More Web Scraping

Stat 220

Bastola

February 16 2022

Get Links

```
bow(url = "https://www.imdb.com/search/title/?groups=best_
   scrape() %>%
  html_elements(css = ".lister-item-header a") %>%
  html_attr(name = "href") %>%
  url_absolute(base = "https://www.imdb.com")
```

```
"https://www.imdb.com/title/tt9770150/?ref =adv li tt
    "https://www.imdb.com/title/tt6751668/?ref =adv li tt
 [2]
    "https://www.imdb.com/title/tt6966692/?ref =adv li tt
 [3]
     "https://www.imdb.com/title/tt5580390/?ref =adv li tt
 [4]
 [5]
    "https://www.imdb.com/title/tt4975722/?ref =adv li tt
    "https://www.imdb.com/title/tt1895587/?ref =adv li tt
 [6]
     "https://www.imdb.com/title/tt2562232/?ref =adv li tt
    "https://www.imdb.com/title/tt2024544/?ref =adv li tt
    "https://www.imdb.com/title/tt1024648/?ref =adv li tt
    "https://www.imdb.com/title/tt1655442/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt1504320/?ref_=adv_li_tt
\lceil 11 \rceil
[12]
    "https://www.imdb.com/title/tt1010048/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0887912/?ref_=adv_li_tt
[14] "https://www.imdb.com/title/tt0477348/?ref =adv li tt
    "https://www.imdb.com/title/tt0407887/?ref =adv li tt
    "https://www.imdb.com/title/tt0405159/?ref =adv li tt
    "https://www.imdb.com/title/tt0375679/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0167260/?ref =adv li tt
Γ18]
    "https://www.imdb.com/title/tt0299658/?ref_=adv_li_tt
Γ19 ]
    "https://www.imdb.com/title/tt0268978/?ref_=adv li tt
[20]
    "https://www.imdb.com/title/tt0172495/?ref_=adv li tt
[21]
    "https://www.imdb.com/title/tt0169547/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0138097/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0120338/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0116209/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0112573/?ref_=adv_li_tt
[27]
    "https://www.imdb.com/title/tt0109830/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0108052/?ref_=adv_li_tt
[28]
    "https://www.imdb.com/title/tt0105695/?ref_=adv_li_tt
[29]
    "https://www.imdb.com/title/tt0102926/?ref =adv li tt
    "https://www.imdb.com/title/tt0099348/?ref_=adv_li_tt
[32]
    "https://www.imdb.com/title/tt0097239/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0095953/?ref_=adv_li_tt
    "https://www.imdb.com/title/tt0093389/?ref_=adv_li_tt
```

Scrape Table

```
table_usafacts <- bow(url = "https://usafacts.org/visualizations/covid-vaccine-tracker-states/state
  scrape() %>% html_elements(css = "table") %>% html_table()
```

knitr::kable(table_usafacts[[3]], format = "html")

State	% of population with at least one dose	% fully vaccinated	% with booster or additional dose
Alabama	61.5%	49.8%	16.7%
Alaska	68.1%	59.9%	24.2%
Arizona	70.8%	59.4%	22.8%
Arkansas	65.3%	53%	19.6%
California	81.1%	69.6%	32.5%
Colorado	77.8%	68.7%	33.6%
Connecticut	93.2%	77.2%	37.1%
Delaware	81%	66.8%	28.6%



05:00

Please clone the repository on advanced web scraping and visualization to your local folder. Go to this webpage and scrape the table that has the latest county-level coronavirus stats for the state of Minnesota.

County	7-day avg. cases	7-day avg. deaths	Cases	Deaths
Aitkin County	7	0	2,836	59
Anoka County	244	2	96,275	757
Becker County	34	0	8,386	86
Beltrami County	35	0	11,038	118
Benton County	53	0	13,555	164
Big Stone County	4	0	1,335	8
Blue Earth County	60	0	17,117	89
Brown County	16	0	6,391	72
Carlton County	57	0	8,500	88
Carver County	71	1	25,772	107
<u> </u>	^^	^		

What are the top 10 counties in Minnesota by the number of COVID cases?

```
all_url <- "https://finance.yahoo.com/losers?count=25&offset="</pre>
```

```
idx <- seq(0, 250, by = 25)

table_new <-data.frame()

df <- data.frame()

for (i in seq_along(idx)) {
   new_webpage <- read_html(str_glue(all_url, {idx[i]}))
   table_new <- html_table(new_webpage)[[1]] %>%
        as_tibble(.name_repair = "unique")
   df <- rbind(df, table_new)
}</pre>
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   df <- rbind(df, table_new)
}</pre>
```

Multiple tables combined

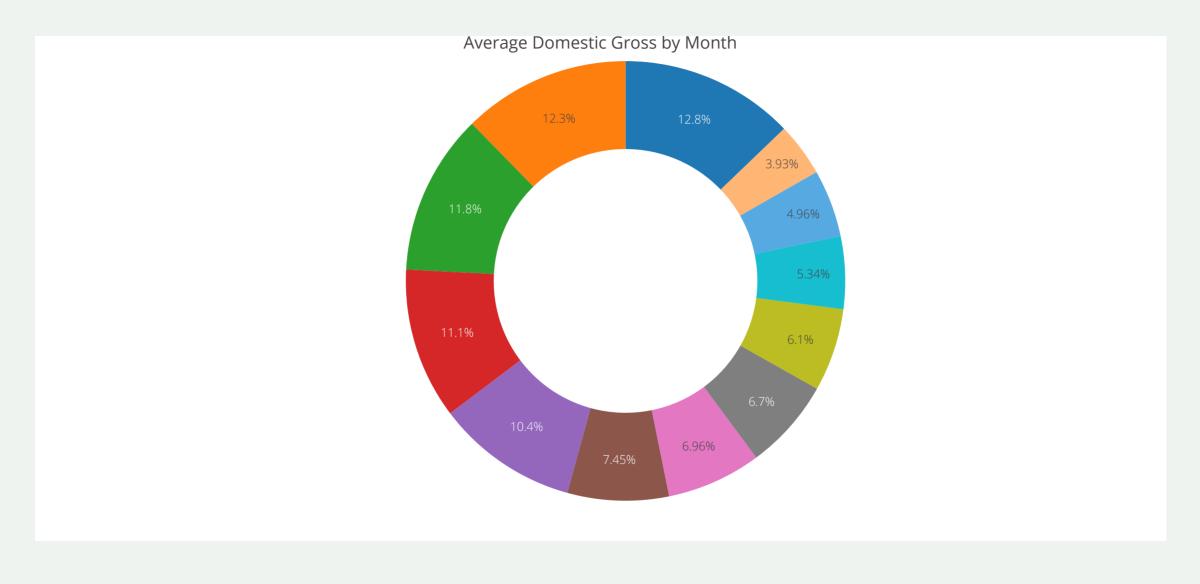
Sho	ow 8 v entries Search:								
	Symbol	Name •	Price (Intraday)	Change •	% Change	Volume •	Avg Vol (3 month)	Market Cap	PE Ratio (TTM)
1	MASI	Masimo Corporation	148.66	-80.18	-35.04%	4.84M	336,161	8.21B	36.97
2	WIX	Wix.com Ltd.	85.23	-30.53	-26.37%	3.408M	898,133	4.854B	N/A
3	RBLX	Roblox Corporation	55.13	-18.17	-24.79%	51.535M	22.686M	31.911B	N/A
4	VIAC	ViacomCBS Inc.	28.12	-7.87	-21.87%	43.836M	14.321M	18.311B	5.50
5	ANGI	Angi Inc.	7.11	-1.74	-19.69%	2.489M	1.367M	3.571B	N/A
6	SCCCF	Sunac China Holdings Limited	1.2103	-0.3097	-20.38%	36,000	0	8.672B	0.97
7	VIACA	ViacomCBS Inc.	31.27	-7.96	-20.29%	368,131	95,858	18.634B	6.11
8	SQSP	Squarespace, Inc.	27.84	-6.47	-18.86%	440,180	310,069	3.865B	N/A
Sho	Showing 1 to 8 of 235 entries Previous 1 2 3 4 5 30 Next								

Tidy further

```
df_movies %>%
  rename(ID = `...1`) %>%
  mutate(ProductionBudget = parse_number(ProductionBudget))
  mutate(DomesticGross = parse_number(DomesticGross)) %>%
  mutate(WorldwideGross = parse_number(WorldwideGross)) %>%
  mutate(ReleaseDate = mdy(ReleaseDate)) %>%
  mutate(ReleaseDate = replace_na(ReleaseDate, make_date())
  mutate(MonthOfRelease = month(ReleaseDate, label = TRUE))
  mutate(YearOfRelease = year(ReleaseDate)) %>%
  select(MonthOfRelease, DomesticGross) %>%
  group_by(MonthOfRelease) %>%
  summarize(AverageByMonth = mean(DomesticGross))
```

```
# A tibble: 12 \times 2
   MonthOfRelease AverageByMonth
                             <dbl>
   <ord>
 1 Jan
                        19586029.
 2 Feb
                        34671367.
 3 Mar
                        37145452.
 4 Apr
                        33383875.
 5 May
                        61400715.
 6 Jun
                        63916536.
 7 Jul
                        55392411.
 8 Aug
                        30375324.
 9 Sep
                        24712010.
10 Oct
                        26629138.
11 Nov
                        52033198.
12 Dec
                        59045166.
```

Interactive Donut Plot



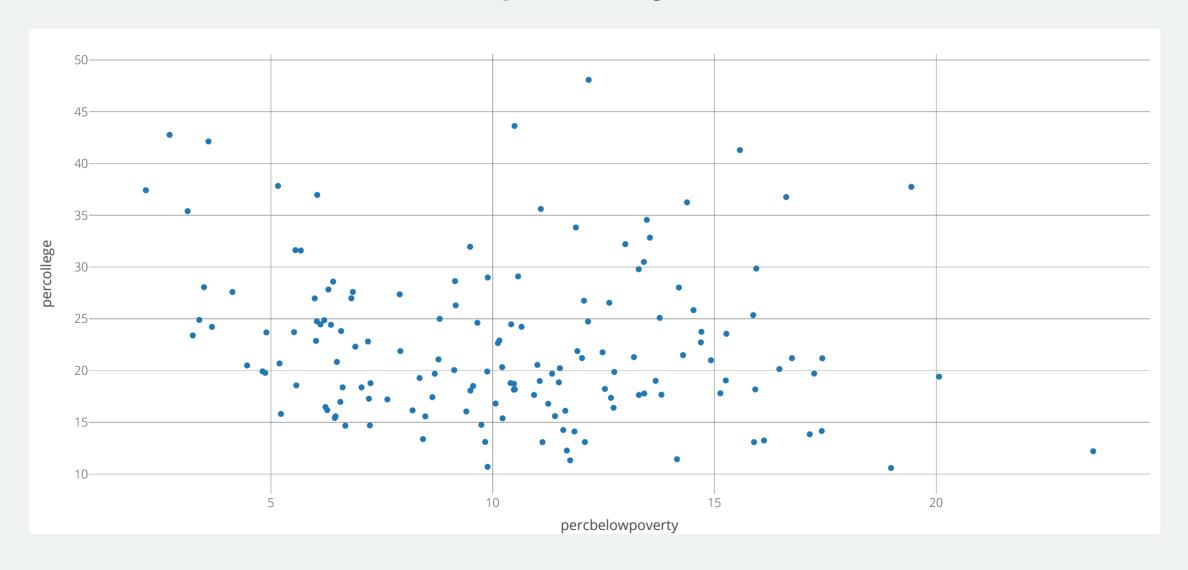
Interactive visualizations using Plotly

```
midwest %>% as tibble()
# A tibble: 437 × 28
     PID county state area poptotal popdensity popwhite popblack popamerindian
   <int> <chr> <chr> <dbl>
                                 <int>
                                             <dbl>
                                                      <int>
                                                                <int>
                                                                               <int>
     561 ADAMS
                 ΤI
                        0.052
                                 66090
                                             1271.
                                                      63917
                                                                 1702
                                                                                  98
                                 10626
                                                                 3496
     562 ALEXAN... IL
                        0.014
                                              759
                                                       7054
                                                                                  19
     563 BOND
                       0.022
                                 14991
                                              681.
                                                      14477
                                                                  429
                                                                                  35
     564 BOONE
                        0.017
                                 30806
                                             1812.
                                                      29344
                                                                  127
                                                                                  46
     565 BROWN
                        0.018
                                  5836
                                                       5264
                                              324.
                                                                  547
                                                                                  14
     566 BUREAU
                       0.05
                                 35688
                                              714.
                                                      35157
                                                                   50
                                                                                  65
     567 CALHOUN IL
                                  5322
                                              313.
                                                       5298
                       0.017
     568 CARROLL IL
                        0.027
                                 16805
                                              622.
                                                      16519
                                                                  111
                                                                                  30
     569 CASS
                       0.024
                                 13437
                                              560.
                                                      13384
                                                                   16
10
     570 CHAMPA... IL
                       0.058
                                173025
                                             2983.
                                                     146506
                                                                16559
                                                                                 331
```

```
library(plotly)

midwest %>%
  filter(inmetro == T) %>%
  plot_ly(x = ~ percbelowpoverty, y = ~ percollege) %>%
  add_markers()
```

Interactive visualizations using Plotly

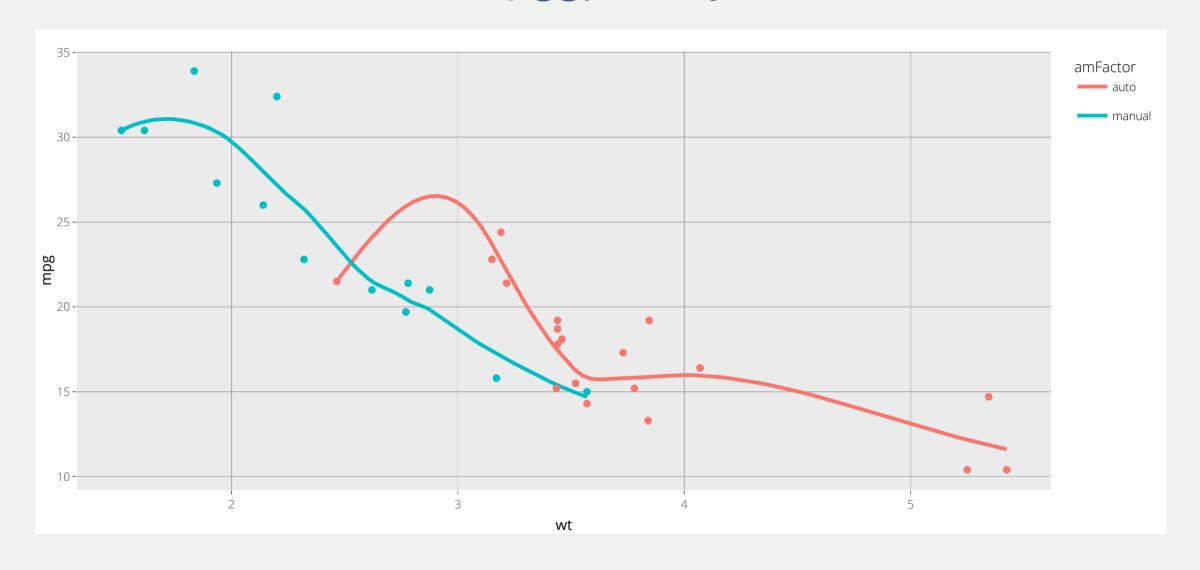


Interactive visualizations using ggplotly

```
mtcars %>% as tibble() %>% head()
 # A tibble: 6 \times 11
                                               cvl disp
                                                                                                                            hp drat wt gsec
                                                                                                                                                                                                                                                                                                                       am gear carb
                       mpg
                                                                                                                                                                                                                                                                         VS
            <dbl> 
                21
                                                                                          160
                                                                                                                             110
                                                                                                                                                         3.9
                                                                                                                                                                                             2.62 16.5
             21
                                                                                          160
                                                                                                                             110 3.9
                                                                                                                                                                                              2.88 17.0
          22.8
                                                                                          108
                                                                                                                           93 3.85
                                                                                                                                                                                             2.32 18.6
          21.4
                                                                                         258
                                                                                                                           110 3.08
                                                                                                                                                                                              3.22 19.4
          18.7 8 360
                                                                                                                           175 3.15 3.44 17.0
          18.1
                                                                      6 225
                                                                                                                                 105 2.76 3.46 20.2
```

```
ggplotly()
```

Interactive visualizations using ggplotly



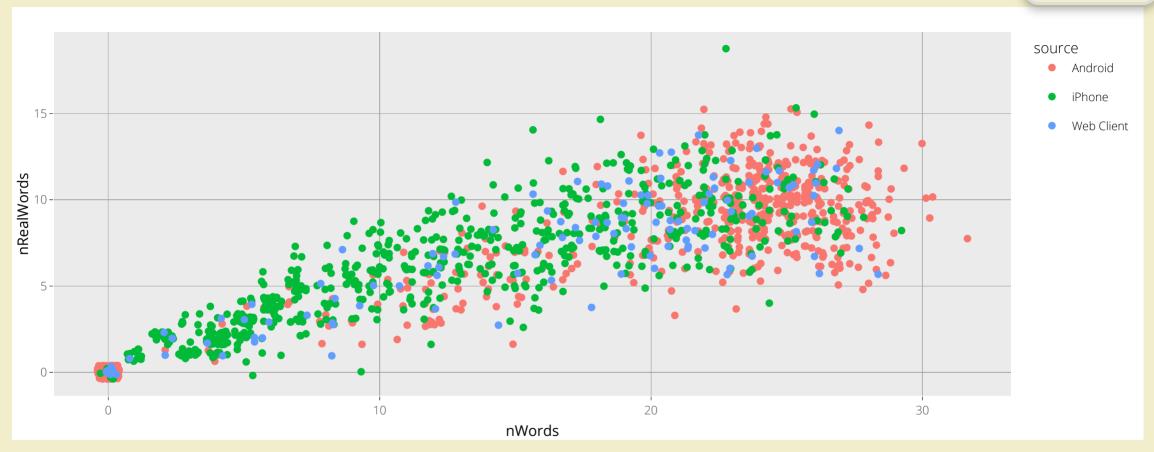
DT: Interactive Data Tables

```
library(ggplot2movies)
movies %>%
  select(1:6) %>%
  filter(rating > 8, !is.na(budget), votes > 1000) %>%
  datatable(fillContainer = FALSE, options = list(pageLength = 6))
```

Show	6 ~	• entries Search:						
		title		year	length	budget	rating	votes
1	12 An	gry Men		1957	96	340000	8.7	29278
2	2001:	A Space Odyssey		1968	156	10500000	8.3	64982
3	Adven	tures of Robin Hood, The		1938	102	1900000	8.2	7359
4	Alien			1979	116	11000000	8.3	63400
5	Aliens			1986	154	18500000	8.3	63961
6	All Qu	iet on the Western Front		1930	147	1200000	8.2	6835
Show	ing 1 to	6 of 149 entries		Previous	1 2	3 4 5	25	Next



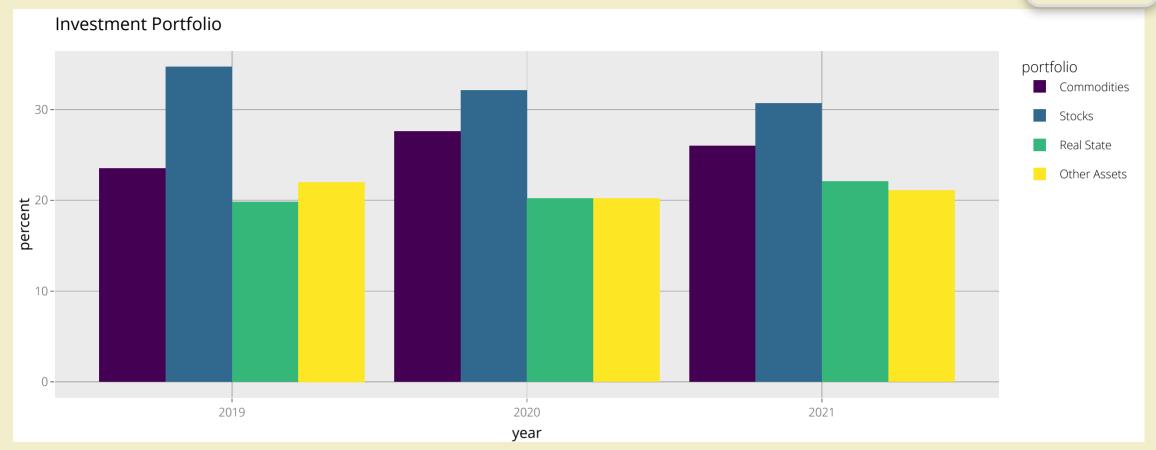




Reproduce the plot using plotly







Reproduce the plot using plotly