In [1]:	<pre>import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt import sklearn</pre>
In [2]:	<pre>import glob path="C:\S1_Dataset" all_files=glob.glob(path+'/*') print(all_files)</pre>
	['C:\\S1_Dataset\\d1p01M', 'C:\\S1_Dataset\\d1p02M', 'C:\\S1_Dataset\\d1p03M', 'C:\\S1_Dataset\\d1p04M', 'C:\\S1_Dataset\\d1p05M', 'C:\\S1_Dataset\\d1p06M', 'C:\\S1_Dataset\\d1p07M', 'C:\\S1_Dataset\\d1p07M', 'C:\\S1_Dataset\\d1p17F', 'C:\\S1_Dataset\\d1p18F', 'C:\\S1_Dataset\\d1p13F', 'C:\\S1_Dataset\\d1p17F', 'C:\\S1_Dataset\\d1p18F', 'C:\\S1_Dataset\\d1p19F', 'C:\\S1_Dataset\\d1p27F', 'C:\\S1_Dataset\\d1p27F', 'C:\\S1_Dataset\\d1p23F', 'C:\\S1_Dataset\\d1p24F', 'C:\\S1_Dataset\\
	'C:\\S1_Dataset\\d1p25F', 'C:\\S1_Dataset\\d1p26F', 'C:\\S1_Dataset\\d1p27F', 'C:\\S1_Dataset\\d1p28F', 'C:\\S1_Dataset\\d1p29F', 'C:\\S1_Dataset\\d1p30F', 'C:\\S1_Dataset\\d1p30F', 'C:\\S1_Dataset\\d1p31F', 'C:\\S1_Dataset\\d1p35F', 'C:\\S1_Dataset\\d1p36M', 'C:\\S1_Dataset\\d1p37M', 'C:\\S1_Dataset\\d1p38M', 'C:\\S1_Dataset\\d1p39M', 'C:\\S1_Dataset\\d1p40M', 'C:\\S1_Dataset\\d1p41M', 'C:\\S1_Dataset\\d1p44M', 'C:\\S1_Dataset\\d1p48M', 'C:\\S1_Dataset\\d1p48M', 'C:\\S1_Dataset\\d1p49F', 'C:\\S1_Dataset\\d1p53F', 'C:\\S1_Dataset\\d1p54F', 'C:\\S1_Dataset\\d1p59F', 'C:\\S1_Dataset\\d1p50F', 'C:\\S1_Dataset\\d1p50F', 'C:\\S1_Dataset\\d1p57F', 'C:\\S1_Dataset\\d1p58F', 'C:\\S1_Dataset\\d1p59F', 'C:\\S1_Dataset\\d1p60F']
In [3]:	<pre>li=[] for file in all_files:     #print(file)     if file.endswith('.txt'):         continue</pre>
In [4]:	<pre>df=pd.read_csv(file, header=None, index_col=None) li.append(df) healthy_op=pd.concat(li, axis=0, ignore_index=True)</pre>
	healthy_op.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 52482 entries, 0 to 52481  Data columns (total 9 columns): # Column Non-Null Count Dtype</class>
	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]: Out[5]:	healthy_op.columns=['Time','Acc.Front','Acc.vert','Acc.Lat','id','RSSI','Phase','Freq','Activity_Label'] healthy_op.head()  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]: Out[6]:	Folio
In [7]:	<pre>cols=len(healthy_op.columns)-1 healthy_op1=healthy_op.values X=healthy_op1[:, :8] Y=healthy_op1[:,8]</pre>
In [8]:	<pre>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</pre>
In [9]:	<pre>normalize=Normalizer() X=normalize.fit_transform(X) print(X)</pre>
	[[ 0.00000000e+00
	3.27354797e-03 9.97637799e-01] [ 4.64054952e-01 3.16054398e-04 9.08811718e-045.78276511e-02 4.78430619e-03 8.83902815e-01] [ 4.65829320e-01 1.14107909e-04 7.86629014e-045.34284525e-02 4.68329008e-03 8.83239105e-01]
In [10]:	[ 4.67376010e-01  6.95187748e-05  1.26064817e-035.40291449e-02  9.91649969e-04  8.82396344e-01]]  healthy_op.shape  (52482, 9)
Out[10]: In [11]:	<pre>train, test = train_test_split(healthy_op, test_size = 0.25) print(train.shape) print(test.shape)</pre> (39361, 9)
In [12]:	<pre>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</pre>
In [13]: Out[13]:	X_train.head()  Time Acc.Front Acc.vert Acc.Lat id RSSI Phase Freq  27023 13.850 0.18995 1.008200 0.054735 4 -57.0 0.62433 925.25
	42145         434.750         1.16320         0.009187         -0.070699         1         -56.0         5.12500         922.25           48453         309.750         1.15140         0.020670         -0.104910         1         -56.5         5.69410         921.25           21562         76.175         1.04590         0.020670         -0.161920         3         -61.5         5.64810         923.25           24070         335.100         1.05760         0.020670         -0.150520         1         -55.5         5.12500         920.75
In [14]: Out[14]:	Y_train.head()  27023
In [15]:	24070 3 Name: Activity_Label, dtype: int64  #normalization normalize=Normalizer() n_X_train=normalize.fit_transform(X_train)
In [16]:	<pre>normalize=Normalizer() n_X_test=normalize.fit_transform(X_test)  # KNN</pre>
	<pre>knn=KNeighborsClassifier(n_neighbors=3) knn.fit(n_X_train, Y_train) y_pred_knn=knn.predict(n_X_test) print("accuracy KNN= ",accuracy_score(Y_test,y_pred_knn)) accuracy KNN= 0.9312552396920967</pre>
In [17]:	<pre>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)</pre>
	<pre>model.fit(n_X_train,Y_train) prediction=model.predict(n_X_test) a=a.append(pd.Series(accuracy_score(prediction,Y_test))) plt.plot(a_index, a) plt.xticks(x)</pre>
Out[17]:	C:\Users\Ganesh\AppData\Local\Temp/ipykernel_10616/3442830369.py:2: DeprecationWarning: The default dtype for empty Series will be 'object' instead of 'float6 4' in a future version. Specify a dtype explicitly to silence this warning.  a=pd.Series()  ([ <matplotlib.axis.xtick 0x27e9141da60="" at="">,</matplotlib.axis.xtick>
	<pre><matplotlib.axis.xtick 0x27e92dec640="" at="">, <matplotlib.axis.xtick 0x27e92decd90="" at="">, <matplotlib.axis.xtick 0x27e92decdc0="" at="">, <matplotlib.axis.xtick 0x27e92df8640="" at="">, <matplotlib.axis.xtick 0x27e92df8640="" at="">, <matplotlib.axis.xtick 0x27e92df8d90="" at="">, <matplotlib.axis.xtick 0x27e92dfd520="" at="">],</matplotlib.axis.xtick></matplotlib.axis.xtick></matplotlib.axis.xtick></matplotlib.axis.xtick></matplotlib.axis.xtick></matplotlib.axis.xtick></matplotlib.axis.xtick></pre>
	[Text(0, 0, ''), Text(0, 0, ''),
	Text(0, 0, ''), Text(0, 0, '')]) 0.95 0.94
	0.93 - 0.92 - 0.91 - 0.90 -
	0.89 - 0.88 - 0.87 - 1 2 3 4 5 6 7 8 9
In [18]:	<pre>lr=LogisticRegression() estimator={'solver':('newton-cg','liblinear','lbfgs','sag')} gsc=GridSearchCV(lr,estimator) gsc.fit(n_X_train,Y_train) y_gsc_pred=gsc.predict(n_X_test) print("accuracy gsc= ",accuracy_score(Y_test,y_gsc_pred))</pre>
	<pre>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.</pre>
	<pre>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( accuracy gsc= 0.5956863043975307 LogisticRegression(solver='newton-cg')</pre>
In [19]:	<pre>LogisticRegression(solver='newton-cg')  from sklearn.ensemble import RandomForestClassifier from sklearn import svm</pre>
In [20]:	model = svm.SVC() #select the algorithm model.fit(n_X_train, Y_train) # we train the algorithm with the training data and the training output prediction=model.predict(n_X_test) #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
In [21]:	The accuracy of the SVM is: 0.5849401722429692  rforest=RandomForestClassifier() rforest.fit(n_X_train, Y_train) y_pred_rforest=rforest.predict(n_X_test) print("accuracy Random Forest= ",accuracy_score(Y_test,y_pred_rforest))
In [ ]:	accuracy Random Forest= 0.9939791174453166