Q1. EM algorithm for binary matrix completion:

(a) Sanity check:

Mean popularity = #Recommended / #Saw

The movies sorted from least to most popular are:

```
Fifty Shades of Grey
                                                0.331
The Last Airbender
                                                0.357
Magic Mike
                                                0.525
Prometheus
                                                0.544
Bridemaids
                                                0.556
World War Z
                                                0.590
Man_of_Steel
                                               0.592
Mad_Max:_Fury_Road
                                               0.633
Drive
                                               0.636
Thor
                                               0.665
Pitch Perfect
                                               0.667
The Hunger Games
                                               0.671
Fast Five
                                               0.678
The Hateful Eight
                                               0.683
                                               0.692
Iron Man 2
The_Perks_of_Being_a_Wallflower
                                               0.695
American Hustle
                                               0.716
The Help
                                               0.725
Avengers: Age of Ultron
                                               0.730
21 Jump Street
                                               0.736
Captain America: The First Avenger
                                               0.736
Les Miserables
                                               0.745
Star Wars: The Force Awakens
                                               0.749
Jurassic World
                                               0.754
The Great Gatsby
                                               0.755
X-Men: First Class
                                               0.764
The Revenant
                                               0.767
Her
                                               0.779
Ex Machina
                                               0.786
                                               0.786
Room
Django Unchained
                                               0.789
                                               0.789
The_Girls_with_the_Dragon_Tattoo
                                               0.790
Frozen
                                               0.798
Midnight_in_Paris
                                               0.801
The Avengers
Wolf of Wall Street
                                                0.804
Harry Potter and the Deathly Hallows: Part 1 0.804
Black Swan
                                                0.810
Toy Story 3
                                                0.810
Harry Potter and the Deathly Hallows: Part 2 0.832
                                               0.839
Gone Girl
The_Theory_of_Everything
                                               0.841
12_Years_a_Slave
                                               0.841
Now_You_See_Me
                                               0.861
The Social Network
                                               0.894
The Martian
                                               0.901
Shutter Island
                                               0.908
Interstellar
                                               0.914
The Dark Knight Rises
                                                0.926
Inception
                                                0.996
```

For the most part the sorted movie list does correspond to my preferences, as I really liked the movies near the far-end of the list like Martian, Inception, The dark knight trilogy etc. Also it is interesting to note that the most popular movies are directed by Christopher Nolan.

The log-likelihood increases at each iteration, verifying the convergence.

Iteration	Log-likelihhod (L)
0	-23.6819
1	-14.3421 (Got -15.9532)
2	-13.4962
4	-12.2653
8	-11.9739
16	-11.6822 (Got -11.5974)
32	-11.2251
64	-10.7815

(f)

(e)

The predictions for movies I am yet to see are (sorted by expectation) are:

Thor	
Harry_Potter_and_the_Deathly_Hallows:_Part_1	
Room	
American_Hustle	
The_Girls_with_the_Dragon_Tattoo	
Toy_Story_3	
The_Great_Gatsby	
Frozen	
Bridemaids	
Drive	
The_Help	
Midnight_in_Paris	
Pitch_Perfect	
Les_Miserables	
Magic_Mike	

The list seems accurate as I just saw Thor this weekend!

(g) Source code attached

Source code:

```
# coding: utf-8
                                                               moviepop.append(temp.count(1)*1.0/(temp.count
# In[133]:
                                                               (1)+temp.count(0)))
                                                               moviespopular=sorted(zip(moviepop,movies),key=
import numpy as np
                                                               lambda i:i[0])
from collections import defaultdict
                                                               for i in moviespopular:
import math
                                                                 print '{0:45s} {1:.3f}'.format(i[1],i[0])
# In[12]:
                                                               # In[119]:
with open('data/hw8_movieTitles.txt','r') as f:
                                                               priorZ=[]
  movies=f.read().split()
                                                               with open('data/hw8_probZ_init.txt','r') as f:
with open('data/hw8_studentPIDs.txt','r') as f:
                                                                 temp=f.read().strip().split('\n')
  students=f.read().split()
                                                                 for i in temp:
                                                                   priorZ.append(float(i))
# In[151]:
                                                               # In[171]:
ratings=[]
count=0
                                                               rgz=[]
with open('data/hw8_ratings.txt','r') as f:
                                                               with open('data/hw8_probRgivenZ_init.txt','r') as
  temp=f.read().split('\n')
  for i in temp:
                                                                 temp=f.read().strip().split('\n')
    i=i.replace('?','-1')
                                                                 for i in temp:
    temp1 = i.strip().split(' ')
                                                                   rgz.append([float(k) for k in i.split()])
    temp1=[int(j) for j in temp1]
                                                               #rgz=np.matrix(rgz)
    ratings.append(temp1)
Iratings=ratings
ratings=np.matrix(ratings)
                                                               # In[176]:
                                                               pzt=defaultdict()
# In[153]:
                                                               for i in range(len(students)):
                                                                 pzt[i]=defaultdict()
vis = defaultdict(list)
                                                                 for j in range(len(priorZ)):
for i,j in enumerate(Iratings):
                                                                   pzt[i][j]=priorZ[j]
  vis[i]=[]
  for k in range(len(j)):
                                                               rgzd=defaultdict()
    if j[k]==1 or j[k]==0:
                                                               for i in range(len(movies)):
       vis[i].append(k)
                                                                 rgzd[i]=defaultdict()
                                                                 for j in range(len(priorZ)):
                                                                   rgzd[i][j]=rgz[i][j]
# In[110]:
                                                               # In[184]:
Is=len(students)
Im=len(movies)
moviepop=[]
                                                               priorZ={0:0.25,1:0.25,2:0.25,3:0.25}
for i in range(lm):
  temp=ratings[:,i].flatten().tolist()[0]
```

```
# In[195]:
                                                                        else:
                                                                          tmp+=pzt[k][i]*rgzd[p][k]
def estep(priorZ,pzt,rgzd,vis):
                                                                     rgzd[i][k]=temp/len(students)/priorZ[k]
  for i in vis:
                                                                 return priorZ,pzt,rgzd,vis
    temp1=0
    for j in priorZ:
                                                              def II(priorZ,pzt,rgzd,vis):
       temp=priorZ[j]
                                                                 temp2=0
                                                                 for i in vis:
      for k in vis[i]:
         temp*=rgzd[i][k]
                                                                   temp1=0
       pzt[i][j]=temp
                                                                   for j in priorZ:
                                                                     temp=priorZ[j]
      temp1+=temp
    for i in pzt:
                                                                     for k in vis[i]:
      for j in pzt[i]:
                                                                        temp*=rgzd[i][k]
         pzt[i][j]/=temp1
                                                                     temp1+=temp
  return priorZ,pzt,rgzd,vis
                                                                   temp2+=math.log(temp1)
                                                                 return temp2/len(students)
def mstep(priorZ,pzt,rgzd,vis):
  for i in priorZ:
    temp=0
                                                              # In[196]:
    for k in pzt:
                                                              for ite in range(64):
       temp+=pzt[k][i]
    priorZ[i]=temp/len(students)
                                                                 priorZ,pzt,rgzd,vis=estep(priorZ,pzt,rgzd,vis)
  for i in rgzd:
                                                                 priorZ,pzt,rgzd,vis=mstep(priorZ,pzt,rgzd,vis)
    for k in rgzd[i]:
                                                                 print II(priorZ,pzt,rgzd,vis)
      tmp=0
      for p in range(len(movies)):
         if p in viz[i]:
                                                              # In[]:
           if ratings[i][p]==1:
              tmp+=pzt[k][i]
```