

045052-p3

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Machine Learning For Managers Project Random Forest

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0.0.1 1. Project Objectives | Problem Statements

1.1. PO1 | PS1: Classification of Consumer Data into Segments | Clusters | Classes using Supervised Learning Classification Algorithms Objective (PO1): The primary objective is to classify consumer data into distinct segments, clusters, or classes using supervised learning classification algorithms.

Problem Statement (PS1): Given the consumer dataset containing various features such as education, marital status, income, etc., the task is to develop a predictive model that accurately classifies consumers into predefined segments or classes based on their characteristics. This involves the application of supervised learning techniques to train a classification model that can effectively distinguish between different consumer groups.

1.2. PO2 | PS2: Determination of an Appropriate Classification Model Objective (PO2): To determine the most suitable classification model for accurately predicting consumer segments based on the dataset features.

Problem Statement (PS2): With several classification algorithms available, including logistic regression, support vector machines, decision trees, and k-nearest neighbors, the goal is to identify the model that yields the highest predictive performance for segmenting consumers. This involves comparing the performance metrics of different classification models and selecting the one that best fits the dataset characteristics.

1.3. PO3 | PS3: Identification of Important | Contributing | Significant Variables or Features and their Thresholds for Classification Objective (PO3): To identify the key variables or features that significantly contribute to the classification of consumers into segments and determine their optimal thresholds.

Problem Statement (PS3): Understanding which features have the most significant impact on segmenting consumers is crucial for model interpretability and performance improvement. Therefore, the objective is to conduct feature importance analysis and determine the thresholds for these important variables, enhancing the classification model's effectiveness in accurately predicting consumer segments.

0.0.2 2. Description of Data

Data Description 1. ID: Unique identifier for each customer. 2. Year_Birth: Year of birth of the customer. 3. Education: Education level of the customer. 4. Marital_Status: Marital status of the customer. 5. Income: Income of the customer. 6. Kidhome: Number of small children in the customer's household. 7. Teenhome: Number of teenagers in the customer's household. 8. Dt_Customer: Date when the customer joined as a customer. 9. Recency: Number of days since the last purchase. 10. MntWines: Amount spent on wines. 11. MntFruits: Amount spent on fruits. 12. MntMeatProducts: Amount spent on meat products. 13. MntFishProducts: Amount spent on fish products. 14. MntSweetProducts: Amount spent on sweet products. 15. MntGoldProds: Amount spent on gold products. 16. NumDealsPurchases: Number of purchases made with discount. 17. NumWebPurchases: Number of purchases made through the website. 18. NumCatalogPurchases: Number of purchases made using a catalog. 19. NumStorePurchases: Number of purchases made directly in stores. 20. NumWebVisitsMonth: Number of visits to the website per month. 21. AcceptedCmp3-5: Binary flags indicating whether the customer accepted marketing campaigns 3 through 5. 22. AcceptedCmp1-2: Binary flags indicating whether the customer accepted marketing campaigns 1 and 2. 23. Complain: Binary flag indicating whether the customer has ever complained. 24. Z_CostContact: Cost of contacting the customer. 25. Z_Revenue: Revenue from contacting the customer. 26. Response: Binary flag indicating whether the customer responded to the campaign.

2.1. Data Source, Size, Shape

- **2.1.1. Data Source (Website Link):** www.Kaggle.com
- **2.1.2. Data Size:** 7.45 MB
- **2.1.3. Data Shape (Dimension: Number of Variables | Number of Records):** (59999, 30)

2.2. Description of Variables

2.2.1. Index Variable(s)

- **Index Variable(s):** S.no

2.2.2. Variables or Features having Categories | Categorical Variables or Features (CV)

- **2.2.2.1. Variables or Features having Nominal Categories | Categorical Variables or Features - Nominal Type:**
 - Year_Birth, Education, Marital_Status, Teenhome, Response, AcceptedCmp3, AcceptedCmp4, AcceptedCmp5, AcceptedCmp1, AcceptedCmp2, Complain

2.2.3. Non-Categorical Variables or Features

- **Non-Categorical Variables or Features:**
 - Income, Kidhome, Recency, MntWines, MntFruits, MntMeatProducts, MntFishProducts, MntSweetProducts, MntGoldProds, NumDealsPurchases, NumWebPurchases, NumCatalogPurchases, NumStorePurchases, NumWebVisitsMonth, Z_CostContact, Z_Revenue

0.0.3 2.3. Descriptive Statistics

2.3.1. Descriptive Statistics: Categorical Variables or Features

2.3.1.1. Count | Frequency Statistics

- **Count | Frequency Statistics:**
 - S.no: Count = 59999
 - Year_Birth: Count = 59999
 - Education: Count = 59999
 - Marital_Status: Count = 59999
 - Teenhome: Count = 59999
 - Response: Count = 59999
 - AcceptedCmp3: Count = 59999
 - AcceptedCmp4: Count = 59999
 - AcceptedCmp5: Count = 59999
 - AcceptedCmp1: Count = 59999
 - AcceptedCmp2: Count = 59999
 - Complain: Count = 59999

2.3.2. Descriptive Statistics: Non-Categorical Variables or Features

2.3.2.1. Measures of Central Tendency

- **Measures of Central Tendency:**
 - S.no: Mean = 30000.0, Min = 1.0, Max = 59999.0
 - Income: Mean = 81359.28, Min = 1000.0, Max = 666480.0
 - Kidhome: Mean = 0.441457, Min = 0.0, Max = 2.0
 - Recency: Mean = 48.698862, Min = 0.0, Max = 103.0
 - MntWines: Mean = 308.099902, Min = 0.0, Max = 1540.0
 - MntFruits: Mean = 26.573010, Min = 0.0, Max = 203.0
 - MntMeatProducts: Mean = 166.544309, Min = 0.0, Max = 1731.0
 - MntFishProducts: Mean = 40.871165, Min = 0.0, Max = 278.0
 - MntSweetProducts: Mean = 28.414324, Min = 0.0, Max = 271.0
 - MntGoldProds: Mean = 43.856964, Min = 0.0, Max = 368.0
 - NumDealsPurchases: Mean = 2.375523, Min = 0.0, Max = 17.0
 - NumWebPurchases: Mean = 3.831597, Min = 0.0, Max = 29.0
 - NumCatalogPurchases: Mean = 2.949816, Min = 0.0, Max = 30.0
 - NumStorePurchases: Mean = 5.341272, Min = 0.0, Max = 15.0
 - NumWebVisitsMonth: Mean = 4.919599, Min = 0.0, Max = 22.0
 - Z_CostContact: Mean = 3.0, Min = 3.0, Max = 3.0
 - Z_Revenue: Mean = 11.0, Min = 11.0, Max = 11.0

0.0.4 3. Analysis of Data

3.1. Data Pre-Processing

3.1.1. Missing Data Statistics and Treatment 3.1.1.1. Missing Data Statistics: Records

The analysis of missing data at the record level indicates a complete dataset with no missing records. This observation is crucial as it ensures the integrity and reliability of the dataset, providing a solid foundation for subsequent analysis and modeling. The absence of missing records simplifies data processing and interpretation, allowing for a comprehensive understanding of the entire dataset without the need for imputation or removal of incomplete records.

3.1.1.2. Missing Data Treatment: Records

Given that no records contain more than 50% missing data, no removal or imputation procedures were deemed necessary. This absence of significant missing data in records alleviates concerns regarding potential biases introduced by incomplete observations. Consequently, the dataset remains intact, preserving the integrity and comprehensiveness of the information contained within each record.

3.1.1.3. Missing Data Statistics and Treatment: Categorical Variables or Features

Analysis of missing data pertaining to categorical variables reveals no instances of missing values across the dataset. This finding underscores the completeness of categorical data, indicating that all categorical variables have been fully populated. As a result, no removal or imputation of missing values was required, ensuring the preservation of categorical data integrity and the reliability of subsequent analyses involving categorical variables.

3.1.2. Numerical Encoding of Categorical Variables or Features (Encoding Schema - Alphanumeric Order) To facilitate further analysis and modeling, categorical variables were numerically encoded using the LabelEncoder from the scikit-learn library. This transformation converts categorical variables into numerical format, allowing for compatibility with machine learning algorithms that require numeric inputs. Each categorical variable was encoded alphabetically, ensuring consistency and ease of interpretation in subsequent analyses.

3.1.3. Outlier Statistics and Treatment (Scaling | Transformation) Detailed analysis and treatment of outliers in non-categorical variables were not provided in the current context. While outliers can impact the performance and accuracy of predictive models, their detection and treatment require careful consideration and domain knowledge. Future analyses may benefit from exploring outlier detection techniques such as z-score, interquartile range, or visual inspection through box plots. Treatment methods such as scaling or t

0.0.5 3.2 Data Analysis

In the data analysis section, we delve into the characteristics of the dataset and gain insights into the features that significantly influence the classification process.

Overview of Features Upon analyzing the dataset, we identified two primary features that play a crucial role in classifying consumer segments: - **Income:** The Random Forest model assigns a high importance score of 0.9906 to income, indicating that income levels have a profound impact on the classification of consumers into segments. This suggests that consumers with varying income levels exhibit distinct behaviors or characteristics that allow for effective segmentation.

- **NumWebVisitsMonth:** While relatively less influential compared to income (importance score: 0.0094), the number of web visits per month also contributes to the classification

process. This feature may capture consumer engagement with online platforms, reflecting different levels of interest or interaction.

Model Understanding The decision trees generated by the Random Forest model provide insights into the hierarchical structure used for classification based on the most discriminative features (Income and NumWebVisitsMonth). By examining the tree structures, we can observe how different thresholds of these features are used to partition the data into distinct segments or classes.

0.0.6 4. Results | Observations

The results and observations section summarizes the performance of the Random Forest model on both the training and testing datasets, highlighting key metrics and outcomes.

Training Set Results The Random Forest model demonstrates high accuracy and robust performance on the training set: - **Accuracy:** The model achieves an impressive accuracy of approximately 95% on the training data, indicating effective learning and generalization from the provided features.

- **Confusion Matrix and Classification Report:** Analysis of the confusion matrix and classification report for the training set reveals consistent high precision, recall, and F1-scores across all target classes (0, 1, 2). This suggests that the model effectively learns to distinguish and classify different consumer segments based on the training data.

Testing Set Results Evaluation of the Random Forest model on the testing dataset reveals important insights: - **Accuracy:** The model's accuracy on the testing set is notably lower at around 34%, highlighting challenges in generalizing to unseen data or potential overfitting to the training set.

- **Confusion Matrix and Classification Report:** Similar performance metrics (precision, recall, F1-score) are observed across classes in the confusion matrix and classification report for the testing set. This indicates the need for further model refinement or exploration of alternative approaches to enhance generalization capability.

0.0.7 5. Managerial Insights

In the managerial insights section, we translate the findings and results into actionable recommendations for strategic decision-making.

Key Insights Based on the analysis and results, several key insights emerge: - **Feature Importance:** The significant influence of income underscores the importance of income-related strategies in targeting specific consumer segments effectively. Tailoring marketing approaches or product offerings based on income brackets could yield substantial benefits.

- **Performance Challenges:** The observed performance gap between training and testing sets necessitates further investigation into model refinement and validation techniques. Exploring feature engineering or alternative ensemble methods could aid in mitigating overfitting and improving generalization.

Recommendations Drawing from the insights gained, actionable recommendations can be formulated: - **Feature Engineering:** Consider augmenting the dataset with additional relevant features or conducting feature transformations to enhance the model's ability to capture underlying patterns and improve predictive performance.

- **Ensemble Model Tuning:** Explore hyperparameter tuning or alternative ensemble methods beyond Random Forest to optimize model performance and robustness.

Strategic Implications Translate the technical findings into strategic implications for decision-makers: - **Strategic Targeting:** Leveraging income-based segmentation strategies can inform targeted marketing campaigns or pricing strategies tailored to different consumer groups.

- **Continuous Improvement:** Emphasize the iterative nature of model refinement and validation to adapt to specific project objectives and stakeholder requirements. Offer customization, and service offerings tailored to the specific needs and preferences of each cluster. Further analysis and refinement of the segmentation strategy can provide valuable insights for maximizing customer satisfaction and business growth.

```
[1]: # Import library
# Required Libraries
import pandas as pd, numpy as np # For Data Manipulation
from sklearn.preprocessing import LabelEncoder, OrdinalEncoder # For Encoding
    ↳ Categorical Data [Nominal | Ordinal]
from sklearn.preprocessing import OneHotEncoder # For Creating Dummy Variables
    ↳ of Categorical Data [Nominal]
from sklearn.impute import SimpleImputer, KNNImputer # For Imputation of
    ↳ Missing Data
from sklearn.preprocessing import StandardScaler, MinMaxScaler, RobustScaler #
    ↳ For Rescaling Data
from sklearn.model_selection import train_test_split # For Splitting Data into
    ↳ Training & Testing Sets
import matplotlib.pyplot as plt
import numpy as np
from scipy.stats import pearsonr
from scipy import stats

# Required Libraries
import pandas as pd, numpy as np # For Data Manipulation
import matplotlib.pyplot as plt, seaborn as sns # For Data Visualization
import scipy.cluster.hierarchy as sch # For Hierarchical Clustering
from sklearn.cluster import AgglomerativeClustering as agclus, KMeans as kmclus
    ↳ # For Agglomerative & K-Means Clustering
from sklearn.metrics import silhouette_score as sscore, davies_bouldin_score as
    ↳ dbscore # For Clustering Model Evaluation

# @title load library { display-mode: "form" }
# Load IPython extension for measuring time
!pip install ipython-autotime
```

```

%reload_ext autotime

# Load IPython extension for memory profiling
!pip install memory-profiler
%reload_ext memory_profiler

# Your imports
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import scipy.cluster.hierarchy as sch
from sklearn.cluster import AgglomerativeClustering as agclus, KMeans as kmclus
from sklearn.metrics import silhouette_score as sscore, davies_bouldin_score as d_
    db_score
from scipy.cluster.hierarchy import dendrogram, linkage
import plotly.graph_objects as go

# Load preprocessing libraries
from sklearn.preprocessing import LabelEncoder, OrdinalEncoder, OneHotEncoder
from sklearn.impute import SimpleImputer, KNNImputer
from sklearn.preprocessing import StandardScaler, MinMaxScaler, RobustScaler
from sklearn.model_selection import train_test_split

from scipy.stats import f_oneway
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.impute import SimpleImputer
from sklearn.preprocessing import OneHotEncoder, StandardScaler
from sklearn.model_selection import train_test_split
import warnings
warnings.filterwarnings('ignore')

```

```

Requirement already satisfied: ipython-autotime in
c:\users\shantanu\anaconda3\lib\site-packages (0.3.2)
Requirement already satisfied: ipython in c:\users\shantanu\anaconda3\lib\site-
packages (from ipython-autotime) (8.15.0)
Requirement already satisfied: backcall in c:\users\shantanu\anaconda3\lib\site-
packages (from ipython->ipython-autotime) (0.2.0)
Requirement already satisfied: decorator in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(5.1.1)
Requirement already satisfied: jedi>=0.16 in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(0.18.1)

```

Requirement already satisfied: matplotlib-inline in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(0.1.6)

Requirement already satisfied: pickleshare in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(0.7.5)

Requirement already satisfied: prompt-toolkit!=3.0.37,<3.1.0,>=3.0.30 in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(3.0.36)

Requirement already satisfied: pygments>=2.4.0 in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(2.15.1)

Requirement already satisfied: stack-data in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(0.2.0)

Requirement already satisfied: traitlets>=5 in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(5.7.1)

Requirement already satisfied: exceptiongroup in
c:\users\shantanu\anaconda3\lib\site-packages (from ipython->ipython-autotime)
(1.0.4)

Requirement already satisfied: colorama in c:\users\shantanu\anaconda3\lib\site-
packages (from ipython->ipython-autotime) (0.4.6)

Requirement already satisfied: parso<0.9.0,>=0.8.0 in
c:\users\shantanu\anaconda3\lib\site-packages (from
jedi>=0.16->ipython->ipython-autotime) (0.8.3)

Requirement already satisfied: wcwidth in c:\users\shantanu\anaconda3\lib\site-
packages (from prompt-toolkit!=3.0.37,<3.1.0,>=3.0.30->ipython->ipython-
autotime) (0.2.5)

Requirement already satisfied: executing in
c:\users\shantanu\anaconda3\lib\site-packages (from stack-
data->ipython->ipython-autotime) (0.8.3)

Requirement already satisfied: asttokens in
c:\users\shantanu\anaconda3\lib\site-packages (from stack-
data->ipython->ipython-autotime) (2.0.5)

Requirement already satisfied: pure-eval in
c:\users\shantanu\anaconda3\lib\site-packages (from stack-
data->ipython->ipython-autotime) (0.2.2)

Requirement already satisfied: six in c:\users\shantanu\anaconda3\lib\site-
packages (from asttokens->stack-data->ipython->ipython-autotime) (1.16.0)

WARNING: Ignoring invalid distribution -atplotlib
(c:\users\shantanu\anaconda3\lib\site-packages)

WARNING: Ignoring invalid distribution -atplotlib
(c:\users\shantanu\anaconda3\lib\site-packages)

WARNING: Ignoring invalid distribution -atplotlib
(c:\users\shantanu\anaconda3\lib\site-packages)

WARNING: Ignoring invalid distribution -atplotlib
(c:\users\shantanu\anaconda3\lib\site-packages)


```
(c:\users\shantanu\anaconda3\lib\site-packages)
```

```
Requirement already satisfied: memory-profiler in  
c:\users\shantanu\anaconda3\lib\site-packages (0.61.0)  
Requirement already satisfied: psutil in c:\users\shantanu\anaconda3\lib\site-  
packages (from memory-profiler) (5.9.0)  
time: 3.84 s (started: 2024-04-13 18:11:00 +05:30)
```

```
[2]: # Load the CSV file into a pandas DataFrame  
df = pd.read_csv(r"C:\Users\Shantanu\Downloads\new_marketing_campaign.csv")
```

```
time: 344 ms (started: 2024-04-13 18:11:04 +05:30)
```

```
[3]: df
```

```
[3]:
```

	ID	Year_Birth	Education	\
0	e20d011e-c904-4301-871d-4120e734f189	1956	2n Cycle	
1	9038499f-e82e-43d3-803d-165dba9afe48	1950	Graduation	
2	a3aaaf2a-c063-45ba-ac1a-ab375d53faa3	1960	PhD	
3	fabc6e91-6e59-4dd6-8d07-572b7f07b9e4	1980	PhD	
4	6bf0b99f-dd6d-4964-a3ca-1e98f500d9e1	1977	Graduation	
...	
59994	befde079-3b10-4bff-ac11-fb4390b9b90c	1967	PhD	
59995	5d7e372b-b857-46b4-be71-be4a46c79c88	1974	Graduation	
59996	22cf1a64-8133-466d-af61-d0475ed92b80	1963	Master	
59997	6304a2fb-4c6e-46cb-ac70-fcacd96278f4	1974	Master	
59998	b67a9fe3-6f44-422c-981d-aa52925c3bb5	1951	Graduation	

	Marital_Status	Income	Kidhome	Teenhome	Dt_Customer	Recency	\
0	Together	40831	0	1	21-07-2022	62	
1	Together	22537	1	1	05-11-2021	42	
2	Together	195755	1	0	24-12-2015	28	
3	Divorced	143143	0	0	14-06-2017	24	
4	Married	84648	0	1	17-06-2021	92	
...	
59994	Married	2688	0	1	19-08-2018	30	
59995	Together	14577	1	0	01-05-2017	43	
59996	Single	3282	1	1	05-03-2022	80	
59997	Single	68946	1	0	18-08-2020	11	
59998	Married	131663	0	0	15-04-2016	50	

	MntWines	...	NumWebVisitsMonth	AcceptedCmp3	AcceptedCmp4	\
0	636	...	5	1	1	
1	6	...	4	0	1	
2	383	...	4	0	1	
3	29	...	5	1	1	
4	133	...	7	1	1	
...	

59994	13	...	7	1	1
59995	10	...	4	0	1
59996	33	...	7	0	0
59997	37	...	6	0	1
59998	19	...	1	1	1

	AcceptedCmp5	AcceptedCmp1	AcceptedCmp2	Complain	Z_CostContact	\
0	1	1	1	0		3
1	1	0	0	0		3
2	0	1	1	0		3
3	0	0	0	0		3
4	1	1	1	0		3
...	
59994	1	1	0	0		3
59995	1	1	0	0		3
59996	1	1	0	0		3
59997	0	1	1	0		3
59998	0	1	0	0		3

	Z_Revenue	Response
0	11	1
1	11	1
2	11	1
3	11	1
4	11	0
...
59994	11	0
59995	11	0
59996	11	0
59997	11	1
59998	11	0

[59999 rows x 29 columns]

time: 94 ms (started: 2024-04-13 18:11:04 +05:30)

```
[4]: # Add a new column named "S.no" as the first column with serial numbers
df.insert(0, "S.no", range(1, len(df) + 1))

# Now the DataFrame df_ppd_subset will have a new column "S.no" as the first_
↪column
print(df.head())
```

	S.no	ID	Year_Birth	Education	\
0	1	e20d011e-c904-4301-871d-4120e734f189	1956	2n Cycle	
1	2	9038499f-e82e-43d3-803d-165dba9afe48	1950	Graduation	
2	3	a3aaaf2a-c063-45ba-ac1a-ab375d53faa3	1960	PhD	

```

3      4  fabc6e91-6e59-4dd6-8d07-572b7f07b9e4      1980      PhD
4      5  6bf0b99f-dd6d-4964-a3ca-1e98f500d9e1      1977  Graduation

```

```

Marital_Status  Income  Kidhome  Teenhome  Dt_Customer  Recency  ...  \
0      Together   40831         0         1  21-07-2022    62  ...
1      Together   22537         1         1  05-11-2021    42  ...
2      Together   195755        1         0  24-12-2015    28  ...
3      Divorced   143143         0         0  14-06-2017    24  ...
4      Married    84648         0         1  17-06-2021    92  ...

```

```

NumWebVisitsMonth  AcceptedCmp3  AcceptedCmp4  AcceptedCmp5  AcceptedCmp1  \
0                  5             1             1             1
1                  4             0             1             0
2                  4             0             1             1
3                  5             1             1             0
4                  7             1             1             1

```

```

AcceptedCmp2  Complain  Z_CostContact  Z_Revenue  Response
0             1         0              3          11          1
1             0         0              3          11          1
2             1         0              3          11          1
3             0         0              3          11          1
4             1         0              3          11          0

```

[5 rows x 30 columns]

time: 16 ms (started: 2024-04-13 18:11:04 +05:30)

```
[5]: df.head()
```

```

[5]:      S.no      ID  Year_Birth  Education  \
0      1  e20d011e-c904-4301-871d-4120e734f189      1956      2n Cycle
1      2  9038499f-e82e-43d3-803d-165dba9afe48      1950  Graduation
2      3  a3aaaf2a-c063-45ba-ac1a-ab375d53faa3      1960      PhD
3      4  fabc6e91-6e59-4dd6-8d07-572b7f07b9e4      1980      PhD
4      5  6bf0b99f-dd6d-4964-a3ca-1e98f500d9e1      1977  Graduation

```

```

Marital_Status  Income  Kidhome  Teenhome  Dt_Customer  Recency  ...  \
0      Together   40831         0         1  21-07-2022    62  ...
1      Together   22537         1         1  05-11-2021    42  ...
2      Together   195755        1         0  24-12-2015    28  ...
3      Divorced   143143         0         0  14-06-2017    24  ...
4      Married    84648         0         1  17-06-2021    92  ...

```

```

NumWebVisitsMonth  AcceptedCmp3  AcceptedCmp4  AcceptedCmp5  AcceptedCmp1  \
0                  5             1             1             1
1                  4             0             1             0
2                  4             0             1             1

```

3	5	1	1	0	0
4	7	1	1	1	1

	AcceptedCmp2	Complain	Z_CostContact	Z_Revenue	Response
0	1	0	3	11	1
1	0	0	3	11	1
2	1	0	3	11	1
3	0	0	3	11	1
4	1	0	3	11	0

[5 rows x 30 columns]

time: 15 ms (started: 2024-04-13 18:11:04 +05:30)

```
[6]: df.info()
list(df.columns)

# Assuming df is your original DataFrame
# Add your normalization or standardization code here

# Display summary statistics
df.describe()

total_records = len(df)
print(f"Total number of records: {total_records}")

# Calculate the total number of filled cells in each column
filled_cells_count = df.count()

# Sum up the counts to get the total number of filled cells in the DataFrame
total_filled_cells = filled_cells_count.sum()

print(f"Total number of filled cells: {total_filled_cells}")

# Assuming df is your DataFrame
unique_counts = df.nunique()

# Display the number of unique values in each column
print(unique_counts)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 59999 entries, 0 to 59998
Data columns (total 30 columns):
#   Column                Non-Null Count  Dtype
---  -
0   S.no                   59999 non-null  int64
1   ID                     59999 non-null  object
2   Year_Birth             59999 non-null  int64
```

3	Education	59999	non-null	object
4	Marital_Status	59999	non-null	object
5	Income	59999	non-null	int64
6	Kidhome	59999	non-null	int64
7	Teenhome	59999	non-null	int64
8	Dt_Customer	59999	non-null	object
9	Recency	59999	non-null	int64
10	MntWines	59999	non-null	int64
11	MntFruits	59999	non-null	int64
12	MntMeatProducts	59999	non-null	int64
13	MntFishProducts	59999	non-null	int64
14	MntSweetProducts	59999	non-null	int64
15	MntGoldProds	59999	non-null	int64
16	NumDealsPurchases	59999	non-null	int64
17	NumWebPurchases	59999	non-null	int64
18	NumCatalogPurchases	59999	non-null	int64
19	NumStorePurchases	59999	non-null	int64
20	NumWebVisitsMonth	59999	non-null	int64
21	AcceptedCmp3	59999	non-null	int64
22	AcceptedCmp4	59999	non-null	int64
23	AcceptedCmp5	59999	non-null	int64
24	AcceptedCmp1	59999	non-null	int64
25	AcceptedCmp2	59999	non-null	int64
26	Complain	59999	non-null	int64
27	Z_CostContact	59999	non-null	int64
28	Z_Revenue	59999	non-null	int64
29	Response	59999	non-null	int64

dtypes: int64(26), object(4)

memory usage: 13.7+ MB

Total number of records: 59999

Total number of filled cells: 1799970

S.no	59999
ID	59999
Year_Birth	83
Education	5
Marital_Status	8
Income	40412
Kidhome	3
Teenhome	3
Dt_Customer	4028
Recency	104
MntWines	1508
MntFruits	204
MntMeatProducts	1033
MntFishProducts	278
MntSweetProducts	226
MntGoldProds	303
NumDealsPurchases	18

```

NumWebPurchases      24
NumCatalogPurchases  26
NumStorePurchases     16
NumWebVisitsMonth     23
AcceptedCmp3          2
AcceptedCmp4          2
AcceptedCmp5          2
AcceptedCmp1          2
AcceptedCmp2          2
Complain              2
Z_CostContact         1
Z_Revenue              1
Response              2
dtype: int64
time: 344 ms (started: 2024-04-13 18:11:04 +05:30)

```

```

[7]: # Assuming df is your DataFrame
columns_list = df.columns.tolist()
columns_list

list(df.columns)

```

```

[7]: ['S.no',
      'ID',
      'Year_Birth',
      'Education',
      'Marital_Status',
      'Income',
      'Kidhome',
      'Teenhome',
      'Dt_Customer',
      'Recency',
      'MntWines',
      'MntFruits',
      'MntMeatProducts',
      'MntFishProducts',
      'MntSweetProducts',
      'MntGoldProds',
      'NumDealsPurchases',
      'NumWebPurchases',
      'NumCatalogPurchases',
      'NumStorePurchases',
      'NumWebVisitsMonth',
      'AcceptedCmp3',
      'AcceptedCmp4',
      'AcceptedCmp5',
      'AcceptedCmp1',

```

```
'AcceptedCmp2',  
'Complain',  
'Z_CostContact',  
'Z_Revenue',  
'Response']
```

time: 16 ms (started: 2024-04-13 18:11:05 +05:30)

```
[8]: # Nominal and Ordinal Columns  
  
# Continuous and Non Continuous Columns  
  
import pandas as pd  
  
# Assuming df is your DataFrame  
continuous_columns = df.select_dtypes(include=['float64', 'int64']).columns  
non_continuous_columns = df.select_dtypes(exclude=['float64', 'int64']).columns  
  
print("Continuous Columns:", list(continuous_columns))  
print("Non-Continuous Columns:", list(non_continuous_columns))  
  
# Assuming df is your DataFrame  
categorical_columns = df.select_dtypes(include=['object', 'category']).columns  
non_categorical_columns = df.select_dtypes(exclude=['object', 'category']).  
    columns  
  
print("Categorical Columns:", list(categorical_columns))  
print("Non-Categorical Columns:", list(non_categorical_columns))
```

```
Continuous Columns: ['S.no', 'Year_Birth', 'Income', 'Kidhome', 'Teenhome',  
'Recency', 'MntWines', 'MntFruits', 'MntMeatProducts', 'MntFishProducts',  
'MntSweetProducts', 'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',  
'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth', 'AcceptedCmp3',  
'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1', 'AcceptedCmp2', 'Complain',  
'Z_CostContact', 'Z_Revenue', 'Response']  
Non-Continuous Columns: ['ID', 'Education', 'Marital_Status', 'Dt_Customer']  
Categorical Columns: ['ID', 'Education', 'Marital_Status', 'Dt_Customer']  
Non-Categorical Columns: ['S.no', 'Year_Birth', 'Income', 'Kidhome', 'Teenhome',  
'Recency', 'MntWines', 'MntFruits', 'MntMeatProducts', 'MntFishProducts',  
'MntSweetProducts', 'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',  
'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth', 'AcceptedCmp3',  
'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1', 'AcceptedCmp2', 'Complain',  
'Z_CostContact', 'Z_Revenue', 'Response']  
time: 62 ms (started: 2024-04-13 18:11:05 +05:30)
```

```
[9]: ### Missing Data Statistics and Treatment
### Missing Data Statistics: Records

# Assuming df is your DataFrame

# Count the missing values in each column
missing_data = df.isnull().sum()

# Create a DataFrame to display missing data statistics
missing_data_stats = pd.DataFrame({
    'Column': missing_data.index,
    'Missing Records': missing_data.values,
    'Percentage Missing': (missing_data / len(df)) * 100
})

# Sort the DataFrame by the percentage of missing values in descending order
missing_data_stats = missing_data_stats.sort_values(by='Percentage Missing',
↪ascending=False)

# Print the missing data statistics
print(missing_data_stats)
```

	Column	Missing Records	Percentage Missing
S.no	S.no	0	0.0
ID	ID	0	0.0
Z_Revenue	Z_Revenue	0	0.0
Z_CostContact	Z_CostContact	0	0.0
Complain	Complain	0	0.0
AcceptedCmp2	AcceptedCmp2	0	0.0
AcceptedCmp1	AcceptedCmp1	0	0.0
AcceptedCmp5	AcceptedCmp5	0	0.0
AcceptedCmp4	AcceptedCmp4	0	0.0
AcceptedCmp3	AcceptedCmp3	0	0.0
NumWebVisitsMonth	NumWebVisitsMonth	0	0.0
NumStorePurchases	NumStorePurchases	0	0.0
NumCatalogPurchases	NumCatalogPurchases	0	0.0
NumWebPurchases	NumWebPurchases	0	0.0
NumDealsPurchases	NumDealsPurchases	0	0.0
MntGoldProds	MntGoldProds	0	0.0
MntSweetProducts	MntSweetProducts	0	0.0
MntFishProducts	MntFishProducts	0	0.0
MntMeatProducts	MntMeatProducts	0	0.0
MntFruits	MntFruits	0	0.0
MntWines	MntWines	0	0.0
Recency	Recency	0	0.0
Dt_Customer	Dt_Customer	0	0.0
Teenhome	Teenhome	0	0.0

Kidhome	Kidhome	0	0.0
Income	Income	0	0.0
Marital_Status	Marital_Status	0	0.0
Education	Education	0	0.0
Year_Birth	Year_Birth	0	0.0
Response	Response	0	0.0

time: 47 ms (started: 2024-04-13 18:11:05 +05:30)

```
[10]: # List of columns to drop
columns_to_drop = ['ID', 'Dt_Customer']

# Drop columns with more than 50% missing values
df_cleaned = df.drop(columns=columns_to_drop)

# Print the cleaned DataFrame
df1 = df_cleaned

# Count the missing values in each column
missing_data = df1.isnull().sum()

# Create a DataFrame to display missing data statistics
missing_data_stats = pd.DataFrame({
    'Column': missing_data.index,
    'Missing Records': missing_data.values,
    'Percentage Missing': (missing_data / len(df)) * 100
})

# Sort the DataFrame by the percentage of missing values in descending order
missing_data_stats = missing_data_stats.sort_values(by='Percentage Missing',
    ↪ascending=False)

# Print the missing data statistics
print(missing_data_stats)
```

S.no	Column	Missing Records	Percentage Missing
S.no	S.no	0	0.0
Year_Birth	Year_Birth	0	0.0
Z_Revenue	Z_Revenue	0	0.0
Z_CostContact	Z_CostContact	0	0.0
Complain	Complain	0	0.0
AcceptedCmp2	AcceptedCmp2	0	0.0
AcceptedCmp1	AcceptedCmp1	0	0.0
AcceptedCmp5	AcceptedCmp5	0	0.0
AcceptedCmp4	AcceptedCmp4	0	0.0
AcceptedCmp3	AcceptedCmp3	0	0.0
NumWebVisitsMonth	NumWebVisitsMonth	0	0.0
NumStorePurchases	NumStorePurchases	0	0.0
NumCatalogPurchases	NumCatalogPurchases	0	0.0
NumWebPurchases	NumWebPurchases	0	0.0

NumDealsPurchases	NumDealsPurchases	0	0.0
MntGoldProds	MntGoldProds	0	0.0
MntSweetProducts	MntSweetProducts	0	0.0
MntFishProducts	MntFishProducts	0	0.0
MntMeatProducts	MntMeatProducts	0	0.0
MntFruits	MntFruits	0	0.0
MntWines	MntWines	0	0.0
Recency	Recency	0	0.0
Teenhome	Teenhome	0	0.0
Kidhome	Kidhome	0	0.0
Income	Income	0	0.0
Marital_Status	Marital_Status	0	0.0
Education	Education	0	0.0
Response	Response	0	0.0

time: 47 ms (started: 2024-04-13 18:11:05 +05:30)

[11]: *### Missing Records (ROWS)*

```
# Count the missing values in each row
missing_rows = df1.isnull().sum(axis=1)

# Count the number of rows with at least one missing value
num_rows_with_missing = len(missing_rows[missing_rows > 0])

# Print the number of rows with missing data
print("Number of rows with missing data:", num_rows_with_missing)

# Calculate the percentage of missing values in each row
missing_percentage_rows = (df1.isnull().sum(axis=1) / len(df1.columns)) * 100

# Count the number of rows with more than 50% missing data
num_rows_more_than_50_percent_missing = \
    len(missing_percentage_rows[missing_percentage_rows > 50])

# Print the number of rows with more than 50% missing data
print("Number of rows with more than 50% missing data:", \
    num_rows_more_than_50_percent_missing)
```

Number of rows with missing data: 0
 Number of rows with more than 50% missing data: 0
 time: 62 ms (started: 2024-04-13 18:11:05 +05:30)

[12]: *# DIVIDING DF1 into Cat and Non Cat*

```
# Assuming df1 is your DataFrame
cat_columns = df1.select_dtypes(include=['object']).columns
```

```
noncat_columns = df1.select_dtypes(exclude=['object']).columns
```

```
# Creating categorical and non-categorical DataFrames
```

```
catdf1 = df1[cat_columns]
```

```
noncatdf1 = df1[noncat_columns]
```

```
#print(list(catdf.columns))
```

```
#print(list(noncatdf.columns))
```

```
print(list(catdf1.columns))
```

```
print(list(noncatdf1.columns))
```

```
#20
```

```
#list(noncatdf.columns)
```

```
['Education', 'Marital_Status']
```

```
['S.no', 'Year_Birth', 'Income', 'Kidhome', 'Teenhome', 'Recency', 'MntWines',  
'MntFruits', 'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts',  
'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases', 'NumCatalogPurchases',  
'NumStorePurchases', 'NumWebVisitsMonth', 'AcceptedCmp3', 'AcceptedCmp4',  
'AcceptedCmp5', 'AcceptedCmp1', 'AcceptedCmp2', 'Complain', 'Z_CostContact',  
'Z_Revenue', 'Response']
```

```
time: 47 ms (started: 2024-04-13 18:11:05 +05:30)
```

```
[13]: # Data Bifurcation
```

```
catdf = df[['S.no', 'Year_Birth', 'Education', 'Marital_Status', 'Teenhome',  
↪ 'Response', 'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1',  
↪ 'AcceptedCmp2', 'Complain']] # Categorical Data [Nominal / Ordinal]
```

```
noncatdf = df[['S.no', 'Income', 'Kidhome', 'Recency', 'MntWines', 'MntFruits',  
↪ 'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts', 'MntGoldProds',  
↪ 'NumDealsPurchases', 'NumWebPurchases', 'NumCatalogPurchases',  
↪ 'NumStorePurchases', 'NumWebVisitsMonth', 'Z_CostContact', 'Z_Revenue']] #  
↪ Non-Categorical Data
```

```
time: 31 ms (started: 2024-04-13 18:11:05 +05:30)
```

```
[14]: ##### STATISTICS OF CAT DATASET
```

```
# Count and frequency statistics for each column in catdf
```

```
catdf_stats = pd.DataFrame()
```

```
for column in catdf.columns:
```

```
    col_count = catdf[column].value_counts().reset_index()
```

```
    col_count.columns = [column, 'Frequency']
```

```
    catdf_stats = pd.concat([catdf_stats, col_count], axis=1)
```

```
# Display the count and frequency statistics
```

```

# print(catdf_stats)

# Summary for each column in catdf
catdf_summary = catdf.describe(include='all').transpose()

# Display the summary
print(catdf_summary)

# Calculate the proportion (relative frequency) for each categorical column
# proportion_stats = catdf.apply(lambda x: x.value_counts(normalize=True).
#                               ↪idxmax() + ': ' + "{:.2%}".format(x.value_counts(normalize=True).max()))

# Display the proportion statistics
# print(proportion_stats)

```

	count	unique	top	freq	mean	std \
S.no	59999.0	NaN	NaN	NaN	30000.0	17320.363738
Year_Birth	59999.0	NaN	NaN	NaN	1968.296488	12.318071
Education	59999	5	Graduation	30137	NaN	NaN
Marital_Status	59999	8	Married	22798	NaN	NaN
Teenhome	59999.0	NaN	NaN	NaN	0.508592	0.545526
Response	59999.0	NaN	NaN	NaN	0.500175	0.500004
AcceptedCmp3	59999.0	NaN	NaN	NaN	0.497058	0.499996
AcceptedCmp4	59999.0	NaN	NaN	NaN	0.499442	0.500004
AcceptedCmp5	59999.0	NaN	NaN	NaN	0.499742	0.500004
AcceptedCmp1	59999.0	NaN	NaN	NaN	0.497642	0.499999
AcceptedCmp2	59999.0	NaN	NaN	NaN	0.495175	0.499981
Complain	59999.0	NaN	NaN	NaN	0.0094	0.096498

	min	25%	50%	75%	max
S.no	1.0	15000.5	30000.0	44999.5	59999.0
Year_Birth	1888.0	1959.0	1969.0	1977.0	2000.0
Education	NaN	NaN	NaN	NaN	NaN
Marital_Status	NaN	NaN	NaN	NaN	NaN
Teenhome	0.0	0.0	0.0	1.0	2.0
Response	0.0	0.0	1.0	1.0	1.0
AcceptedCmp3	0.0	0.0	0.0	1.0	1.0
AcceptedCmp4	0.0	0.0	0.0	1.0	1.0
AcceptedCmp5	0.0	0.0	0.0	1.0	1.0
AcceptedCmp1	0.0	0.0	0.0	1.0	1.0
AcceptedCmp2	0.0	0.0	0.0	1.0	1.0
Complain	0.0	0.0	0.0	0.0	1.0

time: 312 ms (started: 2024-04-13 18:11:05 +05:30)

[15]: ##### STATISTICS OF NONCAT DATASET

```
# Display descriptive statistics for non-categorical variables
noncatdf_descriptive_stats = noncatdf.describe()

# Print the descriptive statistics
print(noncatdf_descriptive_stats)
```

	S.no	Income	Kidhome	Recency	MntWines \
count	59999.000000	59999.000000	59999.000000	59999.000000	59999.000000
mean	30000.000000	81359.282238	0.441457	48.698862	308.099902
std	17320.363738	126077.457422	0.538620	28.975391	333.677239
min	1.000000	1000.000000	0.000000	0.000000	0.000000
25%	15000.500000	3148.000000	0.000000	24.000000	38.000000
50%	30000.000000	43028.000000	0.000000	49.000000	176.000000
75%	44999.500000	90124.000000	1.000000	74.000000	500.000000
max	59999.000000	666480.000000	2.000000	103.000000	1540.000000

	MntFruits	MntMeatProducts	MntFishProducts	MntSweetProducts \
count	59999.000000	59999.000000	59999.000000	59999.000000
mean	26.573010	166.544309	40.871165	28.414324
std	39.336046	225.606608	53.099988	40.453819
min	0.000000	0.000000	0.000000	0.000000
25%	3.000000	16.000000	8.000000	5.000000
50%	8.000000	66.000000	18.000000	10.000000
75%	32.000000	232.000000	50.000000	34.000000
max	203.000000	1731.000000	278.000000	271.000000

	MntGoldProds	NumDealsPurchases	NumWebPurchases	NumCatalogPurchases \
count	59999.000000	59999.000000	59999.000000	59999.000000
mean	43.856964	2.375523	3.831597	2.949816
std	52.005772	2.082430	2.945731	2.731089
min	0.000000	0.000000	0.000000	0.000000
25%	8.000000	1.000000	2.000000	1.000000
50%	24.000000	2.000000	3.000000	2.000000
75%	56.000000	3.000000	6.000000	4.000000
max	368.000000	17.000000	29.000000	30.000000

	NumStorePurchases	NumWebVisitsMonth	Z_CostContact	Z_Revenue
count	59999.000000	59999.000000	59999.0	59999.0
mean	5.341272	4.919599	3.0	11.0
std	3.596753	2.788278	0.0	0.0
min	0.000000	0.000000	3.0	11.0
25%	3.000000	3.000000	3.0	11.0
50%	5.000000	5.000000	3.0	11.0
75%	8.000000	7.000000	3.0	11.0
max	15.000000	22.000000	3.0	11.0

time: 125 ms (started: 2024-04-13 18:11:05 +05:30)

[16]: *# Missing Data Statistics: Non-Categorical Variables or Features*

```
# Calculate missing data statistics for non-categorical columns
missing_data_non_categorical = noncatdf.isnull().sum().reset_index()
missing_data_non_categorical.columns = ['Feature', 'Missing_Records']

# Display the missing data statistics
print(missing_data_non_categorical)
```

	Feature	Missing_Records
0	S.no	0
1	Income	0
2	Kidhome	0
3	Recency	0
4	MntWines	0
5	MntFruits	0
6	MntMeatProducts	0
7	MntFishProducts	0
8	MntSweetProducts	0
9	MntGoldProds	0
10	NumDealsPurchases	0
11	NumWebPurchases	0
12	NumCatalogPurchases	0
13	NumStorePurchases	0
14	NumWebVisitsMonth	0
15	Z_CostContact	0
16	Z_Revenue	0

time: 31 ms (started: 2024-04-13 18:11:05 +05:30)

[17]: *# Missing Data Treatment: Non-Categorical Variables or Features*

```
# Dataset Used : df_noncat

si_noncat = SimpleImputer(missing_values=np.nan, strategy='mean') # Other_
    ↪ Strategy : mean / median / most_frequent / constant
si_noncat_fit = si_noncat.fit_transform(noncatdf)
imputed_data_non_categorical = pd.DataFrame(si_noncat_fit, columns=noncatdf.
    ↪ columns); # Missing Non-Categorical Data Imputed Subset using Simple Imputer
imputed_data_non_categorical.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 59999 entries, 0 to 59998
Data columns (total 17 columns):
#   Column                Non-Null Count  Dtype
---  -
0   S.no                   59999 non-null  float64
1   Income                 59999 non-null  float64
```

```

2 Kidhome          59999 non-null float64
3 Recency          59999 non-null float64
4 MntWines         59999 non-null float64
5 MntFruits        59999 non-null float64
6 MntMeatProducts  59999 non-null float64
7 MntFishProducts  59999 non-null float64
8 MntSweetProducts 59999 non-null float64
9 MntGoldProds     59999 non-null float64
10 NumDealsPurchases 59999 non-null float64
11 NumWebPurchases  59999 non-null float64
12 NumCatalogPurchases 59999 non-null float64
13 NumStorePurchases 59999 non-null float64
14 NumWebVisitsMonth 59999 non-null float64
15 Z_CostContact     59999 non-null float64
16 Z_Revenue        59999 non-null float64

```

dtypes: float64(17)

memory usage: 7.8 MB

time: 78 ms (started: 2024-04-13 18:11:06 +05:30)

```

[18]: # Calculate standard deviation for non-categorical columns
std_deviation_non_categorical = imputed_data_non_categorical.std()

# Creating a DataFrame to display the results
dispersion_non_categorical_df = pd.DataFrame({
    'Variable': imputed_data_non_categorical.columns,
    'Standard Deviation': std_deviation_non_categorical.values
})

print(dispersion_non_categorical_df)

```

	Variable	Standard Deviation
0	S.no	17320.363738
1	Income	126077.457422
2	Kidhome	0.538620
3	Recency	28.975391
4	MntWines	333.677239
5	MntFruits	39.336046
6	MntMeatProducts	225.606608
7	MntFishProducts	53.099988
8	MntSweetProducts	40.453819
9	MntGoldProds	52.005772
10	NumDealsPurchases	2.082430
11	NumWebPurchases	2.945731
12	NumCatalogPurchases	2.731089
13	NumStorePurchases	3.596753
14	NumWebVisitsMonth	2.788278
15	Z_CostContact	0.000000
16	Z_Revenue	0.000000

time: 0 ns (started: 2024-04-13 18:11:06 +05:30)

[19]: # Dataset Used : df_cat

```
si_cat = SimpleImputer(missing_values=np.nan, strategy='most_frequent') #  
    ↳Strategy = median [When Odd Number of Categories Exists]  
si_cat_fit = si_cat.fit_transform(catdf)  
imputed_data_categorical = pd.DataFrame(si_cat_fit, columns=catdf.columns); #  
    ↳Missing Categorical Data Imputed Subset  
imputed_data_categorical.info()  
imputed_data_categorical.head()
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 59999 entries, 0 to 59998

Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	S.no	59999 non-null	object
1	Year_Birth	59999 non-null	object
2	Education	59999 non-null	object
3	Marital_Status	59999 non-null	object
4	Teenhome	59999 non-null	object
5	Response	59999 non-null	object
6	AcceptedCmp3	59999 non-null	object
7	AcceptedCmp4	59999 non-null	object
8	AcceptedCmp5	59999 non-null	object
9	AcceptedCmp1	59999 non-null	object
10	AcceptedCmp2	59999 non-null	object
11	Complain	59999 non-null	object

dtypes: object(12)

memory usage: 5.5+ MB

```
[19]: S.no Year_Birth Education Marital_Status Teenhome Response AcceptedCmp3 \  
0 1 1956 2n Cycle Together 1 1 1  
1 2 1950 Graduation Together 1 1 0  
2 3 1960 PhD Together 0 1 0  
3 4 1980 PhD Divorced 0 1 1  
4 5 1977 Graduation Married 1 0 1  
  
AcceptedCmp4 AcceptedCmp5 AcceptedCmp1 AcceptedCmp2 Complain  
0 1 1 1 1 0  
1 1 1 0 0 0  
2 1 0 1 1 0  
3 1 0 0 0 0  
4 1 1 1 1 0
```

time: 328 ms (started: 2024-04-13 18:11:06 +05:30)


```

[20]: from sklearn.preprocessing import LabelEncoder

# Calculate the number of unique values in each column
unique_values_categorical = imputed_data_categorical.nunique().reset_index()
unique_values_categorical.columns = ['Feature', 'Number_of_Unique_Values']

# Display the number of unique values
print(unique_values_categorical)

# Initialize LabelEncoder
label_encoder = LabelEncoder()

# Create a copy of the imputed_data_categorical dataframe to avoid modifying
↳ the original
encoded_data_categorical = imputed_data_categorical.copy()

# Columns to exclude from encoding
exclude_columns = ['S.no', 'Response', 'AcceptedCmp3', 'AcceptedCmp4',
↳ 'AcceptedCmp5', 'AcceptedCmp1', 'AcceptedCmp2', 'Complain'] # Add column
↳ names you want to exclude here

# Check for missing values
if encoded_data_categorical.isnull().values.any():
    # Handle missing values before encoding
    encoded_data_categorical = encoded_data_categorical.dropna() # Example:
↳ Remove rows with missing values
    # Alternatively, you can impute missing values

# Check data types
for column in encoded_data_categorical.columns:
    if column not in exclude_columns:
        if not encoded_data_categorical[column].apply(lambda x: isinstance(x,
↳ (int, float))).all():
            # If the column contains non-numeric values, convert them to strings
            encoded_data_categorical[column] = encoded_data_categorical[column].
↳ astype(str)

# Initialize LabelEncoder
label_encoder = LabelEncoder()

# Iterate through each column in the dataframe
mapping = {} # To store the mapping of variable names to numeric representation
for column in encoded_data_categorical.columns:
    if column not in exclude_columns:
        # Perform numerical encoding
        encoded_data_categorical[column] = label_encoder.
↳ fit_transform(encoded_data_categorical[column])

```

```

    # Store the mapping information
    mapping[column] = dict(zip(label_encoder.classes_, label_encoder.
↪transform(label_encoder.classes_)))

# Display the mapping
for variable, variable_mapping in mapping.items():
    print(f"\nMapping for {variable}:")
    print(variable_mapping)

# Display the encoded data
print(encoded_data_categorical)

```

	Feature	Number_of_Unique_Values
0	S.no	59999
1	Year_Birth	83
2	Education	5
3	Marital_Status	8
4	Teenhome	3
5	Response	2
6	AcceptedCmp3	2
7	AcceptedCmp4	2
8	AcceptedCmp5	2
9	AcceptedCmp1	2
10	AcceptedCmp2	2
11	Complain	2

Mapping for Year_Birth:

```
{1888: 0, 1889: 1, 1890: 2, 1891: 3, 1892: 4, 1893: 5, 1894: 6, 1895: 7, 1896:
8, 1897: 9, 1898: 10, 1899: 11, 1900: 12, 1901: 13, 1902: 14, 1903: 15, 1904:
16, 1935: 17, 1936: 18, 1937: 19, 1938: 20, 1939: 21, 1940: 22, 1941: 23, 1942:
24, 1943: 25, 1944: 26, 1945: 27, 1946: 28, 1947: 29, 1948: 30, 1949: 31, 1950:
32, 1951: 33, 1952: 34, 1953: 35, 1954: 36, 1955: 37, 1956: 38, 1957: 39, 1958:
40, 1959: 41, 1960: 42, 1961: 43, 1962: 44, 1963: 45, 1964: 46, 1965: 47, 1966:
48, 1967: 49, 1968: 50, 1969: 51, 1970: 52, 1971: 53, 1972: 54, 1973: 55, 1974:
56, 1975: 57, 1976: 58, 1977: 59, 1978: 60, 1979: 61, 1980: 62, 1981: 63, 1982:
64, 1983: 65, 1984: 66, 1985: 67, 1986: 68, 1987: 69, 1988: 70, 1989: 71, 1990:
72, 1991: 73, 1992: 74, 1993: 75, 1994: 76, 1995: 77, 1996: 78, 1997: 79, 1998:
80, 1999: 81, 2000: 82}
```

Mapping for Education:

```
{'2n Cycle': 0, 'Basic': 1, 'Graduation': 2, 'Master': 3, 'PhD': 4}
```

Mapping for Marital_Status:

```
{'Absurd': 0, 'Alone': 1, 'Divorced': 2, 'Married': 3, 'Single': 4, 'Together':
5, 'Widow': 6, 'YOLO': 7}
```

Mapping for Teenhome:

{0: 0, 1: 1, 2: 2}

	S.no	Year_Birth	Education	Marital_Status	Teenhome	Response	\
0	1	38	0	5	1	1	
1	2	32	2	5	1	1	
2	3	42	4	5	0	1	
3	4	62	4	2	0	1	
4	5	59	2	3	1	0	
...	
59994	59995	49	4	3	1	0	
59995	59996	56	2	5	0	0	
59996	59997	45	3	4	1	0	
59997	59998	56	3	4	0	1	
59998	59999	33	2	3	0	0	

	AcceptedCmp3	AcceptedCmp4	AcceptedCmp5	AcceptedCmp1	AcceptedCmp2	\
0	1	1	1	1	1	
1	0	1	1	0	0	
2	0	1	0	1	1	
3	1	1	0	0	0	
4	1	1	1	1	1	
...	
59994	1	1	1	1	0	
59995	0	1	1	1	0	
59996	0	0	1	1	0	
59997	0	1	0	1	1	
59998	1	1	0	1	0	

	Complain
0	0
1	0
2	0
3	0
4	0
...	...
59994	0
59995	0
59996	0
59997	0
59998	0

[59999 rows x 12 columns]

time: 313 ms (started: 2024-04-13 18:11:06 +05:30)

```
[21]: print(imputed_data_non_categorical.columns)
```

```
Index(['S.no', 'Income', 'Kidhome', 'Recency', 'MntWines', 'MntFruits',  
      'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts',
```

```

'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',
'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth',
'Z_CostContact', 'Z_Revenue'],
dtype='object')
time: 0 ns (started: 2024-04-13 18:11:06 +05:30)

```

```

[22]: def identify_outliers(column):
    Q1 = np.percentile(column, 25)
    Q3 = np.percentile(column, 75)
    IQR = Q3 - Q1
    lower_bound = Q1 - 1.5 * IQR
    upper_bound = Q3 + 1.5 * IQR
    outliers = (column < lower_bound) | (column > upper_bound)
    return outliers

# Apply the function to each column to get a DataFrame of True/False values
outliers = imputed_data_non_categorical.apply(identify_outliers)

# Display the number of outliers for each column
outlier_counts = outliers.sum()
print(outlier_counts)

```

```

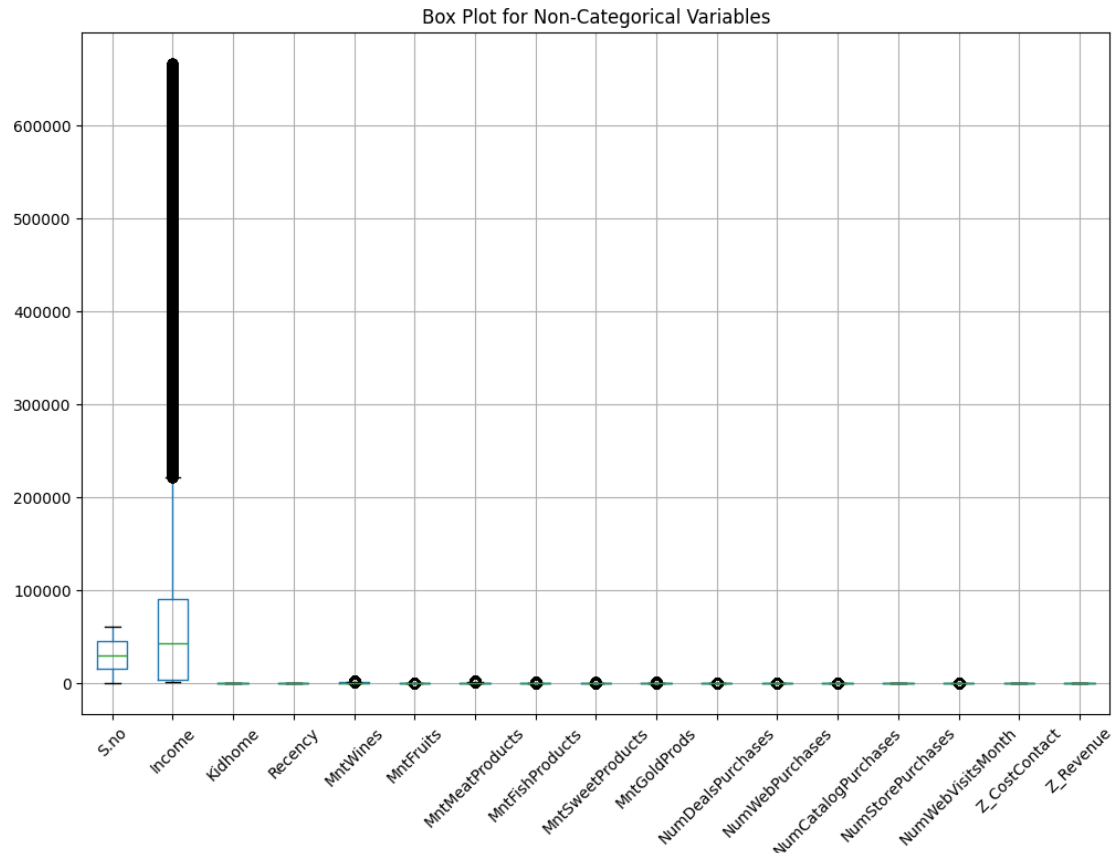
S.no          0
Income        5245
Kidhome       0
Recency       0
MntWines     1219
MntFruits    6872
MntMeatProducts 4700
MntFishProducts 6589
MntSweetProducts 7070
MntGoldProds 5391
NumDealsPurchases 2560
NumWebPurchases 304
NumCatalogPurchases 2813
NumStorePurchases 0
NumWebVisitsMonth 189
Z_CostContact 0
Z_Revenue     0
dtype: int64
time: 94 ms (started: 2024-04-13 18:11:06 +05:30)

```

```

[23]: # Create a box plot for non-categorical variables
imputed_data_non_categorical.boxplot(rot=45, figsize=(12, 8))
plt.title('Box Plot for Non-Categorical Variables')
plt.show()

```



time: 2.14 s (started: 2024-04-13 18:11:06 +05:30)

```
[24]: # Iterate through each column and print count of unique values
for column in imputed_data_non_categorical.columns:
    unique_count = imputed_data_non_categorical[column].nunique()
    print(f"Count of unique values in {column} column: {unique_count}")
```

```
Count of unique values in S.no column: 59999
Count of unique values in Income column: 40412
Count of unique values in Kidhome column: 3
Count of unique values in Recency column: 104
Count of unique values in MntWines column: 1508
Count of unique values in MntFruits column: 204
Count of unique values in MntMeatProducts column: 1033
Count of unique values in MntFishProducts column: 278
Count of unique values in MntSweetProducts column: 226
Count of unique values in MntGoldProds column: 303
Count of unique values in NumDealsPurchases column: 18
Count of unique values in NumWebPurchases column: 24
Count of unique values in NumCatalogPurchases column: 26
```

Count of unique values in NumStorePurchases column: 16
 Count of unique values in NumWebVisitsMonth column: 23
 Count of unique values in Z_CostContact column: 1
 Count of unique values in Z_Revenue column: 1
 time: 62 ms (started: 2024-04-13 18:11:09 +05:30)

```
[25]: # Initialize the StandardScaler
scaler = StandardScaler()

# Apply Standard Scaling to your dataset
scaled_data = scaler.fit_transform(imputed_data_non_categorical)

def identify_outliers(column):
    Q1 = np.percentile(column, 25)
    Q3 = np.percentile(column, 75)
    IQR = Q3 - Q1
    lower_bound = Q1 - 1.5 * IQR
    upper_bound = Q3 + 1.5 * IQR
    outliers = (column < lower_bound) | (column > upper_bound)
    return outliers

# Apply the function to each column in the scaled dataset
outliers_scaled = pd.DataFrame(scaled_data,
    columns=imputed_data_non_categorical.columns).apply(identify_outliers)

# Display the number of outliers for each column in the scaled dataset
outlier_counts_scaled = outliers_scaled.sum()
print(outlier_counts_scaled)
```

S.no	0
Income	5245
Kidhome	0
Recency	0
MntWines	1219
MntFruits	6872
MntMeatProducts	4700
MntFishProducts	6589
MntSweetProducts	7070
MntGoldProds	5391
NumDealsPurchases	2560
NumWebPurchases	304
NumCatalogPurchases	2813
NumStorePurchases	0
NumWebVisitsMonth	189
Z_CostContact	0
Z_Revenue	0

dtype: int64
 time: 125 ms (started: 2024-04-13 18:11:09 +05:30)

```
[26]: # Initialize the RobustScaler
scaler = RobustScaler()

# Apply Robust Scaling to your dataset
scaled_data_robust = scaler.fit_transform(imputed_data_non_categorical)

# Check for outliers in the scaled dataset
outliers_robust = pd.DataFrame(scaled_data_robust,
    ↪ columns=imputed_data_non_categorical.columns).apply(identify_outliers)

# Display the number of outliers for each column in the scaled dataset
outlier_counts_robust = outliers_robust.sum()
print(outlier_counts_robust)
```

```
S.no          0
Income        5245
Kidhome       0
Recency       0
MntWines     1219
MntFruits    6872
MntMeatProducts 4700
MntFishProducts 6589
MntSweetProducts 7070
MntGoldProds 5391
NumDealsPurchases 2560
NumWebPurchases 304
NumCatalogPurchases 2813
NumStorePurchases 0
NumWebVisitsMonth 189
Z_CostContact 0
Z_Revenue     0
dtype: int64
time: 156 ms (started: 2024-04-13 18:11:09 +05:30)
```

```
[27]: # Define columns to exclude from normalization
columns_to_exclude = ['S.no', 'Kidhome', 'Recency', 'MntWines', 'MntFruits',
    ↪ 'MntMeatProducts', 'MntFishProducts',
    ↪ 'MntSweetProducts', 'MntGoldProds', 'NumDealsPurchases',
    ↪ 'NumWebPurchases', 'NumCatalogPurchases', 'NumStorePurchases',
    ↪ 'NumWebVisitsMonth', 'Z_CostContact', 'Z_Revenue']

# Create a copy of the DataFrame with excluded columns
data_to_scale = imputed_data_non_categorical.drop(columns=columns_to_exclude)

# Initialize the MinMaxScaler
scaler = MinMaxScaler()
```

```

# Apply Min-Max Scaling to the selected columns
scaled_data = scaler.fit_transform(data_to_scale)

# Create a DataFrame with scaled data and original column names
scaled_df = pd.DataFrame(scaled_data, columns=data_to_scale.columns)

# Add back the excluded columns to the scaled DataFrame
scaled_df[columns_to_exclude] = imputed_data_non_categorical[columns_to_exclude]

# Display the scaled DataFrame
print(scaled_df)

```

	Income	S.no	Kidhome	Recency	MntWines	MntFruits	\
0	0.059853	1.0	0.0	62.0	636.0	84.0	
1	0.032363	2.0	1.0	42.0	6.0	2.0	
2	0.292653	3.0	1.0	28.0	383.0	51.0	
3	0.213595	4.0	0.0	24.0	29.0	0.0	
4	0.125696	5.0	0.0	92.0	133.0	41.0	
...	
59994	0.002537	59995.0	0.0	30.0	13.0	3.0	
59995	0.020402	59996.0	1.0	43.0	10.0	11.0	
59996	0.003429	59997.0	1.0	80.0	33.0	6.0	
59997	0.102101	59998.0	1.0	11.0	37.0	4.0	
59998	0.196344	59999.0	0.0	50.0	19.0	4.0	

	MntMeatProducts	MntFishProducts	MntSweetProducts	MntGoldProds	\
0	549.0	176.0	86.0	93.0	
1	8.0	18.0	4.0	11.0	
2	130.0	124.0	18.0	37.0	
3	20.0	1.0	6.0	1.0	
4	124.0	28.0	21.0	9.0	
...	
59994	14.0	5.0	1.0	8.0	
59995	4.0	20.0	8.0	12.0	
59996	17.0	7.0	11.0	29.0	
59997	57.0	13.0	3.0	18.0	
59998	9.0	14.0	5.0	6.0	

	NumDealsPurchases	NumWebPurchases	NumCatalogPurchases	\
0	2.0	6.0	11.0	
1	3.0	2.0	2.0	
2	0.0	9.0	2.0	
3	1.0	2.0	2.0	
4	3.0	3.0	1.0	
...	
59994	0.0	0.0	1.0	
59995	2.0	2.0	2.0	

59996	2.0	3.0	1.0
59997	1.0	6.0	0.0
59998	2.0	2.0	1.0

	NumStorePurchases	NumWebVisitsMonth	Z_CostContact	Z_Revenue
0	1.0	5.0	3.0	11.0
1	3.0	4.0	3.0	11.0
2	9.0	4.0	3.0	11.0
3	5.0	5.0	3.0	11.0
4	5.0	7.0	3.0	11.0
...
59994	2.0	7.0	3.0	11.0
59995	1.0	4.0	3.0	11.0
59996	2.0	7.0	3.0	11.0
59997	1.0	6.0	3.0	11.0
59998	6.0	1.0	3.0	11.0

[59999 rows x 17 columns]

time: 63 ms (started: 2024-04-13 18:11:09 +05:30)

```
[28]: def identify_outliers(column):
    Q1 = np.percentile(column, 25)
    Q3 = np.percentile(column, 75)
    IQR = Q3 - Q1
    lower_bound = Q1 - 1.5 * IQR
    upper_bound = Q3 + 1.5 * IQR
    outliers = (column < lower_bound) | (column > upper_bound)
    return outliers

# Apply the function to each column in the scaled dataset
scaled_outliers = scaled_df.apply(identify_outliers)

# Display the number of outliers for each column in the scaled dataset
scaled_outlier_counts = scaled_outliers.sum()
print(scaled_outlier_counts)
```

Income	5245
S.no	0
Kidhome	0
Recency	0
MntWines	1219
MntFruits	6872
MntMeatProducts	4700
MntFishProducts	6589
MntSweetProducts	7070
MntGoldProds	5391
NumDealsPurchases	2560
NumWebPurchases	304

```

NumCatalogPurchases    2813
NumStorePurchases       0
NumWebVisitsMonth       189
Z_CostContact           0
Z_Revenue               0
dtype: int64
time: 78 ms (started: 2024-04-13 18:11:09 +05:30)

```

```

[29]: def identify_outliers(column):
        Q1 = np.percentile(column, 25)
        Q3 = np.percentile(column, 75)
        IQR = Q3 - Q1
        lower_bound = Q1 - 1.5 * IQR
        upper_bound = Q3 + 1.5 * IQR
        outliers1 = (column < lower_bound) | (column > upper_bound)
        return outliers1

# Apply the function to each column to get a DataFrame of True/False values
outliers1 = scaled_df.apply(identify_outliers)

# Display the number of outliers for each column
outlier_counts1 = outliers1.sum()
print(outlier_counts1)

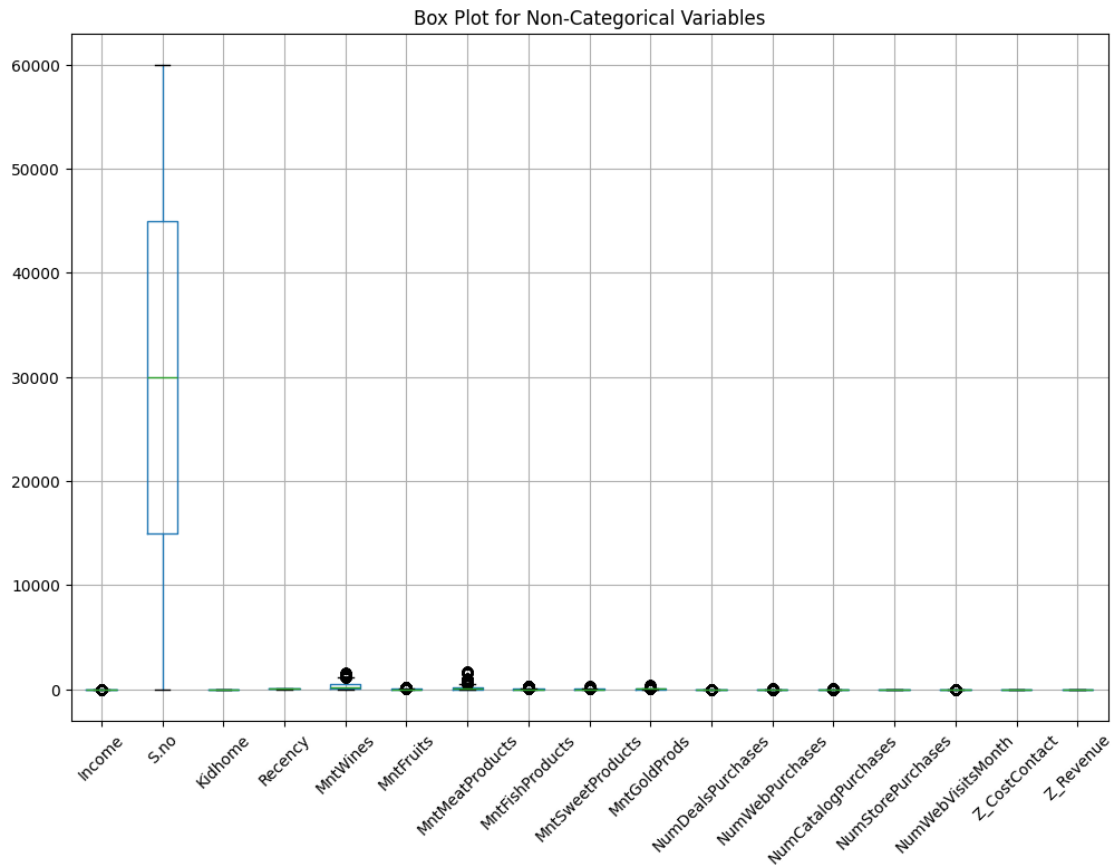
# Create a box plot for non-categorical variables
scaled_df.boxplot(rot=45, figsize=(12, 8))
plt.title('Box Plot for Non-Categorical Variables')
plt.show()

```

```

Income                5245
S.no                  0
Kidhome               0
Recency               0
MntWines              1219
MntFruits              6872
MntMeatProducts        4700
MntFishProducts        6589
MntSweetProducts       7070
MntGoldProds           5391
NumDealsPurchases      2560
NumWebPurchases         304
NumCatalogPurchases    2813
NumStorePurchases       0
NumWebVisitsMonth       189
Z_CostContact           0
Z_Revenue               0
dtype: int64

```



time: 2.91 s (started: 2024-04-13 18:11:09 +05:30)

```
[30]: # Calculate standard deviation for non-categorical columns
std_deviation_non_categorical1 = scaled_df.std()

# Creating a DataFrame to display the results
dispersion_non_categorical_df1 = pd.DataFrame({
    'Variable': scaled_df.columns,
    'Standard Deviation': std_deviation_non_categorical1.values
})

print(dispersion_non_categorical_df1)
```

	Variable	Standard Deviation
0	Income	0.189453
1	S.no	17320.363738
2	Kidhome	0.538620
3	Recency	28.975391
4	MntWines	333.677239
5	MntFruits	39.336046

```

6      MntMeatProducts      225.606608
7      MntFishProducts      53.099988
8      MntSweetProducts     40.453819
9      MntGoldProds         52.005772
10     NumDealsPurchases     2.082430
11     NumWebPurchases       2.945731
12     NumCatalogPurchases   2.731089
13     NumStorePurchases     3.596753
14     NumWebVisitsMonth     2.788278
15     Z_CostContact         0.000000
16     Z_Revenue             0.000000
time: 16 ms (started: 2024-04-13 18:11:12 +05:30)

```

```
[31]: scaled_df
```

```

[31]:      Income      S.no  Kidhome  Recency  MntWines  MntFruits  \
0      0.059853      1.0      0.0      62.0      636.0      84.0
1      0.032363      2.0      1.0      42.0        6.0       2.0
2      0.292653      3.0      1.0      28.0     383.0     51.0
3      0.213595      4.0      0.0      24.0      29.0      0.0
4      0.125696      5.0      0.0      92.0     133.0     41.0
...      ...      ...      ...      ...      ...      ...
59994  0.002537  59995.0      0.0      30.0      13.0       3.0
59995  0.020402  59996.0      1.0      43.0      10.0     11.0
59996  0.003429  59997.0      1.0      80.0      33.0       6.0
59997  0.102101  59998.0      1.0      11.0      37.0       4.0
59998  0.196344  59999.0      0.0      50.0      19.0       4.0

      MntMeatProducts  MntFishProducts  MntSweetProducts  MntGoldProds  \
0              549.0             176.0             86.0             93.0
1               8.0              18.0              4.0             11.0
2             130.0             124.0             18.0             37.0
3              20.0               1.0              6.0              1.0
4             124.0             28.0             21.0              9.0
...      ...      ...      ...      ...
59994             14.0              5.0              1.0             8.0
59995              4.0             20.0              8.0            12.0
59996             17.0              7.0             11.0            29.0
59997             57.0             13.0              3.0            18.0
59998              9.0             14.0              5.0             6.0

      NumDealsPurchases  NumWebPurchases  NumCatalogPurchases  \
0                   2.0                6.0                 11.0
1                   3.0                2.0                 2.0
2                   0.0                9.0                 2.0
3                   1.0                2.0                 2.0
4                   3.0                3.0                 1.0

```

...
59994	0.0	0.0	1.0
59995	2.0	2.0	2.0
59996	2.0	3.0	1.0
59997	1.0	6.0	0.0
59998	2.0	2.0	1.0

	NumStorePurchases	NumWebVisitsMonth	Z_CostContact	Z_Revenue
0	1.0	5.0	3.0	11.0
1	3.0	4.0	3.0	11.0
2	9.0	4.0	3.0	11.0
3	5.0	5.0	3.0	11.0
4	5.0	7.0	3.0	11.0

...	
59994	2.0	7.0	3.0	11.0
59995	1.0	4.0	3.0	11.0
59996	2.0	7.0	3.0	11.0
59997	1.0	6.0	3.0	11.0
59998	6.0	1.0	3.0	11.0

[59999 rows x 17 columns]

time: 94 ms (started: 2024-04-13 18:11:12 +05:30)

```
[32]: # Pre-Processed Dataset
combined_data = pd.merge(encoded_data_categorical, scaled_df, on='S.no')

# Display the Pre-Processed Dataset
%memit
print(combined_data)

list(combined_data.columns)
```

peak memory: 331.95 MiB, increment: 0.09 MiB

	S.no	Year_Birth	Education	Marital_Status	Teenhome	Response	\
0	1	38	0	5	1	1	
1	2	32	2	5	1	1	
2	3	42	4	5	0	1	
3	4	62	4	2	0	1	
4	5	59	2	3	1	0	
...	
59994	59995	49	4	3	1	0	
59995	59996	56	2	5	0	0	
59996	59997	45	3	4	1	0	
59997	59998	56	3	4	0	1	
59998	59999	33	2	3	0	0	

AcceptedCmp3 AcceptedCmp4 AcceptedCmp5 AcceptedCmp1 ... \

0	1	1	1	1	...
1	0	1	1	0	...
2	0	1	0	1	...
3	1	1	0	0	...
4	1	1	1	1	...
...
59994	1	1	1	1	...
59995	0	1	1	1	...
59996	0	0	1	1	...
59997	0	1	0	1	...
59998	1	1	0	1	...

	MntFishProducts	MntSweetProducts	MntGoldProds	NumDealsPurchases	\
0	176.0	86.0	93.0	2.0	
1	18.0	4.0	11.0	3.0	
2	124.0	18.0	37.0	0.0	
3	1.0	6.0	1.0	1.0	
4	28.0	21.0	9.0	3.0	
...	
59994	5.0	1.0	8.0	0.0	
59995	20.0	8.0	12.0	2.0	
59996	7.0	11.0	29.0	2.0	
59997	13.0	3.0	18.0	1.0	
59998	14.0	5.0	6.0	2.0	

	NumWebPurchases	NumCatalogPurchases	NumStorePurchases	\
0	6.0	11.0	1.0	
1	2.0	2.0	3.0	
2	9.0	2.0	9.0	
3	2.0	2.0	5.0	
4	3.0	1.0	5.0	
...	
59994	0.0	1.0	2.0	
59995	2.0	2.0	1.0	
59996	3.0	1.0	2.0	
59997	6.0	0.0	1.0	
59998	2.0	1.0	6.0	

	NumWebVisitsMonth	Z_CostContact	Z_Revenue
0	5.0	3.0	11.0
1	4.0	3.0	11.0
2	4.0	3.0	11.0
3	5.0	3.0	11.0
4	7.0	3.0	11.0
...
59994	7.0	3.0	11.0
59995	4.0	3.0	11.0
59996	7.0	3.0	11.0

59997	6.0	3.0	11.0
59998	1.0	3.0	11.0

[59999 rows x 28 columns]

```
[32]: ['S.no',
       'Year_Birth',
       'Education',
       'Marital_Status',
       'Teenhome',
       'Response',
       'AcceptedCmp3',
       'AcceptedCmp4',
       'AcceptedCmp5',
       'AcceptedCmp1',
       'AcceptedCmp2',
       'Complain',
       'Income',
       'Kidhome',
       'Recency',
       'MntWines',
       'MntFruits',
       'MntMeatProducts',
       'MntFishProducts',
       'MntSweetProducts',
       'MntGoldProds',
       'NumDealsPurchases',
       'NumWebPurchases',
       'NumCatalogPurchases',
       'NumStorePurchases',
       'NumWebVisitsMonth',
       'Z_CostContact',
       'Z_Revenue']
```

time: 2.14 s (started: 2024-04-13 18:11:12 +05:30)

```
[33]: df_ppd_subset = combined_data.copy()
```

time: 16 ms (started: 2024-04-13 18:11:14 +05:30)

```
[37]: list(scaled_df.columns)
```

```
[37]: ['Income',
       'S.no',
       'Kidhome',
       'Recency',
       'MntWines',
       'MntFruits',
```

```
'MntMeatProducts',  
'MntFishProducts',  
'MntSweetProducts',  
'MntGoldProds',  
'NumDealsPurchases',  
'NumWebPurchases',  
'NumCatalogPurchases',  
'NumStorePurchases',  
'NumWebVisitsMonth',  
'Z_CostContact',  
'Z_Revenue']
```

time: 0 ns (started: 2024-04-13 18:23:05 +05:30)

```
[38]: list(encoded_data_categorical.columns)
```

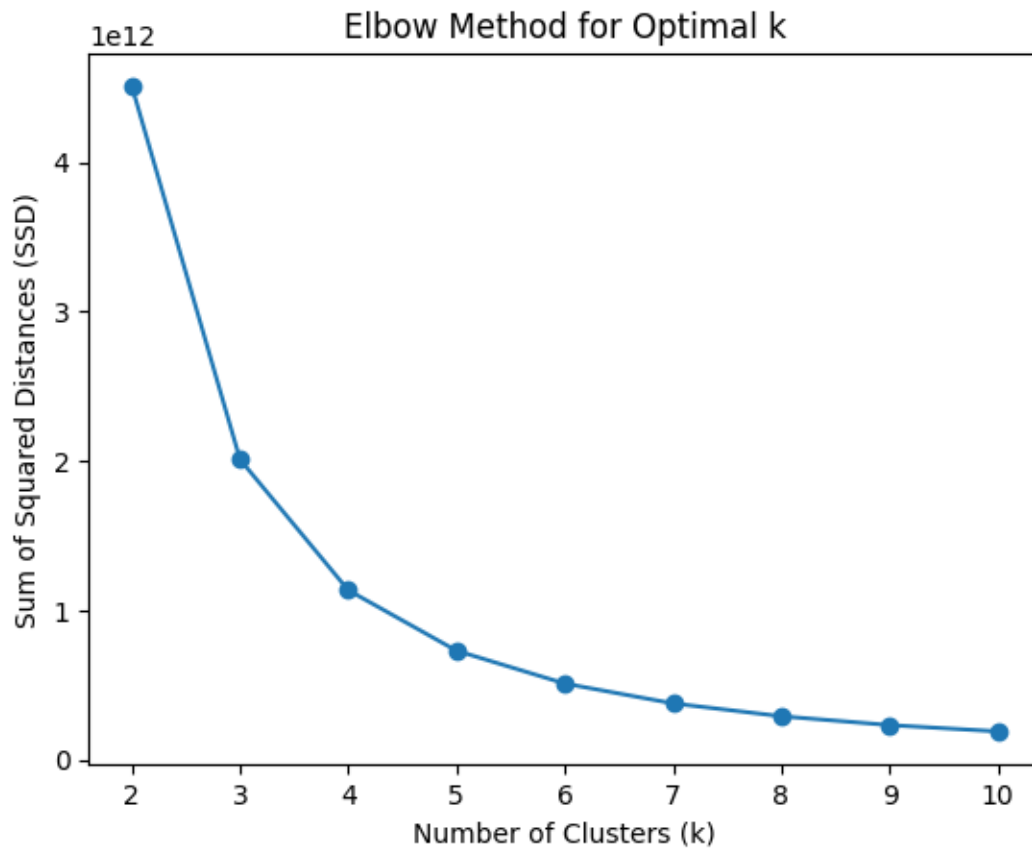
```
[38]: ['S.no',  
      'Year_Birth',  
      'Education',  
      'Marital_Status',  
      'Teenhome',  
      'Response',  
      'AcceptedCmp3',  
      'AcceptedCmp4',  
      'AcceptedCmp5',  
      'AcceptedCmp1',  
      'AcceptedCmp2',  
      'Complain']
```

time: 16 ms (started: 2024-04-13 18:23:17 +05:30)

```
[39]: # Example code for Elbow Method  
import matplotlib.pyplot as plt  
from sklearn.cluster import KMeans  
  
# Assuming 'data' is your feature matrix  
ssd = []  
k_values = range(2, 11)  
  
for k in k_values:  
    kmeans = KMeans(n_clusters=k, random_state=42)  
    kmeans.fit(df_ppd_subset)  
    ssd.append(kmeans.inertia_)  
  
# Plotting the Elbow curve  
plt.plot(k_values, ssd, marker='o')  
plt.title('Elbow Method for Optimal k')
```



```
plt.xlabel('Number of Clusters (k)')
plt.ylabel('Sum of Squared Distances (SSD)')
plt.show()
```



time: 32 s (started: 2024-04-13 18:23:29 +05:30)

```
[40]: # Create K-Means Clusters [K=2]
km_2cluster = kmclus(n_clusters=2, init='random', random_state=333)
df_ppd_subset['Cluster_Label'] = km_2cluster.fit_predict(df_ppd_subset)
km_2cluster_model = df_ppd_subset['Cluster_Label'].values
km_2cluster_model

# K-Means Clustering Model Evaluation [K=2]
# -----

sscore_km_2cluster = sscore(df_ppd_subset, km_2cluster_model)
dbscore_km_2cluster = dbscore(df_ppd_subset, km_2cluster_model);
%memit
print(f"Davies-Bouldin Index for 2 clusters: {dbscore_km_2cluster}")
print(f"Silhouette Score for 2 clusters: {sscore_km_2cluster}")
```

peak memory: 420.21 MiB, increment: 0.00 MiB
Davies-Bouldin Index for 2 clusters: 0.5017148169414763
Silhouette Score for 2 clusters: 0.6253177423151041
time: 1min 34s (started: 2024-04-13 18:24:01 +05:30)

```
[41]: # Create K-Means Clusters [K=3]
km_3cluster = kmclus(n_clusters=3, init='random', random_state=333)
df_ppd_subset['Cluster_Label'] = km_3cluster.fit_predict(df_ppd_subset)
km_3cluster_model = df_ppd_subset['Cluster_Label'].values
km_3cluster_model
```

```
[41]: array([0, 0, 0, ..., 1, 1, 1])
```

time: 1.64 s (started: 2024-04-13 18:25:36 +05:30)

```
[ ]:
```

```
[42]: dff = df_ppd_subset[['S.no', 'Cluster_Label']]
dff
```

```
[42]:
```

	S.no	Cluster_Label
0	1	0
1	2	0
2	3	0
3	4	0
4	5	0
...
59994	59995	1
59995	59996	1
59996	59997	1
59997	59998	1
59998	59999	1

[59999 rows x 2 columns]

time: 16 ms (started: 2024-04-13 18:25:37 +05:30)

```
[43]: df1 = pd.merge(df_ppd_subset, dff, on='S.no', how='left')

# Display the merged DataFrame
df1
```

```
[43]:
```

	S.no	Year_Birth	Education	Marital_Status	Teenhome	Response	\
0	1	38	0	5	1	1	
1	2	32	2	5	1	1	
2	3	42	4	5	0	1	
3	4	62	4	2	0	1	
4	5	59	2	3	1	0	

...
59994	59995	49	4	3	1	0	
59995	59996	56	2	5	0	0	
59996	59997	45	3	4	1	0	
59997	59998	56	3	4	0	1	
59998	59999	33	2	3	0	0	

	AcceptedCmp3	AcceptedCmp4	AcceptedCmp5	AcceptedCmp1	...	MntGoldProds	\
0	1	1	1	1	...	93.0	
1	0	1	1	0	...	11.0	
2	0	1	0	1	...	37.0	
3	1	1	0	0	...	1.0	
4	1	1	1	1	...	9.0	
...		
59994	1	1	1	1	...	8.0	
59995	0	1	1	1	...	12.0	
59996	0	0	1	1	...	29.0	
59997	0	1	0	1	...	18.0	
59998	1	1	0	1	...	6.0	

	NumDealsPurchases	NumWebPurchases	NumCatalogPurchases	\
0	2.0	6.0	11.0	
1	3.0	2.0	2.0	
2	0.0	9.0	2.0	
3	1.0	2.0	2.0	
4	3.0	3.0	1.0	
...	
59994	0.0	0.0	1.0	
59995	2.0	2.0	2.0	
59996	2.0	3.0	1.0	
59997	1.0	6.0	0.0	
59998	2.0	2.0	1.0	

	NumStorePurchases	NumWebVisitsMonth	Z_CostContact	Z_Revenue	\
0	1.0	5.0	3.0	11.0	
1	3.0	4.0	3.0	11.0	
2	9.0	4.0	3.0	11.0	
3	5.0	5.0	3.0	11.0	
4	5.0	7.0	3.0	11.0	
...	
59994	2.0	7.0	3.0	11.0	
59995	1.0	4.0	3.0	11.0	
59996	2.0	7.0	3.0	11.0	
59997	1.0	6.0	3.0	11.0	
59998	6.0	1.0	3.0	11.0	

Cluster_Label_x Cluster_Label_y

0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
...
59994	1	1
59995	1	1
59996	1	1
59997	1	1
59998	1	1

[59999 rows x 30 columns]

time: 157 ms (started: 2024-04-13 18:25:37 +05:30)

```
[44]: # Import
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split, StratifiedShuffleSplit
from sklearn.tree import DecisionTreeClassifier, export_text, plot_tree # For
    ↳Decision Tree Model
from sklearn.metrics import accuracy_score, classification_report,
    ↳confusion_matrix
from sklearn.metrics import confusion_matrix, classification_report # For
    ↳Decision Tree Model Evaluation
from sklearn.neighbors import KNeighborsClassifier
from sklearn.decomposition import PCA
from matplotlib.colors import ListedColormap
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix, accuracy_score
from matplotlib.colors import ListedColormap
```

time: 0 ns (started: 2024-04-13 18:26:00 +05:30)

```
[45]: df1.columns
```

```
[45]: Index(['S.no', 'Year_Birth', 'Education', 'Marital_Status', 'Teenhome',
        'Response', 'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5',
        'AcceptedCmp1', 'AcceptedCmp2', 'Complain', 'Income', 'Kidhome',
        'Recency', 'MntWines', 'MntFruits', 'MntMeatProducts',
        'MntFishProducts', 'MntSweetProducts', 'MntGoldProds',
        'NumDealsPurchases', 'NumWebPurchases', 'NumCatalogPurchases',
        'NumStorePurchases', 'NumWebVisitsMonth', 'Z_CostContact', 'Z_Revenue',
        'Cluster_Label_x', 'Cluster_Label_y'],
        dtype='object')
```

time: 16 ms (started: 2024-04-13 18:26:00 +05:30)

```
[46]: df1_inputs = df1[['Income', 'NumWebVisitsMonth']]; df1_inputs
df1_output = df1[['Cluster_Label_x']]; df1_output

df1_inputs_names = df1_inputs.columns; df1_inputs_names
df1_output_labels = df1_output['Cluster_Label_x'].unique().astype(str);
↳df1_output_labels
```

```
[46]: array(['0', '2', '1'], dtype='<U11')
```

time: 32 ms (started: 2024-04-13 18:26:00 +05:30)

```
[47]: # Initialize StratifiedShuffleSplit with desired test size and random state
stratified_split = StratifiedShuffleSplit(n_splits=1, test_size=0.2,
↳random_state=45007)

# Perform the stratified split to get training and testing indices
for train_index, test_index in stratified_split.split(df1_inputs, df1_output):
    df1_inputs_train, df1_inputs_test = df1_inputs.iloc[train_index],
↳df1_inputs.iloc[test_index]
    df1_output_train, df1_output_test = df1_output.iloc[train_index],
↳df1_output.iloc[test_index]
```

time: 953 ms (started: 2024-04-13 18:26:00 +05:30)

```
[48]: from sklearn.linear_model import LogisticRegression
from sklearn.feature_selection import SelectFromModel
import numpy as np

# Initialize Logistic Regression model with L1 regularization
logreg_l1 = LogisticRegression(penalty='l1', solver='liblinear',
↳random_state=45007)

# Fit the model on the training data
logreg_l1.fit(df1_inputs_train, df1_output_train.values.ravel())

# Get feature importances from the fitted model
feature_importances = np.abs(logreg_l1.coef_).flatten()

# Calculate the threshold as 20% of the maximum feature importance
threshold = 0.2 * np.max(feature_importances)

# Create a selector object to select features based on non-zero coefficients
selector = SelectFromModel(logreg_l1, threshold=threshold)

# Transform the training and testing input data to select features
```

```

df1_inputs_train_selected = selector.transform(df1_inputs_train)
df1_inputs_test_selected = selector.transform(df1_inputs_test)

# Get the selected features
selected_features = df1_inputs_names[selector.get_support()]

# Print the selected features and the calculated threshold
print("Selected Features:", selected_features)
print("Threshold:", threshold)

```

Selected Features: Index(['Income'], dtype='object')
 Threshold: 0.015567165959656204
 time: 437 ms (started: 2024-04-13 18:26:01 +05:30)

```

[57]: # Decision Tree : Model (Training Subset)
dtc = DecisionTreeClassifier(criterion='gini', random_state=45052) # Other
↳ Criteria : Entropy, Log Loss
dtc_model = dtc.fit(df1_inputs_train, df1_output_train); dtc_model

```

[57]: DecisionTreeClassifier(random_state=45052)

time: 1.61 s (started: 2024-04-13 18:26:41 +05:30)

```

[58]: # Decision Tree : Model Rules
dtc_model_rules = export_text(dtc_model, feature_names =
↳ list(df1_inputs_names)); print(dtc_model_rules)

```

```

|--- NumWebVisitsMonth <= 6.50
|   |--- Income <= 0.95
|   |   |--- Income <= 0.89
|   |   |   |--- Income <= 0.89
|   |   |   |   |--- Income <= 0.89
|   |   |   |   |   |--- Income <= 0.12
|   |   |   |   |   |   |--- Income <= 0.11
|   |   |   |   |   |   |   |--- Income <= 0.00
|   |   |   |   |   |   |   |   |--- Income <= 0.00
|   |   |   |   |   |   |   |   |   |--- Income <= 0.00
|   |   |   |   |   |   |   |   |   |   |--- Income <= 0.00
|   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 15
|   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.00
|   |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 27
|   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.00
|   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.00
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 5
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.00
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 30
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.00
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- NumWebVisitsMonth <= 5.00

```

[illegible]

[illegible]

[illegible]

[illegible]

```

| | | | | | | | | | | | |--- truncated branch of depth 7
| | | | | | | | | | | | |--- Income > 0.91
| | | | | | | | | | | | |--- truncated branch of depth 14
| | | | | | | | | | | | |--- Income > 0.94
| | | | | | | | | | | | |--- class: 2
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 1.50
| | | | | | | | | | | | |--- class: 1
| | | | | | | | | | | | |--- NumWebVisitsMonth > 1.50
| | | | | | | | | | | | |--- class: 0
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 4.50
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 1.50
| | | | | | | | | | | | |--- class: 2
| | | | | | | | | | | | |--- NumWebVisitsMonth > 1.50
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 2.50
| | | | | | | | | | | | |--- class: 1
| | | | | | | | | | | | |--- NumWebVisitsMonth > 2.50
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 3.50
| | | | | | | | | | | | |--- class: 2
| | | | | | | | | | | | |--- NumWebVisitsMonth > 3.50
| | | | | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 2.50
| | | | | | | | | | | | |--- class: 0
| | | | | | | | | | | | |--- NumWebVisitsMonth > 2.50
| | | | | | | | | | | | |--- truncated branch of depth 3
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- class: 1
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- class: 2
| | | | | | | | | | | | |--- NumWebVisitsMonth > 4.50
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- class: 2
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- Income <= 0.95
| | | | | | | | | | | | |--- truncated branch of depth 3
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- class: 2
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- class: 1
| | | | | | | | | | | | |--- Income > 0.95
| | | | | | | | | | | | |--- class: 2

```

```

| |---- Income > 0.95
| | |---- Income <= 0.96
| | | |---- NumWebVisitsMonth <= 2.50
| | | | |---- Income <= 0.96
| | | | | |---- NumWebVisitsMonth <= 1.50
| | | | | | |---- class: 2
| | | | | | |---- NumWebVisitsMonth > 1.50
| | | | | | |---- class: 0
| | | | |---- Income > 0.96
| | | | | |---- NumWebVisitsMonth <= 1.50
| | | | | | |---- class: 0
| | | | | | |---- NumWebVisitsMonth > 1.50
| | | | | | |---- Income <= 0.96
| | | | | | | |---- class: 0
| | | | | | |---- Income > 0.96
| | | | | | | |---- Income <= 0.96
| | | | | | | | |---- class: 1
| | | | | | | | |---- Income > 0.96
| | | | | | | | |---- class: 0
| | | |---- NumWebVisitsMonth > 2.50
| | | | |---- Income <= 0.95
| | | | | |---- NumWebVisitsMonth <= 4.50
| | | | | | |---- NumWebVisitsMonth <= 3.50
| | | | | | | |---- class: 0
| | | | | | | |---- NumWebVisitsMonth > 3.50
| | | | | | | | |---- class: 1
| | | | | | | |---- NumWebVisitsMonth > 4.50
| | | | | | | |---- class: 0
| | | |---- Income > 0.95
| | | | |---- Income <= 0.96
| | | | | |---- NumWebVisitsMonth <= 5.50
| | | | | | |---- Income <= 0.95
| | | | | | | |---- class: 1
| | | | | | | |---- Income > 0.95
| | | | | | | |---- Income <= 0.95
| | | | | | | |---- Income <= 0.95
| | | | | | | | |---- class: 0
| | | | | | | | |---- Income > 0.95
| | | | | | | | |---- class: 2
| | | | | | | |---- Income > 0.95
| | | | | | | |---- Income <= 0.96
| | | | | | | |---- Income <= 0.96
| | | | | | | | |---- truncated branch of depth 6
| | | | | | | | |---- Income > 0.96
| | | | | | | | |---- class: 0
| | | | | | | |---- Income > 0.96
| | | | | | | |---- class: 1
| | | | |---- NumWebVisitsMonth > 5.50

```

[illegible]

[illegible]

```
| | | | | | | | | |--- Income <= 0.98
| | | | | | | | | |--- Income <= 0.98
| | | | | | | | | |--- truncated branch of depth 7
| | | | | | | | | |--- Income > 0.98
| | | | | | | | | |--- truncated branch of depth 4
| | | | | | | | | |--- Income > 0.98
| | | | | | | | | |--- Income <= 0.98
| | | | | | | | | |--- class: 0
| | | | | | | | | |--- Income > 0.98
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | | |--- Income > 0.98
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- class: 2
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- class: 0
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- NumWebVisitsMonth <= 5.50
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- class: 2
| | | | | | | | | |--- NumWebVisitsMonth > 5.50
| | | | | | | | | |--- class: 0
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- NumWebVisitsMonth <= 4.50
| | | | | | | | | |--- class: 1
| | | | | | | | | |--- NumWebVisitsMonth > 4.50
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- class: 1
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- class: 2
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- class: 2
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- class: 1
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- Income <= 1.00
| | | | | | | | | |--- Income <= 1.00
| | | | | | | | | |--- class: 0
| | | | | | | | | |--- Income > 1.00
| | | | | | | | | |--- Income <= 1.00
| | | | | | | | | |--- class: 2
| | | | | | | | | |--- Income > 1.00
| | | | | | | | | |--- class: 0
```

[illegible]


```
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 1  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 1  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 2  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 1  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 2  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 8  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 14  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 2  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 1  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- NumWebVisitsMonth <= 8.50
```

[illegible]

```
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 10.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | | | | | | |--- Income <= 0.00
```

[illegible]

[illegible]

```
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- class: 0  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- class: 2  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- class: 0  
| | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 2  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 4  
| | | | | | |--- Income > 0.00  
| | | | | | |--- class: 0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 2  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 3  
| | | | | | |--- Income > 0.00  
| | | | | | |--- class: 2  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 1  
| | | | | | |--- Income > 0.00  
| | | | | | |--- class: 0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 1
```

```
| | | | | | | | | | | Income > 0.00  
| | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 2  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- class: 1  
| | | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 0  
| | | | | | | | | | | | |--- NumWebVisitsMonth > 8.50
```

```
| | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | |--- class: 0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 2  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 7  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | |--- truncated branch of depth 5  
| | | | | |--- Income > 0.00  
| | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | |--- Income <= 0.00  
| | | | | |--- Income <= 0.00  
| | | | | |--- Income <= 0.00  
| | | | | |--- class: 0  
| | | | | |--- Income > 0.00  
| | | | | |--- class: 0  
| | | | | |--- Income > 0.00  
| | | | | |--- class: 0  
| | | | | |--- Income > 0.00  
| | | | | |--- Income <= 0.00  
| | | | | |--- class: 1  
| | | | | |--- Income > 0.00  
| | | | | |--- class: 0  
| | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | |--- Income <= 0.00  
| | | | | |--- class: 0  
| | | | | |--- Income > 0.00  
| | | | | |--- Income <= 0.00  
| | | | | |--- class: 2
```

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```

| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- NumWebVisitsMonth <= 7.50
| | | | | | | | | |--- class: 1
| | | | | | | | | |--- NumWebVisitsMonth > 7.50
| | | | | | | | | |--- Income <= 1.00
| | | | | | | | | |--- class: 1
| | | | | | | | | |--- Income > 1.00
| | | | | | | | | |--- class: 2
| | | | | | | | |--- NumWebVisitsMonth > 8.50
| | | | | | | | |--- Income <= 1.00
| | | | | | | | |--- Income <= 0.99
| | | | | | | | |--- class: 2
| | | | | | | | |--- Income > 0.99
| | | | | | | | |--- class: 0
| | | | | | | | |--- Income > 1.00
| | | | | | | | |--- class: 2

```

time: 188 ms (started: 2024-04-13 18:26:42 +05:30)

```

[59]: # Decision Tree : Feature Importance
dtc_imp_features = pd.DataFrame({'feature': df1_inputs_names, 'importance': np.
    ↳round(dtc_model.feature_importances_, 3)})
dtc_imp_features.sort_values('importance', ascending=False, inplace=True);
    ↳dtc_imp_features

```

```

[59]:          feature  importance
0          Income      0.869
1  NumWebVisitsMonth      0.131

```

time: 32 ms (started: 2024-04-13 18:26:42 +05:30)

```

[60]: # Decision Tree : Model Prediction (Training Subset)
dtc_model_predict = dtc_model.predict(df1_inputs_train); dtc_model_predict

```

```

[60]: array([1, 0, 1, ..., 2, 0, 2])

```

time: 47 ms (started: 2024-04-13 18:26:42 +05:30)

```

[61]: # Decision Tree : Prediction (Testing Subset)
dtc_predict = dtc_model.predict(df1_inputs_test); dtc_predict

```

```

[61]: array([1, 2, 2, ..., 1, 2, 2])

```

time: 32 ms (started: 2024-04-13 18:26:43 +05:30)

```

[62]: # Decision Tree : Model Evaluation (Training Subset)
dtc_model_conf_mat = pd.DataFrame(confusion_matrix(df1_output_train,
    ↳dtc_model_predict)); dtc_model_conf_mat

```

```
dtc_model_perf = classification_report(df1_output_train, dtc_model_predict);  
↳ print(dtc_model_perf)
```

	precision	recall	f1-score	support
0	0.92	0.99	0.96	16035
1	0.96	0.96	0.96	15965
2	0.99	0.92	0.95	15999
accuracy			0.95	47999
macro avg	0.96	0.95	0.95	47999
weighted avg	0.96	0.95	0.95	47999

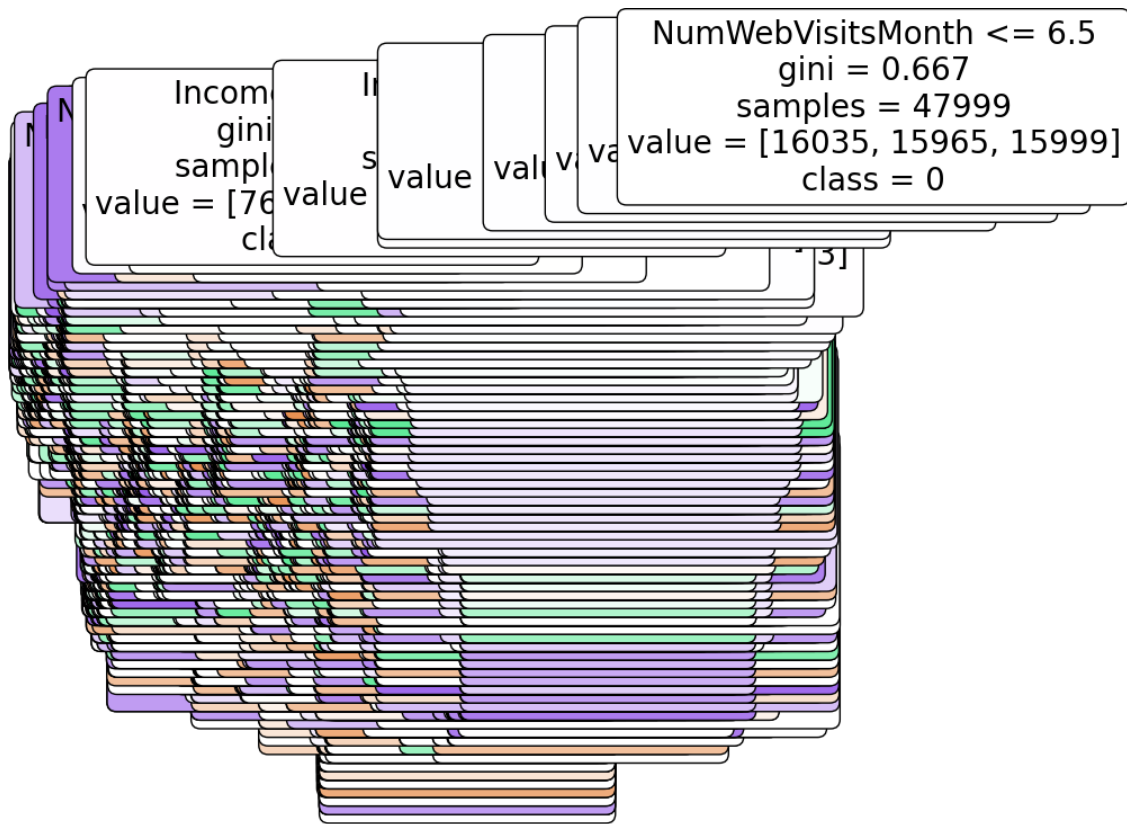
time: 172 ms (started: 2024-04-13 18:26:43 +05:30)

```
[63]: # Decision Tree : Prediction Evaluation (Testing Subset)  
dtc_predict_conf_mat = pd.DataFrame(confusion_matrix(df1_output_test,   
↳ dtc_predict)); dtc_predict_conf_mat  
dtc_predict_perf = classification_report(df1_output_test, dtc_predict);  
↳ print(dtc_predict_perf)
```

	precision	recall	f1-score	support
0	0.34	0.35	0.35	4009
1	0.33	0.33	0.33	3991
2	0.33	0.31	0.32	4000
accuracy			0.33	12000
macro avg	0.33	0.33	0.33	12000
weighted avg	0.33	0.33	0.33	12000

time: 79 ms (started: 2024-04-13 18:26:43 +05:30)

```
[64]: import matplotlib.pyplot as plt  
from sklearn.tree import plot_tree  
  
# Set a larger figure size for better clarity  
plt.figure(figsize=(10, 10))  
  
# Plot the decision tree  
train_subset_dtc_plot = plot_tree(dtc_model, feature_names=df1_inputs_names,   
↳ class_names=df1_output_labels, rounded=True, filled=True, fontsize=20)  
  
# Show the plot  
plt.show()
```



```
-----
KeyboardInterrupt                                Traceback (most recent call last)
Cell In[64], line 11
      8 train_subset_dtc_plot = plot_tree(dtc_model,
    ↪ feature_names=df1_inputs_names, class_names=df1_output_labels, rounded=True,
    ↪ filled=True, fontsize=20)
     10 # Show the plot
--> 11 plt.show()

File ~\anaconda3\lib\site-packages\matplotlib\pyplot.py:527, in show(*args,
    ↪ **kwargs)
     483 """
     484 Display all open figures.
     485
     (...
     524 explicitly there.
     525 """
```

```

526 _warn_if_gui_out_of_main_thread()
--> 527 return _get_backend_mod().show(*args, **kwargs)

File ~\anaconda3\lib\site-packages\matplotlib_inline\backend_inline.py:90, in
↳ show(close, block)
    88 try:
    89     for figure_manager in Gcf.get_all_fig_managers():
--> 90         display(
    91             figure_manager.canvas.figure,
    92             metadata=_fetch_figure_metadata(figure_manager.canvas.figure)
    93         )
    94 finally:
    95     show._to_draw = []

File ~\anaconda3\lib\site-packages\IPython\core\display_functions.py:298, in
↳ display(include, exclude, metadata, transient, display_id, raw, clear, *objs,
↳ **kwargs)
    296     publish_display_data(data=obj, metadata=metadata, **kwargs)
    297 else:
--> 298     format_dict, md_dict = format(obj, include=include, exclude=exclude
    299     if not format_dict:
    300         # nothing to display (e.g. _ipython_display_ took over)
    301         continue

File ~\anaconda3\lib\site-packages\IPython\core\formatters.py:179, in
↳ DisplayFormatter.format(self, obj, include, exclude)
    177 md = None
    178 try:
--> 179     data = formatter(obj)
    180 except:
    181     # FIXME: log the exception
    182     raise

File ~\anaconda3\lib\site-packages\decorator.py:232, in decorate.<locals>.
↳ fun(*args, **kw)
    230 if not kwsyntax:
    231     args, kw = fix(args, kw, sig)
--> 232 return caller(func, *(extras + args), **kw)

File ~\anaconda3\lib\site-packages\IPython\core\formatters.py:223, in
↳ catch_format_error(method, self, *args, **kwargs)
    221 """show traceback on failed format call"""
    222 try:
--> 223     r = method(self, *args, **kwargs)
    224 except NotImplementedError:
    225     # don't warn on NotImplementedError
    226     return self._check_return(None, args[0])

```

```
File ~\anaconda3\lib\site-packages\IPython\core\formatters.py:340, in
↳ BaseFormatter.__call__(self, obj)
```

```
    338     pass
    339 else:
--> 340     return printer(obj)
    341 # Finally look for special method names
    342 method = get_real_method(obj, self.print_method)
```

```
File ~\anaconda3\lib\site-packages\IPython\core\pylabtools.py:152, in
↳ print_figure(fig, fmt, bbox_inches, base64, **kwargs)
```

```
    149     from matplotlib.backend_bases import FigureCanvasBase
    150     FigureCanvasBase(fig)
--> 152 fig.canvas.print_figure(bytes_io, **kw)
    153 data = bytes_io.getvalue()
    154 if fmt == 'svg':
```

```
File ~\anaconda3\lib\site-packages\matplotlib\backend_bases.py:2193, in
↳ FigureCanvasBase.print_figure(self, filename, dpi, facecolor, edgecolor,
```

```
↳ orientation, format, bbox_inches, pad_inches, bbox_extra_artists, backend,
↳ **kwargs)
```

```
    2189 try:
    2190     # _get_renderer may change the figure dpi (as vector formats
    2191     # force the figure dpi to 72), so we need to set it again here.
    2192     with cbook._setattr_cm(self.figure, dpi=dpi):
-> 2193         result = print_method(
    2194             filename,
    2195             facecolor=facecolor,
    2196             edgecolor=edgecolor,
    2197             orientation=orientation,
    2198             bbox_inches_restore=_bbox_inches_restore,
    2199             **kwargs)
    2200 finally:
    2201     if bbox_inches and restore_bbox:
```

```
File ~\anaconda3\lib\site-packages\matplotlib\backend_bases.py:2043, in
↳ FigureCanvasBase._switch_canvas_and_return_print_method.<locals>.
```

```
↳ <lambda>(*args, **kwargs)
    2039     optional_kws = { # Passed by print_figure for other renderers.
    2040         "dpi", "facecolor", "edgecolor", "orientation",
    2041         "bbox_inches_restore"}
    2042     skip = optional_kws - {*inspect.signature(meth).parameters}
-> 2043     print_method = functools.wraps(meth)(lambda *args, **kwargs: meth(
    2044         *args, **{k: v for k, v in kwargs.items() if k not in skip}))
    2045 else: # Let third-parties do as they see fit.
    2046     print_method = meth
```

```
File ~\anaconda3\lib\site-packages\matplotlib\backends\backend_agg.py:497, in
↳ FigureCanvasAgg.print_png(self, filename_or_obj, metadata, pil_kwargs)
```



```

450 def print_png(self, filename_or_obj, *, metadata=None, pil_kwargs=None)
451     """
452     Write the figure to a PNG file.
453     (...)
454     *metadata*, including the default 'Software' key.
455     """
--> 456 self._print_pil(filename_or_obj, "png", pil_kwargs, metadata)

```

```

File ~\anaconda3\lib\site-packages\matplotlib\backends\backend_agg.py:445, in
FigureCanvasAgg._print_pil(self, filename_or_obj, fmt, pil_kwargs, metadata)
440 def _print_pil(self, filename_or_obj, fmt, pil_kwargs, metadata=None):
441     """
442     Draw the canvas, then save it using `.image.imsave` (to which
443     *pil_kwargs* and *metadata* are forwarded).
444     """
--> 445 FigureCanvasAgg.draw(self)
446 mpl.image.imsave(
447     filename_or_obj, self.buffer_rgba(), format=fmt, origin="upper"
448     dpi=self.figure.dpi, metadata=metadata, pil_kwargs=pil_kwargs)

```

```

File ~\anaconda3\lib\site-packages\matplotlib\backends\backend_agg.py:388, in
FigureCanvasAgg.draw(self)
385 # Acquire a lock on the shared font cache.
386 with (self.toolbar._wait_cursor_for_draw_cm() if self.toolbar
387       else nullcontext()):
--> 388 self.figure.draw(self.renderer)
389 # A GUI class may need to update a window using this draw, so
390 # don't forget to call the superclass.
391 super().draw()

```

```

File ~\anaconda3\lib\site-packages\matplotlib\artist.py:95, in
_finalize_rasterization.<locals>.draw_wrapper(artist, renderer, *args,
**kwargs)
93 @wraps(draw)
94 def draw_wrapper(artist, renderer, *args, **kwargs):
--> 95     result = draw(artist, renderer, *args, **kwargs)
96     if renderer._rasterizing:
97         renderer.stop_rasterizing()

```

```

File ~\anaconda3\lib\site-packages\matplotlib\artist.py:72, in
allow_rasterization.<locals>.draw_wrapper(artist, renderer)
69     if artist.get_agg_filter() is not None:
70         renderer.start_filter()
--> 72     return draw(artist, renderer)
73 finally:
74     if artist.get_agg_filter() is not None:

```

File ~\anaconda3\lib\site-packages\matplotlib\figure.py:3154, in Figure.

```
↪ draw(self, renderer)
    3151         # ValueError can occur when resizing a window.
    3153 self.patch.draw(renderer)
-> 3154 mimage._draw_list_compositing_images(
    3155     renderer, self, artists, self.suppressComposite)
    3157 for sfig in self.subfigs:
    3158     sfig.draw(renderer)
```

File ~\anaconda3\lib\site-packages\matplotlib\image.py:132, in

```
↪ _draw_list_compositing_images(renderer, parent, artists, suppress_composite)
    130 if not_composite or not has_images:
    131     for a in artists:
--> 132         a.draw(renderer)
    133 else:
    134     # Composite any adjacent images together
    135     image_group = []
```

File ~\anaconda3\lib\site-packages\matplotlib\artist.py:72, in

```
↪ allow_rasterization.<locals>.draw_wrapper(artist, renderer)
    69     if artist.get_agg_filter() is not None:
    70         renderer.start_filter()
---> 72     return draw(artist, renderer)
    73 finally:
    74     if artist.get_agg_filter() is not None:
```

File ~\anaconda3\lib\site-packages\matplotlib\axes_base.py:3070, in _AxesBase.

```
↪ draw(self, renderer)
    3067 if artists_rasterized:
    3068     _draw_rasterized(self.figure, artists_rasterized, renderer)
-> 3070 mimage._draw_list_compositing_images(
    3071     renderer, self, artists, self.figure.suppressComposite)
    3073 renderer.close_group('axes')
    3074 self.stale = False
```

File ~\anaconda3\lib\site-packages\matplotlib\image.py:132, in

```
↪ _draw_list_compositing_images(renderer, parent, artists, suppress_composite)
    130 if not_composite or not has_images:
    131     for a in artists:
--> 132         a.draw(renderer)
    133 else:
    134     # Composite any adjacent images together
    135     image_group = []
```

File ~\anaconda3\lib\site-packages\matplotlib\artist.py:72, in

```
↪ allow_rasterization.<locals>.draw_wrapper(artist, renderer)
    69     if artist.get_agg_filter() is not None:
    70         renderer.start_filter()
```

```

---> 72     return draw(artist, renderer)
      73 finally:
      74     if artist.get_agg_filter() is not None:

```

File ~\anaconda3\lib\site-packages\matplotlib\text.py:1988, in Annotation.

```

-> draw(self, renderer)
      1986     if self.arrow_patch.figure is None and self.figure is not None:
      1987         self.arrow_patch.figure = self.figure
-> 1988     self.arrow_patch.draw(renderer)
      1989 # Draw text, including FancyBboxPatch, after FancyArrowPatch.
      1990 # Otherwise, a wedge arrowstyle can land partly on top of the Bbox.
      1991 Text.draw(self, renderer)

```

File ~\anaconda3\lib\site-packages\matplotlib\artist.py:39, in

```

-> _prevent_rasterization.<locals>.draw_wrapper(artist, renderer, *args, **kwargs)
      36     renderer.stop_rasterizing()
      37     renderer._rasterizing = False
---> 39 return draw(artist, renderer, *args, **kwargs)

```

File ~\anaconda3\lib\site-packages\matplotlib\patches.py:4380, in

```

-> FancyArrowPatch.draw(self, renderer)
      4376 # FIXME: dpi_cor is for the dpi-dependency of the linewidth. There
      4377 # could be room for improvement. Maybe _get_path_in_displaycoord could
      4378 # take a renderer argument, but get_path should be adapted too.
      4379 self._dpi_cor = renderer.points_to_pixels(1.)
-> 4380 path, fillable = self._get_path_in_displaycoord()
      4382 if not np.iterable(fillable):
      4383     path = [path]

```

File ~\anaconda3\lib\site-packages\matplotlib\patches.py:4355, in

```

-> FancyArrowPatch._get_path_in_displaycoord(self)
      4353     posB = self._convert_xy_units(self._posA_posB[1])
      4354     (posA, posB) = self.get_transform().transform((posA, posB))
-> 4355     _path = self.get_connectionstyle()(posA, posB,
      4356                                     patchA=self.patchA,
      4357                                     patchB=self.patchB,
      4358                                     shrinkA=self.shrinkA * dpi_cor,
      4359                                     shrinkB=self.shrinkB * dpi_cor
      4360                                     )
      4361 else:
      4362     _path = self.get_transform().transform_path(self._path_original)

```

File ~\anaconda3\lib\site-packages\matplotlib\patches.py:2756, in

```

-> ConnectionStyle._Base.__call__(self, posA, posB, shrinkA, shrinkB, patchA,
-> patchB)
      2750 path = self.connect(posA, posB)
      2751 path = self._clip(
      2752     path,

```

```

2753     self._in_patch(patchA) if patchA else None,
2754     self._in_patch(patchB) if patchB else None,
2755 )
-> 2756 path = self._clip(
2757     path,
2758     inside_circle(*path.vertices[0], shrinkA) if shrinkA else None,
2759     inside_circle(*path.vertices[-1], shrinkB) if shrinkB else None
2760 )
2761 return path

```

File ~\anaconda3\lib\site-packages\matplotlib\patches.py:2739, in

```

-> ConnectionStyle._Base._clip(self, path, in_start, in_stop)
2737 if in_stop:
2738     try:
-> 2739         path, _ = split_path_inout(path, in_stop)
2740     except ValueError:
2741         pass

```

File ~\anaconda3\lib\site-packages\matplotlib\bezier.py:370, in

```

-> split_path_inout(path, inside, tolerance, reorder_inout)
367     raise ValueError("The path does not intersect with the patch")
369 bp = bezier_path.reshape((-1, 2))
--> 370 left, right = split_bezier_intersecting_with_closedpath(
371     bp, inside, tolerance)
372 if len(left) == 2:
373     codes_left = [Path.LINETO]

```

File ~\anaconda3\lib\site-packages\matplotlib\bezier.py:333, in

```

-> split_bezier_intersecting_with_closedpath(bezier, inside_closedpath, tolerance)
330 bz = BezierSegment(bezier)
331 bezier_point_at_t = bz.point_at_t
--> 333 t0, t1 = find_bezier_t_intersecting_with_closedpath(
334     bezier_point_at_t, inside_closedpath, tolerance=tolerance)
336 _left, _right = split_de_casteljau(bezier, (t0 + t1) / 2.)
337 return _left, _right

```

File ~\anaconda3\lib\site-packages\matplotlib\bezier.py:169, in

```

-> find_bezier_t_intersecting_with_closedpath(bezier_point_at_t,
-> inside_closedpath, t0, t1, tolerance)
167 # calculate the middle point
168 middle_t = 0.5 * (t0 + t1)
--> 169 middle = bezier_point_at_t(middle_t)
170 middle_inside = inside_closedpath(middle)
172 if start_inside ^ middle_inside:

```

File ~\anaconda3\lib\site-packages\matplotlib\bezier.py:222, in BezierSegment.

```

-> point_at_t(self, t)
218 def point_at_t(self, t):

```

```

219     """
220     Evaluate the curve at a single point, returning a tuple of *d*_
↪ floats.
221     """
--> 222     return tuple(self(t))

File ~\anaconda3\lib\site-packages\matplotlib\bezier.py:216, in BezierSegment.
↪ __call__(self, t)
201     """
202     Evaluate the Bézier curve at point(s) *t* in [0, 1].
203     (...)
212     Value of the curve for each point in *t*.
213     """
214     t = np.asarray(t)
215     return (np.power.outer(1 - t, self._orders[::-1])
--> 216             * np.power.outer(t, self._orders)) @ self._px

```

KeyboardInterrupt:

time: 36min 26s (started: 2024-04-13 18:26:43 +05:30)

```

[ ]: # Set up the plot
ax = plt.axes()

# Plot the confusion matrix with annotations in integer format
sns.heatmap(dtc_predict_conf_mat, annot=True, fmt='d', cmap='Paired')

# Set labels and title
ax.set_xlabel('Predicted Label')
ax.set_ylabel('True Label')
ax.set_title('Decision Tree : Confusion Matrix')

# Show the plot
plt.show()

```

```

[65]: # Cross Validation
from sklearn.model_selection import cross_val_score

# Define your decision tree classifier with desired parameters
dtc_cv = DecisionTreeClassifier(criterion='gini', random_state=45007)

# Perform 5-fold cross-validation
cv_scores = cross_val_score(dtc_cv, df1_inputs, df1_output.values.ravel(),
↪ cv=20)

print("Cross-Validation Scores:", cv_scores)
print("Average Cross-Validation Score:", np.mean(cv_scores))

```

Cross-Validation Scores: [0.34333333 0.326 0.334 0.33666667 0.339
0.32333333
0.33133333 0.32866667 0.351 0.336 0.34066667 0.333
0.33833333 0.34533333 0.352 0.35666667 0.349 0.327
0.33433333 0.33744582]
Average Cross-Validation Score: 0.33815562409692124
time: 37.9 s (started: 2024-04-13 19:03:31 +05:30)

```
[66]: from sklearn.metrics import f1_score

# Compute F1 score
f1 = f1_score(df1_output_test, dtc_predict, average='macro') # or 'weighted'
↳ for weighted F1 score
print("F1 Score:", f1)

# Weighted F1 score
weighted_f1 = f1_score(df1_output_test, dtc_predict, average='weighted')
print("Weighted F1 Score:", weighted_f1)
```

F1 Score: 0.3336188351582074
Weighted F1 Score: 0.33362729475562697
time: 47 ms (started: 2024-04-13 19:04:09 +05:30)

```
[67]: # Specify the number of neighbors (k)
k = 3

# Initialize KNN classifier with k neighbors
knn = KNeighborsClassifier(n_neighbors=k)

# Fit the KNN model using the training data
knn.fit(df1_inputs_train, df1_output_train)
```

```
[67]: KNeighborsClassifier(n_neighbors=3)

time: 94 ms (started: 2024-04-13 19:04:09 +05:30)
```

```
[68]: # Make predictions using the testing data
y_pred = knn.predict(df1_inputs_test)
```

time: 1.14 s (started: 2024-04-13 19:04:09 +05:30)

```
[69]: # Calculate accuracy
accuracy = accuracy_score(df1_output_test, y_pred)
print(f'Accuracy: {accuracy}')

# Generate classification report
classification_rep = classification_report(df1_output_test, y_pred)
print(f'Classification Report:\n{classification_rep}')
```

```
# Generate confusion matrix
conf_mat = confusion_matrix(df1_output_test, y_pred)
print(f'Confusion Matrix:\n{conf_mat}')
```

Accuracy: 0.3358333333333333

Classification Report:

	precision	recall	f1-score	support
0	0.34	0.49	0.40	4009
1	0.34	0.26	0.29	3991
2	0.33	0.26	0.29	4000
accuracy			0.34	12000
macro avg	0.34	0.34	0.33	12000
weighted avg	0.34	0.34	0.33	12000

Confusion Matrix:

```
[[1953 1005 1051]
 [1873 1027 1091]
 [1940 1010 1050]]
```

time: 94 ms (started: 2024-04-13 19:04:10 +05:30)

```
[70]: k_values = [7, 9, 11, 13, 15]
for k in k_values:
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(df1_inputs_train, df1_output_train) # Use your training data here
    y_pred = knn.predict(df1_inputs_test) # Use your testing data here
    accuracy = accuracy_score(df1_output_test, y_pred) # Compare predictions
    with true labels
    print(f'Accuracy for k={k}: {accuracy}')
```

Accuracy for k=7: 0.333416666666666664

Accuracy for k=9: 0.33666666666666667

Accuracy for k=11: 0.33316666666666667

Accuracy for k=13: 0.33783333333333333

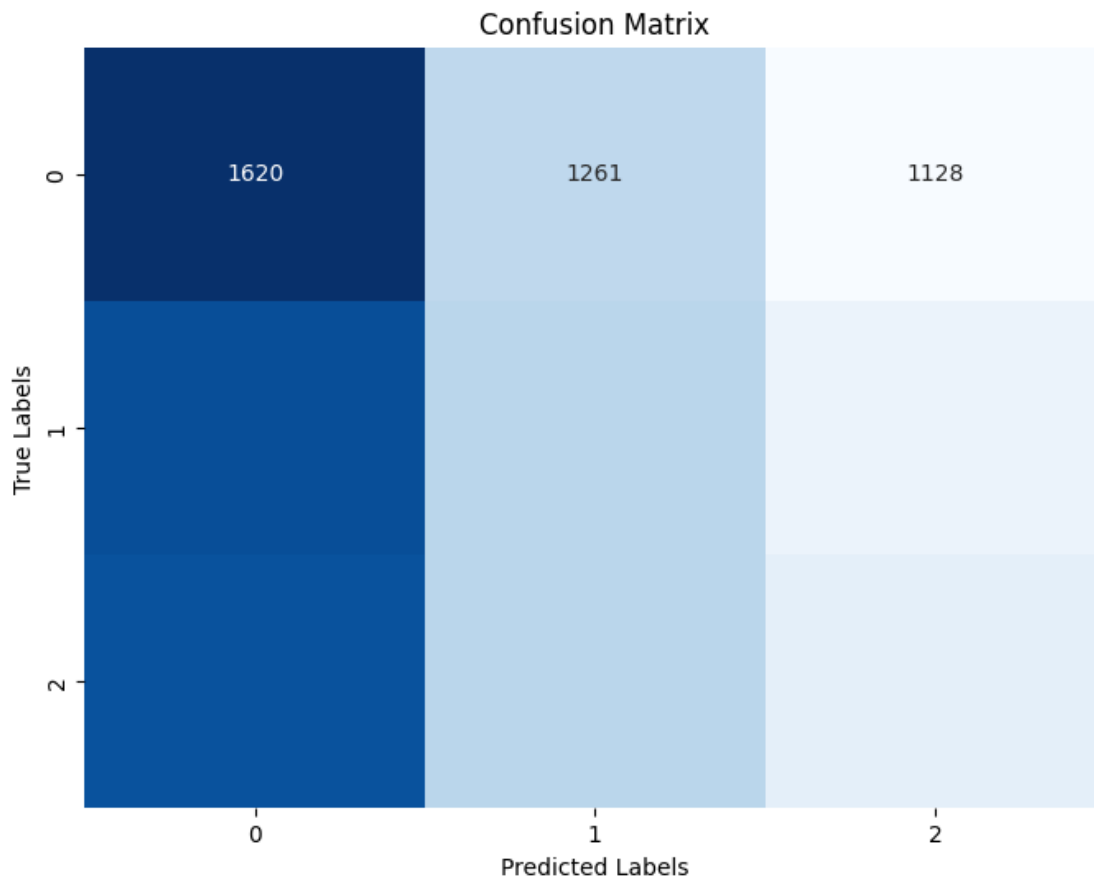
Accuracy for k=15: 0.33875

time: 5.53 s (started: 2024-04-13 19:04:10 +05:30)

```
[71]: # Plot confusion matrix
def plot_confusion_matrix(y_true, y_pred):
    cm = confusion_matrix(y_true, y_pred)
    plt.figure(figsize=(8, 6))
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False)
    plt.title('Confusion Matrix')
    plt.xlabel('Predicted Labels')
    plt.ylabel('True Labels')
```

```
plt.show()

# Assuming df1_output_test and y_pred are the true and predicted labels,
↳ respectively
plot_confusion_matrix(df1_output_test, y_pred)
```



time: 281 ms (started: 2024-04-13 19:04:16 +05:30)

```
[72]: from sklearn.model_selection import cross_val_score
from sklearn.neighbors import KNeighborsClassifier

# Define the KNN classifier with a chosen number of neighbors (k)
knn = KNeighborsClassifier(n_neighbors=13) # Example value, you can adjust this

# Perform cross-validation with 5 folds
cv_scores = cross_val_score(knn, df1_inputs, df1_output.values.ravel(), cv=20)

# Print the cross-validation scores
print("Cross-Validation Scores:", cv_scores)
```



```
# Calculate and print the average accuracy
avg_accuracy = cv_scores.mean()
print("Average Accuracy:", avg_accuracy)
```

```
Cross-Validation Scores: [0.328      0.34333333 0.346      0.331      0.34833333
0.32566667
0.33333333 0.32633333 0.32666667 0.33566667 0.32666667 0.34333333
0.33366667 0.35166667 0.32266667 0.33333333 0.33633333 0.33233333
0.34566667 0.34044682]
Average Accuracy: 0.33552234078026005
time: 7.16 s (started: 2024-04-13 19:04:16 +05:30)
```

```
[73]: from sklearn.svm import SVC
from sklearn.utils.validation import column_or_1d

# Ensure the shape of the target variable is correct
df1_output_train = column_or_1d(df1_output_train)

# Initialize SVC classifier with linear kernel
classifier = SVC(kernel='linear', random_state=45052)

# Fit the classifier to the training data
classifier.fit(df1_inputs_train, df1_output_train)
```

```
[73]: SVC(kernel='linear', random_state=45052)

time: 3min 37s (started: 2024-04-13 19:04:23 +05:30)
```

```
[74]: y_pred = classifier.predict(df1_inputs_test)
```

```
time: 25.6 s (started: 2024-04-13 19:08:01 +05:30)
```

```
[75]: cm = confusion_matrix(df1_output_test, y_pred)
print(cm)
accuracy_score(df1_output_test, y_pred)
```

```
[[3695    1   313]
 [3687    0   304]
 [3722    0   278]]
```

```
[75]: 0.33108333333333334
```

```
time: 31 ms (started: 2024-04-13 19:08:26 +05:30)
```

```
[76]: from sklearn.model_selection import cross_val_score
from sklearn.svm import SVC
```

```

# Perform K-fold cross-validation with 3 folds and enable parallel processing
cv_scores = cross_val_score(classifier, df1_inputs, df1_output.values.ravel(),
    ↪cv=5, n_jobs=-1)

# Print the cross-validation scores
print("Cross-validation scores:", cv_scores)

# Calculate and print the average cross-validation score
avg_cv_score = np.mean(cv_scores)
print("Average Cross-validation score:", avg_cv_score)

```

Cross-validation scores: [0.33283333 0.3345 0.33516667 0.33666667 0.3356113
]
 Average Cross-validation score: 0.33495559352168236
 time: 3min 49s (started: 2024-04-13 19:08:26 +05:30)

```

[77]: from sklearn.model_selection import cross_val_score
      from sklearn.svm import SVC
      from sklearn.metrics import make_scorer, f1_score

# Define a scorer for weighted F1 score
weighted_f1_scorer = make_scorer(f1_score, average='weighted')

# Perform K-fold cross-validation with 5 folds using weighted F1 score as the
    ↪scoring metric
cv_scores_weighted_f1 = cross_val_score(classifier, df1_inputs, df1_output.
    ↪values.ravel(), cv=5, scoring=weighted_f1_scorer)

# Print the cross-validation scores for weighted F1
print("Cross-validation scores (Weighted F1):", cv_scores_weighted_f1)

# Calculate and print the average cross-validation score for weighted F1
avg_cv_score_weighted_f1 = np.mean(cv_scores_weighted_f1)
print("Average Cross-validation score (Weighted F1):", avg_cv_score_weighted_f1)

```

Cross-validation scores (Weighted F1): [0.20192186 0.20399325 0.20243803
 0.20785549 0.25968666]
 Average Cross-validation score (Weighted F1): 0.21517905943747903
 time: 13min 47s (started: 2024-04-13 19:12:16 +05:30)

RANDOM FOREST

```

[82]: import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      from sklearn.model_selection import train_test_split, StratifiedShuffleSplit,
    ↪cross_val_score

```

```

from sklearn.tree import DecisionTreeClassifier, export_text, plot_tree
from sklearn.metrics import accuracy_score, classification_report,
    ↪confusion_matrix, f1_score
from sklearn.neighbors import KNeighborsClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.feature_selection import SelectFromModel
from sklearn.svm import SVC
import numpy as np
import time
import psutil

# Function to measure memory usage
def memory_usage():
    process = psutil.Process()
    return process.memory_info().rss / 1024 ** 2 # Memory usage in MB

# Start time
start_time = time.time()

# Data preprocessing and splitting
# Assuming you have your data loaded into cars_inputs and cars_output
df1_inputs_train, df1_inputs_test, df1_output_train, df1_output_test =
    ↪train_test_split(df1_inputs, df1_output, test_size=0.2, random_state=42)

# End time
end_time = time.time()

# Time taken for data preprocessing and splitting
data_preprocessing_time = end_time - start_time

# Memory usage after data preprocessing
data_preprocessing_memory = memory_usage()

# Decision Tree
dt_start_time = time.time()
dt_model = DecisionTreeClassifier(criterion='gini', random_state=45052,
    ↪max_depth=3)
dt_model.fit(df1_inputs_train, df1_output_train)
dt_training_time = time.time() - dt_start_time
dt_memory_used = memory_usage()
dt_pred = dt_model.predict(df1_inputs_test)
dt_accuracy = accuracy_score(df1_output_test, dt_pred)

# Cross-validation for Decision Tree
dtc_cv_start_time = time.time()
dtc_cv = DecisionTreeClassifier(criterion='gini', random_state=45005)

```

```

cv_scores_dtc = cross_val_score(dtc_cv, df1_inputs, df1_output.values.ravel(),
    ↪cv=20)
dtc_cv_time = time.time() - dtc_cv_start_time
dtc_cv_accuracy = np.mean(cv_scores_dtc)

print("Decision Tree:")
print(f" - Training Time (s): {dt_training_time}")
print(f" - Memory Used (MB): {dt_memory_used}")
print(f" - Single Split Accuracy: {dt_accuracy}")
print(f" - Cross Validation Accuracy: {dtc_cv_accuracy}")
print()

```

Decision Tree:

- Training Time (s): 0.048932790756225586
- Memory Used (MB): 428.61328125
- Single Split Accuracy: 0.3323333333333333
- Cross Validation Accuracy: 0.3388722796487718

time: 23.1 s (started: 2024-04-13 21:58:58 +05:30)

```

[85]: ## Data Visualization Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.graph_objects as go
from collections import Counter
from scipy import stats
from sklearn.tree import plot_tree
from IPython.display import display
from collections import Counter

## Machine Learning Models and Evaluation Metrics
from sklearn.ensemble import RandomForestClassifier
from sklearn.utils.validation import column_or_1d
from sklearn.metrics import accuracy_score, classification_report,
    ↪confusion_matrix, f1_score, precision_recall_fscore_support
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LogisticRegression, Lasso, Ridge
from sklearn.metrics import make_scorer
from sklearn.pipeline import make_pipeline
from sklearn.tree import export_graphviz

```

time: 171 ms (started: 2024-04-13 22:00:16 +05:30)

```

[86]: rf_classifier = RandomForestClassifier(n_estimators=100, random_state=45052)

```

time: 0 ns (started: 2024-04-13 22:00:41 +05:30)

```
[89]: rf_classifier.fit(df1_inputs_train, df1_output_train['Cluster_Label_x'])
```

```
[89]: RandomForestClassifier(random_state=45052)
```

time: 38.1 s (started: 2024-04-13 22:03:13 +05:30)

```
[90]: y_train_pred_rf = rf_classifier.predict(df1_inputs_train)
      y_test_pred_rf = rf_classifier.predict(df1_inputs_test)
```

time: 2.59 s (started: 2024-04-13 22:03:51 +05:30)

```
[93]: # Train the Random Forest classifier
      rf_classifier.fit(df1_inputs_train, df1_output_train['Cluster_Label_x'])

      # Print feature importances
      feature_importances = rf_classifier.feature_importances_
      feature_importance_df = pd.DataFrame({'Feature': df1_inputs_train.columns,
      ↪ 'Importance': feature_importances})
      sorted_feature_importance_df = feature_importance_df.
      ↪ sort_values(by='Importance', ascending=False)
      print("Feature Importances:")
      print(sorted_feature_importance_df)
```

Feature Importances:

	Feature	Importance
0	Income	0.990607
1	NumWebVisitsMonth	0.009393

time: 39.7 s (started: 2024-04-13 22:04:12 +05:30)

```
[ ]: # For training set
      print("Training Set Confusion Matrix:")
      print(confusion_matrix(df1_output_train['Cluster_Label_x'], y_train_pred_rf))

      print("\nTraining Set Classification Report:")
      print(classification_report(df1_output_train['Cluster_Label_x'],
      ↪ y_train_pred_rf))
```

```
[95]: # For testing set
      print("\nTesting Set Confusion Matrix:")
      print(confusion_matrix(df1_output_test['Cluster_Label_x'], y_test_pred_rf))

      print("\nTesting Set Classification Report:")
      print(classification_report(df1_output_test['Cluster_Label_x'], y_test_pred_rf))
```

Testing Set Confusion Matrix:

[[1442	1256	1282]
[1344	1334	1355]

[1314 1364 1309]]

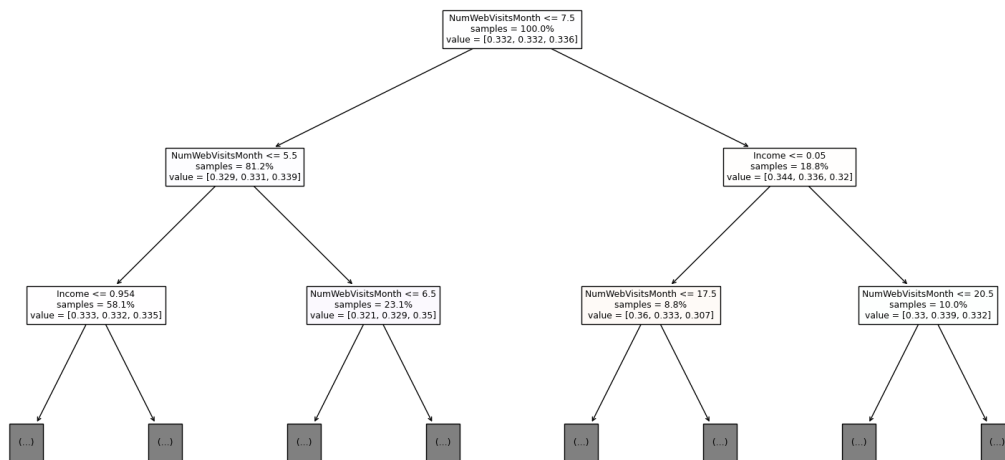
Testing Set Classification Report:

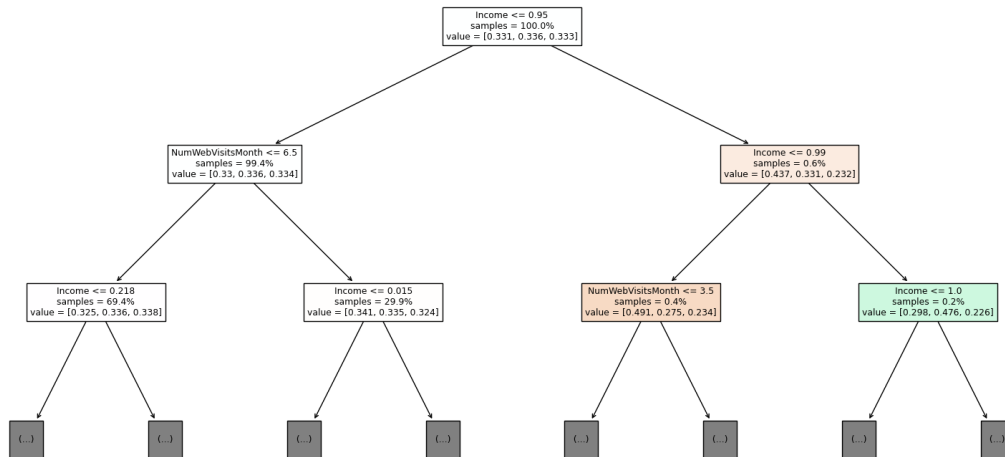
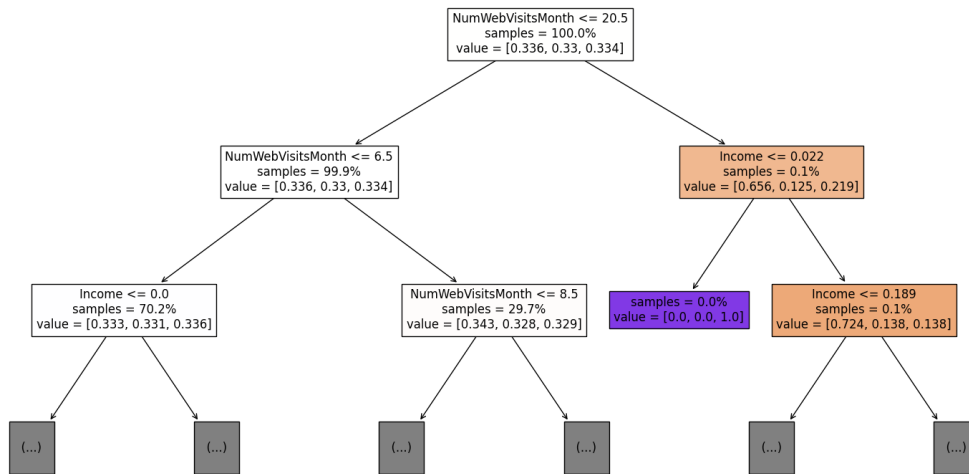
	precision	recall	f1-score	support
0	0.35	0.36	0.36	3980
1	0.34	0.33	0.33	4033
2	0.33	0.33	0.33	3987
accuracy			0.34	12000
macro avg	0.34	0.34	0.34	12000
weighted avg	0.34	0.34	0.34	12000

time: 47 ms (started: 2024-04-13 22:05:01 +05:30)

```
[96]: from sklearn.tree import plot_tree
import matplotlib.pyplot as plt

# Assuming rf is your trained Random Forest classifier
for i in range(3):
    tree = rf_classifier.estimators_[i] # Assuming rf is your trained Random
    ↪Forest model
    plt.figure(figsize=(20,10))
    plot_tree(tree, feature_names=df1_inputs_train.columns, filled=True,
    ↪max_depth=2, impurity=False, proportion=True)
    plt.show()
```





time: 2.92 s (started: 2024-04-13 22:05:12 +05:30)

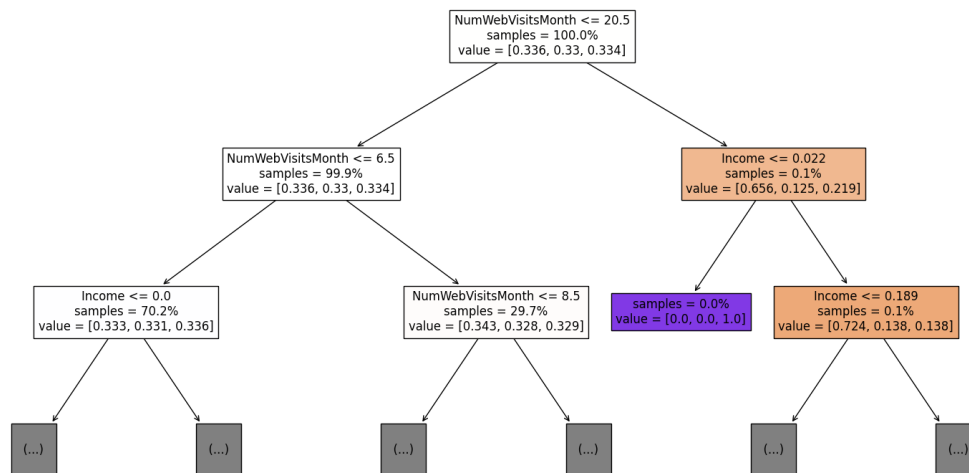
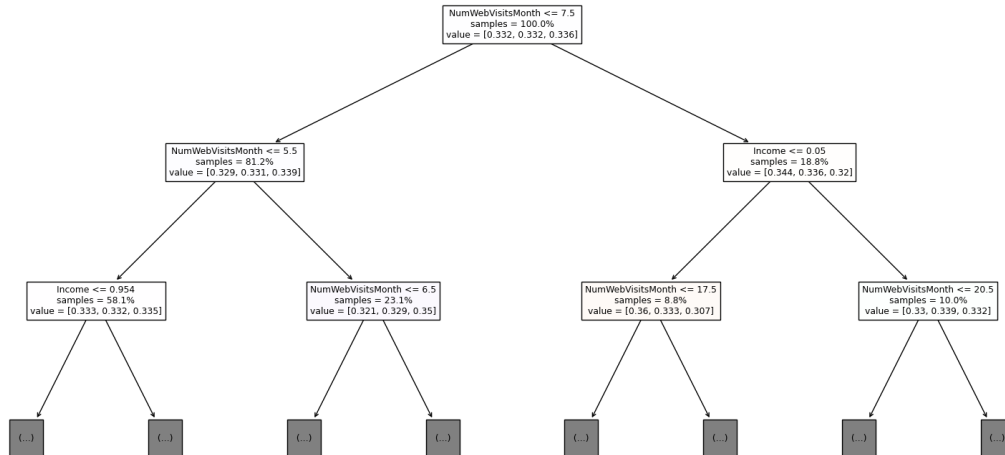
```
[97]: from sklearn.tree import plot_tree
import matplotlib.pyplot as plt

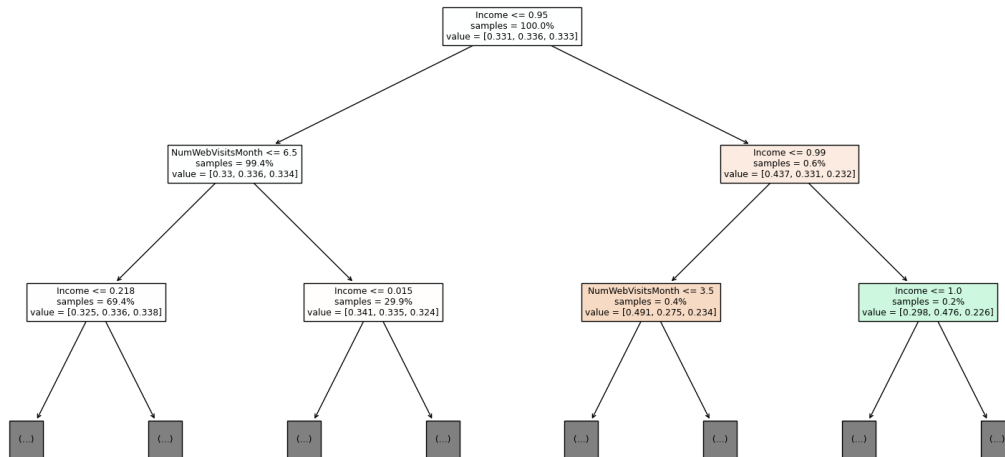
# Export and visualize the first three decision trees
for i in range(3):
    plt.figure(figsize=(20, 10))
    tree = rf_classifier.estimators_[i]
```

```

plot_tree(tree, feature_names=df1_inputs_train.columns, filled=True,
↳max_depth=2, impurity=False, proportion=True)
plt.savefig(f"decision_tree_{i+1}.png") # Save the decision tree as an
↳image file
plt.show()

```





time: 4 s (started: 2024-04-13 22:05:32 +05:30)

```
[98]: # Initialize a dictionary to store tree frequencies and rules
tree_frequency = Counter()
tree_rules = {}

# Loop through the trees and count their frequency while storing rules
for i in range(len(rf_classifier.estimators_)):
    tree = rf_classifier.estimators_[i]
    tree_str = export_text(tree, feature_names=list(df1_inputs_train.columns))
    tree_frequency[tree_str] += 1
    tree_rules[tree_str] = tree

# Get the three most frequent trees
top_trees = tree_frequency.most_common(3)

# Print the rules for the top three trees
for tree_str, frequency in top_trees:
    print(f"Tree Frequency: {frequency}")
    print(tree_str)
    print("\n")
```

```
Tree Frequency: 1
|--- NumWebVisitsMonth <= 7.50
|   |--- NumWebVisitsMonth <= 5.50
|       |--- Income <= 0.95
|           |--- Income <= 0.88
|               |--- NumWebVisitsMonth <= 0.50
|                   |--- Income <= 0.12
```

[illegible]

[illegible]

```
| | | | | | | | | |--- Income > 0.45  
| | | | | | | | | |--- truncated branch of depth 10  
| | | | | | | | | |--- Income > 0.60  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.61  
| | | | | | | | |--- Income <= 0.81  
| | | | | | | | |--- Income <= 0.80  
| | | | | | | | |--- Income <= 0.79  
| | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | |--- Income > 0.79  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.81  
| | | | | | | | |--- Income <= 0.86  
| | | | | | | | |--- Income <= 0.84  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.84  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 0.86  
| | | | | | | | |--- class: 1.0  
| | | | | | |--- NumWebVisitsMonth > 0.50  
| | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | |--- NumWebVisitsMonth <= 1.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 21  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 7  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 5  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 3  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 4  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 3  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.58  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 0.0
```

```
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- Income <= 0.57  
| | | | | | | | |--- truncated branch of depth 73  
| | | | | | | | |--- Income > 0.57  
| | | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.58  
| | | | | | | | |--- Income <= 0.85  
| | | | | | | | |--- Income <= 0.60  
| | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | |--- Income > 0.60  
| | | | | | | | |--- truncated branch of depth 18  
| | | | | | | |--- Income > 0.85  
| | | | | | | | |--- Income <= 0.86  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.86  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | |--- NumWebVisitsMonth > 1.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income<= 0.00  
| | | | | | |--- NumWebVisitsMonth <= 3.50  
| | | | | | |--- truncated branch of depth 4  
| | | | | | |--- NumWebVisitsMonth > 3.50  
| | | | | | |--- truncated branch of depth 6  
| | | | | | |--- Income > 0.00  
| | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | |--- truncated branch of depth 2  
| | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | |--- truncated branch of depth 5  
| | | | | |--- Income > 0.00  
| | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- class: 1.0  
| | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 3  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 5  
| | | | | |--- Income > 0.00  
| | | | | | |--- NumWebVisitsMonth <= 3.50  
| | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 1.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 60  
| | | | | |--- NumWebVisitsMonth > 2.50
```

```
| | | | | | | | | | --- Income <= 0.68  
| | | | | | | | | | |--- truncated branch of depth 55  
| | | | | | | | | | |--- Income > 0.68  
| | | | | | | | | | |--- truncated branch of depth 15  
| | | | | | | | | | |--- NumWebVisitsMonth > 3.50  
| | | | | | | | | | |--- Income <= 0.37  
| | | | | | | | | | |--- Income <= 0.36  
| | | | | | | | | | |--- truncated branch of depth 76  
| | | | | | | | | | |--- Income > 0.36  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.37  
| | | | | | | | | | |--- Income <= 0.48  
| | | | | | | | | | |--- truncated branch of depth 14  
| | | | | | | | | | |--- Income > 0.48  
| | | | | | | | | | |--- truncated branch of depth 28  
| | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | |--- Income <= 0.83  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 13  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 2  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 5  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 7  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- truncated branch of depth 5  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 27  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- truncated branch of depth 2  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.20  
| | | | | | |--- Income <= 0.20  
| | | | | | |--- Income <= 0.06  
| | | | | | |--- truncated branch of depth 53  
| | | | | | |--- Income > 0.06  
| | | | | | |--- truncated branch of depth 54  
| | | | | | |--- Income > 0.20
```

```
| | | | | | | | | | | Income <= 0.20  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.20  
| | | | | | | | | | | |--- truncated branch of depth 8  
| | | | | | | | | | | |--- Income > 0.20  
| | | | | | | | | | | |--- Income <= 0.28  
| | | | | | | | | | | |--- Income <= 0.21  
| | | | | | | | | | | |--- truncated branch of depth 14  
| | | | | | | | | | | |--- Income > 0.21  
| | | | | | | | | | | |--- truncated branch of depth 18  
| | | | | | | | | | | |--- Income > 0.28  
| | | | | | | | | | | |--- Income <= 0.45  
| | | | | | | | | | | |--- truncated branch of depth 19  
| | | | | | | | | | | |--- Income > 0.45  
| | | | | | | | | | | |--- truncated branch of depth 21  
| | | | | | | | | | |--- Income > 0.83  
| | | | | | | | | | |--- Income <= 0.87  
| | | | | | | | | | |--- Income <= 0.85  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.85  
| | | | | | | | | | |--- Income <= 0.85  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.85  
| | | | | | | | | | |--- Income <= 0.86  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.86  
| | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | |--- Income > 0.87  
| | | | | | | | | | |--- Income <= 0.87  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.87  
| | | | | | | | | | |--- Income <= 0.88  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.88  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.88  
| | | | | | | | | | |--- NumWebVisitsMonth <= 3.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 0.50  
| | | | | | | | | | |--- Income <= 0.89  
| | | | | | | | | | |--- Income <= 0.88  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.88  
| | | | | | | | | | |--- Income <= 0.89  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.89  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.89
```

```
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- Income <= 0.92
| | | | | | | | |--- Income <= 0.90
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.90
| | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.92
| | | | | | | | |--- Income <= 0.92
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.92
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.94
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.94
| | | | | | | | |--- class: 0.0
| | | | | | | |--- NumWebVisitsMonth > 0.50
| | | | | | | |--- NumWebVisitsMonth <= 1.50
| | | | | | | |--- Income <= 0.92
| | | | | | | |--- Income <= 0.91
| | | | | | | | |--- Income <= 0.91
| | | | | | | | |--- truncated branch of depth 7
| | | | | | | | |--- Income > 0.91
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.91
| | | | | | | | |--- class: 2.0
| | | | | | | |--- Income > 0.92
| | | | | | | |--- Income <= 0.93
| | | | | | | | |--- class: 1.0
| | | | | | | |--- Income > 0.93
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.94
| | | | | | | | |--- truncated branch of depth 4
| | | | | | | |--- NumWebVisitsMonth > 1.50
| | | | | | | |--- Income <= 0.94
| | | | | | | |--- Income <= 0.93
| | | | | | | | |--- Income <= 0.93
| | | | | | | | |--- truncated branch of depth 6
| | | | | | | | |--- Income > 0.93
| | | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.93
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.94
| | | | | | | | |--- truncated branch of depth 2
| | | | | | | |--- Income > 0.94
| | | | | | | |--- Income <= 0.95
```



```
| | | | | | | | | |--- Income <= 0.95
| | | | | | | | | |--- class: 1.0
| | | | | | | | | |--- Income > 0.95
| | | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.95
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- NumWebVisitsMonth > 2.50
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- Income <= 0.90
| | | | | | | | |--- Income <= 0.88
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.88
| | | | | | | | |--- Income <= 0.88
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.88
| | | | | | | | |--- Income <= 0.89
| | | | | | | | |--- truncated branch of depth 5
| | | | | | | | |--- Income > 0.89
| | | | | | | | |--- truncated branch of depth 4
| | | | | | | | |--- Income > 0.90
| | | | | | | | |--- Income <= 0.92
| | | | | | | | |--- Income <= 0.91
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.91
| | | | | | | | |--- Income <= 0.92
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.92
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.92
| | | | | | | | |--- Income <= 0.93
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.93
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.94
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.94
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- NumWebVisitsMonth > 3.50
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- Income <= 0.94
| | | | | | | | |--- NumWebVisitsMonth <= 4.50
| | | | | | | | |--- Income <= 0.93
| | | | | | | | |--- Income <= 0.89
| | | | | | | | |--- Income <= 0.89
| | | | | | | | |--- truncated branch of depth 4
| | | | | | | | |--- Income > 0.89
| | | | | | | | |--- class: 1.0
```

```
| | | | | | | | | | Income > 0.89  
| | | | | | | | | | |--- Income <= 0.92  
| | | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | | |--- Income > 0.92  
| | | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | | |--- Income > 0.93  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | | | | |--- Income <= 0.92  
| | | | | | | | | | | |--- Income <= 0.89  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.89  
| | | | | | | | | | | |--- Income <= 0.90  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.90  
| | | | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | | |--- Income > 0.92  
| | | | | | | | | | |--- Income <= 0.92  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.92  
| | | | | | | | | | | |--- Income <= 0.93  
| | | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | | |--- Income > 0.93  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.94  
| | | | | | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | | | | |--- Income <= 0.94  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.94  
| | | | | | | | | | | |--- Income <= 0.94  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | | |--- Income > 0.94  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.94  
| | | | | | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | | | | | |--- Income <= 0.95  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.95  
| | | | | | | | | | | |--- Income <= 0.95  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.95  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | | | | |--- Income <= 0.94  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.94
```

```
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |    |--- Income <= 0.95  
| | | | | | | | |    |    |--- class: 2.0  
| | | | | | | | |    |    |--- Income > 0.95  
| | | | | | | | |    |    |    |--- class: 1.0  
| | | | | | | | |    |--- Income > 0.95  
| | | | | | | | |    |    |--- class: 2.0  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |    |--- Income <= 0.95  
| | | | | | | | |    |    |--- class: 1.0  
| | | | | | | | |    |--- Income > 0.95  
| | | | | | | | |    |    |--- class: 0.0  
  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |    |--- Income <= 0.96  
| | | | | | | | |    |    |--- class: 0.0  
| | | | | | | | |    |--- Income > 0.96  
| | | | | | | | |    |    |--- NumWebVisitsMonth <= 4.50  
| | | | | | | | |    |        |--- NumWebVisitsMonth <= 3.50  
| | | | | | | | |    |        |--- Income <= 0.98  
| | | | | | | | |    |        |--- Income <= 0.97  
| | | | | | | | |    |        |--- Income <= 0.97  
| | | | | | | | |    |        |--- Income <= 0.96  
| | | | | | | | |    |            |--- NumWebVisitsMonth <= 1.50  
| | | | | | | | |    |            |    |--- class: 0.0  
| | | | | | | | |    |            |    |--- NumWebVisitsMonth > 1.50  
| | | | | | | | |    |            |    |--- truncated branch of depth 5  
| | | | | | | | |    |--- Income > 0.96  
| | | | | | | | |    |    |--- Income <= 0.97  
| | | | | | | | |    |    |    |--- truncated branch of depth 4  
| | | | | | | | |    |    |    |--- Income > 0.97  
| | | | | | | | |    |    |    |--- truncated branch of depth 3  
| | | | | | | | |--- Income > 0.97  
| | | | | | | | |    |--- NumWebVisitsMonth <= 2.00  
| | | | | | | | |    |    |--- class: 0.0  
| | | | | | | | |    |--- NumWebVisitsMonth > 2.00  
| | | | | | | | |    |    |--- Income <= 0.97  
| | | | | | | | |    |    |    |--- class: 2.0  
| | | | | | | | |    |    |    |--- Income > 0.97  
| | | | | | | | |    |    |    |--- class: 0.0  
  
| | | | | | | | |--- Income > 0.97  
| | | | | | | | |    |--- NumWebVisitsMonth <= 2.50  
| | | | | | | | |    |--- Income <= 0.98  
| | | | | | | | |    |    |--- class: 1.0  
| | | | | | | | |    |--- Income > 0.98  
| | | | | | | | |    |    |--- NumWebVisitsMonth <= 1.50  
| | | | | | | | |    |    |    |--- truncated branch of depth 3  
| | | | | | | | |    |    |    |--- NumWebVisitsMonth > 1.50
```

```
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | | | | |--- Income <= 0.98  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.98  
| | | | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.98  
| | | | | | | | |--- NumWebVisitsMonth <= 1.50  
| | | | | | | | |--- Income <= 1.00  
| | | | | | | | |--- Income <= 0.99  
| | | | | | | | | |--- Income <= 0.98  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.98  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.99  
| | | | | | | | |--- Income <= 1.00  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 1.00  
| | | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 1.00  
| | | | | | | |--- Income <= 1.00  
| | | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 1.00  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- NumWebVisitsMonth > 1.50  
| | | | | | | |--- Income <= 1.00  
| | | | | | | |--- Income <= 0.99  
| | | | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | |--- Income > 0.99  
| | | | | | | | |--- Income <= 0.99  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.99  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 1.00  
| | | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | | | |--- class: 0.0  
| | | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- NumWebVisitsMonth > 3.50  
| | | | | | | |--- Income <= 0.96  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.96  
| | | | | | | |--- Income <= 0.96  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.96
```

```
| | | | | | | | |--- Income <= 0.97
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.97
| | | | | | | | |--- Income <= 0.98
| | | | | | | | |--- Income <= 0.98
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.98
| | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.98
| | | | | | | | |--- Income <= 0.98
| | | | | | | | |--- truncated branch of depth 4
| | | | | | | | |--- Income > 0.98
| | | | | | | | |--- truncated branch of depth 6
| | | | | | | | |--- NumWebVisitsMonth > 4.50
| | | | | | | | |--- Income <= 0.97
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.97
| | | | | | | | |--- Income <= 1.00
| | | | | | | | |--- Income <= 0.98
| | | | | | | | |--- Income <= 0.97
| | | | | | | | |--- Income <= 0.97
| | | | | | | | |--- Income <= 0.97
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.97
| | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.97
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.97
| | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.98
| | | | | | | | |--- Income <= 0.98
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.98
| | | | | | | | |--- Income <= 0.99
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.99
| | | | | | | | |--- Income <= 0.99
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.99
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 1.00
| | | | | | | | |--- class: 0.0
|--- NumWebVisitsMonth > 5.50
| |--- NumWebVisitsMonth <= 6.50
| | |--- Income <= 0.10
| | | |--- Income <= 0.05
| | | |--- Income <= 0.04
| | | |--- Income <= 0.03
```

```

| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- truncated branch of depth 20
| | | | | | | |---- Income > 0.00
| | | | | | | |---- truncated branch of depth 13
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- truncated branch of depth 9
| | | | | | | |---- Income > 0.00
| | | | | | | |---- truncated branch of depth 11
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- truncated branch of depth 3
| | | | | | | |---- Income > 0.00
| | | | | | | |---- class: 1.0
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- truncated branch of depth 19
| | | | | | | |---- Income > 0.00
| | | | | | | |---- truncated branch of depth 28
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- class: 2.0
| | | | | | | |---- Income > 0.00
| | | | | | | |---- class: 0.0
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- class: 2.0
| | | | | | | |---- Income > 0.00
| | | | | | | |---- truncated branch of depth 2
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- truncated branch of depth 10
| | | | | | | |---- Income > 0.00
| | | | | | | |---- truncated branch of depth 8
| | | | | | | |---- Income > 0.00
| | | | | | | |---- Income <= 0.00
| | | | | | | |---- truncated branch of depth 7
| | | | | | | |---- Income > 0.00
| | | | | | | |---- truncated branch of depth 22
| | | | | | | |---- Income > 0.03
| | | | | | | |---- Income <= 0.03

```

[illegible]

[illegible]


```
| | | | | | | | | |--- Income <= 0.06
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | | |--- Income > 0.06
| | | | | | | | | |--- class: 1.0
| | | | | | | | | |--- Income > 0.06
| | | | | | | | | |--- class: 2.0
| | | | | | | |--- Income > 0.07
| | | | | | | |--- Income <= 0.07
| | | | | | | |--- Income <= 0.07
| | | | | | | |--- Income <= 0.07
| | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.07
| | | | | | | |--- Income <= 0.07
| | | | | | | |--- truncated branch of depth 5
| | | | | | | |--- Income > 0.07
| | | | | | | |--- class: 2.0
| | | | | | |--- Income > 0.07
| | | | | | |--- Income <= 0.07
| | | | | | |--- Income <= 0.07
| | | | | | |--- truncated branch of depth 4
| | | | | | |--- Income > 0.07
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.07
| | | | | | |--- Income <= 0.07
| | | | | | |--- truncated branch of depth 9
| | | | | | |--- Income > 0.07
| | | | | | |--- truncated branch of depth 7
| | | | | |--- Income > 0.07
| | | | | |--- Income <= 0.07
| | | | | |--- Income <= 0.07
| | | | | |--- Income <= 0.07
| | | | | |--- truncated branch of depth 3
| | | | | |--- Income > 0.07
| | | | | |--- truncated branch of depth 8
| | | | | |--- Income > 0.07
| | | | | |--- Income <= 0.07
| | | | | |--- class: 1.0
| | | | | |--- Income > 0.07
| | | | | |--- truncated branch of depth 4
| | | | | |--- Income > 0.07
| | | | | |--- class: 0.0
| | | | |--- Income > 0.07
| | | | |--- Income <= 0.10
| | | | |--- Income <= 0.09
| | | | |--- Income <= 0.09
| | | | |--- Income <= 0.09
| | | | |--- Income <= 0.08
| | | | |--- truncated branch of depth 18
```

```
| | | | | | | | | |--- Income > 0.08  
| | | | | | | | | |--- truncated branch of depth 18  
| | | | | | | | | |--- Income > 0.09  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.09  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.09  
| | | | | | | | |--- Income <= 0.09  
| | | | | | | | |--- Income <= 0.09  
| | | | | | | | |--- Income <= 0.09  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.09  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.09  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.09  
| | | | | | | | |--- Income <= 0.09  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.09  
| | | | | | | | |--- Income <= 0.10  
| | | | | | | | |--- truncated branch of depth 11  
| | | | | | | | |--- Income > 0.10  
| | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | |--- Income > 0.10  
| | | | | | | | |--- Income <= 0.10  
| | | | | | | | |--- Income <= 0.10  
| | | | | | | | |--- Income <= 0.10  
| | | | | | | | |--- Income <= 0.10  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.10  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.10  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.10  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.10  
| | | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.10
```

```
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- class: 2.0  
  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- class: 1.0  
  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- class: 2.0  
  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.40  
| | | | | | | | | |--- Income <= 0.39  
| | | | | | | | | |--- Income <= 0.18  
| | | | | | | | | |--- Income <= 0.16  
| | | | | | | | | |--- Income <= 0.16  
| | | | | | | | | |--- Income <= 0.16  
| | | | | | | | | |--- truncated branch of depth 34  
| | | | | | | | | |--- Income > 0.16  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.16  
| | | | | | | | | |--- Income <= 0.16  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.16  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | |--- Income > 0.16  
| | | | | | | | | |--- Income <= 0.17  
| | | | | | | | | |--- Income <= 0.17  
| | | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | | |--- Income > 0.17
```

```
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | |--- Income > 0.17  
| | | | | | | | |--- Income <= 0.17  
| | | | | | | | |--- truncated branch of depth 10  
| | | | | | | | |--- Income > 0.17  
| | | | | | | | |--- truncated branch of depth 10  
| | | | | | | |--- Income > 0.18  
| | | | | | | |--- Income <= 0.24  
| | | | | | | |--- Income <= 0.23  
| | | | | | | |--- Income <= 0.23  
| | | | | | | | |--- truncated branch of depth 20  
| | | | | | | |--- Income > 0.23  
| | | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.23  
| | | | | | | |--- Income <= 0.23  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.23  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.24  
| | | | | | | |--- Income<= 0.24  
| | | | | | | |--- Income <= 0.24  
| | | | | | | | |--- truncated branch of depth 5  
| | | | | | | |--- Income > 0.24  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.24  
| | | | | | | |--- Income <= 0.28  
| | | | | | | | |--- truncated branch of depth 8  
| | | | | | | |--- Income > 0.28  
| | | | | | | | |--- truncated branch of depth 17  
| | | | | | | |--- Income > 0.39  
| | | | | | | |--- class: 0.0  
| | | | | | | --- Income > 0.40  
| | | | | | | |--- Income <= 0.50  
| | | | | | | |--- Income <= 0.48  
| | | | | | | |--- Income <= 0.48  
| | | | | | | |--- Income <= 0.43  
| | | | | | | |--- Income <= 0.43  
| | | | | | | | |--- truncated branch of depth 6  
| | | | | | | |--- Income > 0.43  
| | | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.43  
| | | | | | | |--- Income <= 0.44  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | |--- Income > 0.44  
| | | | | | | | |--- truncated branch of depth 6  
| | | | | | | |--- Income > 0.48  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.48
```

```
| | | | | | | | | | | Income <= 0.49  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.49  
| | | | | | | | | | | |--- Income <= 0.49  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.49  
| | | | | | | | | | | |--- Income <= 0.50  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.50  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.50  
| | | | | | | | | | |--- Income <= 0.50  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.50  
| | | | | | | | | | |--- Income <= 0.50  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.50  
| | | | | | | | | | |--- Income <= 0.51  
| | | | | | | | | | |--- Income <= 0.50  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.50  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.51  
| | | | | | | | | | |--- Income <= 0.53  
| | | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | | | |--- Income > 0.53  
| | | | | | | | | | |--- truncated branch of depth 19  
| | | | | | | | | | |--- NumWebVisitsMonth > 6.50  
| | | | | | | | | | |--- Income <= 0.69  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- class: 0.0
```

```
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00
```

```
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 2  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 4  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 8  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 6  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 4  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 9  
| | | | | | |--- Income > 0.00  
| | | | | | |--- class: 2.0  
| | | | |--- Income > 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- class: 1.0  
| | | | |--- Income > 0.00  
| | | | |--- class: 2.0  
| | | | |--- Income > 0.00  
| | | | |--- class: 1.0  
| | | | |--- Income > 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- class: 0.0  
| | | | |--- Income > 0.00  
| | | | |--- class: 1.0  
| | | | |--- Income > 0.00  
| | | | |--- Income <= 0.08  
| | | | |--- Income <= 0.07  
| | | | |--- Income <= 0.07  
| | | | |--- Income <= 0.07
```

[illegible]


```
| | | | | | | | | |--- Income > 0.09  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.09  
| | | | | | | | | |--- Income <= 0.09  
| | | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | | |--- Income > 0.09  
| | | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | | |--- Income > 0.09  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- truncated branch of depth 16  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- truncated branch of depth 48  
| | | |--- Income > 0.69  
| | | |--- Income <= 0.86  
| | | |--- Income <= 0.85  
| | | |--- Income <= 0.84  
| | | |--- Income <= 0.70  
| | | |--- Income <= 0.69  
| | | |--- class: 1.0  
| | | |--- Income > 0.69  
| | | |--- Income <= 0.70  
| | | |--- class: 2.0  
| | | |--- Income > 0.70  
| | | |--- class: 1.0  
| | | |--- Income > 0.70  
| | | |--- Income <= 0.83  
| | | |--- Income <= 0.81  
| | | |--- Income <= 0.79  
| | | |--- truncated branch of depth 9  
| | | |--- Income > 0.79  
| | | |--- class: 2.0  
| | | |--- Income > 0.81  
| | | |--- Income <= 0.81  
| | | |--- class: 1.0  
| | | |--- Income > 0.81  
| | | |--- truncated branch of depth 6  
| | | |--- Income > 0.83  
| | | |--- class: 2.0  
| | | |--- Income > 0.84  
| | | |--- Income <= 0.85  
| | | |--- class: 1.0  
| | | |--- Income > 0.85
```

```
| | | | | | | |--- Income <= 0.85  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.85  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.85  
| | | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.86  
| | | | | | | |--- Income <= 0.87  
| | | | | | | |--- Income <= 0.86  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.86  
| | | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.87  
| | | | | | | |--- Income <= 0.89  
| | | | | | | |--- Income <= 0.87  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.87  
| | | | | | | |--- Income <= 0.87  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.87  
| | | | | | | |--- Income <= 0.88  
| | | | | | | |--- Income <= 0.87  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.87  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.88  
| | | | | | | |--- class: 1.0  
| | | | | | |--- Income > 0.89  
| | | | | | | |--- Income <= 0.95  
| | | | | | | |--- Income <= 0.95  
| | | | | | | |--- Income <= 0.94  
| | | | | | | |--- Income <= 0.93  
| | | | | | | |--- truncated branch of depth 6  
| | | | | | | |--- Income > 0.93  
| | | | | | | |--- truncated branch of depth 7  
| | | | | | | |--- Income > 0.94  
| | | | | | | |--- Income <= 0.94  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.94  
| | | | | | | |--- truncated branch of depth 5  
| | | | | | | |--- Income > 0.95  
| | | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.95  
| | | | | | | |--- Income <= 0.97  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.97  
| | | | | | | |--- Income <= 0.98  
| | | | | | | |--- Income <= 0.97
```

[illegible]

```
| | | | | | | | | Income <= 0.03  
| | | | | | | | | |--- Income <= 0.03  
| | | | | | | | | | |--- Income <= 0.03  
| | | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | | |--- Income > 0.03  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.03  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.03  
| | | | | | | | | |--- Income <= 0.03  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.03  
| | | | | | | | | | |--- Income <= 0.03  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.03  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.03  
| | | | | | | | | |--- Income <= 0.04  
| | | | | | | | | |--- Income <= 0.04  
| | | | | | | | | | |--- Income <= 0.04  
| | | | | | | | | | | |--- truncated branch of depth 18  
| | | | | | | | | | | |--- Income > 0.04  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.04  
| | | | | | | | | | |--- Income <= 0.04  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.04  
| | | | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | | |--- Income > 0.04  
| | | | | | | | | | |--- Income <= 0.04  
| | | | | | | | | | | |--- Income <= 0.04  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.04  
| | | | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | | |--- Income > 0.04  
| | | | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- truncated branch of depth 13  
| | | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | |--- NumWebVisitsMonth <= 10.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | |--- truncated branch of depth 30  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | |--- truncated branch of depth 21
```

```
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 40  
| | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 8  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 37  
| | | | | | | |--- NumWebVisitsMonth > 10.50  
| | | | | | | |--- NumWebVisitsMonth <= 11.50  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 8  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 7  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 11  
| | | | | | | |--- NumWebVisitsMonth > 11.50  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- NumWebVisitsMonth <= 12.50  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- NumWebVisitsMonth > 12.50  
| | | | | | | |--- truncated branch of depth 5  
| | | | | |--- Income > 0.05  
| | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | |--- Income <= 0.05  
| | | | | |--- class: 1.0  
| | | | | |--- Income > 0.05  
| | | | | |--- Income <= 0.05  
| | | | | |--- Income <= 0.05  
| | | | | |--- truncated branch of depth 4  
| | | | | |--- Income > 0.05  
| | | | | |--- class: 0.0  
| | | | | |--- Income > 0.05  
| | | | | |--- Income <= 0.05
```

```
| | | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | | |--- class: 2.0  
| | | | | | | | |---- NumWebVisitsMonth > 9.50  
| | | | | | | | |---- NumWebVisitsMonth <= 10.50  
| | | | | | | | |---- Income <= 0.05  
| | | | | | | | |---- class: 0.0  
| | | | | | | | |---- Income > 0.05  
| | | | | | | | |---- class: 1.0  
| | | | | | | | |---- NumWebVisitsMonth > 10.50  
| | | | | | | | |---- class: 1.0  
| | | | |---- NumWebVisitsMonth > 15.50  
| | | | |---- NumWebVisitsMonth <= 16.50  
| | | | |---- Income <= 0.02  
| | | | |---- class: 1.0  
| | | | |---- Income > 0.02  
| | | | |---- class: 0.0  
| | | | |---- NumWebVisitsMonth > 16.50  
| | | | |---- Income <= 0.04  
| | | | |---- Income <= 0.02  
| | | | |---- class: 0.0  
| | | | |---- Income > 0.02  
| | | | |---- class: 1.0  
| | | | |---- Income > 0.04  
| | | | |---- class: 0.0  
| | | |---- Income > 0.05  
| | | |---- class: 0.0  
| | |---- NumWebVisitsMonth > 17.50  
| | |---- NumWebVisitsMonth <= 19.50  
| | |---- Income <= 0.02  
| | |---- class: 0.0  
| | |---- Income > 0.02  
| | |---- NumWebVisitsMonth <= 18.50  
| | |---- class: 2.0  
| | |---- NumWebVisitsMonth > 18.50  
| | |---- Income <= 0.03  
| | |---- class: 2.0  
| | |---- Income > 0.03  
| | |---- class: 0.0  
| | |---- NumWebVisitsMonth > 19.50  
| | |---- NumWebVisitsMonth <= 21.50  
| | |---- NumWebVisitsMonth <= 20.50  
| | |---- Income <= 0.04  
| | |---- class: 2.0  
| | |---- Income > 0.04  
| | |---- class: 1.0  
| | |---- NumWebVisitsMonth > 20.50  
| | |---- Income <= 0.02
```

```
| | | | | | |--- class: 1.0  
| | | | | | |--- Income > 0.02  
| | | | | | |--- Income <= 0.03  
| | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.03  
| | | | | | |--- class: 0.0  
| | | | | |--- NumWebVisitsMonth > 21.50  
| | | | | |--- Income <= 0.04  
| | | | | |--- class: 2.0  
| | | | | |--- Income > 0.04  
| | | | | |--- class: 0.0  
|--- Income > 0.05  
| |--- NumWebVisitsMonth <= 20.50  
| | |--- NumWebVisitsMonth <= 9.50  
| | | |--- NumWebVisitsMonth <= 8.50  
| | | |--- Income <= 0.97  
| | | |--- Income <= 0.90  
| | | |--- Income <= 0.89  
| | | |--- Income <= 0.71  
| | | |--- Income <= 0.66  
| | | |--- Income <= 0.46  
| | | | |--- truncated branch of depth 43  
| | | | |--- Income > 0.46  
| | | | |--- truncated branch of depth 17  
| | | | |--- Income > 0.66  
| | | | |--- Income <= 0.68  
| | | | |--- truncated branch of depth 3  
| | | | |--- Income > 0.68  
| | | | |--- truncated branch of depth 9  
| | | | |--- Income > 0.71  
| | | | |--- Income <= 0.76  
| | | | |--- Income <= 0.75  
| | | | |--- truncated branch of depth 5  
| | | | |--- Income > 0.75  
| | | | |--- class: 1.0  
| | | | |--- Income > 0.76  
| | | | |--- Income <= 0.88  
| | | | |--- truncated branch of depth 14  
| | | | |--- Income > 0.88  
| | | | |--- truncated branch of depth 2  
| | | | |--- Income > 0.89  
| | | | |--- class: 0.0  
| | | |--- Income > 0.90  
| | | |--- Income <= 0.96  
| | | |--- Income <= 0.96  
| | | |--- Income <= 0.95  
| | | |--- Income <= 0.95  
| | | |--- truncated branch of depth 5
```

[illegible]


```
| | | | | | | | | | | Income <= 0.05  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- class: 2.0  
  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- Income <= 0.05  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.05  
| | | | | | | | | | | |--- Income <= 0.83  
| | | | | | | | | | | |--- Income <= 0.07  
| | | | | | | | | | | |--- Income <= 0.06  
| | | | | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | | | | |--- Income > 0.06  
| | | | | | | | | | | |--- truncated branch of depth 15  
| | | | | | | | | | | |--- Income > 0.07  
| | | | | | | | | | | |--- Income <= 0.08  
| | | | | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | | | | |--- Income > 0.08  
| | | | | | | | | | | |--- truncated branch of depth 30  
| | | | | | | | | | | |--- Income > 0.83  
| | | | | | | | | | | |--- Income <= 0.94  
| | | | | | | | | | | |--- Income <= 0.92  
| | | | | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | | | | |--- Income > 0.92  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.94  
| | | | | | | | | | | |--- Income <= 0.97  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.97  
| | | | | | | | | | | |--- truncated branch of depth 3  
  
| | | | | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | | | |--- Income <= 0.86  
| | | | | | | | | | | |--- Income <= 0.84  
| | | | | | | | | | | |--- NumWebVisitsMonth <= 19.50  
| | | | | | | | | | | |--- NumWebVisitsMonth <= 15.50  
| | | | | | | | | | | |--- NumWebVisitsMonth <= 13.50
```

[illegible]

```
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- Income <= 0.14  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.16  
| | | | | | | | |--- Income <= 0.18  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.18  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.84  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.86  
| | | | | | | | |--- NumWebVisitsMonth <= 10.50  
| | | | | | | | |--- Income <= 0.93  
| | | | | | | | |--- Income <= 0.86  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.86  
| | | | | | | | |--- Income <= 0.88  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.88  
| | | | | | | | |--- Income <= 0.89  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.89  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.93  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- Income <= 0.99  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.99  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- NumWebVisitsMonth > 10.50  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- NumWebVisitsMonth > 20.50  
| | | | | | | | |--- Income <= 0.22  
| | | | | | | | |--- Income <= 0.16  
| | | | | | | | |--- NumWebVisitsMonth <= 21.50  
| | | | | | | | |--- Income <= 0.15  
| | | | | | | | |--- Income <= 0.13  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- Income <= 0.14  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.15
```

```

| | | | | | | | |--- class: 2.0
| | | | | | | |--- NumWebVisitsMonth > 21.50
| | | | | | | |--- Income <= 0.11
| | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.11
| | | | | | | |--- class: 2.0
| | | | | | | |--- Income > 0.16
| | | | | | | |--- Income <= 0.19
| | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.19
| | | | | | | |--- Income <= 0.20
| | | | | | | |--- class: 1.0
| | | | | | | |--- Income > 0.20
| | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.22
| | | | | | | |--- class: 1.0

```

Tree Frequency: 1

```

|--- NumWebVisitsMonth <= 20.50
| |--- NumWebVisitsMonth <= 6.50
| | |--- Income <= 0.00
| | | |--- Income <= 0.00
| | | |--- Income <= 0.00
| | | |--- NumWebVisitsMonth <= 5.50
| | | | |--- NumWebVisitsMonth <= 3.50
| | | | | |--- Income <= 0.00
| | | | | | |--- NumWebVisitsMonth <= 0.50
| | | | | | |--- class: 1.0
| | | | | | |--- NumWebVisitsMonth > 0.50
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- NumWebVisitsMonth > 3.50
| | | | | | |--- Income <= 0.00
| | | | | | |--- class: 1.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 1.0
| | | | | | |--- NumWebVisitsMonth > 5.50
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 2.0
| | | | | | |--- Income > 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- NumWebVisitsMonth <= 0.50
| | | | | | |--- class: 1.0

```



```
| | | | | | | | | |--- truncated branch of depth 3
| | | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- Income <= 0.00
| | | | | | | | | |--- truncated branch of depth 3
| | | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.00
| | | | | | | | |--- class: 1.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 2.0
| | | | |--- Income > 0.00
| | | | |--- NumWebVisitsMonth <= 1.50
| | | | |--- Income <= 0.00
| | | | |--- NumWebVisitsMonth <= 0.50
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- truncated branch of depth 3
| | | | |--- Income > 0.00
| | | | |--- truncated branch of depth 5
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- class: 1.0
| | | | |--- Income > 0.00
| | | | |--- truncated branch of depth 2
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- truncated branch of depth 4
| | | | |--- Income > 0.00
| | | | |--- class: 0.0
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- class: 2.0
| | | | |--- Income > 0.00
| | | | |--- class: 2.0
| | | | |--- NumWebVisitsMonth > 0.50
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- class: 1.0
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- truncated branch of depth 2
| | | | |--- Income > 0.00
| | | | |--- class: 0.0
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
```

[illegible]

```
| | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 26  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 57  
| | | | | | | |--- NumWebVisitsMonth > 1.50  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- truncated branch of depth 23  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 6  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- truncated branch of depth 21  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- truncated branch of depth 53  
| | | | | | | |--- Income > 0.10  
| | | | | | | |--- truncated branch of depth 10  
| | | | | | | |--- Income > 0.10  
| | | | | | | |--- Income <= 0.10  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.10  
| | | | | | | |--- truncated branch of depth 20  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- Income <= 0.14  
| | | | | | | |--- Income <= 0.14  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- Income <= 0.12
```



```
| | | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | |   |-- Income > 0.12  
| | | | | | | | |     |-- truncated branch of depth 3  
| | | | | | | |       ---- Income > 0.12  
| | | | | | |         ---- Income <= 0.12  
| | | | | |           --- truncated branch of depth 6  
| | | | |             --- Income > 0.12  
| | | |               --- truncated branch of depth 14  
| |                 ---- Income > 0.14  
|                   ---- Income <= 0.14  
|                     -- Income <= 0.14  
|                       - truncated branch of depth 6  
|                         - Income > 0.14  
|                           - truncated branch of depth 7  
|                             - Income > 0.14  
|                               - Income <= 0.14  
|                                 - truncated branch of depth 3  
|                                   - Income > 0.14  
|                                     - truncated branch of depth 8  
|                                       - Income > 0.14  
|                                         - Income <= 0.15  
|                                           - Income <= 0.14  
|                                             - Income <= 0.14  
|                                               - truncated branch of depth 12  
|                                                 - Income > 0.14  
|                                                   - class: 0.0  
|                                                     - Income > 0.14  
|                                                       - Income <= 0.15  
|                                                         - truncated branch of depth 13  
|                                                           - Income > 0.15  
|                                                             - truncated branch of depth 6  
|                                                               - Income > 0.15  
|                                                                 - Income <= 0.16  
|                                                                   - Income <= 0.16  
|                                                                     - truncated branch of depth 6  
|                                                                       - Income > 0.16  
|                                                                         - truncated branch of depth 9  
|                                                                           - Income > 0.16  
|                                                                             - Income <= 0.48  
|                                                                               - truncated branch of depth 31  
|                                                                                 - Income > 0.48  
|                                                                                   - truncated branch of depth 23  
  
|-- NumWebVisitsMonth > 2.50  
| |---- Income <= 0.99  
| |    ---- Income <= 0.97  
| |        ---- NumWebVisitsMonth <= 5.50  
| |            ---- NumWebVisitsMonth <= 3.50  
| |                ---- Income <= 0.18
```

[illegible]

```
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- truncated branch of depth 41  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- truncated branch of depth 35  
| | | | | | | | | |--- Income > 0.07  
| | | | | | | | | |--- Income <= 0.07  
| | | | | | | | | |--- truncated branch of depth 8  
| | | | | | | | | |--- Income > 0.07  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.07  
| | | | | | | | | |--- Income <= 0.07  
| | | | | | | | | |--- Income <= 0.07  
| | | | | | | | | |--- truncated branch of depth 14  
| | | | | | | | | |--- Income > 0.07  
| | | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | | |--- Income > 0.07  
| | | | | | | | | |--- Income <= 0.08  
| | | | | | | | | |--- truncated branch of depth 16  
| | | | | | | | | |--- Income > 0.08  
| | | | | | | | | |--- truncated branch of depth 20  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.10  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | |--- Income > 0.10  
| | | | | | | | | |--- Income <= 0.17  
| | | | | | | | | |--- Income <= 0.16  
| | | | | | | | | |--- truncated branch of depth 40  
| | | | | | | | | |--- Income > 0.16  
| | | | | | | | | |--- truncated branch of depth 13  
| | | | | | | | | |--- Income > 0.17  
| | | | | | | | | |--- Income <= 0.17  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.17  
| | | | | | | | | |--- truncated branch of depth 35  
| | | | | | | | | |--- Income > 0.97  
| | | | | | | | | |--- Income <= 0.99  
| | | | | | | | | |--- Income <= 0.98  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.98
```



```

| | |--- NumWebVisitsMonth <= 8.50
| | | |--- Income <= 0.78
| | | | |--- NumWebVisitsMonth <= 7.50
| | | | |--- Income <= 0.00
| | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 2.0
| | | | | | |--- Income > 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- class: 2.0
| | | | | | |--- Income > 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- class: 1.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 0.0
| | | | | | |--- Income > 0.00
| | | | | | |--- Income <= 0.00
| | | | | | |--- class: 2.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 1.0
| | | | | | |--- Income > 0.00
| | | | | | |--- class: 2.0
| | | | | | |--- Income > 0.00
| | | | | | |--- Income <= 0.14
| | | | | | |--- Income <= 0.13
| | | | | | |--- Income <= 0.13
| | | | | | |--- Income <= 0.12
| | | | | | |--- truncated branch of depth 58
| | | | | | |--- Income > 0.12
| | | | | | |--- truncated branch of depth 13
| | | | | | |--- Income > 0.13
| | | | | | |--- Income <= 0.13

```

```
| | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- Income <= 0.14  
| | | | | | | | |--- Income <= 0.13  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- truncated branch of depth 14  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- Income <= 0.14  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- truncated branch of depth 8  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- Income <= 0.15  
| | | | | | | | |--- Income <= 0.14  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- Income <= 0.14  
| | | | | | | | |--- truncated branch of depth 17  
| | | | | | | | |--- Income > 0.14  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.15  
| | | | | | | | |--- Income <= 0.15  
| | | | | | | | |--- Income <= 0.15  
| | | | | | | | |--- truncated branch of depth 10  
| | | | | | | | |--- Income > 0.15  
| | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | |--- Income > 0.15  
| | | | | | | | |--- Income <= 0.15  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.15  
| | | | | | | | |--- truncated branch of depth 39  
| | | | |--- NumWebVisitsMonth > 7.50  
| | | | |--- Income <= 0.70  
| | | | |--- Income <= 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- class: 1.0  
| | | | |--- Income > 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- class: 0.0  
| | | | |--- Income > 0.00  
| | | | |--- class: 1.0  
| | | | |--- Income > 0.00  
| | | | |--- Income <= 0.00  
| | | | |--- class: 2.0
```

[illegible]

```
| | | | | | | | |--- Income <= 0.71  
| | | | | | | | |--- Income <= 0.71  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.71  
| | | | | | | | |--- Income <= 0.71  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.71  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 0.71  
| | | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.73  
| | | | | | | |--- Income <= 0.78  
| | | | | | | |--- Income <= 0.73  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.73  
| | | | | | | |--- Income <= 0.78  
| | | | | | | |--- Income <= 0.77  
| | | | | | | |--- truncated branch of depth 4  
| | | | | | | |--- Income > 0.77  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.78  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.78  
| | | | | | | |--- Income <= 0.78  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.78  
| | | | | | | |--- Income <= 0.78  
| | | | | | | |--- Income <= 0.78  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.78  
| | | | | | | |--- truncated branch of depth 2  
| | | | | | | |--- Income > 0.78  
| | | | | | | |--- class: 1.0  
| | | | | |--- Income > 0.78  
| | | | | |--- Income <= 0.80  
| | | | | |--- Income <= 0.79  
| | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | |--- Income <= 0.79  
| | | | | |--- Income <= 0.78  
| | | | | |--- Income <= 0.78  
| | | | | |--- Income <= 0.78  
| | | | | |--- class: 2.0  
| | | | | |--- Income > 0.78  
| | | | | |--- class: 1.0  
| | | | | |--- Income > 0.78  
| | | | | |--- class: 2.0  
| | | | | |--- Income > 0.78  
| | | | | |--- class: 1.0
```



```
| | | | | | | | |--- Income > 0.79  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.79  
| | | | | | | | |--- Income <= 0.79  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.79  
| | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | |--- Income <= 0.80  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- Income <= 0.80  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- Income <= 0.80  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- Income <= 0.80  
| | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | |--- Income <= 0.80  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.80  
| | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | |--- Income <= 0.93  
| | | | | | | | |--- Income <= 0.90  
| | | | | | | | |--- Income <= 0.86  
| | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | |--- Income > 0.86  
| | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | |--- Income > 0.90  
| | | | | | | | |--- Income <= 0.91  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.91  
| | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | |--- Income > 0.93  
| | | | | | | | |--- Income <= 0.97  
| | | | | | | | |--- Income <= 0.93  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.93
```

[illegible]

```
| | | | | | | | | |--- Income > 0.12  
| | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | |--- Income > 0.34  
| | | | | | | | | |--- Income <= 0.35  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.35  
| | | | | | | | | |--- Income <= 0.37  
| | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | |--- Income > 0.37  
| | | | | | | | | |--- class: 0.0  
| | | | | | |--- NumWebVisitsMonth > 13.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.20  
| | | | | | | |--- NumWebVisitsMonth <= 14.50  
| | | | | | | |--- truncated branch of depth 4  
| | | | | | | |--- NumWebVisitsMonth > 14.50  
| | | | | | | |--- truncated branch of depth 8  
| | | | | | |--- Income > 0.20  
| | | | | | |--- class: 0.0  
| | | | | |--- NumWebVisitsMonth > 17.50  
| | | | | | |--- Income <= 0.08  
| | | | | | |--- Income <= 0.01  
| | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.01  
| | | | | | |--- Income <= 0.07  
| | | | | | | |--- NumWebVisitsMonth <= 19.50  
| | | | | | | |--- truncated branch of depth 4  
| | | | | | | |--- NumWebVisitsMonth > 19.50  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | |--- Income > 0.07  
| | | | | | | |--- NumWebVisitsMonth <= 19.50  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- NumWebVisitsMonth > 19.50  
| | | | | | | |--- class: 0.0  
| | | | | |--- Income > 0.08  
| | | | | | |--- NumWebVisitsMonth <= 18.50  
| | | | | | |--- Income <= 0.20  
| | | | | | | |--- Income <= 0.12  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.12  
| | | | | | | |--- truncated branch of depth 2  
| | | | | | |--- Income > 0.20  
| | | | | | |--- class: 0.0  
| | | | | |--- NumWebVisitsMonth > 18.50  
| | | | | | |--- Income <= 0.10  
| | | | | | |--- class: 0.0
```

```
| | | | | | | | | | Income > 0.10  
| | | | | | | | | | --- NumWebVisitsMonth <= 19.50  
| | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | --- NumWebVisitsMonth > 19.50  
| | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | |--- Income > 0.37  
| | | | | | | | | | |--- NumWebVisitsMonth <= 11.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 10.50  
| | | | | | | | | | |--- Income <= 0.63  
| | | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | | |--- Income <= 0.45  
| | | | | | | | | | |--- Income <= 0.43  
| | | | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | | | |--- Income > 0.43  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.45  
| | | | | | | | | | |--- Income <= 0.59  
| | | | | | | | | | |--- truncated branch of depth 11  
| | | | | | | | | | |--- Income > 0.59  
| | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | | |--- Income <= 0.46  
| | | | | | | | | | |--- Income <= 0.39  
| | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | | |--- Income > 0.39  
| | | | | | | | | | |--- truncated branch of depth 8  
| | | | | | | | | | |--- Income > 0.46  
| | | | | | | | | | |--- Income <= 0.55  
| | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | |--- Income > 0.55  
| | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | |--- Income > 0.63  
| | | | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | | | |--- Income <= 0.97  
| | | | | | | | | | |--- Income <= 0.69  
| | | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | | |--- Income > 0.69  
| | | | | | | | | | |--- truncated branch of depth 11  
| | | | | | | | | | |--- Income > 0.97  
| | | | | | | | | | |--- Income <= 1.00  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 1.00  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | | |--- Income <= 0.65  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.65  
| | | | | | | | | | |--- Income <= 0.92
```

```

| | | | | | | | | | | | |--- truncated branch of depth 12
| | | | | | | | | | | | |--- Income > 0.92
| | | | | | | | | | | | |--- truncated branch of depth 5
| | | | | | | | | | | | |--- NumWebVisitsMonth > 10.50
| | | | | | | | | | | | |--- Income <= 0.54
| | | | | | | | | | | | |--- Income <= 0.48
| | | | | | | | | | | | |--- Income <= 0.41
| | | | | | | | | | | | |--- class: 2.0
| | | | | | | | | | | | |--- Income > 0.41
| | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | | | |--- Income > 0.48
| | | | | | | | | | | | |--- class: 2.0
| | | | | | | | | | | | |--- Income > 0.54
| | | | | | | | | | | | |--- Income <= 0.72
| | | | | | | | | | | | |--- Income <= 0.70
| | | | | | | | | | | | |--- Income <= 0.59
| | | | | | | | | | | | |--- class: 0.0
| | | | | | | | | | | | |--- Income > 0.59
| | | | | | | | | | | | |--- truncated branch of depth 3
| | | | | | | | | | | | |--- Income > 0.70
| | | | | | | | | | | | |--- class: 0.0
| | | | | | | | | | | | |--- Income > 0.72
| | | | | | | | | | | | |--- Income <= 0.76
| | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | | | |--- Income > 0.76
| | | | | | | | | | | | |--- Income <= 0.92
| | | | | | | | | | | | |--- class: 2.0
| | | | | | | | | | | | |--- Income > 0.92
| | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | | | |--- NumWebVisitsMonth > 11.50
| | | | | | | | | | | | |--- class: 1.0
|--- NumWebVisitsMonth > 20.50
| |--- Income <= 0.02
| | |--- class: 2.0
| |--- Income > 0.02
| | |--- Income <= 0.19
| | | |--- NumWebVisitsMonth <= 21.50
| | | |--- Income <= 0.07
| | | |--- class: 0.0
| | | |--- Income > 0.07
| | | |--- Income <= 0.12
| | | | |--- Income <= 0.10
| | | | |--- class: 2.0
| | | | |--- Income > 0.10
| | | | |--- class: 1.0
| | | | |--- Income > 0.12
| | | | |--- Income <= 0.15
| | | | |--- class: 0.0

```

```

| | | | | | | |--- Income > 0.15
| | | | | | | |--- Income <= 0.16
| | | | | | | |--- class: 2.0
| | | | | | | |--- Income > 0.16
| | | | | | | |--- class: 0.0
| | | |--- NumWebVisitsMonth > 21.50
| | | |--- Income <= 0.04
| | | |--- class: 2.0
| | | |--- Income > 0.04
| | | |--- class: 0.0
| | |--- Income > 0.19
| | |--- Income <= 0.21
| | |--- class: 1.0
| | |--- Income > 0.21
| | |--- Income <= 0.22
| | |--- class: 0.0
| | |--- Income > 0.22
| | |--- class: 2.0

```

Tree Frequency: 1

```

|--- Income <= 0.95
| |--- NumWebVisitsMonth <= 6.50
| | |--- Income <= 0.22
| | | |--- Income <= 0.22
| | | | |--- NumWebVisitsMonth <= 0.50
| | | | |--- Income <= 0.12
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- class: 1.0
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- Income <= 0.00
| | | | |--- class: 2.0
| | | | |--- Income > 0.00
| | | | |--- truncated branch of depth 4
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- truncated branch of depth 6
| | | | |--- Income > 0.00
| | | | |--- truncated branch of depth 6
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00
| | | | |--- class: 2.0
| | | | |--- Income > 0.00
| | | | |--- Income <= 0.00

```

```
| | | | | | | | | |--- class: 1.0
| | | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- Income <= 0.00
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- truncated branch of depth 4
| | | | | | | |--- Income > 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | |--- class: 1.0
| | | | | | | |--- Income > 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- truncated branch of depth 3
| | | | | | | |--- Income > 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | |--- Income > 0.00
| | | | | | | | |--- class: 2.0
| | | | | | | |--- Income > 0.00
| | | | | | | | |--- Income <= 0.00
| | | | | | | | | |--- truncated branch of depth 10
| | | | | | | | |--- Income > 0.00
| | | | | | | | | |--- truncated branch of depth 30
| | | | | | | |--- Income > 0.12
| | | | | | | | |--- Income <= 0.14
| | | | | | | | |--- Income <= 0.13
| | | | | | | | |--- Income <= 0.13
| | | | | | | | |--- Income <= 0.12
| | | | | | | | |--- Income <= 0.12
| | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 0.12
| | | | | | | | | |--- truncated branch of depth 3
| | | | | | | |--- Income > 0.12
```

```
| | | | | | | | | |--- Income <= 0.13  
| | | | | | | | | |--- truncated branch of depth 13  
| | | | | | | | | |--- Income > 0.13  
| | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- Income <= 0.13  
| | | | | | | | |--- Income <= 0.13  
| | | | | | | | |--- truncated branch of depth 12  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- Income <= 0.13  
| | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | |--- Income > 0.13  
| | | | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.13  
| | | | | | |--- Income <= 0.14  
| | | | | | |--- Income <= 0.13  
| | | | | | |--- Income <= 0.13  
| | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.13  
| | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.14  
| | | | | | |--- Income <= 0.14  
| | | | | | |--- Income <= 0.14  
| | | | | | |--- truncated branch of depth 4  
| | | | | | |--- Income > 0.14  
| | | | | | |--- truncated branch of depth 5  
| | | | | | |--- Income > 0.14  
| | | | | | |--- Income <= 0.14  
| | | | | | |--- truncated branch of depth 3  
| | | | | | |--- Income > 0.14  
| | | | | | |--- truncated branch of depth 5  
| | | | |--- Income > 0.14  
| | | | |--- Income <= 0.22  
| | | | |--- Income <= 0.15  
| | | | |--- Income <= 0.15  
| | | | |--- Income <= 0.15  
| | | | |--- truncated branch of depth 10  
| | | | |--- Income > 0.15  
| | | | |--- truncated branch of depth 4  
| | | | |--- Income > 0.15  
| | | | |--- class: 0.0  
| | | | |--- Income > 0.15  
| | | | |--- Income <= 0.18  
| | | | |--- Income <= 0.18
```


[illegible]

```
| | | | | | | | | |--- truncated branch of depth 13  
| | | | | | | | | |--- Income > 0.21  
| | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | | |--- Income > 0.21  
| | | | | | | | | |--- Income <= 0.22  
| | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | | |--- Income > 0.22  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | --- NumWebVisitsMonth > 1.50  
| | | | | | | |--- Income <= 0.22  
| | | | | | | |--- Income <= 0.22  
| | | | | | | |--- NumWebVisitsMonth <= 3.50  
| | | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | | |--- truncated branch of depth 52  
| | | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | | |--- truncated branch of depth 57  
| | | | | | | |--- NumWebVisitsMonth > 3.50  
| | | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | | |--- truncated branch of depth 56  
| | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | |--- truncated branch of depth 64  
| | | | | | | |--- Income > 0.22  
| | | | | | | |--- NumWebVisitsMonth <= 5.50  
| | | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | |--- truncated branch of depth 2  
| | | | | | | |--- NumWebVisitsMonth > 5.50  
| | | | | | | |--- class: 1.0  
| | | | | | |--- Income > 0.22  
| | | | | | | |--- Income <= 0.22  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.22  
| | | | | | | |--- Income <= 0.22  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.22  
| | | | | | | |--- class: 0.0  
| | | | | | --- Income > 0.22  
| | | | | | |--- NumWebVisitsMonth <= 2.50  
| | | | | | |--- NumWebVisitsMonth <= 1.50  
| | | | | | | |--- class: 2.0  
| | | | | | |--- NumWebVisitsMonth > 1.50  
| | | | | | | |--- Income <= 0.22  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.22  
| | | | | | | |--- Income <= 0.22  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.22
```

[illegible]

```

| | | |--- Income <= 0.22
| | | |   |--- Income <= 0.22
| | | |     |--- NumWebVisitsMonth <= 1.50
| | | |       |--- class: 1.0
| | | |     |--- NumWebVisitsMonth > 1.50
| | | |       |--- NumWebVisitsMonth <= 2.50
| | | |         |--- class: 2.0
| | | |       |--- NumWebVisitsMonth > 2.50
| | | |         |--- Income <= 0.22
| | | |           |--- Income <= 0.22
| | | |             |--- class: 1.0
| | | |             |--- Income > 0.22
| | | |               |--- Income <= 0.22
| | | |                 |--- class: 0.0
| | | |                 |--- Income > 0.22
| | | |                   |--- NumWebVisitsMonth <= 3.50
| | | |                     |--- class: 2.0
| | | |                     |--- NumWebVisitsMonth > 3.50
| | | |                       |--- truncated branch of depth 4
| | | |                       |--- Income > 0.22
| | | |                         |--- NumWebVisitsMonth <= 5.50
| | | |                         |--- Income <= 0.22
| | | |                         |--- Income <= 0.22
| | | |                           |--- class: 1.0
| | | |                           |--- Income > 0.22
| | | |                             |--- class: 2.0
| | | |                             |--- Income > 0.22
| | | |                               |--- NumWebVisitsMonth <= 3.50
| | | |                                 |--- class: 1.0
| | | |                                 |--- NumWebVisitsMonth > 3.50
| | | |                                   |--- truncated branch of depth 3
| | | |                                   |--- NumWebVisitsMonth > 5.50
| | | |                                     |--- class: 2.0
| | | |   |--- Income > 0.22
| | | |     |--- NumWebVisitsMonth <= 3.00
| | | |       |--- class: 1.0
| | | |     |--- NumWebVisitsMonth > 3.00
| | | |       |--- Income <= 0.22
| | | |         |--- class: 1.0
| | | |         |--- Income > 0.22
| | | |           |--- class: 2.0
| | | |   |--- Income > 0.22
| | | |     |--- NumWebVisitsMonth <= 0.50
| | | |       |--- Income <= 0.23
| | | |         |--- Income <= 0.23
| | | |         |--- Income <= 0.22
| | | |           |--- class: 0.0
| | | |           |--- Income > 0.22

```

```
| | | | | | | | | Income <= 0.22  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.22  
| | | | | | | | | |--- Income <= 0.23  
| | | | | | | | | |--- Income <= 0.22  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.22  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | |--- Income > 0.23  
| | | | | | | | | |--- Income <= 0.23  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.23  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.23  
| | | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.23  
| | | | | | | |--- Income <= 0.89  
| | | | | | | |--- Income <= 0.79  
| | | | | | | |--- Income <= 0.23  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.23  
| | | | | | | |--- Income <= 0.76  
| | | | | | | |--- Income <= 0.53  
| | | | | | | |--- truncated branch of depth 13  
| | | | | | | |--- Income > 0.53  
| | | | | | | |--- truncated branch of depth 11  
| | | | | | | |--- Income > 0.76  
| | | | | | | |--- Income <= 0.78  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.78  
| | | | | | | |--- truncated branch of depth 2  
| | | | | | | |--- Income > 0.79  
| | | | | | | |--- Income <= 0.87  
| | | | | | | |--- Income <= 0.79  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.79  
| | | | | | | |--- Income <= 0.82  
| | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.82  
| | | | | | | |--- truncated branch of depth 5  
| | | | | | | |--- Income > 0.87  
| | | | | | | |--- Income <= 0.89  
| | | | | | | |--- Income <= 0.88  
| | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.88  
| | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.89  
| | | | | | | |--- class: 1.0
```

[illegible]

```
| | | | | | | | | | --- Income <= 0.59  
| | | | | | | | | | |--- truncated branch of depth 16  
| | | | | | | | | | |--- Income > 0.59  
| | | | | | | | | | |   |-- truncated branch of depth 7  
| | | | | | | | | | |--- Income > 0.63  
| | | | | | | | | | |   |-- Income <= 0.75  
| | | | | | | | | | |   |-- truncated branch of depth 13  
| | | | | | | | | | |--- Income > 0.75  
| | | | | | | | | | |   |-- truncated branch of depth 13  
| | | | | | | | | | |--- NumWebVisitsMonth > 1.50  
| | | | | | | | | | |   |-- NumWebVisitsMonth <= 2.50  
| | | | | | | | | | |     |-- Income <= 0.28  
| | | | | | | | | | |       |-- Income <= 0.22  
| | | | | | | | | | |         |-- truncated branch of depth 2  
| | | | | | | | | | |           |-- Income > 0.22  
| | | | | | | | | | |             |-- truncated branch of depth 13  
| | | | | | | | | | |               |-- Income > 0.28  
| | | | | | | | | | |                 |-- Income <= 0.29  
| | | | | | | | | | |                   |-- class: 2.0  
| | | | | | | | | | |                     |-- Income > 0.29  
| | | | | | | | | | |                       |-- truncated branch of depth 28  
| | | | | | | | | | |--- NumWebVisitsMonth > 2.50  
| | | | | | | | | | |   |-- Income <= 0.24  
| | | | | | | | | | |     |-- Income <= 0.23  
| | | | | | | | | | |       |-- truncated branch of depth 15  
| | | | | | | | | | |         |-- Income > 0.23  
| | | | | | | | | | |           |-- truncated branch of depth 10  
| | | | | | | | | | |             |-- Income > 0.24  
| | | | | | | | | | |               |-- Income <= 0.26  
| | | | | | | | | | |                 |-- truncated branch of depth 4  
| | | | | | | | | | |                   |-- Income > 0.26  
| | | | | | | | | | |                     |-- truncated branch of depth 36  
| | | | | | | | | | |--- NumWebVisitsMonth > 3.50  
| | | | | | | | | | |   |-- NumWebVisitsMonth <= 4.50  
| | | | | | | | | | |     |-- Income <= 0.89  
| | | | | | | | | | |       |-- Income <= 0.45  
| | | | | | | | | | |         |-- Income <= 0.23  
| | | | | | | | | | |           |-- Income <= 0.23  
| | | | | | | | | | |             |-- truncated branch of depth 12  
| | | | | | | | | | |               |-- Income > 0.23  
| | | | | | | | | | |                 |-- class: 2.0  
| | | | | | | | | | |                   |-- Income > 0.23  
| | | | | | | | | | |                     |-- Income <= 0.39  
| | | | | | | | | | |                       |-- truncated branch of depth 21  
| | | | | | | | | | |                         |-- Income > 0.39  
| | | | | | | | | | |                           |-- truncated branch of depth 10  
| | | | | | | | | | |                             |-- Income > 0.45  
| | | | | | | | | | |                               |-- Income <= 0.48
```

```
| | | | | | | | | | Income <= 0.48  
| | | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | | |--- Income > 0.48  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.48  
| | | | | | | | | | |--- Income <= 0.77  
| | | | | | | | | | |--- truncated branch of depth 18  
| | | | | | | | | | |--- Income > 0.77  
| | | | | | | | | | |--- truncated branch of depth 10  
| | | | | | | | | | |--- Income > 0.89  
| | | | | | | | | | |--- Income <= 0.92  
| | | | | | | | | | |--- Income <= 0.92  
| | | | | | | | | | |--- Income <= 0.90  
| | | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | | | | |--- Income > 0.90  
| | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | |--- Income > 0.92  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.92  
| | | | | | | | | | |--- Income<= 0.95  
| | | | | | | | | | |--- Income <= 0.94  
| | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | | |--- Income > 0.94  
| | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | |--- Income > 0.95  
| | | | | | | | | | |--- Income <= 0.95  
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.95  
| | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 5.50  
| | | | | | | | | | |--- Income <= 0.22  
| | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.22  
| | | | | | | | | | |--- Income <= 0.38  
| | | | | | | | | | |--- Income <= 0.33  
| | | | | | | | | | |--- truncated branch of depth 17  
| | | | | | | | | | |--- Income > 0.33  
| | | | | | | | | | |--- truncated branch of depth 9  
| | | | | | | | | | |--- Income > 0.38  
| | | | | | | | | | |--- Income <= 0.43  
| | | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | | | | |--- Income > 0.43  
| | | | | | | | | | |--- truncated branch of depth 25  
| | | | | | | | | | |--- NumWebVisitsMonth > 5.50  
| | | | | | | | | | |--- Income <= 0.23  
| | | | | | | | | | |--- Income <= 0.23  
| | | | | | | | | | |--- Income <= 0.22
```


[illegible]

```
| | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | |--- Income <= 0.00  
| | | | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | | | |--- Income > 0.00  
| | | | | | | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 4  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- class: 2.0  
| | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- class: 2.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | |--- Income <= 0.00  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | | |--- truncated branch of depth 7  
| | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | |--- NumWebVisitsMonth <= 12.00  
| | | | | | | | |--- truncated branch of depth 4
```

```
| | | | | | | | | |--- NumWebVisitsMonth > 12.00  
| | | | | | | | | |--- class: 0.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- Income <= 0.00  
| | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- class: 1.0  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- truncated branch of depth 6  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | |--- Income <= 0.00  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.00  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | |--- Income <= 0.00  
| | | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 1.0  
| | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | |--- class: 0.0  
| | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | |--- class: 2.0  
| | | | | | |--- Income > 0.00  
| | | | | | | |--- class: 0.0  
| | | | | | |--- Income > 0.00  
| | | | | | |--- NumWebVisitsMonth <= 13.50  
| | | | | | |--- NumWebVisitsMonth <= 11.50  
| | | | | | |--- Income <= 0.00  
| | | | | | |--- NumWebVisitsMonth <= 10.50  
| | | | | | |--- Income <= 0.00
```

[illegible]

```

| | | | |--- Income <= 0.01
| | | | |   |--- Income <= 0.01
| | | | |   |   |--- class: 2.0
| | | | |   |--- Income > 0.01
| | | | |   |   |--- class: 0.0
| | | | |   |--- Income > 0.01
| | | | |   |   |--- class: 1.0
| | | |--- Income > 0.02
| | | |   |--- Income <= 0.02
| | | |   |--- NumWebVisitsMonth <= 10.50
| | | |   |   |--- Income <= 0.02
| | | |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |--- NumWebVisitsMonth <= 9.50
| | | |   |   |   |   |   |--- NumWebVisitsMonth <= 8.50
| | | |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |--- NumWebVisitsMonth <= 7.50
| | | |   |   |   |   |   |   |   |   |--- truncated branch of depth 6
| | | |   |   |   |   |   |   |   |   |--- NumWebVisitsMonth > 7.50
| | | |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 7
| | | |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 6
| | | |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 2
| | | |   |   |   |   |   |   |   |   |   |   |   |   |--- NumWebVisitsMonth > 8.50
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- class: 2.0
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 4
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- class: 0.0
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- NumWebVisitsMonth > 9.50
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- class: 2.0
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- class: 1.0
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income <= 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- class: 0.0
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- truncated branch of depth 2
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- class: 2.0
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- Income > 0.02
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- NumWebVisitsMonth <= 9.50
| | | |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |--- NumWebVisitsMonth <= 7.50

```

```
| | | | | | | | | | | Income <= 0.02
| | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- class: 2.0
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 3
| | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- class: 0.0
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 2
| | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 12
| | | | | | | | | | |--- NumWebVisitsMonth > 7.50
| | | | | | | | | | | |--- NumWebVisitsMonth <= 8.50
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 5
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 4
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 7
| | | | | | | | | | |--- NumWebVisitsMonth > 8.50
| | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- truncated branch of depth 7
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- class: 0.0
| | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- class: 2.0
| | | | | | | | | | | | | |--- Income > 0.02
| | | | | | | | | | | | | |--- class: 1.0
| | | | | | | | | | |--- NumWebVisitsMonth > 9.50
| | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- Income <= 0.02
| | | | | | | | | | | | | |--- class: 1.0
```

```
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- NumWebVisitsMonth > 10.50  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- NumWebVisitsMonth <= 16.50  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- NumWebVisitsMonth > 16.50  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- NumWebVisitsMonth <= 20.50  
| | | | | | | | | |--- NumWebVisitsMonth <= 15.50  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- Income <= 0.02  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | | |--- Income > 0.02  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- NumWebVisitsMonth > 15.50  
| | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- NumWebVisitsMonth > 20.50
```

[illegible]

[illegible]

```
| | | | | | | | | | NumWebVisitsMonth > 11.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 18.50  
| | | | | | | | | | | |--- truncated branch of depth 11  
| | | | | | | | | | |--- NumWebVisitsMonth > 18.50  
| | | | | | | | | | | |--- truncated branch of depth 5  
| | | | | | | | | |--- NumWebVisitsMonth > 20.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 21.50  
| | | | | | | | | | | |--- Income <= 0.13  
| | | | | | | | | | | |--- Income <= 0.07  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | | |--- Income > 0.07  
| | | | | | | | | | | |--- Income <= 0.12  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.12  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | |--- Income > 0.13  
| | | | | | | | | | | |--- Income <= 0.15  
| | | | | | | | | | | |--- class: 1.0  
| | | | | | | | | | | |--- Income > 0.15  
| | | | | | | | | | | |--- Income <= 0.21  
| | | | | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | | | | |--- Income > 0.21  
| | | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | |--- NumWebVisitsMonth > 21.50  
| | | | | | | | | | |--- Income <= 0.08  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | | |--- Income > 0.08  
| | | | | | | | | | | |--- Income <= 0.15  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- Income > 0.15  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.76  
| | | | | | | | | |--- NumWebVisitsMonth <= 10.50  
| | | | | | | | | | |--- Income <= 0.77  
| | | | | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | | | |--- NumWebVisitsMonth > 7.50  
| | | | | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | | | | |--- Income <= 0.77  
| | | | | | | | | | | |--- class: 0.0  
| | | | | | | | | | | |--- Income > 0.77  
| | | | | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | | | | |--- NumWebVisitsMonth > 8.50  
| | | | | | | | | | | |--- class: 2.0  
| | | | | | | | | |--- Income > 0.77  
| | | | | | | | | | |--- NumWebVisitsMonth <= 8.50  
| | | | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | | | |--- Income <= 0.78
```


[illegible]

```
| | | | | | | | |--- Income > 0.99  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.99  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- NumWebVisitsMonth > 3.50  
| | | | | | | | |--- Income <= 0.96  
| | | | | | | | |--- NumWebVisitsMonth <= 9.50  
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |--- NumWebVisitsMonth <= 5.50  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- NumWebVisitsMonth > 5.50  
| | | | | | | | |--- NumWebVisitsMonth <= 7.00  
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- NumWebVisitsMonth > 7.00  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |--- NumWebVisitsMonth <= 6.50  
| | | | | | | | |--- NumWebVisitsMonth <= 5.50  
| | | | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | | | |--- Income <= 0.96  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.96  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | | |--- Income <= 0.96  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.96  
| | | | | | | | |--- Income <= 0.96  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.96  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- NumWebVisitsMonth > 5.50  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- NumWebVisitsMonth > 6.50  
| | | | | | | | |--- Income <= 0.95  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.95  
| | | | | | | | |--- Income <= 0.96  
| | | | | | | | |--- class: 1.0
```

```
| | | | | | | | | |--- Income > 0.96  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- NumWebVisitsMonth > 9.50  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.96  
| | | | | | | | |--- Income <= 0.99  
| | | | | | | | | |--- NumWebVisitsMonth <= 4.50  
| | | | | | | | | |--- Income <= 0.97  
| | | | | | | | | |--- Income <= 0.96  
| | | | | | | | | |--- class: 0.0  
| | | | | | | | | |--- Income > 0.96  
| | | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.97  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- NumWebVisitsMonth > 4.50  
| | | | | | | | |--- NumWebVisitsMonth <= 7.50  
| | | | | | | | |--- NumWebVisitsMonth <= 5.50  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- Income <= 0.97  
| | | | | | | | |--- Income <= 0.97  
| | | | | | | | |--- truncated branch of depth 3  
| | | | | | | | |--- Income > 0.97  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 0.97  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- Income <= 0.98  
| | | | | | | | |--- truncated branch of depth 2  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- class: 1.0  
| | | | | | | | |--- Income > 0.98  
| | | | | | | | |--- class: 2.0  
| | | | | | | | |--- NumWebVisitsMonth > 5.50  
| | | | | | | | |--- Income <= 0.97  
| | | | | | | | |--- Income <= 0.96  
| | | | | | | | |--- class: 0.0  
| | | | | | | | |--- Income > 0.96  
| | | | | | | | |--- class: 1.0
```

```
| | | | | | | | |--- Income > 0.97
| | | | | | | | |--- NumWebVisitsMonth <= 6.50
| | | | | | | | |   |--- Income <= 0.98
| | | | | | | | |     |--- truncated branch of depth 4
| | | | | | | | |     |--- Income > 0.98
| | | | | | | | |       |--- class: 2.0
| | | | | | | | |   |--- NumWebVisitsMonth > 6.50
| | | | | | | | |     |--- Income <= 0.97
| | | | | | | | |       |--- truncated branch of depth 3
| | | | | | | | |       |--- Income > 0.97
| | | | | | | | |         |--- truncated branch of depth 4
| | | | | | | | |--- NumWebVisitsMonth > 7.50
| | | | | | | | |   |--- Income <= 0.97
| | | | | | | | |     |--- NumWebVisitsMonth <= 8.50
| | | | | | | | |       |--- Income <= 0.96
| | | | | | | | |         |--- class: 2.0
| | | | | | | | |           |--- Income > 0.96
| | | | | | | | |             |--- Income <= 0.97
| | | | | | | | |               |--- class: 0.0
| | | | | | | | |                 |--- Income > 0.97
| | | | | | | | |                   |--- truncated branch of depth 3
| | | | | | | | |----- NumWebVisitsMonth > 8.50
| | | | | | | | |       |--- Income <= 0.97
| | | | | | | | |         |--- class: 2.0
| | | | | | | | |           |--- Income > 0.97
| | | | | | | | |             |--- class: 1.0
| | | | | | | | |--- Income > 0.97
| | | | | | | | |   |--- NumWebVisitsMonth <= 9.50
| | | | | | | | |     |--- NumWebVisitsMonth <= 8.50
| | | | | | | | |       |--- Income <= 0.98
| | | | | | | | |         |--- class: 0.0
| | | | | | | | |           |--- Income > 0.98
| | | | | | | | |             |--- truncated branch of depth 2
| | | | | | | | |----- NumWebVisitsMonth > 8.50
| | | | | | | | |       |--- class: 0.0
| | | | | | | | |--- NumWebVisitsMonth > 9.50
| | | | | | | | |   |--- Income <= 0.98
| | | | | | | | |     |--- class: 1.0
| | | | | | | | |       |--- Income > 0.98
| | | | | | | | |         |--- Income <= 0.98
| | | | | | | | |           |--- class: 0.0
| | | | | | | | |             |--- Income > 0.98
| | | | | | | | |               |--- class: 2.0
| | | | | | | | |--- Income > 0.99
| | | | | | | | |   |--- class: 0.0
| | | | | | | | |--- Income > 0.99
| | | | | | | | |   |--- Income <= 1.00
| | | | | | | | |   |--- Income <= 1.00
```

```

| | | | | |--- NumWebVisitsMonth <= 8.50
| | | | | | |--- Income <= 0.99
| | | | | | | |--- NumWebVisitsMonth <= 1.50
| | | | | | | | |--- class: 2.0
| | | | | | | |--- NumWebVisitsMonth > 1.50
| | | | | | | | |--- NumWebVisitsMonth <= 6.50
| | | | | | | | | |--- Income <= 0.99
| | | | | | | | | | |--- class: 1.0
| | | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- NumWebVisitsMonth <= 3.50
| | | | | | | | | | |--- class: 2.0
| | | | | | | | | |--- NumWebVisitsMonth > 3.50
| | | | | | | | | | |--- class: 1.0
| | | | | | | | |--- NumWebVisitsMonth > 6.50
| | | | | | | | |--- NumWebVisitsMonth <= 7.50
| | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- class: 0.0
| | | | | | | | |--- Income > 0.99
| | | | | | | | | |--- class: 1.0
| | | | | | | |--- NumWebVisitsMonth > 7.50
| | | | | | | |--- Income <= 0.99
| | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.99
| | | | | | | |--- class: 1.0
| | | | | | |--- Income > 0.99
| | | | | | | |--- Income <= 0.99
| | | | | | | | |--- class: 0.0
| | | | | | | |--- Income > 0.99
| | | | | | | | |--- NumWebVisitsMonth <= 2.50
| | | | | | | | |--- class: 1.0
| | | | | | | | |--- NumWebVisitsMonth > 2.50
| | | | | | | | |--- Income <= 0.99
| | | | | | | | | |--- class: 1.0
| | | | | | | | |--- Income > 0.99
| | | | | | | | |--- NumWebVisitsMonth <= 4.00
| | | | | | | | | |--- class: 0.0
| | | | | | | | |--- NumWebVisitsMonth > 4.00
| | | | | | | | |--- Income <= 1.00
| | | | | | | | | |--- class: 2.0
| | | | | | | | |--- Income > 1.00
| | | | | | | | | |--- class: 0.0
| | | | | | |--- NumWebVisitsMonth > 8.50
| | | | | | |--- NumWebVisitsMonth <= 9.50
| | | | | | | |--- Income <= 0.99
| | | | | | | | |--- class: 2.0
| | | | | | | |--- Income > 0.99
| | | | | | | | |--- class: 0.0
| | | | | | |--- NumWebVisitsMonth > 9.50

```



```
[99]: # Initialize a dictionary to store tree frequencies
tree_frequency = Counter()

# Convert DataFrame column index to a list of strings
feature_names_list = df1_inputs_train.columns.tolist()

# Loop through the trees and count their frequency
for i in range(len(rf_classifier.estimators_)):
    tree = rf_classifier.estimators_[i]
    tree_str = export_text(tree, feature_names=feature_names_list, max_depth=2)
    tree_frequency[tree_str] += 1

# Get the three most frequent trees
top_trees = tree_frequency.most_common(3)

# Print the top three trees
for tree_str, frequency in top_trees:
    print("Tree Frequency:", frequency)
    print(tree_str)
```

```
Tree Frequency: 1
|--- NumWebVisitsMonth <= 7.50
|   |--- NumWebVisitsMonth <= 5.50
|   |   |--- Income <= 0.95
|   |   |   |--- truncated branch of depth 84
|   |   |   |--- Income > 0.95
|   |   |   |--- truncated branch of depth 14
|   |   |--- NumWebVisitsMonth > 5.50
|   |   |   |--- NumWebVisitsMonth <= 6.50
|   |   |   |--- truncated branch of depth 42
|   |   |   |--- NumWebVisitsMonth > 6.50
|   |   |   |--- truncated branch of depth 71
|   |--- NumWebVisitsMonth > 7.50
|   |   |--- Income <= 0.05
|   |   |   |--- NumWebVisitsMonth <= 17.50
|   |   |   |--- truncated branch of depth 48
|   |   |   |--- NumWebVisitsMonth > 17.50
|   |   |   |--- truncated branch of depth 6
|   |   |--- Income > 0.05
|   |   |   |--- NumWebVisitsMonth <= 20.50
|   |   |   |--- truncated branch of depth 51
|   |   |   |--- NumWebVisitsMonth > 20.50
|   |   |   |--- truncated branch of depth 7
```

```
Tree Frequency: 1
|--- NumWebVisitsMonth <= 20.50
|   |--- NumWebVisitsMonth <= 6.50
```

```

|   |   |--- Income <= 0.00
|   |   |   |--- truncated branch of depth 7
|   |   |--- Income > 0.00
|   |   |   |--- truncated branch of depth 100
|   |--- NumWebVisitsMonth > 6.50
|   |   |--- NumWebVisitsMonth <= 8.50
|   |   |   |--- truncated branch of depth 66
|   |   |--- NumWebVisitsMonth > 8.50
|   |   |   |--- truncated branch of depth 55
|--- NumWebVisitsMonth > 20.50
|   |--- Income <= 0.02
|   |   |--- class: 2.0
|   |--- Income > 0.02
|   |   |--- Income <= 0.19
|   |   |   |--- truncated branch of depth 6
|   |   |--- Income > 0.19
|   |   |   |--- truncated branch of depth 3

```

Tree Frequency: 1

```

|--- Income <= 0.95
|   |--- NumWebVisitsMonth <= 6.50
|   |   |--- Income <= 0.22
|   |   |   |--- truncated branch of depth 72
|   |   |--- Income > 0.22
|   |   |   |--- truncated branch of depth 44
|   |--- NumWebVisitsMonth > 6.50
|   |   |--- Income <= 0.02
|   |   |   |--- truncated branch of depth 47
|   |   |--- Income > 0.02
|   |   |   |--- truncated branch of depth 57
|--- Income > 0.95
|   |--- Income <= 0.99
|   |   |--- NumWebVisitsMonth <= 3.50
|   |   |   |--- truncated branch of depth 10
|   |   |--- NumWebVisitsMonth > 3.50
|   |   |   |--- truncated branch of depth 12
|   |--- Income > 0.99
|   |   |--- Income <= 1.00
|   |   |   |--- truncated branch of depth 10
|   |   |--- Income > 1.00
|   |   |   |--- class: 0.0

```

time: 5.61 s (started: 2024-04-13 22:06:58 +05:30)

```

[100]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

```

```

import seaborn as sns
import plotly.graph_objects as go
from collections import Counter
from scipy import stats
from sklearn.tree import plot_tree, export_graphviz
from IPython.display import display
from collections import Counter
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, \
    ↪confusion_matrix
import time
import psutil

# Function to measure memory usage
def memory_usage():
    process = psutil.Process()
    return process.memory_info().rss / 1024 ** 2 # Memory usage in MB

# Start time
start_time = time.time()

# Initialize Random Forest classifier
rf_classifier = RandomForestClassifier(n_estimators=100, random_state=45052)

# Train the Random Forest classifier
rf_classifier.fit(df1_inputs_train, df1_output_train['Cluster_Label_x'])

# Predictions
y_train_pred_rf = rf_classifier.predict(df1_inputs_train)
y_test_pred_rf = rf_classifier.predict(df1_inputs_test)

# End time
end_time = time.time()

# Time taken
execution_time = end_time - start_time

# Memory usage
memory_used = memory_usage()

# Accuracy
accuracy_train = accuracy_score(df1_output_train['Cluster_Label_x'], \
    ↪y_train_pred_rf)
accuracy_test = accuracy_score(df1_output_test['Cluster_Label_x'], \
    ↪y_test_pred_rf)

# Print time, memory usage, and accuracy

```

```

print("Time taken (seconds):", execution_time)
print("Memory used (MB):", memory_used)
print("Training Set Accuracy:", accuracy_train)
print("Testing Set Accuracy:", accuracy_test)

# Print feature importances
feature_importances = rf_classifier.feature_importances_
feature_importance_df = pd.DataFrame({'Feature': df1_inputs_train.columns,
    ↳ 'Importance': feature_importances})
sorted_feature_importance_df = feature_importance_df.
    ↳ sort_values(by='Importance', ascending=False)
print("\nFeature Importances:")
print(sorted_feature_importance_df)

# Print confusion matrix and classification report for training set
print("\nTraining Set Confusion Matrix:")
print(confusion_matrix(df1_output_train['Cluster_Label_x'], y_train_pred_rf))
print("\nTraining Set Classification Report:")
print(classification_report(df1_output_train['Cluster_Label_x'],
    ↳ y_train_pred_rf))

# Print confusion matrix and classification report for testing set
print("\nTesting Set Confusion Matrix:")
print(confusion_matrix(df1_output_test['Cluster_Label_x'], y_test_pred_rf))
print("\nTesting Set Classification Report:")
print(classification_report(df1_output_test['Cluster_Label_x'], y_test_pred_rf))

```

Time taken (seconds): 42.62147378921509
 Memory used (MB): 954.20703125
 Training Set Accuracy: 0.9537490364382591
 Testing Set Accuracy: 0.34041666666666665

Feature Importances:

	Feature	Importance
0	Income	0.990607
1	NumWebVisitsMonth	0.009393

Training Set Confusion Matrix:

```

[[15323  341  400]
 [  358 15195  370]
 [  381  370 15261]]

```

Training Set Classification Report:

	precision	recall	f1-score	support
0	0.95	0.95	0.95	16064
1	0.96	0.95	0.95	15923

2	0.95	0.95	0.95	16012
accuracy			0.95	47999
macro avg	0.95	0.95	0.95	47999
weighted avg	0.95	0.95	0.95	47999

Testing Set Confusion Matrix:

```
[[1442 1256 1282]
 [1344 1334 1355]
 [1314 1364 1309]]
```

Testing Set Classification Report:

	precision	recall	f1-score	support
0	0.35	0.36	0.36	3980
1	0.34	0.33	0.33	4033
2	0.33	0.33	0.33	3987
accuracy			0.34	12000
macro avg	0.34	0.34	0.34	12000
weighted avg	0.34	0.34	0.34	12000

time: 42.8 s (started: 2024-04-13 22:09:50 +05:30)

[]: