October 25, 2021

1 Question 1 - Items which have the highest Jaccard similarity compared to the first item (i.e., the item from the first review, '18471619')? Report both 10 items similarities and item IDs

```
[1]: import gzip
    from collections import defaultdict
    import scipy
    import numpy
    import numpy
    import random

[2]: def parseDataFromFile(fname):
        for l in open(fname):
            yield eval(l)

[3]: data=list(parseDataFromFile("goodreads_reviews_comics_graphic.json"))
    print(data[0])
```

{'user\_id': 'dc3763cdb9b2cae805882878eebb6a32', 'book\_id': '18471619', 'review\_id': '66b2ba840f9bd36d6d27f46136fe4772', 'rating': 3, 'review\_text': 'Sherlock Holmes and the Vampires of London \n Release Date: April 2014 \n Publisher: Darkhorse Comics \n Story by: Sylvain Cordurie \n Art by: Laci \n Colors by: Axel Gonzabo \n Cover by: Jean Sebastien Rossbach \n ISDN: 9781616552664 \n MSRP: \$17.99 Hardcover \n "Sherlock Holmes died fighting Professor Moriarty in the Reichenbach Falls. \n At least, that\'s what the press claims. \n However, Holmes is alive and well and taking advantage of his presumed death to travel the globe. \n Unfortunately, Holmes\'s plans are thwarted when a plague of vampirism haunts Britain. \n This book collects Sherlock Holmes and the Vampires of London Volumes 1 and 2, originally created by French publisher Soleil." - Darkhorse Comics  $\n$  When I received this copy of "Sherlock Holmes and the Vampires of London" I was Ecstatic! The cover art was awesome and it was about two of my favorite things, Sherlock Holmes and Vampires. I couldn\'t wait to dive into this! \n Unfortunately, that is where my excitement ended. The story takes place a month after Sherlock Holmes supposed death in his battle with Professor Moriarty. Sherlock\'s plan to stay hidden and out of site are ruined when on a trip with his brother Mycroft, they stumble on the presence of vampires. That is about as much of Sherlock\'s character that

comes through the book. I can\'t even tell you the story really because nothing and I mean nothing stuck with me after reading it. I never, ever got the sense of Sherlock Holmes anywhere in this graphic novel, nor any real sense of mystery or crime. It was just Sherlock somehow battling vampires that should have had absolutely no trouble snuffing him out in a fight, but somehow always surviving and holding his own against supernatural, super powerful, blazingly fast creatures. In The cover art is awesome and it truly made me excited to read this but everything else feel completely flat for me. I tried telling myself that "it\'s a graphic novel, it would be hard to translate mystery, details, emotion" but then I remembered reading DC Comic\'s "Identity Crisis" and realized that was a load of crap. I know it\'s unfair to compare the two as "Identity Crisis" had popular mystery author Brad Meltzer writing it right? Yeah...no. The standard was set that day and there is more than enough talent out there to create a great story in a graphic novel. \n That being said, it wasn\'t a horrible story, it just didn\'t grip me for feel anything like Sherlock Holmes to me. It was easy enough to follow but I felt no sense of tension, stakes or compassion for any of the characters. \n As far as the vampires go, it\'s hard to know what to expect anymore as there are so many different versions these days. This was the more classic version which I personally prefer, but again I didn\'t find anything that portrayed their dominance, calm confidence or sexuality. There was definitely a presence of their physical prowess but somehow that was lost on me as easily as Sherlock was able to defend himself. I know it, wouldn't do to kill of the main character, but this would have a been a great opportunity to build around the experience and beguiling nature of a vampire that had lived so many years of experience. Another chance to showcase Sherlock\'s intellect in a battle of wits over strength in something more suitable for this sort of story as apposed to trying to make it feel like an action movie. \n Maybe I expected to much and hoped to have at least a gripping premise or some sort of interesting plot or mystery but I didn\'t find it here. This may be a must have for serious Sherlock Holmes fans that have to collect everything about him, but if you are looking for a great story inside a graphic novel, I would have to say pass on this one. \n That artwork is good, cover is great, story is lacking so I am giving it 2.5 out of 5 stars.', 'date\_added': 'Thu Dec 05 10:44:25 -0800 2013', 'date\_updated': 'Thu Dec 05 10:45:15 -0800 2013', 'read\_at': 'Tue Nov 05 00:00:00 -0800 2013', 'started\_at': '', 'n\_votes': 0, 'n\_comments': 0}

```
[4]: userPerItem=defaultdict(set)
   itemPerUser=defaultdict(set)

[5]: itemNames={}
```

```
[6]: for d in data:
    user=d["user_id"]
    item=d["book_id"]
    userPerItem[item].add(user)
    itemPerUser[user].add(item)
```

```
[7]: def Jaccard(s1,s2):
        numer=len(s1.intersection(s2))
        deno=len(s1.union(s2))
        return(numer/deno)
[8]: def mostSimilar(i,N=10):
        users=userPerItem[i]
        similarities=[]
        for j in userPerItem:
            if(i==j):
                continue
            sim=Jaccard(users,userPerItem[j])
            similarities.append((sim,j))
          similarities.sort(reverse=True)
        s=sorted(similarities, key=lambda element:(-element[0],element[1]))
        return(s)
[9]: query=data[0]['book_id']
    print("Similarities and Item IDs")
    mostSimilar(query)[:10]
    Similarities and Item IDs
(0.14285714285714285, '25659811'),
      (0.13793103448275862, '18369278'),
      (0.13157894736842105, '18430205'),
      (0.12903225806451613, '20299669'),
     (0.125, '17995154'),
     (0.12121212121212122, '18853527'),
      (0.121212121212122, '23093378'),
     (0.121212121212122, '23241671'),
      (0.11764705882352941, '18734070')]
```

#### October 25, 2021

- 1 Question 2 There are several ways similar-item recommendations could be used to make personalized recommendations for a particular user. For instance we could: User="dc3763cdb9b2cae805882878eebb6a32"
- 1.1 (a) Choosing the N items most similar to the user's favorite (i.e., highest rated) item.
- 1.2 (b) Finding the N most similar users, and recommending each of their their favorite (highest rated) items.

```
[1]: import gzip
from collections import defaultdict
import scipy
import scipy.optimize
import numpy
import random
```

```
[2]: def parseDataFromFile(fname):
    for 1 in open(fname):
        yield eval(1)
```

```
[3]: data=list(parseDataFromFile("goodreads_reviews_comics_graphic.json"))
print(data[0])
```

{'user\_id': 'dc3763cdb9b2cae805882878eebb6a32', 'book\_id': '18471619', 'review\_id': '66b2ba840f9bd36d6d27f46136fe4772', 'rating': 3, 'review\_text': 'Sherlock Holmes and the Vampires of London \n Release Date: April 2014 \n Publisher: Darkhorse Comics \n Story by: Sylvain Cordurie \n Art by: Laci \n Colors by: Axel Gonzabo \n Cover by: Jean Sebastien Rossbach \n ISDN: 9781616552664 \n MSRP: \$17.99 Hardcover \n "Sherlock Holmes died fighting Professor Moriarty in the Reichenbach Falls. \n At least, that\'s what the press claims. \n However, Holmes is alive and well and taking advantage of his presumed death to travel the globe. \n Unfortunately, Holmes\'s plans are thwarted when a plague of vampirism haunts Britain. \n This book collects Sherlock Holmes and the Vampires of London Volumes 1 and 2, originally created by French publisher Soleil." - Darkhorse Comics \n When I received this copy of "Sherlock Holmes and the Vampires of London" I was Ecstatic! The cover art was

awesome and it was about two of my favorite things, Sherlock Holmes and Vampires. I couldn\'t wait to dive into this! \n Unfortunately, that is where my excitement ended. The story takes place a month after Sherlock Holmes supposed death in his battle with Professor Moriarty. Sherlock\'s plan to stay hidden and out of site are ruined when on a trip with his brother Mycroft, they stumble on the presence of vampires. That is about as much of Sherlock\'s character that comes through the book. I can\'t even tell you the story really because nothing and I mean nothing stuck with me after reading it. I never, ever got the sense of Sherlock Holmes anywhere in this graphic novel, nor any real sense of mystery or crime. It was just Sherlock somehow battling vampires that should have had absolutely no trouble snuffing him out in a fight, but somehow always surviving and holding his own against supernatural, super powerful, blazingly fast creatures. \n The cover art is awesome and it truly made me excited to read this but everything else feel completely flat for me. I tried telling myself that "it\'s a graphic novel, it would be hard to translate mystery, details, emotion" but then I remembered reading DC Comic\'s "Identity Crisis" and realized that was a load of crap. I know it\'s unfair to compare the two as "Identity Crisis" had popular mystery author Brad Meltzer writing it right? Yeah...no. The standard was set that day and there is more than enough talent out there to create a great story in a graphic novel. \n That being said, it wasn\'t a horrible story, it just didn\'t grip me for feel anything like Sherlock Holmes to me. It was easy enough to follow but I felt no sense of tension, stakes or compassion for any of the characters. \n As far as the vampires go, it\'s hard to know what to expect anymore as there are so many different versions these days. This was the more classic version which I personally prefer, but again I didn\'t find anything that portrayed their dominance, calm confidence or sexuality. There was definitely a presence of their physical prowess but somehow that was lost on me as easily as Sherlock was able to defend himself. I know it, wouldn't do to kill of the main character, but this would have a been a great opportunity to build around the experience and beguiling nature of a vampire that had lived so many years of experience. Another chance to showcase Sherlock\'s intellect in a battle of wits over strength in something more suitable for this sort of story as apposed to trying to make it feel like an action movie. \n Maybe I expected to much and hoped to have at least a gripping premise or some sort of interesting plot or mystery but I didn\'t find it here. This may be a must have for serious Sherlock Holmes fans that have to collect everything about him, but if you are looking for a great story inside a graphic novel, I would have to say pass on this one. \n That artwork is good, cover is great, story is lacking so I am giving it 2.5 out of 5 stars.', 'date\_added': 'Thu Dec 05 10:44:25 -0800 2013', 'date\_updated': 'Thu Dec 05 10:45:15 -0800 2013', 'read\_at': 'Tue Nov 05 00:00:00 -0800 2013', 'started\_at': '', 'n\_votes': 0, 'n\_comments': 0}

```
[4]: userPerItem=defaultdict(set) itemPerUser=defaultdict(set)
```

#### [5]: itemNames={}

```
[6]: for d in data:
         user=d["user_id"]
         item=d["book_id"]
         userPerItem[item].add(user)
         itemPerUser[user].add(item)
 [7]: def Jaccard(s1,s2):
         numer=len(s1.intersection(s2))
         deno=len(s1.union(s2))
         return(numer/deno)
 [8]: def mostSimilar(i,N=10):
         users=userPerItem[i]
         similarities=[]
         for j in userPerItem:
             if(i==j):
                 continue
             sim=Jaccard(users,userPerItem[j])
             similarities.append((sim,j))
         s=sorted(similarities, key=lambda element:(-element[0],element[1]))
          #similarities.sort(reverse=True)
         return(s[:10])
 [9]: w=[]
      for d in data:
         if(d["user_id"]=="dc3763cdb9b2cae805882878eebb6a32"):
             o=[]
             o.append(d["book_id"])
             o.append(d["rating"])
             w.append(o)
      rating_u=sorted(w, key=lambda element:(element[1]),reverse=True)
      query=rating_u[0][0]
      print(query)
     18471619
     1.3 (a) Part
[10]: # Part A
      mostSimilar(query)
(0.14285714285714285, '25659811'),
       (0.13793103448275862, '18369278'),
       (0.13157894736842105, '18430205'),
       (0.12903225806451613, '20299669'),
```

```
(0.125, '17995154'),
       (0.1212121212121222, '18853527'),
       (0.12121212121212122, '23093378'),
       (0.121212121212122, '23241671'),
       (0.11764705882352941, '18734070')]
[11]: from operator import itemgetter
     lis_item=[]
     for d in data:
         if(d["user_id"]=="dc3763cdb9b2cae805882878eebb6a32" and d["rating"]):
             item=[]
             item.append(d["rating"])
             item.append(d["book_id"])
             lis_item.append(item)
     w=(sorted(lis_item, key=itemgetter(0),reverse=True))
     print(w)
     mostSimilar(w[0][1])
     [[3, '18471619']]
(0.14285714285714285, '25659811'),
       (0.13793103448275862, '18369278'),
       (0.13157894736842105, '18430205'),
       (0.12903225806451613, '20299669'),
       (0.125, '17995154'),
       (0.121212121212122, '18853527'),
       (0.121212121212122, '23093378'),
       (0.121212121212122, '23241671'),
       (0.11764705882352941, '18734070')]
     1.4 (b) Part
[12]: e1=[]
     def mostSimilarUser(i,N=10):
         item1=itemPerUser[i]
         for j in itemPerUser:
             similarities=[]
             if(i==j):
                 continue
             sim=Jaccard(item1,itemPerUser[j])
             if(sim==1):
                 continue
             similarities.append(sim)
             similarities.append(j)
             e1.append(similarities)
```

```
return(e1)
      # def mostSimilarUser(i, N=10):
            item1=itemPerUser[i]
            similarities=[]
      #
            for j in itemPerUser:
      #
               if(i==j):
      #
                    continue
               sim=Jaccard(item1, itemPerUser[j])
                similarities.append((sim, j))
            similarities.sort(reverse=True)
            return(similarities)
[13]: kab=[]
      kab=mostSimilarUser("dc3763cdb9b2cae805882878eebb6a32")
      kab.sort(key=lambda x:x[0],reverse=True)
      print("Users similar to given ID: ")
      for i in range(10):
          print(kab[i])
     Users similar to given ID:
     [0.33333333333333333, '6470c7f5e3468ba34e9fe628960fbbf1']
     [0.25, '6497ca91df3c182006874c96a8530b37']
     [0.2, '033cf640dfa6f85eb146c39787289628']
     [0.14285714285714285, '5510684ab6c18f2dd493787e66b2722c']
     [0.030303030303030304, 'a39b4249d201ef5ce5ea553bdd013e66']
     [0.023809523809523808, '42519f961f79b61701bda60787b031cf']
     [0.02040816326530612, '65a7975989734fc6e18b7d2bd2bcb49f']
     [0.014925373134328358, 'Ofafb6f0843124383f4e2c5a2090fb09']
     [0.0136986301369863, '071222e19ae29dc9fdbe225d983449be']
[14]: top_user=[]
      for i in range(10):
          top_user.append(kab[i][1])
[15]: # Part B
      # N most similar users, and recommending each of their their favorite (highest,
      \rightarrowrated) items.
      print("Favourite Items recommended by similar User ID")
      for i in top_user:
          q=itemPerUser[i]
          q=list(q)
          w1 = []
          op=0
```

```
for d in data:
    c1=[]
    if(d["user_id"]==i and d["book_id"] in q):
        c1.append(d["user_id"])
        c1.append(d["book_id"])
        c1.append(d["rating"])
        op=max(op,d["rating"])
    if(len(c1)>0):
        w1.append(c1)
abc=(sorted(w1, key=itemgetter(2,1),reverse=False))
for ab in abc:
    if(ab[2]==op):
        print(ab)
        break
#print(abc)
  abc.sort(key=lambda \ x:x[2],reverse=True)
  print(abc)
```

```
Favourite Items recommended by similar User ID ['6470c7f5e3468ba34e9fe628960fbbf1', '10767466', 4] ['6497ca91df3c182006874c96a8530b37', '17570797', 5] ['033cf640dfa6f85eb146c39787289628', '15704307', 5] ['5510684ab6c18f2dd493787e66b2722c', '10138607', 5] ['17f73ea38e97307935c0d3b6ca987b53', '12434747', 5] ['a39b4249d201ef5ce5ea553bdd013e66', '17995248', 5] ['42519f961f79b61701bda60787b031cf', '10105459', 5] ['65a7975989734fc6e18b7d2bd2bcb49f', '10997645', 5] ['0fafb6f0843124383f4e2c5a2090fb09', '10361139', 5] ['071222e19ae29dc9fdbe225d983449be', '10264328', 5]
```

October 25, 2021

1 Question 3 In class we briefly discussed whether the Pearson similarity should be implemented (a) only in terms of shared items (i.e., Ui Uj) in the denominator; or (b) in terms of all items each user consumed (i.e., Ui or Uj for each term in the denominator). (See last slide on Pearson similarity). Implement versions of the Pearson similarity based on both definitions, and report the 10 most similar items to the same query item from Question 1

```
[1]: import gzip
     from collections import defaultdict
     import scipy
     import scipy.optimize
     import numpy
     import random
[2]: def parseDataFromFile(fname):
         for 1 in open(fname):
             yield eval(1)
[3]: data=list(parseDataFromFile("goodreads_reviews_comics_graphic.json"))
[4]: usersPerItem = defaultdict(set) # Maps an item to the users who rated it
     itemsPerUser = defaultdict(set) # Maps a user to the items that they rated
     itemNames = {}
     ratingDict = {} # To retrieve a rating for a specific user/item pair
     for d in data:
         user,item = d['user_id'], d['book_id']
         usersPerItem[item].add(user)
         itemsPerUser[user].add(item)
         ratingDict[(user,item)] = d['rating']
[5]: userAverages = {}
     itemAverages = {}
```

```
for u in itemsPerUser:
    rs = [ratingDict[(u,i)] for i in itemsPerUser[u]]
    userAverages[u] = sum(rs) / len(rs)

for i in usersPerItem:
    rs = [ratingDict[(u,i)] for u in usersPerItem[i]]
    itemAverages[i] = sum(rs) / len(rs)
```

### 1.1 (a) only in terms of shared items

```
[6]: def sharedPearson(i1, i2):
         # Between two items
         iBar1 = itemAverages[i1]
         iBar2 = itemAverages[i2]
         inter = usersPerItem[i1].intersection(usersPerItem[i2])
         numer = 0
         denom1 = 0
         denom2 = 0
         for u in inter:
             numer += (ratingDict[(u,i1)] - iBar1)*(ratingDict[(u,i2)] - iBar2)
         for u in inter: #usersPerItem[i1]:
             denom1 += (ratingDict[(u,i1)] - iBar1)**2
         #for u in usersPerItem[i2]:
             denom2 += (ratingDict[(u,i2)] - iBar2)**2
         denom = math.sqrt(denom1) * math.sqrt(denom2)
         if denom == 0: return 0
         return numer / denom
[7]: def mostSimilar(i, N):
         similarities = []
         users = usersPerItem[i]
         for i2 in usersPerItem:
```

```
[8]: query = data[0]['book_id']
```

```
[9]: import math
ms = mostSimilar(query, 10)
```

```
[10]: ms[0:10]
```

#### 1.2 (b) in terms of all items each user consumed

```
[11]: def unionPearson(i1, i2):
          # Between two items
          iBar1 = itemAverages[i1]
          iBar2 = itemAverages[i2]
          inter1 = usersPerItem[i1].intersection(usersPerItem[i2])
          numer = 0
          denom1 = 0
          denom2 = 0
          for u in inter1:
              numer += (ratingDict[(u,i1)] - iBar1)*(ratingDict[(u,i2)] - iBar2)
          for u in usersPerItem[i1]: #usersPerItem[i1]:
              denom1 += (ratingDict[(u,i1)] - iBar1)**2
          for u in usersPerItem[i2]:
              denom2 += (ratingDict[(u,i2)] - iBar2)**2
          denom = math.sqrt(denom1) * math.sqrt(denom2)
          if denom == 0: return 0
          return numer / denom
```

```
[12]: def mostSimilarUnion(i, N):
    similarities = []
    users = usersPerItem[i]
    for i2 in usersPerItem:
        if i2 == i: continue
        sim = unionPearson(i, i2) # Could use alternate similarity metricsustraightforwardly
        similarities.append((sim,i2))
    #similarities.sort(reverse=True)
    s=sorted(similarities, key=lambda element:(-element[0],element[1]))
    return s
```

```
[13]: query = data[0]['book_id']
print(query)
```

18471619

October 25, 2021

1 Question 4 Implement a rating prediction model based on the similarity function Report the MSE of this rating prediction function when Sim(i, j) = Jaccard(i, j)

```
[1]: import gzip
     from collections import defaultdict
     import scipy
     import scipy.optimize
     import numpy
     import random
[2]: def parseDataFromFile(fname):
         for 1 in open(fname):
             yield eval(1)
[3]: data=list(parseDataFromFile("goodreads_reviews_comics_graphic.json"))
     print(data[1])
    {'user_id': 'bafc2d50014200cda7cb2b6acd60cd73', 'book_id': '6315584',
    'review_id': '72f1229aba5a88f9e72f0dcdc007dd22', 'rating': 4, 'review_text':
    "I've never really liked Spider-Man. I am, however, a huge fan of the Dresden
    Files. Jim Butcher is clever and sarcastic and probably the perfect choice to
    pen a superhero novel. I really enjoyed this book!", 'date_added': 'Wed Aug 10
    06:06:48 -0700 2016', 'date_updated': 'Fri Aug 12 08:49:54 -0700 2016',
    'read_at': 'Fri Aug 12 08:49:54 -0700 2016', 'started_at': 'Wed Aug 10 00:00:00
    -0700 2016', 'n_votes': 0, 'n_comments': 0}
[4]: reviewsPerUser = defaultdict(list)
     reviewsPerItem = defaultdict(list)
[5]: for d in data:
         user,item = d['user_id'], d['book_id']
         reviewsPerUser[user].append(d)
         reviewsPerItem[item].append(d)
[6]: usersPerItem = defaultdict(set) # Maps an item to the users who rated it
     itemsPerUser = defaultdict(set) # Maps a user to the items that they rated
```

```
itemNames = {}
      ratingDict = {} # To retrieve a rating for a specific user/item pair
      for d in data:
          user,item = d['user_id'], d['book_id']
          usersPerItem[item].add(user)
          itemsPerUser[user].add(item)
          ratingDict[(user,item)] = d['rating']
 [7]: userAverages = {}
      itemAverages = {}
      for u in itemsPerUser:
          rs = [ratingDict[(u,i)] for i in itemsPerUser[u]]
          userAverages[u] = sum(rs) / len(rs)
      for i in usersPerItem:
          rs = [ratingDict[(u,i)] for u in usersPerItem[i]]
          itemAverages[i] = sum(rs) / len(rs)
 [8]: ratingMean = sum([d['rating'] for d in data]) / len(data)
 [9]: def Jaccard(s1, s2):
          numer = len(s1.intersection(s2))
          denom = len(s1.union(s2))
          if denom == 0:
              return 0
          return numer / denom
[10]: def predictRating(user,item):
          ratings = []
          similarities = []
          for d in reviewsPerUser[user]:
              i2 = d['book_id']
              if i2 == item: continue
              ratings.append(d['rating'] - itemAverages[i2])
              similarities.append(Jaccard(usersPerItem[item],usersPerItem[i2]))
          if (sum(similarities) > 0):
              weightedRatings = [(x*y) for x,y in zip(ratings,similarities)]
              return itemAverages[item] + sum(weightedRatings) / sum(similarities)
          else:
              # User hasn't rated any similar items
              return itemAverages[item]
```

### 1.1 For 10,000 Items

```
[11]: predictions=[]
      for d in data[:10000]:
          t=predictRating(d["user_id"],d["book_id"])
          predictions.append(t)
[12]: def MSE(predictions, labels):
          differences = [(x-y)**2 \text{ for } x,y \text{ in } zip(predictions,labels)]
          return sum(differences) / len(differences)
[13]: labels = [d['rating'] for d in data[:10000]]
[14]: print("MSE for 10,000 items")
      MSE(predictions, labels)
     MSE for 10,000 items
[14]: 0.7017041185560355
     1.2 For all Items
[15]: predictions=[]
      for d in data:
          t=predictRating(d["user_id"],d["book_id"])
          predictions.append(t)
[16]: def MSE(predictions, labels):
          differences = [(x-y)**2 \text{ for } x,y \text{ in } zip(predictions,labels)]
          return sum(differences) / len(differences)
[17]: labels = [d['rating'] for d in data]
[18]: print("MSE for all items")
      MSE(predictions, labels)
     MSE for all items
[18]: 0.7908367015187353
```

October 25, 2021

### 1 Question 6 Recommender Systems based on Temporal Dynamics.

```
[1]: import gzip
     from collections import defaultdict
     import scipy
     import scipy.optimize
     import numpy
     import random
     import math
[2]: def parseDataFromFile(fname):
         for 1 in open(fname):
             yield eval(1)
[3]: data=list(parseDataFromFile("goodreads_reviews_comics_graphic.json"))
     print(data[1])
    {'user_id': 'bafc2d50014200cda7cb2b6acd60cd73', 'book_id': '6315584',
    'review_id': '72f1229aba5a88f9e72f0dcdc007dd22', 'rating': 4, 'review_text':
    "I've never really liked Spider-Man. I am, however, a huge fan of the Dresden
    Files. Jim Butcher is clever and sarcastic and probably the perfect choice to
    pen a superhero novel. I really enjoyed this book!", 'date_added': 'Wed Aug 10
    06:06:48 -0700 2016', 'date_updated': 'Fri Aug 12 08:49:54 -0700 2016',
    'read_at': 'Fri Aug 12 08:49:54 -0700 2016', 'started_at': 'Wed Aug 10 00:00:00
    -0700 2016', 'n_votes': 0, 'n_comments': 0}
[4]: reviewsPerUser = defaultdict(list)
     reviewsPerItem = defaultdict(list)
[5]: for d in data:
         user,item = d['user_id'], d['book_id']
         reviewsPerUser[user].append(d)
         reviewsPerItem[item].append(d)
[6]: import dateutil.parser
     import time
     d1=data[0]
     dt1 = dateutil.parser.parse(d1["date_added"])
```

```
q1=int(time.mktime(dt1.timetuple()))
      mi=q1
      ma=0
      usersPerItem = defaultdict(set) # Maps an item to the users who rated it
      itemsPerUser = defaultdict(set) # Maps a user to the items that they rated
      itemNames = {}
      ratingDict = {} # To retrieve a rating for a specific user/item pair
      timeStamp = {}
      for d in data:
          user,item = d['user_id'], d['book_id']
          usersPerItem[item].add(user)
          itemsPerUser[user].add(item)
          ratingDict[(user,item)] = d['rating']
          dt = dateutil.parser.parse(d["date_added"])
          q=int(time.mktime(dt.timetuple()))
          mi=min(mi,q)
          ma=max(ma,q)
          timeStamp[(user,item)]=q
[7]: for d in data:
          oab=timeStamp[(d["user_id"],d["book_id"])]
          oab=oab-mi
          ran=ma-mi
          timeStamp[(d["user_id"],d["book_id"])]=(oab/ran)
[8]: userAverages = {}
      itemAverages = {}
      for u in itemsPerUser:
          rs = [ratingDict[(u,i)] for i in itemsPerUser[u]]
          userAverages[u] = sum(rs) / len(rs)
      for i in usersPerItem:
          rs = [ratingDict[(u,i)] for u in usersPerItem[i]]
          itemAverages[i] = sum(rs) / len(rs)
[9]: ratingMean = sum([d['rating'] for d in data]) / len(data)
      avgTimeList=[timeStamp[key] for key in timeStamp]
      avgTime=sum(avgTimeList)/len(avgTimeList)
[10]: def timeFunction(item, i2):
          eConstant=math.e
          co=timeStamp[(user,item)]
          lamdat=-5*(abs(co-avgTime)/avgTime)
          val=math.pow(eConstant,lamdat)
          return(val)
```

```
[11]: def Jaccard(s1, s2):
          numer = len(s1.intersection(s2))
          denom = len(s1.union(s2))
          if denom == 0:
              return 0
          return numer / denom
[12]: def predictRating(user, item):
          ratings = []
          similarities = []
          for d in reviewsPerUser[user]:
              i2 = d['book_id']
              if i2 == item: continue
              ratings.append(d['rating'] - itemAverages[i2])
              ab=Jaccard(usersPerItem[item],usersPerItem[i2])
              \#timefunction
              t1=timeStamp[(user,item)]
              t2=timeStamp[(user,i2)]
              tabs=abs(t1-t2)
              eConstant=math.e
              lamdat=-5*tabs
              val=math.pow(eConstant,lamdat)
              ab=ab*val
              similarities.append(ab)
          if (sum(similarities) > 0):
              weightedRatings = [(x*y) for x,y in zip(ratings,similarities)]
              return itemAverages[item] + sum(weightedRatings) / sum(similarities)
          else:
              # User hasn't rated any similar items
              return itemAverages[item]
```

### 1.1 For 10,000 Items

```
[13]: predictions=[]
    for d in data[:10000]:
        t=predictRating(d["user_id"],d["book_id"])
        predictions.append(t)

[14]: def MSE(predictions, labels):
        differences = [(x-y)**2 for x,y in zip(predictions,labels)]
        return sum(differences) / len(differences)
[15]: labels = [d['rating'] for d in data[:10000]]
```

```
[16]: print("MSE for 10,000 items")
      MSE(predictions, labels)
     MSE for 10,000 items
[16]: 0.6974678265068076
     1.2 For all Items
[17]: predictions=[]
      for d in data:
          t=predictRating(d["user_id"],d["book_id"])
          predictions.append(t)
[18]: def MSE(predictions, labels):
          differences = [(x-y)**2 \text{ for } x,y \text{ in } zip(predictions,labels)]
          return sum(differences) / len(differences)
[19]: labels = [d['rating'] for d in data]
[20]: print("MSE for all items")
      MSE(predictions, labels)
     MSE for all items
```

[20]: 0.7827220281567892