

### Introduction to Audio Content Analysis

module 3.3.2: time-frequency representations — constant Q transform

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### introduction overview



#### corresponding textbook section

section 3.3.2

#### lecture content

- constant-Q transform (CQT)
- learning objectives
  - discussing advantages and disadvantages of different time-frequency transforms
  - explaining the principles of the CQT and auditory filterbanks



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# non-FT time frequency transforms introduction



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- 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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- ⇒ compute DFT-like transform at specific frequencies
  - space frequencies logarithmically (constant Q)
  - resulting abscissa resolution is pitch-related
  - parameter c adjusts number of bins per octave

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

$$X_{\text{CQ}}(k,n) = \frac{1}{\mathcal{K}(k)} \sum_{i=i_{\text{s}}(n)}^{i_{\text{e}}(n)} w_k(i-i_{\text{s}}) \cdot x(i) e^{\mathrm{j}2\pi \frac{\mathcal{Q}\cdot(i-i_{\text{s}})}{\mathcal{K}(k)}} = f(k)$$
: frequency of bin index  $k$ 

$$\mathcal{K}(k) : \text{blocklength for bin index } k$$

$$\mathcal{Q}: \text{ measure of pitch res.}$$

$$\mathcal{W}_k: \text{ window function}$$

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$$\mathbf{w}_k: \text{ start and stop time indices}$$

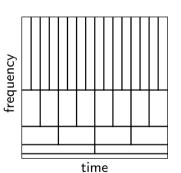
- $\bullet$   $i_s$ ,  $i_e$ : start and stop time indices of block
- $\blacksquare$   $f_{\rm S}$ : sample rate
- long window for low frequencies (high freq res, low time res)
- short window for high frequencies (low freq res, high time res)

CQT oo•o

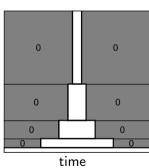
# constant Q transform implementation 2/2

Georgia Center for Music Tech Technology

### non-overlapping



### overlapping



#### differences

- outputs at multiple vs. one time resolution
- multiple different FFT lengths vs. one FFT length (zero-padded)
- dependent vs. independent definition of block and hop length



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- not invertible
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#### 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

