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# INTRODUCTION

- In 2008, heart disease and stroke were responsible for nearly 30.4% in united states. Coronary heart disease is the cause of more than 2/3 of these deaths.
- Coronary artery disease develops when the major blood vessels that supply your heart with blood, oxygen and nutrients (coronary arteries) become damaged or diseased.
- CSD is affected by many factors such as blood pressure, anxiety, cholesterol, diabetes, obesity, smoking , alcohol consumption , lack of physical activities etc.



# DATASET

- In our dataset we have considered the factors given below:
  - ☐ Systolic blood pressure
  - ☐ Tobacco cumulative (kg)
  - ☐ LDL - low density lipoprotein cholesterol
  - ☐ Adiposity - famihist - family history of heart disease (Present, Absent)
  - ☐ Obesity
  - ☐ Alcohol current alcohol consumption - age at onset
  - ☐ CHD

In this project, we are analyzing the most and least significant attributes that helps in predicting whether a person is having CHD or not.

# REPLACING MISSING VALUES

- Using `proc freq`, we find missing values in each of the attributes.
- There are missing values in features tobacco, ldl, obesity, alcohol.
- These missing values are replaced by the mean and median of the corresponding attributes.

The FREQ Procedure

famhist	Frequency
Not Missing	462

sbp	Frequency
Not Missing	462

tobacco	Frequency
Missing	4
Not Missing	458

ldl	Frequency
Missing	5
Not Missing	457

adiposity	Frequency
Missing	1
Not Missing	461

typea	Frequency
Not Missing	462

obesity	Frequency
Missing	2
Not Missing	460

alcohol	Frequency
Missing	6
Not Missing	456

# CORRELATION BETWEEN FEATURES

- Using *proc corr*, we find correlation between each of the attributes.

Pearson Correlation Coefficients, N = 462 Prob >  r  under H0: Rho=0									
	sbp	tobacco	ldl	adiposity	typea	obesity	alcohol	age	chd
sbp	1.00000 1.00000	0.21225 <.0001	0.15830 0.0006	0.35650 <.0001	-0.05745 0.2177	0.23807 <.0001	0.14010 0.0025	0.38877 <.0001	0.19235 <.0001
tobacco	0.21225 <.0001	1.00000	0.15891 0.0006	0.28664 <.0001	-0.01461 0.7542	0.12453 0.0074	0.20081 <.0001	0.45033 <.0001	0.29972 <.0001
ldl	0.15830 0.0006	0.15891 0.0006	1.00000	0.44043 <.0001	0.04405 0.3448	0.33051 <.0001	-0.03340 0.4738	0.31180 <.0001	0.26305 <.0001
adiposity	0.35650 <.0001	0.28664 <.0001	0.44043 <.0001	1.00000	-0.04314 0.3548	0.71656 <.0001	0.10033 0.0311	0.62595 <.0001	0.25412 <.0001
typea	-0.05745 0.2177	-0.01461 0.7542	0.04405 0.3448	-0.04314 0.3548	1.00000	0.07401 0.1122	0.03950 0.3970	-0.10261 0.0274	0.10316 0.0266
obesity	0.23807 <.0001	0.12453 0.0074	0.33051 <.0001	0.71656 <.0001	0.07401 0.1122	1.00000	0.05162 0.2682	0.29178 <.0001	0.10010 0.0315
alcohol	0.14010 0.0025	0.20081 <.0001	-0.03340 0.4738	0.10033 0.0311	0.03950 0.3970	0.05162 0.2682	1.00000	0.10112 0.0298	0.06253 0.1797
age	0.38877 <.0001	0.45033 <.0001	0.31180 <.0001	0.62595 <.0001	-0.10261 0.0274	0.29178 <.0001	0.10112 0.0298	1.00000	0.37297 <.0001
chd	0.19235 <.0001	0.29972 <.0001	0.26305 <.0001	0.25412 <.0001	0.10316 0.0266	0.10010 0.0315	0.06253 0.1797	0.37297 <.0001	1.00000

# REGRESSION ANALYSIS

- Performed stepwise regression analysis using `proc reg`, with `chd` as dependent variable and `sbp`, `tobacco`, `ldl`, `adiposity`, `typea`, `obesity`, `alcohol`, `age`.

Summary of Stepwise Selection								
Step	Variable Entered	Variable Removed	Number Vars In	Partial R-Square	Model R-Square	C(p)	F Value	Pr > F
1	age		1	0.1391	0.1391	33.2702	74.33	<.0001
2	ldl		2	0.0239	0.1630	21.6556	13.08	0.0003
3	tobacco		3	0.0208	0.1838	11.7841	11.67	0.0007
4	typea		4	0.0157	0.1995	4.8181	8.97	0.0029

- We can see that even with `age`, `ldl`, `tobacco` and `typea` entering the model, the maximum variance is only 0.19995.

# PRINCIPAL COMPONENT ANALYSIS

Eigenvalues of the Correlation Matrix				
	Eigenvalue	Difference	Proportion	Cumulative
1	2.99331083	1.77428622	0.3326	0.3326
2	1.21902460	0.12638231	0.1354	0.4680
3	1.09264229	0.09243960	0.1214	0.5894
4	1.00020269	0.23238132	0.1111	0.7006
5	0.76782137	0.07851260	0.0853	0.7859
6	0.68930877	0.08873476	0.0766	0.8625
7	0.60057401	0.13886722	0.0667	0.9292
8	0.46170679	0.28629813	0.0513	0.9805
9	0.17540865		0.0195	1.0000

- Based on Eigenvalue correlation matrix, we select 4 Principal components as their cumulative Eigenvalue is greater than 70% for the first 4 principal components.





# PRINCIPAL COMPONENT ANALYSIS

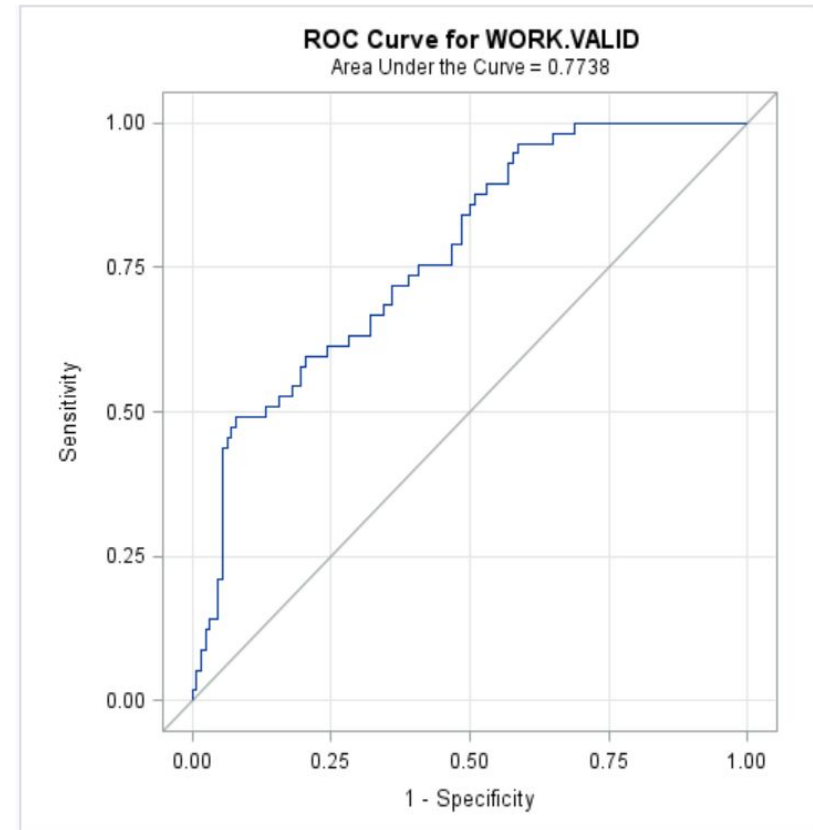
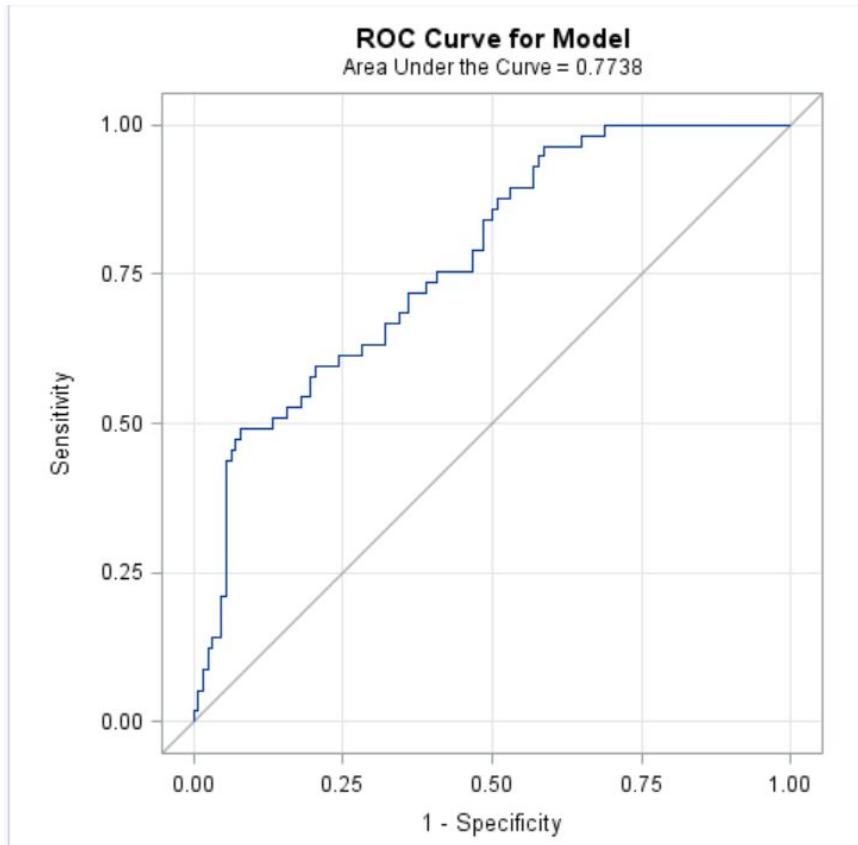
Using PCA on the entire data, we get the below

Eigenvectors									
	Prin1	Prin2	Prin3	Prin4	Prin5	Prin6	Prin7	Prin8	Prin9
sbp	0.318227	0.163177	-.213102	0.154204	0.819826	0.141897	0.268586	0.195882	-.012510
tobacco	0.313298	0.477471	0.059609	-.126550	-.314835	-.494019	0.388476	0.397794	-.044361
ldl	0.329548	-.318762	0.167212	-.247413	-.254987	0.579215	0.543091	0.014290	0.070386
adiposity	0.495637	-.263212	-.109376	0.147778	-.096747	-.113343	-.176545	-.140733	-.760317
typea	-.007083	-.111074	0.860617	0.214080	0.246836	-.254088	0.169562	-.212419	-.041945
obesity	0.374465	-.471809	-.008197	0.355974	-.088370	-.190544	-.253304	0.383013	0.505381
alcohol	0.115181	0.491556	0.093281	0.684144	-.273018	0.423157	-.079607	-.086465	0.030615
age	0.455661	0.166752	-.139171	-.173712	0.011971	-.183447	-.045924	-.722791	0.395842
chd	0.291804	0.264822	0.377244	-.454288	0.104725	0.271827	-.591161	0.248308	-.006564

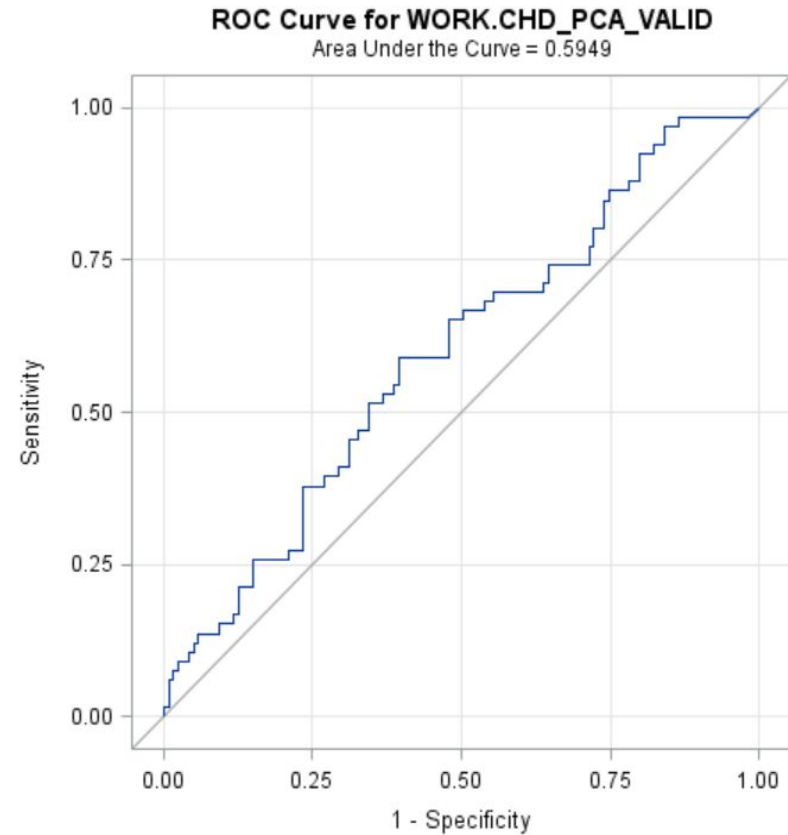
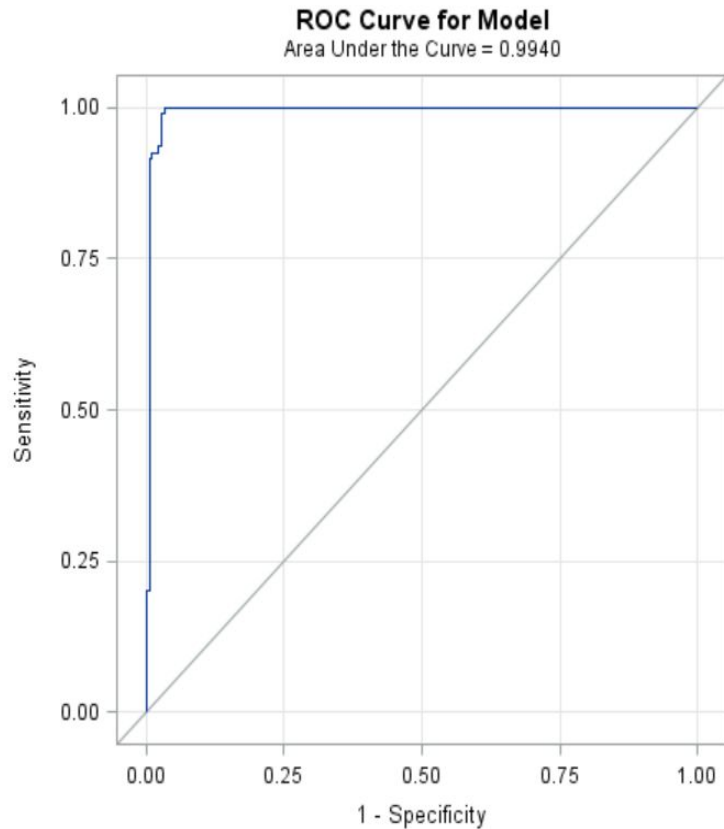
- PC1 has max loading for adiposity and age
- PC2 has max loading for tobacco, alcohol and obesity
- PC3 has max loading for typea and chd
- PC4 has max loading for alcohol and chd



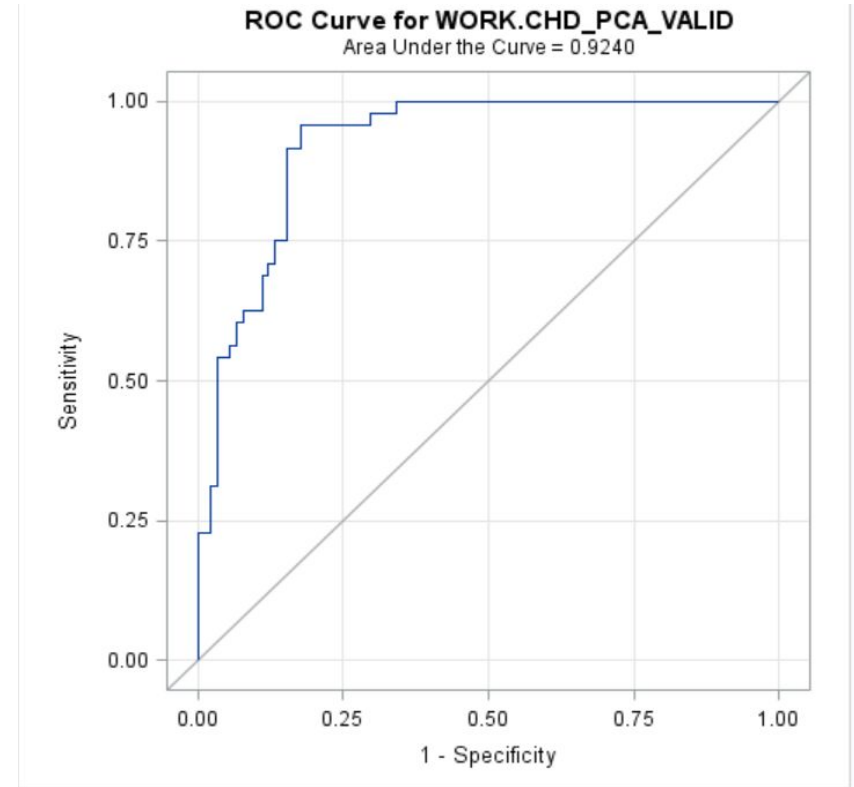
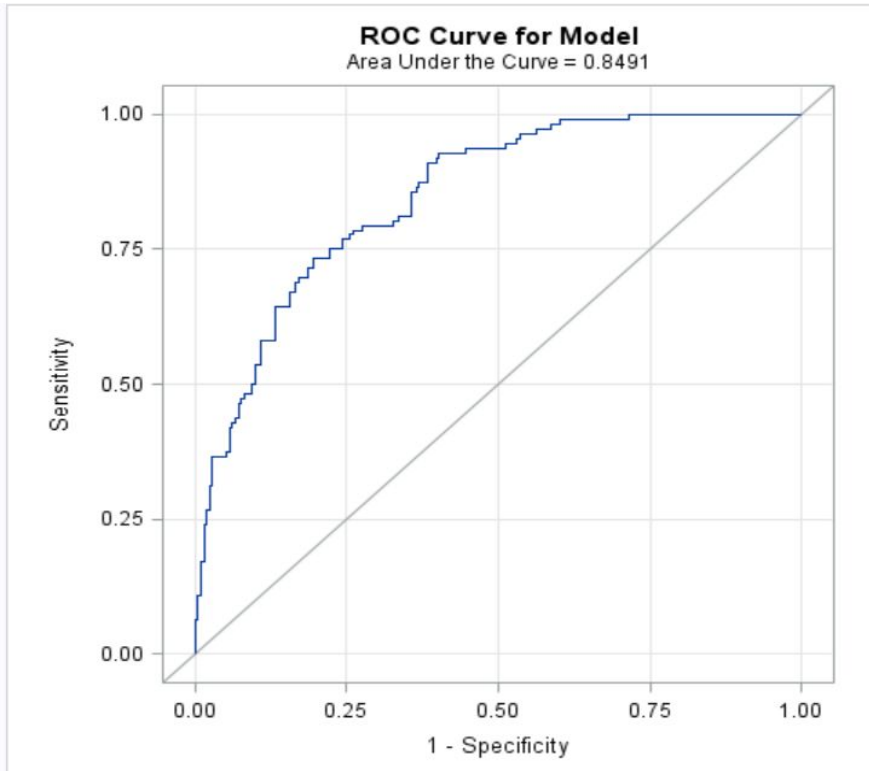
# ROC without PCA



# ROC With PCA (Overfitting)

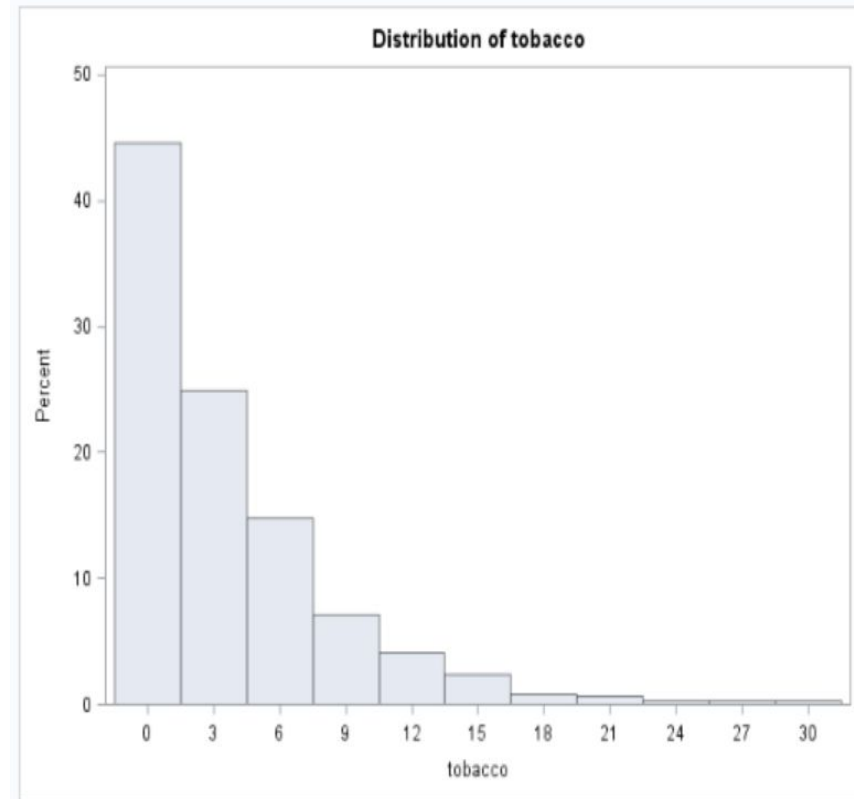
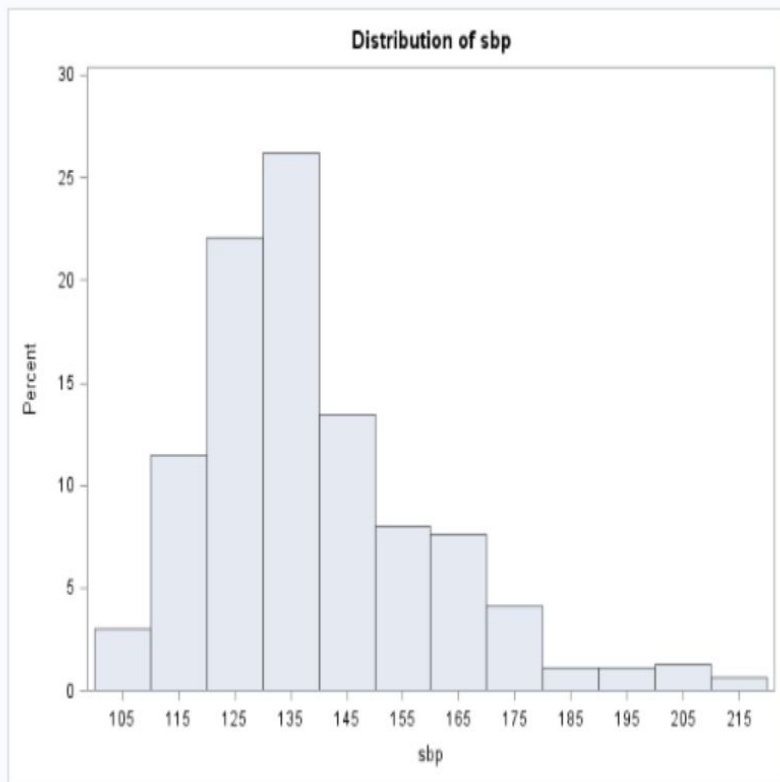


# Cross- Validated ROC

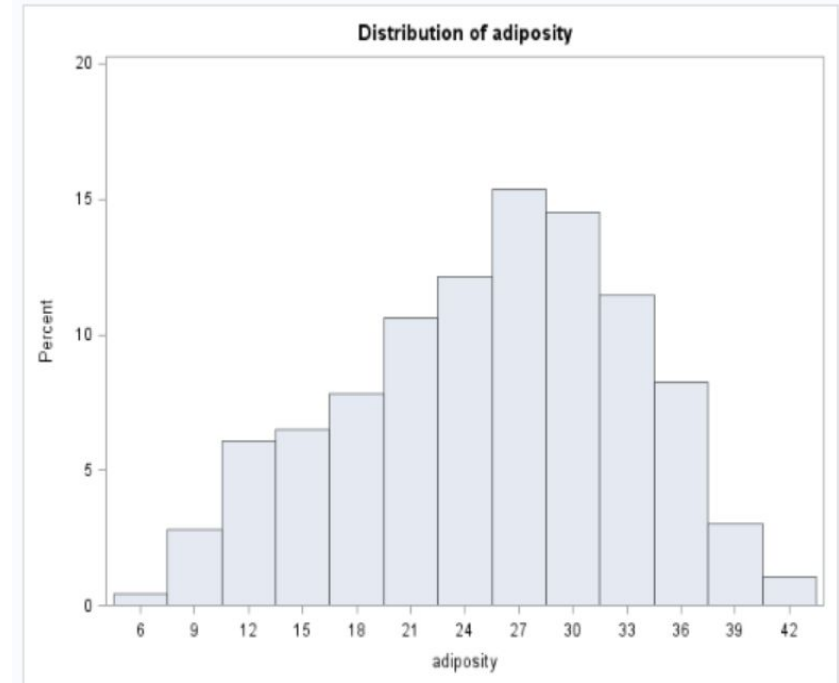
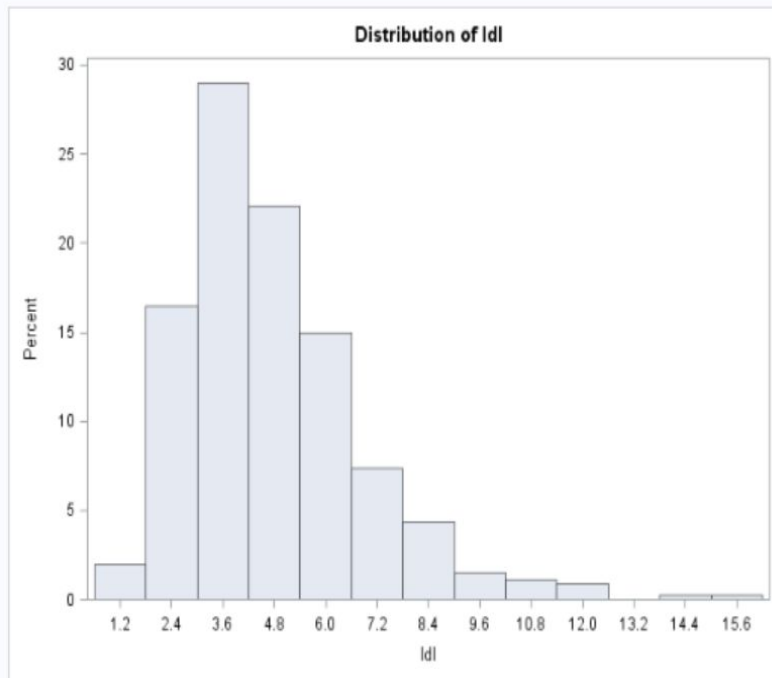


# DISTRIBUTION OF FEATURES

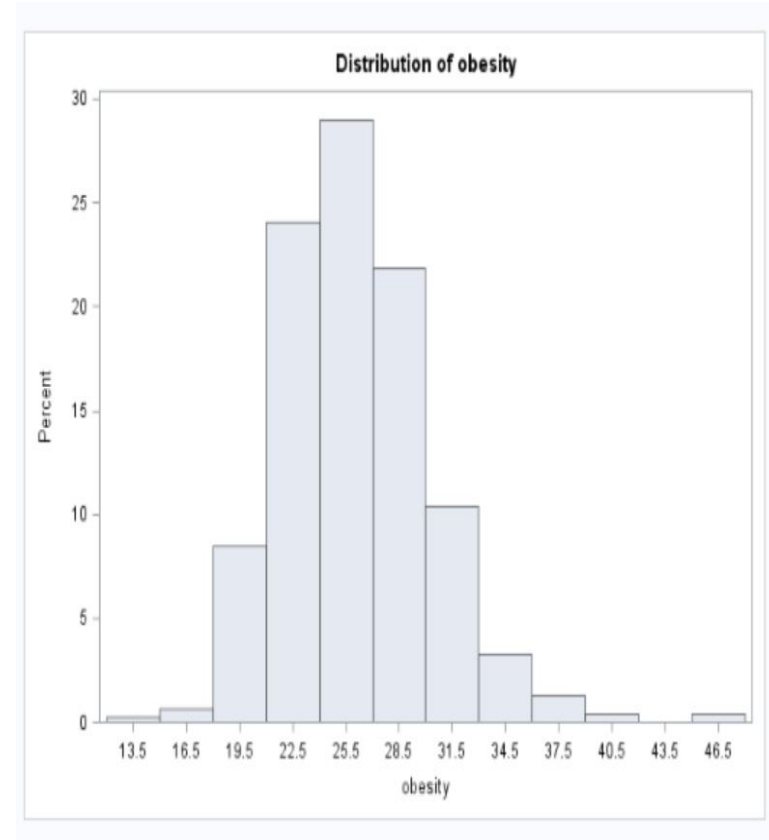
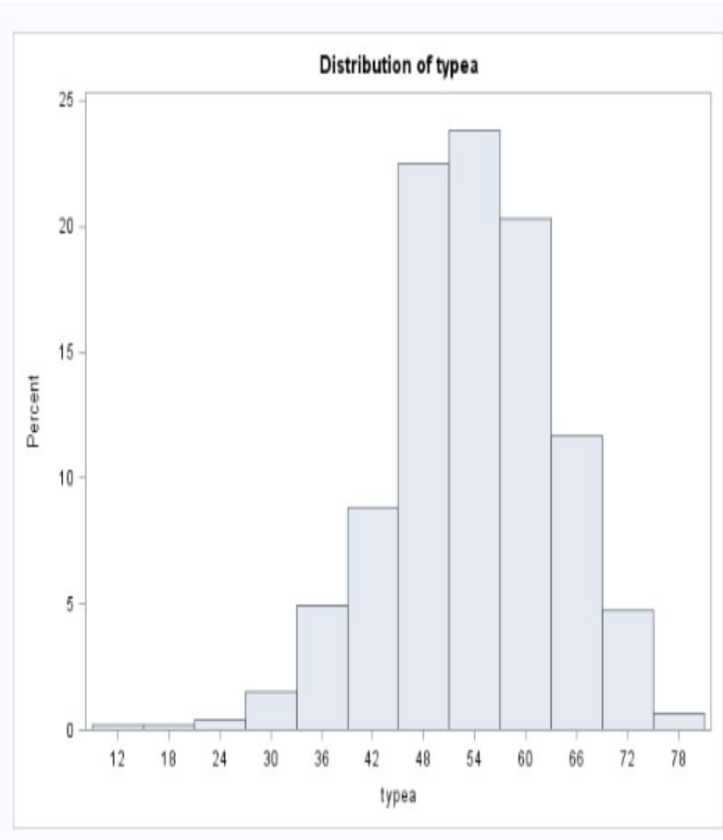
- *Proc univariate* is used to study whether each features are normally distributed



# DISTRIBUTION OF FEATURES

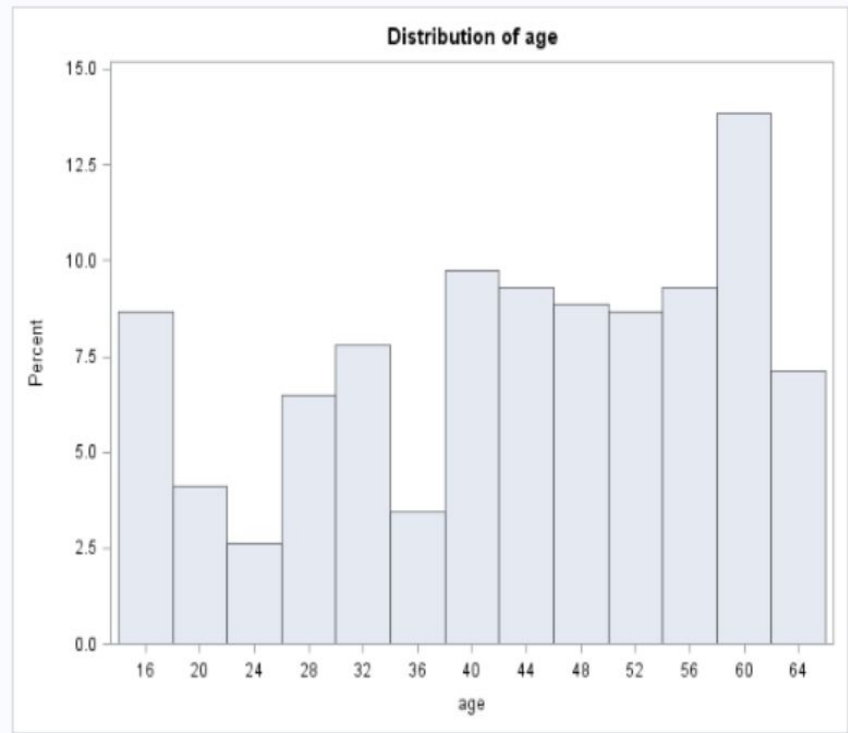
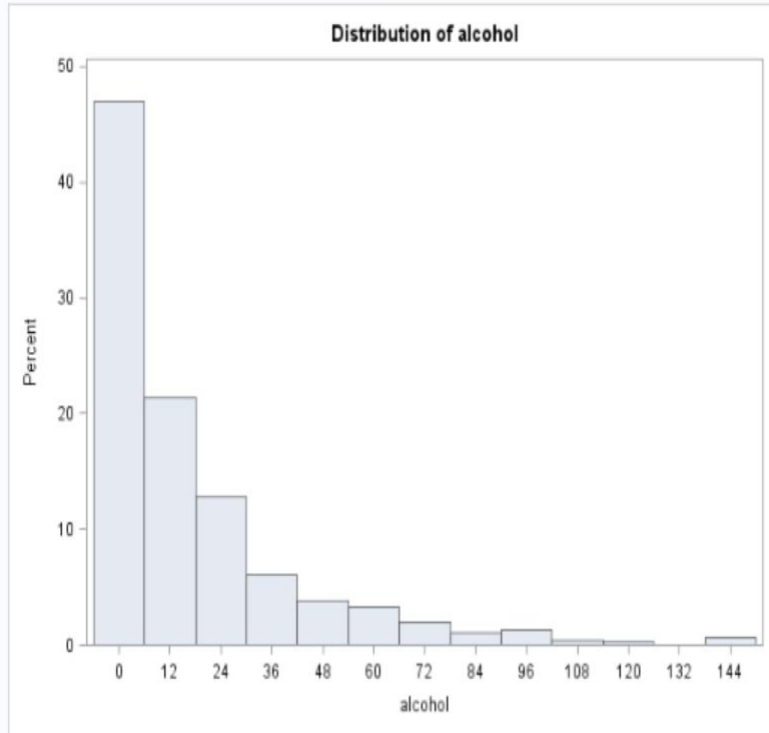


# DISTRIBUTION OF FEATURES





# DISTRIBUTION OF FEATURES



- We can see from the histograms that none of the features are normally distributed as they are either left or right skewed.



# RANDOM FOREST CLASSIFICATION

- All missing values in the datasets are initially replaced with the mean of the corresponding attributes.
- As the features are not normally distributed, logarithmic transformation is performed on each of the 7 attributes.

```
128  /*Transforming data since not normally distributed*/
129  data std_chd_n;
130  set chd1;
131  sbp_n = log10(sbp);
132  tobacco_n = log10(tobacco);
133  ldl_n = log10(ldl);
134  adiposity_n = log10(adiposity);
135  typea_n = log10(typea);
136  alcohol_n = log10(alcohol);
137  age_n = log10(age);
138  run;
```

# RANDOM FOREST CLASSIFICATION

- Used *proc hpforest* on the dataset where missing values are replaced by their mean value.

Fit Statistics					
Number of Trees	Number of Leaves	Average Square Error (Full Data)	Average Square Error (OOB)	Misclassification Rate (Full Data)	Misclassification Rate (OOB)
1	2	0.216	0.253	0.346	0.395
2	4	0.215	0.250	0.346	0.357
3	7	0.211	0.242	0.325	0.370
4	10	0.201	0.230	0.327	0.353
5	13	0.195	0.223	0.331	0.353
6	16	0.200	0.224	0.335	0.339
7	20	0.199	0.227	0.335	0.344
8	22	0.202	0.228	0.335	0.346
9	25	0.200	0.223	0.335	0.336
10	28	0.197	0.222	0.335	0.338
11	31	0.195	0.218	0.335	0.330
12	34	0.192	0.217	0.331	0.343
13	38	0.191	0.217	0.329	0.347
14	41	0.192	0.219	0.329	0.342
15	44	0.190	0.217	0.325	0.346
16	46	0.191	0.217	0.325	0.338

The misclassification rate is lowest at 0.325 and the optimal number of trees in our model is 15

# RANDOM FOREST CLASSIFICATION

Loss Reduction Variable Importance					
Variable	Number of Rules	Gini	OOB Gini	Margin	OOB Margin
adiposity_n	0	0.000000	0.000000	0.000000	0.000000
ldl_n	0	0.000000	0.000000	0.000000	0.000000
alcohol_n	6	0.000428	-0.00087	0.000856	0.00174
age_n	38	0.042035	-0.00136	0.084069	0.01597
sbp_n	16	0.007512	-0.00416	0.015024	-0.00390
typea_n	6	0.002226	-0.00452	0.004452	-0.00261
tobacco_n	31	0.005191	-0.01911	0.010381	0.04427

- The above variables in order contributes the most in prediction of chd in individuals.



# Naive Bayes

The sampling used is Simple random Sampling.

Test and train in the ratio 30- 70

Used Seed in R for Random Sampling.

```
      chd
NBayes_all 0  1
0  231  59
1   71 101
```

```
> NB_error_rate
[1] 0.2813852814
> |
```



# Conclusion

Analyzing various output from Different Classifiers.

- Logistic Regression - Error rate - 20%
- Random Forest - Error Rate- 33%
- Naive Bayes - Error rate- 28 % with high false positive

Hence logistic regression is the best classifier for our dataset. since the data is related to healthcare we need minimum false positive.