All files mentioned in this document should be uploaded into the *github* repository. Answers to all questions will be described into three sections: description, code lines, and output samples. The file *recommendations.py* contains the Python code for solving all 10 questions. start symbol (*), in the output tables, indicates that I choose the marked line as my result.

Problem 1

- To answer this question, we need to know how many ratings each movie has received and the summation of these ratings. After that, we can simply compute the average. The following piece of code will return movies that have the highest average ratings (I got 10 movies with average ratings of 5.0; the first five are marked by (*)):
- Source code

Movie A	Average-ratings
Great Day in Harlem, A (1994) Prefontaine (1997) Aiqing wansui (1994)	5.0 (*) 5.0 (*) 5.0 (*)
Star Kid (1997) Marlene Dietrich: Shadow and Light Entertaining Angels:	5.0 (*) (1996) 5.0 (*)
The Dorothy Day Story (1996) Saint of Fort Washington, The (1993) Someone Else's America (1995) Santa with Muscles (1996)	5.0 5.0 5.0 5.0

They Made Me a Criminal (1939)

5.0

Problem 2

- Here, we are considering only number of ratings each movie has received, so I simply modified the code from question 1 to answer this question:
- Source code

• Output

Movie	Number of ratings	
Star Wars (1977)	583	
Contact (1997)	509	
Fargo (1996)	508	
Return of the Jedi (1983)	507	
Liar Liar (1997)	485	

Problem 3

- This question requires loading more information that was not originally retrieved; I am talking about the genders and ages (ages will be sed in question 9 and 10). After loading, I only added *if* statement to the source code, from question 1, to check whether a user is male (M) or Female (F) (I got 11 movies with average ratings 5.0; the first five are marked by (*)):
- Source code

```
# Load more data: gender and age
gender={}
age={}
for line in open('u.user'):
  (id,ag,g,x,y) = line.split('|')
 gender.setdefault(id,g)
  age.setdefault(id,ag)
 . . .
rating = {}
for user in prefs.keys():
  if gender[user] == 'M':
   continue
  for key,value in prefs[user].iteritems():
            rating.setdefault(key,[])
            rating[key].append(value)
average = {}
for movie in rating.keys():
  avg = np.mean(rating[movie])
  average[movie] = avg
sorted_x = sorted(average.iteritems(), key=operator.itemgetter(1))
sorted_x.reverse()
print (" ##### number [3] solution ##### ")
for (key,value) in sorted_x[0:15]:
print key,' ... ',value
```

Movie	Average ra	tings(Women)	
Prefontaine (1997)		5.0	(*)
Telling Lies in America (199	7)	5.0	(*)
Foreign Correspondent (1940)		5.0	(*)
Faster Pussycat! Kill! Kill!	(1965)	5.0	(*)
Year of the Horse (1997)		5.0	(*)
Mina Tannenbaum (1994)		5.0	
Maya Lin: A Strong Clear Vis:	ion (1994)	5.0	
Everest (1998)		5.0	
Someone Else's America (1995))	5.0	
Visitors, The (Visiteurs, Les	s) (1993)	5.0	
Stripes (1981)		5.0	

- One small change to the previous question source code is sufficient to answer this question –changing 'M'(Male) to 'F' (Female)(I got 15 movies with average ratings of 5.0; the first five are marked by (*):
- Source code

```
rating = {}
for user in prefs.keys():
 if gender[user] == 'F':
   continue
 for key,value in prefs[user].iteritems():
            rating.setdefault(key,[])
            rating[key].append(value)
average = {}
for movie in rating.keys():
 avg = np.mean(rating[movie])
 average[movie] = avg
sorted_x = sorted(average.iteritems(), key=operator.itemgetter(1))
sorted_x.reverse()
print (" ##### number [4] solution ##### ")
for (key,value) in sorted_x[0:15]:
print key,' ... ',value
```

Movie	Average r	atings(M	len)
Delta of Venus (1994)		5.0	(*)
Great Day in Harlem, A (1994)		5.0	(*)
Leading Man, The (1996)		5.0	(*)
Love Serenade (1996)		5.0	(*)
Prefontaine (1997)		5.0	(*)
Aiqing wansui (1994)		5.0	
Little City (1998)		5.0	
Star Kid (1997)		5.0	
Marlene Dietrich: Shadow and Li	ght (1996)	5.0	
Entertaining Angels:			
The Dorothy Day Story (1996)		5.0	
Quiet Room, The (1996)		5.0	
Saint of Fort Washington, The (1993)	5.0	
Letter From Death Row, A (1998)		5.0	
Santa with Muscles (1996)		5.0	
They Made Me a Criminal (1939)		5.0	

- the function topMatches() is considered to answer this question. The function returns the best n matches for person (or movies), but before calling the function, we should swap movies and raters within prefs dictionary using the function it transformPrefs(). I just made a copy of the function body inside my code:
- Source code

```
result={}
for person in prefs:
    for item in prefs[person]:
        result.setdefault(item,{})
        # Flip item and person
        result[item][person]=prefs[person][item]
print (" ##### number [5] solution ##### ")
print('Top matches: ',topMatches(result,'Top Gun (1986)')[0:8])
print('Least matches: ',topMatches(result,'Top Gun (1986)')[-31:])
```

Movie(*) received ratings most like Top Gun

Movie	r	
Wild America (1997)	1.0	(*)
Wedding Gift, The (1994)	1.0	
Underground (1995)	1.0	
Two or Three Things I Know About Her (1966)	1.0	
Two Bits (1995)	1.0	
Total Eclipse (1995)	1.0	
The Innocent (1994)	1.0	
That Old Feeling (1997)	1.0	

Movie(*) received ratings least like Top Gun

Movie	r
Year of the Horse (1997) World of Apu, The (Apur Sansar) (1959) Two Much (1996) Tetsuo II: Body Hammer (1992) Telling Lies in America (1997) Switchback (1997)	-1.0 (*) -1.0 -1.0 -1.0 -1.0

```
Safe Passage (1994)
                                              -1.0
Roseanna's Grave (For Roseanna) (1997)
                                              -1.0
Romper Stomper (1992)
                                              -1.0
Nil By Mouth (1997)
                                              -1.0
Nico Icon (1995)
                                              -1.0
Naked in New York (1994)
                                              -1.0
Midnight Dancers (Sibak) (1994)
                                              -1.0
                                              -1.0
Meet Wally Sparks (1997)
Lover's Knot (1996)
                                              -1.0
Love and Death on Long Island (1997)
                                              -1.0
Loch Ness (1995)
                                              -1.0
Lamerica (1994)
                                              -1.0
Joy Luck Club, The (1993)
                                              -1.0
Heidi Fleiss: Hollywood Madam (1995)
                                              -1.0
Frisk (1995)
                                              -1.0
Everest (1998)
                                              -1.0
                                              -1.0
Carried Away (1996)
Carpool (1996)
                                              -1.0
Caro Diario (Dear Diary) (1994)
                                              -1.0
Broken English (1996)
                                              -1.0
Bitter Sugar (Azucar Amargo) (1996)
                                              -1.0
Bewegte Mann, Der (1994)
                                              -1.0
Beat the Devil (1954)
                                              -1.0
Bad Moon (1996)
                                              -1.0
Babysitter, The (1995)
                                              -1.0
```

- Simply, using the function len() is sufficient to return number of movies rated by a specific user in the dictionary prefs:
- Source code

5 raters rated the most films

ID	Number of movies
405	736
655	678
13	632
450	538
276	516

• By computing r for everyone to everyone, I got 806 raters whose more than 4 raters having r equal to 1. Please, see the file q7.txt for more detail. it has all groups of raters which are similar to each other (r=1). For example, in the forth line, you will find:

```
( 343, 729, 341, 273, and 261 )
```

Which means that any pair of users from this list, with ids 343, 729, 341, 273, and 261, has an r value equal 1.0.

• Source code

```
print (" ##### number [7] solution ##### ")
lis = []
allsim = calculateSimilarUser(prefs,1000)
c = 0
k = 0
lessThanFive = 0
equalZero = 0
grearterThanFive = 0
Max = 0
for user in allsim.keys():
k = k + 1
totalError = 0.0
t = 0
 for (key,value) in allsim[user]:#[0:4]:
 totalError = totalError + (1.0 - float(key))
  if(float(key) == 1.0):
   t = t + 1
lis.append(t)
 if(t == 0):
  equalZero = equalZero + 1
 if(t >= 4):
```

```
grearterThanFive = grearterThanFive + 1
 if(t < 4) & (t > 0):
  lessThanFive = lessThanFive + 1
 if(t > Max):
 Max = t
print('Max = ',Max,'k=',k,' total = ',(equalZero + lessThanFive +
   grearterThanFive))
print('equal zero = ',equalZero,'lessThanFive',lessThanFive,
                   'grearterThanFive', grearterThanFive)
counter = 0
f = open('q7.txt', 'w')
for user in allsim.keys():
  if (lis[counter] > 3):
   s = '(') + user
   f.write(str(s))
   for (key,value) in allsim[user][0:(lis[counter])]:
     f.write(' + '+str(value))
   f.write(' )\n')
  counter = counter + 1
f.close()
```

• Output

5 raters most agreed with each other

The are as many as 806 groups of raters which are most agreed with each other, see q7.txt. I have chosen the following group whose ids are: 343, 729, 341, 273, and 261 (all r values are equal to 1.0).

In the file, q7.txt, you can also find, for example, another group which consists of 96 raters! and all agreed with each other (r = 1.0 between any pair of raters)

Problem 8

• By computing r for everyone to everyone, I got 766 raters whose more than 4 raters having r equal to -1.0. Please, see the file q8.txt for more detail. it has all groups of raters which are not similar to each other (r = -1.0). For example, in the seventeenth line, you will find:

```
(715, 29, 520, 651, and 858)
```

Which means that any pair of users from this list, with ids 715, 29, 520, 651, and 858, has an r value of -1.0.

• Source code

```
print (" ##### number [8] solution ##### ")
lis = []
allsim = calculateSimilarUser2(prefs,1000)
k = 0
lessThanFive = 0
equalZero = 0
grearterThanFive = 0
Max = 0
for user in allsim.keys():
k = k + 1
totalError = 0.0
t = 0
 for (key,value) in allsim[user]:
 totalError = totalError + (1.0 - float(key))
  if(float(key) == (-1.0)):
   t = t + 1
lis.append(t)
 if(t == 0):
  equalZero = equalZero + 1
 if(t >= 4):
 grearterThanFive = grearterThanFive + 1
 if(t < 4) & (t > 0):
  lessThanFive = lessThanFive + 1
 if(t > Max):
 Max = t
print('Max = ',Max,'k=',k,' total = ',(equalZero + lessThanFive +
   grearterThanFive))
print('equal zero =
   ',equalZero,'lessThanFive',lessThanFive,'grearterThanFive',grearterThanFive)
counter = 0
f = open('q8.txt', 'w')
for user in allsim.keys():
 if (lis[counter] > 3):
   s = '(') + user
   f.write(str(s))
   for (key,value) in allsim[user][0:(lis[counter])]:
     f.write(' + '+str(value))
   f.write(' )\n')
  counter = counter + 1
f.close()
```

```
5 raters most disagreed with each other
```

The are as many as 766 groups of raters which are most disagreed with each other, see q8.txt. I have chosen the following group whose ids are: 715, 29, 520, 651, and 858 (all r values are equal to -1.0).

In the file, q8.txt, you can also find, for example, another group which consists of 76 raters! and all disagreed with each other (r = -1.0 between any pair of rates)

Problem 9

- I only made a small change to the source code of question 3 which is a new *if* statement to check the age:
- Source code

```
rating = {}
for user in prefs.keys():
 if gender[user] == 'F':
   continue
 if int(age[user]) <= 40:</pre>
   continue
 for key,value in prefs[user].iteritems():
            rating.setdefault(key,[])
             rating[key].append(value)
average = {}
for movie in rating.keys():
 avg = np.mean(rating[movie])
 average[movie] = avg
sorted_x = sorted(average.iteritems(), key=operator.itemgetter(1))
sorted_x.reverse()
print (" ##### number [9] solution ##### ")
for (key,value) in sorted_x[0:19]:
print key,' ... ', value
```

```
Movie Average-ratings(by men over 40)

Hearts and Minds (1996) 5.0 (*)

Faithful (1996) 5.0

Marlene Dietrich: Shadow and Light (1996) 5.0

Strawberry and Chocolate (Fresa y chocolate) (1993) 5.0
```

```
Late Bloomers (1996)
                                                    5.0
                                                    5.0
Solo (1996)
Grateful Dead (1995)
                                                    5.0
Prefontaine (1997)
                                                    5.0
Rendezvous in Paris (Rendez-vous de Paris, Les) (1995) 5.0
World of Apu, The (Apur Sansar) (1959)
                                                    5.0
Aparajito (1956)
                                                    5.0
Ace Ventura: When Nature Calls (1995)
                                                    5.0
Star Kid (1997)
                                                    5.0
Two or Three Things I Know About Her (1966)
                                                    5.0
Poison Ivy II (1995)
                                                    5.0
Double Happiness (1994)
                                                    5.0
Little City (1998)
                                                    5.0
Boxing Helena (1993)
                                                    5.0
Spice World (1997)
                                                    5.0
They Made Me a Criminal (1939)
                                                    5.0
Great Day in Harlem, A (1994)
                                                    5.0
Little Princess, The (1939)
                                                    5.0
Unstrung Heroes (1995)
                                                    5.0
Leading Man, The (1996)
                                                    5.0
Indian Summer (1996)
                                                    5.0
```

- I only made a small change to the source code of question 9 –changing 'F' to 'M':
- Source code

```
rating = {}
for user in prefs.keys():
 if gender[user] == 'M':
   continue
 if int(age[user]) <= 40:</pre>
   continue
 for key,value in prefs[user].iteritems():
             rating.setdefault(key,[])
             rating[key].append(value)
average = {}
for movie in rating.keys():
 avg = np.mean(rating[movie])
 average[movie] = avg
sorted_x = sorted(average.iteritems(), key=operator.itemgetter(1))
sorted_x.reverse()
print (" ##### number [10] solution ##### ")
```

```
for (key,value) in sorted_x[0:19]:
  print key,' ... ',value
```

• Output

Movie	Average-ratings(by wor	nen ove	r 40)
Shallow Grave (1994)		5.0	(*)
Great Dictator, The (1940))	5.0	
Visitors, The (Visiteurs,	Les) (1993)	5.0	
Shall We Dance? (1937)		5.0	
In the Bleak Midwinter (19	995)	5.0	
Funny Face (1957)		5.0	
Ma vie en rose (My Life in	n Pink) (1997)	5.0	
Swept from the Sea (1997)		5.0	
Best Men (1997)		5.0	
Foreign Correspondent (194	10)	5.0	
Tombstone (1993)		5.0	
Wrong Trousers, The (1993))	5.0	
Top Hat (1935)		5.0	
Quest, The (1996)		5.0	
Balto (1995)		5.0	
Angel Baby (1995)		5.0	
Band Wagon, The (1953)		5.0	
Letter From Death Row, A	(1998)	5.0	
Mina Tannenbaum (1994)		5.0	
Mary Shelley's Frankenste	in (1994)	5.0	
Gold Diggers: The Secret	of Bear Mountain (1995)	5.0	
Nightmare Before Christmas	s, The (1993)	5.0	
Grand Day Out, A (1992)		5.0	
Bride of Frankenstein (193	35)	5.0	
Pocahontas (1995)		5.0	

References

- $[1] \ \ On\mbox{-line GroupLens Research}, \ \mbox{http://grouplens.org/datasets/movielens/}.$
- [2] J. Venge, On-line Stackoverflow, http://stackoverflow.com/questions/783897/truncating-floats-in-python.