

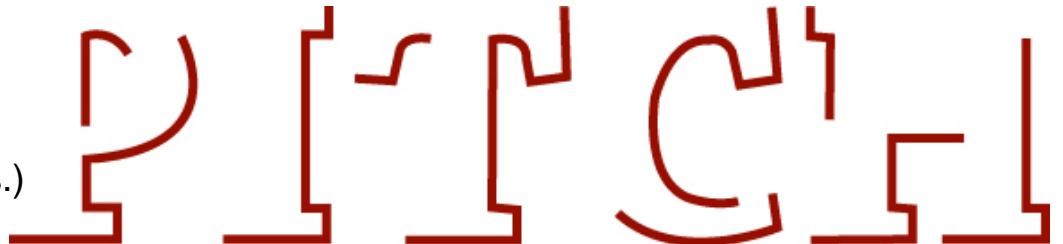
Harvard-MIT Division of Health Sciences and Technology
HST.725: Music Perception and Cognition
Prof. Peter Cariani



HST 725 Lecture 6

Temporal models for musical pitch

(Image removed due to copyright considerations.)



The search for the missing fundamental: theories & models of musical pitch

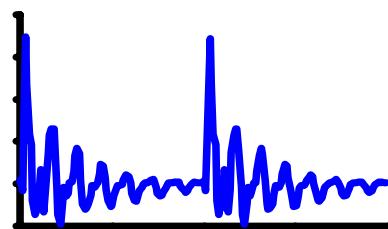
- **Temporal theories of pitch**
 - Residues: Beatings of unresolved harmonics (Schouten, 1940's)
 - Problems with residues and envelopes
 - Temporal autocorrelation models (Licklider, 1951)
 - Interspike interval models (Moore, 1980)
 - Correlogram demonstration (Slaney & Lyon, Apple demo video)
 - Population-interval models (Meddis & Hewitt, Cariani & Delgutte)
- **Interval-based representations of the spectrum**
 - Vowels
- **Competition between interspike interval patterns**
 - Pitch multiplicity
 - Masking
- **Problems & prospects**
 - The fate of temporal information in the auditory system
 - Rethinking the problem of pitch: neural timing nets
 - Pitch and auditory scene analysis
 - Formation and separation of auditory objects using temporal patterns

Synthetic vowels

Waveforms

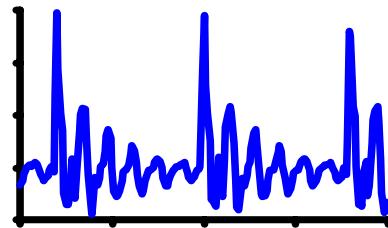
[ae]

F0 = 100 Hz



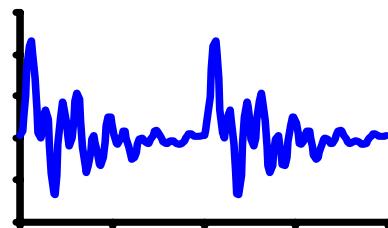
[ae]

F0 = 125 Hz



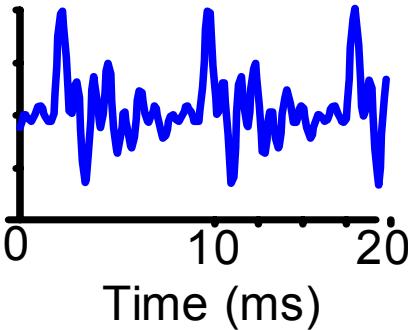
[er]

F0 = 100 Hz



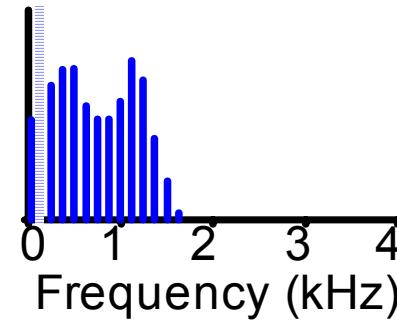
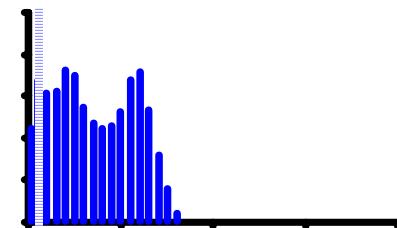
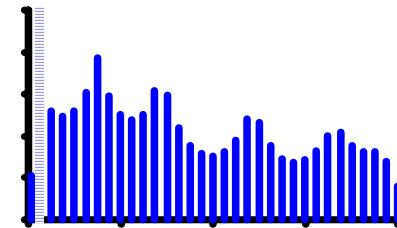
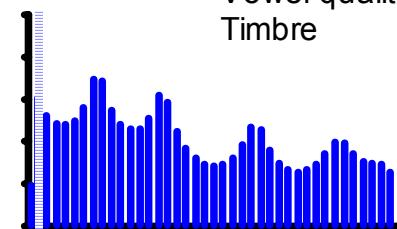
[er]

F0 = 125 Hz



Power Spectra

Formant-related
Vowel quality
Timbre

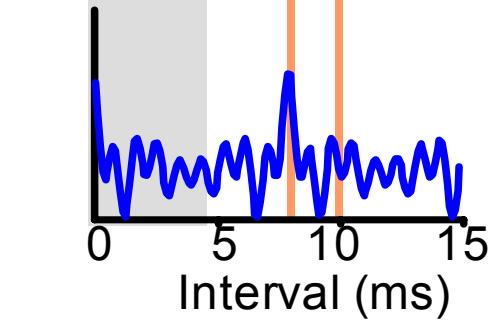
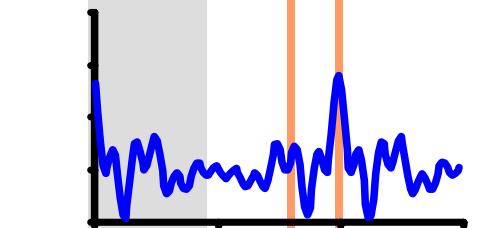
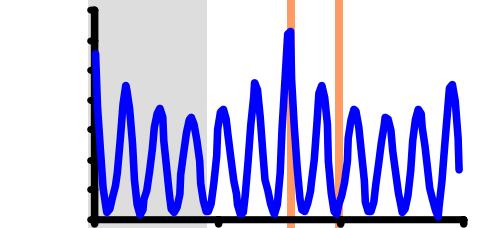
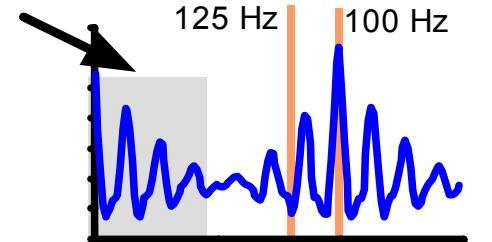


Autocorrelations

Pitch periods, 1/F0

125 Hz

100 Hz



Pitch : basic properties

- **Highly precise percepts**
 - **Musical half step:** 6% change F0
 - **Minimum JND's:** 0.2% at 1 kHz (20 usec time difference, comparable to ITD jnd)
- **Highly robust percepts**
 - **Robust quality** Salience is maintained at high stimulus intensities
 - **Level invariant** (pitch shifts < few % over 40 dB range)
 - **Phase invariant** (largely independent of phase spectrum, $f < 2$ kHz)
- **Strong perceptual equivalence classes**
 - **Octave similarities** are universally shared
 - **Musical tonality** (octaves, intervals, melodies) 30 Hz - 4 kHz
- **Perceptual organization (“scene analysis”)**
 - **Fusion:** Common F0 is a powerful factor for grouping of frequency components
- **Two mechanisms? Temporal (interval-based) & place (rate-based)**
 - **Temporal:** predominates for periodicities < 4 kHz (level-independent, tonal)
 - **Place:** predominates for frequencies > 4 kHz (level-dependent, atonal)

Periodic sounds produce distinct pitches

Many different sounds produce the same pitches

Strong

- Pure tones
- Harmonic complexes
- Iterated noise

Weaker

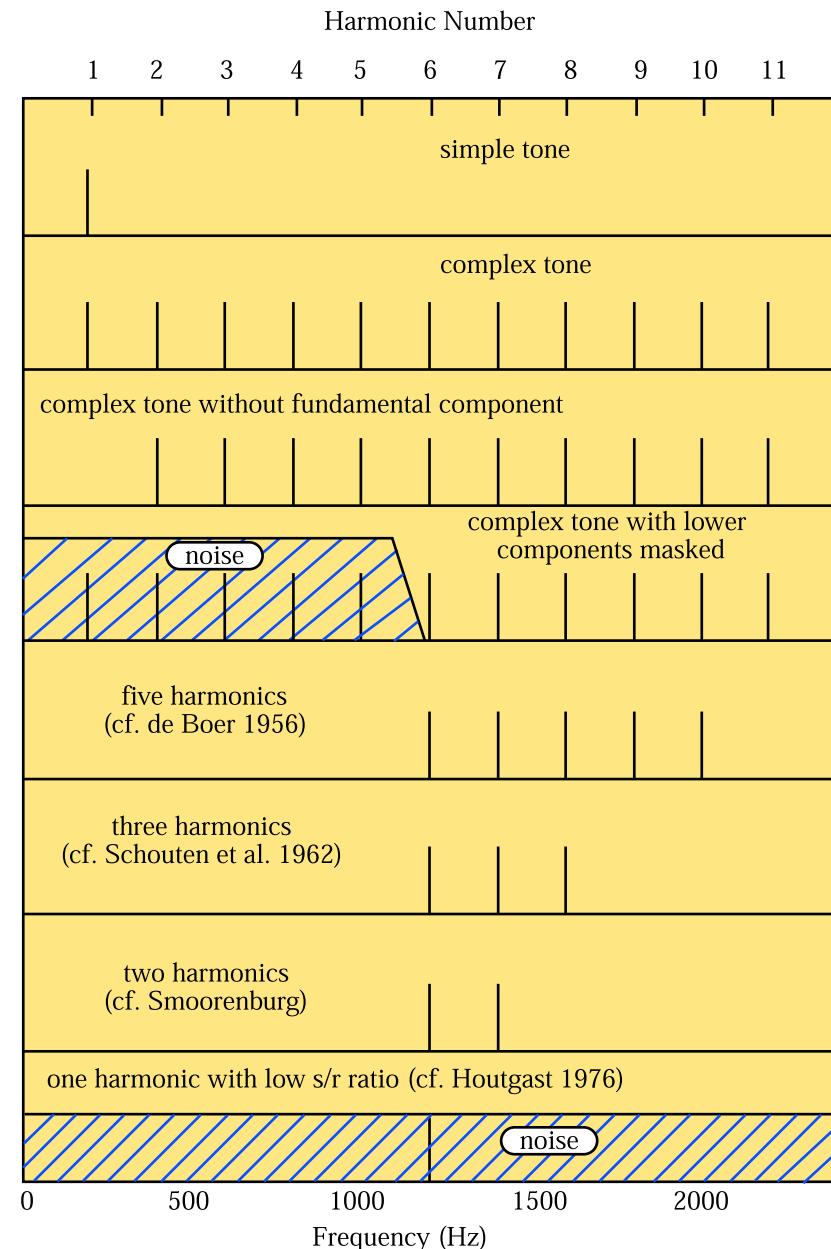
- High harmonics
- Narrowband noise

Very weak

- AM noise
- Repeated noise

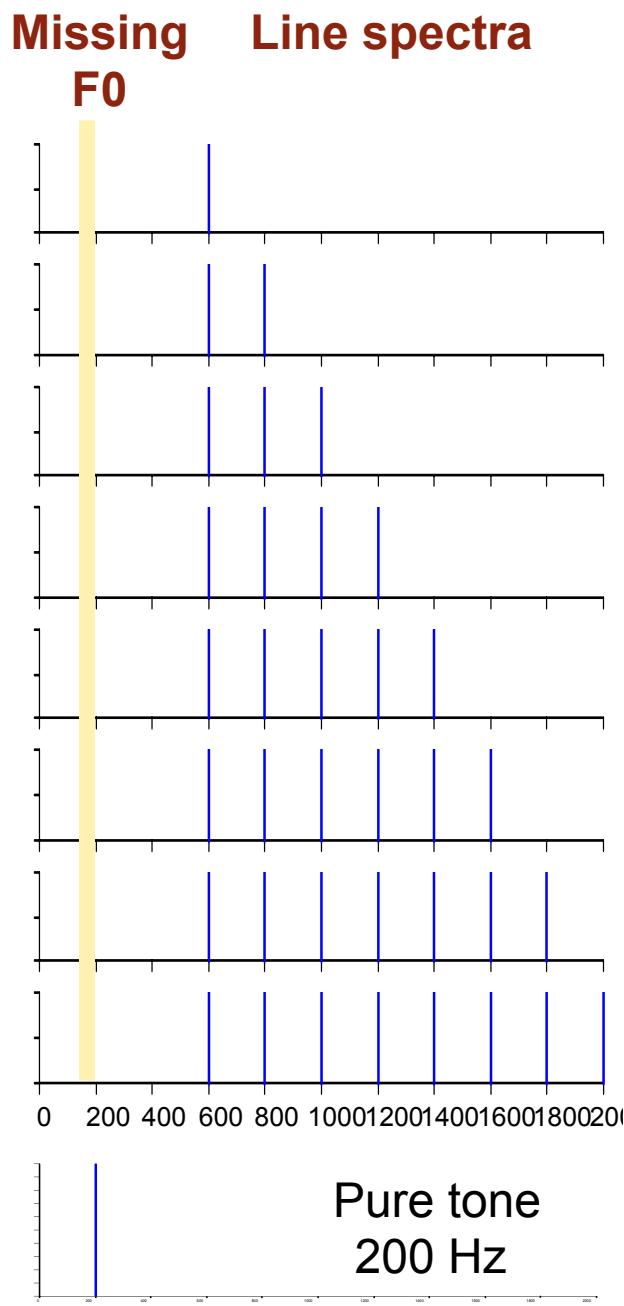
Strong
pitches

Weaker
low
pitches

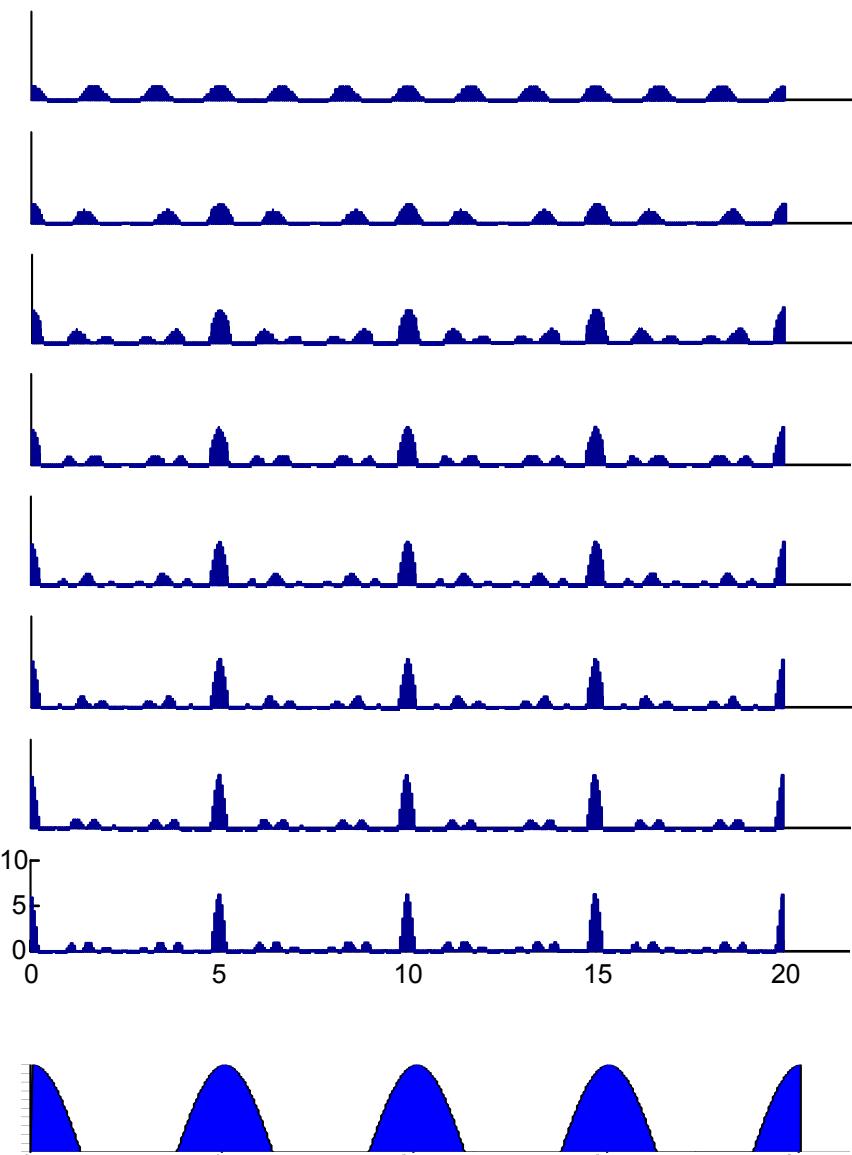


Schematic diagram representing eight signals with the same low pitch.

Pitch as a perceptual emergent

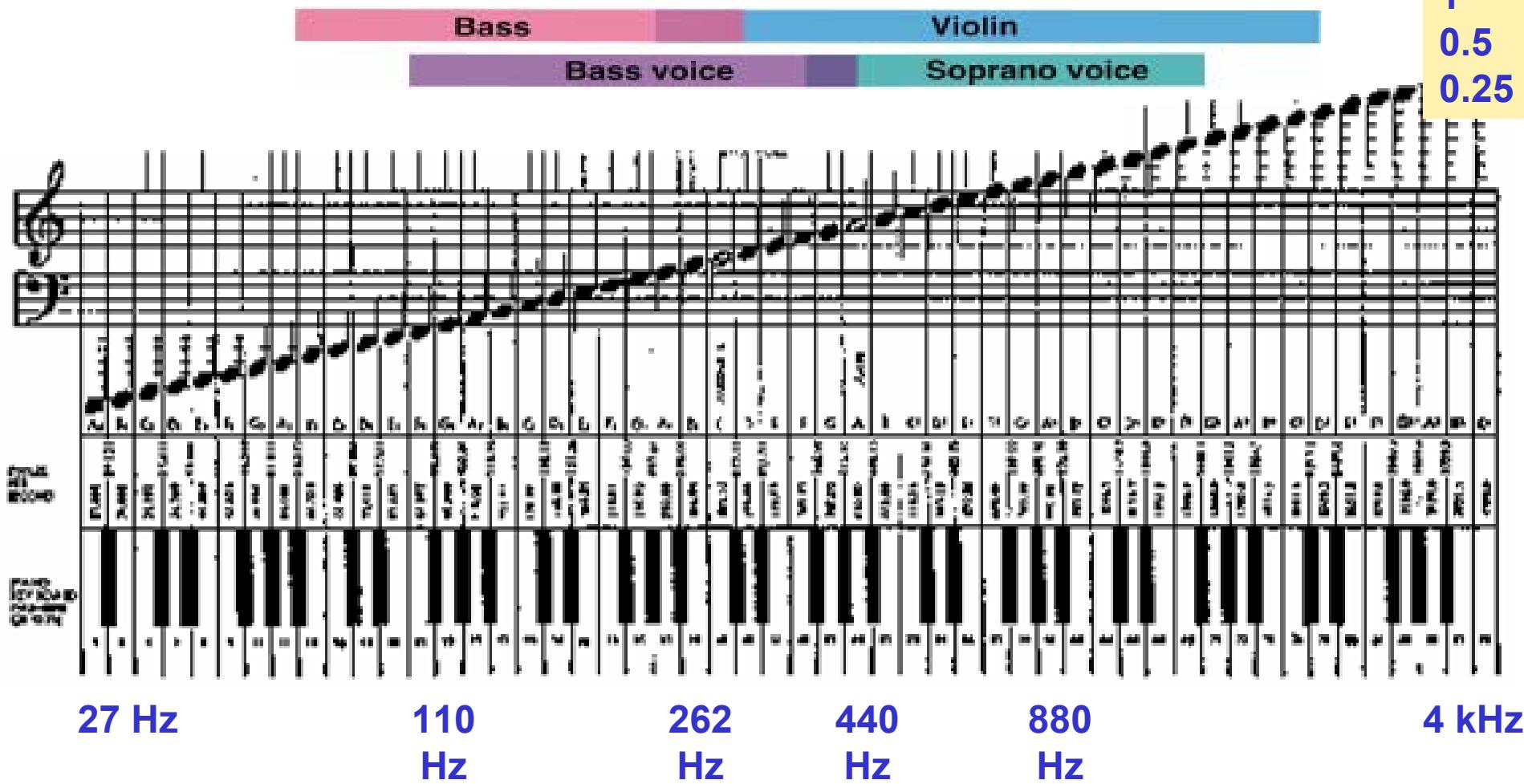


Autocorrelation (positive part)



10k
8
6
5
4
3
2
1
0.5
0.25

Frequency ranges of (tonal) musical instruments



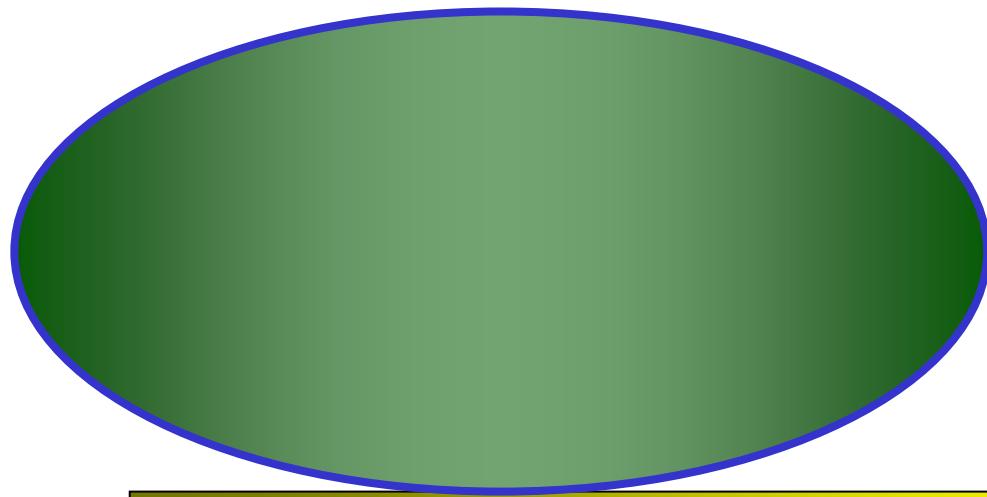
Duplex time-place representations

"Pitch is not simply frequency"

Musical tonality: octaves, intervals, melodies



Strong phase-locking (temporal information)



temporal representation
level-invariant, precise

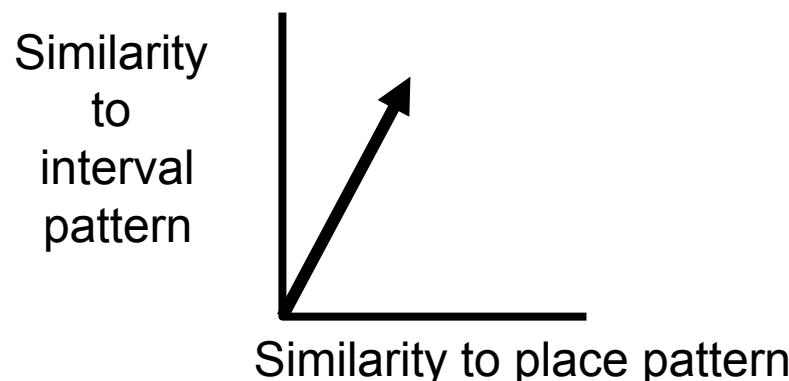
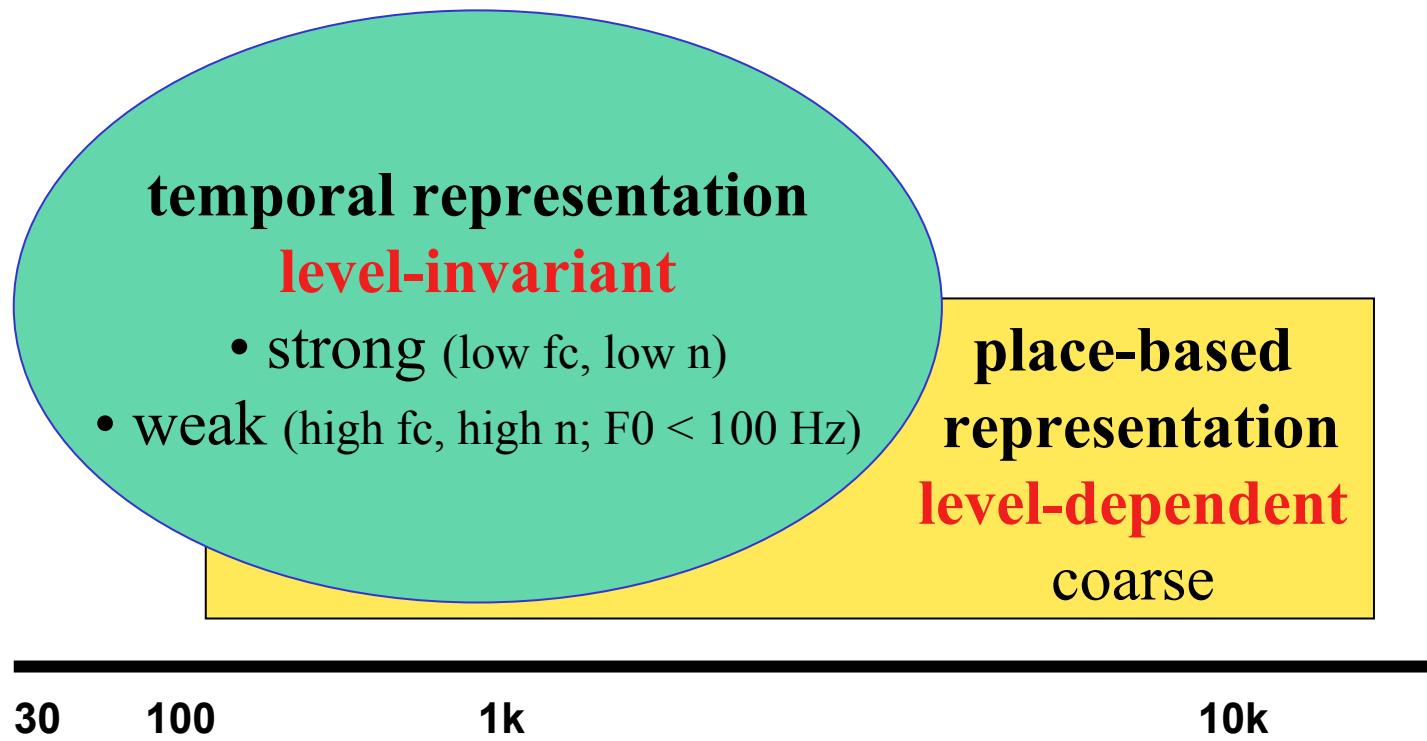
place representation
level-dependent, coarse



Pitch height and pitch chroma

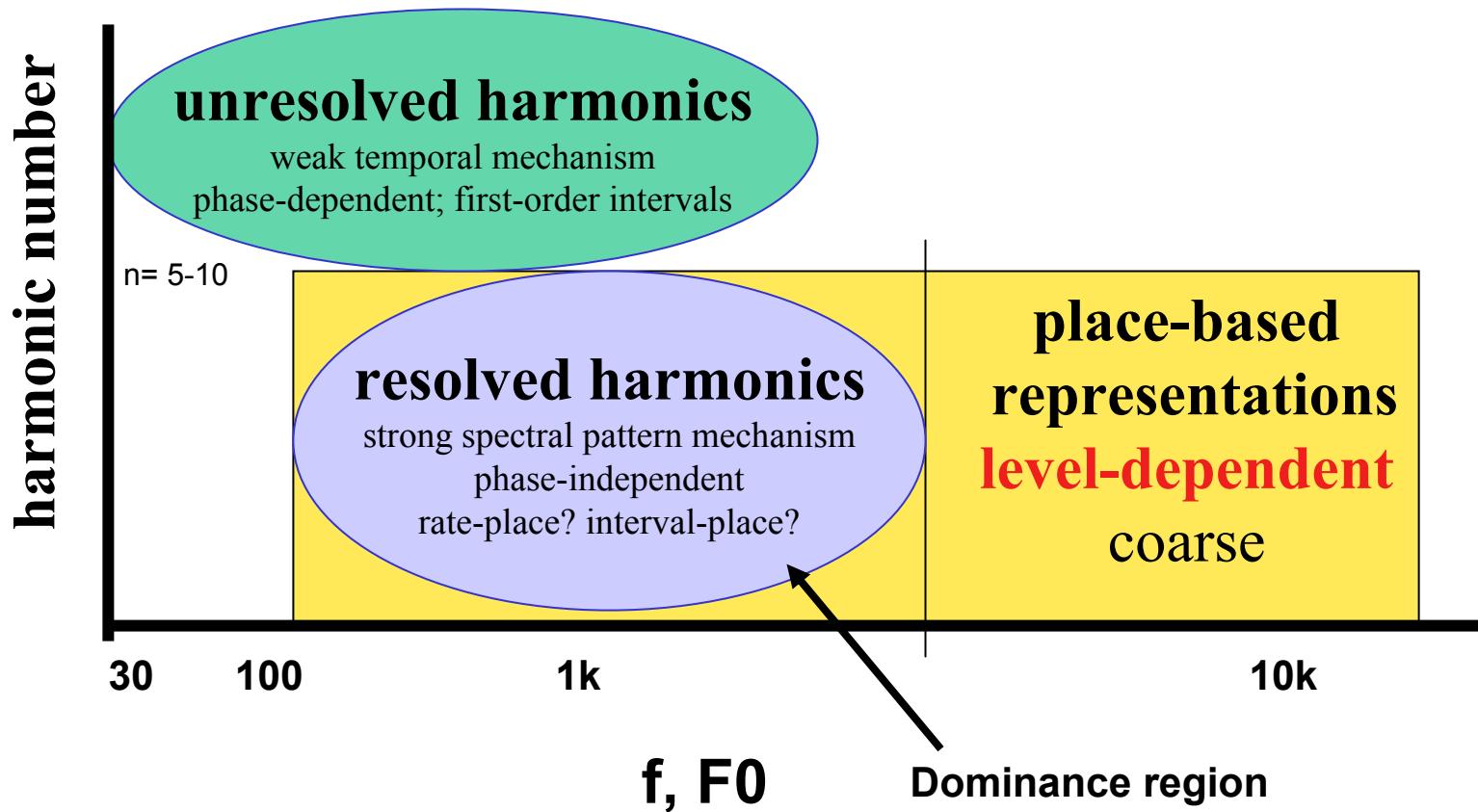
Please see Figure 1, 2, and 7 in Roger N. Shepard. "Geometrical Approximations to the Structure of Musical Pitch." *Psychological Review* 89 no. 4 (1982): 305-322.

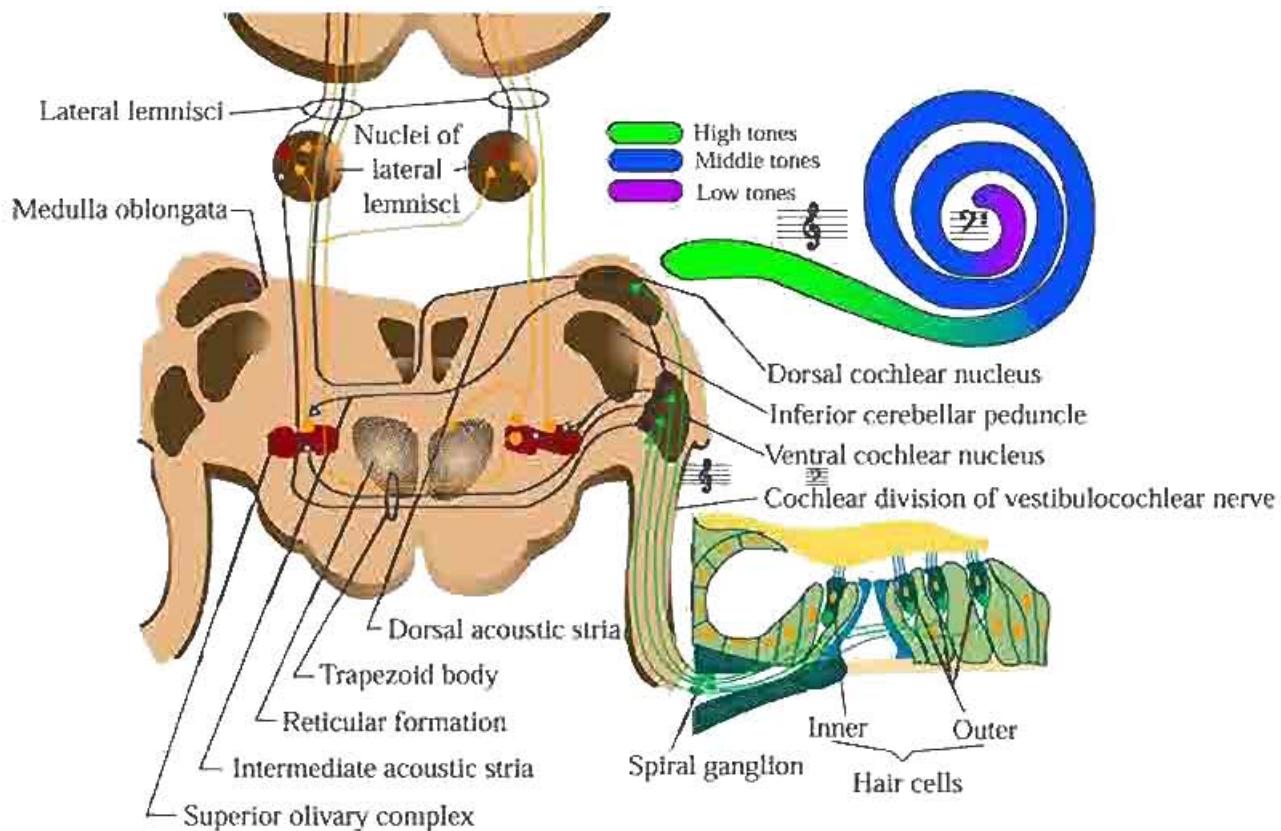
Duplex time-place representations



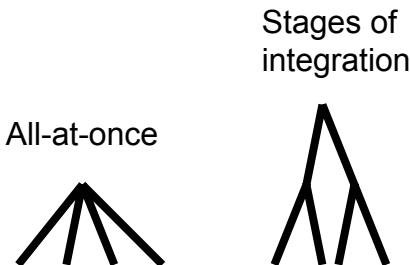
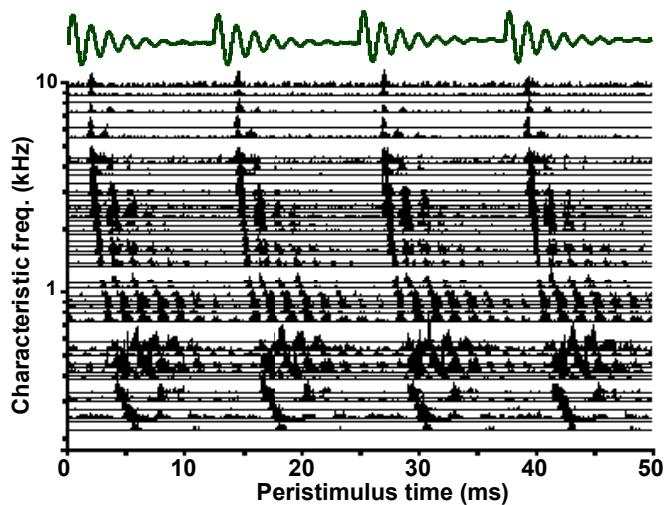
cf. Terhardt's
spectral and virtual pitch

A "two-mechanism" perspective (popular with some psychophysicists)





Some possible auditory representations

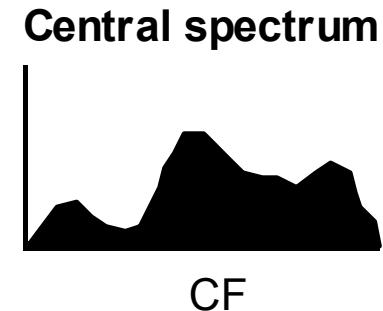


Local

Rate-place



Masking phenomena
Loudness

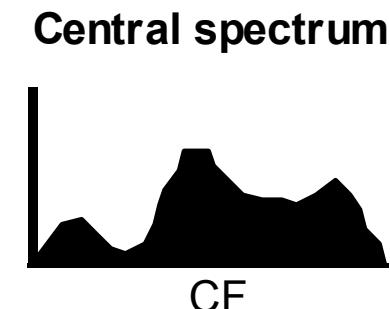


Synchrony-place

Phase-place

Interval-place

Pure tone pitch JNDs: Goldstein



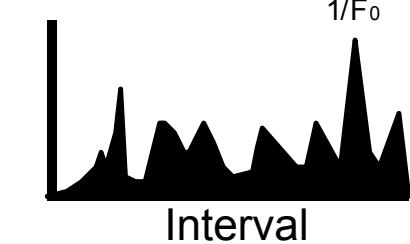
Global

Population-interval

Complex tone pitch



Population interval



General theories of pitch

1. Distortion theories

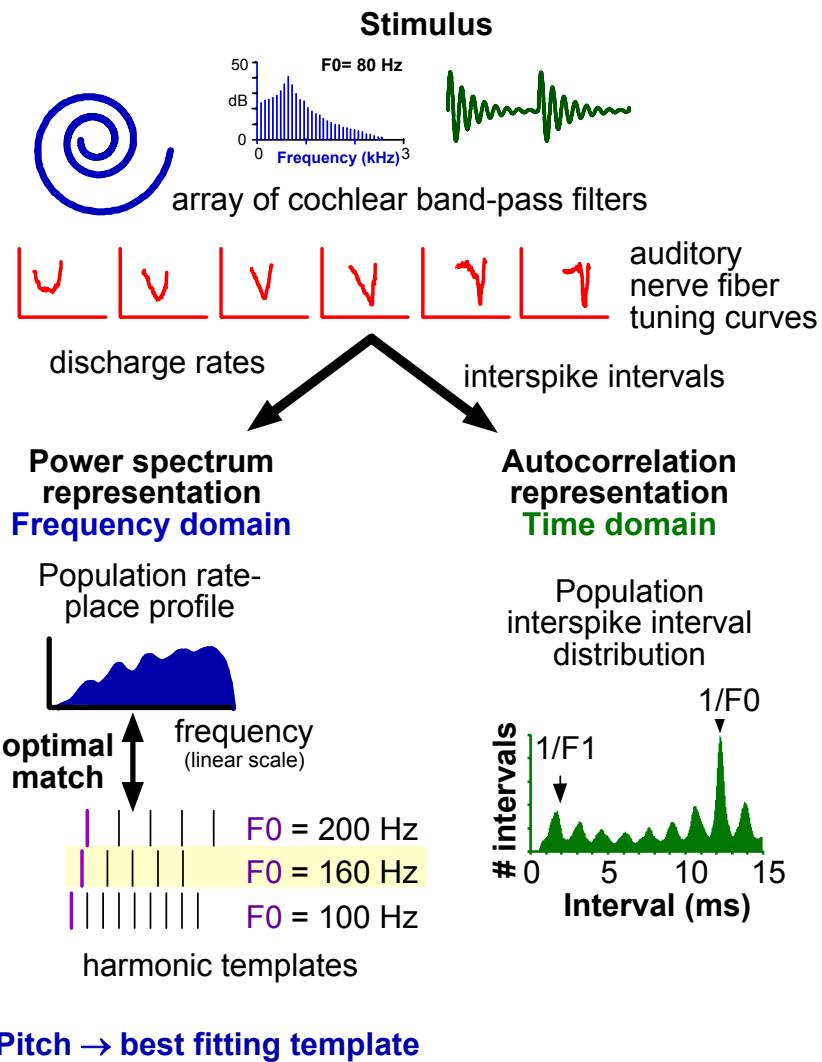
- reintroduce F0 as a cochlear distortion component (Helmholtz)
- sound delivery equipment can reintroduce F0 through distortion
- however, masking F0 region does not mask the low pitch (Licklider)
- low pitch thresholds and growth of salience with level not consistent with distortion processes (Plomp, Small)
- binaurally-created pitches exist

2. Spectral pattern theories

- Operate in frequency domain
- Recognize harmonic relations on resolved components

3. Temporal pattern theories

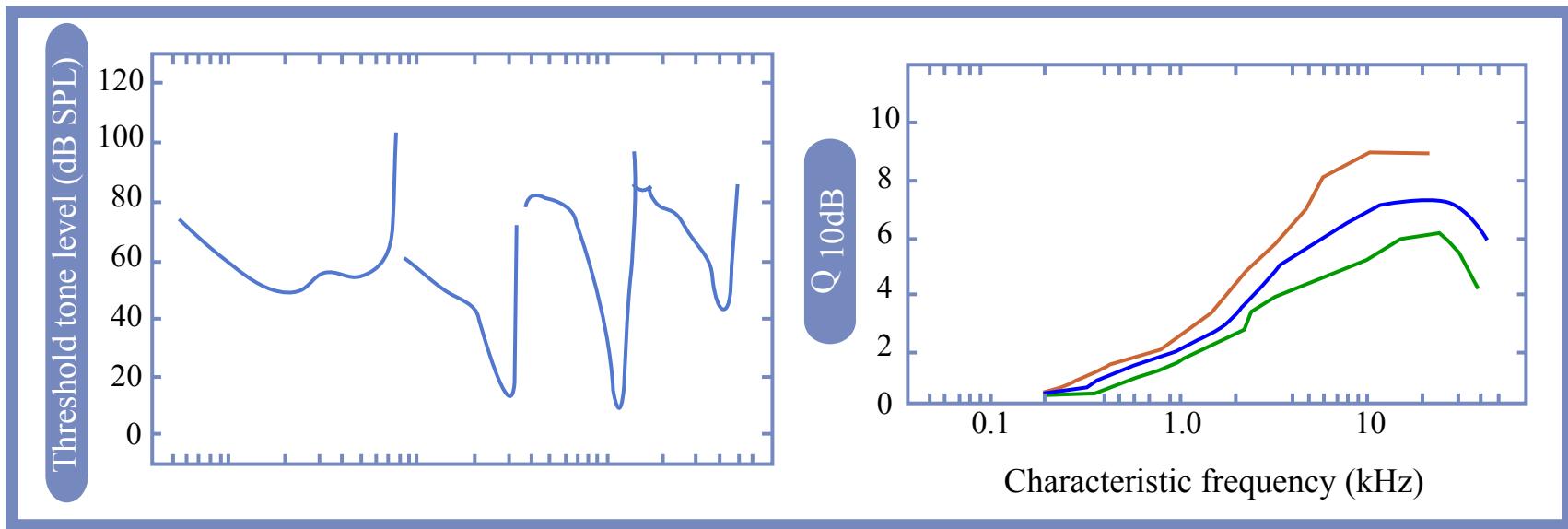
- Operate in time domain
- Analyze interspike interval dists.



Basic problems to be solved

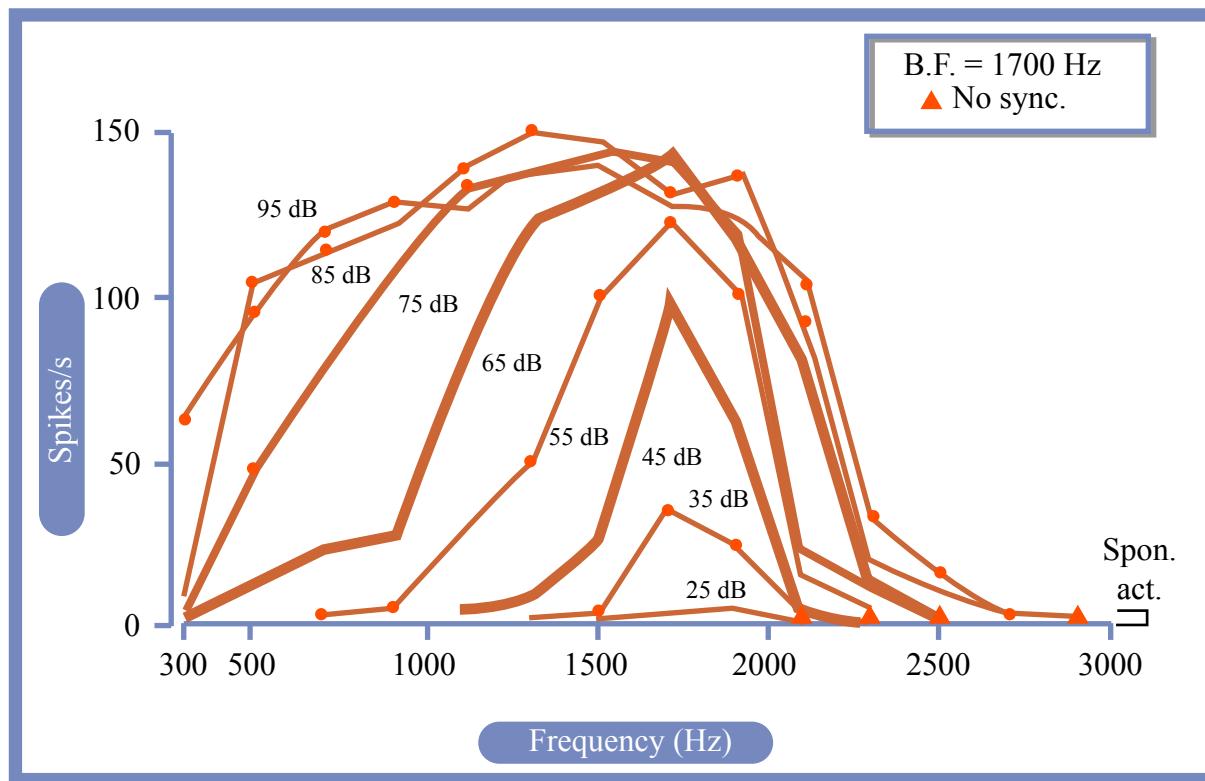
- "Hyperacuity problem"
 - Account for the precision of pitch discriminations given the relatively coarse tunings of auditory neurons (at all levels), especially lower-frequency ones (BFs < 2 kHz)
 - "Dynamic range problem"
 - Account for the ability of listeners to discriminate small fractional changes ($\Delta I/I$) in intensity over a large dynamic range, and especially at high SPLs, where the vast majority of firing rates are saturated.
 - "Level-invariance problem"
 - Account for the invariance (and precision) of auditory percepts over large dynamic ranges given the profound changes in neural response patterns that occur over those ranges (rate saturation, rate non-montonicities).
- Pitch equivalence**
- Account for the ability to precisely match pitches of pure and complex tones (pitch equivalence, metamery) given differences in spectra and under conditions where stimulus levels are roved 20 dB or more
- Multiple pitches**
- Account for the ability of the auditory system to simultaneously represent multiple pitches ($\Delta F_0 > 10\%$ apart)
- Relative nature of pitch & transpositional invariance**
- Account for the ability to precisely match pitches an octave apart (and/or to recognize patterns of pitch sequences) in the absence of an ability to identify absolute frequencies/periodicities. Account for ability to recognize transposed melodies as similar.

Neural tuning as a function of CF

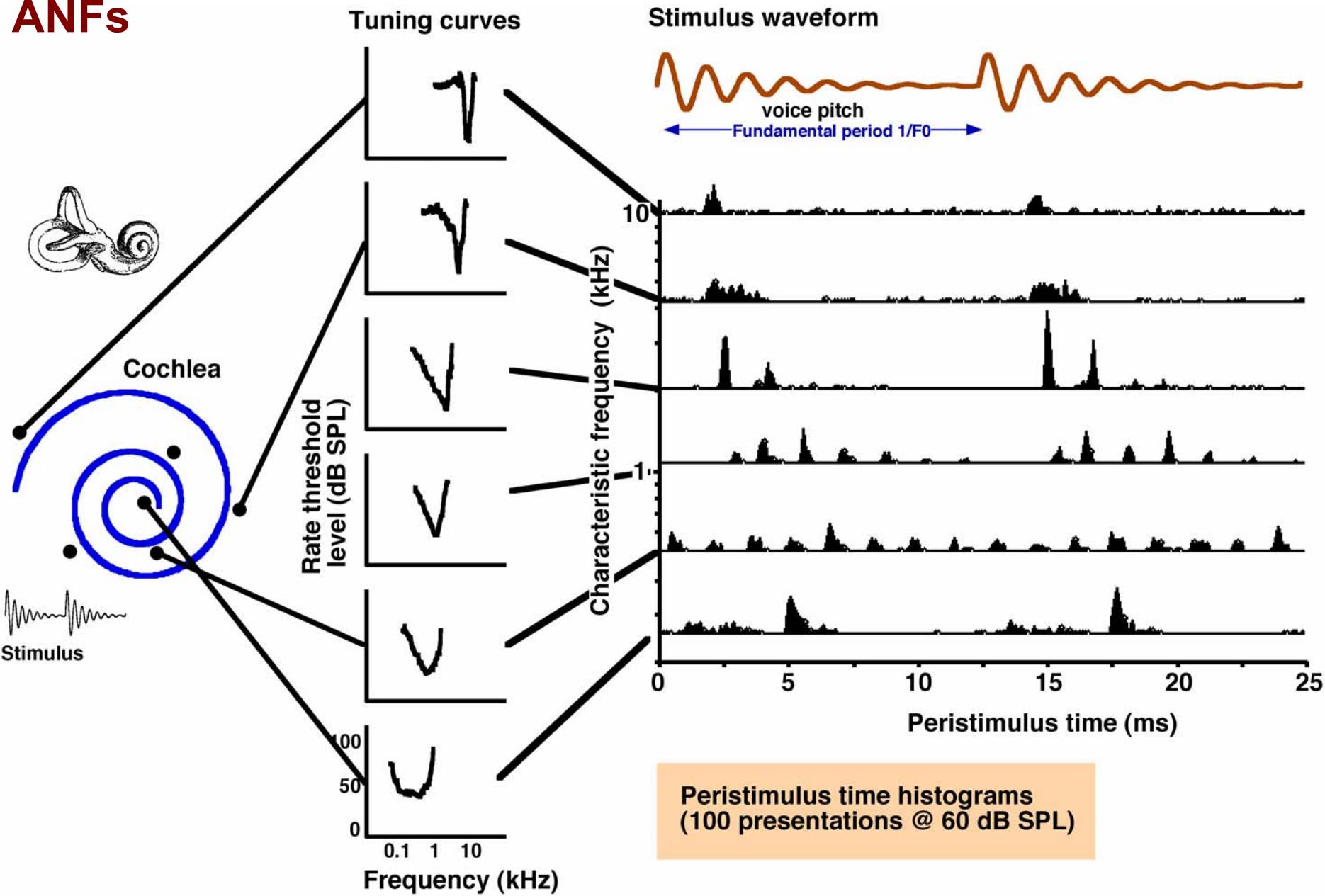


Broad tuning and rate saturation at moderate levels in low-CF auditory nerve fibers confounds rate-based resolution of harmonics.

Low SR auditory nerve fiber



ANFs



Temporal pattern theories

Σ First-order intervals (renewal density)

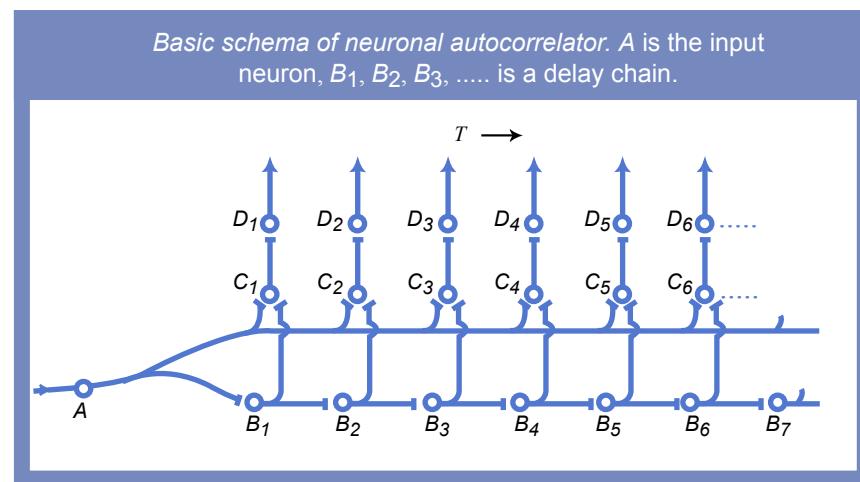
- Schouten's temporal theory (1940's) depended on interactions between unresolved (high) harmonics. It was displaced by discovery of dominance region and binaural combination pitches in the 1960's. The idea persists, however in the form of spectral mechanisms for resolved harmonics and temporal ones for unresolved harmonics.

Please see van Noorden, L. Two channel pitch perception. In Music, Mind and Brain. Edited by M. Clynes. New York: Plenum.

Σ All-order intervals (temporal autocorrelation)

Please see Moore, BCJ. An introduction to the psychology of hearing. 5th ed. San Diego: Academic Press. 2003.

Licklider (1951)



Please see Figure 1 in Meddis, R., and M. J. Hewitt. Virtual pitch and phase sensitivity of a computer model of the Auditory periphery. I. Pitch identification. J. Acoust. Soc. Am. 89 no. 6 (1991): 2866-2882.

Interval-based theories of pitch

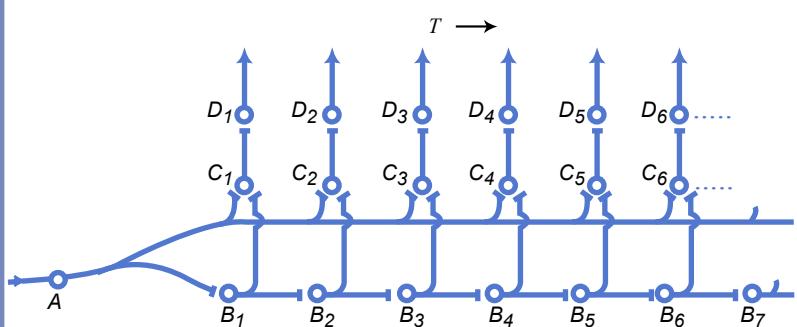
First-order intervals (renewal density)

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All-order intervals (temporal autocorrelation)

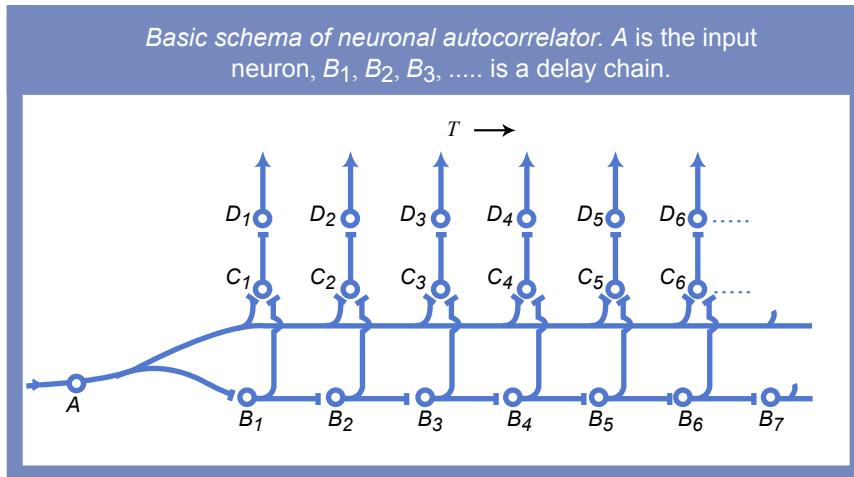
Basic schema of neuronal autocorrelator. A is the input neuron, B_1, B_2, B_3, \dots is a delay chain.



Licklider (1951)

Please see Figure 1 in Meddis, R., and M. J. Hewitt. Virtual pitch and phase sensitivity of a computer model of the Auditory periphery. I. Pitch identification. J. Acoust. Soc. Am. 89 no. 6 (1991): 2866-2882.

Licklider's (1951) duplex model of pitch perception



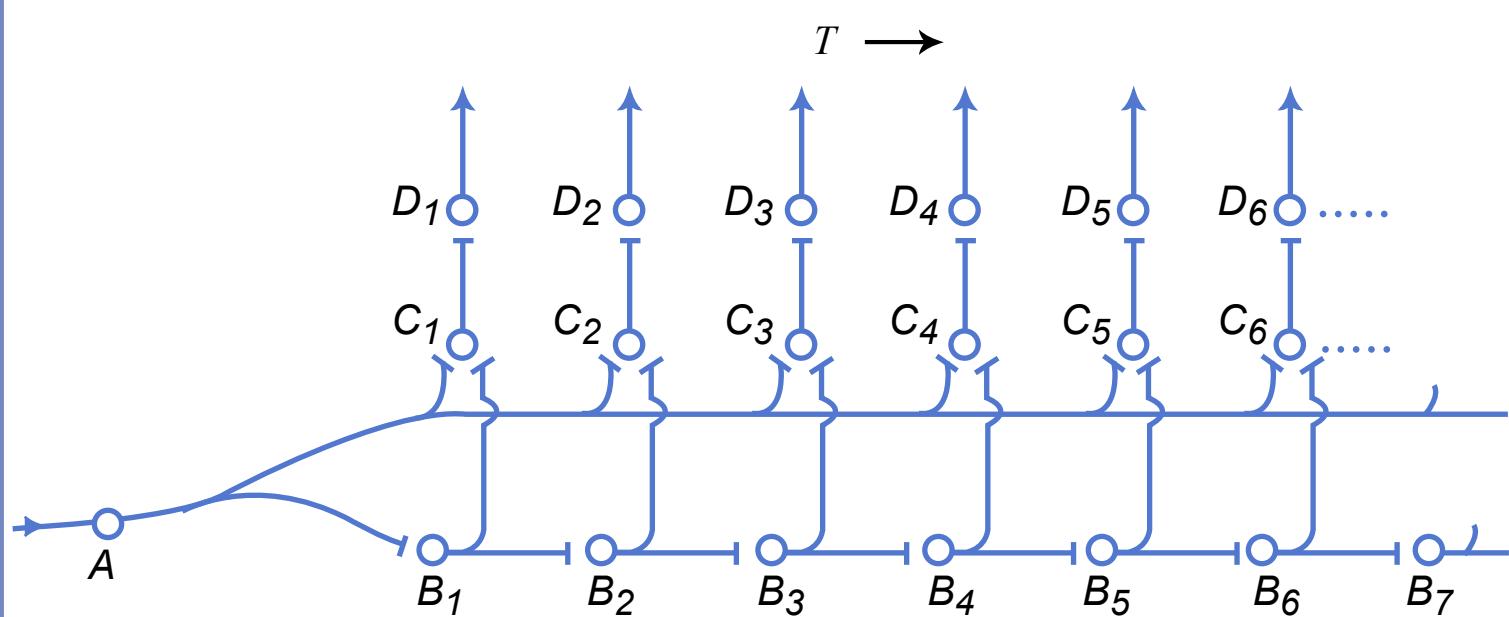
(Image removed due to copyright considerations.)

Licklider's binaural triplex model

**J.C.R. Licklider
“Three
AuditoryTheories”
In Psychology: A
Study of a
Science, Vol. 1, S.
Koch, ed.,
McGraw-Hill, 1959,
41-144.**

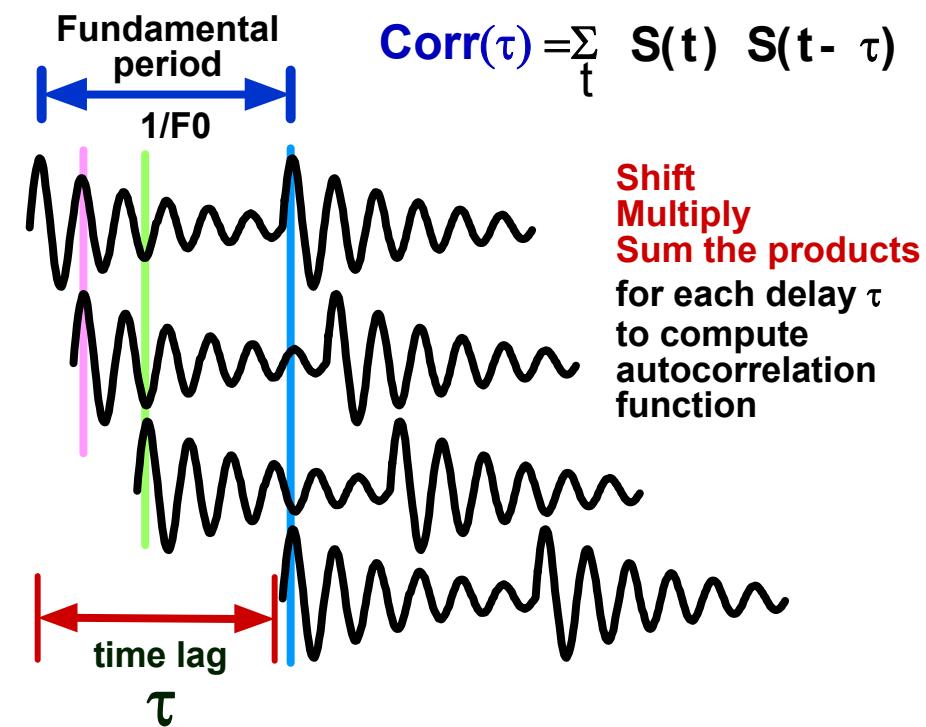
Delay lines

Basic schema of neuronal autocorrelator. A is the input neuron, B_1, B_2, B_3, \dots is a delay chain.

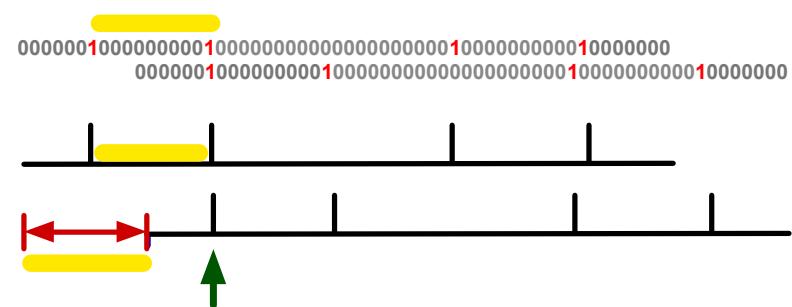


Autocorrelation and interspike intervals

Autocorrelation functions



Autocorrelations of spike trains = Histograms of all-order intervals



Please see Figure 6.16A-D in Lyon, Richard, and Shihab Shamma. "Auditory Representations of Timbre and Pitch." In Auditory Computation. Edited by R. R. Fay. New York: Springer Verlag. 1996.

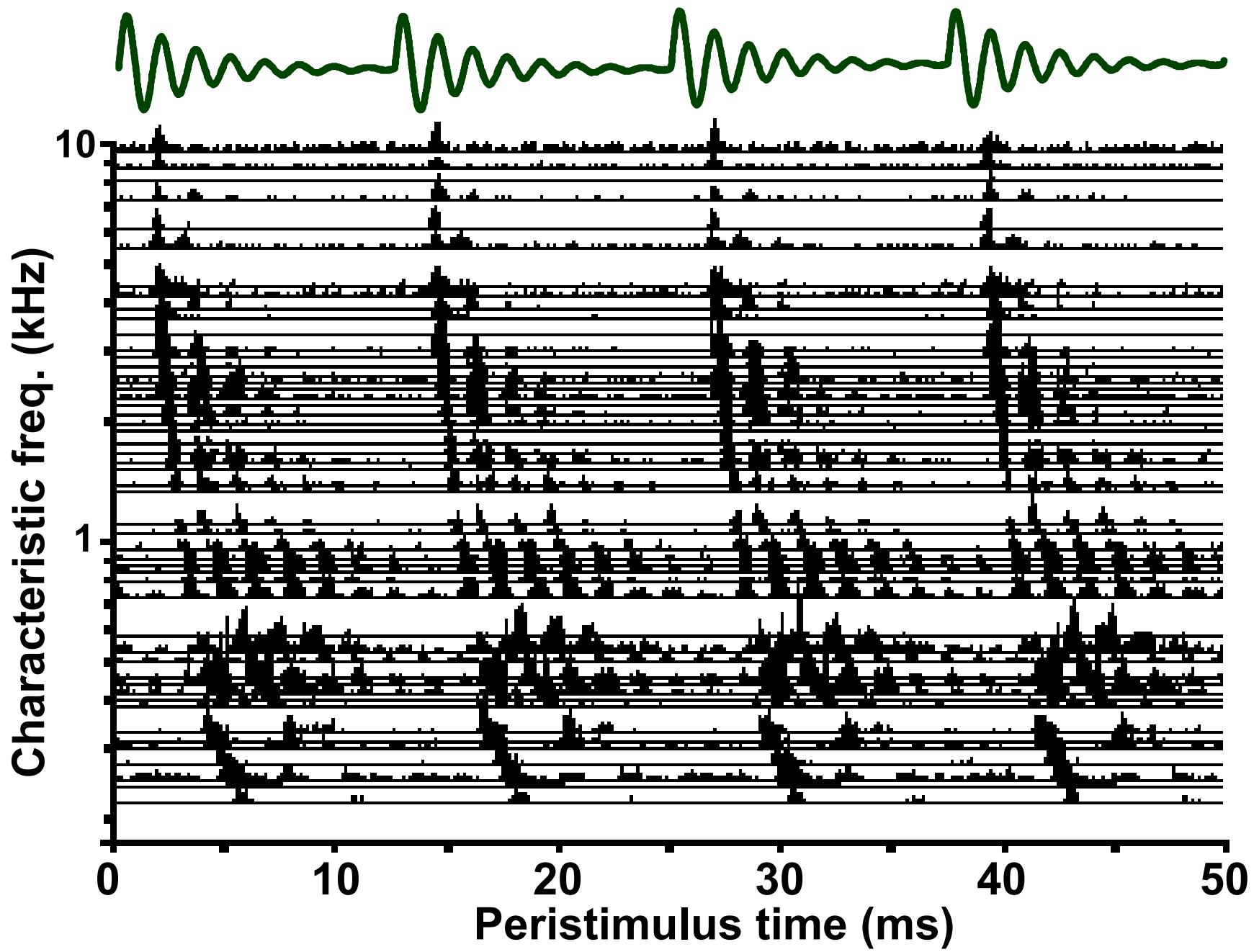
Correlograms: interval-place displays (Slaney & Lyon)

Please see Slaney, Malcolm, and Richard F. Lyon. "On the Importance of Time -A Temporal Representation of Sound." In Visual Representations of Speech Signals. Edited by M. Crawford. New York: John Wiley. 1993.

Correlograms

Please see Figure 6.17 in Slaney, Malcolm, and Richard F. Lyon. "On the Importance of Time - A Temporal Representation of Sound." In Visual Representations of Speech Signals. Edited by M. Crawford. New York: John Wiley. 1993.

Auditory nerve



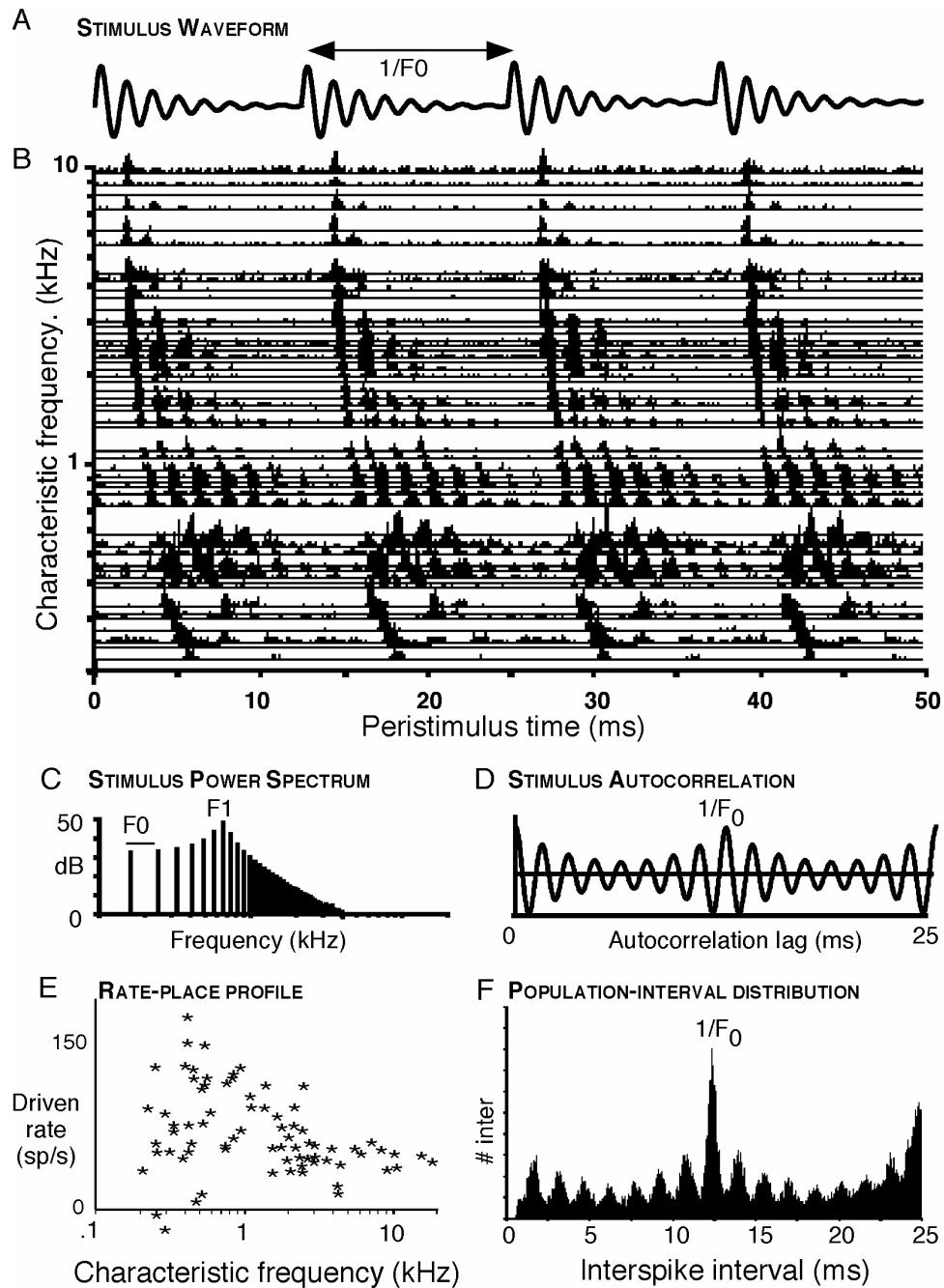
Temporal coding in the auditory nerve

Work with Bertrand Delgutte
Cariani & Delgutte (1996ab)

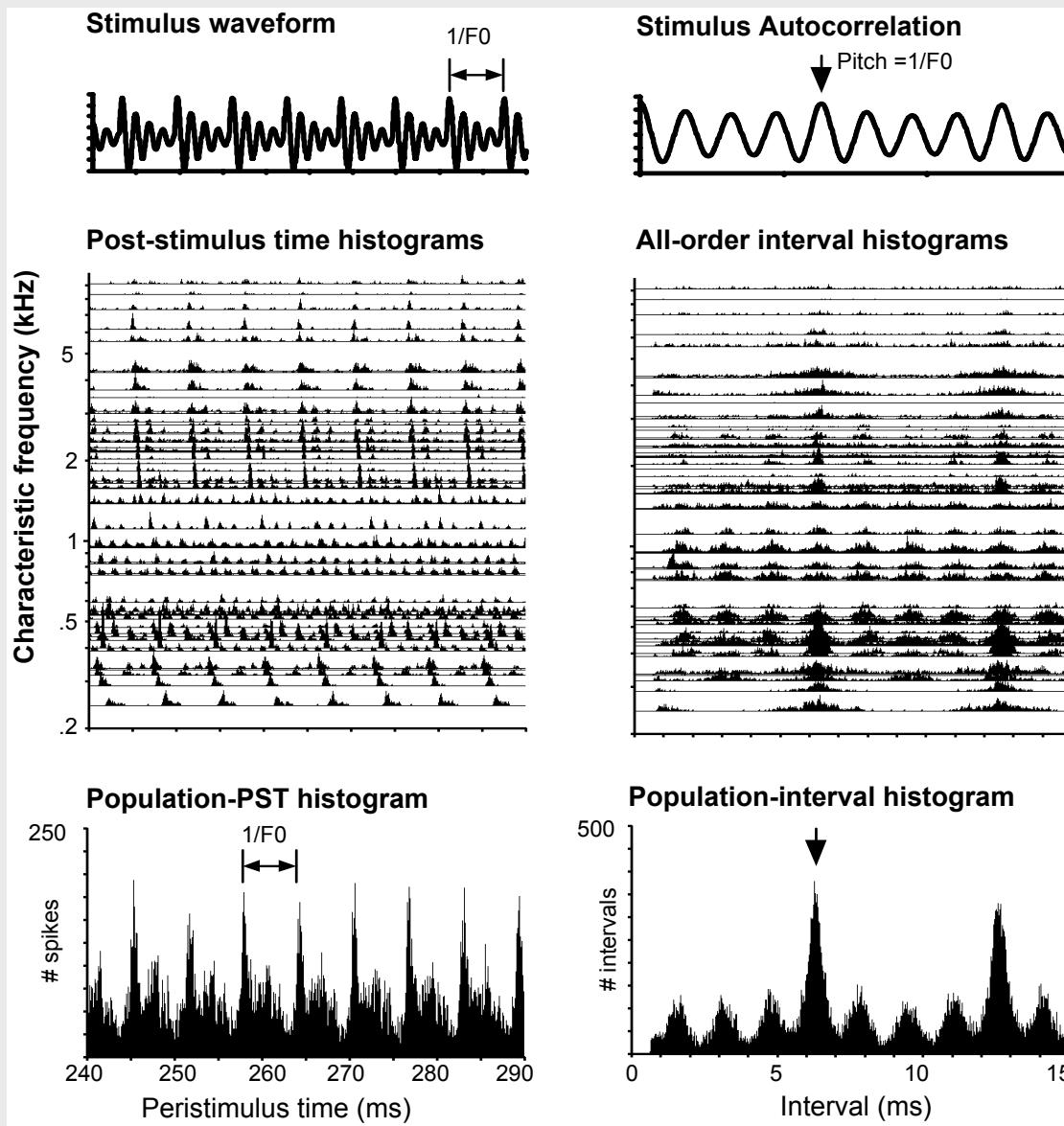
Dial-anesthetized cats.
100 presentations/fiber
60 dB SPL

Population-interval distributions are compiled by summing together intervals from all auditory nerve fibers.

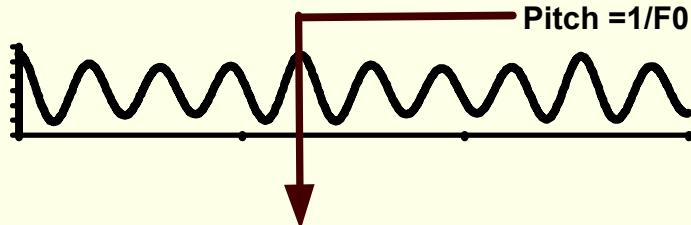
The most common intervals present in the auditory nerve are invariably related to the pitches heard at the fundamentals of harmonic complexes.



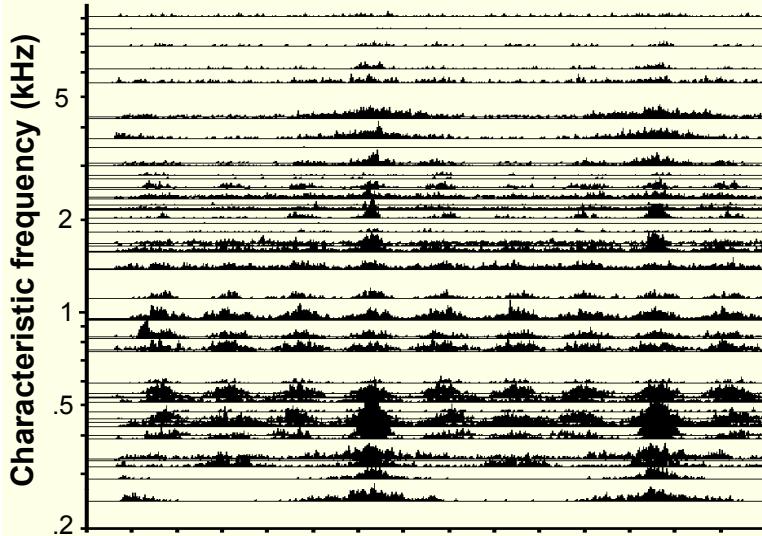
The population-interval distribution of the auditory nerve



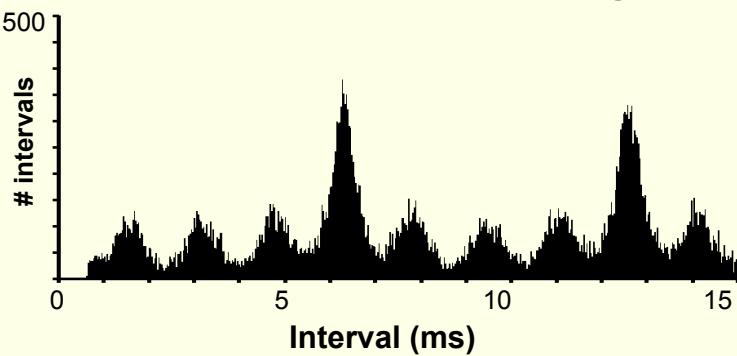
Stimulus Autocorrelation



All-order interval histograms



Population-interval histogram



Autocorrelation functions

Fundamental period

$$\text{Fundamental period} \quad \text{Corr}(\tau) = \sum_t S(t) S(t - \tau)$$

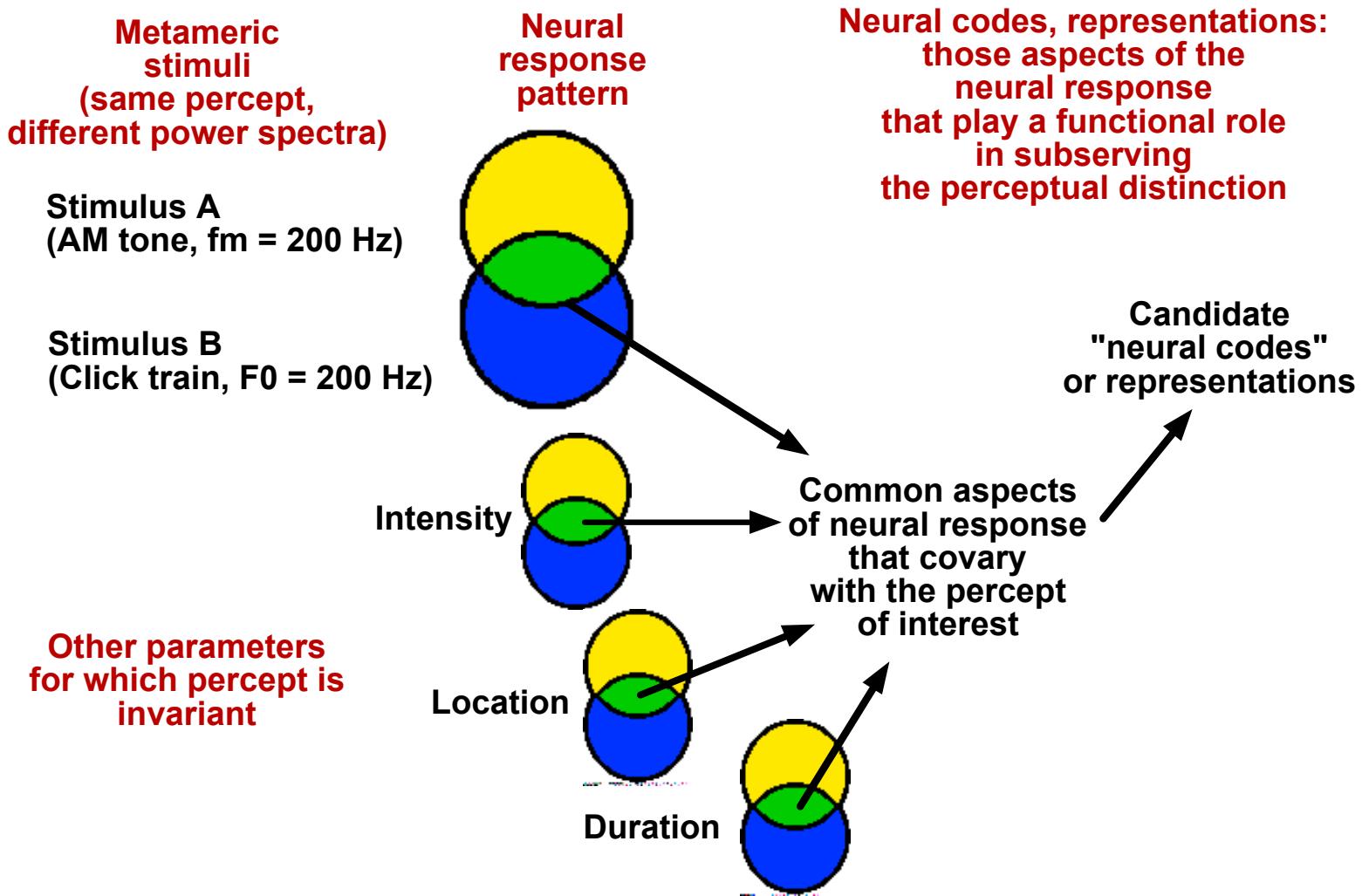
Shift
Multiply
Sum the products
for each delay τ
to compute
autocorrelation
function

The diagram illustrates a time series signal as a black wavy line. A red double-headed arrow at the start of the signal indicates a time interval labeled "time lag" above it. A blue vertical tick mark is placed on the signal at the end of the red arrow's range. The symbol τ is written below the signal, indicating the duration of the time lag.

Autocorrelations of spike trains = Histograms of all-order intervals

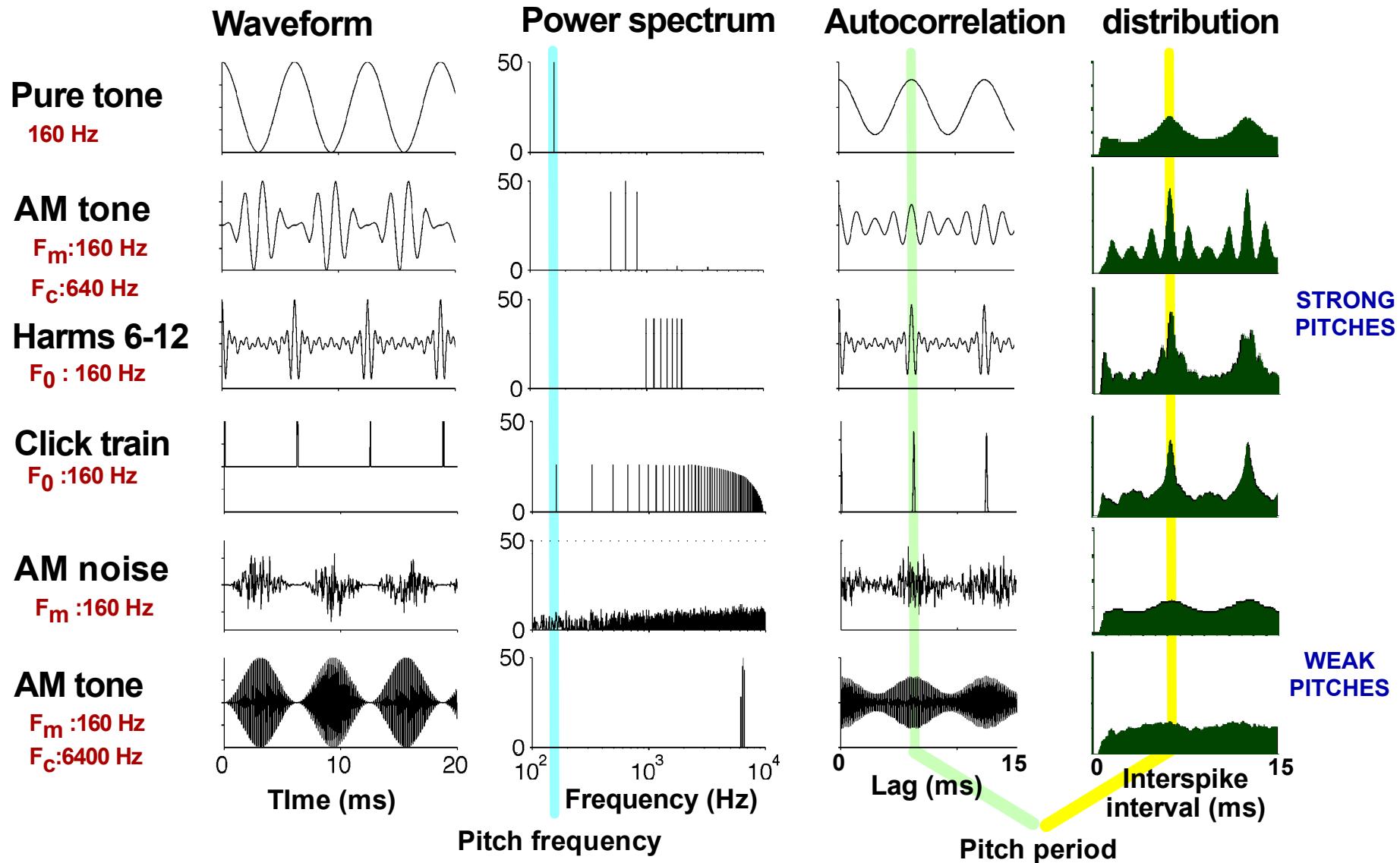
Percept-driven search for neural codes:

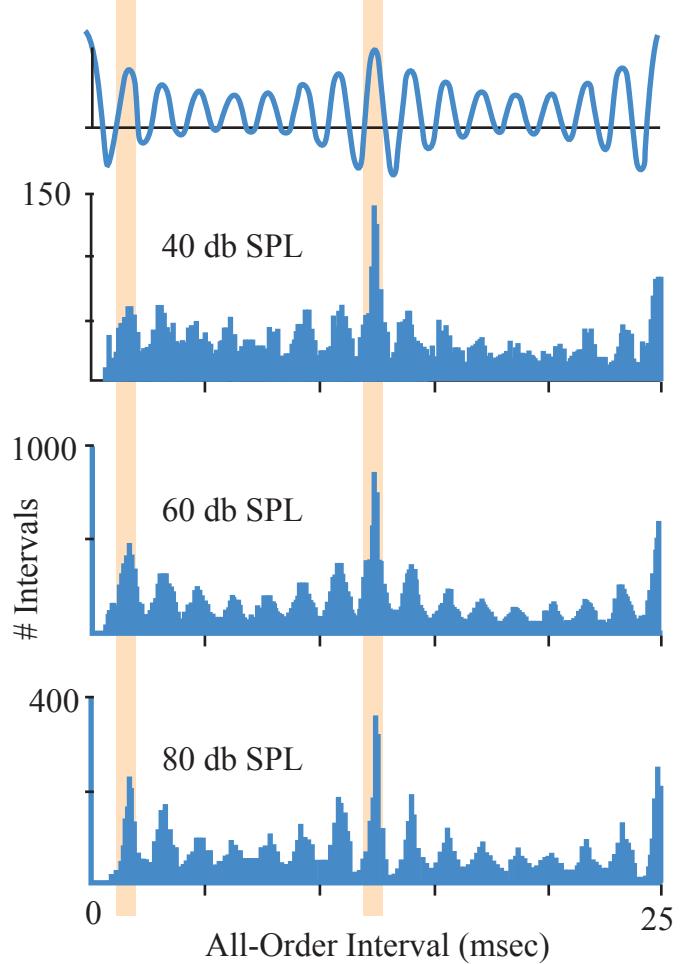
1. Use stimuli that produce equivalent percepts
2. Look for commonalities in neural response
3. Eliminate those aspects that are not invariant



Pitch equivalence

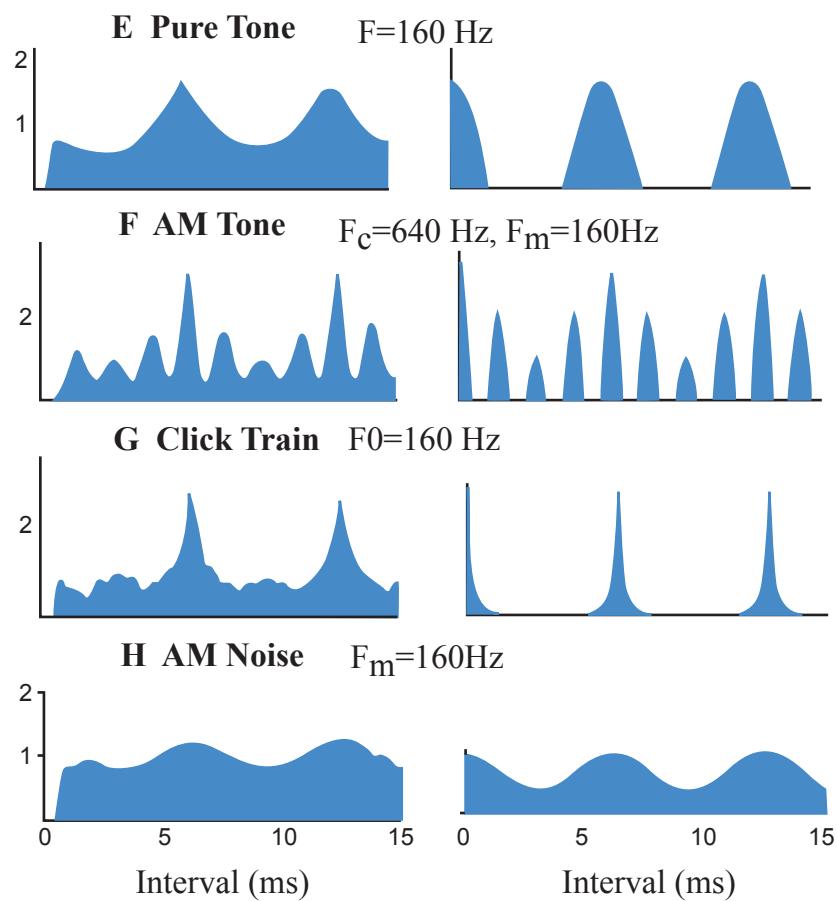
Six stimuli that produce a low pitch at 160 Hz



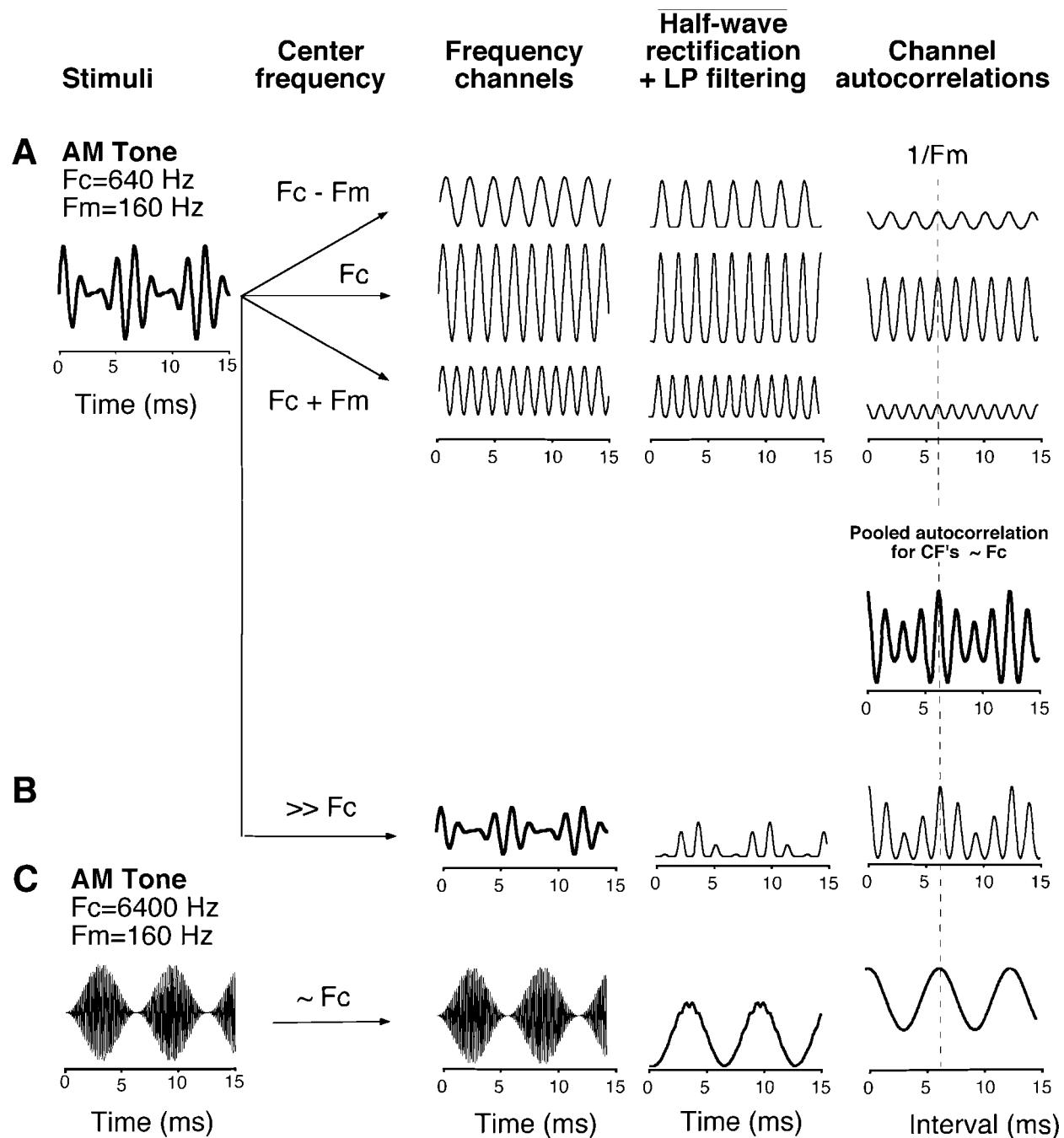


Level-Invariance

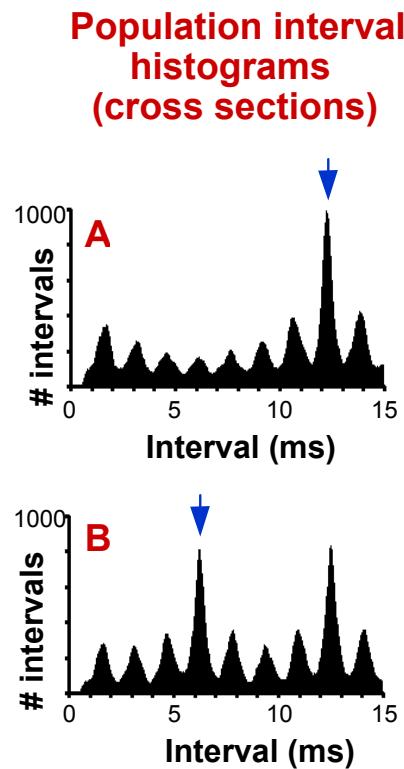
Neural Data Autocorrelation



Pitch Equivalence



The running population-interval distribution

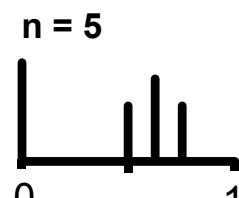
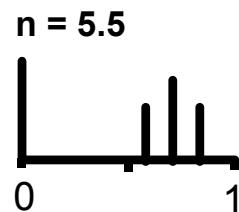
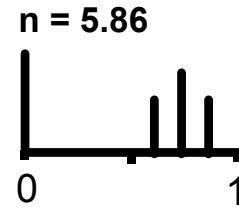
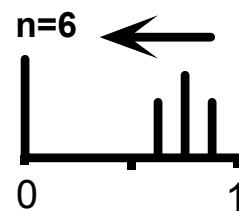


Pitch height and pitch chroma

Please see Figure 1, 2, and 7 in Roger N. Shepard. "Geometrical approximations to the structure of musical pitch." *Psychological Review* 89 no. 4 (1982): 305-322.

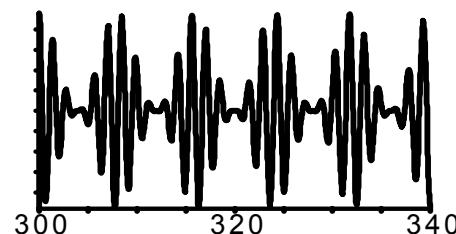
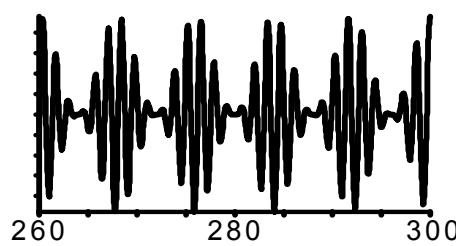
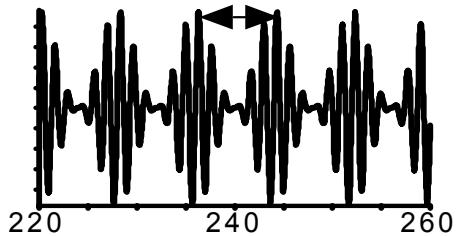
Pitch shift of inharmonic complex tones

$F_m = 125$ Hz
 $F_c = 750$ Hz



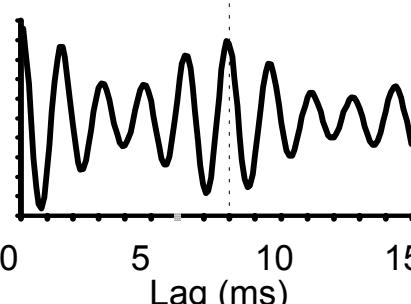
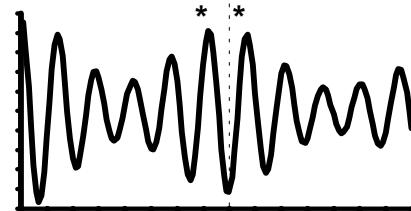
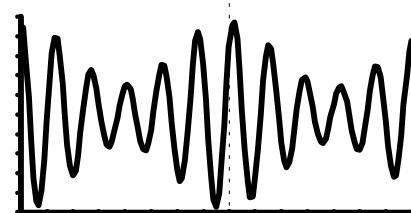
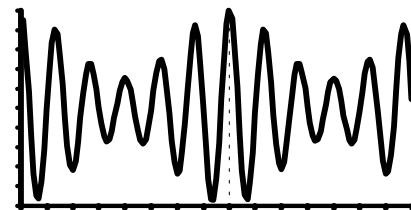
Stimulus waveform

$1/F_m$



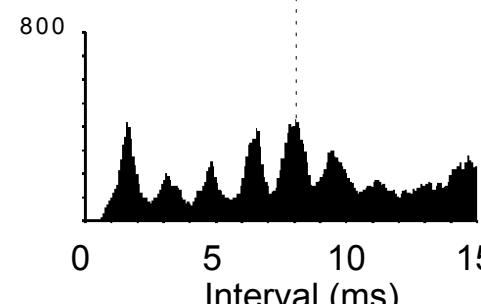
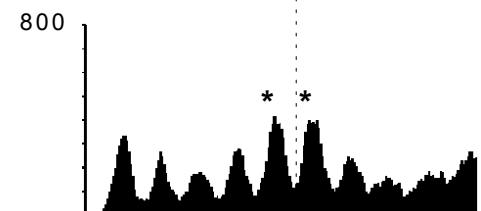
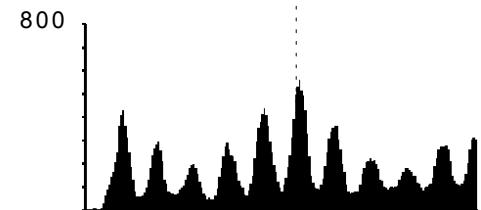
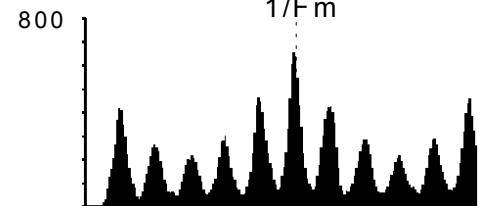
Stimulus autocorrelation

$1/F_m$



Population interval distributions

$1/F_m$

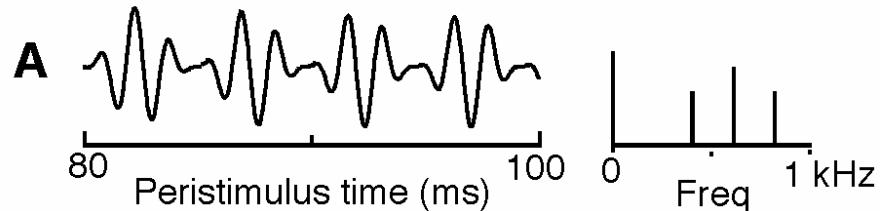


Pitch shift of inharmonic complex tones

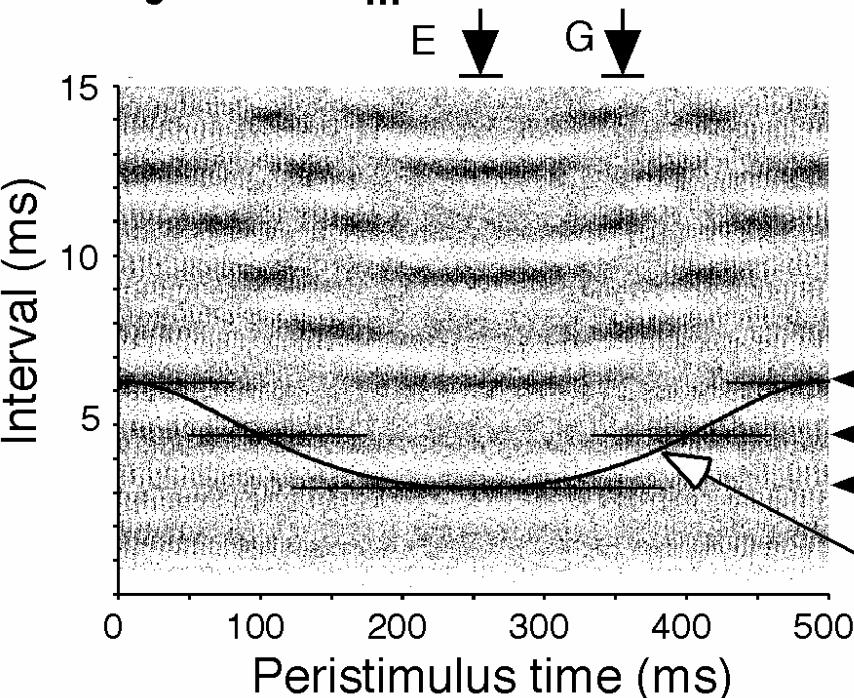
Phase-invariant nature of all-order interval code

AM Tone

$F_c=640$ Hz, $F_m=200$ Hz

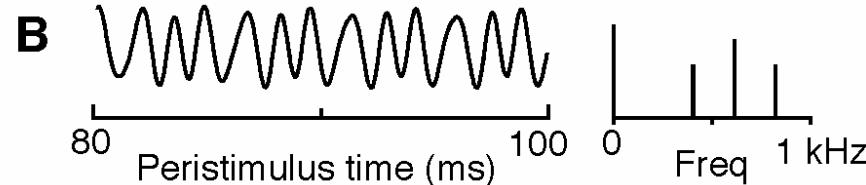


C $F_c=640$ Hz, $F_m=160-320$ Hz

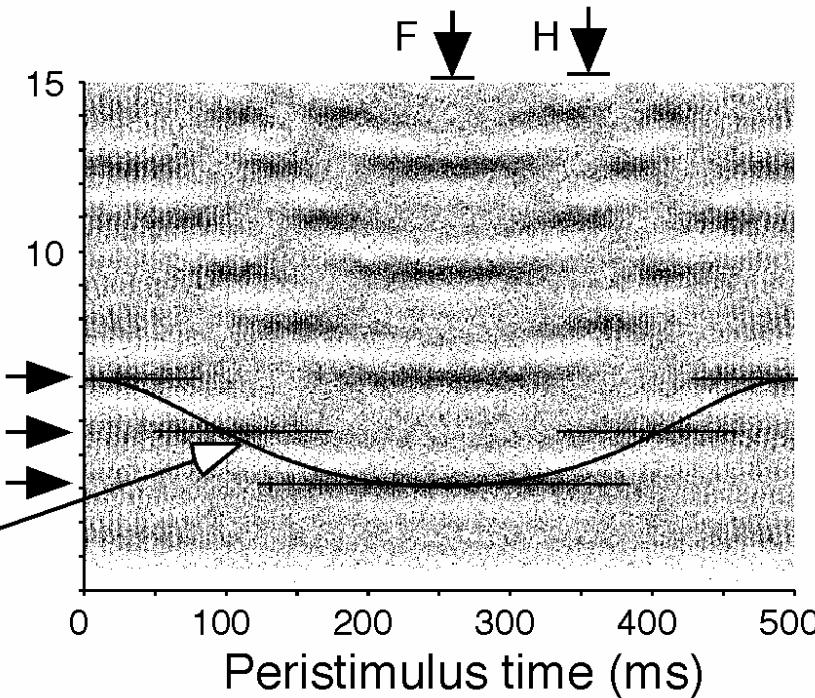


QFM Tone

$F_c=640$ Hz, $F_m=200$ Hz

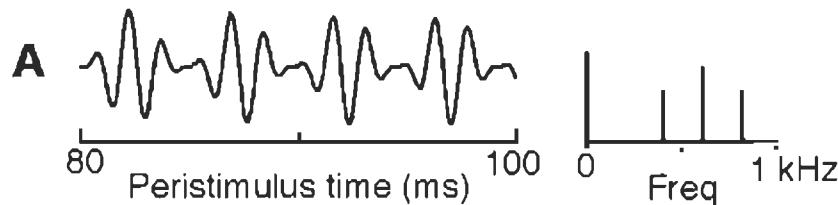


D $F_c=640$ Hz, $F_m=160-320$ Hz

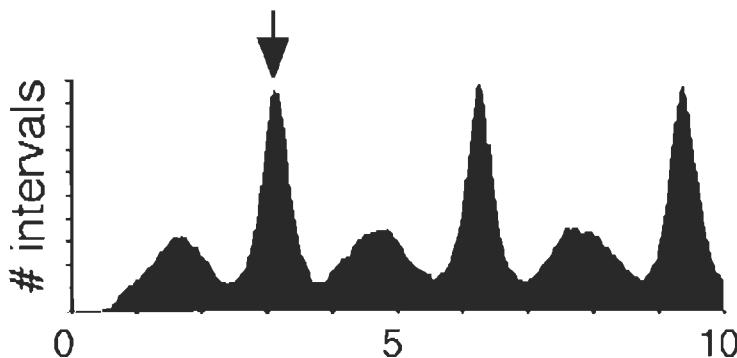


AM Tone

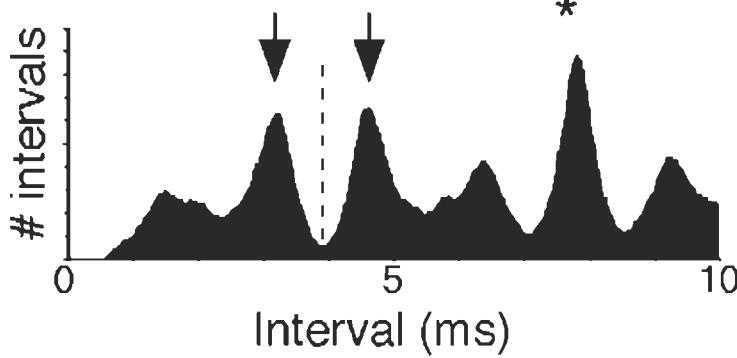
$F_c=640$ Hz, $F_m=200$ Hz



E AM $F_c=640$ Hz, $F_m=320$ Hz
PST=225-275

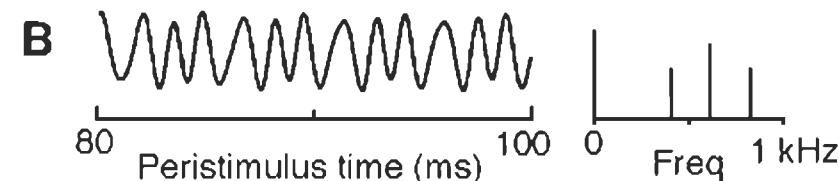


G AM $F_c=640$ Hz, $F_m=256$ Hz
PST=335-385

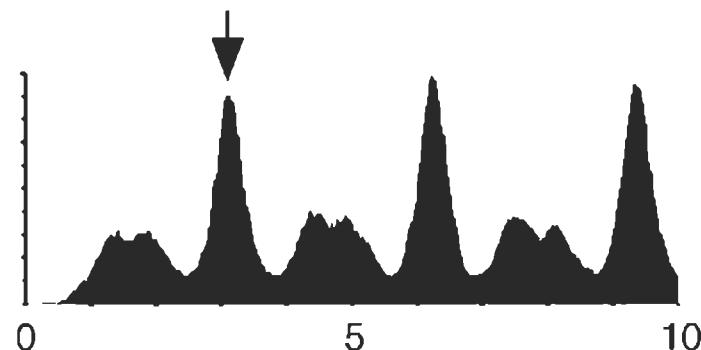


QFM Tone

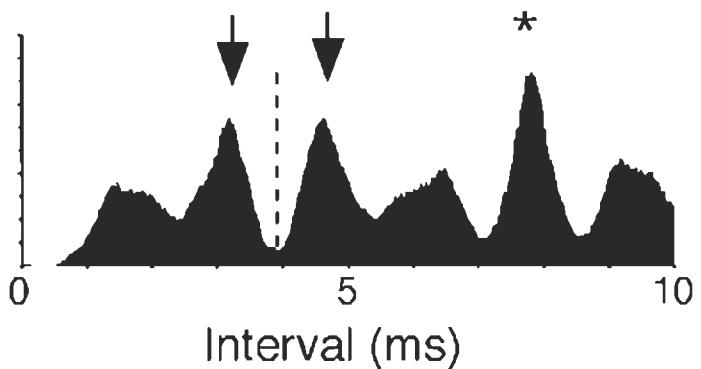
$F_c=640$ Hz, $F_m=200$ Hz



F QFM $F_c=640$ Hz, $F_m=320$ Hz
PST=225-275



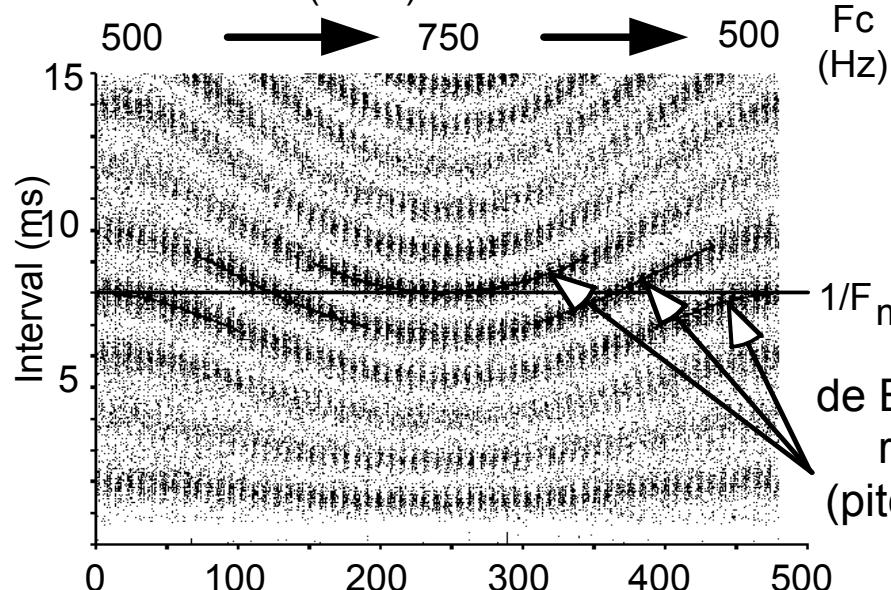
H QFM $F_c=640$ Hz, $F_m=256$ Hz
PST=335-385



Cochlear nucleus IV: Pitch shift

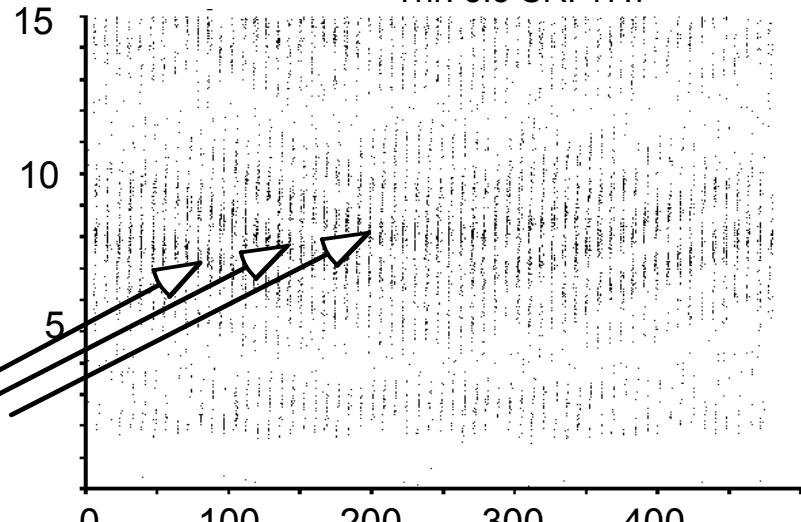
Variable-Fc AM tone $F_m = 125$ Hz $F_c = 500\text{-}750$ Hz **Pitch ~ de Boer's rule**

Pooled ANF (n=47)



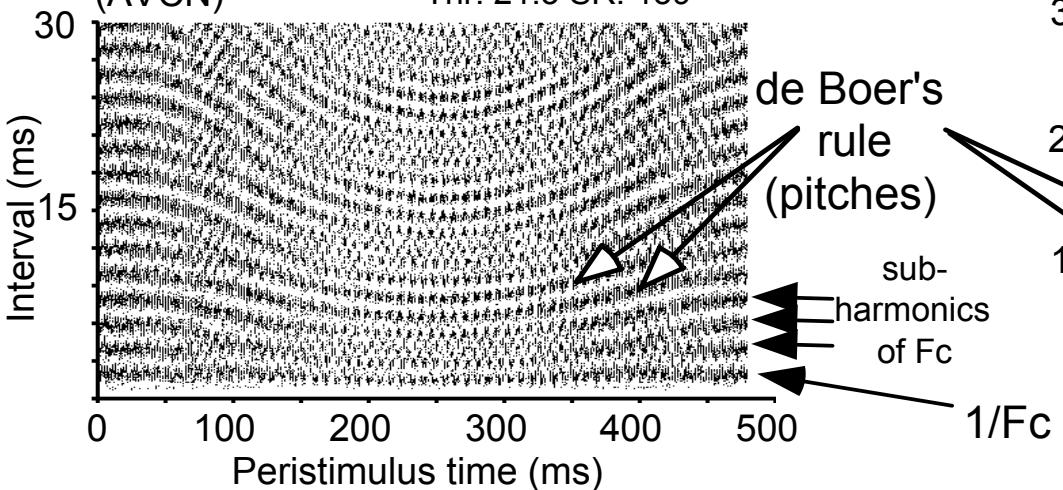
Chop-S (PVCN)

Unit 35-40 CF: 2.1 kHz
Thr: 5.3 SR: 17.7



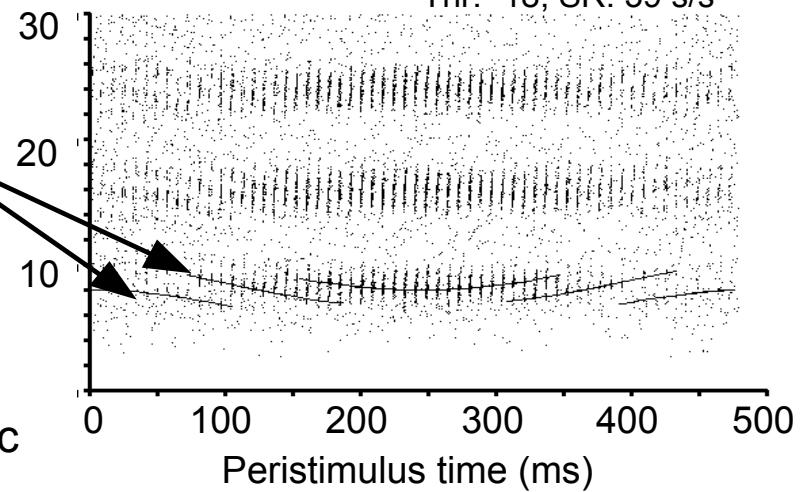
Primarylike (AVCN)

45-17-4 CF: 408 Hz
Thr: 21.3 SR: 159



Pauser (DCN)

45-15-8 CF: 4417
Thr: -18, SR: 39 s/s



Dominance region for pitch (harmonics 3-5 or partials 500-1500 Hz)

$F0_{3-5} = 80$ Hz

Harmonics 3-5 alone

$F0_{3-5} = 160$ Hz

Harmonics 6-12 alone

$F0_{3-5} = 240$ Hz

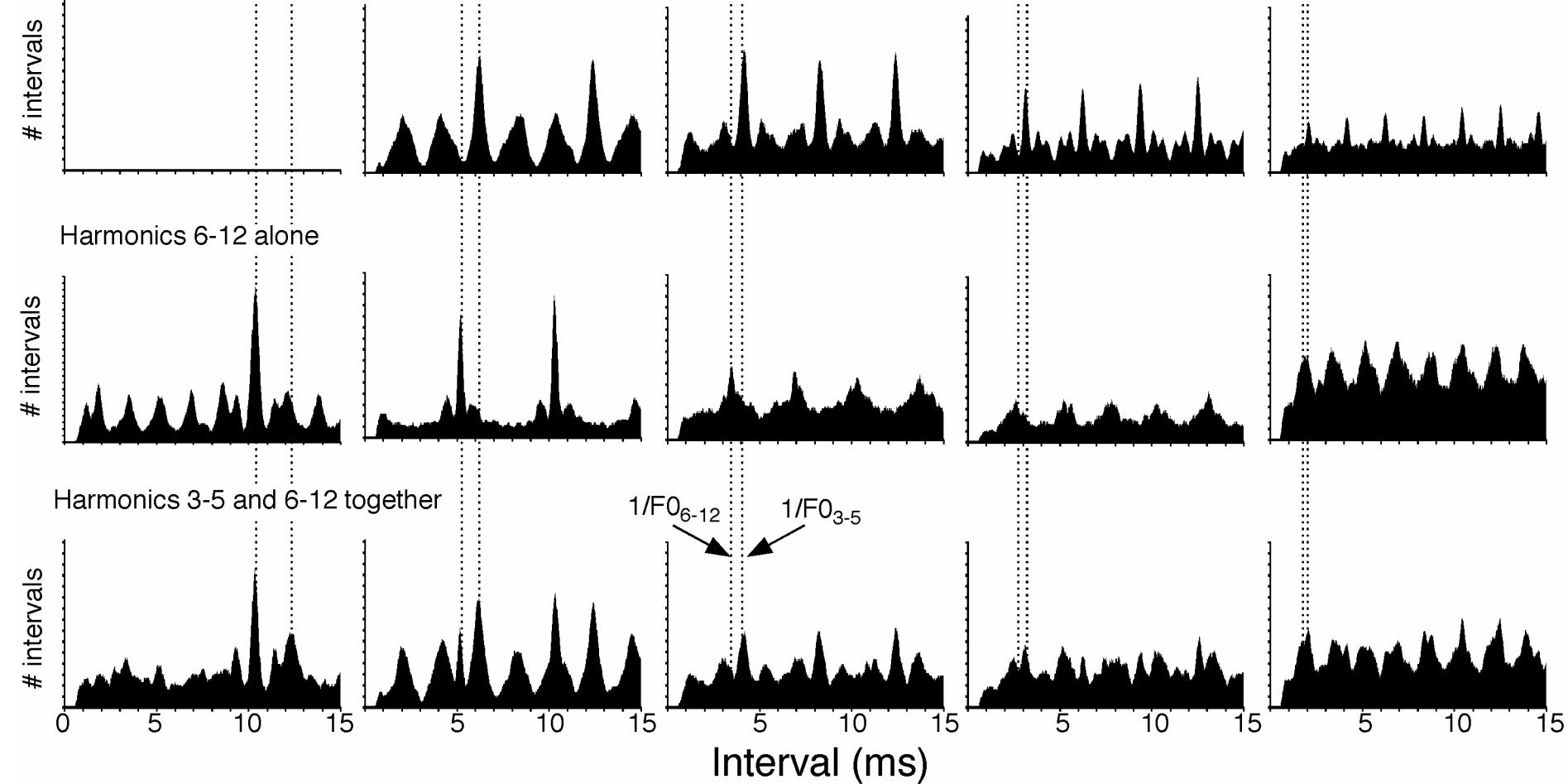
Harmonics 3-5 and 6-12 together

$F0_{3-5} = 320$ Hz

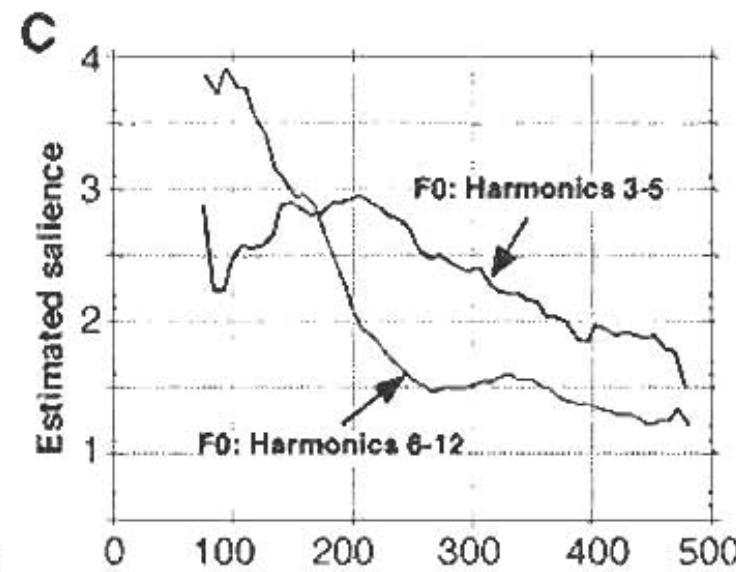
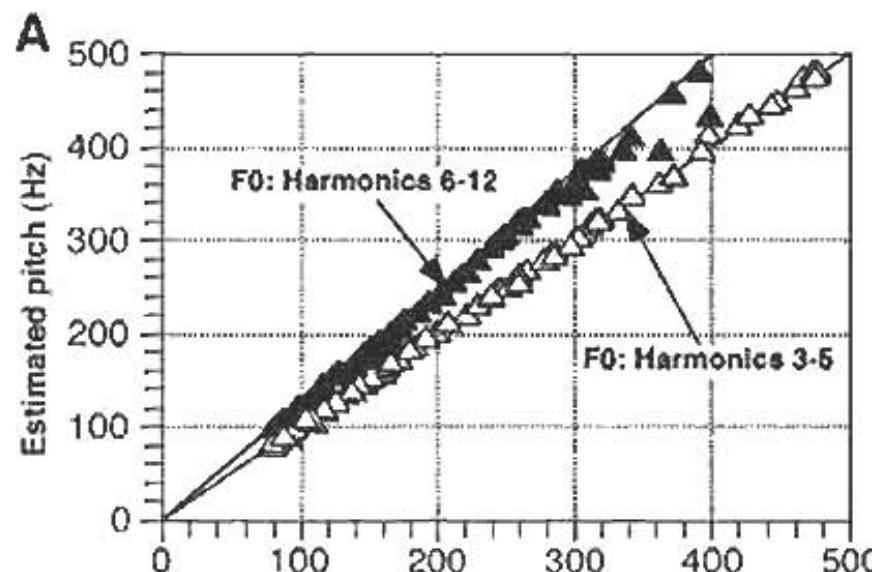
Harmonics 3-5 and 6-12 together

$F0_{3-5} = 480$ Hz

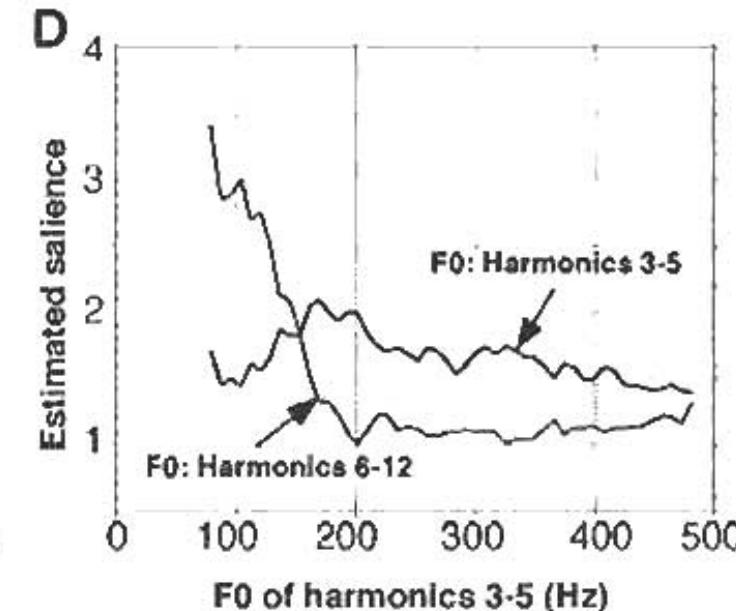
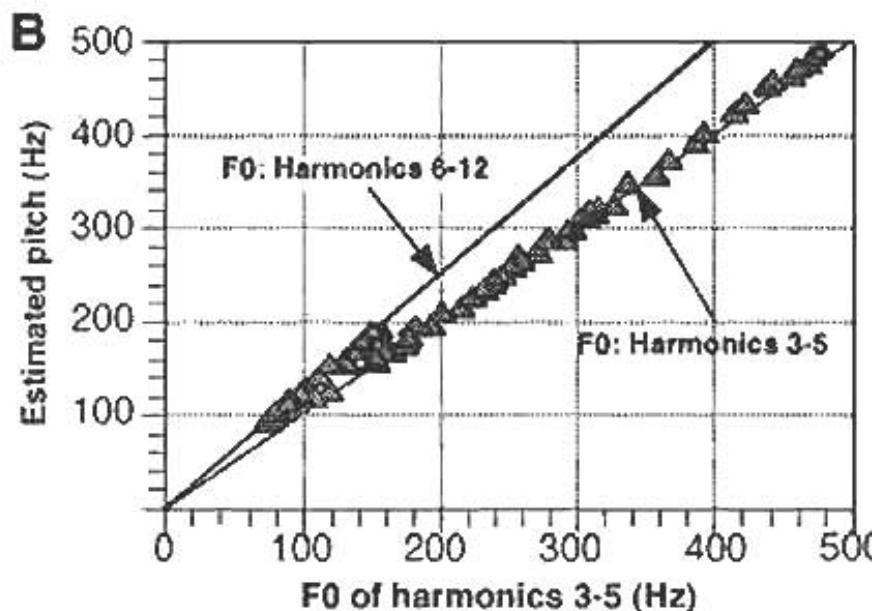
Harmonics 3-5 and 6-12 together



Harmonics 3-5 and 6-12 presented separately



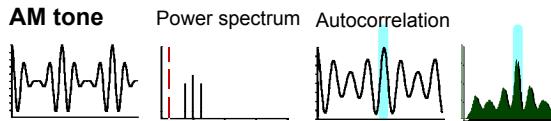
Harmonics 3-5 and 6-12 presented concurrently



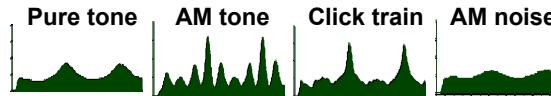
Summary

Population-interval representation of pitch at the level of the auditory nerve

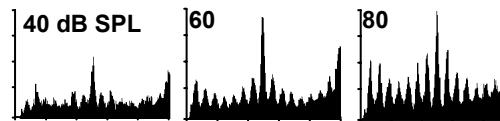
Pitch of the
"missing fundamental"



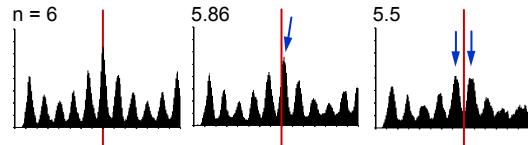
Pitch
Equivalence



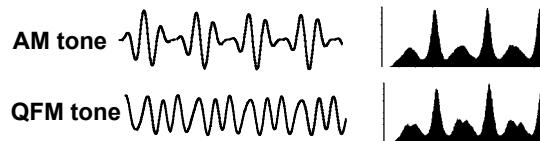
Level invariance



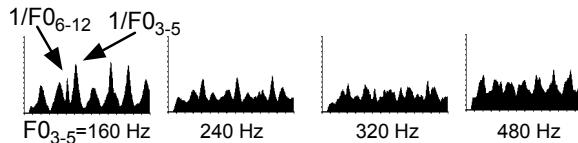
Pitch shift of
inharmonic
AM tones



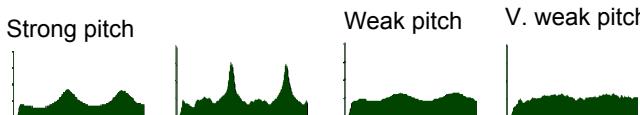
Phase invariance



Dominance region



Pitch salience



Temporal theories - pros & cons

Make use of spike-timing properties of elements in early processing (to midbrain at least)

Interval-information is precise & robust & level-insensitive

No strong neurally-grounded theory of how this information is used

Unified model: account for pitches of perceptually-resolved & unresolved harmonics in an elegant way (dominant periodicity)

Explain well existence region for F0 (albeit with limits on max interval durations)

Do not explain low pitches of unresolved harmonics

Interval analyzers require precise delays & short coincidence windows

Little or no neural evidence for required analyzers

Pitch classes and perceptual similarity

Associationist theory:

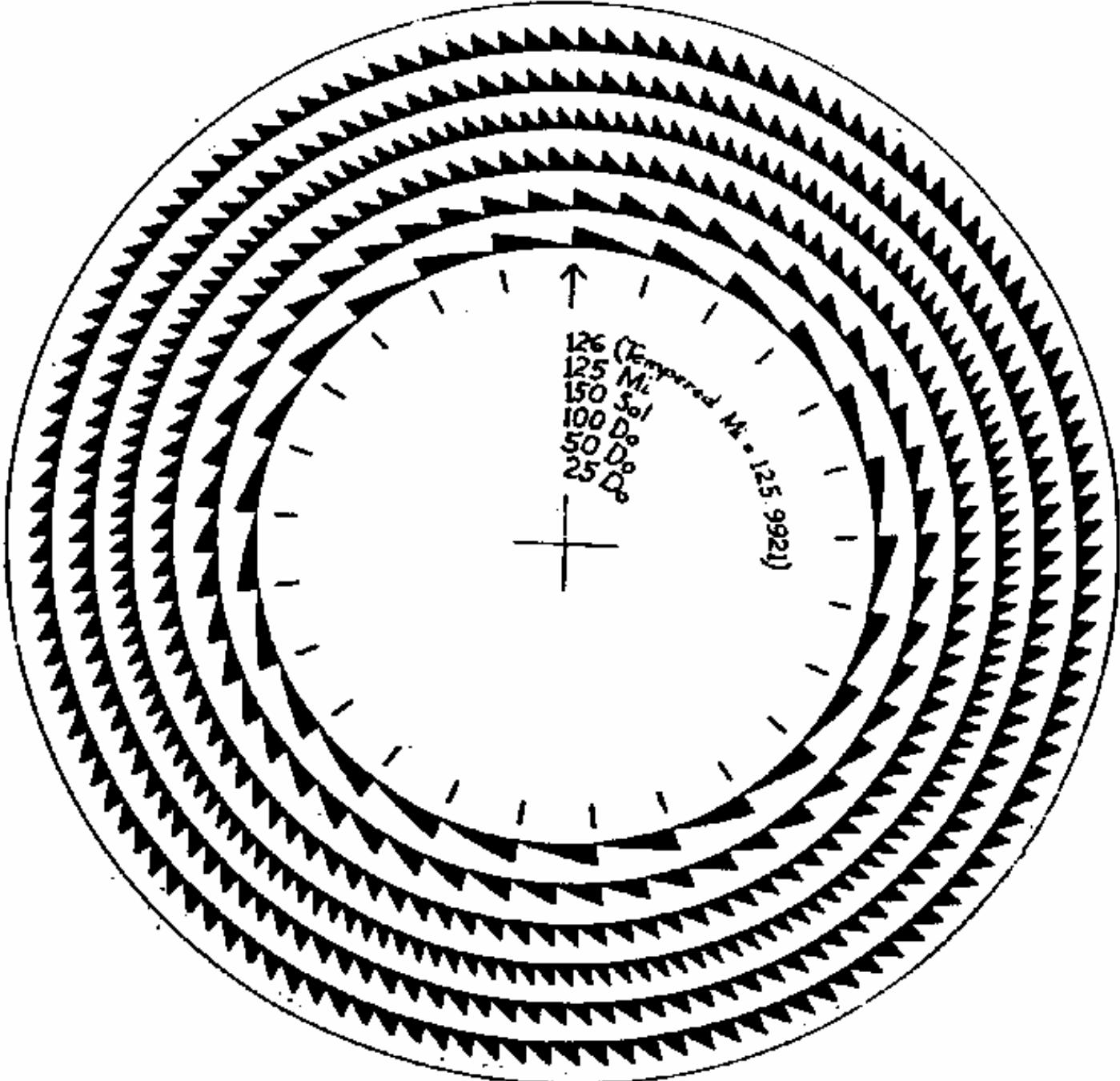
Build up harmonic associations from repeated exposure to harmonic complex tones

Innate mechanisms:

Harmonic similarity relations are direct consequences of the inherent structure of interval codes and the neurocomputational mechanisms used for their analysis

A

Harmony wheel of Boomsliter and Creel (1962)



Temporal coding and music: a few references

- Boomsliter P, Creel W (1962) The long pattern hypothesis in harmony and hearing. *J Music Theory* 5:2-31.
- Cariani P (2002) Temporal codes, timing nets, and music perception. *J New Music Res* 30:107-136.
- McKinney MF, Delgutte B (1999) A possible neurophysiological basis of the octave enlargement effect. *J Acoust Soc Am* 106:2679-2692.
- McKinney MF, Tramo MJ, Delgutte B (2001) Neural correlates of musical dissonance in the inferior colliculus. In: *Physiological and Psychophysical Bases of Auditory Function* (Houtsma AJM, ed), pp 71-77. Maastricht: Shaker Publishing.
- Ohgushi K (1978) On the role of spatial and temporal cues in the perception of the pitch of complex tones. *J Acoust Soc Am* 64:764-771.
- Ohgushi K (1983) The origin of tonality and a possible explanation for the octave enlargement phenomenon. *J Acoust Soc Am* 73:1694-1700.
- Patterson RD (1986) Spiral detection of periodicity and the spiral form of musical scales. *Psychology of Music* 14:44-61.
- Rose JE (1980) Neural correlates of some psychoacoustical experiences. In: *Neural Mechanisms of Behavior* (McFadden D, ed), pp 1-33. New York: Springer Verlag.
- Tramo MJ, Cariani PA, Delgutte B, Braida LD (2001) Neurobiological foundations for the theory of harmony in western tonal music. *Ann N Y Acad Sci* 930:92-116.

INTERVAL DISTRIBUTIONS AND OCTAVE SIMILARITY

Simulated population interval distributions

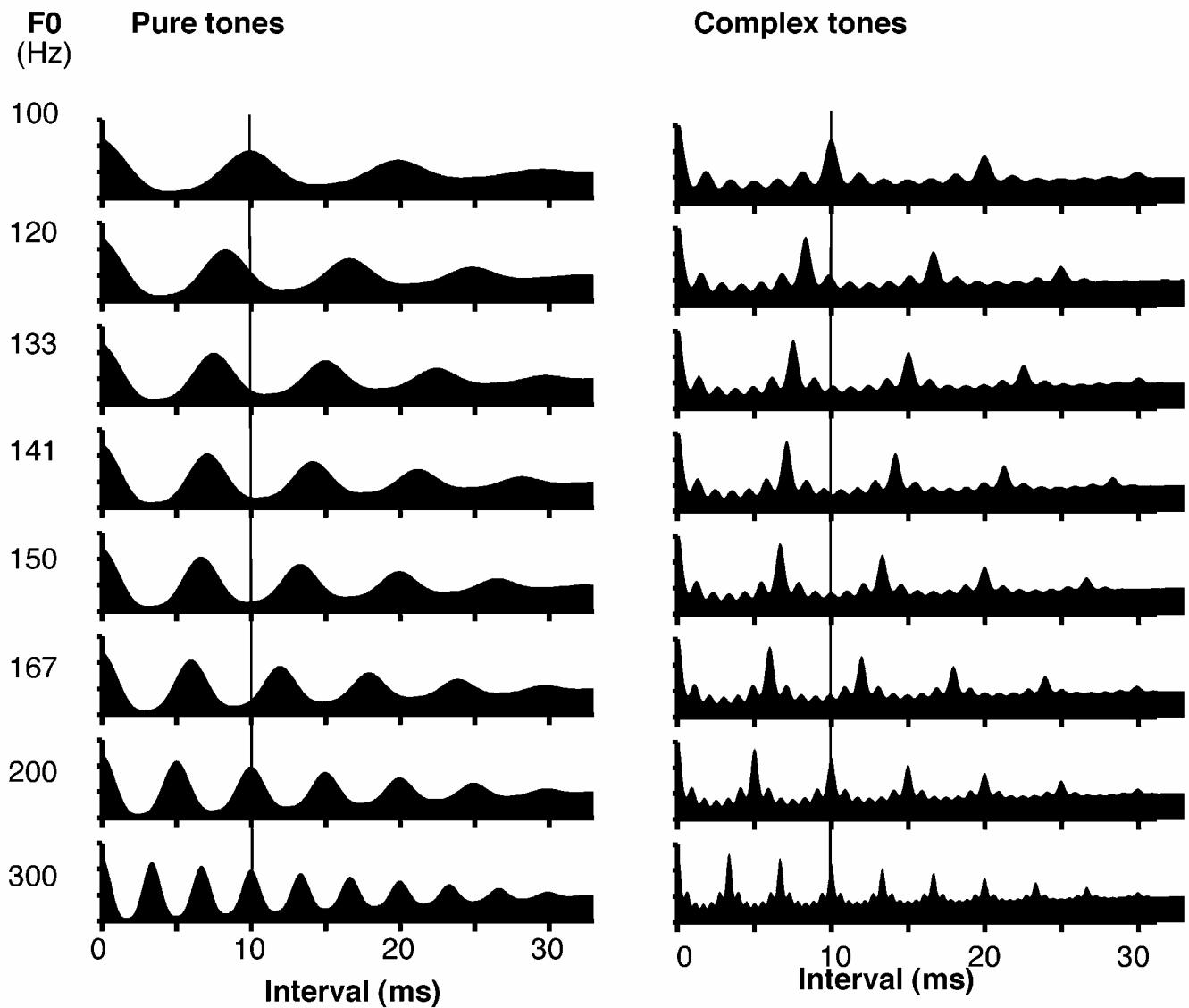


Figure 4. Similarities between population-interval representations associated with different fundamental frequencies. Simulated population-interval distributions for pure tones (left) and complex tones (right) consisting of harmonics 1-6.

Octave similarity

Please see Cariani, Peter. *Journal of New Music Research* 30, no. 2 (2001): 107-135.

Pitch height and pitch chroma

Please see Figure 1, 2, and 7 in Roger N. Shepard. "Geometrical Approximations to the Structure of Musical Pitch." *Psychological Review* 89, no. 4 (1982): 305-322.

Please see Cariani, Peter. *Journal of New Music Research* 30, no. 2 (2001): 107-135.

Musical tonal relations

Please see Shepard, Roger N. "Geometrical Approximations to the Structure of Musical Pitch." *Psychological Review* 89, no. 4 (1982): 305-322.

Please see Cariani, Peter. *Journal of New Music Research* 30, no. 2 (2001): 107-135.

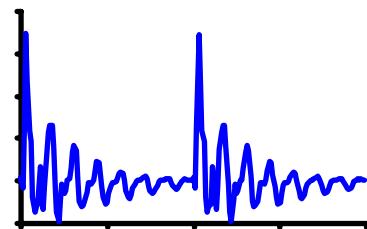
Note-chord relations (Harms 1-12)

Please see Cariani, Peter. *Journal of New Music Research* 30, no. 2 (2001): 107-135.

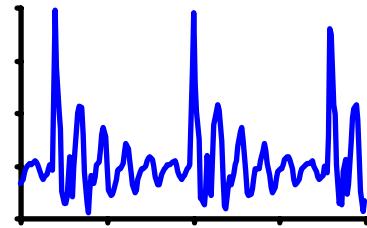
Vowel properties

Waveforms

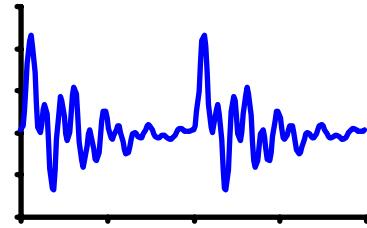
[ae]
 $F_0 = 100 \text{ Hz}$



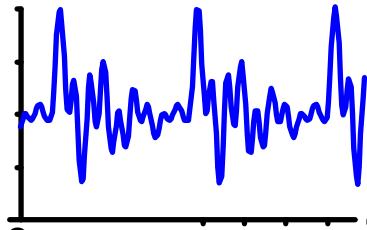
[ae]
 $F_0 = 125 \text{ Hz}$



[er]
 $F_0 = 100 \text{ Hz}$

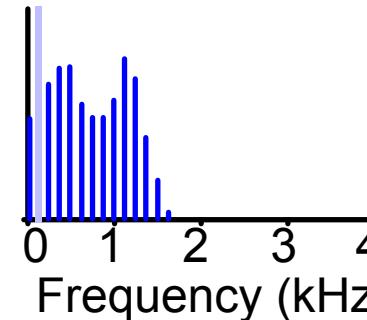
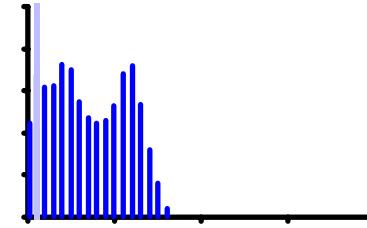
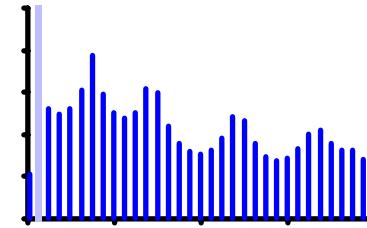
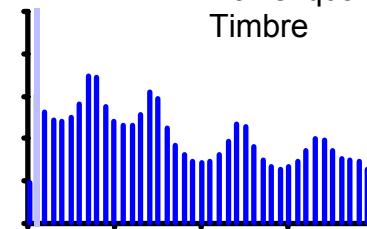


[er]
 $F_0 = 125 \text{ Hz}$



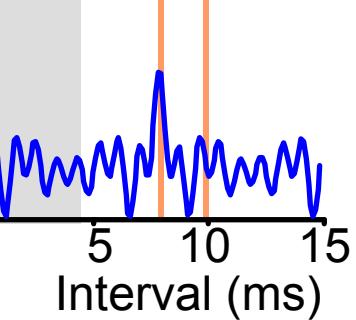
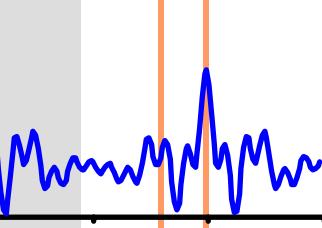
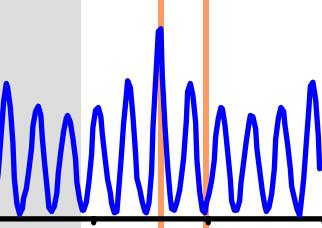
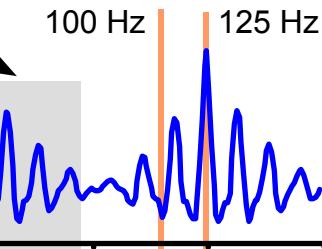
Power Spectra

Formant-related
Vowel quality
Timbre



Autocorrelations

Pitch periods, $1/F_0$

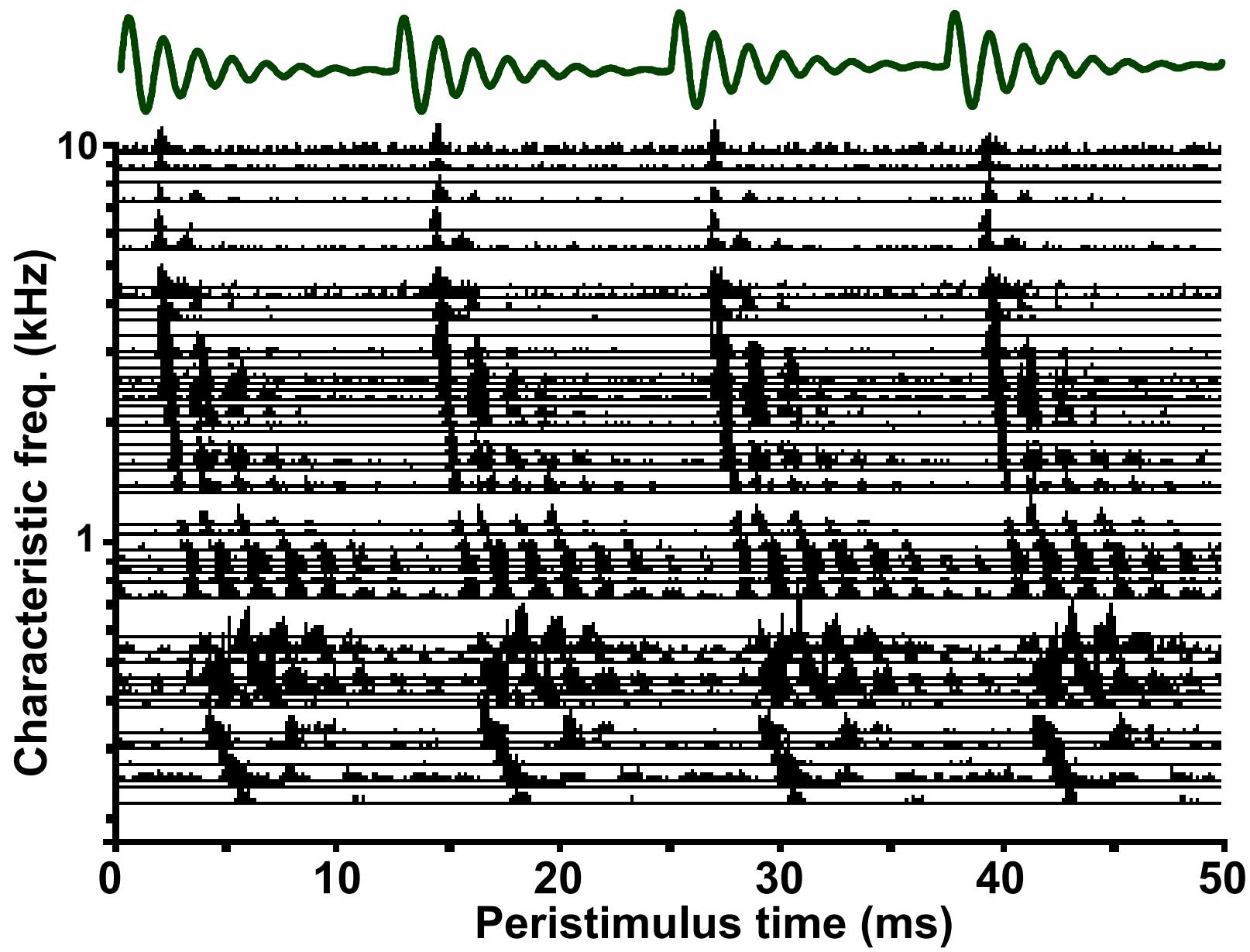


Time (ms)

Frequency (kHz)

Interval (ms)

Phase-locking of discharges in the auditory nerve



Time domain analysis of auditory-nerve fiber firing rates.
Hugh Secker-Walker & Campbell Searle, J. Acoust. Soc. 88(3), 1990
Neural responses to /da/ @ 69 dB SPL from Miller and Sachs (1983)

Reprinted with permission from Secker-Walker HE, Searle CL. 1990. Time-domain analysis of auditory-nerve-fiber firing rates. J. Acoust. Soc. Am. 88 (3): 1427-36. Copyright 1990, Acoustical Society of America. Used with permission.

Low CFs

0.14

0.35

0.44

0.58

0.70

1.00

1.20

1.48

1.67

1.80

2.08

2.32

2.55

2.95

3.64

4.12

7.52

F1

F2

F3

High CFs

10

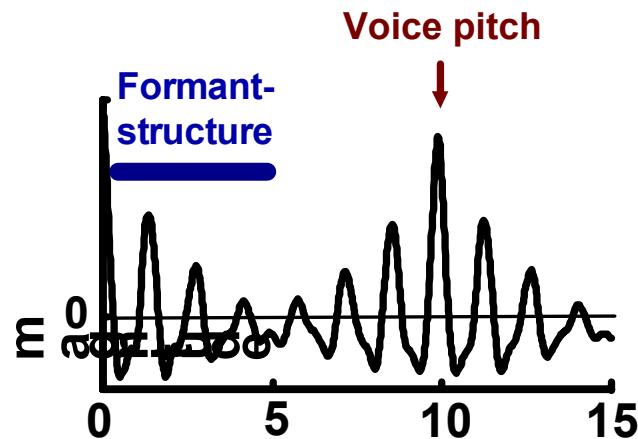
Peristimulus time (ms)

30

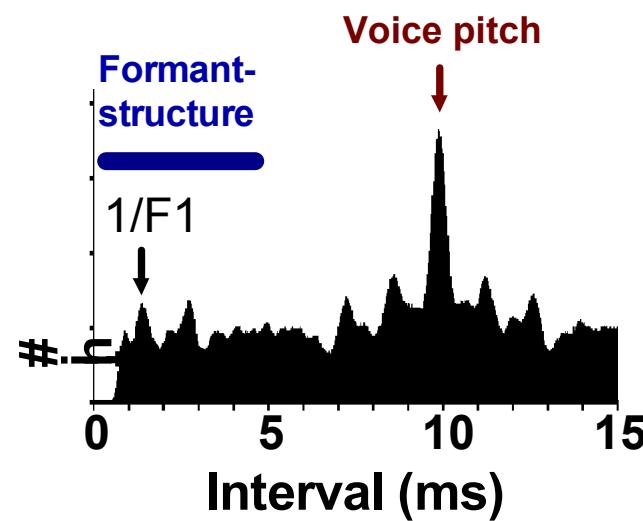
Vowels

Population-interval coding of timbre (vowel formant structure)

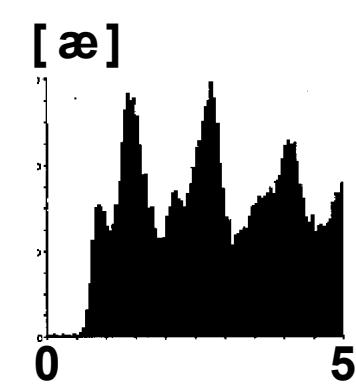
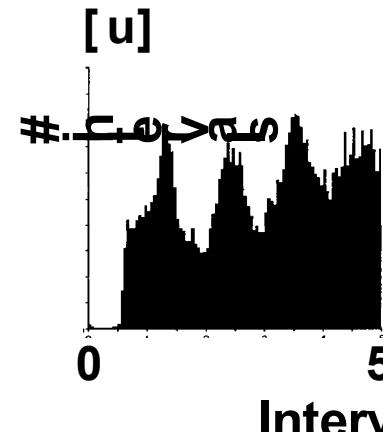
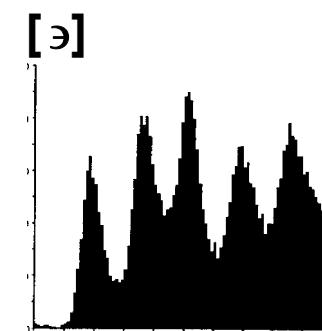
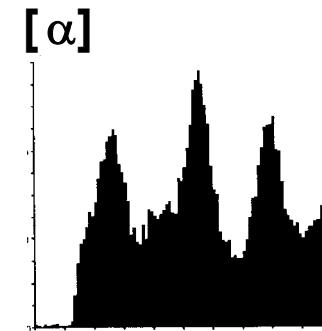
Signal autocorrelation [ae]



Population interval histogram

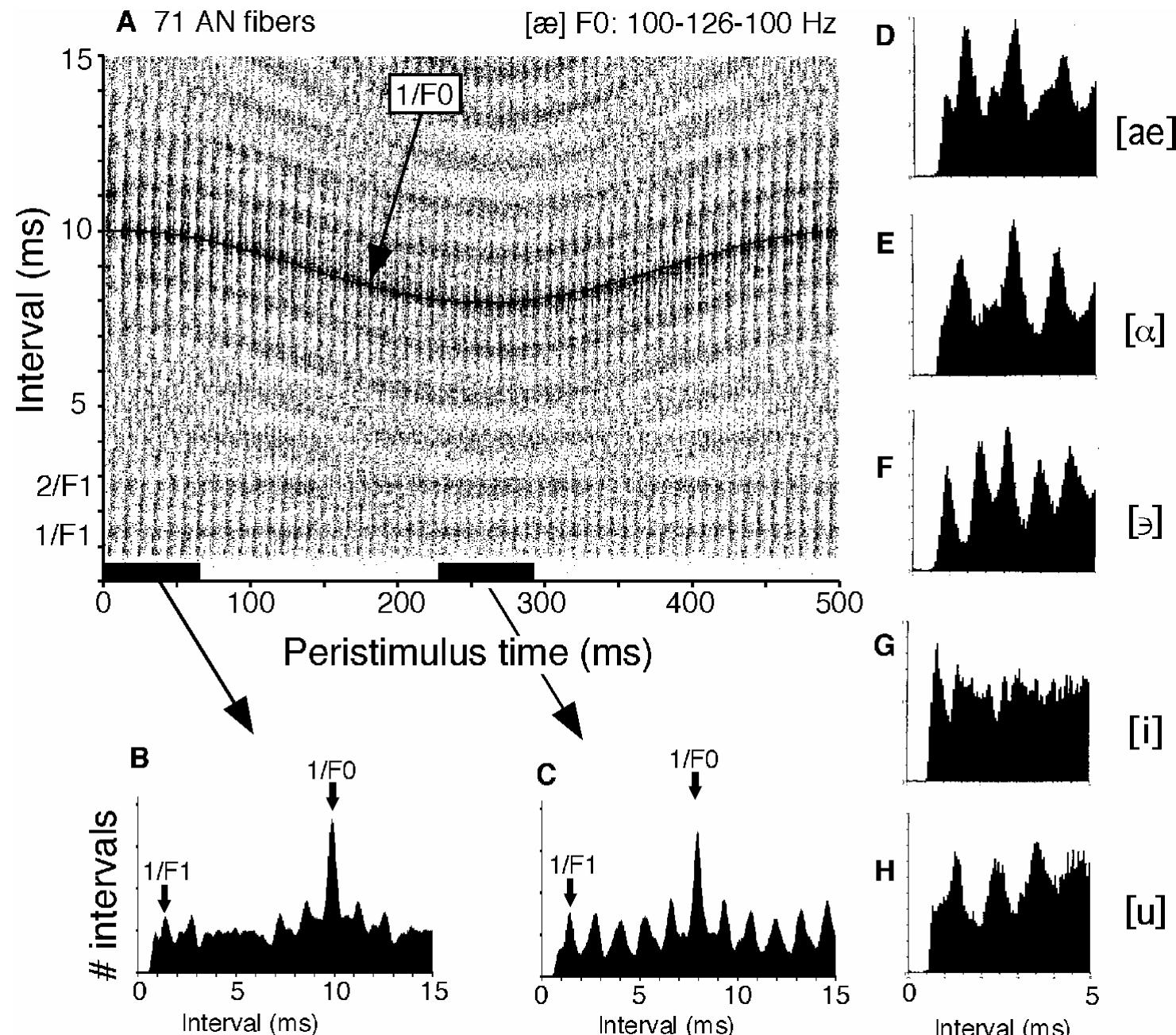


Population-wide distributions of short intervals for 4 vowels



Coding of vowel quality (timbre)

(Reprinted with permission from Secker-Walker HE, Searle CL. 1990. Time-domain analysis of auditory-nerve-fiber firing rates. J. Acoust. Soc. Am. 88 (3): 1427-36. Copyright 1990, Acoustical Society of America.)



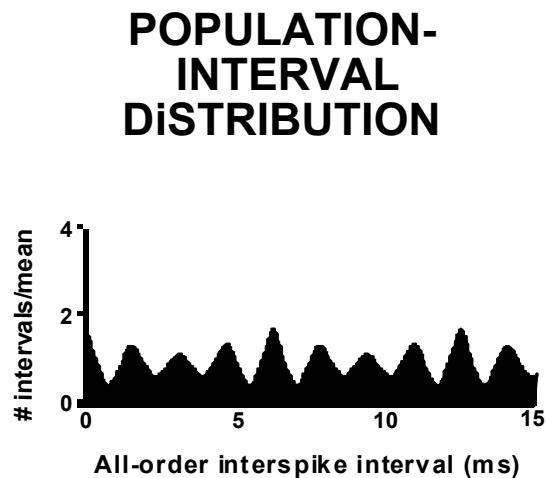
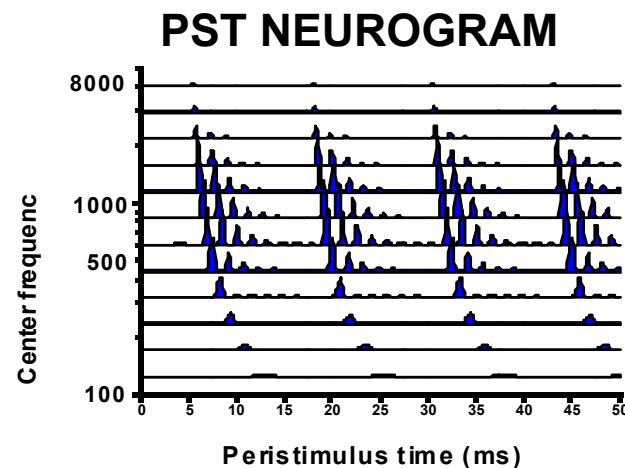
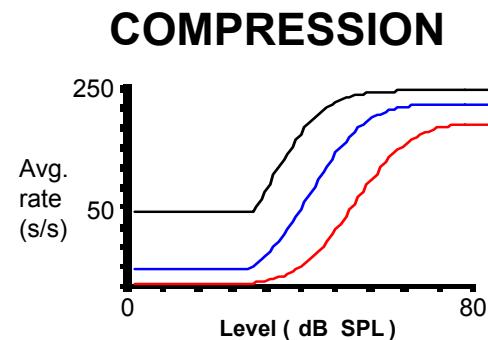
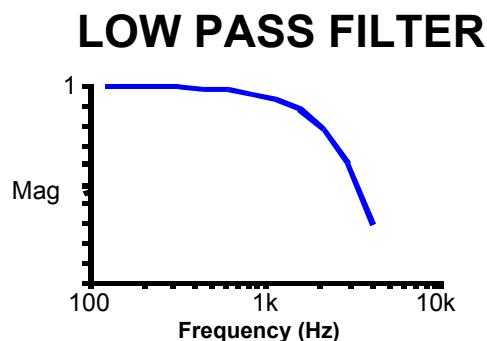
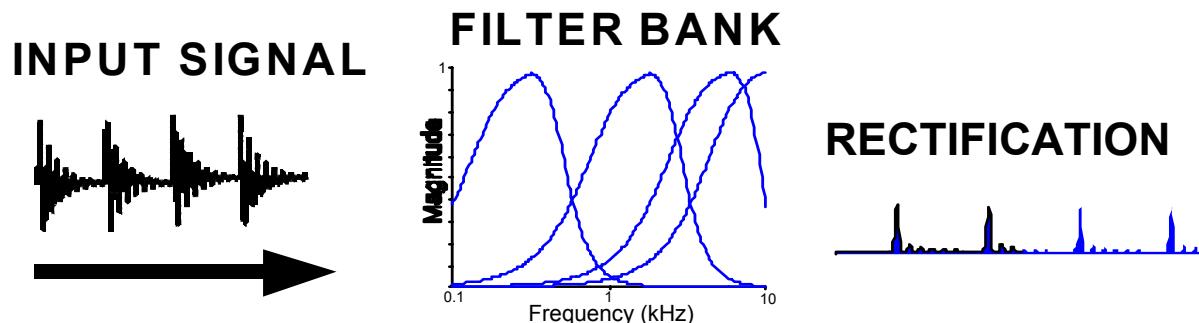
Population-interval representations thus far

- **Periodicity:** capable of accounting for a wide range of pure and complex tone pitches that are heard; salience model less developed.
- **Spectrum:** capable of robustly representing the stimulus spectrum below 4 kHz (our neural data, Palmer neural data Meddis simulations)
- **Sound separation:** multiple interval bands apparent for $\Delta F_0 > 10\%$

Auditory nerve model

Basic features:

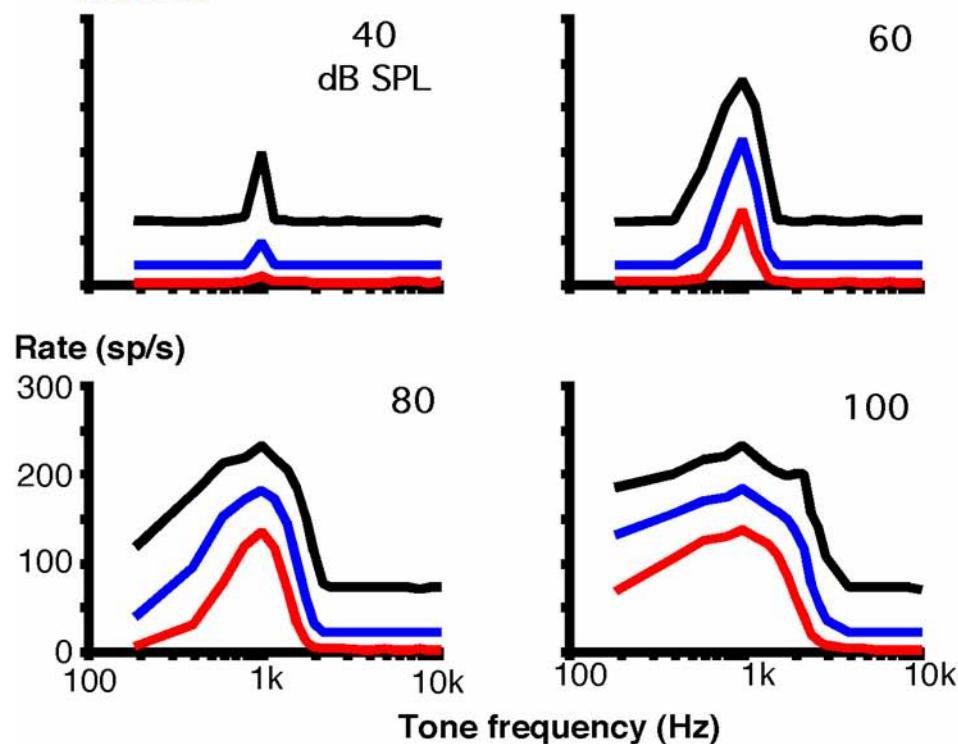
Gammatone filters
Adaptive gain control
Mapping of output
amplitude to firing rates
Spontaneous activity



Please see Figure 60 and 61 in Keidel, D., Wolf., S. Kallert, M. Korth. *The physiological basis of hearing: a review*. Edited by Larry Humes. New York: Thieme-Stratton, Stuttgart, George ThiemeVerlag, 1983. ISBN: 0865770727.

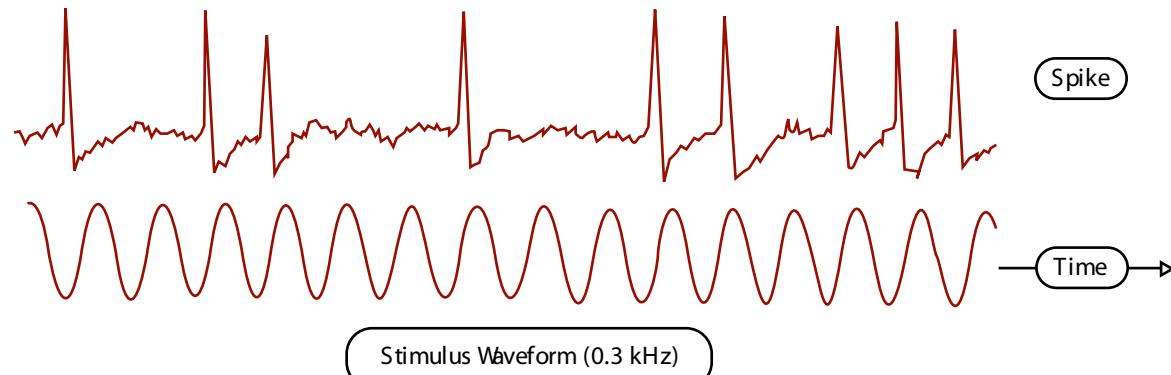
Isointensity profiles for a single ANF

MODEL



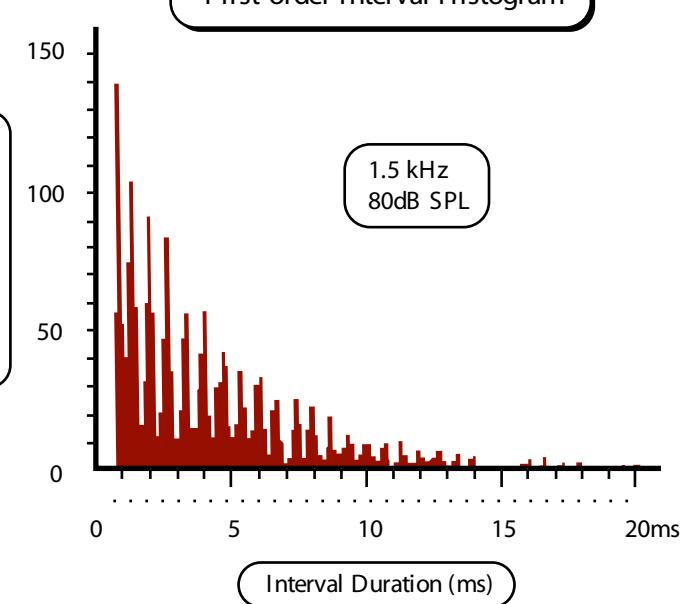
Please see Hind, J. E., D. J. Anderson, J. F. Brugge, and J. E. Rose. "Coding of Information Pertaining to Paired Low-frequency Tones in Single Auditory Nerve Fibers of the Squirrel Monkey." *J Neurophysiol* 30 (July, 1967): 794 -816.

Phase-locking to a 300 Hz Pure Tone

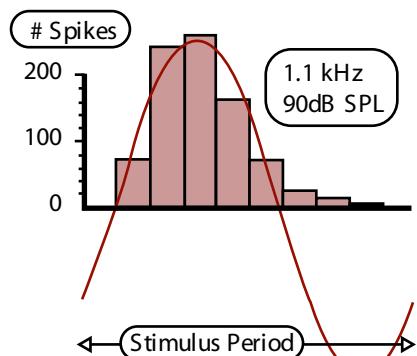


Stimulus Waveform (0.3 kHz)

First-order Interval Histogram



Period Histogram
(1100 Hz)

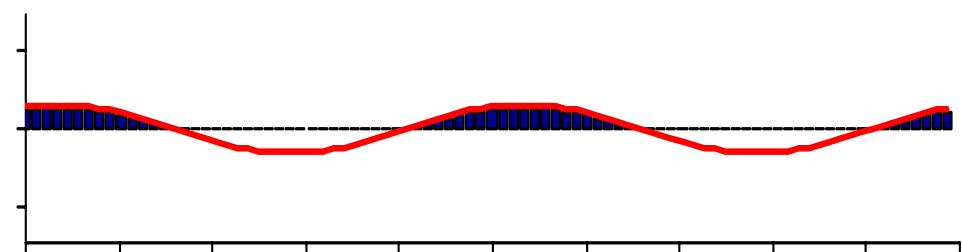


Temporal discharge patterns 40-90 dB SPL

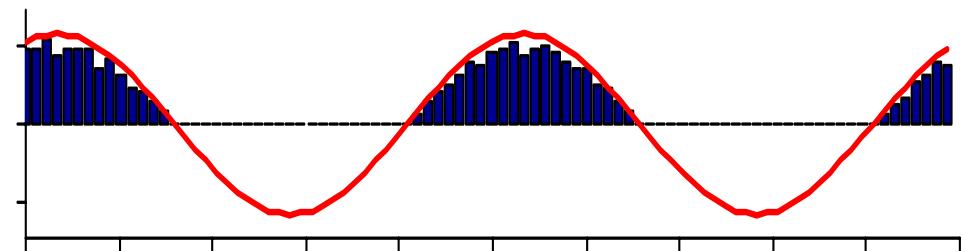
Please see Rose, Hind, Anderson, Brugge J. Neurophysiology 34 (1971): 685-99. Reprinted in D. Keidel, Wolf., S. Kallert, and M. Korth. *The Physiological Basis of Hearing: A Review*. Edited by Larry Humes. New York: Thieme-Stratton, Stuttgart, George ThiemeVerlag, 1983. ISBN: 0865770727.

Medium-SR model fiber

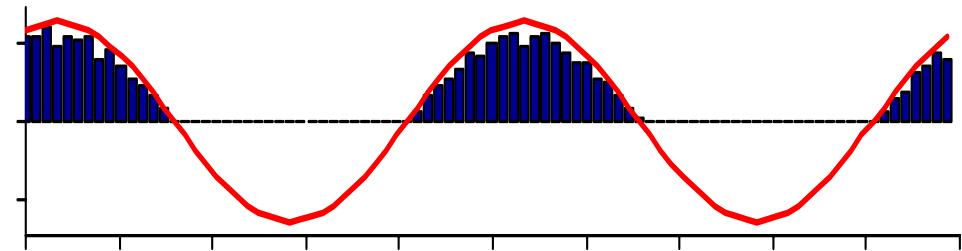
30 dB SPL



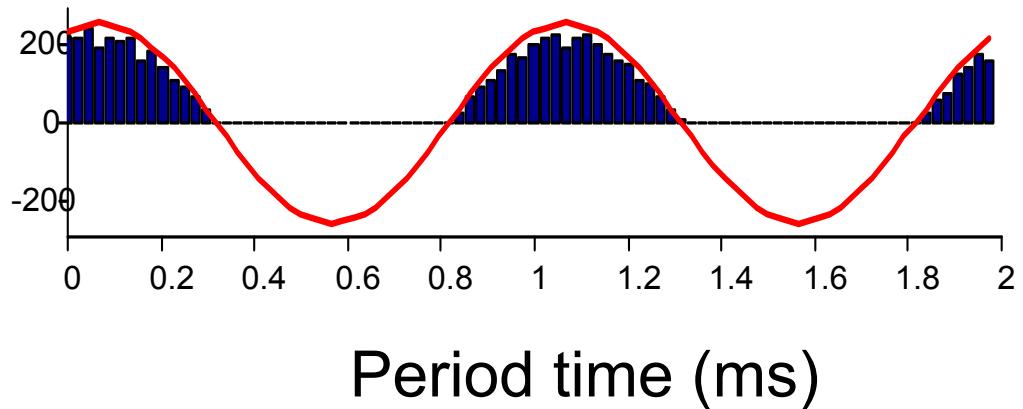
50 dB SPL



70 dB SPL



90 dB SPL



Temporal discharge patterns to pure tone dyad: 600 & 800 Hz

Please see Rose, Hind, Anderson, Brugge J. Neurophysiology 34 (1971): 685-99. Reprinted in D. Keidel, Wolf., S. Kallert, and M. Korth. *The Physiological Basis of Hearing: A Review*. Edited by Larry Humes. New York: Thieme-Stratton, Stuttgart, George ThiemeVerlag, 1983. ISBN: 0865770727.

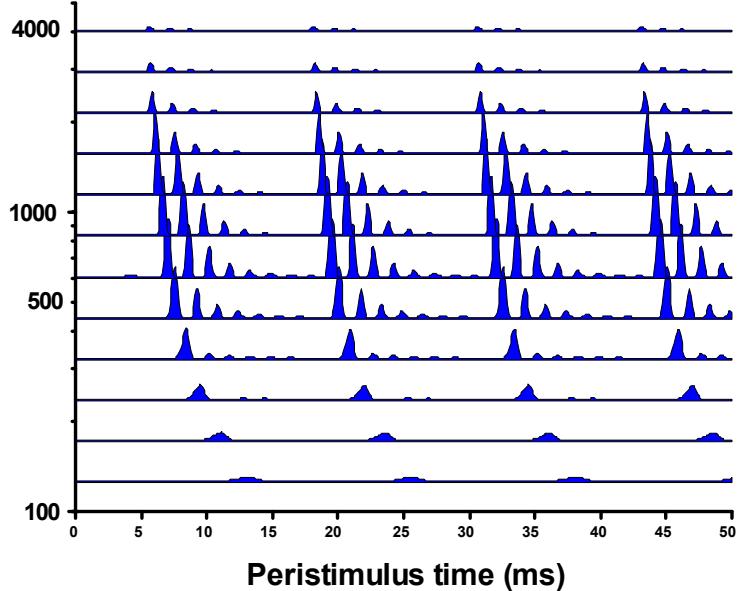
Physiological vs. simulated activity

Simulated vs. observed auditory nerve activity

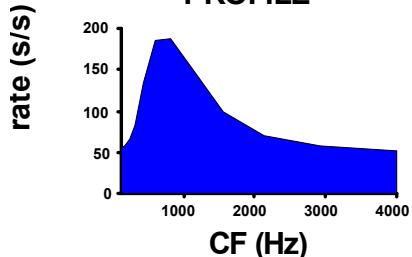
We are mainly interested here in replicating basic phase-locked time-frequency patterns.

SIMULATED RESPONSES

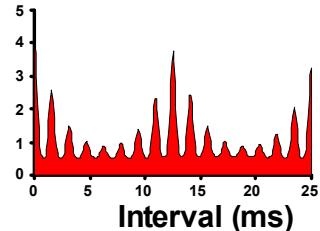
PST NEUROGRAM



RATE-PLACE PROFILE



POPULATION-INTERVAL DISTRIBUTION

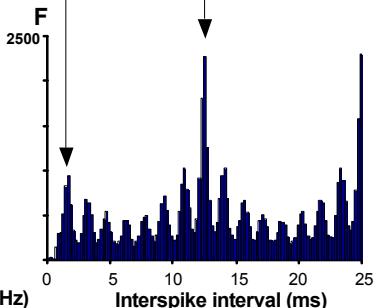
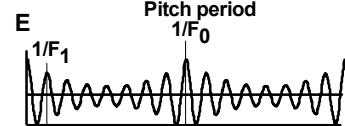
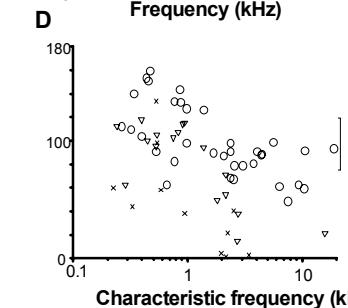
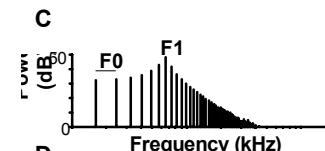
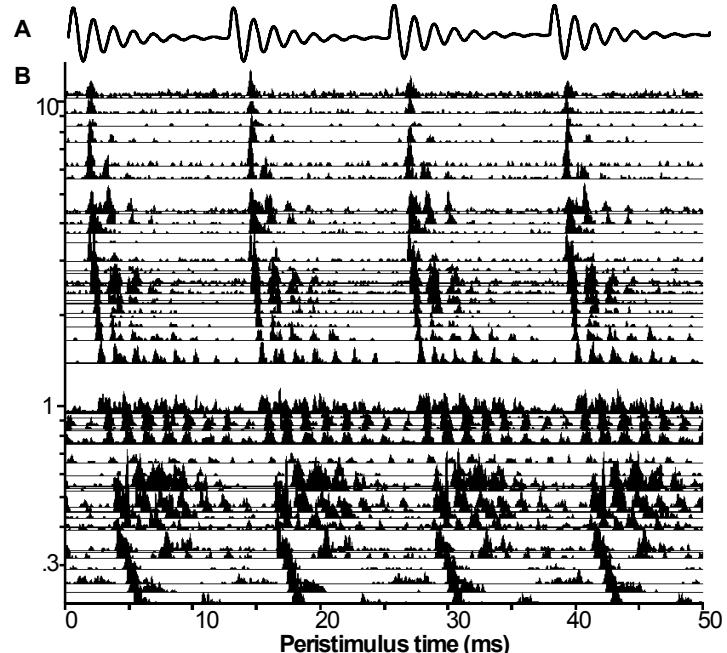


CAT
DATA

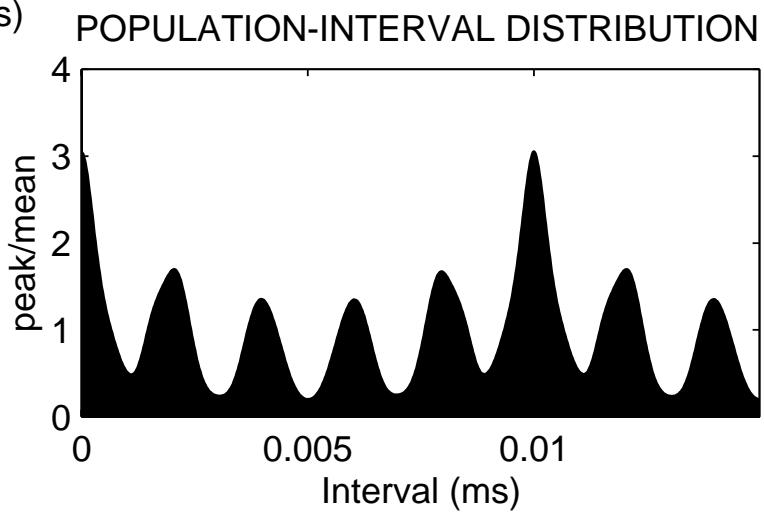
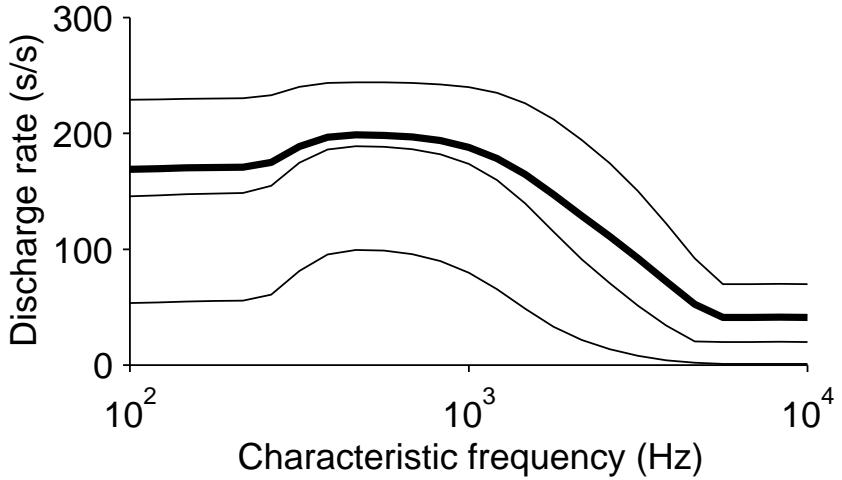
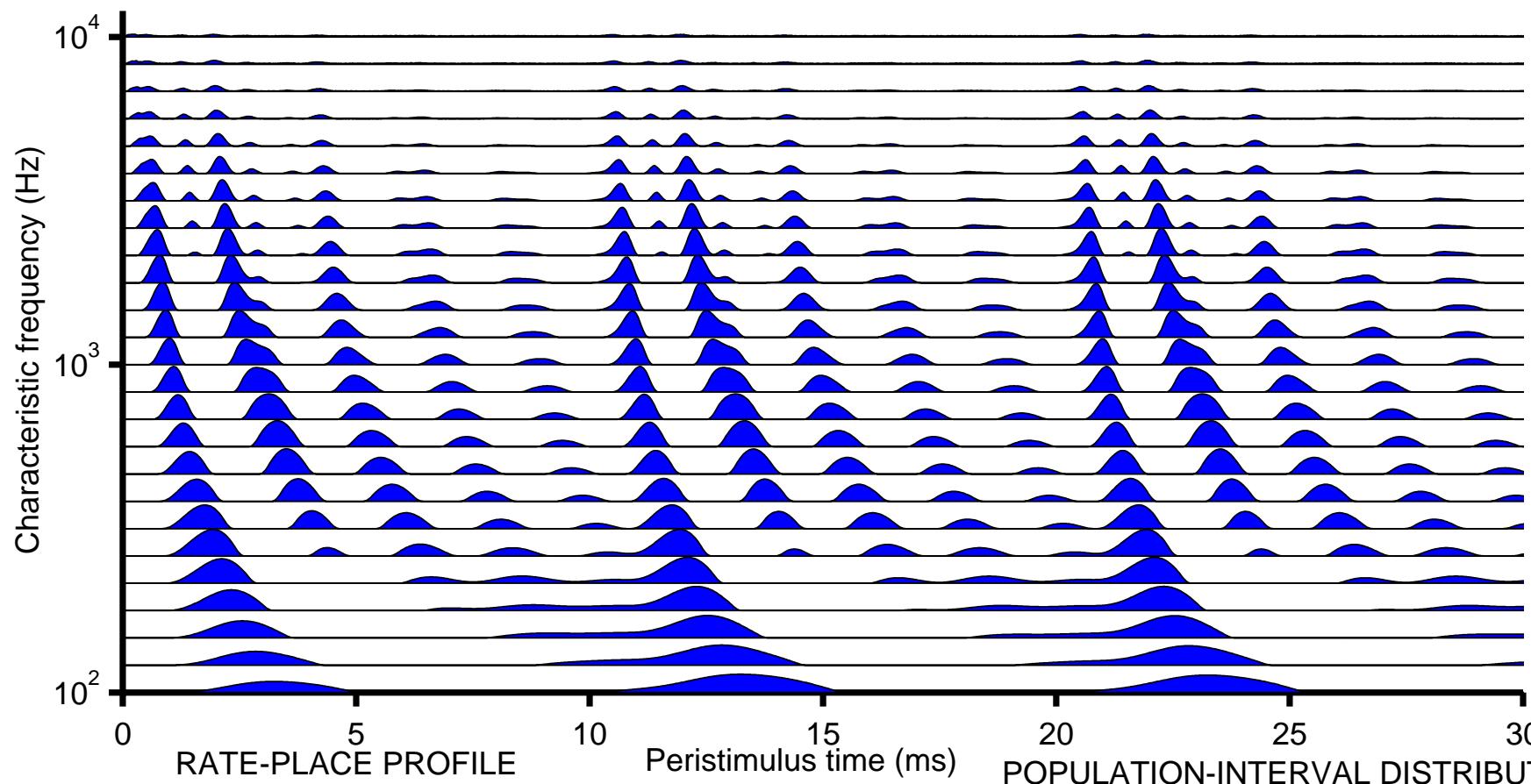
Single-formant vowel

F₀ = 80 Hz
F₁ = 640 Hz
60 dB SPL

1/F₀ = Pitch period



NEUROGRAM



Competition between sounds

Competition between respective interval patterns

Towards interval-based accounts of interacting sounds:

- mutual masking
- pitch multiplicity
- roots of dyads and triads (chords)

Pitch masking

What degree of temporal correlation is necessary for pitch to become audible?

Variable-F0 click train in broad-band noise

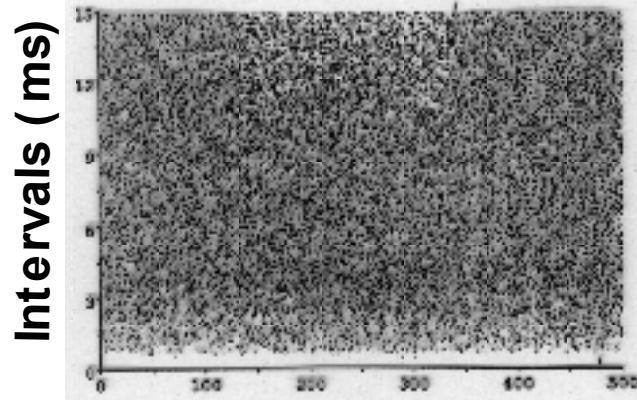
Click train: F0 = 160-320 Hz, positive polarity, 80 dB SPL

Peak/background ratio: intervals @ $1/F_0 \pm 150$ usec/mean over all intervals

Informal pitch thresholds, s/n dB: 12 (MT), 13 (PC), 16 (BD)

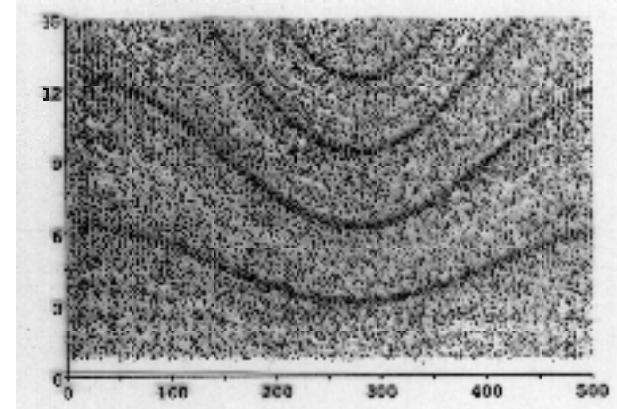
Below threshold (no pitch)

s/n: 8 dB p/b = 1.11



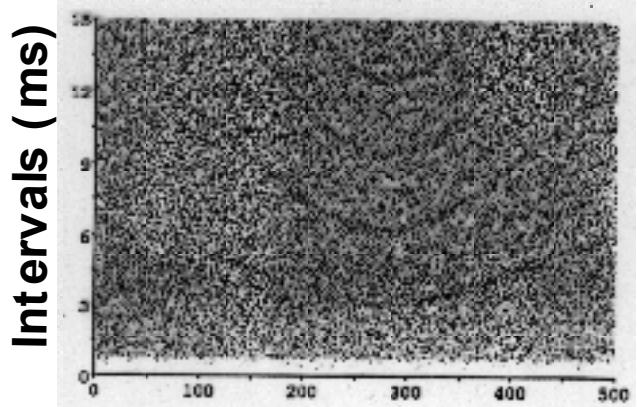
Above threshold (strong pitch)

s/n: 32 dB p/b = 2.38



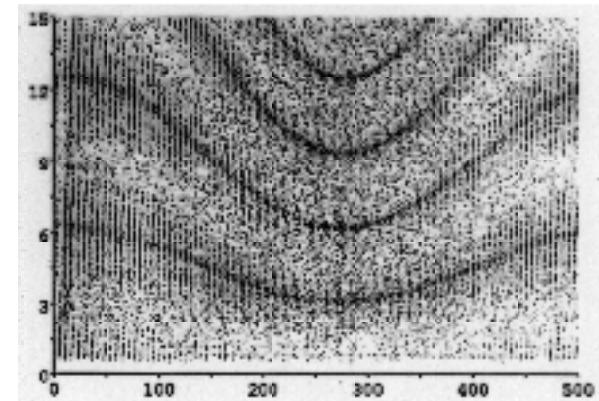
Above threshold (weak pitch)

s/n: 20 dB p/b = 1.57



Above threshold (strong pitch)

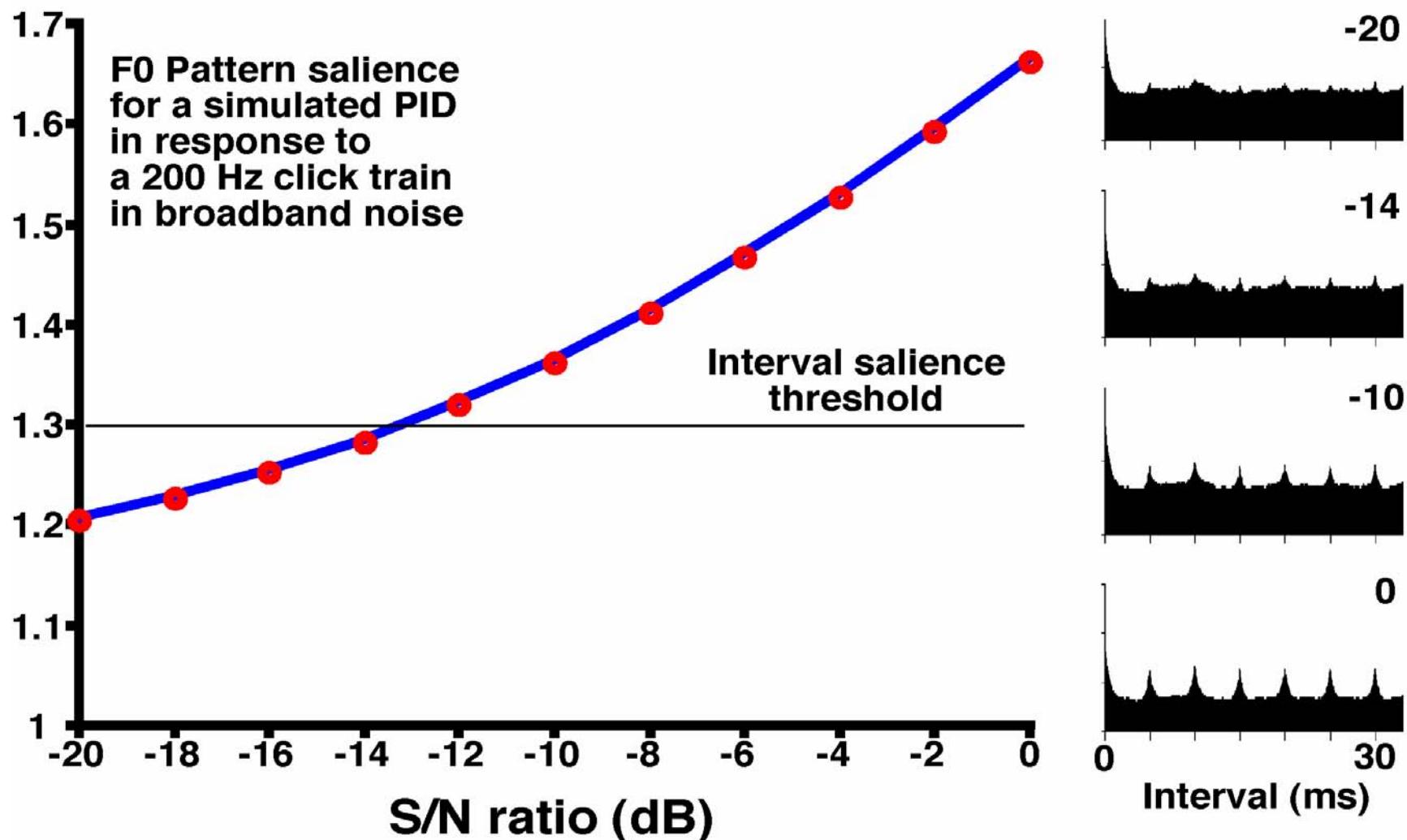
s/n: no noise p/b = 3.26



Peristimulus time (ms)

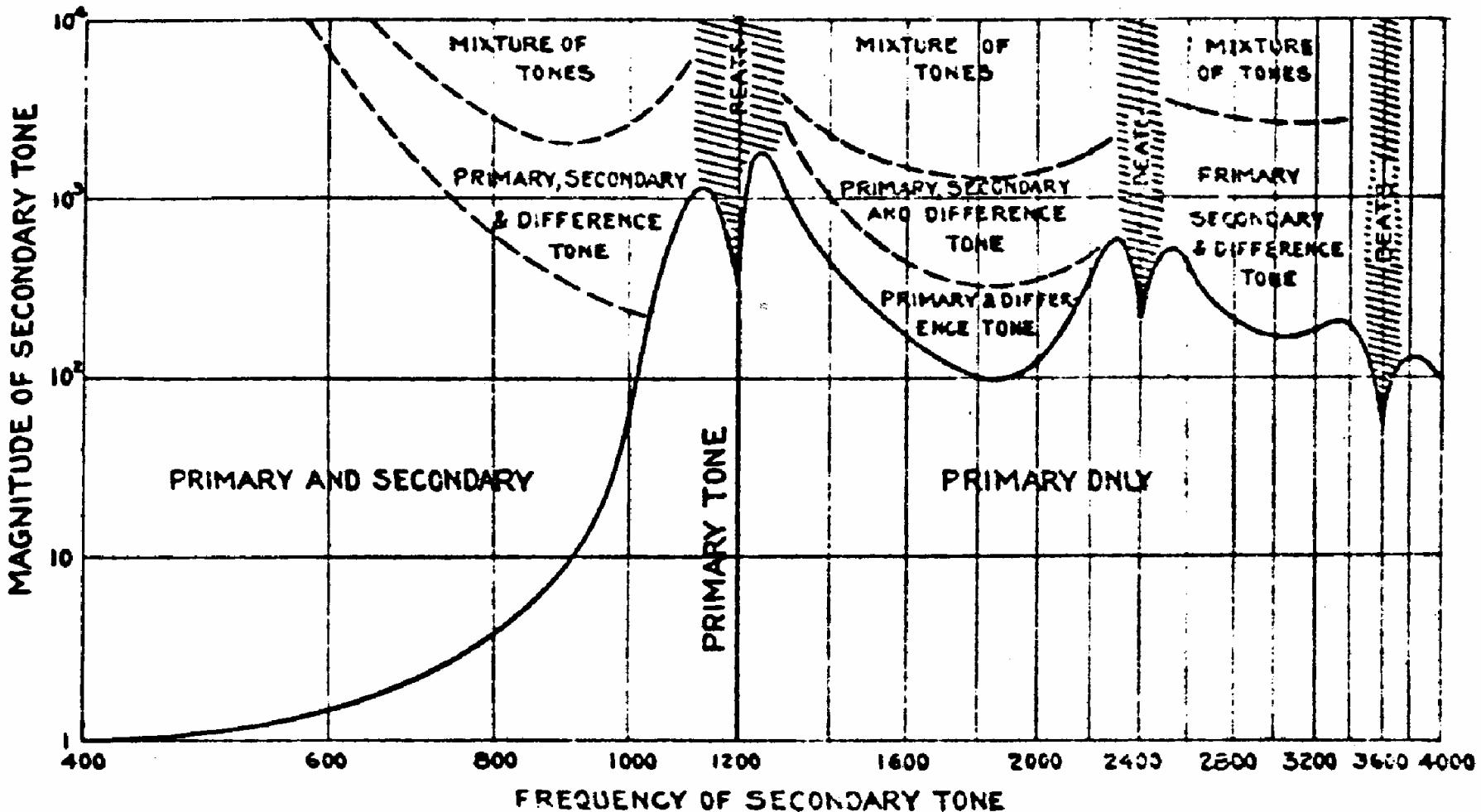
Masking of a click train in broadband noise

AN SIMULATION



Click trains were presented in white noise (0-11 kHz, 70 dB total SPL) at different s/n ratios.
My threshold for pitch detection with low-isolation headphones is -6dB.

Masking audiograms

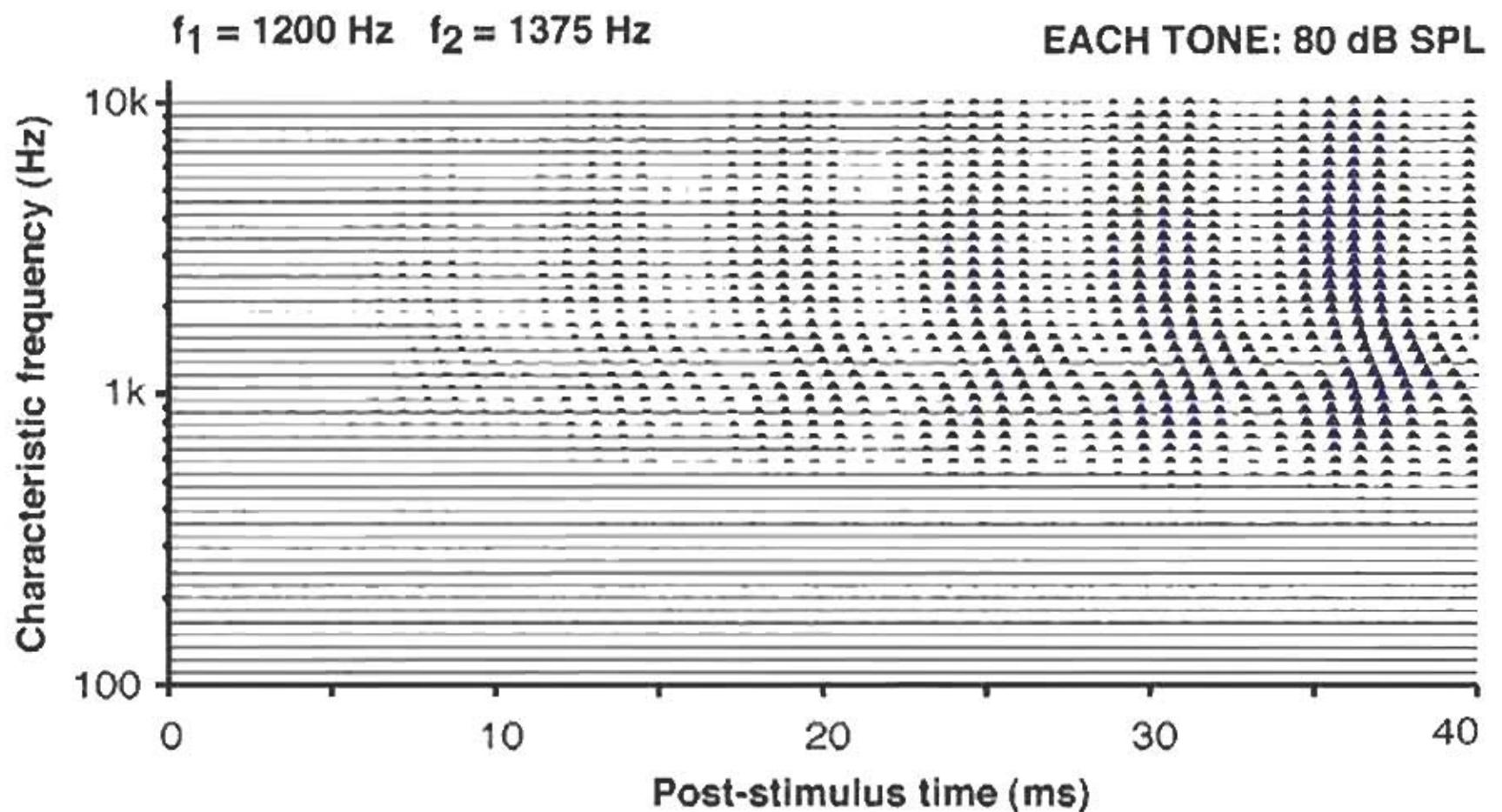


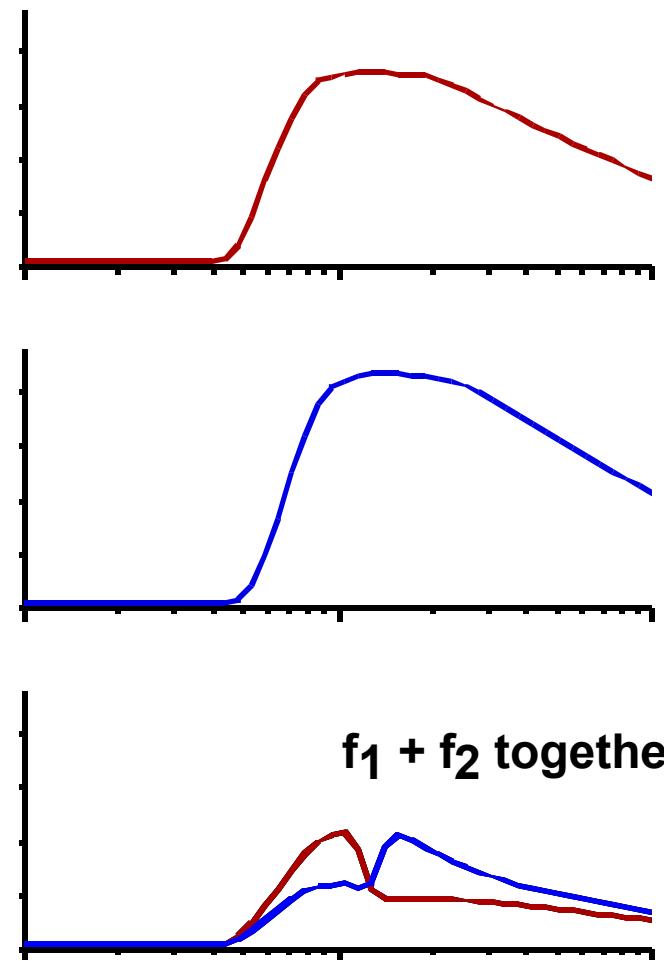
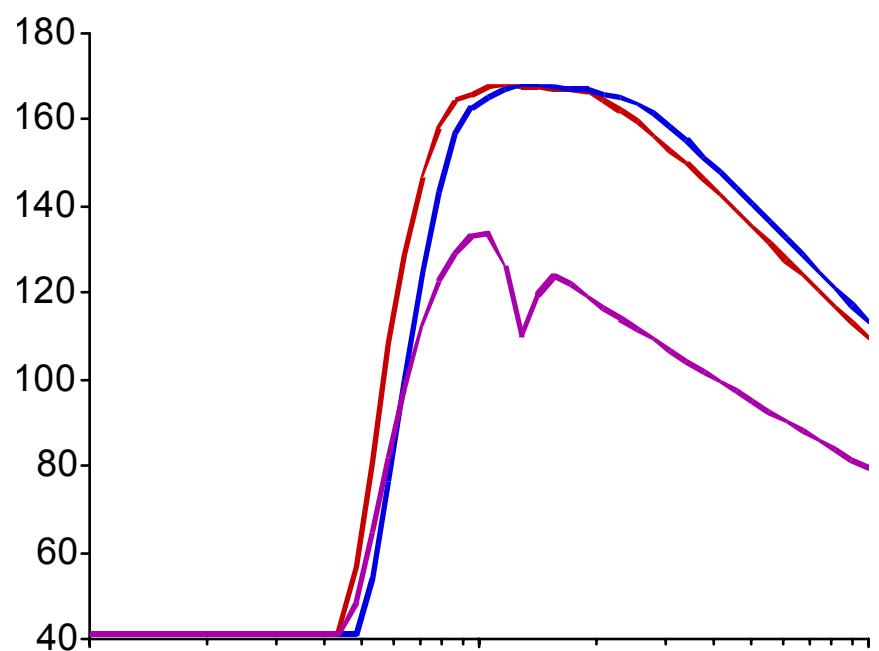
Wegel & Lane, 1924

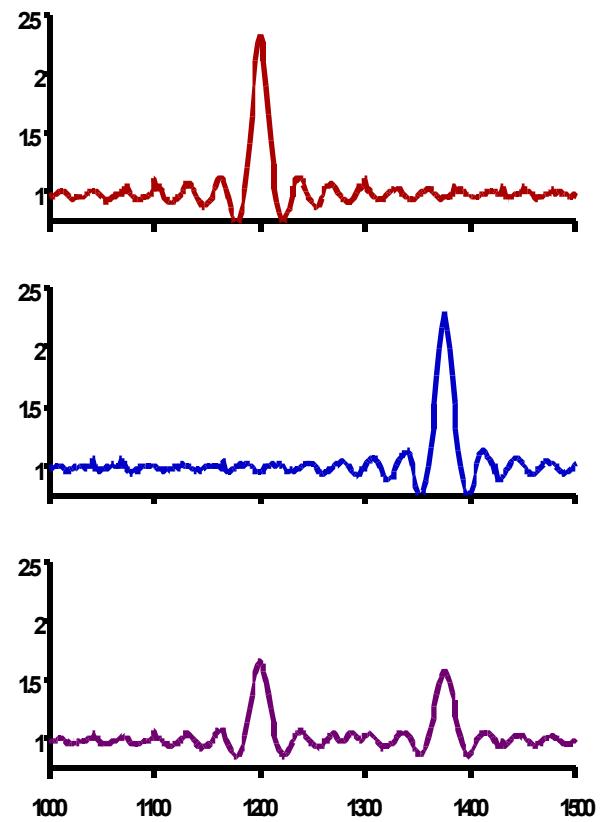
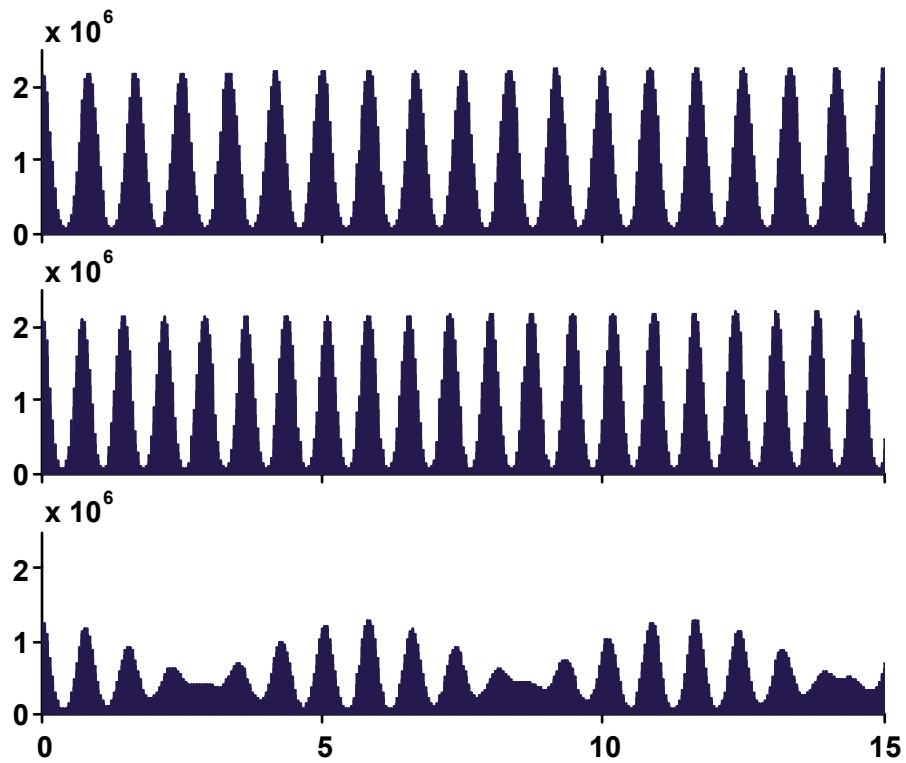
Pitch masking: pure tones

Competition for cochlear territory

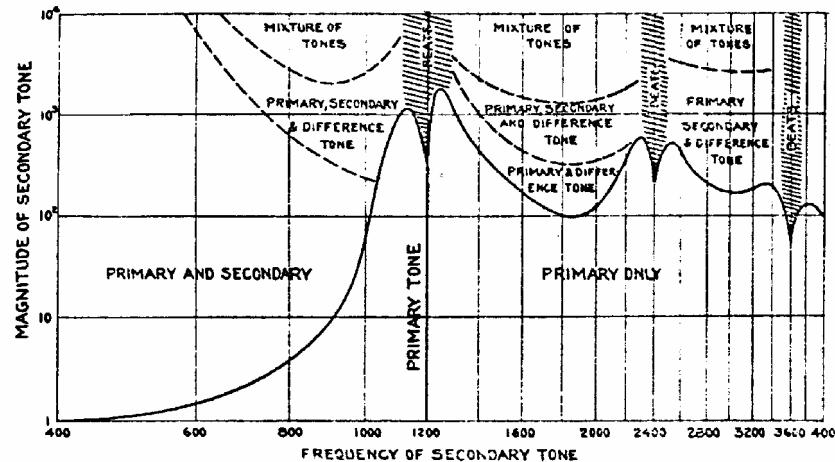
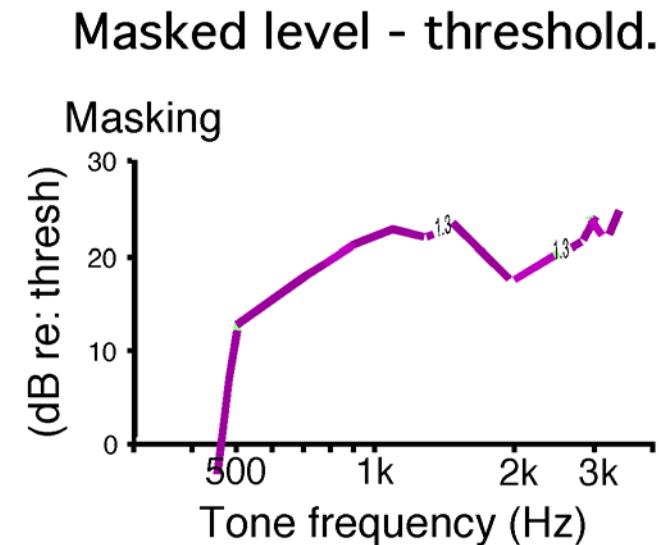
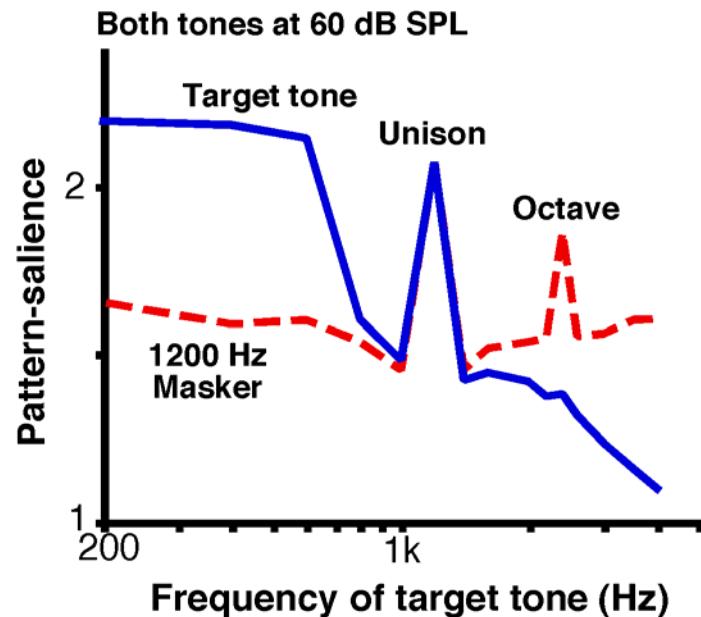
Response of auditory array to two concurrent pure tones



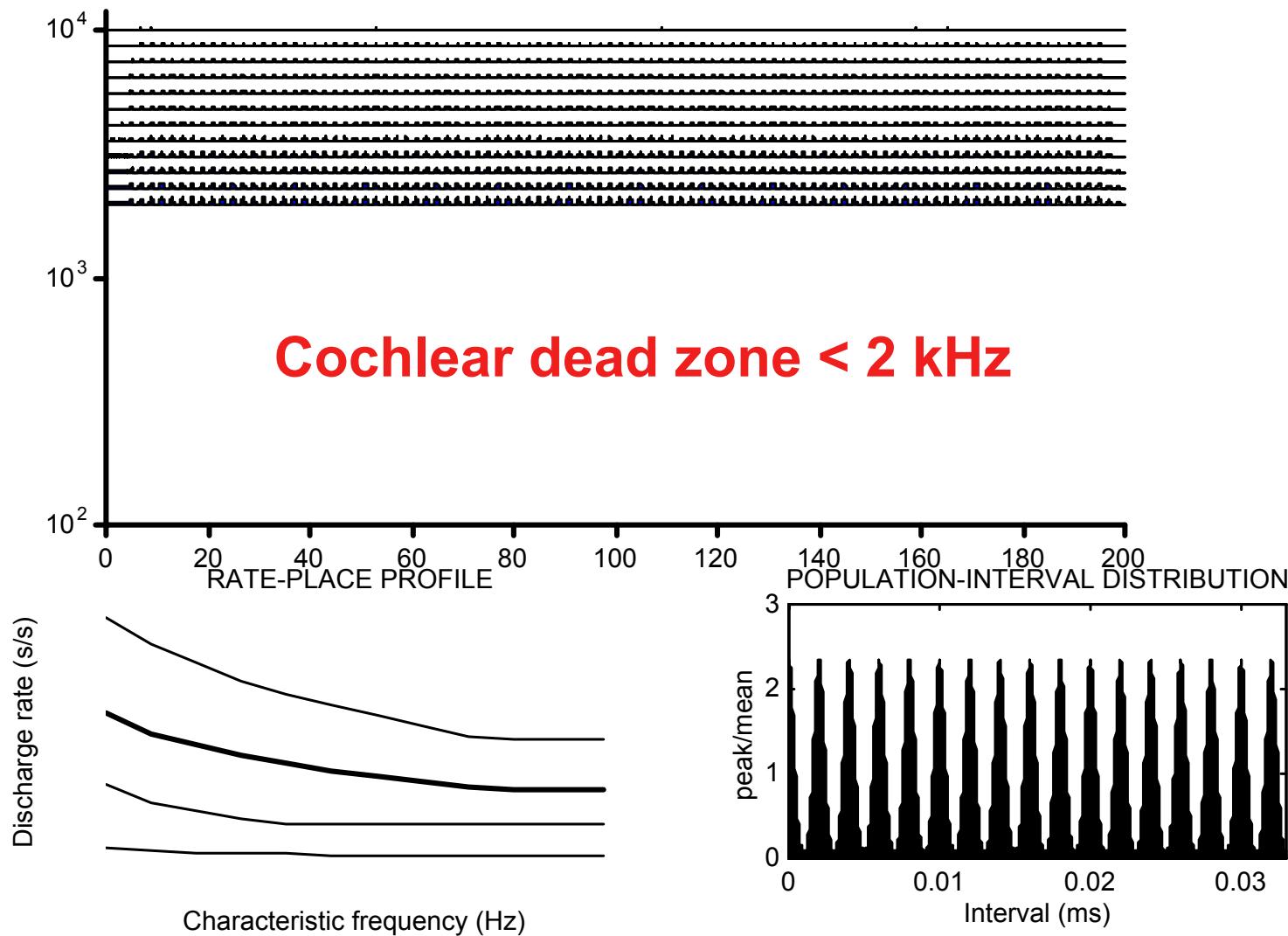


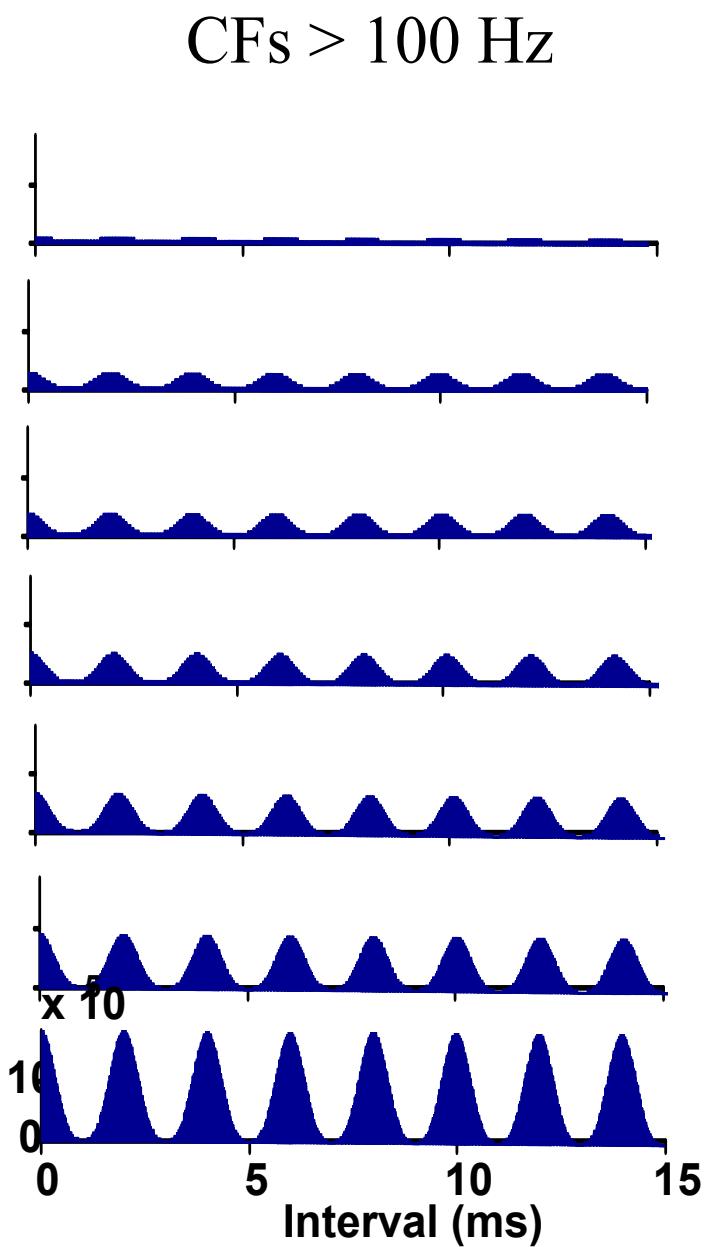
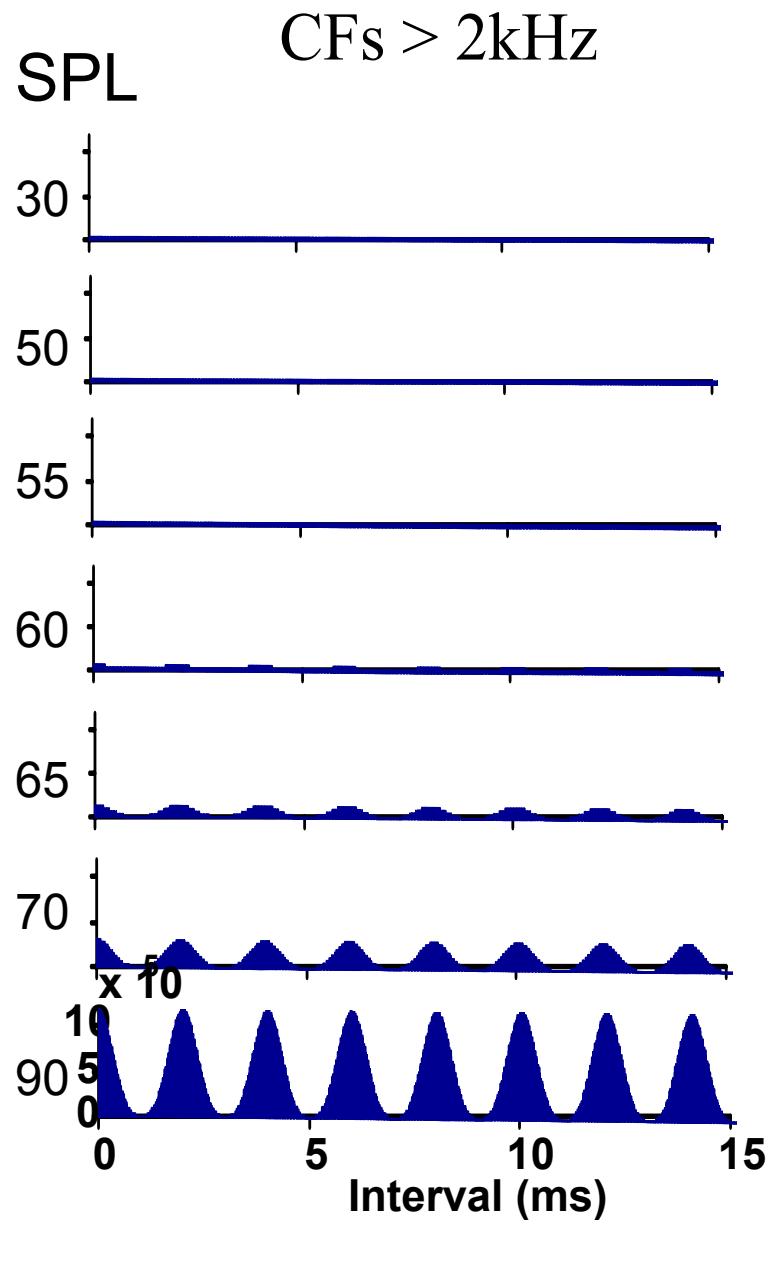


We assume that a pitch can be heard if its salience is >1.3 .
 Masking functions can be then estimated for those conditions
 that involve pitch detection (vs. Δ loudness)

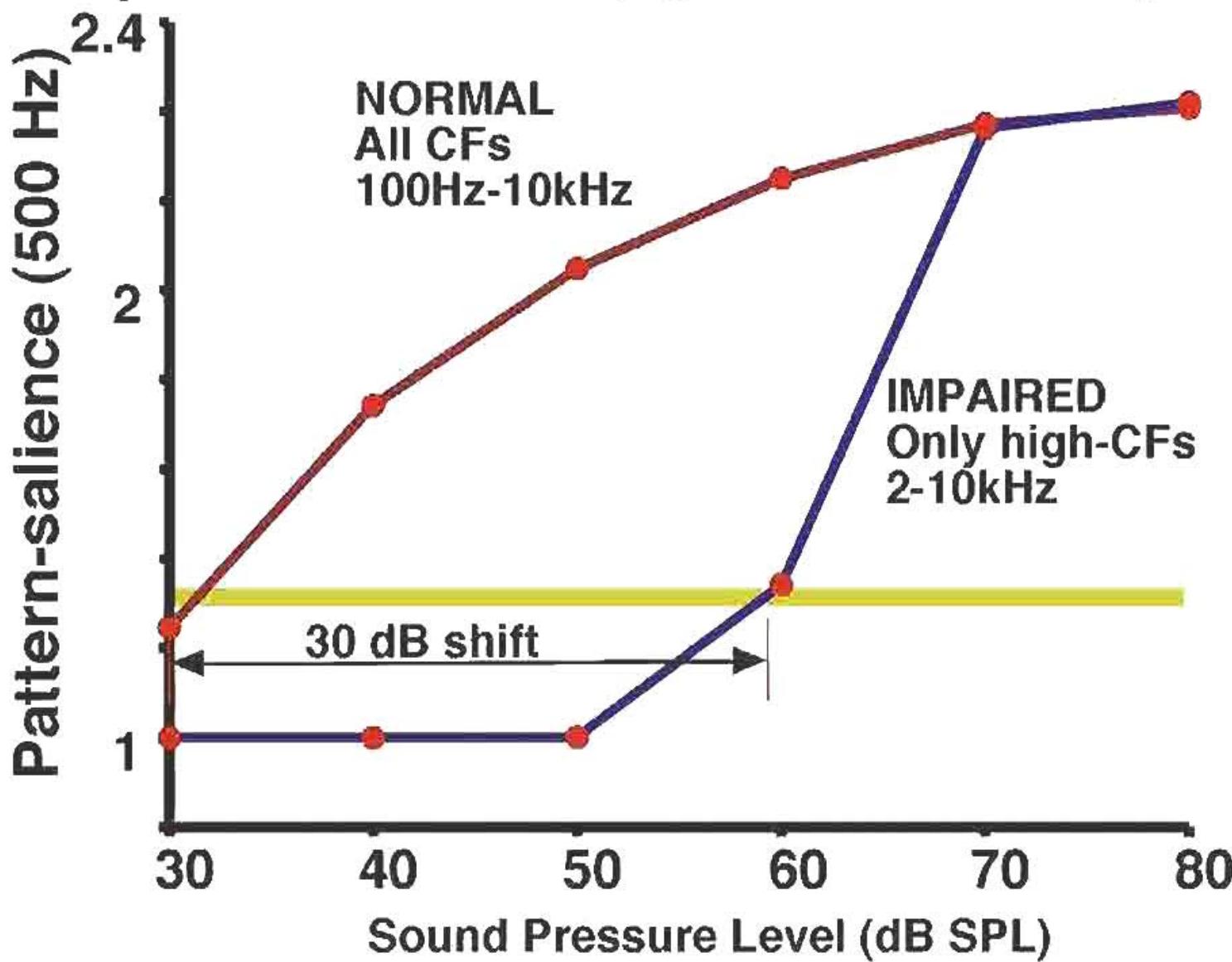


500 Hz tone 65 dB SPL





Effect of “cochlear dead zone” on pitch thresholds (e.g. Florentine & Houtsma, 1983)

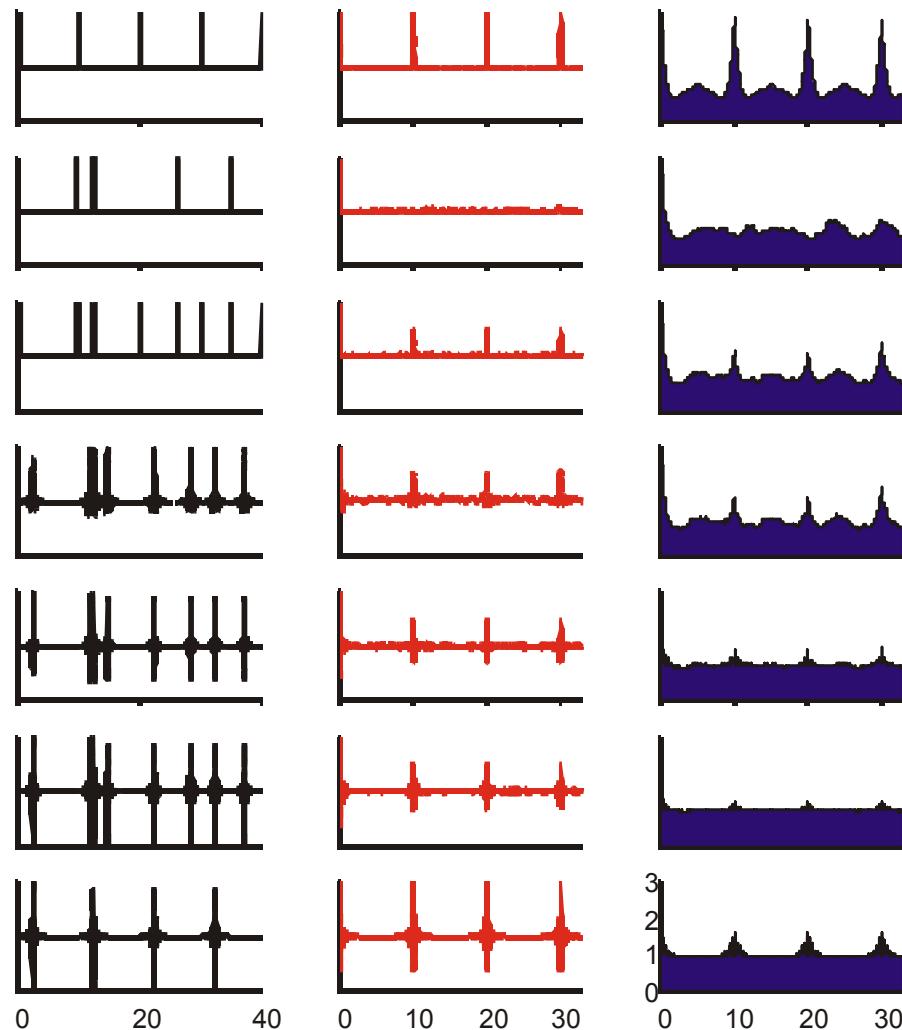


Evolution of population-interval models

- Analysis of peaks -> whole histograms
- Comparisons between whole histograms
 - Pattern similarity (octave, probe-tone:chord, FF timing nets)
- Importance of limited temporal windows
 - Finite frequency resolution of autocorrelation analysis
 - Carlyon's experiments: low F0's of high harmonics
 - Lower limits of F0-pitch (30 Hz, Pressnitzer & Patterson)
- Pitch multiplicity and salience estimation
- CF-dependent temporal analysis windows
- Recurrent timing nets - periodicity & object formation/sep.
are intimately linked mechanisms

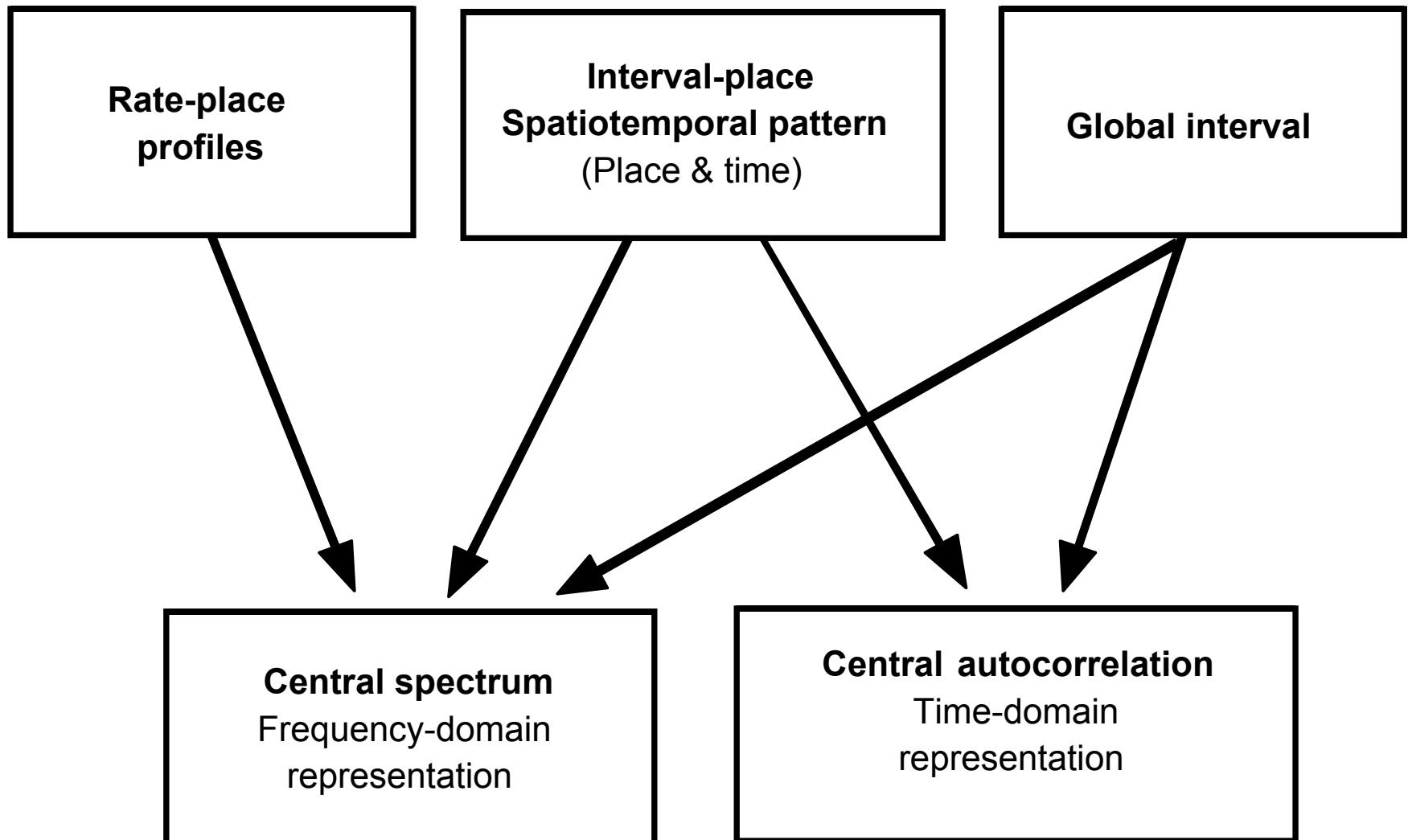
Divergences between autocorrelation & interval models

Masking of click patterns by random clicks

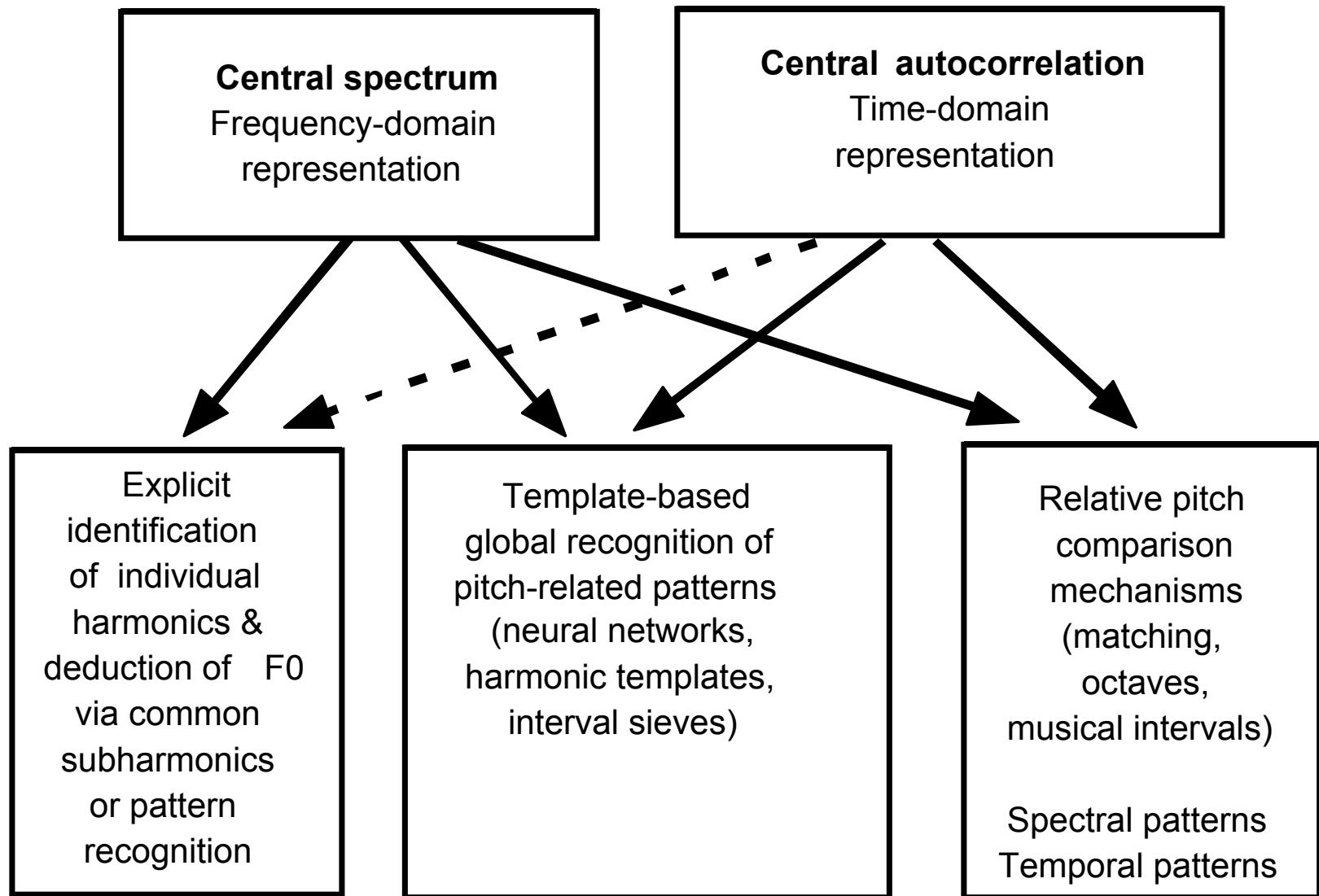


ABX	1.28
KX	1.30
KXX	1.21
KXXX	1.20

Physiological and functional representations



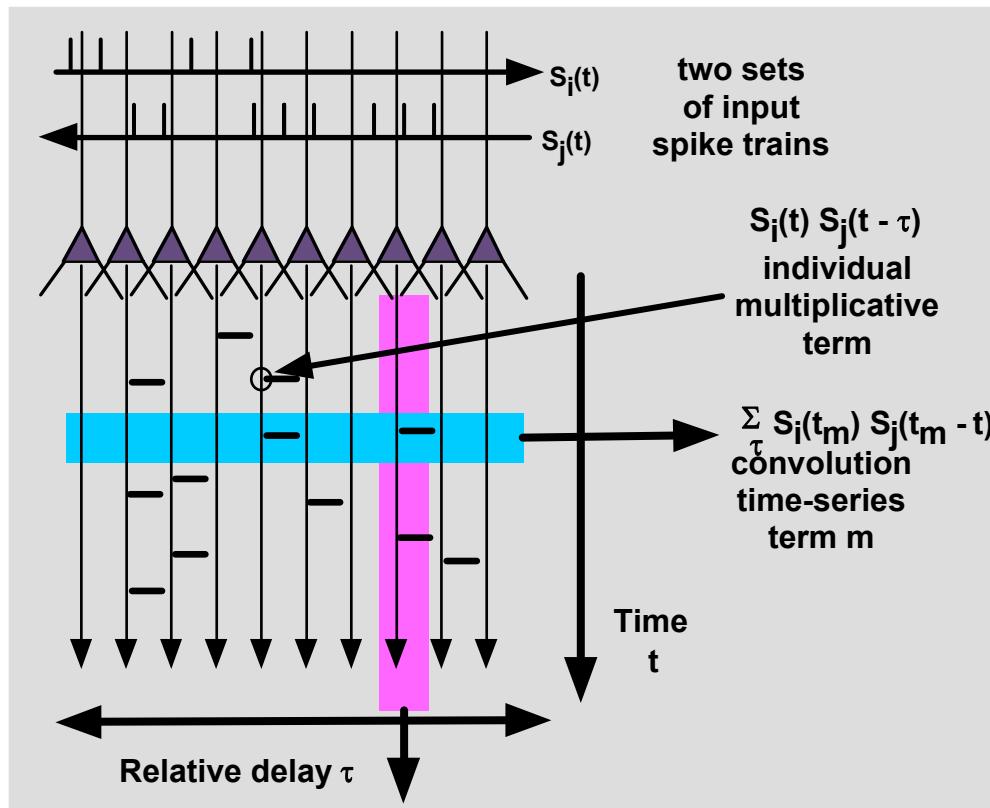
Different representations can support analogous strategies for pitch extraction, recognition, and comparison



Neural timing nets

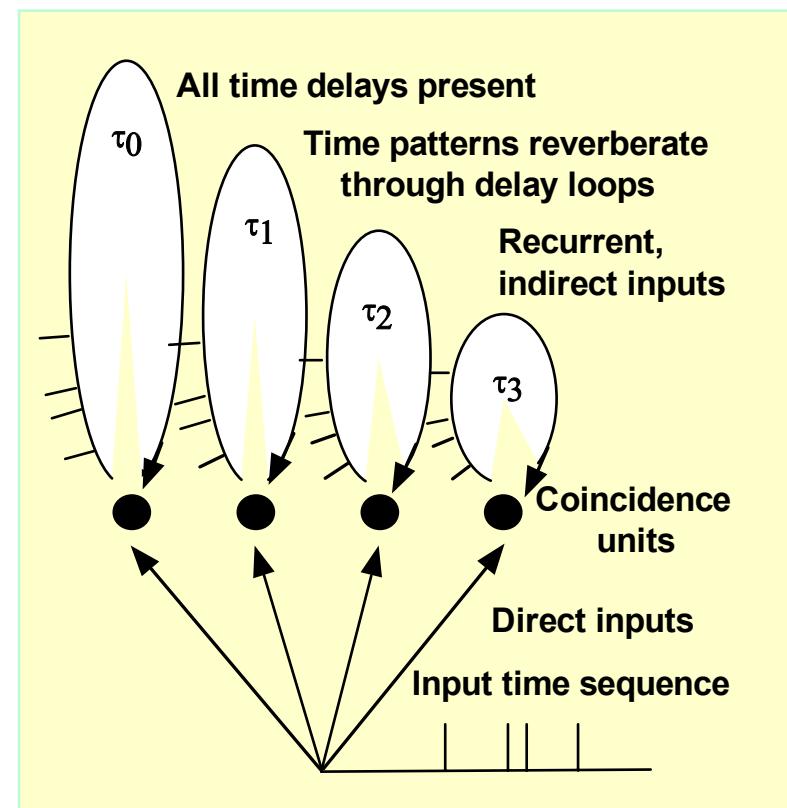
FEED-FORWARD TIMING NETS

- Temporal sieves
- Extract (embedded) similarities
- Multiply autocorrelations

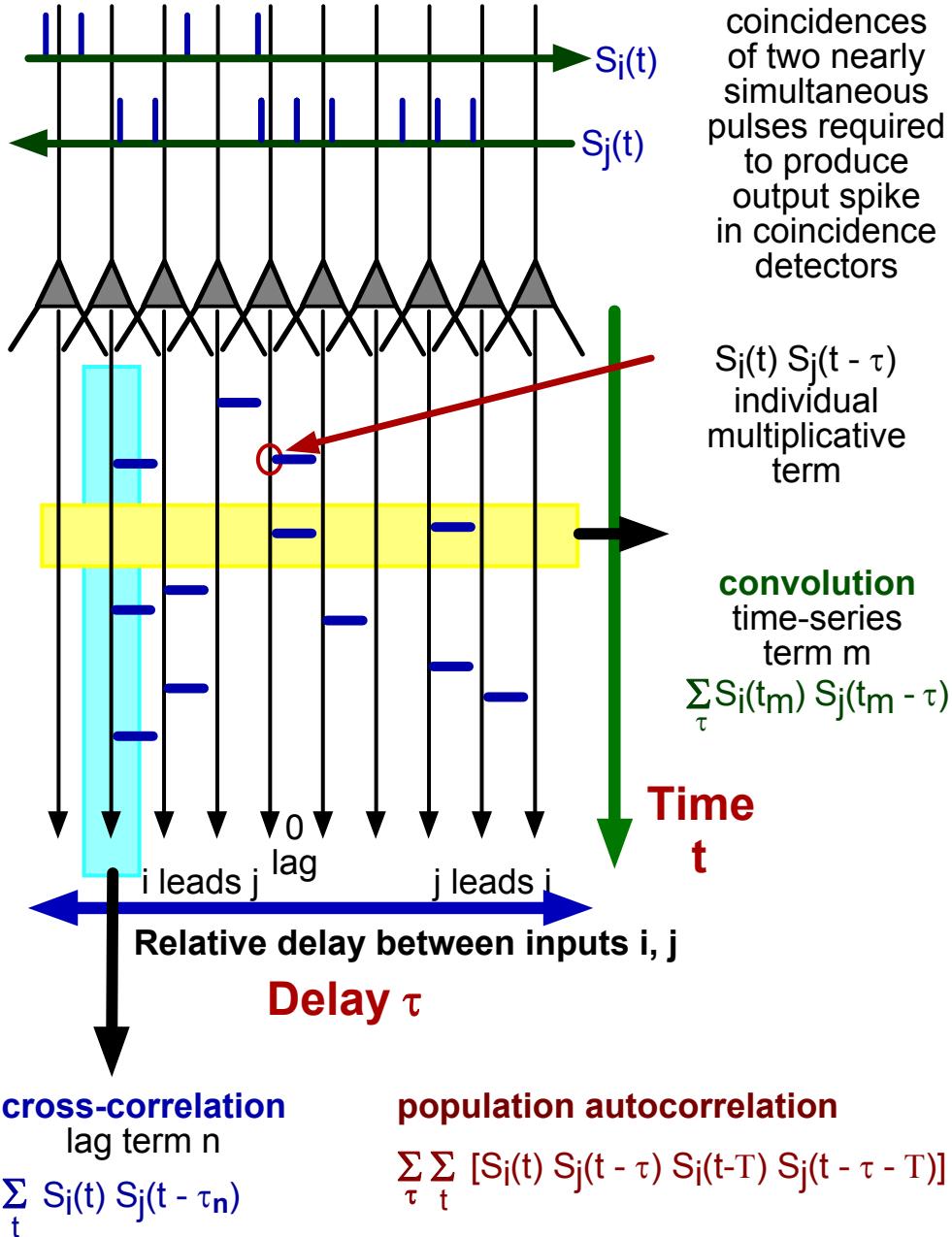


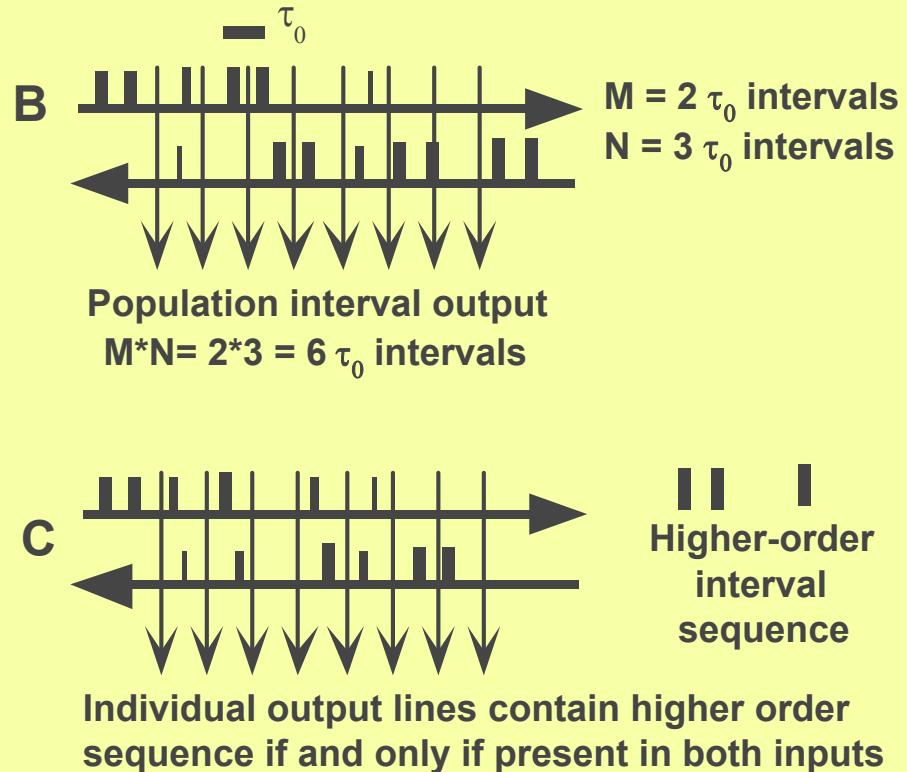
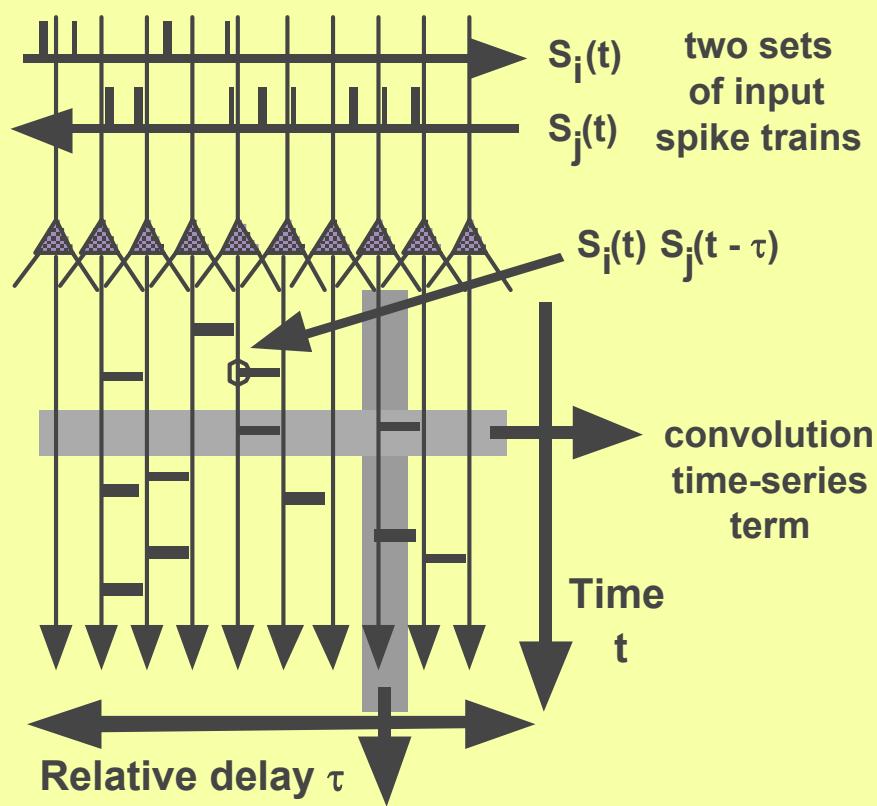
RECURRENT TIMING NETS

- Build up pattern invariances
- Detect periodic patterns
- Separate auditory objects



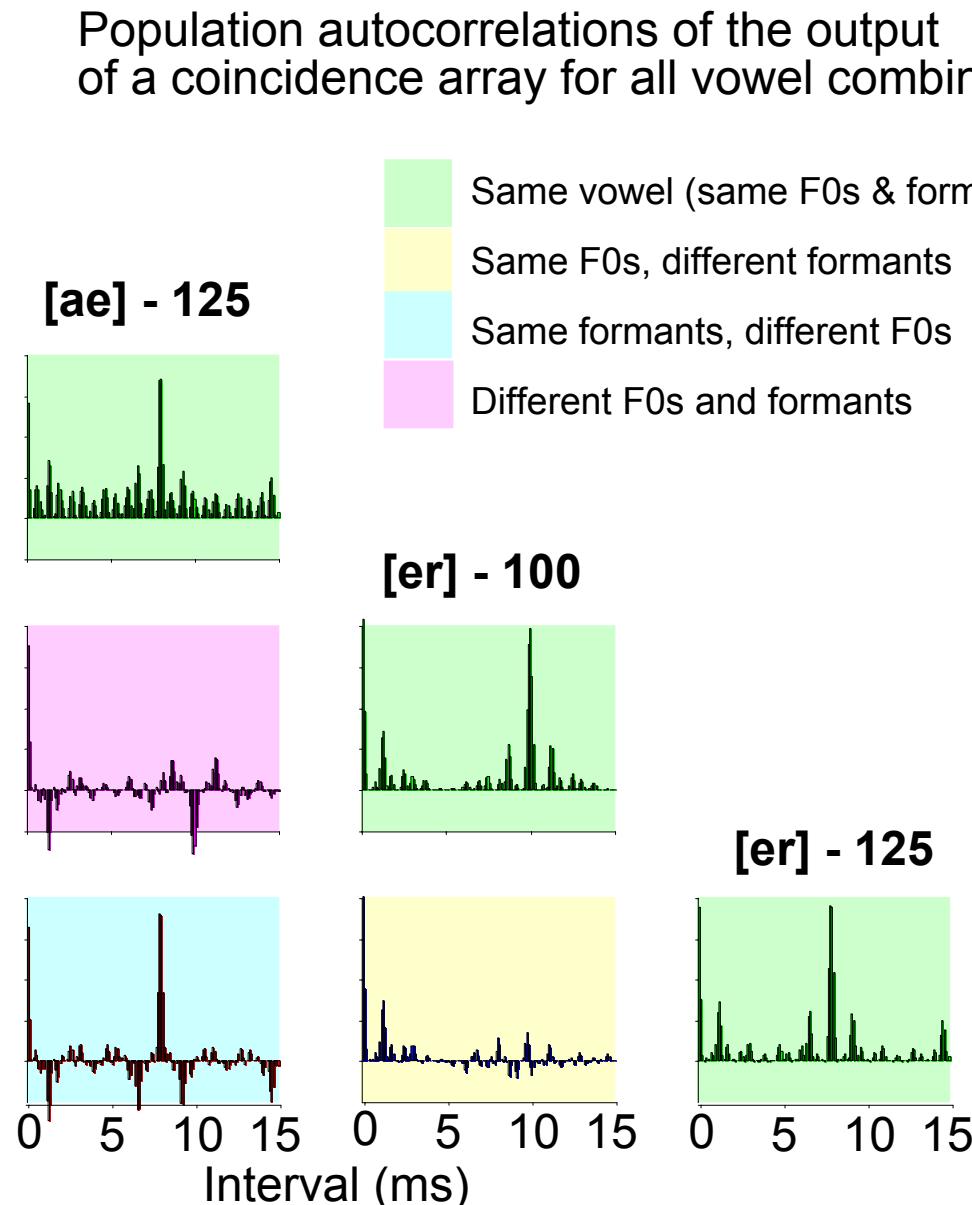
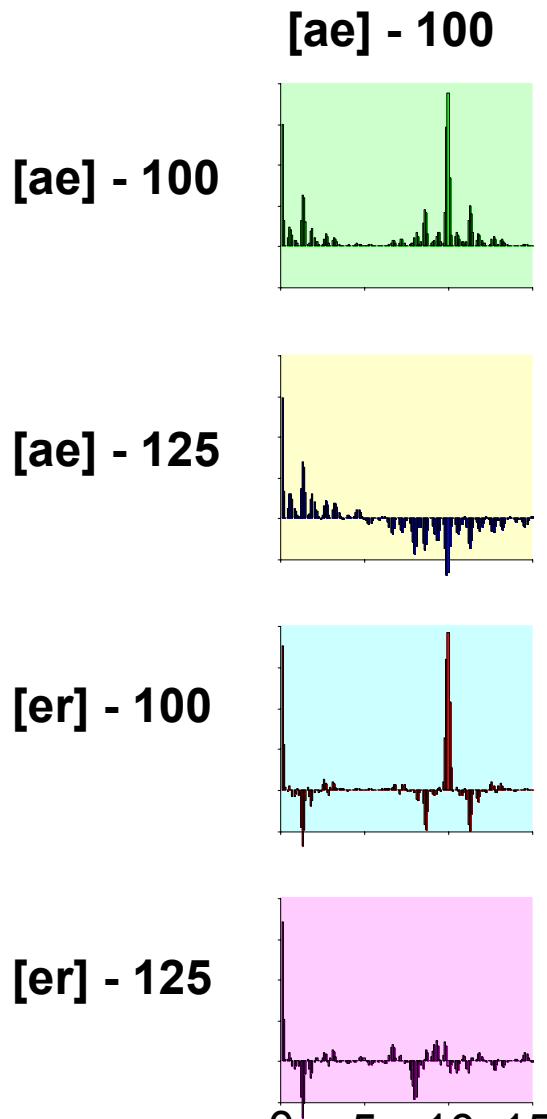
Feedforward coincidence net





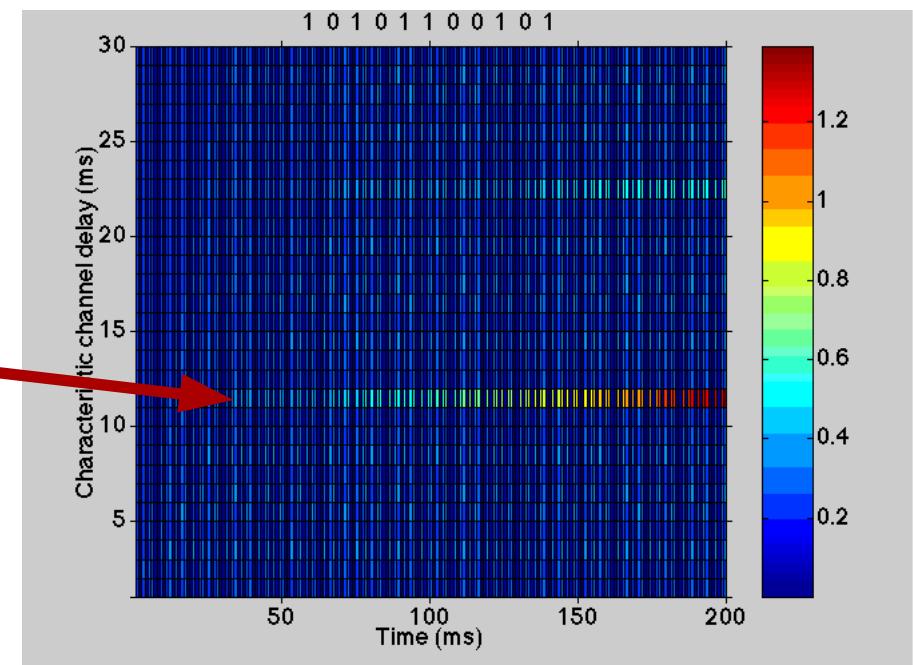
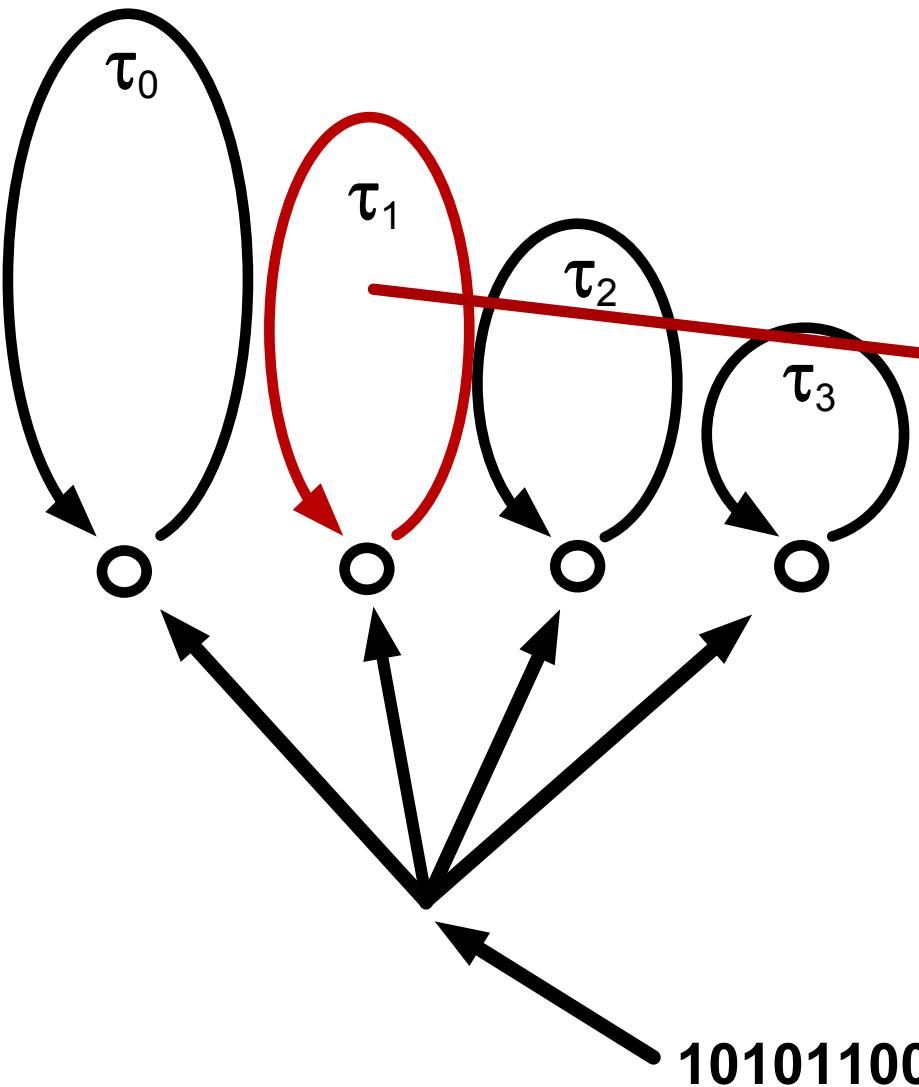
- Networks act as temporal sieves, passing only common patterns
- These are the only networks for which it is irrelevant which particular neurons are activated
- Signals are no longer tied to particular transmission lines
- Processing on the level of ensemble mass-statistics is made possible
- Extraction of complex, embedded patterns is made possible
- Time-domain multiplexing is greatly facilitated

Common timbre



Detection of arbitrary periodic patterns

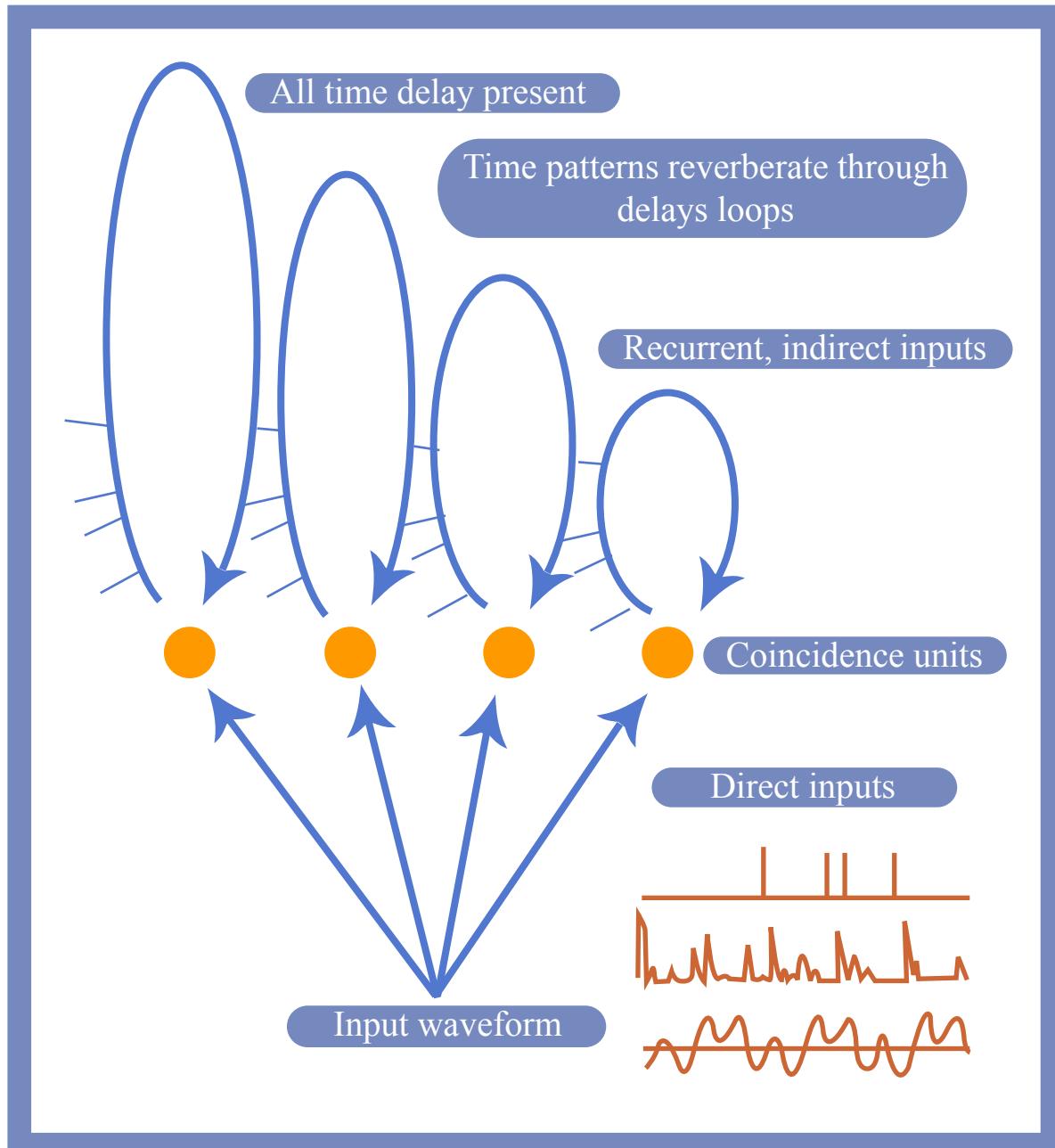
Periodic patterns invariably build up in delay loops whose recurrence times equals the period of the pattern and its multiples.

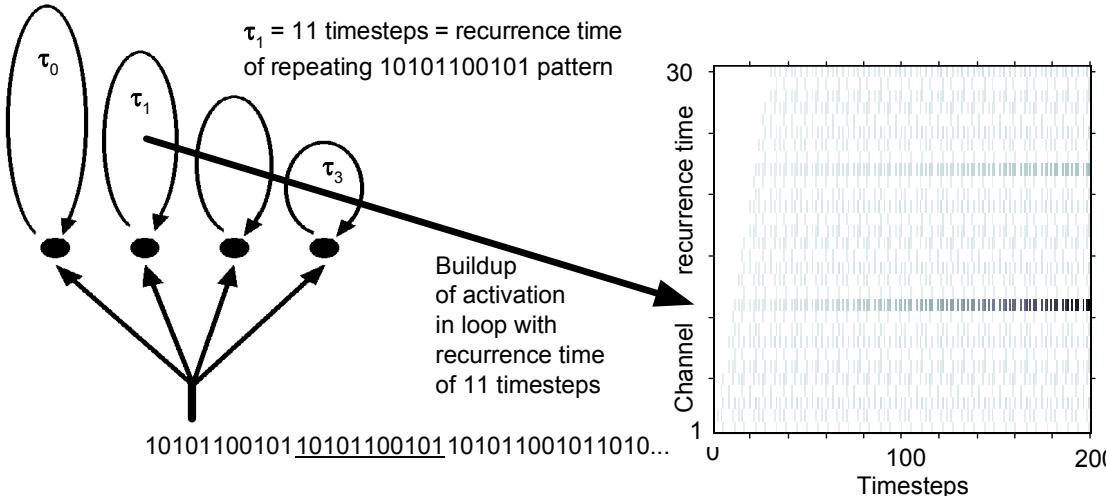


$\tau_1 = 11 \text{ ms} = \text{recurrence time}$
of input pattern 10101100101101011001011010...

Input pattern

10101100101101011001011010...





Periodic pattern builds up

La Marseillaise rhythm

1100110001000100010001000000110011001100010000000100110000000000...

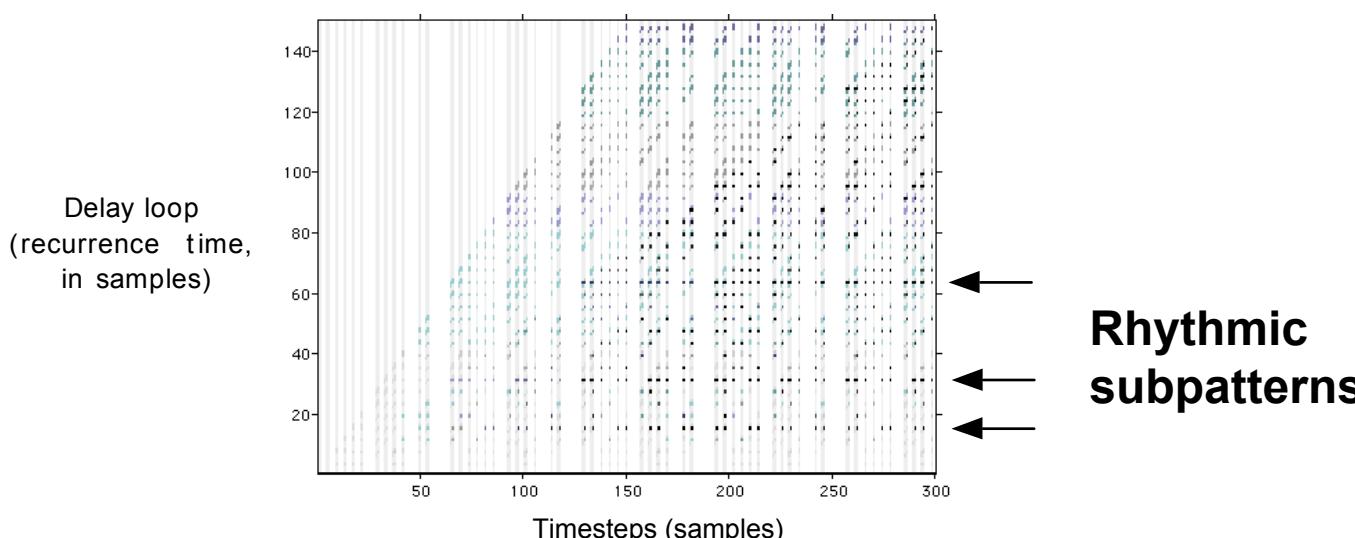
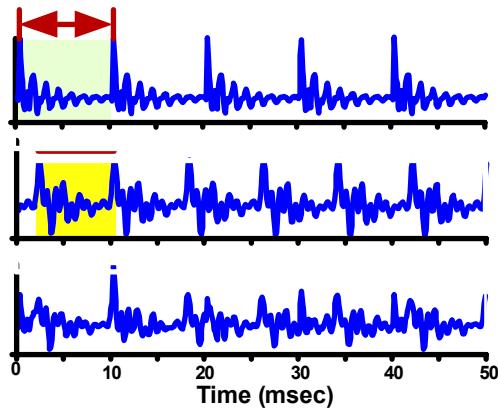


Figure 8. Recurrent timing nets. Top. Behavior of a simple recurrent timing net for periodic pulse train patterns. The network generates many sets of expectations. The delay loop whose recurrence time equals the period of the pattern builds up that pattern. Below. Response of a recurrent timing net to the beat pattern of La Marseillaise. Arrows indicate periodic subpatterns at 16, 32, and 64 timesteps that are built up by the network. The example points to potential applications of recurrent timing nets for rhythm induction and analysis. From Cariani, 1999, working paper on timing nets and rhythm.

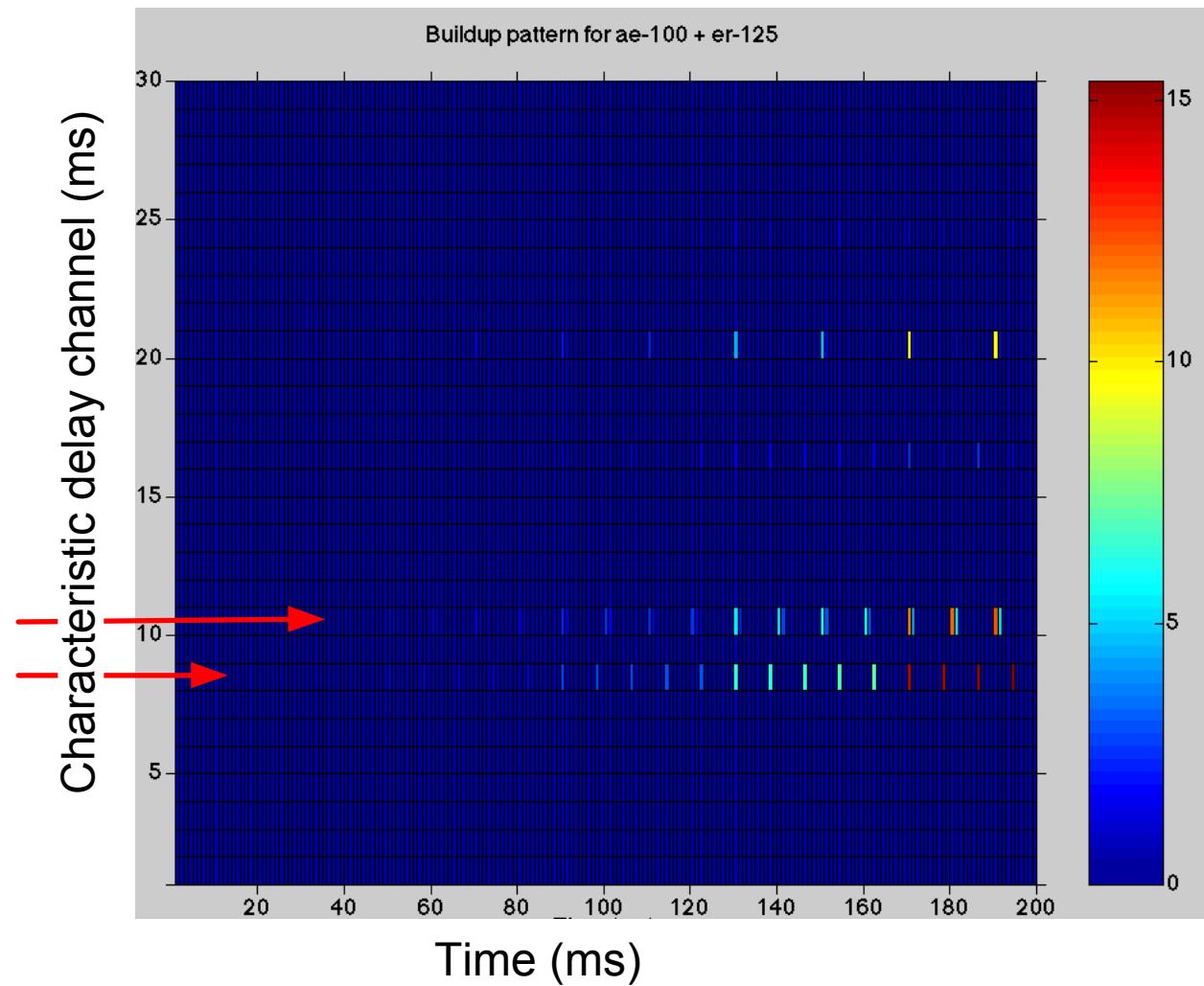
Build-up and separation of two auditory objects

Two vowels with different fundamental frequencies (F0s) are added together and passed through the simple recurrent timing net. The two patterns build up in the delay loops that have recurrence times that correspond to their periods.



Vowel [ae]
F0 = 100 Hz
Period = 10 ms

Vowel [er]
F0 = 125 Hz
Period = 8 ms

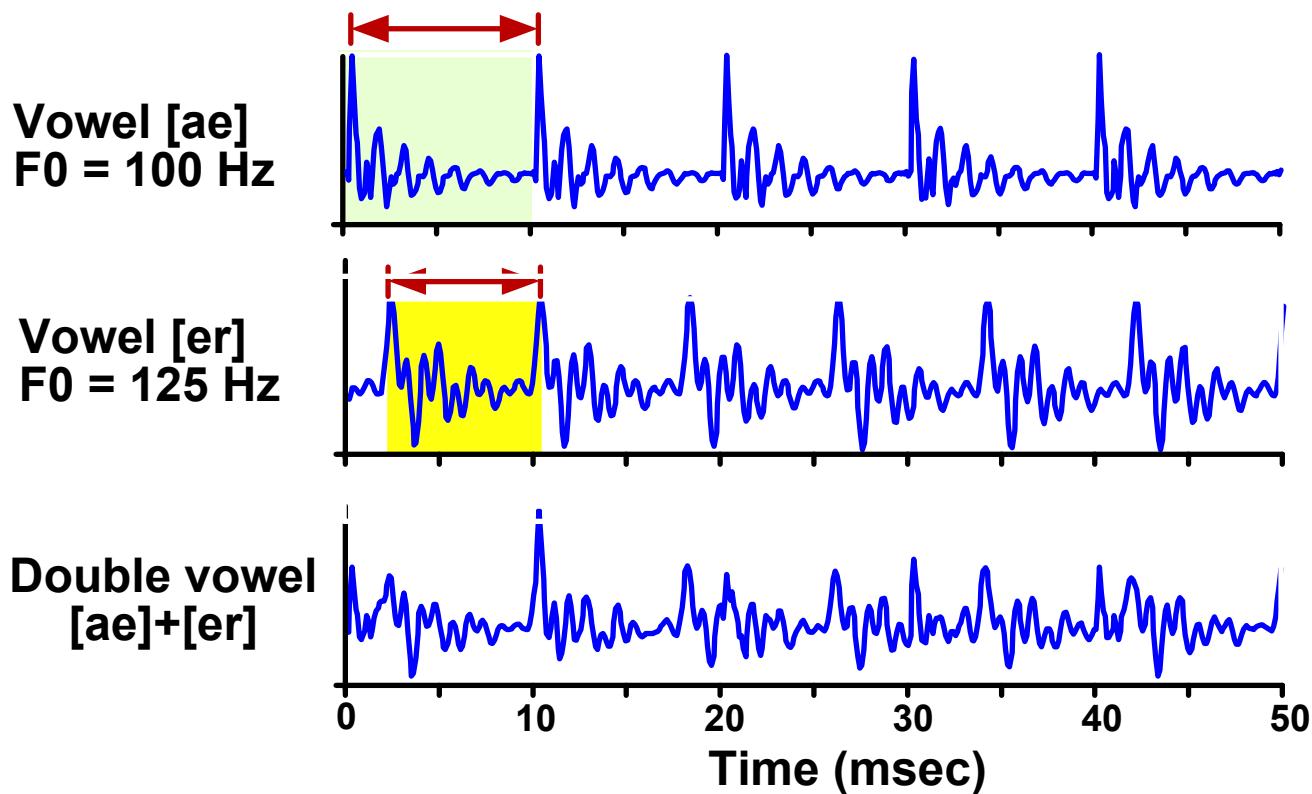


Pitch and auditory scene analysis

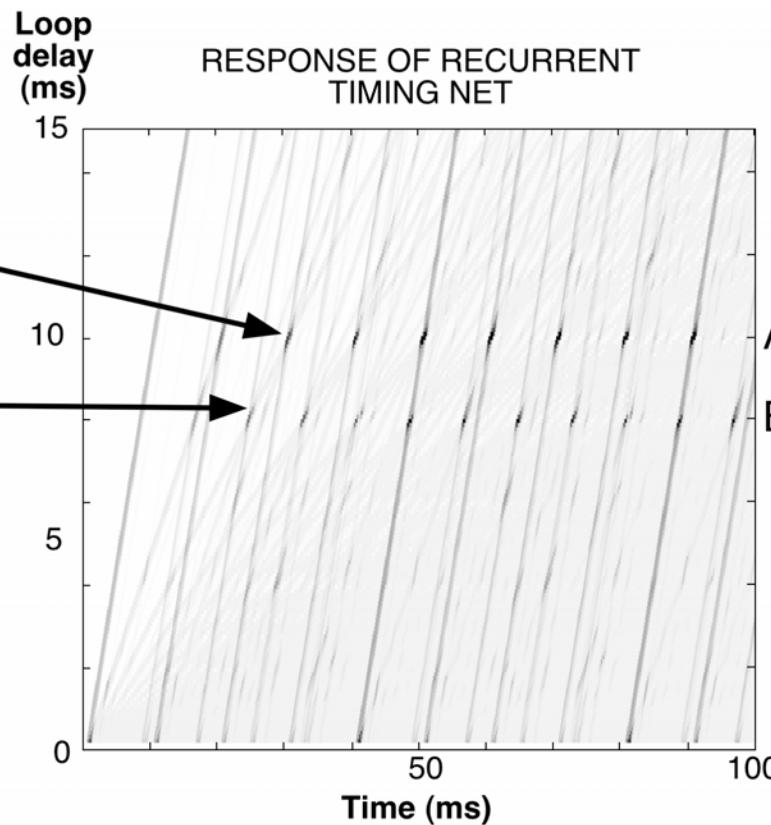
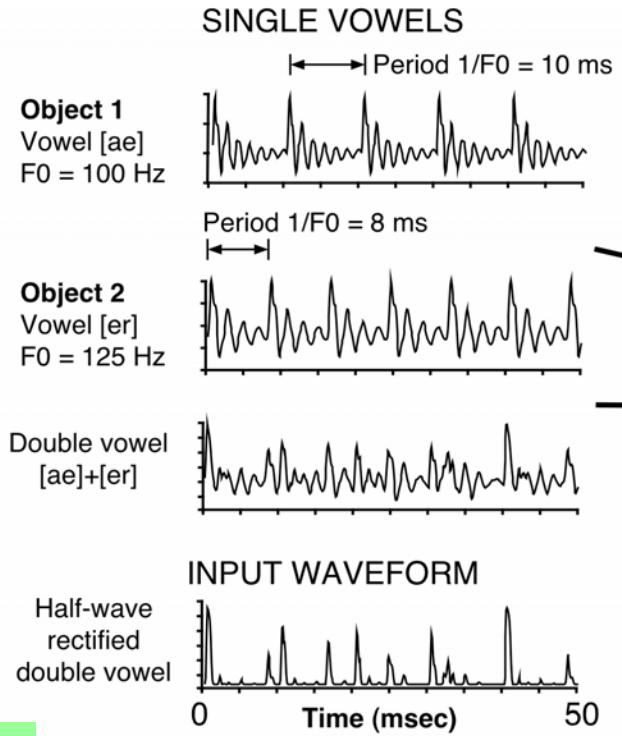
Two different strategies for scene analysis:

1. Group channels by common properties (local features)
2. Build up invariant temporal patterns as objects (invariant relations)

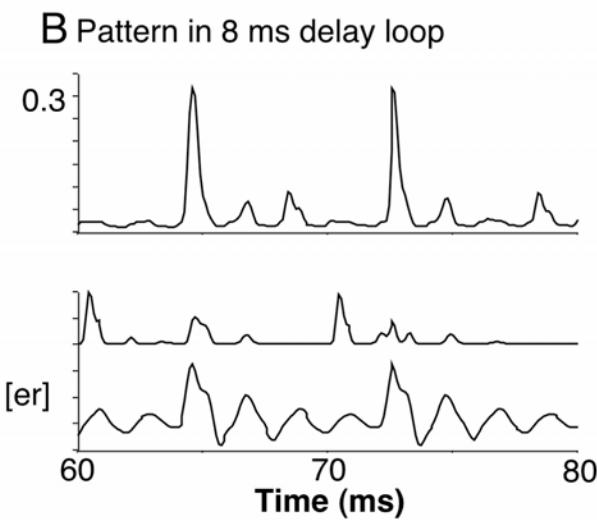
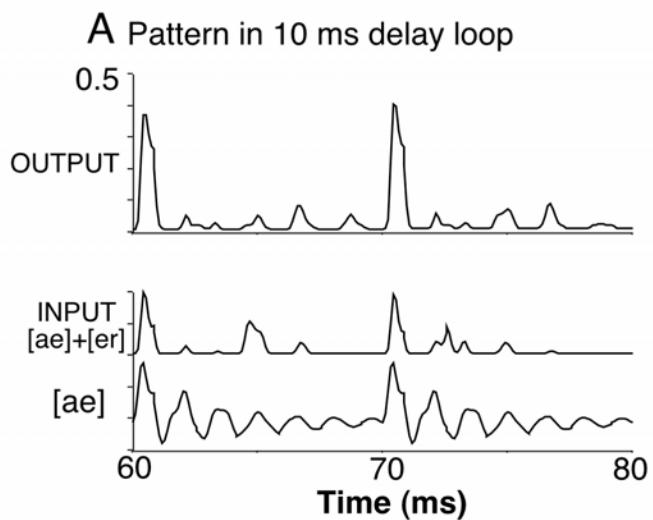
Separation of multiple recurrent time patterns: two vowels with different F0's



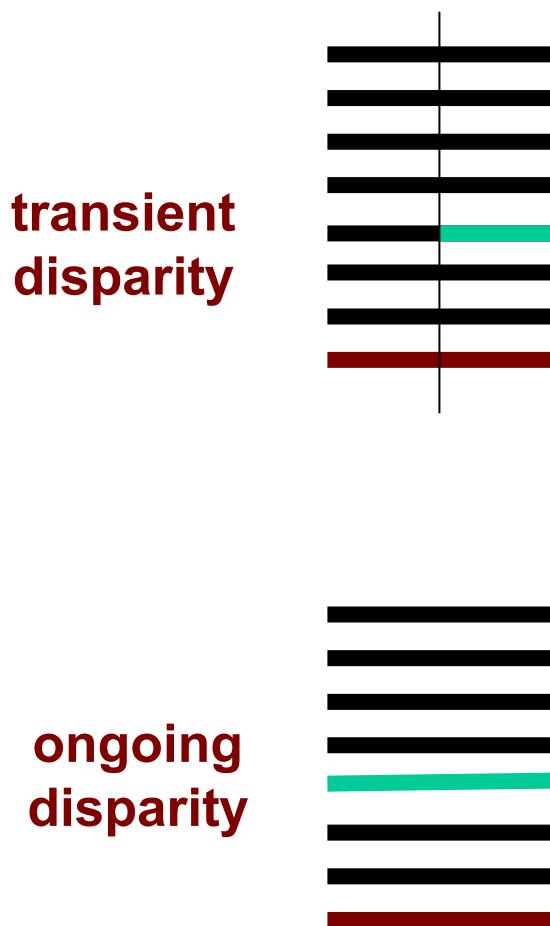
Recurrent timing net with single input channel



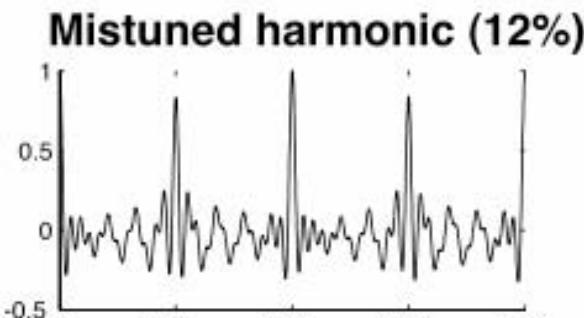
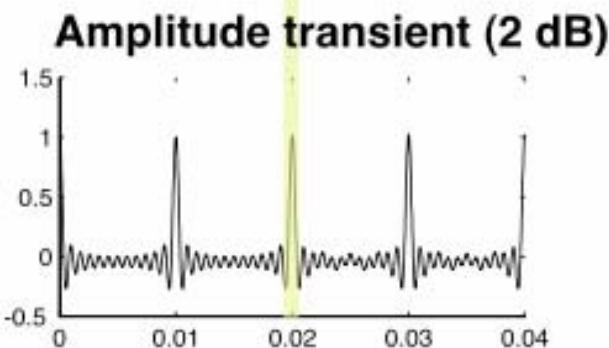
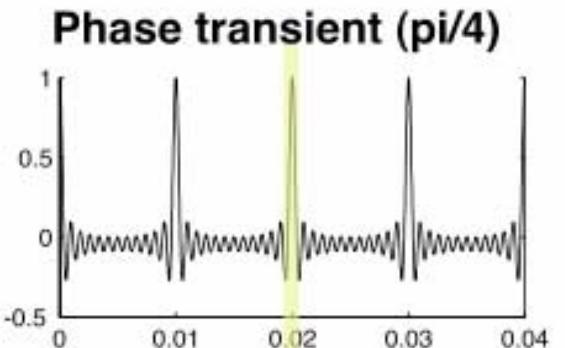
Revised buildup rule:
Min(direct, circulating)
plus a fraction of their
absolute difference



Auditory "pop-out" phenomena suggest a period-by-period waveform comparison

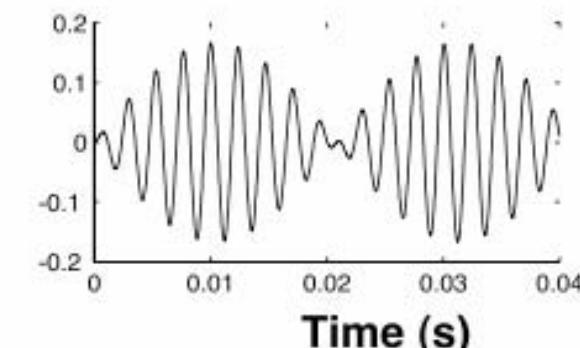
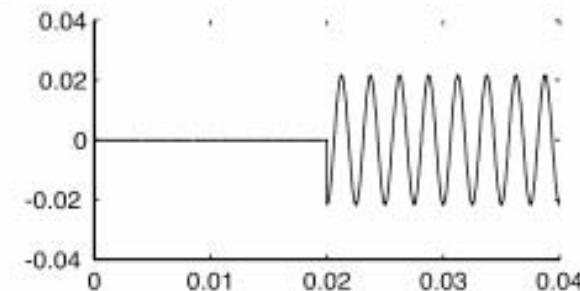
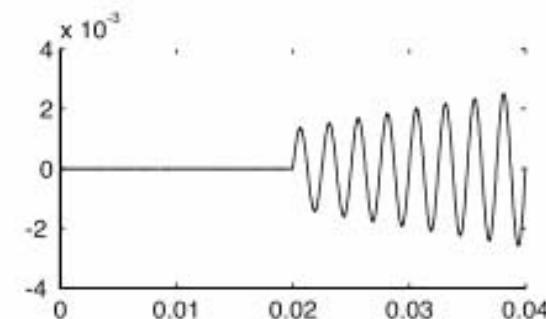


Waveform
F0 = 100; Harmonics 1-8
Δ 4th harmonic



Deviation from periodic pattern

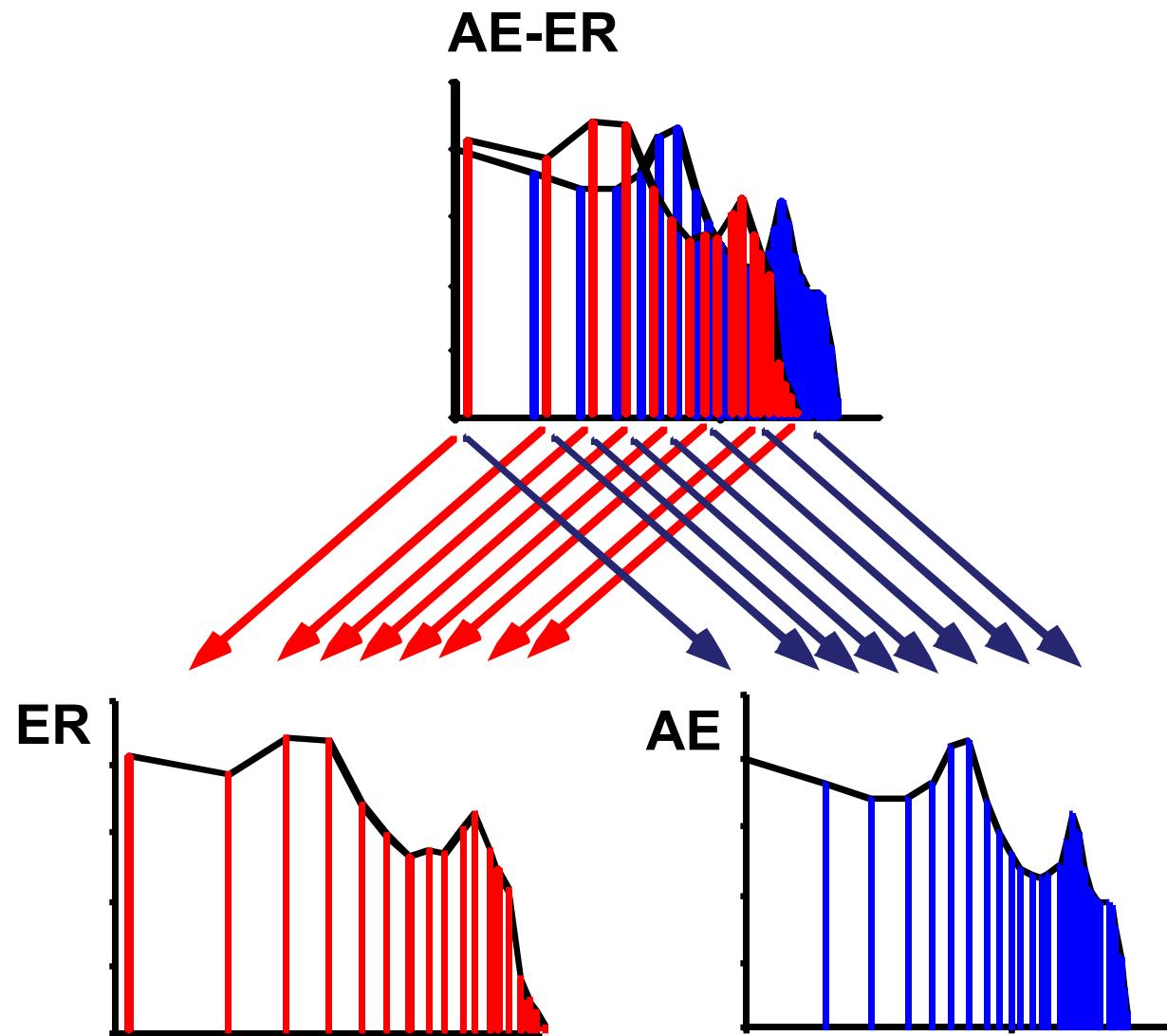
Last 2 periods - first 2



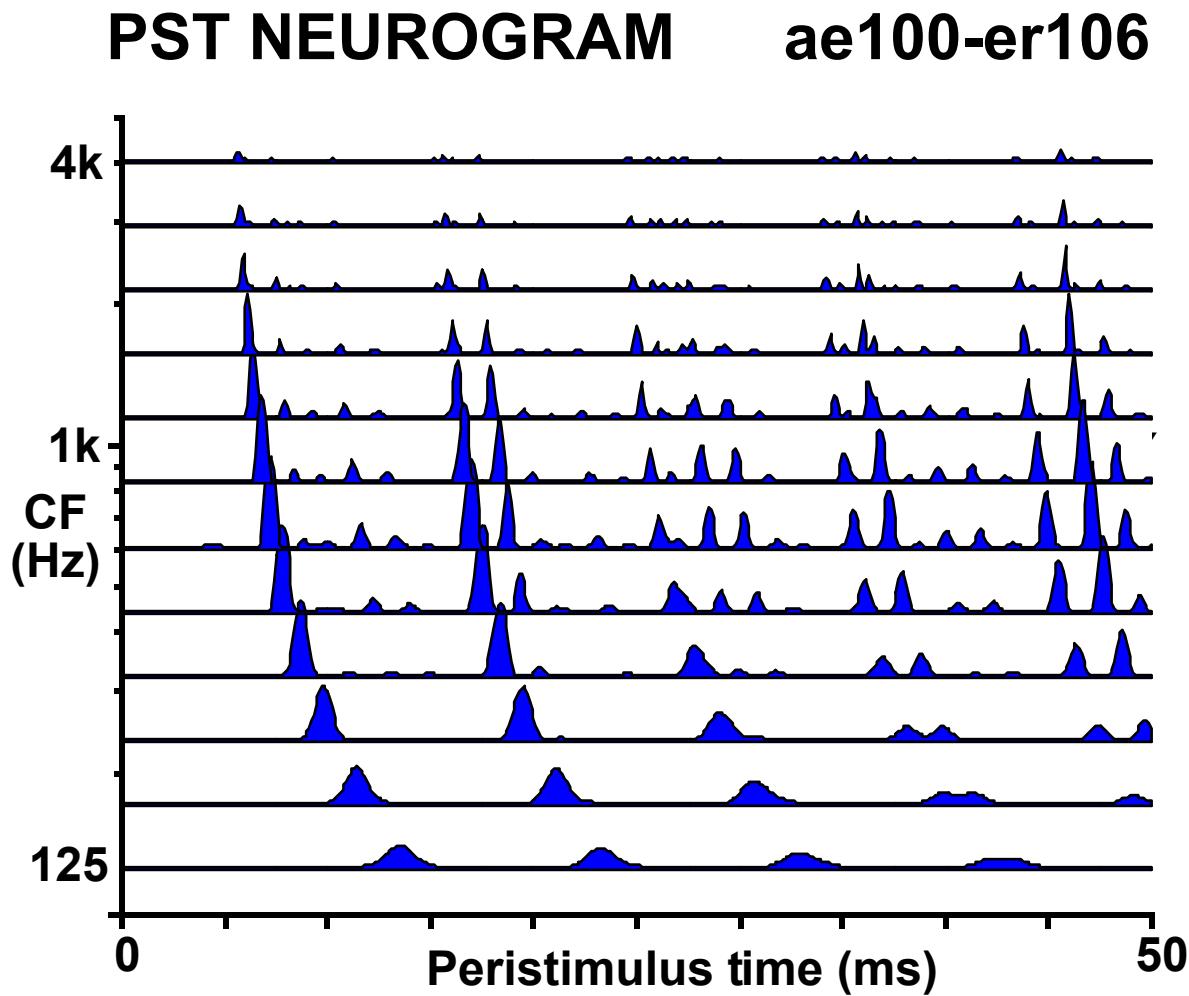
Traditional approach (Frequency domain)

Segregate
frequency
channels

Assign
channels
to objects



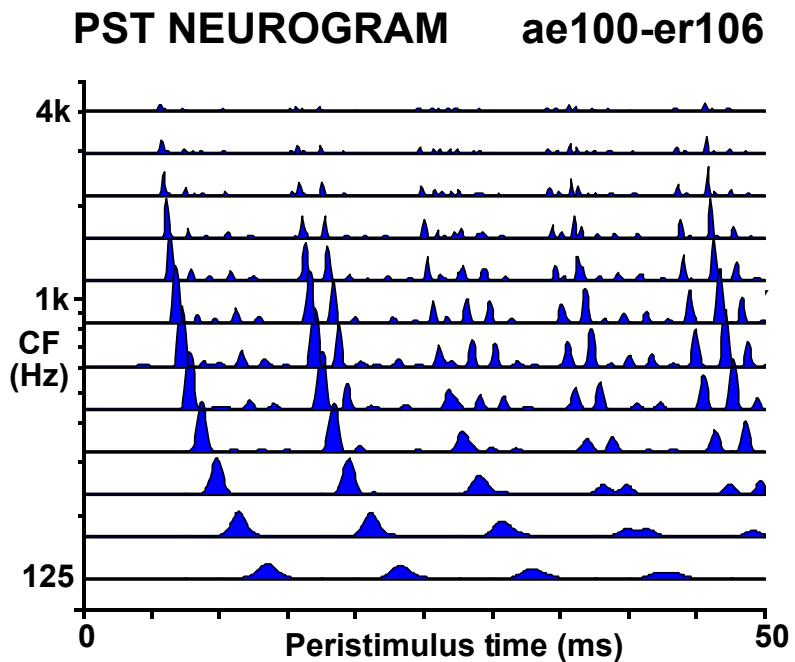
Double vowels drive the same frequency channels in the auditory nerve



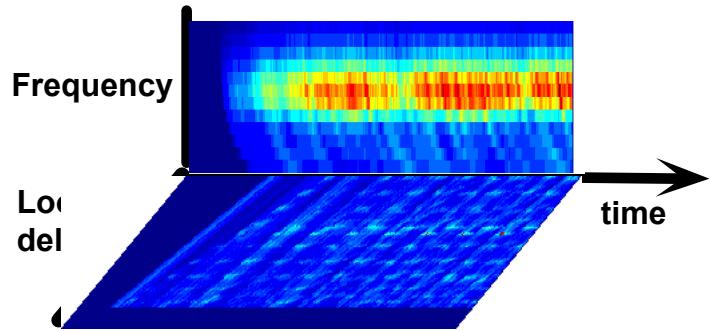
The information needed to separate sounds
on the basis of F0 differences
is multiplexed in the phase-locked fine time structure
of the neural firing patterns
(Traditional spectrographic approaches throw away
this information at an early stage.)

How might a neural system
(e.g. the auditory system)
exploit this information?

Strongly agree that we shouldn't be
limited by ignorance of auditory
system, but we need to reverse-
engineer that system if we are to find
new functional principles



Behavior of the loop array



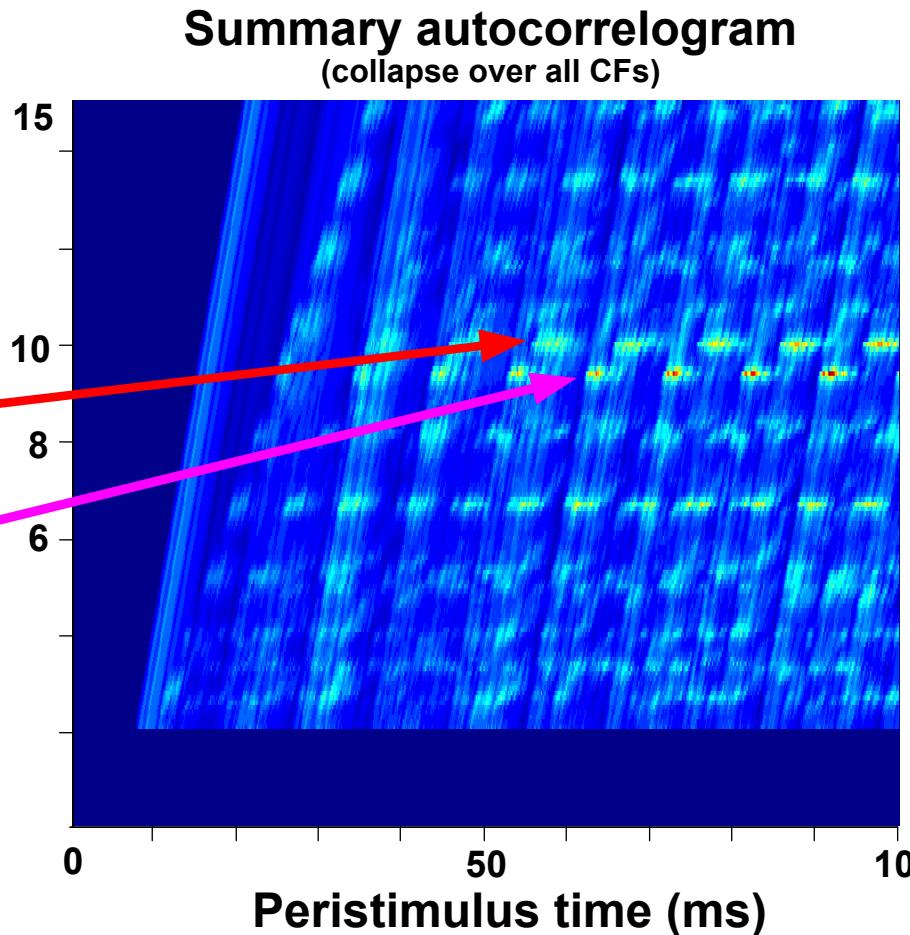
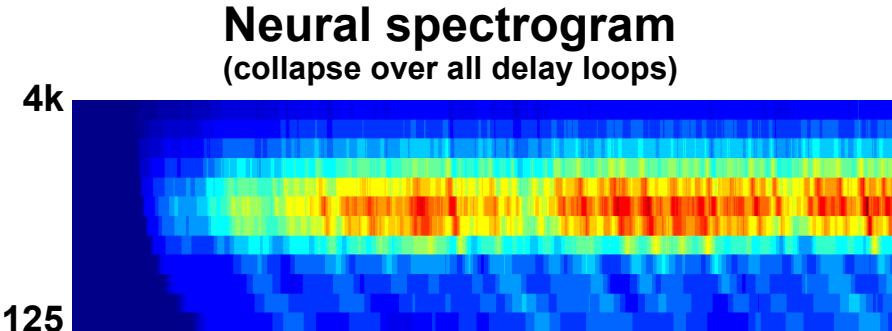
Activity in the loop array is described in terms of 3 dimensions:

- frequency channel
- delay loop
- peristimulus time

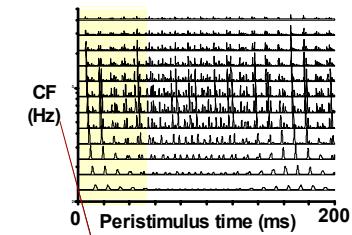
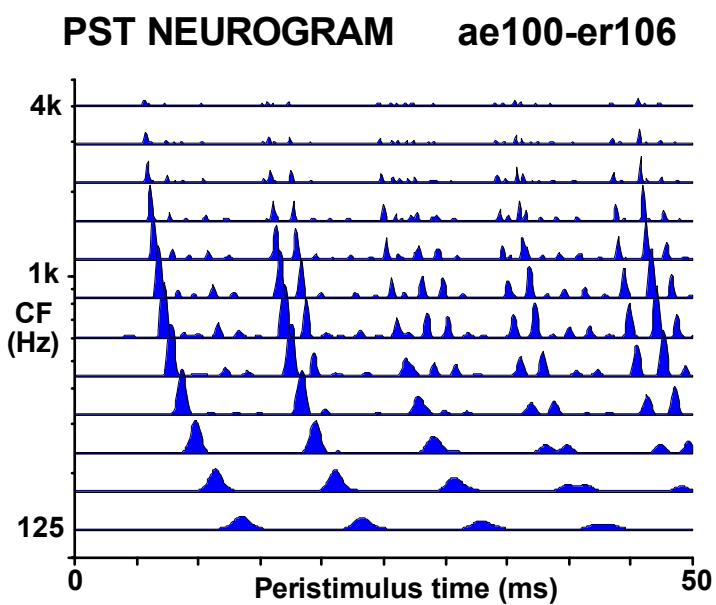
ae100
(10 ms period)

er106
(9.4 ms period)

Loop Delay (ms)



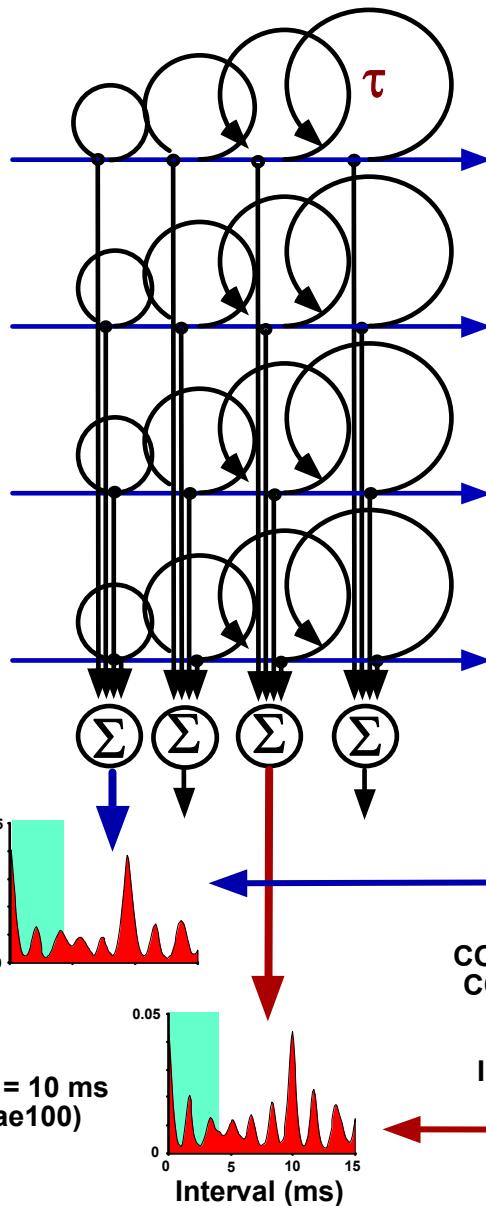
Processing of tonotopically-organized inputs



POPULATION-INTERVAL DISTRIBUTIONS FOR DIFFERENT DELAYS

**Loop delays = 9.7 ms
(period of er106)**

**Loop delays = 10 ms
(period of ae100)**

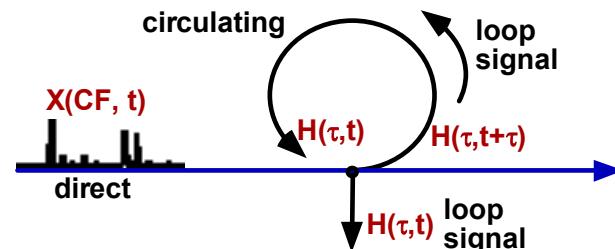


BUILD UP CORRELATION PATTERNS IN DELAY LOOPS FREQUENCY-BY-FREQUENCY

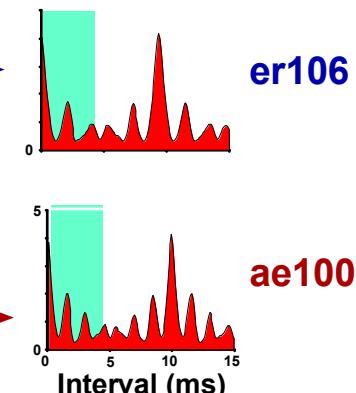
BUILDUP RULE:

pattern entering the delay loop is the minimum of direct & circulating inputs, plus a fraction of their difference

$$H(t,t+t) = \min(X(CF, t), H(t,t)) + 0.1 * \text{abs}(X(CF, t) - H(t,t));$$



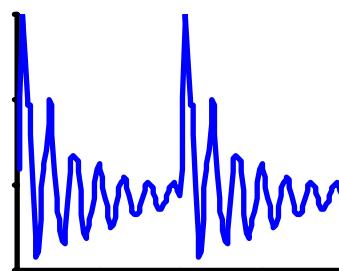
POPULATION-INTERVAL DISTRIBUTIONS FOR CONSTITUENT SINGLE VOWELS PRESENTED INDIVIDUALLY



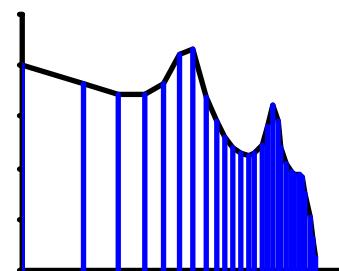
Concurrent vowels

ae-100

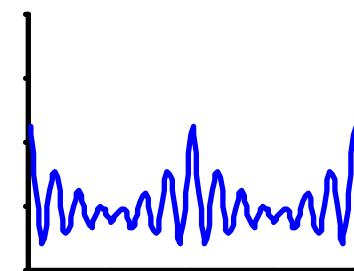
WAVEFORM



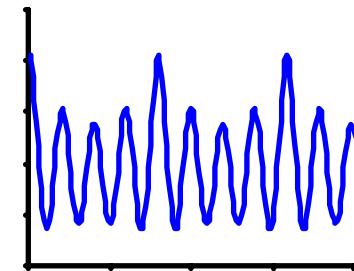
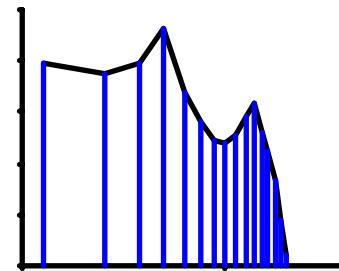
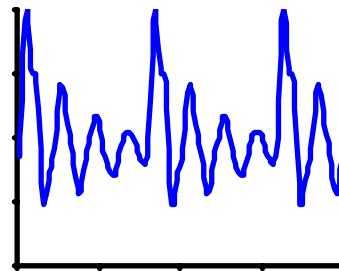
SPECTRUM



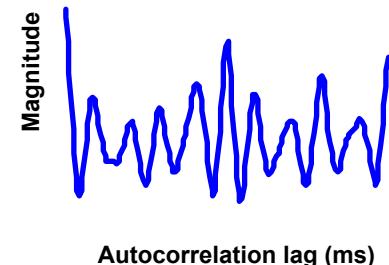
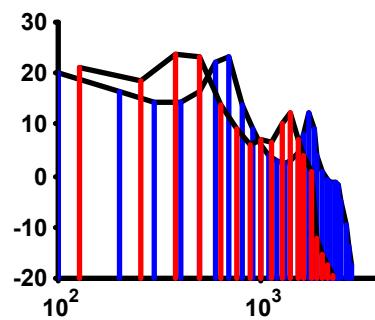
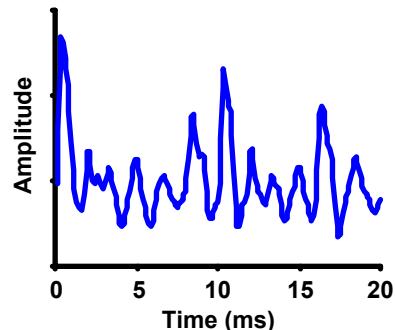
AUTOCORRELATION



er-124



ae-100
+
er-124



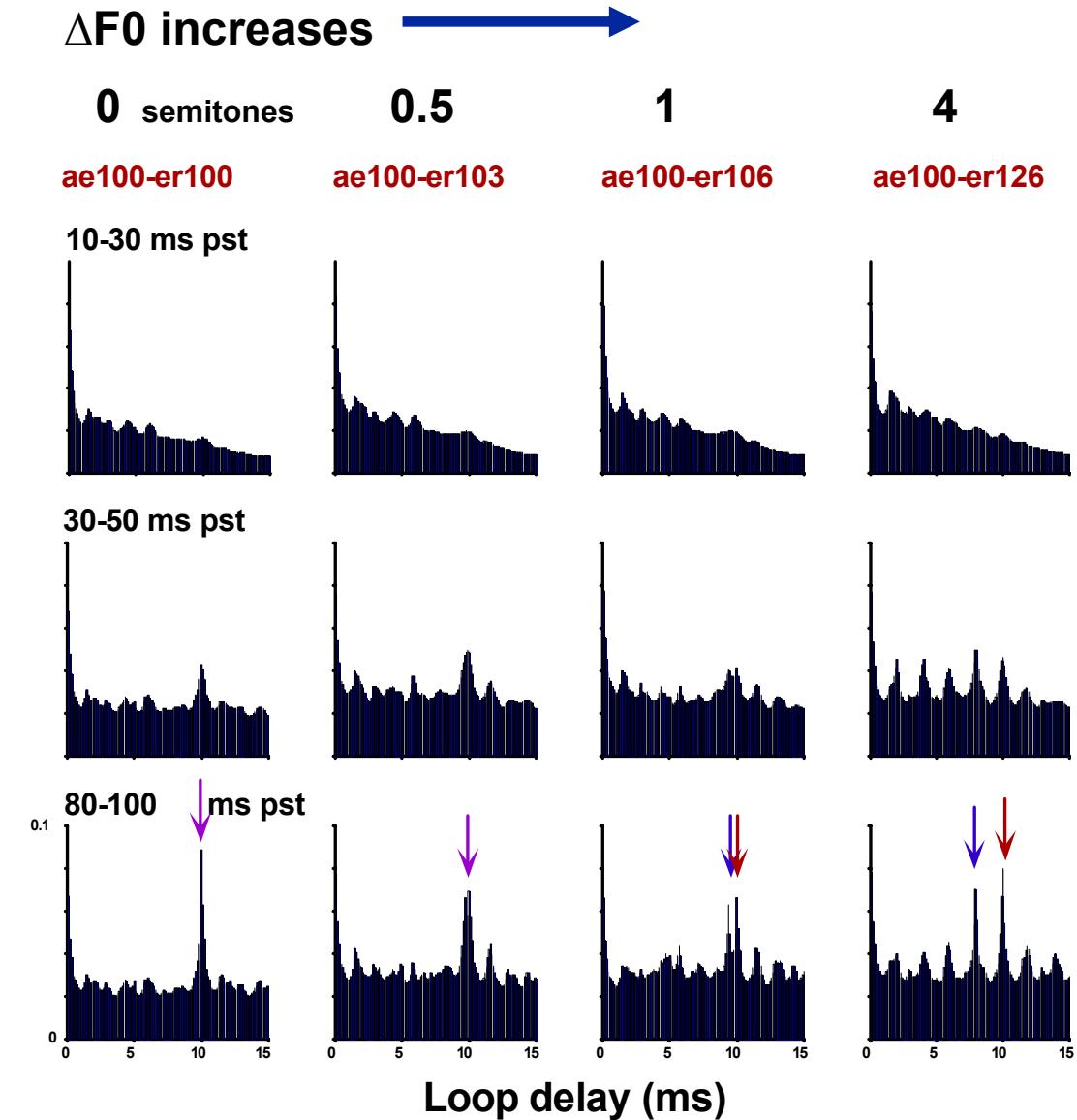
Separation of objects with increasing ΔF_0

Those sets of delay loops whose recurrence times equal the fundamental periods of the two vowels are invariably activated the most.

As time goes on, the signals present in these sets of delay loops are amplified.

When the two vowels have the same fundamental (100 Hz), only one set of delay loops is activated (10 ms delay).

When the two vowels have fundamentals separated by a semitone or more (**ae100-er106** & **ae100-er126**) two sets of delay loops are activated. The recurrence times of these sets of delay loops equal the respective vowel periods (10, 9.4, 7.9 ms).

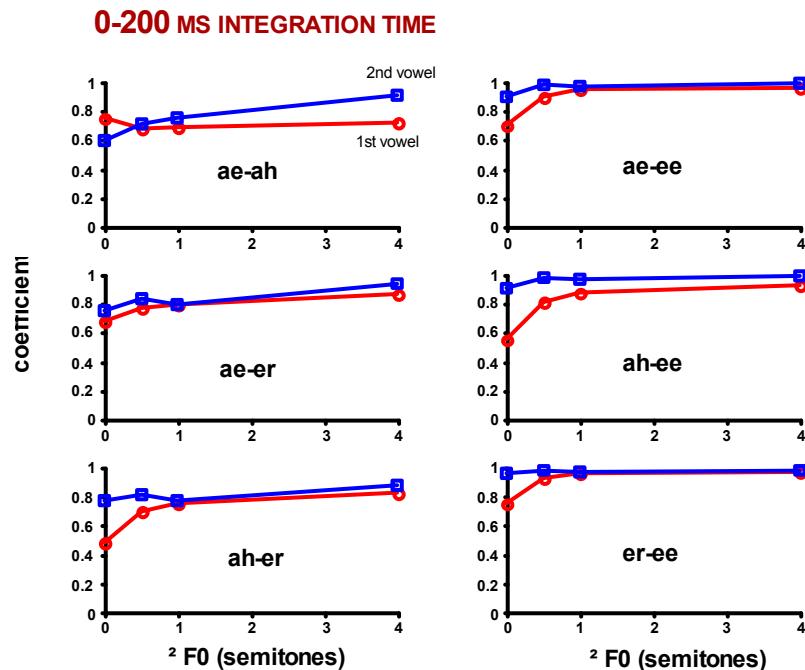


Simulation performance vs. psychophysics

Simulations results

For 200 ms integrations, one (dominant) vowel produced high correlations for all 2F0 's.

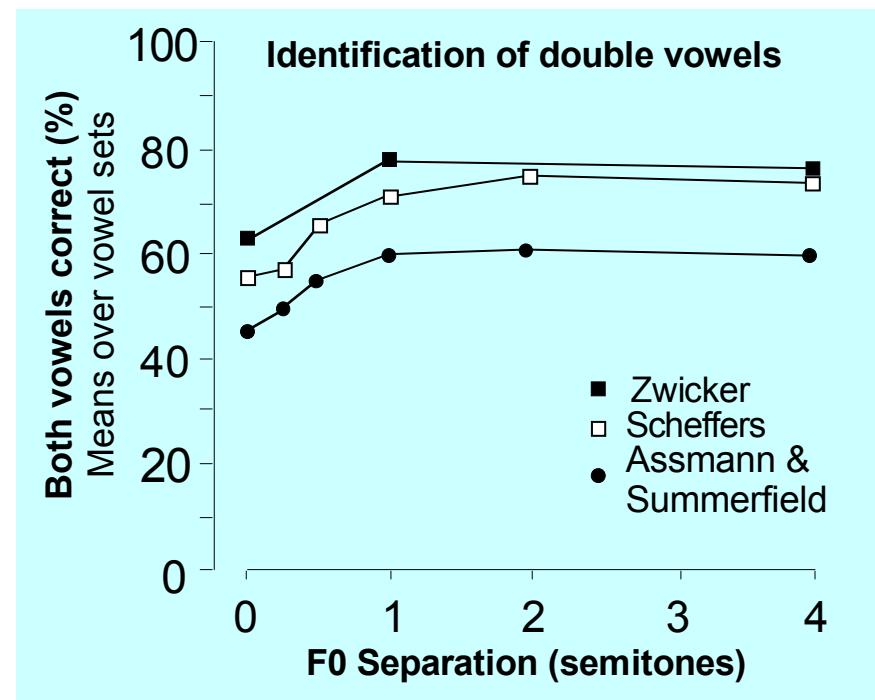
Correlations of nondominant vowels increased as 2F0 's increased to a semitone, then levelling off.



Psychophysical observations

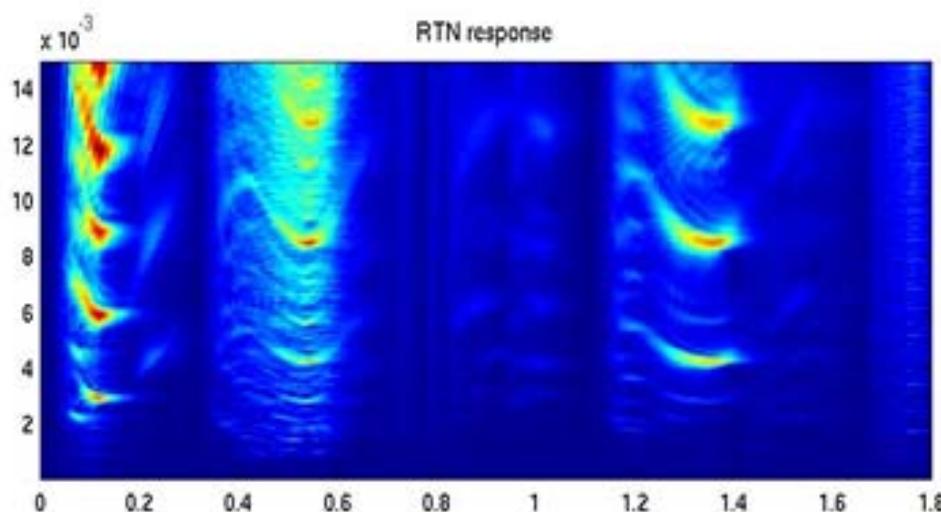
Listeners almost invariably get one vowel correct.

Correct identification of both vowels is slightly enhanced (~15%) by F0 differences of a semitone or more, where performance saturates.



Potential applications

- Pitch tracking of speakers & recovery of waveforms
- Separation of multiple speakers for speech recognition
- Enhance voiced speech segments in background noise
- Separation of tonal music from background noise
- Analogous 2-D nets for image stabilization and object formation from correlational invariants
- Reverberating temporal memories via transiently-facilitated synapses: periodic stimuli dynamically create loops on the fly



Conclusions

- Population-wide distributions of all-order interspike intervals in early stations of the auditory pathway explain many diverse aspects of pitch and timbre perception: level-invariance, discriminatory precision, pitch equivalence, octave similarity, pitch fusion.
- These distributions form autocorrelation-like temporal representations that convey spectral information up to the limits of strong phase-locking (~5 kHz).
- The mechanisms by which the central auditory system might make use of this information remain enigmatic.
- Neural timing nets can be envisioned that operate entirely in the time domain to carry out musical interval processing and scene analysis.

Reading/assignment for next meeting

- Thursday. Feb. 26
- Timbre
- Reading:
 - Moore Chapter 8, pp. 269-273
 - Chapter in Deutsch by Risset & Wessel on Timbre
(first sections up to p. 113-118)
 - Handel, chapter on Identification