

Lecture 6

Rhythm Analysis

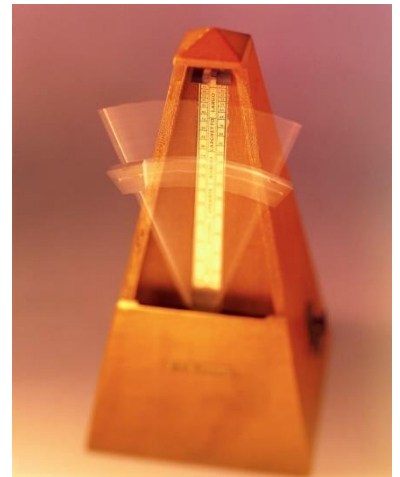
(some slides are adapted from Zafar Rafii and some figures are from Meinard Mueller)

Definitions for Rhythm Analysis

- **Rhythm:** “movement marked by the **regulated succession** of strong and weak elements, or of opposite or different conditions.”

---- Oxford English Dictionary

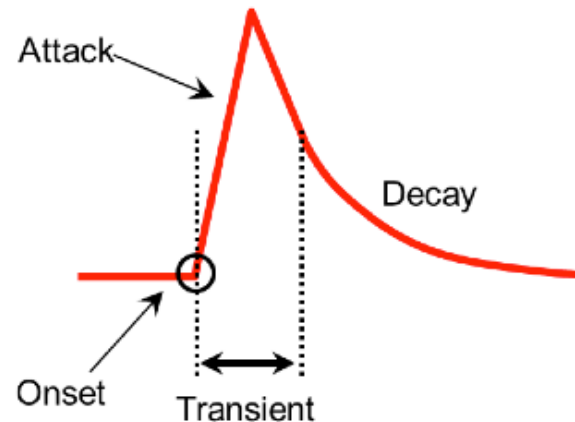
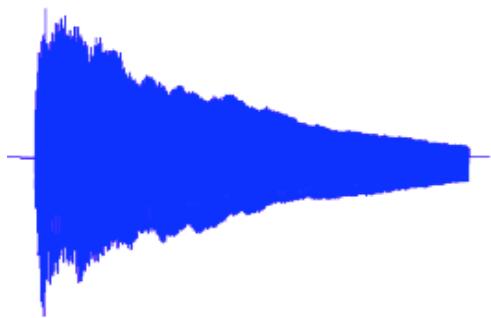
- **Beat:** **basic unit** of time in music
- **Tempo:** speed or pace of a given piece, typically measured in **beats per minute (BPM)**



More Definitions

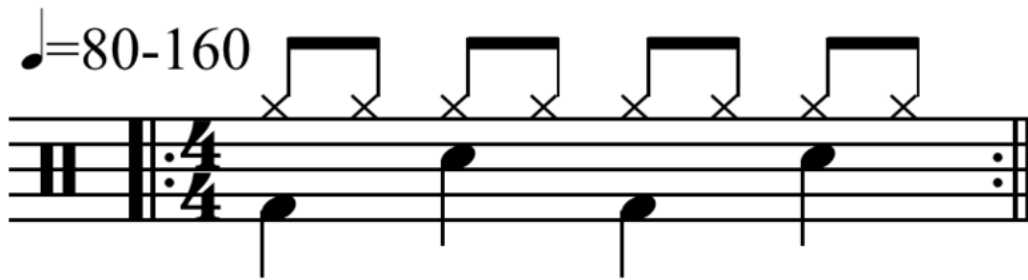
- **Onset:** single instant marking the beginning of transient
 - Onsets often occur on beats.
- **Attack:** sharp increase of energy
- **Transient:** a short duration with high amplitude within which signal evolves quickly

Waveform of one piano note



More Definitions

- **Measure (or bar):** segment of time defined by a given number of beats

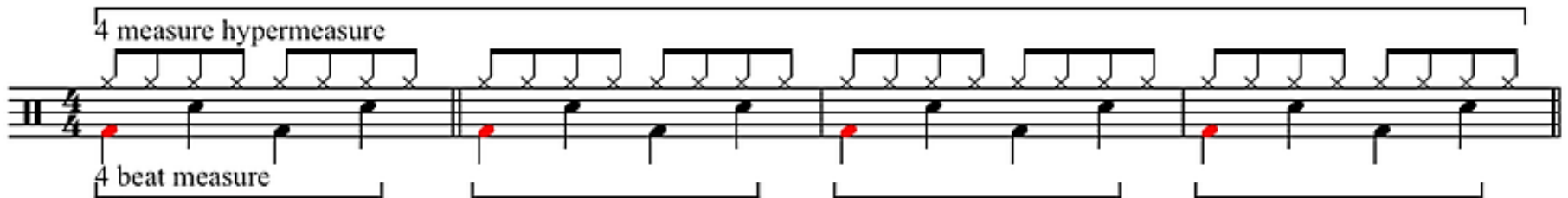


A 4-beat measure drum pattern.

[[http://en.wikipedia.org/wiki/Metre_\(music\)](http://en.wikipedia.org/wiki/Metre_(music))]

More Definitions

- **Meter:** Organization of music into regularly recurring measures of stressed and unstressed beats



Hypermeter: 4-beat measure and 4-measure hypermeasure. Hyperbeats in red.
[[http://en.wikipedia.org/wiki/Metre_\(music\)](http://en.wikipedia.org/wiki/Metre_(music))]

Rhythm Analysis Tasks

- Onset Detection
- Beat Tracking
- Tempo Estimation
- Higher-level Structure Analysis



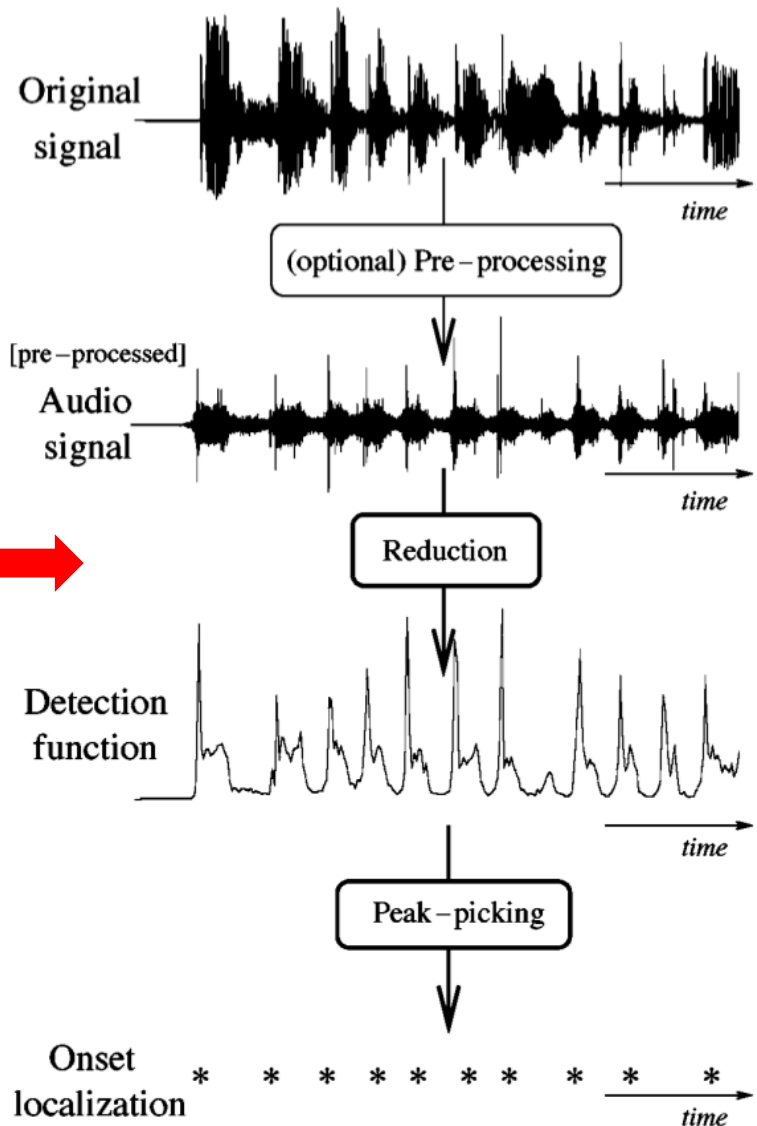
Why is it important?

- Intellectual merit
 - Important component of music understanding
 - Music cognition research

- Broad applications
 - Identify/classify/retrieve by rhythmic similarity
 - Music segmentation/summarization
 - Audio/video synchronization
 - Source separation

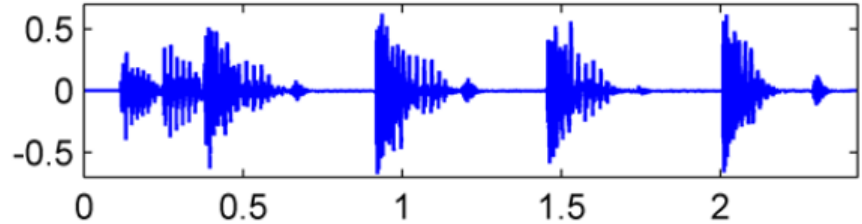
Onset Detection

- Signal processing:
define a detection function
 - Energy-based
 - Spectral-based
 - Phase-based
- Machine Learning:
learn patterns from labeled data
 - Probabilistic models
 - Neural networks



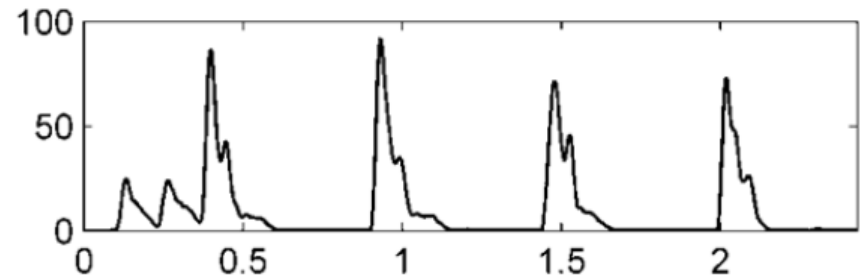
Energy-based Onset Detection

Waveform



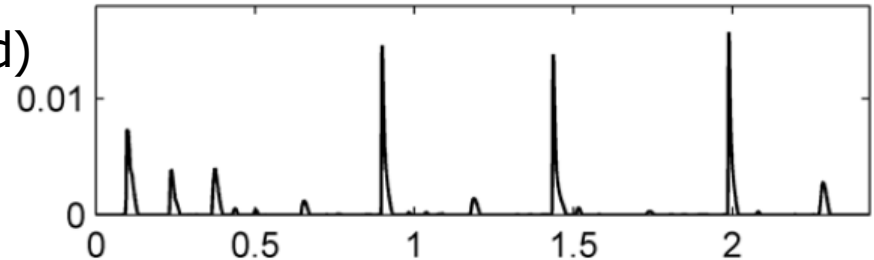
Signal Envelope (energy)

$$E_w^x(n) := \sum_{m=-M}^M |x(n+m)w(m)|^2$$

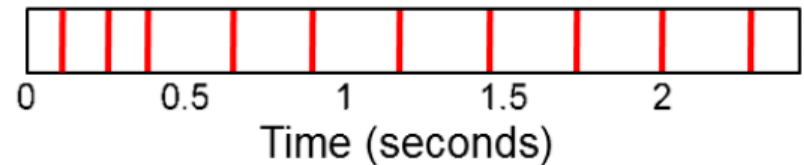


Envelope Derivative (half-wave rectified)

$$\Delta_{\text{Energy}}(n) := |E_w^x(n+1) - E_w^x(n)|_{\geq 0}$$



Thresholding \rightarrow Onsets



Energy-based Onset Detection

- Pros and Cons
 - Simple
 - Works well for percussive sounds
 - Soft onsets by string/wind instruments are hard to detect
 - Tremolo/vibrato can cause false detections
- How to improve
 - Use logarithmic-energy to replace linear energy
 - Perform analysis in different frequency bands, then summarize

Spectral-based Onset Detection

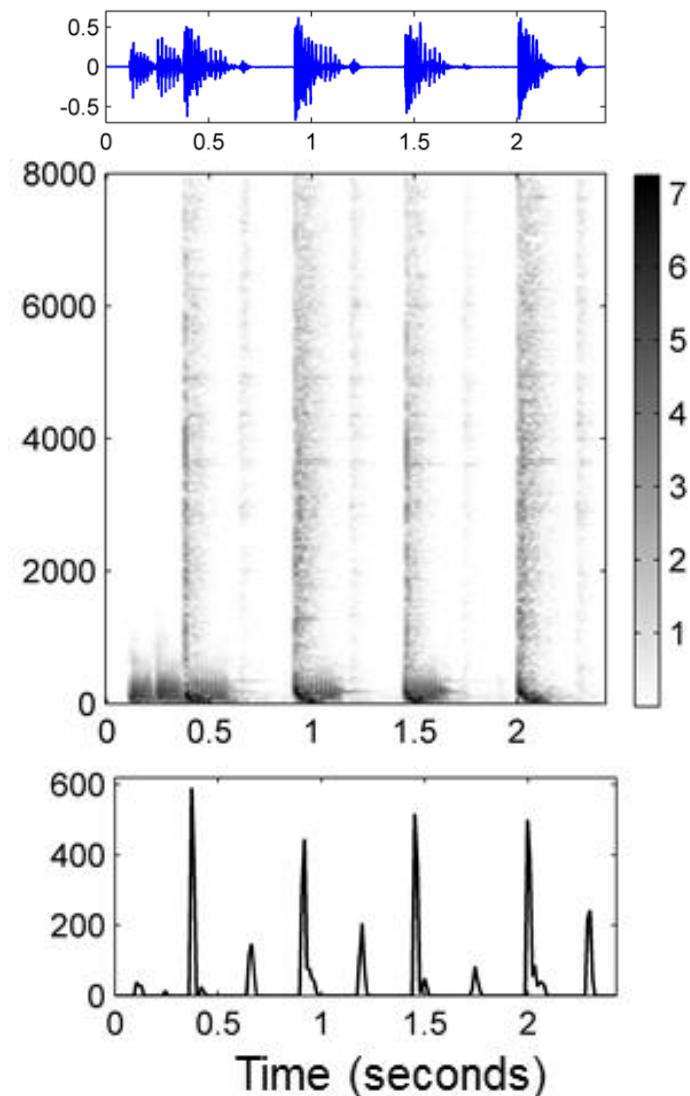
- STFT to get magnitude spectrogram $|\mathcal{X}|$

- (optional) compression

$$\mathcal{Y} := \Gamma_{\gamma}(|\mathcal{X}|) = \log(1 + \gamma \cdot |\mathcal{X}|)$$

- Spectral flux:
 - Take derivative w.r.t. time (half-wave rectified)

$$\Delta_{\text{Spectral}}(n) := \sum_{k=0}^K |\mathcal{Y}(n+1, k) - \mathcal{Y}(n, k)|_{\geq 0}$$



Spectral-based Onset Detection

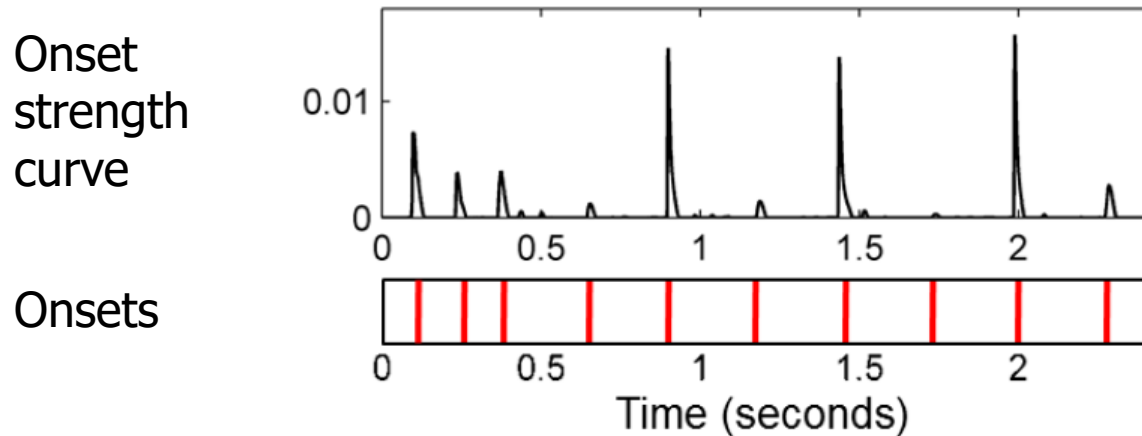
- Pros and Cons
 - More complex than energy-based
 - Can weigh different frequencies differently
 - Works better for soft onsets (e.g., legato notes) and polyphonic music
 - Still doesn't work very well for vibrato

Tempo Estimation

- Tempo = beats / minutes
- Beat tracking is sufficient but not necessary condition for tempo estimation
- How to estimate tempo without tracking beats?

- Idea: look at the **regularity** of onsets
- Assumptions
 - Onsets mostly occur on beats
 - Tempo is constant within a period of time

Tempo Estimation



- Take the onset strength curve and analyze its periodicity
 - Autocorrelation
 - STFT
- Tempogram

Beat Tracking

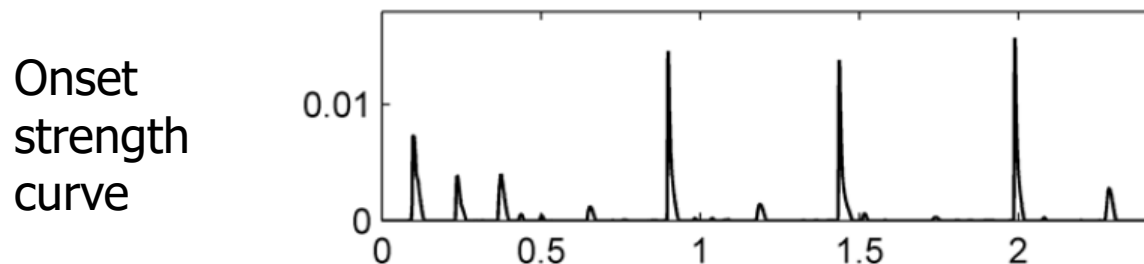
- Identify the beat times, i.e., the times to which we tap our feet
 - Detected onsets provide useful but noisy information, since not all onsets are on beats.
 - Estimated tempo tells us the space between two beats, but not the **exact locations (i.e., phase)**.
- How to identify beats?
- To simplify the problem, we assume
 - Onsets, especially strong ones, are mostly on beats.
 - Tempo is constant.

Beat Tracking

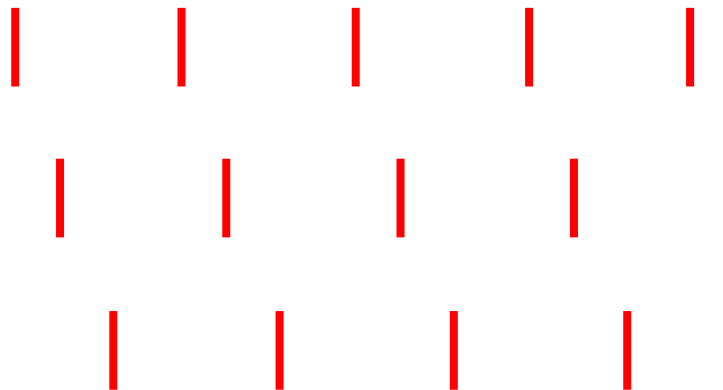
- A 2-step approach
 - Step 1: Tempo estimation
 - Step 2: Identify beats from onsets using the tempo
 - Create an impulse train (i.e., “comb”) with the tempo
 - Cross-correlate the “comb” with the onset strength curve.
 - The lag that gives us the highest cross-correlation value tells us the beat phase.

Beat Tracking

- A 2-step approach, illustration



Combs with the same tempo but different phases



- Problem: **too rigid** about beat spacing

Beat Tracking by Dynamic Programming

- Beat tracking: finding a sequence of beat locations such that
 - 1) are well aligned with strong onsets
 - 2) mostly regularly spaced

[Ellis, 2007]

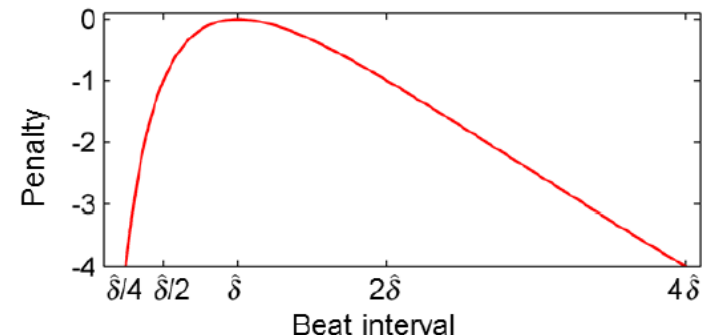
Score function

$$S(B) := \sum_{\ell=1}^L \Delta(b_\ell) + \lambda \sum_{\ell=2}^L P_{\hat{\delta}}(b_\ell - b_{\ell-1})$$

Beat sequence Onset strength Regularity penalty function

Rough estimate of beat spacing

$$P_{\hat{\delta}}(\delta) := -(\log_2(\delta/\hat{\delta}))^2$$



- Find $B = (b_1, b_2, \dots, b_L)$ that maximizes $S(B)$

Beat Tracking by Dynamic Programming

- Suppose beat locations are precise to audio frames, and suppose there are N frames, then **how many possible sequences?**
 - 2^N (although many are bad ones!)
 - Can't enumerate all!
- Key idea: reuse calculations by recursion!

Beat Tracking by Dynamic Programming

- Consider a beat sequence $B_n = (b_1, b_2, \dots, b_L)$ where $b_L = n$.
- Let $D(n)$ be the maximal score over all such sequences ending at n .
- Then

if $L > 1$

$$\mathbf{D}(n) = \Delta(n) + \lambda P_{\hat{\delta}}(n - b_{L-1}) + \mathbf{D}(b_{L-1})$$

if $L = 1$

$$\mathbf{D}(n) = \Delta(n)$$

recursion



Beat Tracking by Dynamic Programming

- Considering the two cases, we have

$$\mathbf{D}(n) = \Delta(n) + \max \begin{cases} 0, \\ \max_{m \in [1:n-1]} \{ \mathbf{D}(m) + \lambda P_{\hat{\delta}}(n-m) \} \end{cases}$$

– We can calculate $D(n)$ from $D(1) = \Delta(1)$.

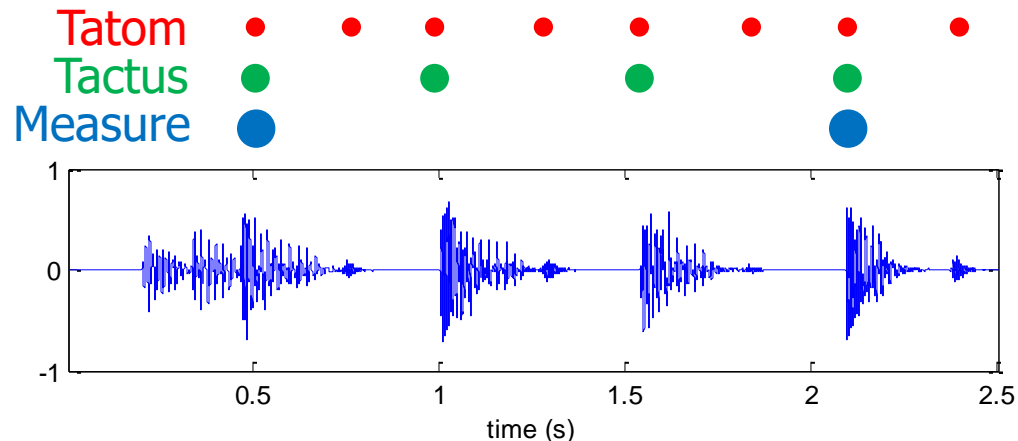
- Record the **preceding beat**

$$\mathbf{P}(n) := \operatorname{argmax}_{m \in [1:n-1]} \{ \mathbf{D}(m) + \lambda P_{\hat{\delta}}(n-m) \}$$

- Best score $\mathbf{S}(B^*) = \max_{n \in [0:N]} \mathbf{D}(n)$

- **Trace back** from $b_L = n^*$ to get the best sequence

Rhythmic Structure

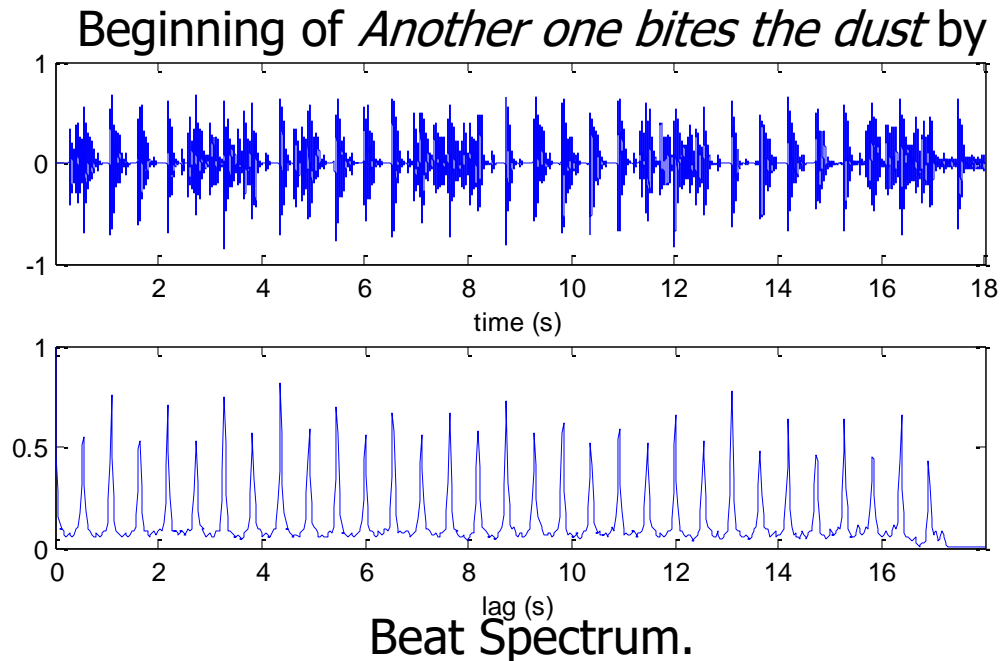


Beginning of *Another one bites the dust* by Queen.

- One approach: detect onsets; analyze tempo and beats at different levels.
- Another approach: analyze repetition of spectral content
 - Beat spectrum

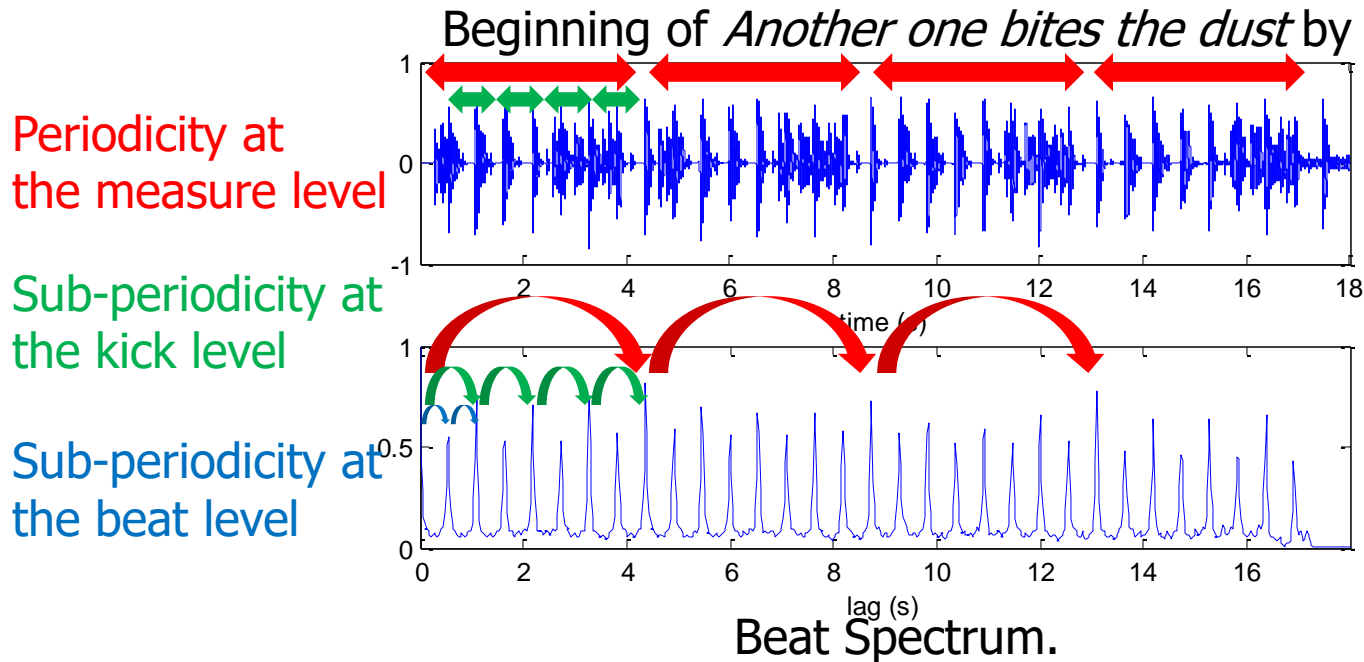
Beat Spectrum

- Definition
 - Using the autocorrelation function, we can derive the beat spectrum [Foote et al., 2001]



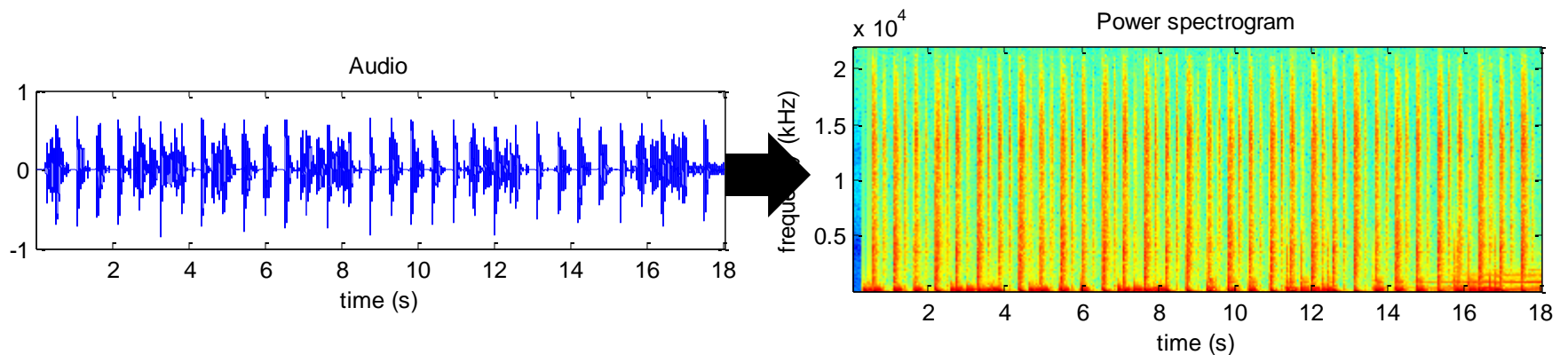
Beat Spectrum

- Use
 - The beat spectrum reveals the hierarchically periodically repeating structure of the audio



Beat Spectrum

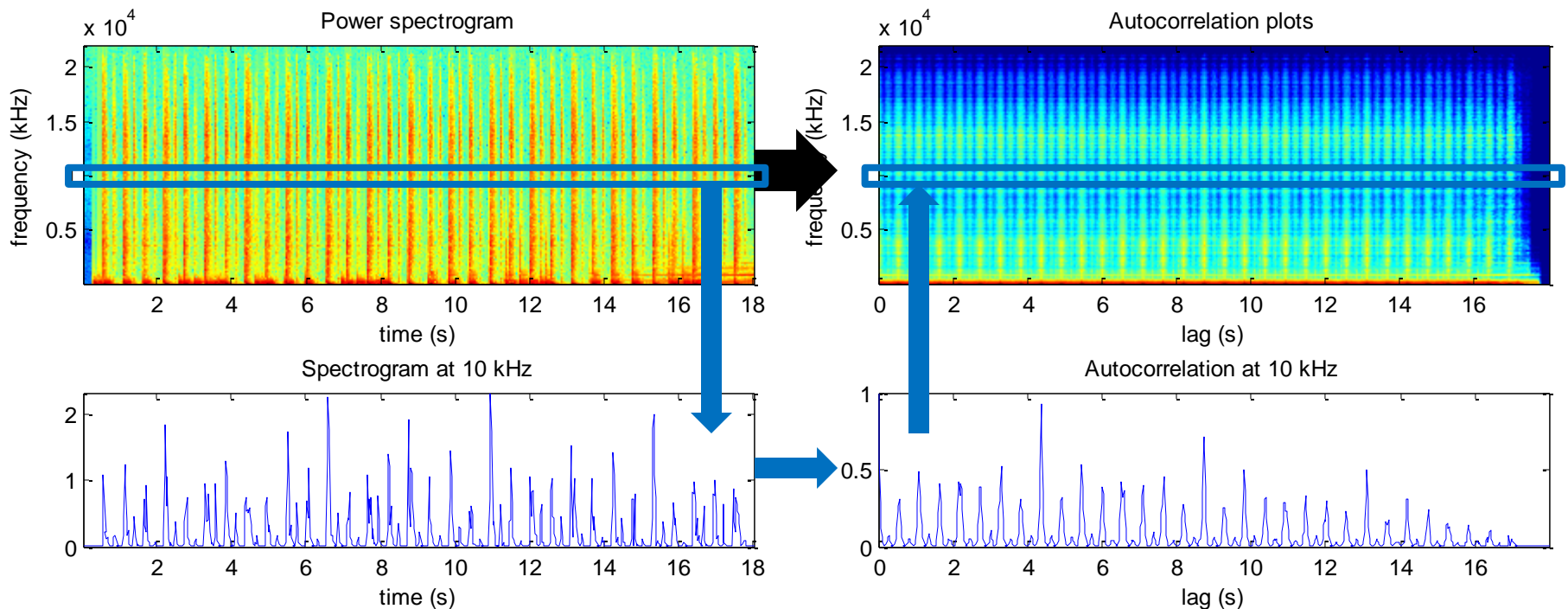
- Calculation
 - Compute the power spectrogram from the audio using the STFT (square of magnitude spectrogram)



Beat Spectrum

- Calculation

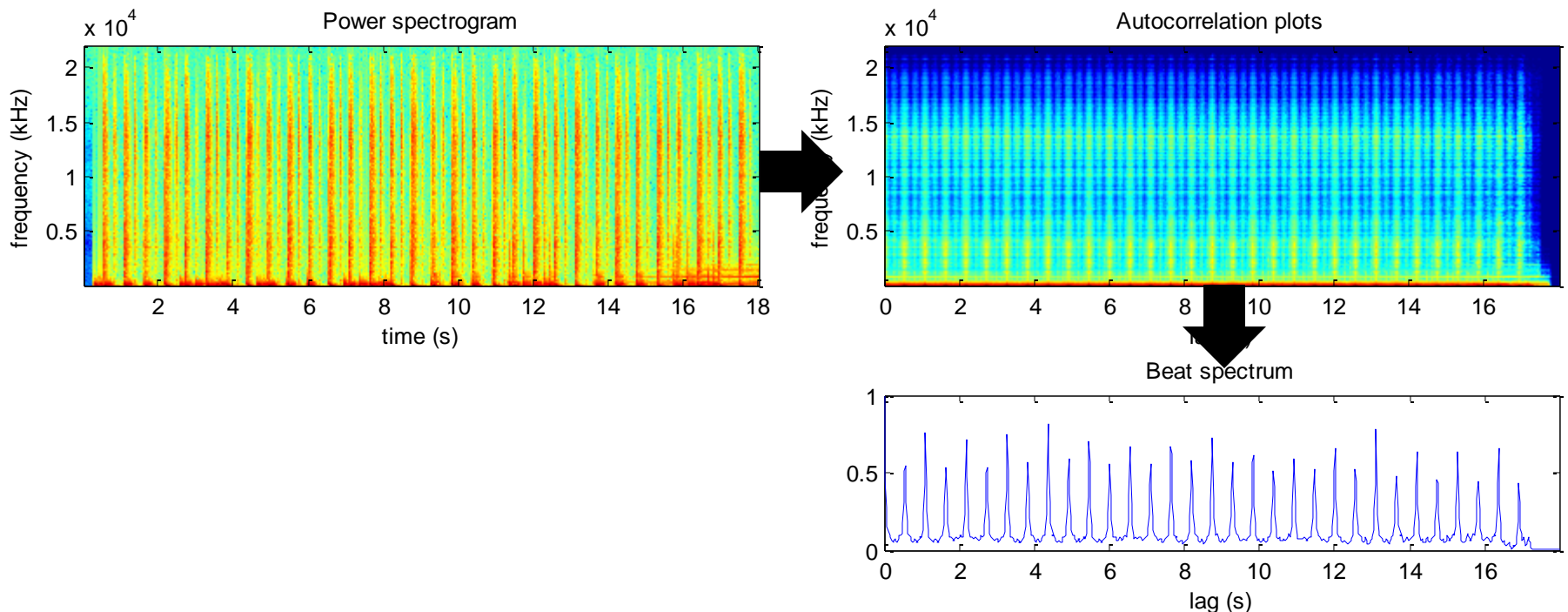
- Compute the autocorrelation of the rows of the spectrogram



Beat Spectrum

- Calculation

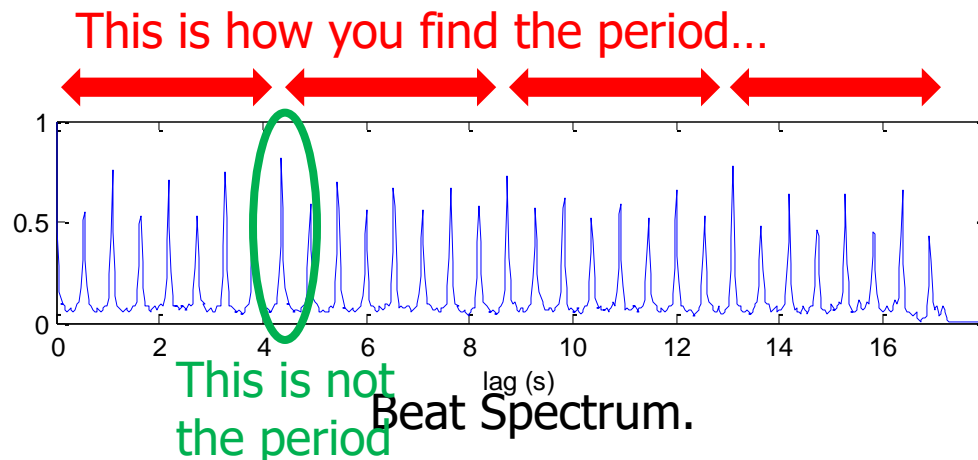
- Compute the mean of the autocorrelations (of the rows)



Beat Spectrum

- Notes

- The first highest peak in the beat spectrum does not always correspond to the repeating period!
- The beat spectrum does not indicate where the beats are or when a measure starts!



Resources

- Some interesting links
 - Dannenberg's articles on beat tracking:
<http://www.cs.cmu.edu/~rbd/bib-beattrack.html>
 - Goto's work on beat tracking:
<http://staff.aist.go.jp/m.goto/PROJ/bts.html>
 - Ellis' Matlab codes for tempo estimation and beat tracking:
<http://labrosa.ee.columbia.edu/projects/beattrack/>
 - MIREX's annual evaluation campaign for Music Information Retrieval (MIR) algorithms, including tasks such as onset detection, tempo extraction, and beat tracking: http://www.music-ir.org/mirex/wiki/MIREX_HOME