

Tempo, Beat, and Rhythm

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Motivation

- Temporal structure is important in music, and it can even exist without melody and harmony
- Intriguing and fun
- MIR goals: detecting the onsets of notes and accents, and tapping the beats like human beings
- Onset detection, tempo estimation, beat tracking, ...



Tempo, beat, and rhythm

- **Tempo**: the speed or pace of music
- How to describe tempo?
 - Italian tempo markings: Largo, Andante, Presto, ...
 - Note-level markings: e.g., $\text{♩}=120$
 - **Beats per minute (BPM)**: how many “pulses” per minutes
- **Beat**: the unit of time of music
 - Beat-level: just “beat”
 - Measure-level: **downbeat** (the first beat in one measure)
每小节第一拍
- **Rhythm**: a characteristic pattern of beats in different level
 - **Meter**: a rhythmic structure identifying the relation between downbeats (accent, measure) and beats (e.g., $\frac{2}{4}, \frac{3}{4}, \dots$)
音較可能出現
的位置
 - Sometimes you can think rhythm as “genre”
 - Cha Cha, Rumba, Tango, Waltz, ...

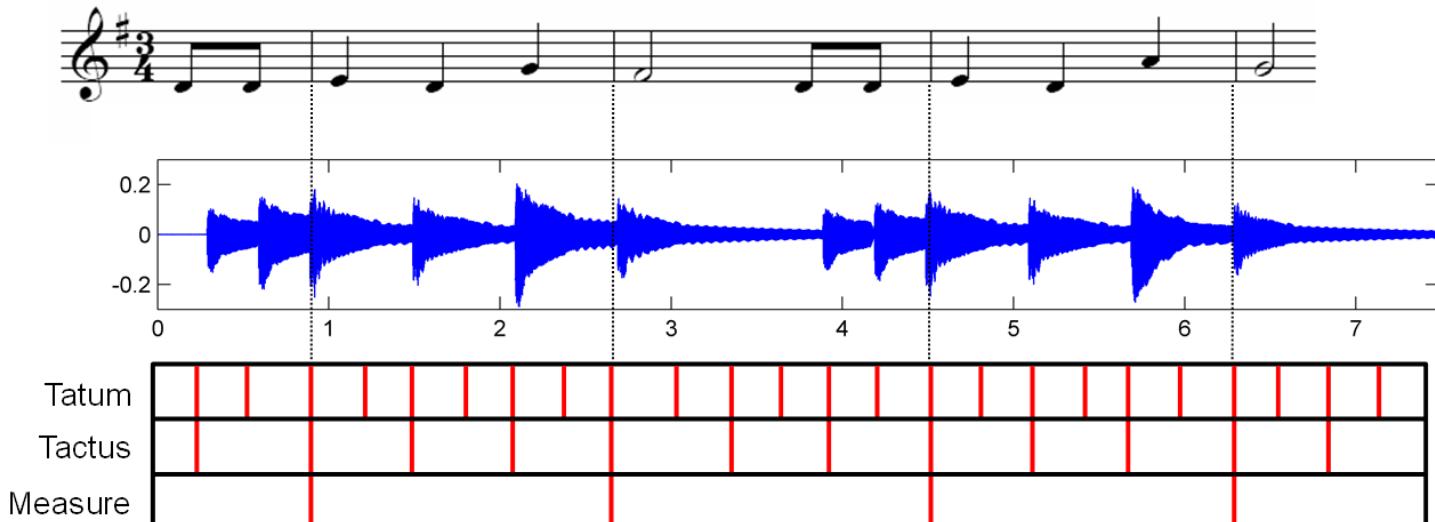
Andante grazioso. ($\text{♩} = 120$)



Beat track = ^①beat + ^②downbeat

Tatum, tactus, and measure

- **Rhythm is hierarchical:** there are various levels presumed to contribute to the human perception of tempo and beat
 - **Tatum:** the fastest repetition rate of musically meaningful accents
 - **Tactus:** typically the foot tapping rate (and quarter note level)
 - **Measure:** typically the rate of one cycle of counting beats
- BPM (beats per minute)



Tasks and problems

- Tempo/meter analysis (on BPM)
BPM = 80, 160, 240 整數倍
 - Same as in pitch detection, tempo also has octave ambiguity
 - “Tempo harmonics” of a song with BPM τ : $2\tau, 3\tau, \dots$
 - “Tempo subharmonics”: $\tau/2, \tau/3, \dots$
 - Basic feature: **tempogram**, an indicator of the local relevance of a specific tempo for a given music recording for each time instance
某時間點可能的哪些 tempo
- Beat tracking and downbeat tracking
 - Backbeat, syncopation
後半拍 切分音
 - Rubato, swing
 - Other genre-/subject-dependent issues
- Rhythm classification

Fourier tempogram (1)

= novelty curve 的 stft

- Given the novelty function $\Delta(m)$ of a musical recording, given a window function $w(m)$, the tempogram , denoted as $F(n, \omega)$, is the STFT of the novelty function:

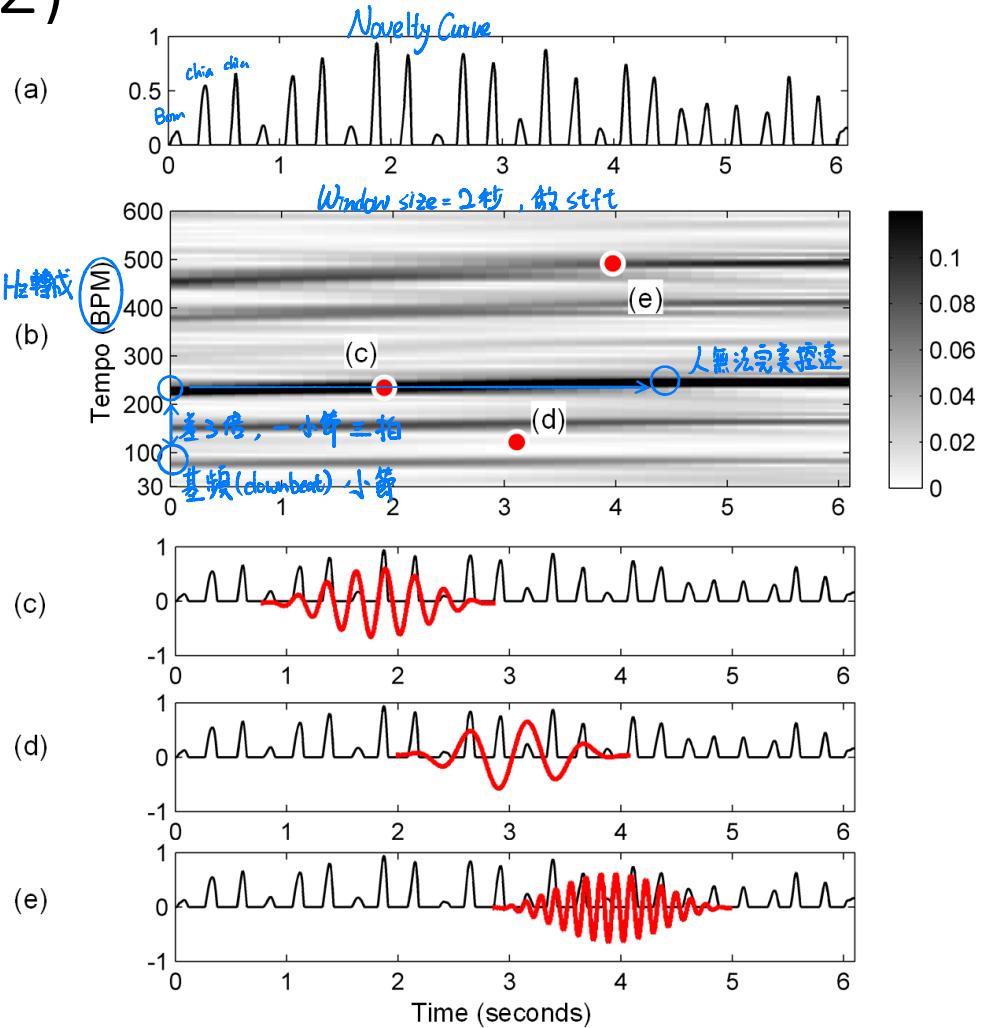
$$F(n, \omega) := \sum_{m \in \mathbb{Z}} \Delta(m) w(m - n) e^{-2\pi i \omega n}$$

- Beat rate $\omega = 1/T$ (beats per second)
- BPM: $\tau = 60\omega$
- Discrete Fourier tempogram

$$T^F(n, \tau) = |F(n, \tau/60)|$$

Fourier tempogram (2)

- Example: Shostakovich's Waltz No. 2



Autocorrelation tempogram (1)

- Recall the issue of finding frequency and finding periodicity in previous lectures in pitch detection
- Alternative way in finding tempo: taking time-varying autocorrelation function (ACF) on the novelty
- Recall ACF:

$$R_{xx}(l) = \sum_{m \in \mathbb{Z}} x(m)x(m - l)$$

不一定用 stft, 亦可用 auto correlation

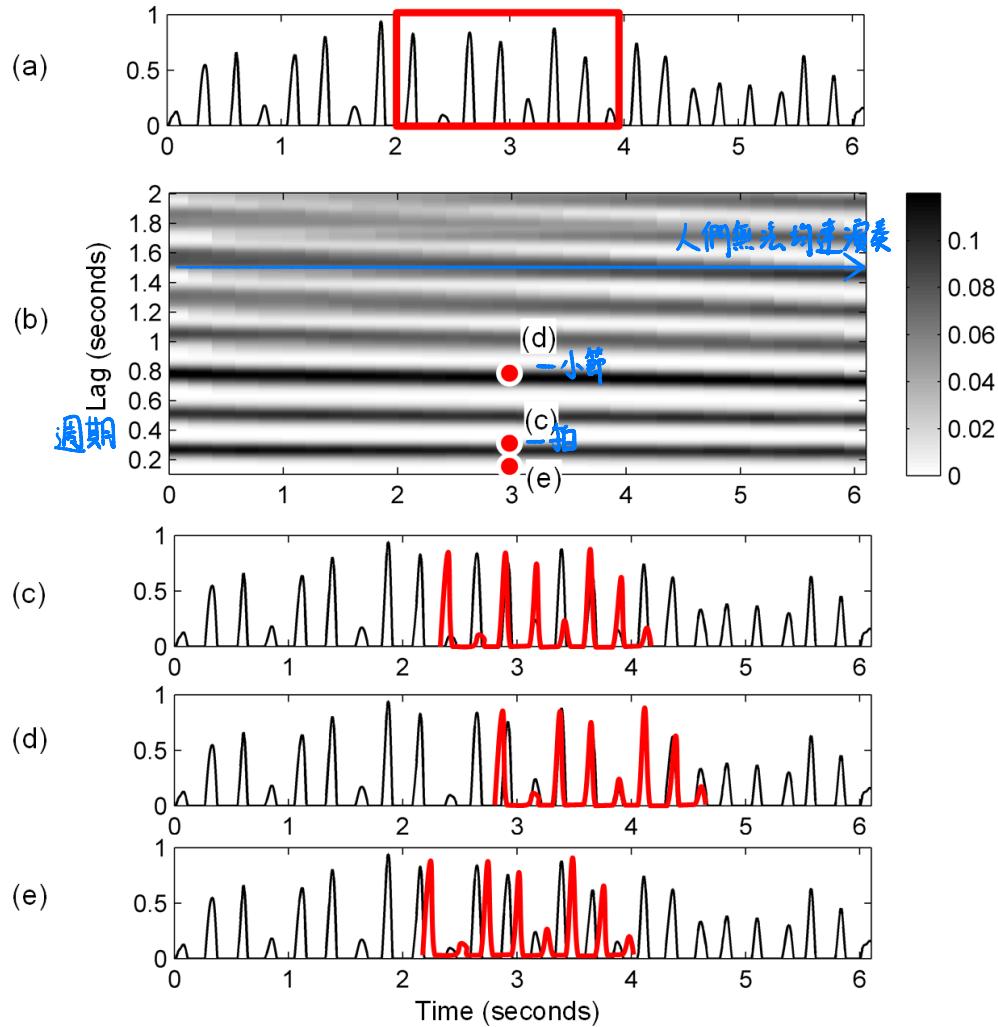
- Short-time ACF of the novelty curve (windowed ACF):

$$A(n, l) := \sum_{m \in \mathbb{Z}} \Delta(m)w(m - n)\Delta(m - l)w(m - n - l)$$

- Alternative way: $A(n, l) = IFFT\{|FFT\{w(m)\Delta(m)\}|^2\}$

Autocorrelation tempogram (2)

- ACF tempogram
 $T^A(n, \tau) := A(n, l)$
- BPM $\tau = 60/(rl)$
- r : the “sampling period” of the novelty curve
- Comparison
 - STFT: Fourier basis
 - ACF: self basis
- Example: Shostakovich’s Waltz No. 2



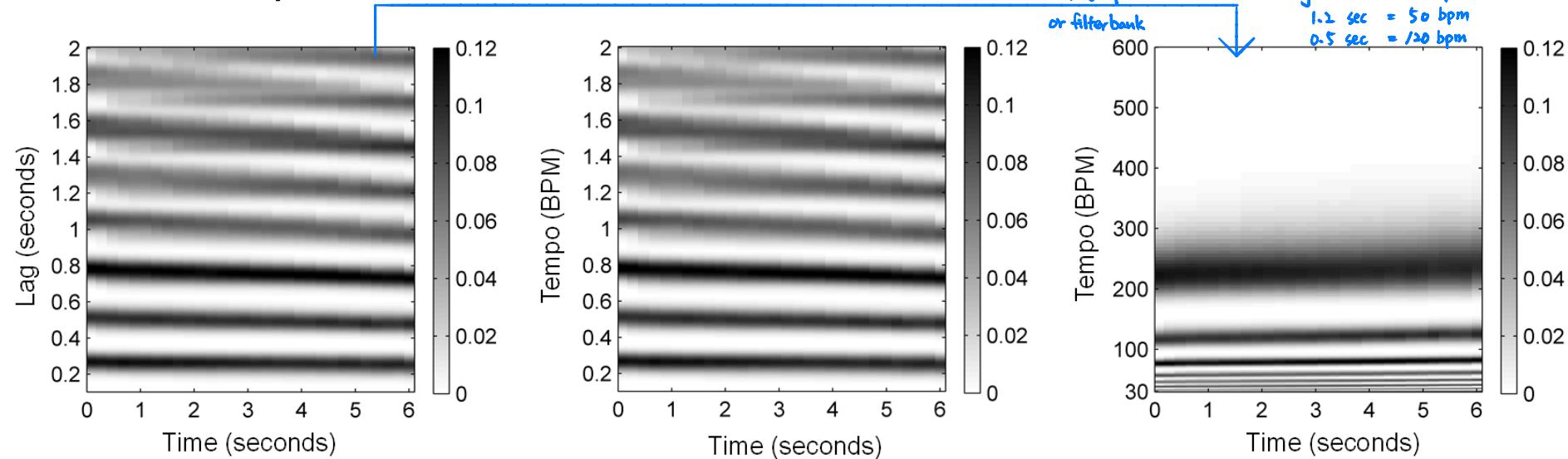
From: M. Mueller, *Fundamentals of Music Processing*, Chapter 6, Springer 2015



stft : 短時(25 ms) 指標
spectral flux : onset detection, novelty curve 由 stft 得到
stft : window size 要能 cover 數拍 (4 ~ 8 sec)
 要抓到 downbeat 要 cover 數小節 (12 ~ sec)
 ↓ by 2x slow

Autocorrelation tempogram (3)

- Conversion from lag to tempo
 - From time-lag representation to time-frequency representation
- Example:



From: M. Mueller, *Fundamentals of Music Processing*, Chapter 6, Springer 2015

Global and local tempo

- For a music recording with N time instances:
 - Average (global) tempo: let Θ the expected interval of tempo

$$T_{\text{Average}}(\tau) := \frac{1}{N} \sum_{n \in [1, N]} T(n, \tau)$$
$$\hat{\tau} := \max \{T_{\text{Average}}(\tau) | \tau \in \Theta\}$$

Cyclic tempogram (1)

來處理 octave ambiguity 看看就好

- Motivation: like the pitch scale, the tempo scale also has “octave equivalence” 像 chroma, 把 octave 合併起來

- A song with BPM=120 can usually be interpreted as BPM=240 (or BPM=60)

- Define $[\tau] := 2^k \tau, k \in \mathbb{Z}$

- Example: for $\tau = 120$ one obtains $[\tau] = \{\dots, 30, 60, 120, 240, 480, \dots\}$

- The cyclic tempogram

$$C(n, [\tau]) := \sum_{\lambda \in [\tau]} T(n, \lambda)$$

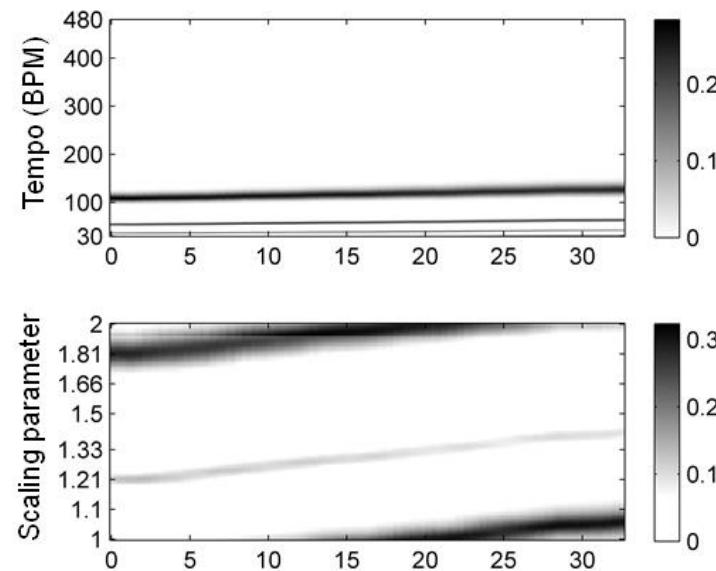
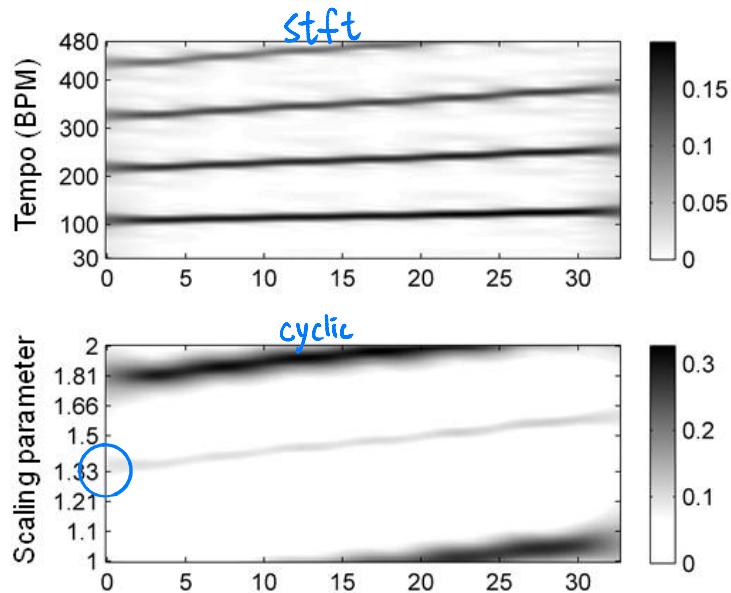
- The cyclic tempogram referred to τ_0 : $C_{\tau_0}(n, s) := C(n, [s, \tau_0])$

- s : scaling parameter

- Notice that $C_{\tau_0}(n, s) := C_{\tau_0}(n, 2^k s)$ for $k \in \mathbb{Z}$

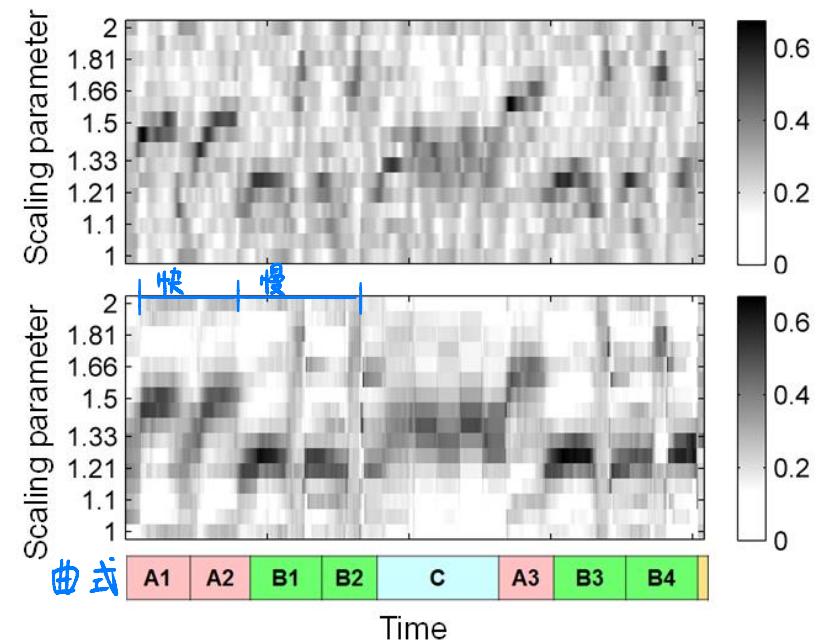
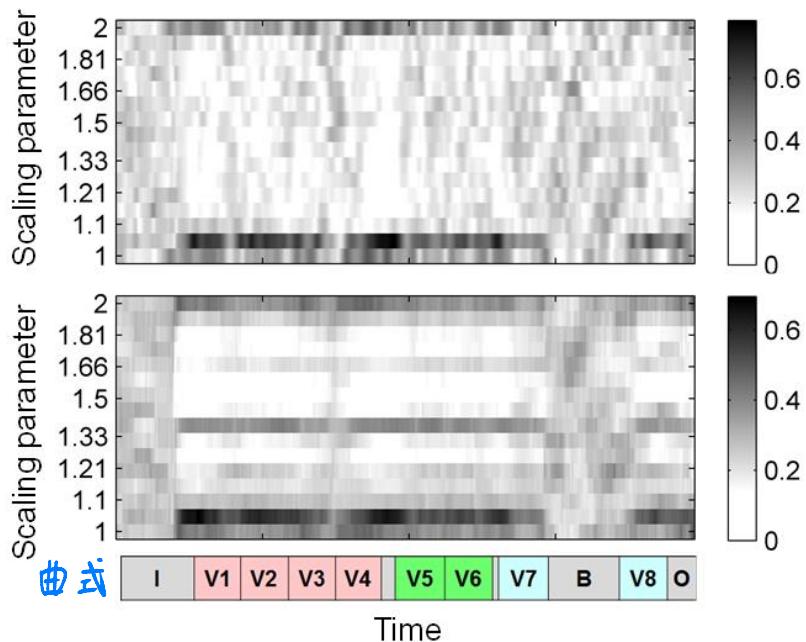
Cyclic tempogram (2)

- Example: a pulse sequence with increasing tempo
- 110 to 130 BPM
- Left: Fourier tempogram
- Right: ACF tempogram



Tempogram and musical structure

- Fourier tempogram (top) and ACF tempogram (bottom)
- “In the year 2525” (Zager and Evans)
- Hungarian Dance No.5 (Johannes Brahms)



From: M. Mueller, *Fundamentals of Music Processing*, Chapter 6, Springer 2015

Comparison

對 novelty curve 做

Fourier tempogram	Autocorrelation tempogram
Comparison of novelty curve with windowed sinusoids with each sinusoid representing a tempo.	Comparison of novelty curve with time-shifted windowed sections of itself with each lag representing a tempo.
Conversion of frequency (Hertz) into tempo (BPM).	Conversion of lag (seconds) into tempo (BPM).
Measurement of novelty periodicities.	Measurement of novelty self-similarities.
Emphasis of tempo harmonics. 很容易高估或低估 tempo	Emphasis of tempo subharmonics. 很容易低估或高估 tempo
Suitable to analyze tempo on tactus and tatum level.	Suitable to analyze tempo on tactus and measure level. or phrase

From: M. Mueller, *Fundamentals of Music Processing*, Chapter 6, Springer 2015

Cyclic tempogram and chromagram

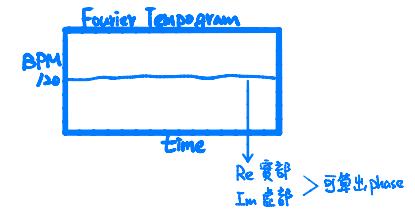
Steps	Chromagram	Cyclic tempogram <small>少被使用，看复习</small>
1.	Analysis of waveform.	Analysis of novelty curve.
2.	Computation of spectrogram.	Computation of tempogram.
3.	Usage of log-frequency axis.	Usage of log-tempo axis.
4.	Cyclic projection.	Cyclic projection.

From: M. Mueller, *Fundamentals of Music Processing*, Chapter 6, Springer 2015

Beat tracking by PLP (1)

- Tempo estimation: identifying the frequency (or periodicity) of the novelty curve
- Beat tracking: identifying the phase of the novelty curve at the local tempo frequency
- Predominant local pulse (PLP)
- Local tempo estimation at time n

$$\tau_n := \operatorname{argmax}_{\tau \in \Theta} T^F(n, \tau)$$



- Local phase estimation: the phase ϕ_n belonging to the windowed sinusoid of tempo τ_n

$$\phi_n = \frac{1}{2\pi} \arccos \left(\frac{\operatorname{Re} \left(F(n, \frac{\tau_n}{60}) \right)}{\left| F(n, \frac{\tau_n}{60}) \right|} \right)$$

Beat tracking by PLP (2)

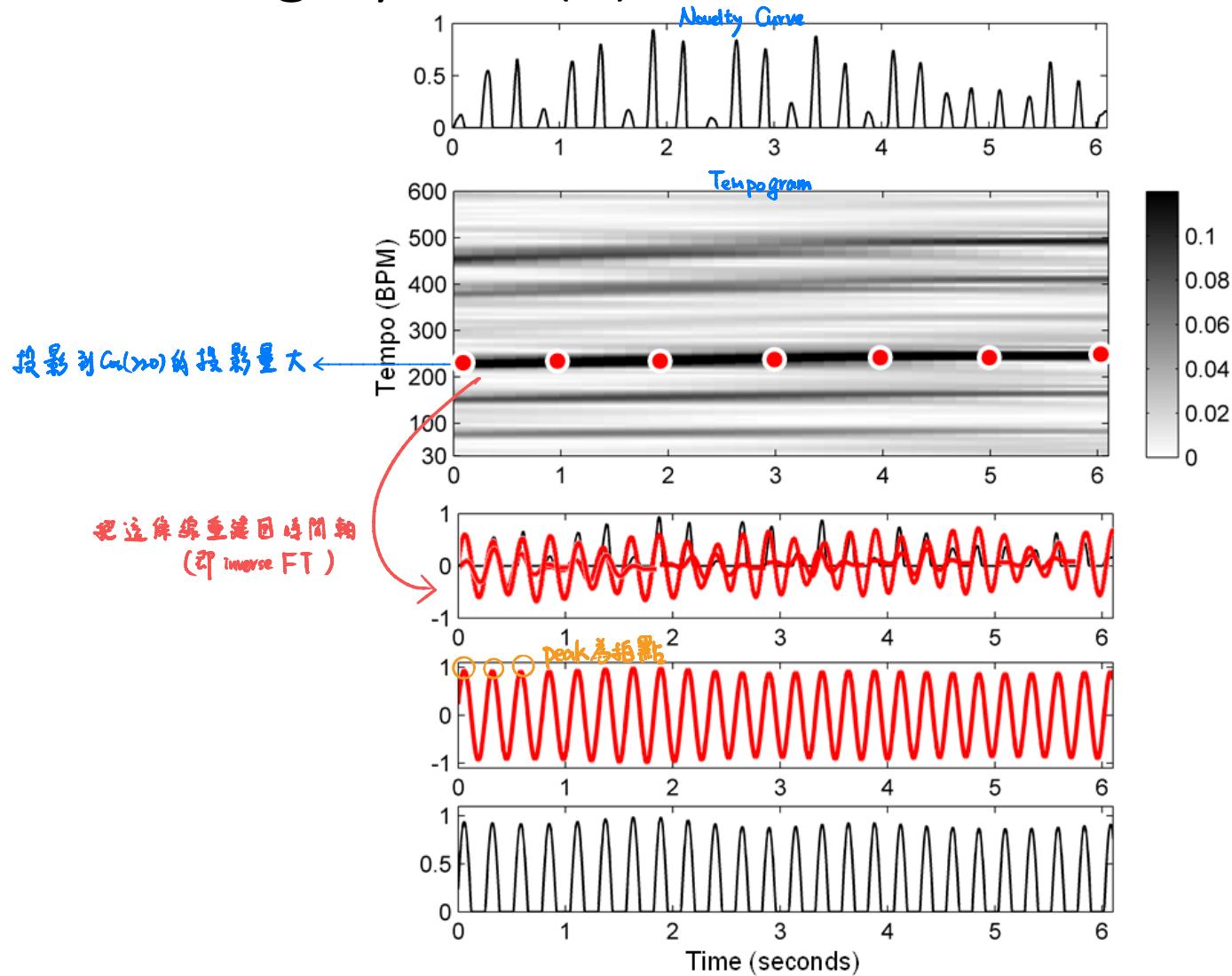
- The optimal window sinusoid “matching” the local novelty curve can be described by a sinusoid **in phase** with the novelty curve

$$\kappa_n(m) := w(m - n) \cos\left(2\pi\left(\frac{\tau_n}{60} \cdot m - \phi_n\right)\right)$$

- The **PLP function**: applying the **overlap-add (OLA)** technique and reconstruct the sinusoids
 - Picking only positive values

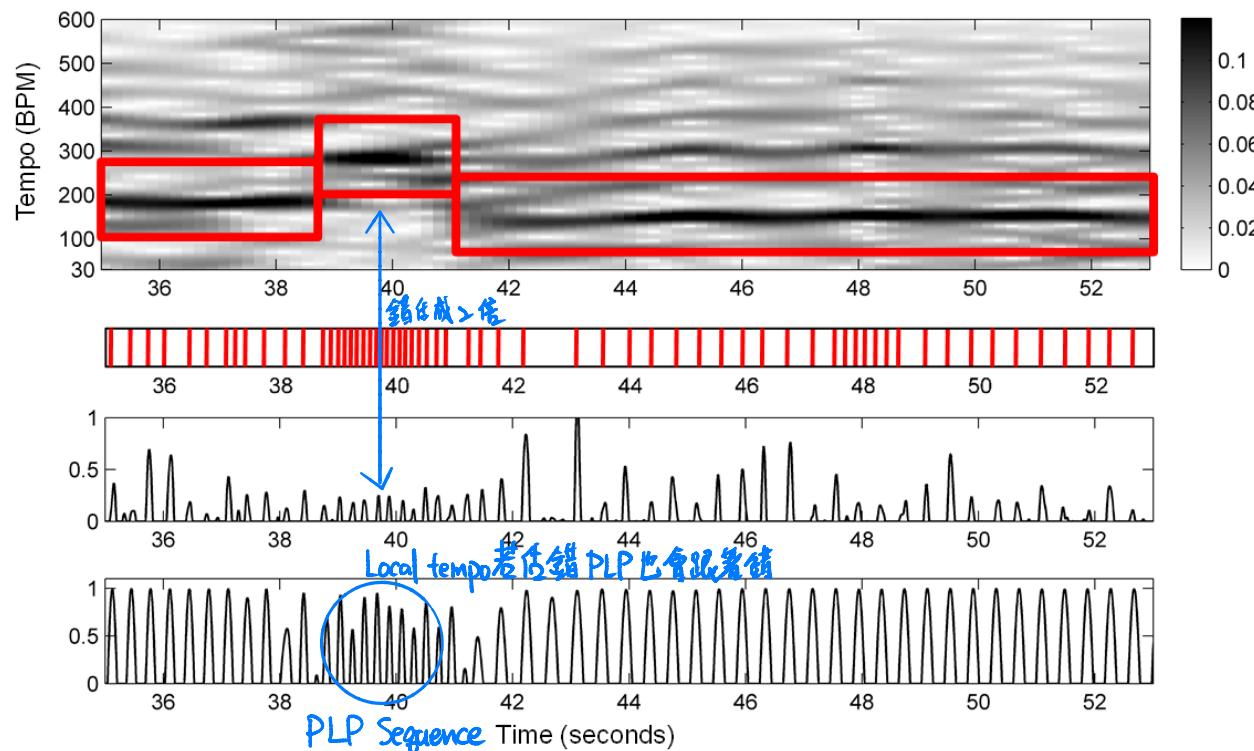
$$\Gamma(m) = \left| \sum_{n \in Z} \kappa_n(m) \right|_{\geq 0}$$

Beat tracking by PLP (3)



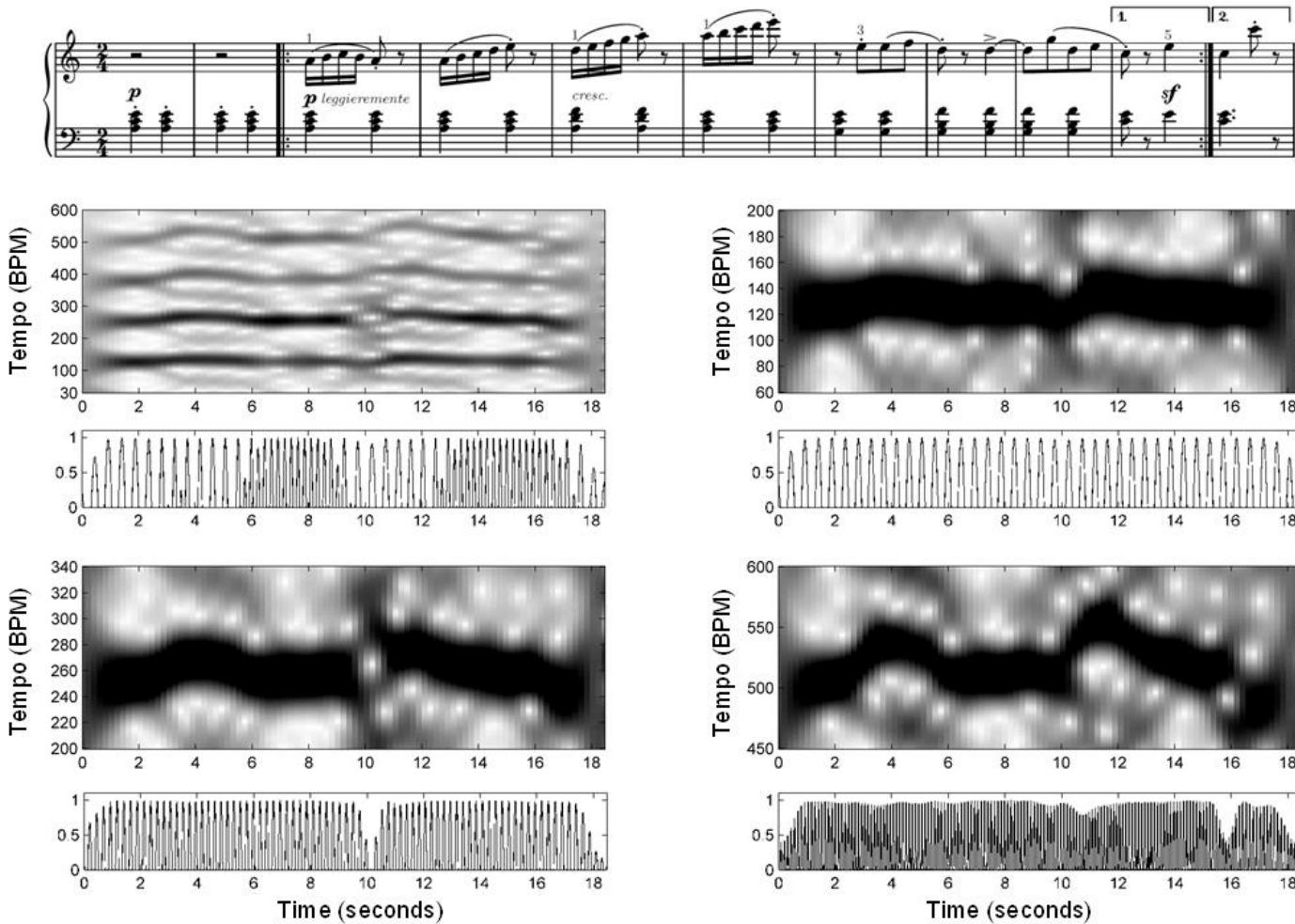
Example of PLP

- Hungarian Dance No.5 (Johannes Brahms)



From: M. Mueller, *Fundamentals of Music Processing*, Chapter 6, Springer 2015

PLP extracted in different bands



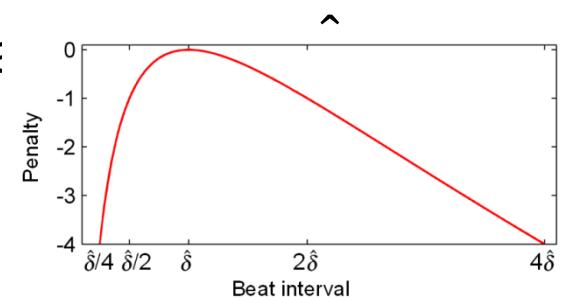
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Dynamic programming for beat tracking

(implemented in librosa)
需給 tempo

- Assumption: the tempo of music is more or less constant
- A beat sequence $B = [b_1, b_2, \dots, b_L]$, estimate $\hat{\delta}$
- Penalty function

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



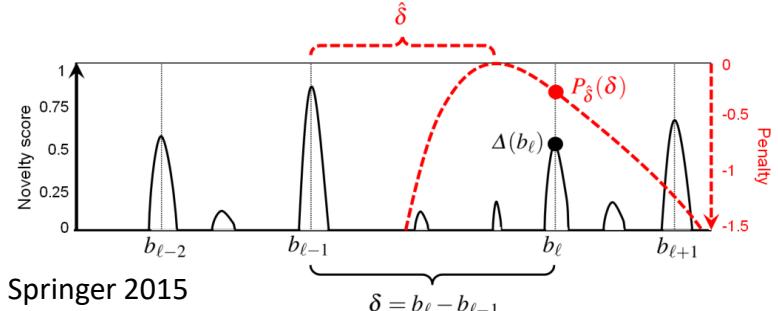
- The objective function

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

novelty curve fit peak
估出來的 tempo , 如果 $b_l - b_{l-1} \approx \hat{\delta}$, 就是 reward
否則就是 penalty

- Let B^N all possible sequence, optimize $B^* :=$

$$\operatorname{argmax}_{B \in B^N} S(B)$$



Algorithm

Algorithm: OPTIMAL BEAT SEQUENCE

Input: Novelty function $\Delta : [1 : N] \rightarrow \mathbb{R}$

Estimate $\hat{\delta}$ for the beat period (given in samples)

Weight parameter $\lambda \in \mathbb{R}$

Output: Optimal beat sequence $B^* = (b_1, b_2, \dots, b_L)$

Procedure: Initialize $\mathbf{D}(0) = 0$ and $\mathbf{P}(0) = 0$. Then compute in a loop for $n = 1, \dots, N$:

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

If $\mathbf{D}(n) = \Delta(n)$ then set $\mathbf{P}(n) = 0$,

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

Set $\ell = 1$ and $a_\ell = \operatorname{argmax}_{n \in [0:N]} \mathbf{D}(n)$. Then repeat the following steps until $\mathbf{P}(a_\ell) = 0$:

Increase ℓ by one.

Set $a_\ell = \mathbf{P}(a_{\ell-1})$.

If $a_\ell = 0$, then set $L = 0$ and return $B^* = \emptyset$.

Otherwise let $L = \ell$ and return $B^* = (a_L, a_{L-1}, \dots, a_1)$.

Recurrent neural networks for beat tracking

annotation 較多
較易取用

- Input: audio features
- Output: the likelihood being beat and downbeat (classification-based output)
- http://www.cp.jku.at/research/papers/Boeck_etal_ISMIR_2016.pdf
- Sebastian Böck, Florian Krebs and Gerhard Widmer, *Joint Beat and Downbeat Tracking with Recurrent Neural Networks*, Proceedings of the 17th International Society for Music Information Retrieval Conference (ISMIR), 2016.
- Böck, Sebastian, et al. "Madmom: A new Python audio and music signal processing library." *Proceedings of the 2016 ACM on Multimedia Conference*. ACM, 2016.

will be used in HW2

Remaining Challenge:

目前只能處理給定 meter (幾拍子)
難處理 dynamic 情況
副拍易難 (拍子的 hierarchy)

Further readings

- Tempogram toolbox
- Beat tracking by dynamic programming (Dan Ellis' code)
 - Dan Ellis, “Beat Tracking by Dynamic Programming,” *J. New Music Research*, Special Issue on Beat and Tempo Extraction, vol. 36 no. 1, March 2007, pp. 51-60.
 - <http://labrosa.ee.columbia.edu/projects/beattrack/>
- Beat tracking with dynamic Bayesian networks
 - A. Srinivasamurthy, A. Holzapfel, A. Cemgil, and X. Serra. Particle filters for efficient meter tracking with Dynamic Bayesian networks. In ISMIR 2015.
- Downbeat tracking with neural networks
 - S. Durand, J. P Bello, B. David, and G. Richard, “Downbeat tracking with multiple features and deep neural networks,” in ICASSP 2015, pp. 409–413.