[2]:	<pre>findspark.init()  from pyspark.sql import Sp spark=SparkSession.builder</pre>	arkSession .appName("BayesTheoremApp")	).getOrCreate()				
[3]: [ t[3]: <b>\$</b>	spark=SparkSession.builder spark  spark  SparkSession - in-memory	payes meoremApp"	, gover of Edte()				
	SparkContext Spark UI						
ı	Version Master AppName	v3.2.1 local[*] BayesTheoremApp					
[46]:	<pre>df=spark.read.option("head df.show()  +</pre>						
	+++	Harris B  male 22.0  Bradley female 38.0  Ba Heikk female 26.0  Bi Heat female 35.0		1  1  1  0  1	0  7.25  0 71.2833  0 7.925  0  53.1		
	0  3  Mr. 3   0  1 Mr. Timoth   0  3 Master. Go   1  3 Mrs. Oscar	mm Henry  male 35.0   James Moran  male 27.0   Dy J McC  male 54.0   Dista Leo  male  2.0   TW (Eli female 27.0   Dlas (Ad female 14.0		0  0  0  3  0  1	0  8.05  0  8.4583  0 51.8625  1  21.075  2 11.1333  0 30.0708		
	1  3 Miss. Marg   1  1 Miss. Eliz   0  3 Mr. Willia   0  3 Mr. Anders	guerite female  4.0  cabeth B female 58.0  m Henry  male 20.0  s Johan  male 39.0  la Amand female 14.0		1  0  0  1  0	1  16.7  0  26.55  0  8.05  5  31.275  0  7.8542		
	1  2 Mrs. (Mary   0  3  Master. E   1  2 Mr. Charle   0  3 Mrs. Juliu	D King female 55.0  Eugene Rice  male  2.0  es Eugen  male 23.0  us (Emel female 31.0  na Masse female 22.0		0  4  0  1  0	0  16.0  1  29.125  0  13.0  0  18.0  0  7.225		
[47]:	only showing top 20 rows  df.printSchema()	+		+	+		
	root   Survived: integer (null)   Pclass: integer (nulla)   Name: string (nullable)   Sex: string (nullable)	uble = true) e = true)					
	Age: double (nullable   SiblingsSpouses Aboard	= true) I: integer (nullable = true I: integer (nullable = true					
[48]:	<pre>from pyspark.ml.feature im from pyspark.sql.types imp from pyspark.sql.functions from pyspark.ml.feature im</pre>	ort * import * port StringIndexer					
[40]		tion import NaiveBayes import BinaryClassificatio	onEvaluator				
[49]:	df.groupBy("Survived").cou	nt().show()					
[50]:	1  342    0  545  ++						
	<pre>input_columns=df.columns input_columns=input_column dependent_var='Survived'  print(input_columns) print(dependent_var)</pre>	s[4:6]					
[51]:	['Age', 'SiblingsSpouses Ab Survived		cast(String)	)))			
			="label")	+			
	Survived Pclass  ++	Name  Sex  Age Sib Harris B  male 22.0  Bradley female 38.0  Ha Heikk female 26.0		ard ParentsChildrer +	Aboard  Fare label + 0  7.25  0 71.2833  0  7.925	_str label  + 0  0.0  1  1.0  1  1.0	
	0  3 Mr. Willia   0  3  Mr. 3   Mr. 3   0  1 Mr. Timoth   0  3 Master. Go	des Heat female 35.0   dm Henry  male 35.0   dames Moran  male 27.0   dy J McC  male 54.0   desta Leo  male  2.0		1  0  0  0  0  3	0  53.1  0  8.05  0  8.4583  0 51.8625  1  21.075	1  1.0  0  0.0  0  0.0  0  0.0  0  0.0	
	1  2 Mrs. Nicho   1  3 Miss. Marg   1  1 Miss. Eliz   0  3 Mr. Willia	W (Eli female 27.0  plas (Ad female 14.0  puerite female  4.0  cabeth B female 58.0  m Henry  male 20.0		0  1  1  0  0	2 11.1333  0 30.0708  1  16.7  0  26.55  0  8.05	1  1.0  1  1.0  1  1.0  1  1.0  0  0.0	
	0  3 Miss. Huld   1  2 Mrs. (Mary   0  3  Master. E   1  2 Mr. Charle   0  3 Mrs. Juliu	s Johan   male   39.0   la Amand   female   14.0   la Amand   female   55.0   la Sension   sensi		1  0  0  4  0  1	5  31.275  0  7.8542  0  16.0  1  29.125  0  13.0  0  18.0	0  0.0  0  0.0  1  1.0  0  0.0  1  1.0  0  0.0	
[52]:	1 3 Mrs. Fatin	na Masse female 22.0		0	0  7.225	1   1.0	
];	indexer=StringInde	s: column].dataType)=='String' xer(inputCol=column, output t(indexed).transform(indexe	tCol=column+"_nu	n")			
	<pre>new_col_name=colum     string_inputs.appe else:     numeric_inputs.app print('numeric_inputs',num</pre>	n+"_num" nd(new_col_name) end(column) eric_inputs)	,				
	<pre>print('string_inputs', stri numeric_inputs ['Age', 'Sik string_inputs []</pre>	ng_inputs)					
[53]:	<pre>for col in numeric_inputs:</pre>	uantile(col,[0.01,0.99],0.2					
	skew=indexed.agg(skewn skew=skew[0][0] if skew>1 : indexed=indexed.wi log(when(df[col]<	<pre>thColumn(col, \ [col][0],d[col][0])\</pre>	()				
	<pre>.otherwise(indexed     print(col+"positiv if skew&lt;-1 :     indexed=indexed.wi</pre>		,skew,")")				
	.when(indexed[col] .otherwise(indexed	<pre>[col][0],d[col][0])\ &gt;d[col][1],d[col][1])\ [col]) +1).alias(col)) e(left) skewness.(skew=)",s</pre>	skew,")")				
	SiblingsSpouses Aboardposit 3.6805221729276023  # Negative value minimums=df soloct([min(s)]	<pre>.ive(right) skewness.(skew= .alias(c) for c in df.colur</pre>		•			
	<pre>min_array=minimums.select( df_minimum=min_array.select) df_minimum=df_minimum[0][0] if df_minimum&lt;0:</pre>	<pre>array(numeric_inputs).alias t(array_min(min_array.mins)</pre>	s("mins")) )).collect()		me as it contain nega	tive value.")	
	else:	ues were found in your data		•		,	
[55]:		ts+string_inputs inputCols=features_list,out (indexed).select('features		s')			
	features	label  + 0.0					
		1.0					
[56]:	<pre># final Data scaler=MinMaxScaler(inputC print("Features scaled to</pre>	ol="features",outputCol="sorange:[%f, %f]" %(scaler.ge					
	scalerModel=scaler.fit(out scaled_data=scalerModel.tr final_data=scaled_data.sel	put)	s')				
	Features scaled to range:[6 ++	+ 					
	1.0 [472.229203317416 1.0 [321.437547122392 1.0 [434.531289268660 0.0 [434.531289268660 0.0 [334.003518471977	 					
	0.0 [673.284744910781 0.0 [19.8542347323448 1.0 [334.003518471977 1.0 [170.645890927368 1.0 [44.9861774315154	 					
	1.0 [723.548630309122   0.0 [246.041719024880   0.0 [484.795174667001   0.0 [170.645890927368   1.0 [685.850716260367   0.0 [19.8542347323448	 					
	0.0 [19.8542347323448   1.0 [283.739633073636   0.0 [384.267403870319   1.0 [271.173661724051 ++	İ İ					
[57]:	# min max Scalar	ol="features",outputCol="so	caler")				
[58]: [59]:	<pre># split train, test=final_data.rand</pre>	omSplit([0.70,0.30])					
[59]: [60]:	<pre>nbclassifier=NaiveBayes() nbModel=nbclassifier.fit(t</pre>	rain)					
[61]:	<pre>predictions.printSchema()</pre>	,'rawPrediction','probabil		').show()			
	label  rawPrediction   0.0 [-992.47852990171   0.0 [-859.01178838804	probability pred. 	iction  + 1.0  1.0				
	0.0 [-996.24369131152   0.0 [-862.77694979784   0.0 [-1003.7740141311   0.0 [-870.30727261746   0.0 [-1011.3043369507   0.0 [-451.55636958911	[5.62684350989519   [1.20484759321544   [1.26863537389421   [2.71646487807285	1.0  1.0  1.0  1.0  1.0  1.0				
	0.0 [-881.60275684689   0.0 [-1134.7264501392   0.0 [-36.570627970569   0.0 [-51.631273609803   0.0 [-1041.4256282292	[4.29481821749465   [6.62687113676345   [0.98695150838391   [0.99740586571205	1.0  1.0  0.0  0.0  1.0				
	0.0 [-1402.9107890707   0.0 [-59.161596429420   0.0 [-485.44282227738   0.0 [-62.926757839228   0.0 [-62.926757839228	[0.99884774699192   [9.80826236515007   [0.99923232209696   [0.99923232209696	1.0  0.0  1.0  0.0  0.0				
	0.0 [-62.926757839228   0.0 [-62.926757839228 ++	[0.99923232209696	0.0  0.0  +				
	label: double (nullable)   features: vector (nulle)   rawPrediction: vector   probability: vector (nulle)   prediction: double (nulle)	able = true) (nullable = true) nullable = true)					
	evaluator=BinaryClassifica	e(predictions) ",accuracy)					
[62]:	<pre>accuracy=evaluator.evaluat print("Accuracy of Model : print("test Error of Model</pre>	, (2 accuracy))					
[62]:							
[62]:	<pre>print("Accuracy of Model : print("test Error of Model  Accuracy of Model : 0.47859</pre>						
[62]:	<pre>print("Accuracy of Model : print("test Error of Model  Accuracy of Model : 0.47859</pre>						