Cover sheet for submission of work for assessment

SWINBURNE UNIVERSITY OF TECHNOLOGY

UNIT DETAILS

Unit name	Data Science Principles			Class day/time	Wed, 8 – 12am	Office use only
Unit code	COS10022	Assignment no.	02	Due date	26/03/2023	
Name of lecturer/teacher	Dr. Pham Thi Kim Dung					
Tutor/marker's name	Dr. Pham Thi Kim Dung					Faculty or school date stamp

STUDENT(S)

Family Name	Given Name	Student ID Number
Hau	Linh Chi	104177160

DECLARATION AND STATEMENT OF AUTHORSHIP

1. For the sake of this evaluation, I have not impersonated anyone or let anyone else to impersonate me.
2. This evaluation is all original work from myself, with the exception of the places where proper credit has been given.
3. Except where such collaboration has been approved by the lecturer or instructor in question, no portion of this evaluation has been prepared for me by anyone else.
4. I have not previously submitted this work for this or any other course/unit.
5. I accept that my assessment response may be duplicated, shared, compared, stored, and used for benchmarking, plagiarism detection, or educational purposes.

I understand that:

6. Plagiarism is known as the act of presenting another person's work, idea, or creativity as your own. It is a sort of academic fraud that might get you kicked out of the university and is considered cheating. Written, graphic, and visual works, computer data, oral presentations, and other forms of presentation can all be used to source and deliver plagiarized content. When the source of the copied material is not properly cited, plagiarism occurs.

Student signature/s

I declare that I have read and understood the declaration and statement of authorship.

Further information relating to the penalties for plagiarism, which range from a formal caution to expulsion from the University is contained on the Current Students website at **www.swin.edu.au/student/**

Copies of this form can be downloaded from the Student Forms web page at **www.swinburne.edu.au/studentforms/**

COS10022 – Data Science Principles – Assignment 1

DATA CLEANING AND ANALYTICS

I. ASSIGNMENT SUMMARY

This assignment provides an overview of data cleaning and prediction model construction, including the key concepts, procedures, and tools involved.
The focus is on:

- Selecting appropriate features and models while gaining some knowledge about prebuilt tools and applying them in a data science project.
- Preparing the dataset using KNIME analytical platform for cleaning and two classification models, Naïve Bayes and Random Forest, are developed.
- Selecting relevant attributes, cleaning the dataset, partitioning the data into training and test sets, developing an effective prediction model, and explaining the results.

The goal of the assignment is to gain hands-on experience with data cleaning and model construction in a real-world setting.

II. INTRODUCTION

The focus of this report is a dataset that was gathered from the real world, consisting of 100,000 tuples that are categorised into three different financial credit score classes. The original data includes 24 attributes in total. The goals of this assignment are: firstly, to carry out the necessary data cleaning and preparation for future use, and secondly, to develop two predictive models that can be used to project the "Credit_Score" classification.
My works are described in detail in this report and include:

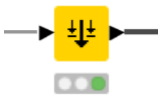
- Preparing the raw data for future use by cleaning and organising it.
- Constructing two models (Naïve Bayes and Random Forest classifier models) to forecast the value and categorise financial credit scores.
- Choosing and implementing suitable features and models for this data project.
- Identifying relevant attributes, splitting the dataset into training and testing sets, developing predictive models, and explaining the results.

III. DATA CLEANING

Question 1.1

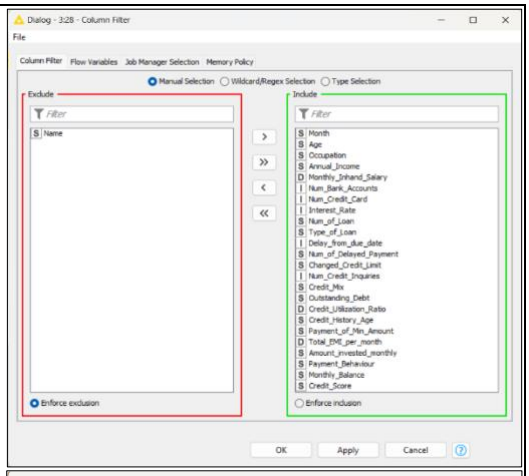
Node	Configuration

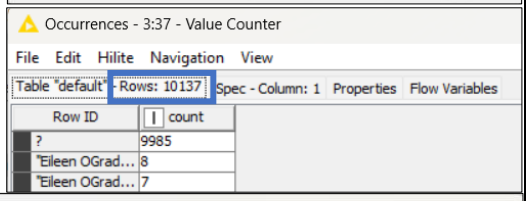
Column Filter

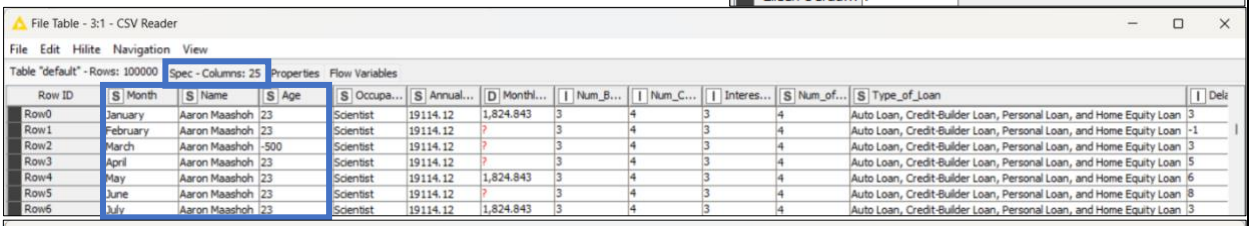


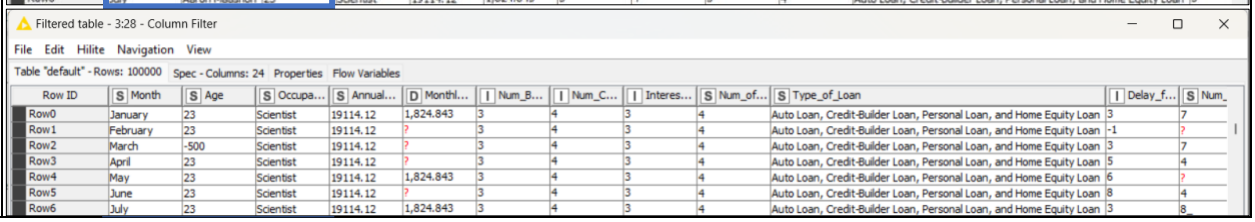
The excluded attribute is *"Name"*. The reason for removing it are:

- Credit scores are determined by a variety of factors such as payment history, credit utilisation, length of credit history, types of credit used, new credit inquiries, etc. Therefore, a person's name does not affect their credit score.
- In the Naïve Bayes learner process, the maximum number of unique nominal values per attribute is set to 600 to limited by the amount of available data, the complexity of the problem, and the performance requirements of the application. However, the number of names in the dataset exceeds the limit number (10137 different names). So, the *"Name"* attribute is filtered out.

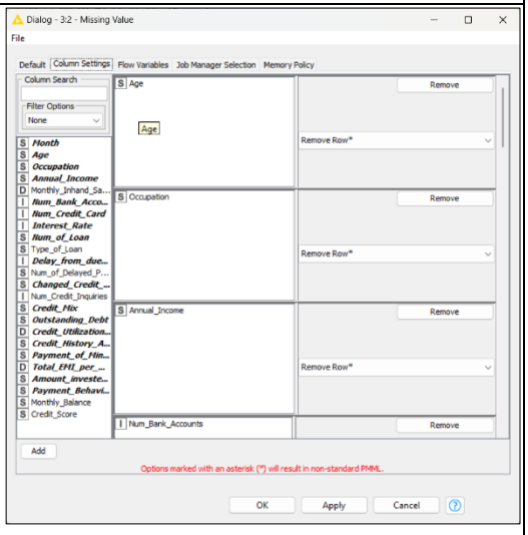


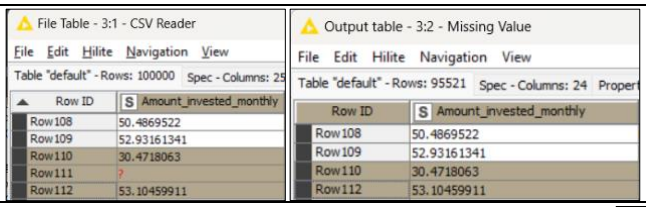
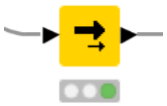
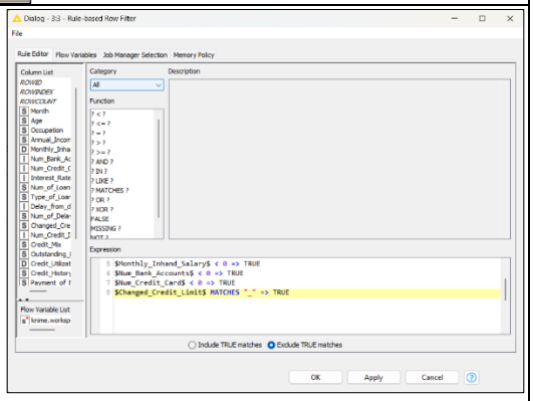
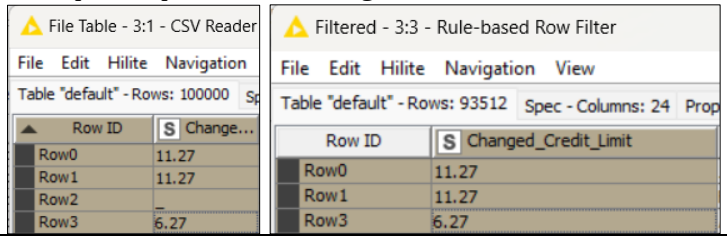




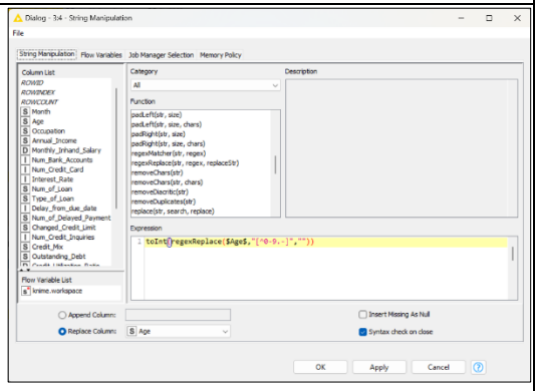


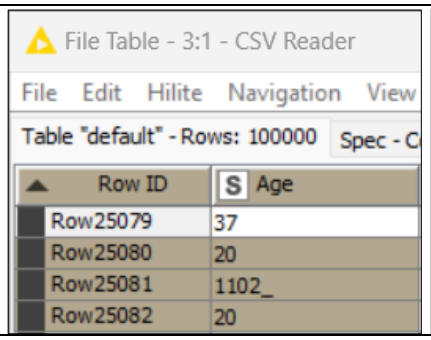
Question 1.2

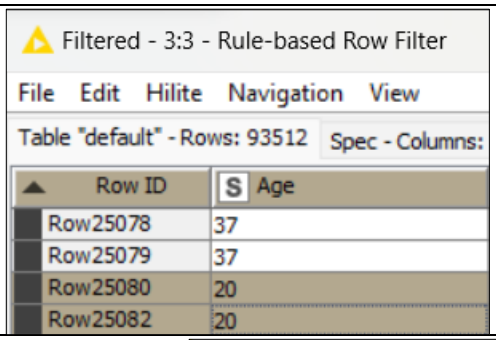
S	Node	Configuration
1	**Missing Value** 	This node is used to remove tuples with missing values from these 18 attributes *"Month"*, *"Age"*, *"Occupation"*, *"Annual_Income"*, *"Num_Bank_Accounts"*, *"Num_Credit_Card"*, *"Interest_Rate"*, *"Num_of_Loan"*, *"Delay_from_due_date"*, *"Changed_Credit_Limit"*, *"Credit_Mix"*, *"Outstanding_debt"*, *"Credit_Utilization_Ratio"*, *"Credit_History_Age"*, *"Payment_of_Min_Amount"*, *"Total_EMI_per_month"*, *"Amount_invested_monthly"* and *"Payment_Behaviour"*. One example output of the *"Amount_invested_monthly" column*: 

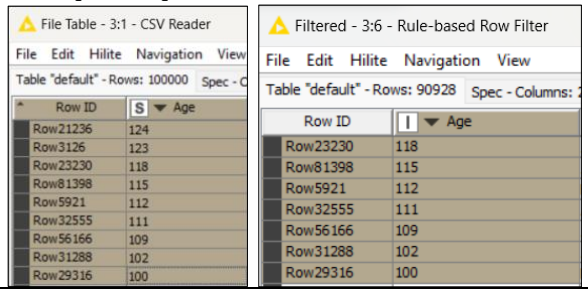
<table>
<tr><td></td><td></td><td> 
</td></tr>
<tr><td>2</td><td>Rule-based Row Filter
  </td><td>Some tuples containing infeasible values, such as:
• “Monthly_Inhand_Salary” < 0
• “Num_Bank_Accounts” < 0
• “Num_Credit_Card” < 0
• “Changed_Credit_Limit” contains “_”
are removed by using this node with the commands (exclude TRUE matches) as follows:
$Monthly_Inhand_Salary$ < 0 => TRUE
$Num_Bank_Accounts$ < 0 => TRUE
Num_Credit_Card < 0 => TRUE
$Changed_Credit_Limit$ MATCHES "_" => TRUE
 
Example output of the “Changed_Credit_Limit” column:
 
</td></tr>
</table>

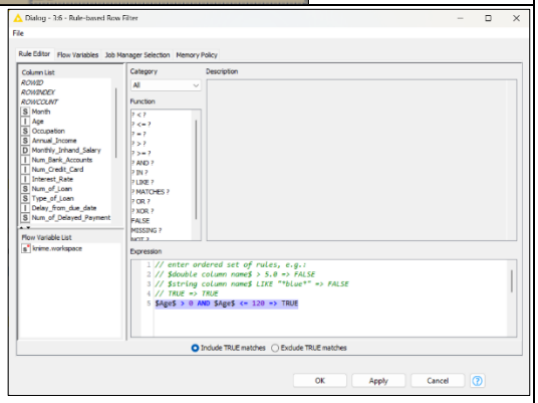
Question 1.3

<table>
<tr><th>S</th><th>Node</th><th>Configuration</th></tr>
<tr><td>1</td><td>String Manipulation
  </td><td>Symbols that are not numbers are eliminated from the “Age” attribute and the data of this attribute is converted into the usual number format (to integer) and replace the new output to the original “Age” column. The “.” and “-” are kept because they represent decimal number and negative number (some of them must be remove, so erasing the “-” can lead to wrong data cleaning, respectively.
Expression:
toInt(regexReplace(Age,"[^0-9.-]",""))
 
Example output:</td></tr>
</table>

<table>
<tr><td></td><td></td><td> 

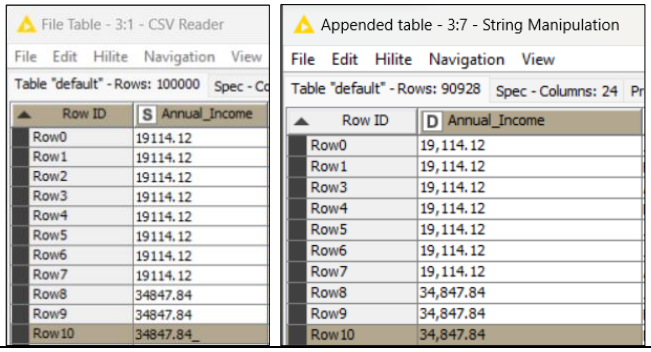

</td></tr>
<tr><td>2</td><td>Rule-based Row Filter</td><td>The tuples having “Age” value lower than or equal to 0 or greater than 120 are dropped using the following command (include TRUE matches):
Age > 0 AND Age <= 120 => TRUE

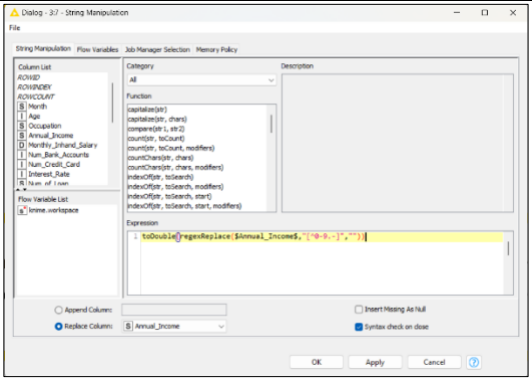
Example output:
 

 </td></tr>
</table>

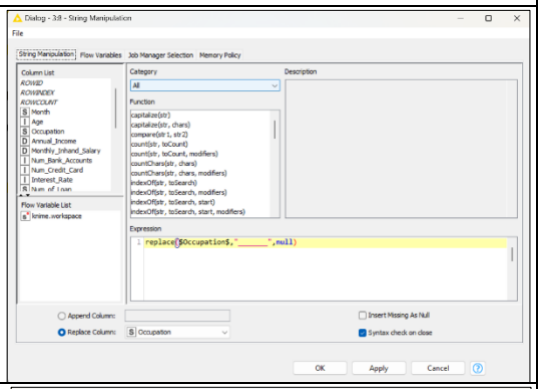
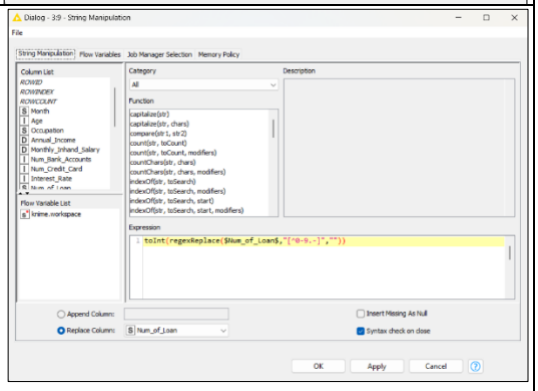
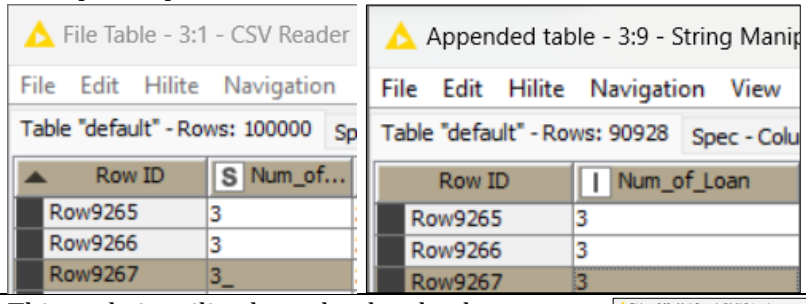
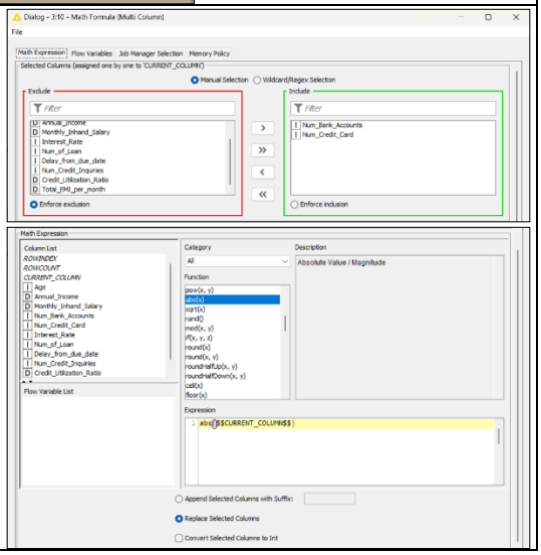
Question 1.4

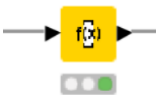
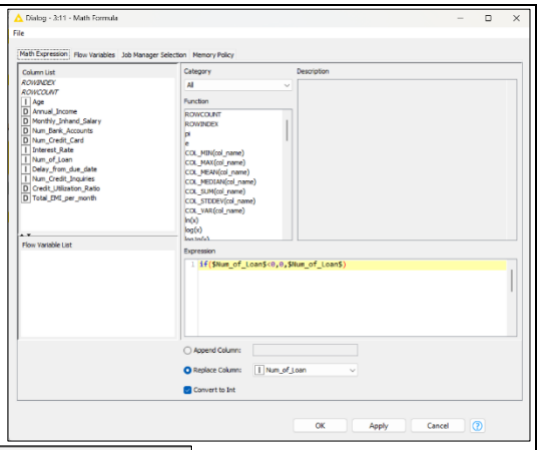
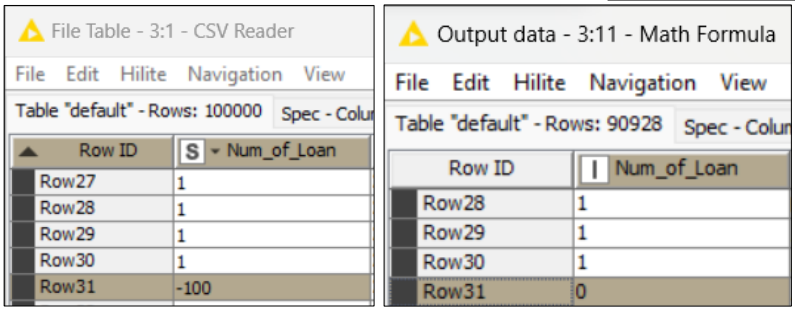
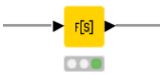
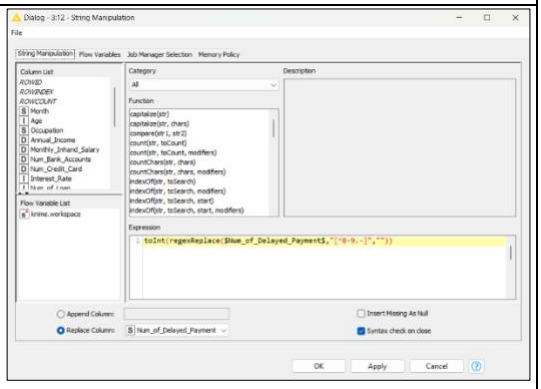
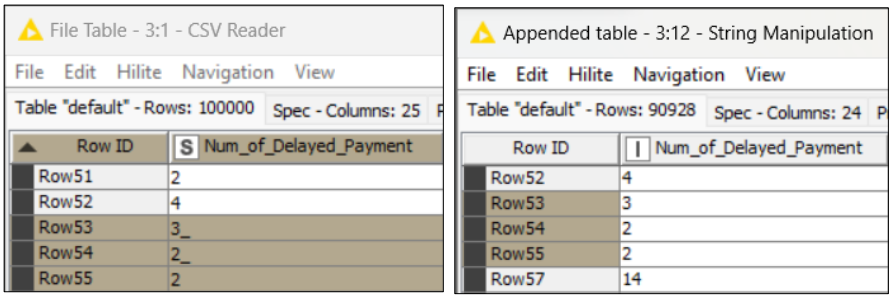
<table>
<tr><th>Node</th><th>Configuration</th></tr>
<tr><td>String Manipulation</td><td>Non-numerical symbols in the “Annual_Income” column are eliminated and the remaining data are converted to the double format. The “.” and “-” are kept because they represent decimal number and negative number, respectively. The new output is replaced to the original “Annual_Income” column.
Expression:
toDouble(regexReplace($Annual_Income$,"[^0-9.-]",""))

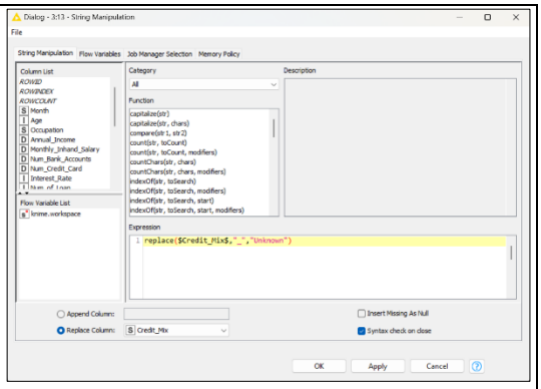
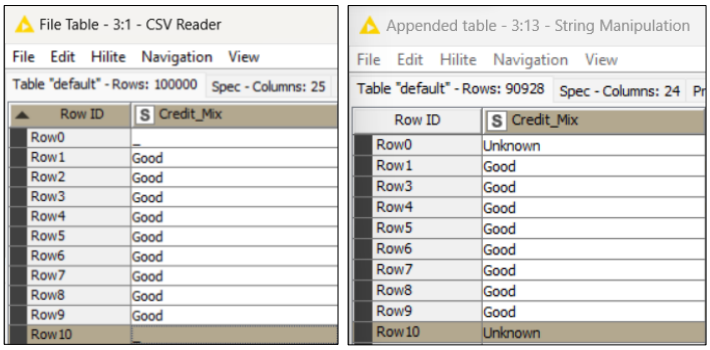
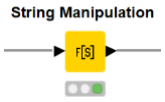
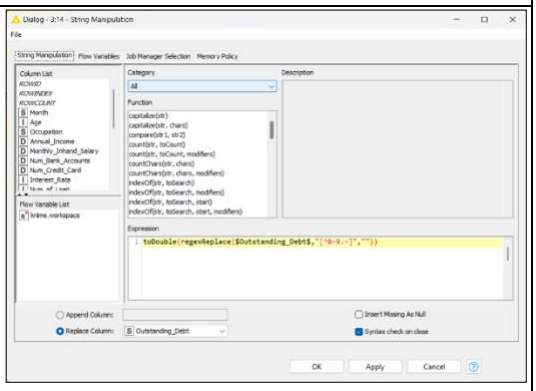
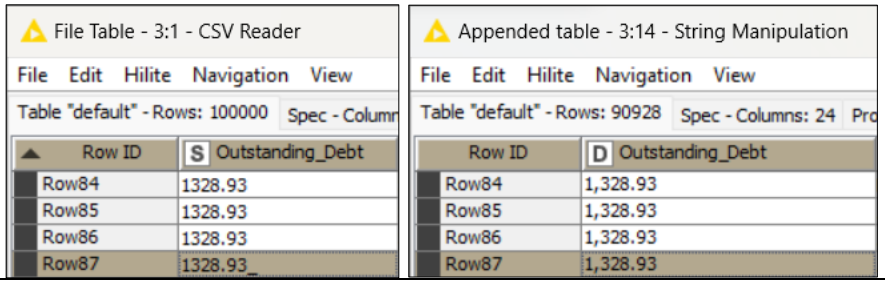
Example output:
 

 </td></tr>
</table>

Question 1.5

<table>
<tr><th>S</th><th>Node</th><th>Configuration</th></tr>
<tr><td>1</td><td>String Manipulation</td><td>“_____” values from the “Occupation” attribute are changed to null. The original “Occupation” column is replaced with the new data.
Expression:
replace($Occupation$,"_____", null)
  </td></tr>
<tr><td>2</td><td>String Manipulation</td><td>Non-numerical symbols in “Num_of_Loan” are removed and the data is converted to integer data type. The “.” and “-” are kept because they represent decimal number and negative number, respectively. The new output is replaced to the original “Num_of_Loan” column.
Expression:
toInt(regexReplace(Num_of_Loan,"[^0-9.-]",""))
 
Example output:
 
</td></tr>
<tr><td>3</td><td>Math Formula (Multi Column)</td><td>This node is utilised to take the absolute values of attributes “Num_Bank_Accounts” and “Num_Credit_Card” using the command as follows: abs($$CURRENT_COLUMN$$)
The original “Num_Bank_Accounts” and “Num_Credit_Card” columns are replaced with the new data.
(negative values of those two attributes are removed by the Rule-based Row Filter node in the question 2)
  </td></tr>
</table>

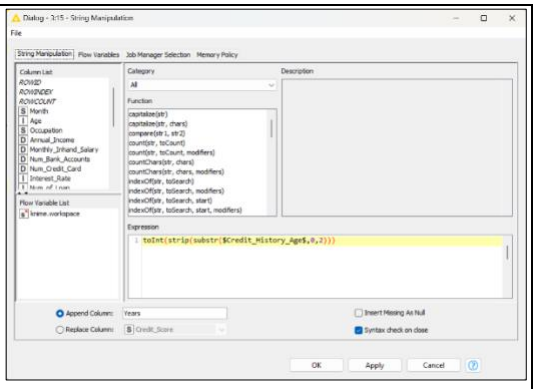
<table>
<tr>
<td>4</td>
<td>Math Formula
  </td>
<td>For the “Num_of_Loan” attribute, if the original values are negative, they are set to 0 and the new data is replaced to the original “Num_of_Loan” column.
Expression:
if(Num_of_Loan<0,0, Num_of_Loan)
 
Example output:
 
File Table - 3:1 - CSV Reader
File Edit Hilite Navigation View
Table "default" - Rows: 100000 Spec - Colu
<table>
<tr><th>Row ID</th><th>S Num_of_Loan</th></tr>
<tr><td>Row27</td><td>1</td></tr>
<tr><td>Row28</td><td>1</td></tr>
<tr><td>Row29</td><td>1</td></tr>
<tr><td>Row30</td><td>1</td></tr>
<tr><td>Row31</td><td>-100</td></tr>
</table>
Output data - 3:11 - Math Formula
File Edit Hilite Navigation View
Table "default" - Rows: 90928 Spec - Colu
<table>
<tr><th>Row ID</th><th>I Num_of_Loan</th></tr>
<tr><td>Row28</td><td>1</td></tr>
<tr><td>Row29</td><td>1</td></tr>
<tr><td>Row30</td><td>1</td></tr>
<tr><td>Row31</td><td>0</td></tr>
</table>
</td>
</tr>
<tr>
<td>5</td>
<td>String Manipulation
  </td>
<td>Non-numerical symbols in “Num_of_Delayed_payment” are taken out and the data is converted to integer. The “.” and “-” are kept because they represent decimal number and negative number, respectively. The new output is replaced to the original “Num_of_Delayed_payment” column.
Expression:
toInt(regexReplace Num_of_Delayed_Payment$,"[^0-9.-]",""))
 
Example output:
 
File Table - 3:1 - CSV Reader
File Edit Hilite Navigation View
Table "default" - Rows: 100000 Spec - Columns: 25
<table>
<tr><th>Row ID</th><th>S Num_of_Delayed_Payment</th></tr>
<tr><td>Row51</td><td>2</td></tr>
<tr><td>Row52</td><td>4</td></tr>
<tr><td>Row53</td><td>3_</td></tr>
<tr><td>Row54</td><td>2_</td></tr>
<tr><td>Row55</td><td>2</td></tr>
</table>
Appended table - 3:12 - String Manipulation
File Edit Hilite Navigation View
Table "default" - Rows: 90928 Spec - Columns: 24
<table>
<tr><th>Row ID</th><th>I Num_of_Delayed_Payment</th></tr>
<tr><td>Row52</td><td>4</td></tr>
<tr><td>Row53</td><td>3</td></tr>
<tr><td>Row54</td><td>2</td></tr>
<tr><td>Row55</td><td>2</td></tr>
<tr><td>Row57</td><td>14</td></tr>
</table>
</td>
</tr>
</table>

<table>
<tr>
<td>6</td>
<td>String Manipulation
  </td>
<td>The “Credit_Mix” values are set to “Unknown” if the original value is “_” using the following expression:
replace($Credit_Mix$, "_","Unknown")
The new output is replaced to the original “Credit_Mix” column.
 
Example output:
 
</td>
</tr>
<tr>
<td>7</td>
<td>String Manipulation
  </td>
<td>The non-numerical symbols in “Outstanding_Debt” attribute are erased and the data is converted to the double format. The “.” and “-” are kept because they represent decimal number and negative number, respectively. The new output is replaced to the original “Outstanding_Debt” column.
Expression:
toDouble(regexReplace Outstanding_Debt$,"[^0-9.-]",""))
 
Example output:
 
</td>
</tr>
</table>

Question 1.6

S	Node	Configuration

String Manipulation



This node is applied to separate the number of year from the *"Credit_History_Age"* string value, erase white spaces and convert it to the number format.
The separated output is appended to the new column named *"Years"*.
Expression:
toInt(strip(substr ($Credit_History_Age$,0,2)))
Two characters counts from the first character (index 0) of a *"Credit_History_Age"* value are taken by using the expression substr(). For example, "9 Years and 6 Months" returns "9 " after substruction. The strip() expression is utilised to remove white spaces. For example, "9 " returns "9" after this function.

Example output:

File Table - 3:1 - CSV Reader

File Edit Hilite Navigation View

Table "default" - Rows: 100000 Spec - Colu

Row ID	Credit_History_...
Row0	22 Years and 1 Months
Row1	NA
Row2	22 Years and 3 Months
Row3	22 Years and 4 Months
Row4	22 Years and 5 Months

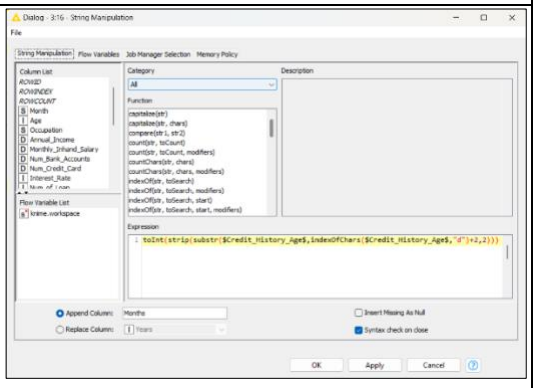
Appended table - 3:15 - Strir

File Edit Hilite Navigation

Table "default" - Rows: 90928 Spe

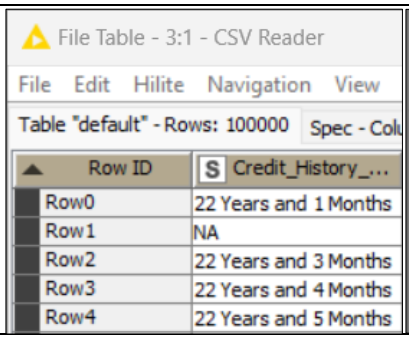
Row ID	Years
Row0	22
Row1	?
Row3	22
Row4	22

String Manipulation



This node is applied to separate the number of month from the *"Credit_History_Age"* string value, erase white spaces and convert it to the number format.
The separated output is appended to the new column named *"Months"*.
Expression:
toInt(strip(substr Credit_History_Age$,indexOfChars ($Credit_History_Age$,"d")+2,2)))
Two characters counts from the first character starting after two characters from the letter "d" (indexOfChars($Credit_History_Age$,"d")+2) of a *"Credit_History_Age"* value are taken by using the expression substr(). For example, "9 Years and 6 Months" returns "6 " after substruction. The strip() expression is utilised to remove white spaces. For example, "6 " returns "6" after this function.

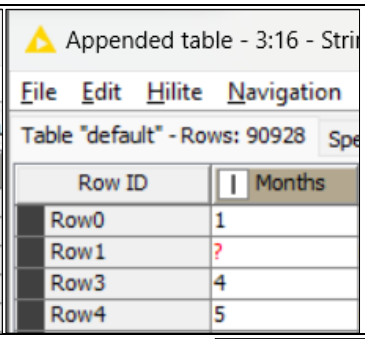
Example output:


File Table - 3:1 - CSV Reader

File Edit Hilite Navigation View

Table "default" - Rows: 100000 Spec - Colu

Row ID	Credit_History_...
Row0	22 Years and 1 Months
Row1	NA
Row2	22 Years and 3 Months
Row3	22 Years and 4 Months
Row4	22 Years and 5 Months

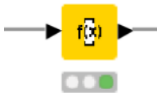

Appended table - 3:16 - Stri

File Edit Hilite Navigation

Table "default" - Rows: 90928 Spe

Row ID	Months
Row0	1
Row1	?
Row3	4
Row4	5

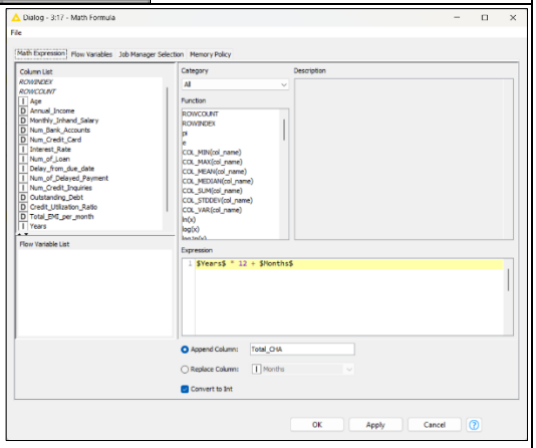
Math Formula



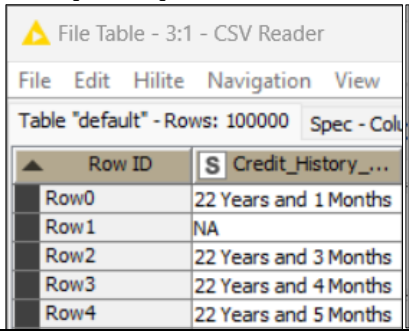
The *"Credit_History_Age"* values are converted to the count of months and store it in the integer format using the following formula:

$Years$ * 12 + $Months$

The converted output is stored in a new column called *"Total_CHA"*.


Dialog - 3:17 - Math Formula

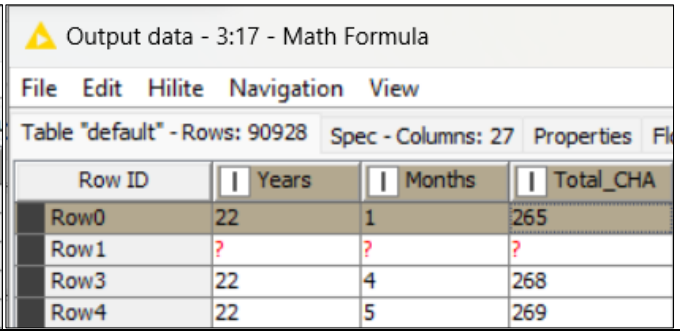
Example output:


File Table - 3:1 - CSV Reader

File Edit Hilite Navigation View

Table "default" - Rows: 100000 Spec - Colu

Row ID	Credit_History_...
Row0	22 Years and 1 Months
Row1	NA
Row2	22 Years and 3 Months
Row3	22 Years and 4 Months
Row4	22 Years and 5 Months

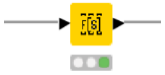
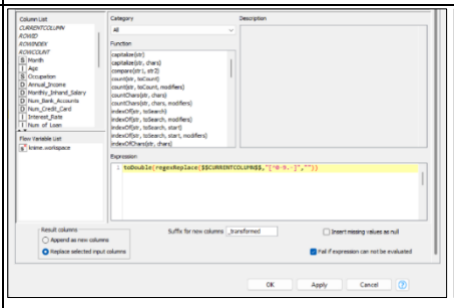
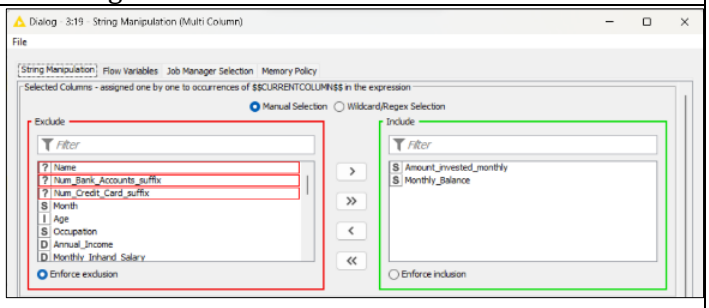

Output data - 3:17 - Math Formula

File Edit Hilite Navigation View

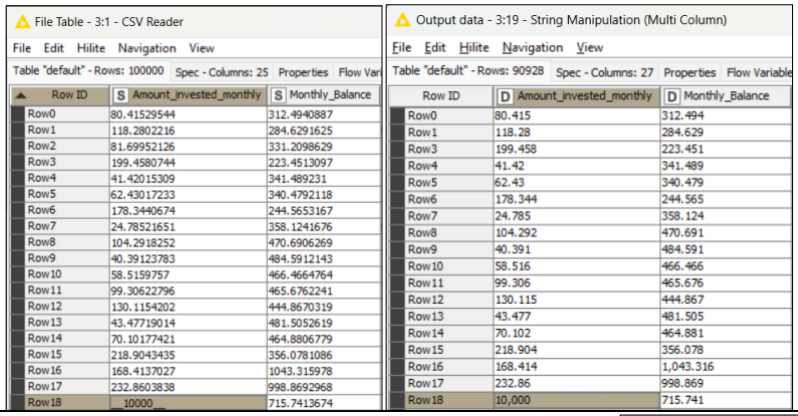
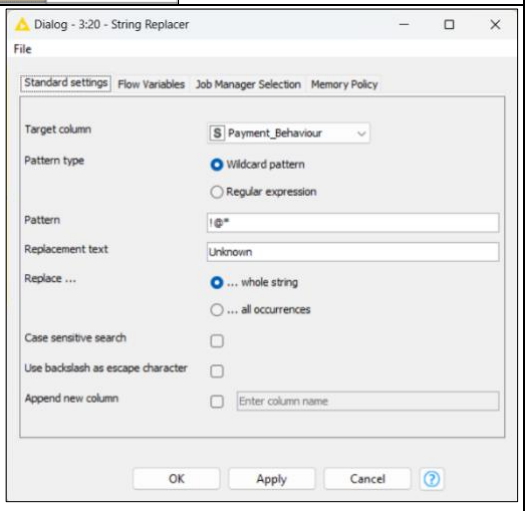
Table "default" - Rows: 90928 Spec - Columns: 27 Properties Fl

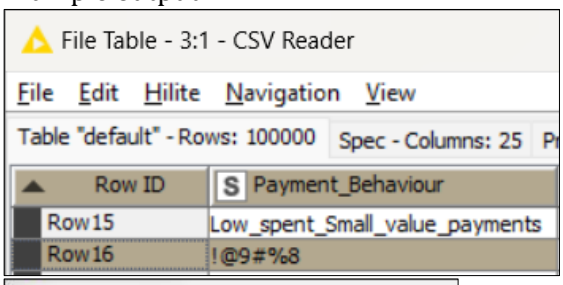
Row ID	Years	Months	Total_CHA
Row0	22	1	265
Row1	?	?	?
Row3	22	4	268
Row4	22	5	269

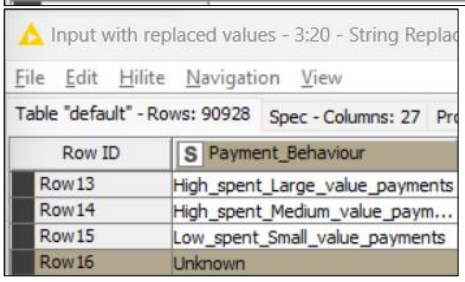
Question 1.7

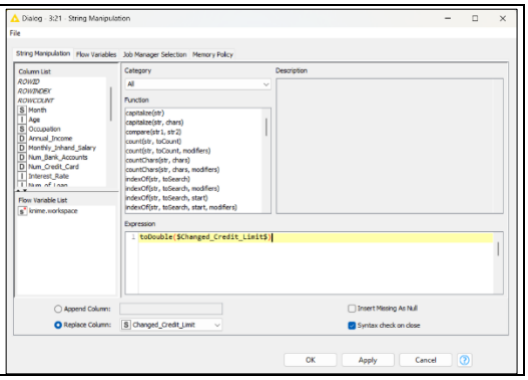
S	Node	Configuration
1	**String Manipulation (Multi Column)** 	  Dialog - 3:19 - String Manipulation (Multi Column) This node is used to remove the non-numerical symbol in two columns: *"Amount_invested_monthly"* and *"Monthly_Balance"* and convert it to the double format. Expression: toDouble(regexReplace($$CURRENTCOLUMN$$,"[^0-9.-]","")) This function is used to apply regex to string and replaces string if regex matches.

<table>
<tr>
<td></td>
<td></td>
<td>The “.” and “-” are kept because they represent decimal number and negative number, respectively. The new output is replaced to their two original columns.

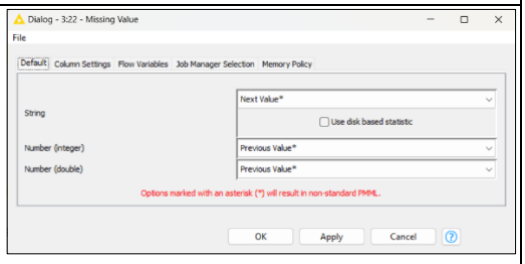
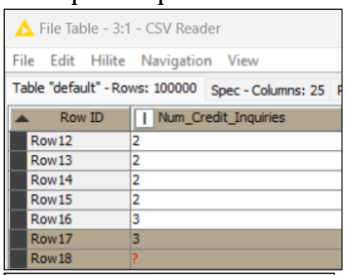
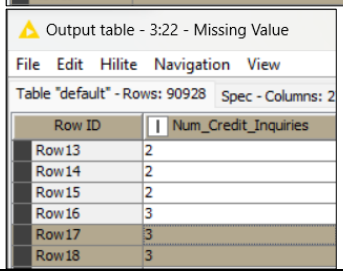
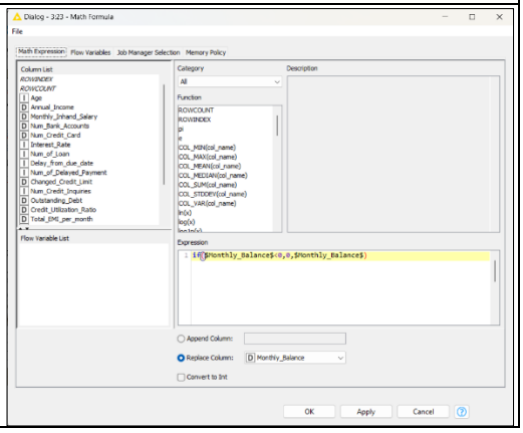
Example output:
 
</td>
</tr>
<tr>
<td>2</td>
<td> 
</td>
<td>This node is used to set the value to “Unknown” if the original value in “Payment_Behaviour” column starts with “!@”. The “*” is applied to find the wildcard pattern in the right order.
 

Example output:
 

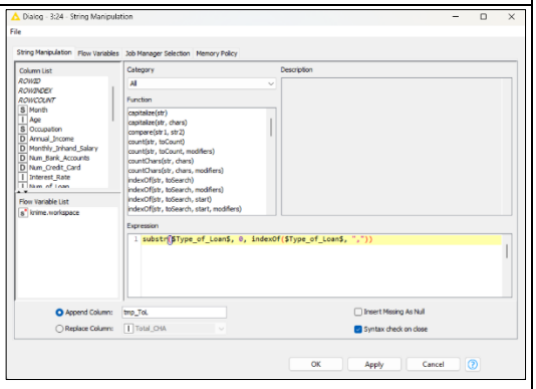
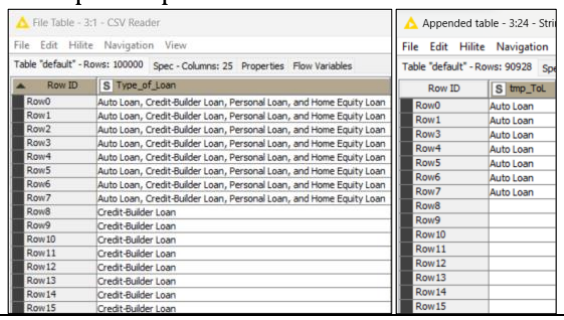
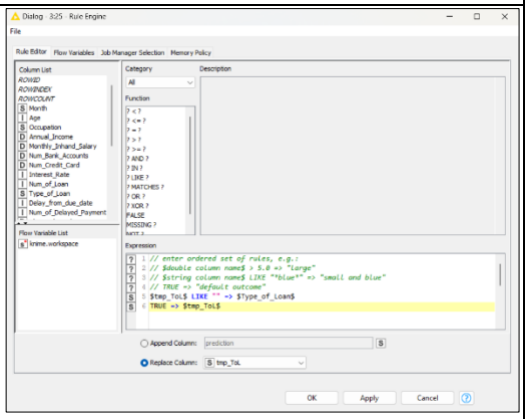
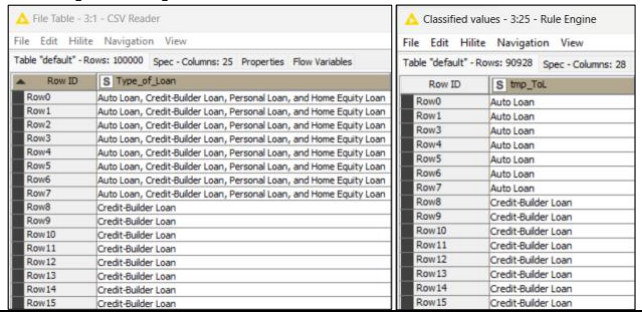
 
</td>
</tr>
</table>

3	String Manipulation	*"Changed_Credit_Limit"* colum is converted to the double format using the expression as follows: toDouble($Changed_Credit_Limit$) The new output is replaced to the original *"Changed_Credit_Limit"* column. 

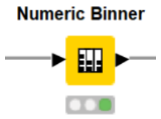
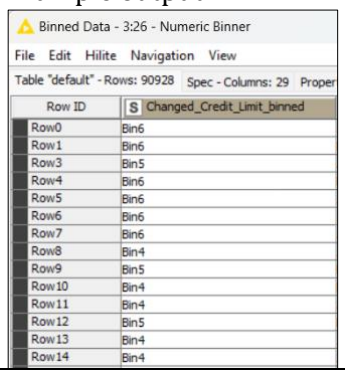
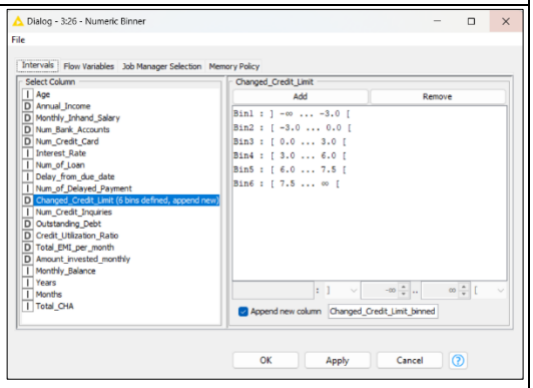
Question 1.8

S	Node	Configuration
1	Missing Value	This node is used to replace the missing values in all string type attributes to the "Next Value*" and all numeric format to the "Previous Value*".  Example output:  File Table - 3:1 - CSV Reader File Edit Hilite Navigation View Table "default" - Rows: 100000 Spec - Columns: 25 Row ID | Num_Credit_Inquiries Row12 2 Row13 2 Row14 2 Row15 2 Row16 3 Row17 3 Row18 ?  Output table - 3:22 - Missing Value File Edit Hilite Navigation View Table "default" - Rows: 90928 Spec - Columns: 2 Row ID | Num_Credit_Inquiries Row13 2 Row14 2 Row15 2 Row16 3 Row17 3 Row18 3
2	Math Formula	The *"Monthly_Balance"* value is replaced with 0 if it is negative using the if condition formula if(condition, true value, false value): if($Monthly_Balance$<0, 0,Num_of_Loan) The new output is replaced to the original *"Monthly_Balance"* columns. 

Question 1.9

<table>
<tr><th>S</th><th>Node</th><th>Configuration</th></tr>
<tr><td>1</td><td>String Manipulation
  </td><td>The “Type_of_Loan” column is simplified using this node. If the original content has more than one type separated by a comma, keep only the first part. Otherwise, keep the full description if there is no comma included. The output result is appended to a new column called “tmp_ToL”.
Expression:
substr($Type_of_Loan$, 0, indexOf($Type_of_Loan$, ","))
This function is applied to substract the string from “Type_of_Loan” column starting from the first character (index 0) to the first comma. For example, “Auto Loan, Auto Loan, and Not Specified” returns “Auto Loan” after substraction and “Credit-Builder Loan” returns missing value because there are not any comma in the value.
 
Example output:
  </td></tr>
<tr><td>2</td><td>Rule Engine
  </td><td>This node is used to recover the missing value from the previous node to its original value from the “Type_of_Loan” column and keep the value subtracted in the previous node.
The new output is replaced to the “tmp_ToL” column.
Expression:
tmp_ToL LIKE "" => $Type_of_Loan$ (the missing values are recovered from the original “Type_of_Loan” attribute)
TRUE => tmp_ToL (the default outcome remains the same)
 
Example output:
  </td></tr>
</table>

Question 1.10

Node	Configuration
Numeric Binner 	This node is utilised to bin the *"Changed_Credit_Limit"* attribute with six bins of ranges: [−∞,−3.0), [−3.0,0), [0,3.0), [3.0,6.0), [6.0,7.5), and [7.5,∞) and put the result into a new attribute called *"Changed_Credit_Limit_binned".* Example output:  

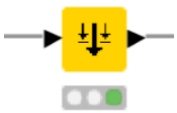
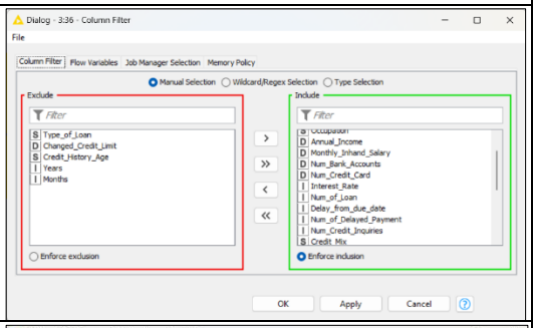
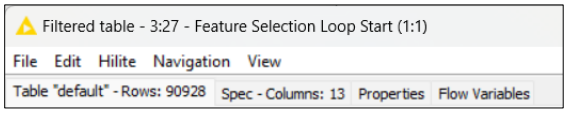
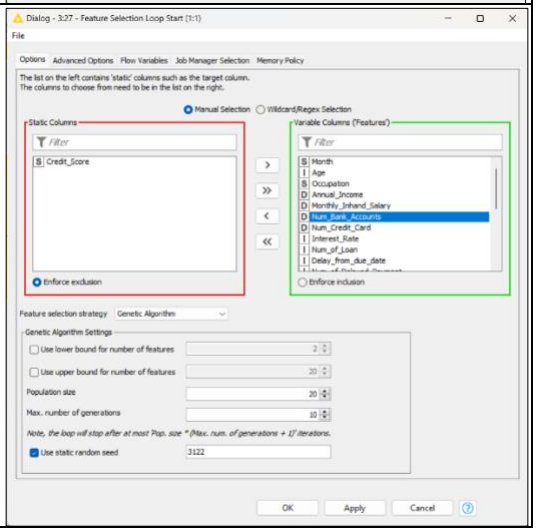
Binned Data - 3:26 - Numeric Binner

File Edit Hilite Navigation View

Table "default" - Rows: 90928 Spec - Columns: 29 Proper

Row ID	Changed_Credit_Limit_binned
Row0	Bin6
Row1	Bin6
Row3	Bin5
Row4	Bin6
Row5	Bin6
Row6	Bin6
Row7	Bin6
Row8	Bin4
Row9	Bin5
Row10	Bin4
Row11	Bin4
Row12	Bin5
Row13	Bin4
Row14	Bin4

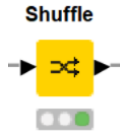
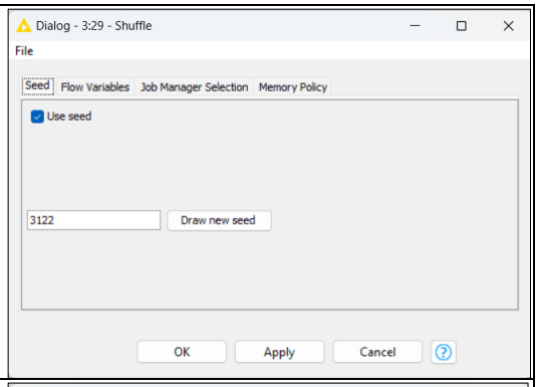
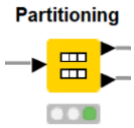
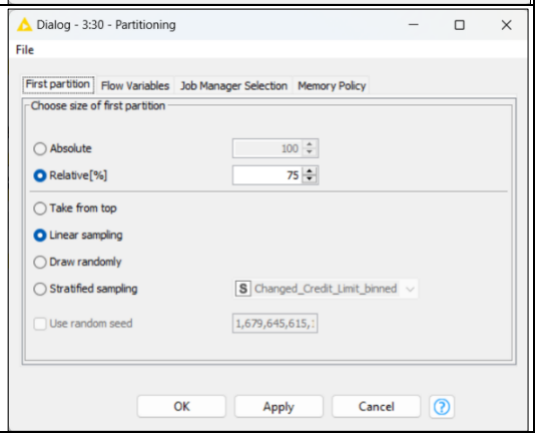
Question 1.11

S	Node	Configuration
1	Column Filter 	The Column Fiter node is used to filter out the original columns that have their new data being appended to new columns (*"Type_of_Loan"*, *"Changed_Credit_Limit"*, *"Credit_History_Age"*) and attributes that are the output of the number of year and month substraction being used for converting the *"Credit_History_Age"* to the count of months (*"Years"*, *"Months"*). 
2	Feature Selection Loop Start (1:1) 	This node is applied to let the algorithm select the attributes using the Genetic Algorithm feature selection strategy with default population size and the maximum number of generations. The class label *"Credit_Score"* is excluded and the static random seed of 3122 is applied. Example output:  

Filtered table - 3:27 - Feature Selection Loop Start (1:1)

File Edit Hilite Navigation View

Table "default" - Rows: 90928 Spec - Columns: 13 Properties Flow Variables

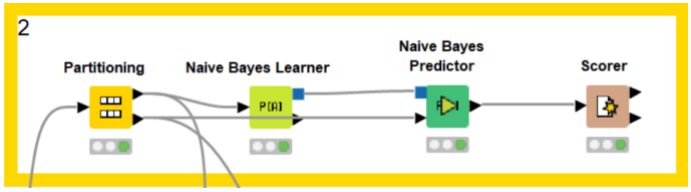
3		The dataset is shuffled with the random seed being set to 3122.	
4		The dataset is partitioned by Linear sampling and a 75:25 ratio, with 75% belongs to the training set and the remaining 25% for the test set.	

IV. DATA ANALYTICS

1. Naïve Bayes Classifier

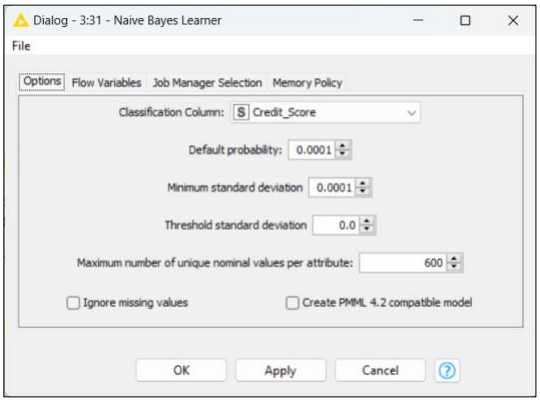
Question 2.1

A screenshot of the Naïve Bayes classifier in the KNIME workflow:



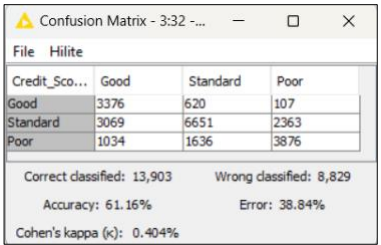
Question 2.2

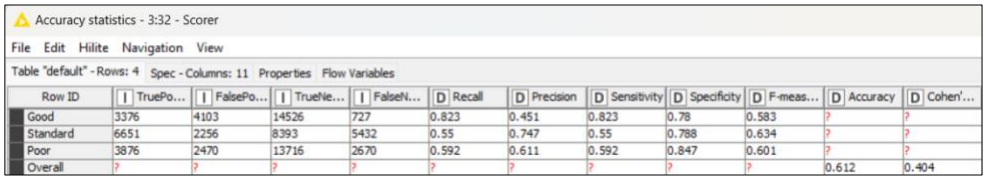
The default probability value and the minimum standard deviation is set to 0.0001. The threshold standard deviation is 0 and the maximum number of unique nominal values per attribute of 600 is applied to the classifier leaner.



Question 2.3

Screenshots of the Confusion Matrix and the Accuracy statistics of the test result:





This Naïve Bayes classifier produces a low precision result of 0.451 which indicates that this classifier performs unsatisfactorily.

Question 2.4

The measurement being looked at to interpret the conclusion in this case is: Precision.

If the bank wants to minimise the risk of lending money to customers, the "Good" in *"Credit_Score"* should be the major target. So, the statistic represents the value being predicted as "Good" are actually "Good" is taken into consideration.

In a confusion matrix, precision is defined as the ratio of true positives (TP: The number of "Good" customers is correctly classified) to the sum of true positives and false positives (FP: The number of "Poor"/"Standard" customers is incorrectly identified as "Good" ones):

$$\text{Precision} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Positive}}$$

Precision measures the proportion of positive predictions (which means the instances that the model classifies as positive) that are actually correct.

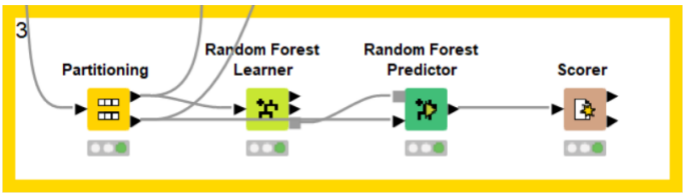
A high precision value indicates that the model is making very few false positive predictions (a few number of "Poor"/"Standard" customer incorrectly classified as "Good" ones) and vice versa.

This Naïve Bayes classifier produces a low precision result of 0.451 which indicates that this classifier perform unsatisfactorily.

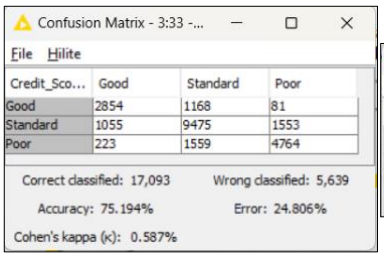
2. Random Forest Classifier

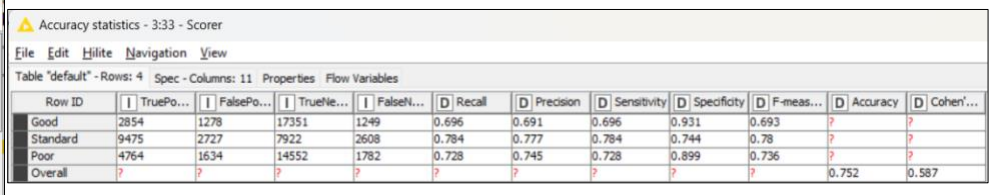
Question 3.1

A screenshot of the Random Forest classifier in the KNIME workflow:



Question 3.2





Question 3.3

The measurement being looked at to interpret the conclusion in this case is: Precision.

If the bank wants to minimise the risk of lending money to customers, the "Good" in *"Credit_Score"* should be the major target. So, the statistic represents the value being predicted as "Good" are actually "Good" is taken into consideration.

In a confusion matrix, precision is defined as the ratio of true positives (TP: The number of "Good" customers is correctly classified) to the sum of true positives and false positives (FP: The number of "Poor"/"Standard" customers is incorrectly identified as "Good" ones):

$$\text{Precision} = \frac{\text{True Positive}}{\text{True Positive + False Positive}}$$

Precision measures the proportion of positive predictions (which means the instances that the model classifies as positive) that are actually correct.

A high precision value indicates that the model is making very few false positive predictions (a few number of "Poor"/"Standard" customer incorrectly classified as "Good" ones) and vice versa.

Random Forest results and Naïve Bayes results Precision comparison:

	Random Forest	Naïve Bayes
Precision (Good)	0.691	0.451
Accuracy (Overall)	0.752	0.612

The Naïve Bayes classifier produces a lower precision (and accuracy) result than the Random Forest one, which indicates that Random Forest model presents a more suitable result compared to Naïve Bayes one.

Question 3.4

Some measurements that should be looked at to find out which class performs the best by the model:

	Precision	Recall	F-measure
Good	0.691	0.696	0.693
Standard	0.777	0.784	0.780
Poor	0.745	0.728	0.736

Precision, Recall and F-measure are common metrics used to evaluate the performance of classification models, including the Random Forest model:

- Precision is used to evaluate how well a model can correctly identify instances of a particular class. It measures the proportion of true positives (instances of the target class that were correctly identified by the model) out of all instances that the model classified as positive for that class. A higher precision means that the model makes fewer false positive errors, which means fewer instances that do not belong to the target class are mistakenly identified as belonging to it.
- Recall measures is used to evaluate how well a model can identify all relevant instances of a particular class. In other words, recall measures the proportion of true positives (instances of the target class that were correctly identified by the model) out of all actual positives (all instances of the target class in the dataset). In this classification problem where all classes are equally important, a higher recall for a particular class would indicate that the model is better at identifying that class.
- F-measure is a metric that combines precision and recall into a single score, which gives equal weight to both measures. It is calculated as the harmonic mean of precision and recall, with values ranging from 0 to 1, where a higher score indicates better performance. Therefore, a higher F-measure for a particular class means that the model is better at identifying instances of that class while also minimising false positives.

Looking at the results, it is obviously that class “Standard” performs the best results.

V. CONCLUSION

In conclusion, this assignment aimed to clean and prepare raw data for future use, and to create two predictive models capable of projecting *"Credit_Score"* classifications. The tasks involved in this project have been thoroughly described in this report, including preparing the data by organising and cleaning it, constructing two models (Naïve Bayes and Random Forest classifier models) to predict and categorise financial credit scores, selecting appropriate features and models for the project, identifying relevant attributes, dividing the dataset into training and testing sets, building predictive models, and providing an explanation of the results. Overall, the report comprehensively showcases and explains the data cleaning and model constructing process.