

In [2]:

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

In [3]:

```
hrv=pd.read_csv("/Users/baskaran/Downloads/HRV csv.csv")
```

In [4]:

```
hrv
```

	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2	LF.HF.ratio.LombScargle	LF.HF.ratio
1							
0	0	1	89.411160	0.130259	0.045349	0.115000	1.996490
1	0	1	86.805760	0.026824	0.009504	0.024437	0.879037
2	0	1	85.507670	0.035203	0.009715	0.033555	2.542225
3	0	1	95.127480	0.013182	0.006829	0.009571	1.596207
4	0	1	93.370410	0.016777	0.006921	0.013585	1.128488
...
4309	0	0	111.683756	0.031518	0.015239	0.018471	0.495158
4310	0	0	111.409998	0.048524	0.021231	0.030251	0.468523
4311	0	0	114.402754	0.053378	0.009238	0.038498	3.641941
4312	0	0	120.986143	0.049403	0.012527	0.032305	1.138312
4313	0	0	109.532809	0.023737	0.007142	0.016945	3.862068

In [5]:

```
hrv.head()
```

Out[5]:

SI	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2	LF.HF.ratio.LombScargle	LF.HF.ratio
1							
0	0	1	89.41116	0.130259	0.045349	0.115000	1.996490
1	0	1	86.80576	0.026824	0.009504	0.024437	0.879037
2	0	1	85.50767	0.035203	0.009715	0.033555	2.542225
3	0	1	95.12748	0.013182	0.006829	0.009571	1.596207
4	0	1	93.37041	0.016777	0.006921	0.013585	1.128488

5 rows × 59 columns

In [6]:

```
hrv.describe()
```

Out[6]:

	SI > 1	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2
count	4314.000000	4314.000000	4314.000000	4314.000000	4314.000000	4314.000000
mean	0.056328	0.124710	89.521243	0.038975	0.012284	0.036459
std	0.230581	0.330428	16.664347	0.023929	0.009302	0.026529
min	0.000000	0.000000	56.485260	0.006346	0.001003	0.004482
25%	0.000000	0.000000	75.815800	0.022457	0.005034	0.018612
50%	0.000000	0.000000	88.548570	0.032952	0.009728	0.028810
75%	0.000000	0.000000	101.961156	0.049546	0.016937	0.046719
max	1.000000	1.000000	183.984092	0.272920	0.063498	0.305502

8 rows × 59 columns

In [7]:

hrv.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 4314 entries, 0 to 4313

Data columns (total 59 columns):

#	Column	Non-Null Count	Dtype
0	SI > 1	4314 non-null	int64
1	Sepsis3	4314 non-null	int64
2	Mean.rate	4314 non-null	float64
3	Coefficient.of.variation	4314 non-null	float64
4	Poincar..SD1	4314 non-null	float64
5	Poincar..SD2	4314 non-null	float64
6	LF.HF.ratio.LombScargle	4314 non-null	float64
7	LF.Power.LombScargle	4314 non-null	float64
8	HF.Power.LombScargle	4314 non-null	float64
9	DFA.Alpha.1	4314 non-null	float64
10	DFA.Alpha.2	4314 non-null	float64
11	Largest.Lyapunov.exponent	4314 non-null	float64
12	Correlation.dimension	4314 non-null	float64
13	Power.Law.Slope.LombScargle	4314 non-null	float64
14	Power.Law.Y.Intercept.LombScargle	4314 non-null	float64
15	DFA.AUC	4314 non-null	float64
16	Multiscale.Entropy	4314 non-null	float64
17	VLF.Power.LombScargle	4314 non-null	float64
18	Complexity	4314 non-null	float64
19	eScaleE	4314 non-null	float64
20	pR	4314 non-null	float64
21	pD	4314 non-null	float64
22	dlmax	4314 non-null	float64
23	sedl	4314 non-null	float64
24	pDpR	4314 non-null	float64
25	pL	4314 non-null	float64
26	vlmax	4314 non-null	float64
27	sevl	4314 non-null	float64
28	shannEn	4314 non-null	float64
29	PSeo	4314 non-null	float64
30	Teo	4314 non-null	float64
31	SymDp0_2	4314 non-null	float64
32	SymDp1_2	4314 non-null	float64
33	SymDp2_2	4314 non-null	float64
34	SymDfw_2	4314 non-null	int64
35	SymDse_2	4314 non-null	float64
36	SymDce_2	4314 non-null	float64
37	formF	4314 non-null	float64
38	gcount	4314 non-null	float64
39	sgridAND	4314 non-null	float64
40	sgridTAU	4314 non-null	float64
41	sgridWGT	4314 non-null	float64
42	aFdP	4314 non-null	float64
43	fFdP	4314 non-null	float64
44	IoV	4314 non-null	float64
45	KLPE	4314 non-null	float64
46	AsymI	4314 non-null	float64
47	CSI	4314 non-null	float64
48	CVI	4314 non-null	float64
49	ARerr	4314 non-null	float64
50	histSI	4314 non-null	float64
51	MultiFractal_c1	4314 non-null	float64

```

52 MultiFractal_c2          4314 non-null    float64
53 SDLEalpha                4314 non-null    float64
54 SDLEmean                 4314 non-null    float64
55 QSE                      4314 non-null    float64
56 Hurst.exponent           4314 non-null    float64
57 mean                     4314 non-null    float64
58 median                   4314 non-null    float64

```

```
dtypes: float64(56), int64(3)
```

```
memory usage: 1.9 MB
```

In [107]:

```
corr=hrv.corr()
corr
```

Correlation.dimension	0.009734	-0.032485	-0.166156	-0.114705	0.244667	-0.111
Power.Law.Slope.LombScargle	-0.042618	0.225378	0.018384	0.035461	0.481598	-0.051
Power.Law.Y.Intercept.LombScargle	-0.020171	0.209579	0.071411	-0.023391	0.323432	-0.081
DFA.AUC	-0.250401	0.255221	-0.660435	0.715520	0.733504	0.741
Multiscale.Entropy	-0.051426	0.023695	-0.325469	-0.220344	0.060554	-0.121
VLF.Power.LombScargle	0.083420	-0.233340	0.347033	-0.149871	-0.701779	-0.091
Complexity	0.106864	-0.177325	0.389119	-0.515126	-0.733795	-0.451
eScaleE	0.018142	-0.049849	0.001698	-0.149347	-0.137451	-0.111
pR	-0.032106	0.059559	0.196255	0.269682	0.183284	0.131
pD	-0.004502	-0.001569	0.298086	0.368837	0.013232	0.251
dlmax	-0.046364	0.032961	0.143503	0.465837	0.098844	0.351
sedl	0.011871	-0.057758	0.326715	0.330435	-0.032931	0.211
pDpR	0.024064	-0.086480	0.108515	0.169715	-0.098566	0.171

SIMPLE FEATURE SCALING

In [110]:

```
hrv=hrv/hrv.max()  
hrv
```

Out[110]:

	SI	>	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2	LF.HF.ratio.Lom
	1							
0	0.0		1.0	0.485972	0.477279	0.714180	0.376430	
1	0.0		1.0	0.471811	0.098285	0.149674	0.079990	
2	0.0		1.0	0.464756	0.128987	0.152997	0.109836	
3	0.0		1.0	0.517042	0.048300	0.107547	0.031329	
4	0.0		1.0	0.507492	0.061472	0.108996	0.044468	
...	
4309	0.0		0.0	0.607029	0.115484	0.239992	0.060461	
4310	0.0		0.0	0.605541	0.177796	0.334357	0.099021	
4311	0.0		0.0	0.621808	0.195581	0.145485	0.126016	
4312	0.0		0.0	0.657590	0.181016	0.197282	0.105744	
4313	0.0		0.0	0.595338	0.086974	0.112476	0.055466	

4314 rows × 59 columns

MIN MAX NORMALISATION

In [111]:

```
hrv_mmn=pd.read_csv("/Users/baskaran/Downloads/HRV csv.csv")
hrv_mmn=(hrv_mmn-hrv_mmn.min())/(hrv_mmn.max()-hrv_mmn.min())
hrv_mmn
```

Out[111]:

	SI	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2	LF.HF.ratio.LombScargle	LF.Pow
0	0.0	1.0	0.258245	0.464835	0.709593	0.367145	0.040955	
1	0.0	1.0	0.237810	0.076819	0.136027	0.066291	0.017583	
2	0.0	1.0	0.227629	0.108251	0.139403	0.096582	0.052369	
3	0.0	1.0	0.303079	0.025644	0.093223	0.016906	0.032583	
4	0.0	1.0	0.289298	0.039130	0.094696	0.030241	0.022800	
...
4309	0.0	0.0	0.432933	0.094428	0.227794	0.046472	0.009554	
4310	0.0	0.0	0.430786	0.158222	0.323674	0.085606	0.008996	
4311	0.0	0.0	0.454259	0.176431	0.131771	0.113002	0.075370	

Z-SCORE NORMALISATION

In [112]:

```
hrv_zs=pd.read_csv("/Users/baskaran/Downloads/HRV csv.csv")
hrv_zs=(hrv_zs-hrv_zs.mean())/hrv_zs.std()
hrv_zs
```

	SI > 1	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2	LF.HF.ratio.LombScargle	
0	-0.244288	2.648955	-0.006606	3.814704	3.554740	2.960603	-0.151281	
1	-0.244288	2.648955	-0.162952	-0.507799	-0.298894	-0.453155	-0.546556	
2	-0.244288	2.648955	-0.240848	-0.157645	-0.276210	-0.109453	0.041760	
3	-0.244288	2.648955	0.336421	-1.077893	-0.586479	-1.013526	-0.292873	
4	-0.244288	2.648955	0.230982	-0.927659	-0.576588	-0.862219	-0.458318	
...
4309	-0.244288	-0.377420	1.329936	-0.311639	0.317666	-0.678042	-0.682344	
4310	-0.244288	-0.377420	1.313508	0.399034	0.961855	-0.233997	-0.691766	
4311	-0.244288	-0.377420	1.493099	0.601881	-0.327491	0.076872	0.430761	
4312	-0.244288	-0.377420	1.888157	0.435767	0.026103	-0.156572	-0.454843	
4313	-0.244288	-0.377420	1.200861	-0.636804	-0.552829	-0.735564	0.508626	

In [113]:

```
import pandas as pd
from sklearn import preprocessing
hrv=pd.read_csv("/Users/baskaran/Downloads/HRV csv.csv")
```

In [114]:

hrv.columns

Out[114]:

```
Index(['SI > 1', 'Sepsis3', 'Mean.rate', 'Coefficient.of.variation',
      'Poincar..SD1', 'Poincar..SD2', 'LF.HF.ratio.LombScargle',
      'LF.Power.LombScargle', 'HF.Power.LombScargle', 'DFA.Alpha.1',
      'DFA.Alpha.2', 'Largest.Lyapunov.exponent', 'Correlation.dimensions',
      'Power.Law.Slope.LombScargle', 'Power.Law.Y.Intercept.LombScargle',
      'DFA.AUC', 'Multiscale.Entropy', 'VLF.Power.LombScargle', 'Complexity',
      'eScaleE', 'pR', 'pD', 'dlmax', 'sedl', 'pDpR', 'pL', 'vlmax',
      'sevl',
      'shannEn', 'PSeo', 'Teo', 'SymDp0_2', 'SymDp1_2', 'SymDp2_2',
      'SymDfw_2', 'SymDse_2', 'SymDce_2', 'formF', 'gcount', 'sgridAND',
      'sgridTAU', 'sgridWGT', 'aFdP', 'fFdP', 'IoV', 'KLPE', 'AsymI',
      'CSI',
      'CVI', 'ARerr', 'histSI', 'MultiFractal_c1', 'MultiFractal_c2',
      'SDLEalpha', 'SDLEmean', 'QSE', 'Hurst.exponent', 'mean', 'median'],
      dtype='object')
```

In [103]:

```
hrv["Sepsis3"].corr(hrv["Sepsis3"])  
for i in hrv.columns:  
    c=hrv[i].corr(hrv["Sepsis3"])  
    print(c)
```

```
-0.02527187721937068  
1.0  
-0.16339079489084518  
0.1677841726215487  
0.22365614440654025  
0.21044018286202543  
-0.07323413923349596  
0.08020929373715716  
0.057125647101376034  
-0.04022189066112259  
-0.1225839432399076  
-0.04395880818673318  
-0.03248454585523562  
0.2253782217742344  
0.20957850071545472  
0.2552208940628494  
0.023695335837259166  
-0.23334038308966018  
-0.17732538232419662  
-0.04984913736416616  
0.05955896505753054  
-0.0015692246137141163  
0.032961392744447794  
-0.057757892311287716  
-0.0864795918199675  
0.01647207525861744  
0.1104639748136656  
-0.014202152841250166  
-0.03753856244824713  
0.22920153852827524  
0.2359670597116792  
-0.16131169746197865  
0.18964678425672235  
0.10120606015202153  
-0.15140349293731684  
0.18099042437881668  
0.13612354757955894  
-0.12117184926613388  
-0.2580152720895615  
-0.26290505085522325  
0.16172489536943382  
0.16790082520811114  
0.3400543647300645  
0.3531992587766918  
0.1796164525390422  
-0.16968590868391098  
-0.1183767046913884  
-0.12070166527698939  
0.2148044585830974  
0.11565868773292132  
-0.09680280266447794  
0.11245506699536517  
-0.04091915207988212
```



```
0.050061979673419806
0.04620081878268628
0.17120638583682043
0.17146838911164766
0.1763162886773732
0.1539233794535976
```

In [116]:

```
hrv.columns
```

Out[116]:

```
Index(['SI > 1', 'Sepsis3', 'Mean.rate', 'Coefficient.of.variation',
      'Poincar..SD1', 'Poincar..SD2', 'LF.HF.ratio.LombScargle',
      'LF.Power.LombScargle', 'HF.Power.LombScargle', 'DFA.Alpha.1',
      'DFA.Alpha.2', 'Largest.Lyapunov.exponent', 'Correlation.dimens
ion',
      'Power.Law.Slope.LombScargle', 'Power.Law.Y.Intercept.LombScarg
le',
      'DFA.AUC', 'Multiscale.Entropy', 'VLF.Power.LombScargle', 'Comp
lexity',
      'eScaleE', 'pR', 'pD', 'dlmax', 'sedl', 'pDpR', 'pL', 'vlmax',
      'sevl',
      'shannEn', 'PSeo', 'Teo', 'SymDp0_2', 'SymDp1_2', 'SymDp2_2',
      'SymDfw_2', 'SymDse_2', 'SymDce_2', 'formF', 'gcount', 'sgridAN
D',
      'sgridTAU', 'sgridWGT', 'aFdP', 'fFdP', 'IoV', 'KLPE', 'AsymI',
      'CSI',
      'CVI', 'ARerr', 'histSI', 'MultiFractal_c1', 'MultiFractal_c2',
      'SDLEalpha', 'SDLEmean', 'QSE', 'Hurst.exponent', 'mean', 'medi
an'],
      dtype='object')
```

K-NEAREST NEIGHBOUR

In [96]:

```
dummies=pd.get_dummies(hrv)
```

In [97]:

```
import cufflinks as cf
```

In [98]:

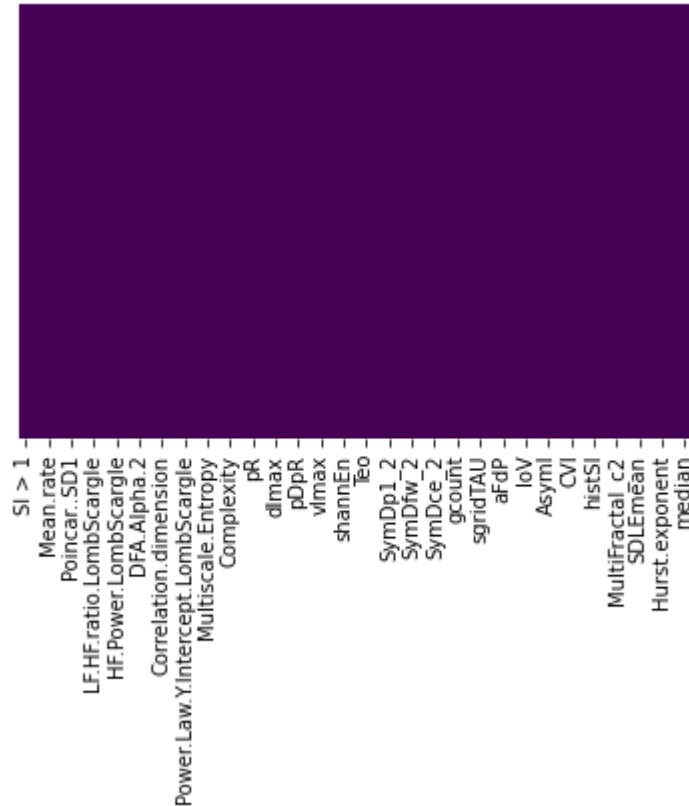
```
cf.go_offline()
```

In [99]:

```
sns.heatmap(hrv.isnull(),yticklabels=False, cbar=False, cmap='viridis')
```

Out[99]:

<AxesSubplot:>



In [101]:

`hrv.keys()`

Out[101]:

```
Index(['SI > 1', 'Sepsis3', 'Mean.rate', 'Coefficient.of.variation',
      'Poincar..SD1', 'Poincar..SD2', 'LF.HF.ratio.LombScargle',
      'LF.Power.LombScargle', 'HF.Power.LombScargle', 'DFA.Alpha.1',
      'DFA.Alpha.2', 'Largest.Lyapunov.exponent', 'Correlation.dimensions',
      'Power.Law.Slope.LombScargle', 'Power.Law.Y.Intercept.LombScargle',
      'DFA.AUC', 'Multiscale.Entropy', 'VLF.Power.LombScargle', 'Complexity',
      'eScaleE', 'pR', 'pD', 'dlmax', 'sedl', 'pDpR', 'pL', 'vlmax', 'sevl',
      'shannEn', 'PSeo', 'Teo', 'SymDp0_2', 'SymDp1_2', 'SymDp2_2',
      'SymDfw_2', 'SymDse_2', 'SymDce_2', 'formF', 'gcount', 'sgridAND',
      'sgridTAU', 'sgridWGT', 'aFdP', 'fFdP', 'IoV', 'KLPE', 'AsymI', 'CSI',
      'CVI', 'ARerr', 'histSI', 'MultiFractal_c1', 'MultiFractal_c2',
      'SDLEalpha', 'SDLEmean', 'QSE', 'Hurst.exponent', 'mean', 'median'],
      dtype='object')
```

In []:

In [186]:

`from sklearn.preprocessing import StandardScaler`

In [187]:

`scaler= StandardScaler()`

In [188]:

`scaler.fit(hrv.drop('Sepsis3',axis=1))`

Out[188]:

`StandardScaler()`

In [189]:

```
hrv['Sepsis3']
```

Out[189]:

```
0      1
1      1
2      1
3      1
4      1
..
4309   0
4310   0
4311   0
4312   0
4313   0
Name: Sepsis3, Length: 4314, dtype: int64
```

In [190]:

```
arr= hrv['Sepsis3'].to_numpy()
arr
```

Out[190]:

```
array([1, 1, 1, ..., 0, 0, 0])
```

In [191]:

```
arr.reshape(-1,1)
```

Out[191]:

```
array([[1],
       [1],
       [1],
       ...,
       [0],
       [0],
       [0]])
```

In [195]:

```
from sklearn.model_selection import train_test_split
```

In [196]:

```
x = hrv
y = np.ravel(arr)
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.30,random_state=58)
```

In [197]:

```
x=hrv.iloc[:,2:].values
y=hrv.iloc[:,1:].values
x
```

Out[197]:

```
array([[8.94111600e+01, 1.30259000e-01, 4.53490000e-02, ...,
        2.93162000e-01, 1.46666670e-02, 1.00000000e-02],
       [8.68057600e+01, 2.68240000e-02, 9.50400000e-03, ...,
        1.89769000e-01, 1.84000000e-02, 1.00000000e-02],
       [8.55076700e+01, 3.52030000e-02, 9.71500000e-03, ...,
        8.42615000e-01, 7.56666700e-03, 1.00000000e-02],
       ...,
       [1.14402754e+02, 5.33780000e-02, 9.23800000e-03, ...,
        1.06483000e-01, 6.01333330e-02, 3.00000000e-02],
       [1.20986143e+02, 4.94030000e-02, 1.25270000e-02, ...,
        5.01250000e-02, 6.96666670e-02, 4.00000000e-02],
       [1.09532809e+02, 2.37370000e-02, 7.14200000e-03, ...,
        8.47800000e-01, 2.31666670e-02, 2.00000000e-02]])
```

In [198]:

```
from sklearn.neighbors import KNeighborsClassifier
```

In [199]:

```
knn= KNeighborsClassifier(n_neighbors=1)
```

In [200]:

```
knn.fit(x_train,y_train)
```

Out[200]:

```
KNeighborsClassifier(n_neighbors=1)
```

In [201]:

```
pred=knn.predict(x_test)
```

In [202]:

```
from sklearn.metrics import classification_report,confusion_matrix,accuracy_score
```

In [203]:

```
print(confusion_matrix(y_test,pred))
```

```
[[1074  60]
 [  92  69]]
```

In [204]:

```
print(classification_report(y_test,pred))
```

	precision	recall	f1-score	support
0	0.92	0.95	0.93	1134
1	0.53	0.43	0.48	161
accuracy			0.88	1295
macro avg	0.73	0.69	0.70	1295
weighted avg	0.87	0.88	0.88	1295

In [205]:

```
error_rate = []

for k in range(1,40):

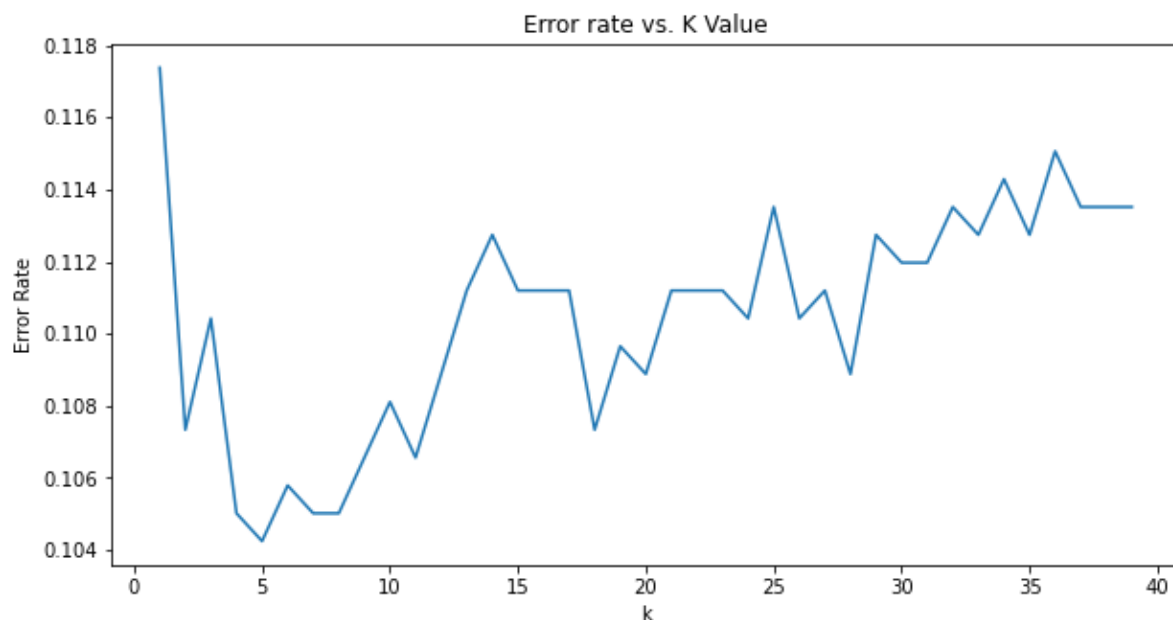
    knn= KNeighborsClassifier(n_neighbors=k)
    knn.fit(x_train,y_train)
    pred_k = knn.predict(x_test)
    error_rate.append(np.mean(pred_k != y_test))
```

In [206]:

```
plt.figure(figsize=(10,5))
plt.plot(range(1,40),error_rate)
plt.title('Error rate vs. K Value')
plt.xlabel('k')
plt.ylabel('Error Rate')
```

Out[206]:

```
Text(0, 0.5, 'Error Rate')
```



In [207]:

```

knn=KNeighborsClassifier(n_neighbors=38)

knn.fit(x_train,y_train)
pred = knn.predict(x_test)

print('with k=38')
print('\n')
print(confusion_matrix(y_test,pred))
print('\n')
print(classification_report(y_test,pred))

```

with k=38

```

[[1131   3]
 [ 144  17]]

```

	precision	recall	f1-score	support
0	0.89	1.00	0.94	1134
1	0.85	0.11	0.19	161
accuracy			0.89	1295
macro avg	0.87	0.55	0.56	1295
weighted avg	0.88	0.89	0.85	1295

In [208]:

```

knn_accuracy = accuracy_score(y_test, pred)
knn_accuracy

```

Out[208]:

0.8864864864864865

SVM

In [139]:

```

dummies=pd.get_dummies(hrv)

```

In [140]:

hrv

Out[140]:

	SI > 1	Sepsis3	Mean.rate	Coefficient.of.variation	Poincar..SD1	Poincar..SD2	LF.HF.ratio.LombScargle	LF.Pov
0	0	1	89.411160	0.130259	0.045349	0.115000	1.996490	
1	0	1	86.805760	0.026824	0.009504	0.024437	0.879037	
2	0	1	85.507670	0.035203	0.009715	0.033555	2.542225	
3	0	1	95.127480	0.013182	0.006829	0.009571	1.596207	
4	0	1	93.370410	0.016777	0.006921	0.013585	1.128488	
...
4309	0	0	111.683756	0.031518	0.015239	0.018471	0.495158	
4310	0	0	111.409998	0.048524	0.021231	0.030251	0.468523	
4311	0	0	114.402754	0.053378	0.009238	0.038498	3.641941	

In [141]:

hrv.keys()

Out[141]:

```
Index(['SI > 1', 'Sepsis3', 'Mean.rate', 'Coefficient.of.variation',
      'Poincar..SD1', 'Poincar..SD2', 'LF.HF.ratio.LombScargle',
      'LF.Power.LombScargle', 'HF.Power.LombScargle', 'DFA.Alpha.1',
      'DFA.Alpha.2', 'Largest.Lyapunov.exponent', 'Correlation.dimen-
      sion',
      'Power.Law.Slope.LombScargle', 'Power.Law.Y.Intercept.LombScarg-
      le',
      'DFA.AUC', 'Multiscale.Entropy', 'VLF.Power.LombScargle', 'Comp-
      lexity',
      'eScaleE', 'pR', 'pD', 'dlmax', 'sedl', 'pDpR', 'pL', 'vlmax',
      'sevl',
      'shannEn', 'PSeo', 'Teo', 'SymDp0_2', 'SymDp1_2', 'SymDp2_2',
      'SymDfw_2', 'SymDse_2', 'SymDce_2', 'formF', 'gcount', 'sgridAN-
      D',
      'sgridTAU', 'sgridWGT', 'aFdP', 'fFdP', 'IoV', 'KLPE', 'AsymI',
      'CSI',
      'CVI', 'ARerr', 'histSI', 'MultiFractal_c1', 'MultiFractal_c2',
      'SDLEalpha', 'SDLEmean', 'QSE', 'Hurst.exponent', 'mean', 'medi-
      an'],
      dtype='object')
```

In [210]:

```
from sklearn.preprocessing import StandardScaler
```

In [211]:

```
scaler= StandardScaler()
```


In [212]:

```
scaler.fit(hrv.drop('Sepsis3',axis=1))
```

Out[212]:

```
StandardScaler()
```

In [213]:

```
hrv['Sepsis3']
```

Out[213]:

```
0      1
1      1
2      1
3      1
4      1
..
4309   0
4310   0
4311   0
4312   0
4313   0
Name: Sepsis3, Length: 4314, dtype: int64
```

In [214]:

```
arr = hrv['Sepsis3'].to_numpy()
arr
```

Out[214]:

```
array([1, 1, 1, ..., 0, 0, 0])
```

In [215]:

```
arr.reshape(-1,1)
```

Out[215]:

```
array([[1],
       [1],
       [1],
       ...,
       [0],
       [0],
       [0]])
```

In [216]:

```
from sklearn.model_selection import train_test_split
```

In [217]:

```
x = hrv
y = np.ravel(arr)
x_test,x_train,y_test,y_train = train_test_split(x,y, test_size=0.5, random_state=20)
```

In [218]:

```
from sklearn.svm import SVC
```

In [219]:

```
model = SVC()
```

In [220]:

```
model.fit(x_train,y_train)
```

Out[220]:

```
SVC()
```

In [221]:

```
predictions = model.predict(x_test)
```

In [222]:

```
from sklearn.metrics import classification_report,confusion_matrix
```

In [223]:

```
print(confusion_matrix(y_test,predictions))
```

```
[[1897    0]
 [ 260    0]]
```

In [224]:

```
print(classification_report(y_test, predictions))
```

	precision	recall	f1-score	support
0	0.88	1.00	0.94	1897
1	0.00	0.00	0.00	260
accuracy			0.88	2157
macro avg	0.44	0.50	0.47	2157
weighted avg	0.77	0.88	0.82	2157

```
/opt/anaconda3/lib/python3.9/site-packages/sklearn/metrics/_classification.py:1248: UndefinedMetricWarning:
```

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

```
/opt/anaconda3/lib/python3.9/site-packages/sklearn/metrics/_classification.py:1248: UndefinedMetricWarning:
```

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

```
/opt/anaconda3/lib/python3.9/site-packages/sklearn/metrics/_classification.py:1248: UndefinedMetricWarning:
```

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

In [225]:

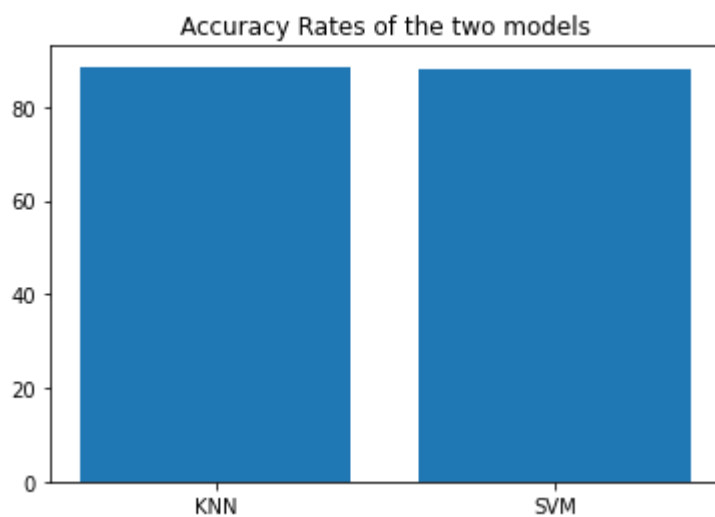
```
svm_accuracy=accuracy_score(y_test, predictions)
svm_accuracy
```

Out[225]:

0.8794622160407974

In [228]:

```
model = ['KNN', 'SVM']  
acc = [88.64864864864865, 87.94622160407974]  
plt.bar(model, acc)  
plt.title("Accuracy Rates of the two models")  
plt.show()
```



In []:

DATA VISUALISATION

In [8]:

```
data = pd.read_csv('/Users/baskaran/Downloads/HRV csv.csv')
```

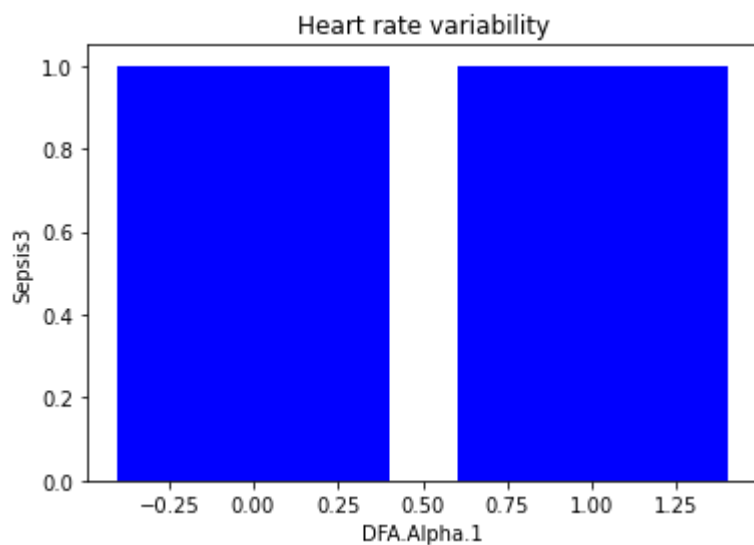
In [9]:

```
df = pd.DataFrame(data)

X = list(df.iloc[:, 0])
Y = list(df.iloc[:, 1])

plt.bar(X, Y, color='b')
plt.title("Heart rate variability")
plt.xlabel("DFA.Alpha.1")
plt.ylabel("Sepsis3")

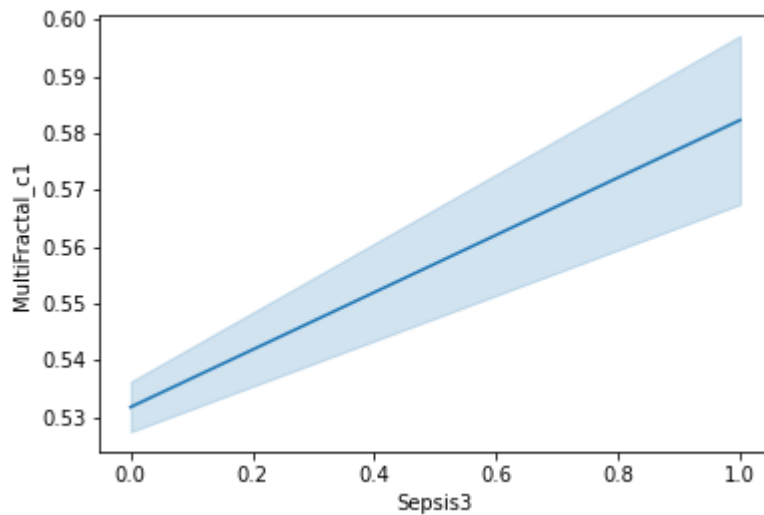
plt.show()
```



In []:

In [11]:

```
import seaborn
import pandas
import matplotlib.pyplot as plt
res = seaborn.lineplot(x="Sepsis3", y="MultiFractal_c1", data=data)
plt.show()
```



In [8]:

```
import seaborn
import pandas
import matplotlib.pyplot as plt
csv = pandas.read_csv("/Users/baskaran/Downloads/HRV csv.csv")
res = seaborn.violinplot(x=csv['Sepsis3'])
plt.show()
```

