#### Winter Institute in Data Science

Ryan T. Moore

2022 - 01 - 02

Goals

Skills

Examples

# Welcome!

► Political methodologist in Dept of Government in SPA

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# Data Science

# Particular intersection of

- ► Statistical practice
- ► Computational tools
- ➤ Substantive knowledge

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- ► Substance: social science

# Computers!

Harvard Business Review

Data | Data Scientist: The Sexiest Job of the 21st Century

# Data Scientist: The Sexiest Job of the 21st Century

by Thomas H. Davenport and D.J. Patil

From the October 2012 Issue

BLS tracks data science now.

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... and salaries are high ...

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- ▶ Do original research using data sci methods. Contribute methods, substance, both.

### Skills

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# Examples

"Keep only non-voters who might be subject to interference"

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```
social <- read_csv("http://j.mp/2Et71U0")
filter(social, (hhsize > 1) & (primary2004 == 0))
```

##

##

##

7 female

9 female

8 male

## 10 male

"Keep only non-voters who might be subject to interference"

```
social <- read_csv("http://j.mp/2Et71U0")</pre>
filter(social, (hhsize > 1) & (primary2004 == 0))
```

```
## # A tibble: 161,275 x 6
```

## yearofbirth primary2004 messages primary2000 <dbl> <dbl> <chr> <dbl: ## <chr>

## 1 male 1941 O Civic Duty 2 female 1947 O Civic Duty ##

1951 0 Hawthorne ## 3 male

## 4 female 1950 0 Hawthorne

## 5 female 1982 0 Hawthorne

##

1959

1956

1968

1967

6 male 1981 0 Control

0 Control

0 Control

0 Control

0 Control

54 / 68

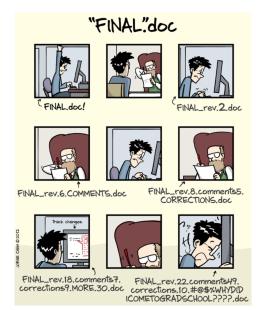
"I need to read these dates from Spanish  $\leadsto$  standard format',

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```
## [1] "2000-01-15"
```

I collaborate, but FinalLAST draft.v.2.doc (1) is painful...

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I need to collaborate, but FinalFinalLAST draft.v.2.doc (1) isn't working for me anymore.

```
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```

git add paper.tex
git commit paper.tex
git push

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- ► How can we fairly estimate probability defendant will appear?
- ➤ Are intersections with new patterns less prone to traffic accidents?
- ► How do we compare models/prediction strategies?

### Course GitHub page:

https://github.com/ryantmoore/winter-inst-2022

(syllabus tour)

#### Installations

- R: https://cran.r-project.org
- ► RStudio (Desktop): https://rstudio.com/products/rstudio/
- ► Anaconda: https://www.anaconda.com/distribution/