prabhudayala@gmail.com_Anonymous Ratings Data from the Jester Online Joke Recommender System

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0.1 Anonymous Ratings Data from the Jester Online Joke Recommender System

0.2 Data source:

Description: Jester dataset is provided by berkeley university, freely available for research. The data contains 4.1 Million continuous ratings (-10.00 to +10.00) of 100 jokes from 73,421 users: collected between April 1999 - May 2003.

web link: https://goldberg.berkeley.edu/jester-data/

Total users: 73421 Total jokes: 100

0.3 Business problem:

User will be recommended jokes from all available jokes in system.

There is no network latency as the jokes can be recommended to users over batch job.

The error metric is decided as NAME by the provider of data.

0.4 Error metric:

As per the requirement of the problem barkley university has choosen NMAE(Normalized Mean Absolute Error)

NMAE:

$$1 \quad \frac{\frac{1}{n}\sum_{i}^{n}|(\widehat{y}-y)|}{\max(y)-\min(y)}$$

```
[1]: #import all required libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import surprise
import os
import itertools
import matplotlib as plt
```

```
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn.metrics.pairwise import cosine_similarity

from surprise import Reader, Dataset, BaselineOnly, KNNBaseline, SVDpp
from surprise.model_selection import cross_validate
from surprise.model_selection.search import GridSearchCV, RandomizedSearchCV

import random
from datetime import datetime
import pickle

from scipy import sparse
from scipy.sparse import csr_matrix

import xgboost as xgb
from xgboost import XGBRegressor
```

1.0.1 1.1 Data collection

```
[2]: #combine all 3 input files
  raw_data = pd.DataFrame()
  for i in os.listdir('./'):
     if(i.endswith('.xls')):
        tmp_data = pd.read_excel(i, header=None)
        print('length of records in %s is %s' %(len(tmp_data),i))
        raw_data = pd.concat([raw_data,tmp_data],axis=0)
  print('Total lenth of records is '+str(len(raw_data)))
```

length of records in 24983 is jester-data-1.xls length of records in 23500 is jester-data-2.xls length of records in 24938 is jester-data-3.xls Total lenth of records is 73421

```
[3]: raw_data_copy = raw_data.copy()
#raw_data = raw_data_copy.copy()
```

[4]: pickle.dump(raw_data_copy, open("dataframe.pickle","wb"))
raw_data.head()

```
[4]:
      0
                                       5
                                                        8
                                                              9
                                                                          91
       74 -7.82
                  8.79 -9.66 -8.16 -7.52 -8.50 -9.85 4.17
                                                            -8.98
                                                                   . . .
                                                                         2.82
   1
     100
          4.08 - 0.29
                         6.36
                                4.37 -2.38 -9.66 -0.73 -5.34
                                                             8.88
                                                                   . . .
                                                                         2.82
   2
       49 99.00 99.00 99.00 9.00 9.03 9.27 9.03 9.27 99.00
                                                                        99.00
                                                                   . . .
   3
       48 99.00
                  8.35 99.00 99.00 1.80 8.16 -2.82 6.21
                                                            99.00
                                                                        99.00
            8.50
                 4.61 -4.17 -5.39 1.36 1.60 7.04 4.61 -0.44
                                                                         5.19
       91
```

```
92
           93
                 94
                        95
                               96
                                     97
                                            98
                                                   99
                                                         100
0 99.00 99.00 99.00
                      99.00 99.00
                                   -5.63 99.00 99.00 99.00
1 - 4.95
        -0.29
                7.86
                      -0.19
                            -2.14
                                    3.06
                                           0.34
                                                 -4.32
                                                       1.07
2 99.00 99.00
                      99.00 99.00
                                                99.00 99.00
                9.08
                                   99.00 99.00
3 99.00
        99.00
                0.53
                      99.00 99.00
                                   99.00 99.00 99.00 99.00
   5.58
          4.27
                5.19
                       5.73
                              1.55
                                    3.11
                                           6.55
                                                  1.80
                                                        1.60
```

[5 rows x 101 columns]

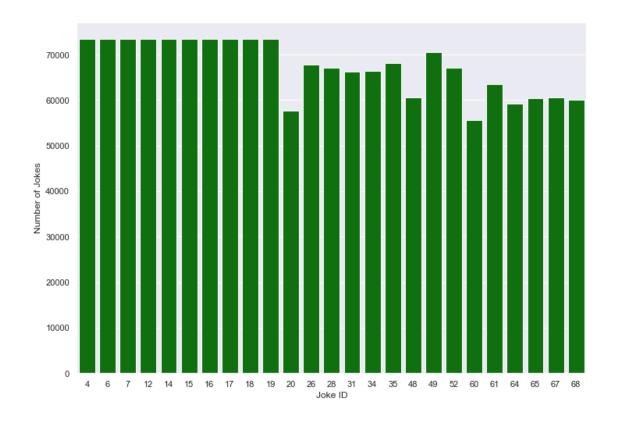
Total number of users who rated all jokes are: 14116

Total number of users who rated no jokes are: 0

```
[7]: jokes_number_of_rating = []
for i in range(1,101):
        jokes_number_of_rating.append(len(raw_data[raw_data[i]!=99]))
#print(len(jokes_number_of_rating))
```

```
[8]: #sort jokes on number of jokes
jokes_number_of_rating = np.array(jokes_number_of_rating)
jokes_number_of_rating_arg_sort = np.argsort(jokes_number_of_rating)
jokes_number_of_rating_arg_sort = jokes_number_of_rating_arg_sort[::-1]
top_10_joke_id = jokes_number_of_rating_arg_sort[:25]
top_10_joke_id_rating = jokes_number_of_rating[top_10_joke_id]
```

```
[9]: sns.set(rc={'figure.figsize':(11.7,8.27)})
ax = sns.barplot(top_10_joke_id,top_10_joke_id_rating, color = 'green')
plot_is = ax.set(xlabel='Joke ID', ylabel='Number of Jokes')
```



1.0.2 1.2 Preparing data in format of ['user', 'joke', 'rating'] and data cleaning

```
[10]: #column 0 contains the number of ratings given by the user which is not useful.
     number_of_rating = raw_data.drop(columns=[0])
[11]: raw_data.drop(columns=[0],inplace=True)
     raw_data.head()
[11]:
                  2
                                                      7
           1
                          3
                                  4
                                         5
                                               6
                                                             8
                                                                    9
                                                                            10
                                                                                       \
     0
        -7.82
                 8.79
                        -9.66
                                -8.16 -7.52 -8.50 -9.85
                                                           4.17
                                                                  -8.98
                                                                          -4.76
         4.08
                -0.29
                         6.36
                                 4.37 -2.38 -9.66 -0.73 -5.34
                                                                           9.22
     1
                                                                   8.88
     2
        99.00
                99.00
                        99.00
                                99.00
                                       9.03
                                              9.27
                                                     9.03
                                                           9.27
                                                                  99.00
                                                                          99.00
     3
        99.00
                 8.35
                                99.00
                                       1.80
                                              8.16 -2.82
                                                           6.21
                        99.00
                                                                  99.00
                                                                           1.84
                                                                                  . . .
                 4.61
                                                     7.04
         8.50
                        -4.17
                                -5.39
                                       1.36
                                              1.60
                                                           4.61
                                                                  -0.44
                                                                           5.73
                                                                                  . . .
          91
                  92
                          93
                                  94
                                          95
                                                 96
                                                         97
                                                                 98
                                                                         99
                                                                                 100
     0
         2.82
                99.00
                        99.00
                                99.00
                                       99.00
                                                       -5.63
                                               99.00
                                                               99.00
                                                                       99.00
                                                                              99.00
                -4.95
     1
         2.82
                        -0.29
                                 7.86
                                       -0.19
                                               -2.14
                                                        3.06
                                                                0.34
                                                                       -4.32
                                                                                1.07
     2
        99.00
                99.00
                        99.00
                                 9.08
                                       99.00
                                               99.00
                                                       99.00
                                                               99.00
                                                                       99.00
                                                                              99.00
     3
        99.00
                99.00
                        99.00
                                 0.53
                                               99.00
                                                       99.00
                                                                       99.00
                                       99.00
                                                               99.00
                                                                              99.00
     4
         5.19
                 5.58
                         4.27
                                 5.19
                                         5.73
                                                 1.55
                                                        3.11
                                                                6.55
                                                                        1.80
                                                                                1.60
```

[5 rows x 100 columns]

4136360

```
[13]: formated_data = pd.DataFrame(list_data, columns=['user','joke','rating'])

[14]: # we will increase the joke id and user id by 1 as it does not feel good as_

suser id 0

formated_data['user'] = formated_data['user'] + 1

formated_data['joke'] = formated_data['joke'] + 1
```

1.0.3 2. Trin test split

Train size: (2895452, 3) test size: (1240908, 3)

1.0.4 3. Feature engineering

3.1 Finding Global average of all movie ratings, Average rating per user, and Average rating per movie

```
[17]: #finding global average train
    print("global average of train is: ", train_df.rating.mean())

global average of train is: 0.739790630271198

[18]: #finding global average test
    print("global average of test is: ", test_df.rating.mean())
```

global average of test is: 0.747643548111544

```
[19]: #find average user rating per user train
train_df_grp_user = train_df.groupby(by='user')
train_average_user_rating={}
for i , j in train_df_grp_user:
    train_average_user_rating[i]=j.rating.mean()
print("Total number of train users: ", len(train_df_grp_user))
```

Total number of train users: 24983

```
[20]: #find average user rating per user test
test_df_grp_user = test_df.groupby(by='user')
test_average_user_rating={}
for i , j in test_df_grp_user:
    test_average_user_rating[i]=j.rating.mean()
print("Total number of train users: ", len(test_df_grp_user))
```

Total number of train users: 24983

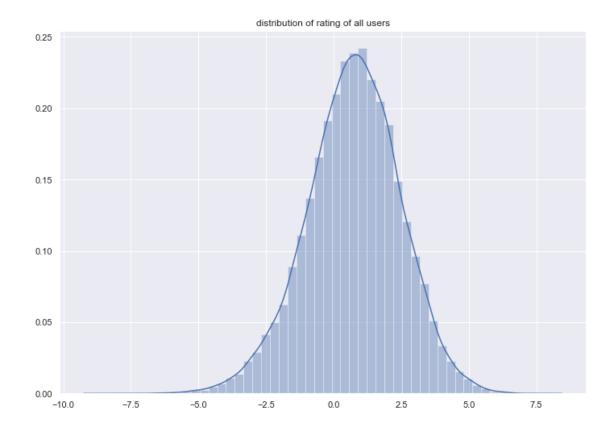
```
[21]: #find average joke rating per joke train
    train_df_grp_joke = train_df.groupby(by='joke')
    train_average_joke_rating={}
    for i , j in train_df_grp_joke:
        train_average_joke_rating[i]=j.rating.mean()
    print("Total number of jokes in train: ", len(train_df_grp_joke))
```

Total number of jokes in train: 100

```
[22]: #find average joke rating per joke test
test_df_grp_joke = test_df.groupby(by='joke')
test_average_joke_rating={}
for i , j in test_df_grp_joke:
    test_average_joke_rating[i]=j.rating.mean()
print("Total number of jokes in test: ", len(test_df_grp_joke))
```

Total number of jokes in test: 100

```
[23]: ax = sns.distplot([*train_average_user_rating.values()])
plot_is = ax.set(title='distribution of rating of all users')
```

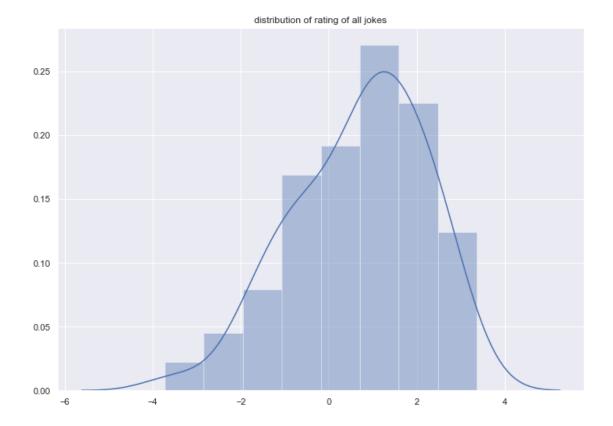


```
[24]: print("Mean value of average rating of users is: ", np.

-mean([*train_average_user_rating.values()]))
```

Mean value of average rating of users is: 0.7277005724312704

```
[25]: ax = sns.distplot([*train_average_joke_rating.values()])
plot_is = ax.set(title='distribution of rating of all jokes')
```



```
[26]: print("Mean value of average rating of jokes is: ", np.

-mean([*train_average_joke_rating.values()]))
```

Mean value of average rating of jokes is: 0.6999030393527355

1.0.5 4. create data frame strure to pass through ML models

```
[27]:
                   joke rating user_avg joke_avg
              user
                                                       gavg
    2064504
              3526
                         -0.24 -2.990463 1.322097 0.73979
    313132
              4371
                     49
                         -5.10 -0.362000 2.526823 0.73979
    555676
              7699
                     25
                          5.87 3.893030 0.434044 0.73979
    4049155
                     7
                          -2.91 1.447450 -0.657305
             21371
                                                    0.73979
    1646072 22719
                          -0.58 3.784375 1.961670 0.73979
                     83
```

```
[28]: pickle.dump(train_df_structured, open("train_df_structured.pickle","wb"))
     train_df_structured.head()
[28]:
                     joke rating user_avg joke_avg
               user
                                                          gavg
     2064504
               3526
                           -0.24 -2.990463 1.322097
                      47
                                                       0.73979
     313132
               4371
                      49
                           -5.10 -0.362000 2.526823
                                                       0.73979
               7699
                      25
     555676
                            5.87
                                   3.893030 0.434044
                                                       0.73979
     4049155
             21371
                       7
                            -2.91 1.447450 -0.657305
                                                       0.73979
     1646072 22719
                            -0.58 3.784375 1.961670
                      83
                                                       0.73979
[29]: # test dataframe
     test df structured = test df.copy()
     test_df_structured['user_avg'] = test_df_structured.user.apply(lambda x:_
     →test_average_user_rating[x])
     test_df_structured['joke_avg'] = test_df_structured.joke.apply(lambda x:u
     →test_average_joke_rating[x])
     test_df_structured['gavg'] = 0.74764
     test_df_structured.head()
[29]:
              user
                    joke rating user_avg joke_avg
                                                          gavg
                                                       0.74764
     1265657
             17469
                      61
                            6.80 -0.077273 2.113085
     1297014 17897
                           -0.78 -2.413220 1.018350 0.74764
                      45
     1567900 21640
                      47
                           -8.16 -1.552368 1.374515
                                                       0.74764
                            5.05 -0.282558 1.741960
     3232032 19559
                      48
                                                       0.74764
     810243
             11231
                      76
                            3.88 0.034872 2.365746 0.74764
[30]: pickle.dump(test_df_structured, open("test_df_structured.pickle","wb"))
     test_df_structured.head()
[30]:
              user
                    joke rating user_avg joke_avg
                                                          gavg
     1265657
                            6.80 -0.077273 2.113085
                                                       0.74764
             17469
                      61
     1297014 17897
                      45
                           -0.78 -2.413220 1.018350
                                                       0.74764
     1567900 21640
                      47
                           -8.16 -1.552368 1.374515
                                                       0.74764
     3232032 19559
                            5.05 -0.282558 1.741960
                       48
                                                       0.74764
     810243
             11231
                      76
                            3.88 0.034872 2.365746
                                                       0.74764
[31]: train df structured.columns
[31]: Index(['user', 'joke', 'rating', 'user_avg', 'joke_avg', 'gavg'],
     dtype='object')
[32]: #create target variable for train
     train_df_structured_target = train_df_structured.rating
     train_df_structured.drop(columns=['rating','user','joke'], inplace = True)
     pickle.dump(train_df_structured_target, open("train_df_structured_target.
     →pickle","wb"))
     #create target variable for test
     test_df_structured_target = test_df_structured.rating
     test_df_structured.drop(columns=['rating', 'user', 'joke'], inplace = True)
```

1.0.6 **5.0 Modeling**

```
[33]: global_model_name={}
```

5.0.1 define erroR metric NAME

```
def _error(actual: np.ndarray, predicted: np.ndarray):
    """ Simple error """
    return actual - predicted

def mae(actual: np.ndarray, predicted: np.ndarray):
    """ Mean Absolute Error """
    return np.mean(np.abs(_error(actual, predicted)))

def nmae(actual: np.ndarray, predicted: np.ndarray):
    """ Normalized Mean Absolute Error """
    return mae(actual, predicted) / (actual.max() - actual.min())

# get rating after building model
def get_ratings(predictions):
    actual = np.array([pred.r_ui for pred in predictions])
    pred = np.array([pred.est for pred in predictions])
    return actual, pred
```

5.0.2 prepare data in surprise way

some sample data for test

```
[36]: [(17469, 61, 6.8), (17897, 45, -0.78), (21640, 47, -8.16)]
[37]: #save the mmodel
     pickle.dump(train_data, open("train_data_sp.pickle","wb"))
     pickle.dump(trainset, open("trainset_sp.pickle","wb"))
     pickle.dump(testset, open("testset.pickle","wb"))
    1.0.7 5.1 Surprise Bseline model
    Let's create a baseline model and do some hyperparameter tuning there.
[34]: param_grid = {
                     'bsl_options' :{
                             'method':['sgd'],
                             'learning_rate': [.001,.01,.1],
                             'n_epochs': [5,7,9,10,20]
                                    },
                     'verbose' : [False]
     rs = GridSearchCV(BaselineOnly, param_grid, cv=3, joblib_verbose=4, n_jobs=-1)
     rs.fit(train_data)
     print(rs.best_params['mae'])
    [Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
    [Parallel(n_jobs=-1)]: Done 13 tasks
                                               | elapsed: 6.3min
    {'bsl_options': {'method': 'sgd', 'learning_rate': 0.01, 'n_epochs': 20},
    'verbose': False}
    [Parallel(n_jobs=-1)]: Done 45 out of 45 | elapsed: 21.6min finished
[38]: bsl_options = {'method': 'sgd',
                    'learning_rate': 0.01,
                    'n_epochs' : 20
     bsl_algo = BaselineOnly(bsl_options=bsl_options)
     bsl_algo.fit(trainset)
     train_preds = bsl_algo.test(trainset.build_testset())
     train_actual_ratings, train_pred_ratings = get_ratings(train_preds)
     test_preds = bsl_algo.test(testset)
     test_actual_ratings, test_pred_ratings = get_ratings(test_preds)
     train_df_structured['BaselineOnly'] = train_pred_ratings
     test_df_structured['BaselineOnly'] = test_pred_ratings
```

global_model_name['Baseline'] = {

```
Estimating biases using sgd...

Result of model is:
{'Train': 0.20336702289913602, 'Test': 0.20505297177649737}
```

This model have 20.5 % of NMAE. We have stored the value we will test the combined effect later.

```
[39]: #save the mmodel pickle.dump(bsl_algo, open("bsl_algo.pickle","wb"))
```

1.0.8 5.2 Surprise KNNBseline model

5.2.1 Surprise KNNBseline model joke joke similarity

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
```

```
{'rmse': {'bsl_options': {'method': 'sgd', 'learning_rate': 0.001},
    'sim_options': {'user_based': False, 'name': 'cosine', 'shrinkage': 100,
    'min_support': 2}, 'k': 70, 'verbose': False}, 'mae': {'bsl_options': {'method': 'sgd', 'learning_rate': 0.001}, 'sim_options': {'user_based': False, 'name': 'cosine', 'shrinkage': 100, 'min_support': 2}, 'k': 60, 'verbose': False}}

[Parallel(n_jobs=-1)]: Done 12 out of 12 | elapsed: 8.4min finished
```

```
[38]: print(rs.best_params['mae'])
```

```
{'bsl_options': {'method': 'sgd', 'learning_rate': 0.001}, 'sim_options':
    {'user_based': False, 'name': 'cosine', 'shrinkage': 100, 'min_support': 2},
    'k': 60, 'verbose': False}
       As the for error mae the k value 60 is declared as optimum we will use that
    value and train the model.
[40]: bsl_options = {'method': 'sgd',
                    'learning_rate': 0.001
     sim options = {'user based' : False,
                    'name': 'cosine',
                    'shrinkage': 100,
                    'min_support': 2
                   }
     knn_bsl_u = KNNBaseline(k=60, sim_options = sim_options, bsl_options = u
      →bsl_options)
     knn_bsl_u.fit(trainset)
     train_preds = knn_bsl_u.test(trainset.build_testset())
     train_actual_ratings, train_pred_ratings = get_ratings(train_preds)
     test_preds = knn_bsl_u.test(testset)
     test_actual_ratings, test_pred_ratings = get_ratings(test_preds)
     train_df_structured['KnnBaseline_joke'] = train_pred_ratings
     test_df_structured['KnnBaseline_joke'] = test_pred_ratings
     global_model_name['KnnBaseline_joke']={
             "Train": nmae(train_pred_ratings, train_actual_ratings),
             "Test": nmae(test pred ratings, test actual ratings)
     print('Result of model is: ')
     print(global_model_name['KnnBaseline_joke'])
    Estimating biases using sgd...
    Computing the cosine similarity matrix...
    Done computing similarity matrix.
```

```
Result of model is:
{'Train': 0.1825198921435078, 'Test': 0.19609971106367624}
```

This model have 19.6 % of NMAE. We have stored the value we will test the combined effect later.

```
[41]: #save the mmodel
     pickle.dump(knn_bsl_u, open("knn_bsl_u.pickle","wb"))
```

1.0.9 5.3 XGBoost model 1

we will use user average, global average, joke average, output of surprise knn joke joke similarity and output of surprise Baseline

```
[42]: train_df_structured.head()
             user_avg joke_avg
                                           BaselineOnly KnnBaseline_joke
[42]:
                                     gavg
     2064504 -2.990463 1.322097 0.73979
                                              -2.420203
                                                                -1.867690
     313132 -0.362000 2.526823 0.73979
                                              -4.659975
                                                                -6.338853
     555676
             3.893030 0.434044 0.73979
                                              -4.530977
                                                                -7.089862
     4049155 1.447450 -0.657305 0.73979
                                              -2.460071
                                                                -1.667773
     1646072 3.784375 1.961670 0.73979
                                              -1.465956
                                                                 0.101095
[42]: from sklearn.model selection import RandomizedSearchCV
     from xgboost import XGBRegressor
     params = {
             'min_child_weight': [1, 3, 5, 10],
             'gamma': [0.5, 1, 1.5, 2, 5],
             'subsample': [0.6, 0.8, 1.0],
             'colsample_bytree': [0.6, 0.8, 1.0],
             'max_depth': [3, 4, 5],
             'eta':[0.02,0.01,0.1],
             'n_estimators' : [100, 200, 400, 600, 800],
             'learning_rate' : [0.001,0.001,0.01,0.1]
             }
     xgb = XGBRegressor()
     random_search = RandomizedSearchCV(xgb, param_distributions=params, n_iter=10,_
      ⇒scoring='neg_mean_squared_error', n_jobs=-1, cv=3, verbose=10,
     →random state=42)
     random_search.fit(train_df_structured, train_df_structured_target)
     print(random_search.best_params_)
    Fitting 3 folds for each of 10 candidates, totalling 30 fits
    [Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
    [Parallel(n_jobs=-1)]: Done
                                  1 tasks
                                               | elapsed: 6.5min
    [Parallel(n_jobs=-1)]: Done
                                  6 tasks
                                               | elapsed: 16.8min
    [Parallel(n_jobs=-1)]: Done 13 tasks
                                               | elapsed: 60.6min
    [Parallel(n_jobs=-1)]: Done 23 out of
                                            30 | elapsed: 91.6min remaining: 27.9min
    [Parallel(n_jobs=-1)]: Done 27 out of
                                            30 | elapsed: 118.4min remaining:
    13.2min
    [Parallel(n_jobs=-1)]: Done 30 out of 30 | elapsed: 121.1min finished
    C:\Users\user\Anaconda3\lib\site-packages\xgboost\core.py:587: FutureWarning:
    Series.base is deprecated and will be removed in a future version
```

C:\Users\user\Anaconda3\lib\site-packages\xgboost\core.py:588: FutureWarning:

if getattr(data, 'base', None) is not None and \

```
Series.base is deprecated and will be removed in a future version
      data.base is not None and isinstance(data, np.ndarray) \
    [00:08:24] WARNING: src/objective/regression_obj.cu:152: reg:linear is now
    deprecated in favor of reg:squarederror.
    {'subsample': 1.0, 'n estimators': 200, 'min child weight': 5, 'max depth': 5,
    'learning_rate': 0.1, 'gamma': 5, 'eta': 0.1, 'colsample_bytree': 0.8}
[43]: # initialize Our first XGBoost model...
     xgb_bsl = XGBRegressor(min_child_weight = 5 ,gamma = 5, subsample = 1.
      \rightarrow0,colsample_bytree = 0.8,
                 max_depth = 5,eta = 0.1,n_estimators = 200,learning_rate = 0.1,
                 objective='reg:squarederror',silent=True, random_state=42)
     xgb bsl.fit(train df structured, train df structured target)
     test_pred_ratings = xgb_bsl.predict(test_df_structured)
     train_pred_ratings = xgb_bsl.predict(train_df_structured)
     global_model_name['First_XGB']={
             "Train": nmae(train_pred_ratings, train_df_structured_target),
             "Test": nmae(test_pred_ratings,test_df_structured_target)
     print('Result of model')
     print(global_model_name['First_XGB'])
    C:\Users\user\Anaconda3\envs\tensorflow2_gpu\lib\site-
    packages\xgboost\core.py:587: FutureWarning: Series.base is deprecated and will
    be removed in a future version
      if getattr(data, 'base', None) is not None and \
    C:\Users\user\Anaconda3\envs\tensorflow2_gpu\lib\site-
    packages\xgboost\core.py:588: FutureWarning: Series.base is deprecated and will
    be removed in a future version
      data.base is not None and isinstance(data, np.ndarray) \
    Result of model
    {'Train': 0.217233447214206, 'Test': 0.19280766965994262}
       This model works well and can have good effects on our model. This is the best
    model till now.
[44]: #save the model
     pickle.dump(xgb_bsl, open("xgb_bsl.pickle","wb"))
    1.0.10 5.3 Surprise SVD model
[45]: from surprise.model_selection.search import GridSearchCV, RandomizedSearchCV
     from surprise import SVD
```

initiallize the model

```
param_distributions = {
                     'n_factors' : [100,250,500,1000,2000,3000,5000],
                     'verbose' : [False],
                     'lr_bu' : [0.001,0.01,0.005],
                     'lr_bi' : [0.001,0.01,0.005],
                     'lr_pu' : [0.001,0.01,0.005],
                     'lr_qi' : [0.001,0.01,0.005],
                     'reg_bu' : [0.01,0.001],
                     'reg_bi' : [0.01,0.001],
                     'reg_pu' : [0.01,0.001],
                     'reg_qi' : [0.01,0.001]
     rs = RandomizedSearchCV(SVD, param_distributions, n_iter=20, cv=3,__
     →joblib_verbose=1, n_jobs=-1,random_state=42)
     rs.fit(train data)
     print(rs.best_params)
    [Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
    [Parallel(n_jobs=-1)]: Done 38 tasks
                                               | elapsed: 399.2min
    {'rmse': {'n factors': 2000, 'verbose': False, 'lr_bu': 0.001, 'lr_bi': 0.01,
    'lr_pu': 0.01, 'lr_qi': 0.01, 'reg_bu': 0.001, 'reg_bi': 0.001, 'reg_pu': 0.001,
    'reg_qi': 0.001}, 'mae': {'n_factors': 2000, 'verbose': False, 'lr_bu': 0.001,
    'lr_bi': 0.01, 'lr_pu': 0.01, 'lr_qi': 0.01, 'reg_bu': 0.001, 'reg_bi': 0.001,
    'reg_pu': 0.001, 'reg_qi': 0.001}}
    [Parallel(n_jobs=-1)]: Done 60 out of 60 | elapsed: 528.4min finished
[46]: print(rs.best_params['mae'])
    {'n_factors': 2000, 'verbose': False, 'lr_bu': 0.001, 'lr_bi': 0.01, 'lr_pu':
    0.01, 'lr_qi': 0.01, 'reg_bu': 0.001, 'reg_bi': 0.001, 'reg_pu': 0.001,
    'reg_qi': 0.001}
       We will use these optimal values in SVD model
[45]: from surprise import SVD
     # initiallize the model
     svd = SVD(n_factors=2000,biased=False, random_state=42, verbose=True,
              lr_bu = 0.001, lr_bi = 0.01, lr_pu = 0.01, lr_qi = 0.01,
               reg_bu = 0.001, reg_bi = 0.001, reg_pu = 0.001, reg_qi = 0.001
     svd.fit(trainset)
     train_preds = svd.test(trainset.build_testset())
```

```
train_actual_ratings, train_pred_ratings = get_ratings(train_preds)

test_preds = svd.test(testset)

test_actual_ratings, test_pred_ratings = get_ratings(test_preds)

train_df_structured['SVD'] = train_pred_ratings

test_df_structured['SVD'] = test_pred_ratings

global_model_name['SVD'] = {
    "Train" : nmae(train_pred_ratings, train_actual_ratings),
    "Test" : nmae(test_pred_ratings, test_actual_ratings)
}

print('Result of model is: ')
print(global_model_name['SVD'])
```

```
Processing epoch 0
Processing epoch 1
Processing epoch 2
Processing epoch 3
Processing epoch 4
Processing epoch 5
Processing epoch 6
Processing epoch 7
Processing epoch 8
Processing epoch 9
Processing epoch 10
Processing epoch 11
Processing epoch 12
Processing epoch 13
Processing epoch 14
Processing epoch 15
Processing epoch 16
Processing epoch 17
Processing epoch 18
Processing epoch 19
Result of model is:
{'Train': 0.17327865608074278, 'Test': 0.21539326049482735}
```

```
[46]: #save the model pickle.dump(svd, open("svd.pickle","wb"))
```

The model performs bad as the NMAE is 0.2153, which is less than our last model NMAE of 0.1928 We will check the effect of the SVD values later in our final model.

1.0.11 5.4 XGBoost model 2

we will use user average, global average, joke average, output of surprise knn joke joke similarity, output of surprise Baseline and output of SVD

```
[47]: train_df_structured.head()
[47]:
             user_avg joke_avg
                                     gavg
                                          BaselineOnly KnnBaseline_joke
                                                                                SVD
     2064504 -2.990463 1.322097 0.73979
                                              -2.420203
                                                                -1.867690 -2.192427
     313132 -0.362000 2.526823 0.73979
                                                                -6.338853 -5.915707
                                              -4.659975
     555676
             3.893030 0.434044 0.73979
                                              -4.530977
                                                                -7.089862 -7.983773
     4049155 1.447450 -0.657305 0.73979
                                              -2.460071
                                                                -1.667773 3.404231
     1646072 3.784375 1.961670 0.73979
                                              -1.465956
                                                                 0.101095 -4.184272
[51]: from sklearn.model selection import RandomizedSearchCV
     from xgboost import XGBRegressor
     params = {
             'min_child_weight': [1, 3, 5, 10],
             'gamma': [0.5, 1, 1.5, 2, 5],
             'subsample': [0.6, 0.8, 1.0],
             'colsample_bytree': [0.6, 0.8, 1.0],
             'max_depth': [3, 4, 5],
             'eta': [0.02,0.01,0.1],
             'n_estimators' : [100, 200, 400, 600, 800],
             'learning_rate' : [0.001,0.001,0.01,0.1]
             }
     xgb = XGBRegressor()
     random_search = RandomizedSearchCV(xgb, param_distributions=params, n_iter=10,_
      ⇒scoring='neg mean squared error', n jobs=-1, cv=3, verbose=10,,,
      →random state=42)
     random_search.fit(train_df_structured, train_df_structured_target)
     print(random_search.best_params_)
    Fitting 3 folds for each of 10 candidates, totalling 30 fits
    [Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
    [Parallel(n_jobs=-1)]: Done
                                  1 tasks
                                               | elapsed: 6.4min
    [Parallel(n_jobs=-1)]: Done
                                  6 tasks
                                               | elapsed: 16.5min
    [Parallel(n_jobs=-1)]: Done 13 tasks
                                               | elapsed: 64.3min
    [Parallel(n_jobs=-1)]: Done 23 out of 30 | elapsed: 93.0min remaining: 28.3min
    [Parallel(n_jobs=-1)]: Done 27 out of
                                            30 | elapsed: 121.2min remaining:
    13.5min
    [Parallel(n_jobs=-1)]: Done 30 out of 30 | elapsed: 121.8min finished
    [12:19:58] WARNING: src/objective/regression_obj.cu:152: reg:linear is now
    deprecated in favor of reg:squarederror.
    {'subsample': 1.0, 'n estimators': 200, 'min child weight': 5, 'max depth': 5,
    'learning_rate': 0.1, 'gamma': 5, 'eta': 0.1, 'colsample_bytree': 0.8}
```

Lets use the best parameters to train the model

```
[48]: # initialize Our first XGBoost model...
     xgb_svd = XGBRegressor(min_child_weight = 5 ,gamma = 5, subsample = 1.
      \rightarrow0,colsample_bytree = 0.8,
                 max_depth = 5,eta = 0.1,n_estimators = 200,learning_rate = 0.1,
                 objective='reg:squarederror',silent=True, random_state=42)
     xgb_svd.fit(train_df_structured, train_df_structured_target)
     test_pred_ratings = xgb_svd.predict(test_df_structured)
     train_pred_ratings = xgb_svd.predict(train_df_structured)
     global_model_name['Second_XGB']={
             "Train": nmae(train_pred_ratings,train_df_structured_target),
             "Test": nmae(test_pred_ratings,test_df_structured_target)
     print('Result of model is: ')
     print(global_model_name['Second_XGB'])
    Result of model is:
    {'Train': 0.2174202221795477, 'Test': 0.18833252907421943}
       This model works well and is the best model till now.
[49]: #save the model
     pickle.dump(xgb_svd, open("xgb_svd.pickle","wb"))
       Lets check the feature importance. This will help us determining the important
    features. We will decide whether to keep the feature engineered features or not.
[50]: print(xgb_bsl.feature_importances_)
     print(train_df_structured.columns)
    [0.5250122 0.4710521 0.
                                      0.00193619 0.00199951]
    Index(['user_avg', 'joke_avg', 'gavg', 'BaselineOnly', 'KnnBaseline_joke',
           'SVD'],
          dtype='object')
       user_avg and joke_avg features seems most important features
    1.0.12 Lets try some feature engineering
    feature engineering 1 We will try a special feature as (user_avg + joke_avg - gavg)
[51]: train_df_structured['special_feature'] = ___
      -train_df_structured['user_avg']+train_df_structured['joke_avg']-train_df_structured['gavg']
     test_df_structured['special_feature'] = __
     →test_df_structured['user_avg']+test_df_structured['joke_avg']-test_df_structured['gavg']
     train_df_structured.head()
[51]:
                                     gavg BaselineOnly KnnBaseline_joke
              user_avg joke_avg
     2064504 -2.990463 1.322097 0.73979
                                                                 -1.867690
                                              -2.420203
```

```
313132 -0.362000 2.526823 0.73979
                                              -4.659975
                                                                -6.338853
     555676
             3.893030 0.434044 0.73979
                                                                -7.089862
                                              -4.530977
     4049155 1.447450 -0.657305 0.73979
                                              -2.460071
                                                                -1.667773
     1646072 3.784375 1.961670 0.73979
                                              -1.465956
                                                                 0.101095
                   SVD special_feature
     2064504 -2.192427
                              -2.408156
     313132 -5.915707
                               1.425033
     555676 -7.983773
                               3.587284
     4049155 3.404231
                               0.050355
     1646072 -4.184272
                               5.006255
[56]: from sklearn.model_selection import RandomizedSearchCV
     from xgboost import XGBRegressor
     params = {
             'min_child_weight': [1, 3, 5, 10],
             'gamma': [0.5, 1, 1.5, 2, 5],
             'subsample': [0.6, 0.8, 1.0],
             'colsample_bytree': [0.6, 0.8, 1.0],
             'max_depth': [3, 4, 5],
             'eta': [0.02,0.01,0.1],
             'n_estimators' : [100, 200, 400, 600, 800],
             'learning_rate' : [0.001,0.001,0.01,0.1]
             }
     xgb = XGBRegressor()
     random search = RandomizedSearchCV(xgb, param distributions=params, n iter=10,11
     ⇒scoring='neg_mean_squared_error', n_jobs=-1, cv=3, verbose=10,
     →random state=42)
     random_search.fit(train_df_structured, train_df_structured_target)
     print(random_search.best_params_)
    Fitting 3 folds for each of 10 candidates, totalling 30 fits
    [Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
    [Parallel(n_jobs=-1)]: Done
                                  1 tasks
                                               | elapsed: 7.8min
```

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.

[Parallel(n_jobs=-1)]: Done 1 tasks | elapsed: 7.8min

[Parallel(n_jobs=-1)]: Done 6 tasks | elapsed: 19.9min

[Parallel(n_jobs=-1)]: Done 13 tasks | elapsed: 75.7min

[Parallel(n_jobs=-1)]: Done 23 out of 30 | elapsed: 110.8min remaining: 33.7min

[Parallel(n_jobs=-1)]: Done 27 out of 30 | elapsed: 143.3min remaining: 15.9min

[Parallel(n_jobs=-1)]: Done 30 out of 30 | elapsed: 145.7min finished

[16:13:03] WARNING: src/objective/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.
```

```
{'subsample': 1.0, 'n_estimators': 200, 'min_child_weight': 5, 'max_depth': 5,
    'learning_rate': 0.1, 'gamma': 5, 'eta': 0.1, 'colsample_bytree': 0.8}
[52]: # initialize Our first XGBoost model...
     xgb_bsl_fe_1 = XGBRegressor(min_child weight = 5 ,gamma = 5, subsample = 1.
      \rightarrow0,colsample_bytree = 0.8,
                 max_depth = 5,eta = 0.1,n_estimators = 200,learning_rate = 0.1,
                 objective='reg:squarederror',silent=True, random_state=42)
     xgb_bsl_fe_1.fit(train_df_structured, train_df_structured_target)
     test_pred_ratings = xgb_bsl_fe_1.predict(test_df_structured)
     train_pred_ratings = xgb_bsl_fe_1.predict(train_df_structured)
     global model name['First XGB FE 1']={
             "Train": nmae(train_pred_ratings, train_df_structured_target),
             "Test": nmae(test_pred_ratings,test_df_structured_target)
             }
     print('Result of model')
     print(global_model_name['First_XGB_FE_1'])
    Result of model
    {'Train': 0.22321715705636846, 'Test': 0.20289328398328535}
       The feature engineering did not improve any thing on the model as the test
    NMAE was previously 0.1883 and is 0.2028 now.
[53]: print(xgb_bsl_fe_1.feature_importances_)
     print(train_df_structured.columns)
    [0.1351127 0.13147087 0.
                                      0.00140833 0.00140642 0.00138204
     0.72921973]
    Index(['user_avg', 'joke_avg', 'gavg', 'BaselineOnly', 'KnnBaseline_joke',
           'SVD', 'special_feature'],
          dtype='object')
    feature
                            2 We will try a special feature as (user_avg + joke_avg -
              engineering
    gavg)^2
[54]: train_df_structured['special_feature_1'] = [

→train_df_structured['special_feature']**2
     test_df_structured['special_feature_1'] = __
     →test_df_structured['special_feature']**2
     train_df_structured.head()
[54]:
             user_avg joke_avg
                                     gavg BaselineOnly KnnBaseline_joke \
     2064504 -2.990463 1.322097 0.73979
                                              -2.420203
                                                                -1.867690
     313132 -0.362000 2.526823 0.73979
                                              -4.659975
                                                                -6.338853
     555676
             3.893030 0.434044 0.73979
                                              -4.530977
                                                                -7.089862
     4049155 1.447450 -0.657305 0.73979
                                              -2.460071
                                                                -1.667773
     1646072 3.784375 1.961670 0.73979
                                              -1.465956
                                                                 0.101095
```

```
SVD special_feature special_feature_1
      2064504 -2.192427
                               -2.408156
                                                   5.799214
      313132 -5.915707
                                1.425033
                                                   2.030718
      555676 -7.983773
                                3.587284
                                                  12.868608
      4049155 3.404231
                                0.050355
                                                   0.002536
      1646072 -4.184272
                                5.006255
                                                  25.062584
[163]: from sklearn.model selection import RandomizedSearchCV
      from xgboost import XGBRegressor
      params = {
              'min_child_weight': [1, 3, 5, 10],
              'gamma': [0.5, 1, 1.5, 2, 5],
              'subsample': [0.6, 0.8, 1.0],
              'colsample_bytree': [0.6, 0.8, 1.0],
              'max_depth': [3, 4, 5],
              'eta': [0.02,0.01,0.1],
              'n_estimators' : [100, 200, 400, 600, 800],
              'learning_rate' : [0.001,0.001,0.01,0.1]
              }
      xgb = XGBRegressor()
      random_search = RandomizedSearchCV(xgb, param_distributions=params, n_iter=10,_
      ⇒scoring='neg_mean_squared_error', n_jobs=-1, cv=3, verbose=10, __
      →random_state=42)
      random_search.fit(train_df_structured, train_df_structured_target)
      print(random_search.best_params_)
```

Fitting 3 folds for each of 10 candidates, totalling 30 fits

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 6 concurrent workers.
[Parallel(n_jobs=-1)]: Done
                                          | elapsed: 9.0min
                             1 tasks
[Parallel(n_jobs=-1)]: Done
                             6 tasks
                                           | elapsed: 21.5min
[Parallel(n_jobs=-1)]: Done 13 tasks
                                          | elapsed: 88.4min
[Parallel(n_jobs=-1)]: Done 23 out of 30 | elapsed: 131.0min remaining:
39.9min
[Parallel(n_jobs=-1)]: Done 27 out of 30 | elapsed: 164.9min remaining:
18.3min
[Parallel(n_jobs=-1)]: Done 30 out of 30 | elapsed: 167.1min finished
[23:06:32] WARNING: src/objective/regression_obj.cu:152: reg:linear is now
deprecated in favor of reg:squarederror.
{'subsample': 1.0, 'n_estimators': 200, 'min_child_weight': 5, 'max_depth': 5,
'learning_rate': 0.1, 'gamma': 5, 'eta': 0.1, 'colsample_bytree': 0.8}
```

```
Result of model {'Train': 0.22556924102141873, 'Test': 0.20872101921358918}
```

This feature engineering also did not improve any thing on the model as the test NMAE was previously 0.1883 and is 0.2087 now.

1.0.13 5.5 Surprise SVD plus plus model

Hyper parameter tuning takes more than 2 days. so i have stopped that code and removed that from notebok.

1.1 Feature engineering is not helping much. So lets try some Deep Neural Network Model.

1.1.1 6. Deep learning model 1

```
es = tf.keras.callbacks.EarlyStopping(monitor='val_loss', mode='min', u
   →verbose=1, patience=40)
   callbacks=[checkpoint,tbCallBack,es]
[95]: if os.path.isfile('weights-improvement.hdf5'):
     pass
   else:
     #define optimizer and train model
     optimizer = tf.keras.optimizers.Adam()
     first_model.compile(loss='mean_absolute_error',optimizer=optimizer)
     first model.
   →fit(train_df_structured,train_df_structured_target,epochs=1000,batch_size=512,validation_sp
   →3, callbacks=callbacks)
  WARNING:tensorflow:Falling back from v2 loop because of error: Failed to find
  data adapter that can handle input: <class 'pandas.core.frame.DataFrame'>,
  <class 'NoneType'>
  Train on 2026816 samples, validate on 868636 samples
  Epoch 1/1000
  Epoch 00001: val_loss improved from inf to 3.85437, saving model to weights-
  improvement.hdf5
  - val_loss: 3.8544
  Epoch 2/1000
  Epoch 00002: val_loss improved from 3.85437 to 3.84818, saving model to weights-
  improvement.hdf5
  - val_loss: 3.8482
  Epoch 3/1000
  Epoch 00003: val_loss improved from 3.84818 to 3.84769, saving model to weights-
  improvement.hdf5
  - val_loss: 3.8477
  Epoch 4/1000
  Epoch 00004: val_loss improved from 3.84769 to 3.84744, saving model to weights-
  improvement.hdf5
  - val_loss: 3.8474
  Epoch 5/1000
  Epoch 00005: val_loss did not improve from 3.84744
  - val_loss: 3.8484
  Epoch 6/1000
```

```
Epoch 00006: val_loss improved from 3.84744 to 3.84681, saving model to weights-
improvement.hdf5
- val loss: 3.8468
Epoch 7/1000
Epoch 00007: val_loss did not improve from 3.84681
- val_loss: 3.8488
Epoch 8/1000
Epoch 00008: val_loss improved from 3.84681 to 3.84592, saving model to weights-
improvement.hdf5
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8490
- val_loss: 3.8459
Epoch 9/1000
Epoch 00009: val_loss did not improve from 3.84592
- val loss: 3.8470
Epoch 10/1000
Epoch 00010: val_loss did not improve from 3.84592
- val_loss: 3.8464
Epoch 11/1000
Epoch 00011: val_loss did not improve from 3.84592
- val_loss: 3.8466
Epoch 12/1000
Epoch 00012: val_loss did not improve from 3.84592
- val loss: 3.8485
Epoch 13/1000
Epoch 00013: val_loss did not improve from 3.84592
- val_loss: 3.8464
Epoch 14/1000
Epoch 00014: val_loss improved from 3.84592 to 3.84590, saving model to weights-
improvement.hdf5
- val_loss: 3.8459
Epoch 15/1000
```

```
Epoch 00015: val_loss did not improve from 3.84590
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8480
- val loss: 3.8461
Epoch 16/1000
Epoch 00016: val loss improved from 3.84590 to 3.84574, saving model to weights-
improvement.hdf5
- val_loss: 3.8457
Epoch 17/1000
Epoch 00017: val_loss did not improve from 3.84574
- val_loss: 3.8482
Epoch 18/1000
Epoch 00018: val_loss improved from 3.84574 to 3.84562, saving model to weights-
improvement.hdf5
- val loss: 3.8456
Epoch 19/1000
Epoch 00019: val_loss did not improve from 3.84562
- val_loss: 3.8468
Epoch 20/1000
Epoch 00020: val_loss improved from 3.84562 to 3.84503, saving model to weights-
improvement.hdf5
- val_loss: 3.8450
Epoch 21/1000
Epoch 00021: val loss did not improve from 3.84503
- val loss: 3.8457
Epoch 22/1000
Epoch 00022: val_loss did not improve from 3.84503
- val_loss: 3.8454
Epoch 23/1000
Epoch 00023: val_loss improved from 3.84503 to 3.84436, saving model to weights-
improvement.hdf5
- val_loss: 3.8444
```

```
Epoch 24/1000
Epoch 00024: val_loss did not improve from 3.84436
- val loss: 3.8450
Epoch 25/1000
Epoch 00025: val_loss did not improve from 3.84436
- val_loss: 3.8455
Epoch 26/1000
Epoch 00026: val_loss did not improve from 3.84436
- val_loss: 3.8456
Epoch 27/1000
Epoch 00027: val_loss improved from 3.84436 to 3.84381, saving model to weights-
improvement.hdf5
- val loss: 3.8438
Epoch 28/1000
Epoch 00028: val_loss did not improve from 3.84381
- val_loss: 3.8443
Epoch 29/1000
Epoch 00029: val_loss did not improve from 3.84381
- val_loss: 3.8443
Epoch 30/1000
Epoch 00030: val_loss improved from 3.84381 to 3.84355, saving model to weights-
improvement.hdf5
- val loss: 3.8435
Epoch 31/1000
Epoch 00031: val_loss did not improve from 3.84355
- val_loss: 3.8437
Epoch 32/1000
Epoch 00032: val_loss did not improve from 3.84355
- val_loss: 3.8449
Epoch 33/1000
```

```
Epoch 00033: val_loss did not improve from 3.84355
- val loss: 3.8451
Epoch 34/1000
Epoch 00034: val_loss did not improve from 3.84355
- val loss: 3.8441
Epoch 35/1000
Epoch 00035: val_loss did not improve from 3.84355
- val loss: 3.8440
Epoch 36/1000
Epoch 00036: val_loss did not improve from 3.84355
- val loss: 3.8444
Epoch 37/1000
Epoch 00037: val loss did not improve from 3.84355
- val_loss: 3.8438
Epoch 38/1000
Epoch 00038: val_loss improved from 3.84355 to 3.84322, saving model to weights-
improvement.hdf5
- val_loss: 3.8432
Epoch 39/1000
Epoch 00039: val_loss did not improve from 3.84322
- val loss: 3.8436
Epoch 40/1000
Epoch 00040: val_loss did not improve from 3.84322
- val_loss: 3.8443
Epoch 41/1000
Epoch 00041: val_loss did not improve from 3.84322
- val_loss: 3.8438
Epoch 42/1000
Epoch 00042: val_loss did not improve from 3.84322
```

```
- val_loss: 3.8434
Epoch 43/1000
Epoch 00043: val_loss improved from 3.84322 to 3.84219, saving model to weights-
improvement.hdf5
- val_loss: 3.8422
Epoch 44/1000
Epoch 00044: val_loss did not improve from 3.84219
- val_loss: 3.8452
Epoch 45/1000
Epoch 00045: val_loss did not improve from 3.84219
- val_loss: 3.8439
Epoch 46/1000
Epoch 00046: val_loss did not improve from 3.84219
- val_loss: 3.8432
Epoch 47/1000
Epoch 00047: val_loss did not improve from 3.84219
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8444
- val_loss: 3.8439
Epoch 48/1000
Epoch 00048: val_loss did not improve from 3.84219
- val_loss: 3.8438
Epoch 49/1000
Epoch 00049: val_loss did not improve from 3.84219
- val_loss: 3.8428
Epoch 50/1000
Epoch 00050: val_loss did not improve from 3.84219
- val_loss: 3.8431
Epoch 51/1000
Epoch 00051: val_loss did not improve from 3.84219
- val_loss: 3.8431
```

```
Epoch 52/1000
Epoch 00052: val_loss did not improve from 3.84219
- val loss: 3.8423
Epoch 53/1000
Epoch 00053: val_loss did not improve from 3.84219
- val_loss: 3.8443
Epoch 54/1000
Epoch 00054: val_loss improved from 3.84219 to 3.84204, saving model to weights-
improvement.hdf5
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8441
- val_loss: 3.8420
Epoch 55/1000
Epoch 00055: val_loss did not improve from 3.84204
- val loss: 3.8427
Epoch 56/1000
Epoch 00056: val_loss did not improve from 3.84204
- val_loss: 3.8424
Epoch 57/1000
Epoch 00057: val_loss did not improve from 3.84204
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8439
- val_loss: 3.8426
Epoch 58/1000
Epoch 00058: val_loss did not improve from 3.84204
- val loss: 3.8422
Epoch 59/1000
Epoch 00059: val_loss did not improve from 3.84204
- val_loss: 3.8424
Epoch 60/1000
Epoch 00060: val_loss did not improve from 3.84204
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8438
- val_loss: 3.8429
Epoch 61/1000
```

```
Epoch 00061: val_loss did not improve from 3.84204
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8438
- val_loss: 3.8428
Epoch 62/1000
Epoch 00062: val_loss did not improve from 3.84204
- val_loss: 3.8444
Epoch 63/1000
Epoch 00063: val_loss did not improve from 3.84204
- val_loss: 3.8426
Epoch 64/1000
Epoch 00064: val_loss did not improve from 3.84204
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8435
- val_loss: 3.8436
Epoch 65/1000
Epoch 00065: val_loss improved from 3.84204 to 3.84193, saving model to weights-
improvement.hdf5
- val_loss: 3.8419
Epoch 66/1000
Epoch 00066: val_loss did not improve from 3.84193
2026816/2026816 [============== ] - 10s 5us/sample - loss: 3.8436
- val_loss: 3.8426
Epoch 67/1000
Epoch 00067: val_loss did not improve from 3.84193
- val_loss: 3.8423
Epoch 68/1000
Epoch 00068: val loss did not improve from 3.84193
- val_loss: 3.8430
Epoch 69/1000
Epoch 00069: val_loss did not improve from 3.84193
- val_loss: 3.8429
Epoch 70/1000
Epoch 00070: val_loss did not improve from 3.84193
```

```
- val_loss: 3.8430
Epoch 71/1000
Epoch 00071: val_loss improved from 3.84193 to 3.84175, saving model to weights-
improvement.hdf5
- val loss: 3.8418
Epoch 72/1000
Epoch 00072: val_loss did not improve from 3.84175
- val_loss: 3.8432
Epoch 73/1000
Epoch 00073: val_loss did not improve from 3.84175
- val_loss: 3.8424
Epoch 74/1000
Epoch 00074: val loss did not improve from 3.84175
- val loss: 3.8423
Epoch 75/1000
Epoch 00075: val_loss did not improve from 3.84175
- val_loss: 3.8447
Epoch 76/1000
Epoch 00076: val_loss did not improve from 3.84175
- val_loss: 3.8427
Epoch 77/1000
Epoch 00077: val loss did not improve from 3.84175
- val loss: 3.8443
Epoch 78/1000
Epoch 00078: val_loss did not improve from 3.84175
- val_loss: 3.8435
Epoch 79/1000
Epoch 00079: val_loss did not improve from 3.84175
- val_loss: 3.8432
Epoch 80/1000
```

```
Epoch 00080: val_loss did not improve from 3.84175
- val loss: 3.8421
Epoch 81/1000
Epoch 00081: val_loss did not improve from 3.84175
- val loss: 3.8422
Epoch 82/1000
Epoch 00082: val_loss did not improve from 3.84175
- val loss: 3.8425
Epoch 83/1000
Epoch 00083: val_loss did not improve from 3.84175
- val_loss: 3.8428
Epoch 84/1000
Epoch 00084: val loss did not improve from 3.84175
- val_loss: 3.8424
Epoch 85/1000
Epoch 00085: val_loss did not improve from 3.84175
- val_loss: 3.8439
Epoch 86/1000
Epoch 00086: val_loss did not improve from 3.84175
- val_loss: 3.8427
Epoch 87/1000
Epoch 00087: val loss did not improve from 3.84175
- val_loss: 3.8430
Epoch 88/1000
Epoch 00088: val_loss did not improve from 3.84175
- val_loss: 3.8421
Epoch 89/1000
Epoch 00089: val_loss did not improve from 3.84175
```

```
- val_loss: 3.8419
Epoch 90/1000
Epoch 00090: val_loss improved from 3.84175 to 3.84146, saving model to weights-
improvement.hdf5
- val loss: 3.8415
Epoch 91/1000
Epoch 00091: val_loss did not improve from 3.84146
- val_loss: 3.8455
Epoch 92/1000
Epoch 00092: val_loss did not improve from 3.84146
- val_loss: 3.8418
Epoch 93/1000
Epoch 00093: val loss did not improve from 3.84146
- val loss: 3.8438
Epoch 94/1000
Epoch 00094: val_loss did not improve from 3.84146
- val_loss: 3.8416
Epoch 95/1000
Epoch 00095: val_loss did not improve from 3.84146
- val_loss: 3.8433
Epoch 96/1000
Epoch 00096: val loss did not improve from 3.84146
- val loss: 3.8417
Epoch 97/1000
Epoch 00097: val_loss did not improve from 3.84146
- val_loss: 3.8452
Epoch 98/1000
Epoch 00098: val_loss did not improve from 3.84146
- val_loss: 3.8433
Epoch 99/1000
```

```
Epoch 00099: val_loss did not improve from 3.84146
- val loss: 3.8418
Epoch 100/1000
Epoch 00100: val_loss did not improve from 3.84146
- val loss: 3.8421
Epoch 101/1000
Epoch 00101: val_loss did not improve from 3.84146
- val_loss: 3.8421
Epoch 102/1000
Epoch 00102: val_loss did not improve from 3.84146
- val loss: 3.8432
Epoch 103/1000
Epoch 00103: val loss did not improve from 3.84146
- val_loss: 3.8424
Epoch 104/1000
Epoch 00104: val_loss did not improve from 3.84146
- val_loss: 3.8422
Epoch 105/1000
Epoch 00105: val_loss did not improve from 3.84146
- val_loss: 3.8428
Epoch 106/1000
Epoch 00106: val loss did not improve from 3.84146
- val_loss: 3.8441
Epoch 107/1000
Epoch 00107: val_loss did not improve from 3.84146
- val_loss: 3.8419
Epoch 108/1000
Epoch 00108: val_loss did not improve from 3.84146
```

```
- val_loss: 3.8419
Epoch 109/1000
Epoch 00109: val_loss did not improve from 3.84146
- val_loss: 3.8422
Epoch 110/1000
Epoch 00110: val_loss did not improve from 3.84146
- val_loss: 3.8463
Epoch 111/1000
Epoch 00111: val_loss did not improve from 3.84146
- val_loss: 3.8427
Epoch 112/1000
Epoch 00112: val_loss did not improve from 3.84146
- val loss: 3.8424
Epoch 113/1000
Epoch 00113: val_loss did not improve from 3.84146
- val_loss: 3.8426
Epoch 114/1000
Epoch 00114: val_loss did not improve from 3.84146
2026816/2026816 [============== ] - 10s 5us/sample - loss: 3.8429
- val_loss: 3.8419
Epoch 115/1000
Epoch 00115: val_loss improved from 3.84146 to 3.84141, saving model to weights-
improvement.hdf5
- val loss: 3.8414
Epoch 116/1000
Epoch 00116: val_loss did not improve from 3.84141
- val_loss: 3.8416
Epoch 117/1000
Epoch 00117: val_loss did not improve from 3.84141
- val_loss: 3.8423
Epoch 118/1000
```

```
Epoch 00118: val_loss did not improve from 3.84141
- val loss: 3.8420
Epoch 119/1000
Epoch 00119: val_loss did not improve from 3.84141
- val loss: 3.8422
Epoch 120/1000
Epoch 00120: val_loss did not improve from 3.84141
- val_loss: 3.8421
Epoch 121/1000
Epoch 00121: val_loss did not improve from 3.84141
- val loss: 3.8441
Epoch 122/1000
Epoch 00122: val loss did not improve from 3.84141
- val_loss: 3.8434
Epoch 123/1000
Epoch 00123: val_loss did not improve from 3.84141
- val_loss: 3.8426
Epoch 124/1000
Epoch 00124: val_loss did not improve from 3.84141
- val_loss: 3.8423
Epoch 125/1000
Epoch 00125: val loss did not improve from 3.84141
- val_loss: 3.8432
Epoch 126/1000
Epoch 00126: val_loss did not improve from 3.84141
- val_loss: 3.8419
Epoch 127/1000
Epoch 00127: val_loss did not improve from 3.84141
```

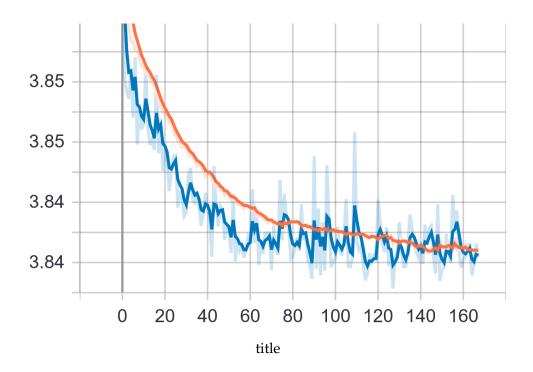
```
- val_loss: 3.8424
Epoch 128/1000
Epoch 00128: val_loss improved from 3.84141 to 3.84113, saving model to weights-
improvement.hdf5
- val loss: 3.8411
Epoch 129/1000
Epoch 00129: val_loss did not improve from 3.84113
- val_loss: 3.8422
Epoch 130/1000
Epoch 00130: val_loss did not improve from 3.84113
- val_loss: 3.8424
Epoch 131/1000
Epoch 00131: val loss did not improve from 3.84113
- val loss: 3.8433
Epoch 132/1000
Epoch 00132: val_loss did not improve from 3.84113
- val_loss: 3.8423
Epoch 133/1000
Epoch 00133: val_loss did not improve from 3.84113
- val_loss: 3.8419
Epoch 134/1000
Epoch 00134: val loss did not improve from 3.84113
- val loss: 3.8419
Epoch 135/1000
Epoch 00135: val_loss did not improve from 3.84113
- val_loss: 3.8426
Epoch 136/1000
Epoch 00136: val_loss did not improve from 3.84113
- val_loss: 3.8436
Epoch 137/1000
```

```
Epoch 00137: val_loss did not improve from 3.84113
- val_loss: 3.8431
Epoch 138/1000
Epoch 00138: val_loss did not improve from 3.84113
- val loss: 3.8427
Epoch 139/1000
Epoch 00139: val_loss did not improve from 3.84113
- val_loss: 3.8423
Epoch 140/1000
Epoch 00140: val_loss did not improve from 3.84113
- val loss: 3.8422
Epoch 141/1000
Epoch 00141: val loss did not improve from 3.84113
- val_loss: 3.8427
Epoch 142/1000
Epoch 00142: val_loss did not improve from 3.84113
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8425
- val_loss: 3.8420
Epoch 143/1000
Epoch 00143: val_loss did not improve from 3.84113
- val_loss: 3.8433
Epoch 144/1000
Epoch 00144: val loss did not improve from 3.84113
- val_loss: 3.8426
Epoch 145/1000
Epoch 00145: val_loss did not improve from 3.84113
- val_loss: 3.8431
Epoch 146/1000
Epoch 00146: val_loss did not improve from 3.84113
```

```
- val_loss: 3.8432
Epoch 147/1000
Epoch 00147: val_loss did not improve from 3.84113
- val_loss: 3.8422
Epoch 148/1000
Epoch 00148: val_loss did not improve from 3.84113
- val_loss: 3.8416
Epoch 149/1000
Epoch 00149: val_loss did not improve from 3.84113
- val_loss: 3.8439
Epoch 150/1000
Epoch 00150: val_loss did not improve from 3.84113
- val loss: 3.8414
Epoch 151/1000
Epoch 00151: val_loss did not improve from 3.84113
- val_loss: 3.8418
Epoch 152/1000
Epoch 00152: val_loss did not improve from 3.84113
- val_loss: 3.8418
Epoch 153/1000
Epoch 00153: val_loss did not improve from 3.84113
- val loss: 3.8429
Epoch 154/1000
Epoch 00154: val_loss did not improve from 3.84113
- val_loss: 3.8423
Epoch 155/1000
Epoch 00155: val_loss did not improve from 3.84113
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8425
- val_loss: 3.8420
Epoch 156/1000
```

```
Epoch 00156: val_loss did not improve from 3.84113
- val_loss: 3.8442
Epoch 157/1000
Epoch 00157: val_loss did not improve from 3.84113
- val_loss: 3.8432
Epoch 158/1000
Epoch 00158: val_loss did not improve from 3.84113
- val_loss: 3.8437
Epoch 159/1000
Epoch 00159: val_loss did not improve from 3.84113
- val_loss: 3.8423
Epoch 160/1000
Epoch 00160: val_loss did not improve from 3.84113
- val_loss: 3.8418
Epoch 161/1000
Epoch 00161: val_loss did not improve from 3.84113
- val_loss: 3.8421
Epoch 162/1000
Epoch 00162: val_loss did not improve from 3.84113
- val_loss: 3.8422
Epoch 163/1000
Epoch 00163: val_loss did not improve from 3.84113
- val_loss: 3.8425
Epoch 164/1000
Epoch 00164: val_loss did not improve from 3.84113
- val_loss: 3.8427
Epoch 165/1000
Epoch 00165: val_loss did not improve from 3.84113
- val_loss: 3.8416
```

```
Epoch 166/1000
Epoch 00166: val_loss did not improve from 3.84113
- val loss: 3.8419
Epoch 167/1000
Epoch 00167: val_loss did not improve from 3.84113
- val_loss: 3.8426
Epoch 168/1000
Epoch 00168: val_loss did not improve from 3.84113
2026816/2026816 [============= ] - 10s 5us/sample - loss: 3.8424
- val_loss: 3.8422
Epoch 00168: early stopping
```



```
first_best_model = tf.keras.models.load_model('weights-improvement.hdf5')

test_pred_ratings = first_best_model.predict(np.asarray(test_df_structured))

train_pred_ratings = first_best_model.predict(np.

asanyarray(train_df_structured))

[97]: # i am getting ther error as: MemoryError: Unable to allocate array with shape

in (1240908, 1240908) and data type float64

# while calculating NMAE

# this is happenning as numpy can't put so many float64 values to memory.
```

[96]: #load model and predict

```
So lets try the python forloop way for calculating NMAE

[98]: test_pred_ratings = test_pred_ratings.tolist()
    train_pred_ratings = train_pred_ratings.tolist()

[99]: test_pred_ratings_list = [i[0] for i in test_pred_ratings]
    train_pred_ratings_list = [i[0] for i in train_pred_ratings]

[100]: test_df_structured_target_list=test_df_structured_target_values
    train_df_structured_target_list=train_df_structured_target_list.tolist()
    train_df_structured_target_list = test_df_structured_target_list.tolist()

1.1.2 redefine error to accept python list

[102]: # re define error metric same as for ML models
    def _error(actual, predicted):
        """ Simple error """
        temp = []
        for i in range(len(actual)):
```

```
[102]: # re define error metric same as for ML models
              temp.append(actual[i] - predicted[i])
          return temp
      def mae(actual, predicted):
          """ Mean Absolute Error """
          temp = _error(actual, predicted)
          sum_error = 0
          for i in temp:
              sum_error += abs(i)
          return sum_error/len(temp)
      def nmae(actual, predicted):
          """ Normalized Mean Absolute Error """
          return mae(actual, predicted) / (max(actual) - min(actual))
[103]: global_model_name['First_NN']={
              "Test": nmae(test_pred_ratings_list,test_df_structured_target_list),
              "Train": nmae(train_pred_ratings_list,train_df_structured_target_list)
              }
      print('Result of model')
      print(global_model_name['First_NN'])
```

```
Result of model {'Test': 0.19394950246246248, 'Train': 0.19275586777667447}
```

1.1.3 6. Deep learning model 2

```
[72]: import pandas as pd
     from sklearn import preprocessing
     min_max_scaler_X = preprocessing.MinMaxScaler()
     train df structured scaled = min max scaler X.fit transform(train df structured)
     test_df_structured_scaled = min_max_scaler_X.transform(test_df_structured)
[108]: import tensorflow as tf
     #define DL model
     first_model = tf.keras.Sequential([
        tf.keras.layers.Dense(64, activation=tf.nn.relu, input_shape=[8]),
        tf.keras.layers.Dense(32, activation=tf.nn.relu),
        tf.keras.layers.Dense(16, activation=tf.nn.relu),
        tf.keras.layers.Dense(8, activation=tf.nn.relu),
        tf.keras.layers.Dense(4, activation=tf.nn.relu),
        tf.keras.layers.Dense(2, activation=tf.nn.relu),
        tf.keras.layers.Dense(1)
       1)
[109]: #define call backs
     filepath="weights-improvement-1.hdf5"
     checkpoint = tf.keras.callbacks.ModelCheckpoint(filepath, monitor='val_loss',_
      →verbose=1, save_best_only=True, mode='min')
     tbCallBack = tf.keras.callbacks.TensorBoard(log_dir='./logs',profile_batch = __
      →100000001)
     es = tf.keras.callbacks.EarlyStopping(monitor='val_loss', mode='min',__
      →verbose=1, patience=40)
     callbacks=[checkpoint,tbCallBack,es]
[110]: if os.path.isfile('weights-improvement-1.hdf5'):
        pass
     else:
         #define optimizer and train model
        optimizer = tf.keras.optimizers.Adam()
        first model.compile(loss='mean absolute error',optimizer=optimizer)
        first_model.fit(np.asarray(train_df_structured_scaled),np.
      -asarray(train_df_structured_target),epochs=500,batch_size=512,validation_split=0.
      →3, callbacks=callbacks)
    Train on 2026816 samples, validate on 868636 samples
    Epoch 1/500
    Epoch 00001: val_loss improved from inf to 3.92380, saving model to weights-
    improvement-1.hdf5
    - val loss: 3.9238
    Epoch 2/500
```

```
Epoch 00002: val_loss improved from 3.92380 to 3.86586, saving model to weights-
improvement-1.hdf5
- val loss: 3.8659
Epoch 3/500
Epoch 00003: val loss improved from 3.86586 to 3.85627, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8563
Epoch 4/500
Epoch 00004: val_loss did not improve from 3.85627
- val_loss: 3.8582
Epoch 5/500
Epoch 00005: val_loss improved from 3.85627 to 3.85119, saving model to weights-
improvement-1.hdf5
- val loss: 3.8512
Epoch 6/500
Os - loss:
Epoch 00006: val_loss improved from 3.85119 to 3.85077, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8508
Epoch 7/500
Epoch 00007: val_loss improved from 3.85077 to 3.84969, saving model to weights-
improvement-1.hdf5
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8533
- val_loss: 3.8497
Epoch 8/500
Epoch 00008: val_loss improved from 3.84969 to 3.84860, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8486
Epoch 9/500
Epoch 00009: val_loss did not improve from 3.84860
- val_loss: 3.8499
Epoch 10/500
Epoch 00010: val_loss improved from 3.84860 to 3.84683, saving model to weights-
```

```
improvement-1.hdf5
- val_loss: 3.8468
Epoch 11/500
Epoch 00011: val_loss did not improve from 3.84683
- val_loss: 3.8507
Epoch 12/500
Epoch 00012: val_loss did not improve from 3.84683
- val_loss: 3.8504
Epoch 13/500
Epoch 00013: val_loss did not improve from 3.84683
- val_loss: 3.8473
Epoch 14/500
Epoch 00014: val_loss improved from 3.84683 to 3.84639, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8464
Epoch 15/500
Epoch 00015: val_loss did not improve from 3.84639
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8508
- val_loss: 3.8479
Epoch 16/500
Epoch 00016: val_loss did not improve from 3.84639
- val_loss: 3.8469
Epoch 17/500
Epoch 00017: val loss did not improve from 3.84639
- val_loss: 3.8478
Epoch 18/500
Epoch 00018: val_loss did not improve from 3.84639
- val_loss: 3.8482
Epoch 19/500
Epoch 00019: val_loss did not improve from 3.84639
```

```
- val_loss: 3.8466
Epoch 20/500
Epoch 00020: val_loss did not improve from 3.84639
- val_loss: 3.8465
Epoch 21/500
Epoch 00021: val_loss improved from 3.84639 to 3.84552, saving model to weights-
improvement-1.hdf5
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8500
- val_loss: 3.8455
Epoch 22/500
Epoch 00022: val_loss did not improve from 3.84552
- val_loss: 3.8468
Epoch 23/500
Epoch 00023: val_loss did not improve from 3.84552
- val loss: 3.8473
Epoch 24/500
Epoch 00024: val_loss did not improve from 3.84552
- val_loss: 3.8481
Epoch 25/500
Epoch 00025: val_loss did not improve from 3.84552
- val_loss: 3.8482
Epoch 26/500
Epoch 00026: val_loss did not improve from 3.84552
- val loss: 3.8470
Epoch 27/500
Epoch 00027: val_loss did not improve from 3.84552
- val_loss: 3.8487
Epoch 28/500
Epoch 00028: val_loss did not improve from 3.84552
- val_loss: 3.8461
```

```
Epoch 29/500
Epoch 00029: val_loss did not improve from 3.84552
- val loss: 3.8459
Epoch 30/500
Epoch 00030: val_loss did not improve from 3.84552
- val_loss: 3.8491
Epoch 31/500
Epoch 00031: val_loss did not improve from 3.84552
- val_loss: 3.8468
Epoch 32/500
Epoch 00032: val_loss did not improve from 3.84552
- val loss: 3.8482
Epoch 33/500
Epoch 00033: val_loss did not improve from 3.84552
- val_loss: 3.8482
Epoch 34/500
Epoch 00034: val_loss did not improve from 3.84552
- val_loss: 3.8473
Epoch 35/500
Epoch 00035: val_loss did not improve from 3.84552
- val loss: 3.8465
Epoch 36/500
Epoch 00036: val_loss did not improve from 3.84552
- val_loss: 3.8512
Epoch 37/500
Epoch 00037: val_loss improved from 3.84552 to 3.84550, saving model to weights-
improvement-1.hdf5
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8489
- val_loss: 3.8455
Epoch 38/500
```

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Epoch 00038: val_loss did not improve from 3.84550
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8490
- val_loss: 3.8485
Epoch 39/500
Epoch 00039: val_loss did not improve from 3.84550
- val_loss: 3.8456
Epoch 40/500
Epoch 00040: val_loss did not improve from 3.84550
- val_loss: 3.8503
Epoch 41/500
Epoch 00041: val_loss did not improve from 3.84550
- val_loss: 3.8459
Epoch 42/500
Epoch 00042: val_loss did not improve from 3.84550
- val_loss: 3.8469
Epoch 43/500
Epoch 00043: val_loss did not improve from 3.84550
- val_loss: 3.8463
Epoch 44/500
Epoch 00044: val_loss did not improve from 3.84550
- val_loss: 3.8456
Epoch 45/500
Epoch 00045: val_loss improved from 3.84550 to 3.84513, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8451
Epoch 46/500
Epoch 00046: val_loss did not improve from 3.84513
- val_loss: 3.8480
Epoch 47/500
Epoch 00047: val_loss did not improve from 3.84513
```

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- val_loss: 3.8460
Epoch 48/500
Epoch 00048: val_loss did not improve from 3.84513
- val loss: 3.8478
Epoch 49/500
Epoch 00049: val_loss did not improve from 3.84513
- val_loss: 3.8453
Epoch 50/500
Epoch 00050: val_loss improved from 3.84513 to 3.84487, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8449
Epoch 51/500
Epoch 00051: val loss did not improve from 3.84487
- val loss: 3.8474
Epoch 52/500
Epoch 00052: val_loss did not improve from 3.84487
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8479
- val_loss: 3.8471
Epoch 53/500
Epoch 00053: val_loss improved from 3.84487 to 3.84479, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8448
Epoch 54/500
Epoch 00054: val_loss did not improve from 3.84479
- val_loss: 3.8450
Epoch 55/500
Epoch 00055: val_loss did not improve from 3.84479
- val_loss: 3.8460
Epoch 56/500
Epoch 00056: val_loss did not improve from 3.84479
- val_loss: 3.8472
```

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Epoch 57/500
Epoch 00057: val_loss did not improve from 3.84479
- val loss: 3.8453
Epoch 58/500
Epoch 00058: val_loss did not improve from 3.84479
- val_loss: 3.8459
Epoch 59/500
Epoch 00059: val_loss did not improve from 3.84479
- val_loss: 3.8449
Epoch 60/500
Epoch 00060: val_loss did not improve from 3.84479
- val loss: 3.8459
Epoch 61/500
Epoch 00061: val_loss improved from 3.84479 to 3.84470, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8447
Epoch 62/500
Epoch 00062: val_loss did not improve from 3.84470
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8473
- val_loss: 3.8474
Epoch 63/500
Epoch 00063: val_loss did not improve from 3.84470
- val loss: 3.8460
Epoch 64/500
Epoch 00064: val_loss did not improve from 3.84470
- val_loss: 3.8459
Epoch 65/500
Epoch 00065: val_loss did not improve from 3.84470
- val_loss: 3.8461
Epoch 66/500
```

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Epoch 00066: val_loss did not improve from 3.84470
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8471
- val_loss: 3.8481
Epoch 67/500
Epoch 00067: val_loss improved from 3.84470 to 3.84465, saving model to weights-
improvement-1.hdf5
- val loss: 3.8446
Epoch 68/500
Epoch 00068: val_loss improved from 3.84465 to 3.84441, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8444
Epoch 69/500
Epoch 00069: val_loss did not improve from 3.84441
- val loss: 3.8459
Epoch 70/500
Epoch 00070: val_loss did not improve from 3.84441
- val_loss: 3.8447
Epoch 71/500
Epoch 00071: val_loss improved from 3.84441 to 3.84406, saving model to weights-
improvement-1.hdf5
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8469
- val_loss: 3.8441
Epoch 72/500
Epoch 00072: val_loss did not improve from 3.84406
- val loss: 3.8473
Epoch 73/500
Epoch 00073: val_loss did not improve from 3.84406
- val_loss: 3.8446
Epoch 74/500
Epoch 00074: val_loss did not improve from 3.84406
- val_loss: 3.8441
Epoch 75/500
```

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Epoch 00075: val_loss did not improve from 3.84406
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8467
- val_loss: 3.8445
Epoch 76/500
Epoch 00076: val_loss improved from 3.84406 to 3.84386, saving model to weights-
improvement-1.hdf5
- val loss: 3.8439
Epoch 77/500
Epoch 00077: val_loss did not improve from 3.84386
- val_loss: 3.8462
Epoch 78/500
Os - loss:
Epoch 00078: val_loss did not improve from 3.84386
- val loss: 3.8442
Epoch 79/500
Epoch 00079: val_loss did not improve from 3.84386
- val_loss: 3.8445
Epoch 80/500
Epoch 00080: val_loss did not improve from 3.84386
- val_loss: 3.8443
Epoch 81/500
Epoch 00081: val_loss improved from 3.84386 to 3.84317, saving model to weights-
improvement-1.hdf5
- val loss: 3.8432
Epoch 82/500
Epoch 00082: val_loss did not improve from 3.84317
- val_loss: 3.8446
Epoch 83/500
Epoch 00083: val_loss did not improve from 3.84317
- val_loss: 3.8449
Epoch 84/500
```

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Epoch 00084: val_loss did not improve from 3.84317
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8464
- val_loss: 3.8452
Epoch 85/500
Epoch 00085: val_loss did not improve from 3.84317
- val_loss: 3.8438
Epoch 86/500
Epoch 00086: val_loss did not improve from 3.84317
- val_loss: 3.8446
Epoch 87/500
Epoch 00087: val_loss did not improve from 3.84317
- val_loss: 3.8432
Epoch 88/500
Epoch 00088: val_loss did not improve from 3.84317
- val_loss: 3.8450
Epoch 89/500
Epoch 00089: val_loss did not improve from 3.84317
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8461
- val_loss: 3.8434
Epoch 90/500
Epoch 00090: val_loss did not improve from 3.84317
- val_loss: 3.8495
Epoch 91/500
Epoch 00091: val_loss did not improve from 3.84317
- val_loss: 3.8432
Epoch 92/500
Epoch 00092: val_loss did not improve from 3.84317
- val_loss: 3.8449
Epoch 93/500
Epoch 00093: val_loss improved from 3.84317 to 3.84304, saving model to weights-
improvement-1.hdf5
```

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- val_loss: 3.8430
Epoch 94/500
Epoch 00094: val_loss did not improve from 3.84304
- val_loss: 3.8444
Epoch 95/500
Epoch 00095: val_loss did not improve from 3.84304
- val_loss: 3.8437
Epoch 96/500
Epoch 00096: val_loss did not improve from 3.84304
- val_loss: 3.8446
Epoch 97/500
Epoch 00097: val_loss did not improve from 3.84304
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8457
- val loss: 3.8441
Epoch 98/500
Epoch 00098: val_loss did not improve from 3.84304
- val_loss: 3.8443
Epoch 99/500
Epoch 00099: val_loss did not improve from 3.84304
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8457
- val_loss: 3.8450
Epoch 100/500
Epoch 00100: val_loss did not improve from 3.84304
- val loss: 3.8442
Epoch 101/500
Epoch 00101: val_loss did not improve from 3.84304
- val_loss: 3.8456
Epoch 102/500
Epoch 00102: val_loss did not improve from 3.84304
- val_loss: 3.8473
Epoch 103/500
```

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Epoch 00103: val_loss improved from 3.84304 to 3.84259, saving model to weights-
improvement-1.hdf5
- val loss: 3.8426
Epoch 104/500
Epoch 00104: val_loss did not improve from 3.84259
- val loss: 3.8449
Epoch 105/500
Epoch 00105: val_loss did not improve from 3.84259
- val_loss: 3.8431
Epoch 106/500
Epoch 00106: val_loss did not improve from 3.84259
- val_loss: 3.8472
Epoch 107/500
Epoch 00107: val_loss did not improve from 3.84259
- val_loss: 3.8453
Epoch 108/500
0s - loss: 3
Epoch 00108: val_loss did not improve from 3.84259
- val_loss: 3.8441
Epoch 109/500
Epoch 00109: val_loss did not improve from 3.84259
- val loss: 3.8441
Epoch 110/500
Epoch 00110: val_loss did not improve from 3.84259
- val_loss: 3.8428
Epoch 111/500
Epoch 00111: val_loss did not improve from 3.84259
- val_loss: 3.8429
Epoch 112/500
Epoch 00112: val_loss did not improve from 3.84259
```

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- val_loss: 3.8485
Epoch 113/500
Epoch 00113: val_loss did not improve from 3.84259
- val loss: 3.8460
Epoch 114/500
Epoch 00114: val_loss did not improve from 3.84259
- val_loss: 3.8434
Epoch 115/500
Epoch 00115: val_loss did not improve from 3.84259
- val_loss: 3.8432
Epoch 116/500
Epoch 00116: val_loss did not improve from 3.84259
- val loss: 3.8474
Epoch 117/500
Epoch 00117: val_loss did not improve from 3.84259
- val_loss: 3.8430
Epoch 118/500
Epoch 00118: val_loss did not improve from 3.84259
- val_loss: 3.8445
Epoch 119/500
Epoch 00119: val loss did not improve from 3.84259
- val loss: 3.8446
Epoch 120/500
Epoch 00120: val_loss did not improve from 3.84259
- val_loss: 3.8450
Epoch 121/500
Epoch 00121: val_loss did not improve from 3.84259
- val_loss: 3.8429
Epoch 122/500
```

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Epoch 00122: val_loss did not improve from 3.84259
- val loss: 3.8444
Epoch 123/500
Epoch 00123: val_loss did not improve from 3.84259
- val loss: 3.8435
Epoch 124/500
Epoch 00124: val_loss did not improve from 3.84259
- val_loss: 3.8450
Epoch 125/500
Epoch 00125: val_loss did not improve from 3.84259
- val_loss: 3.8453
Epoch 126/500
Epoch 00126: val loss did not improve from 3.84259
- val_loss: 3.8448
Epoch 127/500
Epoch 00127: val_loss did not improve from 3.84259
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8450
- val_loss: 3.8441
Epoch 128/500
Epoch 00128: val_loss did not improve from 3.84259
- val_loss: 3.8440
Epoch 129/500
Epoch 00129: val loss did not improve from 3.84259
- val_loss: 3.8427
Epoch 130/500
Epoch 00130: val_loss did not improve from 3.84259
- val_loss: 3.8457
Epoch 131/500
Epoch 00131: val_loss did not improve from 3.84259
```

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- val_loss: 3.8437
Epoch 132/500
Epoch 00132: val_loss did not improve from 3.84259
- val_loss: 3.8446
Epoch 133/500
Epoch 00133: val_loss did not improve from 3.84259
- val_loss: 3.8432
Epoch 134/500
Epoch 00134: val_loss improved from 3.84259 to 3.84232, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8423
Epoch 135/500
Epoch 00135: val loss did not improve from 3.84232
- val loss: 3.8427
Epoch 136/500
Epoch 00136: val_loss did not improve from 3.84232
- val_loss: 3.8443
Epoch 137/500
Epoch 00137: val_loss did not improve from 3.84232
- val_loss: 3.8425
Epoch 138/500
Epoch 00138: val loss did not improve from 3.84232
- val loss: 3.8454
Epoch 139/500
Epoch 00139: val_loss did not improve from 3.84232
- val_loss: 3.8432
Epoch 140/500
Epoch 00140: val_loss did not improve from 3.84232
- val_loss: 3.8438
Epoch 141/500
```

```
Epoch 00141: val_loss did not improve from 3.84232
- val loss: 3.8432
Epoch 142/500
Epoch 00142: val_loss did not improve from 3.84232
- val loss: 3.8436
Epoch 143/500
Epoch 00143: val_loss did not improve from 3.84232
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8447
- val_loss: 3.8469
Epoch 144/500
Epoch 00144: val_loss did not improve from 3.84232
- val_loss: 3.8442
Epoch 145/500
Epoch 00145: val loss did not improve from 3.84232
- val_loss: 3.8436
Epoch 146/500
Epoch 00146: val_loss did not improve from 3.84232
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8447
- val_loss: 3.8423
Epoch 147/500
Epoch 00147: val_loss did not improve from 3.84232
- val_loss: 3.8447
Epoch 148/500
Epoch 00148: val loss did not improve from 3.84232
- val_loss: 3.8435
Epoch 149/500
Epoch 00149: val_loss did not improve from 3.84232
- val_loss: 3.8425
Epoch 150/500
Epoch 00150: val_loss did not improve from 3.84232
```

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- val_loss: 3.8435
Epoch 151/500
Epoch 00151: val loss did not improve from 3.84232
- val loss: 3.8442
Epoch 152/500
Epoch 00152: val_loss did not improve from 3.84232
- val_loss: 3.8443
Epoch 153/500
Epoch 00153: val_loss did not improve from 3.84232
- val_loss: 3.8424
Epoch 154/500
Epoch 00154: val loss did not improve from 3.84232
- val loss: 3.8433
Epoch 155/500
Epoch 00155: val_loss did not improve from 3.84232
- val_loss: 3.8432
Epoch 156/500
Epoch 00156: val_loss improved from 3.84232 to 3.84218, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8422
Epoch 157/500
Epoch 00157: val_loss did not improve from 3.84218
- val_loss: 3.8433
Epoch 158/500
Epoch 00158: val_loss did not improve from 3.84218
- val_loss: 3.8425
Epoch 159/500
Epoch 00159: val_loss did not improve from 3.84218
- val_loss: 3.8430
```

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Epoch 160/500
Epoch 00160: val_loss did not improve from 3.84218
- val loss: 3.8424
Epoch 161/500
Epoch 00161: val_loss did not improve from 3.84218
- val_loss: 3.8434
Epoch 162/500
Epoch 00162: val_loss did not improve from 3.84218
- val_loss: 3.8428
Epoch 163/500
Epoch 00163: val_loss did not improve from 3.84218
- val loss: 3.8452
Epoch 164/500
Epoch 00164: val_loss did not improve from 3.84218
- val_loss: 3.8424
Epoch 165/500
Epoch 00165: val_loss did not improve from 3.84218
- val_loss: 3.8451
Epoch 166/500
Epoch 00166: val_loss did not improve from 3.84218
- val loss: 3.8433
Epoch 167/500
Epoch 00167: val_loss improved from 3.84218 to 3.84211, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8421
Epoch 168/500
Epoch 00168: val_loss did not improve from 3.84211
- val_loss: 3.8425
Epoch 169/500
```

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Epoch 00169: val_loss did not improve from 3.84211
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8440
- val_loss: 3.8427
Epoch 170/500
Epoch 00170: val_loss did not improve from 3.84211
- val_loss: 3.8447
Epoch 171/500
Epoch 00171: val_loss improved from 3.84211 to 3.84176, saving model to weights-
improvement-1.hdf5
- val_loss: 3.8418
Epoch 172/500
Epoch 00172: val_loss did not improve from 3.84176
- val loss: 3.8426
Epoch 173/500
Epoch 00173: val loss did not improve from 3.84176
- val_loss: 3.8440
Epoch 174/500
Epoch 00174: val_loss did not improve from 3.84176
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8441
- val_loss: 3.8437
Epoch 175/500
Epoch 00175: val_loss did not improve from 3.84176
- val_loss: 3.8442
Epoch 176/500
Epoch 00176: val loss did not improve from 3.84176
- val_loss: 3.8449
Epoch 177/500
Epoch 00177: val_loss did not improve from 3.84176
- val_loss: 3.8430
Epoch 178/500
Epoch 00178: val_loss did not improve from 3.84176
```

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- val_loss: 3.8430
Epoch 179/500
Epoch 00179: val_loss did not improve from 3.84176
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8440
- val_loss: 3.8431
Epoch 180/500
Epoch 00180: val_loss did not improve from 3.84176
- val_loss: 3.8436
Epoch 181/500
Epoch 00181: val_loss did not improve from 3.84176
- val_loss: 3.8421
Epoch 182/500
Epoch 00182: val_loss did not improve from 3.84176
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8440
- val loss: 3.8426
Epoch 183/500
Epoch 00183: val_loss did not improve from 3.84176
- val_loss: 3.8421
Epoch 184/500
Epoch 00184: val_loss did not improve from 3.84176
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8439
- val_loss: 3.8425
Epoch 185/500
Epoch 00185: val_loss did not improve from 3.84176
- val loss: 3.8429
Epoch 186/500
Epoch 00186: val_loss did not improve from 3.84176
- val_loss: 3.8429
Epoch 187/500
Epoch 00187: val_loss did not improve from 3.84176
- val_loss: 3.8418
Epoch 188/500
```

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Epoch 00188: val_loss did not improve from 3.84176
- val_loss: 3.8420
Epoch 189/500
Epoch 00189: val_loss did not improve from 3.84176
- val_loss: 3.8429
Epoch 190/500
Epoch 00190: val_loss did not improve from 3.84176
- val_loss: 3.8425
Epoch 191/500
Epoch 00191: val_loss improved from 3.84176 to 3.84171, saving model to weights-
improvement-1.hdf5
- val loss: 3.8417
Epoch 192/500
Epoch 00192: val loss did not improve from 3.84171
- val_loss: 3.8428
Epoch 193/500
Epoch 00193: val_loss did not improve from 3.84171
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8439
- val_loss: 3.8429
Epoch 194/500
Epoch 00194: val_loss improved from 3.84171 to 3.84165, saving model to weights-
improvement-1.hdf5
- val loss: 3.8416
Epoch 195/500
Epoch 00195: val_loss did not improve from 3.84165
- val_loss: 3.8433
Epoch 196/500
Epoch 00196: val_loss did not improve from 3.84165
- val_loss: 3.8420
Epoch 197/500
Epoch 00197: val_loss improved from 3.84165 to 3.84154, saving model to weights-
```

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improvement-1.hdf5
2026816/2026816 [============== ] - 11s 6us/sample - loss: 3.8436
- val_loss: 3.8415
Epoch 198/500
Epoch 00198: val_loss did not improve from 3.84154
- val_loss: 3.8528
Epoch 199/500
Epoch 00199: val_loss did not improve from 3.84154
- val_loss: 3.8421
Epoch 200/500
Epoch 00200: val_loss did not improve from 3.84154
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8436
- val_loss: 3.8435
Epoch 201/500
Epoch 00201: val_loss did not improve from 3.84154
- val_loss: 3.8432
Epoch 202/500
Epoch 00202: val_loss did not improve from 3.84154
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8435
- val_loss: 3.8426
Epoch 203/500
Epoch 00203: val_loss did not improve from 3.84154
- val_loss: 3.8426
Epoch 204/500
Epoch 00204: val_loss did not improve from 3.84154
- val_loss: 3.8434
Epoch 205/500
Epoch 00205: val_loss did not improve from 3.84154
- val_loss: 3.8471
Epoch 206/500
Epoch 00206: val_loss improved from 3.84154 to 3.84140, saving model to weights-
improvement-1.hdf5
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8436
```

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- val_loss: 3.8414
Epoch 207/500
Epoch 00207: val_loss did not improve from 3.84140
- val loss: 3.8423
Epoch 208/500
Epoch 00208: val_loss did not improve from 3.84140
- val_loss: 3.8430
Epoch 209/500
Epoch 00209: val_loss did not improve from 3.84140
- val_loss: 3.8416
Epoch 210/500
Epoch 00210: val_loss did not improve from 3.84140
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8434
- val loss: 3.8422
Epoch 211/500
Epoch 00211: val_loss did not improve from 3.84140
- val_loss: 3.8421
Epoch 212/500
Epoch 00212: val_loss did not improve from 3.84140
- val_loss: 3.8441
Epoch 213/500
Epoch 00213: val_loss did not improve from 3.84140
- val loss: 3.8427
Epoch 214/500
Epoch 00214: val_loss did not improve from 3.84140
- val_loss: 3.8419
Epoch 215/500
Epoch 00215: val_loss did not improve from 3.84140
- val_loss: 3.8428
Epoch 216/500
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Epoch 00216: val_loss did not improve from 3.84140
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8433
- val_loss: 3.8428
Epoch 217/500
Epoch 00217: val_loss did not improve from 3.84140
- val_loss: 3.8442
Epoch 218/500
Epoch 00218: val_loss did not improve from 3.84140
- val_loss: 3.8451
Epoch 219/500
Epoch 00219: val_loss did not improve from 3.84140
- val_loss: 3.8419
Epoch 220/500
Epoch 00220: val_loss did not improve from 3.84140
- val_loss: 3.8424
Epoch 221/500
Epoch 00221: val_loss did not improve from 3.84140
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8432
- val_loss: 3.8448
Epoch 222/500
Epoch 00222: val_loss did not improve from 3.84140
- val_loss: 3.8439
Epoch 223/500
Epoch 00223: val_loss did not improve from 3.84140
- val_loss: 3.8425
Epoch 224/500
Epoch 00224: val_loss did not improve from 3.84140
- val_loss: 3.8446
Epoch 225/500
Epoch 00225: val_loss did not improve from 3.84140
- val_loss: 3.8439
```

```
Epoch 226/500
Epoch 00226: val_loss did not improve from 3.84140
- val loss: 3.8419
Epoch 227/500
Epoch 00227: val_loss did not improve from 3.84140
- val_loss: 3.8418
Epoch 228/500
Epoch 00228: val_loss did not improve from 3.84140
- val_loss: 3.8417
Epoch 229/500
Epoch 00229: val_loss did not improve from 3.84140
- val loss: 3.8437
Epoch 230/500
Epoch 00230: val_loss did not improve from 3.84140
- val_loss: 3.8469
Epoch 231/500
Epoch 00231: val_loss did not improve from 3.84140
- val_loss: 3.8422
Epoch 232/500
Epoch 00232: val_loss improved from 3.84140 to 3.84106, saving model to weights-
improvement-1.hdf5
- val loss: 3.8411
Epoch 233/500
Epoch 00233: val_loss did not improve from 3.84106
- val_loss: 3.8427
Epoch 234/500
Epoch 00234: val_loss did not improve from 3.84106
- val_loss: 3.8425
Epoch 235/500
```

```
Epoch 00235: val_loss did not improve from 3.84106
- val_loss: 3.8432
Epoch 236/500
Epoch 00236: val_loss did not improve from 3.84106
- val_loss: 3.8412
Epoch 237/500
Epoch 00237: val_loss did not improve from 3.84106
- val_loss: 3.8427
Epoch 238/500
Epoch 00238: val_loss did not improve from 3.84106
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8430
- val_loss: 3.8422
Epoch 239/500
Epoch 00239: val_loss did not improve from 3.84106
- val_loss: 3.8413
Epoch 240/500
Epoch 00240: val_loss did not improve from 3.84106
- val_loss: 3.8426
Epoch 241/500
Epoch 00241: val_loss did not improve from 3.84106
- val_loss: 3.8413
Epoch 242/500
Epoch 00242: val_loss did not improve from 3.84106
- val_loss: 3.8418
Epoch 243/500
Epoch 00243: val_loss did not improve from 3.84106
- val_loss: 3.8426
Epoch 244/500
Epoch 00244: val_loss did not improve from 3.84106
- val_loss: 3.8435
```

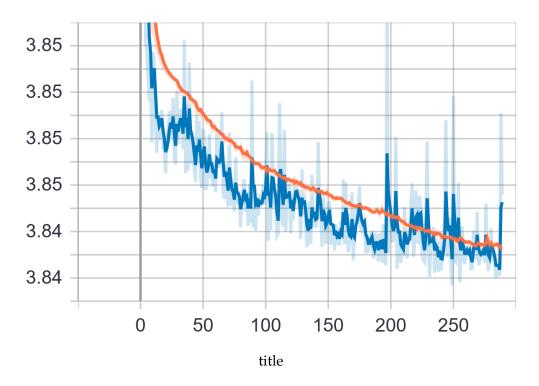
```
Epoch 245/500
Epoch 00245: val_loss did not improve from 3.84106
- val loss: 3.8478
Epoch 246/500
Epoch 00246: val_loss did not improve from 3.84106
2026816/2026816 [============== ] - 11s 6us/sample - loss: 3.8427
- val_loss: 3.8419
Epoch 247/500
Epoch 00247: val_loss did not improve from 3.84106
- val_loss: 3.8422
Epoch 248/500
Epoch 00248: val_loss did not improve from 3.84106
- val loss: 3.8436
Epoch 249/500
Epoch 00249: val_loss did not improve from 3.84106
- val_loss: 3.8415
Epoch 250/500
Epoch 00250: val_loss improved from 3.84106 to 3.84068, saving model to weights-
improvement-1.hdf5
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8427
- val_loss: 3.8407
Epoch 251/500
Epoch 00251: val_loss did not improve from 3.84068
- val loss: 3.8488
Epoch 252/500
Epoch 00252: val_loss did not improve from 3.84068
- val_loss: 3.8417
Epoch 253/500
Epoch 00253: val_loss did not improve from 3.84068
- val_loss: 3.8417
Epoch 254/500
```

```
Epoch 00254: val_loss did not improve from 3.84068
- val_loss: 3.8424
Epoch 255/500
Epoch 00255: val_loss did not improve from 3.84068
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8427
- val_loss: 3.8451
Epoch 256/500
Epoch 00256: val_loss did not improve from 3.84068
- val_loss: 3.8408
Epoch 257/500
Epoch 00257: val_loss did not improve from 3.84068
2026816/2026816 [============= ] - 11s 6us/sample - loss: 3.8428
- val_loss: 3.8417
Epoch 258/500
Epoch 00258: val_loss did not improve from 3.84068
- val_loss: 3.8432
Epoch 259/500
Epoch 00259: val_loss did not improve from 3.84068
- val_loss: 3.8420
Epoch 260/500
Epoch 00260: val_loss did not improve from 3.84068
- val_loss: 3.8416
Epoch 261/500
Epoch 00261: val_loss did not improve from 3.84068
- val_loss: 3.8420
Epoch 262/500
Epoch 00262: val_loss did not improve from 3.84068
- val_loss: 3.8424
Epoch 263/500
Epoch 00263: val_loss did not improve from 3.84068
- val_loss: 3.8419
```

```
Epoch 264/500
Epoch 00264: val_loss did not improve from 3.84068
- val loss: 3.8423
Epoch 265/500
Epoch 00265: val_loss did not improve from 3.84068
- val_loss: 3.8414
Epoch 266/500
Epoch 00266: val_loss did not improve from 3.84068
- val_loss: 3.8414
Epoch 267/500
Epoch 00267: val_loss did not improve from 3.84068
- val loss: 3.8430
Epoch 268/500
Epoch 00268: val_loss did not improve from 3.84068
- val_loss: 3.8429
Epoch 269/500
Epoch 00269: val_loss did not improve from 3.84068
- val_loss: 3.8418
Epoch 270/500
Epoch 00270: val_loss did not improve from 3.84068
- val loss: 3.8438
Epoch 271/500
Epoch 00271: val_loss did not improve from 3.84068
- val_loss: 3.8416
Epoch 272/500
Epoch 00272: val_loss did not improve from 3.84068
- val_loss: 3.8417
Epoch 273/500
Epoch 00273: val_loss did not improve from 3.84068
```

```
- val_loss: 3.8419
Epoch 274/500
Epoch 00274: val loss did not improve from 3.84068
- val loss: 3.8422
Epoch 275/500
Epoch 00275: val_loss did not improve from 3.84068
- val_loss: 3.8424
Epoch 276/500
Epoch 00276: val_loss did not improve from 3.84068
- val_loss: 3.8417
Epoch 277/500
Epoch 00277: val_loss did not improve from 3.84068
- val loss: 3.8442
Epoch 278/500
Epoch 00278: val_loss did not improve from 3.84068
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8430
- val_loss: 3.8418
Epoch 279/500
Epoch 00279: val_loss did not improve from 3.84068
- val_loss: 3.8413
Epoch 280/500
Epoch 00280: val loss did not improve from 3.84068
- val loss: 3.8417
Epoch 281/500
Epoch 00281: val_loss did not improve from 3.84068
- val_loss: 3.8431
Epoch 282/500
Epoch 00282: val_loss did not improve from 3.84068
- val_loss: 3.8423
Epoch 283/500
```

```
Epoch 00283: val_loss did not improve from 3.84068
- val loss: 3.8419
Epoch 284/500
Epoch 00284: val_loss did not improve from 3.84068
- val loss: 3.8416
Epoch 285/500
Epoch 00285: val_loss did not improve from 3.84068
- val loss: 3.8410
Epoch 286/500
Epoch 00286: val_loss did not improve from 3.84068
- val loss: 3.8415
Epoch 287/500
Epoch 00287: val loss did not improve from 3.84068
- val_loss: 3.8415
Epoch 288/500
Epoch 00288: val_loss did not improve from 3.84068
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8424
- val_loss: 3.8411
Epoch 289/500
Epoch 00289: val_loss did not improve from 3.84068
- val_loss: 3.8481
Epoch 290/500
Epoch 00290: val loss did not improve from 3.84068
- val loss: 3.8446
Epoch 00290: early stopping
```



```
[111]: #load model and predict
      first_best_model = tf.keras.models.load_model('weights-improvement-1.hdf5')
      test_pred_ratings = first_best_model.predict(np.
       →asarray(test_df_structured_scaled))
      train_pred_ratings = first_best_model.predict(np.
       →asanyarray(train_df_structured_scaled))
[112]: # i am getting ther error as: MemoryError: Unable to allocate array with shape
      \rightarrow (1240908, 1240908) and data type float64
      # while calculating NMAE
      # this is happenning as numpy can't put so many float64 values to memory.
        So lets try the python forloop way for calculating NMAE
[113]: test_pred_ratings = test_pred_ratings.tolist()
      train_pred_ratings = train_pred_ratings.tolist()
[114]: test_pred_ratings_list = [i[0] for i in test_pred_ratings]
      train_pred_ratings_list = [i[0] for i in train_pred_ratings]
[115]: test_df_structured_target_list=test_df_structured_target.values
      train_df_structured_target_list=train_df_structured_target.values
[116]: test_df_structured_target_list = test_df_structured_target_list.tolist()
      train_df_structured_target_list = train_df_structured_target_list.tolist()
[117]: global_model_name['Second_NN']={
              "Test": nmae(test_pred_ratings_list,test_df_structured_target_list),
              "Train": nmae(train_pred_ratings_list,train_df_structured_target_list)
              }
```

```
print('Result of model')
      print(global_model_name['Second_NN'])
     Result of model
     {'Test': 0.1493320631675459, 'Train': 0.18082728958955502}
        This model has shown best performance till now.
     1.1.4 6. Deep learning model 3
[141]: import pandas as pd
      from sklearn import preprocessing
      min max scaler X = preprocessing.MinMaxScaler()
      train_df_structured_scaled = min_max_scaler_X.fit_transform(train_df_structured)
      test df structured scaled = min max scaler X.transform(test df structured)
[153]: pickle.dump(min_max_scaler_X, open("min_max_scaler_X.pickle","wb"))
[142]: import tensorflow as tf
      #define DL model
      first_model = tf.keras.Sequential([
          tf.keras.layers.Dense(64, activation=tf.nn.relu, input_shape=[8]),
          tf.keras.layers.Dense(32, activation=tf.nn.relu),
          tf.keras.layers.Dropout(0.2),
          tf.keras.layers.Dense(16, activation=tf.nn.relu),
          tf.keras.layers.Dense(8, activation=tf.nn.relu),
          tf.keras.layers.BatchNormalization(),
          tf.keras.layers.Dense(4, activation=tf.nn.relu),
          tf.keras.layers.Dropout(0.2),
          tf.keras.layers.Dense(2, activation=tf.nn.relu),
          tf.keras.layers.Dense(1)
        ])
[144]: #define call backs
      filepath="weights-improvement-2.hdf5"
      checkpoint = tf.keras.callbacks.ModelCheckpoint(filepath, monitor='val_loss',_
       →verbose=1, save_best_only=True, mode='min')
      tbCallBack = tf.keras.callbacks.TensorBoard(log_dir='./logs',profile_batch = __
       →100000002)
      es = tf.keras.callbacks.EarlyStopping(monitor='val_loss', mode='min',_
       →verbose=1, patience=40)
      callbacks=[checkpoint,tbCallBack,es]
[145]: if os.path.isfile('weights-improvement-2.hdf5'):
          pass
      else:
          #define optimizer and train model
          optimizer = tf.keras.optimizers.Adam()
          first_model.compile(loss='mean_absolute_error',optimizer=optimizer)
```

```
first_model.fit(np.asarray(train_df_structured_scaled),np.

asarray(train_df_structured_target),epochs=500,batch_size=512,validation_split=0.

3,callbacks=callbacks)
```

```
Train on 2026816 samples, validate on 868636 samples
Epoch 1/500
Epoch 00001: val_loss improved from inf to 3.96076, saving model to weights-
improvement-2.hdf5
- val_loss: 3.9608
Epoch 2/500
Epoch 00002: val_loss improved from 3.96076 to 3.89359, saving model to weights-
improvement-2.hdf5
- val loss: 3.8936
Epoch 3/500
Epoch 00003: val_loss did not improve from 3.89359
- val_loss: 3.9050
Epoch 4/500
Epoch 00004: val_loss did not improve from 3.89359
- val_loss: 3.9101
Epoch 5/500
Epoch 00005: val_loss did not improve from 3.89359
- val_loss: 3.9476
Epoch 6/500
Epoch 00006: val_loss improved from 3.89359 to 3.89012, saving model to weights-
improvement-2.hdf5
- val_loss: 3.8901
Epoch 7/500
Epoch 00007: val_loss did not improve from 3.89012
- val loss: 3.8963
Epoch 8/500
Epoch 00008: val_loss did not improve from 3.89012
```

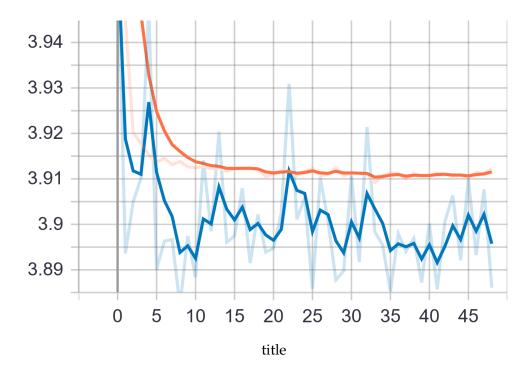
```
- val_loss: 3.8967
Epoch 9/500
Epoch 00009: val_loss improved from 3.89012 to 3.88214, saving model to weights-
improvement-2.hdf5
- val loss: 3.8821
Epoch 10/500
Epoch 00010: val_loss did not improve from 3.88214
2026816/2026816 [============= ] - 14s 7us/sample - loss: 3.9127
- val_loss: 3.8974
Epoch 11/500
Epoch 00011: val_loss did not improve from 3.88214
- val_loss: 3.8884
Epoch 12/500
Epoch 00012: val loss did not improve from 3.88214
- val loss: 3.9142
Epoch 13/500
Epoch 00013: val_loss did not improve from 3.88214
- val_loss: 3.8985
Epoch 14/500
Epoch 00014: val_loss did not improve from 3.88214
- val_loss: 3.9204
Epoch 15/500
Epoch 00015: val loss did not improve from 3.88214
- val loss: 3.8960
Epoch 16/500
Epoch 00016: val_loss did not improve from 3.88214
- val_loss: 3.8974
Epoch 17/500
Epoch 00017: val_loss did not improve from 3.88214
3.9124 - val_loss: 3.9080
Epoch 18/500
```

```
Epoch 00018: val_loss did not improve from 3.88214
2026816/2026816 [============= ] - 21s 10us/sample - loss:
3.9123 - val_loss: 3.8915
Epoch 19/500
Epoch 00019: val_loss did not improve from 3.88214
3.9120 - val_loss: 3.9022
Epoch 20/500
Epoch 00020: val_loss did not improve from 3.88214
3.9105 - val_loss: 3.8939
Epoch 21/500
Epoch 00021: val_loss did not improve from 3.88214
3.9110 - val_loss: 3.8947
Epoch 22/500
Epoch 00022: val loss did not improve from 3.88214
3.9118 - val_loss: 3.9025
Epoch 23/500
Epoch 00023: val_loss did not improve from 3.88214
3.9118 - val_loss: 3.9309
Epoch 24/500
Epoch 00024: val_loss did not improve from 3.88214
3.9107 - val_loss: 3.9011
Epoch 25/500
Epoch 00025: val loss did not improve from 3.88214
3.9116 - val_loss: 3.9058
Epoch 26/500
Epoch 00026: val_loss did not improve from 3.88214
3.9123 - val_loss: 3.8858
Epoch 27/500
Epoch 00027: val_loss did not improve from 3.88214
```

```
3.9107 - val_loss: 3.9102
Epoch 28/500
Epoch 00028: val_loss did not improve from 3.88214
3.9110 - val_loss: 3.9007
Epoch 29/500
Epoch 00029: val_loss did not improve from 3.88214
3.9124 - val_loss: 3.8878
Epoch 30/500
Epoch 00030: val_loss did not improve from 3.88214
3.9107 - val_loss: 3.8897
Epoch 31/500
Epoch 00031: val_loss did not improve from 3.88214
3.9113 - val loss: 3.9108
Epoch 32/500
Epoch 00032: val_loss did not improve from 3.88214
3.9112 - val_loss: 3.8917
Epoch 33/500
Epoch 00033: val_loss did not improve from 3.88214
3.9111 - val_loss: 3.9214
Epoch 34/500
Epoch 00034: val_loss did not improve from 3.88214
3.9092 - val_loss: 3.8984
Epoch 35/500
Epoch 00035: val_loss did not improve from 3.88214
3.9107 - val_loss: 3.8954
Epoch 36/500
Epoch 00036: val_loss did not improve from 3.88214
3.9113 - val_loss: 3.8852
Epoch 37/500
```

```
Epoch 00037: val_loss did not improve from 3.88214
3.9112 - val_loss: 3.8980
Epoch 38/500
Epoch 00038: val_loss did not improve from 3.88214
2026816/2026816 [============== ] - 22s 11us/sample - loss:
3.9100 - val_loss: 3.8941
Epoch 39/500
Epoch 00039: val_loss did not improve from 3.88214
3.9111 - val_loss: 3.8969
Epoch 40/500
Epoch 00040: val_loss did not improve from 3.88214
3.9106 - val_loss: 3.8872
Epoch 41/500
Epoch 00041: val_loss did not improve from 3.88214
3.9107 - val_loss: 3.9002
Epoch 42/500
Epoch 00042: val_loss did not improve from 3.88214
3.9113 - val_loss: 3.8857
Epoch 43/500
Epoch 00043: val_loss did not improve from 3.88214
2026816/2026816 [============= ] - 21s 10us/sample - loss:
3.9111 - val_loss: 3.9007
Epoch 44/500
Epoch 00044: val_loss did not improve from 3.88214
3.9106 - val_loss: 3.9064
Epoch 45/500
Epoch 00045: val_loss did not improve from 3.88214
3.9108 - val_loss: 3.8921
Epoch 46/500
Epoch 00046: val_loss did not improve from 3.88214
3.9104 - val_loss: 3.9101
```

```
Epoch 47/500
Epoch 00047: val_loss did not improve from 3.88214
3.9114 - val loss: 3.8933
Epoch 48/500
Epoch 00048: val_loss did not improve from 3.88214
3.9113 - val_loss: 3.9077
Epoch 49/500
Epoch 00049: val_loss did not improve from 3.88214
3.9122 - val_loss: 3.8860
Epoch 00049: early stopping
```



```
[148]: test_pred_ratings_list = [i[0] for i in test_pred_ratings]
      train_pred_ratings_list = [i[0] for i in train_pred_ratings]
[149]: test_df_structured_target_list=test_df_structured_target.values
      train_df_structured_target_list=train_df_structured_target.values
[150]: test_df_structured_target_list = test_df_structured_target_list.tolist()
      train_df_structured_target_list = train_df_structured_target_list.tolist()
[151]: global_model_name['Third_NN']={
              "Test": nmae(test_pred_ratings_list,test_df_structured_target_list),
              "Train": nmae(train_pred_ratings_list,train_df_structured_target_list)
      print('Result of model')
      print(global_model_name['Third_NN'])
     Result of model
     {'Test': 0.3205020373401373, 'Train': 0.31207603380814686}
        As this model was having less variance of data using Dropout backfired.
        As this model was having less number of deep layers using BatchNormalization
     backfired
     1.1.5 Lets try one model with all simple features
     1.1.6 6. Deep learning model 4
[156]: train_df_structured_simple =
      →train_df_structured[['user_avg','joke_avg','special_feature','gavg']]
      test df structured simple =
       →test_df_structured[['user_avg','joke_avg','special_feature','gavg']]
[159]: import pandas as pd
      from sklearn import preprocessing
      min_max_scaler_X_simple = preprocessing.MinMaxScaler()
      train_df_structured_simple_scaled = min_max_scaler_X.
       →fit_transform(train_df_structured_simple)
      test df_structured_simple_scaled = min_max_scaler_X.
       →transform(test_df_structured_simple)
[160]: pickle.dump(min_max_scaler_X_simple, open("min_max_scaler_X_simple.
       →pickle","wb"))
[163]: import tensorflow as tf
      #define DL model
      first_model = tf.keras.Sequential([
          tf.keras.layers.Dense(16, activation=tf.nn.relu, input_shape=[4]),
          tf.keras.layers.Dense(8, activation=tf.nn.relu),
          tf.keras.layers.Dense(4, activation=tf.nn.relu),
          tf.keras.layers.Dense(2, activation=tf.nn.relu),
          tf.keras.layers.Dense(1)
```

```
])
[164]: #define call backs
    filepath="weights-improvement-3.hdf5"
    checkpoint = tf.keras.callbacks.ModelCheckpoint(filepath, monitor='val_loss', __
    →verbose=1, save_best_only=True, mode='min')
    tbCallBack = tf.keras.callbacks.TensorBoard(log_dir='./logs',profile_batch =__
    →100000003)
    es = tf.keras.callbacks.EarlyStopping(monitor='val_loss', mode='min', u
    →verbose=1, patience=40)
    callbacks=[checkpoint,tbCallBack,es]
[165]: if os.path.isfile('weights-improvement-3.hdf5'):
    else:
      #define optimizer and train model
      optimizer = tf.keras.optimizers.Adam()
      first_model.compile(loss='mean_absolute_error',optimizer=optimizer)
      first_model.fit(np.asarray(train_df_structured_simple_scaled),np.
    -asarray(train_df_structured_target),epochs=500,batch_size=512,validation_split=0.
    →3, callbacks=callbacks)
   Train on 2026816 samples, validate on 868636 samples
   Epoch 1/500
   Epoch 00001: val_loss improved from inf to 3.85266, saving model to weights-
   improvement-3.hdf5
   - val_loss: 3.8527
   Epoch 2/500
   Epoch 00002: val_loss did not improve from 3.85266
   - val_loss: 3.8530
   Epoch 3/500
   Epoch 00003: val_loss improved from 3.85266 to 3.85072, saving model to weights-
   improvement-3.hdf5
   - val loss: 3.8507
   Epoch 4/500
   Epoch 00004: val_loss did not improve from 3.85072
   - val_loss: 3.8523
   Epoch 5/500
   Epoch 00005: val_loss improved from 3.85072 to 3.85045, saving model to weights-
```

```
improvement-3.hdf5
- val_loss: 3.8505
Epoch 6/500
Epoch 00006: val_loss did not improve from 3.85045
- val_loss: 3.8512
Epoch 7/500
Epoch 00007: val_loss did not improve from 3.85045
- val_loss: 3.8603
Epoch 8/500
Epoch 00008: val_loss did not improve from 3.85045
- val_loss: 3.8539
Epoch 9/500
Epoch 00009: val_loss did not improve from 3.85045
- val_loss: 3.8586
Epoch 10/500
Epoch 00010: val_loss improved from 3.85045 to 3.84906, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8491
Epoch 11/500
Epoch 00011: val_loss did not improve from 3.84906
- val_loss: 3.8495
Epoch 12/500
Epoch 00012: val loss did not improve from 3.84906
- val_loss: 3.8578
Epoch 13/500
Epoch 00013: val_loss did not improve from 3.84906
- val_loss: 3.8530
Epoch 14/500
Epoch 00014: val_loss did not improve from 3.84906
```

```
- val_loss: 3.8501
Epoch 15/500
Epoch 00015: val_loss did not improve from 3.84906
- val_loss: 3.8497
Epoch 16/500
Epoch 00016: val_loss did not improve from 3.84906
- val_loss: 3.8491
Epoch 17/500
Epoch 00017: val_loss did not improve from 3.84906
- val_loss: 3.8492
Epoch 18/500
Epoch 00018: val_loss did not improve from 3.84906
- val loss: 3.8501
Epoch 19/500
Epoch 00019: val_loss improved from 3.84906 to 3.84781, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8478
Epoch 20/500
Epoch 00020: val_loss improved from 3.84781 to 3.84722, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8472
Epoch 21/500
Epoch 00021: val_loss did not improve from 3.84722
- val_loss: 3.8486
Epoch 22/500
Epoch 00022: val_loss improved from 3.84722 to 3.84720, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8472
Epoch 23/500
Epoch 00023: val_loss did not improve from 3.84720
```

```
- val_loss: 3.8487
Epoch 24/500
Epoch 00024: val_loss did not improve from 3.84720
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8504
- val_loss: 3.8511
Epoch 25/500
Epoch 00025: val_loss improved from 3.84720 to 3.84707, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8471
Epoch 26/500
Epoch 00026: val_loss did not improve from 3.84707
- val_loss: 3.8503
Epoch 27/500
Epoch 00027: val_loss improved from 3.84707 to 3.84663, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8466
Epoch 28/500
Epoch 00028: val_loss did not improve from 3.84663
2026816/2026816 [============== ] - 11s 6us/sample - loss: 3.8501
- val_loss: 3.8485
Epoch 29/500
Epoch 00029: val_loss improved from 3.84663 to 3.84651, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8465
Epoch 30/500
Epoch 00030: val loss did not improve from 3.84651
- val_loss: 3.8472
Epoch 31/500
Epoch 00031: val_loss did not improve from 3.84651
- val_loss: 3.8480
Epoch 32/500
Epoch 00032: val_loss did not improve from 3.84651
```

```
- val_loss: 3.8467
Epoch 33/500
Epoch 00033: val_loss did not improve from 3.84651
- val_loss: 3.8490
Epoch 34/500
Epoch 00034: val_loss improved from 3.84651 to 3.84649, saving model to weights-
improvement-3.hdf5
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8500
- val_loss: 3.8465
Epoch 35/500
Epoch 00035: val_loss did not improve from 3.84649
- val_loss: 3.8538
Epoch 36/500
Epoch 00036: val loss did not improve from 3.84649
- val loss: 3.8490
Epoch 37/500
Epoch 00037: val_loss did not improve from 3.84649
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8497
- val_loss: 3.8474
Epoch 38/500
Epoch 00038: val_loss did not improve from 3.84649
- val_loss: 3.8506
Epoch 39/500
Epoch 00039: val_loss improved from 3.84649 to 3.84641, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8464
Epoch 40/500
Epoch 00040: val_loss improved from 3.84641 to 3.84619, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8462
Epoch 41/500
Epoch 00041: val_loss did not improve from 3.84619
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8497
```

```
- val_loss: 3.8472
Epoch 42/500
Epoch 00042: val_loss did not improve from 3.84619
- val_loss: 3.8464
Epoch 43/500
Epoch 00043: val_loss did not improve from 3.84619
- val_loss: 3.8500
Epoch 44/500
Epoch 00044: val_loss did not improve from 3.84619
- val_loss: 3.8467
Epoch 45/500
Epoch 00045: val_loss did not improve from 3.84619
- val loss: 3.8477
Epoch 46/500
Epoch 00046: val_loss did not improve from 3.84619
- val_loss: 3.8512
Epoch 47/500
Epoch 00047: val_loss did not improve from 3.84619
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8495
- val_loss: 3.8547
Epoch 48/500
Epoch 00048: val_loss did not improve from 3.84619
- val loss: 3.8481
Epoch 49/500
Epoch 00049: val_loss did not improve from 3.84619
- val_loss: 3.8465
Epoch 50/500
Epoch 00050: val_loss did not improve from 3.84619
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8494
- val_loss: 3.8477
Epoch 51/500
```

```
Epoch 00051: val_loss did not improve from 3.84619
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8494
- val_loss: 3.8468
Epoch 52/500
Epoch 00052: val_loss improved from 3.84619 to 3.84612, saving model to weights-
improvement-3.hdf5
- val loss: 3.8461
Epoch 53/500
Epoch 00053: val_loss did not improve from 3.84612
- val loss: 3.8471
Epoch 54/500
Epoch 00054: val_loss did not improve from 3.84612
- val_loss: 3.8472
Epoch 55/500
Epoch 00055: val_loss did not improve from 3.84612
- val_loss: 3.8468
Epoch 56/500
Epoch 00056: val_loss did not improve from 3.84612
- val_loss: 3.8506
Epoch 57/500
Epoch 00057: val_loss did not improve from 3.84612
- val_loss: 3.8462
Epoch 58/500
Epoch 00058: val loss did not improve from 3.84612
- val_loss: 3.8467
Epoch 59/500
Epoch 00059: val_loss improved from 3.84612 to 3.84607, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8461
Epoch 60/500
Epoch 00060: val_loss did not improve from 3.84607
```

```
- val_loss: 3.8466
Epoch 61/500
Epoch 00061: val loss did not improve from 3.84607
- val loss: 3.8462
Epoch 62/500
Epoch 00062: val_loss did not improve from 3.84607
- val_loss: 3.8477
Epoch 63/500
Epoch 00063: val_loss did not improve from 3.84607
- val_loss: 3.8467
Epoch 64/500
Epoch 00064: val_loss did not improve from 3.84607
- val loss: 3.8472
Epoch 65/500
Epoch 00065: val_loss did not improve from 3.84607
- val_loss: 3.8465
Epoch 66/500
Epoch 00066: val_loss did not improve from 3.84607
- val_loss: 3.8462
Epoch 67/500
Epoch 00067: val loss did not improve from 3.84607
- val loss: 3.8503
Epoch 68/500
Epoch 00068: val_loss did not improve from 3.84607
- val_loss: 3.8461
Epoch 69/500
Epoch 00069: val_loss did not improve from 3.84607
- val_loss: 3.8468
Epoch 70/500
```

```
Epoch 00070: val_loss did not improve from 3.84607
- val_loss: 3.8462
Epoch 71/500
Epoch 00071: val_loss did not improve from 3.84607
- val loss: 3.8465
Epoch 72/500
Epoch 00072: val_loss did not improve from 3.84607
- val_loss: 3.8464
Epoch 73/500
Epoch 00073: val_loss did not improve from 3.84607
- val_loss: 3.8463
Epoch 74/500
Epoch 00074: val loss did not improve from 3.84607
- val_loss: 3.8465
Epoch 75/500
Epoch 00075: val_loss improved from 3.84607 to 3.84606, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8461
Epoch 76/500
Epoch 00076: val_loss did not improve from 3.84606
- val loss: 3.8486
Epoch 77/500
Epoch 00077: val_loss did not improve from 3.84606
- val_loss: 3.8472
Epoch 78/500
Epoch 00078: val_loss did not improve from 3.84606
- val_loss: 3.8462
Epoch 79/500
Epoch 00079: val_loss did not improve from 3.84606
```

```
- val_loss: 3.8472
Epoch 80/500
Epoch 00080: val loss did not improve from 3.84606
- val loss: 3.8463
Epoch 81/500
Epoch 00081: val_loss did not improve from 3.84606
- val_loss: 3.8462
Epoch 82/500
Epoch 00082: val_loss improved from 3.84606 to 3.84601, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8460
Epoch 83/500
Epoch 00083: val_loss did not improve from 3.84601
- val_loss: 3.8461
Epoch 84/500
Epoch 00084: val_loss did not improve from 3.84601
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8490
- val_loss: 3.8490
Epoch 85/500
Epoch 00085: val_loss did not improve from 3.84601
- val_loss: 3.8510
Epoch 86/500
Epoch 00086: val_loss did not improve from 3.84601
- val_loss: 3.8474
Epoch 87/500
Epoch 00087: val_loss did not improve from 3.84601
- val_loss: 3.8468
Epoch 88/500
Epoch 00088: val_loss did not improve from 3.84601
- val_loss: 3.8465
```

```
Epoch 89/500
Epoch 00089: val_loss did not improve from 3.84601
- val loss: 3.8472
Epoch 90/500
Epoch 00090: val_loss did not improve from 3.84601
- val_loss: 3.8462
Epoch 91/500
Epoch 00091: val_loss improved from 3.84601 to 3.84583, saving model to weights-
improvement-3.hdf5
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8490
- val_loss: 3.8458
Epoch 92/500
Epoch 00092: val_loss did not improve from 3.84583
- val loss: 3.8463
Epoch 93/500
Epoch 00093: val_loss improved from 3.84583 to 3.84583, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8458
Epoch 94/500
Epoch 00094: val_loss did not improve from 3.84583
- val_loss: 3.8461
Epoch 95/500
Epoch 00095: val loss did not improve from 3.84583
- val loss: 3.8473
Epoch 96/500
Epoch 00096: val_loss did not improve from 3.84583
- val_loss: 3.8461
Epoch 97/500
Epoch 00097: val_loss did not improve from 3.84583
- val_loss: 3.8464
Epoch 98/500
```

```
Epoch 00098: val_loss did not improve from 3.84583
- val loss: 3.8499
Epoch 99/500
Epoch 00099: val loss did not improve from 3.84583
- val loss: 3.8465
Epoch 100/500
Epoch 00100: val_loss improved from 3.84583 to 3.84572, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8457
Epoch 101/500
Epoch 00101: val_loss did not improve from 3.84572
- val loss: 3.8461
Epoch 102/500
Epoch 00102: val_loss did not improve from 3.84572
- val_loss: 3.8459
Epoch 103/500
Epoch 00103: val_loss did not improve from 3.84572
- val_loss: 3.8463
Epoch 104/500
Epoch 00104: val_loss improved from 3.84572 to 3.84554, saving model to weights-
improvement-3.hdf5
- val loss: 3.8455
Epoch 105/500
Epoch 00105: val_loss did not improve from 3.84554
- val_loss: 3.8469
Epoch 106/500
Epoch 00106: val_loss did not improve from 3.84554
- val_loss: 3.8468
Epoch 107/500
```

```
Epoch 00107: val_loss did not improve from 3.84554
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8488
- val_loss: 3.8473
Epoch 108/500
Epoch 00108: val_loss did not improve from 3.84554
- val_loss: 3.8460
Epoch 109/500
Epoch 00109: val_loss did not improve from 3.84554
- val_loss: 3.8463
Epoch 110/500
Epoch 00110: val_loss did not improve from 3.84554
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8487
- val_loss: 3.8463
Epoch 111/500
Epoch 00111: val_loss did not improve from 3.84554
- val_loss: 3.8460
Epoch 112/500
Epoch 00112: val_loss did not improve from 3.84554
2026816/2026816 [============= ] - 11s 5us/sample - loss: 3.8487
- val_loss: 3.8462
Epoch 113/500
Epoch 00113: val_loss did not improve from 3.84554
- val_loss: 3.8456
Epoch 114/500
Epoch 00114: val_loss did not improve from 3.84554
- val_loss: 3.8458
Epoch 115/500
Epoch 00115: val_loss did not improve from 3.84554
- val_loss: 3.8458
Epoch 116/500
Epoch 00116: val_loss did not improve from 3.84554
- val_loss: 3.8462
```

```
Epoch 117/500
Epoch 00117: val_loss did not improve from 3.84554
- val loss: 3.8467
Epoch 118/500
Epoch 00118: val_loss did not improve from 3.84554
- val_loss: 3.8476
Epoch 119/500
Epoch 00119: val_loss did not improve from 3.84554
- val_loss: 3.8532
Epoch 120/500
Epoch 00120: val_loss did not improve from 3.84554
- val loss: 3.8460
Epoch 121/500
Epoch 00121: val_loss improved from 3.84554 to 3.84547, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8455
Epoch 122/500
Epoch 00122: val_loss did not improve from 3.84547
- val_loss: 3.8456
Epoch 123/500
Epoch 00123: val_loss did not improve from 3.84547
- val loss: 3.8465
Epoch 124/500
Epoch 00124: val_loss did not improve from 3.84547
- val_loss: 3.8455
Epoch 125/500
Epoch 00125: val_loss did not improve from 3.84547
- val_loss: 3.8458
Epoch 126/500
```

```
Epoch 00126: val_loss did not improve from 3.84547
- val_loss: 3.8456
Epoch 127/500
Epoch 00127: val_loss did not improve from 3.84547
- val_loss: 3.8470
Epoch 128/500
Epoch 00128: val_loss did not improve from 3.84547
- val_loss: 3.8472
Epoch 129/500
Epoch 00129: val_loss did not improve from 3.84547
- val_loss: 3.8460
Epoch 130/500
Epoch 00130: val_loss did not improve from 3.84547
- val_loss: 3.8460
Epoch 131/500
Epoch 00131: val_loss did not improve from 3.84547
2026816/2026816 [============== ] - 11s 5us/sample - loss: 3.8485
- val_loss: 3.8465
Epoch 132/500
Epoch 00132: val_loss did not improve from 3.84547
- val_loss: 3.8464
Epoch 133/500
Epoch 00133: val_loss did not improve from 3.84547
- val_loss: 3.8488
Epoch 134/500
Epoch 00134: val_loss improved from 3.84547 to 3.84540, saving model to weights-
improvement-3.hdf5
- val_loss: 3.8454
Epoch 135/500
Epoch 00135: val_loss did not improve from 3.84540
```

```
- val_loss: 3.8457
Epoch 136/500
Epoch 00136: val_loss did not improve from 3.84540
- val_loss: 3.8461
Epoch 137/500
Epoch 00137: val_loss did not improve from 3.84540
- val_loss: 3.8470
Epoch 138/500
Epoch 00138: val_loss did not improve from 3.84540
- val_loss: 3.8465
Epoch 139/500
Epoch 00139: val_loss did not improve from 3.84540
- val loss: 3.8458
Epoch 140/500
Epoch 00140: val_loss did not improve from 3.84540
- val_loss: 3.8455
Epoch 141/500
Epoch 00141: val_loss did not improve from 3.84540
- val_loss: 3.8457
Epoch 142/500
Epoch 00142: val_loss did not improve from 3.84540
- val loss: 3.8456
Epoch 143/500
Epoch 00143: val_loss did not improve from 3.84540
- val_loss: 3.8481
Epoch 144/500
Epoch 00144: val_loss did not improve from 3.84540
- val_loss: 3.8456
Epoch 145/500
```

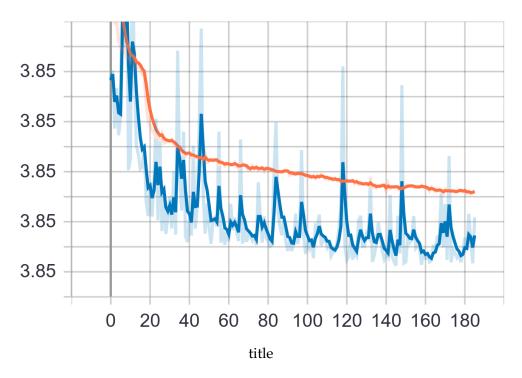
```
Epoch 00145: val_loss did not improve from 3.84540
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8484
- val_loss: 3.8454
Epoch 146/500
Epoch 00146: val_loss improved from 3.84540 to 3.84525, saving model to weights-
improvement-3.hdf5
- val loss: 3.8452
Epoch 147/500
Epoch 00147: val_loss did not improve from 3.84525
- val_loss: 3.8472
Epoch 148/500
Epoch 00148: val_loss did not improve from 3.84525
- val_loss: 3.8457
Epoch 149/500
Epoch 00149: val_loss did not improve from 3.84525
- val_loss: 3.8525
Epoch 150/500
Epoch 00150: val_loss did not improve from 3.84525
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8484
- val_loss: 3.8453
Epoch 151/500
Epoch 00151: val_loss did not improve from 3.84525
- val_loss: 3.8454
Epoch 152/500
Epoch 00152: val loss did not improve from 3.84525
- val_loss: 3.8463
Epoch 153/500
Epoch 00153: val_loss did not improve from 3.84525
- val_loss: 3.8462
Epoch 154/500
Epoch 00154: val_loss did not improve from 3.84525
```

```
- val_loss: 3.8463
Epoch 155/500
Epoch 00155: val_loss did not improve from 3.84525
- val_loss: 3.8472
Epoch 156/500
Epoch 00156: val_loss did not improve from 3.84525
- val_loss: 3.8459
Epoch 157/500
Epoch 00157: val_loss did not improve from 3.84525
- val_loss: 3.8460
Epoch 158/500
Epoch 00158: val_loss did not improve from 3.84525
- val loss: 3.8459
Epoch 159/500
Epoch 00159: val_loss did not improve from 3.84525
- val_loss: 3.8453
Epoch 160/500
Epoch 00160: val_loss did not improve from 3.84525
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8484
- val_loss: 3.8459
Epoch 161/500
Epoch 00161: val_loss did not improve from 3.84525
- val_loss: 3.8455
Epoch 162/500
Epoch 00162: val_loss did not improve from 3.84525
- val_loss: 3.8458
Epoch 163/500
Epoch 00163: val_loss did not improve from 3.84525
- val_loss: 3.8454
Epoch 164/500
```

```
Epoch 00164: val_loss did not improve from 3.84525
2026816/2026816 [=============== ] - 11s 5us/sample - loss: 3.8484
- val_loss: 3.8455
Epoch 165/500
Epoch 00165: val_loss did not improve from 3.84525
- val_loss: 3.8461
Epoch 166/500
Epoch 00166: val_loss did not improve from 3.84525
- val_loss: 3.8458
Epoch 167/500
Epoch 00167: val_loss did not improve from 3.84525
- val_loss: 3.8464
Epoch 168/500
Epoch 00168: val_loss did not improve from 3.84525
- val_loss: 3.8462
Epoch 169/500
Epoch 00169: val_loss did not improve from 3.84525
- val_loss: 3.8482
Epoch 170/500
Epoch 00170: val_loss did not improve from 3.84525
- val_loss: 3.8460
Epoch 171/500
Epoch 00171: val_loss did not improve from 3.84525
- val_loss: 3.8477
Epoch 172/500
Epoch 00172: val_loss did not improve from 3.84525
- val_loss: 3.8455
Epoch 173/500
Epoch 00173: val_loss did not improve from 3.84525
- val_loss: 3.8496
```

```
Epoch 174/500
Epoch 00174: val_loss did not improve from 3.84525
- val loss: 3.8454
Epoch 175/500
Epoch 00175: val_loss did not improve from 3.84525
- val_loss: 3.8459
Epoch 176/500
Epoch 00176: val_loss did not improve from 3.84525
- val_loss: 3.8458
Epoch 177/500
Epoch 00177: val_loss did not improve from 3.84525
- val loss: 3.8456
Epoch 178/500
Epoch 00178: val_loss did not improve from 3.84525
- val_loss: 3.8457
Epoch 179/500
Epoch 00179: val_loss did not improve from 3.84525
- val_loss: 3.8454
Epoch 180/500
Epoch 00180: val_loss did not improve from 3.84525
- val loss: 3.8458
Epoch 181/500
Epoch 00181: val_loss did not improve from 3.84525
- val_loss: 3.8464
Epoch 182/500
Epoch 00182: val_loss did not improve from 3.84525
- val_loss: 3.8459
Epoch 183/500
Epoch 00183: val_loss did not improve from 3.84525
```

```
- val_loss: 3.8473
Epoch 184/500
Epoch 00184: val_loss did not improve from 3.84525
- val loss: 3.8462
Epoch 185/500
Epoch 00185: val_loss did not improve from 3.84525
- val_loss: 3.8453
Epoch 186/500
Epoch 00186: val_loss did not improve from 3.84525
2026816/2026816 [============== ] - 11s 6us/sample - loss: 3.8483
- val_loss: 3.8472
Epoch 00186: early stopping
```



```
[169]: test_pred_ratings = test_pred_ratings.tolist()
    train_pred_ratings = train_pred_ratings.tolist()

[170]: test_pred_ratings_list = [i[0] for i in test_pred_ratings]
    train_pred_ratings_list = [i[0] for i in train_pred_ratings]

[171]: test_df_structured_target_list=test_df_structured_target.values
    train_df_structured_target_list=train_df_structured_target.values

[172]: test_df_structured_target_list = test_df_structured_target_list.tolist()
    train_df_structured_target_list = train_df_structured_target_list.tolist()

[173]: global_model_name['Fourth_NN']={
        "Test": nmae(test_pred_ratings_list,test_df_structured_target_list),
        "Train": nmae(train_pred_ratings_list,train_df_structured_target_list)
        }
        print('Result of model')
        print(global_model_name['Fourth_NN'])
```

Result of model {'Test': 0.19761617961530226, 'Train': 0.19061504214534597}

Though this model dont have 14.9 % NMAE it has very similar accuracy both in train and test similar to First NN

1.1.7 7 Results of ML models

+		-+-		+-		+
	Model	' 	Trin error	 -	Test error	
İ	Baseline	İ	0.20336702289913602		0.20505297177649737	İ
-	KnnBaseline_joke		0.1825198921435078		0.19609971106367624	
-	First_XGB	1	0.217233447214206		0.19280766965994262	1
-	SVD	1	0.17327865608074278		0.21539326049482735	1
-	${\tt Second_XGB}$	1	0.2174202221795477		0.18833252907421943	1
-	First_XGB_FE_1	1	0.22321715705636846		0.20289328398328535	
-	First_XGB_FE_2		0.22556924102141873		0.20872101921358918	
-	${\tt Second_NN}$		0.18082728958955502		0.1493320631675459	
-	First_NN	1	0.19275586777667447		0.19394950246246248	1

As we can see the Second_NN model is the best model with NMAE as 14.9 percentage.