

```

from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import Nadam

from sklearn.model_selection import train_test_split
import pandas as pd
from sklearn.neighbors import KNeighborsClassifier
import numpy as np
import seaborn as sns
from sklearn.metrics import confusion_matrix, accuracy_score
import matplotlib.pyplot as plt

from google.colab import files

uploaded = files.upload()

<IPython.core.display.HTML object>

Saving Simulink_ДПТ_10B.csv to Simulink_ДПТ_10B.csv
df = pd.read_csv('Simulink_ДПТ_10B.csv', delimiter=';')

/usr/local/lib/python3.8/dist-packages/IPython/core/
interactiveshell.py:3326: DtypeWarning: Columns
(1,2,3,4,5,6,7,8,9,10,11,12,13,14,15) have mixed types.Specify dtype
option on import or set low_memory=False.
  exec(code_obj, self.user_global_ns, self.user_ns)

df

```

	T	Удс(вход2)	Unnamed: 2 \
0	NaN	U1	U2_diff
1	0.000000e+00	0	0
2	1.169967e-66	1.16996669923822e-66	3.15544362088405e-30
3	1.843200e-21	1.843200000000000e-21	1.600000000000000e-10
4	3.981307e-19	3.98130696143209e-19	9.600000000000000e-10
...	...	...	...
1000028	9.999999e+00	9.999999	0.999997
1000029	9.999999e+00	9.999999	0.999998
1000030	9.999999e+00	9.999999	0.999999
1000031	9.999999e+00	9.999999	1.0
1000032	9.999999e+00	9.999999	1.0

	Unnamed: 3	Удт(вход3)
Unnamed: 5 \		
0	U3_int	U4
U5_diff		
1	0	0
0		
2	-1.14702617601537e-70	-1.14702617601537e-70
2.36658271566097e-22		
3	5.81611368201348e-51	5.81611368201348e-51

0.0119999934720000		
4	4.93977239027873e-29	4.93977239027873e-29
0.0839997192963500		
...	...	...
...		
1000028	9.995423	9.995423
0.000002		
1000029	9.995433	9.995433
0.000002		
1000030	9.995443	9.995443
0.000002		
1000031	9.995453	9.995453
0.000002		
1000032	9.995456	9.995456
0.000002		

	Unnamed: 6	Урт (выход)	Unnamed: 8
\			
0	U6_int	U7	U8_ditt
1	0	0	0
2	2.36658271566097e-22	6.49981499390246e-68	0
3	0.0119999934720000	1.024000000000000e-22	198.988003852207
4	0.0839997192963500	2.21183720079561e-20	198.915999227722
...	...	...	...
1000028	0.000002	0.555556	-0.000001
1000029	0.000002	0.555556	-0.000001
1000030	0.000002	0.555556	-0.000001
1000031	0.000002	0.555556	-0.000001
1000032	0.000002	0.555556	-0.0

	Unnamed: 9	Иобщ	Unnamed:
11 \			
0	U9_int	U10	
U11_diff			
1	0	0	
0			
2	3.15544362088405e-29	3.15544362088405e-29	
1.49352366668537e-51			

3	1.60000000254710e-09	1.60000000254710e-09	
3.83999791104000e-12			
4	9.60000009167708e-09	9.60000009167708e-09	
1.38239548785200e-10			
...	...	...	
...			
1000028	6.815096	6.815096	-
0.0			
1000029	6.815103	6.815103	-
0.0			
1000030	6.815109	6.815109	-
0.0			
1000031	6.815116	6.815116	-
0.0			
1000032	6.815118	6.815118	-
0.0			
	Unnamed: 12	wn	Unnamed:
14 \			
0	U12_int	U13	
U14_diff			
1	0	0	
0			
2	4.73316543132195e-22	1.29996299878049e-67	
3.89988899746073e-65			
3	0.0239999869440000	2.04800000000000e-22	
6.14400000000000e-20			
4	0.167999438592700	4.42367440159121e-20	
1.32710232047736e-17			
...	...	...	
...			
1000028	0.000004	1.111111	
333.333314			
1000029	0.000004	1.111111	
333.333314			
1000030	0.000004	1.111111	
333.333314			
1000031	0.000004	1.111111	
333.333314			
1000032	0.000004	1.111111	
333.333314			
	Unnamed: 15		
0	U15_int		
1	0		
2	1.23592415711364e-35		
3	3.84000000000000e-10		
4	1.65119790059671e-08		
...	...		
1000028	-0.000053		

```

1000029      -0.000053
1000030      -0.000053
1000031      -0.000053
1000032      -0.000053

```

```
[1000033 rows x 16 columns]
```

```
df=df.drop(index=0)
```

```
df
```

	T	Удс(вход2)	Unnamed: 2 \
1	0.000000e+00	0	0
2	1.169967e-66	1.16996669923822e-66	3.15544362088405e-30
3	1.843200e-21	1.843200000000000e-21	1.600000000000000e-10
4	3.981307e-19	3.98130696143209e-19	9.600000000000000e-10
5	5.491037e-17	5.49103709210333e-17	4.960000000000000e-09
...	...	...	...
1000028	9.999999e+00	9.999999	0.999997
1000029	9.999999e+00	9.999999	0.999998
1000030	9.999999e+00	9.999999	0.999999
1000031	9.999999e+00	9.999999	1.0
1000032	9.999999e+00	9.999999	1.0

	Unnamed: 3	Удт(вход3)
Unnamed: 5 \		
1	0	0
0		
2	-1.14702617601537e-70	-1.14702617601537e-70
2.36658271566097e-22		
3	5.81611368201348e-51	5.81611368201348e-51
0.0119999934720000		
4	4.93977239027873e-29	4.93977239027873e-29
0.0839997192963500		
5	3.92430326568353e-26	3.92430326568353e-26
0.443992277431536		
...	...	...
...		
1000028	9.995423	9.995423
0.000002		
1000029	9.995433	9.995433
0.000002		
1000030	9.995443	9.995443
0.000002		
1000031	9.995453	9.995453
0.000002		
1000032	9.995456	9.995456
0.000002		

```
Unnamed: 6
```

```
Урт(выход)
```

```
Unnamed: 8
```

\			
1	0	0	0
2	2.36658271566097e-22	6.49981499390246e-68	0
3	0.0119999934720000	1.02400000000000e-22	198.988003852207
4	0.0839997192963500	2.21183720079561e-20	198.915999227722
5	0.443992277431536	3.05057616227963e-18	198.556007635631
...	...	...	...
1000028	0.000002	0.555556	-0.000001
1000029	0.000002	0.555556	-0.000001
1000030	0.000002	0.555556	-0.000001
1000031	0.000002	0.555556	-0.000001
1000032	0.000002	0.555556	-0.0

	Unnamed: 9	Иобщ	Unnamed:
11 \			
1	0	0	
0			
2	3.15544362088405e-29	3.15544362088405e-29	
1.49352366668537e-51			
3	1.60000000254710e-09	1.60000000254710e-09	
3.83999791104000e-12			
4	9.60000009167708e-09	9.60000009167708e-09	
1.38239548785200e-10			
5	4.96000024448086e-08	4.96000024448086e-08	
3.69017776823749e-09			
...	...	...	
...			
1000028	6.815096	6.815096	-
0.0			
1000029	6.815103	6.815103	-
0.0			
1000030	6.815109	6.815109	-
0.0			
1000031	6.815116	6.815116	-
0.0			
1000032	6.815118	6.815118	-
0.0			

	Unnamed: 12	wn	Unnamed:
14 \			
1	0	0	
0			
2	4.73316543132195e-22	1.29996299878049e-67	
3	3.89988899746073e-65		
3	0.0239999869440000	2.04800000000000e-22	
6	1.14400000000000e-20		
4	0.167999438592700	4.42367440159121e-20	
1	3.2710232047736e-17		
5	0.887984554863073	6.10115232455925e-18	
1	8.3034569736778e-15		
...	...	...	
1000028	0.000004	1.111111	
333.333314			
1000029	0.000004	1.111111	
333.333314			
1000030	0.000004	1.111111	
333.333314			
1000031	0.000004	1.111111	
333.333314			
1000032	0.000004	1.111111	
333.333314			

	Unnamed: 15
1	0
2	1.23592415711364e-35
3	3.84000000000000e-10
4	1.65119790059671e-08
5	4.54268668540751e-07
...	...
1000028	-0.000053
1000029	-0.000053
1000030	-0.000053
1000031	-0.000053
1000032	-0.000053

[1000032 rows x 16 columns]

```
from sklearn.metrics import max_error
```

```
from sklearn.preprocessing import StandardScaler
```

```
df1=df[['T','Unnamed: 9', 'Io6u', 'Unnamed: 11', 'Unnamed: 12',
        'wn', 'Unnamed: 14', 'Unnamed: 15']]
```

```
scaler = StandardScaler().fit(df1)
```

```
df1=scaler.transform(df1)
```

```

df2=df[['Удс(вход2)', 'Unnamed: 2', 'Unnamed: 3',
        'Удт(вход3)', 'Unnamed: 5', 'Unnamed: 6', 'Урт(выход)']]

scaler = StandardScaler().fit(df2)

df2=scaler.transform(df2)

from sklearn.model_selection import train_test_split
X_train1 ,X_test1,y_train1,y_test1=train_test_split(
df1,
df2,
test_size=0.20
)

model = Sequential([
    Dense(13,activation='sigmoid', input_shape=[8]),
    Dense(11, activation='sigmoid', input_shape=[8]),
    Dense(9, activation='sigmoid', input_shape=[8]),
    Dense(7, activation='sigmoid', input_shape=[8]),
    ])
model.compile(loss='mean_squared_error', optimizer=Nadam(lr=0.0001),
metrics=['mean_absolute_error'])

X_train1 = np.asarray(X_train1).astype(np.float32)

y_train1 = np.asarray(y_train1).astype(np.float32)

X_test1 = np.asarray(X_test1).astype(np.float32)

history=model.fit(X_train1, y_train1, batch_size=200, epochs=20)

res = model.predict(X_test1, verbose = 0)
res = np.array(res)
detres = res * scaler.scale_[-1] + scaler.mean_[-1]
detsource = y_test1 * scaler.scale_[-1] + scaler.mean_[-1]

print(detres)
print(model.layers[0].get_weights())
plt.plot(history.history['loss'])
plt.grid(True)
plt.show()

Epoch 1/20

/usr/local/lib/python3.8/dist-packages/keras/optimizers/optimizer_v2/
nadam.py:78: UserWarning: The `lr` argument is deprecated, use
`learning_rate` instead.
  super(Nadam, self).__init__(name, **kwargs)

4001/4001 [=====] - 6s 1ms/step - loss:
1.1012 - mean_absolute_error: 0.6454

```

Epoch 2/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.9634 - mean\_absolute\_error: 0.5198  
Epoch 3/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.8443 - mean\_absolute\_error: 0.4197  
Epoch 4/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7849 - mean\_absolute\_error: 0.3743  
Epoch 5/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7599 - mean\_absolute\_error: 0.3481  
Epoch 6/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7509 - mean\_absolute\_error: 0.3369  
Epoch 7/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7464 - mean\_absolute\_error: 0.3293  
Epoch 8/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7442 - mean\_absolute\_error: 0.3257  
Epoch 9/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7432 - mean\_absolute\_error: 0.3249  
Epoch 10/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7427 - mean\_absolute\_error: 0.3245  
Epoch 11/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7425 - mean\_absolute\_error: 0.3242  
Epoch 12/20  
4001/4001 [=====] - 6s 1ms/step - loss:  
0.7423 - mean\_absolute\_error: 0.3239  
Epoch 13/20  
4001/4001 [=====] - 6s 1ms/step - loss:  
0.7423 - mean\_absolute\_error: 0.3237  
Epoch 14/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7422 - mean\_absolute\_error: 0.3234  
Epoch 15/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7422 - mean\_absolute\_error: 0.3232  
Epoch 16/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7422 - mean\_absolute\_error: 0.3229  
Epoch 17/20  
4001/4001 [=====] - 5s 1ms/step - loss:  
0.7422 - mean\_absolute\_error: 0.3226  
Epoch 18/20  
4001/4001 [=====] - 5s 1ms/step - loss:

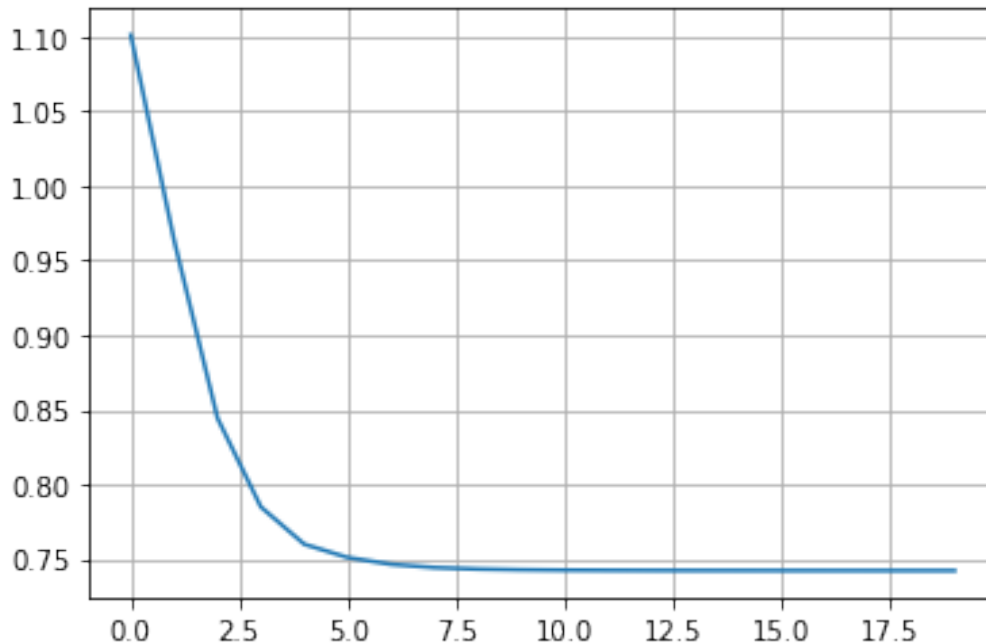


```
0.7422 - mean_absolute_error: 0.3223
Epoch 19/20
4001/4001 [=====] - 5s 1ms/step - loss:
0.7421 - mean_absolute_error: 0.3221
Epoch 20/20
4001/4001 [=====] - 5s 1ms/step - loss:
0.7421 - mean_absolute_error: 0.3220
[[0.55758274 0.55530196 0.55530196 ... 0.55530196 0.55530196
0.55747634]
 [0.5553021 0.6297774 0.6297726 ... 0.55530214 0.5553025 0.5553021
]
 [0.5752796 0.55530196 0.55530196 ... 0.55530196 0.55530196
0.57527846]
...
 [0.555302 0.6570569 0.65700257 ... 0.555305 0.55530685 0.555302
]
 [0.555302 0.653774 0.65354085 ... 0.55530256 0.5553034 0.555302
]
 [0.555302 0.6572921 0.6572858 ... 0.55531836 0.5553198
0.55530196]]
[array([[-0.38773367, -0.19059043, -2.8898568 , 0.2512025 , -
0.6072035 ,
0.91008675, -1.3847708 , 1.2638124 , 0.24831836, -
0.65324146,
-0.3785339 , -1.4666837 , 0.0652221 ],
[ 1.6788678 , 1.5311464 , 1.0213298 , -1.1211855 , -
0.13667743,
-0.8148778 , -1.276097 , -0.5602227 , -1.1077905 ,
0.12833633,
1.2376262 , -1.3899841 , 0.83659065],
[ 1.6719214 , 1.447256 , 0.98517746, -1.0694246 , -
0.07866496,
-0.3462141 , -0.97524697, -0.8573471 , -0.8313965 , -
0.13685569,
1.4421283 , -1.6009767 , 1.3436425 ],
[-0.20821501, -0.20049605, 0.4791772 , -0.17485228, -
0.06406721,
-0.4185824 , -0.54062533, -0.75162685, -0.81762785,
0.10748944,
0.20270288, -0.1469431 , -0.2510594 ],
[-0.05586456, -0.04715582, 0.03088687, -0.08127035, -
0.03251934,
-0.15885721, -0.02782091, -0.32128105, -0.10629448, -
0.03724622,
-0.08350988, -0.15395367, -0.10278604],
[-0.02215724, 0.03952196, -2.7825553 , -0.1335898 , -
0.81433785,
0.2394147 , -0.86800057, 1.2414325 , 0.36269435, -0.6360786
,
-0.07879678, -1.7482498 , -0.45204934],
```

```

        [-0.12330174, -0.22578602, -3.415827 , -0.12134934, -0.9423826
,
        0.5538462 , -1.487874 , 1.3873376 , -0.40463126, -
0.26731324,
        -0.10876487, -2.3290699 , -0.27660993],
        [-0.23315041, 0.10517702, 1.0526392 , -0.71298105, -
0.46748102,
        -0.45379454, 0.34570205, -0.0255143 , 0.08112771,
0.12865318,
        -0.6555678 , -0.23296534, -0.10000278]], dtype=float32),
array([ 0.0728178 , -0.22849657, -0.5368915 , 1.2274625 ,
1.0987822 ,
        0.06047882, 1.3142412 , 0.41828984, 1.014615 ,
0.78299713,
        -0.00743077, 2.430956 , 0.06408177]), dtype=float32)]

```



```

res = model.predict(X_test1, verbose = 0)
res = np.array(res)
detres = res * scaler.scale_[-1] + scaler.mean_[-1]
detsource = y_test1 * scaler.scale_[-1] + scaler.mean_[-1]

x=[]

from numpy import arange
for i in arange(-10.0, 10.0, 0.1):
    x.append(i)

myarray = np.asarray(x)
v=myarray.reshape(-1,8)

```

```
detres.shape
```

```
(200007, 7)
```

```
X_test1.shape
```

```
(200007, 8)
```

```
sss=model.predict(v)
```

```
y_test1[:25].shape
```

```
1/1 [=====] - 0s 14ms/step
```

```
(25, 7)
```

```
from sklearn import metrics
```

```
print("Максимальная ошибка на тестовом наборе:
```

```
{:.2f}".format(metrics.mean_absolute_error(sss,detsource[:25])))
```

```
Максимальная ошибка на тестовом наборе: 0.53
```

```
y_test1
```

```
array([[ 2.75967146e-02, -5.09462352e-01, -4.84004402e-01, ...,
         2.24690394e-03,  2.24690394e-03,  2.75967146e-02],
       [ 2.50237898e-03,  7.29532961e-01,  7.29013591e-01, ...,
         4.49731304e-05,  4.49731304e-05,  2.50237898e-03],
       [ 1.91009093e-01, -8.55276983e-01, -8.27392991e-01, ...,
         1.64429316e-02,  1.64429316e-02,  1.91009093e-01],
       ...,
       [ 2.48811447e-03,  1.03857339e+00,  1.03138294e+00, ...,
         4.36791216e-05,  4.36791216e-05,  2.48811447e-03],
       [ 2.48970184e-03,  9.60809269e-01,  9.55297526e-01, ...,
         4.38238854e-05,  4.38238854e-05,  2.48970184e-03],
       [ 2.48679346e-03,  1.15180924e+00,  1.14217431e+00, ...,
         4.35582774e-05,  4.35582774e-05,  2.48679346e-03]])
```

```
sss
```

```
array([[ -3.4593062 , -1.6511676 , -1.557725  , -1.5189536 ,
        -17.639482  , -17.36426  , -3.4362965 ],
       [ -3.572784  , -1.6854769 , -1.5922087 , -1.5532039 ,
        -17.093292  , -16.825266 , -3.5504932 ],
       [ -3.700366  , -1.7234143 , -1.6277994 , -1.5886106 ,
        -16.422525  , -16.163864 , -3.6789112 ],
       [ -3.8316143 , -1.7640986 , -1.6632187 , -1.6240796 ,
        -15.612964  , -15.366903 , -3.8112826 ],
       [ -3.9462328 , -1.8052561 , -1.6961787 , -1.657631  ,
        -14.656165  , -14.427353 , -3.9276018 ],
       [ -4.0115204 , -1.8423026 , -1.7227623 , -1.6857778 ,
        -13.548343  , -13.343042 , -3.995555  ],
       [ -3.9828923 , -1.8669822 , -1.7362865 , -1.7022777 ,
        -12.2810545 , -12.106896 , -3.9708586 ],
```

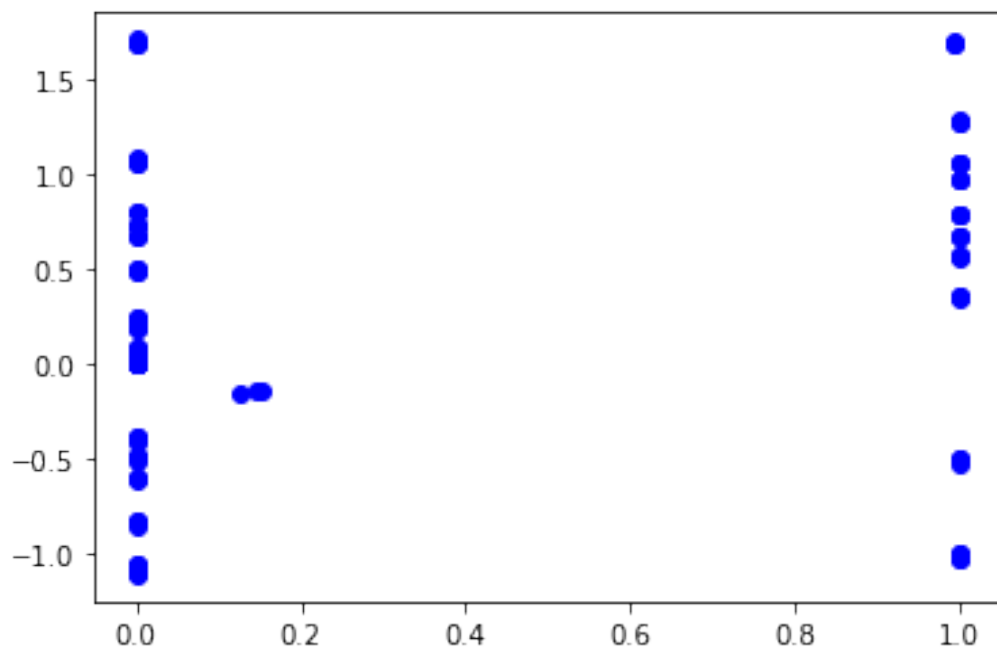
```
[ -3.8110592 , -1.8650113 , -1.7250017 , -1.6955305 ,
  -10.820709 , -10.685295 , -3.803937 ],
[ -3.4571664 , -1.8112957 , -1.6676459 , -1.6436409 ,
  -9.088382 , -8.995593 , -3.4542594 ],
[ -2.905838 , -1.6612338 , -1.5259517 , -1.5067674 ,
  -6.9812875 , -6.925237 , -2.9032147 ],
[ -2.1545892 , -1.3465536 , -1.2390392 , -1.2231706 ,
  -4.489319 , -4.4531436 , -2.1468353 ],
[ -1.1974071 , -0.817021 , -0.7536267 , -0.7432533 ,
  -1.8859826 , -1.859961 , -1.1872523 ],
[ -0.06935954, -0.16137516, -0.14235628, -0.14493406,
  0.22121954, 0.2224834 , -0.07257009],
[ 1.0983648 , 0.3739648 , 0.34805143, 0.32998514,
  1.2844368 , 1.2523206 , 1.072848 ],
[ 2.0987089 , 0.66050637, 0.5783168 , 0.54886794,
  1.5133954 , 1.4583608 , 2.0557013 ],
[ 2.7692637 , 0.77200997, 0.64254844, 0.60504746,
  1.7117981 , 1.6456212 , 2.7137666 ],
[ 3.129013 , 0.7974435 , 0.6506351 , 0.6066923 ,
  2.2634864 , 2.194923 , 3.06314 ],
[ 3.3016186 , 0.77741396, 0.63687694, 0.58778524,
  3.0299835 , 2.9644618 , 3.227045 ],
[ 3.3890758 , 0.73023283, 0.6059027 , 0.55293226,
  3.8126593 , 3.7527528 , 3.307662 ],
[ 3.4443917 , 0.66725886, 0.56078804, 0.50500655,
  4.5213366 , 4.4678283 , 3.357799 ],
[ 3.488679 , 0.5959858 , 0.50589824, 0.4480989 ,
  5.140953 , 5.093733 , 3.3981752 ],
[ 3.5281005 , 0.5211717 , 0.4453684 , 0.3861133 ,
  5.6844015 , 5.6429887 , 3.434599 ],
[ 3.5631166 , 0.44579136, 0.38241816, 0.32210684,
  6.169419 , 6.1332006 , 3.46727 ],
[ 3.592548 , 0.37170565, 0.31932497, 0.25825155,
  6.610944 , 6.579281 , 3.4948397 ],
[ 3.6151366 , 0.30006564, 0.2575997 , 0.19599378,
  7.019755 , 6.992018 , 3.5159435 ]], dtype=float32)
```

```
import matplotlib.cm as cm
import matplotlib.pyplot as plt
import matplotlib.colors as mcolors
import numpy as np

cmap, norm = mcolors.from_levels_and_colors([0, 2, 5, 6], ['red',
'green', 'blue'])

plt.scatter(sss,y_test1[:25],c='blue')

<matplotlib.collections.PathCollection at 0x7f6d65220a90>
```



```
y_train1.shape  
(800025, 7)  
X_test1.dtype  
dtype('float32')
```