[2]:	incompared and small(UE) Data asiana torining) D. AND. DYTHON KDNO) etct. and all ACCTONNENT Concess call incompared all incompared and all incompared all incompared all incompared and all incompared all incompared all incompared and all incompared all incompared and all incompa
	<pre>insurance=pd.read_excel("E:\Data science training\R AND PYTHON KPMG\stat and ml\ASSIGNMENT\Cross sell insurance.xlsx") insurance.head() id Gender Age Driving_License Previously_Insured Vehicle_Age Vehicle_Damage Annual_Premium Vintage Score Response</pre>
	0 1 Male 44 1 0 > 2 Years Yes 40454 217 1 1 2 Male 76 1 0 1-2 Year No 33536 183 0
	2 3 Male 47 1 0 > 2 Years Yes 38294 27 1 3 4 Male 21 1 1 <1 Year No 28619 203 0 4 5 Female 29 1 1 <1 Year No 27496 39 0
[3]:	insurance.info()
	cclass 'pandas.core.frame.DataFrame'> RangeIndex: 3999 entries, 0 to 3998 Data columns (total 10 columns):
	# Column Non-Null Count Dtype 0 id 3999 non-null int64
	1 Gender 3999 non-null object 2 Age 3999 non-null int64 3 Driving_License 3999 non-null int64 4 Previously_Insured 3999 non-null int64
	5 Vehicle_Age 3999 non-null object 6 Vehicle_Damage 3999 non-null object 7 Annual_Premium 3999 non-null int64 8 Vintage Score 3999 non-null int64
	8 Vintage Score 3999 non-null int64 9 Response 3999 non-null int64 Htypes: int64(7), object(3) nemory usage: 312.5+ KB
[4]:	# Checking the null value. insurance.isnull().sum()
[4]:	Ld 0 Sender 0 Age 0
	Priving_License 0 Previously_Insured 0 Vehicle_Age 0
	/ehicle_Damage 0 Annual_Premium 0 /intage Score 0 Response 0
[5]:	Htype: int64 # create a seperate object column.
[6]:	<pre>cat_col = ['Gender','Vehicle_Age','Vehicle_Damage'] ## creating Dummy of the Data</pre>
	insurance_dummy = pd.get_dummies(insurance,columns=cat_col,drop_first= True)
[7]: :[7]:	insurance_dummy.shape 3999, 11)
[8]:	insurance_dummy.head()
[8]:	id Age Driving_License Previously_Insured Annual_Premium Vintage Score Response Gender_Male Vehicle_Age_<1 Year Vehicle_Age_> 2 Years Vehicle_Damage_Yes 1 44 1 0 40454 217 1 1 0 1 1
	1 2 76 1 0 33536 183 0 1 0 0 0 0 2 3 47 1 0 38294 27 1 1 0 1 1 1
	3 4 21 1 1 28619 203 0 1 1 0 0 4 5 29 1 1 27496 39 0 0 1 0 0
[9]:	# here in this data i don't need "id" column so exclude from the Data #Identifying the input and output variables Y = insurance_dummy[['Response']]
[<u>1</u> ቦ ^{ገ -}	X = insurance_dummy.drop(columns=['Response','id'])
[10]: [10]:	X.head() Age Driving_License Previously_Insured Annual_Premium Vintage Score Gender_Male Vehicle_Age_< 1 Year Vehicle_Age_> 2 Years Vehicle_Damage_Yes
	0 44 1 0 40454 217 1 0 1 1 1 76 1 0 33536 183 1 0 0 0 0
	2 47 1 0 38294 27 1 0 1 1 3 21 1 1 28619 203 1 1 0 0 4 29 1 1 27496 39 0 1 0 0
[11]:	##Spillting the data into train and test
	<pre>from sklearn.model_selection import train_test_split x_train,x_test,y_train,y_test = train_test_split(X,Y,test_size=0.2,random_state=42)</pre>
[11]:	len(x_train),len(x_test),len(y_train),len(y_test) 3199, 800, 3199, 800)
	uilding the Random Forest model:
-1.	<pre>from sklearn.ensemble import RandomForestClassifier model = RandomForestClassifier(n_estimators=1000, random_state=42) rf = model.fit(x_train, y_train) print('The model has been built successfully!! yeah')</pre>
	::\Users\Pratik\AppData\Local\Temp/ipykernel_8264/3858015865.py:3: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Pleasinge the shape of y to (n_samples,), for example using ravel().
[13]:	rf = model.fit(x_train,y_train) The model has been built successfully!! yeah
	# Predicting the model on test data y_test Response
[13]:	Response 1760 0 3326 1
	1.770 0 0 2.176 0
	2099 0
	2510 0 2752 0 1869 1
	1 423 0 2990 0
	00 rows × 1 columns
[14]:	<pre>y_test['Prediction'] = model.predict(x_test) y_test</pre>
[14]:	Response Prediction 1760 0 0
	3326 1 0 1770 0 0
	2099 0 0
	2752 0 0 0 1869 1 0 423 0 1
	2990 0 0 00 rows × 2 columns
[15]:	from sklearn.metrics import confusion_matrix, accuracy_score
	print(confusion_matrix(y_test['Response'],y_test['Prediction'])) [672 32]
[16]:	<pre>[81 15]] print(accuracy_score(y_test['Response'],y_test['Prediction']))</pre>
[17]:	0.85875 ## My Model Accuracy Score is 0.85875 or 85.87% correctly predicted
	What is the accuracy of Support Vector Machine? How does changing the model affect the accuracy
[18]:	#Identifying the input and output variables Y1 = insurance_dummy[['Response']] X1 = insurance_dummy.drop(columns=['Response','id'])
[19]:	X1.head()
[19]:	Age Driving_License Previously_Insured Annual_Premium Vintage Score Gender_Male Vehicle_Age_<1 Year Vehicle_Age_>2 Years Vehicle_Damage_Yes 1 1 0 40454 217 1 0 1 1
	1 76 1 0 33536 183 1 0 0 0 0 2 47 1 0 38294 27 1 0 1 1 3 21 1 1 28619 203 1 1 0 0
	3 21 1 1 28619 203 1 1 0 0 4 29 1 1 27496 39 0 1 0 0
[20]:	#Spillting the data into train and test from sklearn.model_selection import train_test_split
	<pre>from sklearn.model_selection import train_test_split x1_train,x1_test,y1_train,y1_test = train_test_split(X1,Y1,test_size=0.2,random_state=42) len(x1_train),len(x1_test),len(y1_train),len(y1_test)</pre>
[20]: [21]:	3199, 800, 3199, 800)
[21]:	<pre>from sklearn.svm import SVC model1 = SVC(random_state=42)</pre>
[22]:	<pre># fitting the Model svc_model = model1.fit(x1_train,y1_train) print('The model has been built successfully!! yeah')</pre>
	:\Users\Pratik\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was passed when a 1d array was expectlease change the shape of y to (n_samples,), for example using ravel().
[23]:	y = column_or_1d(y, warn=True) The model has been built successfully!! yeah
	#PRediction of test data y1_test Response
[23]:	Response 1760 0 0 1326 1
	3326 1 1770 0 3176 0
	2099 0
	 2510 0 2752 0
	 2510 0
[24]: [25]:	2510 0 2752 0 2869 1 423 0 2990 0 200 rows × 1 columns y1_test['Prediction'] = svc_model.predict(x1_test)
[24]:	
[24]: [25]:	
[24]: [25]:	Seponse Prediction
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[24]: [25]: [25]:	1510 0 15752 0 15899 1 1423 0 15990 0 150 TOVEX * 1 Columns 170
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