## SmileRecognition

February 17, 2021

Get current working directory.

```
[1]: import os
orig_dir = os.getcwd()
```

Import libraries.

```
[2]: import numpy as np
import pandas as pd
import seaborn as sn
import matplotlib.pyplot as plt
from src.recognition.svm import SVM
```

Load data and labels.

```
[3]: samples_path = 'datasets/processed/landmarks.txt'
labels_path = 'datasets/processed/valences.txt'

X = np.loadtxt(samples_path, dtype=np.float32)
y = np.loadtxt(labels_path).astype(int)
```

Let's see how sample data is represented...

```
[4]: Xdf = pd.DataFrame(X)
print(Xdf.head())
```

```
2
      0
            1
                         3
                                4
                                      5
                                             6
                                                   7
                                                           8
                                                                 9
  400.0
          67.0
                407.0
                       62.0 416.0
                                    61.0
                                          425.0
                                                 67.0
                                                       416.0
                                                               70.0
  400.0
          67.0
                407.0
                       62.0 416.0
                                    61.0
                                          424.0
                                                 67.0
                                                       416.0
                                                               71.0
1
  400.0
                407.0
                       62.0 416.0
                                          424.0
          67.0
                                    61.0
                                                 67.0
                                                       416.0
                                                               71.0
  400.0
          68.0
                407.0
                       62.0 416.0
                                    61.0
                                          424.0
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3
                                                       416.0
                                                               71.0
  400.0
          67.0
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                       62.0 416.0
                                   62.0
                                          424.0
                                                 67.0
                                                       416.0
                                                               71.0
      54
             55
                    56
                           57
                                  58
                                         59
                                                60
                                                               62
                                                       61
                                                                      63
  452.0
                467.0
                               452.0
                                      140.0
                                             446.0
          130.0
                        133.0
                                                    141.0
                                                           439.0
                                                                   140.0
1
  451.0
          131.0
                467.0
                        134.0
                               452.0
                                      142.0
                                             446.0
                                                    143.0
                                                           439.0
                                                                   142.0
                                             446.0
  451.0
                        135.0
                               453.0
                                      145.0
                                                    146.0
                                                                  146.0
2
          132.0
                467.0
                                                           439.0
3
  452.0
          132.0
                 467.0
                        136.0
                               453.0
                                      146.0
                                             446.0
                                                    148.0
                                                           439.0 147.0
  452.0
          132.0
                468.0
                        135.0
                               453.0
                                      145.0
                                             447.0 146.0
                                                           439.0 146.0
```

[5 rows x 64 columns]

...and the same for responses.

[5]: ydf = pd.DataFrame(y)
print(ydf.head())

0

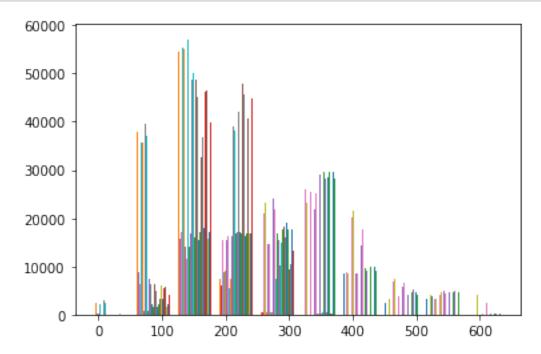
- 0 1
- 1 1
- 2 1
- 3 1
- 4 1

Split data in train and test.

[6]: from sklearn.model\_selection import train\_test\_split
X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y)

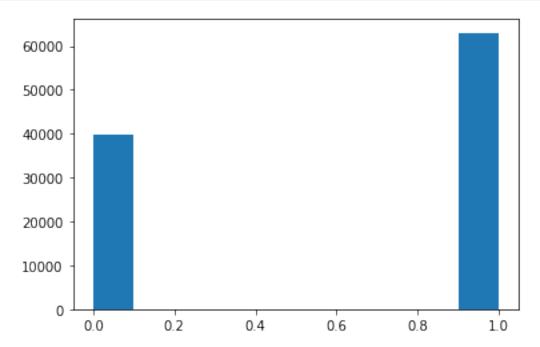
See sample data's distribution.

[7]: plt.hist(X\_train) plt.show()



See response data's distribution

```
[8]: plt.hist(y_train)
plt.show()
```



Initialize classifiers, one for scikit-learn and one for OpenCV.

```
[9]: rec_skl = SVM('skl')
svm = rec_skl.load('datasets/processed/svm.pkl')
```

Predict data

```
[10]: prediction = rec_skl.predict(svm, X_test)

data = [ [p, a] for p,a in zip(y_test, prediction)]

df = pd.DataFrame(data, columns=['Actual', 'Predicted'])
```

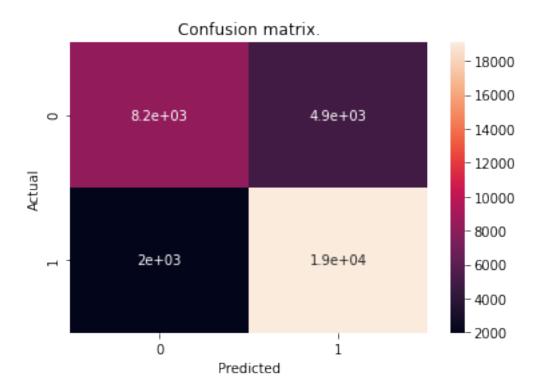
Prediction for the scikit-learn SVM classifier

## [11]: print(df.head())

	Actual	Predicted
0	1	1
1	1	0
2	0	1
3	0	0
4	1	1

Confusion matrixes.

```
[12]: n = y_test.shape[0]
      confusion_matrix = pd.crosstab(df['Actual'], df['Predicted'],__
      →rownames=['Actual'], colnames=['Predicted'])
      sn.heatmap(confusion_matrix, annot=True)
      plt.title('Confusion matrix.')
      plt.show()
      tp = confusion_matrix[0][0]
      fp = confusion_matrix[0][1]
      fn = confusion_matrix[1][0]
      tn = confusion_matrix[1][1]
      tpr = tp/(tp+fn)
      tnr = tn/(fp+tn)
      fpr = 1-tnr
      fnr = 1-tpr
      print("True positives:\t", tp)
      print("False positives:", fp)
      print("True negatives:\t", tn)
      print("False negatives:", fn)
      print("TPR: {:.3f}".format(tpr))
      print("FPR: {:.3f}".format(fpr))
      print("TNR: {:.3f}".format(tnr))
      print("FNR: {:.3f}".format(fnr))
```



True positives: 8231
False positives: 1962
True negatives: 19094
False negatives: 4941

TPR: 0.625 FPR: 0.093 TNR: 0.907 FNR: 0.375