

## 2. Isolated word speech recognition & HMMs

Task: Recognition of isolated spoken words from a given vocabulary

Problems:

- variable speed
- speaker independence
- prosody etc.

How do we (mammals) hear?

→ tympanic membrane → ossicles →

→ cochlea:

basilar membrane (scala media)

inner & outer hair cells

→ auditory cortex

### A. Signal pre-processing

- Sample the pressure-time function  $f(t)$ , digitise  
highest frequency in speech signal  $< 10$  KHz  
→ Nyquist theorem → sample with 20 KHz

- Frequency analysis: apply digital Fourier transform with sliding window

$$C(\omega, t) = \int_{-\infty}^{\infty} W(t-t') f(t') e^{i\omega t'} dt'$$

simplest window function  $W(t) = \begin{cases} 1 & \text{if } |t| \leq b \\ 0 & \text{otherwise} \end{cases}$

width  $b$ : lowest freq. vs. time resolution

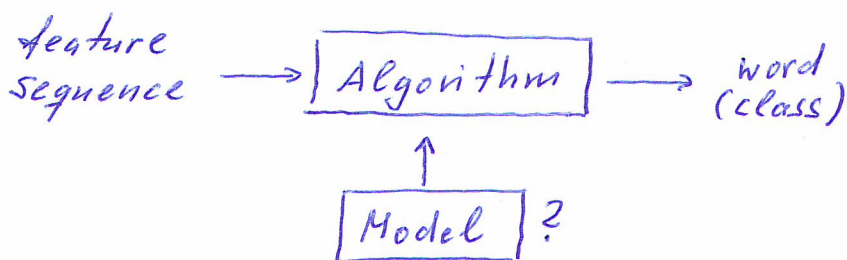
- Energy in spectra (logarithmic, dB)

$$S(\omega, t) = 20 \cdot \log_{10} \sqrt{\text{Re}^2 C(\omega, t) + \text{Im}^2 C(\omega, t)}$$

discretise domain of  $\omega$  into  $\sim 20$  frequency channels with freq. dependent width

- possibly cluster spectral vectors
  - pro: small number of feature vectors
  - con: dominance of stationary parts

B. Dynamic time warping & word recognition



Model: a set of prototypes (i.e. feature sequences) for each word (i.e. class)

Algorithm: nearest neighbour classifier

We need a distance measure for sequences of feature vectors

prototype  $x = (\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n)$ , signal  $y = (\vec{y}_1, \vec{y}_2, \dots, \vec{y}_m)$

$\vec{x}_i, \vec{y}_j \in \mathbb{R}^{20}$ . Distance  $D(x, y) = ?$

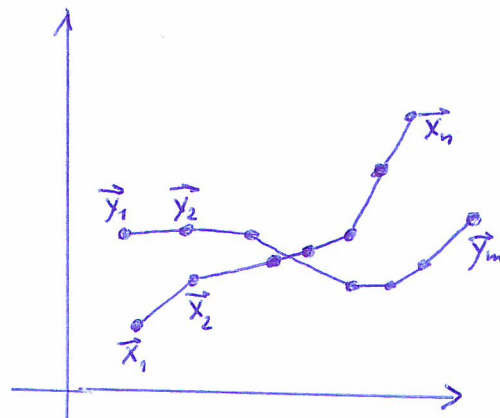
Monotonous matching (aka time warping)

$$\mathcal{T} = ((i_1, j_1), (i_2, j_2), \dots, (i_L, j_L))$$

$\mathcal{T} \in \mathcal{T}$  if

(1)  $(i_1, j_1) = (1, 1)$ ,  $(i_L, j_L) = (n, m)$

(2)  $i_{k-1} \leq i_k \leq i_{k-1} + 1$   
and similar for  $j$



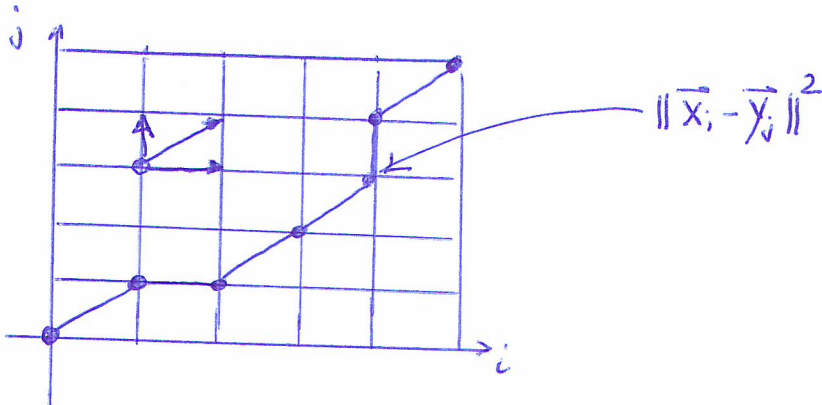
distance for a fixed matching  $\tau \in \mathcal{T}$

$$D(x, y; \tau) = \sum_{k=1}^{|\tau|} \|\bar{x}_{i_k} - \bar{y}_{j_k}\|^2$$

distance

$$D(x, y) = \min_{\tau \in \mathcal{T}} D(x, y; \tau)$$

How to compute it efficiently?



i.e. shortest path, here by dynamic programming, complexity  $O(nm)$

### Discussion model & algorithm

- inference has high time complexity  $O(n^2p)$ , where  $p$  - total number of prototypes
- learning: how to choose optimal prototypes?

Better: model each word (class) by an HMM

$x = (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)$  - sequence of features

$s = (s_1, \dots, s_n)$  - sequence of hidden states (e.g. phonemes)

$$p(x, s) = p(s_1) \prod_{i=2}^n p(s_i | s_{i-1}) \prod_{i=1}^n p(\bar{x}_i | s_i)$$

- fast inference (linear in  $n$ )
- feasible learning of model parameters