

### Introduction to Audio Content Analysis

module 9.3: onset detection

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## introduction overview



#### corresponding textbook section

section 9.3

#### lecture content

- detection of the start of musical events
- fundamental methods for generating a novelty function
- fundamental methods for peak picking

#### ■ learning objectives

- describe the term onset
- implement an automatic onset detection system



module 9.3: onset detection  $1 \ / \ 1$ 

## introduction overview



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module 9.3: onset detection  $1 \ / \ 1$ 

## onset detection problem statement



- onset: begin of musical event
- goal: detect the point in time of an onset
- challenges:
  - which time stamp of the initial attack time actually marks the onset time?
  - polyphonic audio signals:
    - unknown number of voices and events
    - multiple onsets occur at "the same" time
    - onset might be obfuscated by other musical content

#### note onset time:

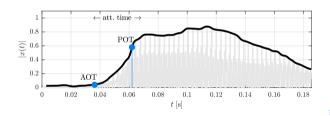
• time the instrument is triggered

#### acoustic onset time:

• time of first *measurable* instrument output

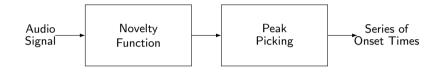
#### perceptual onset time:

 time the event is perceived by listener



## onset detection overview





### 1 novelty function

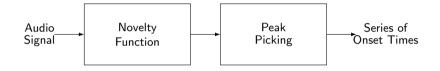
measure of probability for new events/signal change over time

#### 2 peak picking

identify the most likely locations for onsets

## onset detection overview





### novelty function

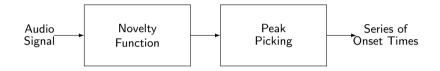
• measure of probability for new events/signal change over time

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identify the most likely locations for onsets

## onset detection overview





### novelty function

• measure of probability for new events/signal change over time

#### 2 peak picking

identify the most likely locations for onsets



- alternative **terms** for *novelty function* 
  - detection function
  - difference function

#### processing steps

- 1 extract features
- 2 compute derivative
- 3 smooth result
- 4 apply Half-Wave-Rectification (HWR)



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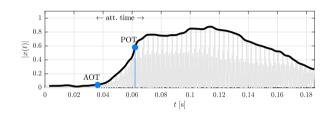
overview intro **novelty function** peak picking eval summary

# onset detection novelty function examples 1/3

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#### 1 time domain example

- feature: time domain envelope
- *derivative*: slope of envelope
  - HWR: only interested in onsets, not offsets



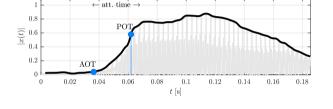
#### 2 pitch domain:

- feature: pitch contour
- derivative: changes in pitch

# onset detection novelty function examples 1/3

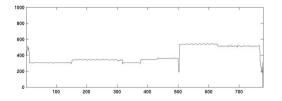
#### 1 time domain example

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### 2 pitch domain:

- feature: pitch contour
- derivative: changes in pitch



<sup>&</sup>lt;sup>1</sup>N. Collins, "Using a pitch detector for onset detection," in ISMIR, 2005, pp. 100–106.

matlab source: plotOnset.m

## onset detection novelty function examples 2/3



#### 3 STFT-based: compute block difference

flux

- cosine distance
- complex

## onset detection novelty function examples 2/3



#### 3 STFT-based: compute block difference

flux

$$\begin{array}{l} \blacktriangleright \ \, d_{\mathrm{hai}}(n) = \sum\limits_{k=0}^{K/2-1} \log_2 \left( \frac{|X(k,n)|}{|X(k,n-1)|} \right) \\ \\ \blacktriangleright \ \, d_{\mathrm{lar}}(n) = \sum\limits_{k=k(f_{\mathrm{min}})}^{k(f_{\mathrm{max}})} \sqrt{|X(k,n)|} - \sqrt{|X(k,n-1)|} \end{array}$$

cosine distance

$$b \ d_{\text{foo}}(n) = 1 - \frac{\sum\limits_{k=0}^{\mathcal{K}/2-1} |X(k,n)| \cdot |X(k,n-1)|}{\sqrt{\left(\sum\limits_{k=0}^{\mathcal{K}/2-1} |X(k,n)|^2\right) \cdot \left(\sum\limits_{k=0}^{\mathcal{K}/2-1} |X(k,n-1)|^2\right)} }$$

complex

## onset detection novelty function examples 2/3

#### 3 STFT-based: compute block difference

flux

$$d_{\text{hai}}(n) = \sum_{k=0}^{K/2-1} \log_2\left(\frac{|X(k,n)|}{|X(k,n-1)|}\right)$$

$$d_{\text{lar}}(n) = \sum_{k=0}^{K(f_{\text{max}})} \sqrt{|X(k,n)|} - \sqrt{|X(k,n-1)|}$$

novelty function

cosine distance

complex

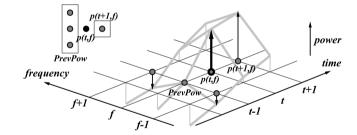
$$d_{\text{dux}}(n) = \sum_{k=0}^{K/2-1} |X(k,n) - X(k,n-1)|$$

## onset detection novelty function examples 3/3



#### 3 STFT-based cont'd

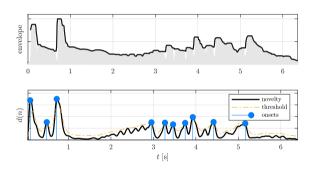
- Goto-distance<sup>2</sup>
  - higher power than closest preceding and following bins



<sup>&</sup>lt;sup>2</sup>M. Goto and Y. Muraoka, "Music Understanding At The Beat Level – Real-time Beat Tracking For Audio Signals," in *Proceedings of the Workshop on Computational Auditory Scene Analysis (IJCAI)*, Aug. 1995.

#### ■ typical criteria

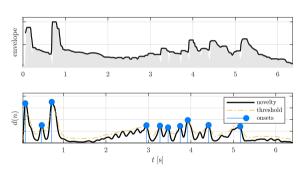
- local maximum & salientt peak
- higher than minimum likelihood
- not too close to maxima with higher likelihood
- other options: high attack slope, distance to prev. min. . . .



## onset detection peak picking: introduction

Georgia Center for Music Tech Technology

- detect onsets in the smoothed novelty function
- typical criteria
  - local maximum & salient peak
  - higher than minimum likelihood
  - not too close to maxima with higher likelihood
  - other options: high attack slope, distance to prev. min, . . .



# onset detection peak picking: thresholding

Georgia Center for Music Tech (1) Technology Callege of Design

- options for thresholding
  - fixed threshold

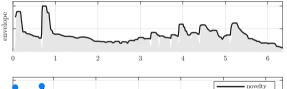
$$G_{d,c} = \lambda_1$$

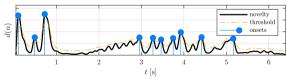
• smoothed threshold

$$G_{d,\mathrm{ma}} = \lambda_2 + \sum_{j=0}^{\mathcal{O}-1} b(j) \cdot d(i-j)$$

• median threshold

$$G_{d,\mathrm{me}} = \lambda_2 + \hat{Q}_d(0.5)$$





erview intro novelty function peak picking **eval** summar

## onset detection evaluation



#### ■ goal

 compare a series of ground truth onset time stamps with a series of predicted time stamps

#### ■ ground truth annotation problems

- deviations between annotators
- how to annotate quasi-synchronous onsets

#### metrics

- measure TP with tolerance range  $\Rightarrow$  TP, FN, FP (TN only implicitly)
- Precision, Recall, F-Measure
- other metrics
  - ► mean (absolute) deviation
  - standard deviation
  - max deviation

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module 9.3: onset detection  $11 \ / \ 1$ 

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module 9.3: onset detection  $11 \ / \ 1$ 

### summary lecture content



#### novelty function

- measure of unexpectedness likelihood of an event
  - often a measure similar to flux

#### peak picking

- detecting peaks (onsets) in the novelty function
- usually done by smoothing and adaptive thresholding

