

Part 4. Feature extraction

Questions A

A1. Explain the structure of the prediction error filter and of its inverse.

- What is the difference of the two filters p on slide 9?
- Which cost function is minimized?

A2. Cost function / Distances of spectral envelopes

- What are the cost function (distance function) requirements?
- Does the equation on slide 16 comply with these requirements?
- Describe the plots on pages 17 and 18 and compare them with regard to their influence on a subsequent pattern recognition.

A3. Cepstral coefficients

- What is the statement of Parseval's theorem used on slide 19?
- What is the influence of the cepstral coefficients c_i (slide 26) on the cost function (slide 16)?

A4. MFCCs: Logarithm and IDCT

- How is the logarithm in the MFCC calculation motivated?
- Explain the equation on slide 40. What meaning does the matrix P have?

Part 4. Feature extraction

Answers B

B1. What does speech recognition expect from feature extraction (slide 6)?

- Speech signals are assumed to be approximately stationary in frames of 10 to 20 ms.
- The subsequent pattern recognition is most efficient if only few coefficients contain every relevant information.

B2. Linear prediction

- See slide 14.
- See definition of error signal on slide 11 and of s_{yy} on slide 14, $p_{opt} = s_{yy}^{-1} s_{yy}(k)$.

B3. MFCCs: Overview

- See slide 28.
- See slide 31.

B4. MFCCs: Squared absolute, mel filtering

- The phase information is lost. It is assumed to be not relevant for speech recognition.
- The human perception of pitch motivates the Mel filtering. At high frequencies, a lower frequency resolution is sufficient.
- Influence of mel filtering: Reduction of the data rate, “smoothing” of the spectral envelope, neglect of the pitch and its harmonics.

Part 4. Feature extraction

Questions B

B1. What does speech recognition expect from feature extraction (slide 6)?

- Why are new features extracted every 10 to 20 ms?
- Why is it the goal to calculate as few coefficients as possible?

B2. Linear prediction

- Explain the equation $\mathbf{p}_{opt} = \mathbf{S}_{yy}^{-1} \mathbf{s}_{yy}(1)$.
- How is \mathbf{s}_{yy} defined? How would the equation of a predictor with a delay of z^{-k} instead of z^{-1} look like? Where in the derivation does the delay appear?

B3. MFCCs: Overview

- Give an overview of the calculation of MFCCs.
- Which influence does the windowing have?

B4. MFCCs: Squared absolute, mel filtering

- Which information is lost in the process of calculating the absolute of the spectrum?
- Why is the mel filtering applied?
- Which influence does the mel filtering have on the data rate, on the spectral envelope, on the pitch and on the harmonics?

Part 4. Feature extraction

Answers A

A1. Explain the structure of the prediction error filter and of its inverse.

- There is no difference. Only how the filter is connected (as FIR or IIR) makes the difference.
- Cost function: See slide 8.

A2. Cost function / Distances of spectral envelopes

- See slide 16.
- Yes. The requirement to be invariant to level/gain is ensured by the use of linear prediction.
- The cepstral distance relativizes the large distances in areas of high energy (low frequencies) and increases the influence of deviations in areas of low energy (high frequencies)

A3. Cepstral coefficients

- Interpretation of Parseval's theorem: The power of a time-domain signal is equal to the power of its Fourier coefficients.
- See equation on slide 19, sum over the squared differences of the first $3N/2$ coefficients.

A4. MFCCs: Logarithm and IDCT

- The human perception of loudness motivates the use of the logarithm (slide 37).
- The matrix P is described in the last passage; it performs a reduction of dimensions.