

BAN210 - Final Assessment

Using Predictive Modeling on the Breast Cancer Dataset in SAS Enterprise Miner

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STUDENT NAME: April Tolosa

STUDENT NUMBER: 131785214

STUDENT SIGNATURE: April Tolosa

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Introduction

In this assessment paper, the breast cancer data set will be analyzed using exploratory and modeling techniques using SAS Enterprise Miner. The goal of this paper is to show the steps and the powerful insights to be gained from the dataset using simple statistical analysis. Steps will be explained as well as the results to show ability to use the tool and to demonstrate data analysis skills.

About the Data

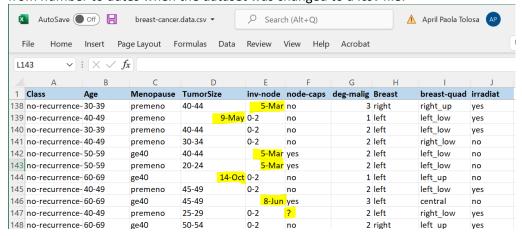
The Breast Cancer data set was obtained from the University Medical Centre, Institute of Oncology, Ljubljana, Yugoslavia. It includes 201 instances of one class and 85 instances of another class. The instances are described by 10 attributes, some of which are linear, and some are nominal.

Data set Characteristics:	Multivariate	Number of Instances:	286	Area:	Life
Attribute Characteristics:	Categorical	Number of Attributes:	10	Date Donated	1988-07-11
Associated Tasks:	Classification	Missing Values?	Yes	Number of Web Hits:	622435

Attribute Information:

- 1. Class: no-recurrence-events, recurrence-events
- 2. age: 10-19, 20-29, 30-39, 40-49, 50-59, 60-69, 70-79, 80-89, 90-99.
- 3. menopause: lt40, ge40, premeno.
- 4. tumor-size: 0-4, 5-9, 10-14, 15-19, 20-24, 25-29, 30-34, 35-39, 40-44, 45-49, 50-54, 55-59.
- 5. inv-nodes: 0-2, 3-5, 6-8, 9-11, 12-14, 15-17, 18-20, 21-23, 24-26, 27-29, 30-32, 33-35, 36-39.
- 6. node-caps: yes, no.
- 7. deg-malig: 1, 2, 3.
- 8. breast: left, right.
- 9. breast-quad: left-up, left-low, right-up, right-low, central.
- 10. irradiat: yes, no.

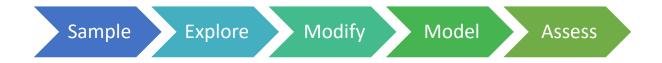
Before analysis, labels were added on the top row and the data cleaned as there were conversion errors from number to dates when the dataset was changed to a .csv file:



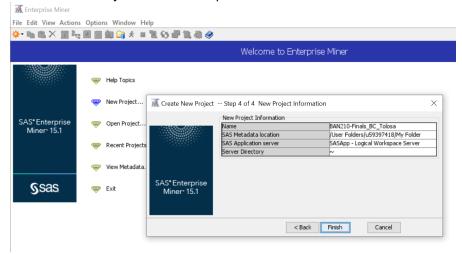


Analysis

This section will demonstrate how the dataset is analyzed using SAS Miner guided by the data mining process learned from this course

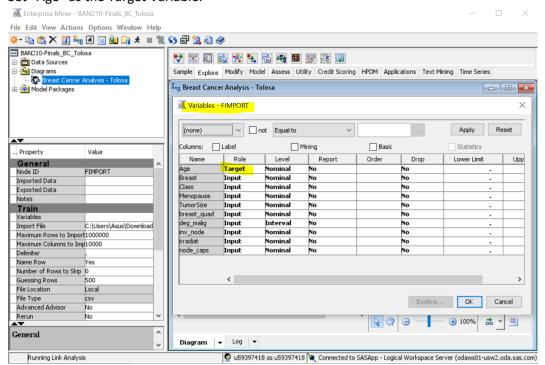


Create a New Project on SAS Enterprise Miner



Import the file using File Import node.

Right click on the File Import node and click Variables to open the Variables window. Set "Age" as the Target Variable.





In this window we can click Explore to know more about the data of each variable.

For example, for the Age variable, below is a screenshot of the data: SEXPLORE - EMWS1.FIMPORT_DATA File View Actions Window *∰* ■ 🖰 🇳 Sample Properties Sample Statistics
 Obs #
 Variable ...
 Label
 Type
 Percent ...
 Number o...
 Mode Per...
 Mode

 1 Age
 CLASS
 06
 33.5664350-59
 Property Value Rows
Columns
Library
Member
Type
Sample Method
Fetch Size
Fetched Rows
Random Seed 286 10 EMWS1 FIMPORT_DATA
DATA Apply Plot... EMWS1.FIMPORT_DATA **Ⅲ** Age Obs # Age 130-39 240-49 340-49 460-69 540-49 660-69 750-59 860-69 940-49 Frequ 1040-49 1140-49 1250-59 1360-69 1450-59 1540-49 60-69 20-29

The data tells us that there are 6 classifications for age, with the age range of 50-59 being the highest frequency at 33.56%.

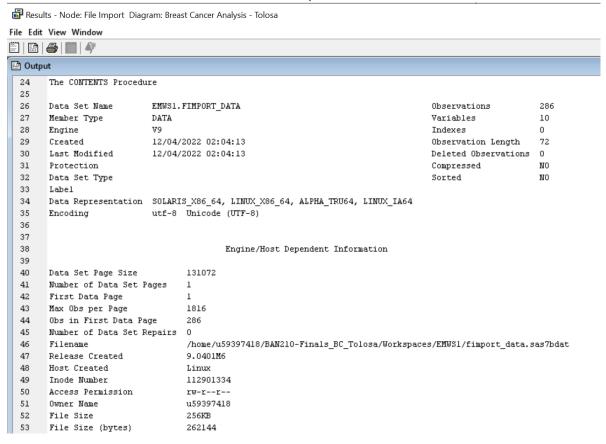
Exploring all variables, below are the results derived:





The File Import node is ran to check about the data. Then right click to check the Results.

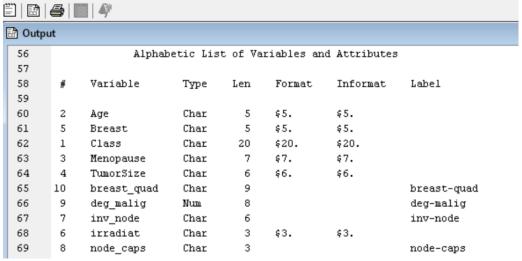
We can validate we have 286 observations, and 10 variables:



Information about the variables is shown like the type, length, and format:

Results - Node: File Import Diagram: Breast Cancer Analysis - Tolosa

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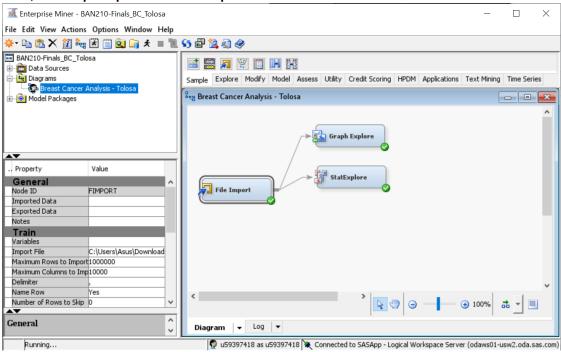




Exploratory Data Analysis

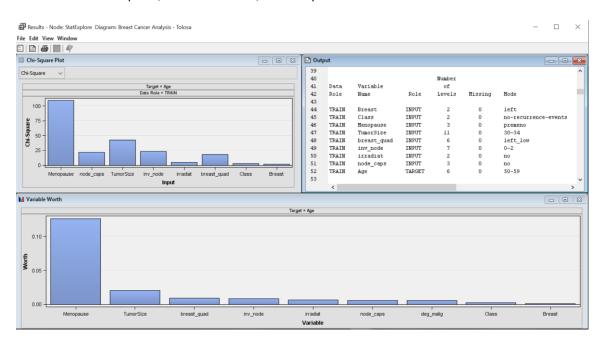
Next is the Exploratory Data Analysis (EDA) to discover patterns, spot anomalies, test hypothesis and to check assumptions with the help of summary statistics and graphical representations.

Here, the Graph Explorer and StatExplorer nodes are used.



StatExplorer Results

The StatExplorer is a versatile tool that gives the variable distributions and statistical information of the dataset. The Chi-Square, Variable Worth, and Output is shown here.

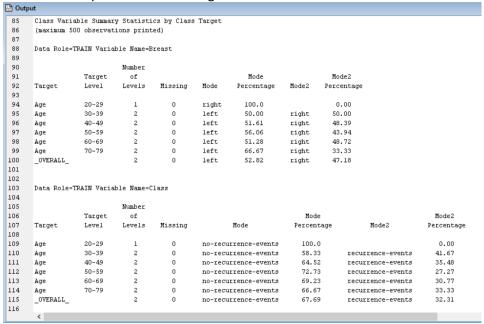




Concentrating on the Output, more insight is gained on the statistics of how the target Age is compared with other variables.

An example below is the statistics for Age vs. Breast and Age vs. Class.

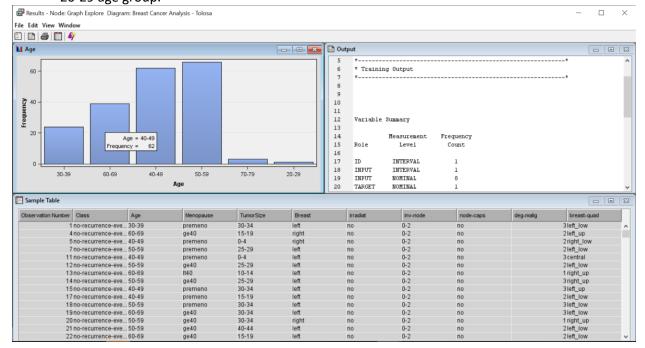
We can say that for Age 40-49, 51.61% developed tumor on the left breast and 48.39% developed tumor on the right breast.



Graph Explorer Results

The graph explorer shows more statistics about the target variable Age.

This confirms that the most frequent data is that from the 50-59 age group and the least for the 20-29 age group.



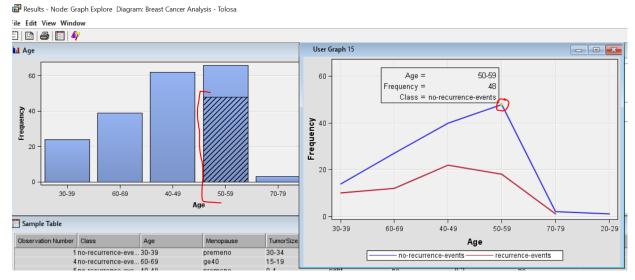


Graphs can also be made by going to View and the Plot. As an example, below is a Line graph with the Age set to Category role and the Class as the Group role

❖ The line graph shows us that for all Age range, the tumor being no-recurring (blue line) is greater than recurring tumors.



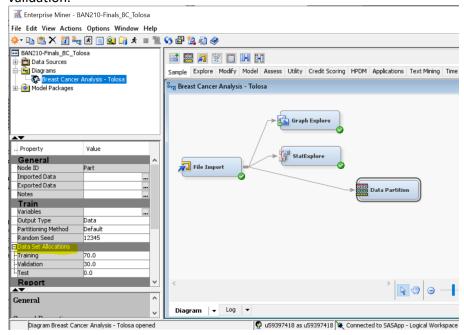
❖ There are 48 observations of no-reccurence at the age 50-59 which is the highest frequency. As this is pointed, the frequency graph on the left also shades the corresponding bar that represents this figure.





Predictive Models

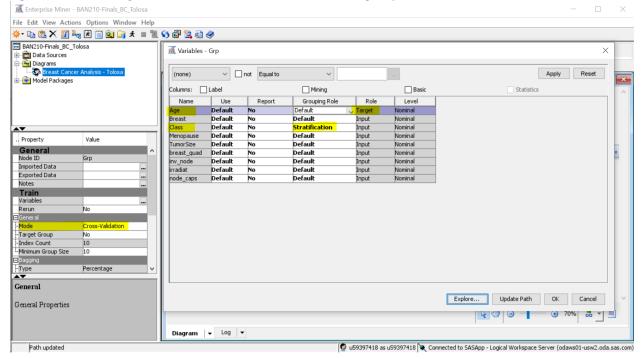
Data Partition node was added, and the Data Set Allocation was set to 70 for training and 30 for validation.



K-Fold Validation

Start Group node is added as a way to do the **k-fold validation**, to resample the dataset to ensure generalizability of the predictive models. It is on **Cross Validation** mode which exports the complement of the groups specified, as opposed to the groups themselves.

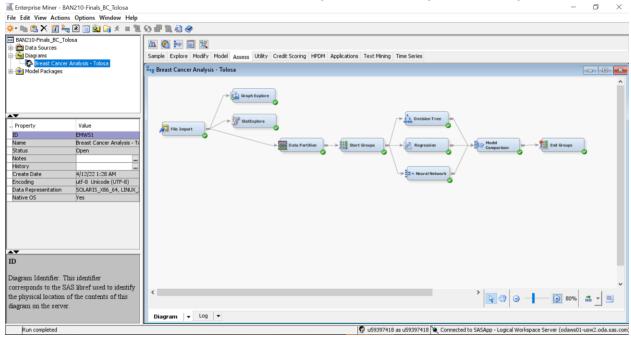
Age is set as the target role and Class is set to stratification group role.





Next is the completion of the model by connecting the 3 predictive models used namely, the **Decision Tree, Regression, and Neural Network** model nodes.

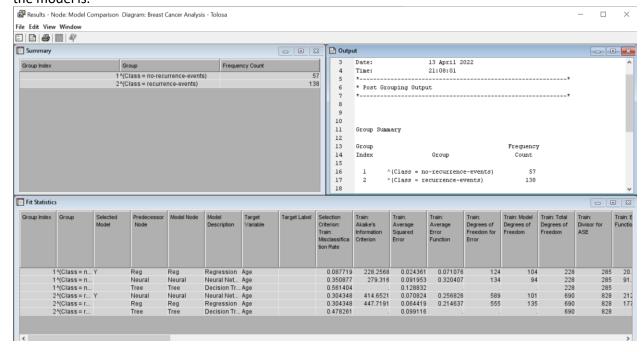
These three nodes are connected to the **Model Comparison and finally the End Groups** node.



Run the End Group Node.

Model Comparison

First, the **Model Comparison Results** is examined— focusing on the Misclassification Rate to see how fit the model is.

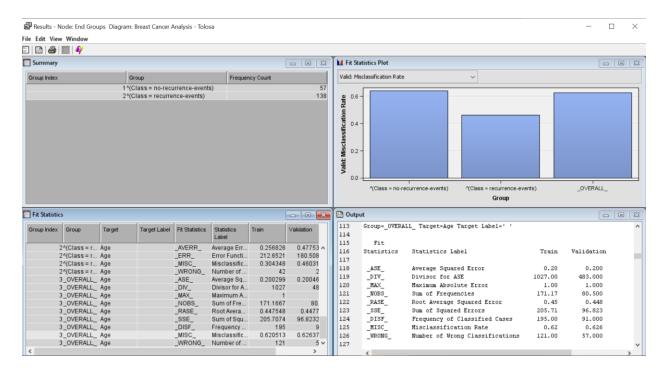




Based on the Fit Statistics, the Regression is the selected model for the no-recurring group and Neural Network is the selected model for the recurring group, both with the Age as target variable.

These were picked to be the best fit because they had the lowest misclassification rate and the lower, the better because it means that the forecast is closer to the actual.

The Results of the three models is shown when clicking on the End Group Results. It shows statistics of all three models and their overall results.



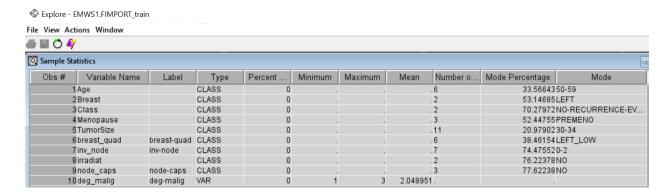
The Fit Statistics plot can be navigated to show the bar graph for each fit statistic label.

Conclusion

In this report, a simple exploratory data analysis was run wherein Age vs. Breast was first analyzed and a sample conclusion was mentioned as seen from the result, that is, for Age 40-49, 51.61% developed tumor on the left breast and 48.39% developed tumor on the right breast. This is a statistical analysis which can be presented to the business about the likelihood of the location of the tumor for those age group. With insights like this, looking at the bigger picture, medical diagnosis can be shifted when the patient's age is known as well as testing and prescription that might result to savings in time and money.



Another sample statistic that can be looked at is the mode of each variable, to see their likelihood:



Analyzing the chart above, it can be said to the business that the likelihood of developing breast cancer is at the age of 50-59 (33.5%) based on the data set. The tumor is also 53% likely to be on the left breast. It is, however 70% not recurring, and 52% in the premeno stage. There is a 20.9% change that the tumor size is 30-34, 38.5% chance it is in the left lower breast quadrant, 74% chance it is in the 0-2 inv_node, and 76% not irradiat and 77% likelihood that it has no node_caps.

When it comes to predictive analysis, three models were ran, namely, the Decision Tree, Logistic Regression, and Neural Network. Age was set as Target variable and Class as the Stratification in the grouping role.

After running the model comparison and checking the Fit Statistics, the **Regression** is the selected model for the no-recurring group and **Neural Network** is the selected model for the recurring group. Knowing which model to use is important because the goal is to minimize the error or difference in the prediction to the actual, in this case, the class.

Work Cited

Zwitter, Matjaz and & Soklic, Milan. "Breast Cancer Data Set." *UCI Machine Learning*. https://archive.ics.uci.edu/ml/datasets/Breast+Cancer.