

In [3]: `import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import itertools
plt.style.use('fivethirtyeight')`

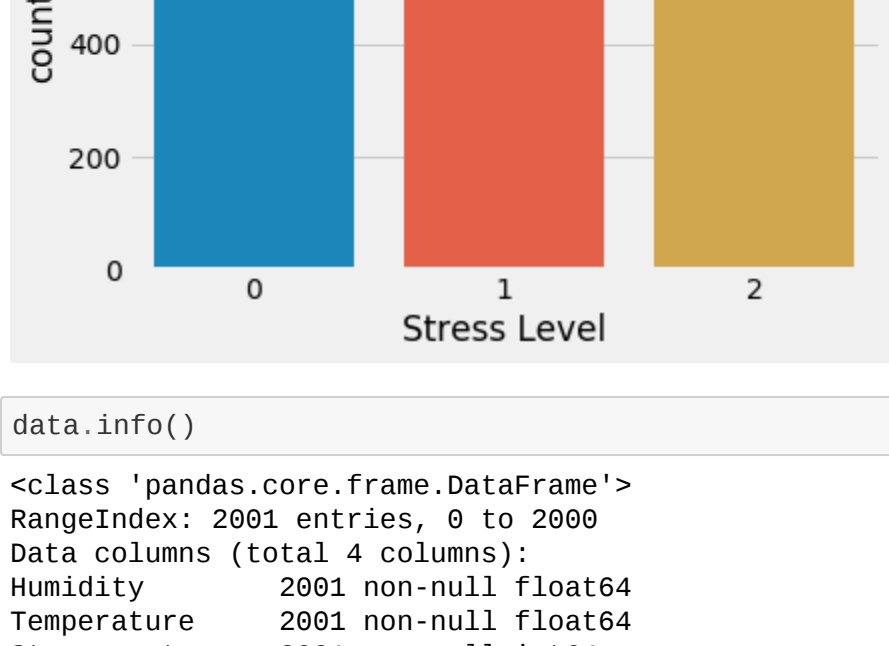
In [5]: `data=pd.read_csv('E:\ML Project\Stress-Lysis.csv')`

In [6]: `data.isnull().sum()`

Out[6]:

Humidity	0
Temperature	0
Step count	0
Stress Level	0
dtype:	int64

In [7]: `sns.countplot(data["Stress Level"])
plt.show()`



In [8]: `data.info()`

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2881 entries, 0 to 2880
Data columns (total 4 columns):
Humidity 2881 non-null float64
Temperature 2881 non-null float64
Step count 2881 non-null int64
Stress level 2881 non-null int64
dtypes: float64(2), int64(2)
memory usage: 62.6 KB

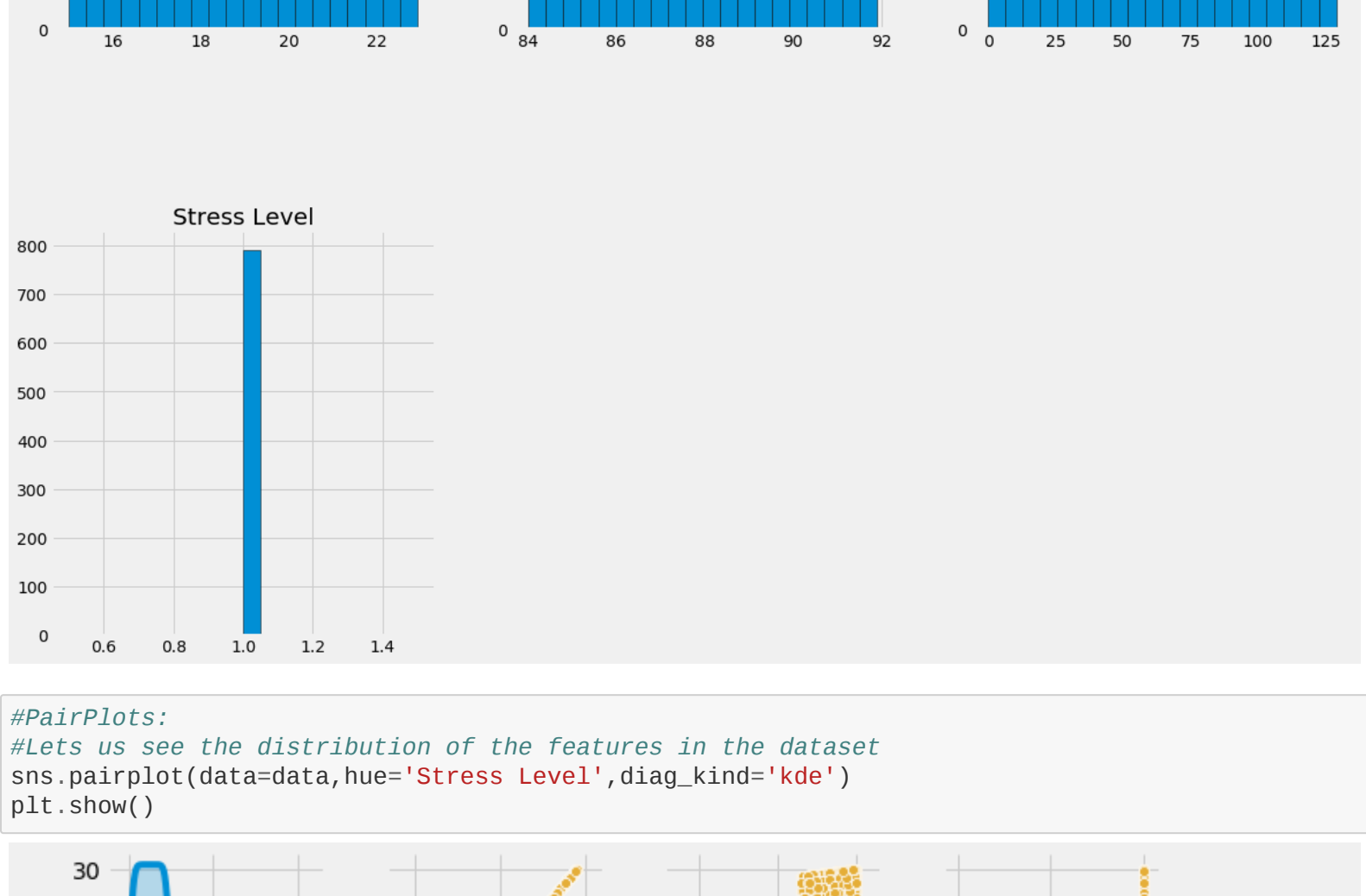
In [9]: `data.columns`

Out[9]: `Index(['Humidity', 'Temperature', 'Step count', 'Stress Level'], dtype='object')`

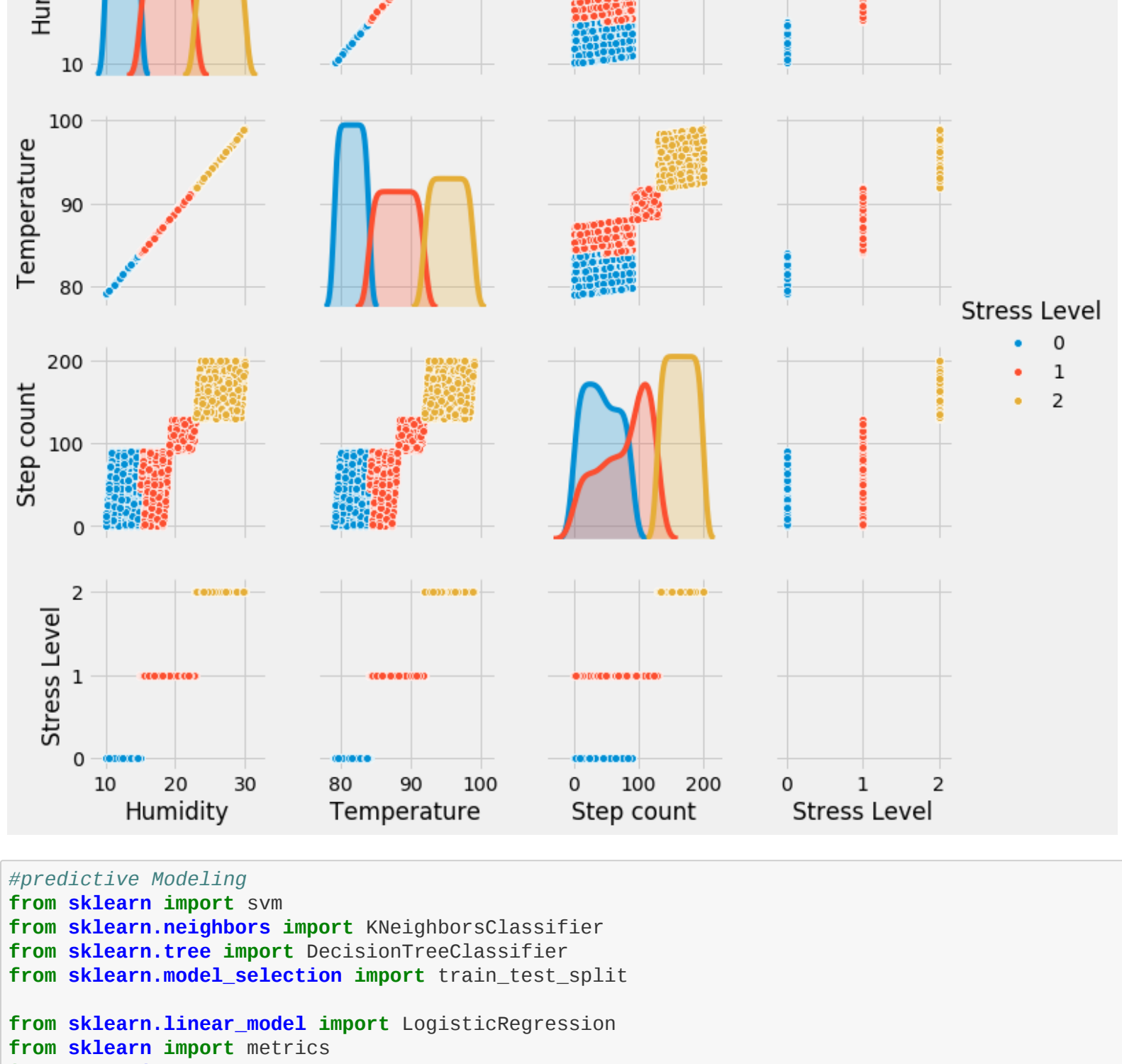
In [10]: `columns=data.columns[:8]
plt.subplots(figsize=(18,15))
length=len(columns)
for i,j in itertools.zip_longest(columns,range(length)):
 plt.subplot((length/2),3,i+1)
 plt.subplots_adjust(wspace=0.2,hspace=0.5)
 data[i].hist(bins=20,edgecolor='black')
 plt.title(i)
plt.show()`



In [11]: `# Analysis of human stress cases
data=data[data["Stress Level"]==1]
columns=data.columns[:8]
plt.subplots(figsize=(18,15))
length=len(columns)
for i,j in itertools.zip_longest(columns,range(length)):
 plt.subplot((length/2),3,i+1)
 plt.subplots_adjust(wspace=0.2,hspace=0.5)
 data[i].hist(bins=20,edgecolor='black')
 plt.title(i)
plt.show()`



In [16]: `#PairPlots:
#lets us see the distribution of the features in the dataset
sns.pairplot(data=data,hue='Stress Level',diag_kind='kde')
plt.show()`



In [17]: `#predictive Modeling
from sklearn import svm
from sklearn.neighbors import KNeighborsClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split

from sklearn.linear_model import LogisticRegression
from sklearn import metrics
import warnings
warnings.filterwarnings('ignore')`

In [18]: `outcome=data['Stress Level']
data=data[data.columns[:8]]
train,test=train_test_split(data,test_size=0.25,random_state=0,stratify=data['Stress Level'])
#y= stratify the outcome
train_X=train[train.columns[:8]]
test_X=test[test.columns[:8]]
train_Y=train['Stress Level']
test_Y=test['Stress Level']`

In [19]: `train_X.head(2)`

Out[19]:

	Humidity	Temperature	Step count	Stress Level
1337	15.59	64.59	13	1
1448	11.85	80.85	3	0

In [20]: `train_Y.head(2)`

Out[20]:

1337	1
1448	0

Name: Stress Level, dtype: int64

In [21]: `#SVM
types=['rbf','linear']
for i in types:
 model=svm.SVC(kernel=i)
 model.fit(train_X,train_Y)
 prediction=model.predict(test_X)
 print('Accuracy for SVM kernel= %i'%i,metrics.accuracy_score(prediction,test_Y))`

Accuracy for SVM kernel= rbf is 1.0
Accuracy for SVM kernel= linear is 1.0

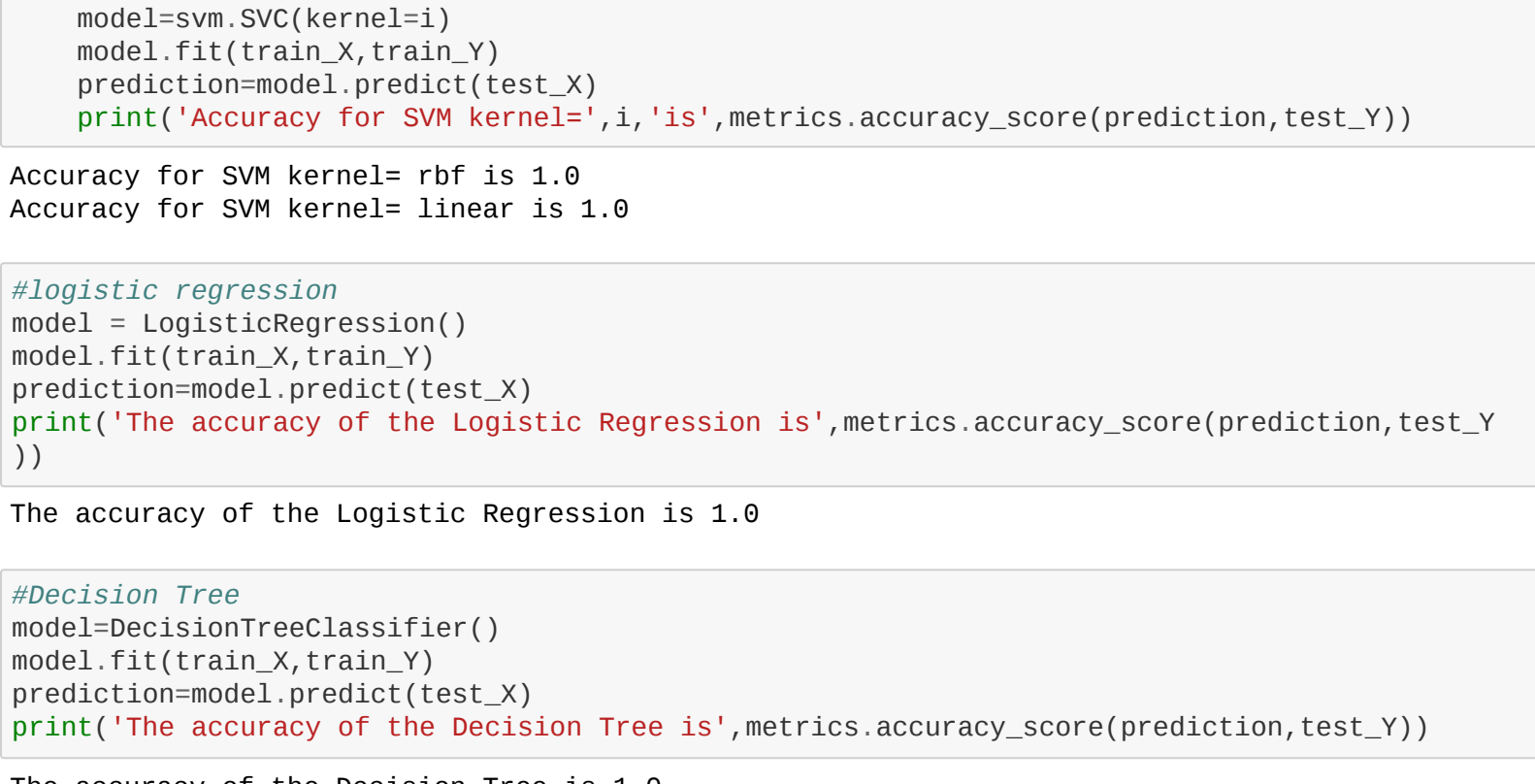
In [22]: `#logistic regression
model = LogisticRegression()
model.fit(train_X,train_Y)
prediction=model.predict(test_X)
print('The accuracy of the Logistic Regression is',metrics.accuracy_score(prediction,test_Y))`

The accuracy of the Logistic Regression is 1.0

In [23]: `#Decision Tree
model=DecisionTreeClassifier()
model.fit(train_X,train_Y)
prediction=model.predict(test_X)
print('The accuracy of the Decision Tree is',metrics.accuracy_score(prediction,test_Y))`

The accuracy of the Decision Tree is 1.0

In [24]: `#K-Nearest Neighbours
a_index=list(range(1,11))
a=pd.Series(a_index)
x=[0,1,2,3,4,5,6,7,8,9,10]
for i in list(range(1,11)):
 model=KNeighborsClassifier(n_neighbors=i)
 model.fit(train_X,train_Y)
 prediction=model.predict(test_X)
 a=a.append(pd.Series(metrics.accuracy_score(prediction,test_Y)))
plt.plot(a_index, a)
plt.xticks(x)
plt.show()
print('Accuracies for different values of n are:',a.values)`



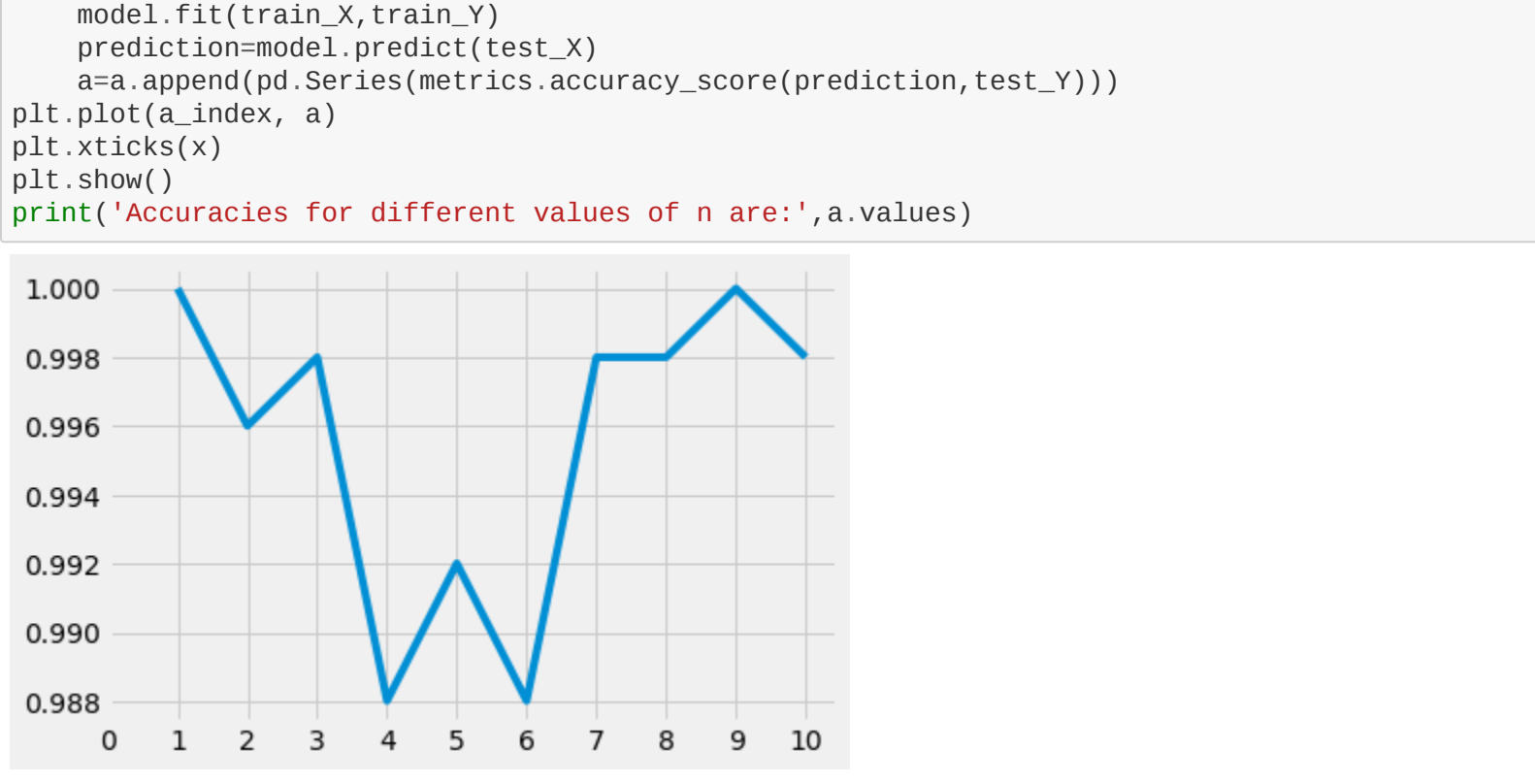
Accuracies for different values of n are: [1. 0.99600798 0.99800399 0.98802395 0.99201587 0.98802395 0.99800399 1. 0.99800399]

In [25]: `abc=[]
classifiers=['Linear SVM','Radial SVM','Logistic Regression','KNN','Decision Tree']
models=[svm.SVC(kernel='linear'),svm.SVC(kernel='rbf'),LogisticRegression(),KNeighborsClassifier(n_neighbors=5),DecisionTreeClassifier()]
for i in models:
 model = i
 model.fit(train_X,train_Y)
 prediction=model.predict(test_X)
 abc.append(metrics.accuracy_score(prediction,test_Y))
models_dataframe=pd.DataFrame(abc,index=classifiers)
models_dataframe.columns=['Accuracy']
models_dataframe`

Out[25]:

	Accuracy
Linear SVM	1.000000
Radial SVM	1.000000
Logistic Regression	1.000000
KNN	0.998004
Decision Tree	1.000000

In [26]: `#Correlation Matrix
sns.heatmap(data[data.columns[:8]].corr(),annot=True,cmap="YlOrRd")
fig=plt.gcf()
fig.set_size_inches(10,8)
plt.show()`



In [27]: `from sklearn.ensemble import RandomForestClassifier
model=RandomForestClassifier(n_estimators=100,random_state=0)
X=data[data.columns[:8]]
Y=data['Stress Level']
model.fit(X,Y)
pd.Series(model.feature_importances_,index=X.columns.sort_values(ascending=False))`

Out[27]:

Stress Level	0.315796
Temperature	0.278227
Humidity	0.250384
Step count	0.148584
dtype:	float64

In [28]: `data2=data[['Stress Level','Temperature','Humidity','Step count']]
from sklearn.preprocessing import StandardScaler #Standardisation
features=data2[data2.columns[:4]]
features_standard=StandardScaler().fit_transform(features)# Gaussian Standardisation
x=pd.DataFrame(features_standard,columns=['Stress Level','Temperature','Humidity','Step count'])
x['Stress Level']=data2['Stress Level']
outcome=x['Stress Level']
train,test=train_test_split(x,test_size=0.25,random_state=0,stratify=x['Stress Level'])
train_X=train[train.columns[:4]]
test_X=test[test.columns[:4]]
train_Y=train['Stress Level']
test_Y=test['Stress Level']`

In [29]: `abc=[]
classifiers=['Linear SVM','Radial SVM','Logistic Regression','KNN','Decision Tree']
models=[svm.SVC(kernel='linear'),svm.SVC(kernel='rbf'),LogisticRegression(),KNeighborsClassifier(n_neighbors=5),DecisionTreeClassifier()]
for i in models:
 model = i
 model.fit(train_X,train_Y)
 prediction=model.predict(test_X)
 abc.append(metrics.accuracy_score(prediction,test_Y))
new_models_dataframe=pd.DataFrame(abc,index=classifiers)
new_models_dataframe.columns=['New Accuracy']`

In [30]: `new_models_dataframe=new_models_dataframe.merge(models_dataframe,left_index=True,right_index=True,how='left')
new_models_dataframe['Increase']=new_models_dataframe['New Accuracy']-new_models_dataframe['Accuracy']
new_models_dataframe`

Out[30]:

	New Accuracy	Accuracy	Increase
Linear SVM	1.0	1.000000	0.000000
Radial SVM	1.0	1.000000	0.000000
Logistic Regression	1.0	1.000000	0.000000
KNN	1.0	0.998004	0.001996
Decision Tree	1.0	1.000000	0.000000

In [31]: `#scale dataset
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
train_X1=sc.fit_transform(train_X)
test_X1=sc.transform(test_X)`

In [32]: `from sklearn.linear_model import LogisticRegression
clf = LogisticRegression(C=1.0, penalty='l2', random_state=0)
clf.fit(train_X1, train_Y1)`

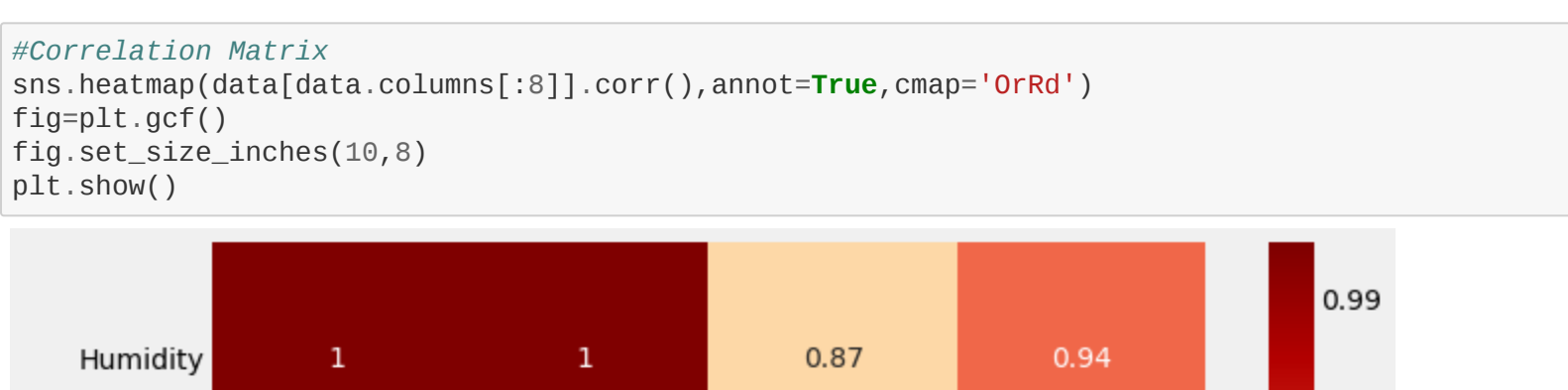
Out[32]: `LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
intercept_scaling=1, max_iter=1000, multi_class='warn',
n_jobs=None, penalty='l2', random_state=0, solver='warn',
tol=0.0001, verbose=0, warm_start=False)`

In [33]: `from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
pred = clf.predict(test_X1)
cm = confusion_matrix(test_Y1,pred)
print(cm)
print(accuracy_score(test_Y1,pred))`

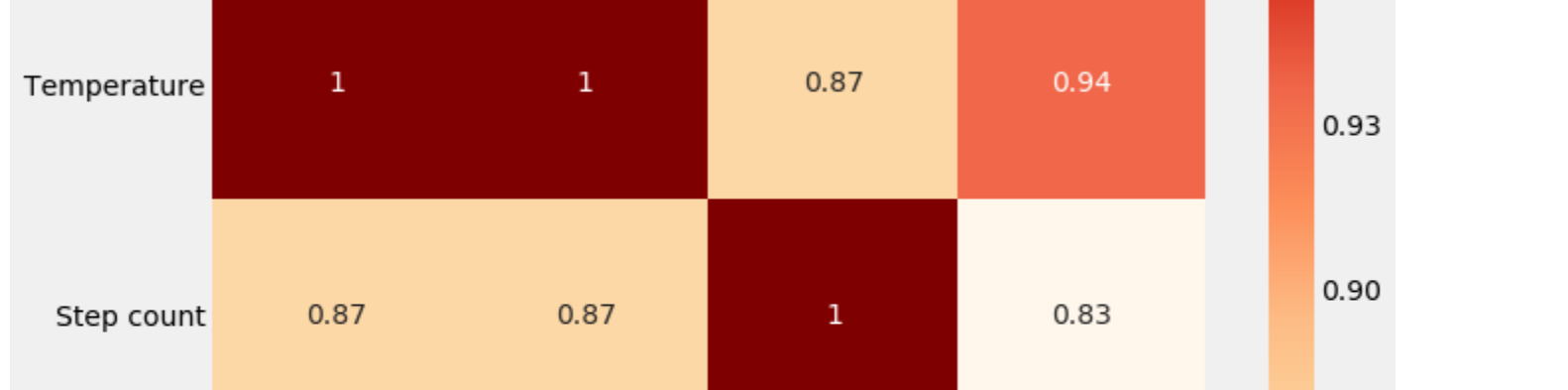
[[125 0 0]
[0 198 0]
[0 0 178]]
1.0

In [34]: `plt.scatter(data["Stress Level"], data["Temperature"])`

Out[34]: `<matplotlib.collections.PathCollection at 0x9c8410>`



In [35]: `X1 = data[data["Stress Level"] == 0]["Step count"]
X2 = data[data["Stress Level"] == 0]["Temperature"]
plt.scatter(X1, X2)
print(min(X1), max(X1))`

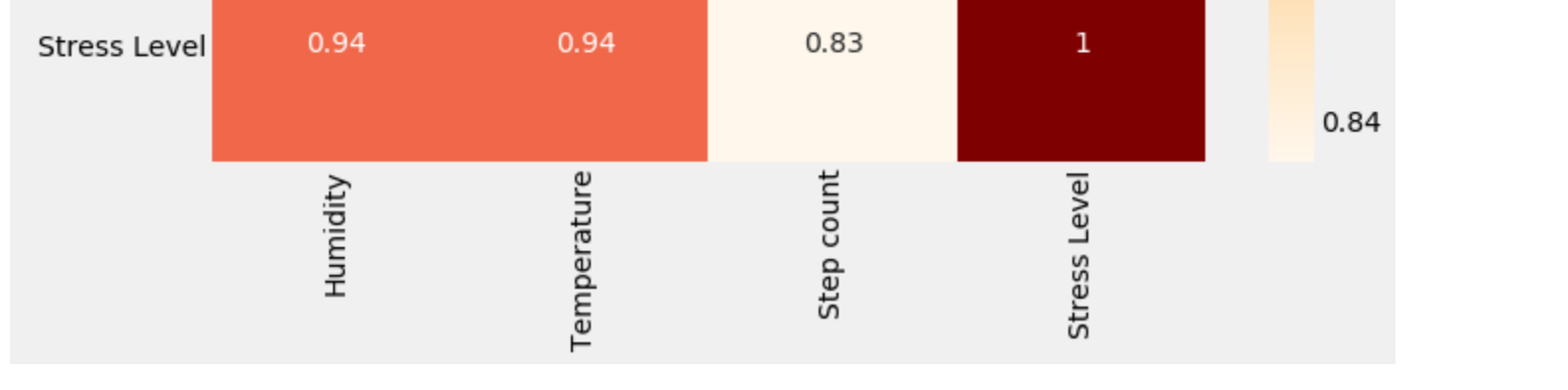


In [36]: `data_gen = []
humidity, temp, steps = 10., 79., 0.
for i in range(95):
 data_gen.append([humidity, temp, steps, 0])
 steps = steps % 90
 steps = 1
 temp += 0.01

data_gen = np.array(data_gen)
plt.scatter(data_gen[:, 2], data_gen[:, 1], alpha=0.4)
data_gen = []

for i in range(400):
 data_gen.append([humidity, temp, steps, 1])
 steps = steps % 90
 steps = 1
 temp += 0.01

data_gen = np.array(data_gen)
plt.scatter(data_gen[:, 2], data_gen[:, 1], alpha=0.4)
print(temp)`



In [37]: `from sklearn.model_selection import KFold #for K-fold cross validation`

In [38]: `kfold = KFold(n_splits=10, random_state=None) # k=10, split the data into 10 equal parts`

In [41]: `xyz=[]
accuracy=[]
classifiers=['Linear SVM','Radial SVM','Logistic Regression','KNN','Decision Tree']
models=[svm.SVC(kernel='linear'),svm.SVC(kernel='rbf'),LogisticRegression(),KNeighborsClassifier(n_neighbors=5),DecisionTreeClassifier()]
for i in models:
 model = i
 cv_result = cross_val_score(model,x[:,columns[:4]],x['Stress Level'], cv = kfold,scoring = "accuracy")
 cv_result=cv_result
 xyz.append(cv_result.mean())
 accuracy.append(cv_result)
new_models_dataframe=pd.DataFrame(abc,index=classifiers)
new_models_dataframe2.columns=['CV Mean']
new_models_dataframe2`

Out[41]:

	CV Mean
Linear SVM	1.0
Radial SVM	1.0
Logistic Regression	1.0
KNN	1.0
Decision Tree	1.0

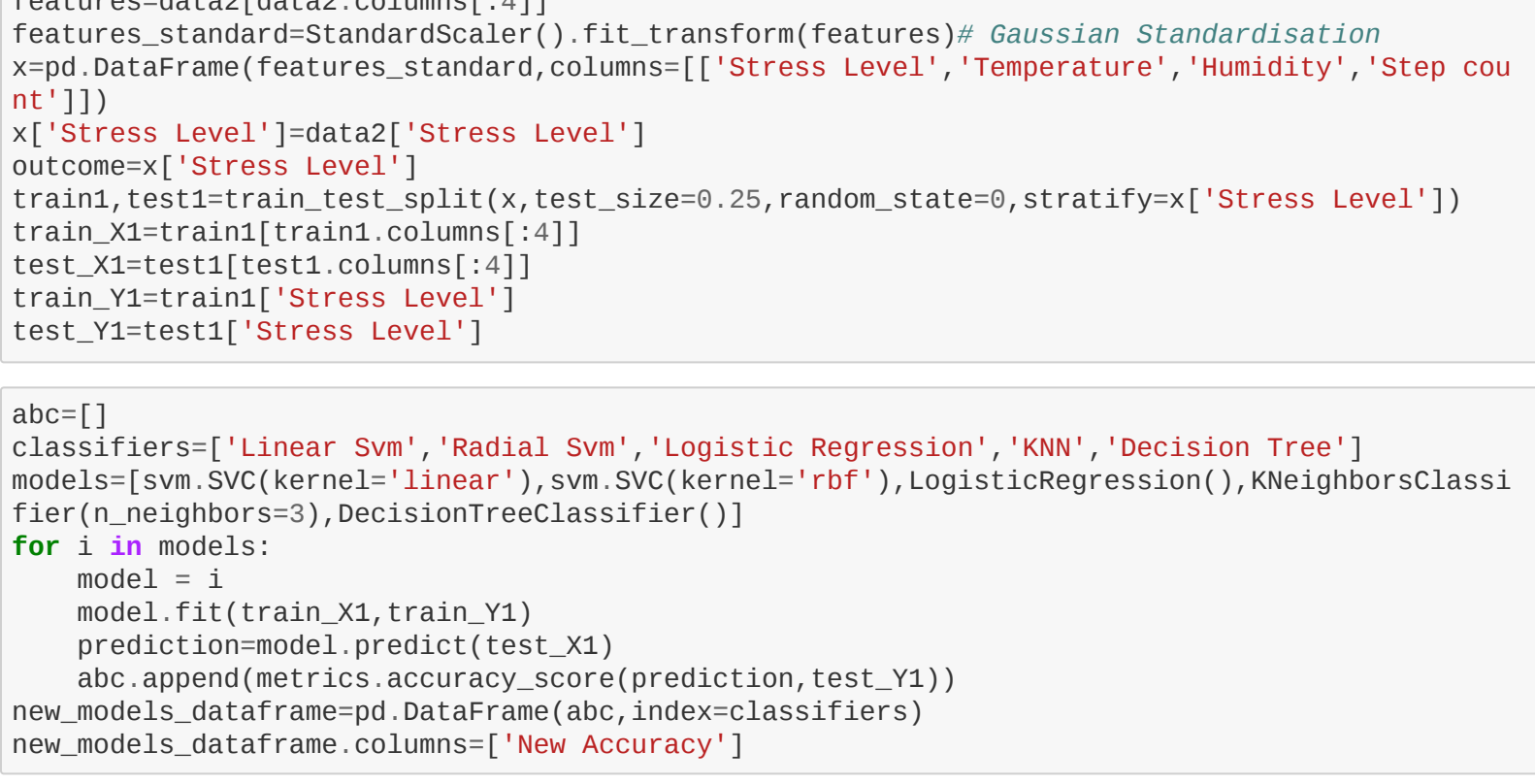
In [42]: `box=pd.DataFrame(accuracy,index=classifiers))
box1=box.T
box1`

Out[42]:

	Linear SVM	Radial SVM	Logistic Regression	KNN	Decision Tree
0	1.0	1.0	1.0	1.0	1.0
1	1.0	1.0	1.0	1.0	1.0
2	1.0	1.0	1.0	1.0	1.0
3	1.0	1.0	1.0	1.0	1.0
4	1.0	1.0	1.0	1.0	1.0
5	1.0	1.0	1.0	1.0	1.0
6	1.0	1.0	1.0	1.0	1.0
7	1.0	1.0	1.0	1.0	1.0
8	1.0	1.0	1.0	1.0	1.0
9	1.0	1.0	1.0	1.0	1.0

In [43]: `box1.boxplot()`

Out[43]: `fig=plt.gcf()
fig.set_size_inches(10,8)
plt.show()`



In [44]: `linear_svc=svm.SVC(kernel='linear',C=0.1,gamma=10,probability=True)
radial_svc=svm.SVC(kernel='rbf',C=0.1,gamma=10,probability=True)
lr=LogisticRegression(C=0.1)`

In [45]: `from sklearn.ensemble import VotingClassifier #for Voting Classifier`

In [46]: `# Linear and Radial SVM
ensemble_lr_rf=VotingClassifier(estimators=[('Linear_svm', linear_svc), ('Radial_svm', radial_svc)],
voting='soft', weights=[2,1]).fit(train_X1,train_Y1)
print('The accuracy for Linear and Radial SVM is:',ensemble_lr_rf.score(test_X1,test_Y1))`

The accuracy for Linear and Radial SVM is: 1.0

In [47]: `# Linear SVM with Logistic Regression
ensemble_lr_lr=VotingClassifier(estimators=[('Linear_svm', linear_svc), ('Logistic Regression', lr)],
voting='soft', weights=[2,1]).fit(train_X1,train_Y1)
print('The accuracy for Linear SVM and Logistic Regression is:',ensemble_lr_lr.score(test_X1,test_Y1))`

The accuracy for Linear SVM and Logistic Regression is: 1.0

In [48]: `# Logistic Regression with Radial SVM
ensemble_lr_rf=VotingClassifier(estimators=[('Radial_svm', radial_svc), ('Logistic Regression', lr)],
voting='soft', weights=[2,1]).fit(train_X1,train_Y1)
print('The accuracy for Radial SVM and Logistic Regression is:',ensemble_lr_rf.score(test_X1,test_Y1))`

The accuracy for Radial SVM and Logistic Regression is: 1.0

In [49]: `ensemble_rad_lr_lr=VotingClassifier(estimators=[('Radial_svm', radial_svc), ('Logistic Regression', lr)],
voting='soft', weights=[2,1]).fit(train_X1,train_Y1)
print('The ensemble model with all the 3 classifiers is:',ensemble_rad_lr_lr.score(test_X1,test_Y1))`

The ensemble model with all the 3 classifiers is: 1.0

In []: