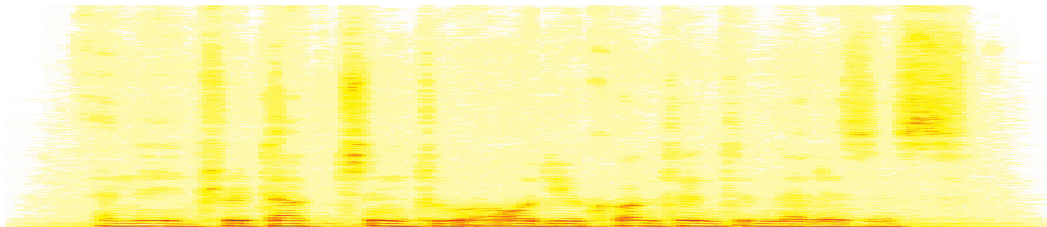


# Introduction to Audio Content Analysis

## Module 2.6: Fundamentals — Non-Fourier Time-Frequency Transforms

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# introduction

## overview

### corresponding textbook section

Chapter 2 — Fundamentals: pp. 9–11

- **lecture content**

- constant-Q transform (CQT)
- Gammatone filterbank

- **learning objectives**

- discussing the advantages and disadvantages of different time-frequency transforms
- explaining the principles of the CQT and auditory filterbanks



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# other time frequency transforms

## introduction

- Fourier transform continues to be much-used tool in audio signal processing and MIR
  - but there are disadvantages, e.g.
    - frequency axis does not directly map to (perceptual) pitch axis
    - frequency and time resolution inversely related
- ⇒ **alternative transforms** can be used

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- ⇒ **alternative transforms** can be used

# constant-Q transform

## introduction

- DFT has a *linear* frequency axis:
  - not perceptually meaningful: *logarithmic* is better match
  - low frequency resolution at low frequencies

⇒ compute DFT-like transform at specific frequencies

- space frequencies logarithmically (constant  $Q$ )
- resulting abscissa resolution is pitch-related

# constant-Q transform

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- DFT has a *linear* frequency axis:
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- ⇒ compute DFT-like transform **at specific frequencies**
- space frequencies logarithmically (constant  $Q$ )
  - resulting abscissa resolution is pitch-related

$$Q = \frac{f}{\Delta f} = \frac{1}{2^{1/c} - 1}$$

# constant Q transform

## implementation 1/2

$$X_{\text{CQ}}(k, n) = \frac{1}{\mathcal{K}(k)} \sum_{i=i_s(n)}^{i_e(n)} w_k(i - i_s) \cdot x(i) e^{j2\pi \frac{Q \cdot (i - i_s)}{\mathcal{K}(k)}}$$

$$\mathcal{K}(k) = \frac{f_s}{f(k)} Q$$

- $f(k)$ : frequency of bin index  $k$
- $\mathcal{K}(k)$ : blocklength for bin index  $k$
- $Q$ : measure of pitch res.
- $w_k$ : window function
- $i_s, i_e$ : start and stop time indices of block
- $f_s$ : sample rate

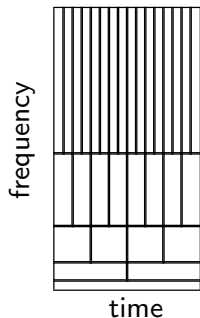
- long window for low frequencies (high freq res, low time res)
- short window for high frequencies (low freq res, high time res)



# constant Q transform

## implementation 2/2

### non-overlapping



### overlapping

- define transformation matrix with maximum window length
  - zeropad higher frequencies (left & right)
- ⇒ independent definition of block and hop length

# constant Q transform

## CQT vs. DFT

### CQT:

- + perceptually/musically adapted frequency resolution
- time resolution depends on frequency
- not invertible
- no optimized implementation (compare FFT)

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# auditory filterbanks

## introduction

FT and related transforms bad models of physiological properties of the human ear:

- frequency resolution (critical bands)
- frequency scale (pitch resolution)
- loudness & masking
- event perception & time integration

⇒ **auditory filterbanks**

not as widely used as one might think because

- computationally inefficient
- analysis only: no invertibility (mostly)
- not proven to be superior

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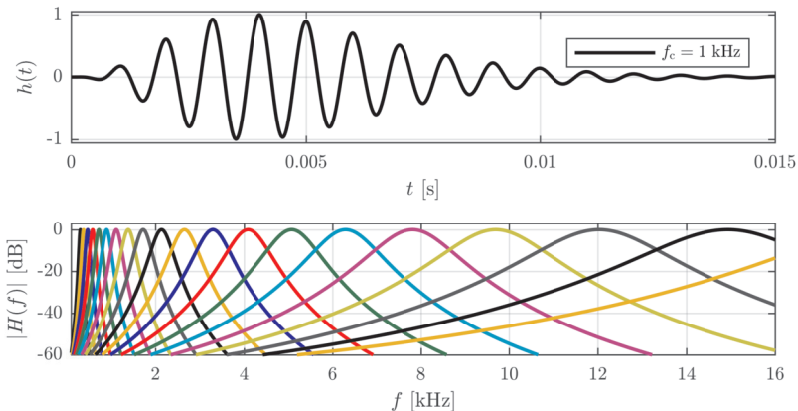
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# auditory filterbanks

## gammatone filterbank

$$h(i) = \frac{a \cdot (i/f_s)^{\mathcal{O}-1} \cdot \cos\left(2\pi \cdot f_c \frac{i}{f_s}\right)}{e^{2\pi i \Delta f / f_s}}$$



# summary

## lecture content

- **DFT has disadvantages**
  - low frequency resolution for low pitches
  - non-logarithmic/perceptually relevant pitch resolution
- **CQT**
  - similar to Fourier Transform but logarithmically spaced frequency bins
  - not invertible and inefficient
- **Filterbanks**
  - good model of human physiology
  - not invertible and inefficient
  - not proven to be superior

