

Disease Prediction

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```
library(tidyr)
```

```
## Warning: package 'tidyr' was built under R version 4.1.2
```

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 4.1.2
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
## intersect, setdiff, setequal, union
```

```
library(magrittr)
```

```
## Warning: package 'magrittr' was built under R version 4.1.2
```

```
##  
## Attaching package: 'magrittr'
```

```
## The following object is masked from 'package:tidyr':  
##  
## extract
```

```
library(caret)
```

```
## Warning: package 'caret' was built under R version 4.1.2
```

```
## Loading required package: ggplot2
```

```
## Warning: package 'ggplot2' was built under R version 4.1.2
```

```
## Loading required package: lattice
```

```
library(Boruta)
```

```
## Warning: package 'Boruta' was built under R version 4.1.2
```

```
library(rpart)
```

```
## Warning: package 'rpart' was built under R version 4.1.2
```

```
library(rpart.plot)
```

```
## Warning: package 'rpart.plot' was built under R version 4.1.2
```

```
library(pROC)
```

```
## Type 'citation("pROC")' for a citation.
```

```
##  
## Attaching package: 'pROC'
```

```
## The following objects are masked from 'package:stats':  
##  
## cov, smooth, var
```

D. Data And Experiment

Data Preparation and Cleaning

The most important part of this project is to import and clean the data as needed. The dataset contains the variables as various clinical symptoms and prognosis as a result of combination of symptoms. The data is originally taken from Kaggle data source: <https://www.kaggle.com/datasets/kaushil268/disease-prediction-using-machine-learning> (<https://www.kaggle.com/datasets/kaushil268/disease-prediction-using-machine-learning>)

Importing data

We set the working directory as we have already downloaded the ‘Disease.csv’ data in my folder from the website.

```
setwd("/Users/pranjalsrivastava/Desktop/Projects/Disease_Prediction_Model")
```

After setting the working directory, we imported the csv data file and generating the raw data frame “Disease1”

```
Disease1 <- read.csv('Disease.csv')  
  
Disease1 <- Disease1 %>% select(-X)  
  
Disease1[] <- lapply(Disease1, as.factor)
```

We can clearly see that there no missing values in our final dataframe. Prognosis has 42 unique categorical values names as various diseases. All other variables are valued either 1 or 0

Feature Selection (Boruta)

```
Boruta <- Boruta(prognosis~., data = Disease1, doTrace = 2)  
  
## 1. run of importance source...  
  
## 2. run of importance source...  
  
## 3. run of importance source...  
  
## 4. run of importance source...  
  
## 5. run of importance source...  
  
## 6. run of importance source...  
  
## 7. run of importance source...  
  
## 8. run of importance source...
```

```

## 9. run of importance source...

## 10. run of importance source...

## 11. run of importance source...

## 12. run of importance source...

## 13. run of importance source...

## 14. run of importance source...

## After 14 iterations, +60 secs:

## confirmed 130 attributes: abdominal_pain, abnormal_menstruation, acidity, acute_liver_failure, alte

## rejected 1 attribute: fluid_overload;

## still have 1 attribute left.

## 15. run of importance source...

## 16. run of importance source...

## 17. run of importance source...

## 18. run of importance source...

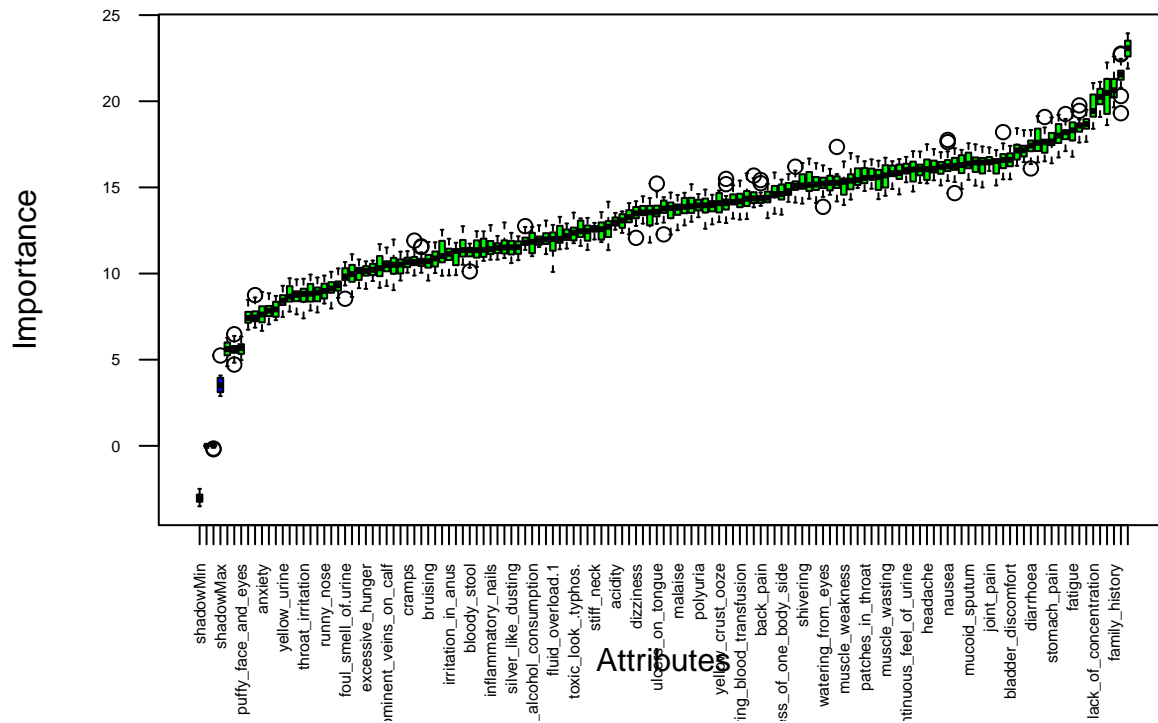
## After 18 iterations, +1.4 mins:

## confirmed 1 attribute: puffy_face_and_eyes;

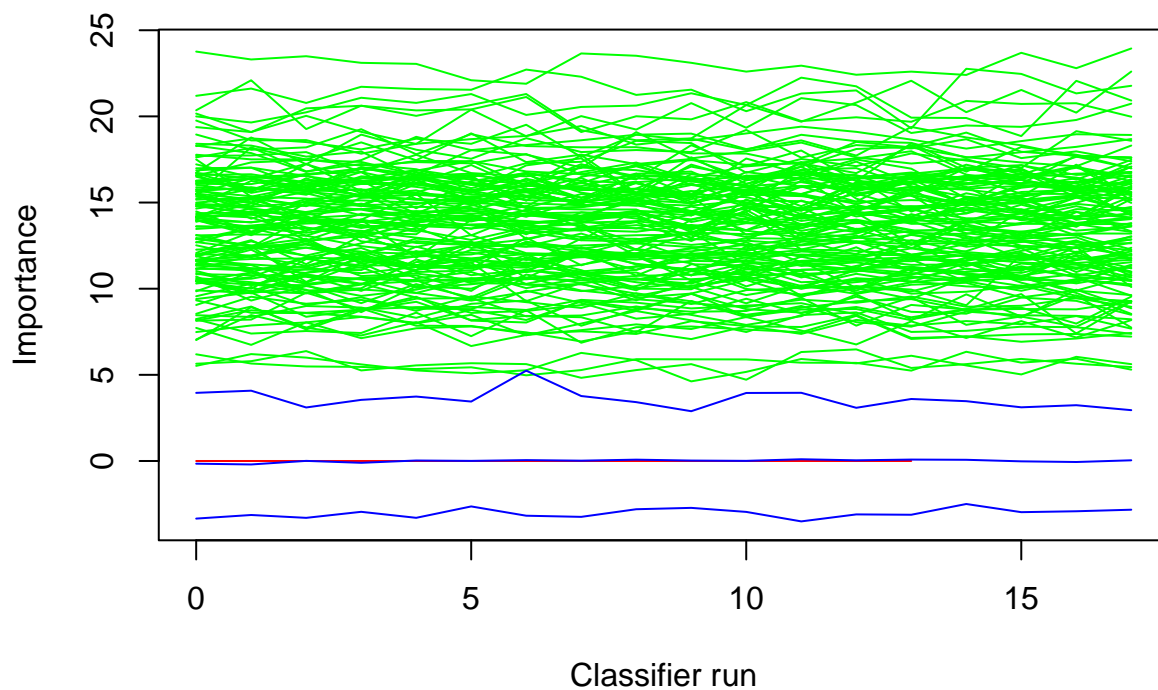
## no more attributes left.

plot(Boruta, las = 2, cex.axis = 0.5)

```



```
plotImpHistory(Boruta)
```



```
attStats(Boruta)
```

```
##               meanImp medianImp   minImp   maxImp
## itching      20.205122 20.255305 19.088752 21.125202
## skin_rash    14.687857 14.612328 13.471400 15.867553
```

## nodal_skin_eruptions	18.046187	18.007509	16.737949	19.212937
## continuous_sneezing	13.623133	13.538891	12.898583	14.486071
## shivering	15.289066	15.107457	14.061299	16.413457
## chills	14.870945	14.784199	14.162112	15.773883
## joint_pain	16.516806	16.470960	15.958053	17.320315
## stomach_pain	17.792447	17.671431	16.396868	19.058945
## acidity	13.015820	12.960748	12.544515	13.518716
## ulcers_on_tongue	13.694797	13.584306	12.889317	15.221930
## muscle_wasting	15.669884	15.696239	14.513739	16.721540
## vomiting	20.399151	20.487740	18.614565	22.248255
## burning_micturition	15.181647	15.175284	14.099351	16.411376
## spotting_.urination	17.685016	17.595413	16.453652	19.139541
## fatigue	18.207508	18.314792	16.779014	19.387773
## weight_gain	5.599838	5.612910	4.621594	6.274480
## anxiety	7.643983	7.626452	6.672173	8.930135
## cold_hands_and_feets	5.622558	5.622808	4.717834	6.476577
## mood_swings	7.474511	7.417686	6.741075	8.476038
## weight_loss	15.096487	15.094446	14.119256	16.201103
## restlessness	9.127263	9.136143	7.973190	10.102890
## lethargy	10.240826	10.214530	9.567898	10.756153
## patches_in_throat	15.634125	15.544447	14.787463	16.517632
## irregular_sugar_level	7.940304	7.920313	7.293936	8.703909
## cough	14.054976	14.087236	12.932815	15.494457
## high_fever	18.106121	18.200258	17.103848	19.255985
## sunken_eyes	16.236823	16.249102	14.666995	17.272204
## breathlessness	16.317504	16.444527	15.103378	17.500590
## sweating	16.313065	16.300652	14.913816	17.178280
## dehydration	15.845092	15.834670	14.594233	16.640620
## indigestion	11.496122	11.533146	10.775938	12.161114
## headache	16.145158	16.057377	15.236091	17.504047
## yellowish_skin	13.948319	13.905084	13.006408	15.182143
## dark_urine	17.189538	17.157915	16.469907	18.319764
## nausea	16.327940	16.225515	15.378590	17.754205
## loss_of_appetite	14.492596	14.361077	13.844740	15.683865
## pain_behind_the_eyes	16.138872	16.081900	15.410335	17.227718
## back_pain	14.417281	14.361880	13.931657	15.427047
## constipation	12.227932	12.183266	11.452068	12.903030
## abdominal_pain	15.279370	15.262140	14.572451	16.075340
## diarrhoea	17.315956	17.425682	16.100275	18.329986
## mild_fever	18.676055	18.678714	17.647450	19.514111
## yellow_urine	8.454498	8.421938	7.476941	9.289412
## yellowing_of_eyes	16.573832	16.596089	15.466303	18.206840
## acute_liver_failure	9.963223	9.967905	8.639921	11.122202
## fluid_overload	0.000000	0.000000	0.000000	0.000000
## swelling_of_stomach	11.499691	11.373280	10.737522	12.448024
## swelled_lymph_nodes	11.410447	11.362209	10.386636	12.455620
## malaise	13.721903	13.813762	12.515341	14.561331
## blurred_and_distorted_vision	9.332043	9.319725	8.596628	10.296197
## phlegm	11.913484	11.789089	11.375928	12.750886
## throat_irritation	8.761289	8.817528	7.399713	9.698219
## redness_of_eyes	8.821902	8.893422	7.745286	9.769752
## sinus_pressure	8.715955	8.665477	7.793139	9.726287
## runny_nose	9.004320	8.983501	8.260576	9.881359
## congestion	8.884158	8.828043	7.854155	9.941998

## chest_pain	17.184605	17.140626	16.253962	18.435625
## weakness_in_limbs	13.035959	13.078596	12.358488	13.602600
## fast_heart_rate	14.414107	14.401584	13.213824	15.201792
## pain_during_bowel_movements	10.900027	10.879197	10.027363	11.811599
## pain_in_anal_region	11.032559	11.309525	9.907723	11.785625
## bloody_stool	11.163199	11.360943	10.123528	11.734532
## irritation_in_anus	11.064031	11.141835	9.889266	12.019273
## neck_pain	15.713004	15.433796	14.967011	16.796518
## dizziness	13.443720	13.492474	12.071526	14.196190
## cramps	10.655583	10.663356	10.048511	11.300151
## bruising	10.693411	10.720675	9.592241	11.833389
## obesity	9.986171	10.143199	9.151278	10.816632
## swollen_legs	10.724181	10.681036	10.046512	11.576241
## swollen_blood_vessels	7.867412	7.844359	7.052510	8.854811
## puffy_face_and_eyes	5.636393	5.675927	4.973212	6.339953
## enlarged_thyroid	10.430402	10.504996	9.035734	11.975813
## brittle_nails	10.493181	10.523431	9.596620	11.269826
## swollen_extremities	10.425132	10.306166	9.171227	11.707818
## excessive_hunger	10.091864	10.153678	9.110888	10.706669
## extra_marital_contacts	15.307479	15.263276	14.187765	17.349163
## drying_and_tingling_lips	7.607925	7.452505	6.863179	8.749858
## slurred_speech	13.362931	13.553875	11.794823	14.377605
## knee_pain	12.625743	12.716100	11.359525	14.171124
## hip_joint_pain	12.603086	12.577032	11.824837	13.567874
## muscle_weakness	15.186200	15.315523	13.974821	16.505238
## stiff_neck	12.580109	12.566958	11.977757	12.941711
## swelling_joints	12.070423	11.978734	11.129363	13.075532
## movement_stiffness	13.245900	13.271972	12.274580	14.087172
## spinning_movements	12.513484	12.458010	11.602743	13.199110
## loss_of_balance	16.280082	16.524004	15.115403	17.167485
## unsteadiness	17.530084	17.620246	16.514808	19.085475
## weakness_of_one_body_side	14.636659	14.616553	13.416115	15.451447
## loss_of_smell	8.700061	8.766726	7.633280	9.661264
## bladder_discomfort	16.564354	16.598509	15.472464	17.409955
## foul_smell_of_urine	9.873646	9.826023	8.537531	10.666400
## continuous_feel_of_urine	16.051859	16.010428	15.296768	16.770063
## passage_of_gases	15.443631	15.389359	14.294165	16.997260
## internal_itching	15.475607	15.591822	14.315492	16.547014
## toxic_look_.typhos.	12.384441	12.370139	11.762782	13.537984
## depression	10.726836	10.669993	9.883358	11.911043
## irritability	12.247945	12.016205	11.409135	13.289802
## muscle_pain	23.020866	23.077932	21.894102	23.940344
## altered_sensorium	21.501727	21.547001	19.305500	22.762548
## red_spots_over_body	15.260059	15.149664	14.343851	16.673148
## belly_pain	12.360365	12.459704	11.370678	13.230159
## abnormal_menstruation	13.840982	13.986719	12.854193	14.570824
## dischromic_.patches	18.541540	18.582844	17.604486	19.759630
## watering_from_eyes	15.241204	15.225155	13.871586	16.279664
## increased_appetite	13.675523	13.800079	12.263150	14.700928
## polyuria	13.967749	13.954439	13.268863	14.614368
## family_history	20.747934	20.642540	19.646492	22.602127
## mucoid_sputum	16.485717	16.375826	15.441880	17.565464
## rusty_sputum	16.326319	16.446016	14.998539	17.375789
## lack_of_concentration	19.708693	19.432235	18.398341	21.061195

## visual_disturbances	11.085682	11.014722	9.872130	12.540892
## receiving_blood_transfusion	14.266945	14.187624	13.135380	15.398582
## receiving_unsterile_injections	14.331192	14.343640	13.159805	15.817755
## coma	15.951794	16.039159	14.760000	17.101353
## stomach_bleeding	16.146746	16.191078	15.115360	17.045835
## distention_of_abdomen	11.622757	11.511123	10.899158	12.965698
## history_of_alcohol_consumption	11.810475	11.842004	11.007505	12.706576
## fluid_overload.1	11.806383	12.002379	10.094805	12.820788
## blood_in_sputum	13.903652	13.960802	12.602411	14.873257
## prominent_veins_on_calf	10.474254	10.488302	9.274715	11.511526
## palpitations	13.736151	13.770773	12.273593	14.430074
## painful_walking	11.976898	11.905232	11.537846	12.802369
## pus_filled_pimples	15.868446	15.834287	14.903849	16.520320
## blackheads	16.030117	16.053630	15.259784	17.083785
## scurring	15.694316	15.548758	14.924326	16.626781
## skin_peeling	11.488694	11.492882	10.811451	12.181334
## silver_like_dusting	11.505301	11.520791	10.604971	12.361083
## small_dents_in_nails	11.477019	11.360780	10.514161	12.747409
## inflammatory_nails	11.532988	11.417046	10.804546	12.812734
## blister	14.262035	14.145707	13.190027	15.113263
## red_sore_around_nose	13.869620	13.868637	12.796019	14.685415
## yellow_crust_ooze	14.046284	14.085010	12.986261	14.814348
##	normHits	decision		
## itching	1.0000000	Confirmed		
## skin_rash	1.0000000	Confirmed		
## nodal_skin_eruptions	1.0000000	Confirmed		
## continuous_sneezing	1.0000000	Confirmed		
## shivering	1.0000000	Confirmed		
## chills	1.0000000	Confirmed		
## joint_pain	1.0000000	Confirmed		
## stomach_pain	1.0000000	Confirmed		
## acidity	1.0000000	Confirmed		
## ulcers_on_tongue	1.0000000	Confirmed		
## muscle_wasting	1.0000000	Confirmed		
## vomiting	1.0000000	Confirmed		
## burning_micturition	1.0000000	Confirmed		
## spotting_.urination	1.0000000	Confirmed		
## fatigue	1.0000000	Confirmed		
## weight_gain	1.0000000	Confirmed		
## anxiety	1.0000000	Confirmed		
## cold_hands_and_feets	1.0000000	Confirmed		
## mood_swings	1.0000000	Confirmed		
## weight_loss	1.0000000	Confirmed		
## restlessness	1.0000000	Confirmed		
## lethargy	1.0000000	Confirmed		
## patches_in_throat	1.0000000	Confirmed		
## irregular_sugar_level	1.0000000	Confirmed		
## cough	1.0000000	Confirmed		
## high_fever	1.0000000	Confirmed		
## sunken_eyes	1.0000000	Confirmed		
## breathlessness	1.0000000	Confirmed		
## sweating	1.0000000	Confirmed		
## dehydration	1.0000000	Confirmed		
## indigestion	1.0000000	Confirmed		

## headache	1.0000000	Confirmed
## yellowish_skin	1.0000000	Confirmed
## dark_urine	1.0000000	Confirmed
## nausea	1.0000000	Confirmed
## loss_of_appetite	1.0000000	Confirmed
## pain_behind_the_eyes	1.0000000	Confirmed
## back_pain	1.0000000	Confirmed
## constipation	1.0000000	Confirmed
## abdominal_pain	1.0000000	Confirmed
## diarrhoea	1.0000000	Confirmed
## mild_fever	1.0000000	Confirmed
## yellow_urine	1.0000000	Confirmed
## yellowing_of_eyes	1.0000000	Confirmed
## acute_liver_failure	1.0000000	Confirmed
## fluid_overload	0.0000000	Rejected
## swelling_of_stomach	1.0000000	Confirmed
## swelled_lymph_nodes	1.0000000	Confirmed
## malaise	1.0000000	Confirmed
## blurred_and_distorted_vision	1.0000000	Confirmed
## phlegm	1.0000000	Confirmed
## throat_irritation	1.0000000	Confirmed
## redness_of_eyes	1.0000000	Confirmed
## sinus_pressure	1.0000000	Confirmed
## runny_nose	1.0000000	Confirmed
## congestion	1.0000000	Confirmed
## chest_pain	1.0000000	Confirmed
## weakness_in_limbs	1.0000000	Confirmed
## fast_heart_rate	1.0000000	Confirmed
## pain_during_bowel_movements	1.0000000	Confirmed
## pain_in_anal_region	1.0000000	Confirmed
## bloody_stool	1.0000000	Confirmed
## irritation_in_anus	1.0000000	Confirmed
## neck_pain	1.0000000	Confirmed
## dizziness	1.0000000	Confirmed
## cramps	1.0000000	Confirmed
## bruising	1.0000000	Confirmed
## obesity	1.0000000	Confirmed
## swollen_legs	1.0000000	Confirmed
## swollen_blood_vessels	1.0000000	Confirmed
## puffy_face_and_eyes	0.9444444	Confirmed
## enlarged_thyroid	1.0000000	Confirmed
## brittle_nails	1.0000000	Confirmed
## swollen_extremities	1.0000000	Confirmed
## excessive_hunger	1.0000000	Confirmed
## extra_marital_contacts	1.0000000	Confirmed
## drying_and_tingling_lips	1.0000000	Confirmed
## slurred_speech	1.0000000	Confirmed
## knee_pain	1.0000000	Confirmed
## hip_joint_pain	1.0000000	Confirmed
## muscle_weakness	1.0000000	Confirmed
## stiff_neck	1.0000000	Confirmed
## swelling_joints	1.0000000	Confirmed
## movement_stiffness	1.0000000	Confirmed
## spinning_movements	1.0000000	Confirmed

## loss_of_balance	1.0000000	Confirmed
## unsteadiness	1.0000000	Confirmed
## weakness_of_one_body_side	1.0000000	Confirmed
## loss_of_smell	1.0000000	Confirmed
## bladder_discomfort	1.0000000	Confirmed
## foul_smell_of.urine	1.0000000	Confirmed
## continuous_feel_of_urine	1.0000000	Confirmed
## passage_of_gases	1.0000000	Confirmed
## internal_itching	1.0000000	Confirmed
## toxic_look_.typhos.	1.0000000	Confirmed
## depression	1.0000000	Confirmed
## irritability	1.0000000	Confirmed
## muscle_pain	1.0000000	Confirmed
## altered_sensorium	1.0000000	Confirmed
## red_spots_over_body	1.0000000	Confirmed
## belly_pain	1.0000000	Confirmed
## abnormal_menstruation	1.0000000	Confirmed
## dischromic_.patches	1.0000000	Confirmed
## watering_from_eyes	1.0000000	Confirmed
## increased_appetite	1.0000000	Confirmed
## polyuria	1.0000000	Confirmed
## family_history	1.0000000	Confirmed
## mucoid_sputum	1.0000000	Confirmed
## rusty_sputum	1.0000000	Confirmed
## lack_of_concentration	1.0000000	Confirmed
## visual_disturbances	1.0000000	Confirmed
## receiving_blood_transfusion	1.0000000	Confirmed
## receiving_unsterile_injections	1.0000000	Confirmed
## coma	1.0000000	Confirmed
## stomach_bleeding	1.0000000	Confirmed
## distention_of_abdomen	1.0000000	Confirmed
## history_of_alcohol_consumption	1.0000000	Confirmed
## fluid_overload.1	1.0000000	Confirmed
## blood_in_sputum	1.0000000	Confirmed
## prominent_veins_on_calf	1.0000000	Confirmed
## palpitations	1.0000000	Confirmed
## painful_walking	1.0000000	Confirmed
## pus_filled_pimples	1.0000000	Confirmed
## blackheads	1.0000000	Confirmed
## scurring	1.0000000	Confirmed
## skin_peeling	1.0000000	Confirmed
## silver_like_dusting	1.0000000	Confirmed
## small_dents_in_nails	1.0000000	Confirmed
## inflammatory_nails	1.0000000	Confirmed
## blister	1.0000000	Confirmed
## red_sore_around_nose	1.0000000	Confirmed
## yellow_crust_ooze	1.0000000	Confirmed

Boruta

```
## Boruta performed 18 iterations in 1.352149 mins.
## 131 attributes confirmed important: abdominal_pain,
## abnormal_menstruation, acidity, acute_liver_failure, altered_sensorium
## and 126 more;
```

```
## 1 attributes confirmed unimportant: fluid_overload;
```

Removing “fluid_overload” as it has been rejected in feature selection

```
Disease <- Disease1 %>% select(-fluid_overload)
```

Unique Diseases(prognosis) in the dataset.

```
unique(Disease$prognosis)
```

```
## [1] Fungal infection
## [2] Allergy
## [3] GERD
## [4] Chronic cholestasis
## [5] Drug Reaction
## [6] Peptic ulcer disease
## [7] AIDS
## [8] Diabetes
## [9] Gastroenteritis
## [10] Bronchial Asthma
## [11] Hypertension
## [12] Migraine
## [13] Cervical spondylosis
## [14] Paralysis (brain hemorrhage)
## [15] Jaundice
## [16] Malaria
## [17] Chicken pox
## [18] Dengue
## [19] Typhoid
## [20] hepatitis A
## [21] Hepatitis B
## [22] Hepatitis C
## [23] Hepatitis D
## [24] Hepatitis E
## [25] Alcoholic hepatitis
## [26] Tuberculosis
## [27] Common Cold
## [28] Pneumonia
## [29] Dimorphic hemmorhoids(piles)
## [30] Heart attack
## [31] Varicose veins
## [32] Hypothyroidism
## [33] Hyperthyroidism
## [34] Hypoglycemia
## [35] Osteoarthritis
## [36] Arthritis
## [37] (vertigo) Paroymsal Positional Vertigo
## [38] Acne
## [39] Urinary tract infection
```

```
## [40] Psoriasis
## [41] Impetigo
## 41 Levels: (vertigo) Paroymsal Positional Vertigo Acne ... Varicose veins
```

Splitting the Data

We can now split the data into 70% training and 30% testing data. I used createDataPartition() function of library caret for random splitting resulting in balanced outcome classes.

```
## Training and Testing Dataset

Train_index <- createDataPartition(Disease$prognosis, p = .7, list = FALSE, times = 1)

train <- Disease[ Train_index,]
test <- Disease[-Train_index,]
```

E.Modelling and Results

After splitting the data into training and testing data, we will apply various machine learning algorithms to make various models, and compare them using various metrics.

Decision Tree

Decision Trees are a type of Supervised Machine Learning (that is you explain what the input is and what the corresponding output is in the training data) where the data is continuously split according to a certain parameter. The tree can be explained by two entities, namely decision nodes and leaves. The leaves are the decisions or the final outcomes. And the decision nodes are where the data is split.

I trained the decision tree on the training dataset using all variables of symptoms as predictors and prognosis as response variable. We also did pruning of tree to make the better visualized tree.

```
Disease_tree <- rpart(prognosis ~ ., data = train, method = "class")

# Plot the tree
rpart.plot(Disease_tree, extra = 0, yesno = TRUE)
```

```
## Warning: All boxes will be white (the box.palette argument will be ignored) because
## the number of classes in the response 41 is greater than length(box.palette) 6.
## To silence this warning use box.palette=0 or trace=-1.
```

```
## Warning: labs do not fit even at cex 0.15, there may be some overplotting
```



```

## Specificity

# Initialize sums
sum_TN <- 0
sum_FP <- 0

# Calculate the sum of true negatives and false positives for each class
for (i in 1:nrow(table)) {
  TN <- sum(table[-i, -i])
  FP <- sum(table[-i, i])
  sum_TN <- sum_TN + TN
  sum_FP <- sum_FP + FP
}

# Calculate Micro-averaged Specificity
Micro_Specificity <- sum_TN / (sum_TN + sum_FP)

cat(paste("Specificity = ", Micro_Specificity), "\n")

```

```
## Specificity = 0.996341463414634
```

Results

The decision tree model achieved the accuracy of 89.70%

Naive Bayes

```

## Naive Bayes
library(e1071)

model <- naiveBayes(prognosis ~ ., data = train)
print("Prediction on Test data")

## [1] "Prediction on Test data"

predictions <- predict(model, newdata = test)

CM_dt1 <- confusionMatrix(predictions, test$prognosis)
CM_dt1$overall

```

```

##      Accuracy      Kappa  AccuracyLower  AccuracyUpper  AccuracyNull
## 1.00000000 1.00000000 0.99750388 1.00000000 0.02439024
## AccuracyPValue McNemarPValue
## 0.00000000      NaN

```

```

# Extract the table from the confusion matrix
table <- as.table(CM_dt1$table)

```

```
## Sensitivity
```

```

# Calculate the sum of true positives
TP <- sum(diag(table))

# Calculate the sum of false negatives
FN <- sum(rowSums(table)) - TP

# Calculate Micro-averaged Recall
Micro_Recall <- TP / (TP + FN)

# Print Micro-averaged Recall
cat(paste("Sensitivity = ", Micro_Recall), "\n")

```

```
## Sensitivity = 1
```

```

## Specificity

# Initialize sums
sum_TN <- 0
sum_FP <- 0

# Calculate the sum of true negatives and false positives for each class
for (i in 1:nrow(table)) {
  TN <- sum(table[-i, -i])
  FP <- sum(table[-i, i])
  sum_TN <- sum_TN + TN
  sum_FP <- sum_FP + FP
}

# Calculate Micro-averaged Specificity
Micro_Specificity <- sum_TN / (sum_TN + sum_FP)

cat(paste("Specificity = ", Micro_Specificity), "\n")

```

```
## Specificity = 1
```

Results

The Naive Bayes model has achieved an accuracy of 1 which is far better than that of decision tree model.

Support Vector Machine

```

### Support Vector Machine

model_svm <- svm(prognosis ~ ., data = train)

predictions_svm <- predict(model_svm, newdata = test)

CM_dt1 <- confusionMatrix(predictions_svm, test$prognosis)
CM_dt1$overall

```

```
##      Accuracy      Kappa AccuracyLower AccuracyUpper AccuracyNull
##      1.00000000      1.00000000      0.99750388      1.00000000      0.02439024
## AccuracyPValue McNemarPValue
##      0.00000000      NaN
```

```
# Extract the table from the confusion matrix
table <- as.table(CM_dt1$table)

## Sensitivity

# Calculate the sum of true positives
TP <- sum(diag(table))

# Calculate the sum of false negatives
FN <- sum(rowSums(table)) - TP

# Calculate Micro-averaged Recall
Micro_Recall <- TP / (TP + FN)

# Print Micro-averaged Recall
cat(paste("Sensitivity = ", Micro_Recall), "\n")
```

```
## Sensitivity = 1
```

```
## Specificity

# Initialize sums
sum_TN <- 0
sum_FP <- 0

# Calculate the sum of true negatives and false positives for each class
for (i in 1:nrow(table)) {
  TN <- sum(table[-i, -i])
  FP <- sum(table[-i, i])
  sum_TN <- sum_TN + TN
  sum_FP <- sum_FP + FP
}

# Calculate Micro-averaged Specificity
Micro_Specificity <- sum_TN / (sum_TN + sum_FP)

cat(paste("Specificity = ", Micro_Specificity), "\n")
```

```
## Specificity = 1
```

Results of Support Vector Machine

The SVM model also achieved accuracy of 1 which is same as Naive Bayes

F. Summary

After, comparing the models, it is shown that SVM and Naive Bayes are better performing than Decision tree in terms of accuracy, specificity and sensitivity.

However, in real world scenario, it is not practically possible to achieve an accuracy of 1. So it is derived that the models are overfitting due to unavailability of accurate data.