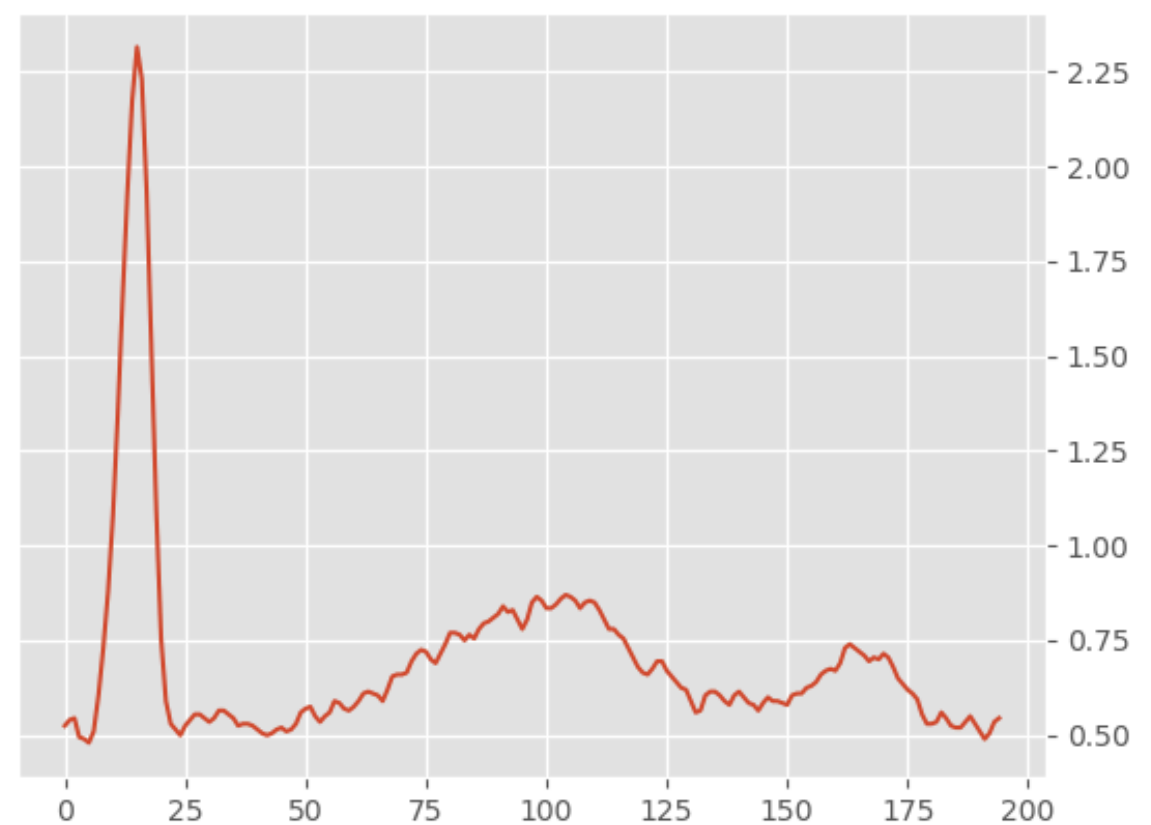
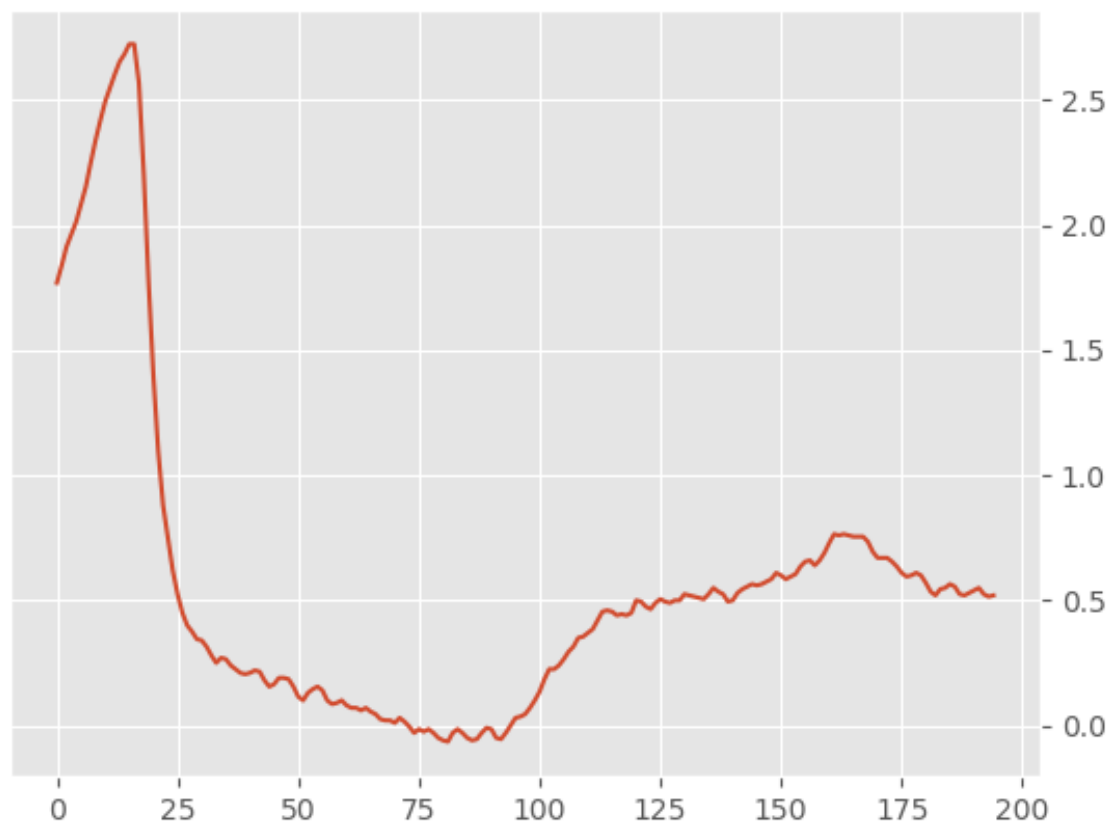


# Dynamic Time Warping

by Dr. Tan Jen Hong

# Problem

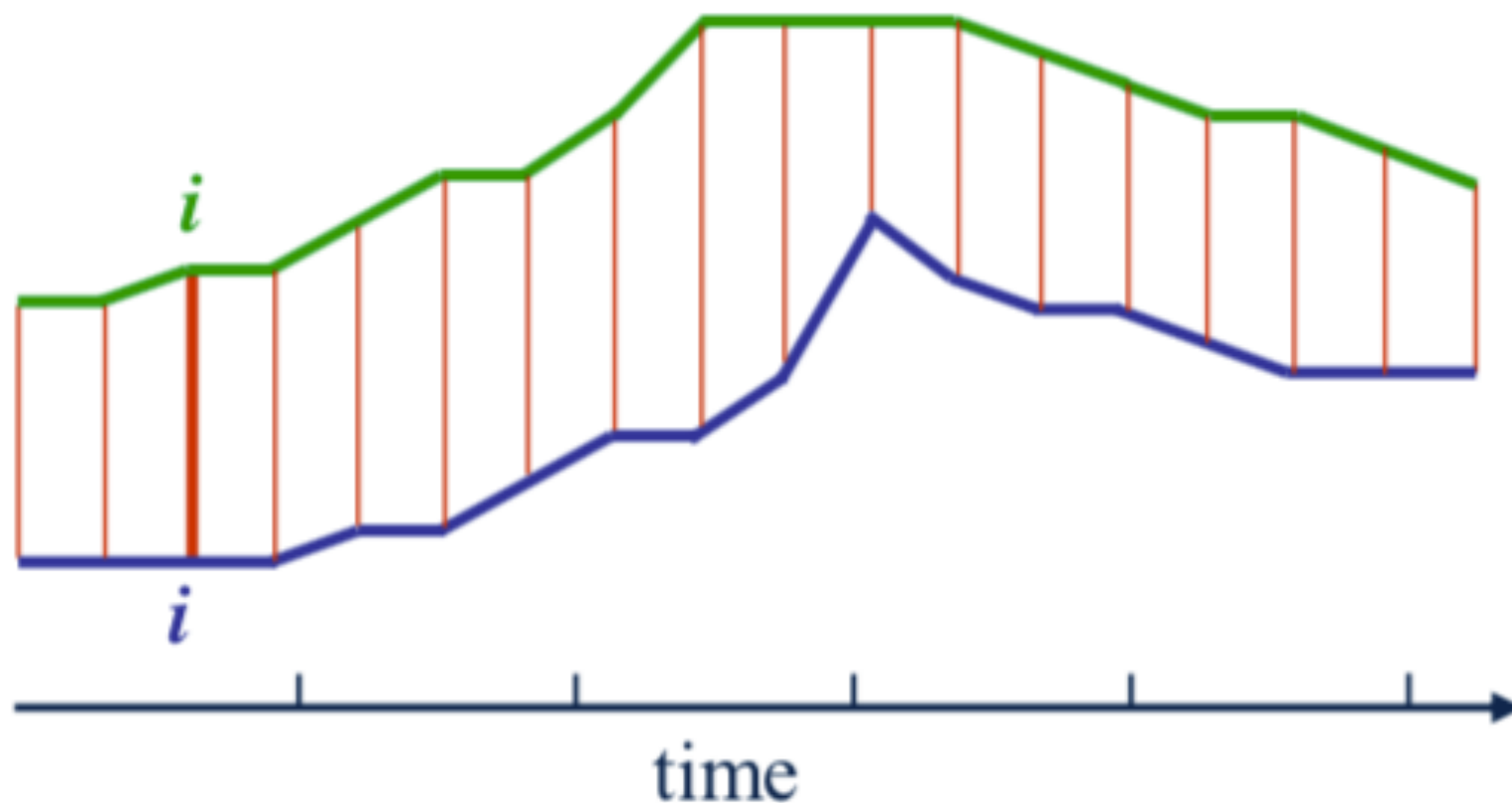
- How similar are these two signals?
- In which manners are they similar?



## Can we try ...

Euclidean, Manhattan .....

- We can measure the similarity of two signals by calculating the distance between the  $i$ -th point on one signal and the  $i$ -th point on another signal
- Simple concept, but could not capture the similarity in shape

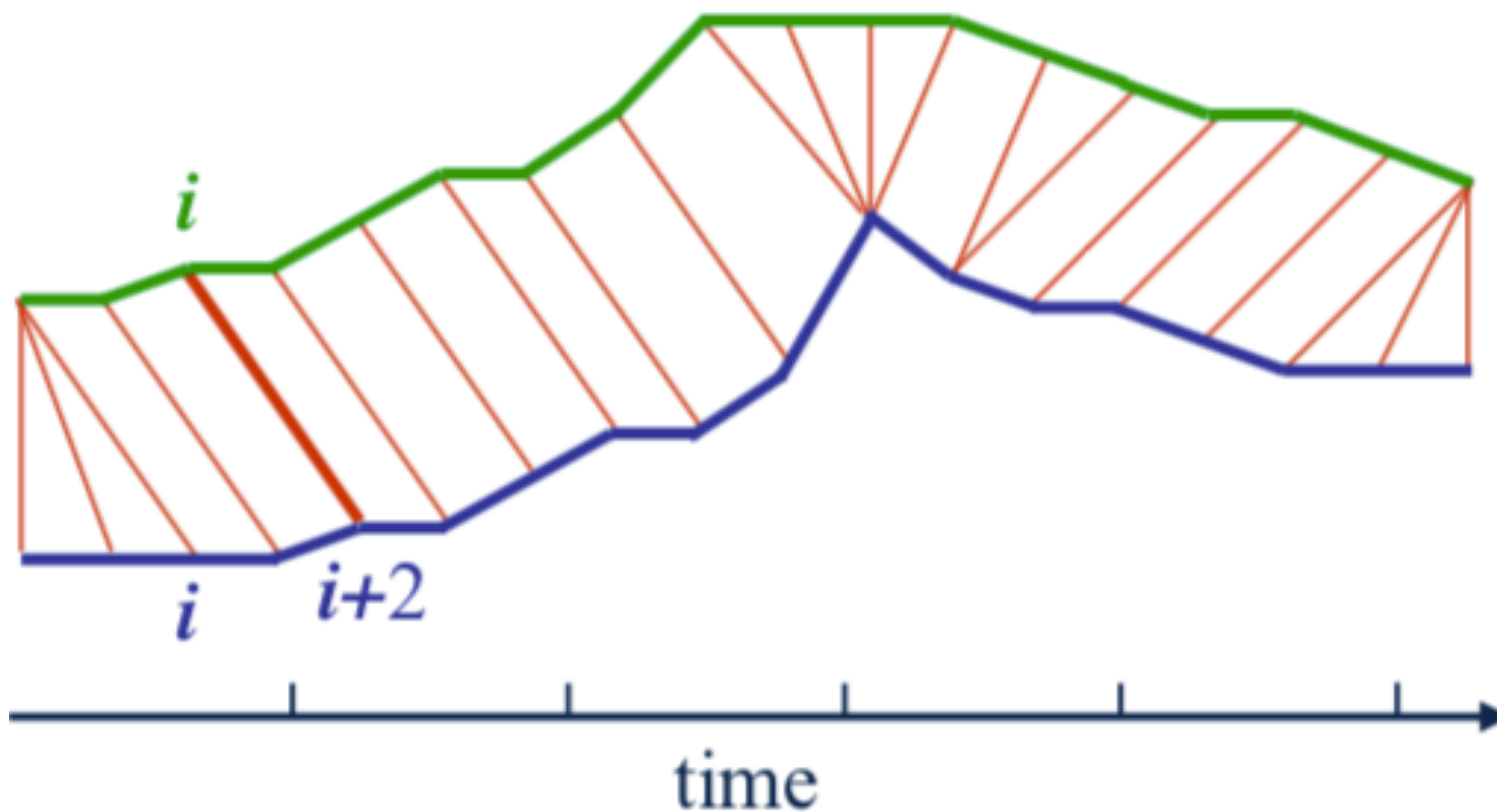


Source: "Dynamic Time Warping Algorithm", by Elena Tsiporkova

## How about ...

non-linear alignment .....?

- Elastic alignment between points of two signals produces a better, more intuitive similarity measure
- Allow similar shapes to match even if they are out of phase



Source: "Dynamic Time Warping Algorithm", by Elena Tsiporkova

# Distance

Another term to say 'similarity'

- Consider two distinct signals

$$\mathbf{x} = [x_1, x_2, \dots, x_i, \dots, x_m]$$

$$\mathbf{y} = [y_1, y_2, \dots, y_j, \dots, y_n]$$

- The distance between the two signals is defined as

$$d_s(\mathbf{x}, \mathbf{y})$$

- Euclidean distance between two signals:

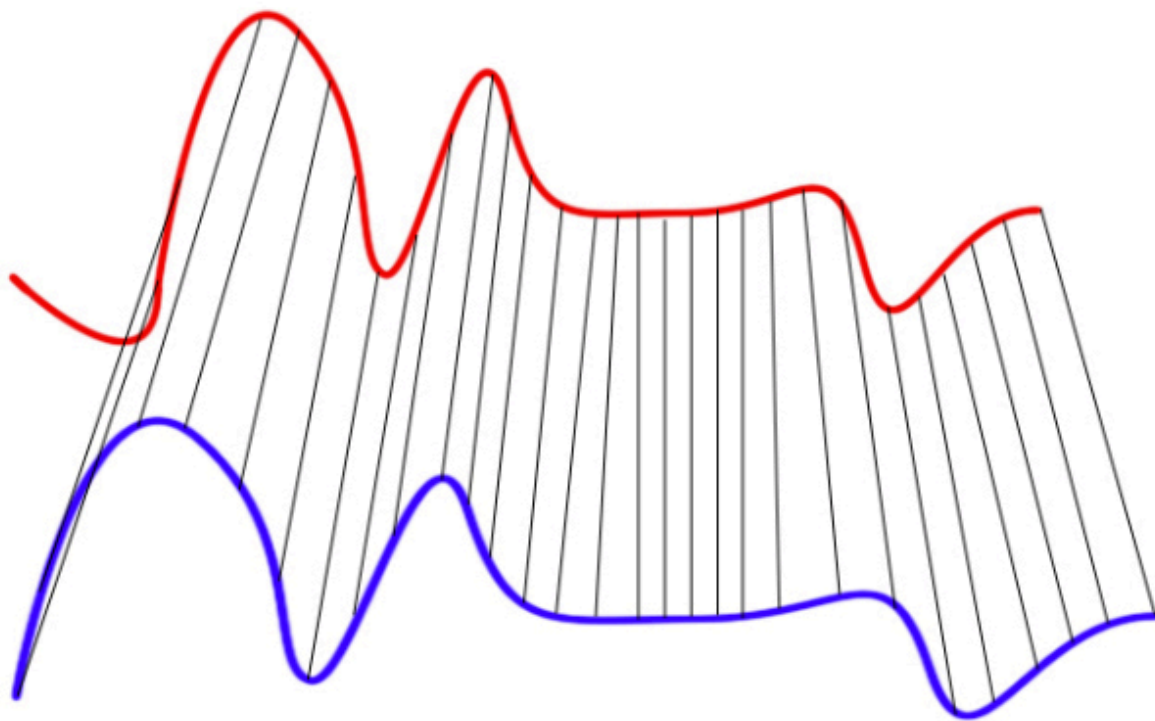
$$d_s(\mathbf{x}, \mathbf{y}) = \sqrt{\sum_i (x_i - y_i)^2}$$

- Problem: two signals must be of same length!

# Dynamic Time Warping

DTW

- An algorithm to measure similarity between two temporal sequences (signal), which may vary in speed
- DTW calculates an optimal match between two given sequences
- Sequences are warped along time dimension to determine similarity independent of variations in time
- DTW produces warping path, which enables alignment between two signals



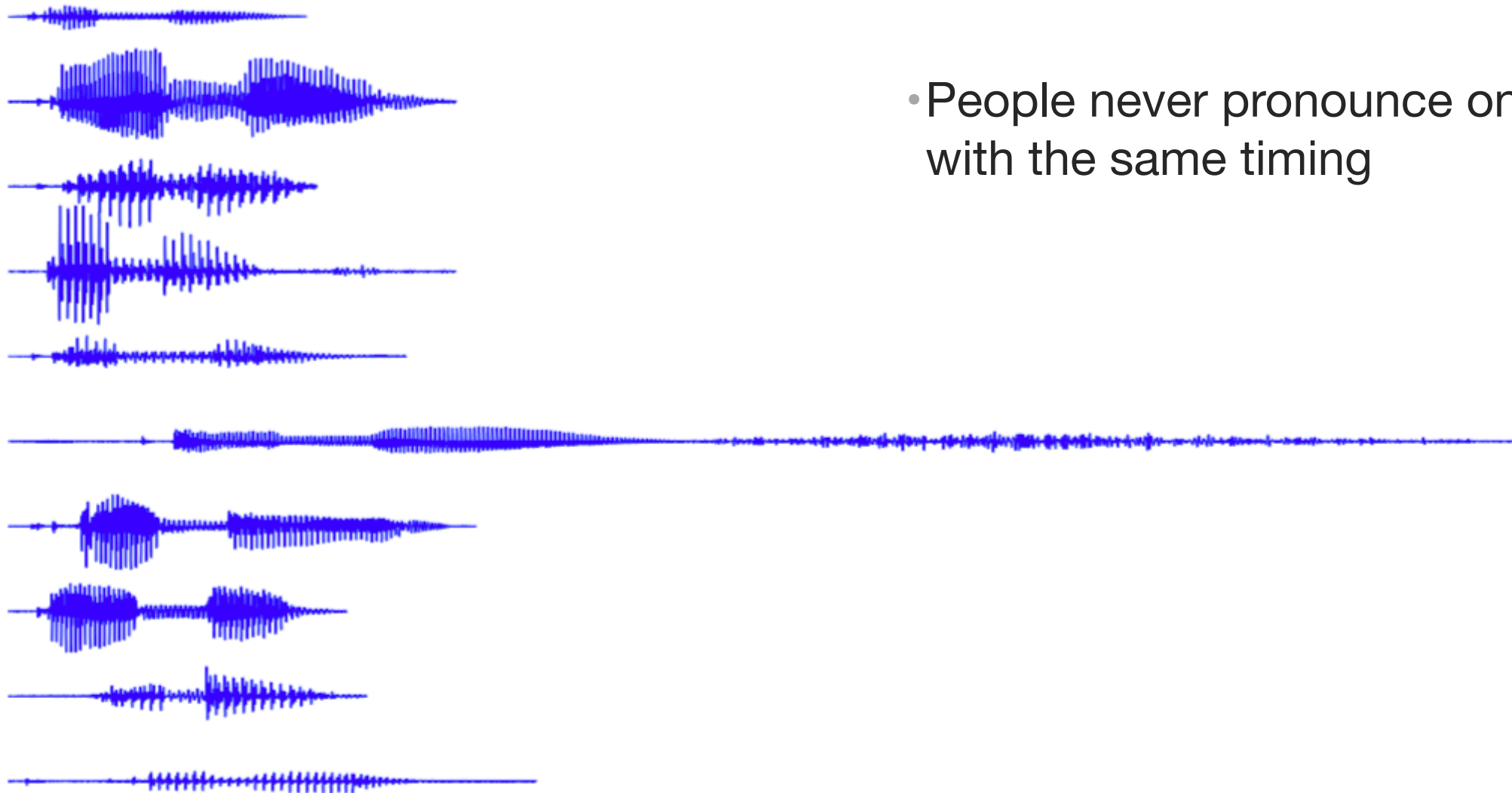
Source: [https://th.wikipedia.org/wiki/Dynamic\\_time\\_warping#/media/File:Euclidean\\_vs\\_DTW.jpg](https://th.wikipedia.org/wiki/Dynamic_time_warping#/media/File:Euclidean_vs_DTW.jpg)

# Dynamic Time Warping

## Usage

- Commonly used in speech recognition
- Individual never pronounces one word twice in exact way
- People never pronounce one word with the same timing

"timing"

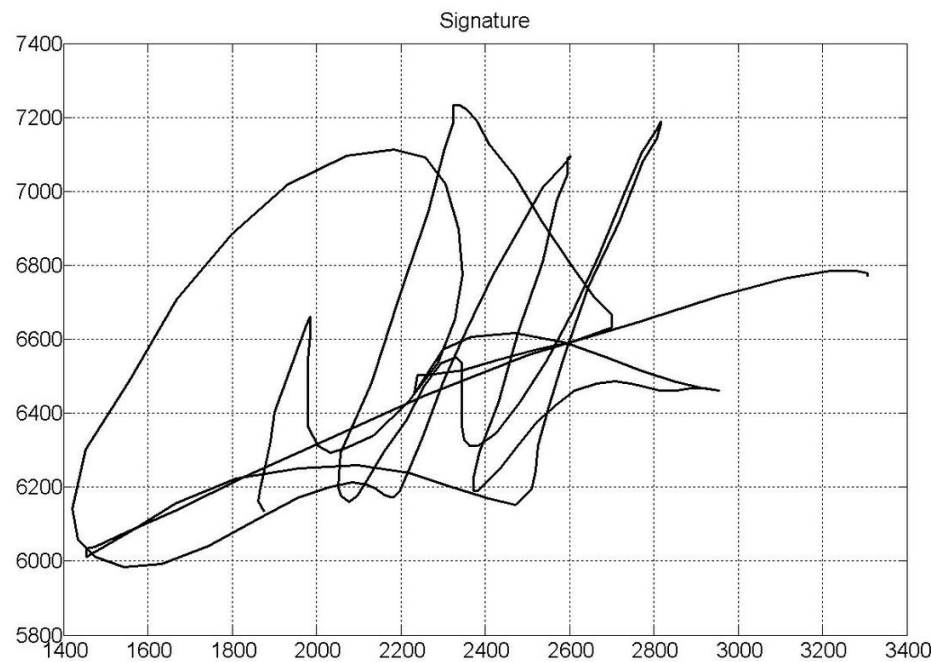


Source: "Speech Recognition - Intro and DTW", by Jan Černocký

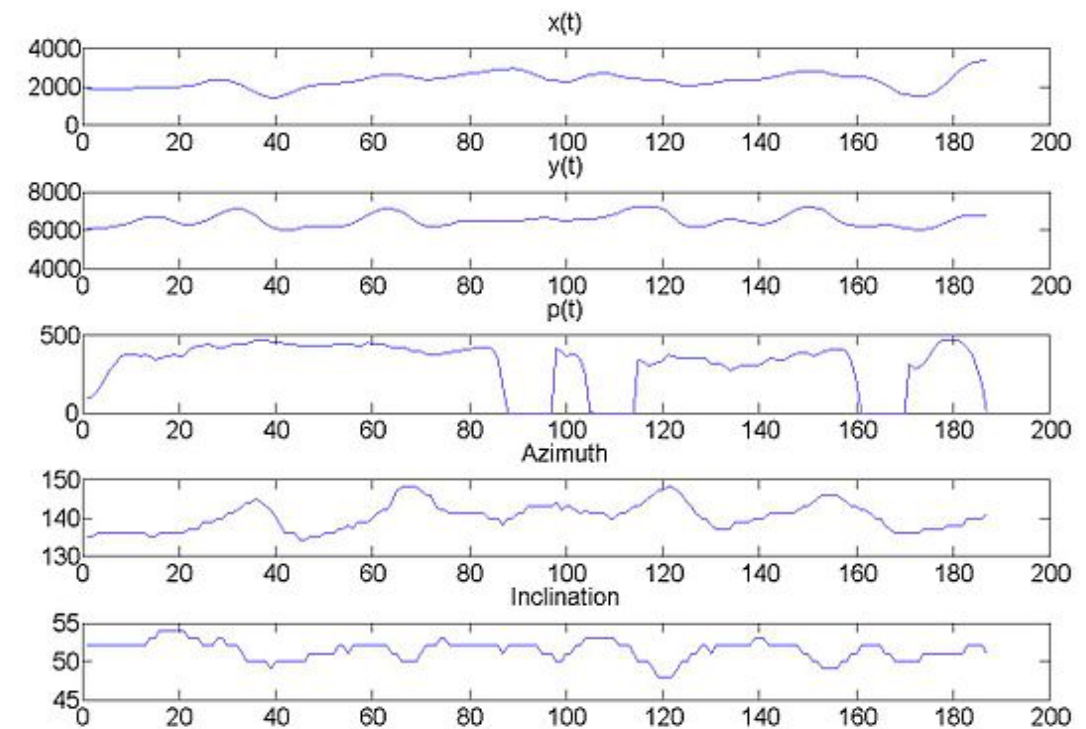
# Dynamic Time Warping

## Dynamic signature recognition

- Users sign their signature on digital tablet
- Dynamic information captured:
  - x position
  - y position
  - pressure
  - azimuth
  - inclination
- Use DTW to check / match signature



Source: "Speech Recognition - Intro and DTW", by Jan Černocký

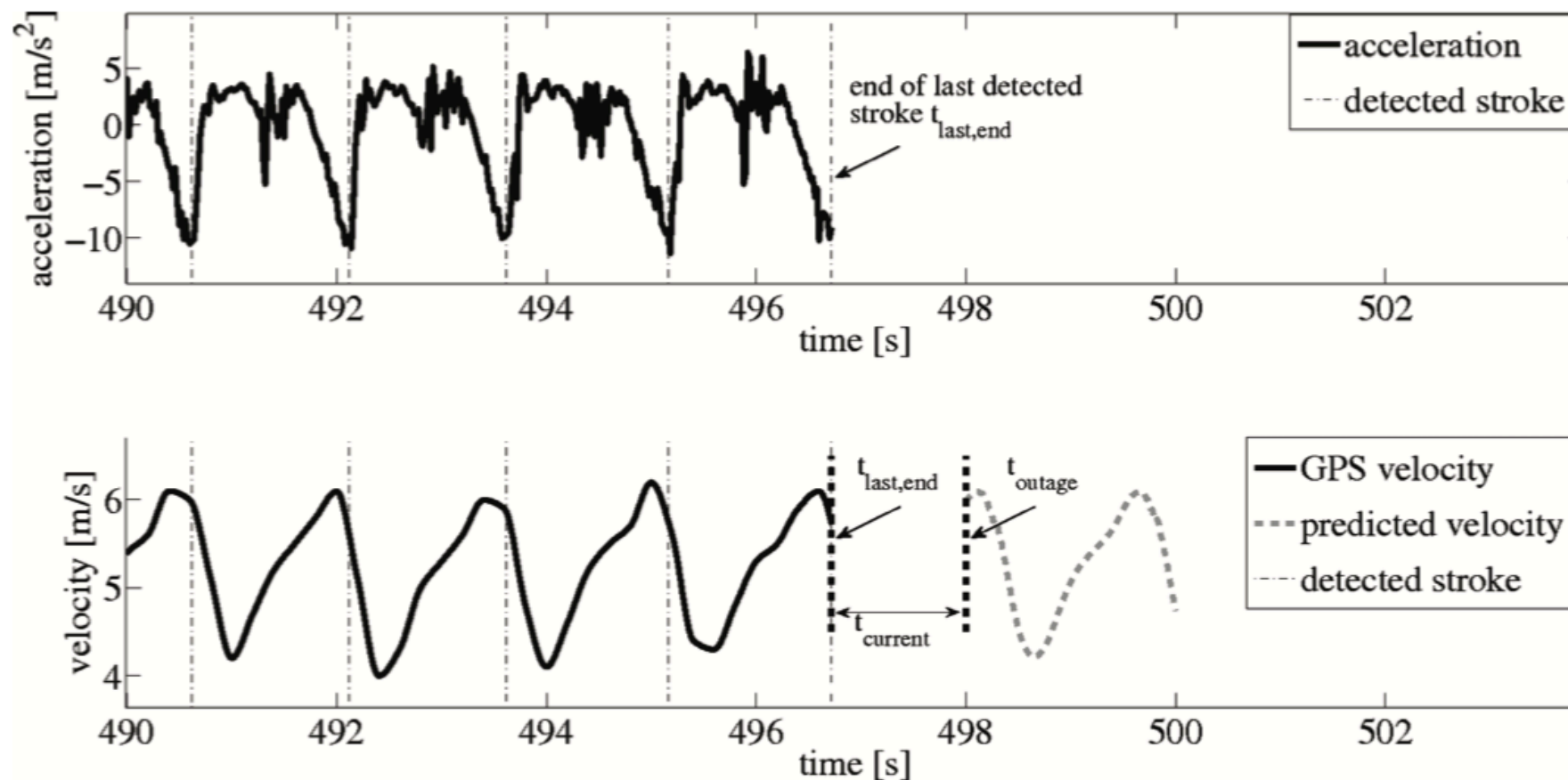




# Dynamic Time Warping

## Stroke detection (rowing)

- Use DTW to detect stroke (used in rowing competitions)
- With strokes detected, predict boat's movement and position when sensor transmission lost

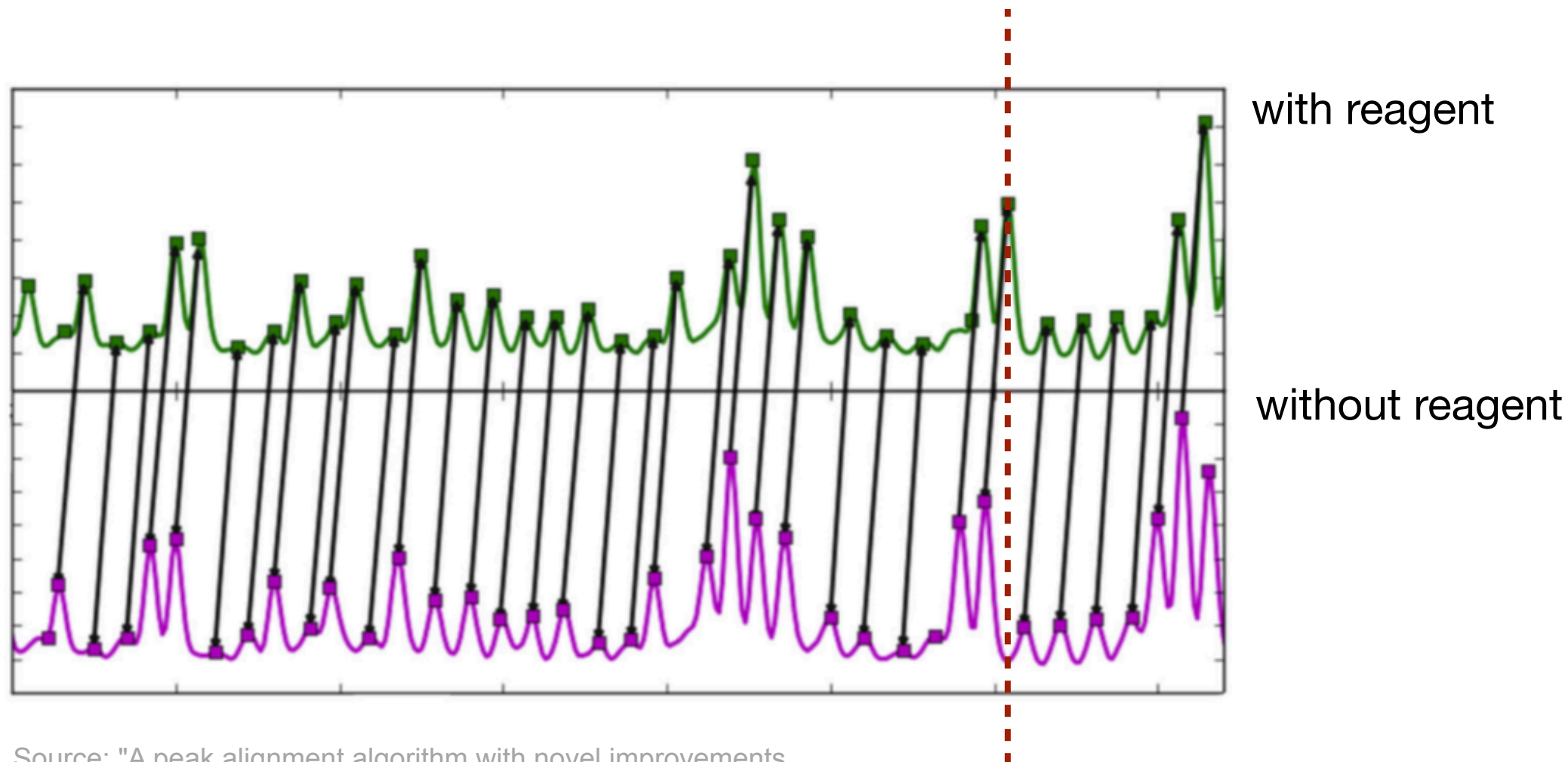


Source: "Movement prediction in rowing using a dynamic time warping base stroke detection", by Groh et al.

# Dynamic Time Warping

Peak alignment in DNA sequencing

- Use DTW to align peaks in electropherogram (a plot generated by DNA sequencer)
- Accurate alignment gives better interpretation (e.g. better RNA secondary structure prediction)



Source: "A peak alignment algorithm with novel improvements in application to electropherogram analysis", by Karabiber

# Dynamic Time Warping

## Overview of algorithm

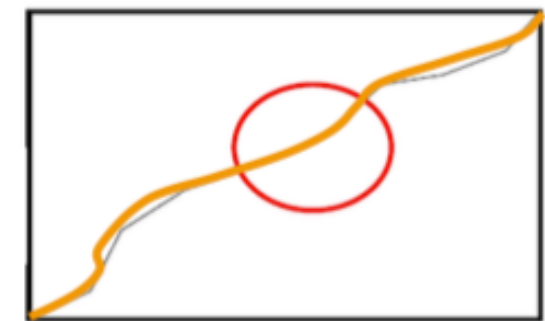
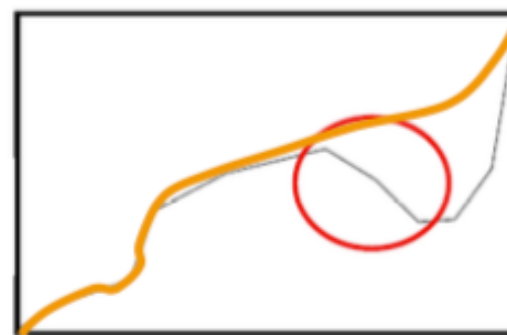
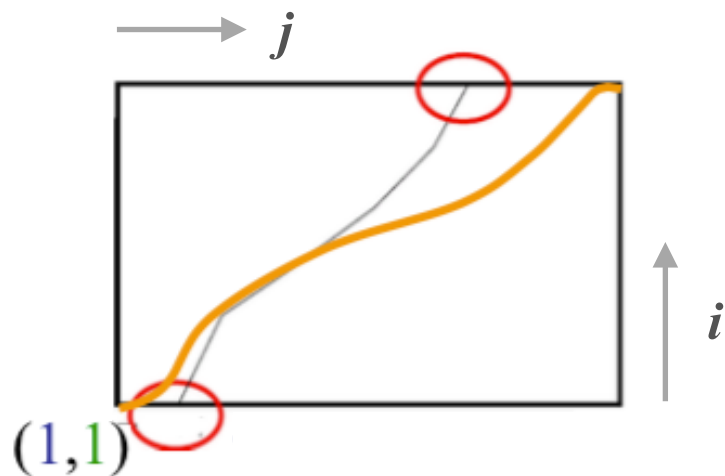
- Start by constructing  $n \times m$  matrix  $D$ , in which

$$D_{i,j} = d_s(y_i, x_j)$$

- where

$$d_s(y_i, x_j) = (y_i - x_j)^2$$

- Create a warping path  $w$  that maps points between  $x$  and  $y$ , the path  $w$  must satisfy the following:
  - Boundary conditions
  - Monotonicity
  - Continuity



Source: "Dynamic time warping algorithm", by Elena Tsiporkova

# Dynamic Time Warping

## Overview of algorithm

- DTW algorithm consists of mainly 3 parts:

- 1.Compute distance matrix
- 2.Compute accumulated cost matrix
- 3.Search the optimal path

- To start the code, import the necessary libraries, and setup a bit

```
> import numpy as np
> import matplotlib.pyplot as plt
> import pandas as pd

> plt.style.use('ggplot')
> plt.rcParams['ytick.right'] = True
> plt.rcParams['ytick.labelright'] = True
> plt.rcParams['ytick.left'] = False
> plt.rcParams['ytick.labelleft'] = False
```

# Dynamic Time Warping

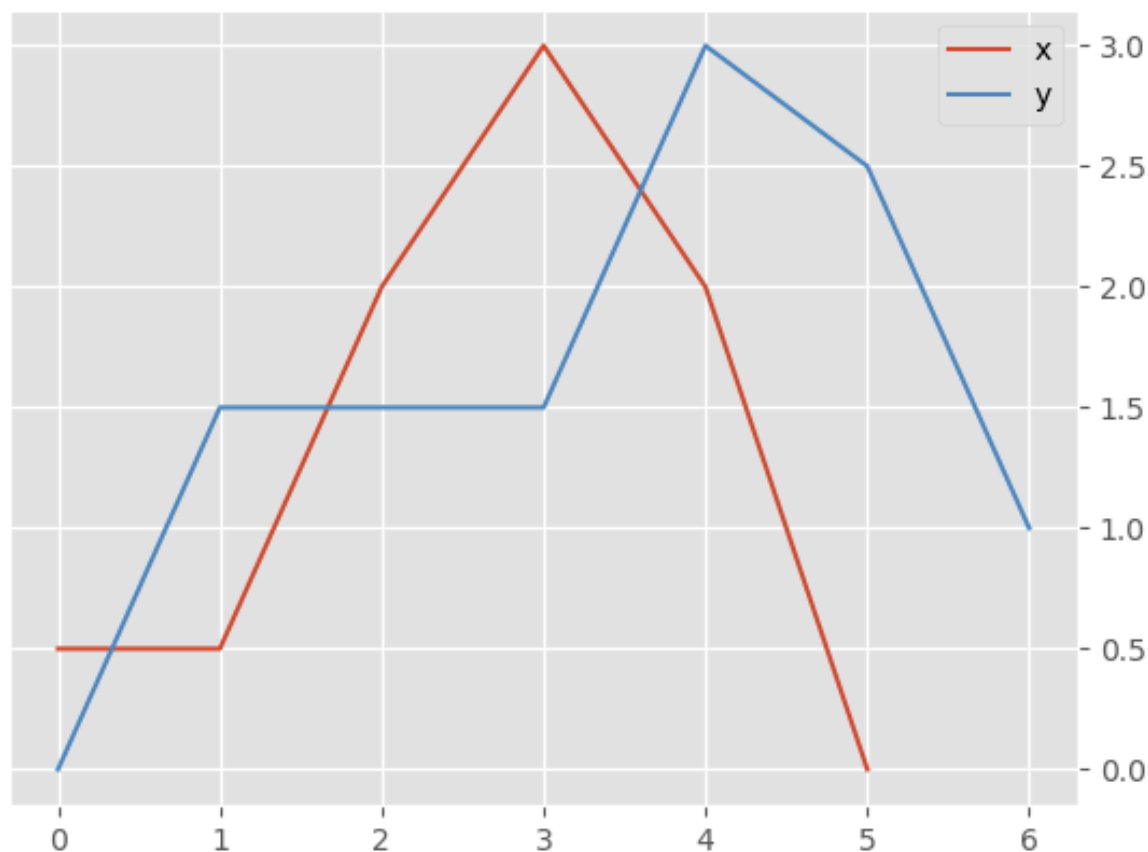
## 1. Compute distance matrix

- Define two simple short signals:

```
> x = np.array([0.5, 0.5, 2.0, 3.0, 2.0, 0.0])  
> y = np.array([0.0, 1.5, 1.5, 1.5, 3.0, 2.5, 1.0])
```

- Plot the two signals

```
> plt.figure()  
> plt.plot(x,  
           color="C0",  
           label='x')  
> plt.plot(y,  
           color="C1",  
           label='y')  
> plt.legend()
```



# Dynamic Time Warping

## 1. Compute distance matrix

- Compute the distance matrix is straightforward, since the matrix is defined as

$$D_{i,j} = d_s(y_i, x_j)$$

- and

$$d_s(y_i, x_j) = (y_i - x_j)^2$$

- The corresponding code

```
> dists = np.zeros((len(y), len(x)))
```

```
> for i in range(len(y)):
    for j in range(len(x)):
        dists[i,j] = (y[i]-x[j])**2
```

dists

```
[[0.25, 0.25, 4. , 9. , 4. , 0. ],
 [1. , 1. , 0.25, 2.25, 0.25, 2.25],
 [1. , 1. , 0.25, 2.25, 0.25, 2.25],
 [1. , 1. , 0.25, 2.25, 0.25, 2.25],
 [6.25, 6.25, 1. , 0. , 1. , 9. ],
 [4. , 4. , 0.25, 0.25, 0.25, 6.25],
 [0.25, 0.25, 1. , 4. , 1. , 1. ]]
```

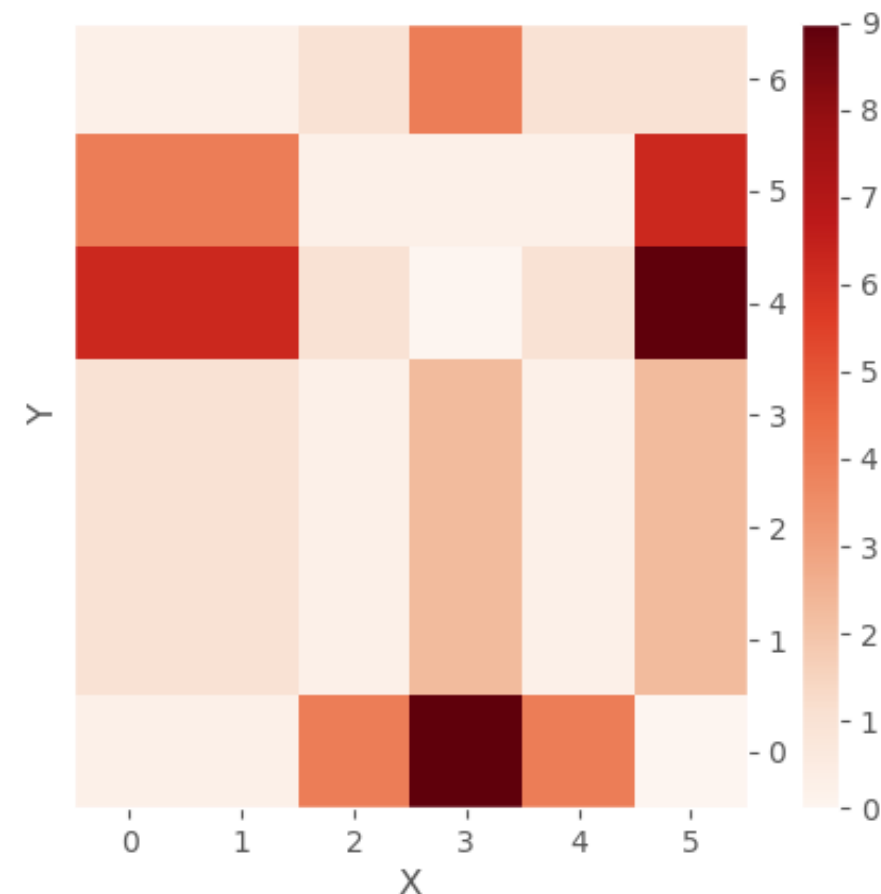
# Dynamic Time Warping

## 1. Compute distance matrix

- Create a function to do a plot on the distance matrix

```
> def pltDistances(dists,xlab="X",ylab="Y",clrmap="viridis"):  
    imgplt = plt.figure()  
    plt.imshow(dists,  
               interpolation='nearest',  
               cmap=clrmap)  
  
    plt.gca().invert_yaxis()  
    plt.xlabel(xlab)  
    plt.ylabel(ylab)  
    plt.grid()  
    plt.colorbar()  
  
    return imgplt
```

```
> pltDistances(dists,clrmap='Reds')
```

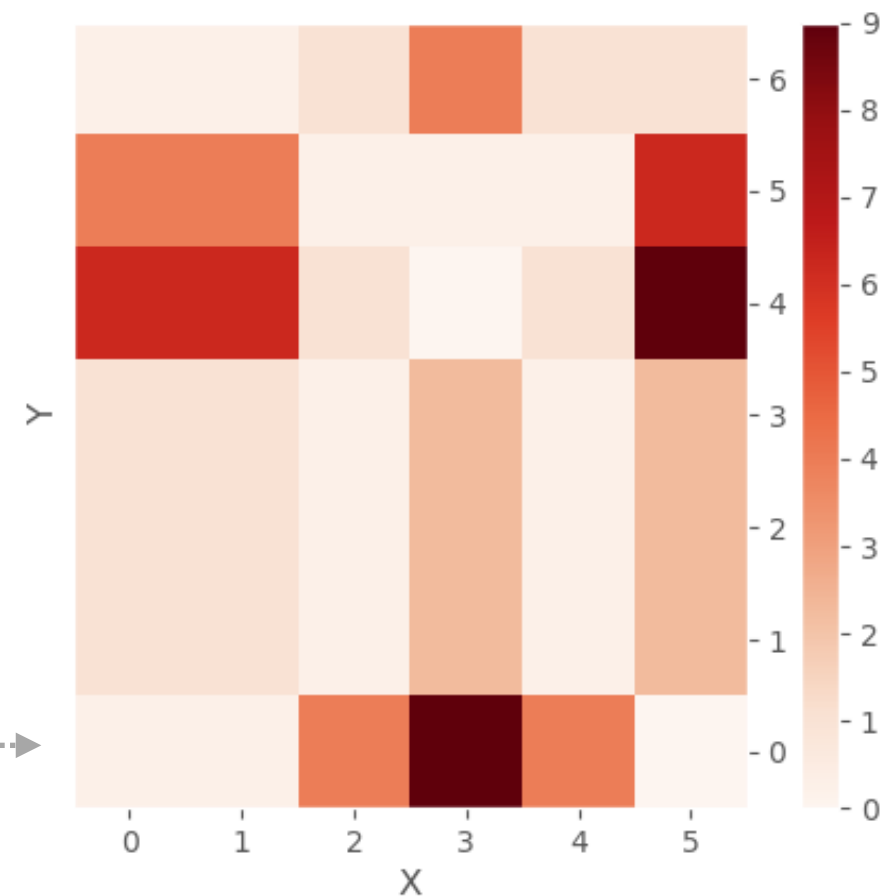


# Dynamic Time Warping

## 1. Compute distance matrix

```
[[0.25, 0.25, 4. , 9. , 4. , 0. ],  
 [1. , 1. , 0.25, 2.25, 0.25, 2.25],  
 [1. , 1. , 0.25, 2.25, 0.25, 2.25],  
 [1. , 1. , 0.25, 2.25, 0.25, 2.25],  
 [6.25, 6.25, 1. , 0. , 1. , 9. ],  
 [4. , 4. , 0.25, 0.25, 0.25, 6.25],  
 [0.25, 0.25, 1. , 4. , 1. , 1. ]]
```

- Do take note, in the plot, the y axis is inverted
- Thus, first row of the matrix corresponds to the last row in the figure





# Dynamic Time Warping

## 2. Compute accumulated cost matrix

- The accumulated cost matrix is defined

$$A_{i,j} = D_{i,j} + \min(A_{i-1,j}, A_{i,j-1}, A_{i-1,j-1})$$

- When  $i$  and  $j$  equals to 0

$$A_{0,0} = D_{0,0}$$

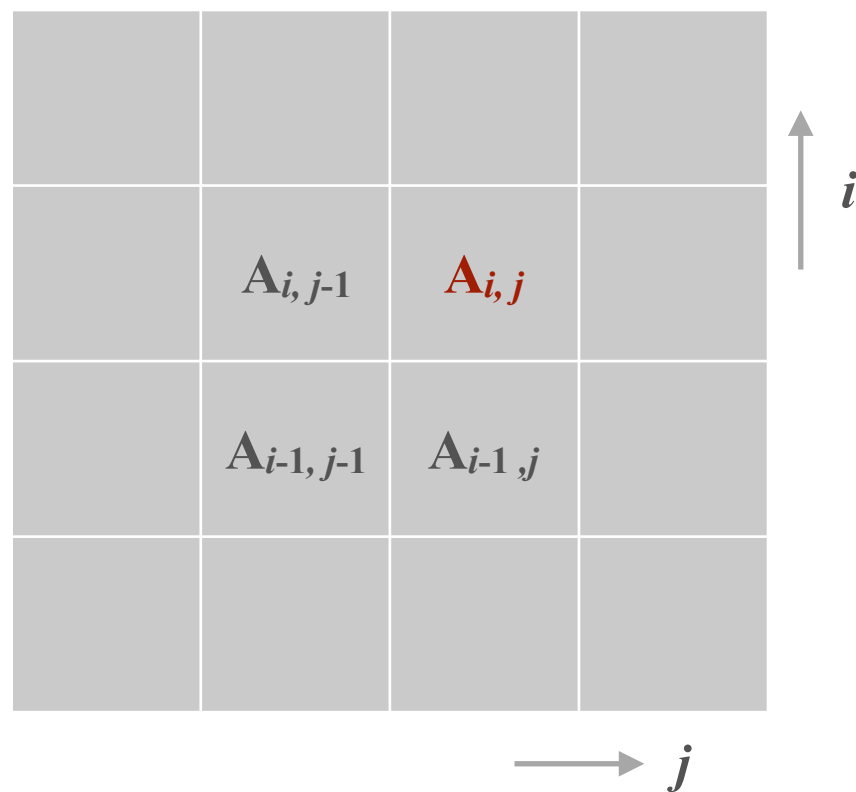
- When  $i$  equals to 0 (first row)

$$A_{0,j} = D_{0,j} + A_{0,j-1}$$

- When  $j$  equals to 0 (first column)

$$A_{i,0} = D_{i,0} + A_{i-1,0}$$

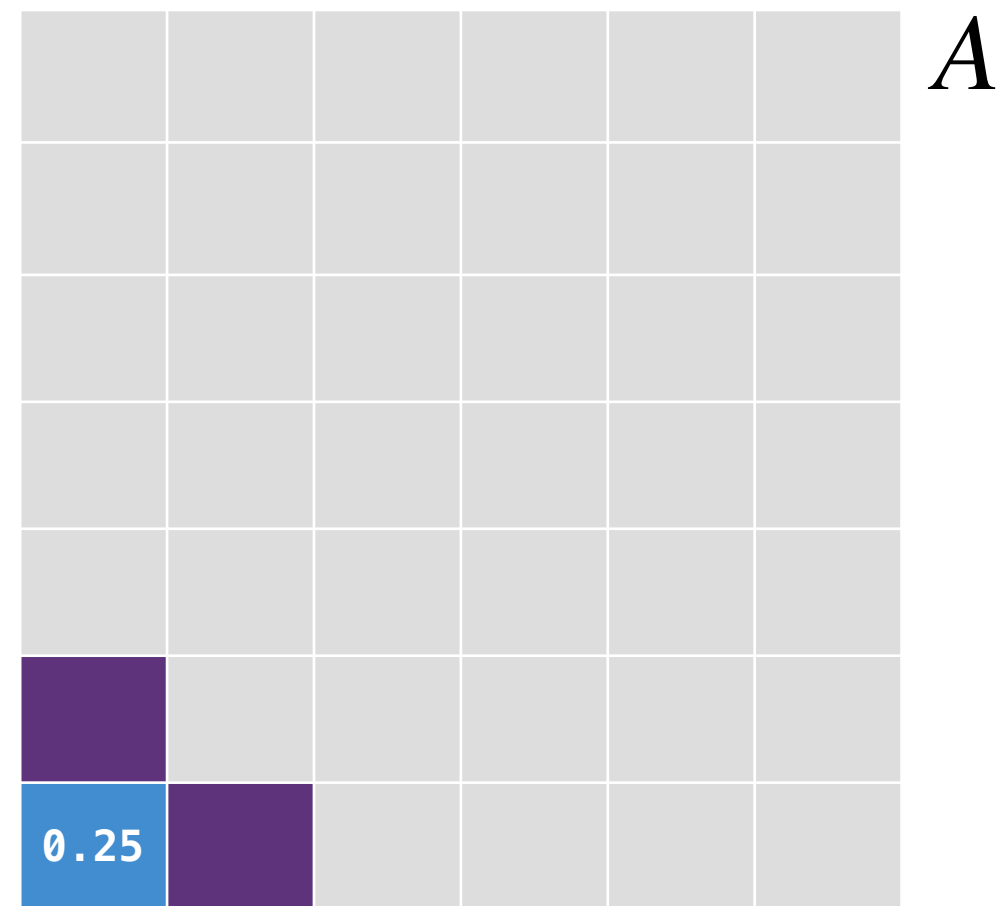
$A_{i,j}$  equals to  $D_{i,j}$  plus either  $A_{i-1,j-1}$ ,  $A_{i,j-1}$  or  $A_{i-1,j}$ , whichever has the lowest value



## 2. Compute accumulated cost matrix

- When  $j$  equals to 0 (first column)

$$A_{i,0} = D_{i,0} + A_{i-1,0}$$



# Dynamic Time Warping

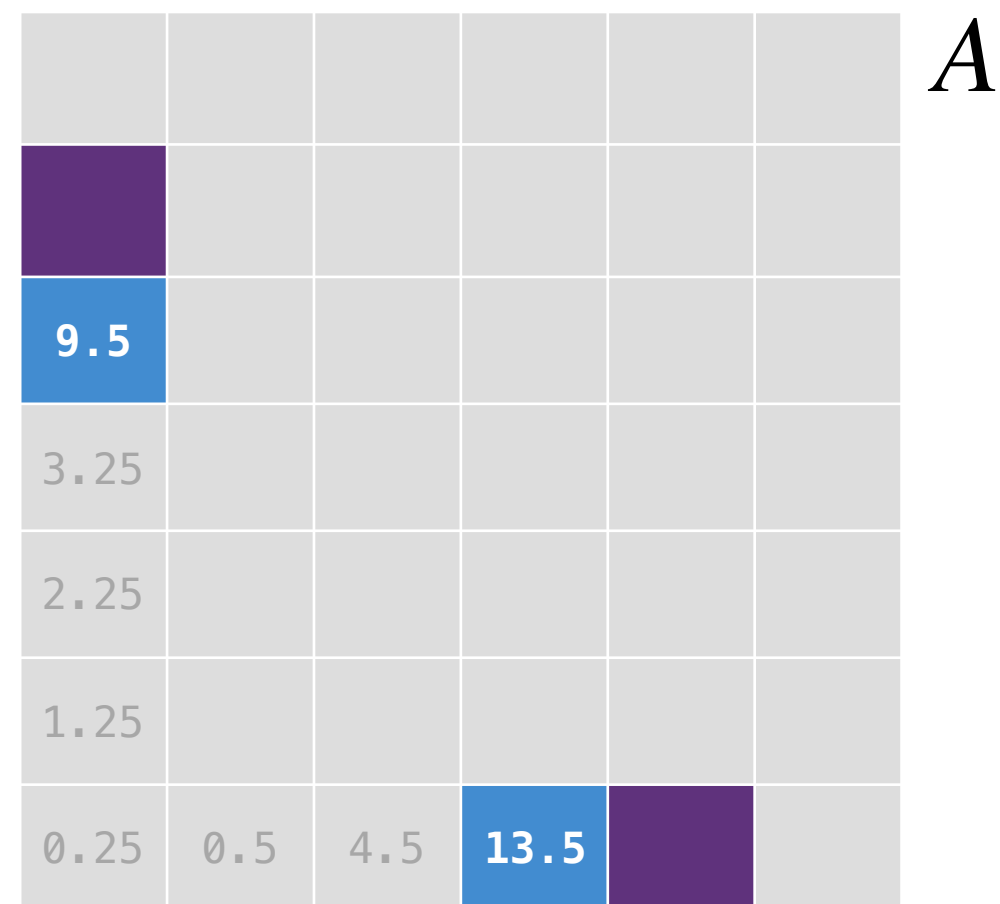
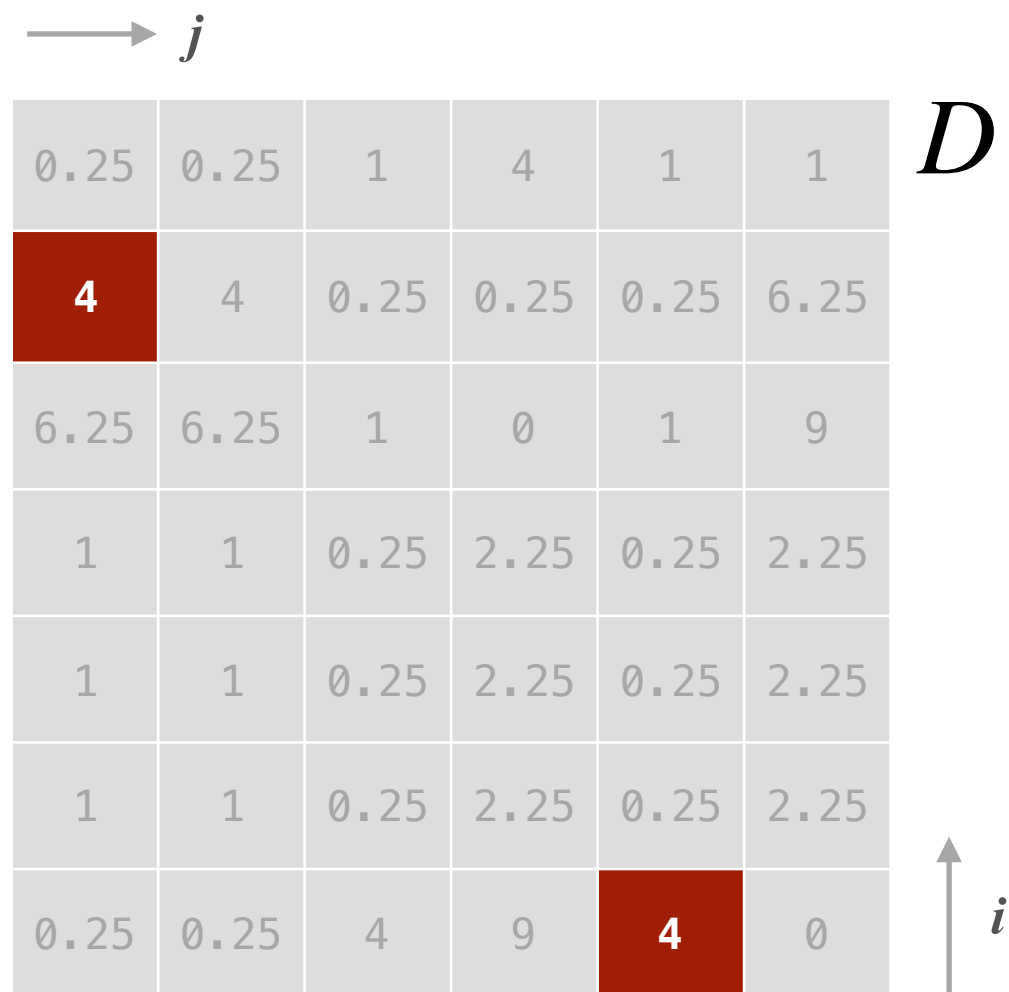
2. Compute accumulated cost matrix

- When  $i$  equals to 0 (first row)

$$A_{0,j} = D_{0,j} + A_{0,j-1}$$

- When  $j$  equals to 0 (first column)

$$A_{i,0} = D_{i,0} + A_{i-1,0}$$

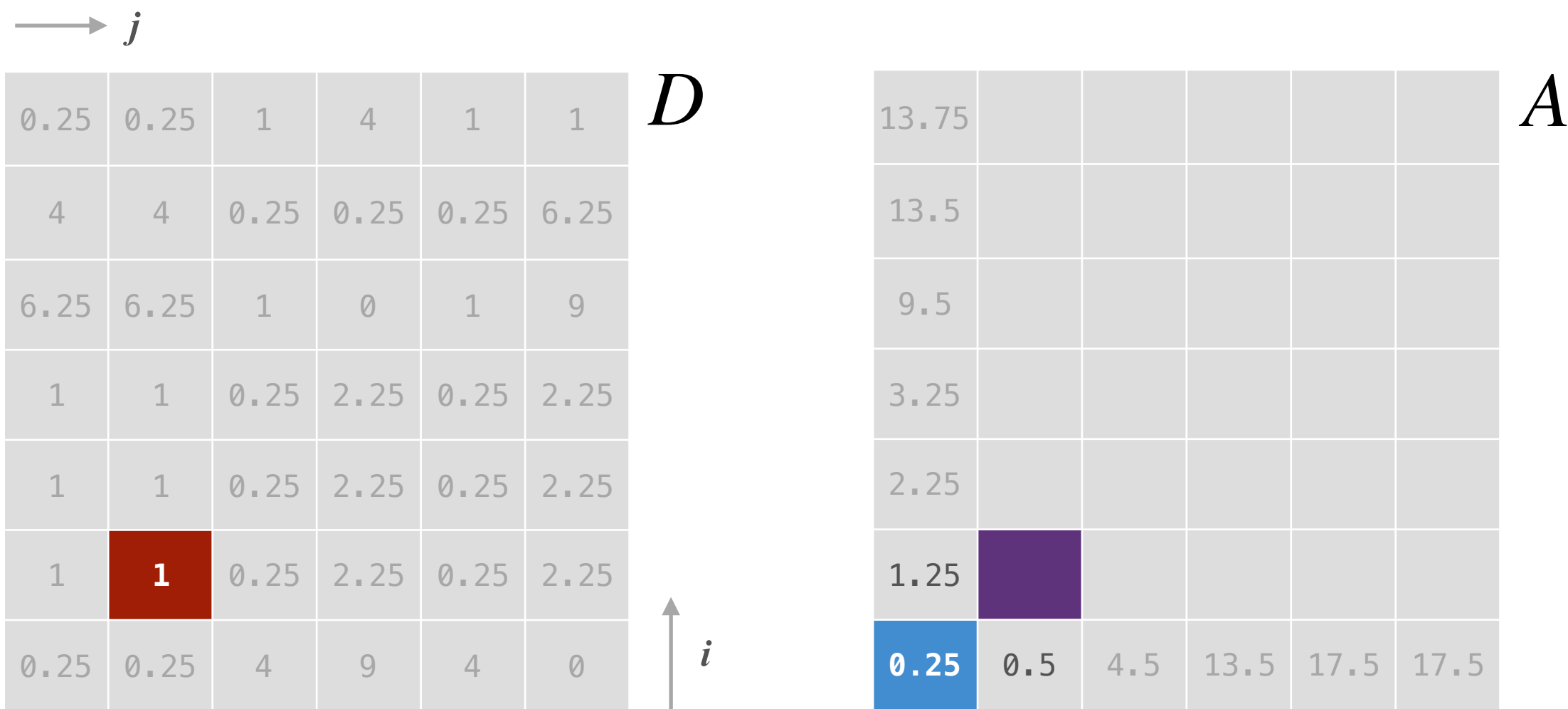


# Dynamic Time Warping

2. Compute accumulated cost matrix

•Else

$$A_{i,j} = D_{i,j} + \min (A_{i-1,j}, A_{i,j-1}, A_{i-1,j-1})$$



# Dynamic Time Warping

2. Compute accumulated cost matrix

•Else

$$A_{i,j} = D_{i,j} + \min(A_{i-1,j}, A_{i,j-1}, A_{i-1,j-1})$$

→ *j*

0.25	0.25	1	4	1	1
4	4	0.25	0.25	0.25	6.25
6.25	6.25	1	0	1	9
1	1	0.25	2.25	0.25	2.25
1	1	0.25	2.25	0.25	2.25
1	1	0.25	2.25	0.25	2.25
0.25	0.25	4	9	4	0

*D*

↑  
*i*

13.75					
13.5					
9.5	9.5	2.25	1.25	2.25	
3.25	3.25	1.25	3.25	3.25	5.5
2.25	2.25	1	3	3.25	5.5
1.25	1.25	0.75	3	3.25	5.5
0.25	0.5	4.5	13.5	17.5	17.5

*A*

# Dynamic Time Warping

2. Compute accumulated cost matrix

- Name the accumulated cost matrix as `acuCost`. It has the same shape as `dists`.

$$A_{0,0} = D_{0,0}$$

$$A_{0,j} = D_{0,j} + A_{0,j-1}$$

*first row*

$$A_{i,0} = D_{i,0} + A_{i-1,0}$$

*first column*

$$A_{i,j} = D_{i,j} + \min(A_{i-1,j}, A_{i,j-1}, A_{i-1,j-1})$$

```
> acuCost = np.zeros(dists.shape)
> acuCost[0,0] = dists[0,0]

> for j in range(1,dists.shape[1]):
    acuCost[0,j] = dists[0,j]+acuCost[0,j-1]

> for i in range(1,dists.shape[0]):
    acuCost[i,0] = dists[i,0]+acuCost[i-1,0]

> for i in range(1,dists.shape[0]):
    for j in range(1,dists.shape[1]):
        acuCost[i,j] = min(acuCost[i-1,j-1],
                           acuCost[i-1,j],
                           acuCost[i,j-1])+dists[i,j]

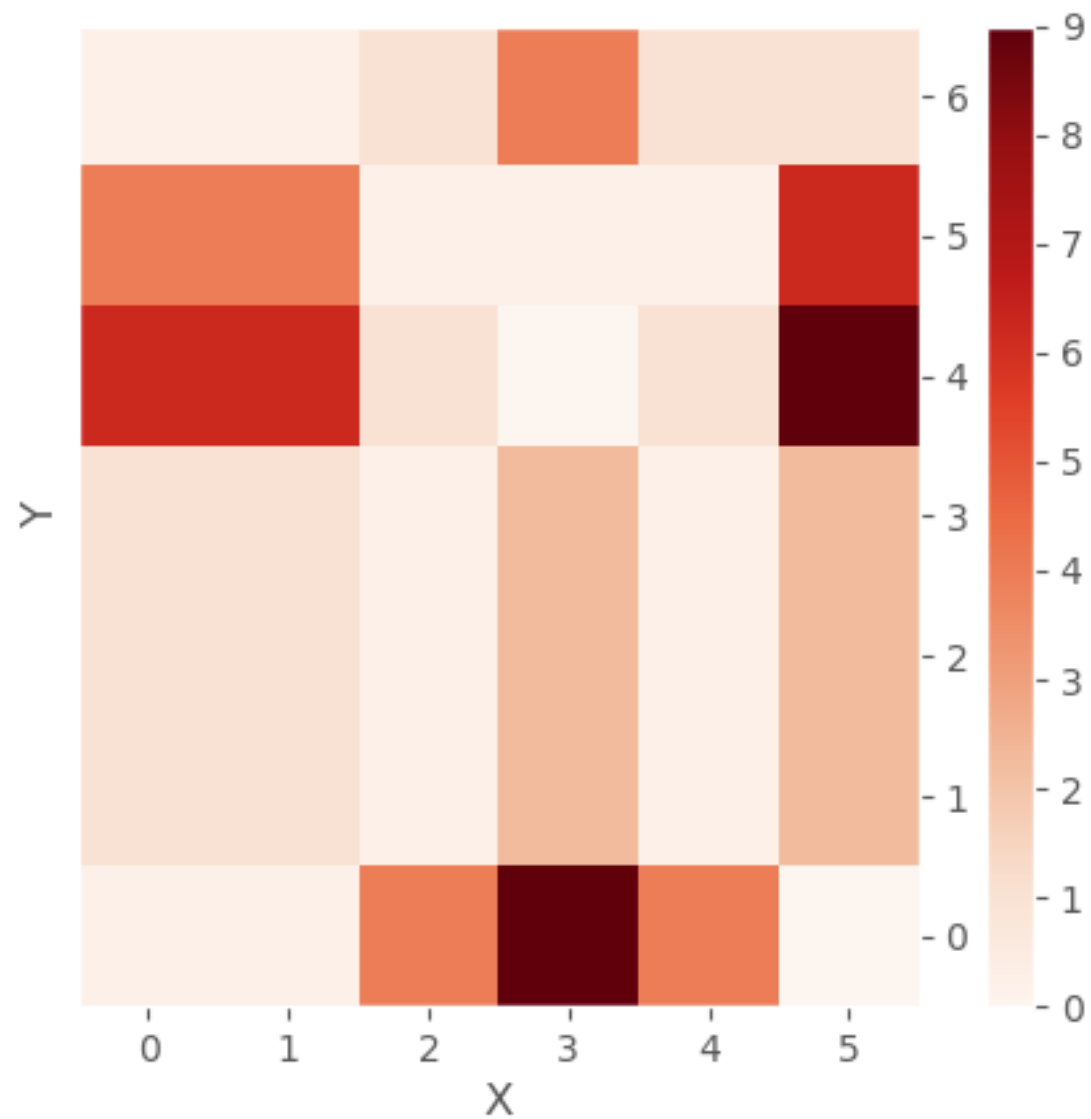
> pltDistances(acuCost,clrmap='Reds')
```

# Dynamic Time Warping

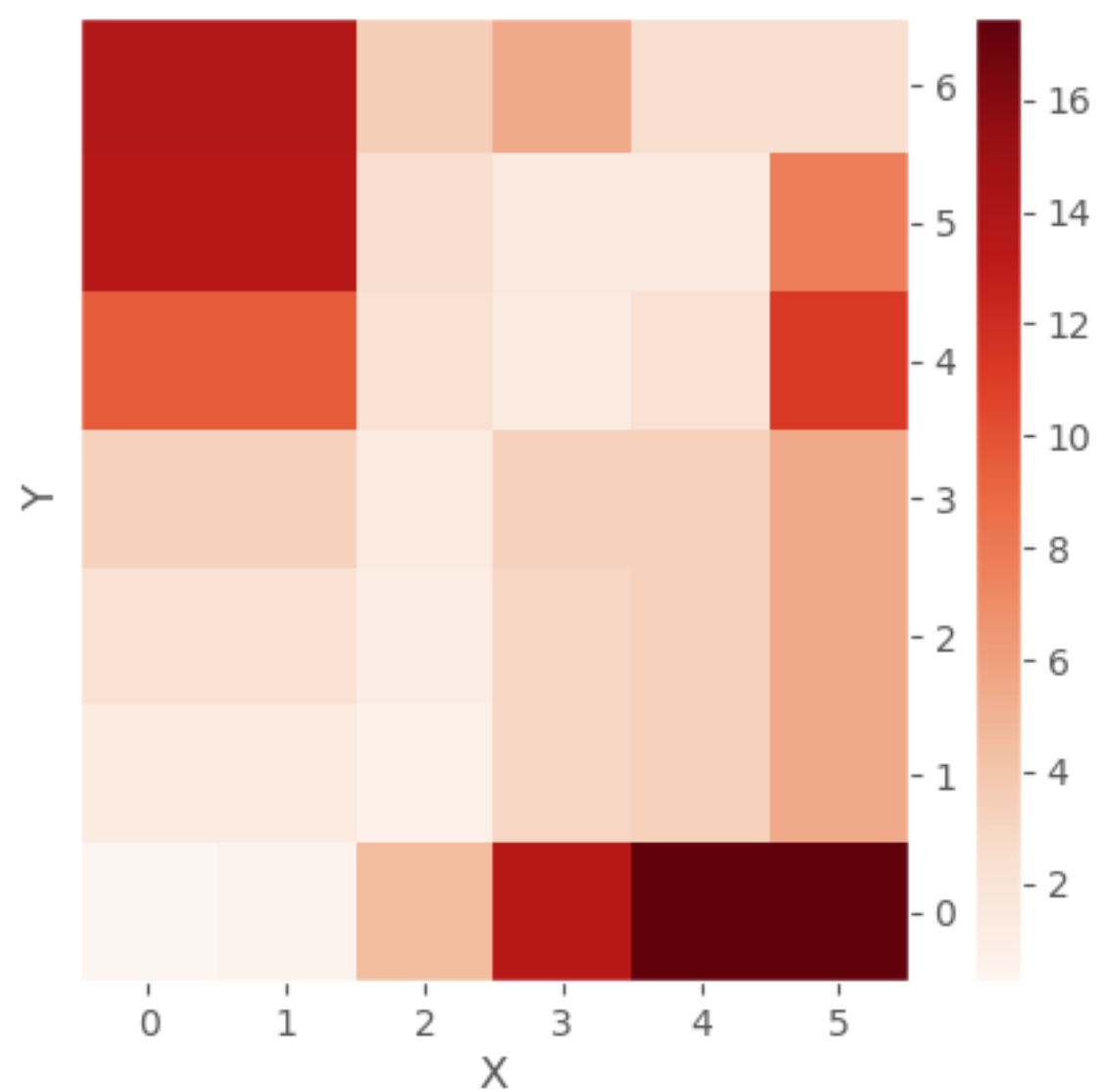
2. Compute accumulated cost matrix

- Can you see the warping / optimal path?

dists



acuCost

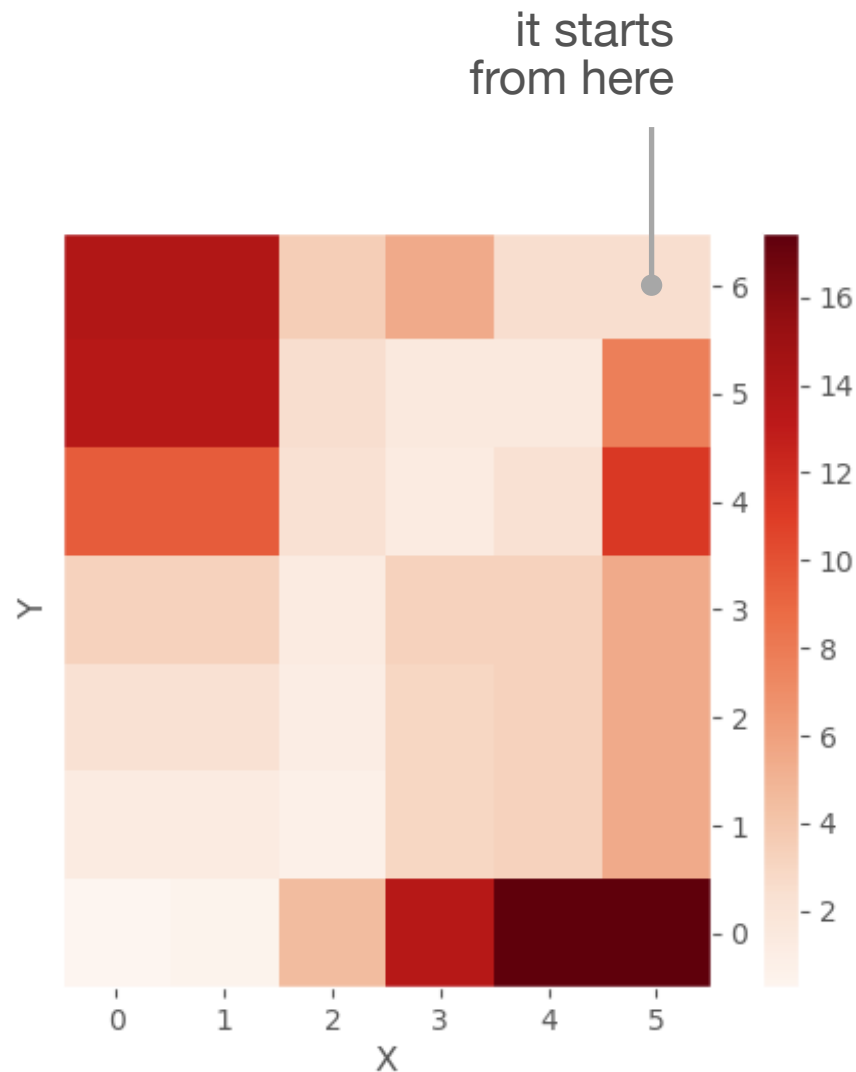


# Dynamic Time Warping

## 3. Search the optimal path

- Name the warping path as `path`.

```
> i      = len(y)-1
> j      = len(x)-1
> path   = [[j,i]]
> while (i > 0) and (j > 0):
    if i==0:
        j    = j-1
    elif j==0:
        i    = i-1
    else:
        if acuCost[i-1,j] == min(acuCost[i-1,j-1],
                                   acuCost[i-1,j],
                                   acuCost[i,j-1]):
            i    = i-1
        elif acuCost[i,j-1] == min(acuCost[i-1,j-1],
                                     acuCost[i-1,j],
                                     acuCost[i,j-1]):
            j    = j-1
        else:
            i    = i-1
            j    = j-1
        path.append([j,i])
> path.append([0,0])
```



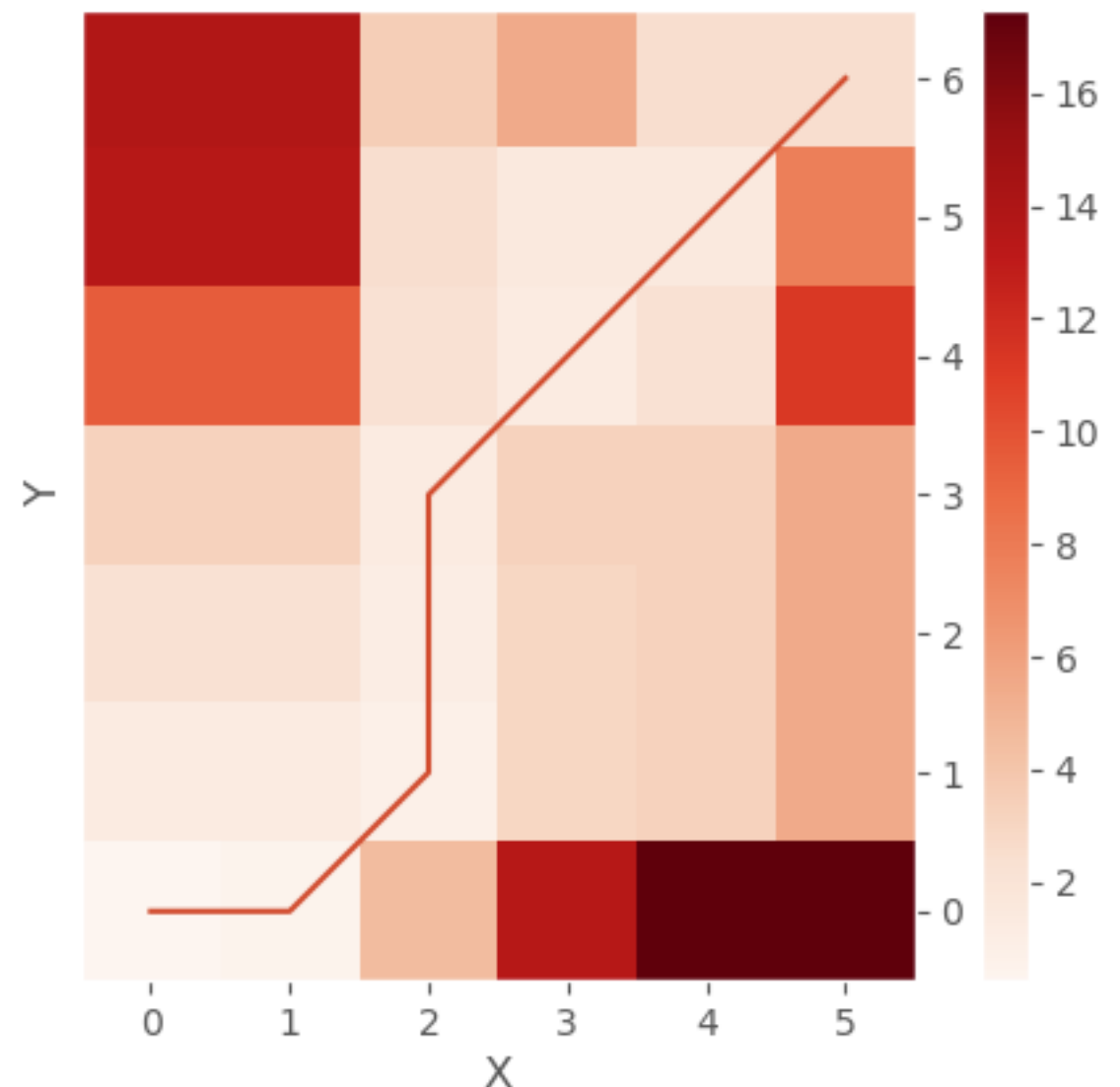


# Dynamic Time Warping

## 3. Search the optimal path

- Create a function that plots the path

```
> def pltCostAndPath(acuCost,path,xlab="X",ylab="Y",clrmap="viridis"):  
    px      = [pt[0] for pt in path]  
    py      = [pt[1] for pt in path]  
  
    imgplt  = pltDistances(acuCost,  
                           xlab=xlab,  
                           ylab=ylab,  
                           clrmap=clrmap)  
  
    plt.plot(px,py)  
  
    return imgplt  
  
> pltCostAndPath(acuCost,path,clrmap='Reds')
```



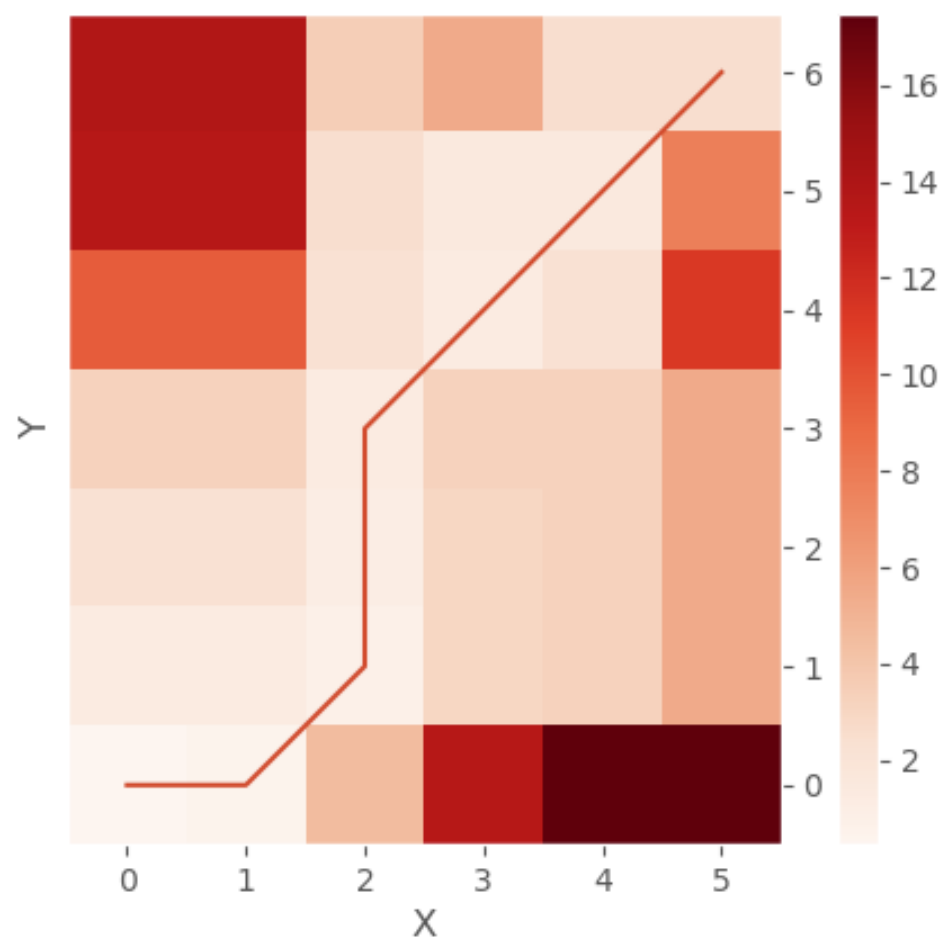
# Dynamic Time Warping

## 3. Search the optimal path

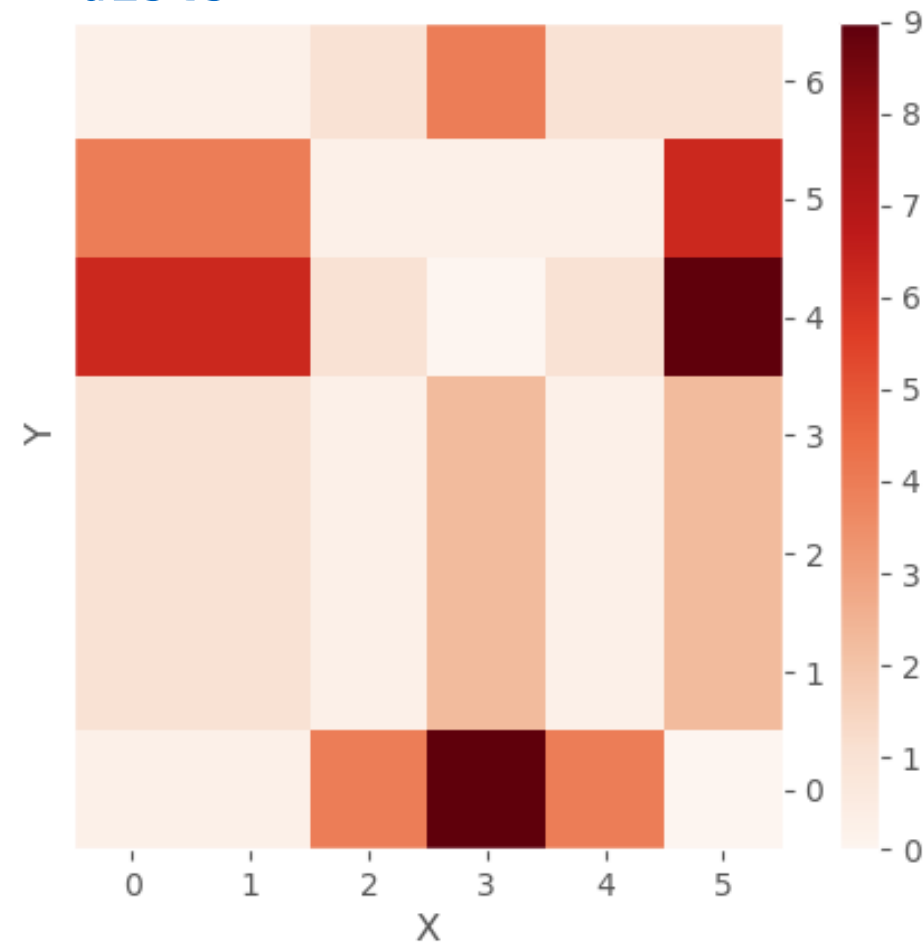
- Calculate the cost based on **dists**, which can be considered as a measure for similarity / distance

```
> cost          = 0
> for [j,i] in path:
    cost      = cost+dists[i,j]
> cost
: 2.5
```

acuCost



dists

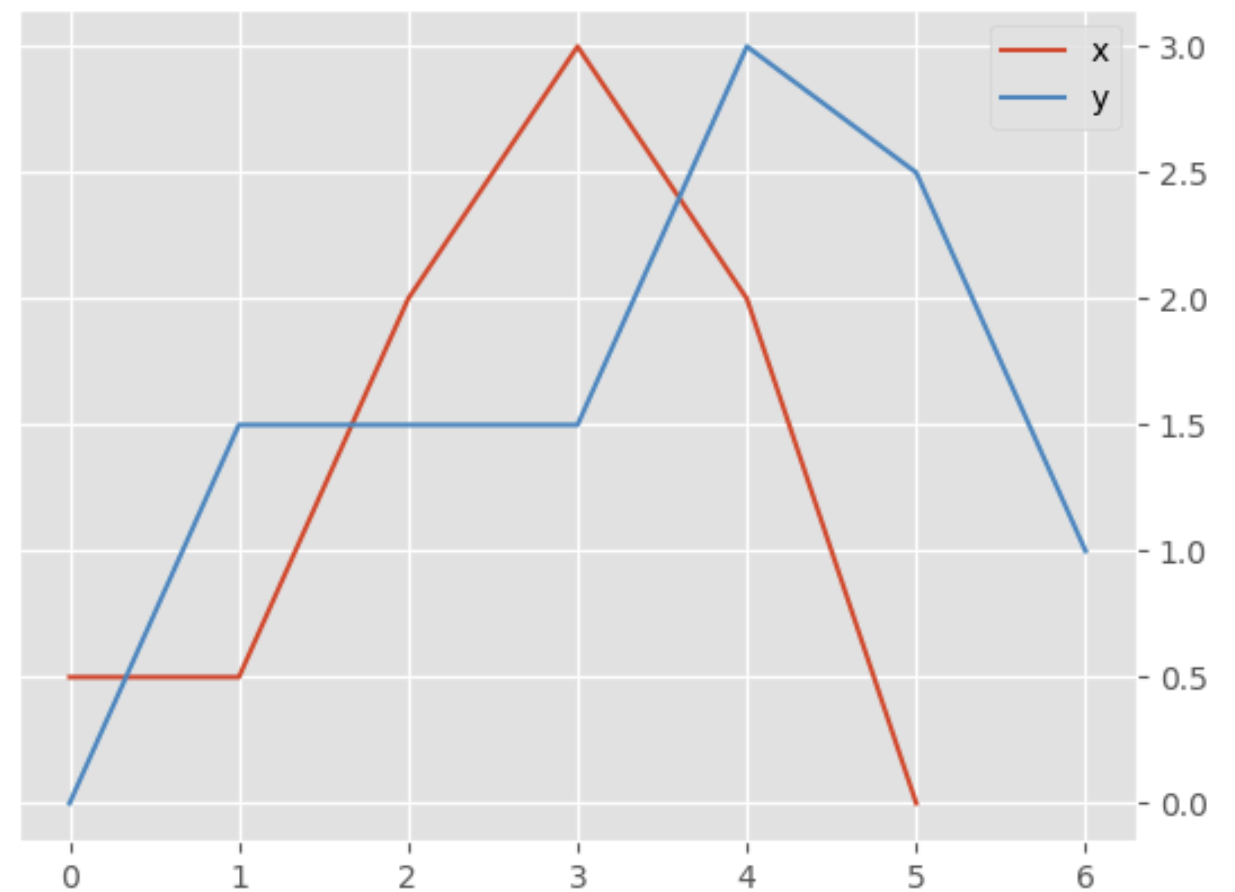
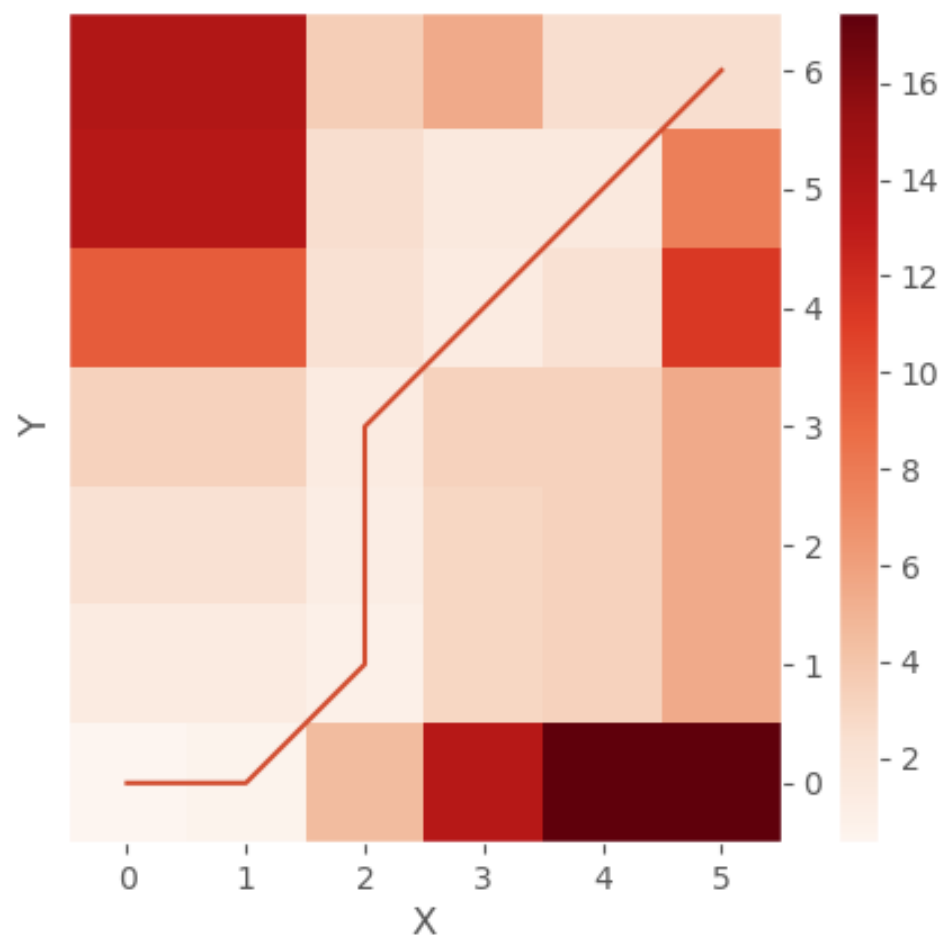


# Dynamic Time Warping

## 3. Search the optimal path

- The implication

acuCost



# Dynamic Time Warping

## 3. Search the optimal path

- Plot the mapping of points between two signals

```
> def pltWarp(s1,s2,path,xlab="idx",ylab="Value"):  
    imgplt      = plt.figure()
```

```
    for [idx1,idx2] in path:  
        plt.plot([idx1,idx2],[s1[idx1],s2[idx2]],  
                 color="C4",  
                 linewidth=2)
```

```
    plt.plot(s1,  
             'o-',  
             color="C0",  
             markersize=3)
```

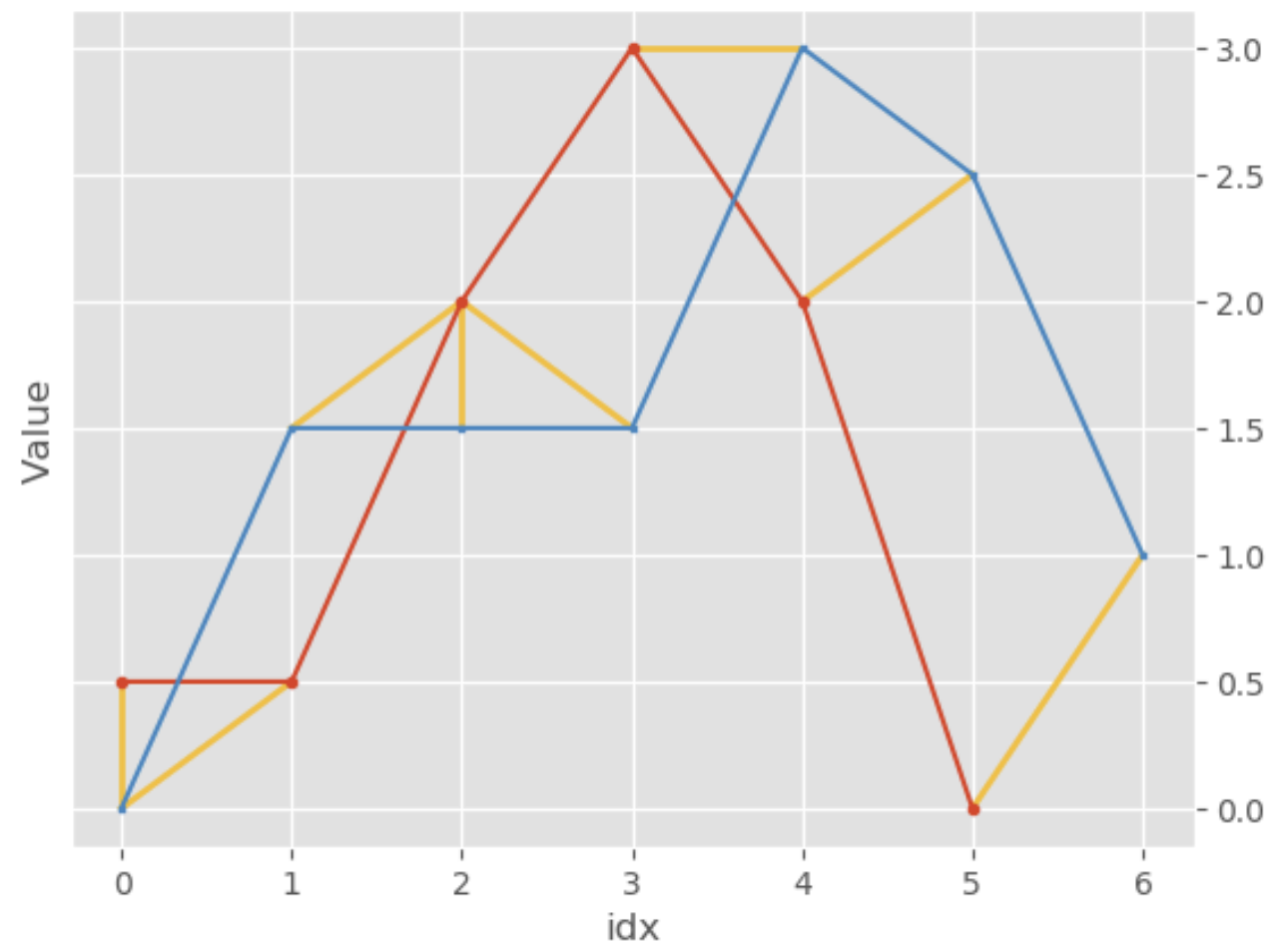
```
    plt.plot(s2,  
             's-',  
             color="C1",  
             markersize=2)
```

```
    plt.xlabel(xlab)  
    plt.ylabel(ylab)
```

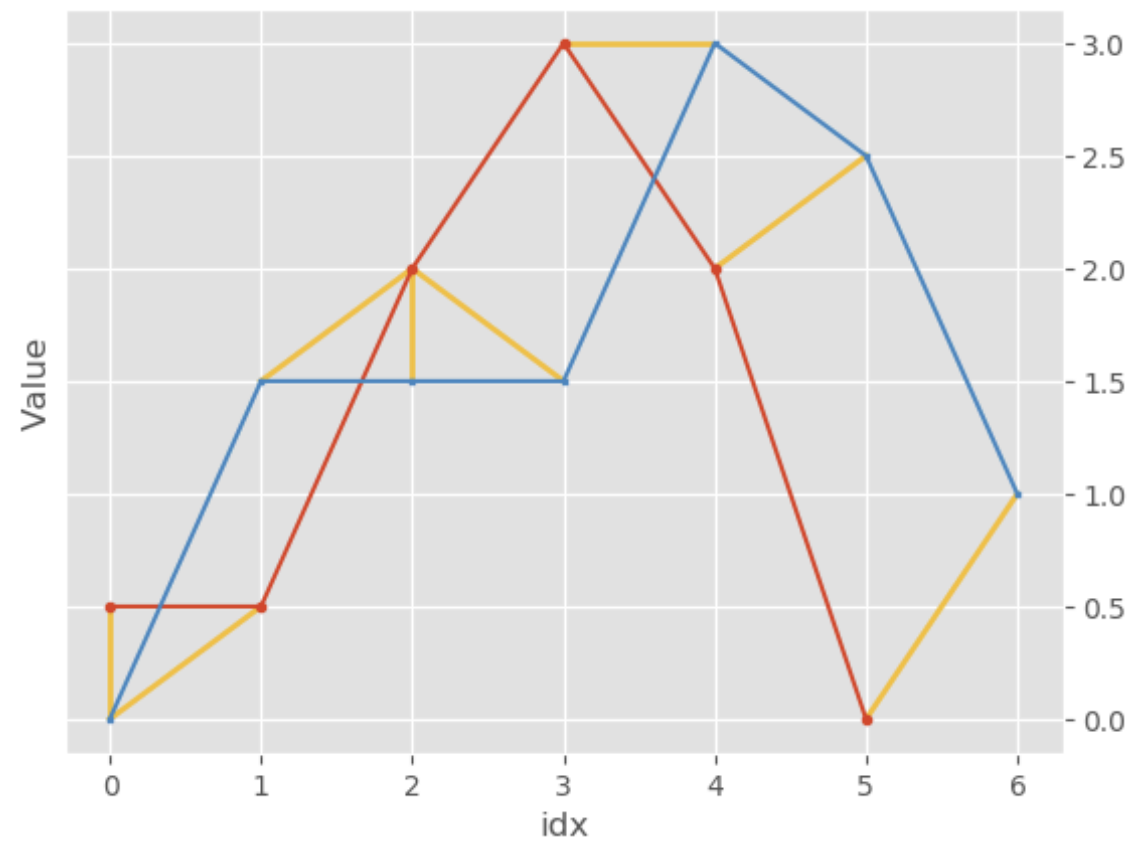
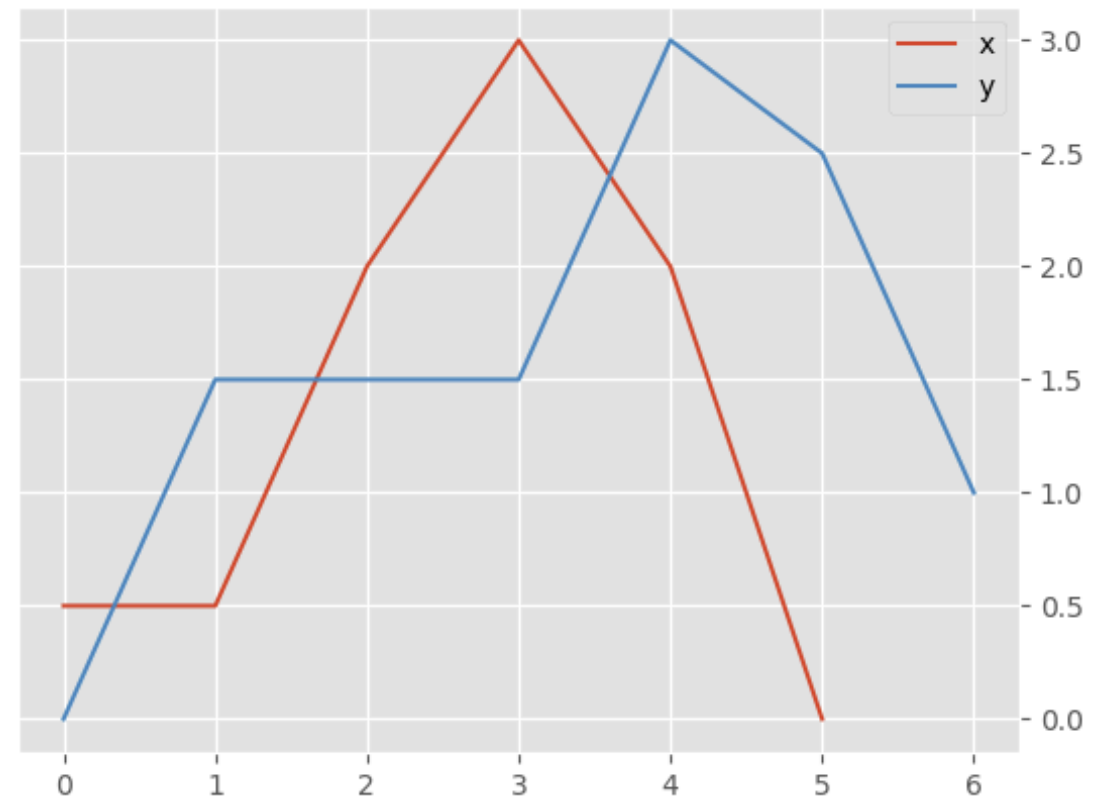
```
    return imgplt
```

```
> pltWarp(x,y,path)
```

*Plot the connections  
between s1 and s2 (yellow  
lines)*



# Dynamic Time Warping



# Dynamic Time Warping

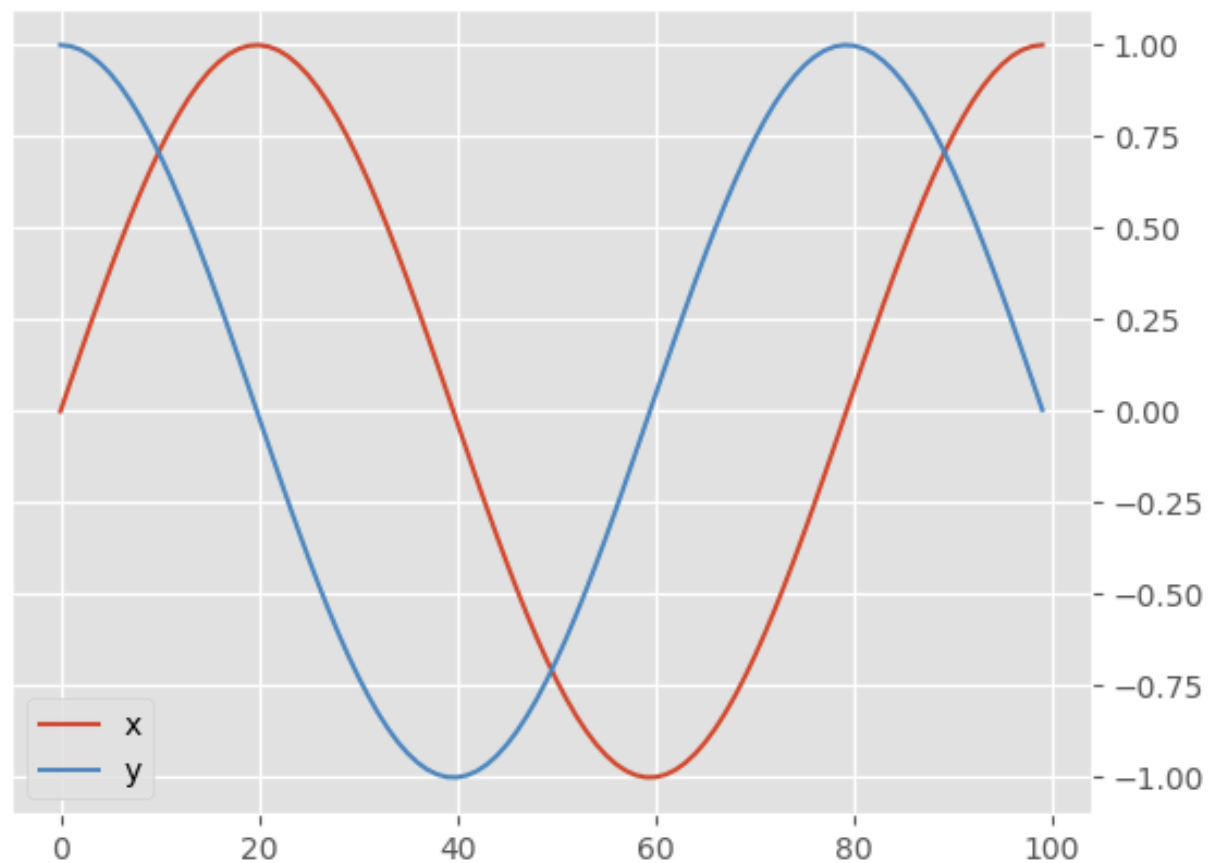
Another example

- Define two signals as:

```
> x = np.sin(np.linspace(0,7.85,100))  
> y = np.cos(np.linspace(0,7.85,100))
```

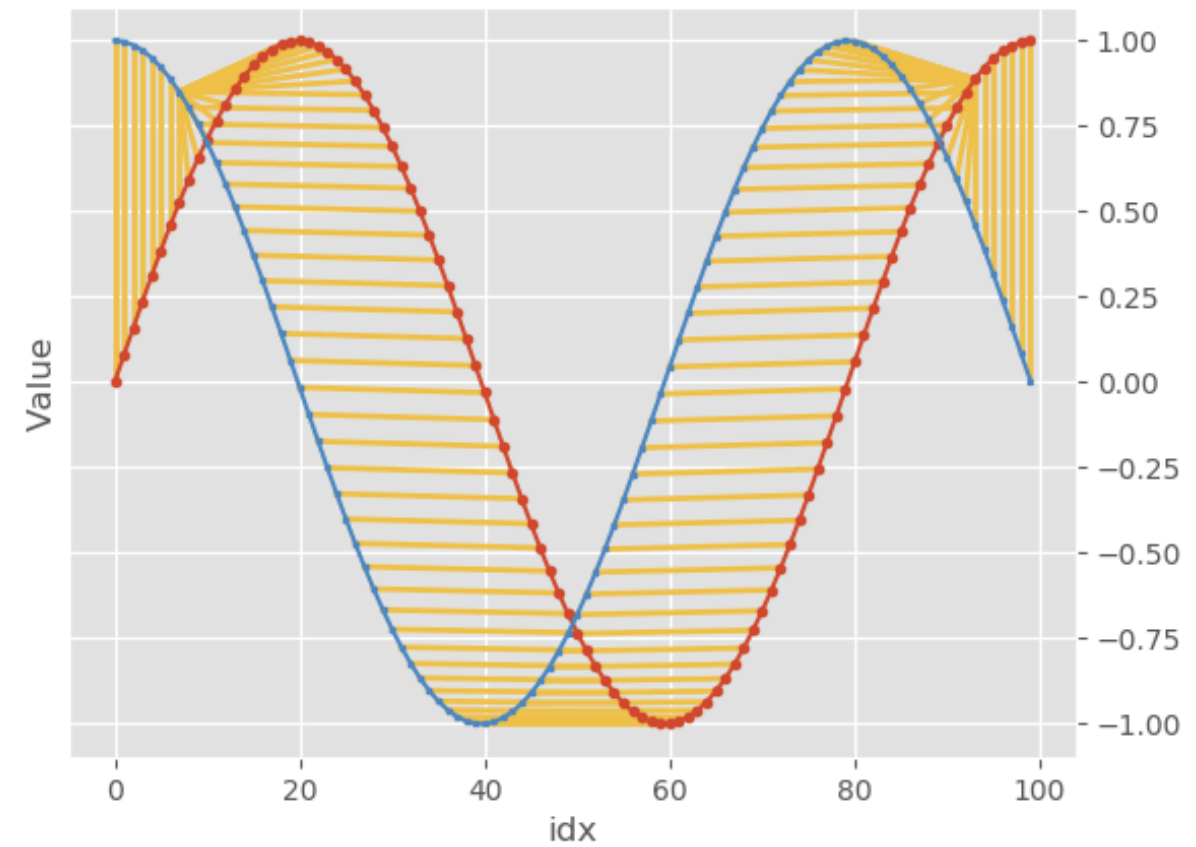
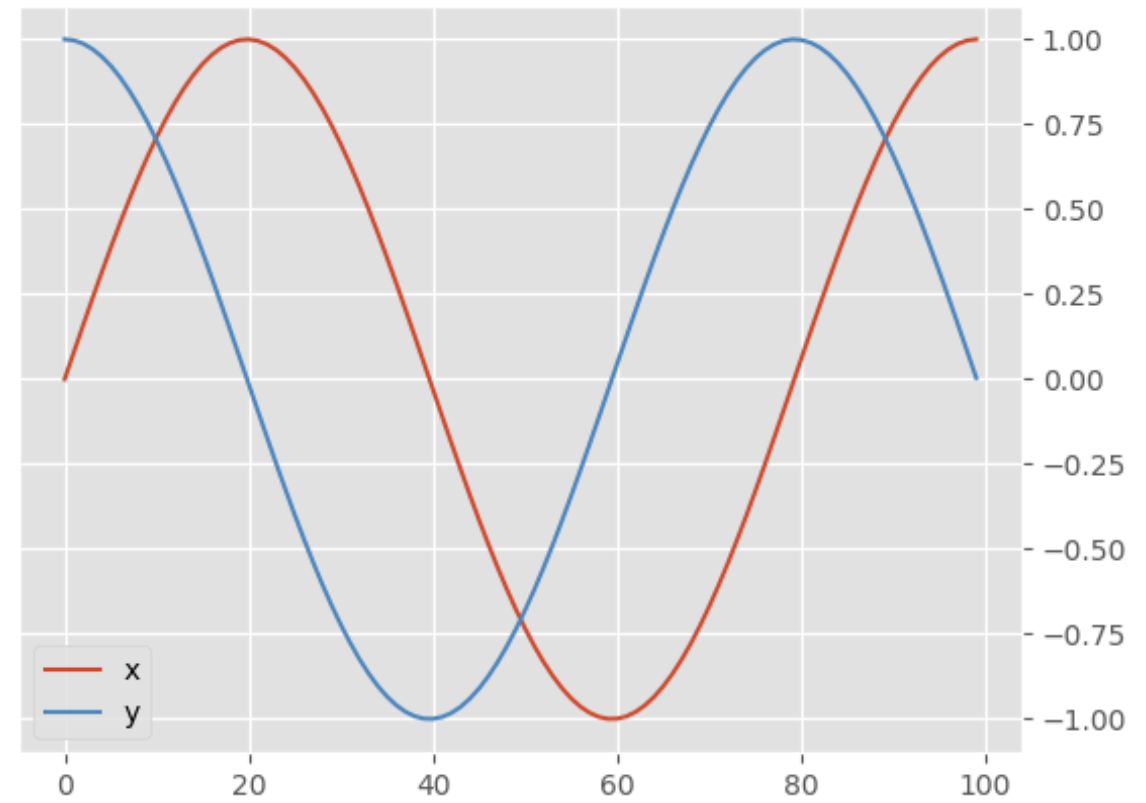
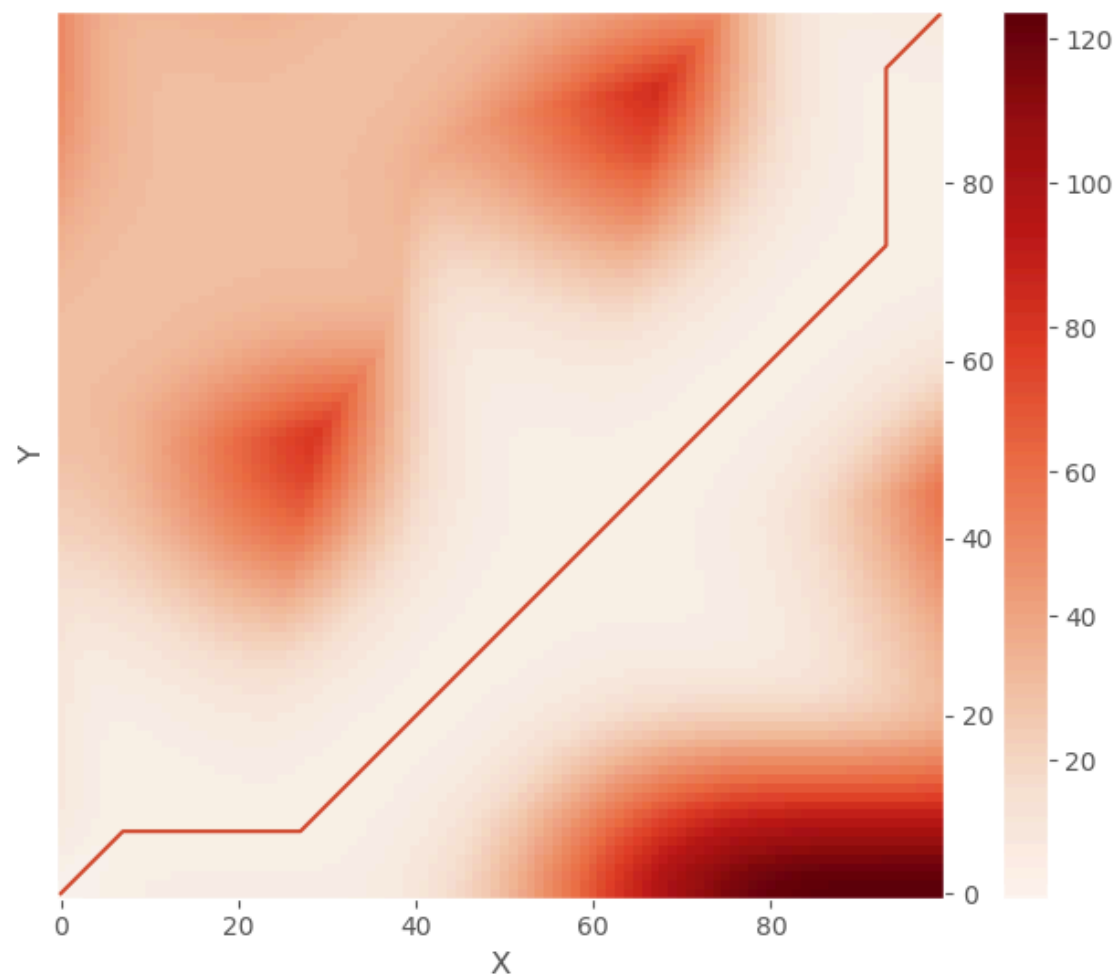
- Plot the two signals

```
> plt.figure()  
> plt.plot(x,  
           color="C0",  
           label='x')  
> plt.plot(y,  
           color="C1",  
           label='y')  
> plt.legend()
```



# Dynamic Time Warping

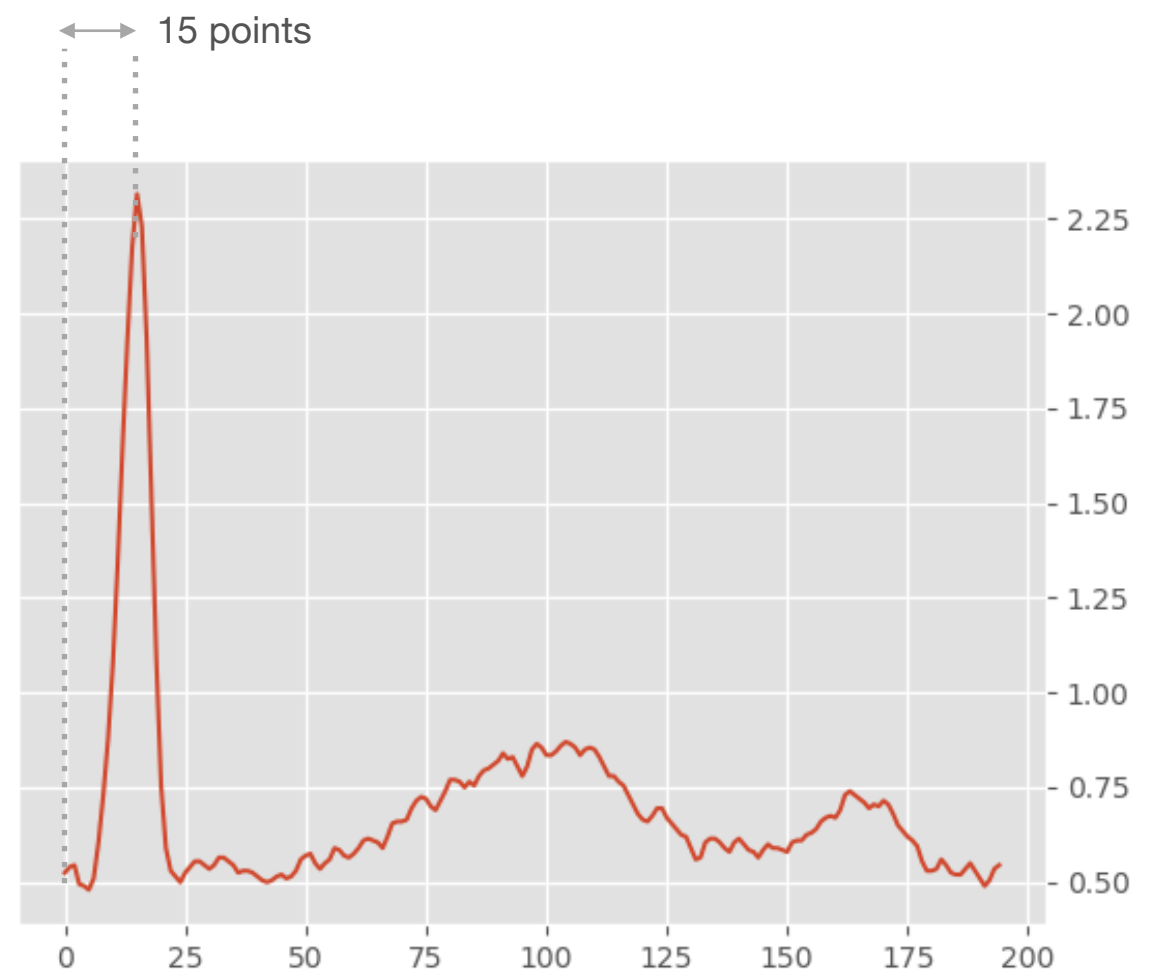
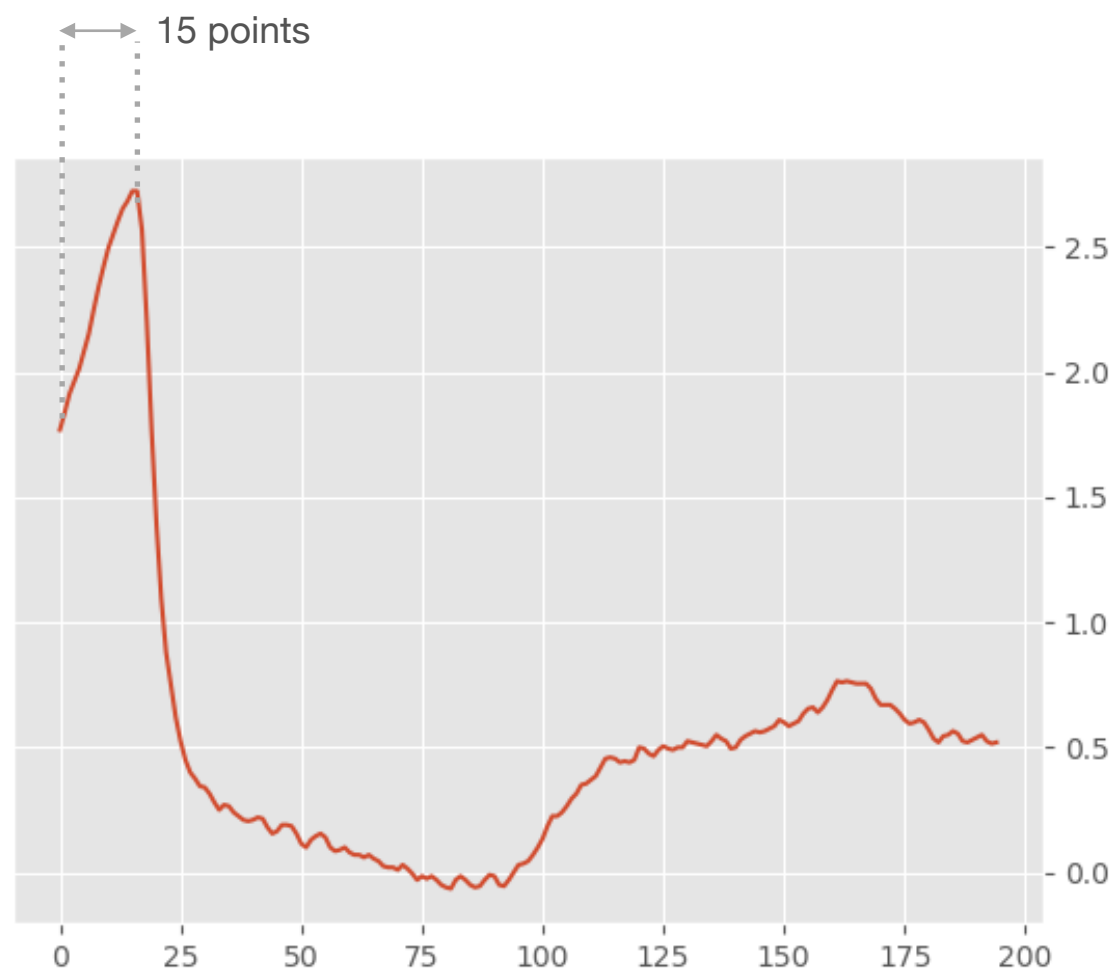
Another example



# Back to the problem

How to start?

- Before we compute similarity, must segment individual heartbeat signal
- With signal segmented, perform DTW





# Workshop

To start

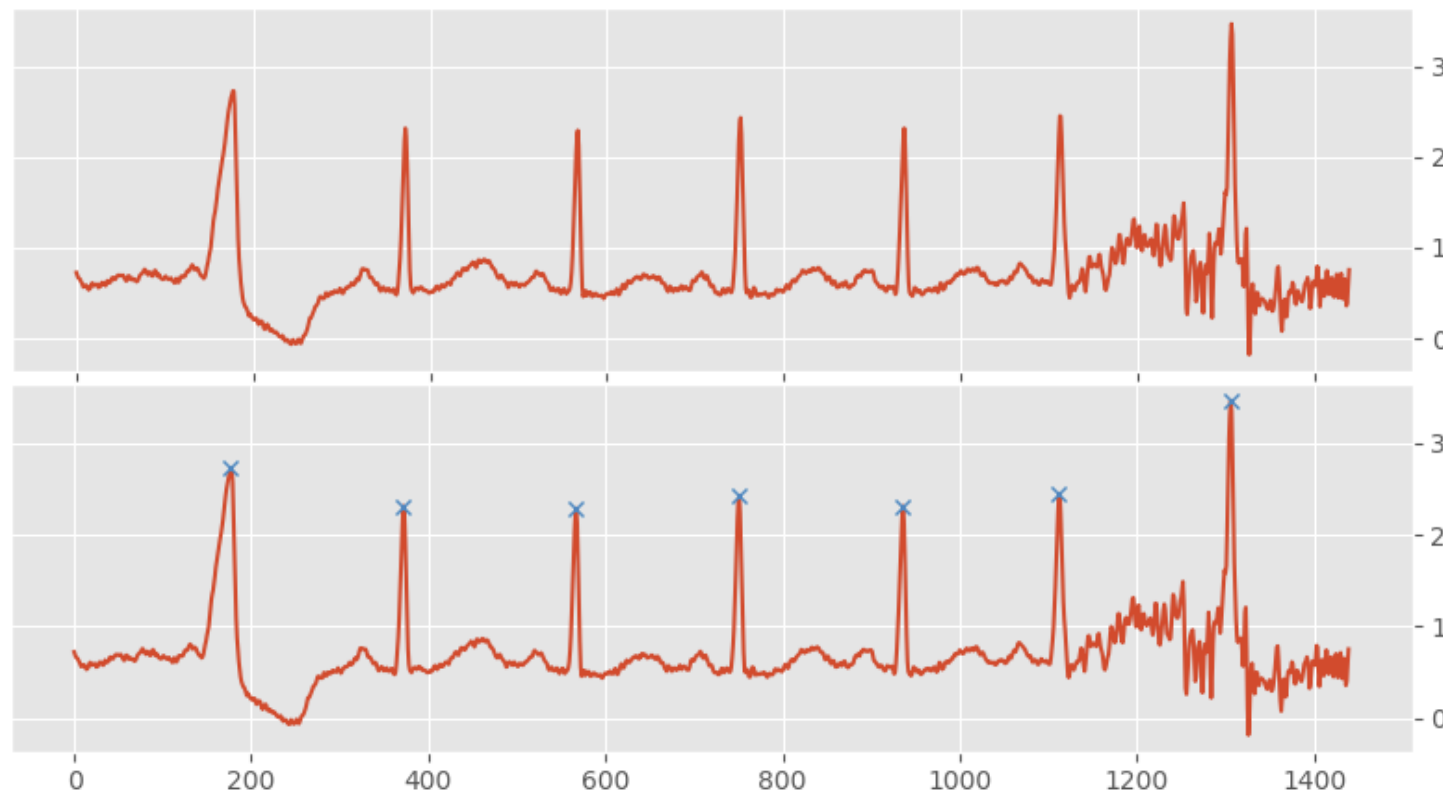
- Load the data

```
> l2D      = pd.read_csv('ecg2D.csv',  
                        header=None)
```

```
> ECGs     = l2D[1].values
```

- Create a function with the below signature. The output is a list consists of all the ECG segments in a ECG signal

```
def extractECG(ecg,pks,offset=15):
```



# Workshop

Compute the accumulated costs,  
plot the optimal paths and warp

- Make the comparisons between segment

1 and 2

2 and 3

2 and 6

