

## Results For Dataset data-40-20-2 Dataset:

RMSE for Baseline model on data-40-20-2 dataset: 1.6038626145141004

RMSE for Matrix factorization algorithm on data-40-20-2 dataset: 1.3325094901844552

Time Taken: 0.4574759006500244

We can see that Matrix decomposition algorithm performs better than baseline model.

For time saving algorithm was run on multiple notebooks simultaneously.

## Part1: Experiments on data-500-500-3 Dataset:

Results for different iterations:

RMSE for Baseline model on data-500-500-3 dataset:  
1.8085931611785533

Trainset RMSE for Matrix factorization algorithm on data-500-500-3 dataset:

[0.2350533238219942, 0.18651915195041363, 0.18653752665902773, 0.18649137254510384, 0.18648475872645406, 0.18648231692572093, 0.18648868486518597, 0.18647868221033498, 0.18648484576289592, 0.18647362085452987]

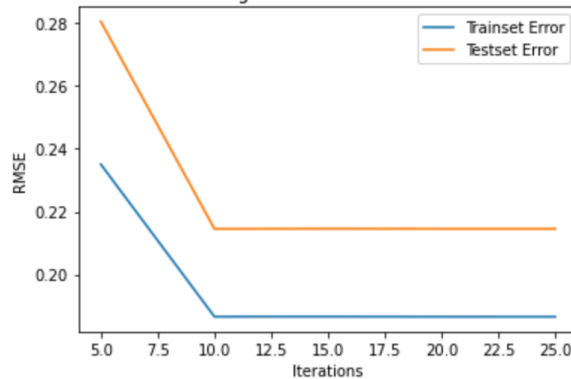
Testset RMSE for Matrix factorization algorithm on data-500-500-3 dataset:

[0.2805822726867546, 0.21449882923560504, 0.2145747104860485, 0.21452263044957218, 0.21451989099456598, 0.21453492728169743, 0.21454668961126128, 0.214521145852022, 0.21453186677499447, 0.21452204697207156]

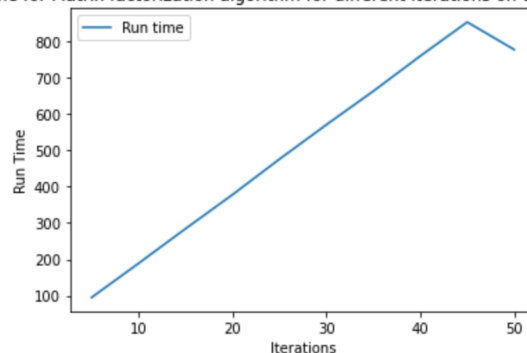
Time taken for each iterations on data-500-500-3 dataset:

[95.00797533988953, 188.72331619262695, 284.1239380836487, 377.6744041442871, 475.4104208946228, 570.1608538627625, 662.579155921936, 759.1655380725861, 852.5149669647217, 776.8515269756317]

Plot of RMSE for Matrix factorization algorithm for different iterations on data-500-500-3 Dataset



Plot of Run Time for Matrix factorization algorithm for different iterations on data-500-500-3 Dataset



Above results shows that Matrix decomposition model outperforms the baseline model. From the plot we can see that error rate decreases with increase in the number of iterations, and also model converges at 15 iterations. Error on test set is higher compared to training set due to prediction on unseen data. Model doesn't overfit with the increased number of iterations. Run time of the model increases almost linearly with higher iterations as shown in the plot.

Results for different Ranks:

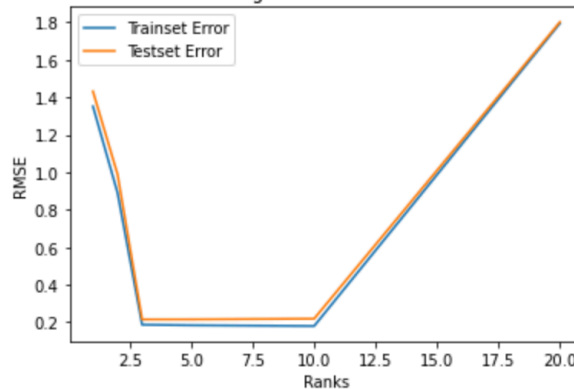
RMSE for Baseline model on data-500-500-3 dataset:  
1.8085931611785533

Trainset RMSE for Matrix factorization algorithm on data-500-500-3 dataset:  
[1.351636701144638, 0.8901490287602056, 0.18646247604032903, 0.1834770163038688, 0.17930531803251487, 1.7947885122906055]

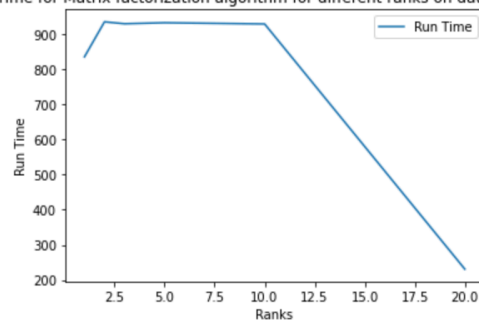
Testset RMSE for Matrix factorization algorithm on data-500-500-3 dataset:  
[1.4317217382532492, 0.9891146654528372, 0.21450686732159147, 0.21484491688130747, 0.2184340988586387, 1.8020866041221522]

Time taken for each ranks on data-500-500-3 dataset:  
[835.5827040672302, 935.2584528923035, 929.7639081478119, 932.6789197921753, 929.161426782608, 230.908264875412]

Plot of RMSE for Matrix factorization algorithm for different ranks on data-500-500-3 Dataset



Plot of Run Time for Matrix factorization algorithm for different ranks on data-500-500-3 Dataset



From the above results we can see that error decreases with the increase in rank, and coverages at rank 3. The large spike in at rank 20 is due to termination of algorithm based on  $\|u^- - u^{-old}\| < 0.01$  following condition which relates to the less run time. Since normally run time increases with increase in rank, which is reflected until rank 10.

## Part2: Experiments on ml-100k Dataset:

Results for different iterations:

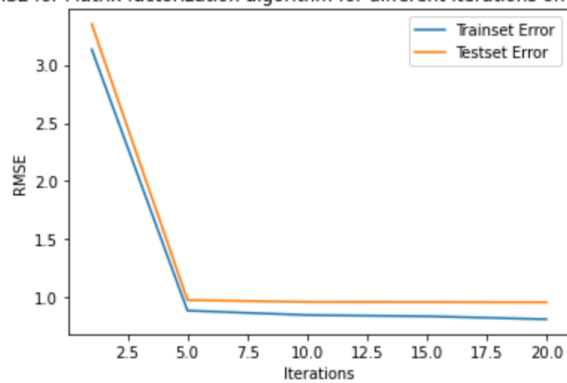
RMSE for Baseline model on ml-100k dataset:  
0.987077661785431

Trainset RMSE for Matrix factorization algorithm on ml-100k dataset:  
[3.1346149379000994, 0.8856999821064543, 0.8475142178717391, 0.8371820833241175, 0.8114342626193253]

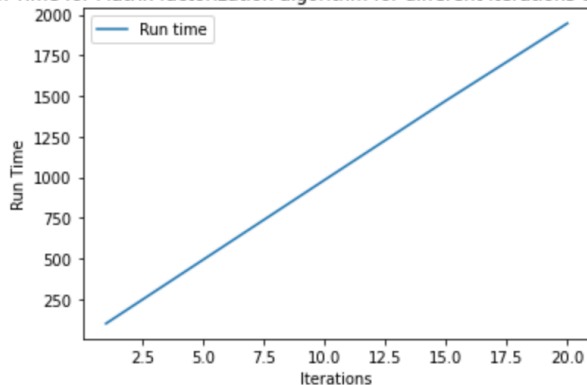
Testset RMSE for Matrix factorization algorithm on ml-100k dataset:  
[3.3523981254162605, 0.9764807805268304, 0.960245923231339, 0.9596497376779604, 0.9572013363790863]

Time taken for each iterations on ml-100k dataset:  
[97.9582929611206, 491.2160429954529, 982.4216120243073, 1471.5200171470642, 1947.568200826645]

Plot of RMSE for Matrix factorization algorithm for different iterations on ml-100k Dataset



Plot of Run Time for Matrix factorization algorithm for different iterations on ml-100k Dataset



Above results shows that Matrix decomposition model performs better the baseline model. From the plot we can see that error rate decreases with increase in the number of iterations, and also model converges at 10 iterations. Error on test set is higher compared to training set due to prediction on unseen data. Model doesn't overfit with the increased number of iterations.

Run time of the model increases almost linearly with higher iterations as shown in the plot.

Results for different Ranks:

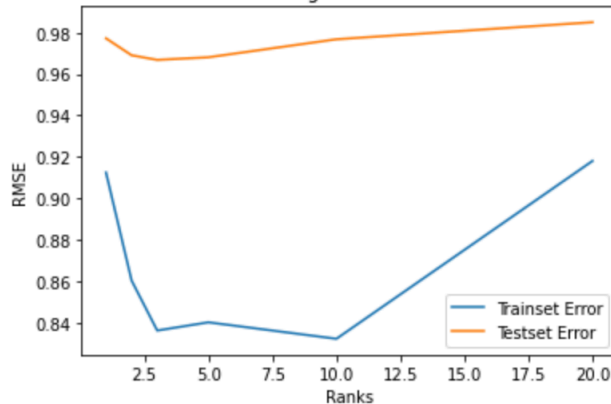
RMSE for Baseline model on ml-100k dataset:  
0.987077661785431

Trainset RMSE for Matrix factorization algorithm on ml-100k dataset:  
[0.9124310093663421, 0.8602739410091855, 0.8361557031762931, 0.8401142798960715, 0.8321635916508645, 0.918029674117959]

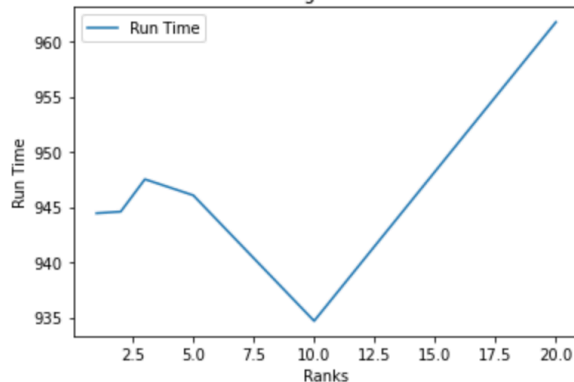
Testset RMSE for Matrix factorization algorithm on ml-100k dataset:  
[0.9772580111982806, 0.9690775491187654, 0.9668531204629545, 0.9681123906224746, 0.9768082803871406, 0.9850338182377717]

Time taken for each ranks on ml-100k dataset:  
[944.4395041465759, 944.579163312912, 947.5134961605072, 946.0610370635986, 934.6771709918976, 961.7496240139008]

Plot of RMSE for Matrix factorization algorithm for different ranks on ml-100k Dataset



Plot of Run Time for Matrix factorization algorithm for different ranks on ml-100k Dataset



From the above results we can see that error decreases with the increase in rank, and coverages at rank 10. The large spike in at rank 20 is due to termination of algorithm based on  $\|u^- - u^{old}\| < 0.01$  following condition which relates to the less run time. Since normally run time increases with increase in rank, which is reflected until rank 10.

### Extra Credits:

Matrix U and V were initialized to the results obtained after one iteration of EM SVD algorithm.

Results For Dataset data-40-20-2 Dataset:

RMSE for Baseline model on data-40-20-2 dataset: 1.6038626145141004

RMSE for Matrix factorization algorithm on data-40-20-2 dataset: 1.3319674674990984

Time Taken: 0.5028479099273682

We can see that Matrix decomposition algorithm performs better than baseline model. But does make much difference from the normal random initialization on small dataset.

Results for different iterations on data-500-500-3 Dataset:

RMSE for Baseline model on data-500-500-3 dataset:  
1.8085931611785533

Trainset RMSE for Matrix factorization algorithm on data-500-500-3 dataset:

[0.1917241670442899, 0.18652944236658786, 0.1865204994990351, 0.18651267829751308, 0.18650551219712438, 0.18649892510298466, 0.1864928528145436, 0.18648724007344605, 0.1864820390740893, 0.18647720826378794]

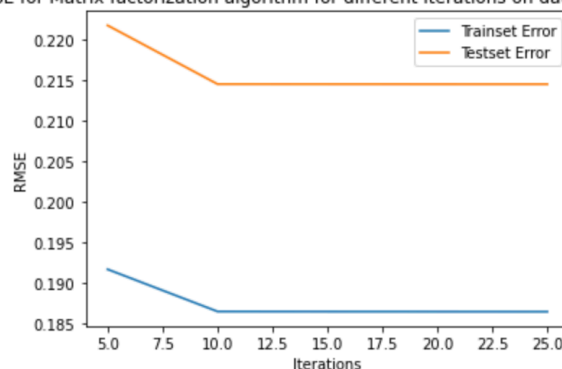
Testset RMSE for Matrix factorization algorithm on data-500-500-3 dataset:

[0.22178431829595627, 0.214550589316961, 0.21454790739431567, 0.21454287203400296, 0.2145383376961641, 0.21453425211015406, 0.2145305599279776, 0.2145272139873927, 0.21452417394484627, 0.21452140512354212]

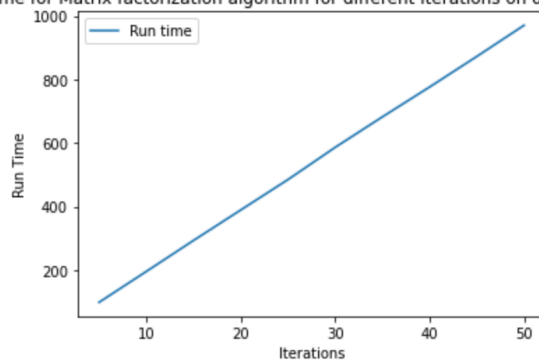
Time taken for each iterations on data-500-500-3 dataset:

[98.27072596549988, 195.39876198768616, 293.1208381652832, 389.08395195007324, 484.8011908531189, 586.3320031166077, 682.4391779899597, 776.4937698841095, 872.7533288002014, 970.8264882564545]

Plot of RMSE for Matrix factorization algorithm for different iterations on data-500-500-3 Dataset



Plot of Run Time for Matrix factorization algorithm for different iterations on data-500-500-3 Dataset



Results shows that the matrix decomposition algorithm the EM SVD Initialization converges faster at iteration 10 compared to the model with normal random initialization. And also we see lower initial error rates when compared to the other model. But the outcome for both models is same.