## lec05

## September 7, 2021

```
[1]: from datascience import *
     import numpy as np
     %matplotlib inline
     import matplotlib.pyplot as plots
     plots.style.use('fivethirtyeight')
    0.1 Creating a Table from Scratch
[2]: Table()
[2]:
[3]: streets = make_array('Bancroft', 'Durant', 'Channing', 'Haste')
[4]: streets
[4]: array(['Bancroft', 'Durant', 'Channing', 'Haste'], dtype='<U8')
[5]: Table().with_column('Street name', streets)
[5]: Street name
     Bancroft
    Durant
     Channing
    Haste
[6]: southside = Table().with_column('Street name', streets)
[7]: # creates a new table with the specified column
     southside.with_column('Blocks away from campus', np.arange(4))
[7]: Street name | Blocks away from campus
    Bancroft
                 10
    Durant
                 1
                 | 2
    Channing
    Haste
                 13
```

```
[8]: southside
 [8]: Street name
      Bancroft
      Durant
      Channing
      Haste
 [9]: southside = southside.with_column('Blocks away from campus', np.arange(4))
[10]: southside
[10]: Street name | Blocks away from campus
      Bancroft
      Durant
                  1 1
      Channing
                  1 2
      Haste
                  13
     0.2 Reading a Table from a File
[11]: minard = Table.read_table('minard.csv')
[12]: minard
[12]: Longitude | Latitude | City
                                          | Direction | Survivors
      32
                | 54.8
                            | Smolensk
                                          Advance
                                                      | 145000
      33.2
                | 54.9
                            | Dorogobouge | Advance
                                                      | 140000
      34.4
                            | Chjat
                | 55.5
                                          | Advance
                                                      | 127100
      37.6
                I 55.8
                            | Moscou
                                          | Advance
                                                      100000
      34.3
                | 55.2
                            | Wixma
                                          | Retreat
                                                      | 55000
      32
                I 54.6
                            | Smolensk
                                          | Retreat
                                                      1 24000
      30.4
                | 54.4
                            | Orscha
                                          | Retreat
                                                      | 20000
      26.8
                | 54.3
                            | Moiodexno
                                          | Retreat
                                                      | 12000
     0.3 Selecting data in a column
[13]: minard.select('Survivors')
[13]: Survivors
      145000
      140000
      127100
      100000
      55000
      24000
      20000
```

12000

[14]: minard.column('Survivors')

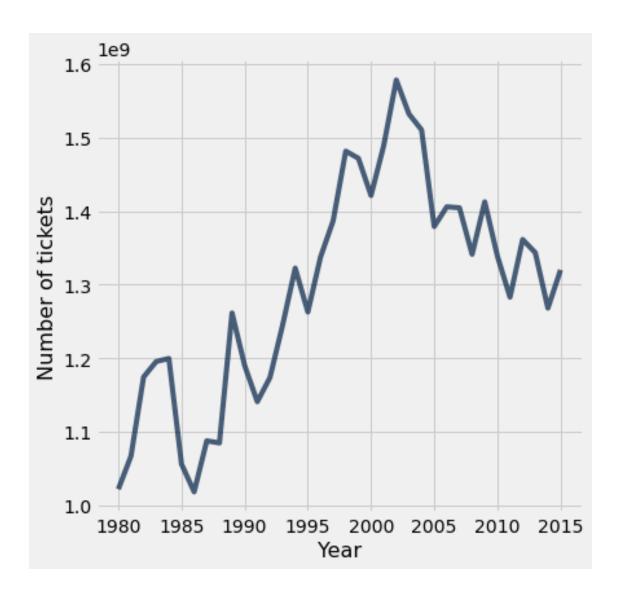
```
[14]: array([145000, 140000, 127100, 100000, 55000,
                                                      24000,
                                                              20000.
                                                                       12000])
[15]: minard.column('Survivors').item(0)
[15]: 145000
     0.4 Extending a table with a new column
[16]: initial_count = minard.column('Survivors').item(0)
[17]: proportion_surviving = minard.column('Survivors')/initial_count
[18]: minard = minard.with_column('Percent surviving', proportion_surviving)
[19]: minard
[19]: Longitude | Latitude | City
                                         | Direction | Survivors | Percent surviving
                                         | Advance
      32
                1 54.8
                           Smolensk
                                                      I 145000
                                                                  1 1
      33.2
                | 54.9
                           | Dorogobouge | Advance
                                                     | 140000
                                                                  0.965517
      34.4
                I 55.5
                           | Chjat
                                         | Advance
                                                     1 127100
                                                                  1 0.876552
                | 55.8
      37.6
                           | Moscou
                                         Advance
                                                     100000
                                                                  0.689655
      34.3
                | 55.2
                           | Wixma
                                                     | 55000
                                         | Retreat
                                                                  1 0.37931
      32
                1 54.6
                           | Smolensk
                                         | Retreat
                                                      1 24000
                                                                  0.165517
      30.4
                | 54.4
                           | Orscha
                                                      1 20000
                                                                  0.137931
                                         | Retreat
      26.8
                I 54.3
                           | Moiodexno
                                         | Retreat
                                                     l 12000
                                                                  0.0827586
[20]: minard.set_format('Percent surviving', PercentFormatter)
[20]: Longitude | Latitude | City
                                         | Direction | Survivors | Percent surviving
      32
                | 54.8
                           Smolensk
                                         Advance
                                                      | 145000
                                                                  | 100.00%
      33.2
                1 54.9
                           | Dorogobouge | Advance
                                                      1 140000
                                                                  1 96.55%
      34.4
                I 55.5
                           | Chjat
                                         Advance
                                                     1 127100
                                                                  1 87.66%
                           l Moscou
                                                     100000
      37.6
                | 55.8
                                         Advance
                                                                  1 68.97%
      34.3
                | 55.2
                           | Wixma
                                         | Retreat
                                                     l 55000
                                                                  1 37.93%
      32
                | 54.6
                                                                  | 16.55%
                           Smolensk
                                         | Retreat
                                                     1 24000
      30.4
                | 54.4
                           | Orscha
                                         | Retreat
                                                      1 20000
                                                                  | 13.79%
      26.8
                | 54.3
                           | Moiodexno
                                                                  | 8.28%
                                         | Retreat
                                                      1 12000
          Working with Columns
     0.5
[21]: movies = Table.read_table('movies_by_year_with_ticket_price.csv')
```

```
[22]: movies.show()
     <IPython.core.display.HTML object>
[23]: movies.labels
[23]: ('Year', 'Average Ticket Price', 'Total Gross', 'Number of Movies', '#1 Movie')
[24]: movies.num_rows
[24]: 36
[25]: number_of_tix = movies.column('Total Gross') * (10 ** 6) / movies.
      [26]: movies = movies.with_column('Number of tickets', number_of_tix)
[27]: movies
[27]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie
     | Number of tickets
     2015 | 8.43
                                 | 11128.5
                                               I 702
                                                                  | Star Wars: The
     Force Awakens
                          | 1.32011e+09
     2014 | 8.17
                                 1 10360.8
                                                                  | American Sniper
                                               702
     | 1.26815e+09
     2013 | 8.13
                                 | 10923.6
                                               | 688
                                                                  | Catching Fire
     | 1.34362e+09
     2012 | 7.96
                                                                  | The Avengers
                                 | 10837.4
                                               I 667
     | 1.36148e+09
     2011 | 7.93
                                 | 10174.3
                                               | 602
                                                                  | Harry Potter /
     Deathly Hallows (P2) | 1.28301e+09
     2010 | 7.89
                                                                  | Toy Story 3
                                 | 10565.6
                                               | 536
     | 1.33911e+09
     2009 | 7.5
                                 10595.5
                                               | 521
                                                                  | Avatar
     | 1.41273e+09
     2008 | 7.18
                                 | 9630.7
                                               | 608
                                                                  | The Dark Knight
     1.34132e+09
     2007 | 6.88
                                 9663.8
                                               631
                                                                  | Spider-Man 3
     I 1.40462e+09
     2006 | 6.55
                                 | 9209.5
                                                                  | Dead Man's Chest
                                               608
     | 1.40603e+09
     ... (26 rows omitted)
[28]: movies.set_format(5, NumberFormatter)
```

[28]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie | Number of tickets

2015   8.43	11128.5	702	Star Wars: The
Force Awakens	1,320,106,761.57		
2014   8.17	10360.8	702	American Sniper
1,268,151,774.79			
2013   8.13	10923.6	688	Catching Fire
1,343,616,236.16			
2012   7.96	10837.4	667	The Avengers
1,361,482,412.06			_
2011   7.93	10174.3	602	Harry Potter /
Deathly Hallows (P2)	1,283,013,871.37		,
2010   7.89	10565.6	536	Toy Story 3
1,339,112,801.01			
2009   7.5	10595.5	521	Avatar
1,412,733,333.33			
2008   7.18	9630.7	608	The Dark Knight
1,341,323,119.78			_
2007   6.88	9663.8	631	Spider-Man 3
1,404,622,093.02			-
2006   6.55	9209.5	608	Dead Man's Chest
1,406,030,534.35			
(26 rows omitted)			
(20 10 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0			

[29]: movies.plot('Year', 'Number of tickets')



## 0.6 Rows

[30]: movies.where('Year', are.between(2000, 2005)) [30]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie | Number of tickets | Shrek 2 2004 | 6.21 | 9380.5 | 551 | 1,510,547,504.03 2003 | 6.03 | 9239.7 | 506 | Return of the | 1,532,288,557.21 King | Spider-Man 2002 | 5.8 9155 | 479 | 1,578,448,275.86 | Harry Potter / 2001 | 5.65 | 8412.5 | 482 Sorcerer's Stone | 1,488,938,053.10

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2000 | 5.39
                                  | 7661
                                           | 478
                                                                   | The Grinch
      | 1,421,335,807.05
[31]: movies.where('#1 Movie', are.equal_to('Avatar'))
[31]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie | Number
     of tickets
      2009 | 7.5
                                  | 10595.5
                                                | 521
                                                                   | Avatar
                                                                              Τ
      1,412,733,333.33
[32]: movies.where('#1 Movie', 'Avatar')
[32]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie | Number
      of tickets
      2009 | 7.5
                                  I 10595.5
                                                l 521
                                                                   | Avatar
      1,412,733,333.33
[33]: movies.where('#1 Movie', are.containing('Harry Potter'))
[33]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie
      | Number of tickets
     2011 | 7.93
                                                602
                                                                   | Harry Potter /
                                  | 10174.3
     Deathly Hallows (P2) | 1,283,013,871.37
      2001 | 5.65
                                  8412.5
                                                482
                                                                   | Harry Potter /
      Sorcerer's Stone
                           | 1,488,938,053.10
[34]: movies.where('Number of Movies', are.below(450))
[34]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie
      | Number of tickets
      1995 | 4.35
                                  I 5493.5
                                                                   | Toy Story
                                                | 411
      1,262,873,563.22
                                               | 410
      1990 | 4.22
                                  | 5021.8
                                                                   | Home Alone
      | 1,190,000,000.00
      1982 | 2.94
                                  3453
                                                | 428
                                                                   I E.T.
      | 1,174,489,795.92
      1981 | 2.78
                                                                   | Raiders / Lost
                                  1 2966
                                                | 173
      Ark
               1,066,906,474.82
                                                                   | The Empire
      1980 | 2.69
                                  | 2749
                                                | 161
      Strikes Back | 1,021,933,085.50
[35]: movies.where('Year', are.above(2010))
[35]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie
      | Number of tickets
      2015 | 8.43
                                  | 11128.5
                                                | 702
                                                                   | Star Wars: The
      Force Awakens
                           | 1,320,106,761.57
```

```
2014 | 8.17
                                  | 10360.8
                                                 702
                                                                    | American Sniper
      | 1,268,151,774.79
      2013 | 8.13
                                  | 10923.6
                                                 | 688
                                                                    | Catching Fire
      | 1,343,616,236.16
      2012 | 7.96
                                  | 10837.4
                                                 | 667
                                                                    | The Avengers
      | 1,361,482,412.06
                                                                    | Harry Potter /
      2011 | 7.93
                                  | 10174.3
                                                 602
      Deathly Hallows (P2) | 1,283,013,871.37
[36]: movies.take(3)
[36]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie
      Number of tickets
                                  | 10837.4
                                                                    | The Avengers |
      2012 | 7.96
                                                 | 667
      1,361,482,412.06
[37]: movies.take(np.arange(4))
[37]: Year | Average Ticket Price | Total Gross | Number of Movies | #1 Movie
      | Number of tickets
      2015 | 8.43
                                  | 11128.5
                                                 | 702
                                                                    | Star Wars: The
      Force Awakens | 1,320,106,761.57
      2014 | 8.17
                                  | 10360.8
                                                 702
                                                                    | American Sniper
      | 1,268,151,774.79
      2013 | 8.13
                                  | 10923.6
                                                                    | Catching Fire
                                                 | 688
      | 1,343,616,236.16
      2012 | 7.96
                                                                    | The Avengers
                                  | 10837.4
                                                 | 667
      | 1,361,482,412.06
 []:
```