

# **Data Science**

# **Survival Skills**

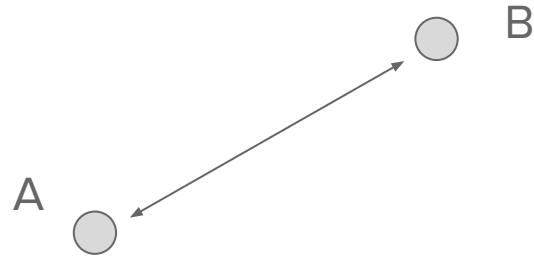
## Exercise 6

# Agenda

- Metrics (Distance + Evaluation)
- Audio signal processing → Data Augmentation

# Distance metrics

Computing a distance



# Distance metrics

- **L1 Norm (Manhattan distance)**

- Distance a taxi would travel along the grid-like streets of Manhattan from one point to another
- Measures the absolute sum of the differences between corresponding coordinates of two vectors
- Scale invariance: not affected by the magnitude of the vectors

$$\sum_{i=1}^n |x_i - y_i|$$

# Distance metrics

- **L2 Norm (Euclidean distance)**

- Measures the straight-line distance between two points in Euclidean space
- Scale sensitivity: sensitive to the magnitude of the vectors → takes into account both the direction and the magnitude of the differences between corresponding coordinates

$$\sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

# Distance metrics

- **$L^\infty$  Norm (Chebyshev distance)**

- Maximum absolute difference: focuses in the dimension with the maximum absolute difference between corresponding coordinates → measures the “worst-case” in terms of dissimilarity
- Scale invariance

$$\max_i (|x_i - y_i|)$$

# Distance metrics

- **Lp Norm (Minkowski distance)**
  - Generalization

$$\left(\sum_{i=1}^n |x_i - y_i|^p\right)^{\frac{1}{p}}$$

# Distance metrics

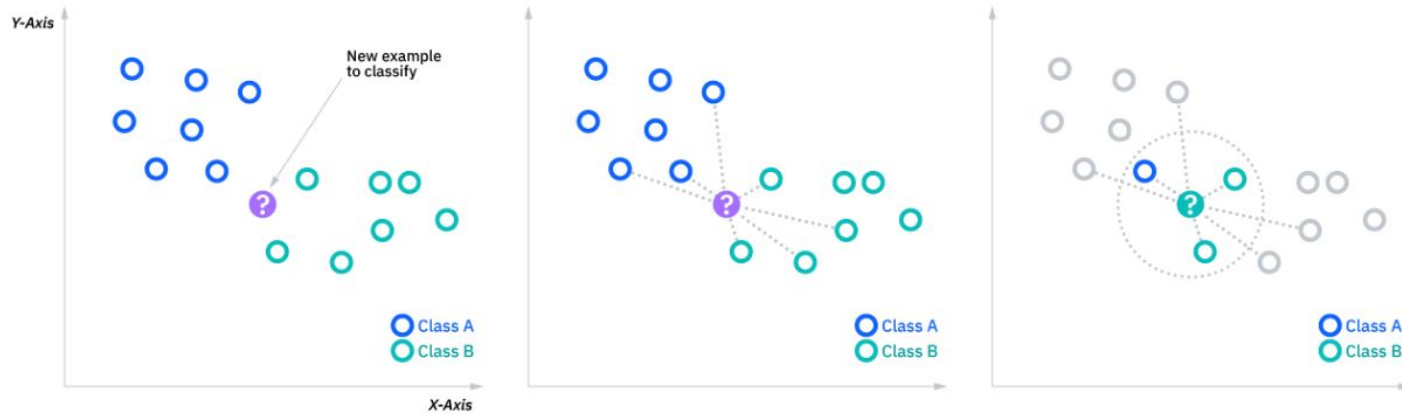
- **Cosine similarity**

- Scale invariance
- High-dimensional data: concept of distance becomes less meaningful → CS can be more effective in capturing the relationship between data points

$$\cos(\theta) = \frac{\mathbf{a} \cdot \mathbf{b}}{\|\mathbf{a}\| \|\mathbf{b}\|} = \frac{\mathbf{a} \cdot \mathbf{b}}{\sqrt{\mathbf{a}^2 \cdot \mathbf{b}^2}} = \frac{\sum_{i=1}^n a_i \cdot b_i}{\sqrt{\sum_{i=1}^n (a_i)^2} \cdot \sqrt{\sum_{i=1}^n (b_i)^2}}$$

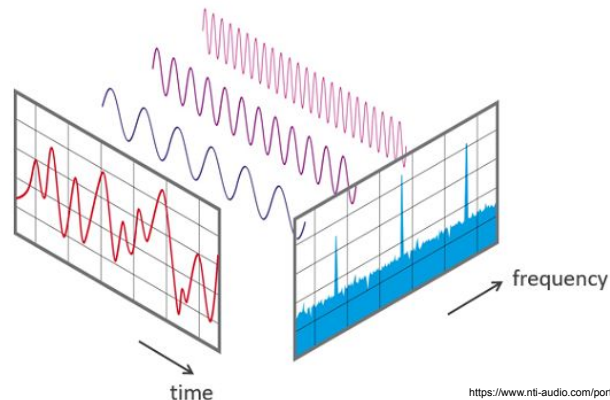


# kNN - k-Nearest-Neighbors



# Fourier transform

- Takes a function in the time (or spatial) domain and represents it in the frequency domain
- Result of the Fourier Transform provides information about the frequency components present in the original function
- Convolution Theorem: powerful property → states that the convolution of two functions in the time domain is equivalent to the multiplication of their Fourier Transforms in the frequency domain
- Fast Fourier Transform (FFT): algorithm for efficiently computing the fourier transform



# STFT - Short-term fourier transform

- Provides a time-dependent frequency analysis by dividing the signal into smaller, overlapping segments
- Time and Frequency Resolution: provides a time-varying frequency analysis, allowing for the examination of changes in frequency content over time → time and frequency resolution can be adjusted by choosing different window sizes

