# 16.1) K-Nearest Neighbors

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December 2019

# Reference

Tables, Graphics, and Figures from

# Computational and Inferential Thinking: The Foundations of Data Science

Adhikari & DeNero (2019): Ch 17.1 Nearest Neighbors

17.2 Training and Testing

17.3 Rows of Tables

https://www.inferentialthinking.com/

## **Chronic Kidney Disease (CKD)**

```
import numpy as np
from datascience import *
path_data = 'https://github.com/data-8/textbook/raw/gh-pages/data/'
data = Table.read_table(path_data + 'ckd.csv')
ckd = data.relabeled('Blood Glucose Random', 'Glucose')
```

Red Blood Cells	Sugar	Albumin	Specific Gravity	Blood Pressure	Age
normal	0	4	1.005	70	48
abnormal	0	2	1.02	90	53
abnormal	0	3	1.01	70	63

### 1 = Chronic Kidney Disease (CKD)

```
def standard_units(x):
    return (x - np.mean(x))/np.std(x)

ckd = Table().with_columns(
    'Hemoglobin', standard_units(ckd.column('Hemoglobin')),
    'Glucose', standard_units(ckd.column('Glucose')),
    'White Blood Cell Count',
    standard_units(ckd.column('White Blood Cell Count')),
    'Class', ckd.column('Class'))
```

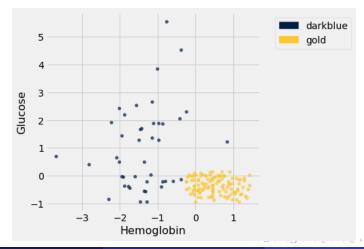
	Hemoglobin	Glucose	White Blood Cell Count	Class
	-0.865744	-0.221549	-0.569768	1
	-1.45745	-0.947597	1.16268	1
	-1.00497	3.84123	-1.27558	1

### Blue dots are patients with CKD

```
color_table = Table().with_columns(
    'Class', make_array(1, 0),
    'Color', make_array('darkblue', 'gold'))
ckd = ckd.join('Class', color_table)
```

Color	White Blood Cell Count	Glucose	Hemoglobin	Class
gold	0.617283	0.133751	0.456884	0
gold	0.424788	-0.947597	1.153	0
gold	0.200211	-0.762223	0.770138	0

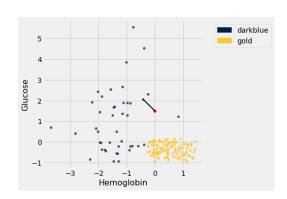
```
%matplotlib inline
import matplotlib.pyplot as plt
plt.style.use('fivethirtyeight')
ckd.scatter('Hemoglobin', 'Glucose', colors='Color')
```



#### **Functions to Calculate Distances**

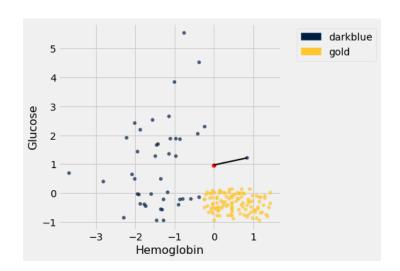
```
def distance(point1, point2):
    """The distance between two arrays of numbers."""
    return np.sqrt(np.sum((point1 - point2)**2))
def all distances(training, point):
    """The distance between p (an array of numbers)
    and the numbers in row i of attribute table."""
    attributes = training.drop('Class')
    def distance from point(row):
        return distance(point, np.array(row))
    return attributes.apply(distance from point)
def table with distances(training, point):
    """A copy of the training table with
    the distance from each row to array p."""
    return training.with column('Distance',
            all distances(training, point))
```

```
def closest(training, point, k):
    """A table containing the k closest
    rows in the training table to array p."""
    with_dists = table_with_distances(training, point)
    sorted_by_distance = with_dists.sort('Distance')
    topk = sorted_by_distance.take(np.arange(k))
    return topk
```



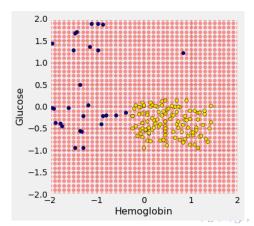
alice = make\_array(0, 1.5)
show closest(alice)

# alice = make\_array(0, 0.97) show\_closest(alice)



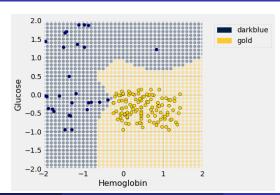
#### Create a Grid

```
x array = make array()
y_array = make_array()
for x in np.arange(-2, 2.1, 0.1):
    for y in np.arange(-2, 2.1, 0.1):
        x = np.append(x = nray, x)
        y array = np.append(y array, y)
test grid = Table().with columns(
    'Hemoglobin', x array,
    'Glucose', y array)
```

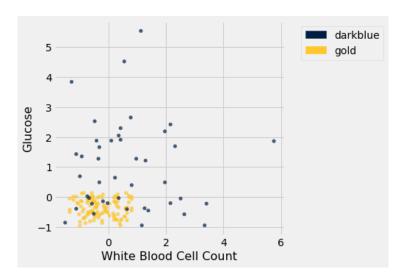


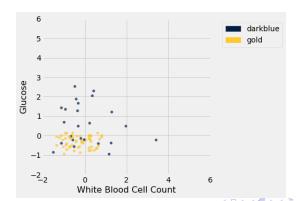
### **Nearest Neighbors Method**

```
def majority(topkclasses):
    """1 if the majority of the "Class" column is 1s, and 0 otherwise."""
    ones = topkclasses.where('Class', are.equal to(1)).num rows
    zeros = topkclasses.where('Class', are.equal to(0)).num rows
    if ones > zeros:
        return 1
    else:
        return 0
def classify(training, p, k):
    """Classify an example with attributes p using k-nearest
     neighbor classification with the given training table."""
    closestk = closest(training, p, k)
    topkclasses = closestk.select('Class')
    return majority(topkclasses)
def classify_grid(training, test, k):
    c = make array()
    for i in range(test.num rows):
        c = np.append(c, classify(training,
              make array(test.row(i)), k))
    return c
```



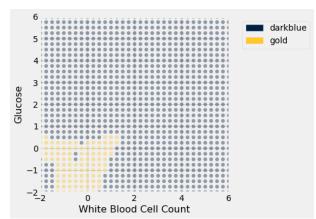
### 

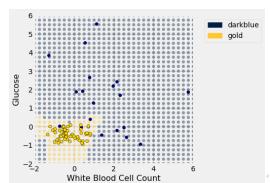




### Run k-Nearest Neighbors

```
x array = make array()
y array = make array()
for x in np.arange(-2, 6.1, 0.25):
    for y in np.arange(-2, 6.1, 0.25):
        x = np.append(x = nray, x)
        y array = np.append(y array, y)
test grid = Table().with columns(
    'Glucose', x array,
    'White Blood Cell Count', y array)
c = classify grid(training.drop('Hemoglobin',
                      'Color'), test grid, 1)
```





```
Row(Class=0, Hemoglobin=0.4568837017159849, Glucose=0.133750854517
ckd.row(0).item(1)
                    0.4568837017159849
ckd attributes = ckd.select('Hemoglobin', 'Glucose')
ckd attributes.row(3)
Row(Hemoglobin=0.5961076648232668, Glucose=-0.190653630343277
patient3 = np.array(ckd attributes.row(3))
alice = make array(0, 1.1)
alice, patient3
  (array([0., 1.1]), array([0.59610766, -0.19065363]))
```

### t = ckd\_attributes.take(np.arange(5))

Hemoglobin	Glucose
0.456884	0.133751
1.153	-0.947597
0.770138	-0.762223
0.596108	-0.190654
-0.239236	-0.49961

```
def max_abs(row):
    return np.max(np.abs(np.array(row)))
max_abs(t.row(4))
```

0.4996102825918697

```
t.apply(max_abs)
```

array([0.4568837 , 1.15300352, 0.77013762, 0.59610766, 0.49961028])

### Distance between Alice and another point

$$D = \sqrt{(x_0 - x_1)^2 + (y_0 - y_1)^2}$$

distance = np.sqrt(np.sum((alice - patient3)\*\*2))

1.421664918881847

```
def distance(point1, point2):
    return np.sqrt(np.sum((point1 - point2)**2))
def distance_from_alice(row):
    return distance(alice, np.array(row))
distance_from_alice(ckd_attributes.row(3))
```

1.421664918881847

distances = ckd\_attributes.apply(distance\_from\_alice)
ckd with distances = ckd.with column('Distance from Alice', distances)

ı	Hemoglobin	Glucose	White Blood Cell Count	Color	Distance from Alice
	0.456884	0.133751	0.617283	gold	1.06882
	1.153	-0.947597	0.424788	gold	2.34991
	0.770138	-0.762223	0.200211	gold	2.01519

sorted\_by\_distance = ckd\_with\_distances.sort('Distance from Alice')

Hemogrobin	GIucose	Muite Blood Cell Count	Color	Distance from Alice
0.83975	1.2151	1.29101	darkblue	0.847601
-0.970162	1.27689	-0.345191	darkblue	0.986156
-0.0304002	0.0874074	-0.184779	gold	1.01305

