

asg2-sol

May 23, 2018

1 Introduction to Data Science - Homework 2 - Solutions

CENG 499

Adapted from University of Utah - Data Science Course

Due: Sunday, Mar 11, 11:59pm.

In this homework you will read in and analyze a movies dataset. First we'll do some basic analysis with vanilla Python, then we'll move on to doing more advanced analysis with [Pandas](#).

1.1 Your Data

Fill out the following information:

First Name:

Last Name:

E-mail:

Student ID:

1.2 Part 1: Analyzing Data The Hard Way

In this part we'll do some manual analysis of a movies dataset.

1.2.1 Task 1.1: Read in the data

Parse the file [movies.csv](#) using the [csv library](#). Lecture 6 and/or Homework 2 might be a good inspiration for this.

We recommend that you store the header in a separate array. Make sure that at least the "ratings" and the "votes" columns are cast to the appropriate data types for doing calculations with them.

Print the header, a row of the table, and the number of rows and columns.

```
In [2]: # import the csv library
import csv

# initialize the top-level array
movies = []
header = []

cnt = 0
```

```

with open('movies.csv', newline='') as f:
    reader = csv.reader(f)
    for row in reader:    # iterate over the lines in the file
        if cnt == 0:      # first line
            header = row  # get the headers
        else:
            # append to movies list, change the following line
            movies.append(row)
        cnt += 1
        # print(len(movies))
header

```

```

Out[2]: ['',
'title',
'year',
'length',
'budget',
'rating',
'votes',
'r1',
'r2',
'r3',
'r4',
'r5',
'r6',
'r7',
'r8',
'r9',
'r10',
'mpaa',
'Action',
'Animation',
'Comedy',
'Drama',
'Documentary',
'Romance',
'Short']

```

1.2.2 Task 1.2: Calculate movie stats

In this task, you will calculate some statistics about movies. We suggest you implement your solutions for Tasks 1.2.1-1.2.3 in one code cell - you should be able to calculate this in a single iteration over the dataset.

Task 1.2.1 Calcualte average ratings Compute the average rating for the movies and print the output. Also print the number of movies. Your output could look like this:

Average rating: xxx.xxxx, Number of movies: xx

Task 1.2.2: Calculate average rating for major movies Compute the average rating for the movies that have more than 500 votes in your loaded dataset and print the output. (We'll call these movies with more than 500 votes *major movies* from now on).

Your output could look like this:

Average rating of movies with more than 500 votes: xxx.xxxx, Number of major movies: xx

Task 1.2.3: Find the highest rated major movie Find out which of the movies with more than 500 votes has the highest rating.

Your output could look like this:

Highest rating: xxx.xxxx, Title: MOVIE TITLE

Task 1.2.4: Interpret the data

- What's the size relationship of major movies to all movies?
- Are major movies usually better than the average movies?

```
In [19]: # Task 1.2.1 Calcualte average ratings
sum = 0
cnt = 0
for row in movies:
    sum += float(row[5])
    cnt += 1
print('Average rating: ' + str(round(sum/len(movies),4)) + ', Number of movies: ' + str(len(movies)))

# Task 1.2.2: Calculate average rating for major movies
sum = 0
cnt_major = 0
for row in movies:
    try:
        if int(row[6]) >= 500:
            cnt_major += 1
            sum += float(row[5])
    except ValueError:
        continue
print('Average rating of movies with more than 500 votes: ' + str(round(sum/cnt_major,4)))

# Task 1.2.3: Find the highest rated major movie
high_rate = 0
high_movie = 0
cnt = 0
for row in movies:
    try:
        if int(row[5]) > high_rate:
            high_rate = int(row[5])
            high_movie = cnt
    except ValueError:
        continue
```

```

        continue
    cnt += 1
    print('Highest rating: ' + str(high_rate) + ', Title: ')

```

Average rating: 5.9329, Number of movies: 58788

Average rating of movies with more than 500 votes: 6.3705, Number of major movies: 6462

Highest rating: 10, Title:

Your Interpretation:

About 10% of the movies are major movies.

Major movies have a slightly lower average rating than all movies.

1.3 Part 2: Pandas

In this part we will use the [Pandas library](#) for our analysis.

1.3.1 Task 2.1: Loading data

Read in the data again. This time you should create a [Pandas DataFrame](#). Print the head of the dataset. * How many data rows did you load? How many columns? * Of which data types are the columns? * Do you have to do manual data type conversions?

In [22]: # Task 2.1: Loading data

```

# import pandas and numpy
import pandas as pd

movies_pd = pd.read_csv("movies.csv")
movies_pd.head(2)

```

```

Out[22]:
   Unnamed: 0  title  year  length  budget  rating  votes  r1  \
0           1      $  1971     121     NaN    6.4    348  4.5
1           2 $1000 a Touchdown  1939      71     NaN    6.0    20  0.0

   r2  r3  ...  r9  r10  mpaa  Action  Animation  Comedy  Drama  \
0  4.5  4.5  ...  4.5  4.5   NaN      0         0        1      1
1 14.5  4.5  ...  4.5 14.5   NaN      0         0        1      0

   Documentary  Romance  Short
0             0         0      0
1             0         0      0

[2 rows x 25 columns]

```

In [23]: # print data types

```

movies_pd.dtypes

```

```

Out[23]: Unnamed: 0      int64
         title         object
         year          int64
         length        int64
         budget        float64
         rating        float64
         votes         int64
         r1            float64
         r2            float64
         r3            float64
         r4            float64
         r5            float64
         r6            float64
         r7            float64
         r8            float64
         r9            float64
         r10           float64
         mpaa          object
         Action         int64
         Animation      int64
         Comedy         int64
         Drama          int64
         Documentary    int64
         Romance        int64
         Short          int64
         dtype: object

```

Your Interpretation: Looks line no, no need for data conversion.

1.3.2 Task 2.2: Calculate the average rating

Compute the average rating for all movies and print the output in a formatted way.

Your output could look like this:

Average rating: xxx.xxxx

```

In [24]: # Task 2.2: Calculate the average rating
         print('Average rating: ' + str("%.4f" % movies_pd['rating'].mean()))

```

Average rating: 5.9329

1.3.3 Task 2.3: Compare the runtime

Measure the runtime of the mean calculation using [Pandas](#) and compare it to the computation time for calculating the mean using a for loop (you can copy the relevant parts from part one).

You can use [time.clock\(\)](#) to set timestamps before and after the execution of the code you want to measure, then you simply subtract end time from start time.

Print your results in a human readable way and add a metric to the output. Calculate the factor of the difference and print it. Note that the exact times and the factors will vary when you re-run this and especially between machines.

E.g.:

time using own code: xxx.xxx s

time using Pandas: xxx.xxx s

difference factor: xxx

In [28]: *# Task 2.3: Compare the runtime*

```
import time

start = time.clock()

sum = 0
for row in movies:
    try:
        sum += float(row[5])
    except ValueError:
        continue

print('Average rating: ' + str(round(sum/len(movies),4)) + ', Number of movies: ' + str(len(movies)))
time_loop = time.clock() - start
print("time using own code",time_loop,"s")

start = time.clock()
print('Average rating: ' + str("%.4f" % movies_pd['rating'].mean()))
time_pandas = time.clock() - start
print("time using Pandas",time_pandas,"s")

print("difference factor: ",str("%.4f" % (time_pandas/time_loop)))
```

Average rating: 5.9329, Number of movies: 58788

time using own code 0.034253999999999785 s

Average rating: 5.9329

time using Pandas 0.0005880000000000477 s

difference factor: 0.0172

1.3.4 Task 2.4: Filter out rows

The whole movies dataset has about 60k entries. Use pandas to filter your dataframe to contain only the major movies with more than 500 votes.

Count and print the number of movies with more than 500 votes.

E.g.: xxx.xxx movies have more than 500 votes.

In [29]: *# Task 2.4: Filter out rows*

```
majorMovies = movies_pd[movies_pd.votes>500]
print(len(majorMovies), "movies have more than 500 votes")
```

6458 movies have more than 500 votes

1.3.5 Task 2.5: Calculate the average rating for major movies

Compute the average rating for the major movies. Your output could look like this:

Average rating of movies with more than 500 votes: xxx.xxxx

```
In [51]: # Task 2.5: Calculate the average rating for major movies
         avg_rate_major_movies = movies_pd[movies_pd.votes > 500].rating.mean()
         print('Average rating of movies with more than 500 votes:', str("%.4f" % avg_rate_majo
```

Average rating of movies with more than 500 votes: 6.3706

1.3.6 Task 2.6: Find the highest rated major movie

Find the highest rated major movie in the dataframe. Hint: `idxmax()` could be a helpful function.

Print the title and the rating.

Your output could look like this:

Highest rated movie: TITLE, rating: x.x

```
In [31]: # Task 2.6: Find the highest rated major movie
         maxrate = majorMovies.rating.max()
         maxMovies = majorMovies[majorMovies.rating == maxrate]
         print("Highest rated movies:", maxMovies.title.values, " rating:", maxrate)
```

Highest rated movies: ['Godfather, The' 'Shawshank Redemption, The'] rating: 9.1

1.3.7 Task 2.7: Filter out rows and count specific features

From the major movies you filtered out in a previous task, find out * How many are comedies. *

How many are dramas * How many are both, comedies and dramas

Create new dataframes for each of these subsets.

Hint: you can combine broadcasting statements with a boolean AND: `&`.

Print the results, e.g.:

xxx.xxx major movies are comedies.

xxx.xxx major movies are dramas.

xxx.xxx major movies are both, comedies and dramas.

```
In [107]: # Task 2.7: Filter out rows and count specific features
         movComedy = majorMovies[majorMovies.Comedy==1]
         movDrama = majorMovies[majorMovies.Drama==1]
         movBoth = majorMovies[(majorMovies.Comedy==1) & (majorMovies.Drama==1)]
         print("%6d" % len(movComedy), "major movies are comedies.")
         print("%6d" % len(movDrama), "major movies are dramas.")
         print("%6d" % len(movBoth), "major movies are both, comedies and dramas.")
```

2553 major movies are comedies.
3370 major movies are dramas.
814 major movies are both, comedies and dramas.

1.3.8 Task 2.8: Compare ratings of different categories

Now we want to compare the ratings for comedies and dramas for major movies.

Compute the average rating, maximum rating, minimum rating, standard deviation, and the median for each category. Hint: there is a function that does all of this in one line.

What do these numbers tell us? Provide an interpretation.

```
In [108]: # Task 2.8: Compare ratings of different categories
print('Comedy Ratings:')
print("Average\t", "%8.4f" % movComedy.rating.mean())
print("Min\t", "%8.4f" % movComedy.rating.min())
print("Max\t", "%8.4f" % movComedy.rating.max())
print("Std.Dev\t", "%8.4f" % movComedy.rating.std())
print("Median\t", "%8.4f" % movComedy.rating.median())
print()
print('Drama Ratings:')
print("Average\t", "%8.4f" % movDrama.rating.mean())
print("Min\t", "%8.4f" % movDrama.rating.min())
print("Max\t", "%8.4f" % movDrama.rating.max())
print("Std.Dev\t", "%8.4f" % movDrama.rating.std())
print("Median\t", "%8.4f" % movDrama.rating.median())
```

Comedy Ratings:

Average	6.1214
Min	1.7000
Max	8.7000
Std.Dev	1.1961
Median	6.2000

Drama Ratings:

Average	6.7496
Min	1.3000
Max	9.1000
Std.Dev	0.9910
Median	6.9000

Your Interpretation: Drama rating variation is smaller than comedy rating variation (see plots at the end)

1.3.9 Task 2.9: Movies per year

Calculate how many major movies were made in each year.

Print the number like this, sorted by year:


```

year    number of movies
1902    xxx
1903    xxx
...      ...

```

Use this data to render a line chart of the number of movies per year.

```
In [109]: # Task 2.9: Movies per year
```

```

mYear = movies_pd.groupby('year')['title'].count().reset_index(name='number of movies')
mYear.head(2)

```

```

Out[109]:   year  number of movies
0  1893                1
1  1894                9

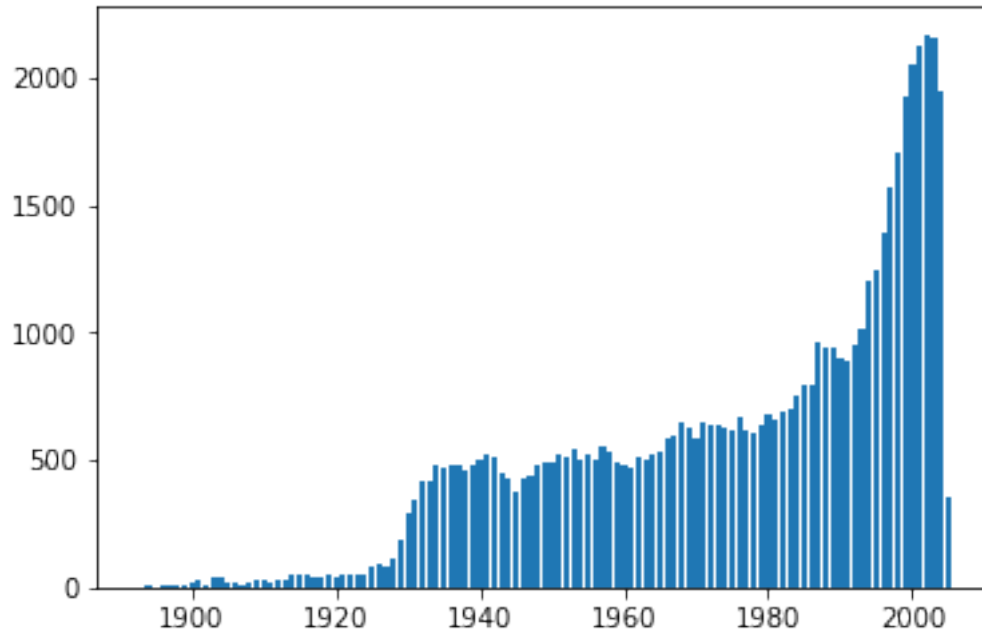
```

```

In [110]: %matplotlib inline
           # create the plot here
           import matplotlib.pyplot as plt
           year = mYear['year']
           nummov = mYear['number of movies']
           plt.bar(year,nummov)

```

```
Out[110]: <Container object of 113 artists>
```



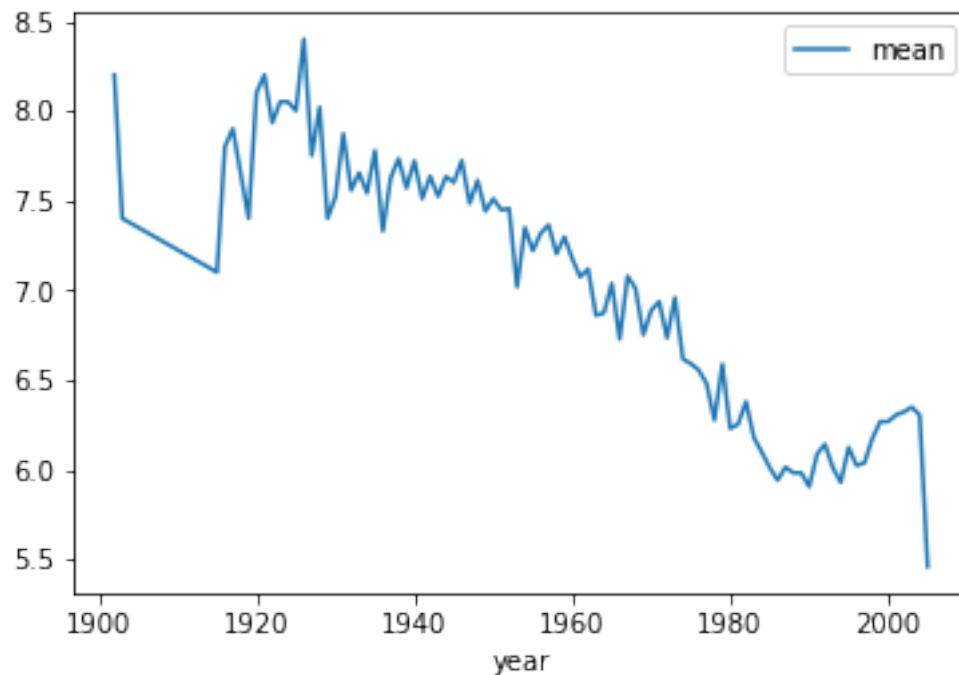
1.3.10 Task 2.10: Yearly average

Compute the average rating per year for all major movies.

Use the numbers you computed to plot a line chart. Plot the year on the x-axis and the average rating on the y axis.

```
In [111]: # Task 2.10: Yearly average
avg_year=majorMovies.groupby('year')['rating'].agg(['mean'])
avg_year.plot()
```

```
Out[111]: <matplotlib.axes._subplots.AxesSubplot at 0x12f863ba8>
```



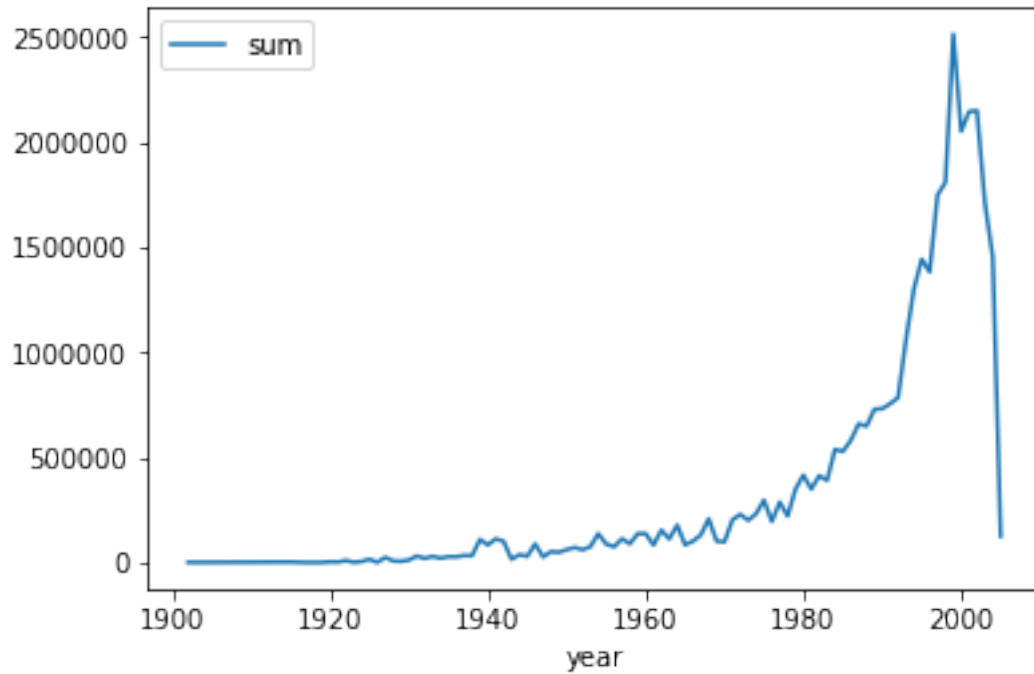
1.3.11 Task 2.11: Explore and Interpret

Are old movies better? How could you explain this? Are there differences between the rating of major movies and all movies over time? Continue to explore and use plots to inform your answer. Interesting measures to consider are the total number of votes per year, the average number of votes for a movie in a particular year, etc.

TODO: your code and your interpretation

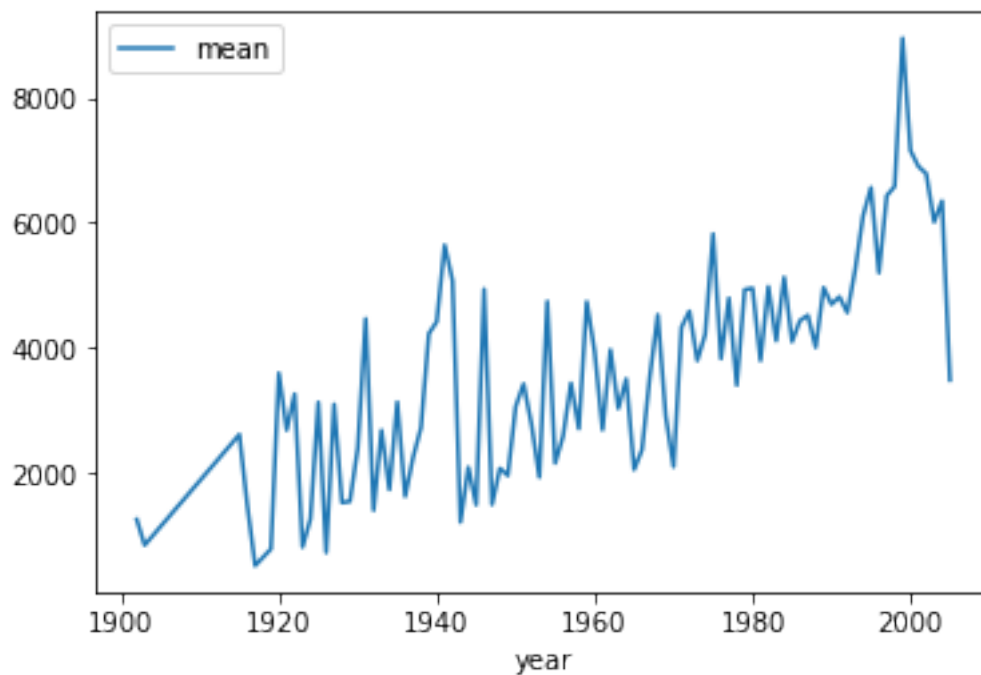
```
In [112]: # total number of votes for movies by year
avg_year=majorMovies.groupby('year')['votes'].agg(['sum'])
avg_year.plot()
```

```
Out[112]: <matplotlib.axes._subplots.AxesSubplot at 0x12f597f98>
```



```
In [113]: # average number of votes for movies by years
avg_year=majorMovies.groupby('year')['votes'].agg(['mean'])
avg_year.plot()
```

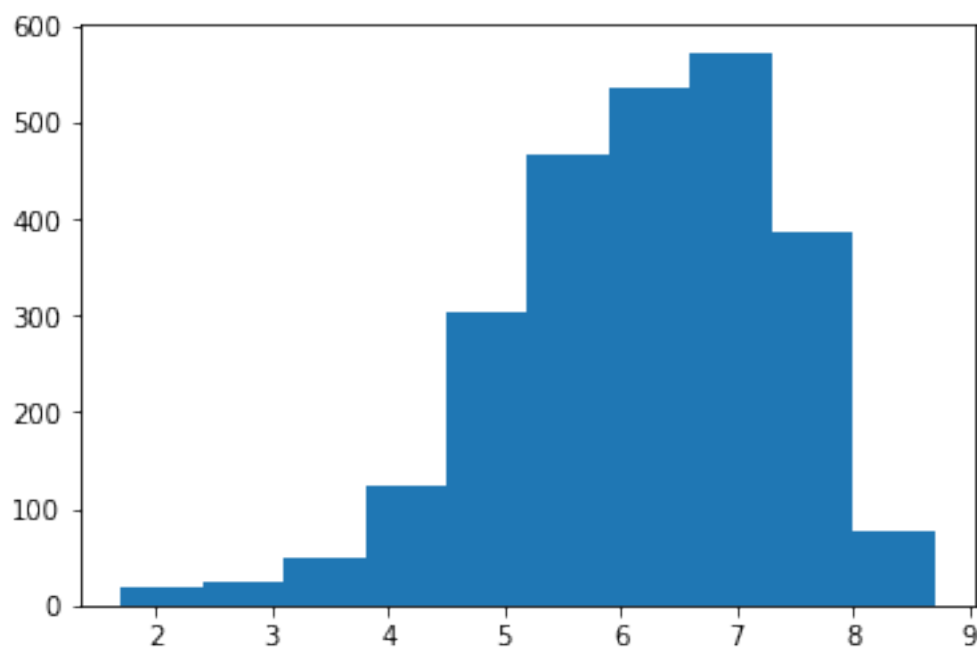
```
Out[113]: <matplotlib.axes._subplots.AxesSubplot at 0x12fc84710>
```



1.3.12 Note: Looks like the average number of votes are increasing by years, and the average rating of the movies are decreasing.

```
In [114]: import matplotlib.pyplot as plt
```

```
In [115]: n, bins, patches = plt.hist(movComedy['rating'])  
plt.show()
```



```
In [116]: n, bins, patches = plt.hist(movDrama['rating'])  
plt.show()
```

