## tqlf3yadu

April 29, 2024

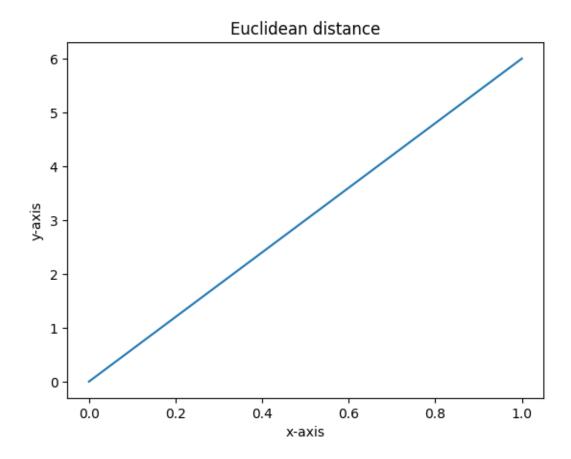
#### **EUCLIDEAN DISTANCE**

Euclidean distance represent the shortest distance between two points into the vector space.

The formula for calculating the distance is: Let suppose two points P(p1, p2, p3), Q(q1, q2, q3)The euclidean distance between these two points is thus represented as:  $d = ((q1 - p1)^2 + (q2 - p2)^2 + (q3 - p3)^2)(0.5)$ 

```
[]: # importing the library
     from scipy.spatial import distance
     import numpy as np
     import matplotlib.pyplot as plt
     # defining the points
     point_1 = np.random.randint(0,10,size=2)
     point_2 = np.random.randint(0,10,size=2)
     #calculating the euclidean distance between the given two points
     euclidean_distance = distance.euclidean(point_1,point_2)
     print('Euclidean Distance b/w', point_1, 'and', point_2, 'is: ',u
      →euclidean_distance)
     # plotting the points
     plt.plot(point_1,point_2)
     plt.xlabel("x-axis")
     plt.ylabel("y-axis")
     plt.title("Euclidean distance")
    plt.show()
```

Euclidean Distance b/w [0 1] and [0 6] is: 5.0



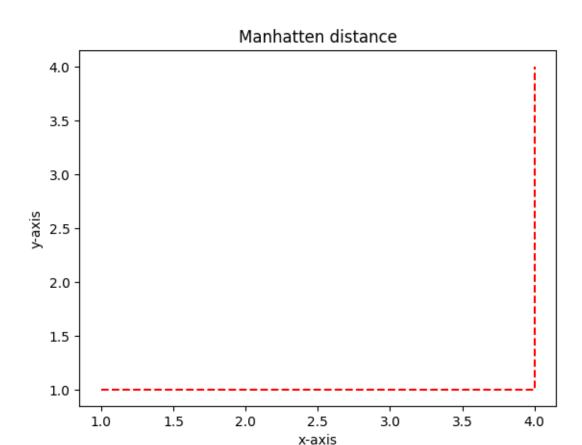
#### MANHATTEN DISTANCE

Manhatten Distance is also known as city block distance

which is given in terms of the coordinates as:

$$d = |(p1 - q1)| + |(p2 - q2)| + |(p3 - q3)|$$

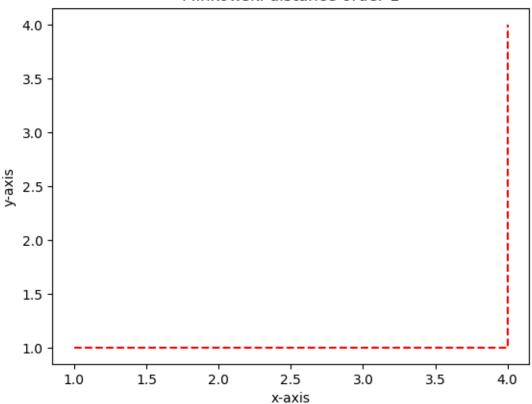
Manhatten Distance b/w [0 1] and [0 6] is: 5



#### MINKOWSKI DISTANCE ORDER 1:

Minkowski Distance of order 1: 6.0 Manhattan Distance: 6

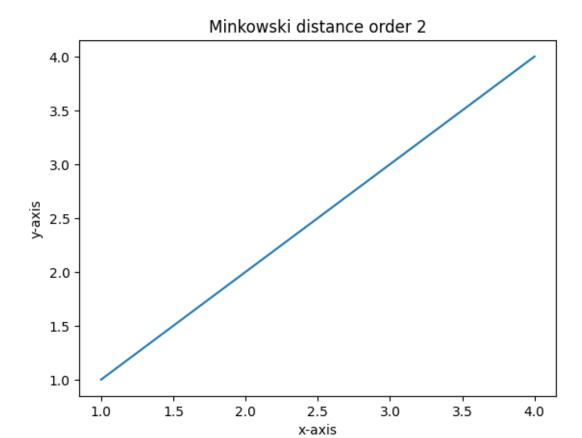




#### MINKOWSKI DISTANCE ORDER 2:

Minkowski Distance of order 2: 4.242640687119285

Euclidean Distance: 4.242640687119285



# []:

### COSINE DISTANCE/SIMILARITY

```
[]: # import required libraries
import numpy as np
from numpy.linalg import norm

# define two lists or array
A = np.array([2,5,0,7,2,0,0,3])
B = np.array([1,2,0,2,2,0,2,4])

print("A:", A)
print("B:", B)

# compute cosine similarity
cosine = np.dot(A,B)/(norm(A)*norm(B))
print("Cosine Similarity:", cosine)
```

A: [2 1 2 3 2 9]

B: [3 4 2 4 5 5]

Cosine Similarity: 0.8188504723485274