

A Deductive Learning of Heart Disease Dataset by using K Means Clustering

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Abstract

Cardiovascular diseases is one of the most significant causes of mortality in today's world. Cardiovascular diseases are the number one cause of death globally with 17.9 million death cases each year. CVDs are concerted by hypertension, diabetes, overweight and unhealthy lifestyles. Exploratory Data Analysis is a pre-processing step to understand the data. There are numerous methods and steps in performing EDA, however, most of them are specific, focusing on visualization and distribution. If the number of cluster is 2, this model has 43% & 57% of cluster instances for full training set and 46% & 54% of cluster instances for 66% training set, if the number of cluster is 3, this model has 18% 48% & 34% of cluster instances for full training set and 25%,50% & 25% of cluster instances for 66% training set, if the number of cluster is 4, this model has 21%,40%,10% & 28% of cluster instances for full training set and 24%,13%,26% and 37% of cluster instances for 66% training set, If the number of cluster is 5, this model has 17%,31%,11%,19% & 21% of cluster instances for full training set and 23%,14%,20%,33% &11% of cluster instances for 66% training set, If the number of cluster is 6, this model has 10%,31%,15%,20%,6% &18% of cluster instances for full training set and 16%,18%,15%,22%,13% &15% of cluster instances for 66% training set. In this system proposes the optimal results for build the deductive learning model. Based on the time consumption the system recommends that cluster 2, 3 and 5 have zero second taken the time consumption for build the model in 66% training set. 0.01 seconds for cluster 6 and 0.03 seconds for cluster 4 in 66% training set models. Cluster 5 and 6 have low sum of squared errors for full training and 66% training set comparatively other models.

Keywords: K Means clustering, Centroids, Sum of Squared Errors, Iterations.

Introduction

In this section presents introduction of this research work. 17.9 million people die every year due to heart diseases accounting for 31% of all the deaths in the world. [1]Thus, it is important for early and accurate detection of heart diseases.[2] 4 out of 5 Heart disease patients die due to a heart attack or a stroke, and one-third of the deaths occur prematurely in patients below 70 years of age.[3] Heart disease proves to be the leading cause of death for both men and women.[4] This affects the human life very badly. The diagnosis of heart disease in most cases depends on a complex combination and huge volume of clinical and pathological data. Machine learning has been shown to be effective assisting in making decisions and predictions from the large quantity of data produced by the health care industry.[5] Patients with a high risk of heart diseases demonstrate

raised blood pressure, glucose, and lipids along with overweight, and obesity. [6]Lifestyle also plays an important factor in heart diseases along with physiological factors. [7]Tobacco use, unhealthy diet, excessive alcohol intake, and inadequate physical activity are leading reasons for heart diseases.[8]Identifying such people and ensuring they are given appropriate treatment could prevent premature deaths.

In this paper presents section 2 of this paper explains the detail on the related works. In section 3 presents the materials and methods adopted and section 4 presents the details of the experiments and discussions. Finally section 5 concludes the paper by sharing our inferences and future plans.

Related Works

In this section presents focuses the related works of this research work. The accuracy and precision statistics for different algorithms such as Support Vector machines, KNN, Decision Trees, and Neural networks being most popular.[9]The UCI dataset for comparison of different classifiers such as Multilayer perceptron, Naive Bayes, KNN etc. and validated that SVM with boosting hyper parameters outperformed others.[10] The machine learning techniques providing the accuracy of 88.7% in prediction of cardiovascular diseases with a hybrid random forest and linear model.[11]New selection features and methods can be adopted to get broader perception of performance.[12]The traditional machine learning algorithms that aim in improving the accuracy of heart disease prediction. [13]The UK Biobank dataset observed that rather than complex models, information gain was better by consideration of different risk factors. a south African dataset consisting of 462 instances for analyzing algorithms such as Naive Bayes, SVM, and decision trees.[14]Naive Bayes obtained good accuracy results however specificity and sensitivity results can be improved with more instances.[15]the accuracy of decision trees in the prediction of heart diseases with the help of a dataset consisting of 573 instances. More number of attributes and hyper parameters can result in better performance classification.[16] Association rules, clustering and other data mining algorithms prove to be useful to mine huge amounts of unstructured data. [17]Various kernel implementations with certain rule-based classifiers.[18]It concludes that the RBF kernel is best for infinite data and Hyper parameter tuning can be added to make the model more effective. [19]

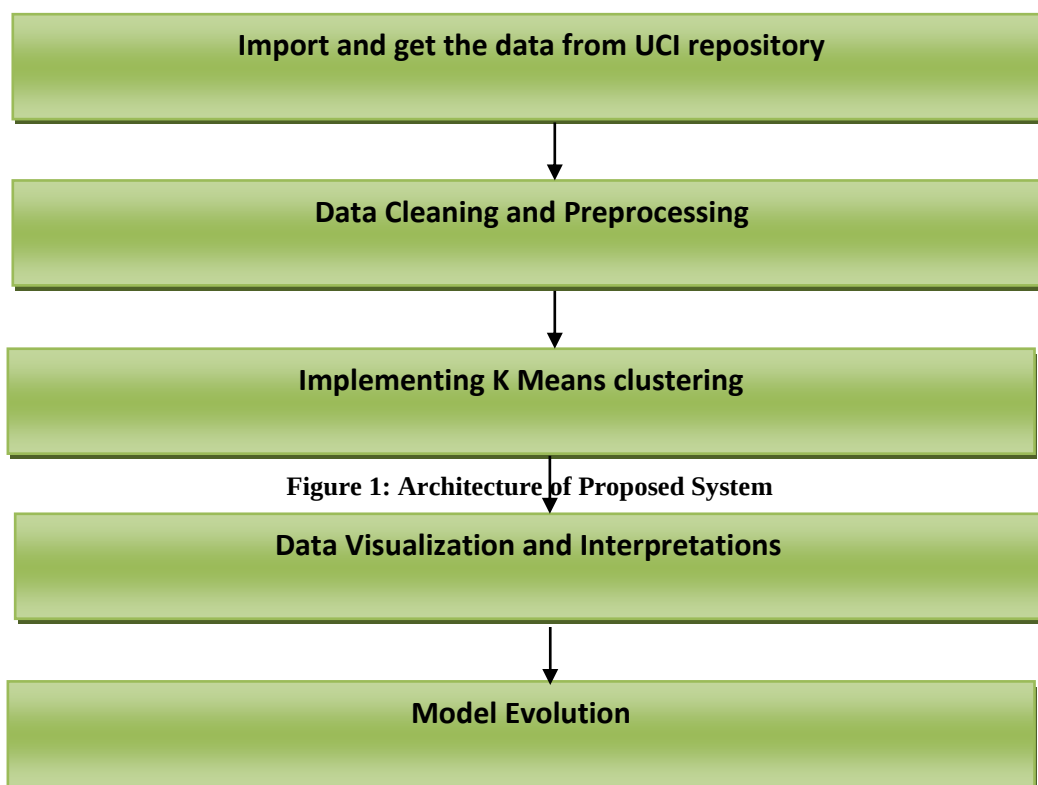
Materials and Methods

In this section presents the materials and methods of this research work. This research work focuses exploratory data analysis and using Weka 3.8.3. The dataset used in this work is UCI Heart Disease dataset. It has 76 features (attributes) from 303 patients. This work uses the dataset consisting of 270 patients with 14 features set.

Table 1: Meta Data Description

S.No	Attribute	Description of the Attribute	Type of the Attribute	Range
1	age	Age in years	Continuous	Minimum = 29 Maximum = 77
2	sex	0=Female ; 1=Male	Binary	Male=183

				Female=87
3	chest pain type (cp)	Type of the chest pain	Categorical	Asymtomatic=129 Non Angina=79 Atypical Angina=42 Typical Angina=20
4	resting blood pressure (restbps)	in mm Hg on admission to the hospital	Continuous	Minimum=94 Maximum=200 Mean= 131.34 StdDeviation=17.86
5	serum cholestoral (chol)	serum cholestoral in mg/dl	Continuous	Minimum=126 Maximum=564 Mean= 249.66 StdDeviation=51.69
6	fasting blood sugar (fbs)	0=false; 1=true	Binary	False=230 True=40
7	Resting ECG (restecg)	(fbs>120 mg/dl) 0=Normal ; 1=Having ST-T wave abnormality; 2=Showing probable or define left ventricular hypertrophy	Categorical	Normal=131 ST-T Wave Abnormality=2 Left Ventricular Hypertrophy=137
8	maximum heart rate achieved (thalach)	maximum heart rate reached	Continuous	Minimum=71 Maximum=202 Mean= 149.68 StdDeviation=23.17
9	exercise induced angina (exang)	0=No; 1=Yes	Binary	No=181 Yes=89
10	oldpeak	ST depreve to restserelatision induced by exercise relative to rest	Continuous	Minimum=0 Maximum=6.2 Mean= 1.05 StdDeviation=1.145
11	slope	the slope of the peak exercise ST segment 0=upsloping; 1=Flat; 2=Downsloping	Categorical	Flat=122 Upsloping=130 Downsloping=18
12	ca	number of major vessels(0-3) colored by flourosopy 0=Typical Angina 1=Atypical Angina 2=Non Anginal Pain 3=Asymptomatic	Categorical	Typical Angina=160 Atypical Angina=58 Non AnginalPain=33 Asymptomatic=19
13	thal	1=normal; 2=fixed defect; 3=reversible defect	Categorical	Normal=152 Fixed Defect=14 Reversible Defect=104
14	diagnosis of heart disease (Target or Class)	angiographic disease status 0=Disease; 1=No Disease	Binary	Disease=151 No Disease=119



Results and Discussion

In this section focuses the results and discussions of this research work. This project covers exploratory data analysis like data visualization and implementing K means clustering approaches by using Weka 3.8.3.

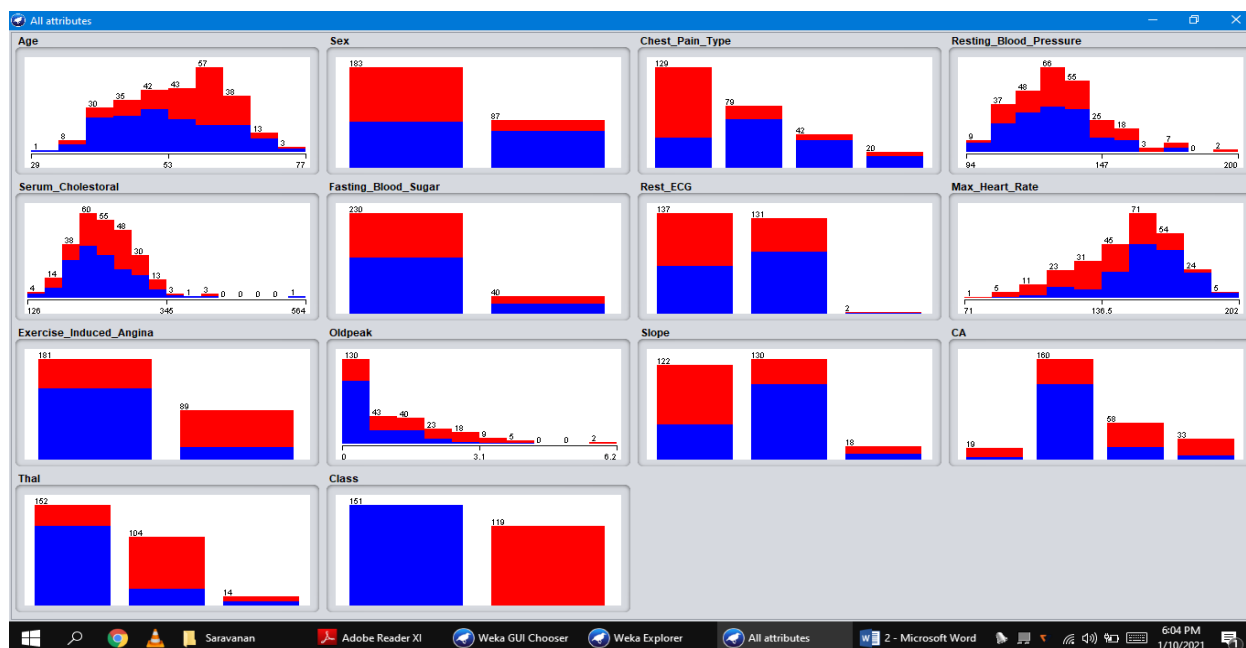


Figure 2: Data Visualization for all Attributes

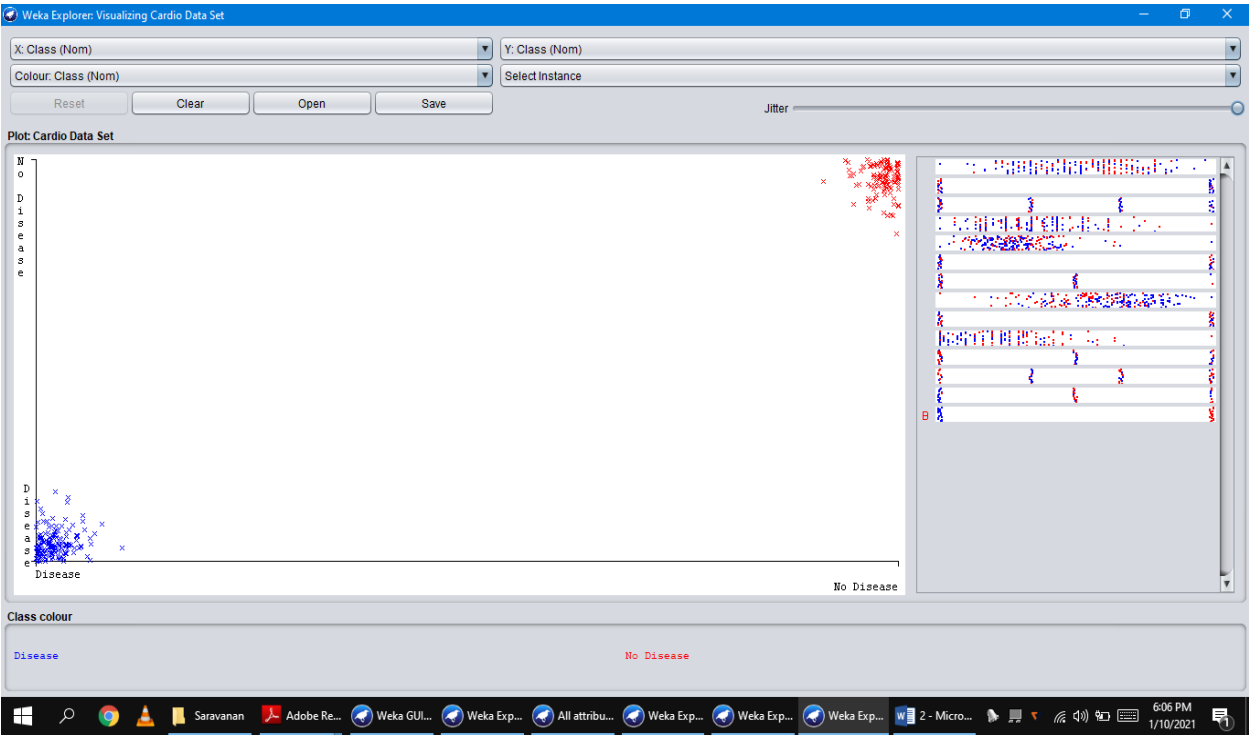


Figure 3: Visualization of Target Attribute

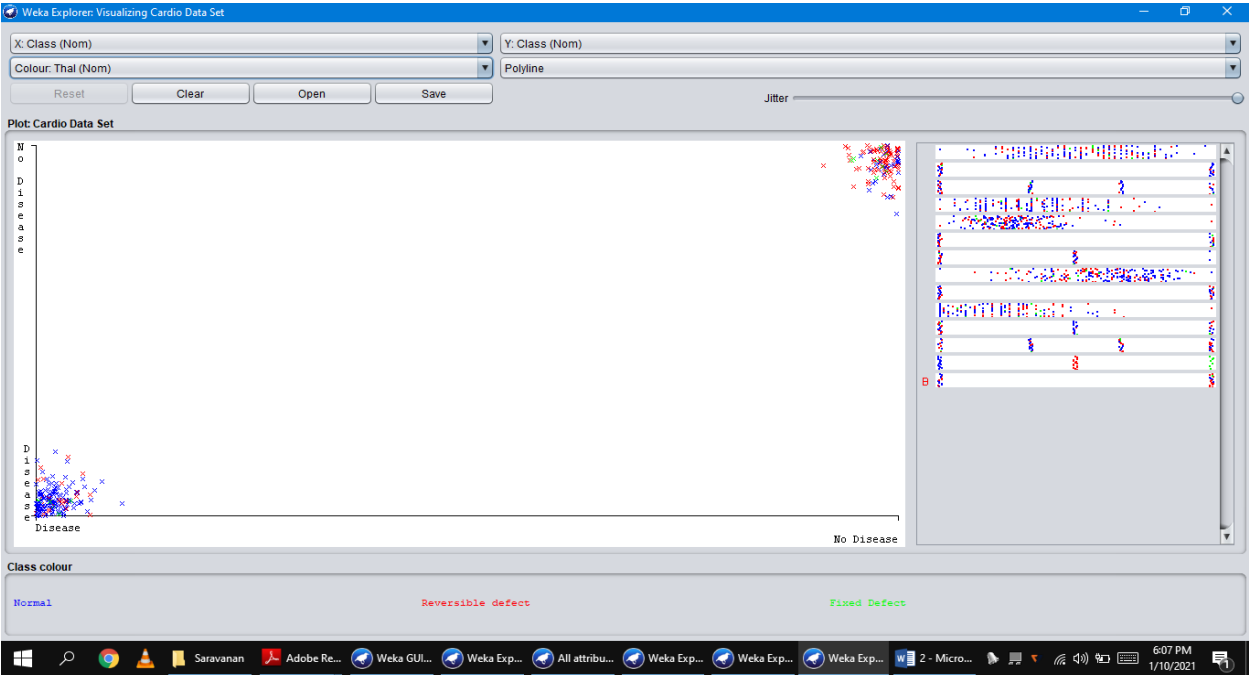


Figure 4: Visualization of ThalAttribute

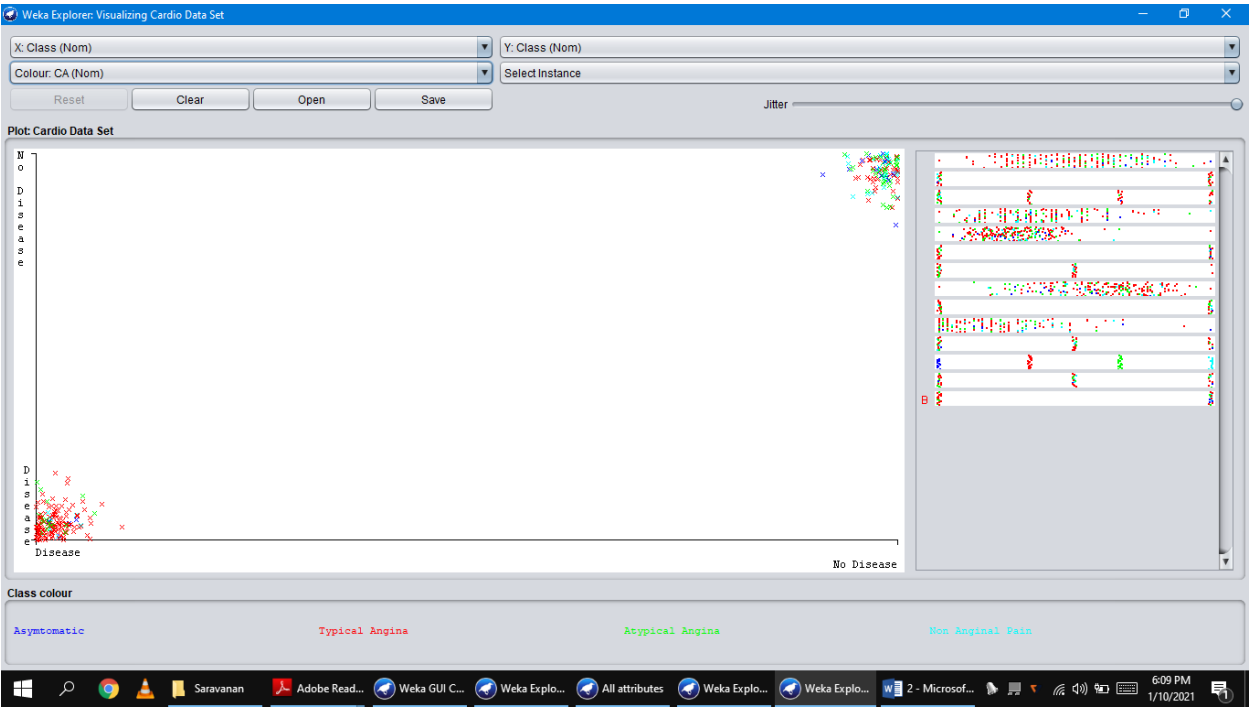


Figure 5: Visualization of CA Attribute

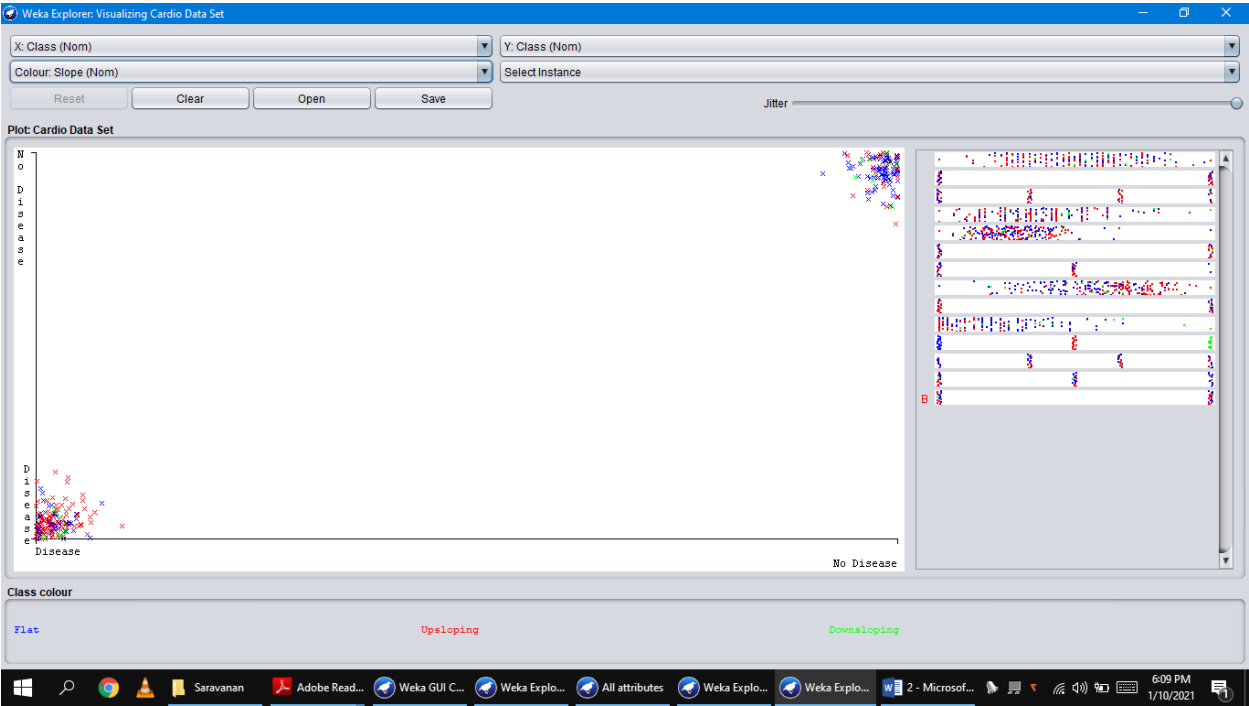


Figure 6: Visualization of Slope Attribute

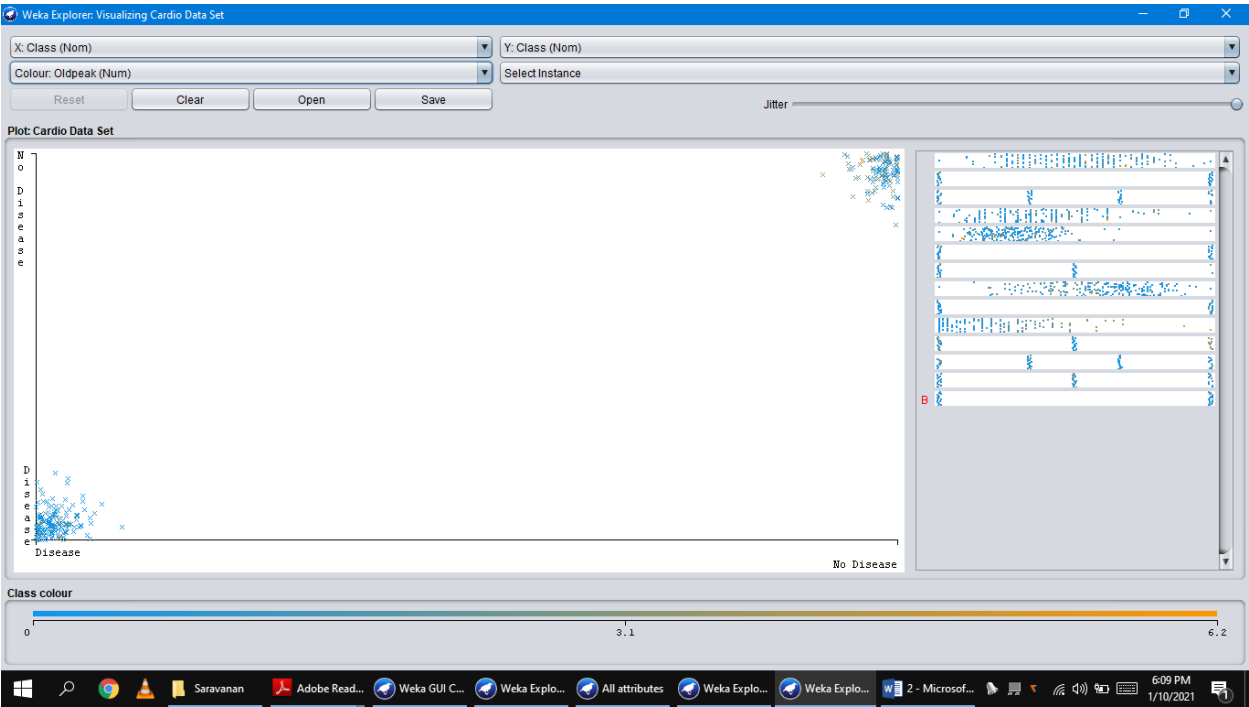


Figure 7: Visualization of Old Peak Attribute

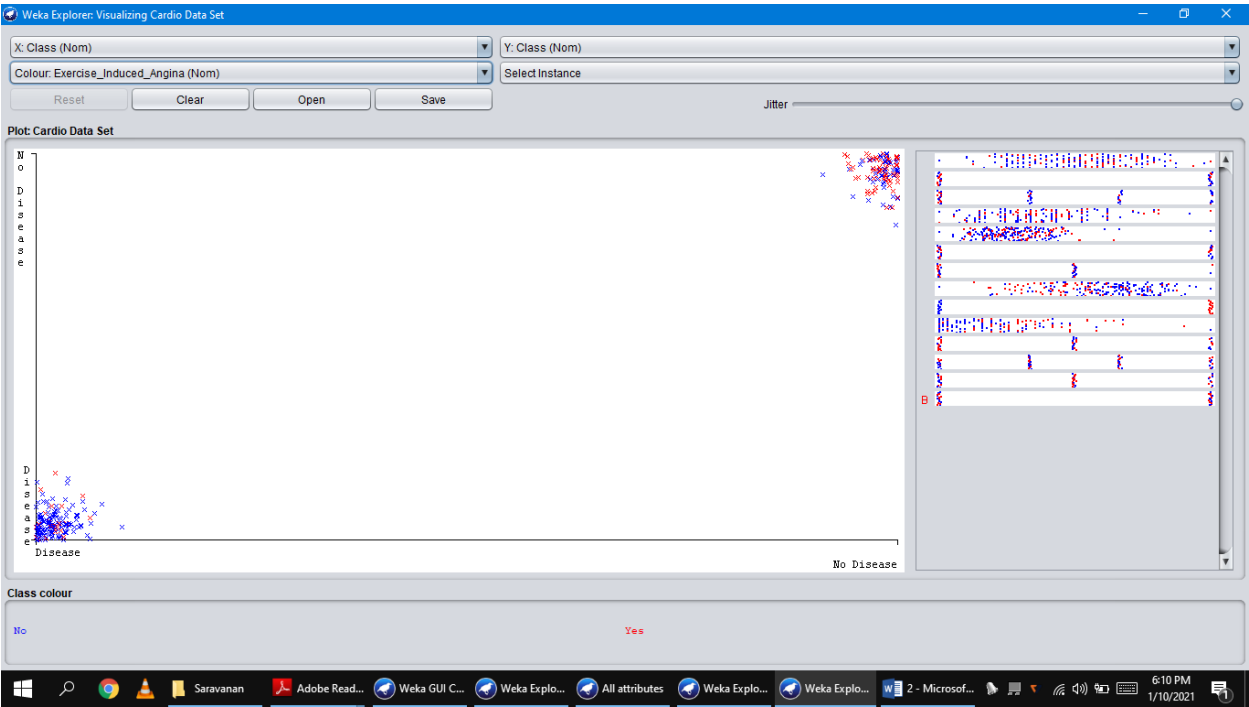


Figure 8: Visualization of Exercise_Induced_Angina Attribute

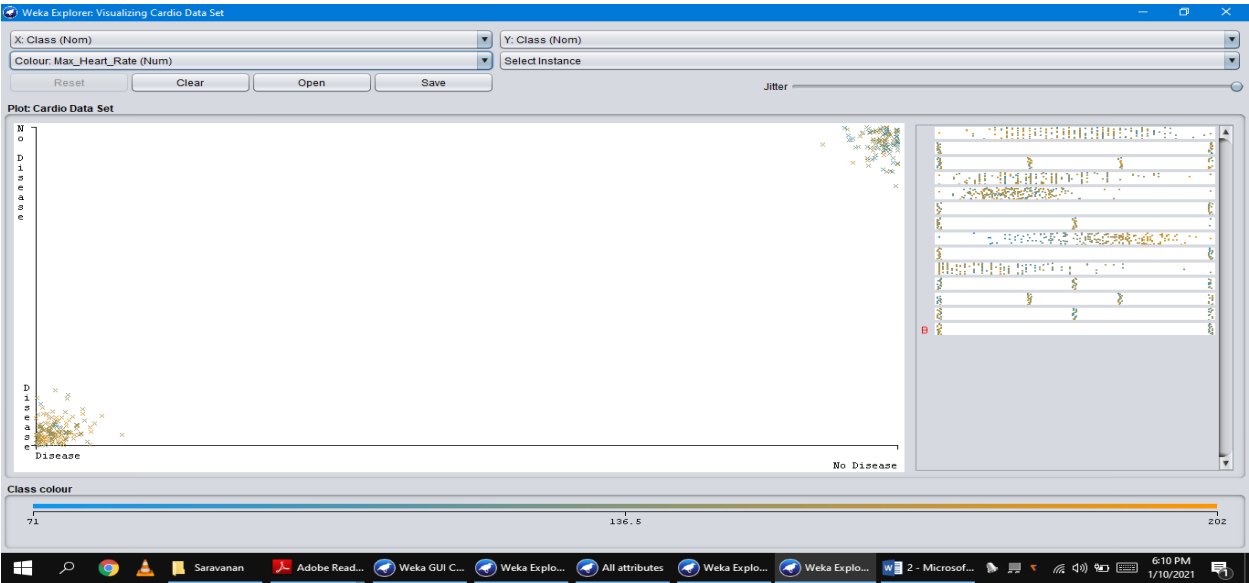


Figure 9: Visualization of Max_Heart Rate Attribute

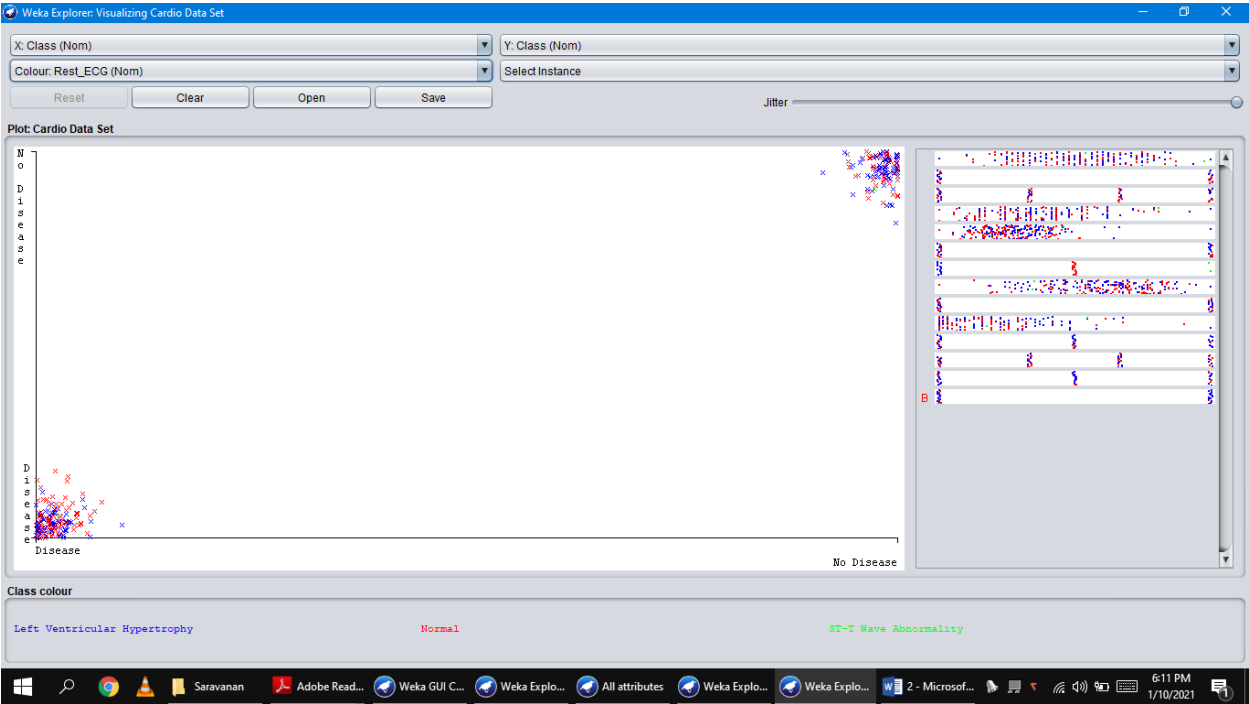


Figure 10: Visualization of Rest_ECGAttribute

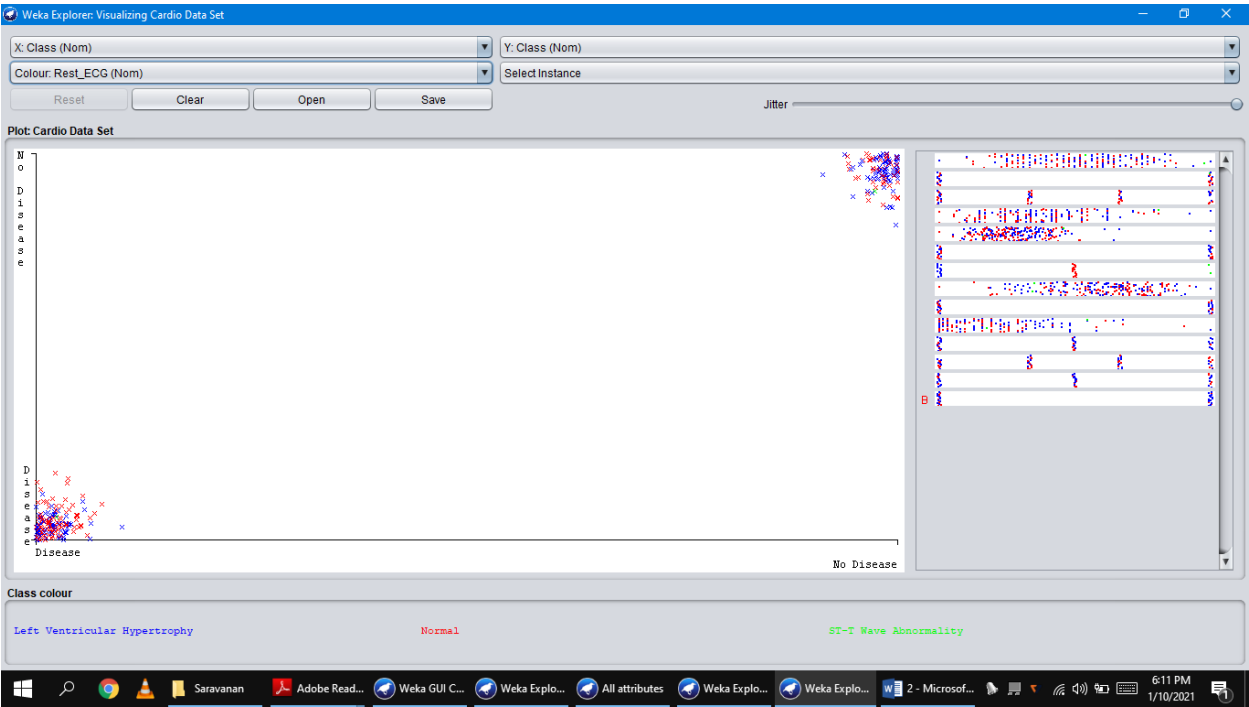


Figure 11: Visualization of Fasting_ECGAttribute

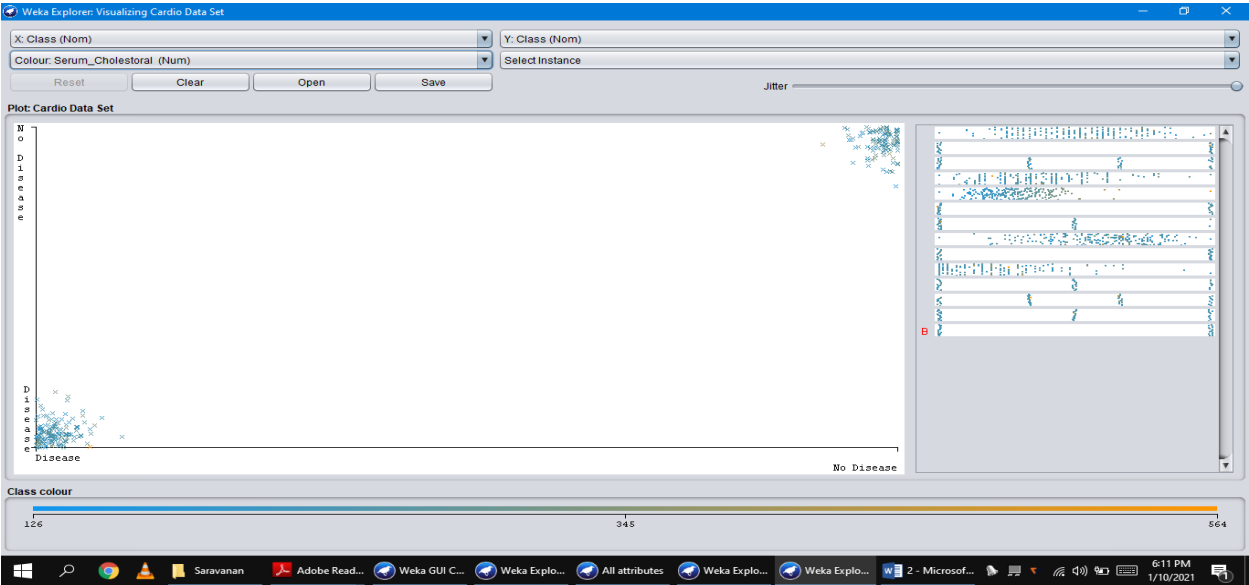


Figure 12: Visualization of Serum_CholastralAttribute

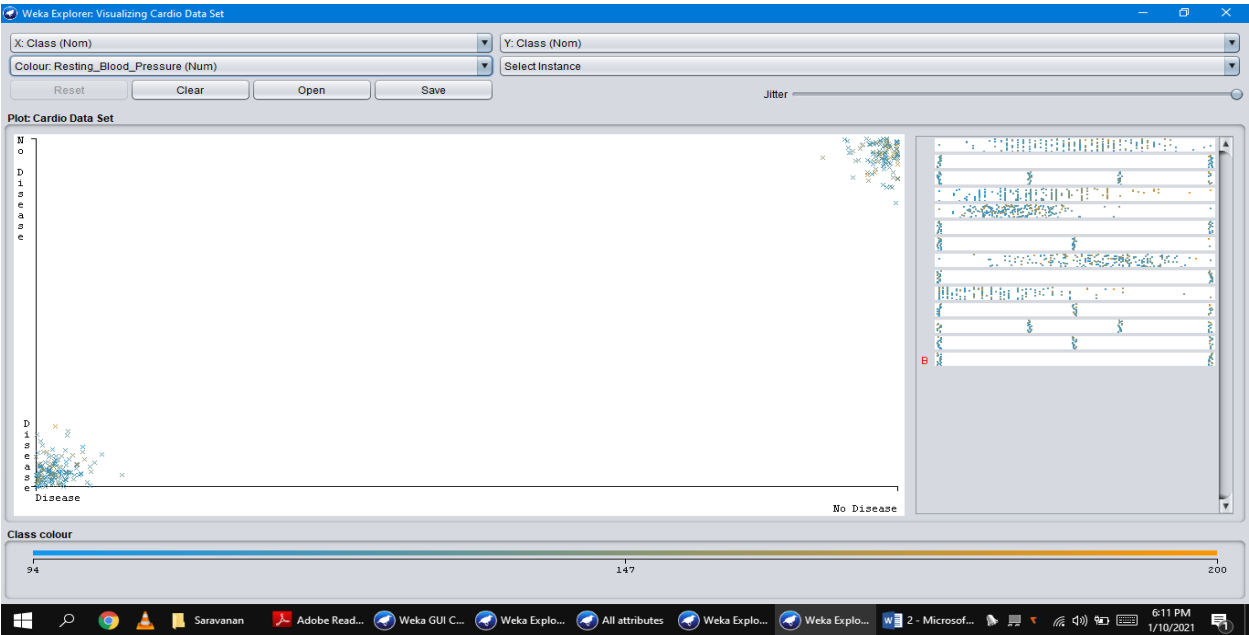


Figure 13: Visualization of Resting_Blood Pressure Attribute

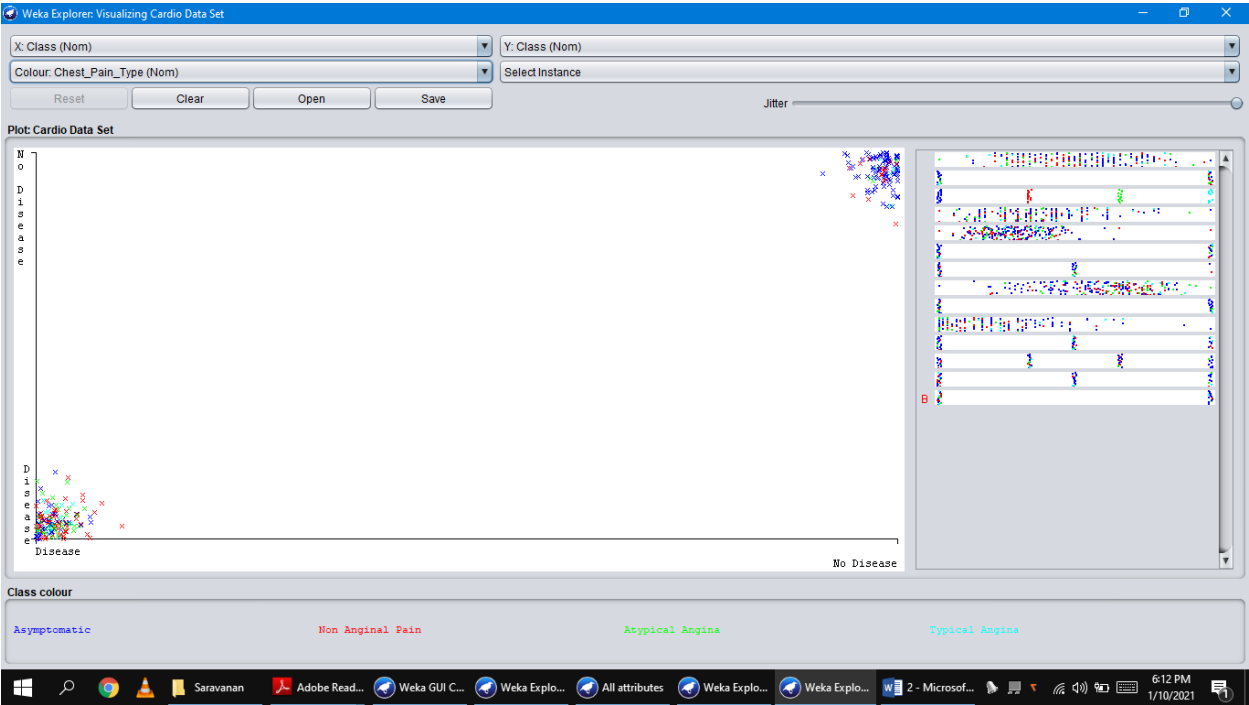


Figure 14: Visualization of Chest PainAttribute

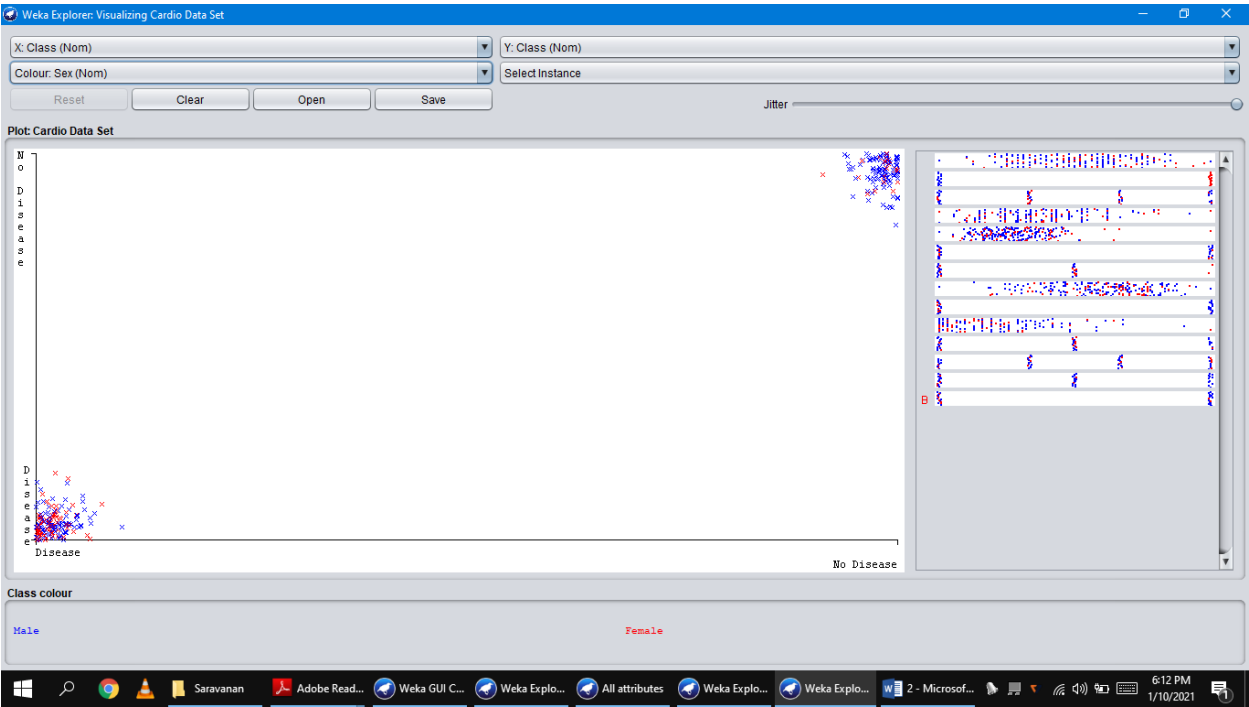


Figure 15: Visualization of Sex Attribute

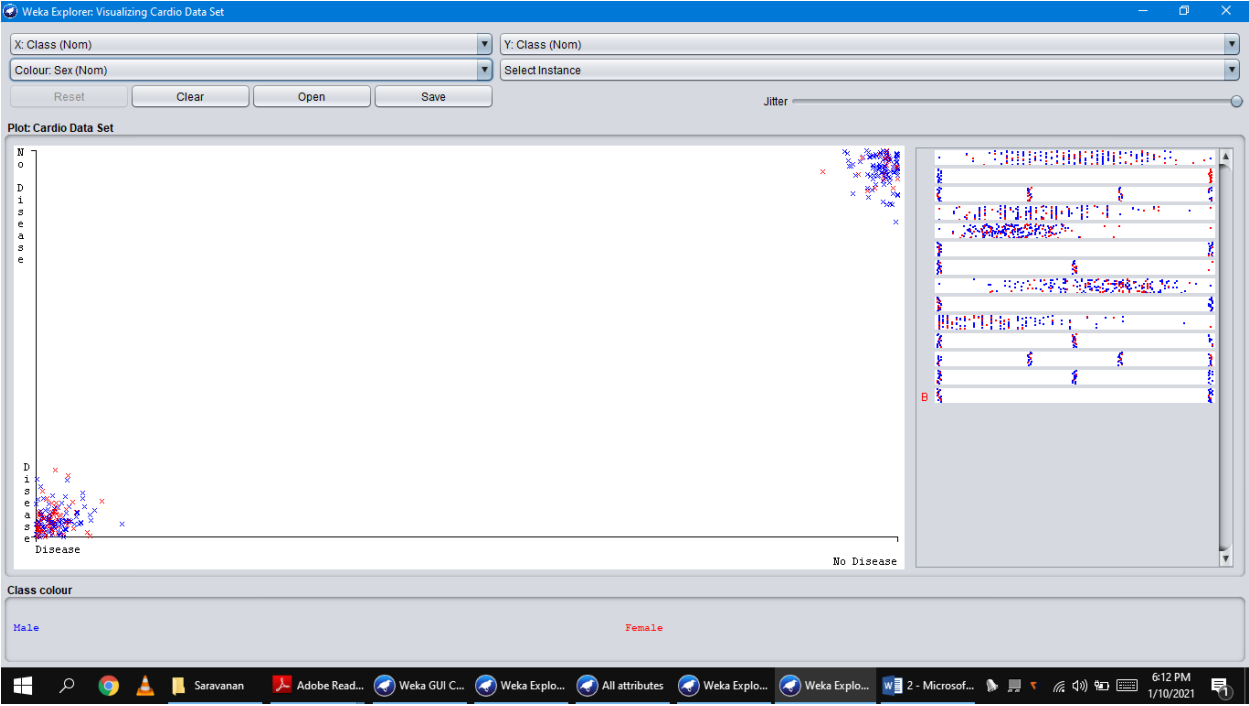


Figure 16: Visualization of Age Attribute

The diagrams shows that the data visualizations of all the attributes like there are 14 attributes lies on the disease and no disease classes.

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kMeans
=====

Number of iterations: 4
Within cluster sum of squared errors: 710.4670812208781

Initial starting points (random):

Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular Hypertrophy',160,No,3.6,Downsloping,'Non Anginal Pain',Normal,'No Disease'
Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease

Missing values globally replaced with mean/mode

Final cluster centroids:

Attribute                                Full Data                                Cluster#                                0                                1
(270.0)                                (115.0)                                (155.0)
=====
Age                                54.4333                                56.8348                                52.6516
Sex                                Male                                Male                                Male
Chest_Pain_Type                                Asymptomatic                                Asymptomatic                                Non Anginal Pain
Resting_Blood_Pressure                                131.3444                                133.8783                                129.4645
Serum_Cholesterol                                249.6593                                254.9913                                245.7032
Fasting_Blood_Sugar                                FALSE                                FALSE                                FALSE
Rest_ECG                                Left Ventricular Hypertrophy Left Ventricular Hypertrophy                                Normal
Max_Heart_Rate                                149.6778                                136.9826                                159.0968
Exercise_Induced_Angina                                No                                Yes                                No
Oldpeak                                1.05                                1.6157                                0.6303
Slope                                Upsloping                                Flat                                Upsloping
CA                                Typical Angina                                Atypical Angina                                Typical Angina
Thal                                Normal                                Reversible defect                                Normal
Class                                Disease                                No Disease                                Disease

Time taken to build model (full training data) : 0.01 seconds

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Figure 17: K Means cluster No=2

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kMeans
=====

Number of iterations: 4
Within cluster sum of squared errors: 648.2251333482561

Initial starting points (random):

Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular Hypertrophy',160,No,3.6,Downsloping,'Non Anginal Pain',Normal,'No Disease'
Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease
Cluster 2: 60,Male,Asymptomatic,117,230,TRUE,Normal,160,Yes,1.4,Upsloping,'Non Anginal Pain','Reversible defect ','No Disease'

Missing values globally replaced with mean/mode

Final cluster centroids:

Attribute                                Full Data                                Cluster#                                0                                1                                2
(270.0)                                (49.0)                                (130.0)                                (91.0)
=====
Age                                54.4333                                59                                51.4                                56.3077
Sex                                Male                                Female                                Male                                Male
Chest_Pain_Type                                Asymptomatic                                Asymptomatic                                Non Anginal Pain                                Asymptomatic
Resting_Blood_Pressure                                131.3444                                134.6735                                129.3308                                132.4286
Serum_Cholesterol                                249.6593                                271.6735                                240.4308                                250.989
Fasting_Blood_Sugar                                FALSE                                FALSE                                FALSE                                FALSE
Rest_ECG                                Left Ventricular Hypertrophy Left Ventricular Hypertrophy                                Normal Left Ventricular Hypertrophy
Max_Heart_Rate                                149.6778                                145.4694                                161.9308                                134.4396
Exercise_Induced_Angina                                No                                No                                No                                Yes
Oldpeak                                1.05                                1.1816                                0.5792                                1.6516
Slope                                Upsloping                                Flat                                Upsloping                                Flat
CA                                Typical Angina                                Typical Angina                                Typical Angina                                Atypical Angina
Thal                                Normal                                Normal                                Normal                                Reversible defect
Class                                Disease                                No Disease                                Disease                                No Disease

Time taken to build model (full training data) : 0.01 seconds

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Figure 18: K Means cluster=3

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KMeans
=====

Number of iterations: 8
Within cluster sum of squared errors: 608.7121457525952

Initial starting points (random):

Cluster 0: 62, Female, Asymptomatic, 140, 268, FALSE, 'Left Ventricular Hypertrophy', 160, No, 3.6, Downsloping, 'Non Anginal Pain', Normal, 'No Disease'
Cluster 1: 42, Male, Asymptomatic, 140, 226, FALSE, Normal, 178, No, 0, Upsloping, 'Typical Angina', Normal, Disease
Cluster 2: 60, Male, Asymptomatic, 117, 230, TRUE, Normal, 160, Yes, 1.4, Upsloping, 'Non Anginal Pain', 'Reversible defect ', 'No Disease'
Cluster 3: 64, Male, Asymptomatic, 128, 263, FALSE, Normal, 105, Yes, 0.2, Flat, 'Atypical Angina', 'Reversible defect ', Disease

Missing values globally replaced with mean/mode

Final cluster centroids:

Attribute          Full Data          Cluster#          1          2          3
                   (270.0)          (57.0)          (109.0)          (28.0)          (76.0)
=====
Age                54.4333            58.5263            50.3303            56            56.6711
Sex                Male              Female             Male              Male              Male
Chest_Pain_Type    Asymptomatic      Asymptomatic      Non Anginal Pain  Asymptomatic      Asymptomatic
Resting_Blood_Pressure 131.3444          134.0877          128.1193          132.0357          133.6579
Serum_Cholesterol    249.6593          263.193           237.9817          261.0714          252.0526
Fasting_Blood_Sugar  FALSE            FALSE            FALSE            FALSE            FALSE
Rest_ECG           Left Ventricular Hypertrophy Left Ventricular Hypertrophy Normal Left Ventricular Hypertrophy Left Ventricular Hypertrophy
Max_Heart_Rate      149.6778          146.9123          162.9633          155.8929          130.4079
Exercise_Induced_Angina No                No                No                Yes              Yes
Oldpeak            1.05              1.0667           0.511             0.7679           1.9145
Slope              Upsloping         Flat              Upsloping         Upsloping         Flat
CA                 Typical Angina     Typical Angina     Typical Angina     Non Anginal Pain  Atypical Angina
Thal               Normal            Normal            Normal            Reversible defect Reversible defect
Class              Disease           Disease           Disease           No Disease        No Disease

```

Figure 19: K Means cluster=4

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ons: 5
m of squared errors: 581.9453678366942

points (random):

ale, Asymptomatic, 140, 268, FALSE, 'Left Ventricular Hypertrophy', 160, No, 3.6, Downsloping, 'Non Anginal Pain', Normal, 'No Disease'
e, Asymptomatic, 140, 226, FALSE, Normal, 178, No, 0, Upsloping, 'Typical Angina', Normal, Disease
e, Asymptomatic, 117, 230, TRUE, Normal, 160, Yes, 1.4, Upsloping, 'Non Anginal Pain', 'Reversible defect ', 'No Disease'
e, Asymptomatic, 128, 263, FALSE, Normal, 105, Yes, 0.2, Flat, 'Atypical Angina', 'Reversible defect ', Disease
ale, Asymptomatic, 128, 303, FALSE, 'Left Ventricular Hypertrophy', 159, No, 0, Upsloping, 'Atypical Angina', Normal, Disease

obally replaced with mean/mode

troids:

Full Data          Cluster#          0          1          2          3          4
                   (270.0)          (45.0)          (85.0)          (31.0)          (51.0)          (58.0)
=====
54.4333            59.5778            47.8118            56.6129            55.2549            58.2586
Male              Male              Male              Male              Male              Female
Asymptomatic      Asymptomatic      Non Anginal Pain  Asymptomatic      Asymptomatic      Non Anginal Pain
131.3444          139.7556          125.7882          134.6774          129.3725          132.9138
249.6593          250.8444          231.1176          264.8387          246.3529          270.7069
FALSE            FALSE            FALSE            FALSE            FALSE            FALSE
Left Ventricular Hypertrophy Left Ventricular Hypertrophy Normal Left Ventricular Hypertrophy Normal Left Ventricular Hypertrophy
149.6778          142.4667          164.8235          145.7742          129.098            153.2586
No                No                No                Yes              Yes              No
1.05              1.9333           0.4741            1.371            1.6431            0.5155
Upsloping         Flat              Upsloping         Upsloping         Flat              Upsloping
Typical Angina     Typical Angina     Typical Angina     Non Anginal Pain  Atypical Angina     Typical Angina
Normal            Normal            Normal            Reversible defect Reversible defect  Normal
Disease           No Disease        Disease           No Disease        No Disease        Disease

```

Figure 20: K Means cluster=5

The above pictures shown that the K Means clusters of all attributes (14 attributes) in the heart disease dataset for implementing deductive learning process.

Table 2: Metrics of K Means Clusters

S.No	Number of Clusters	Number of Iterations	Sum of Squared Errors (Within cluster)	Clustered Instances	Time Taken to Build the Model (In
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								seconds)	
		A	B	A	B	A	B	A	B
1	2	3	4	710.46	466.36	0-115(43%) 1-155(57%)	0-42 (46%) 1-50 (54%)	0.01	0
2	3	4	4	648.26	426.63	0-49(18%) 1-130(48%) 2-91(34%)	0-23(25%) 1-46(50%) 2-23(25%)	0.01	0
3	4	5	8	608.71	398.56	0-57(21%) 1-109(40%) 2-28(10%) 3-76(28%)	0-22(24%) 1-12(13%) 2-24(26%) 3-34(37%)	0.01	0.03
4	5	6	5	581.95	379.77	0-45(17%) 1-85(31%) 2-31(11%) 3-51(19%) 4-58(21%)	0-21(23%) 1-13(14%) 2-18(20%) 3-30(33%) 4-10(11%)	0.01	0
5	6	7	9	572.62	355.02	0-27(10%) 1-83(31%) 2-40(15%) 3-55(20%) 4-17(6%) 5-48(18%)	0 -15 (16%) 1-17(18%) 2-14(15%) 3-20(22%) 4-12(13%) 5-14(15%)	0.02	0.01

- Cluster Model (Full Training Set) =A

- Cluster Model(66% Split)=B

* Implementing Euclidean distance (or similarity) function.

The above table represents that the various measurements producing while implementing full training and 66% training set of the heart disease dataset.

The below table represents that the centroid clusters of K means clusters for full and 66% training set in Weka 3.8.3 tool.

Table 3: Centroid clusters of K Means Clusters for Full / 66% Training set

Cluster Centroids / Clustering model (full training set)		
S.No	Nu mbe r of Clus ters	Initial starting points (random)
1	2	Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular

		Hypertrophy',160,No,3.6,Downsloping,'Non AnginalPain',Normal,'No Disease' Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease
2	3	Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular Hypertrophy',160,No,3.6,Downsloping,'Non AnginalPain',Normal,'No Disease' Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease Cluster 2: 60,Male,Asymptomatic,117,230,TRUE,Normal,160,Yes,1.4,Upsloping,'Non AnginalPain','Reversible defect ','No Disease'
3	4	Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular Hypertrophy',160,No,3.6,Downsloping,'Non AnginalPain',Normal,'No Disease' Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease Cluster 2: 60,Male,Asymptomatic,117,230,TRUE,Normal,160,Yes,1.4,Upsloping,'Non AnginalPain','Reversible defect ','No Disease' Cluster 3: 64,Male,Asymptomatic,128,263,FALSE,Normal,105,Yes,0.2,Flat,'Atypical Angina','Reversible defect ',Disease
4	5	Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular Hypertrophy',160,No,3.6,Downsloping,'Non AnginalPain',Normal,'No Disease' Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease Cluster 2: 60,Male,Asymptomatic,117,230,TRUE,Normal,160,Yes,1.4,Upsloping,'Non AnginalPain','Reversible defect ','No Disease' Cluster 3: 64,Male,Asymptomatic,128,263,FALSE,Normal,105,Yes,0.2,Flat,'Atypical Angina','Reversible defect ',Disease Cluster 4: 57,Female,Asymptomatic,128,303,FALSE,'Left Ventricular Hypertrophy',159,No,0,Upsloping,'Atypical Angina',Normal,Disease
5	6	Cluster 0: 62,Female,Asymptomatic,140,268,FALSE,'Left Ventricular Hypertrophy',160,No,3.6,Downsloping,'Non AnginalPain',Normal,'No Disease' Cluster 1: 42,Male,Asymptomatic,140,226,FALSE,Normal,178,No,0,Upsloping,'Typical Angina',Normal,Disease Cluster 2: 60,Male,Asymptomatic,117,230,TRUE,Normal,160,Yes,1.4,Upsloping,'Non AnginalPain','Reversible defect ','No Disease' Cluster 3: 64,Male,Asymptomatic,128,263,FALSE,Normal,105,Yes,0.2,Flat,'Atypical Angina','Reversible defect ',Disease Cluster 4: 57,Female,Asymptomatic,128,303,FALSE,'Left Ventricular Hypertrophy',159,No,0,Upsloping,'Atypical Angina',Normal,Disease

		Cluster 5: 50,Female,Asymptomatic,110,254,FALSE,'Left Ventricular Hypertrophy',159,No,0,Upsloping,'Typical Angina',Normal,Disease
Cluster Centroids / Clustering model (66% Slit)		
1	2	Cluster 0: 48,Male,'Non Anginal Pain',124,255,TRUE,Normal,175,No,0,Upsloping,'Non Anginal Pain',Normal,Disease Cluster 1: 38,Male,'Typical Angina',120,231,FALSE,Normal,182,Yes,3.8,Flat,'Typical Angina','Reversible defect ','No Disease'
2	3	Cluster 0: 48,Male,'Non Anginal Pain',124,255,TRUE,Normal,175,No,0,Upsloping,'Non Anginal Pain',Normal,Disease Cluster 1: 38,Male,'Typical Angina',120,231,FALSE,Normal,182,Yes,3.8,Flat,'Typical Angina','Reversible defect ','No Disease' Cluster 2: 44,Male,'Atypical Angina',120,263,FALSE,Normal,173,No,0,Upsloping,'Typical Angina','Reversible defect ',Disease
3	4	Cluster 0: 48,Male,'Non Anginal Pain',124,255,TRUE,Normal,175,No,0,Upsloping,'Non Anginal Pain',Normal,Disease Cluster 1: 38,Male,'Typical Angina',120,231,FALSE,Normal,182,Yes,3.8,Flat,'Typical Angina','Reversible defect ','No Disease' Cluster 2: 44,Male,'Atypical Angina',120,263,FALSE,Normal,173,No,0,Upsloping,'Typical Angina','Reversible defect ',Disease Cluster 3: 61,Male,Asymptomatic,120,260,FALSE,Normal,140,Yes,3.6,Flat,'Atypical Angina','Reversible defect ','No Disease'
4	5	Cluster 0: 48,Male,'Non Anginal Pain',124,255,TRUE,Normal,175,No,0,Upsloping,'Non Anginal Pain',Normal,Disease Cluster 1: 38,Male,'Typical Angina',120,231,FALSE,Normal,182,Yes,3.8,Flat,'Typical Angina','Reversible defect ','No Disease' Cluster 2: 44,Male,'Atypical Angina',120,263,FALSE,Normal,173,No,0,Upsloping,'Typical Angina','Reversible defect ',Disease Cluster 3: 61,Male,Asymptomatic,120,260,FALSE,Normal,140,Yes,3.6,Flat,'Atypical Angina','Reversible defect ','No Disease' Cluster 4: 58,Male,Asymptomatic,150,270,FALSE,'Left Ventricular Hypertrophy',111,Yes,0.8,Upsloping,'Typical Angina','Reversible defect ','No Disease'
5	6	Cluster 0: 48,Male,'Non Anginal Pain',124,255,TRUE,Normal,175,No,0,Upsloping,'Non Anginal Pain',Normal,Disease Cluster 1: 38,Male,'Typical Angina',120,231,FALSE,Normal,182,Yes,3.8,Flat,'Typical Angina','Reversible defect ','No Disease'

	Cluster 2: 44,Male,'Atypical Angina',120,263,FALSE,Normal,173,No,0,Upsloping,'Typical Angina','Reversible defect ',Disease
	Cluster 3: 61,Male,Asymptomatic,120,260,FALSE,Normal,140,Yes,3.6,Flat,'Atypical Angina','Reversible defect ', 'No Disease'
	Cluster 4: 58,Male,Asymptomatic,150,270,FALSE,'Left Ventricular Hypertrophy',111,Yes,0.8,Upsloping,'Typical Angina','Reversible defect ', 'No Disease'
	Cluster 5: 67,Male,Asymptomatic,120,237,FALSE,Normal,71,No,1,Flat,'Typical Angina','Normal','No Disease'

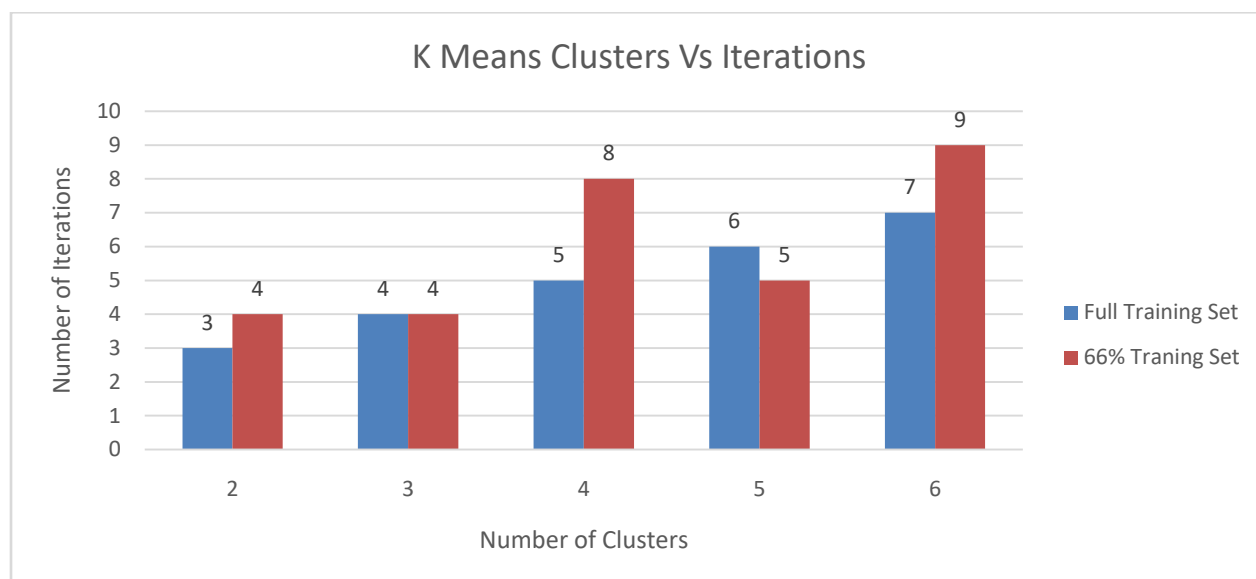


Figure 21: K Means cluster Vs Iterations

The above diagram clearly shows that number of cluster is 2, the model produces 3 iterations for full training set and 4 iterations for 66% training set, if the number of cluster is 3, the model produces 4 iterations for full training set and 66% training set, if the number of cluster is 4, the model produces 5 iterations for full training set and 8 iterations for 66% training set, If the number of cluster is 5, the model produces the 6 iterations for full training set and 5 iterations for 66% training set, If the number of cluster is 6, the model produces 7 iterations for full training set and 9 iterations for 66% training set.

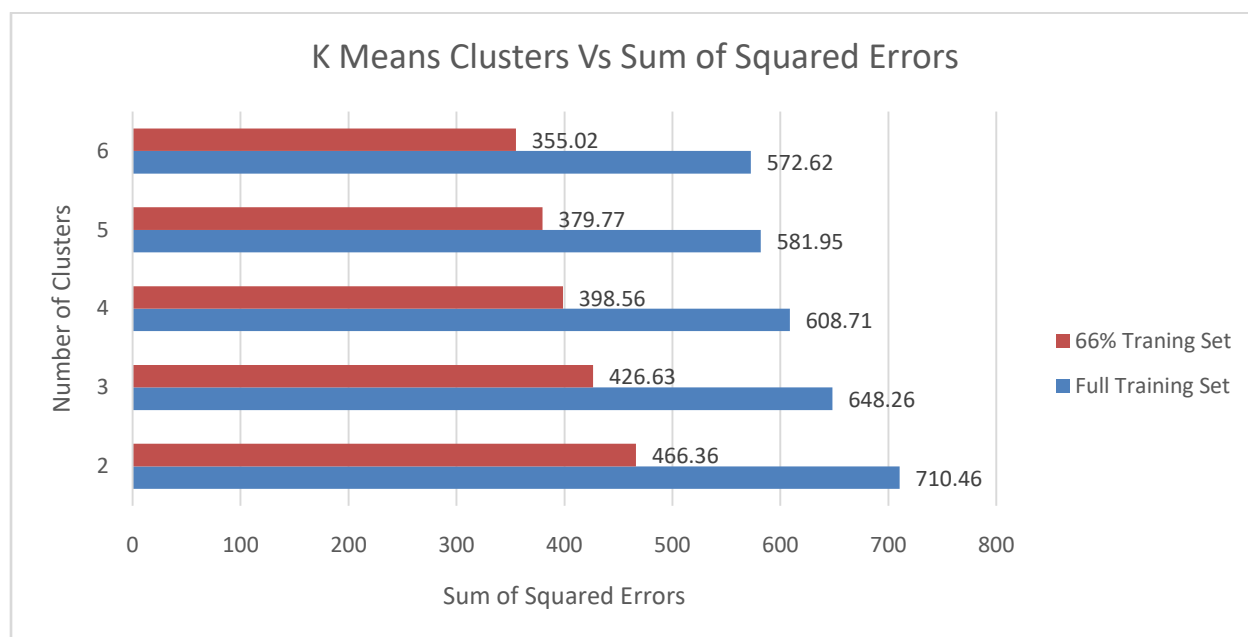


Figure 22: K Means cluster Vs SSE

The above diagram clearly shows that number of cluster is 2, this model has 710.46 sum of squared errors for full training set and 466.36 sum of squared errors for 66% training set, if the number of cluster is 3, this model has 648.26 sum of squared errors for full training set and 426.63 sum of squared errors for 66% training set, if the number of cluster is 4, this model has 608.71 sum of squared errors for full training set and 398.56 sum of squared errors for 66% training set, If the number of cluster is 5, this model has 581.95 sum of squared errors for full training set and 379.77 sum of squared errors for 66% training set, If the number of cluster is 6 this model has 572.62 sum of squared errors for full training set and 355.02 sum of squared errors for 66% training set.

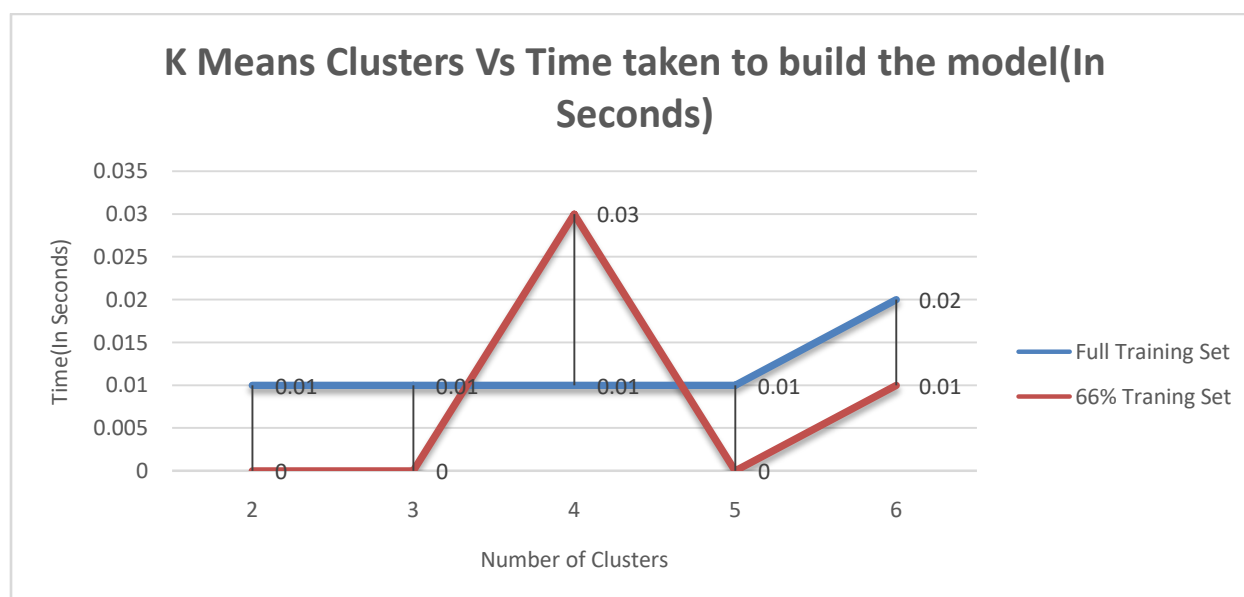


Figure 23: K Means cluster VsTime(Seconds)

The above diagram clearly shows that number of cluster is 2, this model has taken the time to build the model is 0.01 seconds for full training set and zero second for 66% training set, if the number of cluster is 3, this model has taken the time to build the model is 0.01 seconds for full training set and zero second for 66% training set, if the number of cluster is 4, this model has taken the time to build the model is 0.01 seconds for full training set and 0.03 seconds for 66% training set, If the number of cluster is 5, this model has taken the time to build the model is zero second for full training set and 0.01 seconds for 66% training set ,If the number of cluster is 6,this model has taken the time to build the model is 0.02 seconds for full training set and 0.01 seconds for 66% training set.

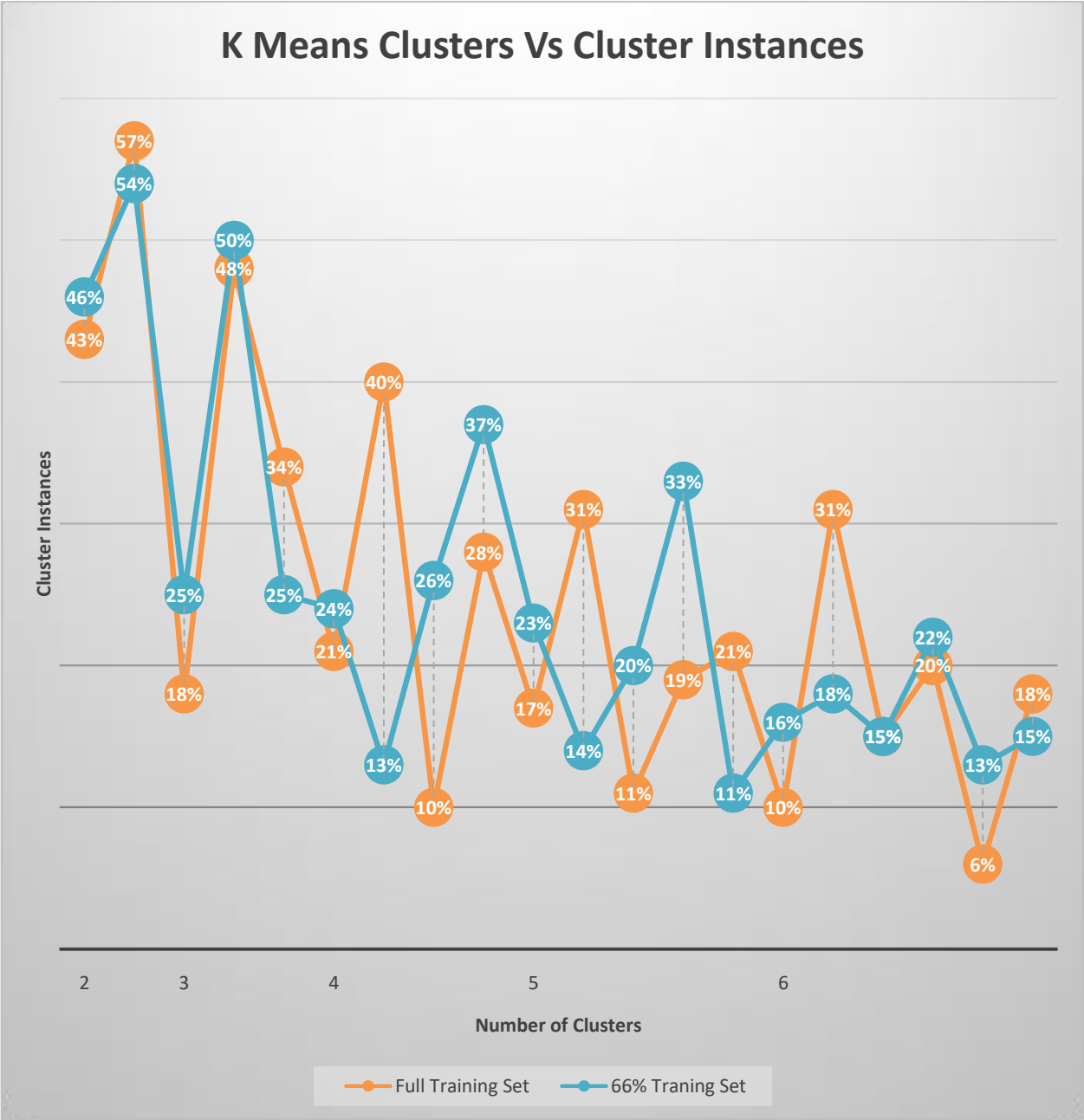


Figure 24: K Means cluster Vs Cluster Instances

The above diagram clearly shows that number of cluster is 2, this model has 43% & 57% of cluster instances for full training set and 46% & 54% of cluster instances for 66% training set, if the number of cluster is 3, this model has 18% 48% & 34% of cluster instances for full training set and 25%,50% & 25% of cluster instances for 66% training set, if the number of cluster is 4, this model has 21%,40%,10% & 28% of cluster instances for full training set and 24%,13%,26% and 37% of cluster instances for 66% training set, If the number of cluster is 5, this model has 17%,31%,11%,19% & 21% of cluster instances for full training set and 23%,14%,20%,33% &11% of cluster instances for 66% training set, If the number of cluster is 6, this model has 10%,31%,15%,20%,6% &18% of cluster instances for full training set and 16%,18%,15%,22%,13% &15% of cluster instances for 66% training set. Based on the time consumption the system recommends that cluster 2, 3 and 5 have zero second taken the time consumption for build the model in 66% training set. 0.01 seconds for cluster 6 and 0.03 seconds for cluster 4 in 66% training set models. Cluster 5 and 6 have low sum of squared errors for full training and 66% training set comparatively other models.

Conclusion

Finally this work concludes that when the proposed model has 6 clusters, it has more number of iteration to build the model like full training set has 7 iterations and 44% testing test has 9 iterations with 572.62 sum of squared error for full training set and 355.02 for 44% test set. It has taken the time to build the model 0.02 seconds for full training set and 0.01 second for 66% training set. This model produces the low sum of squared errors comparatively other models.

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