

Digit Recognition Using Multi Layered Perceptrons

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Aim: Digit recognition using MLPClassifier.

Packages and Libraries Used:

1. Pandas
2. Numpy
3. Matplotlib
4. Sklearn(scikit learn)

Importing the packages:

```
In [1]: import pandas as pd
import numpy as np
import random
from sklearn.neural_network import MLPClassifier
import matplotlib.pyplot as plt
from sklearn.metrics import accuracy_score
from sklearn.model_selection import RandomizedSearchCV
```

Importing the Dataset:

```
In [2]: X_train=pd.read_csv("Data/image_data_training_without_label.csv")
X_train.set_index(X_train["id"])
X_test=pd.read_csv("Data/image_data_test.csv")
X_test.set_index(X_test["id"])
Y_train=pd.read_csv("Data/training_label.csv")
Y_train.set_index(Y_train["id"])
Y_train=Y_train.drop("id",axis=1)
Y_test=pd.read_csv("Data/test_label.csv")
Y_test.set_index(Y_test["id"])
Y_test=Y_test.drop("id",axis=1)
Y_test=np.array(Y_test)
Y_test=Y_test.ravel()
#Since X_train and Y_train are too large, lets try reducing their size to reduce training time
l=[]

for i in range(3000):
    l+=random.randint(0,60010)
Xtf=pd.DataFrame(columns=X_train.columns)
Ytf=pd.DataFrame(columns=Y_train.columns)
for i in range(3000):
    Xtf.loc[i]=X_train.iloc[l[i]]
    Ytf.loc[i]=Y_train.iloc[l[i]]
Xtf=Xtf.drop("id",axis=1)
```

Now, lets look into the data:

```
In [3]: X_train.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 60010 entries, 0 to 60009
Columns: 786 entries, id to 784
dtypes: int64(786)
memory usage: 359.9 MB
```

Since X_train has 786 columns, with the first column being labeled as 'id', we can convert the DataFrame into a Numpy array. By excluding the 'id' and '784' columns, we can reshape the array into 28x28 so that we can display examples as an image using Matplotlib.

```
In [4]: show=pd.DataFrame(columns=X_train.columns)
show.loc[0]=X_train.loc[3000] #Random index for visualization
show.loc[1]=X_train.loc[15000] #Random index for visualization
show.loc[2]=X_train.loc[58000] #Random index for visualization
show=show.drop("id",axis=1)
show=show.drop("784",axis=1)
show=np.array(show).astype("float")
plt.figure(figsize=(6,6))
for i in range(0,3):
    plt.subplot(1,3,i+1)
    plt.imshow(show[i].reshape(28,28))
    plt.axis("off")
```



Now lets optimize the parameters of MLP Classifier using RandomizedSearchCV:

```
In [5]: parameters = {'hidden_layer_sizes':[64,128,256], 'alpha':[0.01,0.001,0.0001], 'max_iter': [200, 500, 800], 'learning_rate_init':[0.0001, 0.001, 0.01, 0.1] }
model = MLPClassifier()
abc=RandomizedSearchCV(estimator=model,param_distributions=parameters,cv=5)
#To remove warnings:
Ytf=np.array(Ytf)
Ytf=Ytf.ravel()

abc.fit(Xtf,Ytf)
print(f"Best parameters {abc.best_params_}")

best_model=abc.best_estimator_#Selects the best model
```

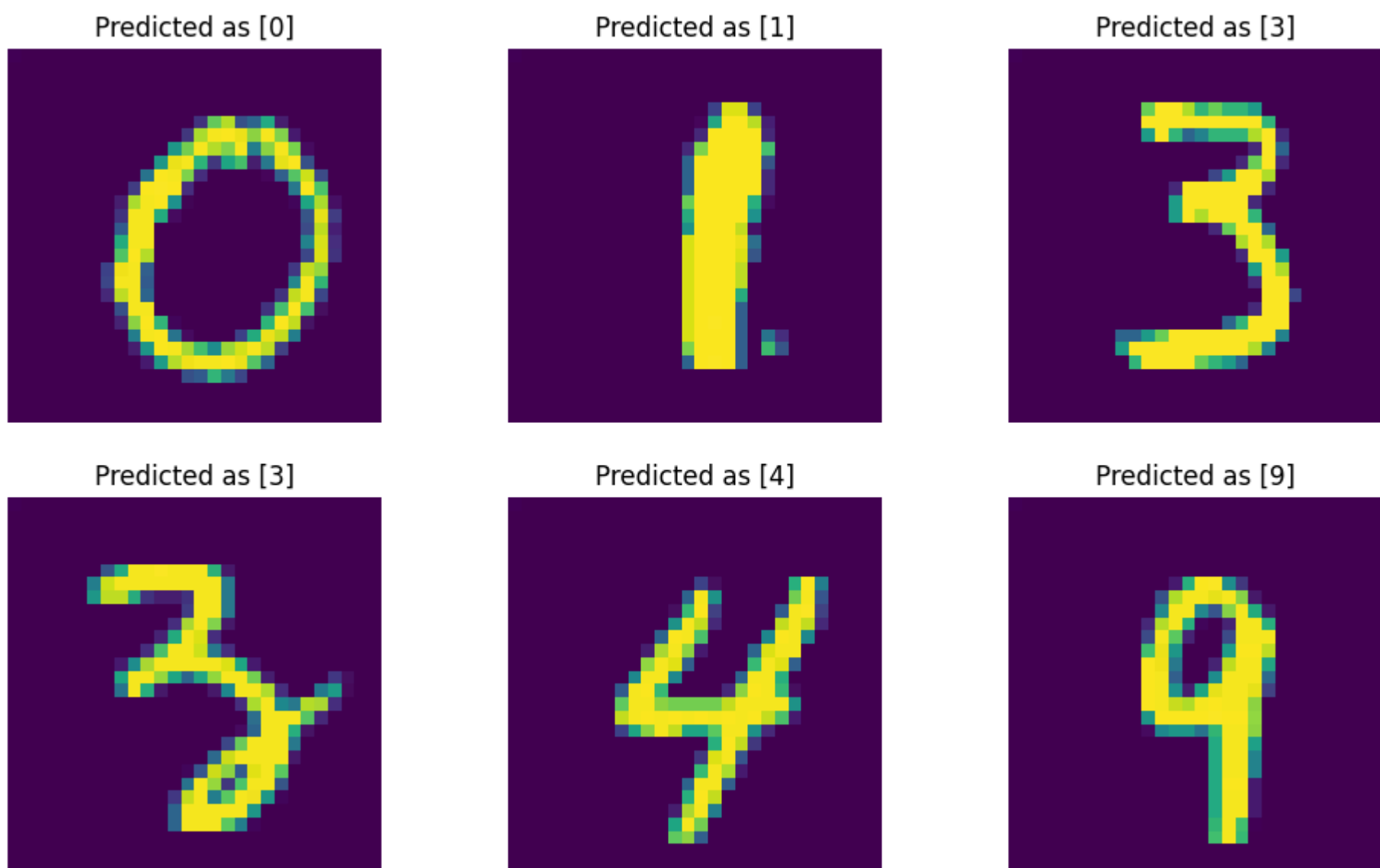
Best parameters {'max_iter': 200, 'learning_rate_init': 0.001, 'hidden_layer_sizes': 256, 'alpha': 0.001}

The cross validation training was carried out with the help of RandomizedSearchCV from sklearn.model_selection . Now lets evaluate the performance of the model with the help of performance metrics. Here we are using accuracy_score from sklearn.metrics .

```
In [6]: X_test.set_index("id")
X_test=X_test.drop("id",axis=1)
Y_pred=best_model.predict(X_test)
print(f"The accuracy score of the Best Model is : {accuracy_score(Y_pred, Y_test)}")
```

The accuracy score of the Best Model is : 0.893

```
In [7]: plt.figure(figsize=(12, 7))
plt.subplot(2, 3, 1)
sample_indices = [3, 2000, 3000, 4003, 4600, 9322]
for i, idx in enumerate(sample_indices, start=1):
    sample = pd.DataFrame(columns=X_test.columns)
    sample.loc[0] = X_test.loc[idx]
    pred = best_model.predict(sample)
    plt.subplot(2, 3, i)
    sample=sample.drop("784",axis=1)
    plt.imshow(np.array(sample).reshape(28, 28))
    plt.title(f"Predicted as {pred}")
    plt.axis("off")
plt.show()
```



About me

Hello! I'm Durvank Gade, a passionate CSE AI ML undergraduate student at VIT Chennai. With a background in Data Science, I thrive on solving complex problems and creating innovative solutions.

I'm deeply interested in Data Science, especially Machine Learning. Whether it's diving into machine learning algorithms or exploring data science, I'm always eager to learn and grow. My curiosity drives me to stay up-to-date with the latest trends and technologies.

I'm always open to new opportunities, discussions and collaborations. Feel free to connect via:

