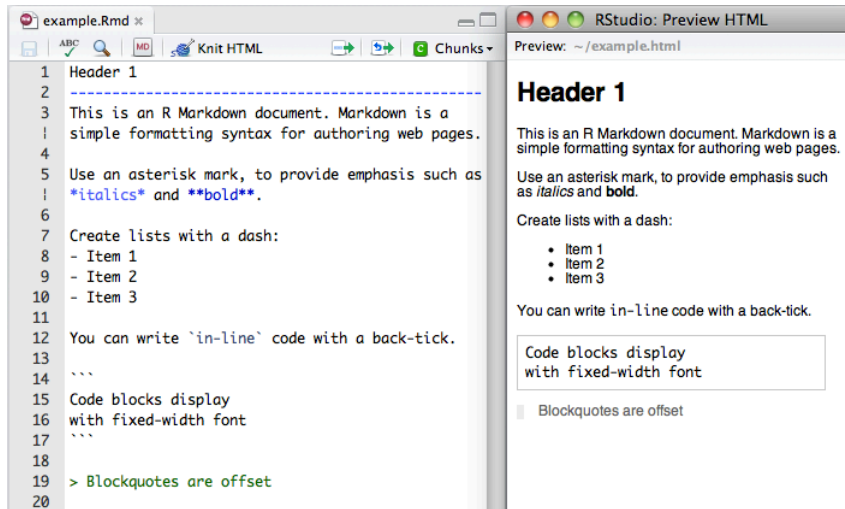
⁵ <https://jamanetwork.com/journals/jamanetworkopen/fullarticle/2766229>

Copy paste ad nauseam

The screenshot illustrates a workflow for copying data from Excel to Word. In the background, the Excel window 'Import Sample.xlsx' shows a table with columns: Designer, Project, Due Date, and Budget. The data includes Anderson (Christmas Village, Nov 28, \$26,000.00), Cochran (Arts Festival, Sep 15, \$31,000.00), Fotou (Botanical Gardens, Aug 10, \$18,000.00), Heath (LeMieux Galleries, Oct 15, \$22,000.00), and Hill (Beau Rivage, Nov 01, \$24,500.00). A red box highlights the 'Paste' button in the Excel ribbon, with a red arrow pointing to it. In the foreground, the Word window 'Select Table from Graph.doc [Compatibility Mode] - Word' shows the 'Paste' button in the ribbon, also highlighted with a red box. A 'Paste Options' menu is open, showing icons for pasting as text, text with formatting, and a table. The 'Table' icon is selected, and the data from the Excel table is pasted into the Word document as a table.

Designer	Project	Due Date	Budget
Anderson	Christmas Village	Nov 28	\$ 26,000.00
Cochran	Arts Festival	Sep 15	\$ 31,000.00
Fotou	Botanical Gardens	Aug 10	\$ 18,000.00
Heath	LeMieux Galleries	Oct 15	\$ 22,000.00
Hill	Beau Rivage	Nov 01	\$ 24,500.00

Markdown: HTML without knowing HTML



The screenshot displays the RStudio interface with two windows. The left window, titled 'example.Rmd', shows a Markdown document with the following content:

```
1 Header 1
2 -----
3 This is an R Markdown document. Markdown is a
4 | simple formatting syntax for authoring web pages.
5 Use an asterisk mark, to provide emphasis such as
6 | italics and bold.
7 Create lists with a dash:
8 - Item 1
9 - Item 2
10 - Item 3
11
12 You can write `in-line` code with a back-tick.
13
14 ```
15 Code blocks display
16 with fixed-width font
17 ```
18
19 > Blockquotes are offset
20
```

The right window, titled 'RStudio: Preview HTML', shows the rendered HTML output of the document. The preview URL is '~ / example.html'. The rendered content is as follows:

Header 1

This is an R Markdown document. Markdown is a simple formatting syntax for authoring web pages.

Use an asterisk mark, to provide emphasis such as *italics* and **bold**.

Create lists with a dash:

- Item 1
- Item 2
- Item 3

You can write `in-line` code with a back-tick.

```
Code blocks display
with fixed-width font
```

Blockquotes are offset

R + Markdown = RMarkdown

chunks.Rmd

Knit HTML

Chunks

```
1 R Code Chunks
2 =====
3
4 With R Markdown, you can insert R code
5 chunks including plots:
6
7 ```{r qplot, fig.width=4, fig.height=3,
8 message=FALSE}
9 # quick summary and plot
10 library(ggplot2)
11 summary(cars)
12 qplot(speed, dist, data=cars) +
13   geom_smooth()
```

RStudio: Preview HTML

Preview: ~/chunks.html

Save As

Publish

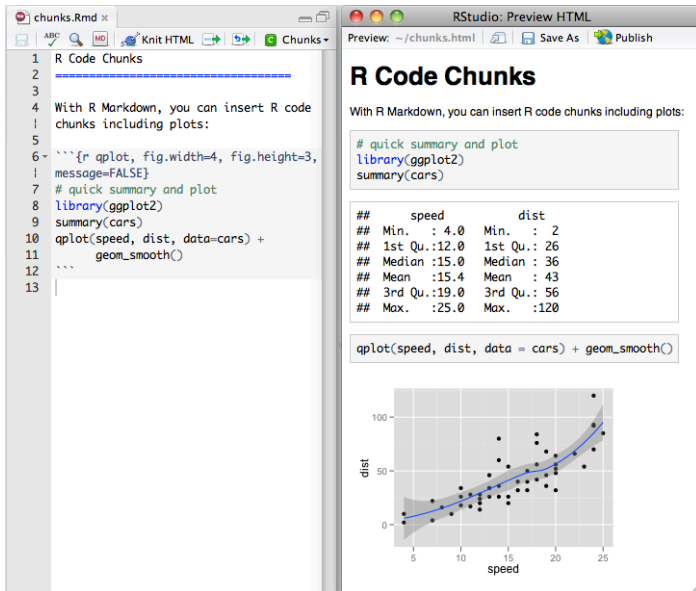
R Code Chunks

With R Markdown, you can insert R code chunks including plots:

```
# quick summary and plot
library(ggplot2)
summary(cars)
```

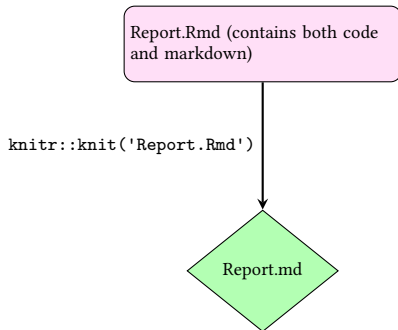
##	speed	dist
##	Min. : 4.0	Min. : 2
##	1st Qu.: 12.0	1st Qu.: 26
##	Median : 15.0	Median : 36
##	Mean : 15.4	Mean : 43
##	3rd Qu.: 19.0	3rd Qu.: 56
##	Max. : 25.0	Max. : 120

```
qplot(speed, dist, data = cars) + geom_smooth()
```



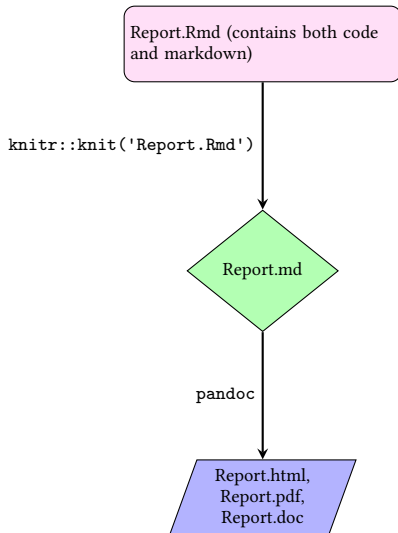
What rmarkdown does

RMarkdown example:



What rmarkdown does

RMarkdown example:

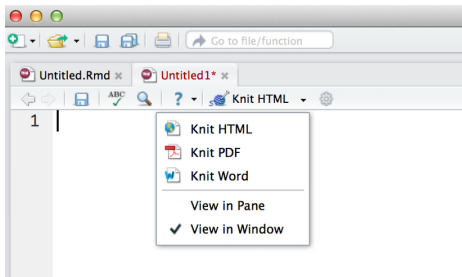


Compiling a .Rmd document

The two steps on previous slide can be executed in one command:

```
rmarkdown::render()
```

or in RStudio:



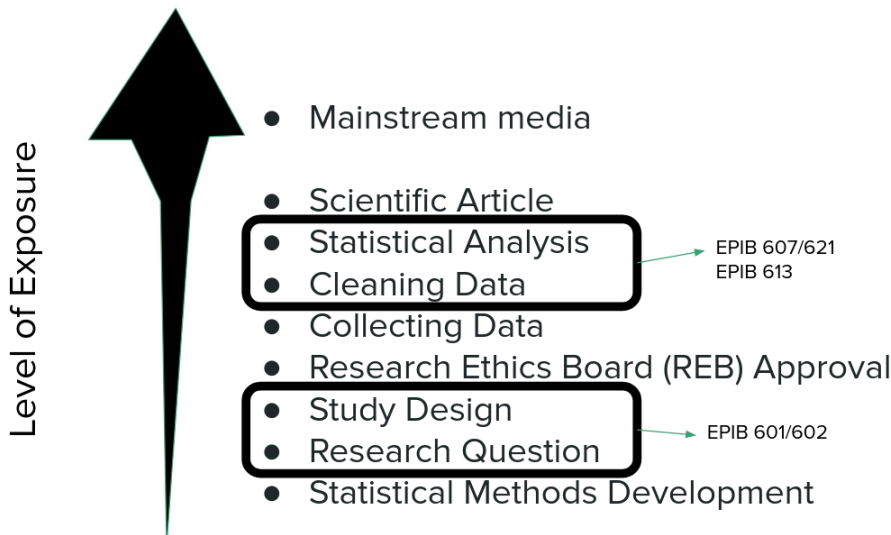
Topics by level of exposure

Level of Exposure



- Mainstream media
- Scientific Article
- Statistical Analysis
- Cleaning Data
- Collecting Data
- Research Ethics Board (REB) Approval
- Study Design
- Research Question
- Statistical Methods Development

First year courses



My area of research



Session Info

```
R version 3.6.2 (2019-12-12)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Pop!_OS 19.10

Matrix products: default
BLAS:   /usr/lib/x86_64-linux-gnu/openblas/libblas.so.3
LAPACK: /usr/lib/x86_64-linux-gnu/libopenblas-p-r0.3.7.so

attached base packages:
[1] tools      stats      graphics  grDevices  utils      datasets  methods
[8] base

other attached packages:
[1] NCStats_0.4.7      FSA_0.8.30          forcats_0.5.0       stringr_1.4.0
[5] dplyr_1.0.2        purrr_0.3.4         readr_1.3.1         tidyr_1.1.2
[9] tibble_3.0.3       ggplot2_3.3.2.9000 tidyverse_1.3.0     knitr_1.29

loaded via a namespace (and not attached):
[1] Rcpp_1.0.4.6        lubridate_1.7.4     lattice_0.20-38     assertthat_0.2.1
[5] digest_0.6.25       R6_2.4.1            cellranger_1.1.0    plyr_1.8.6
[9] backports_1.1.9     reprex_0.3.0        evaluate_0.14       httr_1.4.1
[13] highr_0.8           pillar_1.4.6        TeachingDemos_2.12 rlang_0.4.7
[17] readxl_1.3.1        rstudioapi_0.11     minqa_1.2.4         nloptr_1.2.2.1
[21] Matrix_1.2-18       labeling_0.3         splines_3.6.2       lme4_1.1-21
[25] munsell_0.5.0       broom_0.7.0         compiler_3.6.2      modelr_0.1.5
[29] xfun_0.16           pkgconfig_2.0.3     tidyselect_1.1.0    fansi_0.4.1
[33] crayon_1.3.4        dbplyr_1.4.2        withr_2.2.0         ggpubr_0.2.5
[37] MASS_7.3-51.5       grid_3.6.2          nlme_3.1-143        jsonlite_1.7.0
[41] gtable_0.3.0        lifecycle_0.2.0     DBI_1.1.0           magrittr_1.5
[45] scales_1.1.1        cli_2.0.2           stringi_1.4.6       farver_2.0.3
[49] ggsignif_0.6.0      fs_1.3.2            xml2_1.3.0          ellipsis_0.3.1
[53] generics_0.0.2      vctrs_0.3.4         boot_1.3-24         glue_1.4.2
[57] hms_0.5.3           colorspace_1.4-1    rvest_0.3.5         haven_2.3.1
```