

Biologically inspired robust onset detection

Leslie Smith, Department of Computing Science and mathematics,
University of Stirling, Stirling FK9 4LA, Scotland, UK

lss@cs.stir.ac.uk

- Why onsets
 - and robust against what?
- How to detect onsets robustly

Why onsets?

- All sound sources start up
 - and often contain internal onsets as well
- Useful grouping cue
 - most sounds are wide-band
 - the energy in different parts of the spectrum generally starts at the same time
 - so onsets that occur at about the same time in different parts of the spectrum can be used to group sound energies together
 - co-occurring onsets are like coincident collinear edge segments
- First sound arrival is from direct path
 - so sound energy at onset arrives direct from sound source
 - usable for source direction finding (both ITD and IID)
- Robust onset detection provides an ecologically useful cue for source grouping and direction finding.

Robustness and onsets

- Detecting a signal onset from a zero base is straightforward
 - but noise can add false onsets
- However:
 - there is normally a noise level from which the signal rises
 - particularly when there are concurrent sound sources
 - direct path onsets are often followed by reflection mediated increases in level
- A robust onset detector would
 - find onsets in noise, and
 - not be confused by reflections

Onset detection techniques

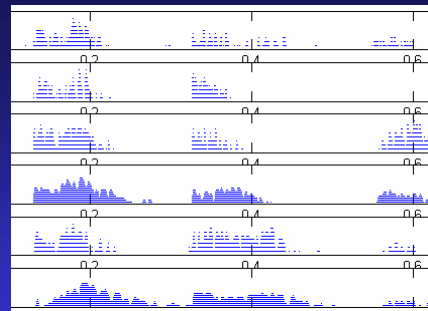
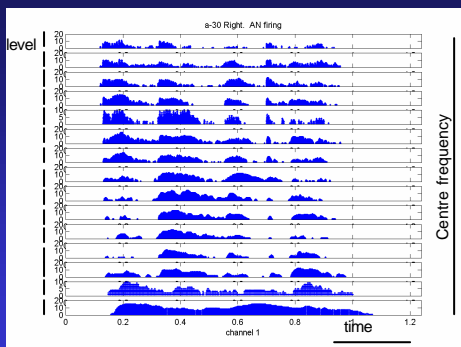
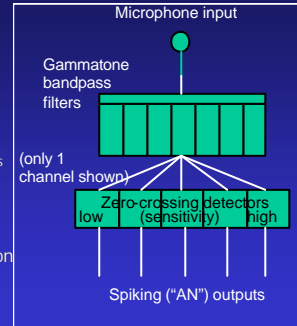
- Simple "first difference techniques"
 - often used in music research
 - Bilmes 93, Goto 95, 96, Schreier 96
 - Optimal filter techniques
 - extension (improvement) to first difference
 - Smith 96, Tait 97, Majoli 2001
 - Expectation based techniques
 - different extension to first difference
 - Huang et al 97
- Problems:
- what exactly constitutes an onset?
 - we know what we mean psychophysically, hard to define in a signal processing sense!
 - real onsets are very variable in form
 - can be short (1-2ms) or longer (up to 50ms)

Biologically inspired techniques

- The early auditory system is sensitive to onsets
 - AN type 1 fibers
 - CN onset cells (of various forms)
- Precedence effect and our sensitivity to onsets illustrates robust capability
- What underlies this?
 - wide dynamic range of the middle ear/inner ear/auditory nerve system
 - spectrum based filtering, allowing onsets in one part of the spectrum not to be swamped
 - characteristics of the neurons of the CN
 - e.g. non-constant leakiness
 - aspects of cell morphology
- We model the wide dynamic range, the filtering, and the high (but not non-constant) leakiness

Modelling the wide dynamic range

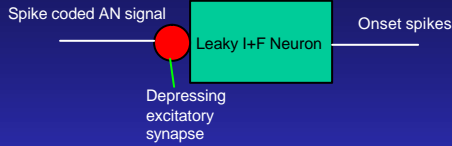
- Each side input starts with a transducer (microphone)
- followed by a gammatone filterbank
 - multichannel, cochlea-like response (static)
- followed by spike generation
 - on positive-going zero-crossings
 - geometric range of sensitivities
 - spike generation predicated on pre-zero-crossing level
 - levels increase geometrically
- result is AN-like representation of sound signal
 - with a form of log compression
 - similar to Ghizta 1986



Although signal is "log-compressed", representation is made up from sequences of spikes (horizontal lines), which can be processed independently

Detecting onsets from these AN signals (1)

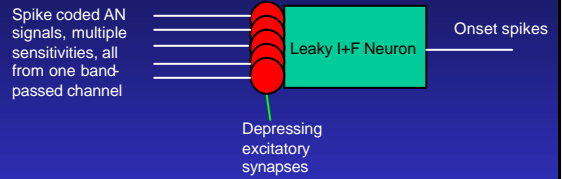
First attempt: 1 onset cell per "AN" signal



Generates multiple onset signals, in phase with AN signal, low latency, but rather sensitive to noise.

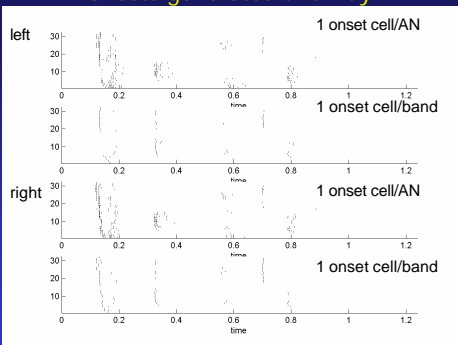
Detecting onsets from these AN signals (2)

Second attempt: 1 onset cell per band-passed channel

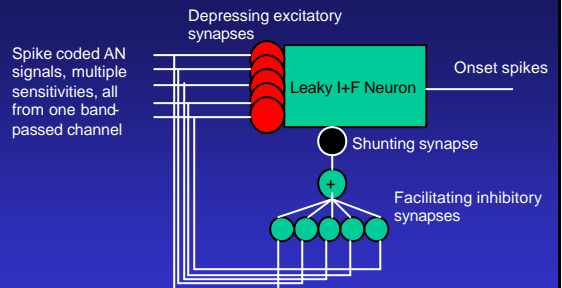


Generates a single onset signal, except at low frequencies, where multiple onsets are sometimes generated. Tends to miss onsets that start from a medium level.

Onsets generated this way



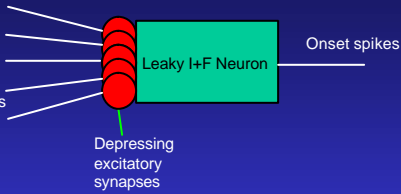
Detecting onsets from these AN signals (3)



Generates a single onset signal even at low frequencies. Still tends to miss onsets that start from a medium level. Complicated as well. Precise time constants in shunting synapse matter.

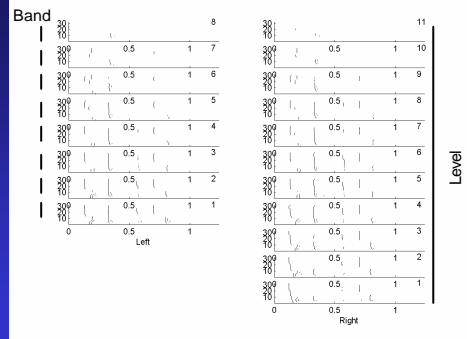
Detecting onsets from these AN signals (4)

Spike coded AN signals, single sensitivity, from a number of adjacent band-passed channels (spread = 2 here)



Generates onset spikes for each band, and for each sensitivity level. Picks up onsets even though they start from a medium level.

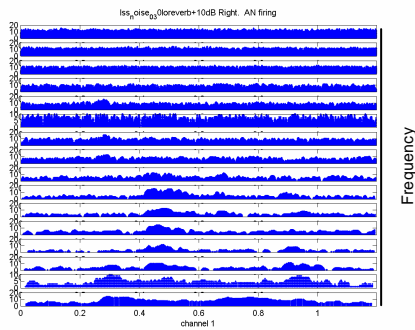
Onsets generated (type 4)



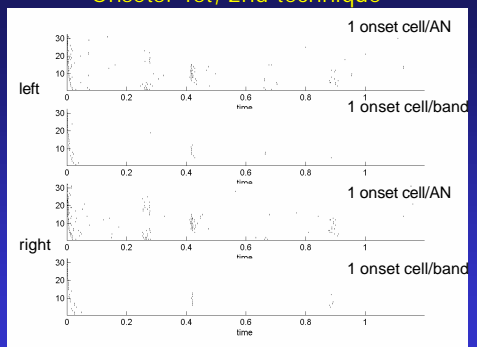
Noisy signals

Speech signal plus white noise.

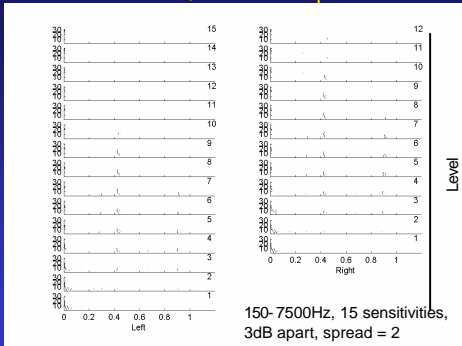
SNR approx 0dB overall



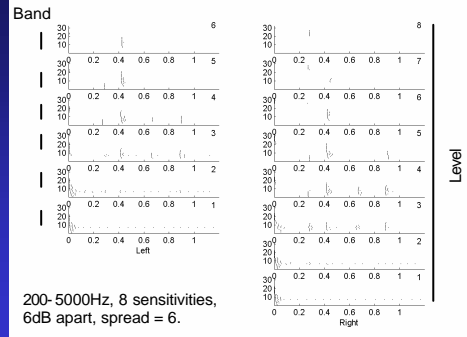
Onsets: 1st, 2nd technique



Onsets, 4th technique

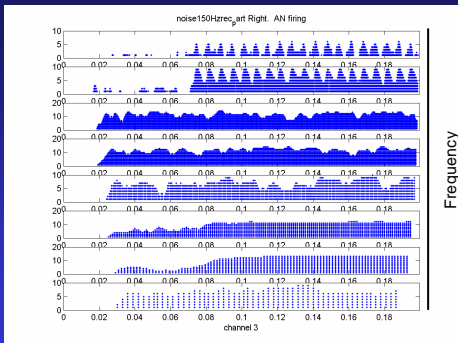


...and again (different parameters)

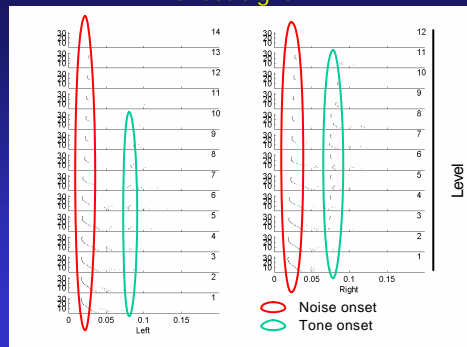


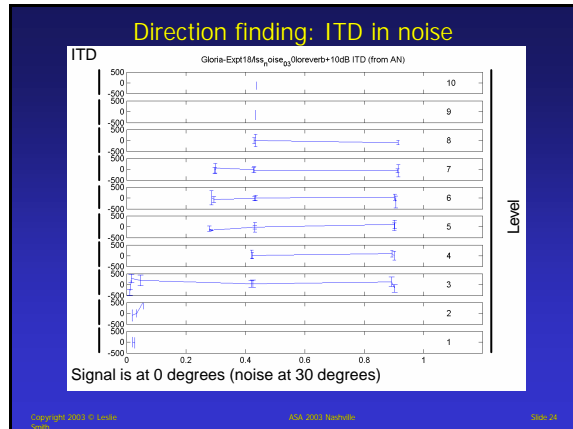
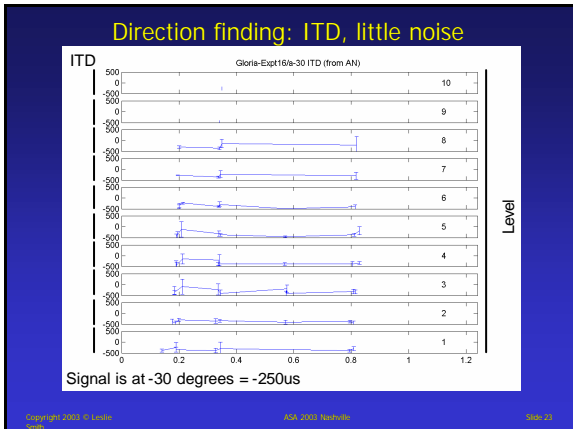
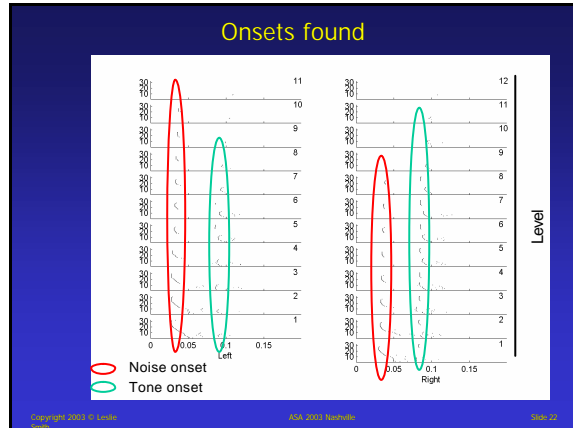
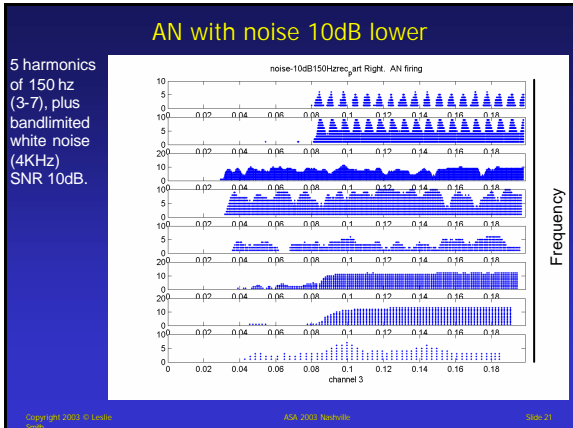
Synthesised sounds: 'AN' signal

5 harmonics
of 150 hz
(3-7), plus
bandlimited
white noise
(4KHz)
SNR 0dB.

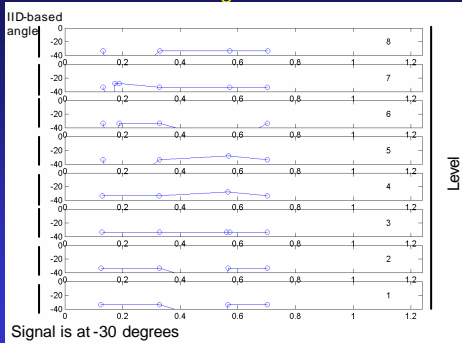


Onset signal

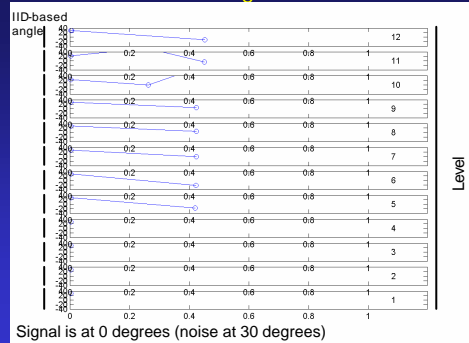




Direction finding: IID, little noise



Direction finding: IID in noise



Conclusions

- We have developed a biologically -inspired onset detection technique which is
 - able to deal with wide dynamic range
 - able to cope with onsets starting from a non-zero level
- This has been used to cluster onsets
 - and hence to find the directions of concurrent sound sources
- Issues: on onsets
 - Would probabilistic AN spike generation widen the dynamic range?
 - Would incorporating inhibition into type 4 onset detection assist?
 - Need to characterise the precise capabilities of this technique fully.
- On onset-feature based streaming
 - how to string grouped features together over time
 - use direction of sound?
 - Use characteristics of the features themselves: level, signal energy structure, post-onset envelope