

In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import sklearn
```

In [2]:

```
import glob
path="C:\\S1_Dataset"
all_files=glob.glob(path+'/*')
print(all_files)
```

C:\\S1_Dataset\\dip01M', 'C:\\S1_Dataset\\dip02M', 'C:\\S1_Dataset\\dip03M', 'C:\\S1_Dataset\\dip04M', 'C:\\S1_Dataset\\dip05M', 'C:\\S1_Dataset\\dip06M', 'C:\\S1_Dataset\\dip07M', 'C:\\S1_Dataset\\dip08F', 'C:\\S1_Dataset\\dip09F', 'C:\\S1_Dataset\\dip10F', 'C:\\S1_Dataset\\dip11F', 'C:\\S1_Dataset\\dip12F', 'C:\\S1_Dataset\\dip13F', 'C:\\S1_Dataset\\dip14F', 'C:\\S1_Dataset\\dip15F', 'C:\\S1_Dataset\\dip16F', 'C:\\S1_Dataset\\dip17F', 'C:\\S1_Dataset\\dip18F', 'C:\\S1_Dataset\\dip19F', 'C:\\S1_Dataset\\dip20F', 'C:\\S1_Dataset\\dip21F', 'C:\\S1_Dataset\\dip22F', 'C:\\S1_Dataset\\dip23F', 'C:\\S1_Dataset\\dip24F', 'C:\\S1_Dataset\\dip25F', 'C:\\S1_Dataset\\dip26F', 'C:\\S1_Dataset\\dip27F', 'C:\\S1_Dataset\\dip28F', 'C:\\S1_Dataset\\dip29F', 'C:\\S1_Dataset\\dip30F', 'C:\\S1_Dataset\\dip31F', 'C:\\S1_Dataset\\dip32F', 'C:\\S1_Dataset\\dip33F', 'C:\\S1_Dataset\\dip34F', 'C:\\S1_Dataset\\dip35F', 'C:\\S1_Dataset\\dip36M', 'C:\\S1_Dataset\\dip37M', 'C:\\S1_Dataset\\dip38M', 'C:\\S1_Dataset\\dip39M', 'C:\\S1_Dataset\\dip40M', 'C:\\S1_Dataset\\dip41M', 'C:\\S1_Dataset\\dip42M', 'C:\\S1_Dataset\\dip43M', 'C:\\S1_Dataset\\dip44M', 'C:\\S1_Dataset\\dip45M', 'C:\\S1_Dataset\\dip46M', 'C:\\S1_Dataset\\dip47M', 'C:\\S1_Dataset\\dip48M', 'C:\\S1_Dataset\\dip49F', 'C:\\S1_Dataset\\dip50F', 'C:\\S1_Dataset\\dip51F', 'C:\\S1_Dataset\\dip52F', 'C:\\S1_Dataset\\dip53F', 'C:\\S1_Dataset\\dip54F', 'C:\\S1_Dataset\\dip55F', 'C:\\S1_Dataset\\dip56F', 'C:\\S1_Dataset\\dip57F', 'C:\\S1_Dataset\\dip58F', 'C:\\S1_Dataset\\dip59F', 'C:\\S1_Dataset\\dip60F']

In [3]:

```
li=[]
for file in all_files:
    #print(file)
    if file.endswith('.txt'):
        continue
    df=pd.read_csv(file,header=None,index_col=None)
    li.append(df)
healthy_op=pd.concat(li,axis=0,ignore_index=True)
```

In [4]:

```
healthy_op.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 52482 entries, 0 to 52481
Data columns (total 9 columns):
 # Column Non-Null Count Dtype
--- -
 0 0 52482 non-null float64
 1 1 52482 non-null float64
 2 2 52482 non-null float64
 3 3 52482 non-null float64
 4 4 52482 non-null int64
 5 5 52482 non-null float64
 6 6 52482 non-null float64
 7 7 52482 non-null float64
 8 8 52482 non-null int64
dtypes: float64(7), int64(2)
memory usage: 3.6 MB

In [5]:

```
healthy_op.columns=['Time','Acc.Front','Acc.vert','Acc.Lat','id','RSSI','Phase','Freq','Activity_Label']
healthy_op.head()
```

Out[5]:

	Time	Acc.Front	Acc.vert	Acc.Lat	id	RSSI	Phase	Freq	Activity_Label
0	0.00	0.27203	1.00820	-0.082102	1	-63.5	2.4252	924.25	1
1	0.50	0.27203	1.00820	-0.082102	1	-63.0	4.7369	921.75	1
2	1.50	0.44791	0.91636	-0.013684	1	-63.5	3.0311	923.75	1
3	1.75	0.44791	0.91636	-0.013684	1	-63.0	2.0371	921.25	1
4	2.50	0.34238	0.96229	-0.059296	1	-63.5	5.8920	920.25	1

In [6]:

```
healthy_op.isnull().values.any()
```

Out[6]:

False

In [7]:

```
cols=len(healthy_op.columns)-1
healthy_op1=healthy_op.values
X=healthy_op1[:, :8]
Y=healthy_op1[:,8]
```

In [8]:

```
from sklearn.preprocessing import Normalizer
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split,GridSearchCV
from sklearn.metrics import accuracy_score
from sklearn.neighbors import KNeighborsClassifier
```

In [9]:

```
normalize=Normalizer()
X=normalize.fit_transform(X)
print(X)
```

[[0.00000000e+00 2.93631563e-04 1.08825991e-03 ... -6.85424559e-02
 2.61778211e-03 9.97643541e-01]
 [5.41175984e-04 2.94432206e-04 1.09122725e-03 ... -6.81881739e-02
 5.12699303e-03 9.97657926e-01]
 [1.61998019e-03 4.83736884e-04 9.89656697e-04 ... -6.85791613e-02
 3.27354797e-03 9.97637799e-01]
 ...
 [4.64054952e-01 3.16054398e-04 9.08811718e-04 ... -5.78276511e-02
 4.78430619e-03 8.83902815e-01]
 [4.65829320e-01 1.14107909e-04 7.86629014e-04 ... -5.34284525e-02
 4.68329008e-03 8.83239105e-01]
 [4.67376010e-01 6.95187748e-05 1.26064817e-03 ... -5.40291449e-02
 9.91649969e-04 8.82396344e-01]]

In [10]:

```
healthy_op.shape
```

Out[10]:

(52482, 9)

In [11]:

```
train, test = train_test_split(healthy_op, test_size = 0.25)
print(train.shape)
print(test.shape)
```

(39361, 9)
(13121, 9)

In [12]:

```
X_train = train[['Time','Acc.Front','Acc.vert','Acc.Lat','id','RSSI','Phase','Freq']]
Y_train =train.Activity_Label
X_test = test[['Time','Acc.Front','Acc.vert','Acc.Lat','id','RSSI','Phase','Freq']]
Y_test =test.Activity_Label
```

In [13]:

```
X_train.head()
```

Out[13]:

	Time	Acc.Front	Acc.vert	Acc.Lat	id	RSSI	Phase	Freq
24690	110.50	1.16320	-0.059712	-0.093505	3	-59.0	4.16320	920.75
40747	343.45	0.40101	0.962290	-0.013684	1	-58.0	0.65348	924.25
502	82.25	1.10450	-0.266410	-0.150520	3	-56.5	1.07840	922.75
5953	348.00	1.24520	0.146980	0.054735	4	-58.0	4.88730	923.25
10519	451.75	1.26870	-0.151580	0.089944	3	-50.5	4.13560	922.75

In [14]:

```
Y_train.head()
```

Out[14]:

24690 3
40747 1
502 3
5953 3
10519 3
Name: Activity_Label, dtype: int64

In [15]:

```
#standardization
from sklearn.preprocessing import StandardScaler
X_train_stand = X_train.copy()
X_test_stand = X_test.copy()

num_cols = ['Time','Acc.Front','Acc.vert','Acc.Lat','id','RSSI','Phase','Freq']

for i in num_cols:
    scale = StandardScaler().fit(X_train_stand[[i]])
    X_train_stand[i] = scale.transform(X_train_stand[[i]])
    X_test_stand[i] = scale.transform(X_test_stand[[i]])
```

In [16]:

```
# KNN
knn=KNeighborsClassifier(n_neighbors=3)
knn.fit(X_train_stand,Y_train)
y_pred_knn=knn.predict(X_test_stand)
print("accuracy KNN= ",accuracy_score(Y_test,y_pred_knn))
```

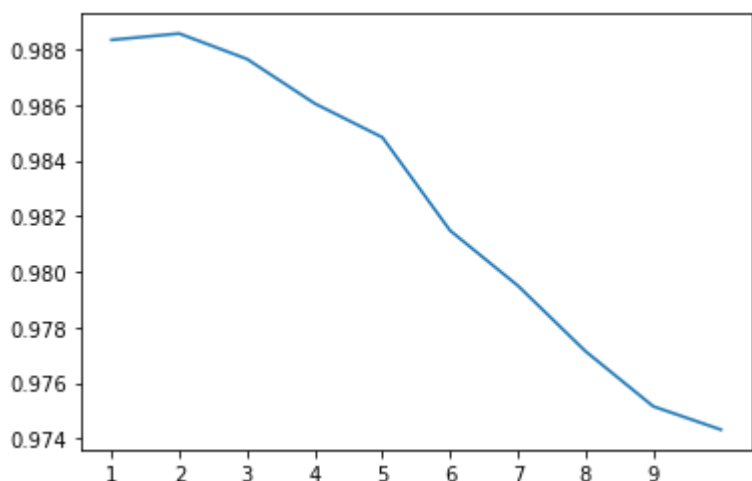
accuracy KNN= 0.987653380077738

In [17]:

```
a_index=list(range(1,11))
a=pd.Series()
x=[1,2,3,4,5,6,7,8,9]
for i in list(range(1,11)):
    model=KNeighborsClassifier(n_neighbors=i)
    model.fit(X_train_stand,Y_train)
    prediction=model.predict(X_test_stand)
    a=a.append(pd.Series(accuracy_score(prediction,Y_test)))
plt.plot(a_index, a)
plt.xticks(x)
```

C:\Users\Ganesh\AppData\Local\Temp\ipykernel_1568\4091055183.py:2: DeprecationWarning: The default dtype for empty Series will be 'object' instead of 'float64' in a future version. Specify a dtype explicitly to silence this warning.
a=pd.Series()

([<matplotlib.axis.XTick at 0x2c3cc4dcee0>,
 <matplotlib.axis.XTick at 0x2c3cc4dceb0>,
 <matplotlib.axis.XTick at 0x2c3cc4dc580>,
 <matplotlib.axis.XTick at 0x2c3cc760b50>,
 <matplotlib.axis.XTick at 0x2c3cc77a0d0>,
 <matplotlib.axis.XTick at 0x2c3cc77a760>,
 <matplotlib.axis.XTick at 0x2c3cc77aeb0>,
 <matplotlib.axis.XTick at 0x2c3cc781640>,
 <matplotlib.axis.XTick at 0x2c3cc781d90>],
 [Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, ''),
 Text(0, 0, '')])



In [18]:

```
lr=LogisticRegression()
estimator=('solver':('newton-cg','liblinear','lbfgs','sag'))
gsc=GridSearchCV(lr,estimator)
gsc.fit(X_train_stand,Y_train)
y_gsc_pred=gsc.predict(X_test_stand)
print("accuracy gsc= ",accuracy_score(Y_test,y_gsc_pred))
print(gsc.best_estimator_)
```

C:\Users\Ganesh\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:763: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
n_iter_i = _check_optimize_result(
C:\Users\Ganesh\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:763: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

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Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
n_iter_i = _check_optimize_result(
C:\Users\Ganesh\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:763: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
n_iter_i = _check_optimize_result(
LogisticRegression(solver='newton-cg')

In [19]:

```
from sklearn.ensemble import RandomForestClassifier
from sklearn import svm
```

In [20]:

```
model = svm.SVC() #select the algorithm
model.fit(X_train_stand,Y_train) # we train the algorithm with the training data and the training output
prediction=model.predict(X_test_stand) #now we pass the testing data to the trained algorithm
print("The accuracy of the SVM is:",accuracy_score(prediction,Y_test))#now we check the accuracy of the algorithm.
#we pass the predicted output by the model and the actual output
```

The accuracy of the SVM is: 0.9777455986586389

In [21]:

```
rforest=RandomForestClassifier()
rforest.fit(X_train_stand,Y_train)
y_pred_rforest=rforest.predict(X_test_stand)
print("accuracy Random Forest= ",accuracy_score(Y_test,y_pred_rforest))
```

accuracy Random Forest= 0.991921347458273

In []: