
EECS 225D

Audio Signal Processing in Humans and Machines

Lecture 15 – Pitch

2012-3-12

Professor Nelson Morgan
today's lecture by **John Lazzaro**

www.icsi.berkeley.edu/eecs225d/



Today's lecture: Pitch

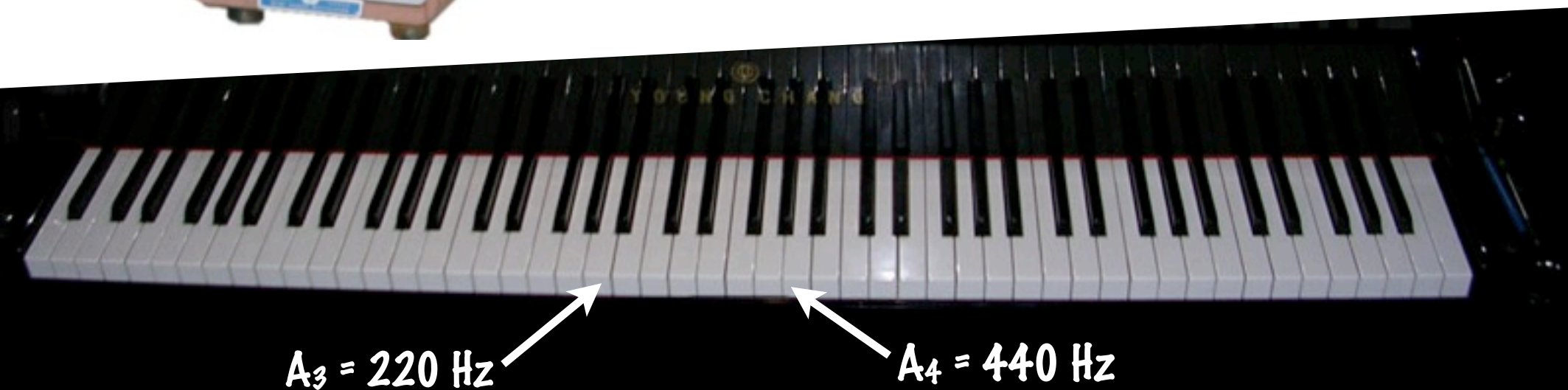
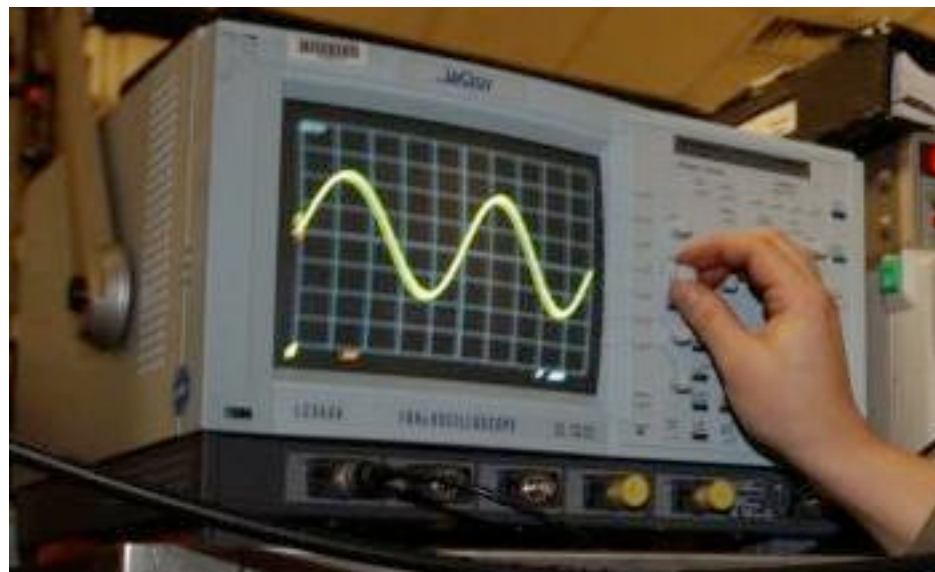
✱ **Basic concepts in pitch detection**

✱ **Practical issues in pitch tracking**

✱ **The quiz ...**

Musical pitch: an experimental definition

Pitch (unit: Hz). The frequency of a sine wave whose pitch is heard to be the same as a played note.



Timbre and Pitch

Why are the timbres different?

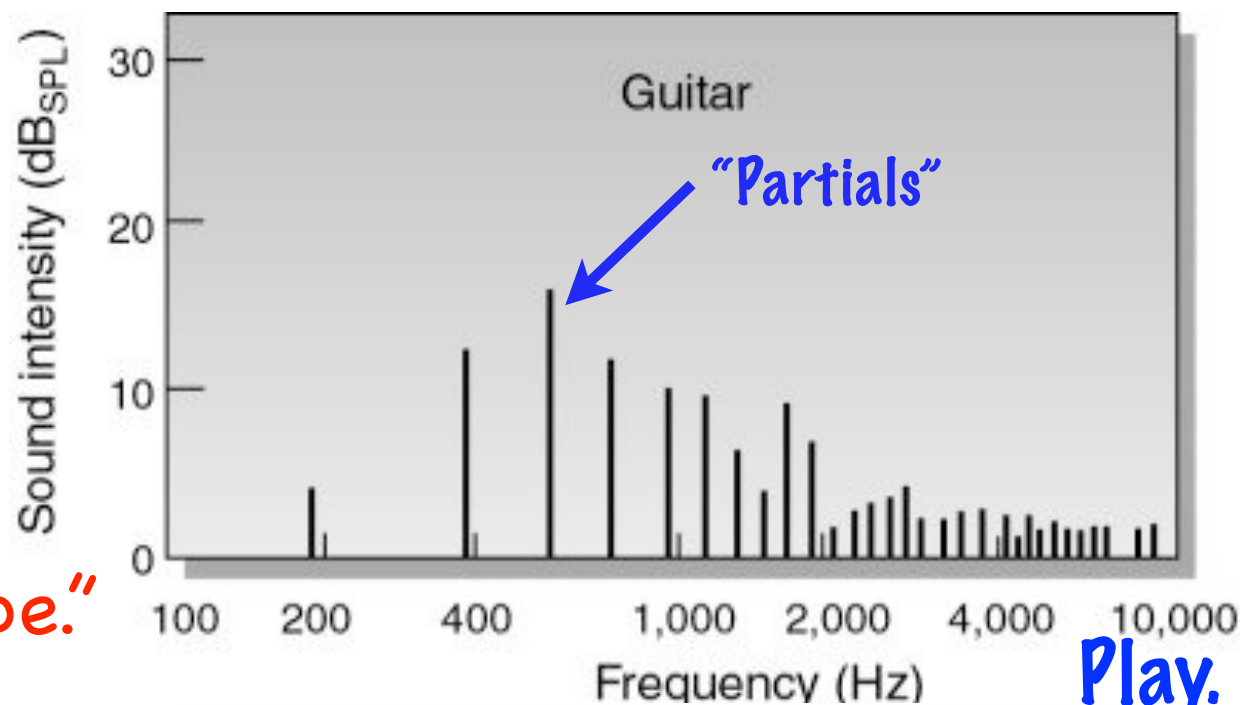
Contributing factor: Partial heights differ, and evolve differently over time. "Spectral Shape."

Why are both sounds pitched?

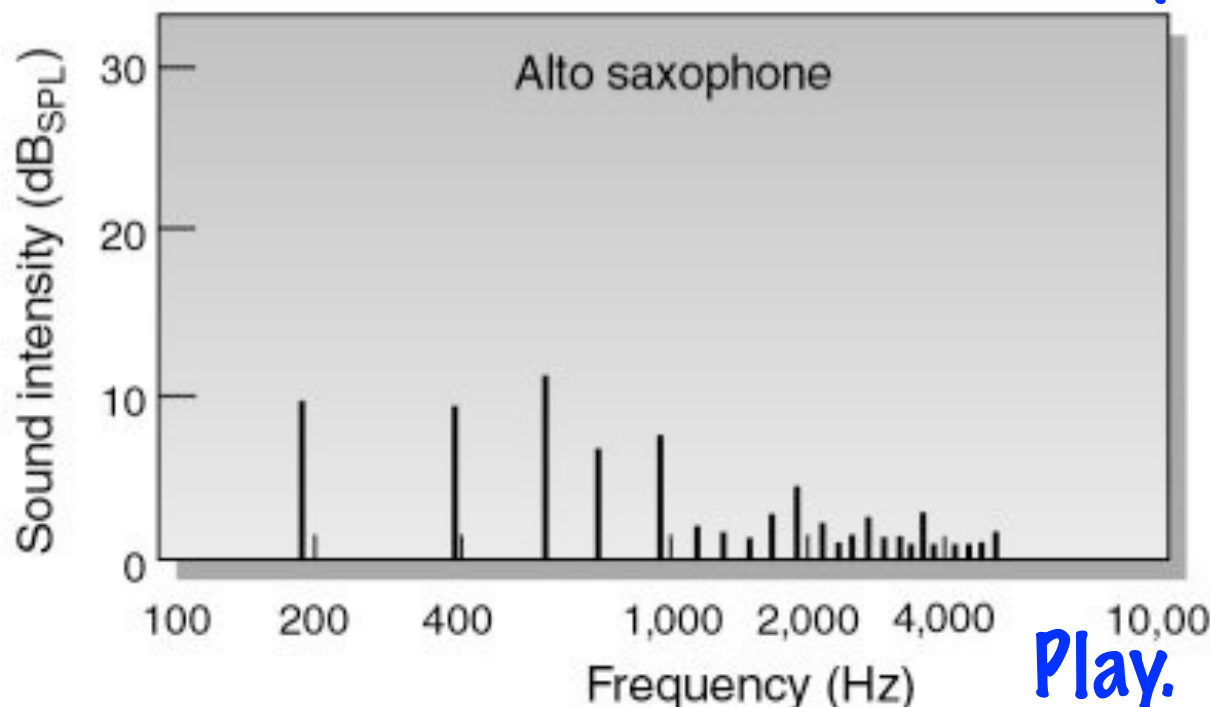
Why is the pitch the same?

Frequency placement of partials share a common structure.

Same pitch, different timbre

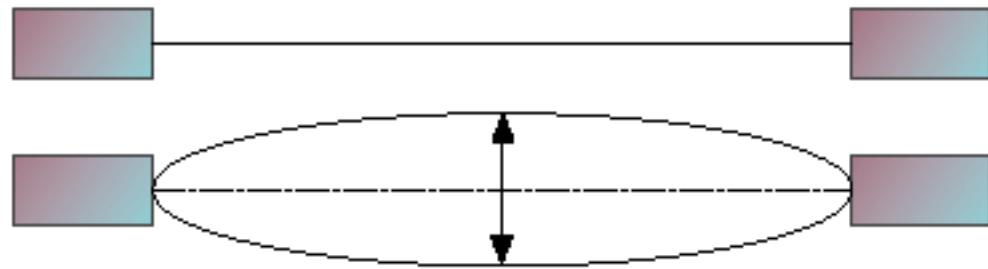
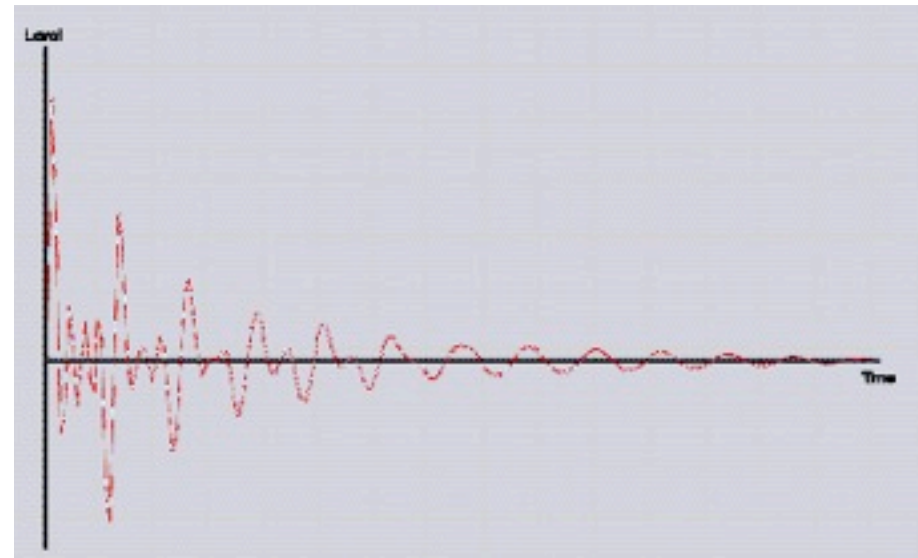


Play.

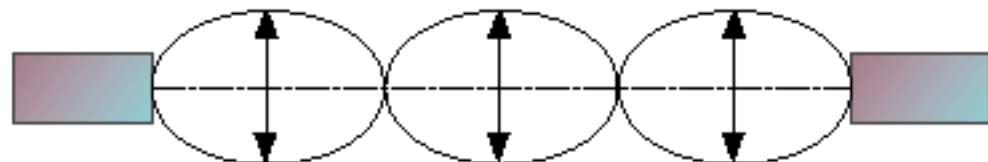
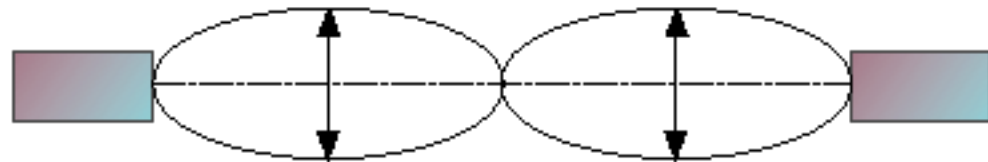
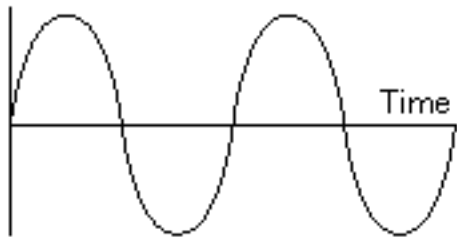


Play.

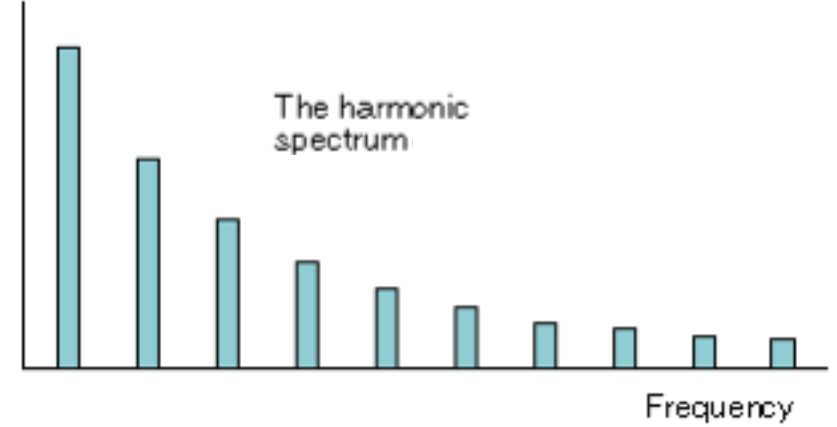
EXAMPLE BY KENNETH STEELE, APPALACHIAN STATE .

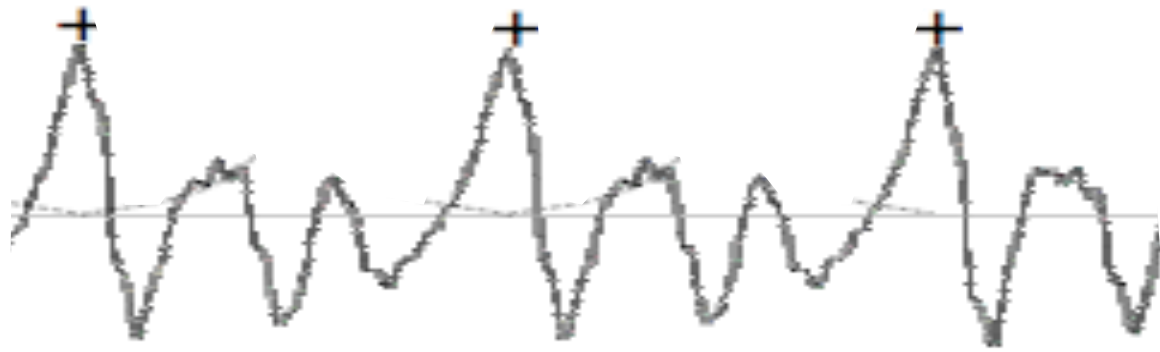


Displacement

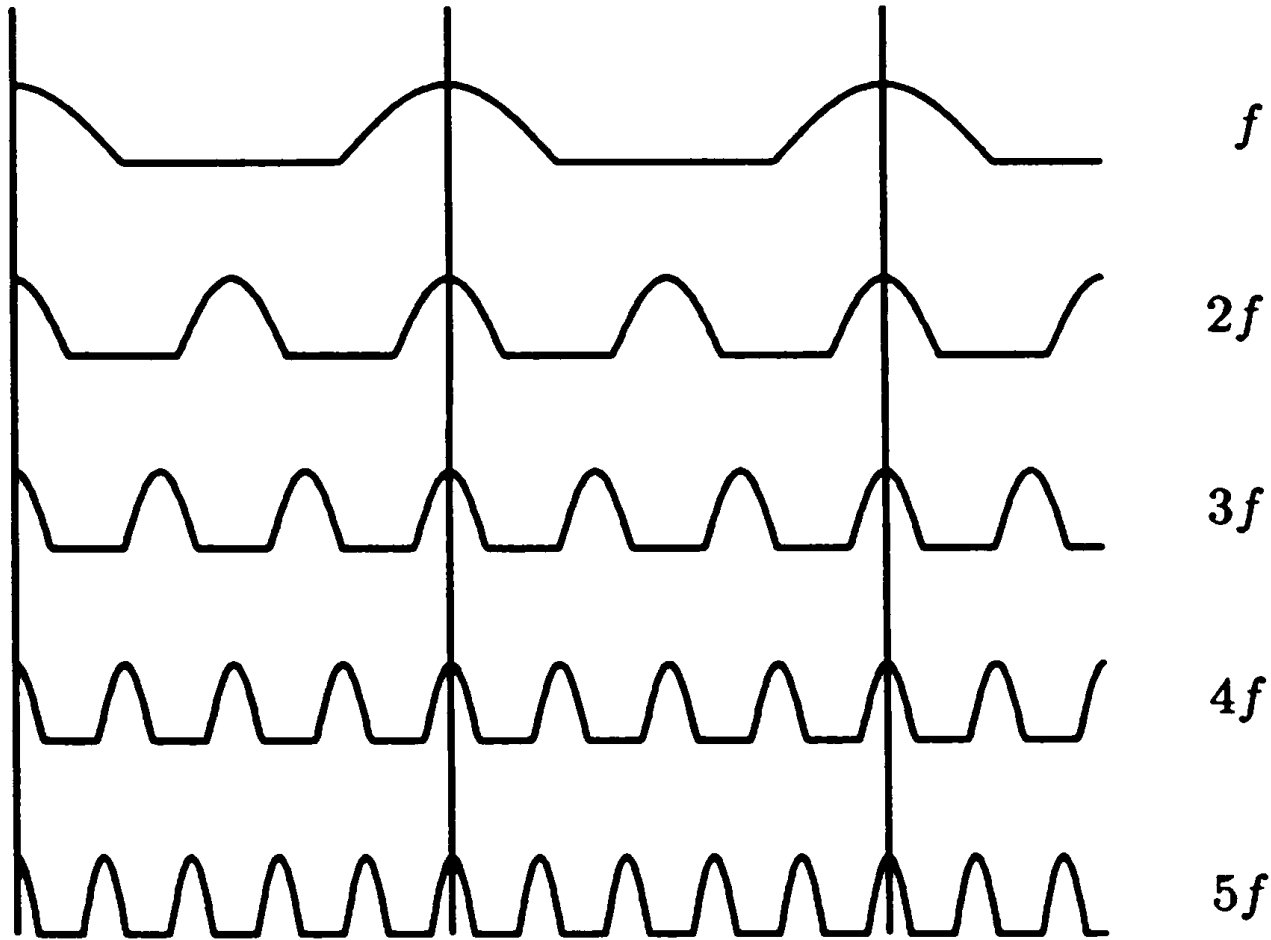


Amplitude





Summed waveform repeats at pitch frequency.

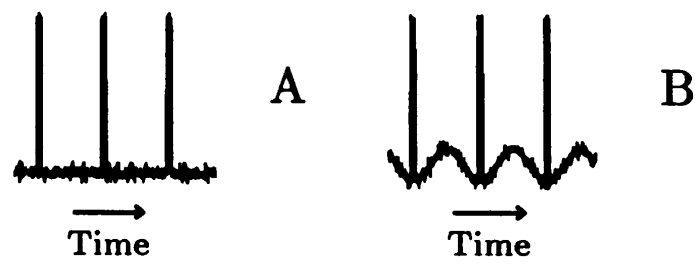


Frequencies of partials are integer multiples of an underlying fundamental.

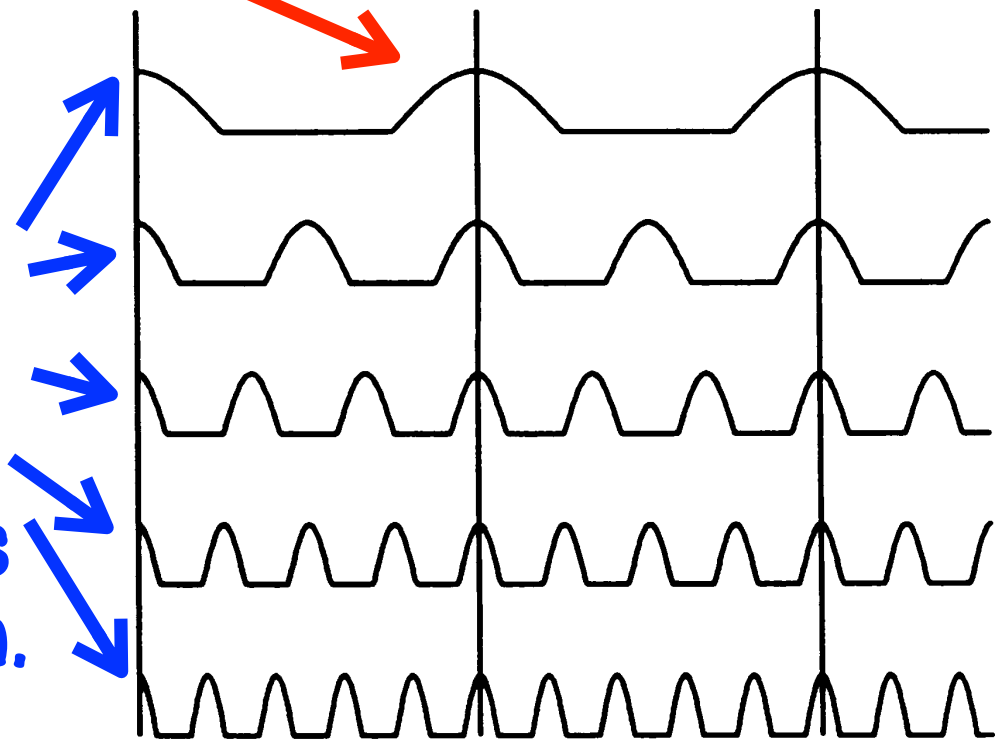


Pitch Period = $1 / (\text{Pitch Frequency})$

Caveats: First partial not necessary to detect pitch - A and B → are heard with same pitch.



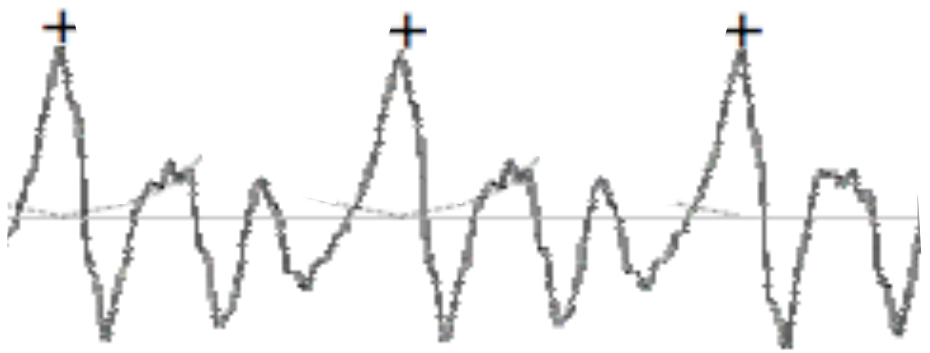
Relative phases of partials need not be aligned - any phase relation yields a strong pitch.



f
 $2f$
 $3f$
 $4f$
 $5f$

Sounds whose partials are not quite integer-related still yield a sense of pitch -

Thus ... repeating shape may be subtle to detect directly.

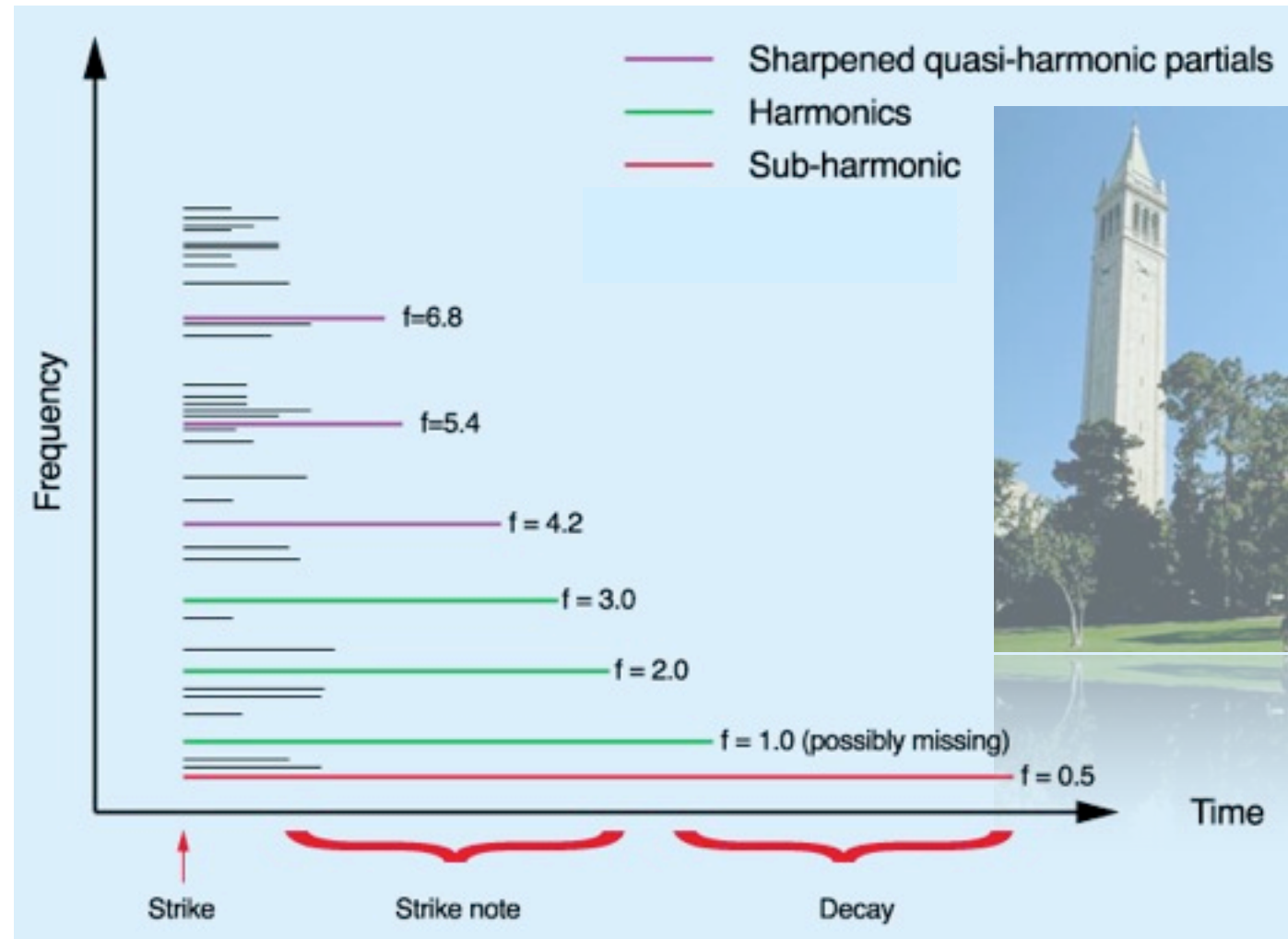


Bells

Lowest partials are exact integers, but higher partials are quasi-harmonic (4.2, 5.4, 6.8).

We still hear the bells as having a definite pitch.

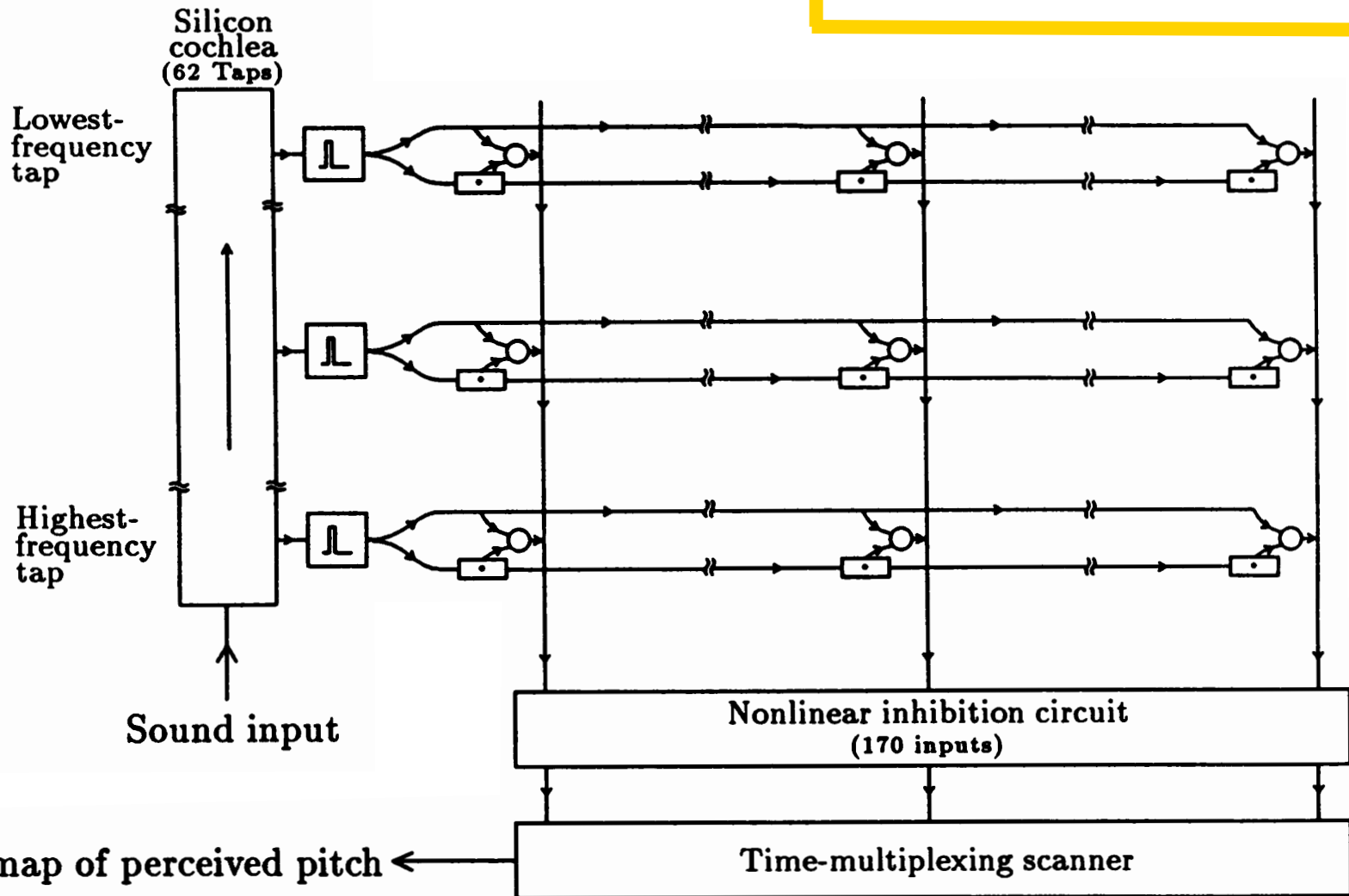
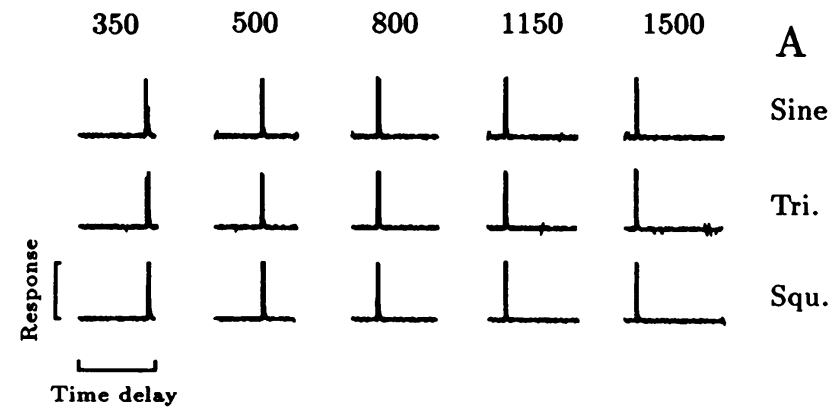
Play.



But bell "chords" often sound **atonal**.

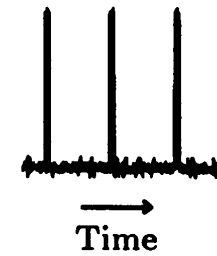
Licklider Pitch Model

Autocorrelate filtered versions of the audio waveform.

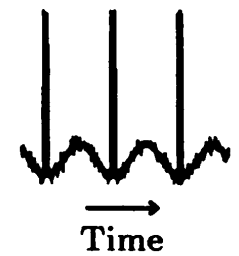


Recall:

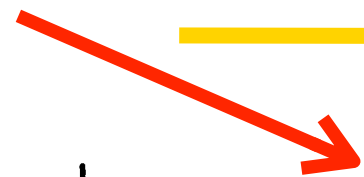
First partial not necessary to detect pitch - A and B → are heard with same pitch.



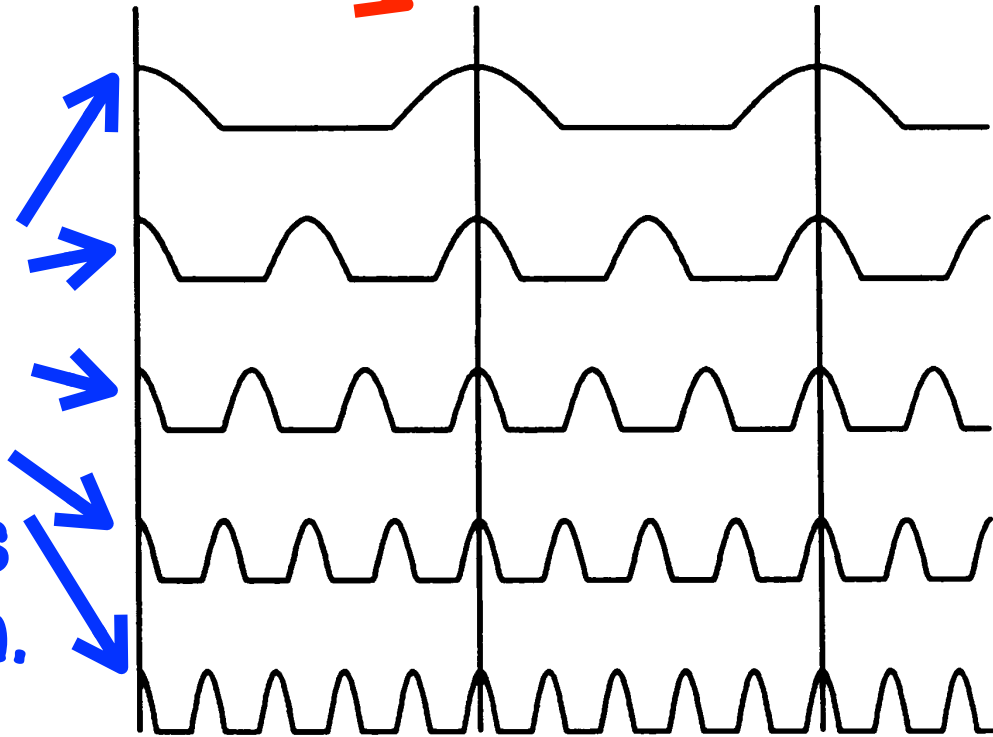
A



B



Relative phases of partials need not be aligned - any phase relation yields a strong pitch.



f
 $2f$
 $3f$
 $4f$
 $5f$

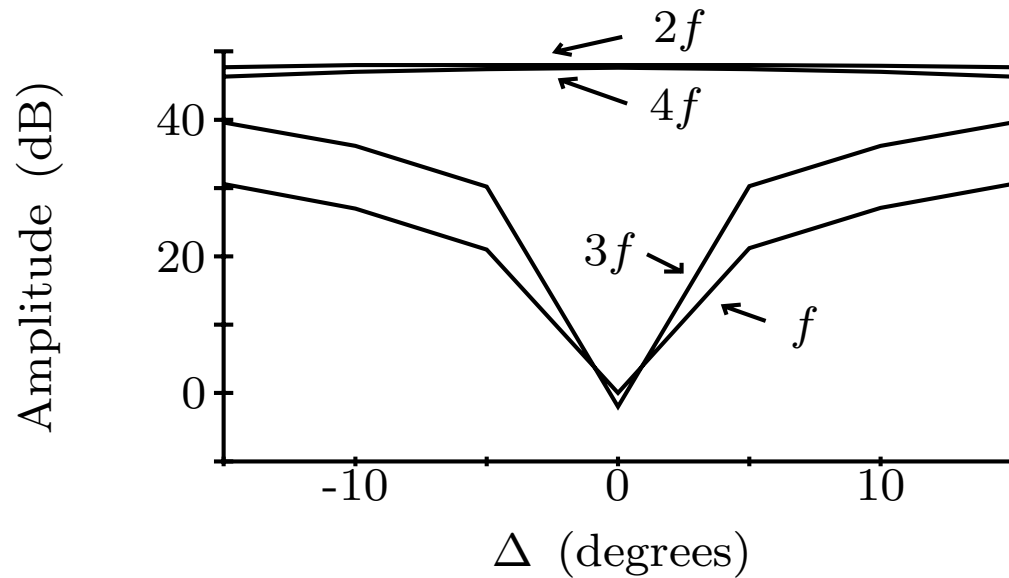
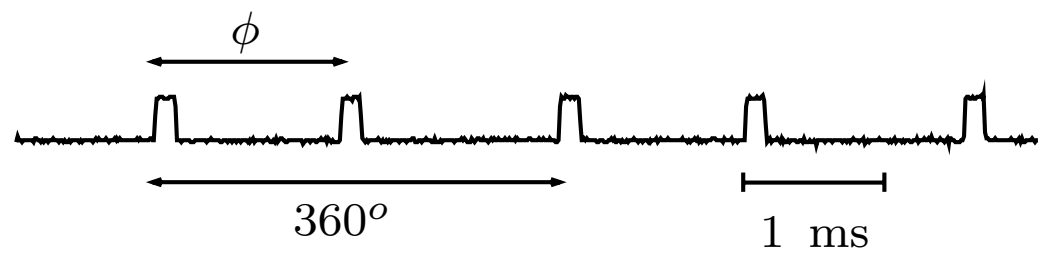
Sounds whose partials are not quite integer-related still yield a sense of pitch -

Autocorrelation model addresses all of these issues.

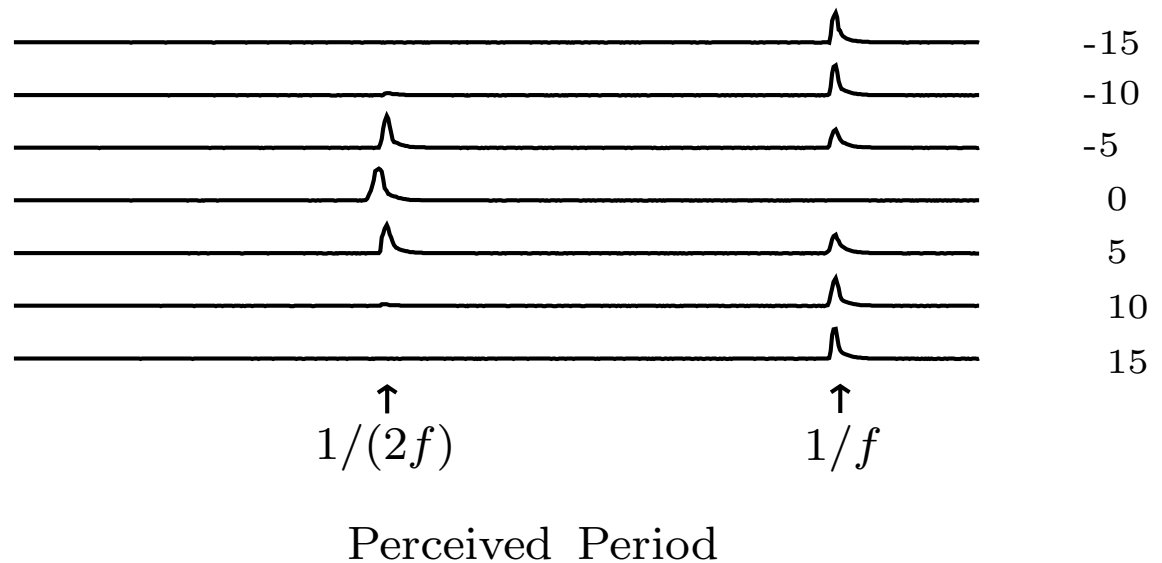
Seebeck's Siren

When ϕ is offset slightly from 180° , the pitch **drops** one octave. Why? The repetition period of the waveform **doubles**.

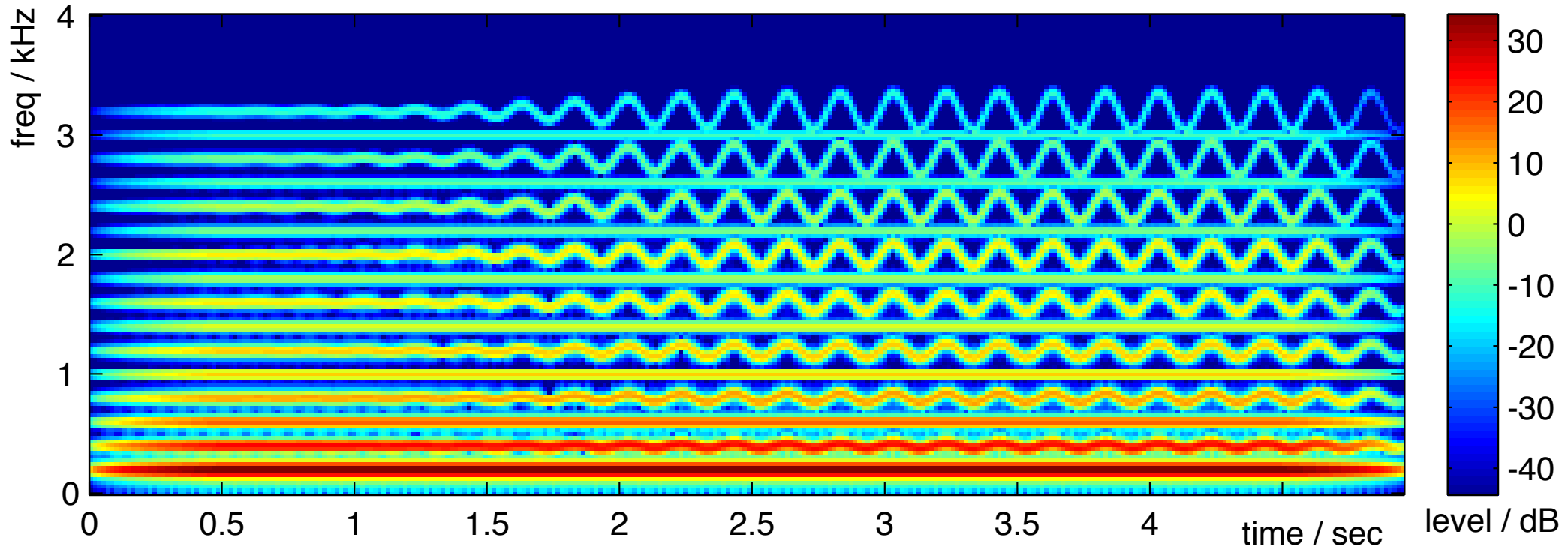
Autocorrelation "hears" this signal the **same way humans do**.



Play.



Related: Reynolds/McAdams Oboe



Adding vibrato to even partials
"creates" a second sound
whose pitch is raised by one
octave (spectrum by Dan Ellis).

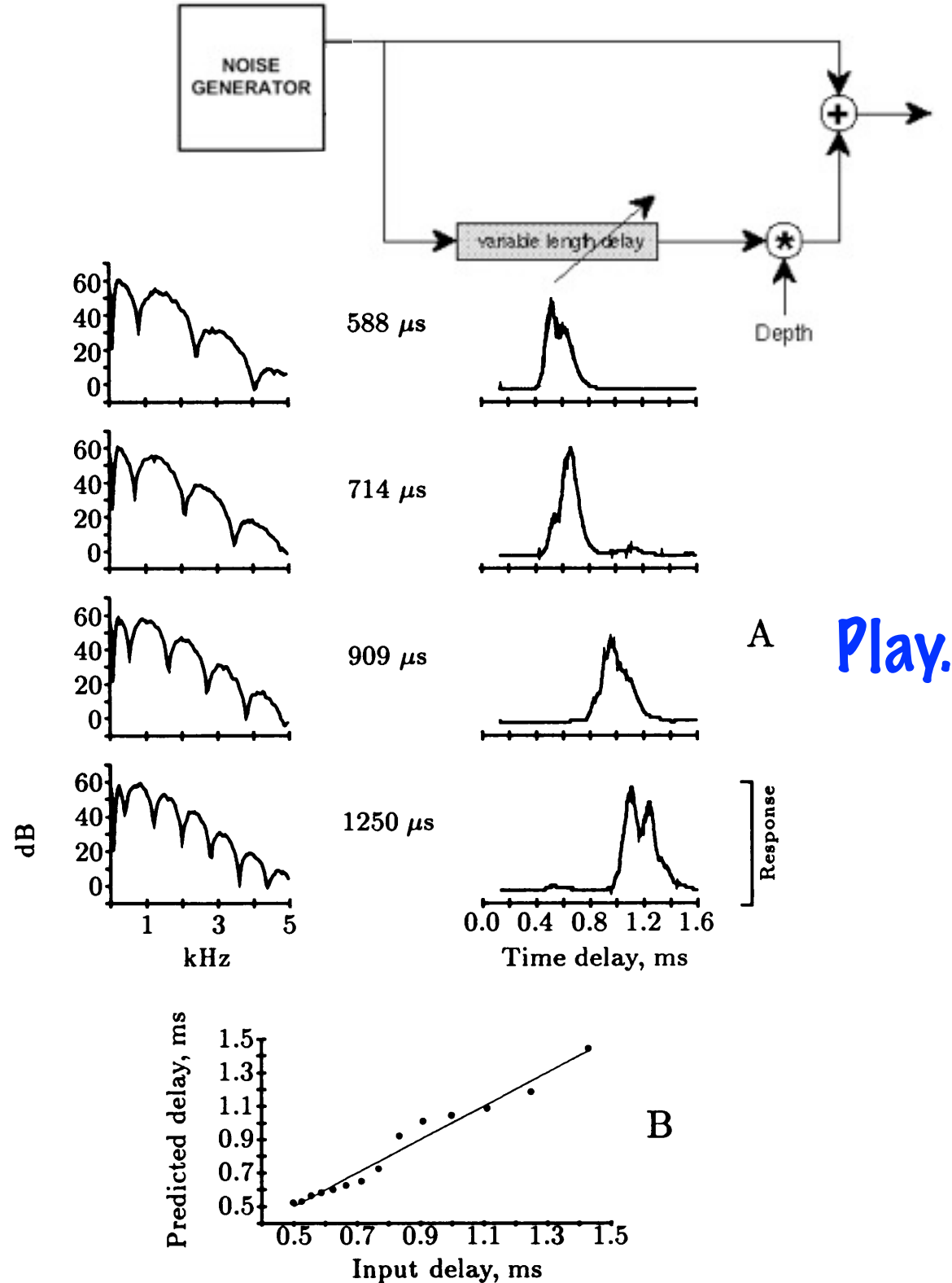
Play

Comb-Filtered Noise

We hear a weak pitch in **comb-filtered noise**, corresponding to the period of the comb delay.

This is unusual because the **waveform** of each **period** is **unique**.

Autocorrelation **"hears"** the same pitch humans perceive.



Play.

Amplitude Modulation

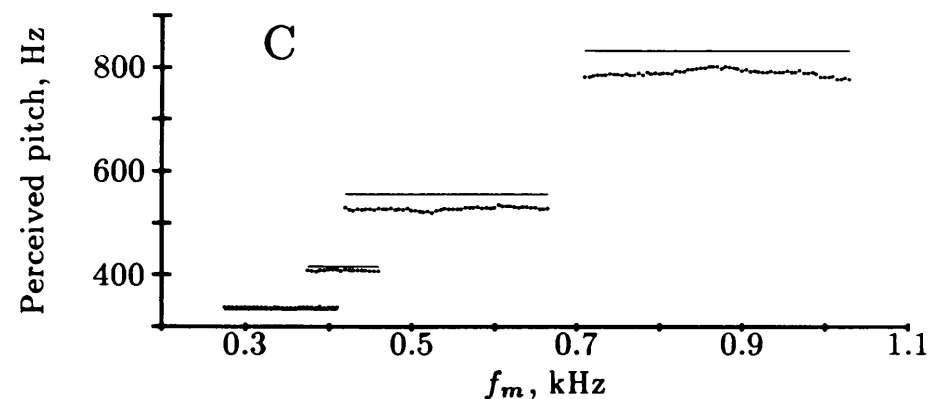
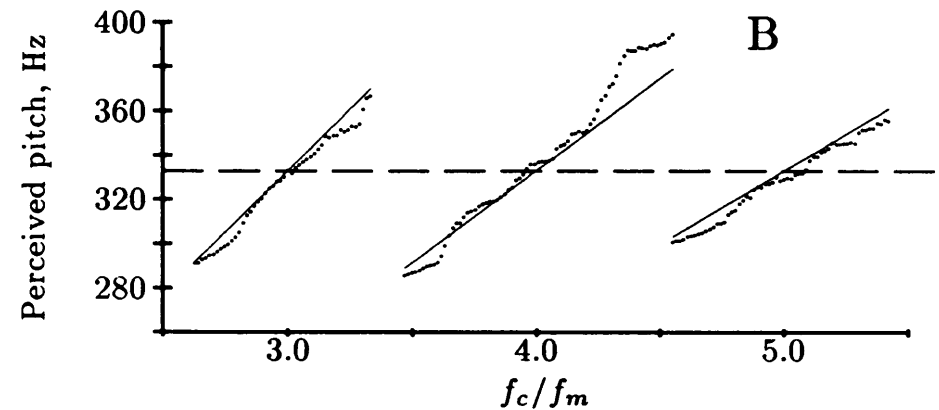
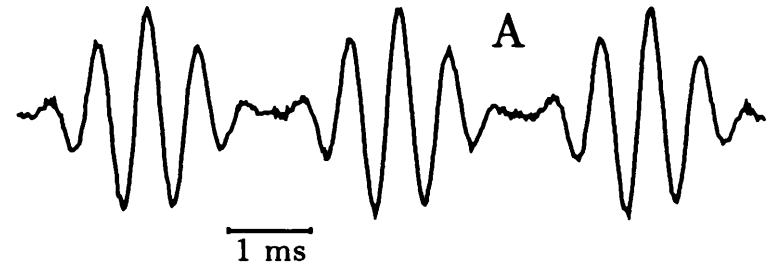
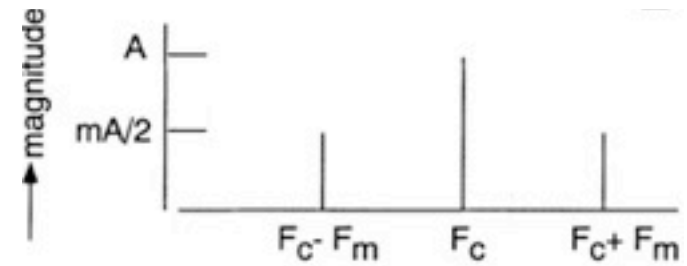
This sound has components at

$F_c - F_m$, F_c , and $F_c + F_m$.

To a first order, the pitch we hear is F_m (the repetition frequency).

Autocorrelation matches this **first-order** result.

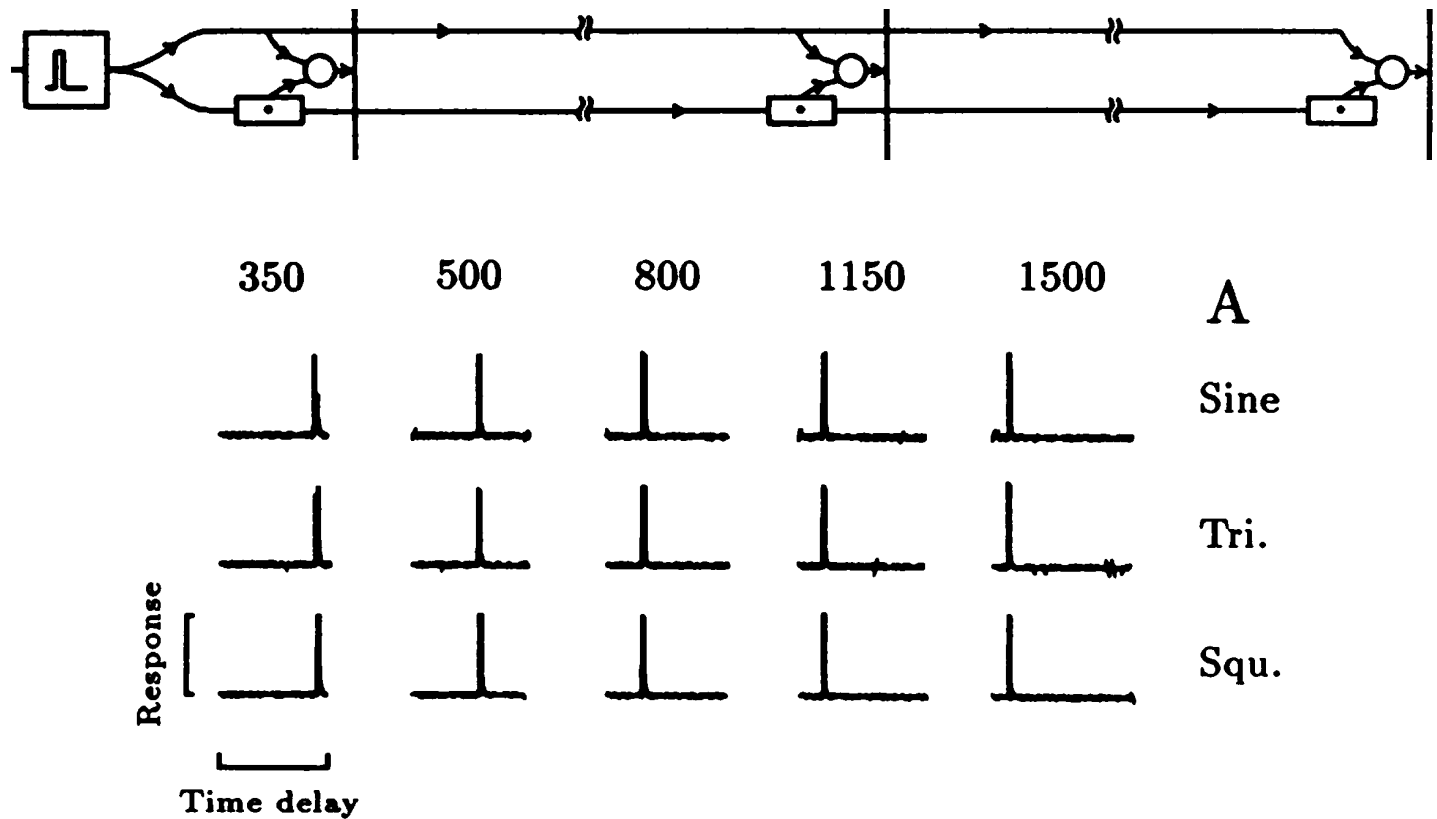
But fails on **second-order** phenomena.
(slope of pitch for small changes in F_c , F_m).



Today's lecture: Pitch

- * **Basic concepts in pitch detection**
- * **Practical issues in pitch tracking**
- * **The quiz ...**

Modern engineering pitch trackers are based on **multi-tap autocorrelation** (or similar operators).



Algorithm enhancements mostly take the form of **pre-processing** the audio input or **post-processing** the auto-correlation tap outputs, to better handle the **"difficult"** pitch signals we see in **real-world** engineering applications.

Real-World Issues



Dynamic range of pitch

(a)



Pitch variations in time

(b)

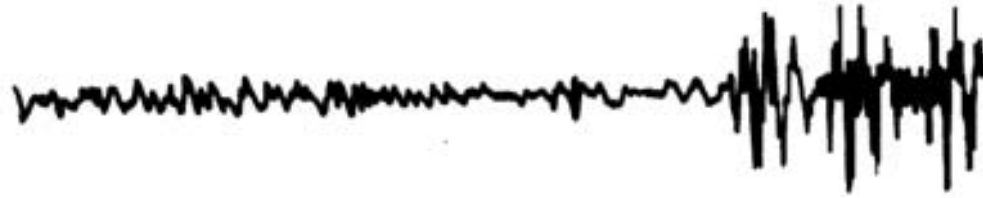


Vocal tract variations in time

(c)

10 ms

Real-World Issues



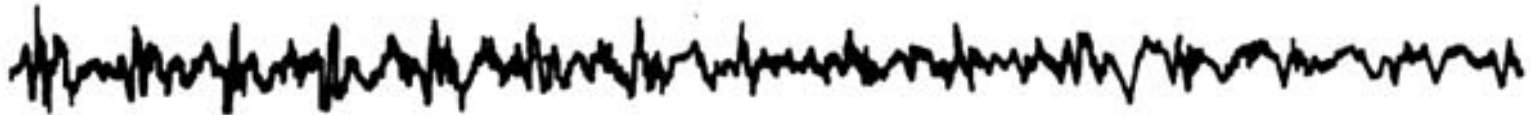
Voiced-unvoiced transition

(*d*)



Telephone speech

(*e*)

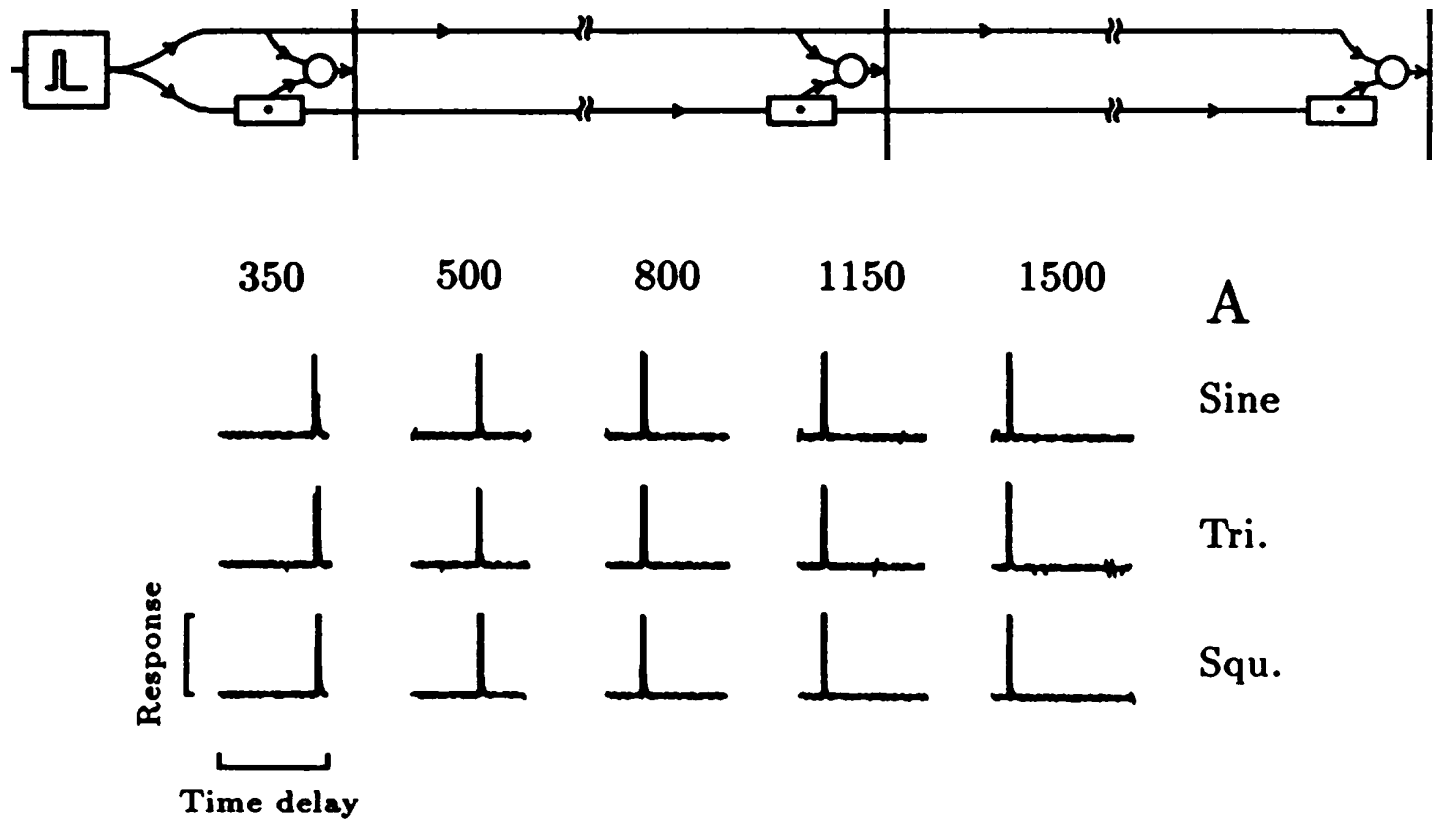


Acoustic noise background

(*f*)

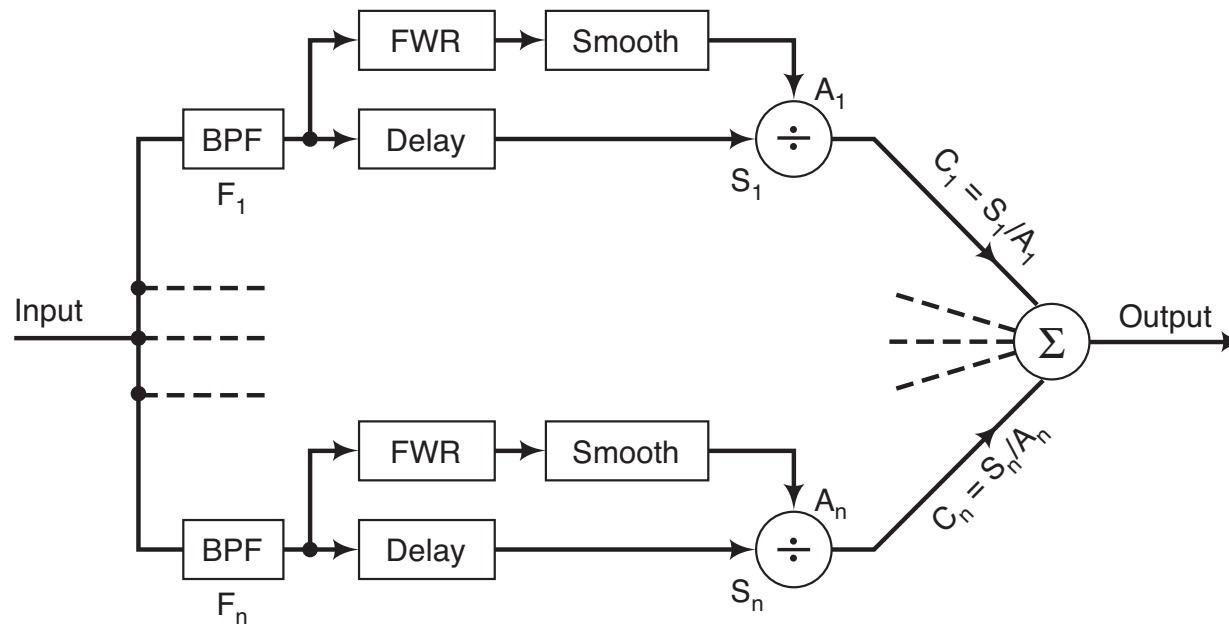
10 ms

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Pre-Processing : Spectral Flattening



Pre-Processing : Low-pass Filtering

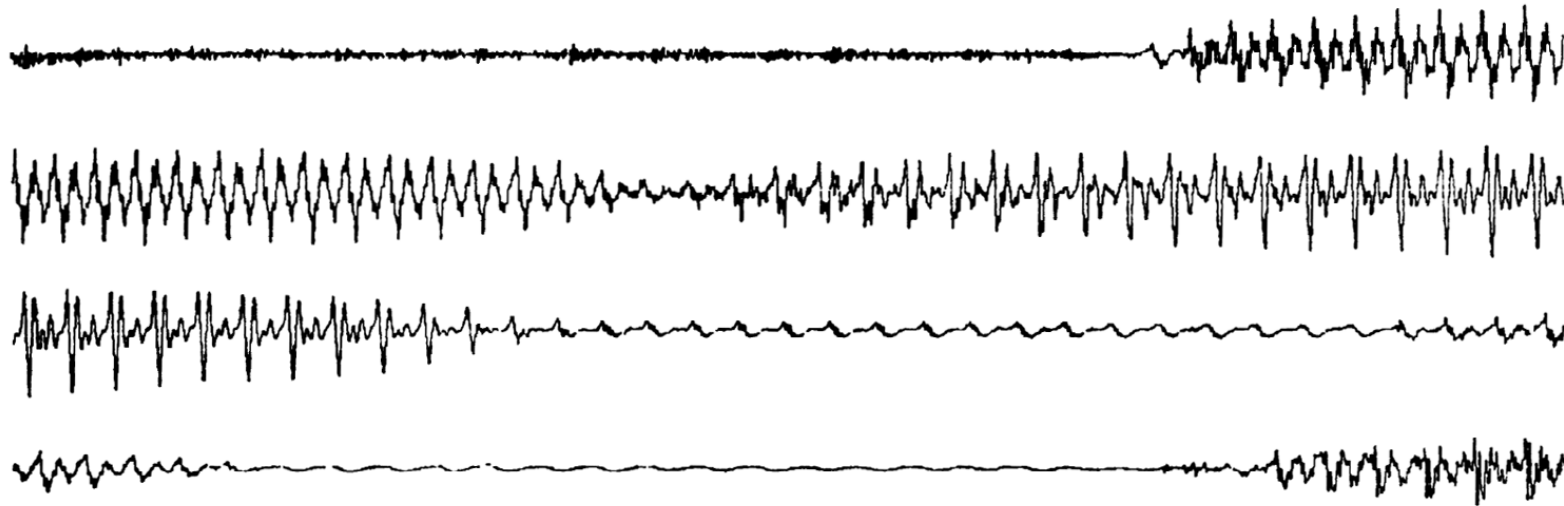


FIGURE 31.5 Full-band speech signal.

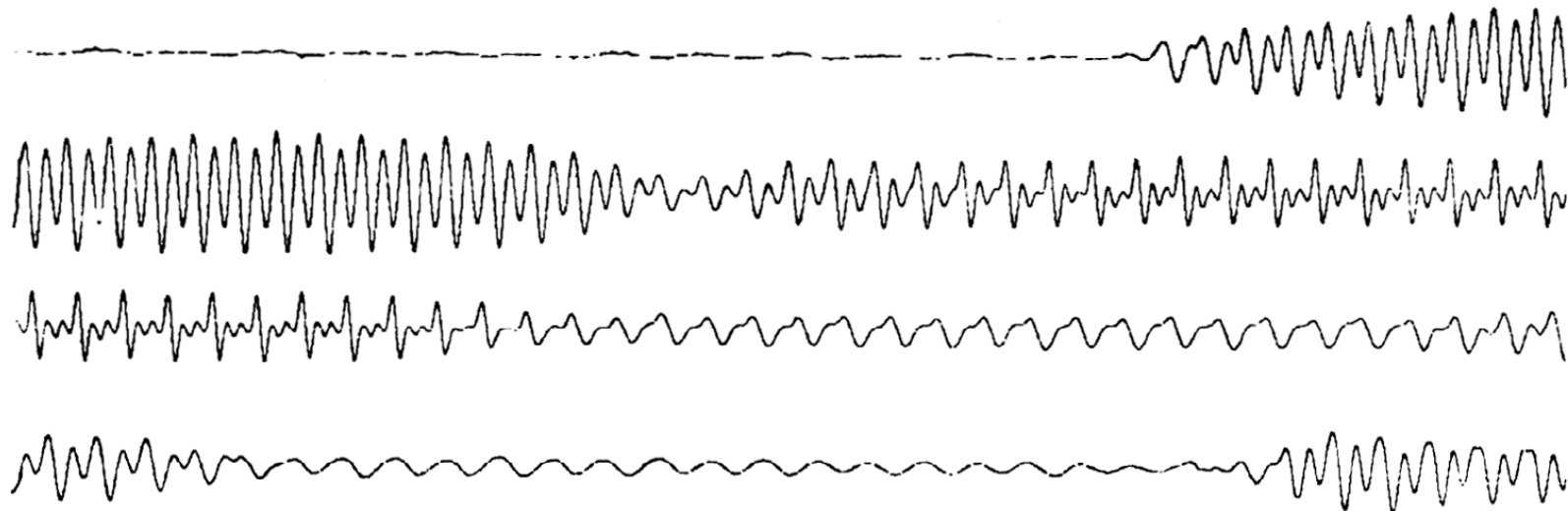
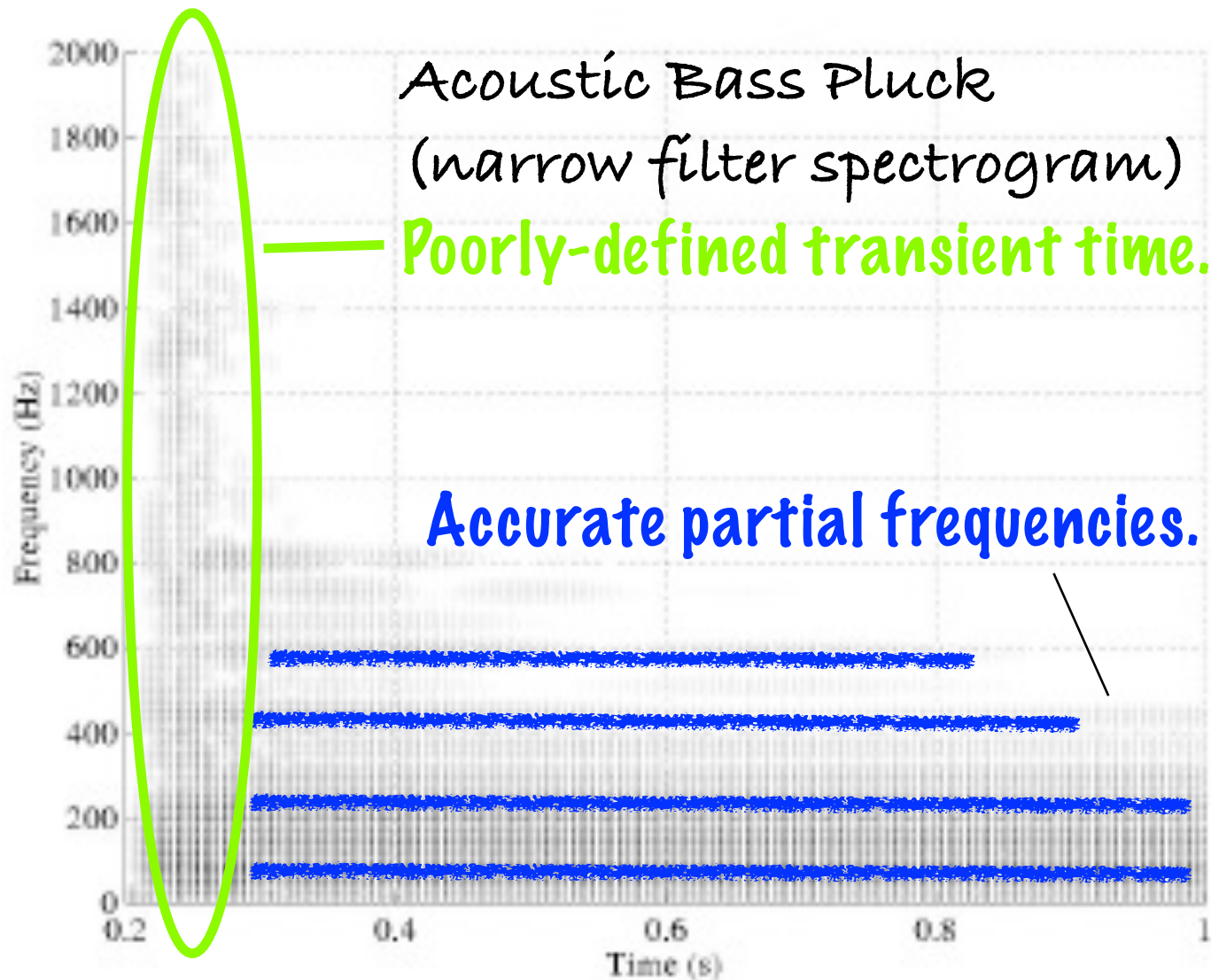


FIGURE 31.6 Low-pass filtered speech signal.

Time/Frequency Tradeoffs

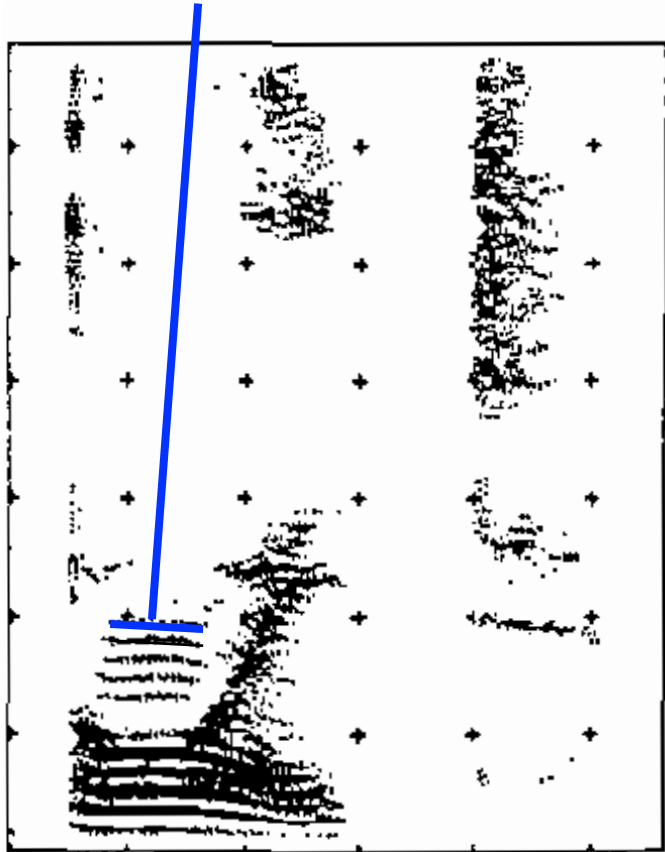
Narrow linear filters are slow
Fast linear filters are wide



Filters: Wide vs Narrow

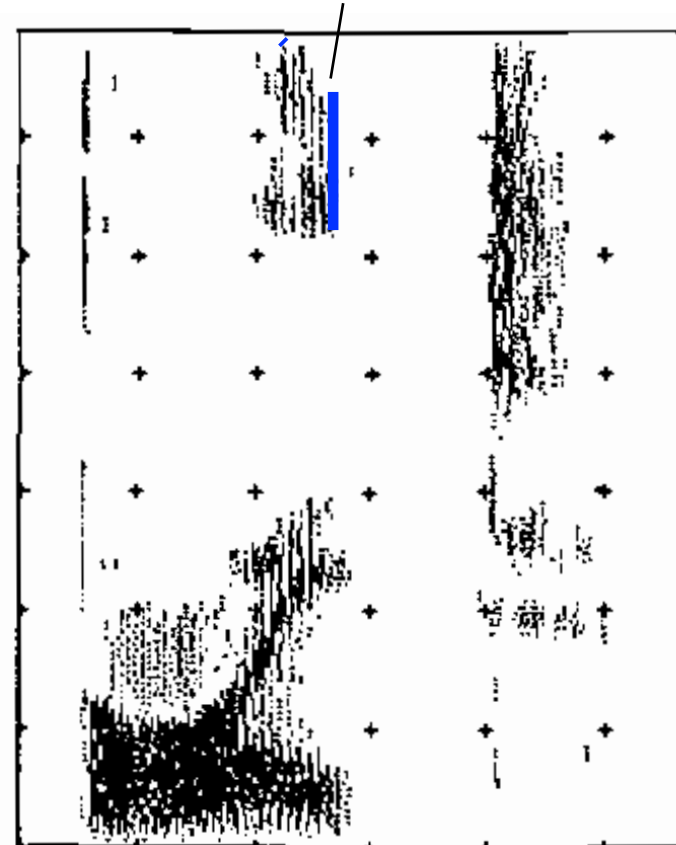
Spoken word "boyt"
processed by **two**
filterbanks.

Accurate pitch harmonics.



Narrow filters

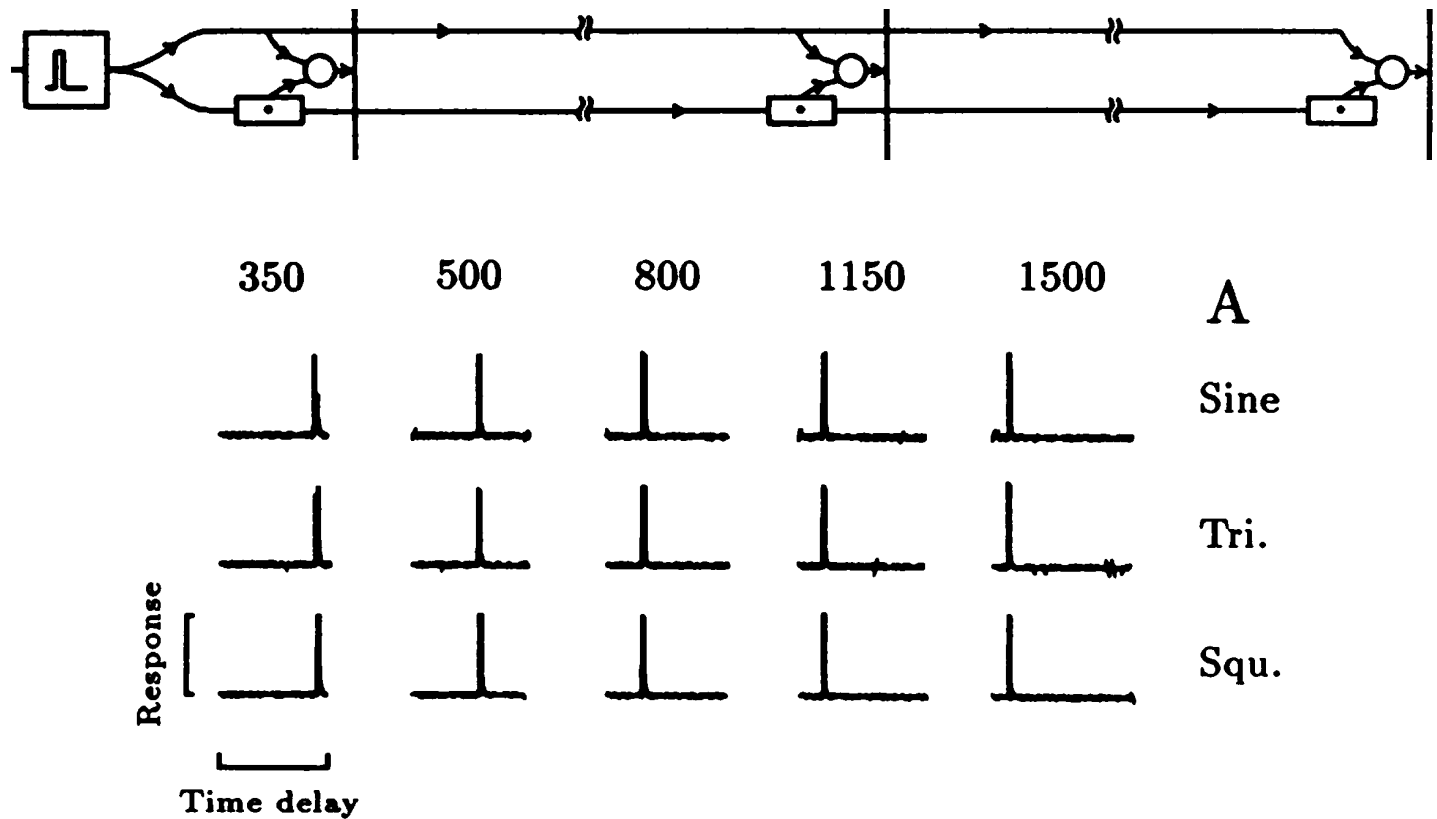
Good glottal pulse timing.



Fast filters

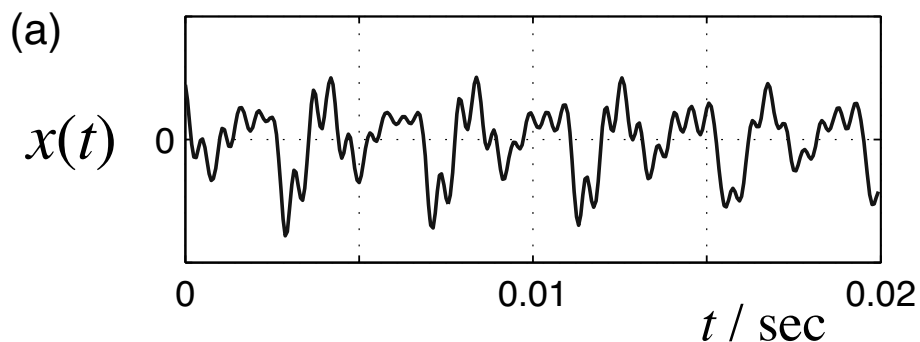
Data from Hong Leung and Victor Zue

Modern engineering pitch trackers are based on **multi-tap autocorrelation** (or similar operators).

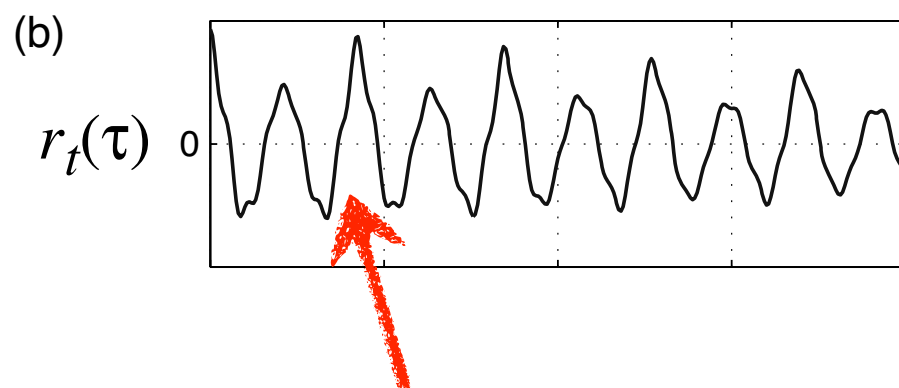


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Post-Processing : Yin Normalization

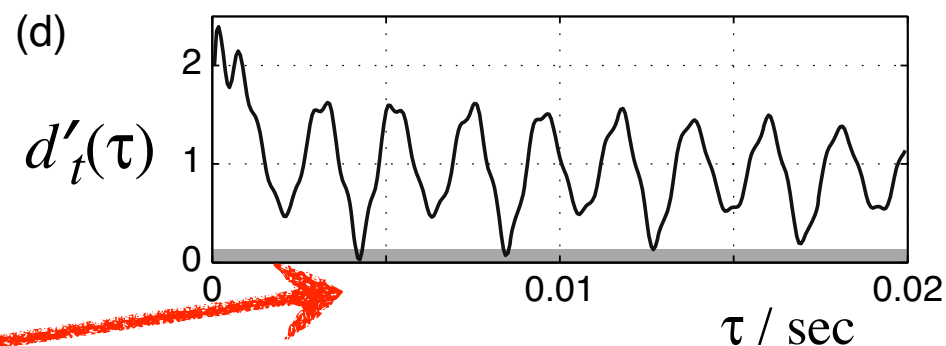


Female voiced speech,
time domain waveform



Raw autocorrelation.
Requires "peak" picking.

Yin normalization:
autocorrelation peaks
fall in a small
threshold around "0"



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