Winter Institute in Data Science

Ryan T. Moore

2021-01-02

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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Highest concentrations: DC, WA, MD, VA, IL!

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... and salaries are high ...

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

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 - text as data

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

10 male

8 male

9 female

##

##

"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> ## <chr> <dbl> <chr> <dbl: 1 male ## 1941 O Civic Duty

2 female 1947 O Civic Duty 3 male 1951 0 Hawthorne

1950 ## 4 female 0 Hawthorne

5 female 0 Hawthorne ##

1982

1959

1956

1968

1967

... with 161,265 more rows

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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- ▶ What characteristics of rodent complaints actually lead to successful abatement?
- ► 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/