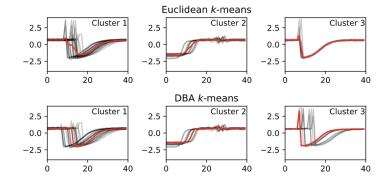
# **Time Series Clustering**

K-means

#### k-means clustering for time series: Similarity function

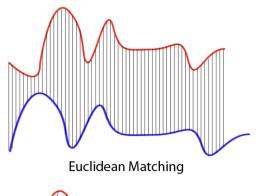
- Many variants of the algorithm are available, two of them:
  - Euclidean k-means
  - DBA-*k*-means (for DTW Barycenter Averaging)

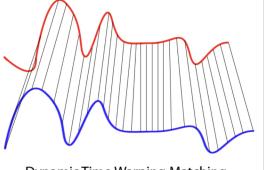


Training set series for a cluster (in black) and centroid of the cluster (in red).

## **Dynamic Time Warping(DTW)**

- Euclidean distance not invariant to time shifts.
- Alternate similarity measure between time series, which may vary in speed.
- Based on the Levenshtein distance (edit distance)





Dynamic Time Warping Matching

### **Similarity Functions**

Let A =  $< a_1, a_2, ..., a_T >$  and B =  $< b_1, b_2, ..., b_T >$  be two sequences, and let  $\delta$  be a distance between elements (or coordinates) of sequences.

- Euclidean Distance:  $D(A,B) = \sqrt{\delta(a_1,b_1)^2 + \cdots + \delta(a_T,b_T)^2}$
- Dynamic time warping:  $D(A_i, B_j) = \delta(a_i, b_j) + \min \begin{cases} D(A_{i-1}, B_{j-1}) \\ D(A_i, B_{j-1}) \\ D(A_{i-1}, B_j) \end{cases}$
- Numerical Example [YT video]

### k -means and Dynamic Time Warping

- Cluster centers (aka centroids) are computed with respect to DTW.
- Clusters gather time series of similar shapes
  - DTW deals with time shifts.
- Capable of retrieving sensible average shape of the temporal shifts in the cluster.

### **Further Readings**

- https://www.cs.unm.edu/~mueen/DTW.pdf
- https://dl.acm.org/doi/10.1145/2939672.2945383