



Institute of
Business Administration
Karachi

Leadership and Ideas for Tomorrow

**INTRODUCTION TO DATA MINING
FINAL PROJECT
AIRLINE PASSENGER SATISFACTION - CLASSIFICATION**

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Introduction:

Recently, there has been a surge in interest in the field of data mining, where the goal is to predict correct and useful knowledge for users. We used a summary of our study and exploration of the Airline passenger satisfaction Data in this project to come up with useful, important, and intriguing data properties. The CRISP-DM data mining methodology was heavily used in the development of our project model to predict as accurate as possible statistics for our users.

Problem Description:

Passenger satisfaction is a high concern for airlines' management as their prime concern is whether the customer will choose their airlines for their next trip or not. Airline passenger satisfaction is affected by many factors, but at its root, this type of customer satisfaction is no different from that of any other business. Therefore, various data mining techniques are utilized in this project to build a classification model for Airlines management to determine if the customer is satisfied or not depending on various factors.

Data Collection:

The dataset, primarily known as "Airline Passenger Satisfaction" was found on Kaggle, <https://www.kaggle.com/teejmahal20/airline-passenger-satisfaction>, while we were searching for a classification data set for our Data Mining project. This data set was collected by TJ Klein in 2019 through a survey.

Data Description:

The data set has two csv files named Test and Train. The test.csv file contains 26k records while the train.csv file contains 104k records. There are 25 attributes in total. The following are the given attributes:

Gender: Gender of the passengers (Female, Male)

Customer Type: The customer type (Loyal customer, disloyal customer)

Age: The actual age of the passengers

Type of Travel: Purpose of the flight of the passengers (Personal Travel, Business Travel)

Class: Travel class in the plane of the passengers (Business, Eco, Eco Plus)

Flight distance: The flight distance of this journey

Inflight wifi service: Satisfaction level of the inflight wifi service (0:Not Applicable;1-5)

Departure/Arrival time convenient: Satisfaction level of Departure/Arrival time convenient

Ease of Online booking: Satisfaction level of online booking

Gate location: Satisfaction level of Gate location

Food and drink: Satisfaction level of Food and drink

Online boarding: Satisfaction level of online boarding

Seat comfort: Satisfaction level of Seat comfort

Inflight entertainment: Satisfaction level of inflight entertainment

On-board service: Satisfaction level of On-board service

Leg room service: Satisfaction level of Leg room service

Baggage handling: Satisfaction level of baggage handling

Check-in service: Satisfaction level of Check-in service

Inflight service: Satisfaction level of inflight service

Cleanliness: Satisfaction level of Cleanliness

Departure Delay in Minutes: Minutes delayed when departure

Arrival Delay in Minutes: Minutes delayed when Arrival

Satisfaction: Airline satisfaction level(Satisfaction, neutral or dissatisfaction)

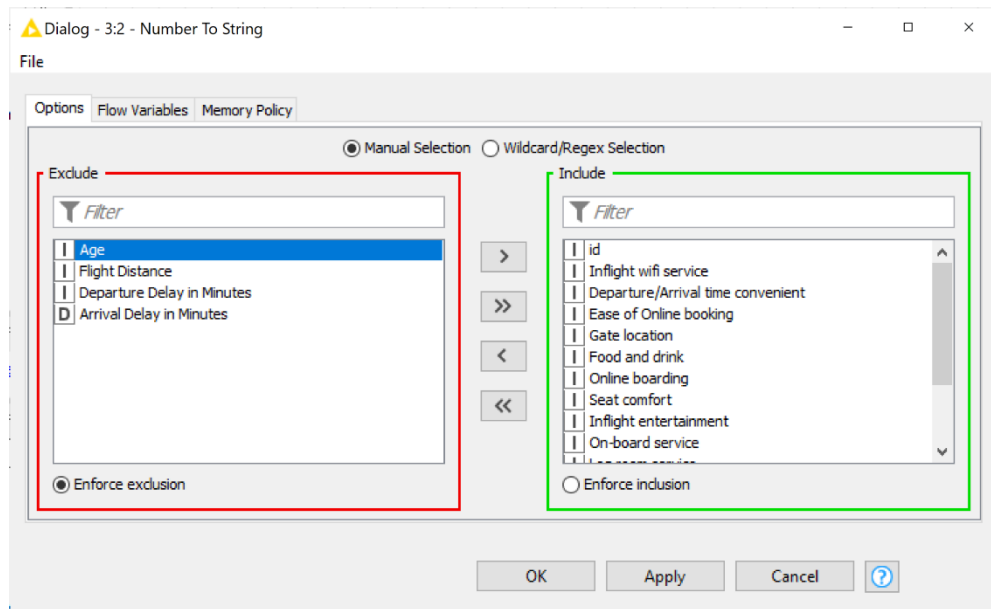
Attempt 1:

Data Preprocessing:

The steps taken to prepare the data are as follow:

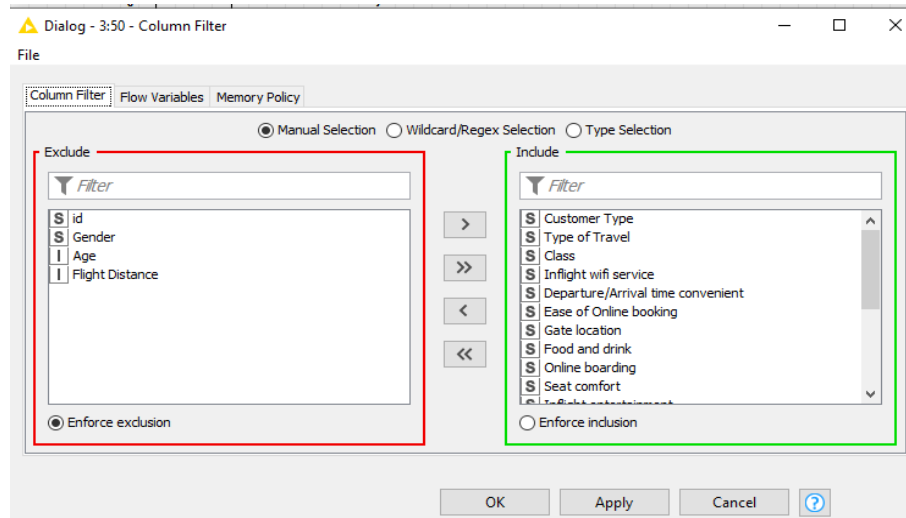
1. Number to String:

All the attributes except the continuous attributes such as Age, Flight Distance, Departure Delay in Minutes, Arrival Delay in Minutes are converted to String data type.



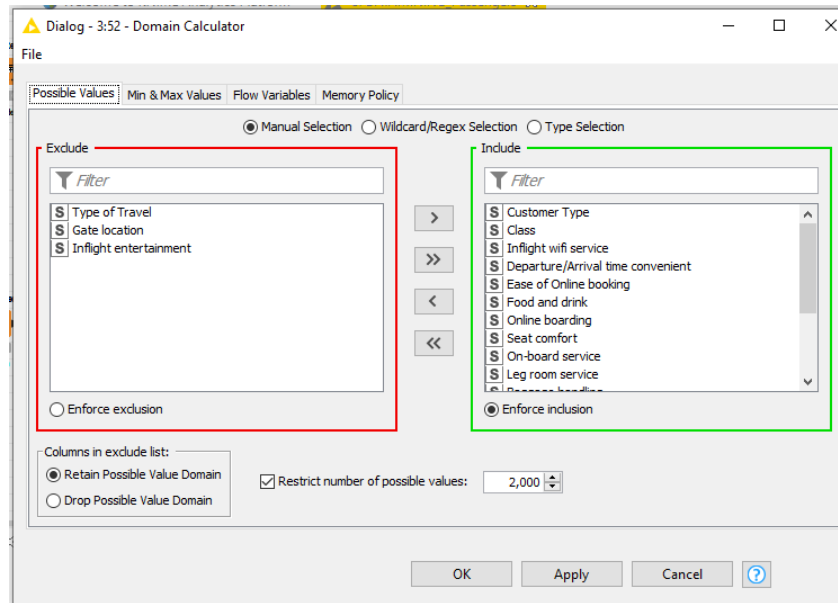
2. Column Filter:

The 4 columns ID, Gender, Age, and Flight Distance are filtered out using Column Filter as including these columns did not have much effect on our prediction/ accuracy.



3. Domain Calculator:

After applying the previous techniques on the dataset, the domain information of the data has changed. So we have used the Domain Calculator for this purpose.



Model Building:

After preprocessing the entire data, we used four classifiers to build our model. They are as follow:

1. **Decision Tree:** We have used the following configuration on the Decision Tree Learner

Dialog - 3:3 - Decision Tree Learner

File

Options | PMMLSettings | Flow Variables

General

Class column: satisfaction

Quality measure:

Pruning method:

☒ Reduced Error Pruning

Min number records per node:

Number records to store for view:

☒ Average split point

Number threads:

☒ Skip nominal columns without domain information

Root split

☐ Force root split column

Root split column: Arrival Delay in Minutes

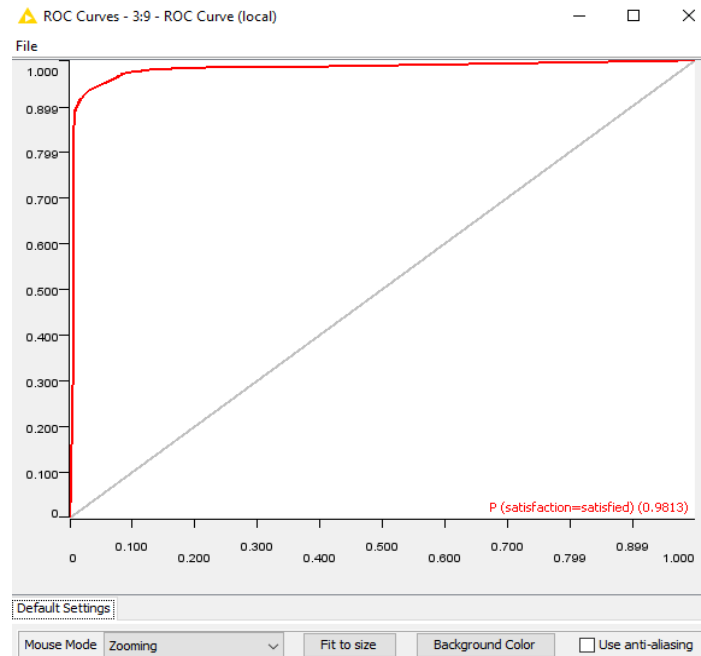
Binary nominal splits

☐ Binary nominal splits

Max #nominal:

☐ Filter invalid attribute values in child nodes

ROC:



Confusion Matrix - 3:56 - Scorer

File Hilite

There were missing values in the reference or in the prediction...

satisfactio...	satisfied	neutral or ...
satisfied	10561	830
neutral or di...	356	14206

Correct classified: 24,767 Wrong classified: 1,186

Accuracy: 95.43 % Error: 4.57 %

Cohen's kappa (κ) 0.907

2. Random Forest: We have used the following configuration on the Random Forest Learner

Dialog - 3:15 - Random Forest Learner

File

Options | Flow Variables | Memory Policy

Target Column: [S] satisfaction

Attribute Selection

☐ Use fingerprint attribute [no] <no valid fingerprint input>

☒ Use column attributes

☒ Manual Selection ☐ Wildcard/Regex Selection

Exclude

Filter

No columns in this list

☒ Enforce exclusion

Include

Filter

- [S] Customer Type
- [S] Type of Travel
- [S] Class
- [S] Inflight wifi service
- [S] Departure/Arrival time convenient
- [S] Ease of Online booking
- [S] Gate location
- [S] Food and drink

☐ Enforce inclusion

Misc Options

☐ Enable Highlighting (#patterns to store) 2,000

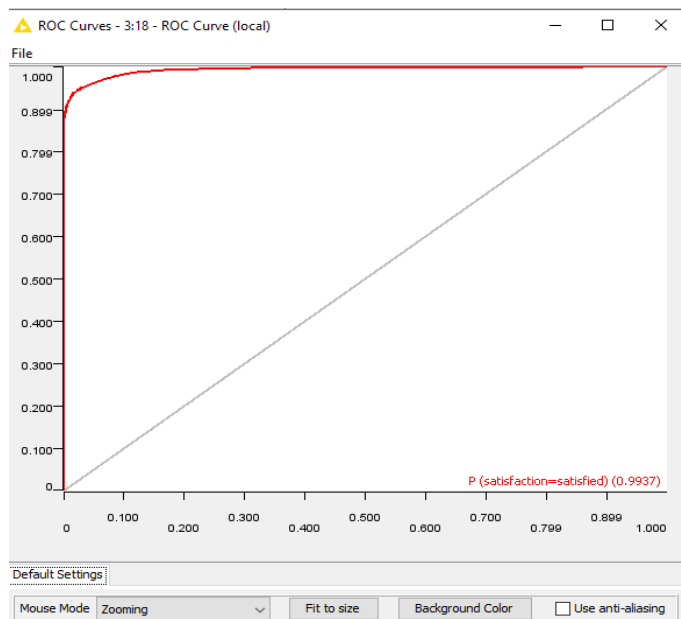
☐ Save target distribution in tree nodes (memory expensive - only important for tree view and PMML export)

Tree Options

Split Criterion: Information Gain Ratio

OK Apply Cancel ?

ROC:



Confusion Matrix - 3:57 - Scorer

File Hilite

satisfactio...	satisfied	neutral or ...
satisfied	10695	708
neutral or di...	251	14322

Correct classified: 25,017

Wrong classified: 959

Accuracy: 96.308 %

Error: 3.692 %

Cohen's kappa (κ) 0.925

1. Naive Bayes: The following Configuration was used for Naïve Bayes

Dialog - 3:25 - Naive Bayes Learner

File

Options Flow Variables Memory Policy

Classification Column: **S** satisfaction

Default probability: 0.0001

Minimum standard deviation 0.0001

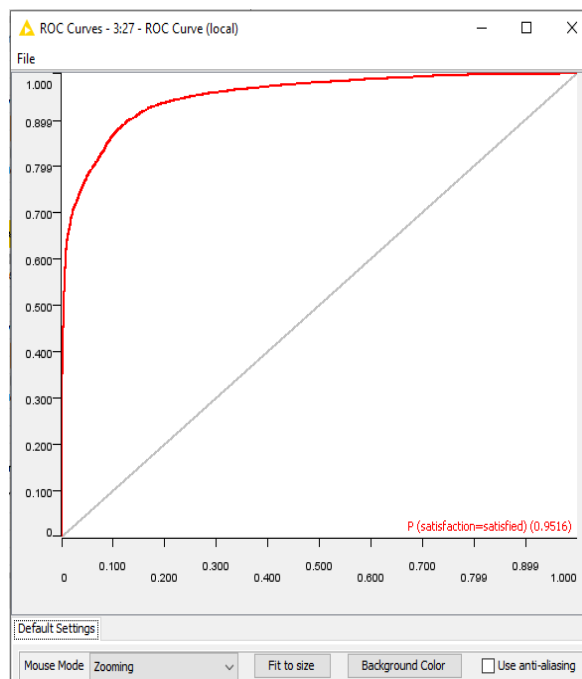
Threshold standard deviation 0.0

Maximum number of unique nominal values per attribute: 20

☐ Ignore missing values ☐ Create PMML 4.2 compatible model

OK Apply Cancel ?

ROC:



Confusion Matrix - 3:58 - Scorer

File Hilite

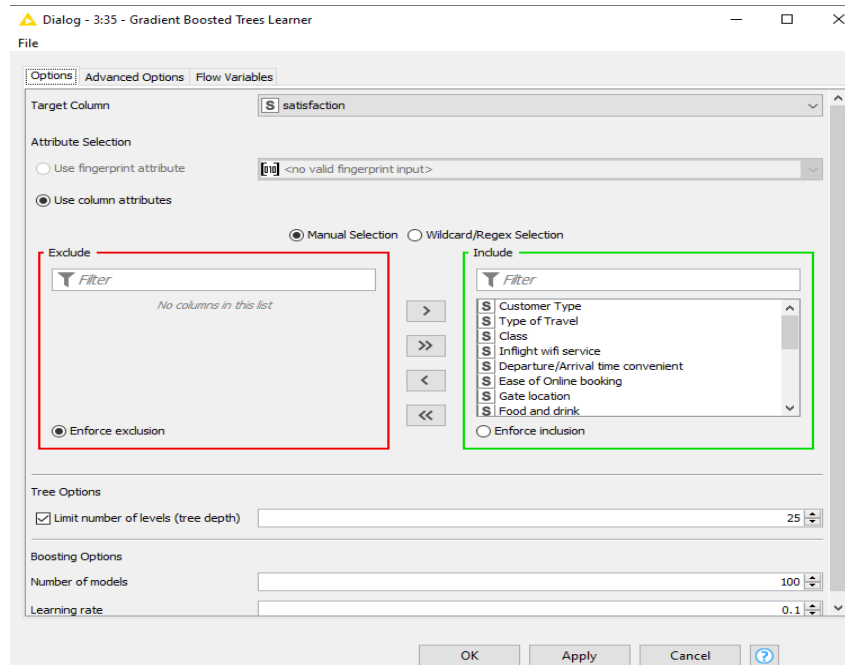
satisfacio...	satisfied	neutral or ...
satisfied	10042	1361
neutral or di...	1629	12944

Correct classified: 22,986 Wrong classified: 2,990

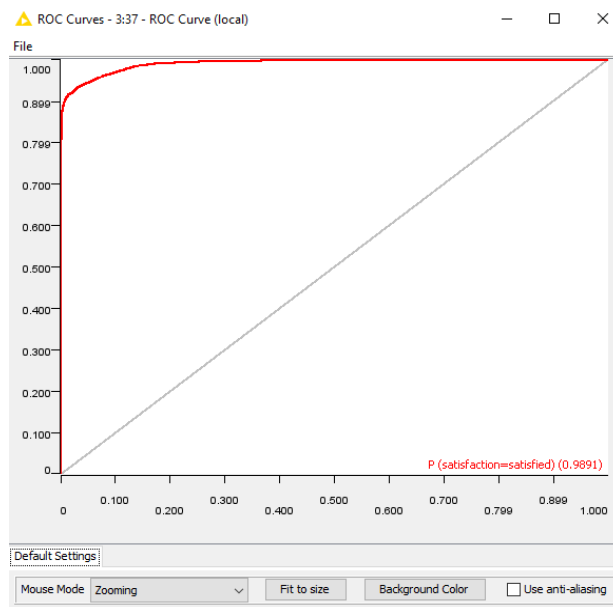
Accuracy: 88.489 % Error: 11.511 %

Cohen's kappa (κ) 0.767

2. Gradient Boost: The following configuration was used for gradient boosting



ROC:



Confusion Matrix - 3:59...

File Hilite

satisfacio...	satisfied	neutral or ...
satisfied	10735	668
neutral or di...	669	13904

Correct classified: 24,639 Wrong classified: 1,337

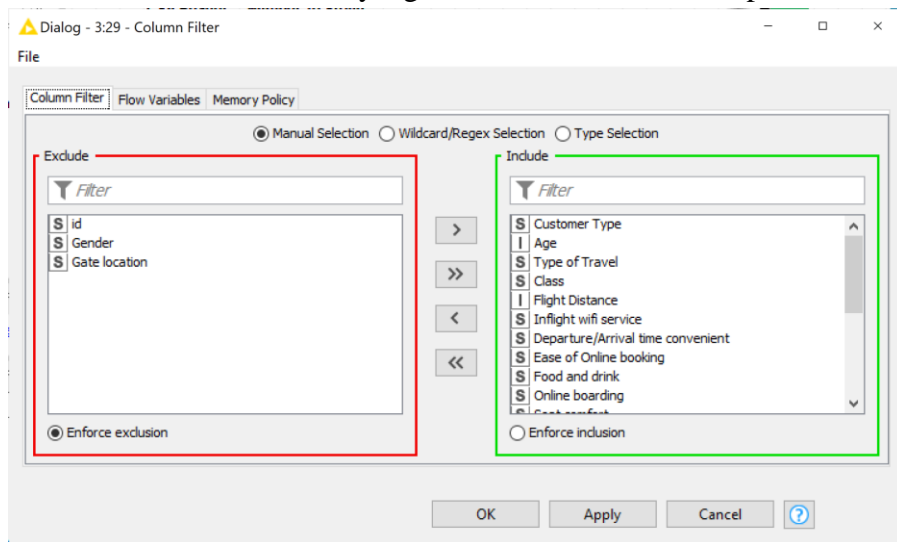
Accuracy: 94.853 % Error: 5.147 %

Cohen's kappa (κ) 0.896

Attempt 2:

Data Preprocessing:

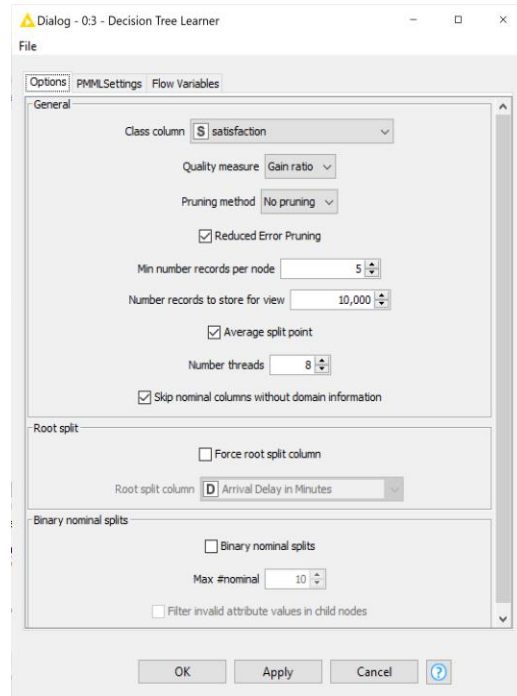
1. **Column Filter:** From our previous attempt we found out that the flight distance and age are important factors therefore their columns should be included and gate location was excluded since it did not make any significant difference in our prediction.



Model Building:

After changing the column filter's configuration, we again checked the accuracy using four classifiers. They are as follow:

1. Decision Tree:



ROC: The accuracy decreased a little by 0.01 percent from approximately 98% to 97%.

Confusion Matrix - 0:39 - Scorer

File Hilite

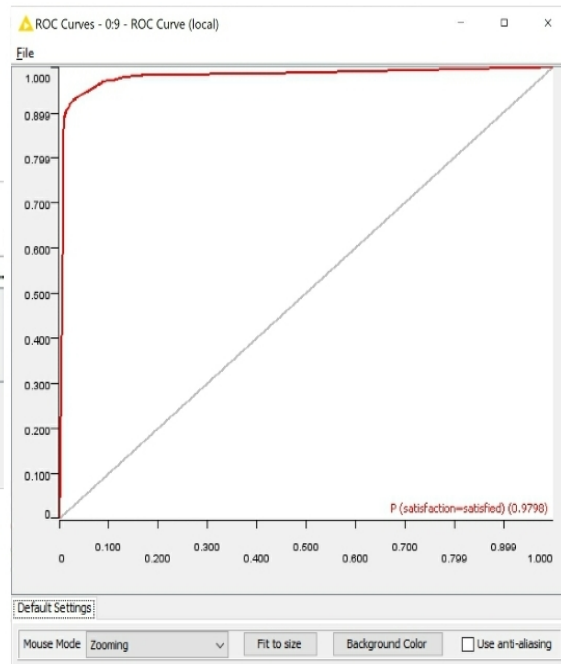
⚠ There were missing values in the reference or in the prediction class column...

satisfacio...	satisfied	neutral or ...
satisfied	10581	804
neutral or di...	446	14110

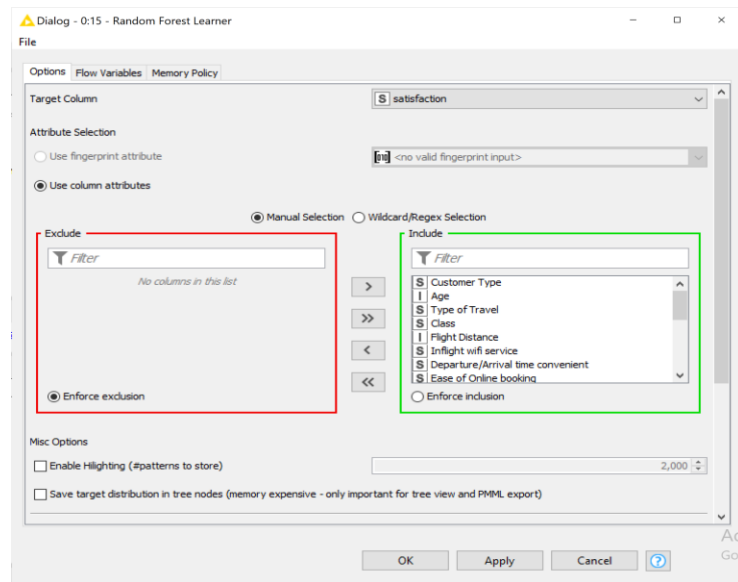
Correct classified: 24,691 Wrong classified: 1,250

Accuracy: 95.181 % Error: 4.819 %

Cohen's kappa (κ) 0.902



2. Random Forest:



ROC & Scorer: The accuracy increased.

Confusion Matrix - 0:38 - Scorer

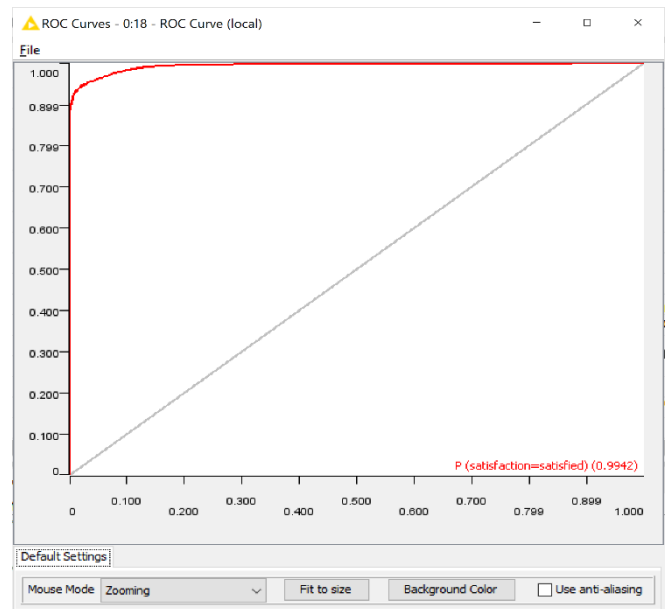
File Hilite

satisfacio...	satisfied	neutral or ...
satisfied	10699	704
neutral or di...	236	14337

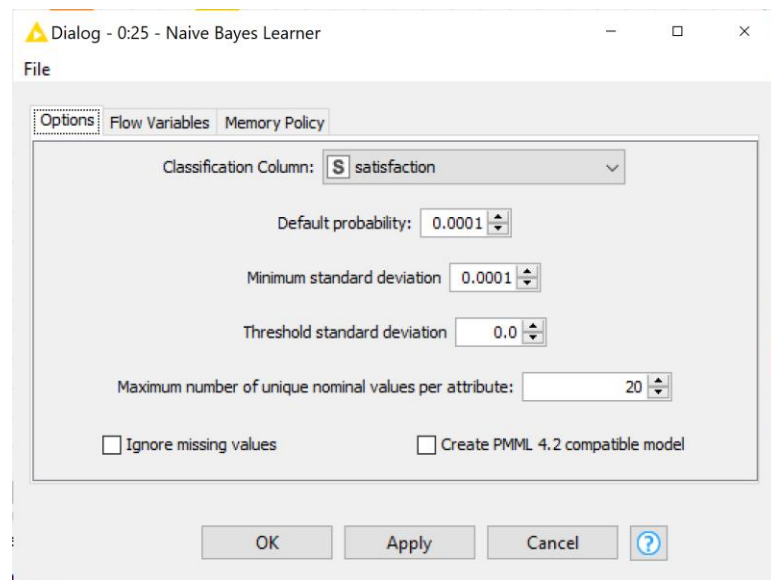
Correct classified: 25,036 Wrong classified: 940

Accuracy: 96.381 % Error: 3.619 %

Cohen's kappa (κ) 0.926



3. Naive Bayes:



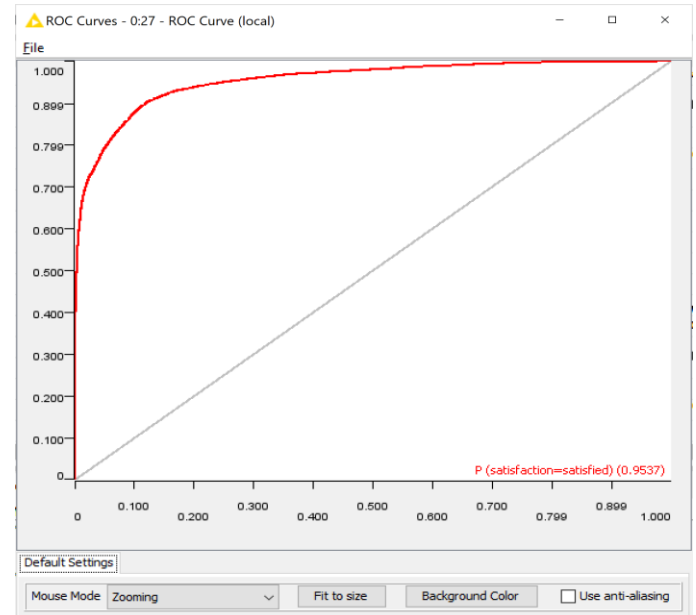
ROC & Scorer:

The screenshot shows the 'Confusion Matrix - 0:40 - Scorer' window. It has a 'File' menu and a 'Hilite' button. The confusion matrix is as follows:

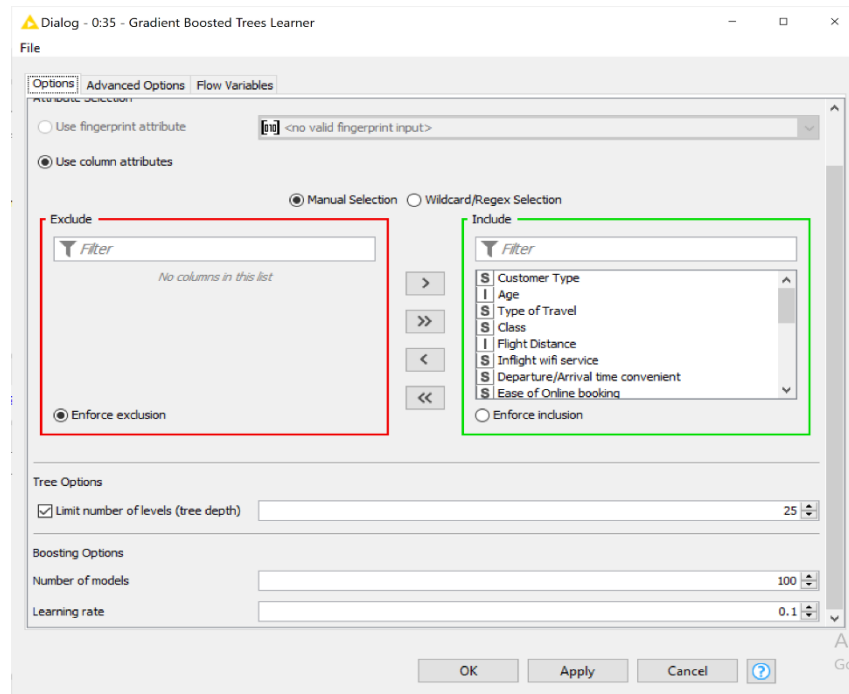
	satisfac...	satisfied	neutral or ...
satisfied	9978	1425	
neutral or di...	1440	13133	

Below the matrix, the following statistics are displayed:

- Correct classified: 23,111
- Wrong classified: 2,865
- Accuracy: 88.971 %
- Error: 11.029 %
- Cohen's kappa (κ) 0.776



4. Gradient Boosting:



ROC & Scorer:

Confusion Matrix - 0:41 - Scorer

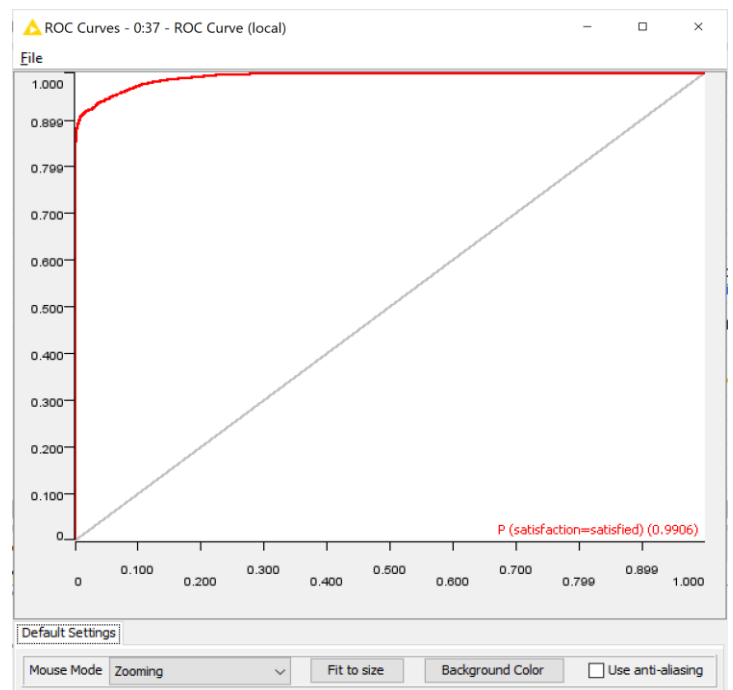
File Hilite

	satisfied	neutral or ...
satisfied	10740	663
neutral or d...	701	13872

Correct classified: 24,612 Wrong classified: 1,364

Accuracy: 94.749 % Error: 5.251 %

Cohen's kappa (κ) 0.893

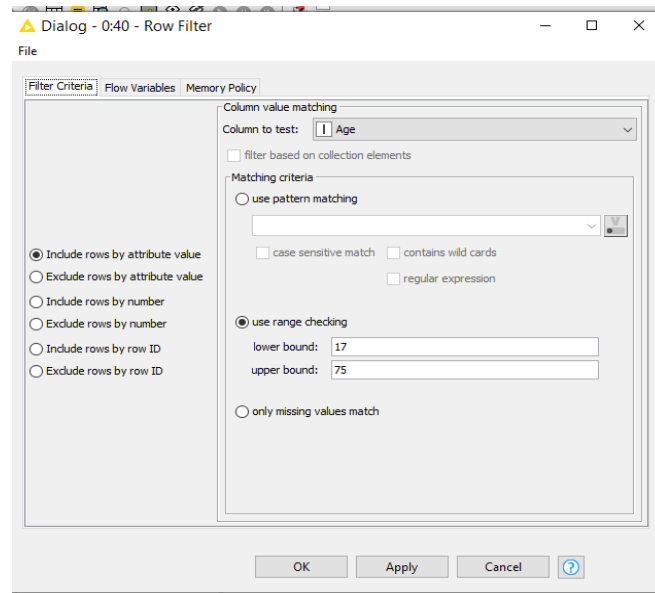


Attempt 3:

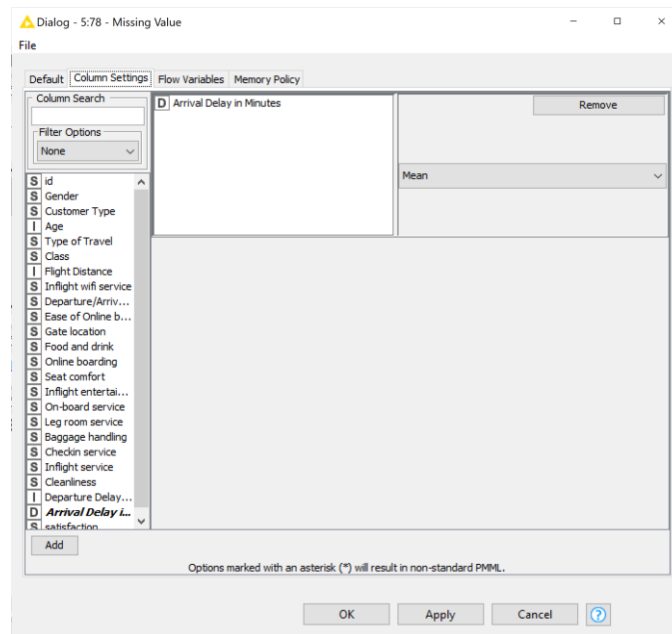
Data Preprocessing:

1. Row Filter

Keeping all the above conditions applied constant we applied a new filter -row filter- where we set the range of the age from 17 to 75 i.e. only the responses of people whose age lies in this range matter. People less than 18 are considered kids whose response does not hold much value.



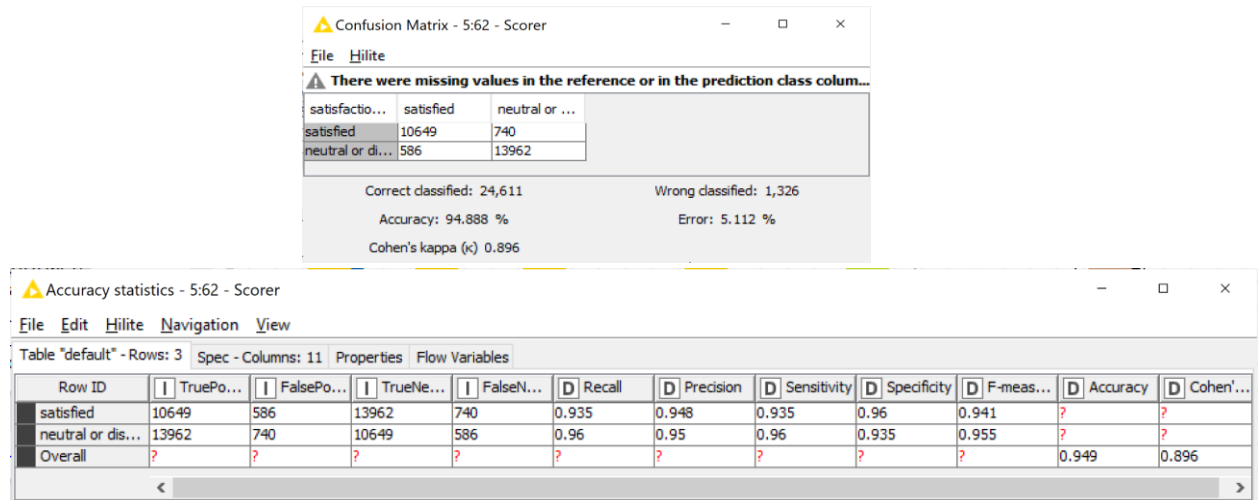
2. **Missing Value:** It was found that we had 310 missing data in arrival delay time so we used missing value filter.



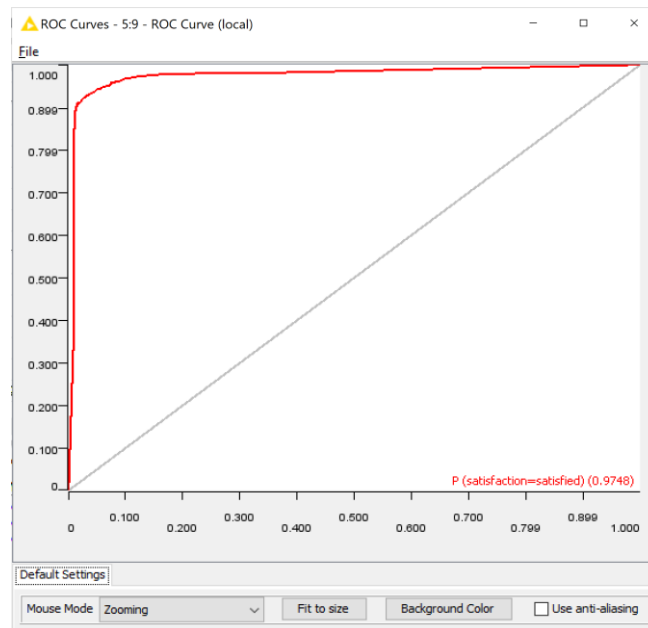
Models: The configurations for all the models were kept constant, only a few filters mentioned above were added to our models. However the accuracy of all models were effected at most 0.01. Therefore it was concluded that applying the row filter did not have much of an effect.

1. Decision Tree

Scorer



ROC:



2. Random Forest

ROC & Scorer:

Confusion Matrix - 5:63 - Scorer

	satisfio...	satisfied	neutral or ...
satisfied		10687	716
neutral or di...		246	14327

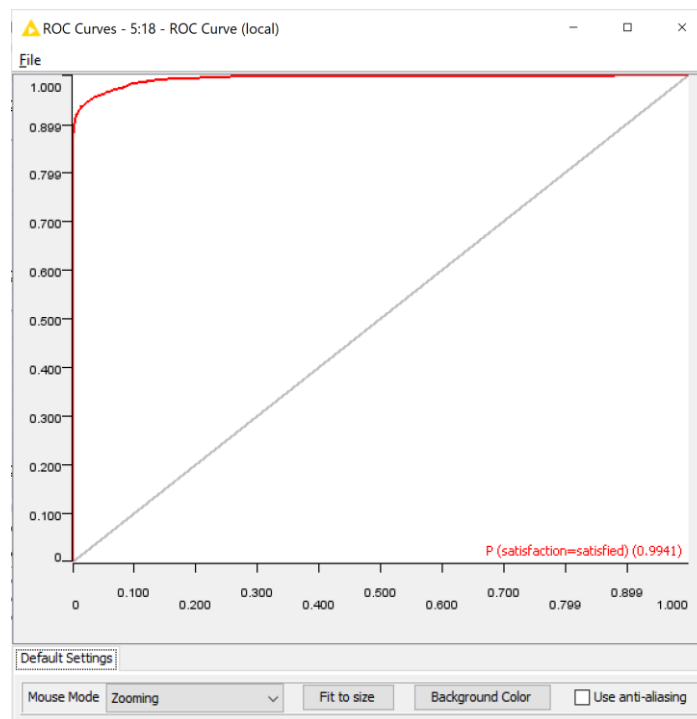
Correct classified: 25,014 Wrong classified: 962
Accuracy: 96.297 % Error: 3.703 %
Cohen's kappa (κ) 0.924

Accuracy statistics - 5:63 - Scorer

File Edit Hilite Navigation View

Table "default" - Rows: 3 Spec - Columns: 11 Properties Flow Variables

Row ID	TruePo...	FalsePo...	TrueNe...	FalseNe...	Recall	Precision	Sensitivity	Specificity	F-meas...	Accuracy	Cohen'...
satisfied	10687	246	14327	716	0.937	0.977	0.937	0.983	0.957	?	?
neutral or dis...	14327	716	10687	246	0.983	0.952	0.983	0.937	0.968	?	?
Overall	?	?	?	?	?	?	?	?	?	0.963	0.924



3. Naive Bayes

ROC & Scorer:

Confusion Matrix - 5:64 - Scorer

	satisfio...	satisfied	neutral or ...
satisfied	9905		1498
neutral or di...	1432		13141

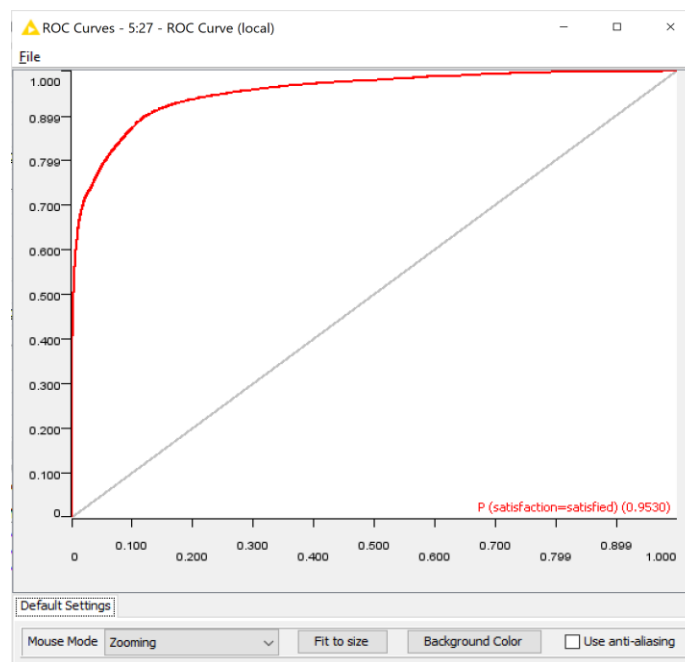
Correct classified: 23,046 Wrong classified: 2,930
Accuracy: 88.72 % Error: 11.28 %
Cohen's kappa (κ) 0.771

Accuracy statistics - 5:64 - Scorer

File Edit Hilite Navigation View

Table "default" - Rows: 3 Spec - Columns: 11 Properties Flow Variables

Row ID	I TruePo...	I FalsePo...	I TrueNe...	I FalseNe...	D Recall	D Precision	D Sensitivity	D Specificity	D F-meas...	D Accuracy	D Cohen's...
satisfied	9905	1432	13141	1498	0.869	0.874	0.869	0.902	0.871	?	?
neutral or dis...	13141	1498	9905	1432	0.902	0.898	0.902	0.869	0.9	?	?
Overall	?	?	?	?	?	?	?	?	?	0.887	0.771



4. Gradient Boosting

ROC And Scorer:

Confusion Matrix - 5:65 - Scorer

File Hilite

	satisfied	neutral or ...
satisfied	10719	684
neutral or di...	694	13879

Correct classified: 24,598 Wrong classified: 1,378

Accuracy: 94.695 % Error: 5.305 %

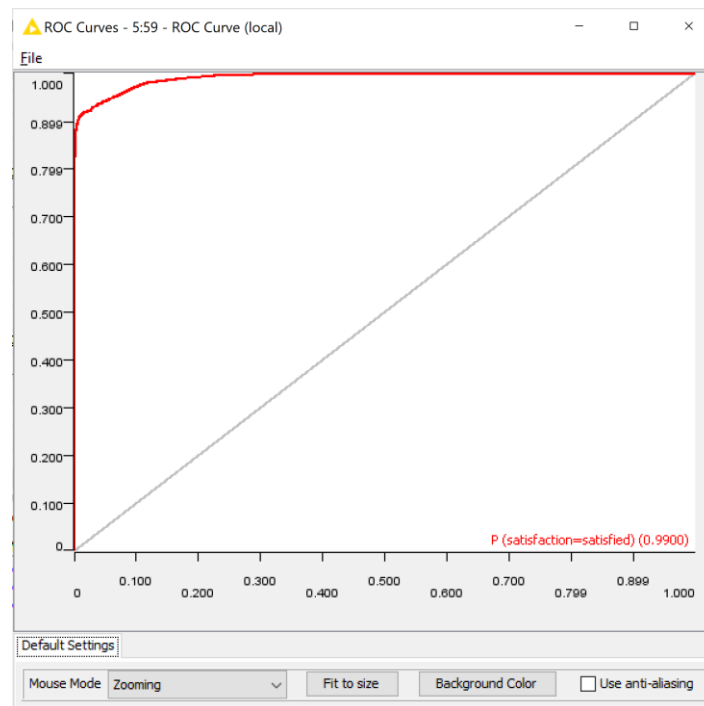
Cohen's kappa (κ) 0.892

Accuracy statistics - 5:65 - Scorer

File Edit Hilite Navigation View

Table "default" - Rows: 3 Spec - Columns: 11 Properties Flow Variables

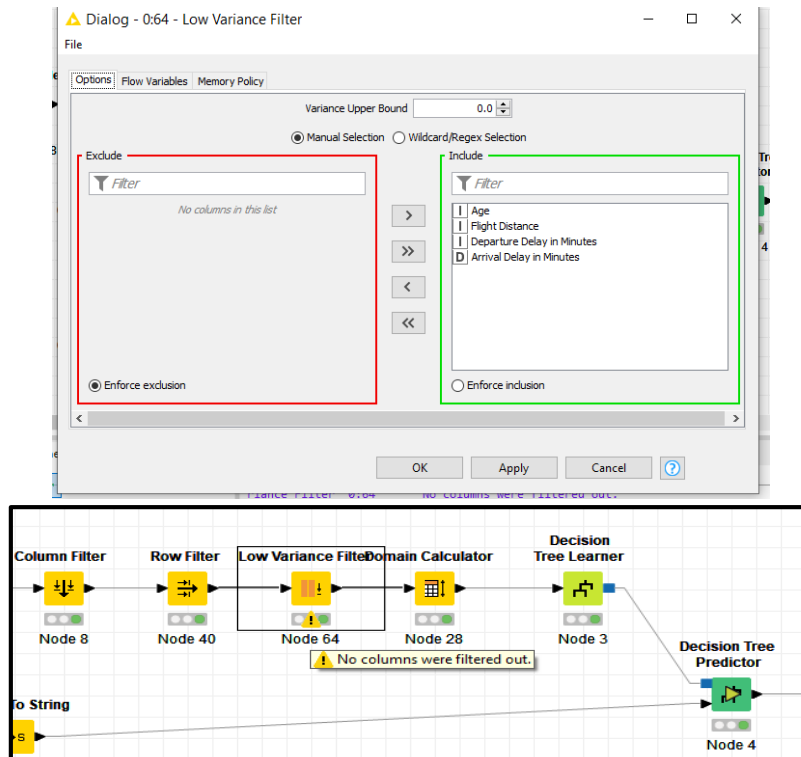
Row ID	TruePo...	FalsePo...	TrueNe...	FalseN...	Recall	Precision	Sensitivity	Specificity	F-meas...	Accuracy	Cohen'...
satisfied	10719	694	13879	684	0.94	0.939	0.94	0.952	0.94	?	?
neutral or dis...	13879	684	10719	694	0.952	0.953	0.952	0.94	0.953	?	?
Overall	?	?	?	?	?	?	?	?	?	0.947	0.892



Attempt 4:

Data Preprocessing

Low Variance Filter-In this attempt we applied low variance filter that ignores values that are below a certain range, however in our data set there were no columns that can be filtered out using low variance filter. hence , nodes' configurations were not changed a bit. nodes' configurations were not changed a bit.



Model Building

Since the low variance filter did not affect any of the columns therefore the accuracy was also not affected. Accuracy is the same as in attempt 3 for all the classification models.

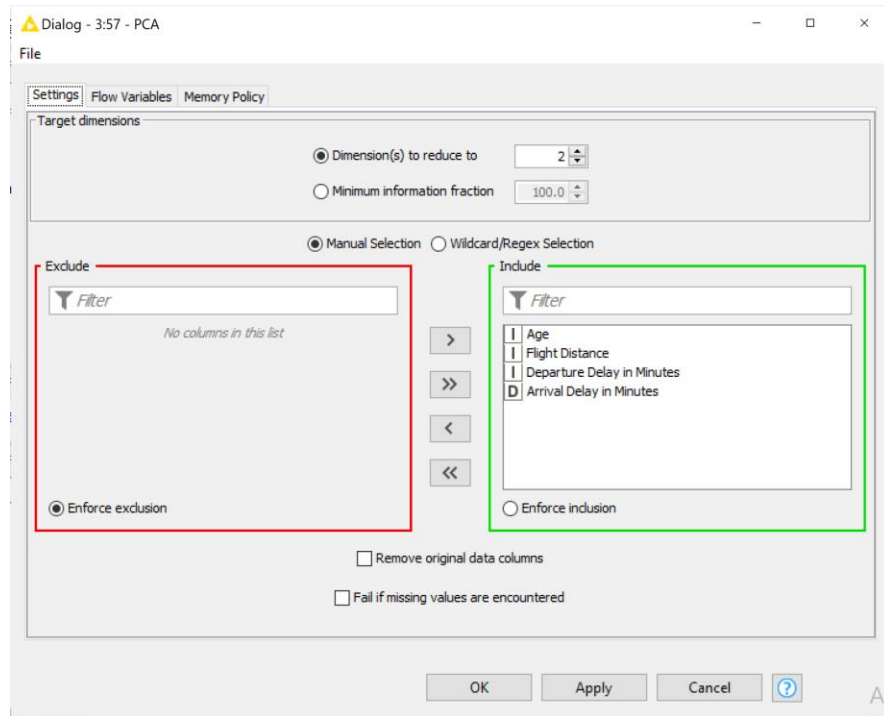
Confusion Matrix - 0:41 - Scorer		
File Hilite		
⚠ There were missing values in the reference or in the prediction class column...		
satisfactio...	satisfied	neutral or ...
satisfied	10562	802
neutral or di...	424	14098
Correct classified: 24,660		
Wrong classfied: 1,226		
Accuracy: 95.264 %		
Error: 4.736 %		
Cohen's kappa (κ) 0.903		

Attempt 5:

Data Preprocessing:

1. PCA:

Keeping the same configuration as Attempt 3, we added PCA to data preprocessing. Four numerical columns were reduced to two.



Model Building

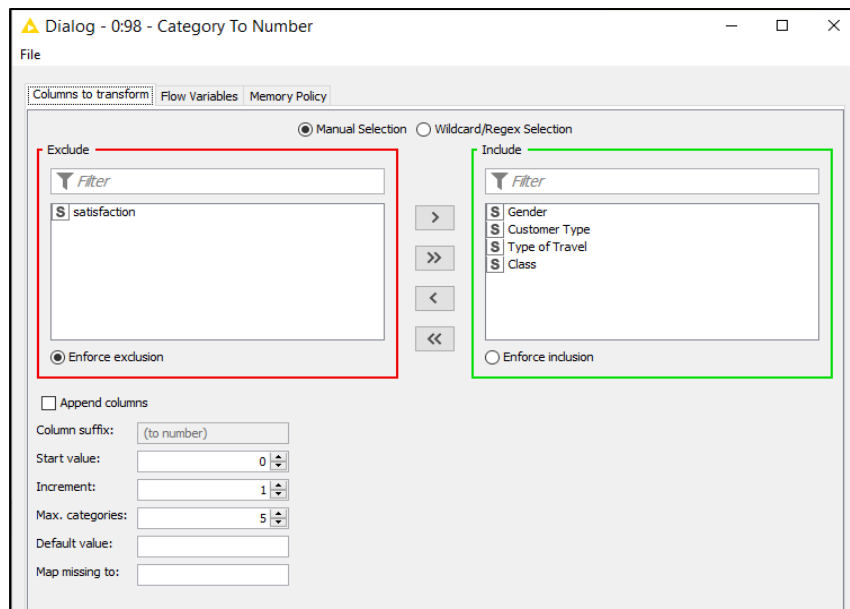
The configuration of all the classification models is kept the same as attempt 2 and 3. We found that the accuracy of each model declined after using PCA therefore we did not use it in the next attempts.

Attempt 6:

Data Preprocessing

1. Category to Number:

In this attempt we converted all the string columns to numerical values using category to number converter as the model used required all the values to be numerical. All the other filters used above were used with the same configuration.

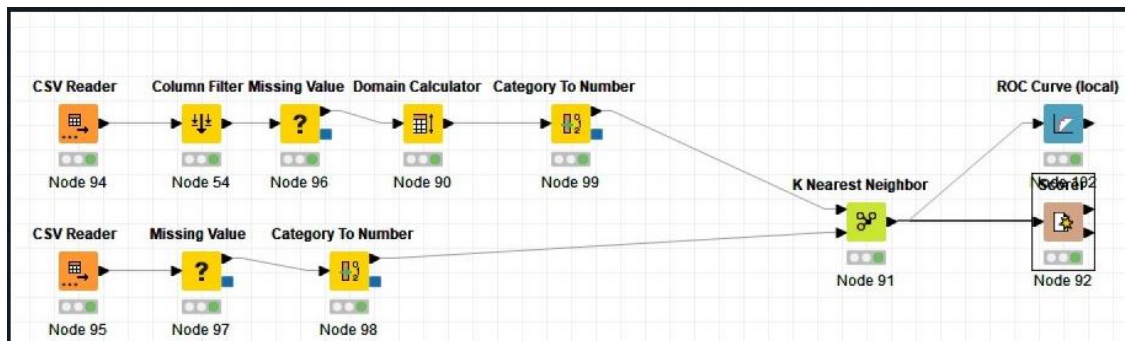
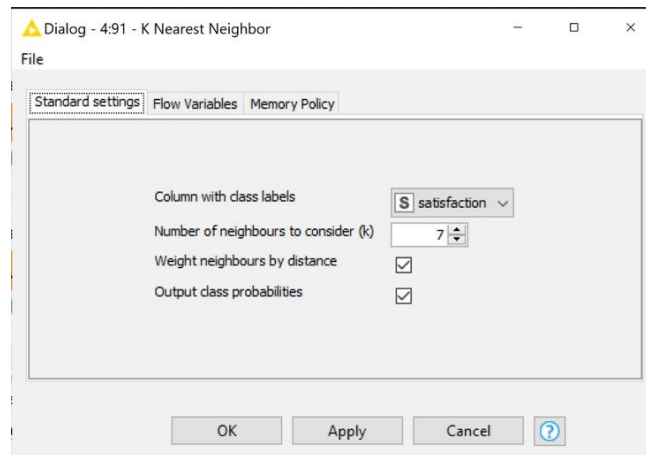


Row ID	Custom...	Age	Type of...	Class	Flight D...	Inflight...	Depart...	Ease of...	Food a...	Online ...	Seat co...	Inflight...	On-bos...	Leg roo...	Baggag...
0	0	13	0	0	460	3	4	3	5	3	5	5	4	3	4
1	1	25	1	1	235	3	2	3	1	3	1	1	1	5	3
2	0	26	1	1	1142	2	2	2	5	5	5	5	4	3	4
3	0	25	1	1	562	2	5	5	2	2	2	2	5	3	1
4	0	61	1	1	214	3	3	3	4	5	5	3	3	4	3
5	0	26	0	2	1180	3	4	2	1	2	1	1	3	4	4
6	0	47	0	2	1276	2	4	2	2	2	2	2	3	3	4
7	0	52	1	1	2035	4	3	4	5	5	5	5	5	5	4
8	0	41	1	1	853	1	2	2	4	3	3	1	1	2	1
9	1	20	1	2	1061	3	3	3	2	3	2	2	3	3	4
10	1	24	1	2	1162	4	5	5	2	5	2	2	3	3	3
11	0	12	0	0	308	2	4	2	1	2	1	1	1	2	5
12	0	53	1	0	834	1	4	4	1	1	1	1	1	1	3
13	0	33	0	2	946	4	2	4	4	4	4	4	4	5	2
14	0	26	0	2	453	3	2	3	2	3	2	2	4	3	2
15	1	13	1	2	486	2	1	2	4	2	1	4	2	1	4
16	0	26	1	1	2123	3	3	3	4	4	4	4	3	3	4
17	0	41	1	1	2075	4	4	2	4	4	4	5	5	5	3
18	0	45	1	1	2486	4	4	4	3	4	3	5	5	5	3
19	0	38	0	2	460	2	3	3	5	3	5	5	1	2	4
20	0	9	1	2	1174	2	4	2	2	2	1	2	1	5	3
21	0	17	0	2	208	3	1	3	5	3	5	5	2	5	3
22	0	43	0	2	752	3	5	3	5	4	5	3	3	3	3
23	0	58	0	2	2139	4	5	4	4	3	4	4	4	4	2
24	1	23	1	2	452	5	0	5	1	5	1	1	4	5	3
25	0	57	0	2	719	4	4	4	4	5	4	5	3	2	4
26	0	33	1	1	1561	1	1	1	1	5	3	4	4	4	3
27	0	49	1	0	315	4	4	4	2	2	1	4	4	4	2
28	0	26	1	1	3347	3	1	1	1	2	1	3	3	3	2
29	0	22	0	2	2342	3	2	3	3	3	1	3	2	4	4
30	0	31	1	1	819	4	4	4	4	5	5	5	4	4	1
31	1	15	1	2	1043	2	2	2	3	5	2	5	3	1	4
32	0	35	1	1	2611	4	5	4	4	4	4	3	3	4	5
33	0	67	0	2	1192	4	5	4	2	4	5	5	4	3	5
34	1	37	1	1	1182	3	3	3	1	3	1	1	4	1	3
35	0	40	1	2	349	1	4	4	1	1	1	1	3	3	3
36	1	34	1	1	883	3	4	4	5	4	2	5	5	4	5

Model Building:

K Nearest Neighbor

For this attempt we used KNN model, with the following configuration.



Scorer

Using KNN the accuracy of our model decline from approximately 95% to 80%.

Confusion Matrix - 4:92 - Scorer

File Hilite

satisfactio...	satisfied	neutral or ...
satisfied	7482	3921
neutral or di...	2881	11692

Correct classified: 19,174 Wrong classified: 6,802

Accuracy: 73.814 % Error: 26.186 %

Cohen's kappa (κ) 0.463

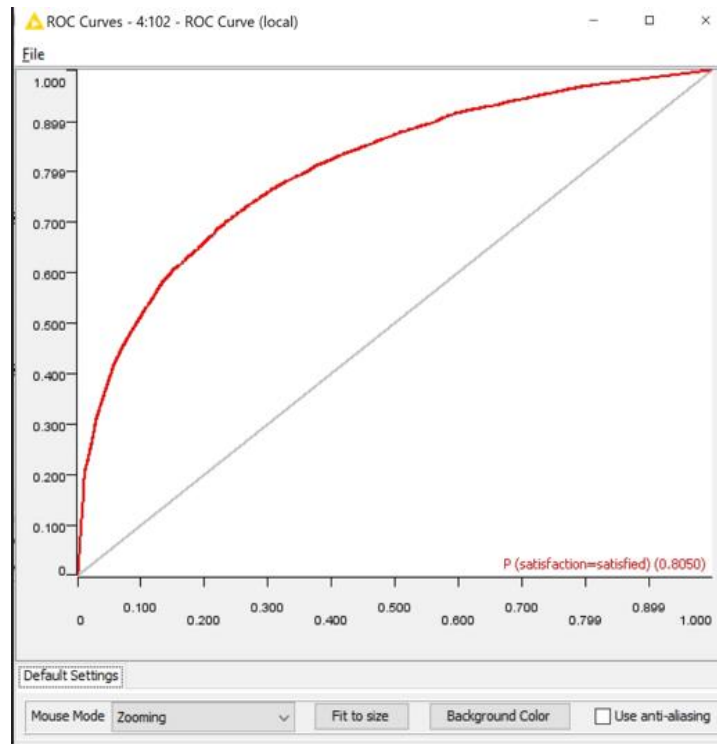
Accuracy statistics - 4:92 - Scorer

File Edit Hilite Navigation View

Table "default" - Rows: 3 Spec - Columns: 11 Properties Flow Variables

Row ID	TruePos...	FalsePos...	TrueNe...	FalseNe...	D Recall	D Precision	D Sensitivity	D Specificity	D F-meas...	D Accuracy	D Cohen...
satisfied	7482	2881	11692	3921	0.656	0.722	0.656	0.802	0.687	?	?
neutral or dis...	11692	3921	7482	2881	0.802	0.749	0.802	0.656	0.775	?	?
Overall	?	?	?	?	?	?	?	?	?	0.738	0.463

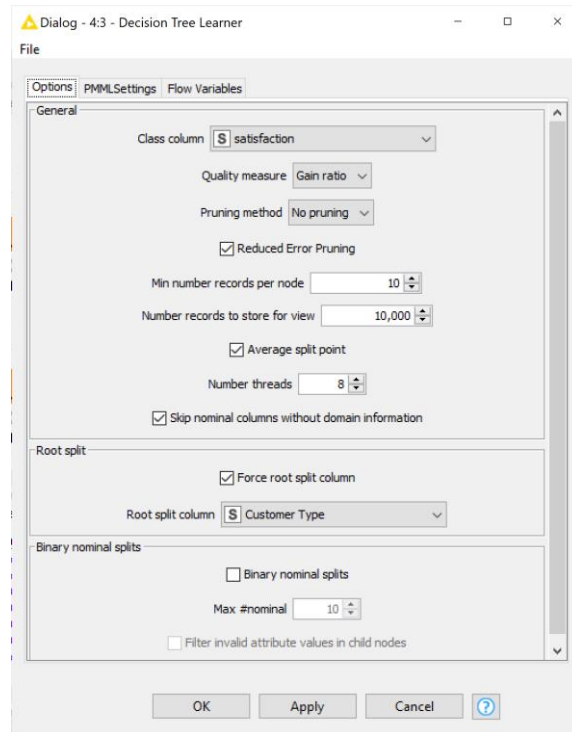
ROC:



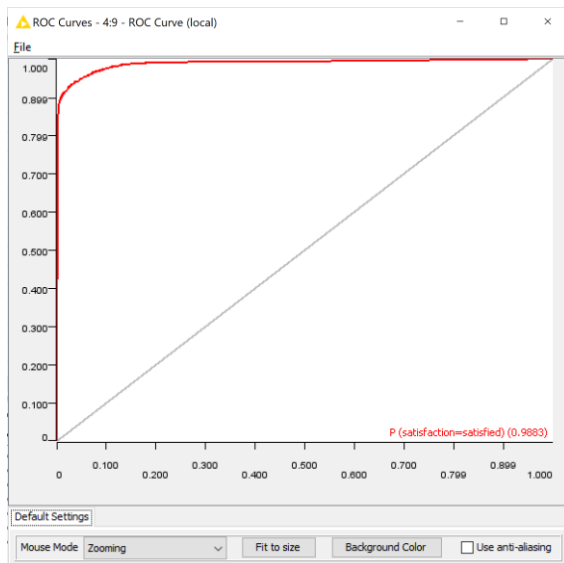
Attempt 7:


Changing configurations in Decision Tree:

7.1- For this attempt, we changed the minimum number of records per node from 5 to 10 and Root split the columns from Customer Type (Loyal or Disloyal Customers). We tried root split using other attributes as well but we got the highest accuracy using Customer Type.




ROC And Scorer:




Confusion Matrix - 4:62 - Scorer

File

Hilite


There were missing values in the reference or in the prediction class column...

satisfactio...	satisfied	neutral or ...	
satisfied	10637	764	
neutral or di...	451	14115	

Correct classified: 24,752

Wrong classified: 1,215

Accuracy: 95.321 %

Error: 4.679 %

Cohen's kappa (κ) 0.905

7.2- Changing the Quality measure from gain ratio to gini index reduced our accuracy.

Dialog - 4:91 - Decision Tree Learner

File

Options PMMLSettings Flow Variables

General

Class column **S** satisfaction

Quality measure Gini index

Pruning method No pruning

☒ Reduced Error Pruning

Min number records per node 10

Number records to store for view 10,000

☒ Average split point

Number threads 8

☒ Skip nominal columns without domain information

Root split

☐ Force root split column

Root split column **S** Customer Type

Binary nominal splits

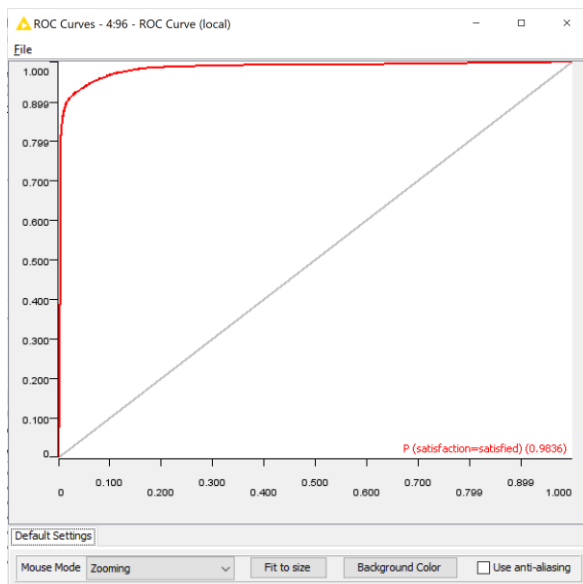
☐ Binary nominal splits

Max #nominal 10

☐ Filter invalid attribute values in child nodes

OK Apply Cancel ?

ROC And Scorer:



Confusion Matrix - 4:99 - Scorer

File Hilite

There were missing values in the reference or in the prediction class column...

satisfactio...	satisfied	neutral or ...
satisfied	10563	833
neutral or di...	559	14000

Correct classified: 24,563

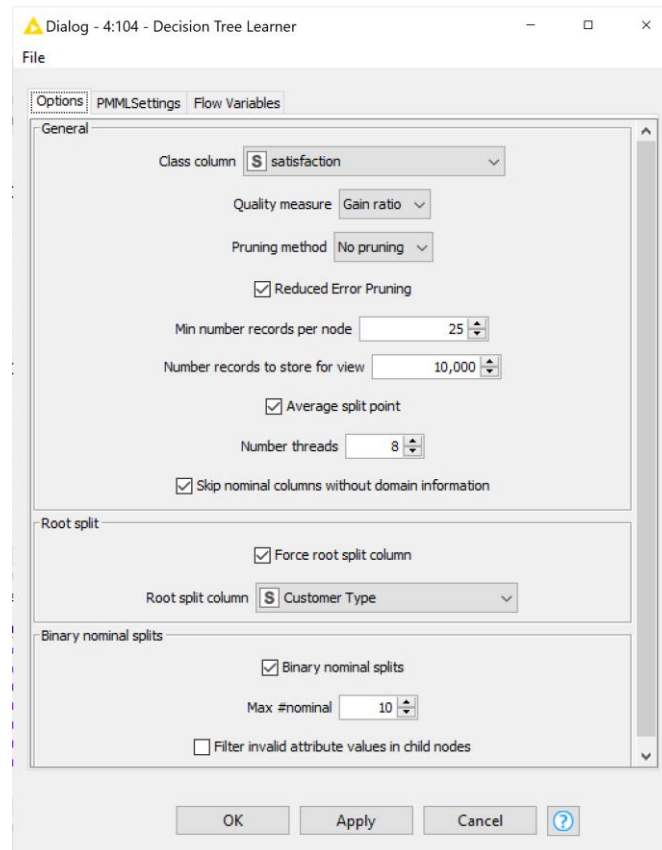
Wrong classified: 1,392

Accuracy: 94.637 %

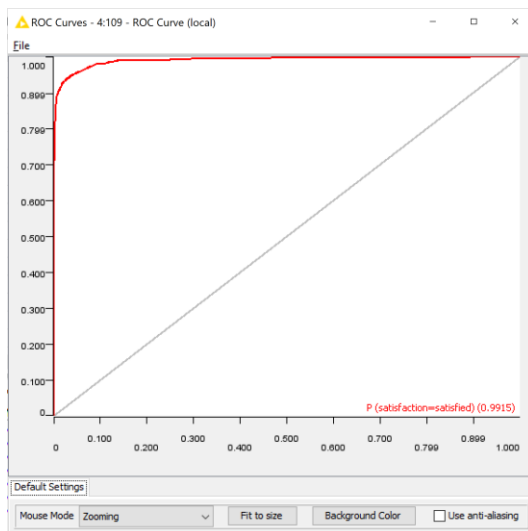
Error: 5.363 %

Cohen's kappa (κ) 0.891

7.3- Changed the Quality Measure back to Gain ratio with 25 as the minimum number of records per node, force splitting the column at customer type and binary nominal splits.



ROC and Scorer:



Confusion Matrix - 4:112 - Scorer

	satisfied	neutral or ...
satisfied	10745	658
neutral or di...	474	14099

Correct classified: 24,844 Wrong classified: 1,132

Accuracy: 95.642 % Error: 4.358 %

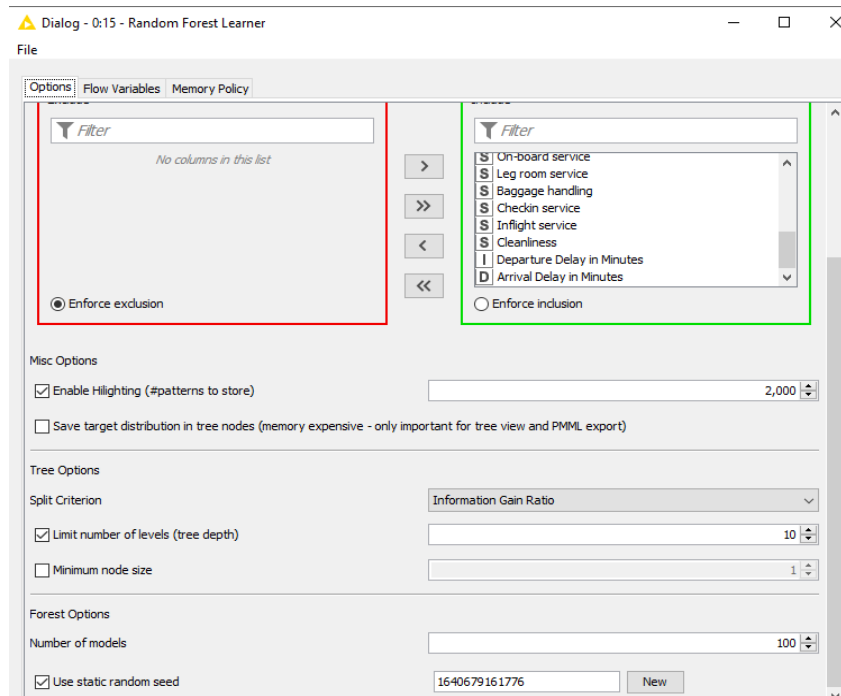
Cohen's kappa (κ) 0.911

For attempts 8-10 we did not made changes in our data preprocessing but we made changes in our model building to check if it had any effect on our accuracy or not.

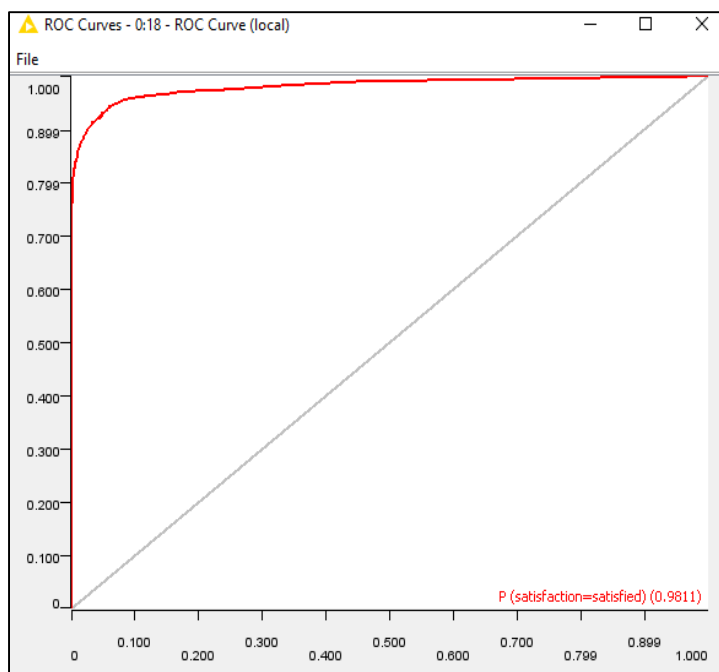
Attempt 8:

Changing configurations in Random Forest:

8.1- I limited the number of levels (tree depth) to 10.



ROC And Scorer:



Confusion Matrix - 4:63 - Sco...

File Hilite

satisfactio...	satisfied	neutral or ...
satisfied	10571	832
neutral or di...	704	13869

Correct classified: 24,440 Wrong classified: 1,536

Accuracy: 94.087 % Error: 5.913 %

Cohen's kappa (κ) 0.88

8.2- Reset the changes made in 8.1. Then tried excluding on-board service and checkin service. Just to see if it will affect the results.

Options | Flow Variables | Memory Policy

Target Column:

Attribute Selection

☐ Use fingerprint attribute

☒ Use column attributes

☒ Manual Selection ☐ Wildcard/Regex Selection

Exclude

- ☒ On-board service
- ☒ Checkin service

☒ Enforce exclusion

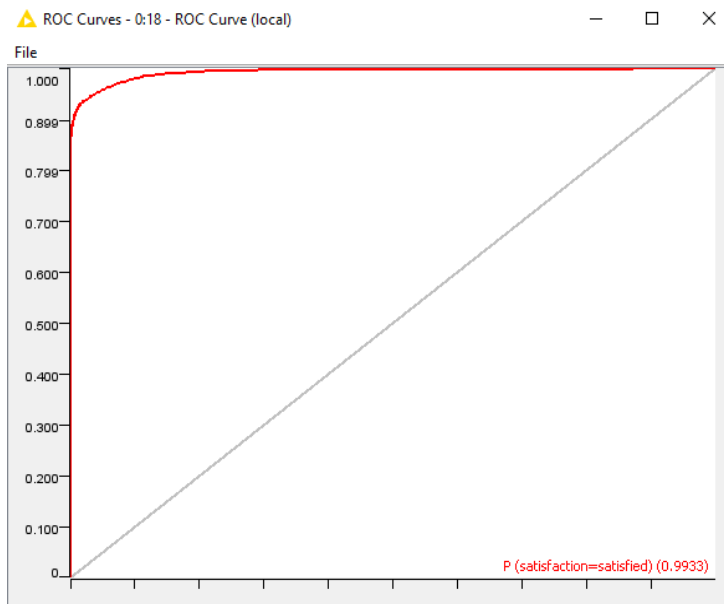
Include

- ☒ On-board service
- ☒ Seat comfort
- ☒ Inflight entertainment
- ☒ Leg room service
- ☒ Baggage handling
- ☒ Inflight service
- ☒ Cleanliness
- ☐ Departure Delay in Minutes

☐ Enforce inclusion

Micro Online

ROC And Scorer:



Confusion Matrix - 4:...

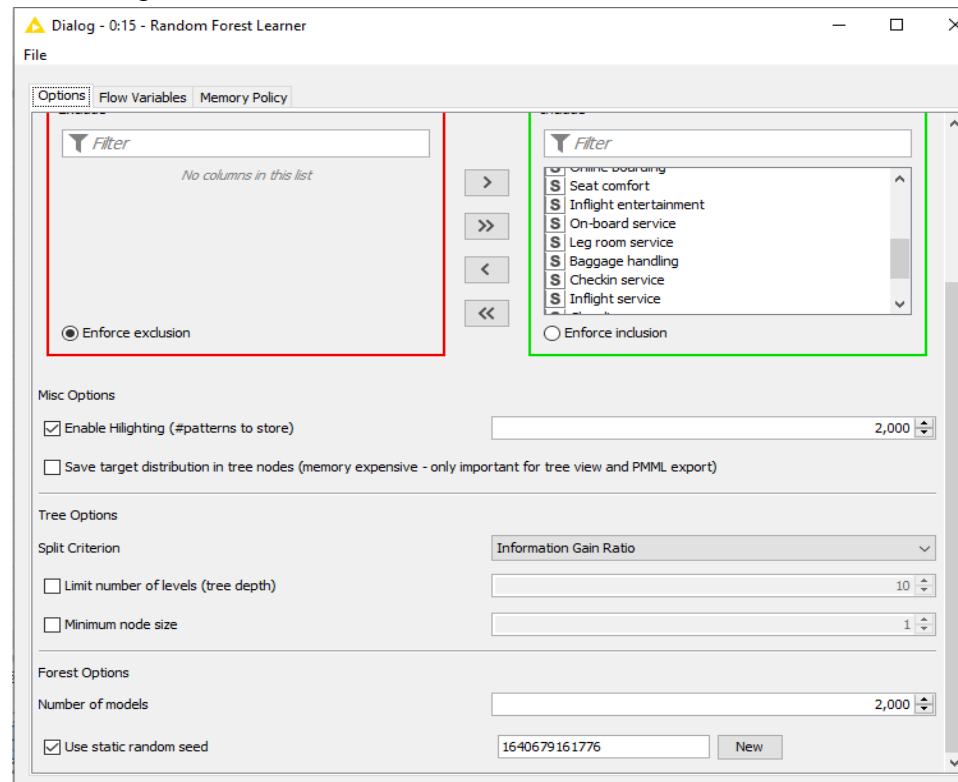
	satisfied	neutral or ...
satisfacio...	10649	754
neutral or di...	271	14302

Correct classified: 24,951 Wrong classified: 1,025

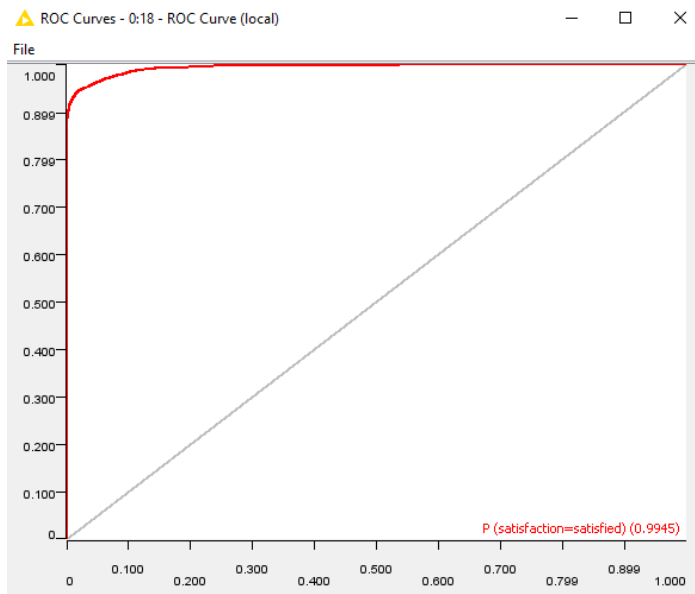
Accuracy: 96.054 % Error: 3.946 %

Cohen's kappa (κ) 0.92

8.3- Reset the changes made in 8.2. Increased the number of models from 100 to 2000.



ROC And Scorer:



Confusion Matrix - 3:38 - Scorer

File Hilight

satisfactio...	satisfied	neutral or ...
satisfied	10584	819
neutral or di...	694	13879

Correct classified: 24,463 Wrong classified: 1,513

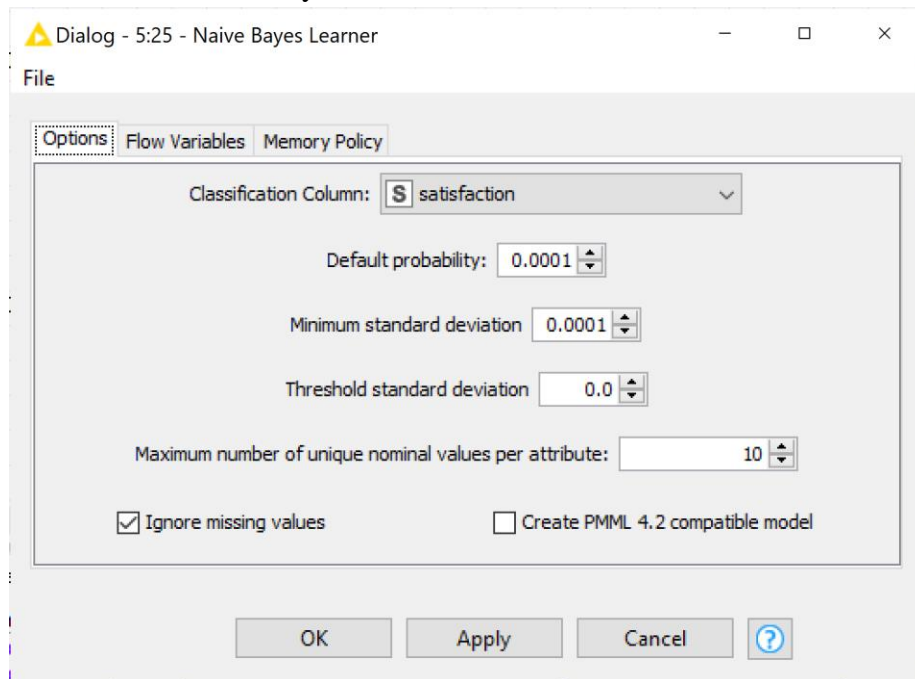
Accuracy: 94.175 % Error: 5.825 %

Cohen's kappa (κ) 0.882

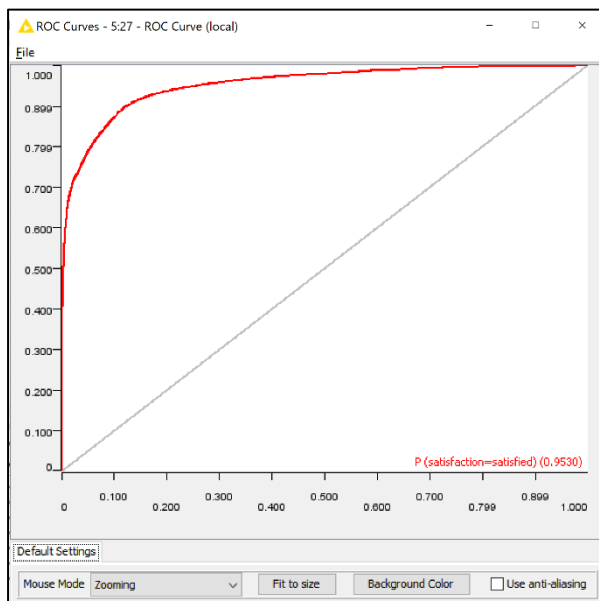
Attempt 9:

Changing configurations in Naive Bayes Learner:

9.1- I changed the maximum number of unique nominal values per attribute from 20 to 10, to check whether it affects the accuracy or not.



ROC And Scorer:



The screenshot shows the 'Confusion Matrix - 5:64 - Scorer' window. It displays a confusion matrix for the 'satisfaction' variable. The matrix shows 9905 correct classifications for 'satisfied' and 1498 misclassifications. The overall accuracy is 88.72%, and the error rate is 11.28%. The Cohen's kappa is 0.771.

	satisfied	neutral or ...
satisfied	9905	1498
neutral or di...	1432	13141

Correct classified: 23,046 Wrong classified: 2,930
Accuracy: 88.72 % Error: 11.28 %
Cohen's kappa (κ) 0.771

9.2- Reset the changes made in 9.1, realized if we go below 6 in the maximum number of unique nominal values per attribute the accuracy drops sharply.

Dialog - 5:87 - Naive Bayes Learner

File

Options Flow Variables Memory Policy

Classification Column: **S** satisfaction

Default probability: 0.0001

Minimum standard deviation 0.0001

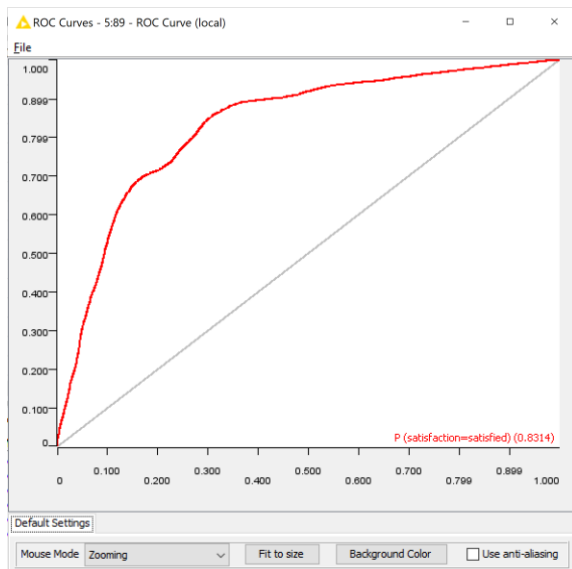
Threshold standard deviation 0.0

Maximum number of unique nominal values per attribute: 4

☒ Ignore missing values ☐ Create PMML 4.2 compatible model

OK Apply Cancel ?

ROC And Scorer:



Confusion Matrix - 5:93 - Scorer

File Hilite

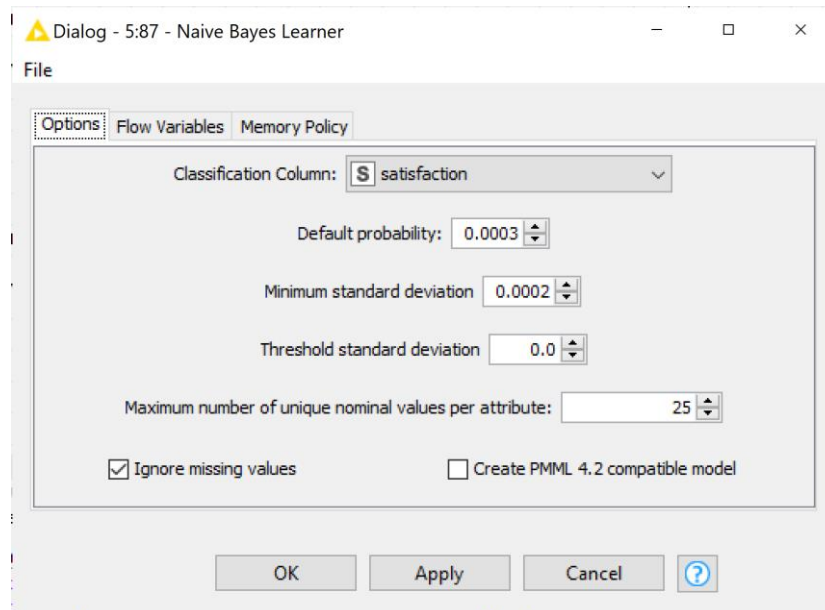
satisfactio...	satisfied	neutral or ...
satisfied	8190	3213
neutral or di...	3003	11570

Correct classified: 19,760 Wrong classified: 6,216

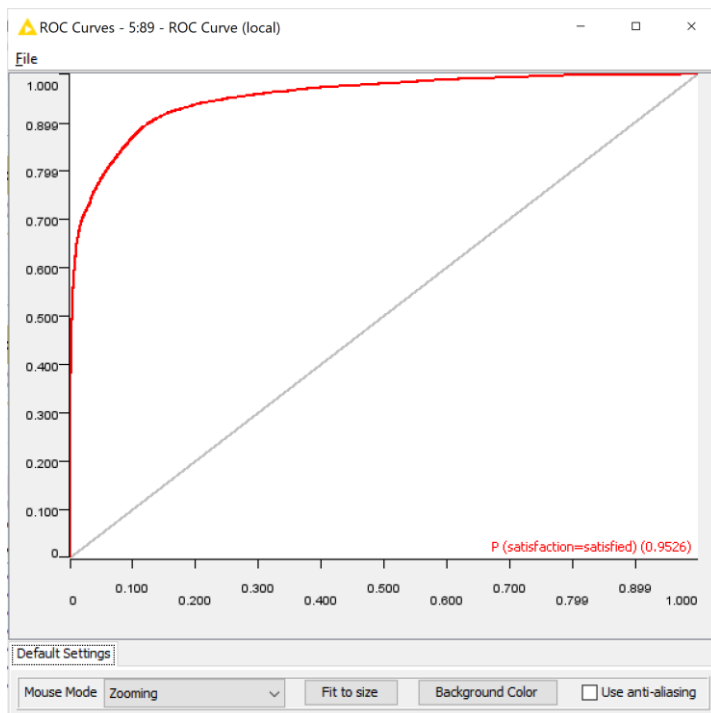
Accuracy: 76.07 % Error: 23.93 %

Cohen's kappa (κ) 0.513

9.3: Changed the probability to 0.0003, minimum S.D to 0.0002 and increased the nominal values per attribute to 25. Even if we increased it more the accuracy did not change.



ROC And Scorer:



Confusion Matrix - 5:93 - Scorer

File Hilite

	satisfac...	satisfied	neutral or ...
satisfied	9922	1481	
neutral or di...	1472	13101	

Correct classified: 23,023 Wrong classified: 2,953

Accuracy: 88.632 % Error: 11.368 %

Cohen's kappa (κ) 0.769

Attempt 10:

Changing configurations in Gradient Boosting Learner:

10.1: Unticked the option of limiting the number of levels(tree depth)

Dialog - 0:61 - Gradient Boosted Trees Learner

File

Options Advanced Options Flow Variables

Target Column [S] satisfaction

Attribute Selection

☐ Use fingerprint attribute [no] <no valid fingerprint input>

☒ Use column attributes

☒ Manual Selection ☐ Wildcard/Regex Selection

Exclude

Filter

No columns in this list

☒ Enforce exclusion

Include

Filter

[S] Customer Type
[I] Age
[S] Type of Travel
[S] Class
[I] Flight Distance
[S] Inflight wifi service
[S] Departure/Arrival time convenient
[S] Ease of Online booking

☐ Enforce inclusion

Tree Options

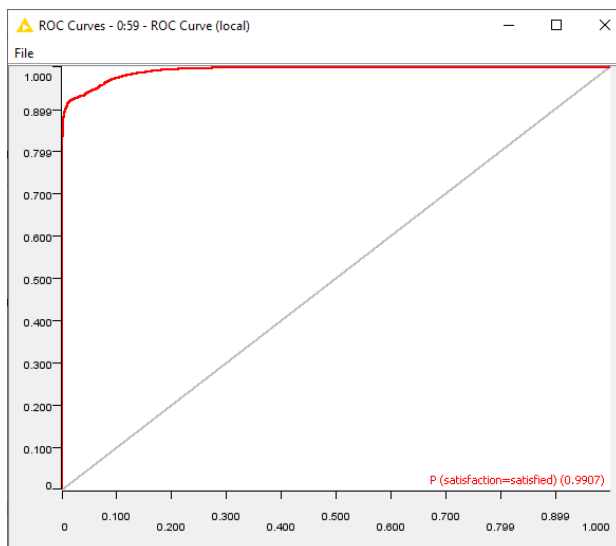
☐ Limit number of levels (tree depth) 10

Boosting Options

Number of models 100

Learning rate 0.1

ROC And Scorer:



Confusion Matrix - 6:65 - Scorer

File Hilite

satisfactio...	satisfied	neutral or ...
satisfied	10743	660
neutral or di...	757	13816

Correct classified: 24,559 Wrong classified: 1,417

Accuracy: 94.545 % Error: 5.455 %

Cohen's kappa (κ) 0.889

10.2: Increased the number of models to 500.

Dialog - 0:61 - Gradient Boosted Trees Learner

File

Options | Advanced Options | Flow Variables

Target Column:

Attribute Selection

☐ Use fingerprint attribute

☒ Use column attributes

☒ Manual Selection ☐ Wildcard/Regex Selection

Exclude

No columns in this list

☒ Enforce exclusion

Include

☒ Enforce inclusion

Tree Options

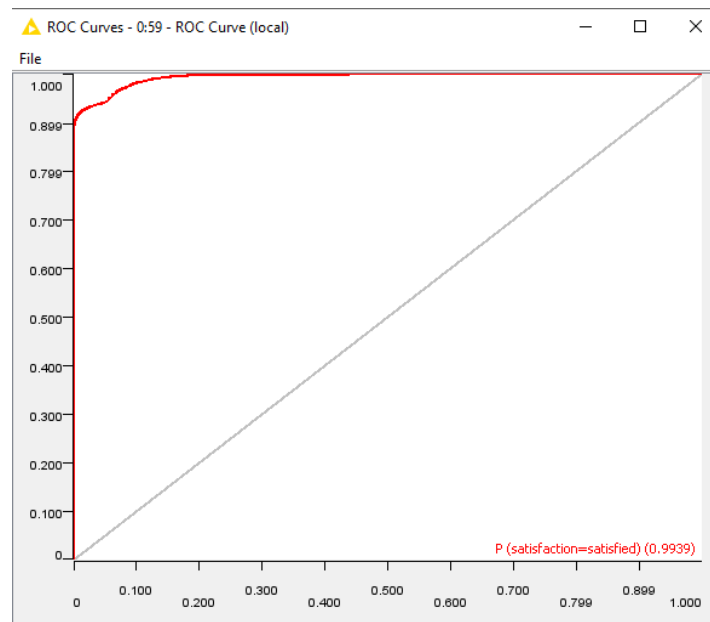
☐ Limit number of levels (tree depth)

Boosting Options

Number of models:

Learning rate:

ROC And Scorer:



Confusion Matrix - 6:65 - Scorer

File | Hilit

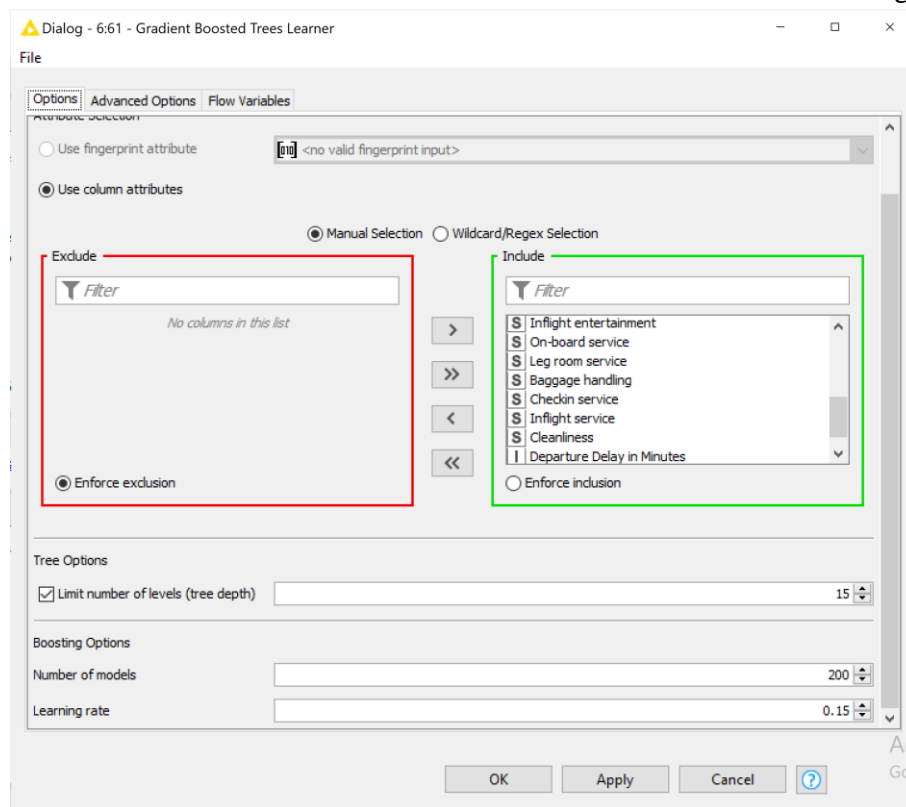
satisfactio...	satisfied	neutral or ...
satisfied	10742	661
neutral or di...	726	13847

Correct classified: 24,589 Wrong classified: 1,387

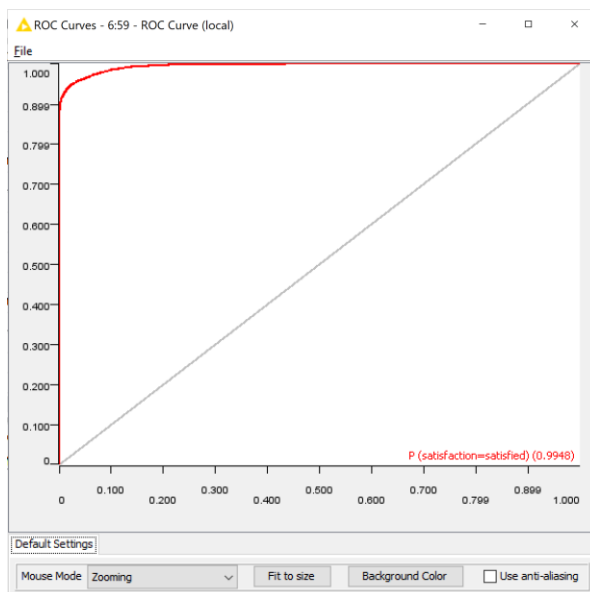
Accuracy: 94.66 % Error: 5.34 %

Cohen's kappa (κ) 0.892

10.3: Limited the number of levels to 15 with 200 number of models and a learning rate of 0.15



ROC And Scorer:

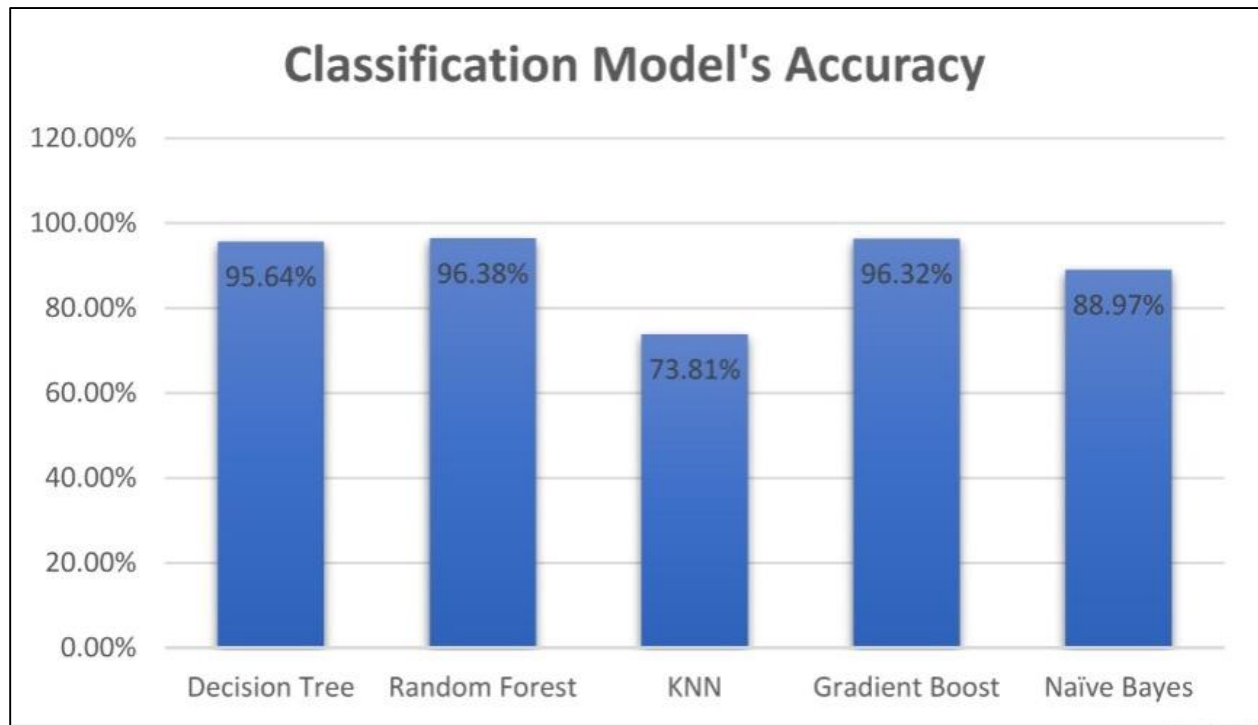


Confusion Matrix - 6:65 - Scorer		
File	Highlight	
satisfactio...	satisfied	neutral or ...
satisfied	10779	624
neutral or di...	333	14240
Correct classified: 25,019		
Wrong classified: 957		
Accuracy: 96.316 %		
Error: 3.684 %		
Cohen's kappa (κ) 0.925		

Findings

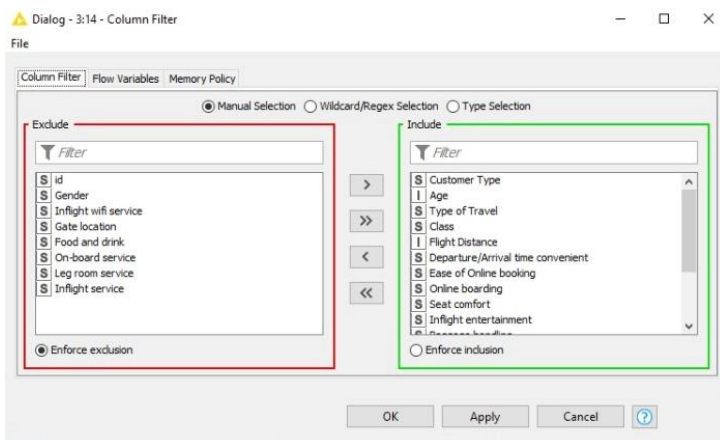
Finding Best Algorithm

We used various classification model for airlines to identify critical bottleneck to raise passenger satisfaction. As seen in the bar chart below, random forest gave us the best accuracy in attempt 2.



Factors Effecting Passenger Satisfaction

To figure out which features effect our accuracy the most, we took our best model and tried excluding certain features using column filter.

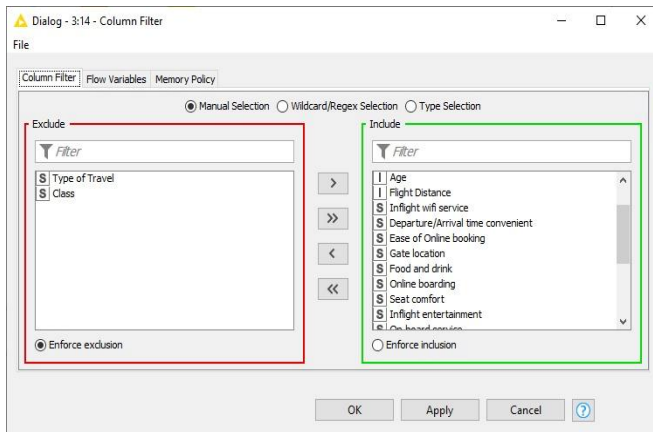


Confusion Matrix - 3:46 - Scorer

File Hilite

satisfactio...	satisfied	neutral or ...
satisfied	10337	1066
neutral or di...	593	13980

Correct classified: 24,317	Wrong classified: 1,659
Accuracy: 93.613 %	Error: 6.387 %
Cohen's kappa (κ) 0.87	



Confusion Matrix - 3:46 - Scorer		
File Hilite		
satisfactio...	satisfied	neutral or ...
satisfied	10544	859
neutral or di...	333	14240
Correct classified: 24,784		
Wrong classified: 1,192		
Accuracy: 95.411 %		
Error: 4.589 %		
Cohen's kappa (κ) 0.906		

From some of the simulations, it was seen that airlines should focus on improving the Services provided in the flight. For instance, airlines could develop better software to allow easier access to inflight wi-fi, food & etc. Excluding the “class” feature we observed how it effected the overall satisfaction of the passenger. Business class passenger are more satisfied with the flight so the airlines should initiate cheaper packages for business class.

Concluding remarks

The dramatic drop in demand for passenger air transport due to the COVID-19 pandemic has caused tremendous decline in the viability in the aviation industry. As a result, in order to resurrect the industry in the face of the crisis, it is critical to understand client pain points and increase their satisfaction with the services supplied. The industry should involve a third party market research company to provide them with unique, fresh, and unbiased recommendations to continuously improve their services attracting more people to choose their airlines for their trip.

There may be various options for gathering client feedback. The most straightforward and traditional method is to use the customer feedback form that is provided throughout the trip. However, the majority of passengers are uninterested in completing feedback forms. Other methods for collecting client input include the airline's internet website or online mobile applications. After the travel, the passenger can be sent an email with a link asking for feedback.

Finally, we hope that the model will provide a reference and be utilized for business value to the airlines, and will help them improve their services to allure more passengers towards them.