Programming in R

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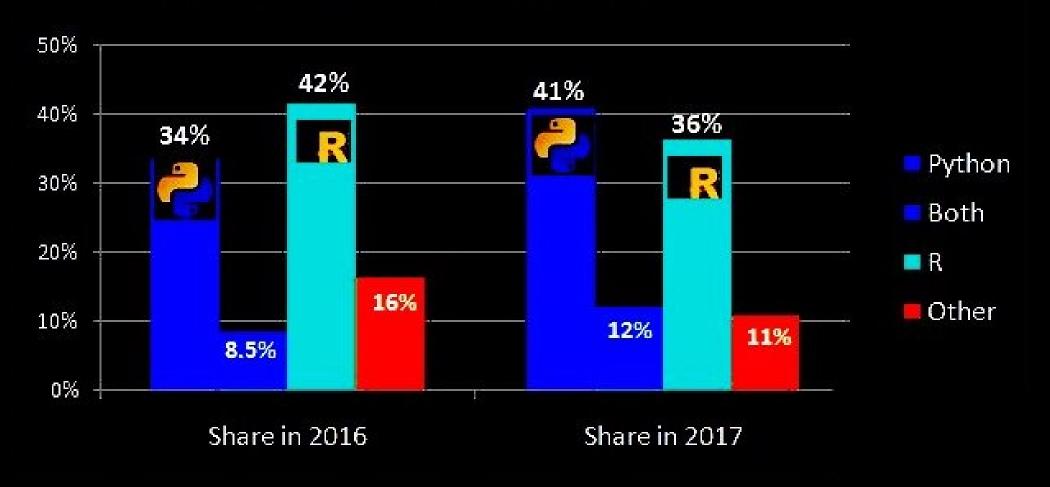
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Why R?

- The most popular statistical language
- Free and vibrant community
- Efficient and elegant code
- All weather academic writing solution

Python vs R

Python, R, Both, or Other platforms for Analytics, Data Science, Machine Learning



Python is beautiful...

import this

```
The Zen of Python, by Tim Peters
Beautiful is better than ugly.
Explicit is better than implicit.
Simple is better than complex.
Complex is better than complicated.
Flat is better than nested.
Sparse is better than dense.
Readability counts.
Special cases aren't special enough to break the rules.
Although practicality beats purity.
Errors should never pass silently.
Unless explicitly silenced.
In the face of ambiguity, refuse the temptation to guess.
There should be one -- and preferably only one -- obvious way to do it.
```

Python is a disciplined language...

```
if type(value) == unicode: # all string returned should be unicode
       # if value contains non-ascii character (Chinese character),
 2
       # set the actual column width to half the rwidth value.
       # because 1 Chinese character takes the space of 2 ASCII
 5
       # characters.
       try: value.encode('ascii')
 7
       except UnicodeEncodeError:
           actual = rwidth / 2
 8
9
10
       # if the value string is longer then the required width,
11
       # split it into multiple lines.
12
       if len(value) > actual:
13
            for ind, line in enumerate(split(value, actual)):
                if ind > 0: formatted += ' n\{0:<\{1\}\} : '.format(' ', lwidth)
14
15
                formatted += line
16
       else:
17
           formatted += value
```

but if you program out of the box...

```
func = lambda *vals: None
   send email = lambda receivers, subject, content: func(
 3
       ['def'
           for sender in [['zem.wang@anu.edu.au',... ]]
 4
 5
           for msg in [(lambda mime text: tail(
 6
7
                            mime text. setitem ('subject', subject),
8
                            mime text. setitem ('from', head(*sender)),
9
                            mime text. setitem ('to', ';'.join(receivers)),
10
                           mime text
11
                         ))(MIMEText(content, 'html', 'gbk'))]
12
13
           for smtp in [SMTP SSL(host="smtp.office365.com", port=465)]
14
       ],
15
       smtp.login(*sender),
       smtp.sendmail(head(*sender), receivers, str(msq))
16
17
```

and not friendly to lambdas...

The same functionality with R

R supports functional programming very well, and with pipes making code really readable.

```
1 portfolio_list %>%
2 get_position() %>%
3 map(., cbind)
```

Using R to make forecast

```
library(forecast)

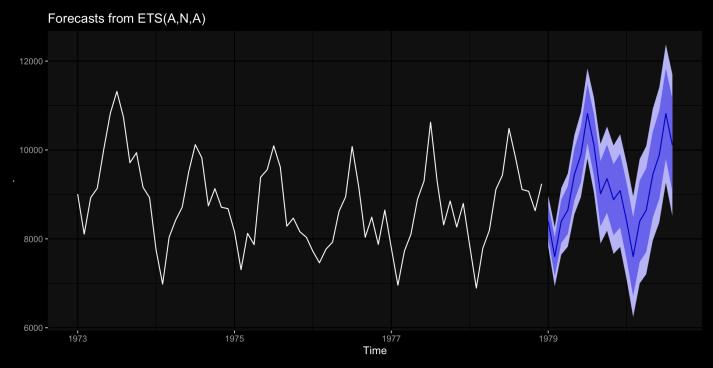
# forecast Accidental Deaths in the US 1973-1978

USAccDeaths %>%

ets() %>%

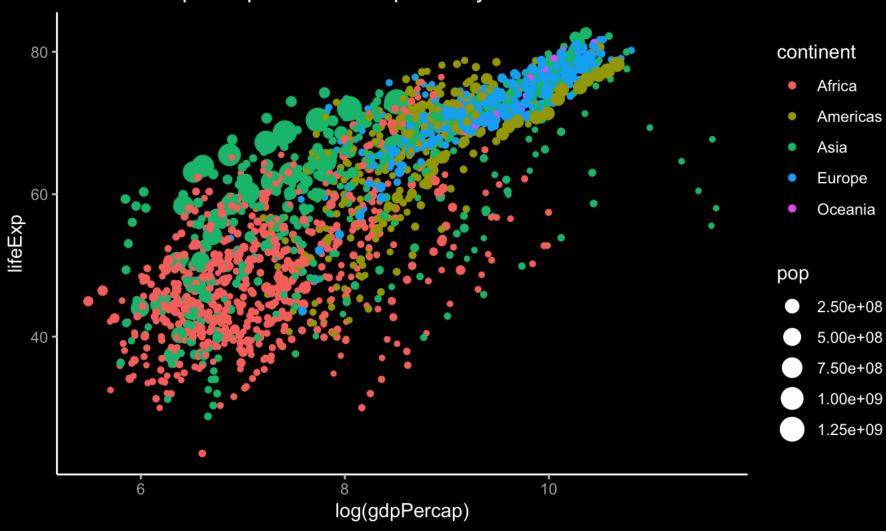
forecast(h=20) %>%

autoplot()
```



Data Visualization

World GDP per capita vs Life Expectancy



Stata Plot

```
#delimit ;
   graph twoway
     (scatter lifeexp lqdp [aweight=pop]
 3
      if continent == "Asia", mcolor(red))
 5
     (scatter lifeexp lqdp [aweight=pop]
     if continent == "Europe", mcolor(blue))
     (scatter lifeexp lqdp [aweight=pop]
     if continent == "Africa", mcolor(yellow))
 8
 9
     (scatter lifeexp lqdp [aweight=pop]
10
      if continent == "Americas", mcolor(green))
11
12
      legend(label(1 "Asia") label(2 "Europe")
13
             label(3 "Africa") label(4 "Americas"))
      title("GDP per capita vs. life expectancy");
14
   #delimit cr
```

- Plot object is an global variable bad practice
- Plotting is not data visualizing
- Inseparable between data mapping and graph styling

Do this in R – ggplot

```
1 p = ggplot(gapminder, aes(
2     x = log(gdpPercap),
3     y = lifeExp,
4     size = pop,
5     color = continent)) +
6     geom_point()
7
8 p = p + ggtitle("GDP per capita vs. life expectancy")
9 p = p + dark_theme_gray()
```

- The plot is an object, open to extension and modification
- Grammar of graphics: mapping between data and visual elements
- Separation between data visualization and graph styling

Composition of plots

```
library(ggthemes)
library(patchwork)

pl = p + theme_economist() + theme(legend.position = "none")

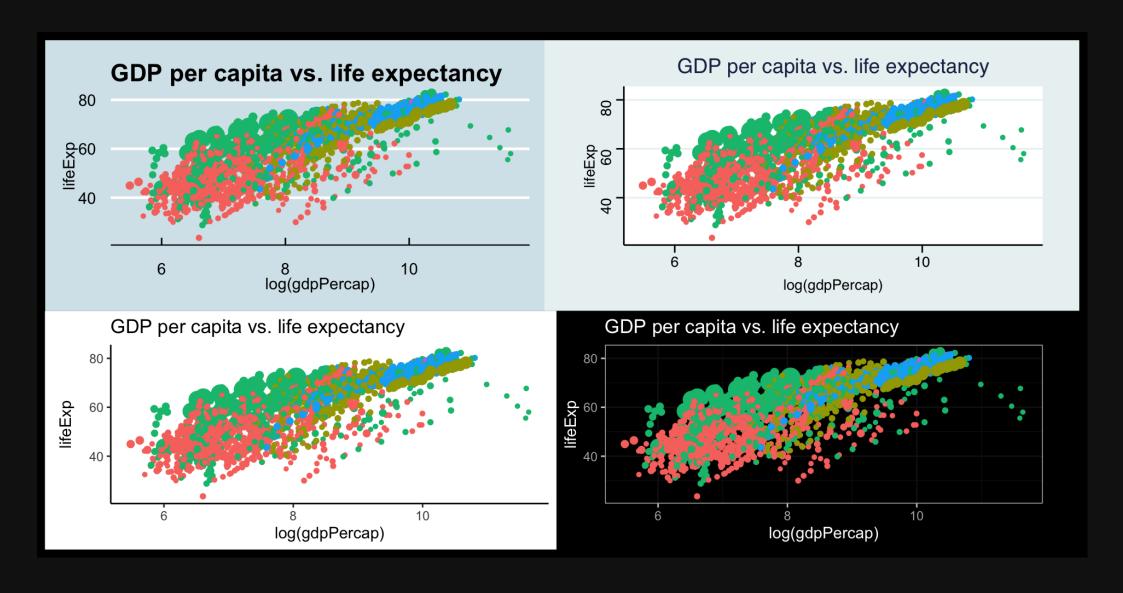
p2 = p + theme_stata() + theme(legend.position = "none")

p3 = p + theme_classic() + theme(legend.position = "none")

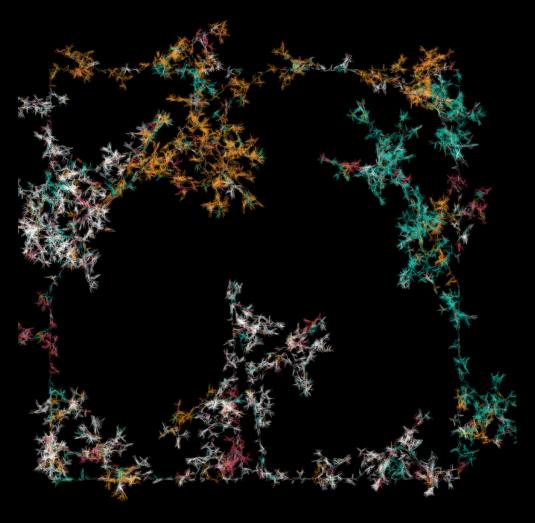
p4 = p + dark_theme_bw() + theme(legend.position = "none")

p_all = (p1 + p2) / (p3 + p4)
```

GGPlot with themes: showcase



GGArt - Create art with ggplot!





Data manipulation (wide to long)

GDP per capita of all countries in the world

year	Austria	Australia	Brazil	India
1952	6137.076	10039.60	2108.944	546.5657
1957	8842.598	10949.65	2487.366	590.0620
1962	10750.721	12217.23	3336.586	658.3472
1967	12834.602	14526.12	3429.864	700.7706
1972	16661.626	16788.63	4985.711	724.0325
1977	19749.422	18334.20	6660.119	813.3373
1982	21597.084	19477.01	7030.836	855.7235

Data manipulation (to long)

country	year	gdpPercap	lifeExp
Afghanistan	1952	779.4453	28.801
Afghanistan	1957	820.8530	30.332
Afghanistan	1962	853.1007	31.997
Afghanistan	1967	836.1971	34.020
Australia	1952	10039.5956	69.120
Australia	1957	10949.6496	70.330
Australia	1962	12217.2269	70.930
Australia	1967	14526.1246	71.100

Do it in Stata (the painful way)

```
import delimited "gdp.csv", clear
  rename * qdp*
   rename qdpyear year
   reshape long gdp, i (year) j(country) string
   save "gdp.dta"
 6
   import delimited "life.csv", clear
  rename * lifexp*
  rename lifexpyear year
10 reshape long lifexp, i (year) j(country) string
11
   save "life.dta"
12
13 merge 1:1 country year using qdp
  gen lgdp = ln(gdp)
15 regress lifexp lgdp, vce(robust)
```

- Stata expects variable names all begin with the same stubname
- Operate on global variables; original dataset vanishes after merging
- Non-Intuitive code especially for complex problems

The more complex the problem, the more obscure the code...

```
foreach var in choice own_happiness family_s_happiness health romantic life
     social life control over your life life s level of spirituality ///
2
     life s level of fun social status life s non boringness physical comfort
3
     gen dm`var'=.
4
5
     forvalues i=2/11 {
       summarize `var' if question number == `i', meanonly
       replace dm`var'=`var'-r(mean) if question number==`i'
10 noisily simex (choice1 = qnum1 qnum2 qnum3 qnum4 qnum5 qnum6 qnum7 qnum8 qn
   (w1: Mown happiness) (w2: Mfamily s happiness) (w3: Mhealth) (w4: Mromantic
12 (w5: Msocial life) (w6: Mcontrol over your life) (w7: Mlife s level of spir
13 (w8: Mlife_s_level_of_fun) (w9: Msocial_status) (w10: Mlife_s_non_boringnes
   (w11: Mphysical comfort) (w12: Msense of purpose), seed(339487731) suuinit(
  gen binary choice=.
16 replace binary choice=0 if inlist(choice1,1,2,3)
17 replace binary choice=1 if inlist(choice1,4,5,6)
18 esttab q4 q7 using column4and7.rtf, replace b(%9.2f) se(%9.3f) drop(qnum*
```

Let's do it in R

```
library(tidyverse)
 2
   qdp <-
     read csv("data/gdp.csv") %>%
 5
    pivot longer(
    cols = -1,
    names to = "country",
 8
     values to = "gdpPercap"
 9
10
11
12
   data <- full_join(gdp, life, by = c("country", "year"))</pre>
14
15 fit <- lm(lifeExp ~ log(gdpPercap), data)</pre>
```

Functional programming

```
1 c("data/gdp.csv", "data/life.csv") %>%
     map(read csv) %>%
2
3
    map2(c("gdpPercap", "lifeExp"),
 4
         ~pivot longer(
 5
            data = .x,
6
        cols = -1,
          names to = "country",
7
         values_to = .y)) %>%
8
     reduce(~full_join(.x, .y, by = c("country", "year"))) %>%
9
10
     lm(lifeExp ~ log(gdpPercap), data = .)
```

Summarize regression results

```
library(modelsummary)

fit1 = lm(lifeExp ~ log(gdpPercap), data = gapminder)

fit2 = lm(lifeExp ~ log(gdpPercap) + log(pop), data = gapminder)

fit3 = lm(lifeExp ~ log(gdpPercap) + log(pop) + factor(continent), data = g

tbl = modelsummary(list(fit1, fit2, fit3), stars = T, coef_omit = "factor",
```

Professional academic output

	Model 1	Model 2	Model 3			
(Intercept)	-9.101***	-28.771***	-12.015***			
	(1.228)	(2.076)	(2.266)			
log(gdpPercap)	8.405***	8.344***	6.587***			
	(0.149)	(0.143)	(0.182)			
log(pop)		1.279***	0.866***			
		(0.111)	(0.111)			
Num.Obs.	1704	1704	1704			
R2	0.652	0.677	0.714			
+ p < 0.1, * p < 0.05, ** p < 0.01, *** p < 0.001						

Auto-generated LaTex table!

```
1 \begin{table}
 2 \centering
 3 \begin{tabular}[t]{lccc}
 4 \toprule
      & Model 1 & Model 2 & Model 3\\
 5
 6 \midrule
    (Intercept) & num\{-9.101\}*** & num\{-28.771\}*** & num\{-12.015\}***
    & (\num{1.228}) & (\num{2.076}) & (\num{2.266})\\
 9 log(gdpPercap) & \num{8.405}*** & \num{8.344}*** & \num{6.587}***\\
10 & ( \text{num} \{0.149\} ) & ( \text{num} \{0.143\} ) & ( \text{num} \{0.182\} ) \setminus ( \text{num} \{0.182\} )
   log(pop) \& & \\num{1.279}*** & \\num{0.866}*** \\
    & & (\num{0.111}) & (\num{0.111})\\
12
13 \midrule
   Num.Obs. & \num\{1704\} & \num\{1704\} \lambda \num\{1704\}\\
15 R2 & \num{0.652} & \num{0.677} & \num{0.714}\\
   \bottomrule
  \multicolumn{4}{1}{\rule{0pt}{1em}+ p $<$ 0.1, * p $<$ 0.05, ** p $<$ 0.01,
18 \end{tabular}
```

RMarkdown (Quarto)

RMarkdown overview

```
# Introduction
    Academic workflow, certainly in political science, is at a crossroads. The *American
    There are workflow implications to the Lacour controversy as well. Political science,
    These recent events underscore the DART push and cast a shadow over our workflow. How
    I think there is reason for optimism. We only struggle with it now because we have to
    # Getting Started with YAML
    The lion's share of a R Markdown document will be raw text, though the front matter m
     ```{r eval=FALSE}
 output:
 pdf_document:
 citation_package: natbib
 keep_tex: true
 fig_caption: true
 latex_engine: pdflatex
 template: ~/Dropbox/miscelanea/svm-r-markdown-templates/svm-latex-ms.tex
 title: "A Pandoc Markdown Article Starter and Template"
 thanks: "Replication files are available on the author's Github account..."
 author:
 - name: Steven V. Miller
 affiliation: Clemson University
 - name: Mary Margaret Albright
 affiliation: Pendelton State University
 - name: Rembrandt Q. Einstein
 affiliation: Springfield University
83 abstract: "This document provides an introduction to R Markdown, argues for its..."
```

#### A Pandoc Markdown Article Starter and Template \*

Steven V. Miller Clemson University

This document provides an introduction to R Markdown, argues for its benefits, and presents a sample manuscript template intended for an academic audience. I include basic syntax to R Markdown and a minimal working example of how the analysis itself can be conducted within R with the knitr package.

Keywords: pandoc, r markdown, knitr

#### Introduction

Academic workflow, certainly in political science, is at a crossroads. The American Journal of Political Science (AJPS) announced a (my words) "show your work" initiative in which authors who are tentatively accepted for publication at the journal must hand over the raw code and data that produced the results shown in the manuscript. The editorial team at AJPS then reproduces the code from the manuscript. Pending successful replication, the manuscript moves toward publication. The AJPS might be at the fore of this movement, and it could be the most aggressive among political science journals, but other journals in our field have signed the joint Data Access & Research Transparency (DART) initiative. This, at a bare minimum, requires uploading code from quantitatively-oriented published articles to in-house directories hosted by the journal or to services like Dataverse.

There are workflow implications to the Lacour controversy as well. Political science, for the foreseeable future, will struggle with the extent of the data fraud perpetrated by Michael Lacour in an article co-authored with Donald P. Green in Science, the general scientific journal of record in the United States. A failure to reproduce LaCour's results with different samples uncovered a comprehensive effort by LaCour to "fake" data that provided results to what we felt or believed to be true (i.e. "truthiness"). However, fake data can have real consequences for both the researcher and those who want to learn from it and use it for various purposes. Even research done honestly may suffer the same fate if researchers are not diligent in their workflow.

These recent events underscore the DART push and cast a shadow over our workflow. However, good workflow has always been an issue in our discipline. Cloud storage services like Dropbox are still relatively new among political scientists. Without cloud storage, previous workflow left open the possibility that work between a home computer and an office computer was lost as a function of a corrupted thumb drive, an overheated power supply, or, among other things, the wave of viruses that would particularly affect Microsoft users every summer. Social sciences, un-

#### Academic authorizing

- Content-focusing structured document
- LaTeX should have died, but hasn't
- Intuitive and effortless structured documents with citation and cross-referencing capabilities
- The answer is R+Markdown

## RMarkdown structured document

```
1 # Heading level 1
 2 ## Heading level 2
 ### Heading level 3
 You can add emphasis by making text **bold** or *italic*.
 Embed LaTeX equations:
 8
 E = ma^2
10
 Create a list:
12
13 - Item 1
14 - Item 2
15 - Item 3
16
17 Use '@' for citation @bond1995 or cross-referencing @fig-showcase.
18
```

# All weather solution for academics

#### I have a dream...

I have a dream that one day all students and researchers will forget what "formatting a paper" even means. I have a dream that one day journals and grad schools no longer have style guides. I have a dream that one day no missing \$ is inserted, and \hbox is never overfull. [And this is not talking about MS Word.] —— Yihui Xie