

Team Esh: Chris Muro, Andrew Greensweight, Marc DiGeronimo

Music Recommender

Submit Prediction

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This leaderboard is calculated with approximately 60% of the test data. The final results will be based on the other 40%, so the final standings may be different.

#	Team	Members	Score	Entries	Last	Join
1	Anthony Paolantonio		0.876	6	18h	
2	TEAM ROCKET		0.870	7	14h	
3	Shashankk Shekar Chaturvedi		0.869	4	12d	
4	Derick Miller		0.868	25	2d	
5	jdawg		0.868	11	16d	
6	Brandon Tai		0.865	34	16h	
7	N-R Team		0.856	24	12d	
8	esh		0.848	36	1d	
<div> Your Best Entry! Your submission scored 0.624, which is not an improvement of your previous score. Keep trying!</div>						
9	Apex		0.846	5	12d	
10	Eric Liotta		0.803	11	16d	

Current Score = .848 Total Submissions = 36

Part 2: For part 2 of the lab, `re_u.data` file was used. It contains 3 columns of data being the same as the `trainItem.data` and `testItem.data` files in the previous part which are in order from left to right, `userID`, `itemID` and the rating. First the environment must be configured properly on google colab. This is done by installing the proper programs and setting the environment variables and uploading the file. Also `re_u.data` needs to be split into the train and test set. To do this the dataset was split using `np.array_split()` function. For 1. `maxIter` is set to 20 and the rank sizes of 5,7,10 and 20 are tested. With Constant `maxIter` 20 and rank = 5 the mse was 1.1893121240129518. Rank 7 mse = 1.2858819425771069, Rank = 10, mse = 1.322679774938557, Rank =20 mse = 1.402295299515129. Therefore for these observations we can conclude that at same `maxIter` and `data_size` values, as rank increases, mse also increases. This would lead us to using lower rank number for possible models. For 2. rank is set to equal a constant 10 as `maxIter` takes on the possible values of 2,5,7,10. For `maxIter` of 2, mse = 1.3911330043443335. `MaxIter` = 5, mse = 1.315439824961866, `MaxIter` = 7, mse = 1.3192493847018245. `MaxIter` = 10, mse = 1.3202044450326698. The results are very interesting. It is assumed that the more iterations the more accurate the model and the less the mse. That is true for this data only to a point however. The lowest mse was at `MaxIter` 5 and then the mse gradually increased. This may be explained by overfitting as the more iterations

the more closely the model will resemble the training set and lose precision on the test set. This would lead us to choose a lower MaxIter value such as 5 but not too low as 2. However this does not seem to have a drastic impact overall on the mse. For 3, rank is fixed to constant 10, and maxIter is set to constant 20, Here different sizes of data are taken for the size of the train and test set, the values being 2000, 5000, 10000, 20000, 50000 and 100000. For data_size = 2000, mse = 2.9512606448263443. Data_size = 5000, mse = 2.1864230858059064, Data_size = 10000, mse = 2.0322522978298445, Data_size = 20000, mse = 1.6178100700769171. Data_size = 50000, mse = 1.322679774938557. Data_size = 100000, mse = 1.322679774938557. From the results of our data, we can conclude that for the dataset increase in data size is inversely proportional to mse. This would lead us to choose the highest possible data size for our model. Which is consistent with our prediction as more data means more trends and patterns can be identified and averaged out and outliers become less impactful. From the parameters tested in this HW, we conclude that the data_size will change MSE value most significantly with MaxIter having the least significant impact for our data and models. There are more parameter sets tested with their corresponding mse in the provided .ipynb file for part 2. Highest MSE = 12.89519086944308 occurs with rank = 7 maxIter = 2 and data_size = 2000. Lowest MSE = 1.1884060045633205 occurs with rank = 5, maxIter = 10 and data_size = 100000.





