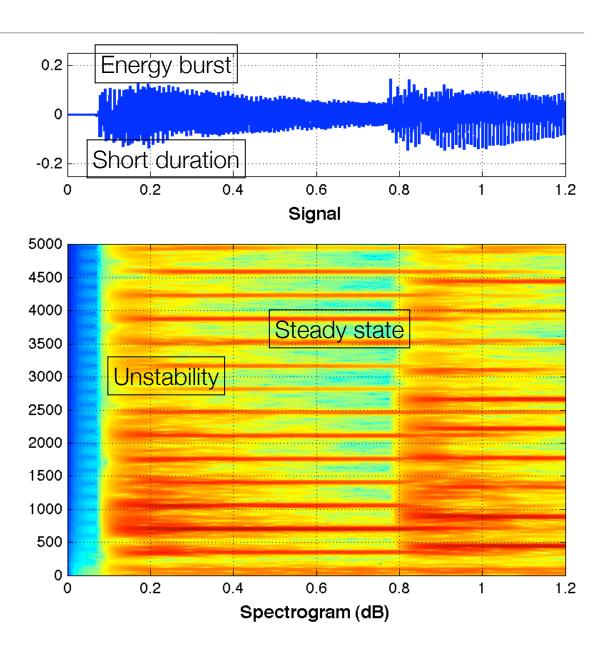
## Novelty detection

Juan Pablo Bello EL9173 Selected Topics in Signal Processing: Audio Content Analysis NYU Poly

## Novelty detection

• Find the start time (onset) of new events in the audio signal.

 Onset: single instant chosen to mark the start of the (attack) transient.

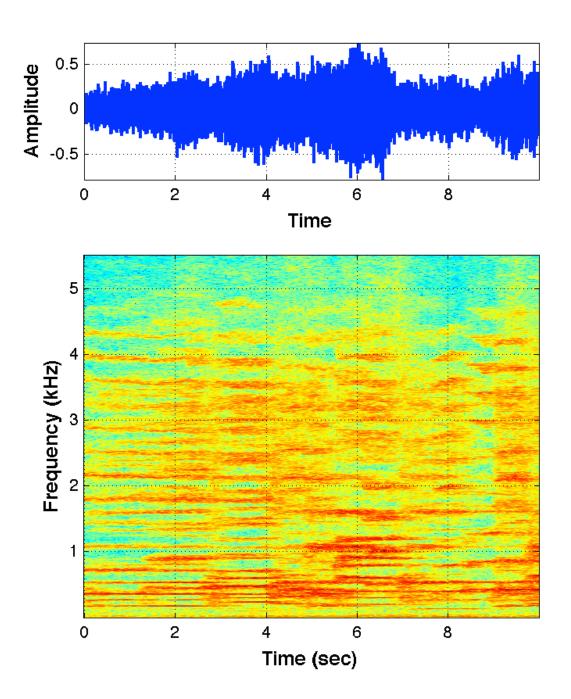


# Applications

- Identifying regions of interest in environmental recordings
- Segmentation of word/phonemes in speech, notes in music
- First layer of rhythm analysis
- Sound manipulation and synthesis: <u>http://www.music.mcgill.ca/~hockman/</u> projects/ARTMA/index.html
- Computational biology?!! <u>http://isophonics.net/content/calcium-signal-analyser</u>

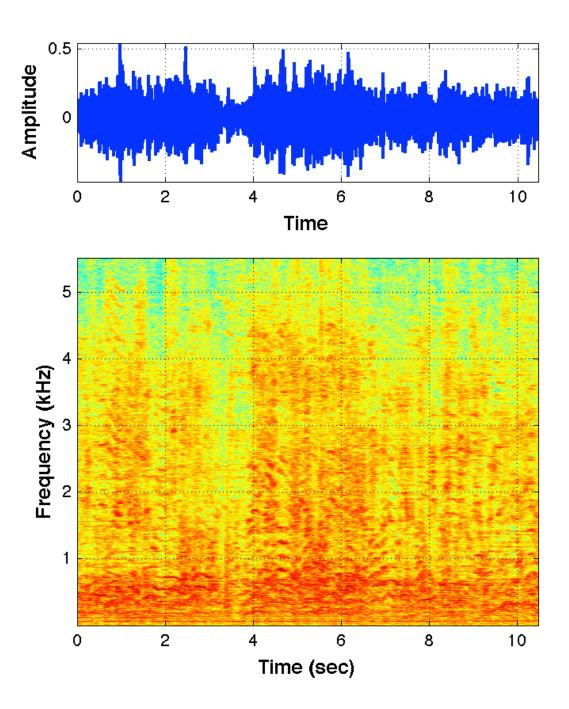
- transient extended in time
- multiple voices ->

   (a)synchronous onsets
- ambiguous events (vibrato, tremolo, glissandi)
- perceptual vs physical



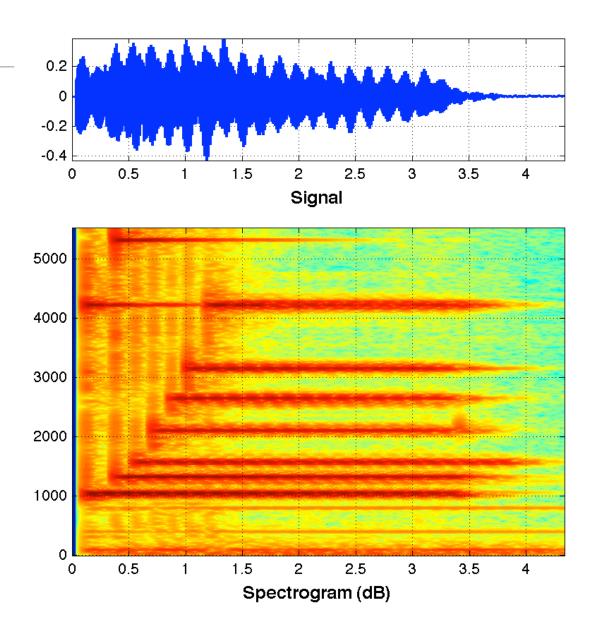
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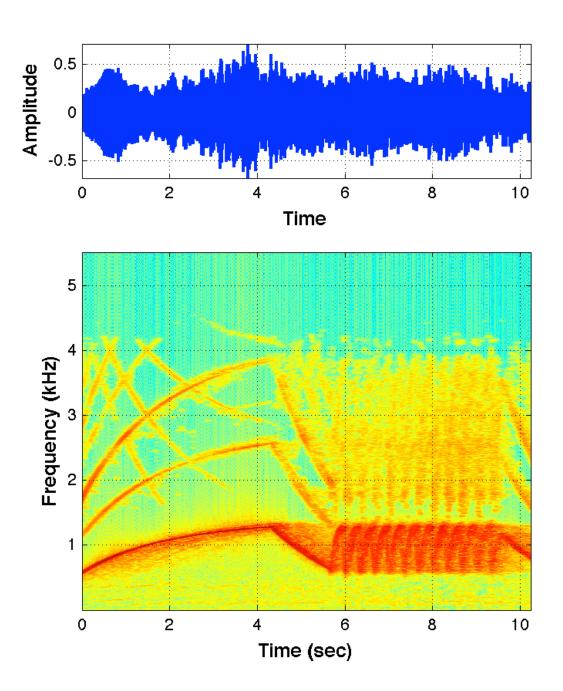
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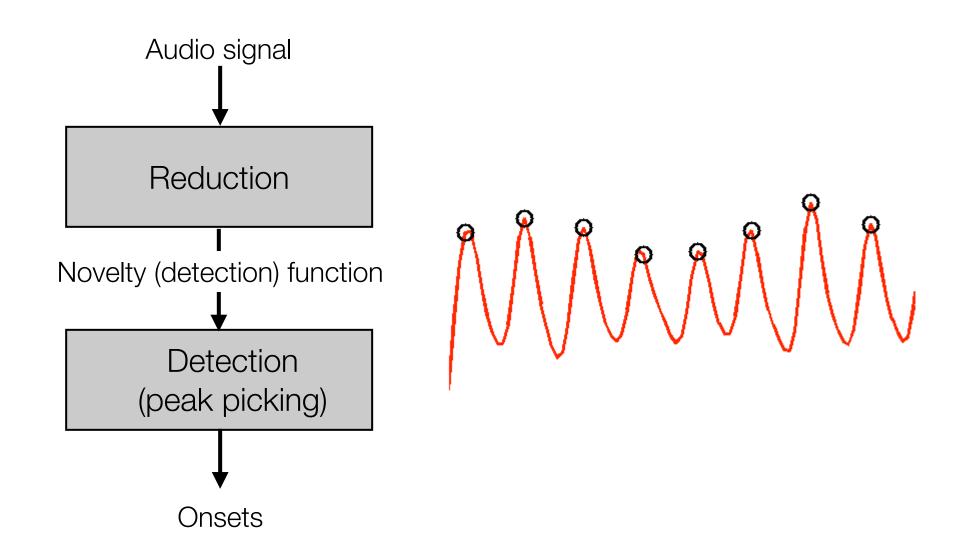


- transient extended in time
- multiple voices ->

   (a)synchronous onsets
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#### Architecture

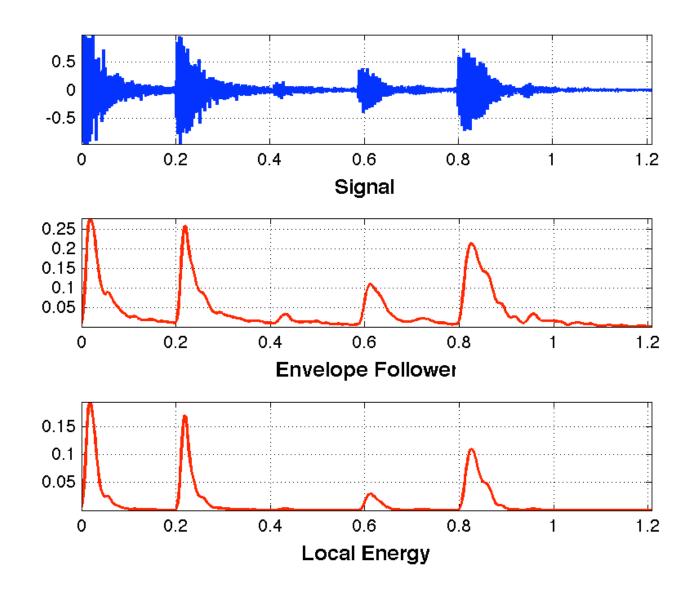


- Onsets: often characterized by an amplitude increase
- Envelope following (full-wave rectification + smoothing):

$$E_0(m) = \frac{1}{N} \sum_{n=-N/2}^{N/2} |x(n+mh)| w(n)$$

• Squaring instead of rectifying, results in the local energy:

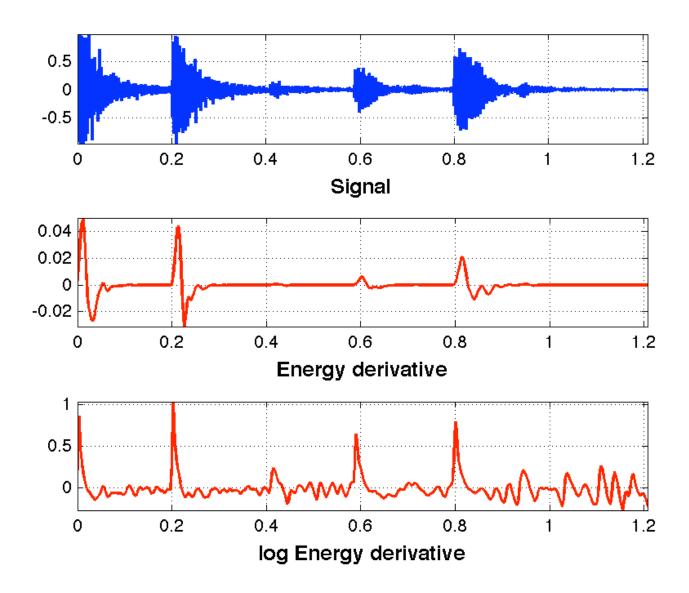
$$E(m) = \frac{1}{N} \sum_{n=-N/2}^{N/2} \left( x(n+mh) \right)^2 w(n)$$



- We can use the derivative of energy w.r.t. time -> sharp peaks during energy rise
- Detectable changes in loudness are proportional to the overall loudness of the sound.

$$\frac{\partial E(m)/\partial m}{E(m)} = \frac{\partial log(E(m))}{\partial m}$$

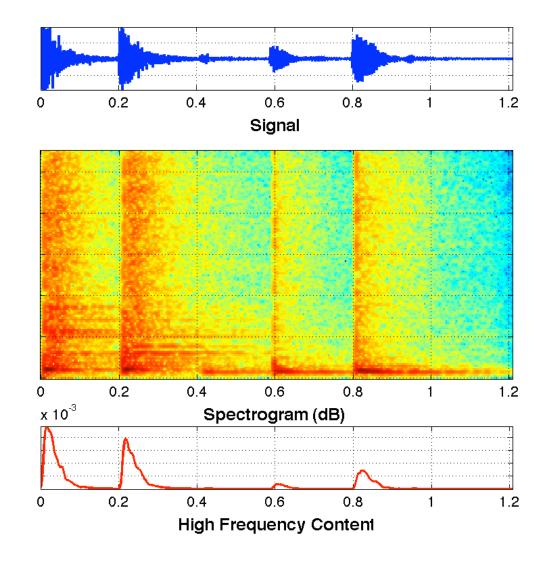
• Simulates the ear's perception of loudness (Klapuri, 1999)



### Frequency-domain

- Impulsive noise in time -> wide band noise in frequency
- More noticeable in high frequencies.
- Linear weighting of Energy

$$HFC(m) = \frac{2}{N} \sum_{k=0}^{N/2} |X_k(m)|^2 k$$



#### Frequency-domain

• We can also measure change (flux) in spectral content (Duxbury, 02).

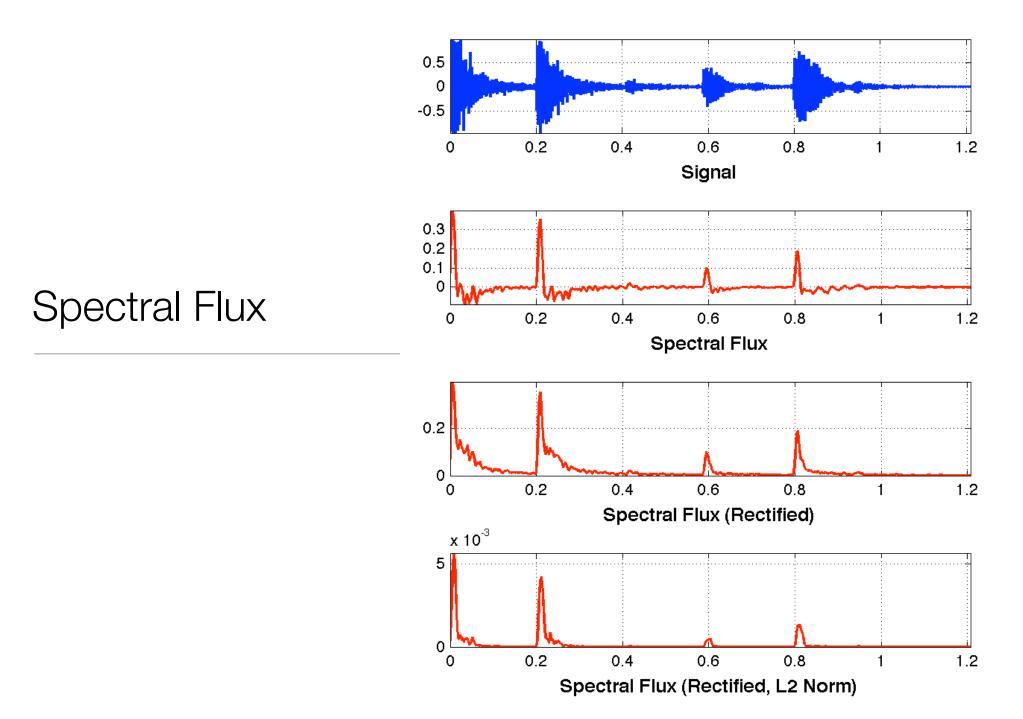
$$SF(m) = \frac{2}{N} \sum_{k=0}^{N/2} \left( |X_k(m)| - |X_k(m-1)| \right)$$

Use half-wave rectification to only take energy increases into account

$$SF_R(m) = \frac{2}{N} \sum_{k=0}^{N/2} H\left(|X_k(m)| - |X_k(m-1)|\right)$$
$$H(x) = (x + |x|)/2$$

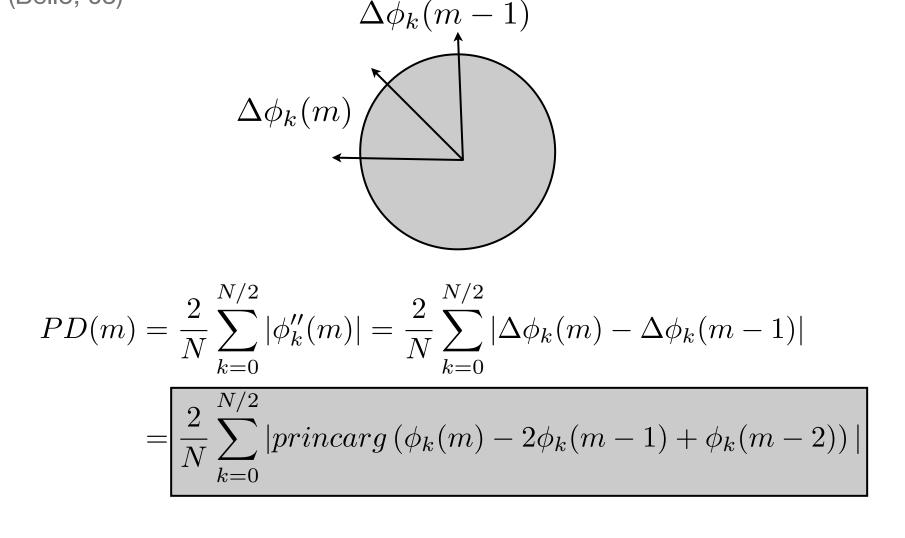
• Use the (squared) L<sub>2</sub> norm of the flux vector

$$SF_{RL_2}(m) = \frac{2}{N} \sum_{k=0}^{N/2} \left[ H\left( |X_k(m)| - |X_k(m-1)| \right) \right]^2$$



### Phase deviation

 The change in instantaneous frequency in bin k can be used to detect onsets (Bello, 03)



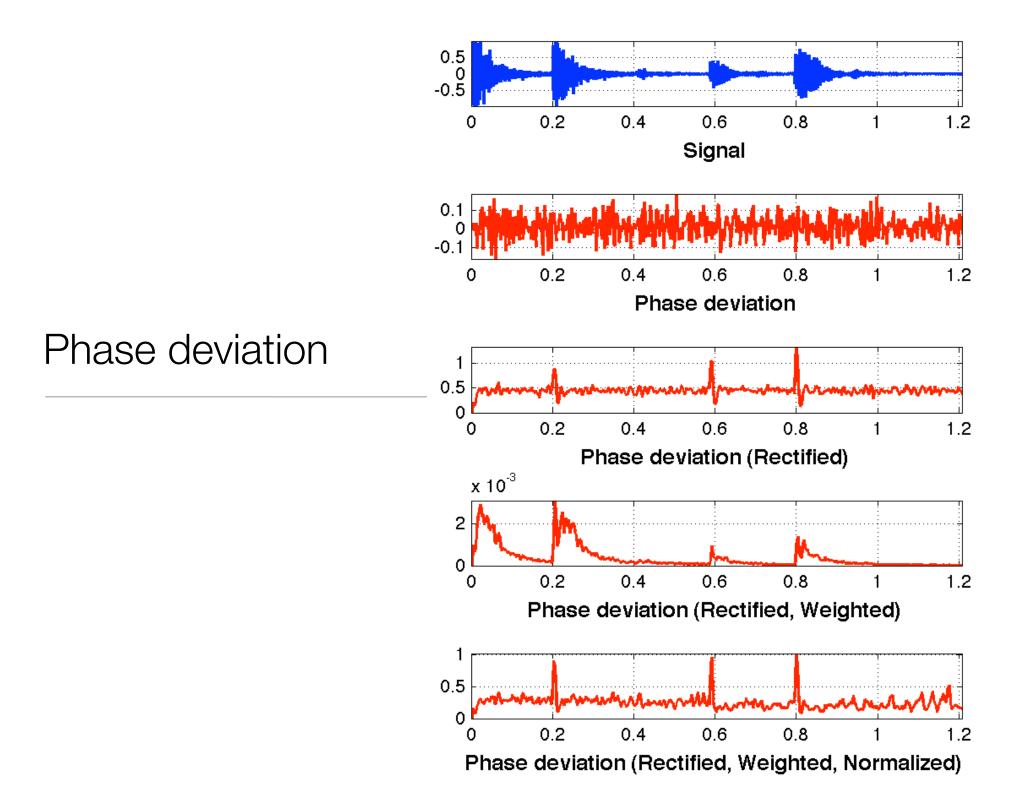
#### Phase deviation

• This function can be improved by weighting frequency bins by their magnitude (Dixon, 06):

$$PD_W(m) = \frac{2}{N} \sum_{k=0}^{N/2} |X_k(m)\phi_k''(m)|$$

• and normalized:

$$PD_{WN}(m) = \frac{\sum_{k=0}^{N/2} |X_k(m)\phi_k''(m)|}{\sum_{k=0}^{N/2} |X_k(m)|}$$



#### Complex domain

• We can combine the spectral flux and phase deviation strategies, such that:

$$\hat{X}_k(m) = |\hat{X}_k(m)| e^{j\hat{\phi}_k(m)}$$

where,

$$\begin{aligned} |\hat{X}_k(m)| &= |X_k(m-1)|\\ \hat{\phi}_k(m) &= princarg\left(2\phi_k(m-1) - \phi_k(m-2)\right) \end{aligned}$$

such that,

$$CD(m) = \frac{2}{N} \sum_{k=0}^{N/2} |X_k(m) - \hat{X}_k(m)|$$

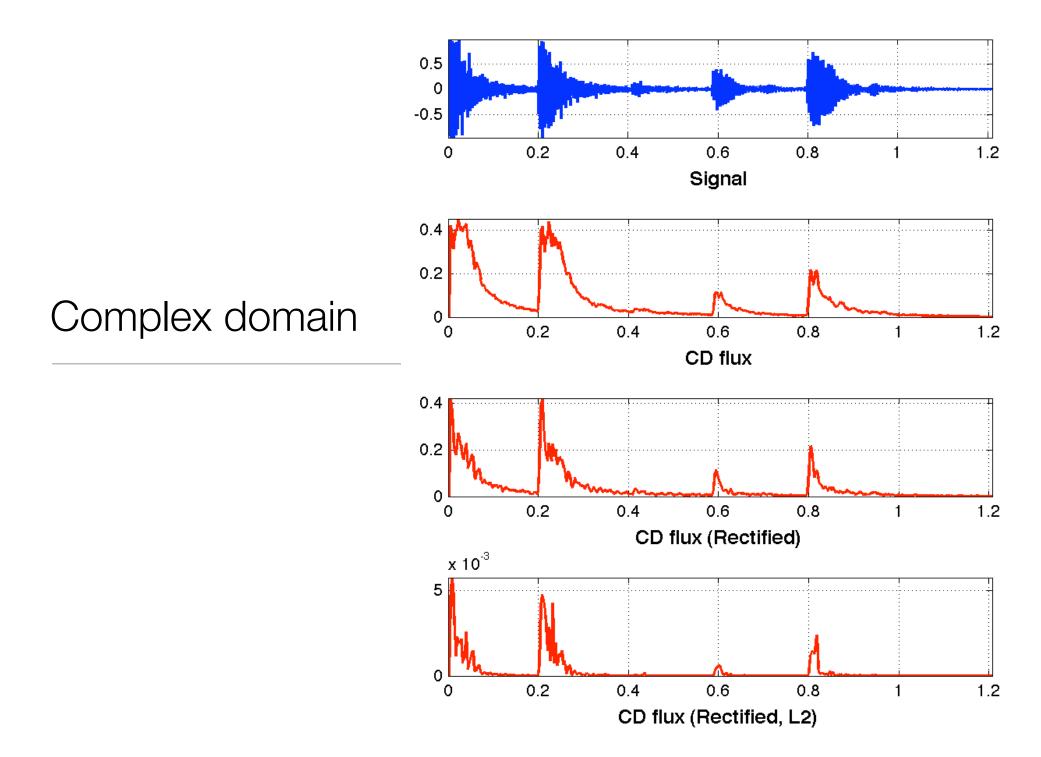
# Complex domain

As before, we can use half-wave rectification to improve the function (Dixon, 06):

$$CD(m) = \frac{2}{N} \sum_{k=0}^{N/2} RCD_k(m)$$

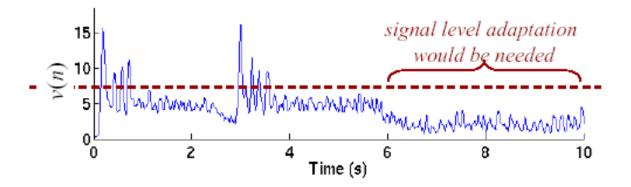
where,

$$RCD_k(m) = \begin{cases} |X_k(m) - \hat{X}_k(m)| & \text{if } |X_k(m)| \ge |X_k(m-1)| \\ 0 & \text{otherwise} \end{cases}$$



## Peak picking

- The function is post-processed to facilitate peak picking:
  - Smoothing -> decrease jaggedness
  - Normalization -> generalization of threshold values
  - Thresholding -> eliminate spurious peaks



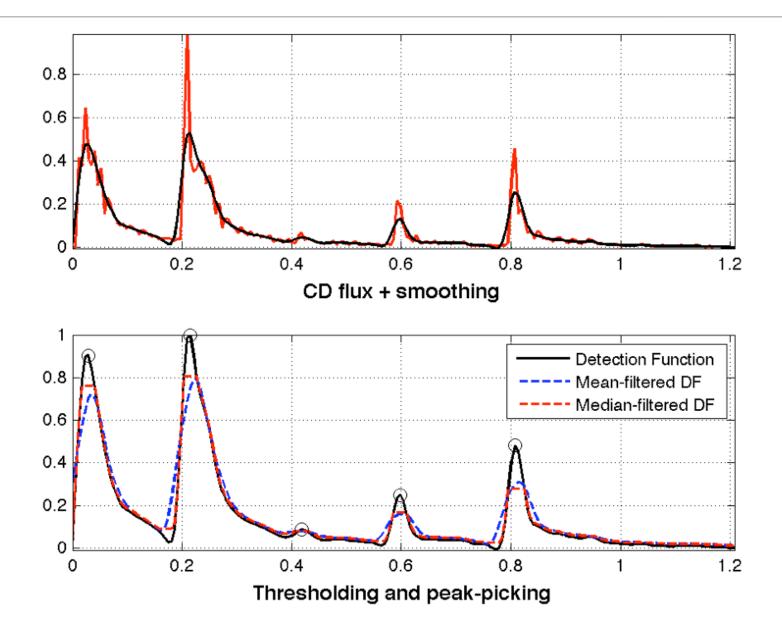
### Peak picking

• Adaptive thresholding is a more robust choice, typically defined as:

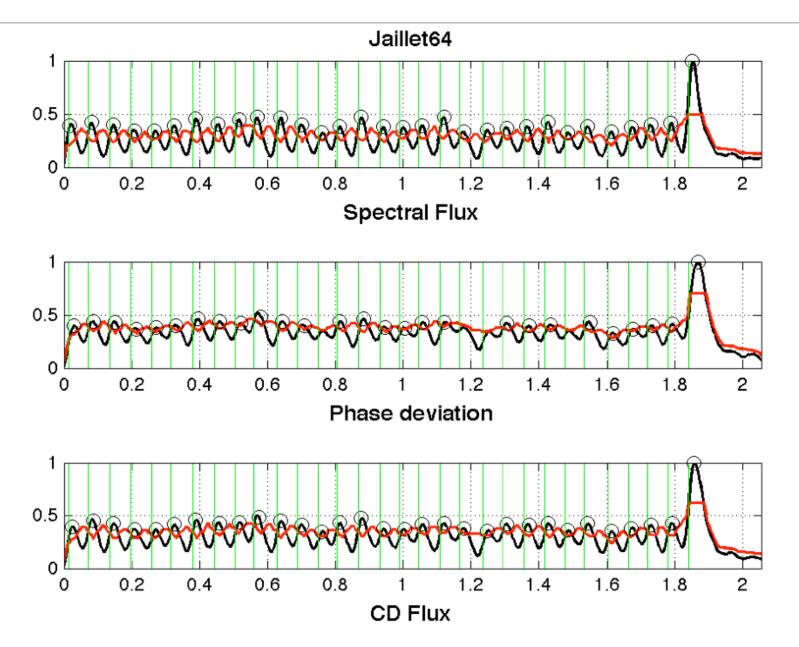
$$\delta(m) = \alpha + f(\bar{m}), \quad m - \beta L \le \bar{m} \le m + L$$

- where *f* is a function, e.g. the local mean or median, of the detection function;
   *β* increases the window length before the peak; and *a* is an offset value
- Peak picking reduces to selecting local maxima above the threshold

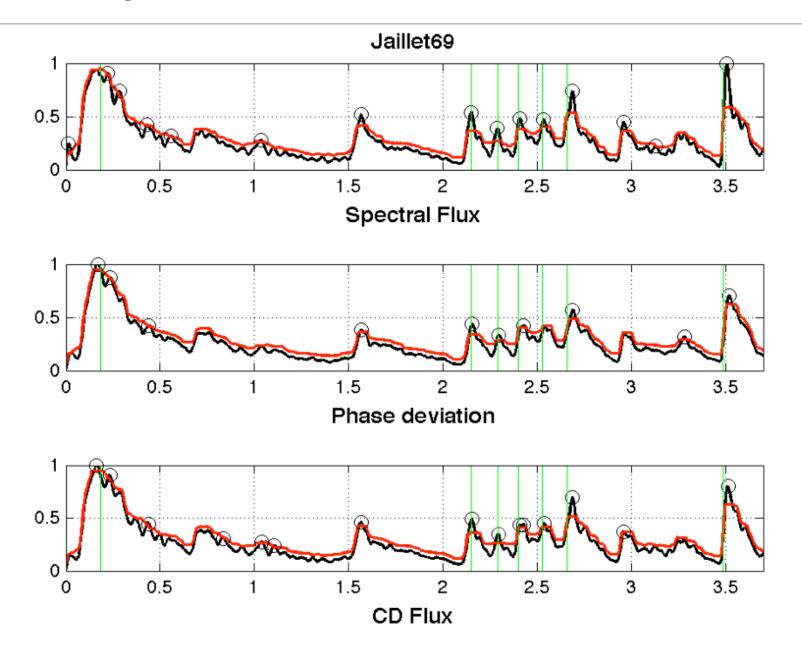
# Peak picking



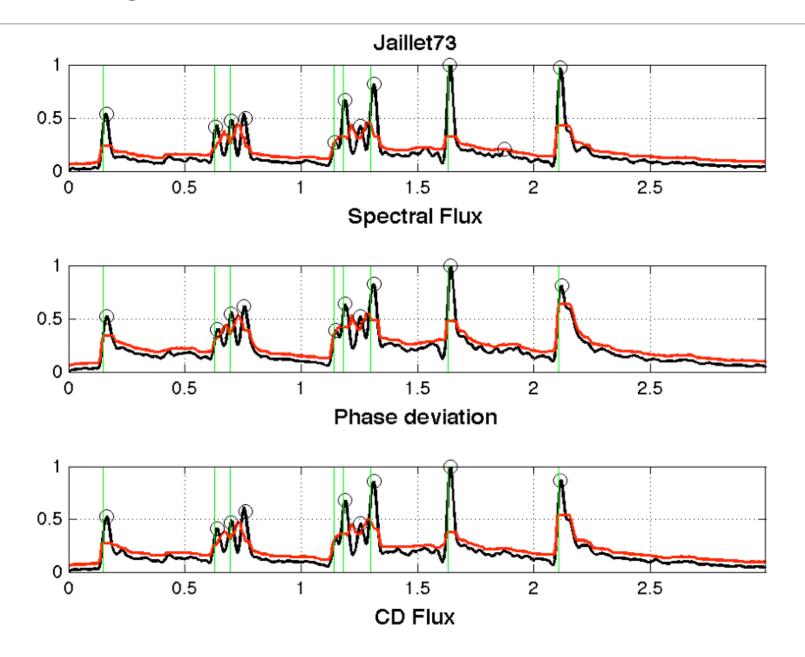
### Comparing detection functions



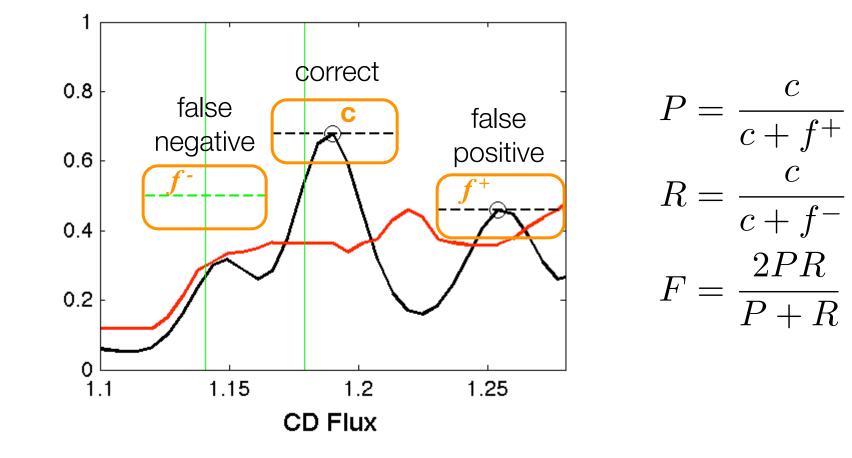
### Comparing detection functions



### Comparing detection functions



#### Benchmarking



### References

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