

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

Prepare/collect data

```
In [2]: import os

path = os.listdir('../data/brain_tumor/Training/')
classes = {'no_tumor':0, 'pituitary_tumor':1}
```

```
In [3]: import cv2
X = []
Y = []
for cls in classes:
    pth = '../data/brain_tumor/Training/'+cls
    for j in os.listdir(pth):
        img = cv2.imread(pth+'/'+j, 0)
        img = cv2.resize(img, (200,200))
        X.append(img)
        Y.append(classes[cls])
```

```
In [5]: X = np.array(X)
Y = np.array(Y)
```

```
In [ ]: np.unique(Y)
```

```
In [6]: pd.Series(Y).value_counts()
```

```
Out[6]: 1      827
0       395
dtype: int64
```

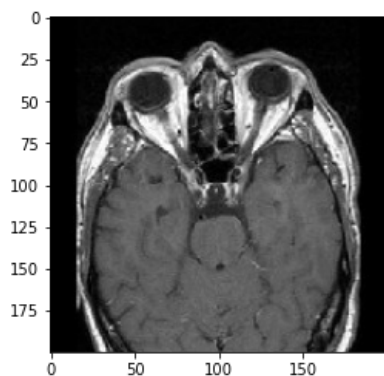
```
In [7]: X.shape
```

```
Out[7]: (1222, 200, 200)
```

Visualize data

```
In [8]: plt.imshow(X[0], cmap='gray')
```

```
Out[8]: <matplotlib.image.AxesImage at 0x1d2466d4400>
```



Prepare data

```
In [9]: X_updated = X.reshape(len(X), -1)
X_updated.shape
```

```
Out[9]: (1222, 40000)
```

Split Data

```
In [10]: xtrain, xtest, ytrain, ytest = train_test_split(X_updated, Y, random_state=10,
test_size=.20)
```

```
In [11]: xtrain.shape, xtest.shape
```

```
Out[11]: ((977, 40000), (245, 40000))
```

Feature Scaling

Feature Scaling

```
In [12]: print(xtrain.max(), xtrain.min())
print(xtest.max(), xtest.min())
xtrain = xtrain/255
xtest = xtest/255
print(xtrain.max(), xtrain.min())
print(xtest.max(), xtest.min())

255 0
255 0
1.0 0.0
1.0 0.0
```

Feature Selection: PCA

```
In [14]: from sklearn.decomposition import PCA

In [15]: print(xtrain.shape, xtest.shape)

pca = PCA(.98)
# pca_train = pca.fit_transform(xtrain)
# pca_test = pca.transform(xtest)
pca_train = xtrain
pca_test = xtest

(977, 40000) (245, 40000)

In [ ]: # print(pca_train.shape, pca_test.shape)
# print(pca.n_components_)
# print(pca.n_features_)
```

Train Model

```
In [16]: from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC

In [20]: import warnings
warnings.filterwarnings('ignore')

lg = LogisticRegression(C=0.1)
lg.fit(pca_train, ytrain)

Out[20]: LogisticRegression(C=0.1)

In [21]: sv = SVC()
sv.fit(pca_train, ytrain)

Out[21]: SVC()
```

Evaluation

```
In [22]: print("Training Score:", lg.score(pca_train, ytrain))
print("Testing Score:", lg.score(pca_test, ytest))

Training Score: 1.0
Testing Score: 0.9591836734693877

In [23]: print("Training Score:", sv.score(pca_train, ytrain))
print("Testing Score:", sv.score(pca_test, ytest))

Training Score: 0.9938587512794268
Testing Score: 0.963265306122449
```

Prediction

```
In [24]: pred = sv.predict(pca_test)
np.where(ytest!=pred)

Out[24]: (array([ 36,  51,  68, 120, 212, 214, 220, 227, 239], dtype=int64),)

In [27]: pred[36]

Out[27]: 0

In [28]: ytest[36]

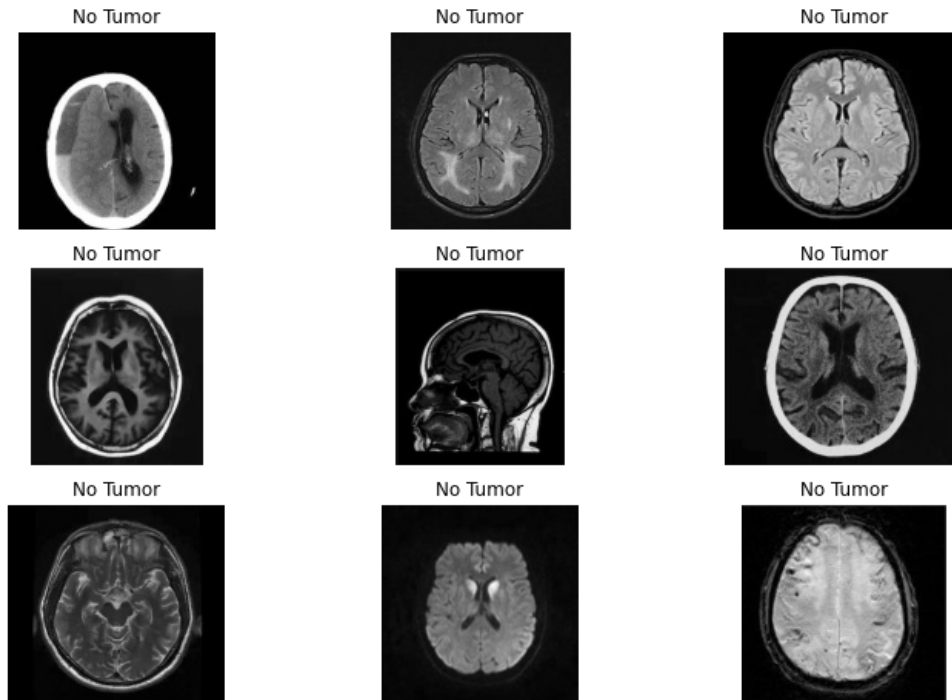
Out[28]: 1
```

TEST MODEL

```
In [29]: dec = {0: 'No Tumor', 1: 'Positive Tumor'}
```

```
In [30]: plt.figure(figsize=(12,8))
p = os.listdir('../data/brain_tumor/Testing/')
c=1
for i in os.listdir('../data/brain_tumor/Testing/no_tumor/')[:9]:
    plt.subplot(3,3,c)

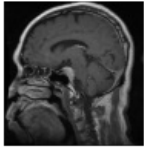
    img = cv2.imread('../data/brain_tumor/Testing/no_tumor/'+i,0)
    img1 = cv2.resize(img, (200,200))
    img1 = img1.reshape(1,-1)/255
    p = sv.predict(img1)
    plt.title(dec[p[0]])
    plt.imshow(img, cmap='gray')
    plt.axis('off')
    c+=1
```



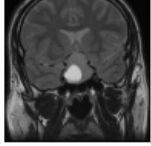
```
In [31]: plt.figure(figsize=(12,8))
p = os.listdir('../data/brain_tumor/Testing/')
c=1
for i in os.listdir('../data/brain_tumor/Testing/pituitary_tumor/')[:16]:
    plt.subplot(4,4,c)

    img = cv2.imread('../data/brain_tumor/Testing/pituitary_tumor/'+i,0)
    img1 = cv2.resize(img, (200,200))
    img1 = img1.reshape(1,-1)/255
    p = sv.predict(img1)
    plt.title(dec[p[0]])
    plt.imshow(img, cmap='gray')
    plt.axis('off')
    c+=1
```

Positive Tumor



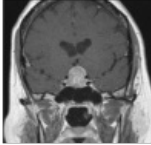
Positive Tumor



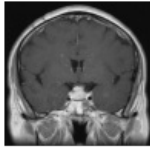
Positive Tumor



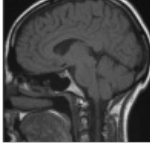
No Tumor



Positive Tumor



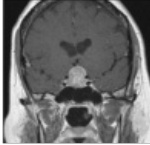
Positive Tumor



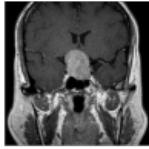
Positive Tumor



No Tumor



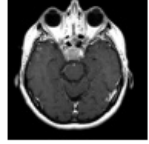
No Tumor



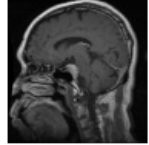
Positive Tumor



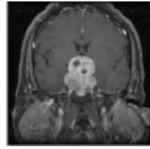
Positive Tumor



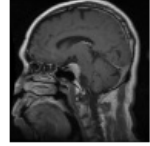
Positive Tumor



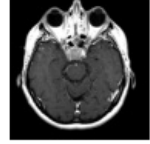
Positive Tumor



Positive Tumor



Positive Tumor



Positive Tumor

