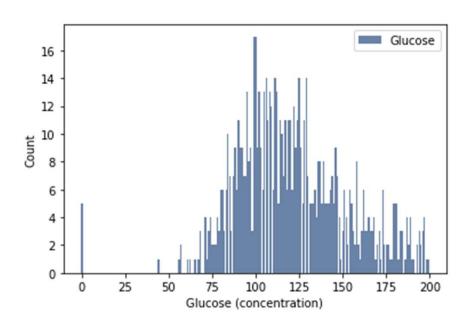


### Introduction

- Aims to identify diagnostic parameters that can be used to predict diabetes.
- •Are other variables correlated?
- •Utilizes data from the National Institute of Diabetes and Digestive and Kidney Diseases, India (Source: <a href="Pima Indians Diabetes Database">Pima Indians Diabetes Database</a> | Kaggle).
  - Dataset included 8 predictor variables such as number of times pregnant, Glucose levels, Blood Pressure, Skin Thickness, Insulin, BMI, Diabetes pedigree function, and Age. Outcome variable was Boolean (O for diabetes and 1 for non-diabetic).

### Selected Variables

- •Following variables were selected in the analysis based on generally associated diagnostic parameters with existing diabetic cases:
  - Glucose Plasma glucose concentration at 2 hours in an oral glucose tolerance test.
  - BloodPressure Diastolic blood pressure (mm Hg)
  - Insulin 2-Hour serum insulin (mu U/ml)
  - BMI Body mass index (weight in kg/(height in m)^2)
  - Age Age in years

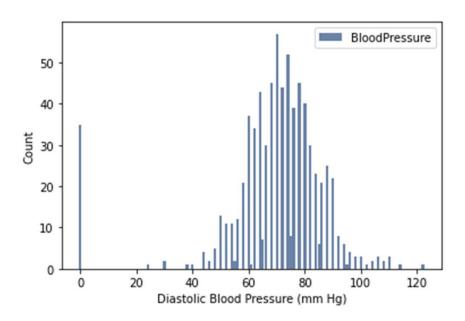


```
df.Glucose.describe()
count
          768.000000
mean
          120.894531
          31.972618
std
min
            0.000000
25%
          99.000000
50%
         117.000000
75%
          140.250000
          199.000000
Name: Glucose, dtype: float64
df.Glucose.mode()
      99
     100
Name: Glucose, dtype: int64
df.Glucose.tail()
763
       101
764
       122
765
       121
766
       126
767
Name: Glucose, dtype: int64
```

### Glucose

Glucose levels below 70 are outliers as glucose levels are typically over 70 for general population.

A check on the frequency of occurrence produced 16 data points out of 768 below 70.

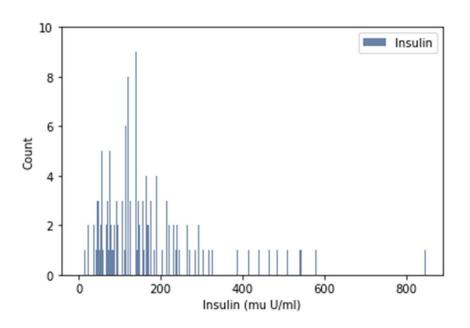


```
df.BloodPressure.describe()
         768.000000
count
          69.105469
mean
std
          19.355807
min
           0.000000
25%
          62.000000
50%
          72.000000
75%
          80.000000
         122.000000
Name: BloodPressure, dtype: float64
df.BloodPressure.mode()
     70
Name: BloodPressure, dtype: int64
df.BloodPressure.tail()
763
       76
764
       70
765
       72
766
       60
767
       70
Name: BloodPressure, dtype: int64
```

#### Blood Pressure

Blood Pressure below 50 and over 100 are outliers as these typically are considered too low or too high for diastolic levels.

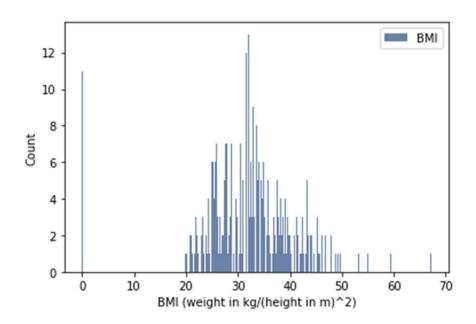
Latest and smallest 20 numbers indicated single digit frequencies.



```
df.Insulin.describe()
count
         768.000000
          79.799479
mean
std
         115.244002
            0.000000
min
25%
           0.000000
50%
          30.500000
         127.250000
75%
         846.000000
Name: Insulin, dtype: float64
df.Insulin.mode()
Name: Insulin, dtype: int64
df.Insulin.tail()
763
       180
764
         0
765
       112
766
767
Name: Insulin, dtype: int64
```

Insulin

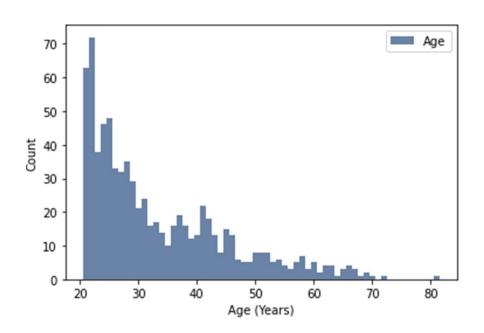
Insulin test results less than 50 and higher than 600 were considered outliers as these typically fall under too low and too high for physically active individuals.

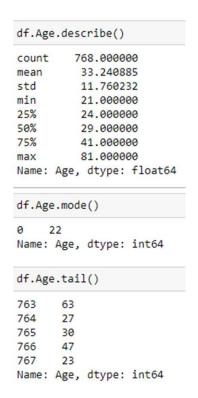


```
df.BMI.describe()
         768.000000
count
mean
          31.992578
           7.884160
std
           0.000000
min
25%
          27.300000
50%
          32.000000
75%
          36.600000
          67.100000
max
Name: BMI, dtype: float64
df.BMI.mode()
     32.0
Name: BMI, dtype: float64
df.BMI.tail()
763
       32.9
764
       36.8
765
       26.2
766
       30.1
767
       30.4
Name: BMI, dtype: float64
```

### Histogram - BMI

BMI less than 18.2 are outliers and only zero showed as the value which is not a correct representation for BMI,

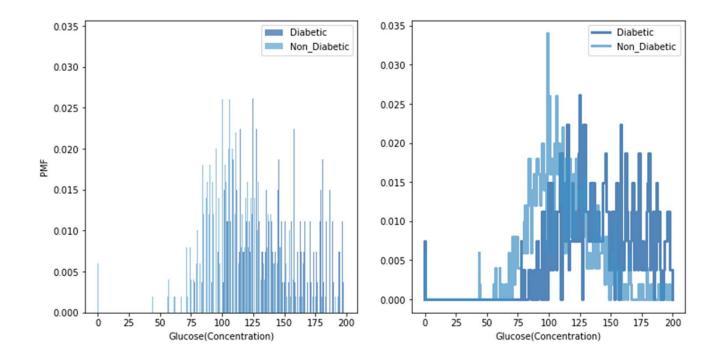




Histogram - Age

No outliers were observed as 20 to 80+ years can have diabetic people.

## PMF - Glucose

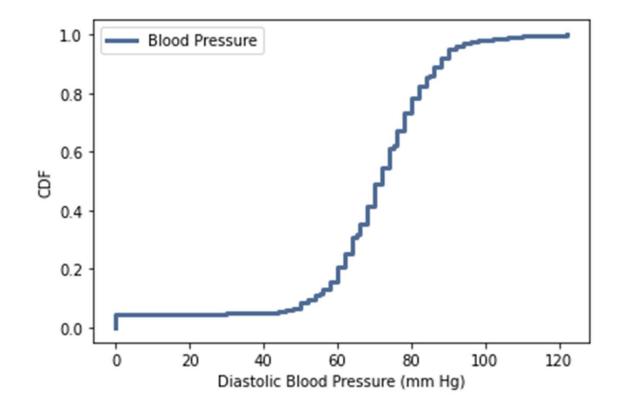


### CDF – Blood Pressure

Very few people have their diastolic Blood Pressure below 50. Therefore, it is flat.

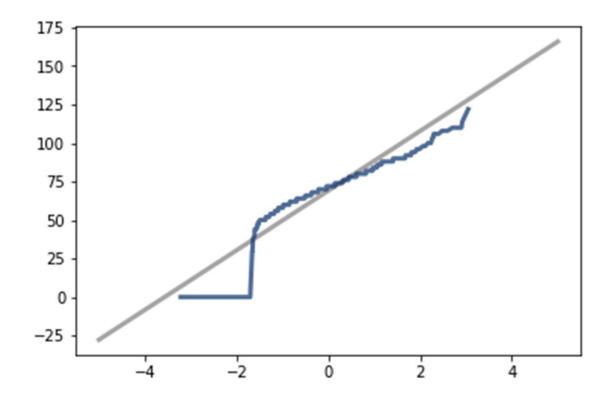
About 80% of the people have their diastolic Blood Pressure in the normal range, i.e., 80 mm Hg per recommended limit.

Therefore about 20% of the people are expected to have diabetes.



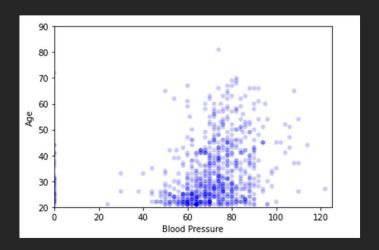
# Normal Distribution – Blood Pressure

The normal probability plot shows data is almost normal as it matches the model seen as the gray line



## Scatterplot – Blood Pressure / Age

Blood Pressure and Age are +vely correlated but strength of relationship is weak. It can be non-linearly correlated.



Cov(df.BloodPressure, df.Age)

54.45245869954427

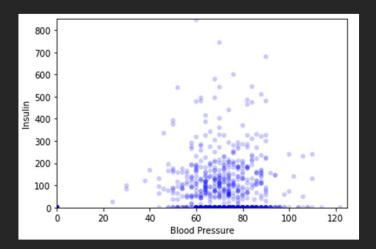
Corr(df.BloodPressure, df.Age)

0.23952794642136355

# Scatterplot – Insulin / Blood Pressure

Blood Pressure and Insulin levels seem to be +vely correlated but strength of relationship is very less.

It can be non-linearly correlated.



Cov(df.BloodPressure, df.Insulin)

198.12010701497397

Corr(df.BloodPressure, df.Insulin)

0.08893337837319301

# Hypothesis Test

- •Difference in means for BMI in diabetic and non-diabetic people was tested.
- •pvalue indicated that we expect to see a difference as big as the observed effect about 0% of the time.
- Result is statistically significant.

	coef	std err	Z	P> z	[0.025	0.975]
Intercept	-8.0932	0.692	-11.698	0.000	-9.449	-6.737
df.Glucose	0.0344	0.004	9.621	0.000	0.027	0.041
df.BloodPressure	-0.0122	0.005	-2.388	0.017	-0.022	-0.002
df.Insulin	-0.0010	0.001	-1.196	0.232	-0.003	0.001
df.BMI	0.0908	0.014	6.404	0.000	0.063	0.119
df.Age	0.0331	0.008	4.133	0.000	0.017	0.049

# Regression Analysis

Logistical Regression was performed for predicting diabetes with glucose, blood pressure, insulin, BMI, and Age as predictor variables.

As evident from the pvalues Glucose, BloodPressure, BMI, and Age are statistically significant in predicting diabetes if we consider alpha to be 5%.

# The End