

Problem 1.

Installed MatLab.

Problem 2.

a.)

Attribute 1: minimum = 0, maximum = 17, range = 17

Attribute 2: minimum = 0, maximum = 199, range = 199

Attribute 3: minimum = 0, maximum = 122, range = 122

Attribute 4: minimum = 0, maximum = 99, range = 99

Attribute 5: minimum = 0, maximum = 846, range = 846

Attribute 6: minimum = 0, maximum = 67.1, range = 67.1

Attribute 7: minimum = 0.078, maximum = 2.42, range = 2.342

Attribute 8: minimum = 21, maximum = 81, range = 60

Attribute 9: minimum = 0, maximum = 1, range = 1

b.)

Attribute 1: mean = 3.8451, variance = 11.3541

Attribute 2: mean = 120.8945, variance = 1022.2483

Attribute 3: mean = 69.1055, variance = 374.6473

Attribute 4: mean = 20.5365, variance = 254.4732

Attribute 5: mean = 79.7995, variance = 13281.1801

Attribute 6: mean = 31.9926, variance = 62.16

Attribute 7: mean = 0.4719, variance = 0.1098

Attribute 8: mean = 33.2409, variance = 138.3030

Attribute 9: mean = 0.3490, variance = 0.2275

c.)

Attribute 1: correlation = 0.2219

Attribute 2: correlation = 0.4666

Attribute 3: correlation = 0.0651

Attribute 4: correlation = 0.0748

Attribute 5: correlation = 0.1305

Attribute 6: correlation = 0.2927

Attribute 7: correlation = 0.1738

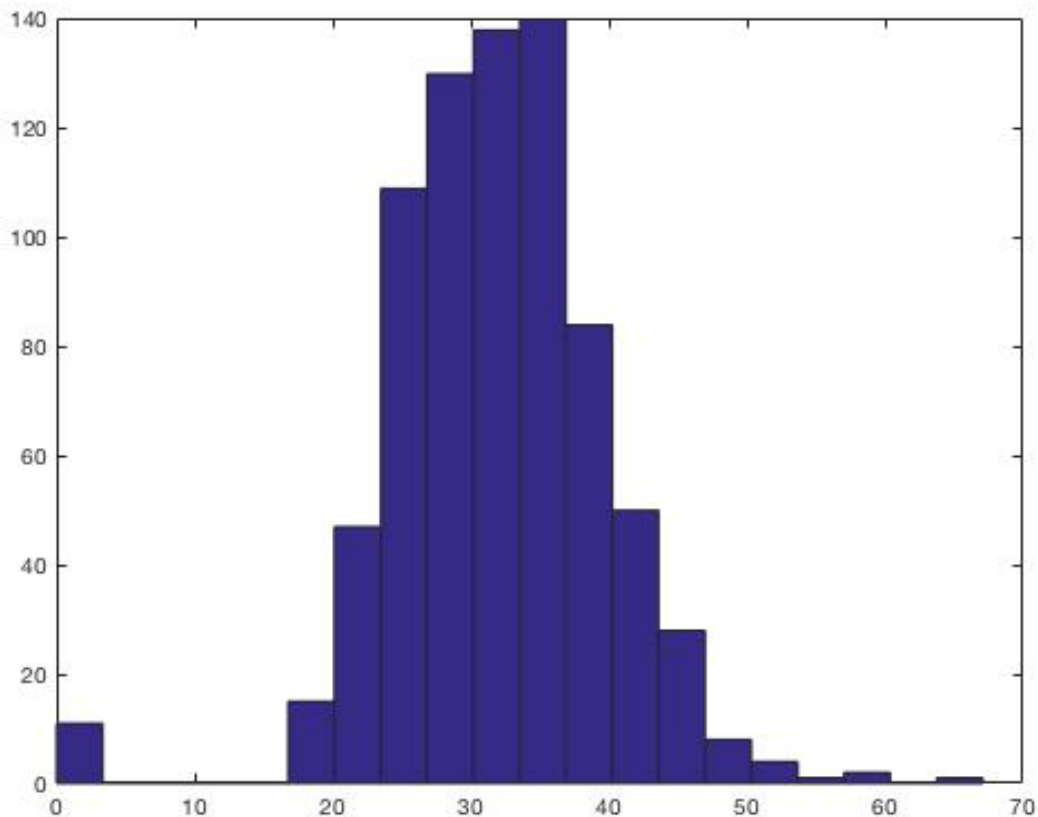
Attribute 8: correlation = 0.2384

The attribute with the highest positive correlation to the target attribute is attribute 2 (Plasma glucose concentration a 2 hours in an oral glucose tolerance test). I would think that this is the most helpful attribute in predicting the target class due to its correlation being significantly higher than any other attribute (the second highest is attribute 6 whose correlation is lower by 0.1739), and given the knowledge that diabetes is a disease which directly affects blood sugar concentrations.

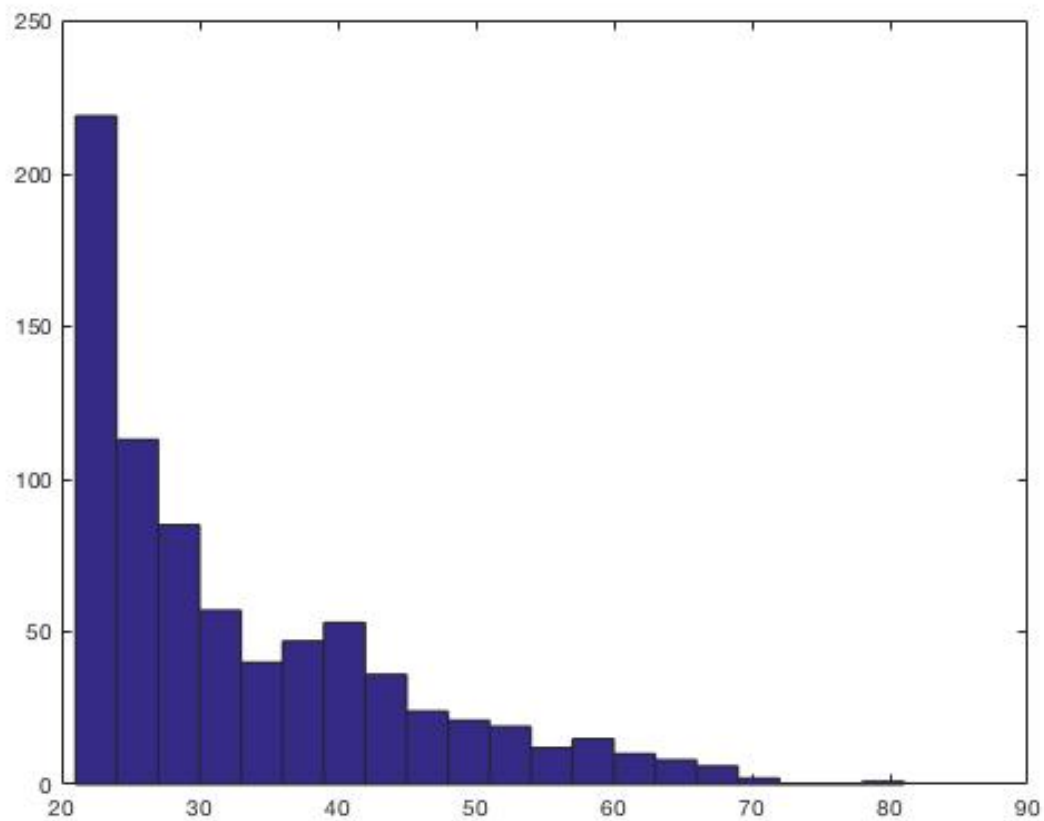
d.) Attribute 1 (Number of times pregnant) and attribute 8 (Age in years) have the largest mutual correlation at 0.5443

e.) I do **not** think having two attributes that are fully correlated would help in the prediction of the target class. While using one to predict the other could be accomplished since the two attributes are fully correlated, I believe this correlation between the two has no effect on the two attributes individual correlation/effect on the target class.

f.) I believe that the histogram representing attribute 6 (Body mass index) resembles the normal distribution the most.

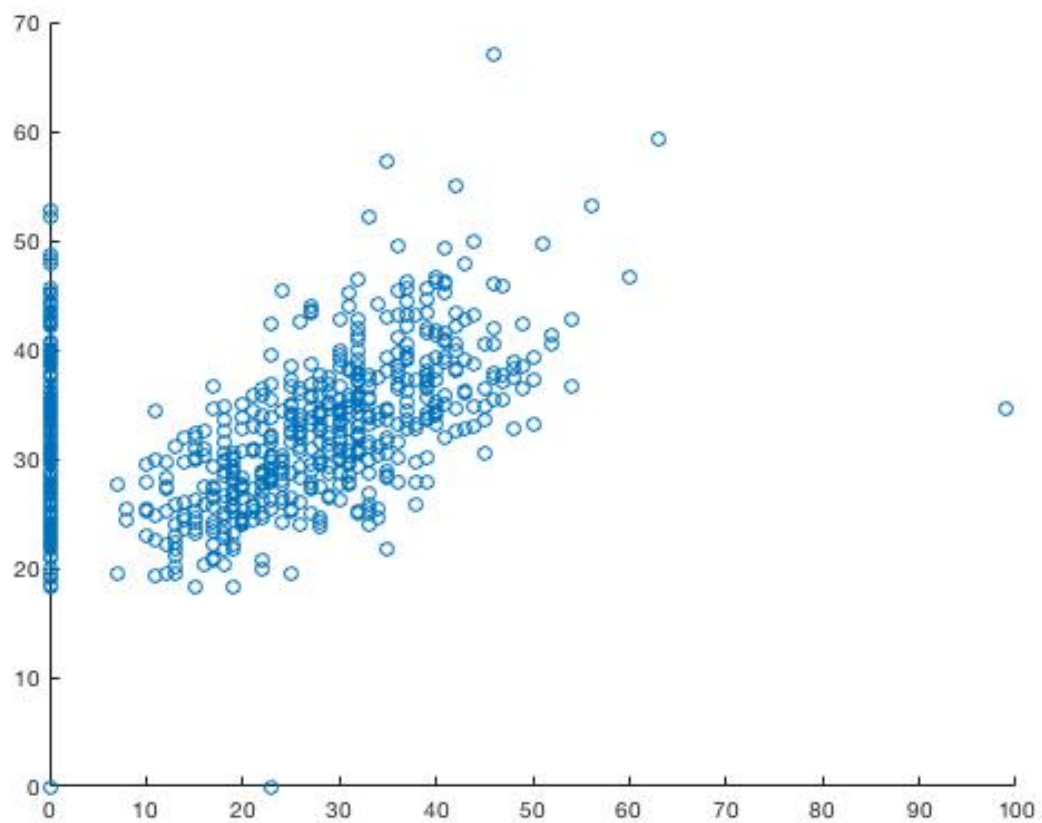


Above is the histogram of attribute 6 (Body mass index), which I believe most resembles the normal distribution.

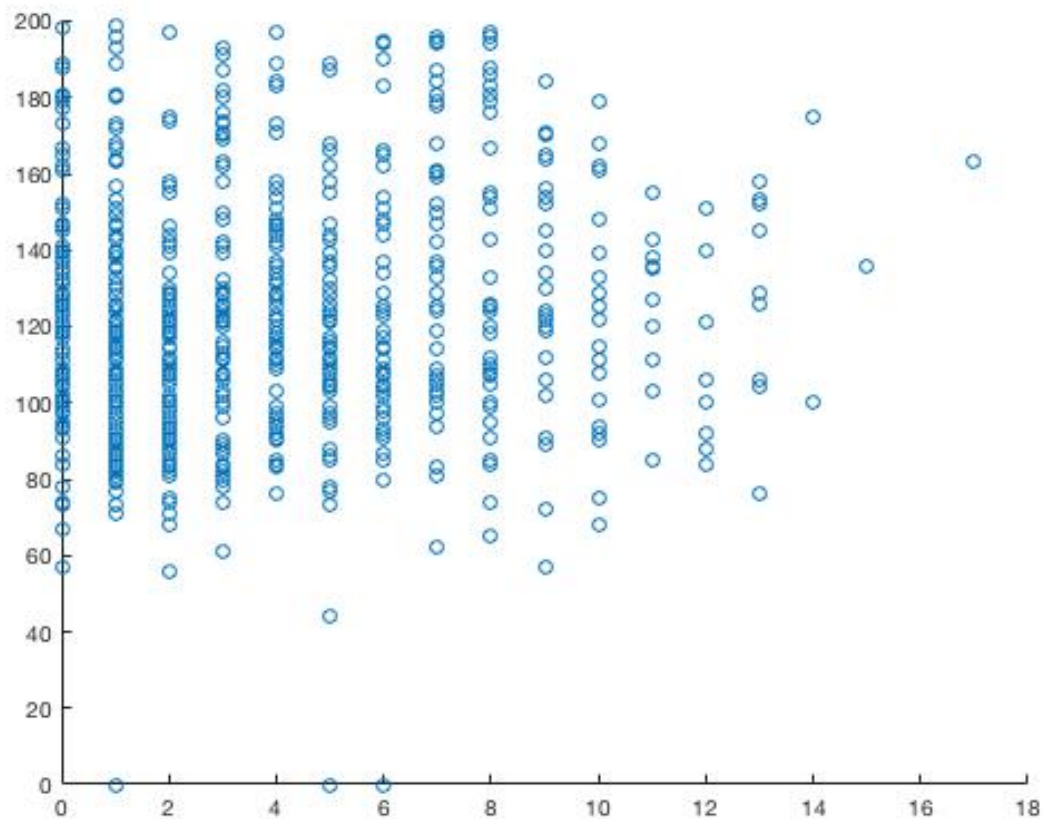


Above is the histogram of attribute 8 (Age in years) which is a good example of a histogram which does not resemble the normal distribution.

g.) I believe that the scatterplot of attribute 4 (Triceps skin fold thickness) against attribute 6 (Body mass index) indicates possible linear dependency between the two variables.



Above is attribute 4 (Triceps skin fold thickness) plotted against attribute 6 (Body mass index). A clear positive linear dependency can be observed between the two variables.



Above is attribute 1 (Number of times pregnant) plotted against attribute 2 (Plasma glucose concentration a 2 hours in an oral glucose tolerance test). It can be observed that there is no clear linear dependency between the two variables.

Problem 3.

a.) Normalized values for the first five entries of attribute 3 are listed below:

Entry1: 0.1495
 Entry2: -0.1604
 Entry3: -0.2638
 Entry4: -0.1604
 Entry5: -1.5037

b.) Discretized values for the first five entries of attribute 3 are listed below:

Entry1: 6
 Entry2: 6
 Entry3: 6

Entry4: 6
Entry5: 4

Problem 4.

a.)

Class label "0"

Attribute 1: mean = 3.298, standard deviation = 3.0172
Attribute 2: mean = 109.98, standard deviation = 26.1412
Attribute 3: mean = 68.184, standard deviation = 18.0631
Attribute 4: mean = 19.664, standard deviation = 14.8899
Attribute 5: mean = 68.792, standard deviation = 98.8653
Attribute 6: mean = 30.3042, standard deviation = 7.6899
Attribute 7: mean = 0.4297, standard deviation = 0.2991
Attribute 8: mean = 31.19, standard deviation = 11.6677

Class label "1"

Attribute 1: mean = 4.8657, standard deviation = 3.7412
Attribute 2: mean = 141.2575, standard deviation = 31.9396
Attribute 3: mean = 70.8246, standard deviation = 21.4918
Attribute 4: mean = 22.1642, standard deviation = 17.6797
Attribute 5: mean = 100.3358, standard deviation = 138.6891
Attribute 6: mean = 35.1425, standard deviation = 7.263
Attribute 7: mean = 0.5505, standard deviation = 0.3724
Attribute 8: mean = 37.0672, standard deviation = 10.9683

b.)

The average length of the training dataset: 504

c.)

When running `divideset2` with `p_train = 0.66` I get unique, random data sets in which my training set has 507 rows and my testing set has 261 rows.

$507/768 = 0.6602$ which is very close to 0.66 so I believe this function is working properly.