

FMP Notebooks

Teaching and Learning Fundamentals of Music Processing

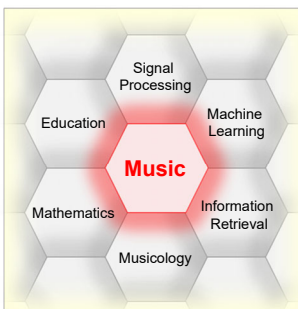
Meinard Müller

International Audio Laboratories Erlangen
meinard.mueller@audiolabs-erlangen.de

Music Processing

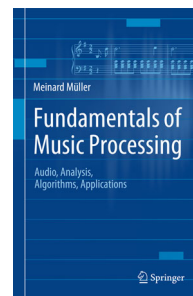


Music Processing: A Multifaceted Research Area



- Music is a ubiquitous and vital part of our lives
- Digital music services: Spotify, Pandora, iTunes, ...
- Music yields intuitive entry point to support and motivate education in technical disciplines
- Music bridges the gap between engineering, computer science, mathematics, and the humanities

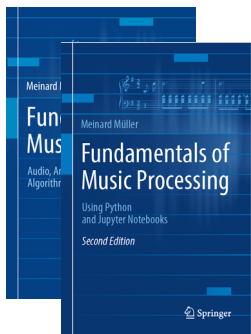
Fundamentals of Music Processing (FMP)



Meinard Müller
Fundamentals of Music Processing
Audio, Analysis, Algorithms, Applications
Springer, 2015

Accompanying website:
www.music-processing.de

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2nd edition
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Fundamentals of Music Processing
Using Python and Jupyter Notebooks
Springer, 2021

Fundamentals of Music Processing (FMP)

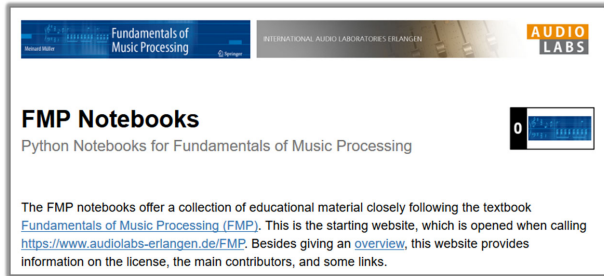
| Chapter | Music Processing Scenario |
|---------|--|
| 1 | Music Representations |
| 2 | Fourier Analysis of Signals |
| 3 | Music Synchronization |
| 4 | Music Structure Analysis |
| 5 | Chord Recognition |
| 6 | Tempo and Beat Tracking |
| 7 | Content-Based Audio Retrieval |
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FMP Notebooks: Education & Research



<https://www.audiolabs-erlangen.de/FMP>

FMP Notebooks: Education & Research

- ... provide educational material for teaching and learning fundamentals of music processing.
- ... combine textbook-like explanations, technical concepts, mathematical details, Python code examples, illustrations, and sound examples.
- ... bridge the gap between theory and practice being based on interactive Jupyter notebook framework.
- ... are freely accessible under a Creative Commons license.

<https://www.audiolabs-erlangen.de/FMP>

FMP Notebooks

| Part | Title | Notions, Techniques & Algorithms | HTML | IPYNB |
|------|--|--|----------------------|-----------------------|
| B | Basics | Basic information on Python, Jupyter notebooks, Anaconda package management system, Python environments, visualizations, and other topics | html | ipynb |
| 0 | Overview | Overview of the notebooks (https://www.audiolabs-erlangen.de/FMP) | html | ipynb |
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- Part B: Basic introductions to
 - Jupyter notebook framework
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 - Other technical concepts underlying these notebooks

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- Part 1 to Part 8: Different music processing scenarios

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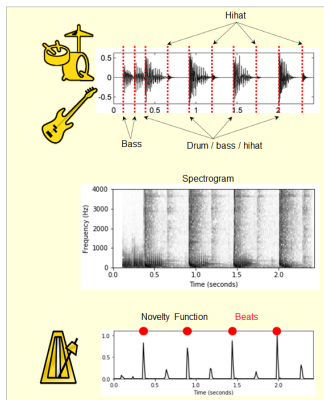
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Part 6: Tempo and Beat Tracking

Part 6: Tempo and Beat Tracking

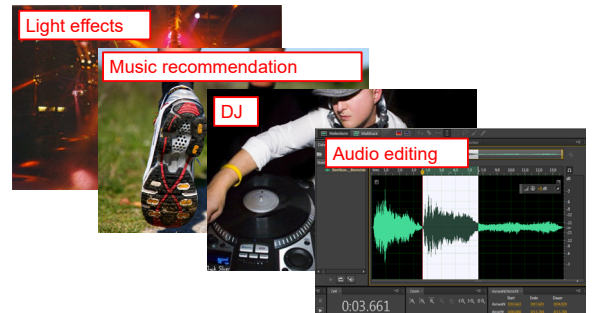


- When listening to a piece of music, we as humans are often able to tap along with the musical beat
- Automated beat tracking: Simulate this cognitive process by a computer



Tempo and Beat Tracking

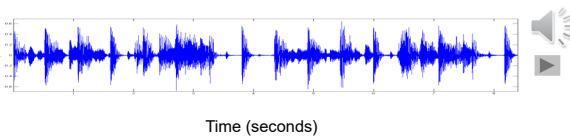
Basic task: "Tapping the foot when listening to music"



Tempo and Beat Tracking

Basic task: "Tapping the foot when listening to music"

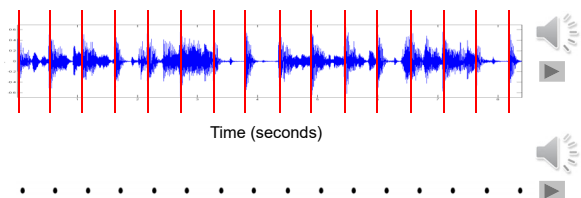
Example: Queen – Another One Bites The Dust



Tempo and Beat Tracking

Basic task: "Tapping the foot when listening to music"

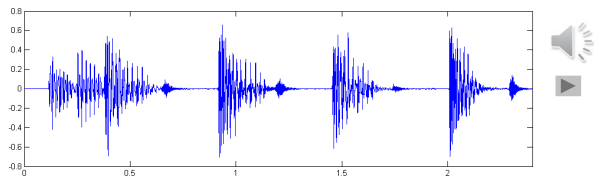
Example: Queen – Another One Bites The Dust



Tempo and Beat Tracking

Tasks

- Onset detection
- Beat tracking
- Tempo estimation



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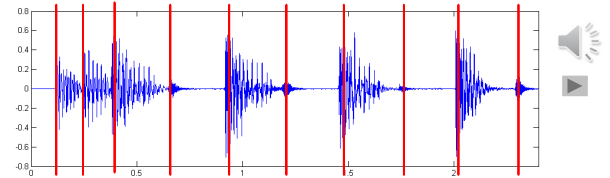
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Tempo and Beat Tracking

Tasks

- Onset detection
- Beat tracking
- Tempo estimation



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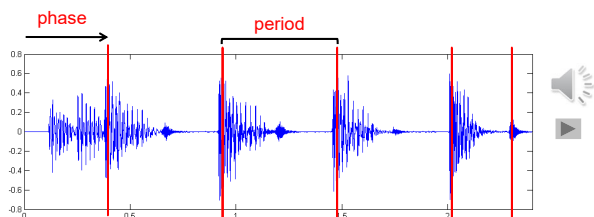
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Tempo and Beat Tracking

Tasks

- Onset detection
- Beat tracking
- Tempo estimation



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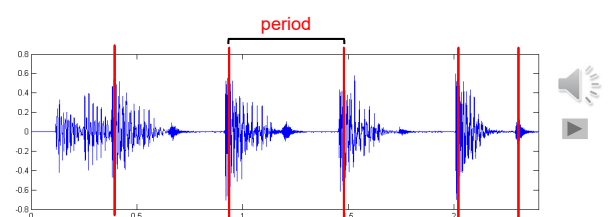
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Tempo and Beat Tracking

Tasks

- Onset detection
 - Beat tracking
 - Tempo estimation
- Tempo := 60 / period
Beats per minute (BPM)



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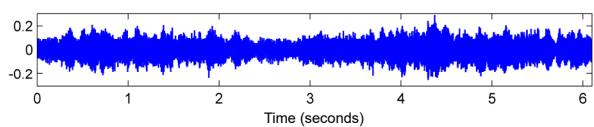
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Onset Detection (Spectral Flux)



Audio recording

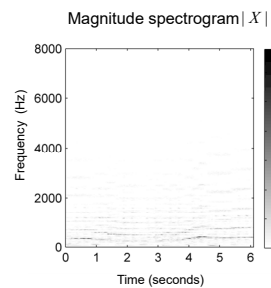


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Onset Detection (Spectral Flux)



Steps:

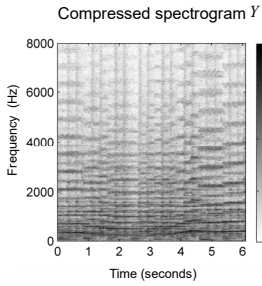
- Spectrogram

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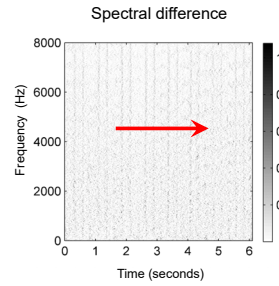
Onset Detection (Spectral Flux)



Steps:

1. Spectrogram
2. Logarithmic compression

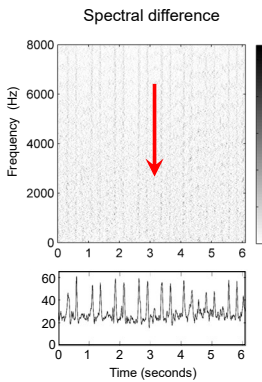
Onset Detection (Spectral Flux)



Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation & half wave rectification

Onset Detection (Spectral Flux)

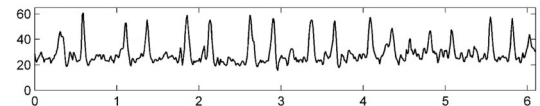


Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation & half wave rectification
4. Accumulation

Onset Detection (Spectral Flux)

Novelty function



Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation & half wave rectification
4. Accumulation

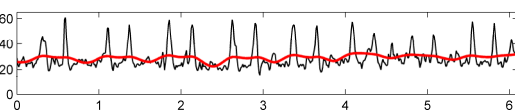
Onset Detection (Spectral Flux)

Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation & half wave rectification
4. Accumulation
5. Normalization

Novelty function

Substraction of local average

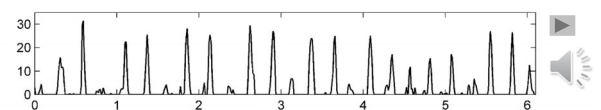


Onset Detection (Spectral Flux)

Steps:

1. Spectrogram
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4. Accumulation
5. Normalization

Normalized novelty function



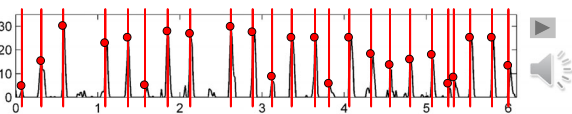
Onset Detection (Spectral Flux)

Steps:

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4. Accumulation
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Normalized novelty function

Peak positions indicate beat candidates



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Onset Detection (Spectral Flux)

Deep Learning

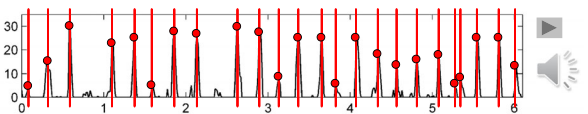
1. Input representation
2. Sigmoid activation
3. Convolution & rectified linear unit (ReLU)
4. Pooling
5. Convolution & ReLU

Steps:

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3. Differentiation & half wave rectification
4. Accumulation
5. Normalization

Normalized novelty function

Peak positions indicate beat candidates



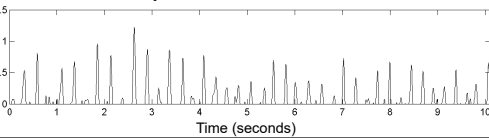
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Local Pulse and Tempo Tracking

Normalized novelty function



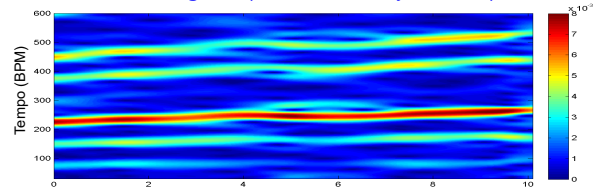
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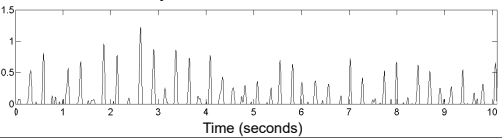
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Local Pulse and Tempo Tracking

Fourier temogram (STFT of novelty function)



Normalized novelty function



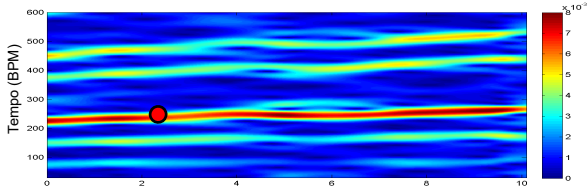
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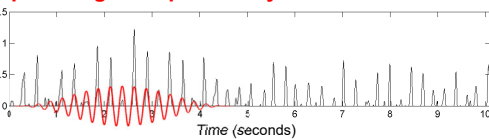
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Fourier temogram (STFT of novelty function)



Optimizing local periodicity kernel



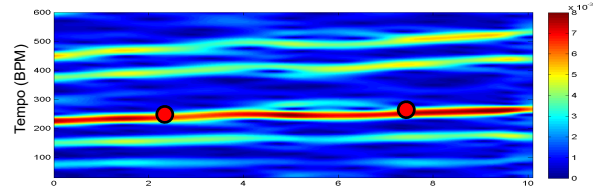
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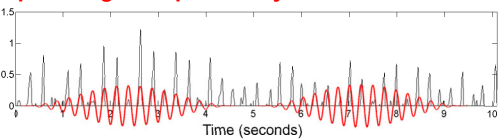
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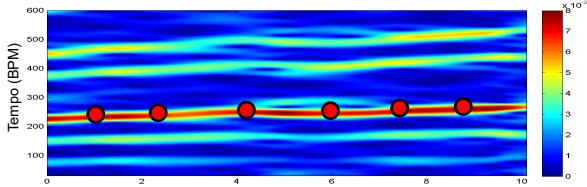
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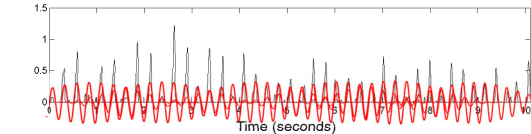
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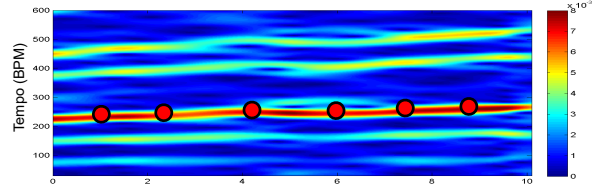
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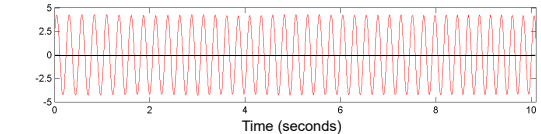


Local Pulse and Tempo Tracking

Fourier temogram (STFT of novelty function)



Accumulation of kernels



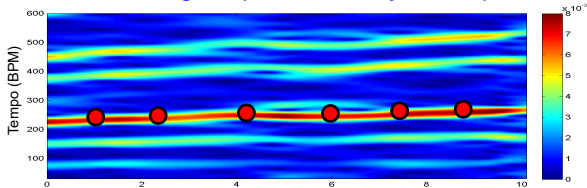
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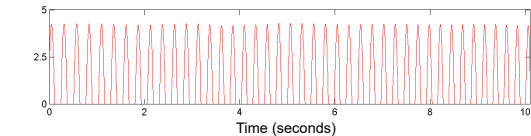


Local Pulse and Tempo Tracking

Fourier temogram (STFT of novelty function)



Halfwave rectification



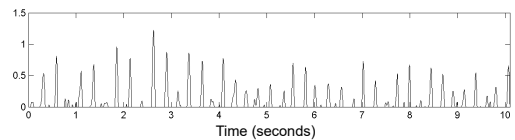
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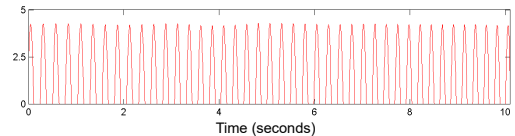


Local Pulse and Tempo Tracking

Novelty Curve



Predominant Local Pulse (PLP)



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| 2 | Fourier Analysis of Signals | Discrete-time signal, sinusoid, exponential, Fourier transforms, Fourier representation, DFT, FFT, STFT | html | ipynb |
| 3 | Music Synchronization | Chroma features, dynamic programming, dynamic time warping (DTW), alignment, user interface | html | ipynb |
| 4 | | Similarity matrix, repetition | html | ipynb |
| 5 | | | html | ipynb |
| 6 | Tempo and Beat Tracking | Onset, novelty, tempo, tempogram, beat, periodicity, Fourier analysis, autocorrelation | html | ipynb |
| 7 | Content-Based Audio Retrieval | Identification, fingerprint, indexing, inverted list, matching, version, cover song | html | ipynb |
| 8 | Musically Informed Audio Decomposition | Harmonic/percussive separation, signal reconstruction, instantaneous frequency, fundamental frequency (F0), trajectory, non-negative matrix factorization (NMF) | html | ipynb |

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FMP Notebooks



Part B: Basics

| Topic | Description |
|--|---|
| Get Started | Explanation on how to install and use the FMP notebooks |
| Installation | Installation of Python using Conda |
| Jupyter Notebook | Usage of Jupyter notebook framework |
| Python Basics | Introduction of data types, control structures, and functions |
| Python Style Guide | Recommendations for programming style |
| Multimedia | Integration of multimedia objects into notebooks |
| Python Visualization | Generation of figures and images |
| Python Audio | Reading and writing audio files |
| Numba | Acceleration of Python functions via JIT compilation |
| Annotation Visualization | Visualization of annotations (single value, segments) |
| Sonification | Sonification methods (onsets, F0 trajectories, pitch, chroma) |
| libfmp | Library of FMP-specific Python functions |
| MIR Resources | Links to resources that are useful for MIR |

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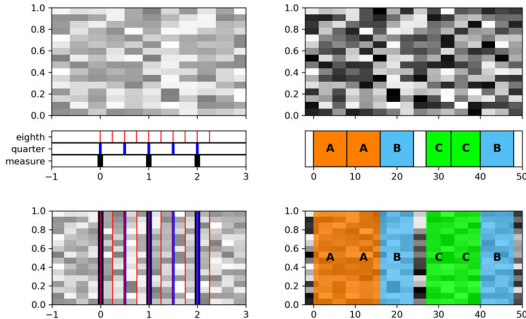
FMP Notebooks



Part B: Basics

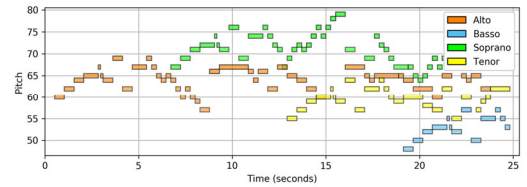
Annotation Visualization

Examples for visualizing annotations of time positions and segments.



Part 1: Music Representations

Symbolic Format: CSV

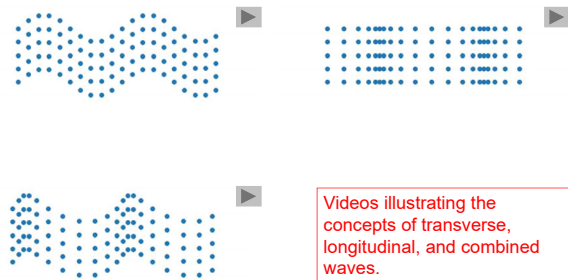


Visualization of a piano-roll representation (Fugue BWV 846 by Bach).



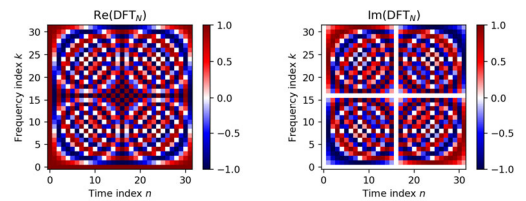
Part 1: Music Representations

Waves and Waveforms



Part 2: Fourier Analysis of Signals

Discrete Fourier Transform (DFT)

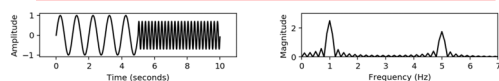


The matrix DFT_N and a visualization of its real and imaginary parts for the case $N = 32$

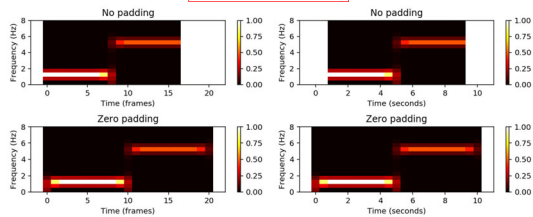
Part 2: Fourier Analysis of Signals

STFT: Padding

Time-domain signal and magnitude Fourier transform.



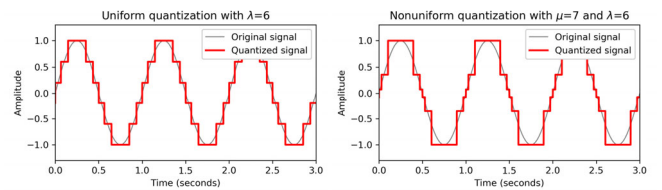
Magnitude STFT.



Part 2: Fourier Analysis of Signals

Digital Signals: Quantization

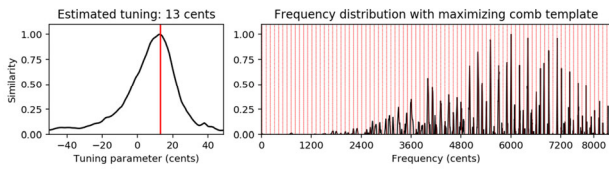
Uniform and nonuniform quantization (based on μ -law encoding) using $\lambda = 6$ quantization levels.



Part 3: Music Synchronization

Transposition and Tuning

Tuning procedure using a comb-filter approach.



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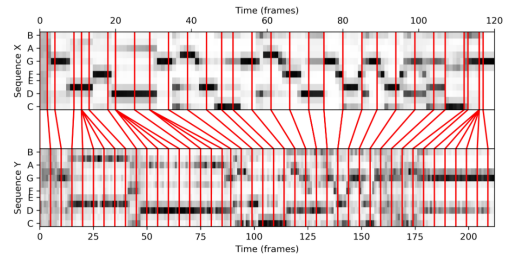
FMP Notebooks



Part 3: Music Synchronization

Music Synchronization

Music synchronization result obtained for two input chromagrams (obtained from two recordings of the beginning of Beethoven's Fifth Symphony).



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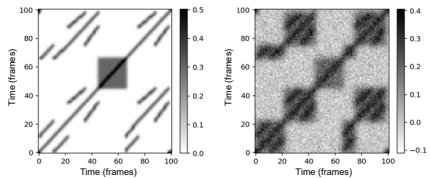
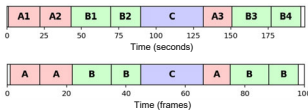


Part 4: Music Structure Analysis

SSM: Synthetic Generation

Structure annotation and different synthetically generated SSMs.

| | start | end | label |
|---|--------|--------|-------|
| 0 | 0.00 | 1.01 | |
| 1 | 1.01 | 22.11 | A1 |
| 2 | 22.11 | 43.06 | A2 |
| 3 | 43.06 | 69.42 | B1 |
| 4 | 69.42 | 89.57 | B2 |
| 5 | 89.57 | 131.64 | C |
| 6 | 131.64 | 150.84 | A3 |
| 7 | 150.84 | 176.96 | B3 |
| 8 | 176.96 | 196.90 | B4 |
| 9 | 196.90 | 199.64 | |



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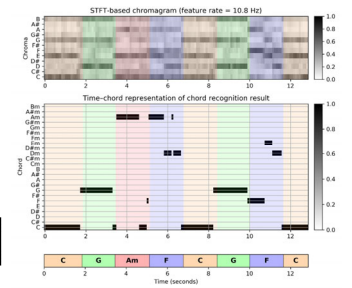
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Part 5: Chord Recognition

Template-Based Chord Recognition

Chord recognition task illustrated by the first measures of the Beatles song "Let It Be."



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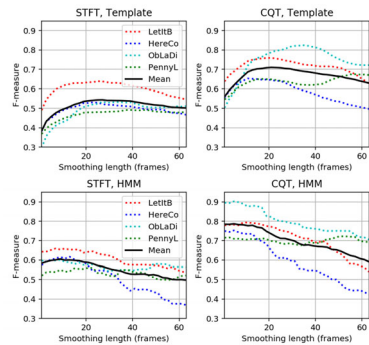


Part 5: Chord Recognition

Experiments: Beatles Collection

Prefiltering experiments for a template-based and an HMM-based chord recognizer applied to two different input chroma representations (SFTF, CQT).

The evaluation is performed on the basis of four Beatles songs (LetItB, HereCo, ObLaDi, PennyL).



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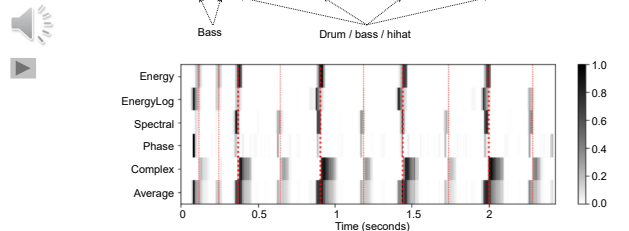
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Part 6: Tempo and Beat Tracking

Novelty: Comparison of Approaches

Comparison of novelty detectors using a matrix-based visualization.



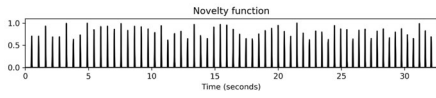
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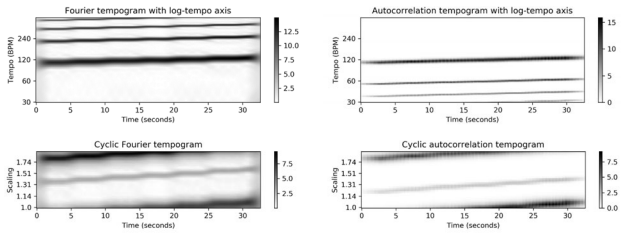


Part 6: Tempo and Beat Tracking

Cyclic Tempogram



Different tempogram representations of a click track with increasing tempo.



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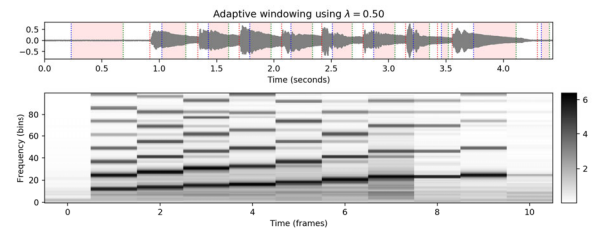
FMP Notebooks



Part 6: Tempo and Beat Tracking

Adaptive Windowing

Example of adaptive windowing using a parameter λ to control the neighborhood's relative size to be excluded.



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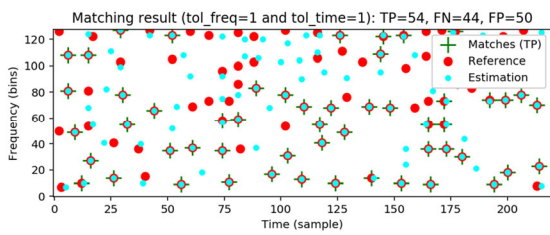
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Part 7: Content-Based Audio Retrieval

Audio Identification

Evaluation measures that indicate the agreement between two constellation maps computed for an original version (Reference) and a noisy version (Estimation).



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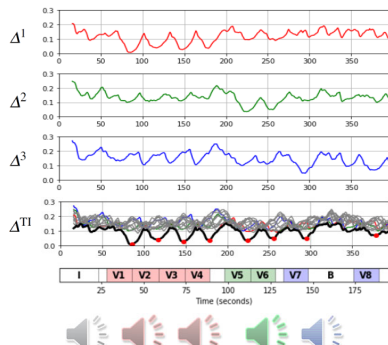
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Part 7: Content-Based Audio Retrieval

Audio Matching

Transposition-invariant matching function illustrated by Zager and Evans' song "In the Year 2525."



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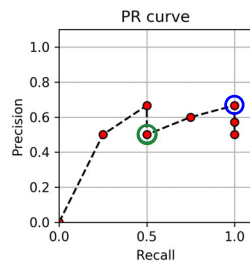
Part 7: Content-Based Audio Retrieval

Evaluation Measures

Various evaluation metrics applied to a toy example.

| Rank | ID | Score | χ_0 | P(r) | R(r) | F(r) |
|------|----|-------|----------|------|------|------|
| 1 | 6 | 3.70 | False | 0.00 | 0.00 | 0.00 |
| 2 | 3 | 3.60 | True | 0.50 | 0.25 | 0.33 |
| 3 | 4 | 3.50 | True | 0.67 | 0.50 | 0.57 |
| 4 | 5 | 3.20 | False | 0.50 | 0.50 | 0.50 |
| 5 | 8 | 3.10 | True | 0.60 | 0.75 | 0.67 |
| 6 | 2 | 2.60 | True | 0.67 | 1.00 | 0.80 |
| 7 | 7 | 1.50 | False | 0.57 | 1.00 | 0.73 |
| 8 | 1 | 0.70 | False | 0.50 | 1.00 | 0.67 |

Break-even point = 0.50
 $F_{max} = 0.80$
 Average precision = 0.60833



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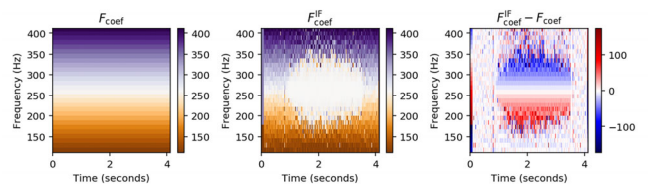
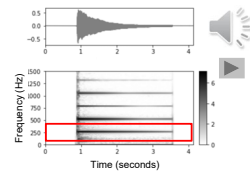
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Part 8: Audio Decomposition

Instantaneous Frequency Estimation

Interpretation of time-frequency bins of an STFT using (frame-dependent) instantaneous frequency values.



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Part 8: Audio Decomposition

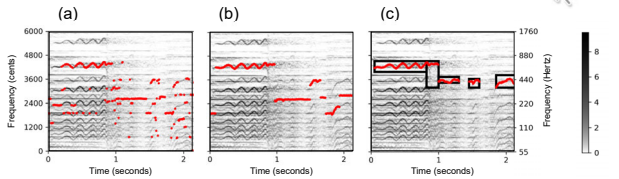
Fundamental Frequency Tracking

Saliency representation with trajectories computed by

- a frame-wise approach,
- an approach using continuity constraints, and
- a score-informed approach.



Figure 8.10a from [Müller, FMP, Springer 2015]



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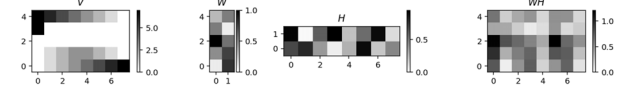


Part 8: Audio Decomposition

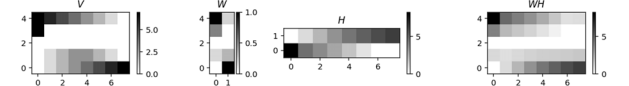
Nonnegative Matrix Factorization (NMF)

NMF procedure applied to a toy example.

Matrix V and randomly initialized matrices W and H .



Matrix V and matrices W and H after training.



Error terms over iteration.



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FMP Notebooks



Browser screenshot showing the FMP Notebooks website. The URL is <https://www.audiolabs-erlangen.de/resources/MIR/FMP/CO/CO.html>. The page title is 'FMP Notebooks' and the subtitle is 'Python Notebooks for Fundamentals of Music Processing'. A navigation bar shows 9 numbered thumbnails.

<https://www.audiolabs-erlangen.de/FMP>

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Basics + 8 Chapters

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Basics + 8 Chapters

Tempo and Beat Tracking

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Tempo and Beat Tracking

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FMP Notebooks



Tempo and Beat Tracking

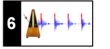


Definition

We assume that we are given a discrete-time novelty function $\Delta : \mathbb{Z} \rightarrow \mathbb{R}$ indicate note onset candidates. The idea of Fourier analysis is to detect local periodicity in novelty curve by comparing it with windowed sinusoids. A high correlation of Δ with a windowed sinusoid indicates a periodicity of the novelty curve (given a suitable phase). This correlation (along with the phase) can be computed using the short-time Fourier transform. To this end, we fix a window function $w : \mathbb{Z} \rightarrow \mathbb{R}$ of length centered at $n = 0$ (e.g., a sampled Hann window). Then, for a frequency parameter $\omega \in \mathbb{R}_{>0}$ and time parameter $n \in \mathbb{Z}$, the complex Fourier coefficient is defined by

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

Tempo and Beat Tracking



Definition

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Mathematics

Explanations

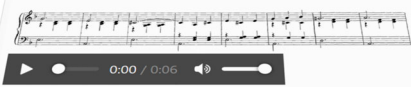
Theory

Tempo and Beat Tracking



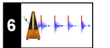
Example: Shostakovich

In the following example, we consider an excerpt of a recording of Dimitri Shostakovich's Suite for Variety Orchestra No. 1. The score version of the excerpt.



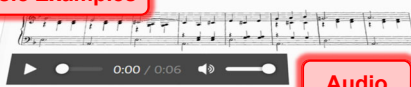
We start with a spectral-based novelty function resampled to F_s^{Δ} . Furthermore, we use a window size corresponding to 5 seconds (1

Tempo and Beat Tracking



Example: Shostakovich

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We start with a spectral-based novelty function resampled to F_s^{Δ} . Furthermore, we use a window size corresponding to 5 seconds (1

Music Examples

Annotations

Audio

Links

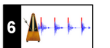
Tempo and Beat Tracking



```
In [2]: def compute_sinusoid_optimal(c, tempo, n, Fs, N)
        """Compute windowed sinusoid with optimal phase
        Notebook: C6/C6S2_TempogramFourier.ipynb

        Args:
        c: Coefficient of tempogram (c=X(k,n))
        tempo: Tempo parameter corresponding to
        _coef_BPM[k]
        n: Frame parameter of c
        Fs: Sampling rate
        N: Window length
        H: Hop size
```

Tempo and Beat Tracking



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```

Python Code

Algorithms

Functions

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FMP Notebooks

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References

- Meinard Müller: Fundamentals of Music Processing – Using Python and Jupyter Notebooks. 2nd Edition, Springer, 2021. <https://www.springer.com/gp/book/9783030698072>
- Meinard Müller and Frank Zalkow: libfmp: A Python Package for Fundamentals of Music Processing. Journal of Open Source Software (JOSS), 6(63): 1–5, 2021. <https://joss.theoj.org/papers/10.21105/joss.03326>
- Meinard Müller: An Educational Guide Through the FMP Notebooks for Teaching and Learning Fundamentals of Music Processing. Signals, 2(2): 245–285, 2021. <https://www.mdpi.com/2624-6120/2/2/18>
- Meinard Müller and Frank Zalkow: FMP Notebooks: Educational Material for Teaching and Learning Fundamentals of Music Processing. Proc. International Society for Music Information Retrieval Conference (ISMIR): 573–580, 2019. <https://zenodo.org/record/3527872#.YQhEQOqzaUk>
- Meinard Müller, Brian McFee, and Katherine Kinnaird: Interactive Learning of Signal Processing Through Music: Making Fourier Analysis Concrete for Students. IEEE Signal Processing Magazine, 38(3): 73–84, 2021. <https://ieeexplore.ieee.org/document/9418542>

Resources (Group Meinard Müller)

- FMP Notebooks: <https://www.audiolabs-erlangen.de/FMP>
- libfmp: <https://github.com/meinardmueller/libfmp>
- synctoolbox: <https://github.com/meinardmueller/synctoolbox>
- libtsm: <https://github.com/meinardmueller/libtsm>
- Preparation Course Python (PCP) Notebooks: <https://www.audiolabs-erlangen.de/resources/MIR/PCP/PCP.html>
<https://github.com/meinardmueller/PCP>

Resources

- librosa: <https://librosa.org/>
- madmom: <https://github.com/CPJKU/madmom>
- Essentia Python tutorial: https://essentia.upf.edu/essentia_python_tutorial.html
- mirdata: <https://github.com/mir-dataset-loaders/mirdata>
- open-unmix: <https://github.com/sigsep/open-unmix-pytorch>
- Open Source Tools & Data for Music Source Separation: <https://source-separation.github.io/tutorial/landing.html>

