

Import the libraries

```
In [1]: #importing the libraries
import tensorflow as tf
from tensorflow import keras
import numpy as np
import pandas as pd
```

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s\r\losses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is deprecated. Please use tf.compat.v1.losses.sparse_softmax_cross_entropy instead.

Load the data and data pre-processing

```
In [2]: #load the datasets
baseline_df = pd.read_excel('extracted_features_baseline.xlsx')
toolwear_df = pd.read_excel('extracted_features_toolwear.xlsx')
```

```
In [3]: #labelling the datasets. 0 for baseline, 1 for toolwear. This will be the variable the model tries to predict
baseline_df["state"] = 0
toolwear_df["state"] = 1
```

```
In [4]: #concatenate the datasets
combined_df = pd.concat([baseline_df, toolwear_df], axis=0)
print(combined_df.shape)
```

(840, 67)

```
In [5]: #getting the y label
state = combined_df["state"].values
print(state.shape)
```

(840,)

```
In [6]: #getting the features to train the model
features = combined_df.drop('state', axis=1).values
print(features.shape)
```

(840, 66)

```
In [7]: #train test split
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(features, state, test_size=0.2, random_state=40)
```

```
In [8]: #data scaling
from sklearn.preprocessing import StandardScaler

sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

Training and Evaluation of Classification Models

1) Support Vector Machine (SVM)

```
In [9]: #Import svm model from scikit learn
from sklearn import svm

#create the svm classifier model
SVM = svm.SVC(kernel="linear")

#train the model
SVM.fit(X_train, Y_train)
```

```
Out[9]: ▼ SVC
SVC(kernel='linear')
```

```
In [10]: #test the model
predict_SVM = SVM.predict(X_test)

#crosstabs
pd.crosstab(predict_SVM, Y_test)
```

```
Out[10]:
```

	col_0	0	1
row_0			
0	81	0	
1	0	87	

```
In [11]: from sklearn.metrics import classification_report
```

```
#print detailed report for SVM
```

```
print(classification_report(Y_test, predict_SVM))
```

```

              precision    recall  f1-score   support

     0       1.00        1.00        1.00        81
     1       1.00        1.00        1.00        87

 accuracy          1.00          1.00          1.00        168
 macro avg          1.00          1.00          1.00        168
weighted avg          1.00          1.00          1.00        168

```

Timing analysis for SVM

```
In [12]: #data is fed in 1 by 1 in the for loop and the time taken for each prediction is summed in "time_passed"
```

```

result = []
time_passed = 0
import time

for j in range(0, X_test.shape[0]):
    #process starts so record the start time
    start = time.time()

    #classify with SVM
    prediction = SVM.predict(np.array( [X_test[j],] ))

    #process ends so record the end time
    end = time.time()

    #storing the result
    result.append(prediction[0])
    time_passed = time_passed + (end-start)

```

```
In [13]: pd.crosstab(result, Y_test)
```

```
Out[13]:
```

	col_0	0	1
row_0			
0	81	0	
1	0	87	

```
In [14]: print(time_passed)
```

```
0.03381824493408203
```

```
In [15]: #The average time is calculated by dividing the total time with the number of predictions
```

```

avg_time = time_passed / len(result)
print(avg_time)

```

```
0.00020129907698858353
```

2) Naive Bayes

```
In [16]: #Import NaiveBayes model from scikit learn
```

```
from sklearn.naive_bayes import GaussianNB
```

```
#create the NB classifier model
```

```
GAU_NB = GaussianNB()
```

```
#train the model
```

```
GAU_NB.fit(X_train, Y_train)
```

```
Out[16]:
```

```

▼ GaussianNB
GaussianNB()

```

```
In [17]: #test the model
```

```
predict_NB = GAU_NB.predict(X_test)
```

```
#crosstabs  
pd.crosstab(predict_NB, Y_test)
```

```
Out[17]:
```

	col_0	0	1
row_0			
0	79	0	
1	2	87	

```
In [18]: from sklearn.metrics import classification_report
```

```
#print detailed report for NB  
print(classification_report(Y_test, predict_NB))
```

	precision	recall	f1-score	support
0	1.00	0.98	0.99	81
1	0.98	1.00	0.99	87
accuracy			0.99	168
macro avg	0.99	0.99	0.99	168
weighted avg	0.99	0.99	0.99	168

3) K-Nearest Neighbor

```
In [19]: #Import KNN model from scikit learn  
from sklearn.neighbors import KNeighborsClassifier
```

```
#create the KNN classifier model  
KNN = KNeighborsClassifier(n_neighbors=2)
```

```
#train the model  
KNN.fit(X_train, Y_train)
```

```
Out[19]:
```

▼

KNeighborsClassifier

KNeighborsClassifier(n_neighbors=2)

```
In [20]: #test the model  
predict_KNN = KNN.predict(X_test)
```

```
#crosstabs  
pd.crosstab(predict_KNN, Y_test)
```

```
Out[20]:
```

	col_0	0	1
row_0			
0	81	0	
1	0	87	

```
In [21]: from sklearn.metrics import classification_report
```

```
#print detailed report for KNN  
print(classification_report(Y_test, predict_KNN))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	81
1	1.00	1.00	1.00	87
accuracy			1.00	168
macro avg	1.00	1.00	1.00	168
weighted avg	1.00	1.00	1.00	168