Data Science with Python Project

Building user-based recommendation model for Amazon

DESCRIPTION

The dataset provided contains movie reviews given by Amazon customers. Reviews were given between May 1996 and July 2014.

analysis

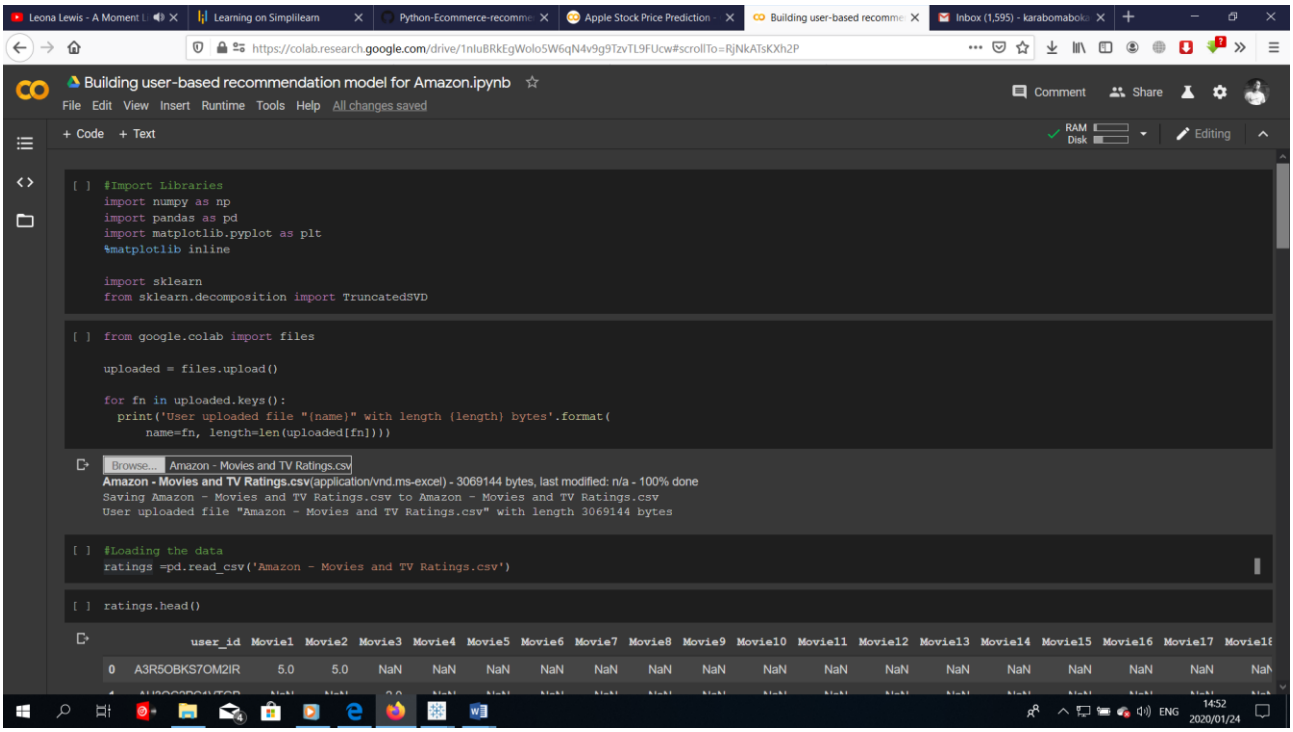
- Exploratory Data Analysis:

- Which movies have maximum views/ratings?
- What is the average rating for each movie? Define the top 5 movies with the maximum ratings.
- Define the top 5 movies with the least audience.

- Recommendation Model: Some of the movies hadn't been watched and therefore, are not rated by the users. Netflix would like to take this as an opportunity and build a machine learning recommendation algorithm which provides the ratings for each of the users.

- Divide the data into training and test data
- Build a recommendation model on training data
- Make predictions on the test data

Screenshots


Building user-based recommendation model for Amazon.ipynb
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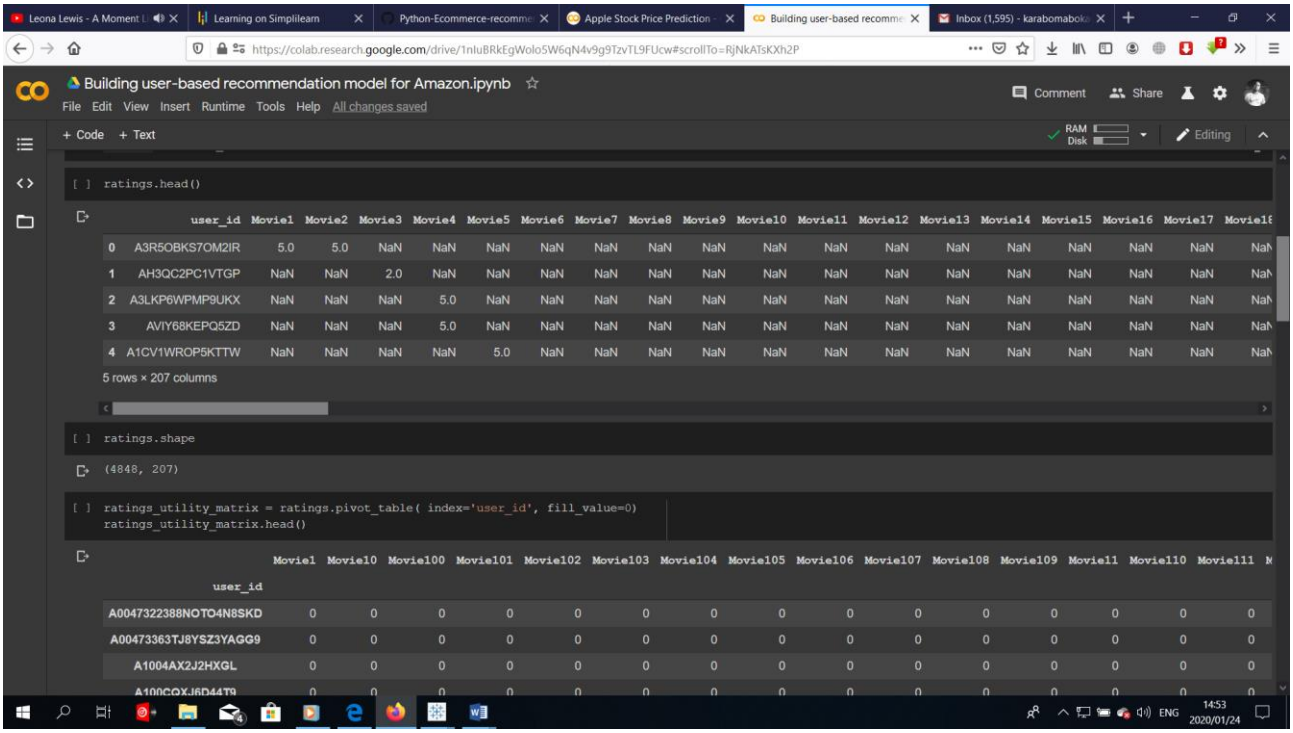
#Import Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import sklearn
from sklearn.decomposition import TruncatedSVD
from google.colab import files
uploaded = files.upload()
for fn in uploaded.keys():
print('User uploaded file "{name}" with length {length} bytes'.format(
name=fn, length=len(uploaded[fn])))
Browse... Amazon - Movies and TV Ratings.csv
Amazon - Movies and TV Ratings.csv(application/vnd.ms-excel) - 3069144 bytes, last modified: n/a - 100% done
Saving Amazon - Movies and TV Ratings.csv to Amazon - Movies and TV Ratings.csv
User uploaded file "Amazon - Movies and TV Ratings.csv" with length 3069144 bytes
#Loading the data
ratings =pd.read_csv('Amazon - Movies and TV Ratings.csv')
ratings.head()
user_id Movie1 Movie2 Movie3 Movie4 Movie5 Movie6 Movie7 Movie8 Movie9 Movie10 Movie11 Movie12 Movie13 Movie14 Movie15 Movie16 Movie17
0 A3R5OBKS7OM2IR 5.0 5.0 NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
ENG 14:52 2020/01/24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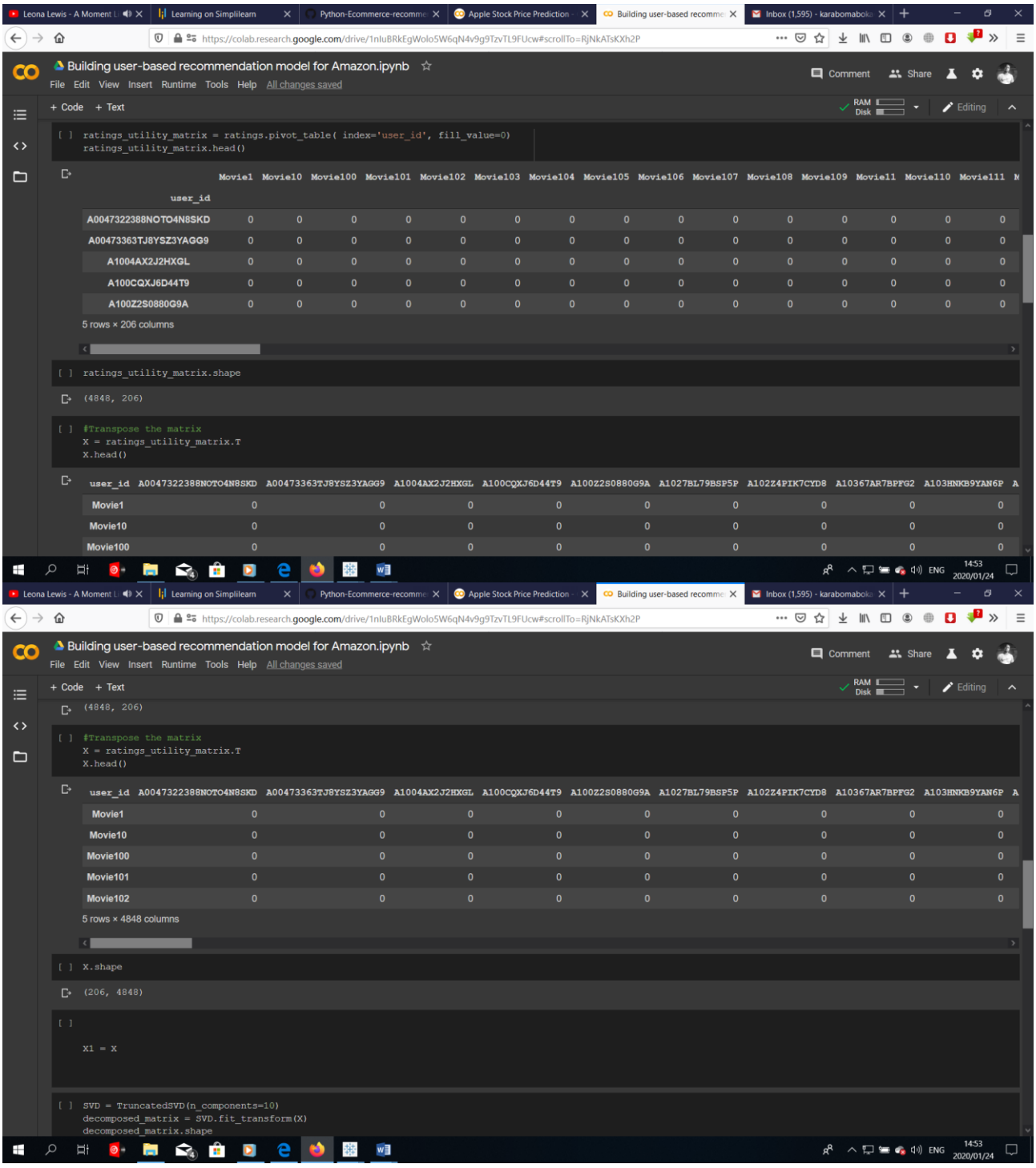
user_id Movie1 Movie2 Movie3 Movie4 Movie5 Movie6 Movie7 Movie8 Movie9 Movie10 Movie11 Movie12 Movie13 Movie14 Movie15 Movie16 Movie17
0 A3R5OBKS7OM2IR 5.0 5.0 NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
1 AH3QC2PC1VTGP NaN NaN 2.0 NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
2 A3LKP6WPMP9UKX NaN NaN NaN 5.0 NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
3 AVIY68KEPQ5ZD NaN NaN NaN 5.0 NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
4 A1CV1WROP5KTTW NaN NaN NaN NaN 5.0 NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
5 rows × 207 columns
ratings.shape
(4848, 207)
ratings_utility_matrix = ratings.pivot_table(index='user_id', fill_value=0)
ratings_utility_matrix.head()
Movie1 Movie10 Movie100 Movie101 Movie102 Movie103 Movie104 Movie105 Movie106 Movie107 Movie108 Movie109 Movie11 Movie110 Movie111
user_id
A0047322388NOTO4N8SKD 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
A00473363TJ8YSZ3YAGG9 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
A1004AX2J2HXGL 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
ENG 14:53 2020/01/24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[ ] ratings_utility_matrix = ratings.pivot_table(index='user_id', fill_value=0)
ratings_utility_matrix.head()
user_id Movie1 Movie10 Movie100 Movie101 Movie102 Movie103 Movie104 Movie105 Movie106 Movie107 Movie108 Movie109 Movie11 Movie110 Movie111
A0047322388NOTO4N8SKD 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
A00473363TJ8YSZ3YAGG9 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
A1004AX2J2HXGL 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
A100CQXJ6D44T9 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
A100Z2S0880G9A 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
5 rows × 206 columns
[ ] ratings_utility_matrix.shape
(4848, 206)
[ ] #Transpose the matrix
X = ratings_utility_matrix.T
X.head()
user_id A0047322388NOTO4N8SKD A00473363TJ8YSZ3YAGG9 A1004AX2J2HXGL A100CQXJ6D44T9 A100Z2S0880G9A A1027BL79BSP5P A102Z4PIK7CYD8 A10367AR7BPFG2 A103HNKB9YAN6P
Movie1 0 0 0 0 0 0 0 0 0
Movie10 0 0 0 0 0 0 0 0 0
Movie100 0 0 0 0 0 0 0 0 0
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X.head()
user_id A0047322388NOTO4N8SKD A00473363TJ8YSZ3YAGG9 A1004AX2J2HXGL A100CQXJ6D44T9 A100Z2S0880G9A A1027BL79BSP5P A102Z4PIK7CYD8 A10367AR7BPFG2 A103HNKB9YAN6P
Movie1 0 0 0 0 0 0 0 0 0
Movie10 0 0 0 0 0 0 0 0 0
Movie100 0 0 0 0 0 0 0 0 0
Movie101 0 0 0 0 0 0 0 0 0
Movie102 0 0 0 0 0 0 0 0 0
5 rows × 4848 columns
[ ] X.shape
(206, 4848)
[ ]
X1 = X
[ ] SVD = TruncatedSVD(n_components=10)
decomposed_matrix = SVD.fit_transform(X)
decomposed_matrix.shape
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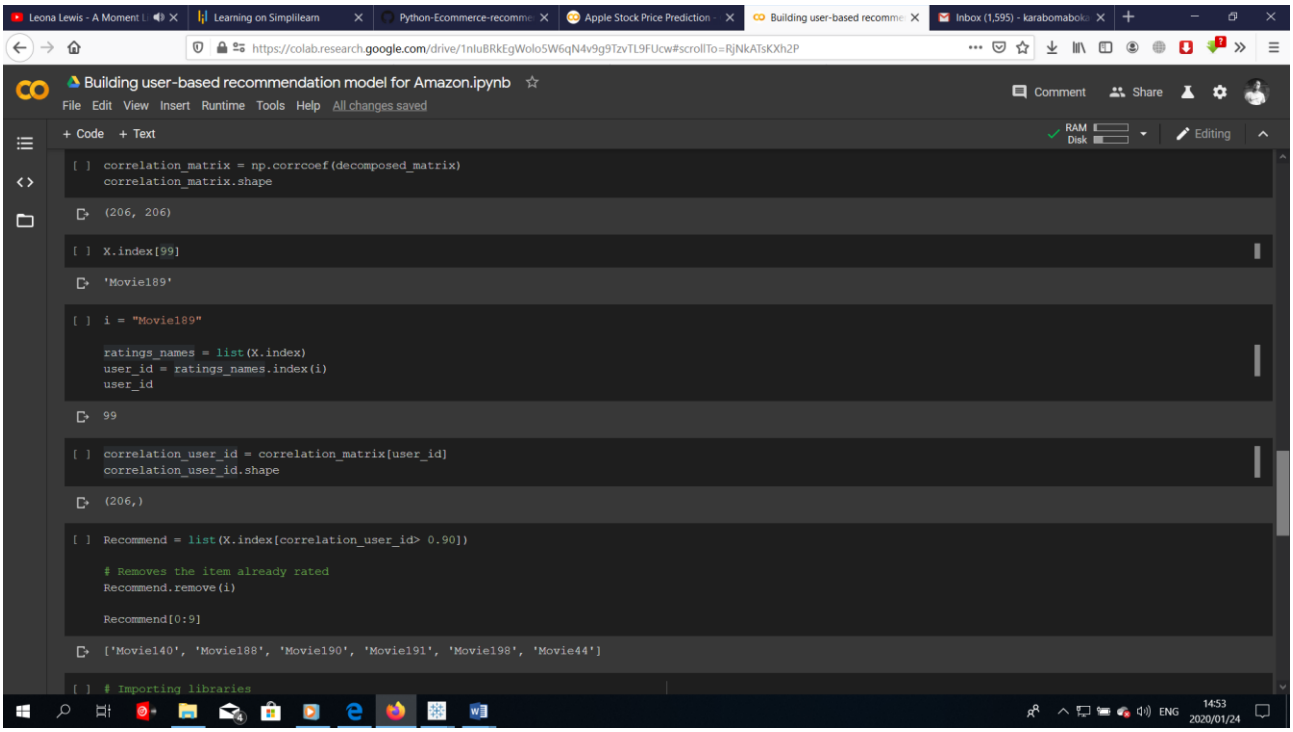
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[ ] correlation_matrix = np.corrcoef(decomposed_matrix)
correlation_matrix.shape
(206, 206)
[ ] X.index[99]
'Movie189'
[ ] i = "Movie189"
ratings_names = list(X.index)
user_id = ratings_names.index(i)
user_id
99
[ ] correlation_user_id = correlation_matrix[user_id]
correlation_user_id.shape
(206,)
[ ] Recommend = list(X.index[correlation_user_id> 0.90])
# Removes the item already rated
Recommend.remove(i)
Recommend[0:9]
['Movie140', 'Movie188', 'Movie190', 'Movie191', 'Movie198', 'Movie44']
[ ] # Importing libraries
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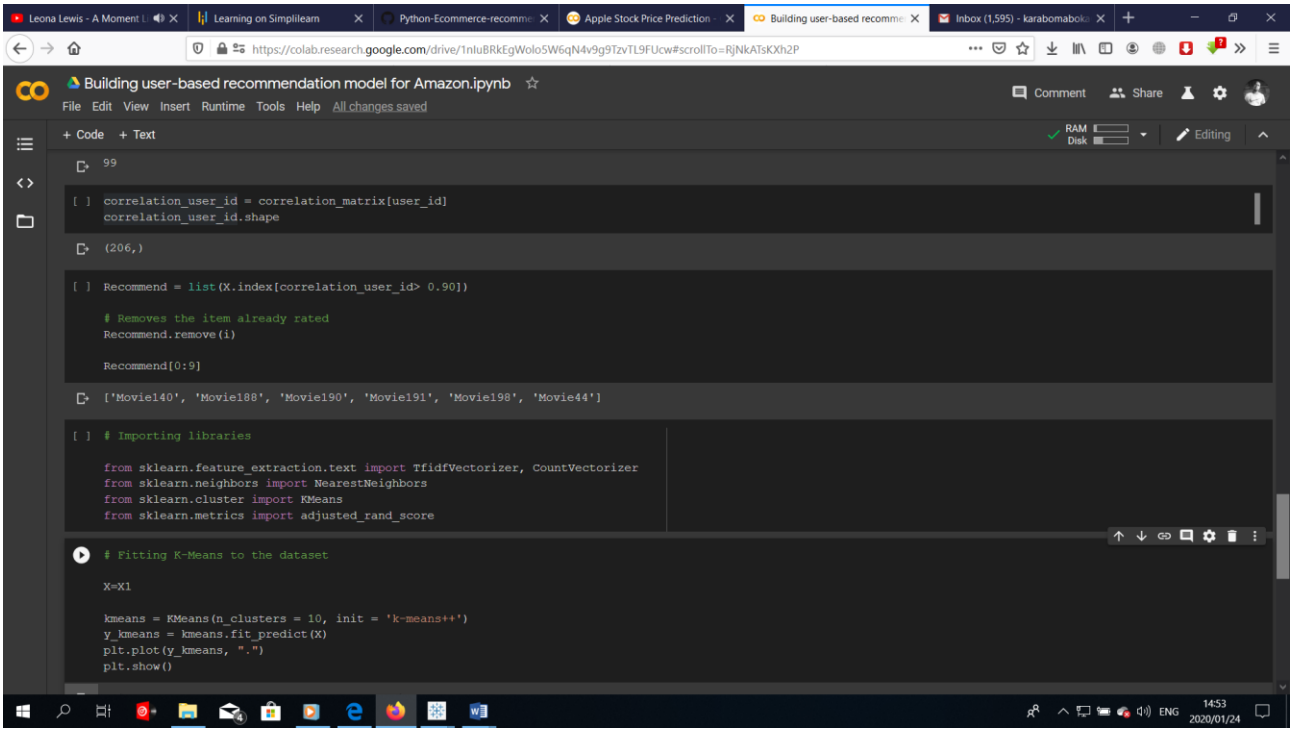
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99
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correlation_user_id.shape
(206,)
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# Removes the item already rated
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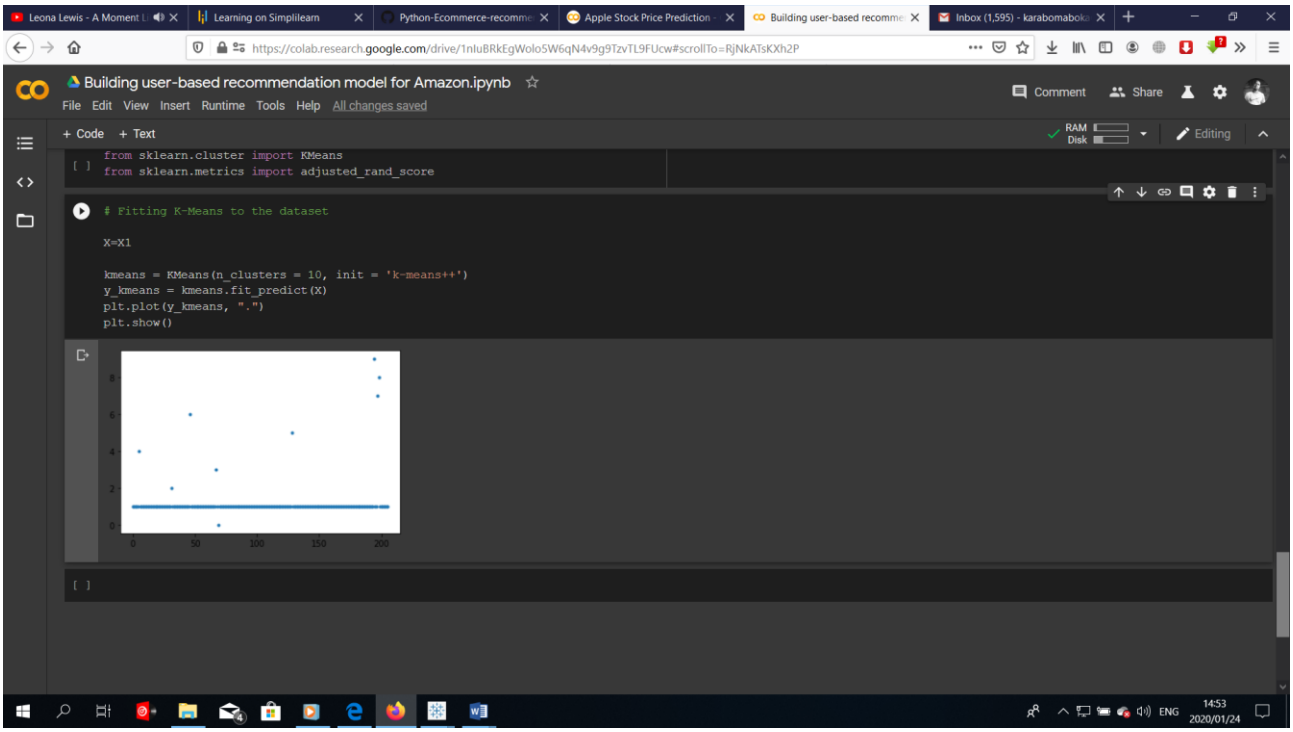
Recommend[0:9]
['Movie140', 'Movie188', 'Movie190', 'Movie191', 'Movie198', 'Movie44']
[ ] # Importing libraries
from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer
from sklearn.neighbors import NearestNeighbors
from sklearn.cluster import KMeans
from sklearn.metrics import adjusted_rand_score
# Fitting K-Means to the dataset
X=X1
kmeans = KMeans(n_clusters = 10, init = 'k-means++')
y_kmeans = kmeans.fit_predict(X)
plt.plot(y_kmeans, ".")
plt.show()
ENG
14:53
2020/01/24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plt.plot(y_kmeans, ".")
plt.show()
0
2
4
6
8
0
50
100
150
200
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