

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")

from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
```

```
df=pd.read_csv("MiningProcess_Flotation_Plant_Database.csv")
df
```

	date	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	...	Flotation Column 07 Air Flow	Flotation Column 01 Level
0	2017-03-10 01:00:00	55,2	16,98	3019,53	557,434	395,713	10,0664	1,74	249,214	253,235	...	250,884	457,396
1	2017-03-10 01:00:00	55,2	16,98	3024,41	563,965	397,383	10,0672	1,74	249,719	250,532	...	248,994	451,891
2	2017-03-10 01:00:00	55,2	16,98	3043,46	568,054	399,668	10,068	1,74	249,741	247,874	...	248,071	451,24
3	2017-03-10 01:00:00	55,2	16,98	3047,36	568,665	397,939	10,0689	1,74	249,917	254,487	...	251,147	452,441
4	2017-03-10 01:00:00	55,2	16,98	3033,69	558,167	400,254	10,0697	1,74	250,203	252,136	...	248,928	452,441
...	...	...	...	...	...	...	...	...	...	...	...	...	...
737448	2017-09-09 23:00:00	49,75	23,2	2710,94	441,052	386,57	9,62129	1,65365	302,344	298,786	...	313,695	392,16
737449	2017-09-09 23:00:00	49,75	23,2	2692,01	473,436	384,939	9,62063	1,65352	303,013	301,879	...	236,7	401,505
737450	2017-09-09 23:00:00	49,75	23,2	2692,2	500,488	383,496	9,61874	1,65338	303,662	307,397	...	225,879	408,899
737451	2017-09-09 23:00:00	49,75	23,2	1164,12	491,548	384,976	9,61686	1,65324	302,55	301,959	...	308,115	405,107
737452	2017-09-09 23:00:00	49,75	23,2	1164,12	468,019	384,801	9,61497	1,6531	300,355	292,865	...	308,115	413,754

737453 rows × 24 columns



```
df=df.drop(columns=['date'])

df.head()
```

	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	Flotation Column 03 Air Flow	...	Flotation Column 07 Air Flow	Flotation Column 01 Level	Flotat
0	55,2	16,98	3019,53	557,434	395,713	10,0664	1,74	249,214	253,235	250,576	...	250,884	457,396	43
1	55,2	16,98	3024,41	563,965	397,383	10,0672	1,74	249,719	250,532	250,862	...	248,994	451,891	4
2	55,2	16,98	3043,46	568,054	399,668	10,068	1,74	249,741	247,874	250,313	...	248,071	451,24	46

```
def convert(x):
    return float(x.replace(',','.'))

# rows x 23 columns
for i in df.columns:
    df[i]=df[i].apply(lambda x: convert(x))

df.head()
```

	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	Flotation Column 03 Air Flow	...	Flotation Column 07 Air Flow	Flotation Column 01 Level	Flotat
0	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	250.576	...	250.884	457.396	43
1	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	250.862	...	248.994	451.891	42
2	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	250.313	...	248.071	451.240	46
3	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	250.049	...	251.147	452.441	45
4	55.2	16.98	3033.69	558.167	400.254	10.0697	1.74	250.203	252.136	249.895	...	248.928	452.441	45

5 rows x 23 columns



```
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 737453 entries, 0 to 737452
Data columns (total 23 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   % Iron Feed                               737453 non-null float64
1   % Silica Feed                             737453 non-null float64
2   Starch Flow                               737453 non-null float64
3   Amina Flow                                737453 non-null float64
4   Ore Pulp Flow                             737453 non-null float64
5   Ore Pulp pH                               737453 non-null float64
6   Ore Pulp Density                           737453 non-null float64
7   Flotation Column 01 Air Flow               737453 non-null float64
8   Flotation Column 02 Air Flow               737453 non-null float64
9   Flotation Column 03 Air Flow               737453 non-null float64
10  Flotation Column 04 Air Flow               737453 non-null float64
11  Flotation Column 05 Air Flow               737453 non-null float64
12  Flotation Column 06 Air Flow               737453 non-null float64
13  Flotation Column 07 Air Flow               737453 non-null float64
14  Flotation Column 01 Level                   737453 non-null float64
15  Flotation Column 02 Level                   737453 non-null float64
16  Flotation Column 03 Level                   737453 non-null float64
17  Flotation Column 04 Level                   737453 non-null float64
18  Flotation Column 05 Level                   737453 non-null float64
19  Flotation Column 06 Level                   737453 non-null float64
20  Flotation Column 07 Level                   737453 non-null float64
21  % Iron Concentrate                         737453 non-null float64
22  % Silica Concentrate                       737453 non-null float64
dtypes: float64(23)
memory usage: 129.4 MB

df.head()
```

	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	Flotation Column 03 Air Flow	...	Flotation Column 07 Air Flow	Flotation Column 01 Level	Flot:
0	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	250.576	...	250.884	457.396	43
1	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	250.862	...	248.994	451.891	42
2	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	250.313	...	248.071	451.240	46
3	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	250.049	...	251.147	452.441	45

```
x=df.iloc[:, :-1]
```

```
y=df['% Silica Concentrate']
```

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

```
sc=StandardScaler()
```

```
x=sc.fit_transform(x)
```

```
from sklearn.model_selection import train_test_split
```

```
x_train,x_test,y_train,y_test =train_test_split(x,y,test_size=0.3,random_state=1)
```

```
import tensorflow as tf
```

```
from tensorflow.keras.callbacks import EarlyStopping
```

```
from tensorflow.keras import Sequential
```

```
from tensorflow.keras.layers import Dense
```

```
from tensorflow.keras.layers import Dropout
```

```
from sklearn.metrics import r2_score
```

```
es=EarlyStopping(monitor='val_loss',mode='min',verbose=1,patience=75,min_delta=0.2)
```

```
ann=Sequential()
```

```
ann.add(Dense(units=280,activation='relu'))
```

```
ann.add(Dense(units=1))
```

```
ann.compile(optimizer='adam',loss='mse',metrics=['accuracy'])
```

```
ann.fit(x_train,y_train,epochs=256,verbose=1,validation_data=(x_test,y_test),batch_size=1000,callbacks=[es])
```

```
loss=ann.history.history
```

```
loss_df=pd.DataFrame(loss)
```

```
plt.figure(figsize=(15,10))
```

```
plt.plot(loss_df['loss'])
```

```
plt.grid()
```

```
plt.show()
```

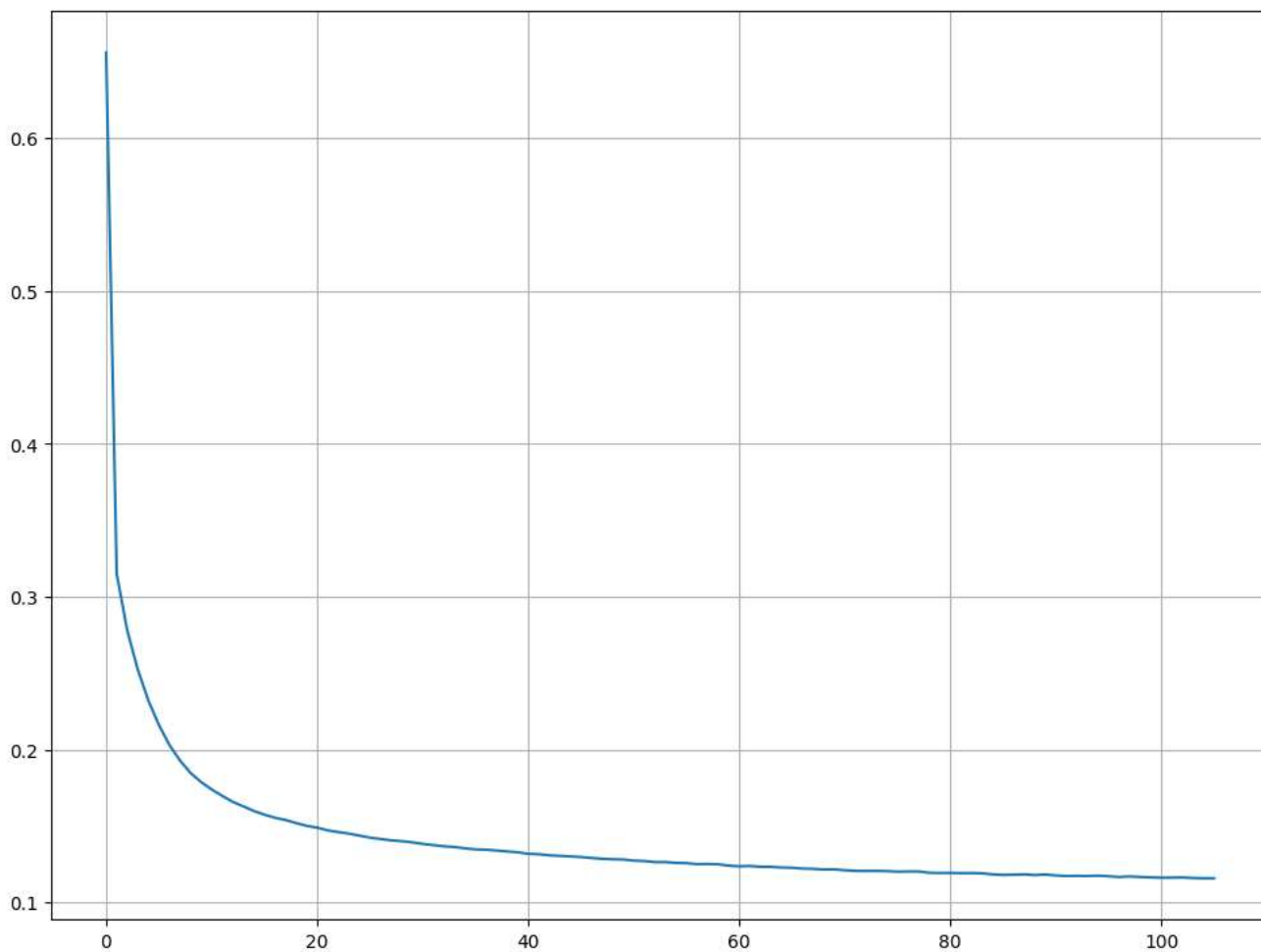
```
y_pred_test=ann.predict(x_test)
```

```
print("\nAccuracy:\n")
```

```
print(r2_score(y_test,y_pred_test))
```



```
517/517 [=====] - 4s 8ms/step - loss: 0.1162 - accuracy: 0.0036 - val_loss: 0.1236 - val_accuracy: 0.0037
Epoch 103/256
517/517 [=====] - 5s 10ms/step - loss: 0.1163 - accuracy: 0.0036 - val_loss: 0.1270 - val_accuracy: 0.0037
Epoch 104/256
517/517 [=====] - 3s 6ms/step - loss: 0.1160 - accuracy: 0.0036 - val_loss: 0.1200 - val_accuracy: 0.0037
Epoch 105/256
517/517 [=====] - 3s 6ms/step - loss: 0.1158 - accuracy: 0.0036 - val_loss: 0.1198 - val_accuracy: 0.0037
Epoch 106/256
517/517 [=====] - 5s 10ms/step - loss: 0.1158 - accuracy: 0.0036 - val_loss: 0.1202 - val_accuracy: 0.0037
Epoch 106: early stopping
```



```
6914/6914 [=====] - 15s 2ms/step
```

Accuracy:

0.9053582588311376