

Midterm 1

March 10, 2022

```
[235]: from datascience import *
import numpy as np
import warnings
warnings.filterwarnings("ignore")

import matplotlib.pyplot as plt
%matplotlib inline
import matplotlib.pyplot as plots
plots.style.use('fivethirtyeight')
plots.rcParams["patch.force_edgecolor"] = True
```

1 Question 1

```
[236]: food = Table.read_table('caloriesMcD.csv')
food
```

```
[236]: Category | Item |
Calories | Calories from Fat | Total Fat | Saturated Fat | Saturated Fat (%)
Daily Value) | Trans Fat | Cholesterol | Sodium | Carbohydrates | Dietary Fiber
| Sugars | Protein
Breakfast | Egg McMuffin | 300
| 120 | 13 | 5 | 25
| 0 | 260 | 750 | 31 | 4 | 3 | 17
Breakfast | Egg White Delight | 250
| 70 | 8 | 3 | 15
| 0 | 25 | 770 | 30 | 4 | 3 | 18
Breakfast | Sausage McMuffin | 370
| 200 | 23 | 8 | 42
| 0 | 45 | 780 | 29 | 4 | 2 | 14
Breakfast | Sausage McMuffin with Egg | 450
| 250 | 28 | 10 | 52
| 0 | 285 | 860 | 30 | 4 | 2 | 21
Breakfast | Sausage McMuffin with Egg Whites | 400
| 210 | 23 | 8 | 42
| 0 | 50 | 880 | 30 | 4 | 2 | 21
Breakfast | Steak & Egg McMuffin | 430
| 210 | 23 | 9 | 46
```

	1	300	960	31	4	3	26
Breakfast	Bacon, Egg & Cheese Biscuit (Regular Biscuit)						460
	230	26	13	65			
	0	250	1300	38	2	3	19
Breakfast	Bacon, Egg & Cheese Biscuit (Large Biscuit)						520
	270	30	14	68			
	0	250	1410	43	3	4	19
Breakfast	Bacon, Egg & Cheese Biscuit with Egg Whites (Regular Bis ...						410
	180	20	11	56			
	0	35	1300	36	2	3	20
Breakfast	Bacon, Egg & Cheese Biscuit with Egg Whites (Large Biscuit)						470
	220	25	12	59			
	0	35	1420	42	3	4	20

... (250 rows omitted)

```
[237]: #A
sum(food.column('Calories'))
```

[237]: 95750

```
[238]: #B
fat = food.where('Calories from Fat',are.above(200))
fat
```

```
[238]: Category | Item
Calories | Calories from Fat | Total Fat | Saturated Fat | Saturated Fat (%)
Daily Value | Trans Fat | Cholesterol | Sodium | Carbohydrates | Dietary Fiber
| Sugars | Protein
Breakfast | Sausage McMuffin with Egg | 450
| 250 | 28 | 10 | 52
| 0 | 285 | 860 | 30 | 4 | 2 | 21
Breakfast | Sausage McMuffin with Egg Whites | 400
| 210 | 23 | 8 | 42
| 0 | 50 | 880 | 30 | 4 | 2 | 21
Breakfast | Steak & Egg McMuffin | 430
| 210 | 23 | 9 | 46
| 1 | 300 | 960 | 31 | 4 | 3 | 26
Breakfast | Bacon, Egg & Cheese Biscuit (Regular Biscuit) | 460
| 230 | 26 | 13 | 65
| 0 | 250 | 1300 | 38 | 2 | 3 | 19
Breakfast | Bacon, Egg & Cheese Biscuit (Large Biscuit) | 520
| 270 | 30 | 14 | 68
| 0 | 250 | 1410 | 43 | 3 | 4 | 19
Breakfast | Bacon, Egg & Cheese Biscuit with Egg Whites (Large Biscuit) | 470
| 220 | 25 | 12 | 59
| 0 | 35 | 1420 | 42 | 3 | 4 | 20
Breakfast | Sausage Biscuit (Regular Biscuit) | 430
```

	240		27		12		62			
	0		30		1080		34		2	2 11
Breakfast	Sausage Biscuit (Large Biscuit)									480
	280		31		13		65			
	0		30		1190		39		3	3 11
Breakfast	Sausage Biscuit with Egg (Regular Biscuit)									510
	290		33		14		71			
	0		250		1170		36		2	2 18
Breakfast	Sausage Biscuit with Egg (Large Biscuit)									570
	330		37		15		74			
	0		250		1280		42		3	3 18

... (50 rows omitted)

```
[239]: #C
sugar = food.sort("Sugars", descending = True)
sugar
```

```
[239]: Category | Item |
Calories | Calories from Fat | Total Fat | Saturated Fat | Saturated Fat (%)
Daily Value) | Trans Fat | Cholesterol | Sodium | Carbohydrates | Dietary Fiber
| Sugars | Protein
Smoothies & Shakes | McFlurry with M&M's Candies (Medium) | 930
| 290 | 33 | 20 | 102
| 1 | 75 | 260 | 139 | 2 | 128 | 20
Smoothies & Shakes | Strawberry Shake (Large) | 850
| 210 | 24 | 15 | 75
| 1 | 90 | 260 | 140 | 0 | 123 | 18
Smoothies & Shakes | Chocolate Shake (Large) | 850
| 210 | 23 | 15 | 74
| 1 | 85 | 380 | 141 | 2 | 120 | 19
Smoothies & Shakes | Shamrock Shake (Large) | 820
| 210 | 23 | 15 | 73
| 1 | 90 | 260 | 135 | 0 | 115 | 18
Smoothies & Shakes | McFlurry with Reese's Peanut Butter Cups (Medium) | 810
| 290 | 32 | 15 | 76
| 1 | 60 | 400 | 114 | 2 | 103 | 21
Smoothies & Shakes | Vanilla Shake (Large) | 820
| 210 | 23 | 15 | 73
| 1 | 90 | 260 | 135 | 0 | 101 | 18
Smoothies & Shakes | Strawberry Shake (Medium) | 690
| 180 | 20 | 13 | 63
| 1 | 75 | 210 | 114 | 0 | 100 | 15
Coffee & Tea | Frappé Chocolate Chip (Large) | 760
| 280 | 31 | 20 | 101
| 1.5 | 95 | 200 | 111 | 1 | 99 | 12
Smoothies & Shakes | Chocolate Shake (Medium) | 700
| 180 | 20 | 12 | 62
```

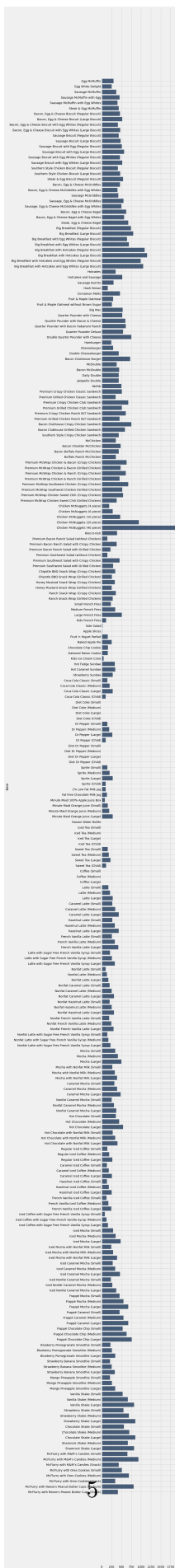
1	75	300	114	2	97	15
Smoothies & Shakes	Shamrock Shake (Medium)				660	
170	19	12	61			
1	75	210	109	0	93	14

... (250 rows omitted)

```
[240]: #D
a = food.group('Category', np.max).select(0,3)
a
```

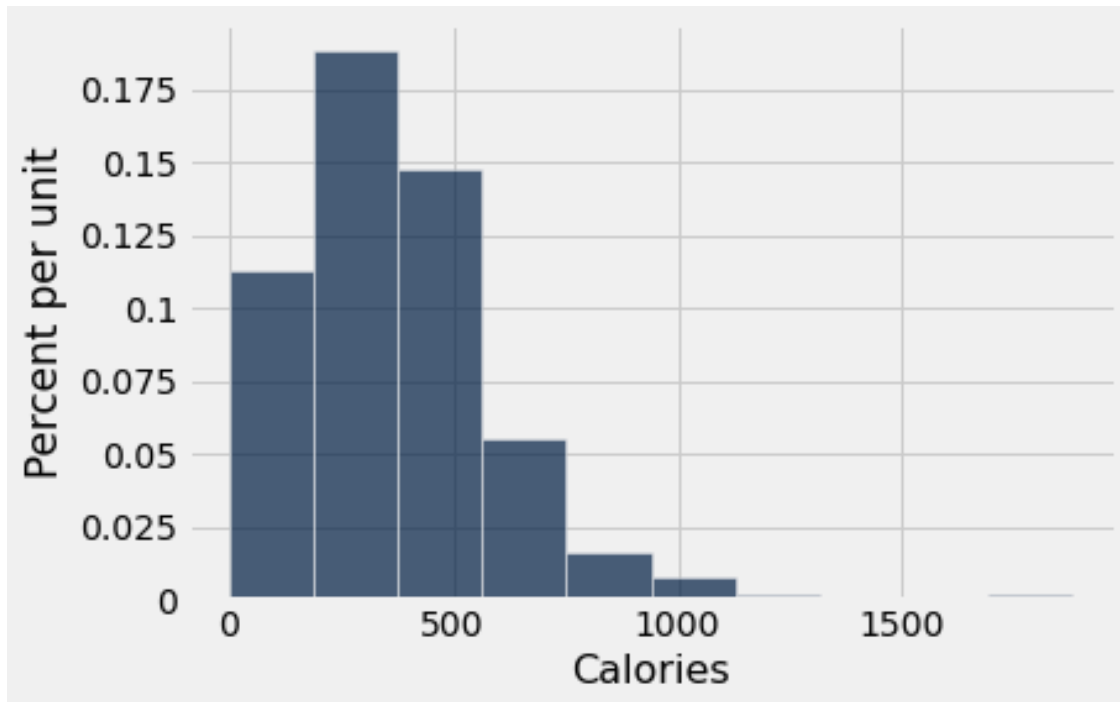
```
[240]: Category      | Calories from Fat amax
Beef & Pork        | 380
Beverages          | 20
Breakfast          | 540
Chicken & Fish     | 1060
Coffee & Tea       | 280
Desserts           | 110
Salads             | 190
Smoothies & Shakes | 290
Snacks & Sides     | 220
```

```
[241]: #E
food.barh('Item', 'Calories')
```



Chicken McNuggets(40 piece) has the most calories and the diet sodas and ice tea have no calories

```
[242]: food.hist(2)
```



```
[243]: #F
category_item = food.select('Category', 'Item')
```

```
[244]: item_distribution = category_item.group('Item')
item_distribution
```

```
[244]: Item | count
1% Low Fat Milk Jug | 1
Apple Slices | 1
Bacon Buffalo Ranch McChicken | 1
Bacon Cheddar McChicken | 1
Bacon Clubhouse Burger | 1
Bacon Clubhouse Crispy Chicken Sandwich | 1
Bacon Clubhouse Grilled Chicken Sandwich | 1
Bacon McDougle | 1
Bacon, Egg & Cheese Bagel | 1
Bacon, Egg & Cheese Bagel with Egg Whites | 1
... (250 rows omitted)
```

```
[245]: sum(item_distribution.column('count'))
```

```
[245]: 260
```

```
[246]: item_distribution.sort('count', descending=True).barh('Item')
```

[illegible]


```
[ ]:
```

2 Question 2

```
[339]: comp = Table.read_table('raw_compensation.csv')
comp
```

```
[339]: Rank | Name | Company (Headquarters) | Total Pay | %
Change | Cash Pay | Equity Pay | Other Pay | Ratio of CEO pay to
average industry worker pay
1 | Mark V. Hurd* | Oracle (Redwood City) | $53.25 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
2 | Safra A. Catz* | Oracle (Redwood City) | $53.24 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
3 | Robert A. Iger | Walt Disney (Burbank) | $44.91 | -3%
| $24.89 | $17.28 | $2.74 | 477
4 | Marissa A. Mayer | Yahoo! (Sunnyvale) | $35.98 | -15%
| $1.00 | $34.43 | $0.55 | 342
5 | Marc Benioff | salesforce.com (San Francisco) | $33.36 | -16%
| $4.65 | $27.26 | $1.45 | 338
6 | John H. Hammergren | McKesson (San Francisco) | $24.84 | -4%
| $12.10 | $12.37 | $0.37 | 222
7 | John S. Watson | Chevron (San Ramon) | $22.04 | -15%
| $4.31 | $14.68 | $3.05 | 183
8 | Jeffrey Weiner | LinkedIn (Mountain View) | $19.86 | 27%
| $2.47 | $17.26 | $0.13 | 182
9 | John T. Chambers** | Cisco Systems (San Jose) | $19.62 | 19%
| $5.10 | $14.51 | $0.01 | 170
10 | John G. Stumpf | Wells Fargo (San Francisco) | $19.32 | -10%
| $6.80 | $12.50 | $0.02 | 256
... (92 rows omitted)
```

```
[340]: #A
comp5 = Table.read_table('raw_compensation.csv').show(5)
comp5
```

<IPython.core.display.HTML object>

```
[341]: #B
def string_num(Total_Pay):
    return int(float(Total_Pay.strip('$')))
```

```
[342]: string_num('$2')
```

[342]: 2

```
[343]: Total_Pay = comp.apply(string_num,"Total Pay")
Total_Pay
```

```
[343]: array([53, 53, 44, 35, 33, 24, 22, 19, 19, 19, 18, 18, 18, 18, 17, 16, 16,
        16, 16, 16, 15, 14, 14, 14, 14, 14, 14, 14, 13, 12, 12, 12, 12,
        12, 11, 11, 11, 11, 11, 11, 10, 10, 10, 10, 10, 10, 10, 10, 9, 9,
        9, 9, 9, 9, 9, 9, 8, 8, 8, 8, 8, 8, 8, 7, 7, 7,
        7, 7, 7, 7, 6, 6, 6, 6, 6, 5, 5, 5, 5, 5, 5, 4, 4,
        4, 4, 4, 3, 3, 2, 2, 2, 2, 1, 1, 1, 0, 0, 0, 0, 0])
```

```
[344]: #C
comp = comp.with_column("Total_Pay", Total_Pay)
comp
```

```
[344]: Rank | Name | Company (Headquarters) | Total Pay | %
Change | Cash Pay | Equity Pay | Other Pay | Ratio of CEO pay to
average industry worker pay | Total_Pay
1 | Mark V. Hurd* | Oracle (Redwood City) | $53.25 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
| 53
2 | Safra A. Catz* | Oracle (Redwood City) | $53.24 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
| 53
3 | Robert A. Iger | Walt Disney (Burbank) | $44.91 | -3%
| $24.89 | $17.28 | $2.74 | 477
| 44
4 | Marissa A. Mayer | Yahoo! (Sunnyvale) | $35.98 | -15%
| $1.00 | $34.43 | $0.55 | 342
| 35
5 | Marc Benioff | salesforce.com (San Francisco) | $33.36 | -16%
| $4.65 | $27.26 | $1.45 | 338
| 33
6 | John H. Hammergren | McKesson (San Francisco) | $24.84 | -4%
| $12.10 | $12.37 | $0.37 | 222
| 24
7 | John S. Watson | Chevron (San Ramon) | $22.04 | -15%
| $4.31 | $14.68 | $3.05 | 183
| 22
8 | Jeffrey Weiner | LinkedIn (Mountain View) | $19.86 | 27%
| $2.47 | $17.26 | $0.13 | 182
| 19
9 | John T. Chambers** | Cisco Systems (San Jose) | $19.62 | 19%
| $5.10 | $14.51 | $0.01 | 170
| 19
10 | John G. Stumpf | Wells Fargo (San Francisco) | $19.32 | -10%
```

```
| $6.80      | $12.50      | $0.02      | 256
| 19
... (92 rows omitted)
```

```
[345]: compnew = comp.with_column("Total Pay", comp.apply(string_num,"Total Pay"))
compnew
```

```
[345]: Rank | Name | Company (Headquarters) | Total Pay | %
Change | Cash Pay | Equity Pay | Other Pay | Ratio of CEO pay to
average industry worker pay | Total_Pay
1 | Mark V. Hurd* | Oracle (Redwood City) | 53 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
| 53
2 | Safra A. Catz* | Oracle (Redwood City) | 53 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
| 53
3 | Robert A. Iger | Walt Disney (Burbank) | 44 | -3%
| $24.89 | $17.28 | $2.74 | 477
| 44
4 | Marissa A. Mayer | Yahoo! (Sunnyvale) | 35 | -15%
| $1.00 | $34.43 | $0.55 | 342
| 35
5 | Marc Benioff | salesforce.com (San Francisco) | 33 | -16%
| $4.65 | $27.26 | $1.45 | 338
| 33
6 | John H. Hammergren | McKesson (San Francisco) | 24 | -4%
| $12.10 | $12.37 | $0.37 | 222
| 24
7 | John S. Watson | Chevron (San Ramon) | 22 | -15%
| $4.31 | $14.68 | $3.05 | 183
| 22
8 | Jeffrey Weiner | LinkedIn (Mountain View) | 19 | 27%
| $2.47 | $17.26 | $0.13 | 182
| 19
9 | John T. Chambers** | Cisco Systems (San Jose) | 19 | 19%
| $5.10 | $14.51 | $0.01 | 170
| 19
10 | John G. Stumpf | Wells Fargo (San Francisco) | 19 | -10%
| $6.80 | $12.50 | $0.02 | 256
| 19
... (92 rows omitted)
```

```
[346]: #D
average_total_pay = np.average(comp.column("Total_Pay"))
average_total_pay
```

```
[346]: 10.970588235294118
```

Because of the \$ sign being a string it wont let us compute the average.

```
[364]: #E
CEOavg = comp.sort(Total_Pay,are.above(15)).show(21)
CEOavg
```

<IPython.core.display.HTML object>

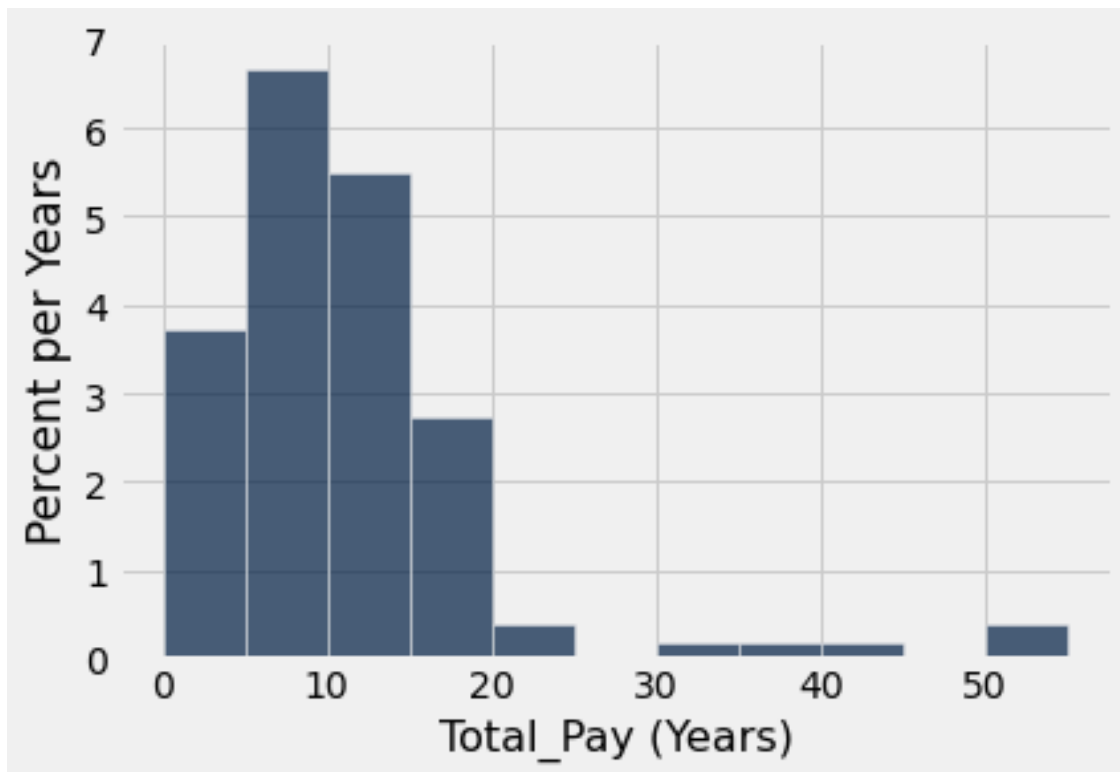
21 CEO made more then 15 million

```
[365]: #F
my_bins = make_array(0, 5, 10, 15, 20, 25, 30, 35, 40, 45, 50, 55)
```

```
[366]: comp.bin('Total_Pay', bins = my_bins)
```

```
[366]: bin | Total_Pay count
0      | 19
5      | 34
10     | 28
15     | 14
20     | 2
25     | 0
30     | 1
35     | 1
40     | 1
45     | 0
... (2 rows omitted)
```

```
[367]: comp.hist('Total_Pay', bins = my_bins, unit = 'Years')
```



I would say the histogram is skewed to the left since most of the data is in the beginning of the histogram

```
[387]: #G
(50-30)*0.25
```

```
[387]: 5.0
```

```
[388]: #H
comp.where('Total_Pay', are.between(30, 50)).num_rows
```

```
[388]: 3
```

```
[389]: comp.num_rows
```

```
[389]: 102
```

```
[390]: 3*100/(102)
```

```
[390]: 2.9411764705882355
```

```
[351]: comp.where(Total_Pay,are.above(30))
```

```
[351]: Rank | Name | Company (Headquarters) | Total Pay | % Change
| Cash Pay | Equity Pay | Other Pay | Ratio of CEO pay to average industry
worker pay | Total_Pay
1 | Mark V. Hurd* | Oracle (Redwood City) | $53.25 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
| 53
2 | Safra A. Catz* | Oracle (Redwood City) | $53.24 | (No
previous year) | $0.95 | $52.27 | $0.02 | 362
| 53
3 | Robert A. Iger | Walt Disney (Burbank) | $44.91 | -3%
| $24.89 | $17.28 | $2.74 | 477
| 44
4 | Marissa A. Mayer | Yahoo! (Sunnyvale) | $35.98 | -15%
| $1.00 | $34.43 | $0.55 | 342
| 35
5 | Marc Benioff | salesforce.com (San Francisco) | $33.36 | -16%
| $4.65 | $27.26 | $1.45 | 338
| 33
```

```
[352]: #I
type(comp.column("Cash Pay")[0])
```

```
[352]: numpy.str_
```

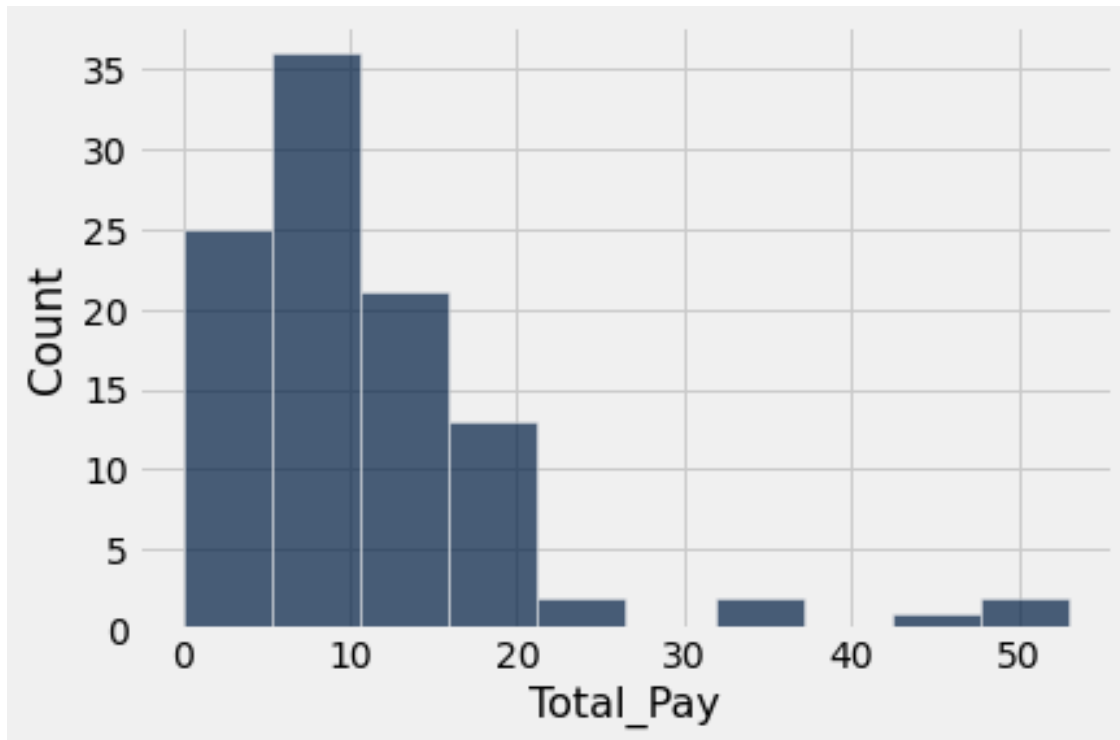
```
[353]: cash_proportion = comp.apply(string_num, "Cash Pay") / comp.column("Total_Pay")
cash_proportion
```

```
[353]: array([0.          , 0.          , 0.54545455, 0.02857143, 0.12121212,
0.5          , 0.18181818, 0.10526316, 0.26315789, 0.31578947,
0.27777778, 0.22222222, 0.11111111, 0.16666667, 0.17647059,
0.0625       , 0.125       , 0.25       , 0.3125       , 0.125       ,
0.26666667, 0.21428571, 0.28571429, 0.21428571, 0.21428571,
0.14285714, 0.28571429, 0.21428571, 0.21428571, 0.15384615,
0.33333333, 0.25        , 0.16666667, 0.41666667, 0.41666667,
0.36363636, 0.18181818, 0.18181818, 0.36363636, 0.18181818,
0.18181818, 0.1        , 0.3        , 0.1        , 0.2        ,
1.          , 0.2        , 0.2        , 0.3        , 0.22222222,
0.77777778, 0.33333333, 0.22222222, 0.22222222, 0.11111111,
0.22222222, 0.33333333, 0.33333333, 0.5          , 0.125       ,
0.375       , 0.25       , 0.25       , 0.125       , 0.375       ,
0.28571429, 0.42857143, 0.28571429, 0.14285714, 1.          ,
0.14285714, 0.85714286, 0.33333333, 0.83333333, 0.16666667,
0.16666667, 0.16666667, 0.4          , 0.2          , 0.2          ,
0.6          , 0.2          , 0.          , 0.5          , 0.5          ,
0.25         , 0.          , 0.25         , 0.33333333, 0.33333333,
1.          , 0.5         , 0.5         , 0.5         , 0.          ,
1.          , 1.          ,          nan,          nan,          nan,
```

```
nan, nan])
```

```
[354]: def percent_string_to_num(percent_string): return float(percent_string.  
↳ strip("%"))
```

```
[355]: comp.hist("Total_Pay", normed= False)
```



3 Question 3

```
[415]: imdb = Table.read_table('imdb.csv')  
imdb
```

```
[415]: Votes | Rating | Title | Year | Decade  
88355 | 8.4 | M | 1931 | 1930  
132823 | 8.3 | Singin' in the Rain | 1952 | 1950  
74178 | 8.3 | All About Eve | 1950 | 1950  
635139 | 8.6 | Léon | 1994 | 1990  
145514 | 8.2 | The Elephant Man | 1980 | 1980  
425461 | 8.3 | Full Metal Jacket | 1987 | 1980  
441174 | 8.1 | Gone Girl | 2014 | 2010  
850601 | 8.3 | Batman Begins | 2005 | 2000  
37664 | 8.2 | Judgment at Nuremberg | 1961 | 1960  
46987 | 8 | Relatos salvajes | 2014 | 2010
```

... (240 rows omitted)

```
[229]: #A
forties = imdb.where('Decade', are.equal_to(1940))
forties
```

```
[229]: Votes | Rating | Title | Year | Decade
55793 | 8.1 | The Grapes of Wrath | 1940 | 1940
86715 | 8.3 | Double Indemnity | 1944 | 1940
101754 | 8.1 | The Maltese Falcon | 1941 | 1940
71003 | 8.3 | The Treasure of the Sierra Madre | 1948 | 1940
35983 | 8.1 | The Best Years of Our Lives | 1946 | 1940
81887 | 8.3 | Ladri di biciclette | 1948 | 1940
66622 | 8 | Notorious | 1946 | 1940
350551 | 8.5 | Casablanca | 1942 | 1940
59578 | 8 | The Big Sleep | 1946 | 1940
78216 | 8.2 | Rebecca | 1940 | 1940
... (4 rows omitted)
```

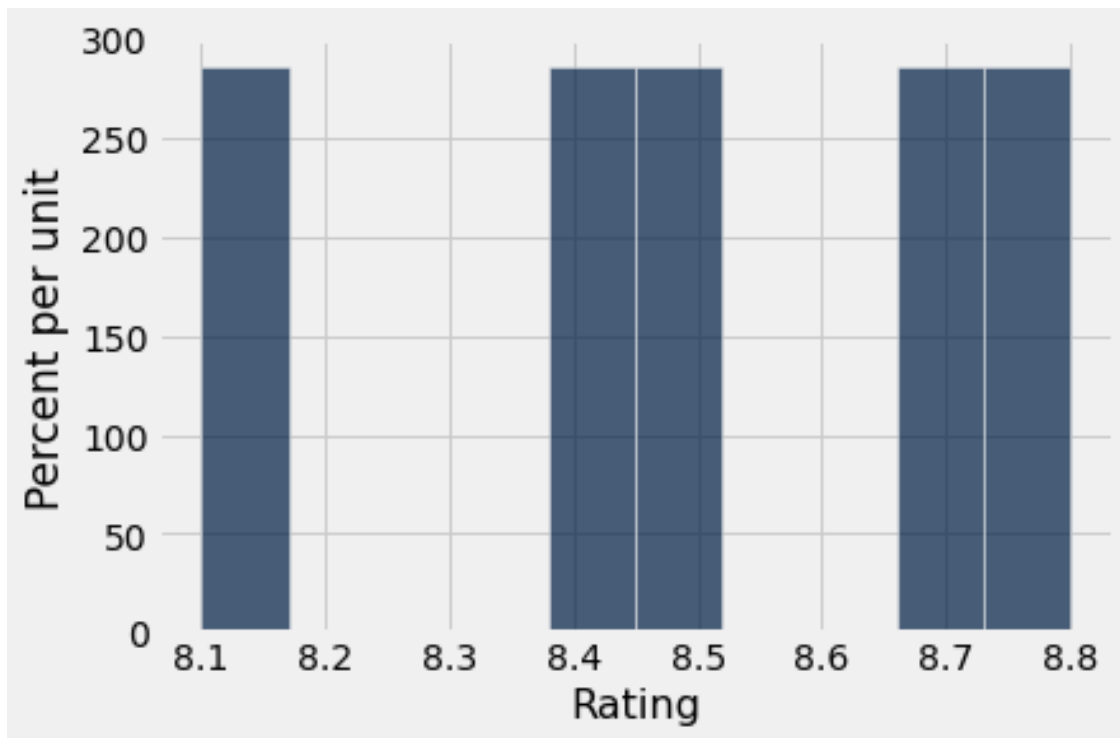
```
[230]: #B
average_rating_in_forties = np.average(forties.column("Rating"))
average_rating_in_forties
```

```
[230]: 8.257142857142856
```

```
[231]: #C
ninety_nine = imdb.where('Year', are.equal_to(1999))
ninety_nine
```

```
[231]: Votes | Rating | Title | Year | Decade
1177098 | 8.8 | Fight Club | 1999 | 1990
735056 | 8.4 | American Beauty | 1999 | 1990
630994 | 8.1 | The Sixth Sense | 1999 | 1990
1073043 | 8.7 | The Matrix | 1999 | 1990
672878 | 8.5 | The Green Mile | 1999 | 1990
```

```
[232]: ninety_nine.hist('Rating')
```

yes but we need frequencies of rating if we are to make a bar chart for Ratings

```
[233]: #D
really_highly_rated = imdb.where('Rating', are.above(8.5))
really_highly_rated
```

```
[233]: Votes    | Rating | Title                                     | Year | Decade
635139   | 8.6    | Léon                                     | 1994 | 1990
1027398  | 9.2    | The Godfather                           | 1972 | 1970
767224   | 8.6    | The Silence of the Lambs                | 1991 | 1990
1498733  | 9.2    | The Shawshank Redemption                | 1994 | 1990
447875   | 8.9    | Il buono, il brutto, il cattivo (1966) | 1966 | 1960
967389   | 8.7    | The Lord of the Rings: The Two Towers   | 2002 | 2000
689541   | 8.6    | Interstellar                            | 2014 | 2010
1473049  | 8.9    | The Dark Knight                         | 2008 | 2000
192206   | 8.6    | C'era una volta il West                 | 1968 | 1960
1271949  | 8.7    | Inception                              | 2010 | 2010
... (19 rows omitted)
```

```
[394]: #E
rating = imdb.column("Rating")
rating
```

```
[394]: array([8.4, 8.3, 8.3, 8.6, 8.2, 8.3, 8.1, 8.3, 8.2, 8. , 8.1, 8.2, 8.3,
            8.3, 8.1, 8.4, 8.5, 8.2, 8.1, 8.4, 8.1, 8.1, 9.2, 8. , 8.2, 8.1,
            8.2, 8.5, 8. , 8.3, 8.1, 8. , 8. , 8.3, 8.1, 8. , 8. , 8.3, 8.4,
            8.1, 8.1, 8.5, 8.5, 8. , 8.3, 8.1, 8. , 8.6, 8.5, 8.3, 8.3, 8. ,
            8.2, 9.2, 8.2, 8.5, 8. , 8.9, 8.4, 8.2, 8.1, 8.3, 8.1, 8.1, 8.1,
            8.3, 8.2, 8.3, 8.7, 8.3, 8.6, 8. , 8.1, 8.2, 8.5, 8.3, 8.9, 8. ,
            8.6, 8.3, 8.1, 8.7, 8.4, 8.1, 8.4, 8. , 8.5, 8.8, 8.2, 8.2, 8.5,
            9. , 8. , 8. , 8.3, 8.4, 8.6, 8.5, 8.7, 8.4, 8.1, 8.1, 8.1, 8.7,
            8.4, 8.9, 8.1, 8.2, 8. , 8.5, 8.5, 8. , 8. , 8.4, 8.1, 8.1, 8. ,
            8. , 8.3, 8.1, 8. , 8.3, 8. , 8. , 8. , 8. , 8. , 8. , 8.7,
            8.3, 8. , 8. , 8.5, 8. , 8.1, 8.1, 8.1, 8.3, 8.2, 8.3, 8.9, 8.2,
            8.2, 8. , 8.3, 8.2, 8.9, 8.5, 8.5, 8.1, 8.1, 8.5, 8.3, 8. , 8.2,
            8.7, 8.3, 8.5, 8.1, 8.3, 8.2, 8.4, 8.1, 8.1, 8.1, 8. , 8.2, 8. ,
            8.6, 8.3, 8.2, 8. , 8.3, 8. , 8.2, 8. , 8.2, 8.8, 8.1, 8. , 8.1,
            8. , 8.2, 8.5, 8.1, 8.4, 8.1, 8.1, 8.7, 8.2, 8. , 8. , 8. , 8.3,
            8.4, 8. , 8.5, 8.1, 8.1, 8.2, 8.2, 8.4, 8.3, 8.6, 8.2, 8. , 8.1,
            8.2, 8.1, 8.3, 8.4, 8.5, 8.6, 8. , 8.3, 8.5, 8.5, 8.3, 8.5, 8.4,
            8. , 8.1, 8.7, 8.9, 8.3, 8.1, 8.1, 8. , 8.2, 8.4, 8.4, 8.1, 8.3,
            8.4, 8.2, 8.5, 8. , 8.2, 8.1, 8.4, 8.1, 8.6, 8.4, 8.1, 8.7, 8.1,
            8.2, 8.1, 8.3])
```

```
[ ]:
```

```
[393]: imdb.sort("Rating", descending=True)
```

```
[393]: Votes    | Rating | Title                                     | Year | Decade
1027398 | 9.2    | The Godfather                           | 1972 | 1970
1498733 | 9.2    | The Shawshank Redemption                 | 1994 | 1990
692753  | 9      | The Godfather: Part II                   | 1974 | 1970
447875  | 8.9    | Il buono, il brutto, il cattivo (1966) | 1966 | 1960
1473049 | 8.9    | The Dark Knight                         | 2008 | 2000
384187  | 8.9    | 12 Angry Men                           | 1957 | 1950
1074146 | 8.9    | The Lord of the Rings: The Return of the King | 2003 | 2000
761224  | 8.9    | Schindler's List                        | 1993 | 1990
1166532 | 8.9    | Pulp Fiction                           | 1994 | 1990
1177098 | 8.8    | Fight Club                             | 1999 | 1990
... (240 rows omitted)
```

```
[400]: def rating_year(Rating):
        return int(float(Rating))
```

```
[404]: rating_year(9.2)
```

```
[404]: 9
```

```
[409]: Total_Pay = imdb.apply(rating_year, "Rating")
        Total_Pay
```

[illegible]

```
[417]: compnew = imdb.with_column("Rating", imdb.apply(rating_year, "Rating"))
      compnew
```

```
[417]: Votes | Rating | Title | Year | Decade
88355 | 8 | M | 1931 | 1930
132823 | 8 | Singin' in the Rain | 1952 | 1950
74178 | 8 | All About Eve | 1950 | 1950
635139 | 8 | Léon | 1994 | 1990
145514 | 8 | The Elephant Man | 1980 | 1980
425461 | 8 | Full Metal Jacket | 1987 | 1980
441174 | 8 | Gone Girl | 2014 | 2010
850601 | 8 | Batman Begins | 2005 | 2000
37664 | 8 | Judgment at Nuremberg | 1961 | 1960
46987 | 8 | Relatos salvajes | 2014 | 2010
... (240 rows omitted)
```

```
[419]: type(compnew.column("Rating")[0])
```

```
[419]: numpy.int64
```

```
[422]: compnew.barh('Year', 'Rating')
```



#F

The plot shows that most of the years recieved an 8 rating but 1974, 1994 and 1972 received a rating higher then 8

```
[431]: #G
imdb.sort("Title", descending=True)
```

```
[431]: Votes | Rating | Title | Year | Decade
65370 | 8.3 | Yôjinbô | 1961 | 1960
138240 | 8 | Yip Man | 2008 | 2000
427099 | 8 | X-Men: Days of Future Past | 2014 | 2010
53186 | 8.3 | Witness for the Prosecution | 1957 | 1950
50208 | 8 | Who's Afraid of Virginia Woolf? | 1966 | 1960
264333 | 8.5 | Whiplash | 2014 | 2010
287727 | 8.2 | Warrior | 2011 | 2010
618914 | 8.4 | WALL·E | 2008 | 2000
218430 | 8.4 | Vertigo | 1958 | 1950
700999 | 8.2 | V for Vendetta | 2005 | 2000
... (240 rows omitted)
```

```
[430]: imdb.barh('Title', 'Year')
```

	1	to
	2	Bright (1974) 10
	3	Al-Mualla (1974) 10
	4	
	5	The Elephant (1974) 10
	6	Andromeda (1974) 10
	7	Al-Mualla (1974) 10
	8	Al-Mualla (1974) 10
	9	Al-Mualla (1974) 10
	10	Al-Mualla (1974) 10
	11	Al-Mualla (1974) 10
	12	Al-Mualla (1974) 10
	13	Al-Mualla (1974) 10
	14	Al-Mualla (1974) 10
	15	Al-Mualla (1974) 10
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	99	Al-Mualla (1974) 10
	100	Al-Mualla (1974) 10

[]: