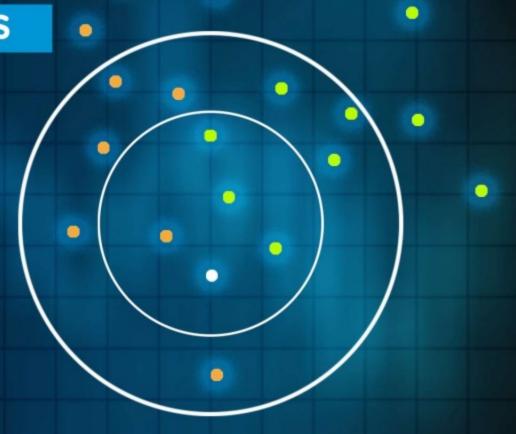
# K-NEAREST NEIGHBORS ALGORITHM TUTORIAL







## What's in it for you?

- Why do we need KNN?
- What is KNN?
- How do we choose the factor 'K'?
- When do we use KNN?
- How does KNN Algorithm work?
- Use Case: Predict whether a person will have diabetes or not

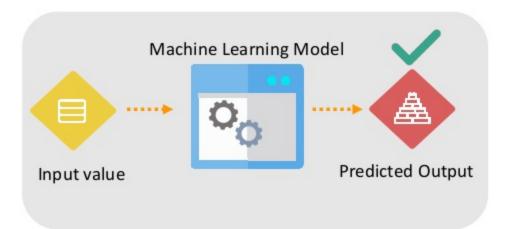




## Why KNN?

By now, we all know
Machine learning models
makes predictions by
learning from the past
data available











CATS



Sharp Claws, uses to climb

Smaller length of ears

Meows and purrs

Doesn't love to play around

DOGS



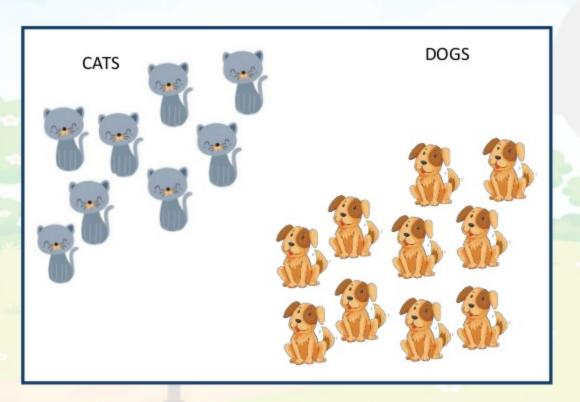
**Dull Claws** 

Bigger length of ears

Barks

Loves to run around

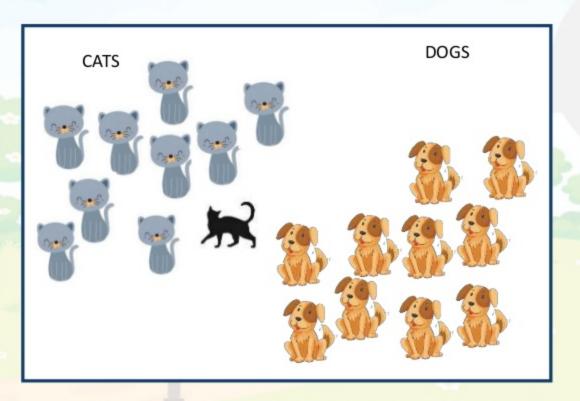
No dear, you can differentiate between a cat and a dog based on their characteristics



No dear, you can differentiate between a cat and a dog based on their characteristics

Length of ears →

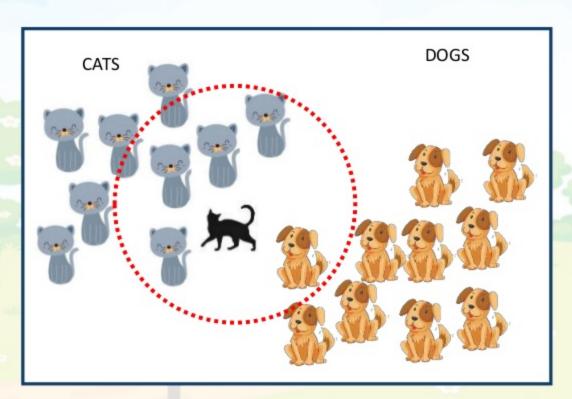




Now tell me if it's a cat or a dog?

Length of ears →



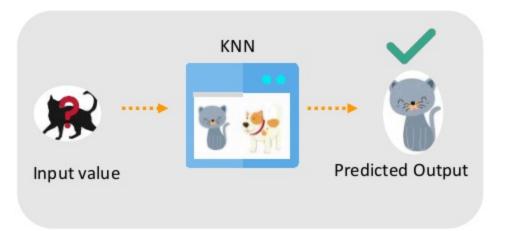


Length of ears →

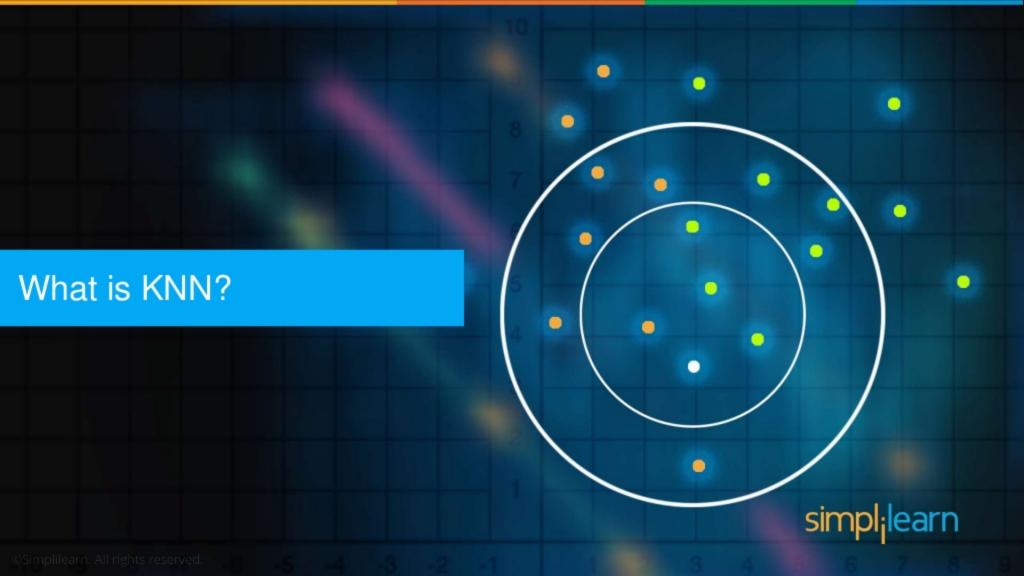
## Why KNN?

Because KNN is based on feature similarity, we can do classification using KNN Classifier!









KNN – K Nearest Neighbors, is one of the simplest **Supervised** Machine Learning algorithm mostly used for

#### Classification

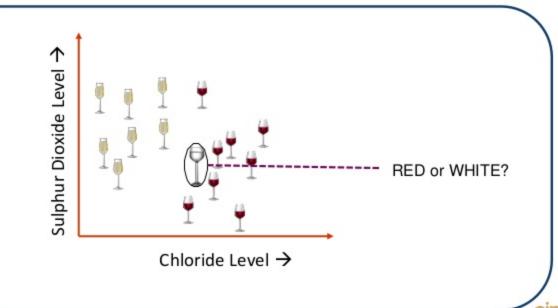


It classifies a data point based on how its neighbors are classified



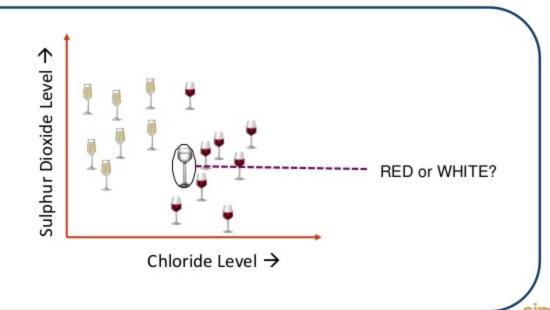
KNN stores all available cases and classifies new cases based on a similarity measure





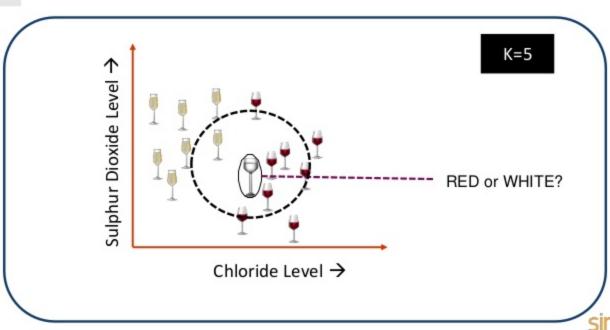
But, what is K?





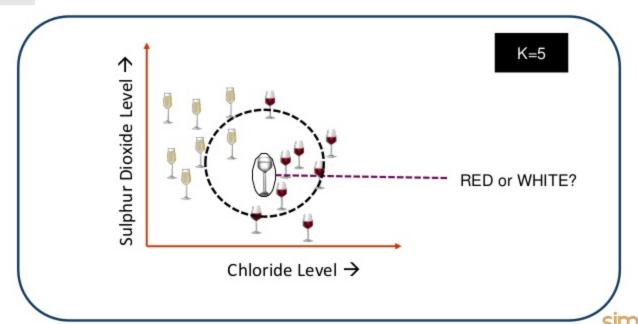
k in KNN is a parameter that refers to the number of nearest neighbors to include in the majority voting process





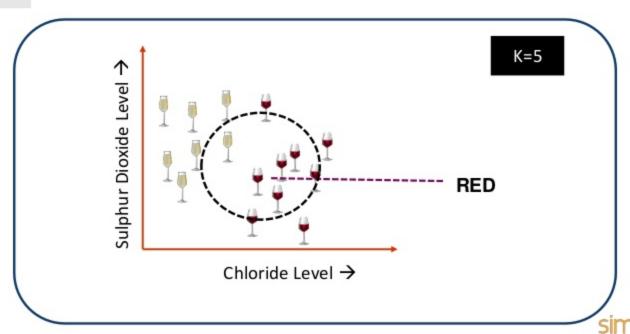
A data point is classified by majority votes from its 5 nearest neighbors





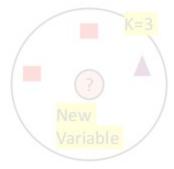
Here, the unknown point would be classified as red, since 4 out of 5 neighbors are red





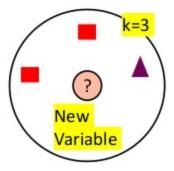
How do we choose 'k'? simpl<sub>i</sub>learn

KNN Algorithm is based on **feature similarity**: Choosing the right value of *k* is a process called parameter tuning, and is important for better accuracy





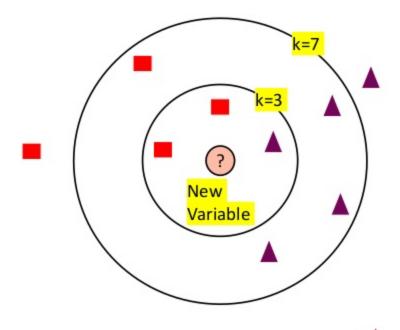
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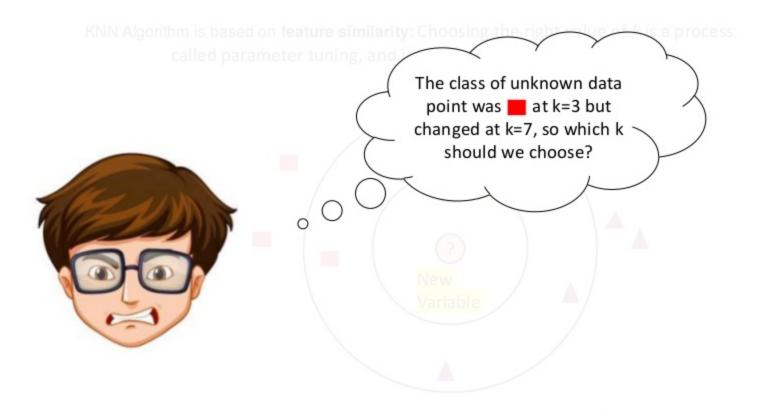


KNN Algorithm is based on **feature similarity**: Choosing the right value of *k* is a process called parameter tuning, and is important for better accuracy

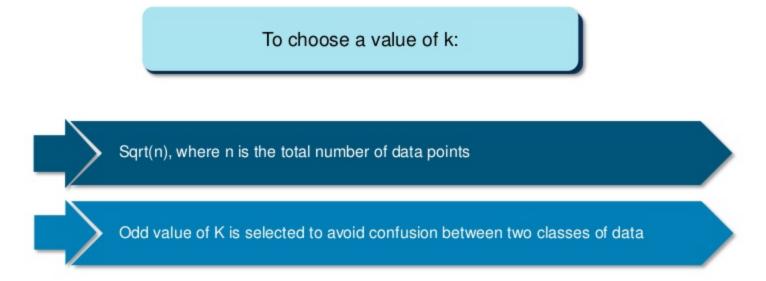




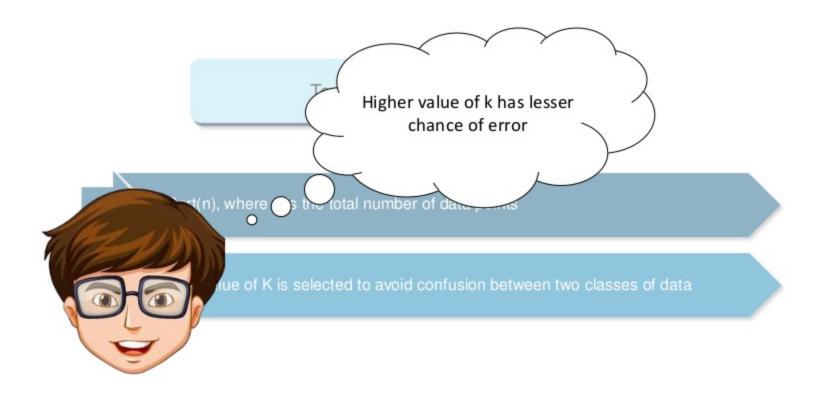








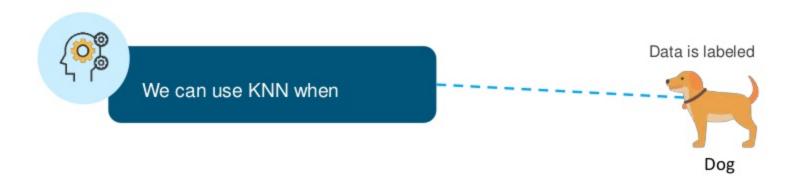






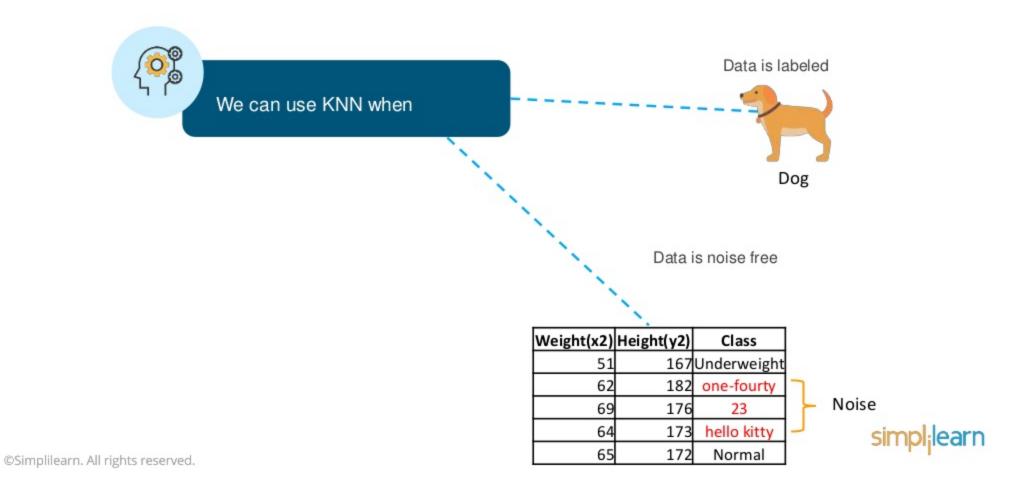
When do we use KNN?

## When do we use KNN Algorithm?

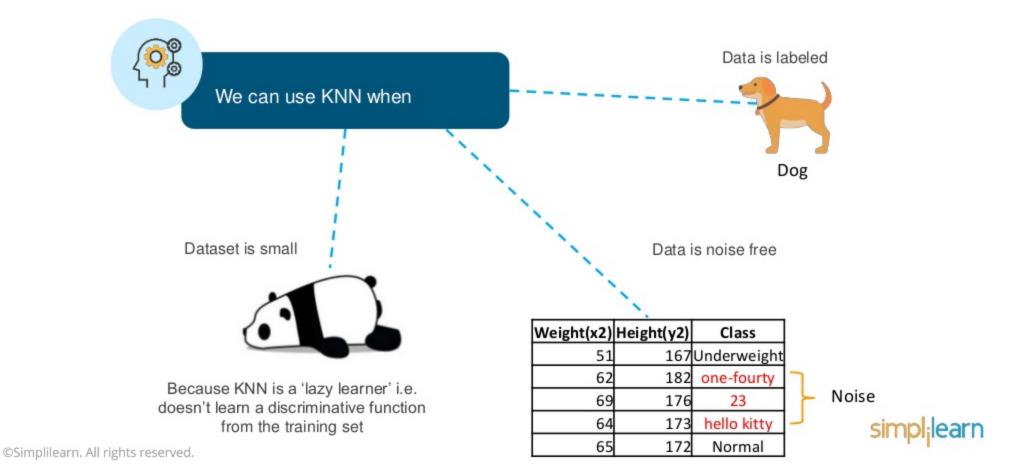


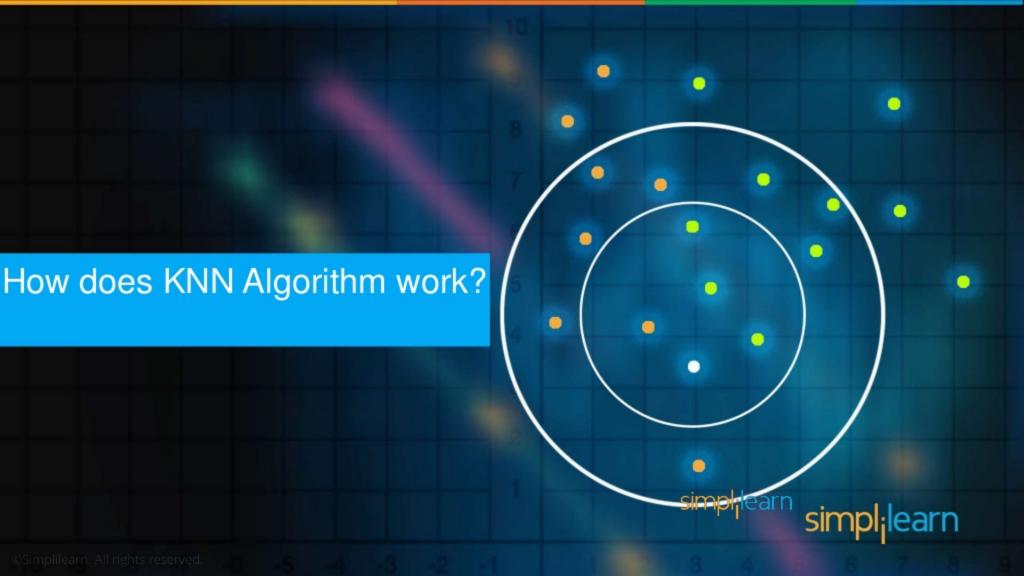


## When do we use KNN Algorithm?



## When do we use KNN Algorithm?







Consider a dataset having two variables: height (cm) & weight (kg) and each point is classified as Normal or Underweight

Weight(x2)	Height(y2)	Class
51	167	Underweight
62	182	Normal
69	176	Normal
64	173	Normal
65	172	Normal
56	174	Underweight
58	169	Normal
57	173	Normal
55	170	Normal





On the basis of the given data we have to classify the below set as Normal or Underweight using KNN

57 kg 170 cm ?





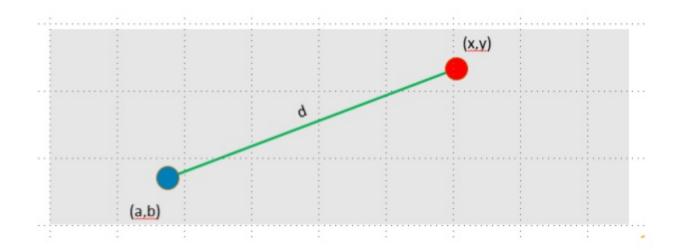
To find the nearest neighbors, we will calculate Euclidean distance





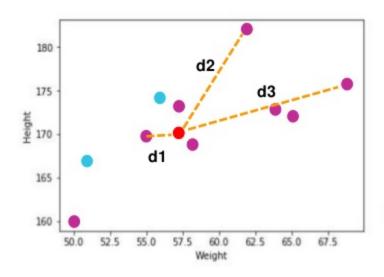
According to the **Euclidean distance** formula, the **distance** between two points in the plane with coordinates (x, y) and (a, b) is given by:

dist(d)= 
$$\sqrt{(x - a)^2 + (y - b)^2}$$





Let's calculate it to understand clearly:



dist(**d1**)= 
$$\sqrt{(170-167)^2 + (57-51)^2} \sim 6.7$$

dist(**d2**)= 
$$\sqrt{(170-182)^2 + (57-62)^2} \sim = 13$$

dist(d3)= 
$$\sqrt{(170-176)^2 + (57-69)^2} \sim = 13.4$$

Similarly, we will calculate Euclidean distance of unknown data point from all the points in the dataset

Unknown data point



Hence, we have calculated the Euclidean distance of unknown data point from all the points as shown:

Where (x1, y1) = (57, 170) whose class we have to classify

Weight(x2)	Height(y2)	Class	Euclidean Distance
51	167	Underweight	6.7
62	182	Normal	13
69	176	Normal	13.4
64	173	Normal	7.6
65	172	Normal	8.2
56	174	Underweight	4.1
58	169	Normal	1.4
57	173	Normal	3
55	170	Normal	2

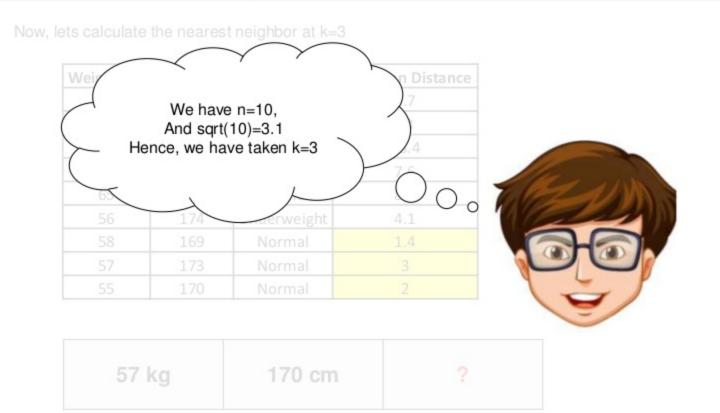


Now, lets calculate the nearest neighbor at k=3

Weight(x2)	Height(y2)	Class	Euclidean Distance	
51	167	Underweight	6.7	8
62	182	Normal	13	
69	176	Normal	13.4	
64	173	Normal	7.6	
65	172	Normal	8.2	
56	174	Underweight	4.1	
58	169	Normal	1.4	k = 3
57	173	Normal	3	
55	170	Normal	2	

57 kg
-------









Class	<b>Euclidean Distance</b>	
Underweight	6.7	
Normal	13	
Normal	13.4	
Normal	7.6	
Normal	8.2	
Underweight	4.1	
Normal	1.4	k=3
Normal	3	
Normal	2	

So, majority neighbors are pointing towards 'Normal'

Hence, as per KNN algorithm the class of (57, 170) should be 'Normal'



# Recap of KNN

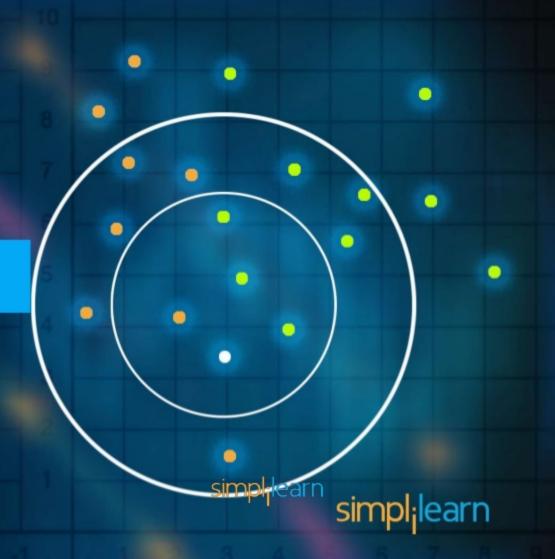


### Recap of KNN

- A positive integer k is specified, along with a new sample
- We select the k entries in our database which are closest to the new sample
- We find the most common classification of these entries
- This is the classification we give to the new sample

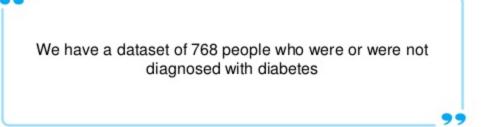


**USE CASE: Predict Diabetes** 





Objective: Predict whether a person will be diagnosed with diabetes or not





### Import the required Scikit-learn libraries as shown:

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import confusion_matrix
from sklearn.metrics import f1_score
from sklearn.metrics import accuracy_score
```



Load the dataset and have a look:

at	ataset.head()								
	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Age	Outcome
0	6	148	72	35	0	33.6	0.627	50	1
1	1	85	66	29	0	26.6	0.351	31	0
2	8	183	64	0	0	23.3	0.672	32	1
3	1	89	66	23	94	28.1	0.167	21	0
4	0	137	40	35	168	43.1	2.288	33	1



Values of columns like 'Glucose', BloodPressure' cannot be accepted as zeroes because it will affect the outcome

We can replace such values with the mean of the respective column:

```
# Replace zeroes
zero_not_accepted = ['Glucose', 'BloodPressure', 'SkinThickness', 'BMI', 'Insulin']

for column in zero_not_accepted:
    dataset[column] = dataset[column].replace(0, np.NaN)
    mean = int(dataset[column].mean(skipna=True))
    dataset[column] = dataset[column].replace(np.NaN, mean)
```



Before proceeding further, let's split the dataset into train and test:

```
# split dataset
X = dataset.iloc[:, 0:8]
y = dataset.iloc[:, 8]
X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=0, test_size=0.2)
```



Rule of thumb: Any algorithm that computes distance or assumes normality, scale your features!

Feature Scaling:



```
# Feature scaling
sc_X = StandardScaler()
X_train = sc_X.fit_transform(X_train)
X_test = sc_X.transform(X_test)
```



N\_neighbors here is 'K' p is the power parameter to define the metric used, which is 'Euclidean' in our case

Then define the model using KNeighborsClassifier and fit the train data in the model









### Let's predict the test results:



It's important to evaluate the model, let's use confusion matrix to do that:

```
# Evaluate Model
cm = confusion_matrix(y_test, y_pred)
print (cm)
print(f1_score(y_test, y_pred))

[[94 13]
  [15 32]]
0.6956521739130436
```



Calculate accuracy of the model:

print(accuracy\_score(y\_test, y\_pred))
0.8181818181818182





So, we have created a model using KNN which can predict whether a person will have diabetes or not



print(accuracy\_score(y\_test, y\_pred))
0.8181818181818182

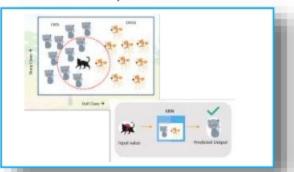


And accuracy of 80% tells us that it is a pretty fair fit in the model!

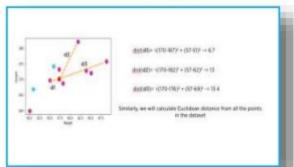


# **Summary**

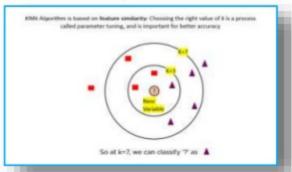
#### Why we need knn?



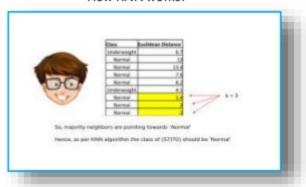
#### Eucledian distance



#### Choosing the value of k



#### How KNN works?



#### Knn classifier for diabetes prediction





