

1. Consider the toy dataset below which shows if 4 subjects have diabetes or not, along with two diagnostic measurements.

Preg	BP	HasDiabetes	Preg.Norm	BP.Norm
2	74	No	0.5	1.0
3	58	Yes	1.0	0.2
2	58	Yes	0.5	0.2
1	54	No	0.0	0.0
p2	70	?	0.5	0.8

a. Which variable is the “Class” variable?

The Class variable is HasDiabetes

b. Normalize the Preg and BP values by scaling the minimum-maximum range of each column to 0-1. Fill in the empty columns in the table.

c. Predict whether a subject with Preg=2, BP=70 will have diabetes using the 1-NN algorithm and

i. Using Euclidean distance on the original variables

Nearest distance: 0

Nearest neighbor: Row 5

ii. Using Euclidean distance on the normalized variables

Nearest distance: 69.0163

Nearest neighbor: Row 1

For each of these cases, give the nearest distance, nearest neighbor (e.g., “Row 1” or “Row 2”), and prediction.

```
1 library(class)
2 library(tidyverse)
3
4 first <- c(2,3,2,1,2)
5 second <- c(74,58,58,54,70)
6
7 df <- data.frame(first, second)
8
9 normalized<-function(y) {
10
11   x<-y[!is.na(y)]
12
13   x<-(x - min(x)) / (max(x) - min(x))
14
15   y[!is.na(y)]<-x
16
17   return(y)
18 }
19
20 apply(df[,c(1,2)],2,normalized)
21
22 a0 <- c(2, 70)
23 a1 <- c(2, 74)
24 a2 <- c(3, 58)
25 a3 <- c(2, 58)
26 a4 <- c(1, 54)
27 a5 <- c(2, 70)
28
29 d1 <- sqrt(sum((a0-a1)^2))
30 d1
31 d2 <- sqrt(sum((a0-a2)^2))
32 d2
33 d3 <- sqrt(sum((a0-a3)^2))
34 d3
35 d4 <- sqrt(sum((a0-a4)^2))
36 d4
37 d5 <- sqrt(sum((a0-a5)^2))
38 d5
39
40 a0 <- c(2, 70)
41 a1 <- c(0.5, 1.0)
42 a2 <- c(1.0, 0.2)
43 a3 <- c(0.5, 0.2)
44 a4 <- c(0.0, 0.0)
45 a5 <- c(0.5, 0.8)
46
47 d1 <- sqrt(sum((a0-a1)^2))
48 d1
49 d2 <- sqrt(sum((a0-a2)^2))
50 d2
51 d3 <- sqrt(sum((a0-a3)^2))
52 d3
53 d4 <- sqrt(sum((a0-a4)^2))
54 d4
55 d5 <- sqrt(sum((a0-a5)^2))
56 d5
57
```

```
58 pima <- read_csv("pima-indians-diabetes-resampled.csv")
59 pima
60
61
62
63 pima <- filter(pima, Glucose > 0)
64
65 normalize <- function(x) { return ((x-min(x)) / (max(x)-min(x))) }
66 pimaNorm <- pima %>%
67   mutate(Preg.norm=normalize(Preg),Pedigree.norm=normalize(Pedigree),Glucose.norm=normalize(Glucose))
68
69
70 trainIndex <- sample(1:500)
71
72 trainfeatures <- pimaNorm[trainIndex, c(1,2,3,4,5,6,7,8,9,10,11,12)]
73 trainlabels <- pimaNorm[trainIndex, c(1,2,3,4,5,6,7,8,9,10,11,12)]
74
75 testIndex <- setdiff(1:nrow(pimaNorm), trainIndex)
76
77 testfeatures <- pimaNorm[testIndex, c(1,2,3,4,5,6,7,8,9,10,11,12)]
78 testlabels <- pimaNorm[testIndex, c(1,2,3,4,5,6,7,8,9,10,11,12)]
79
80 trainfeatures
81 trainlabels
82 testfeatures
83 testlabels
84
85 trainfeatures <- pimaNorm[trainIndex, c(10,11,12)]
86 trainlabels <- pimaNorm[trainIndex, c(10,11,12)]
87
88 testfeatures <- pimaNorm[testIndex, c(10,11,12)]
89 testlabels <- pimaNorm[testIndex, c(10,11,12)]
90
91 trainfeatures
92 trainlabels
93 testfeatures
94 testlabels
95
96 k1 <- knn(train = trainfeatures, test = testfeatures,
97           cl = trainlabels, k=1)
98 table(testlabels, k1)
99
100 k5 <- knn(train = trainfeatures, test = testfeatures,
101           cl = trainlabels, k=5)
102 table(testlabels, k5)
103
104 k11 <- knn(train = trainfeatures, test = testfeatures,
105            cl = trainlabels, k=11)
106 table(testlabels, k11)
```