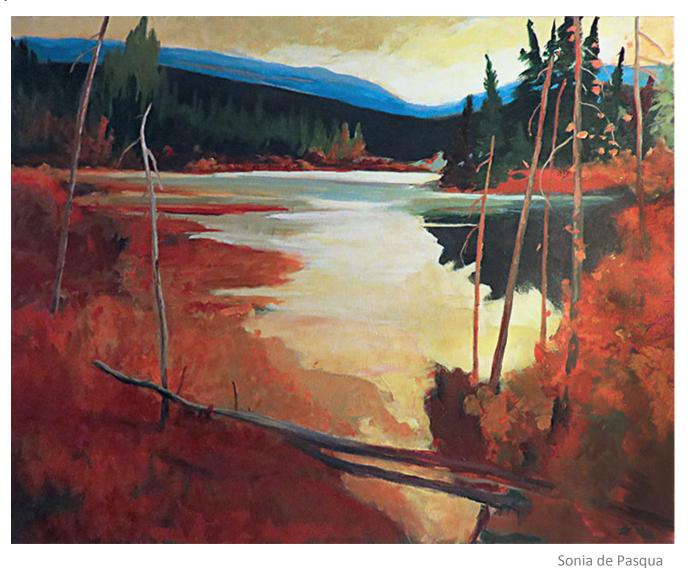
Interdisciplinary viewpoints of resonance and its role in cochlear mechanics

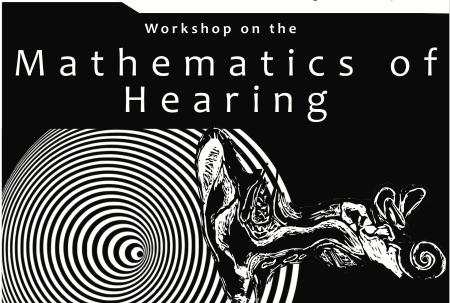


Christopher Bergevin

http://www.d-sho.com/ Toronto, ON Canada



THE FIELDS INSTITUTE
JUNE 16, 2017



The ear is a remarkable detector, encoding sound pressure into neural signals that carry myriad pieces of information about the world around us to the brain. It is also highly selective, decomposing sound into constituent frequency components by virtue of acting as a hydrodynamic Fourier analyzer. Interestingly, not only does the ear respond to sound, but emits it as well, a facet that has revolutionized pediatric audiology.

Yet much still remains unknown about this system we all have a pair of built in to our heads. This one-day workshop aims to bring together a variety of researchers from different backgrounds to explore, from both mathematical and biological viewpoints, a wide range of topics related to the ear.

Bard Ermentrout (University of Pittsburgh) – nonlinear dynamics
Andre Longtin (University of Ottawa) – neural coding
Laura Miller (University of North Carolina) – fluid dynamics
Christopher Shera (University of Southern California) – inverse problems
Sarah Verhulst (Ghent University) – cochlear neurobiology and psychoacoustical modeling
George Zweig (MIT) – cochlear modeling

ORGANIZING COMMITTEE: CHRISTOPHER BERGEVIN (YORK UNIVERSITY) AND SUNIL PURIA (HARVARD MEDICAL SCHOOL)

For registration and more information, please visit: www.fields.utoronto.ca/activities/16-17/mathhearing









13th Mechanics of Hearing conference

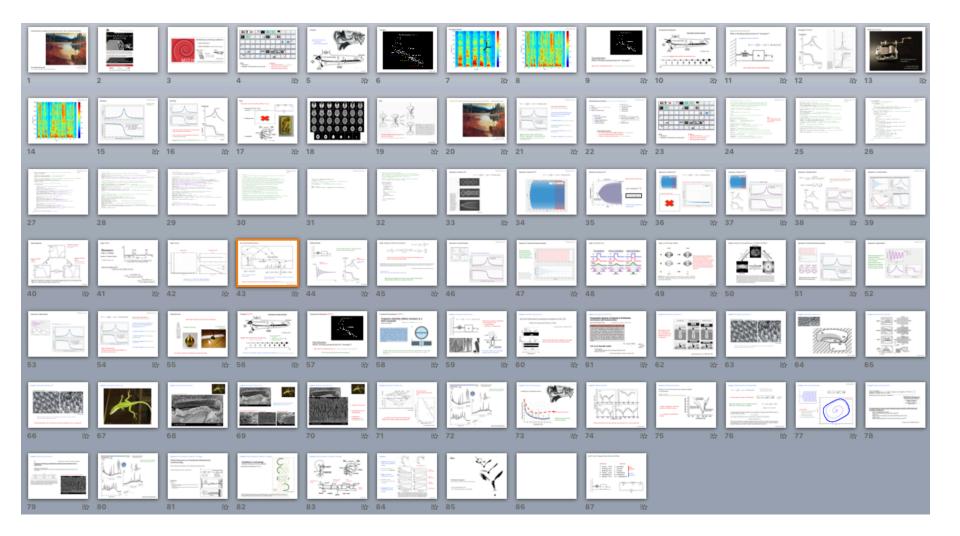
- ◆ June 19-24, 2017
- ◆ Brock University (St. Catherines, Ontario, Canada)

Niagara region Waves

Hair cells

... & much more

www.mechanics of hearing.org/moh 2017/

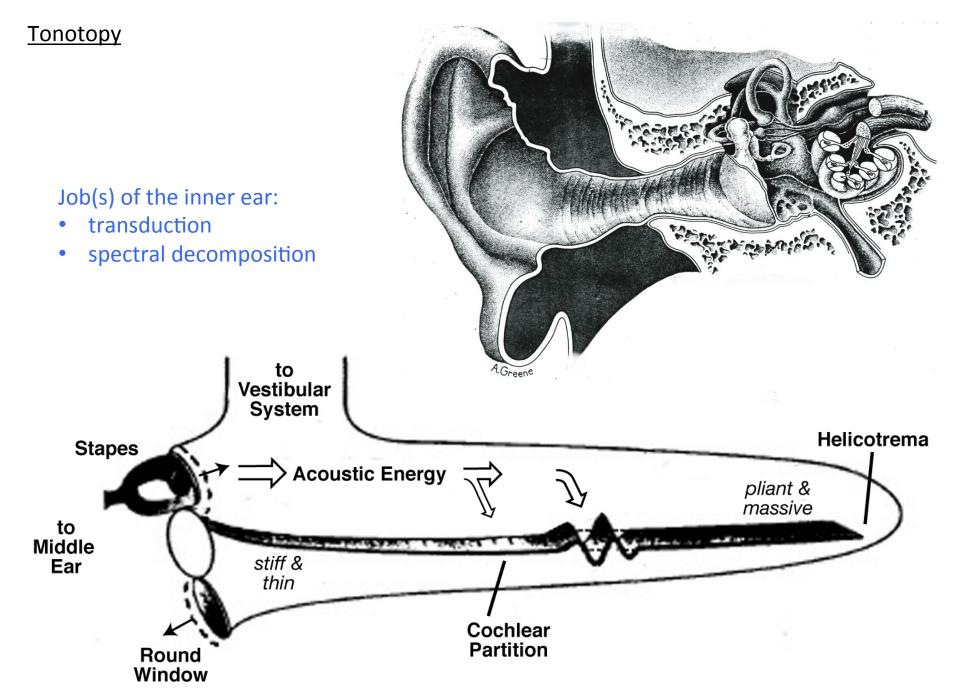


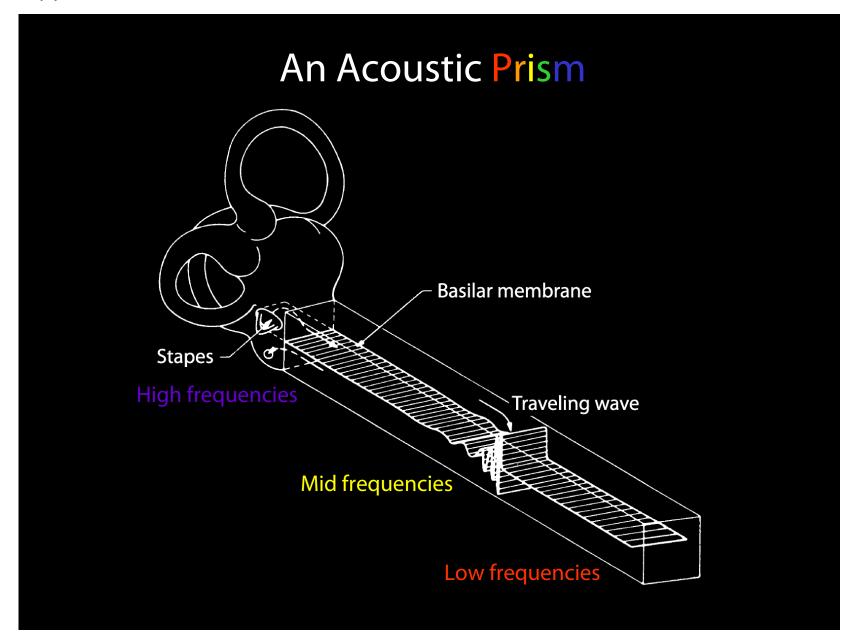
To Do

- Resonance
- Highlight "interdisciplinary" approaches

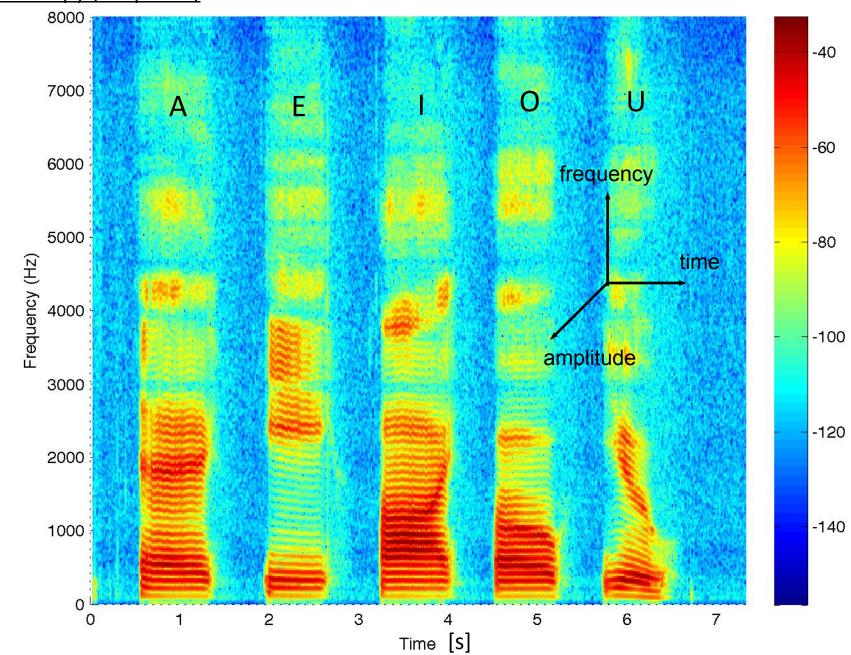
Tangents

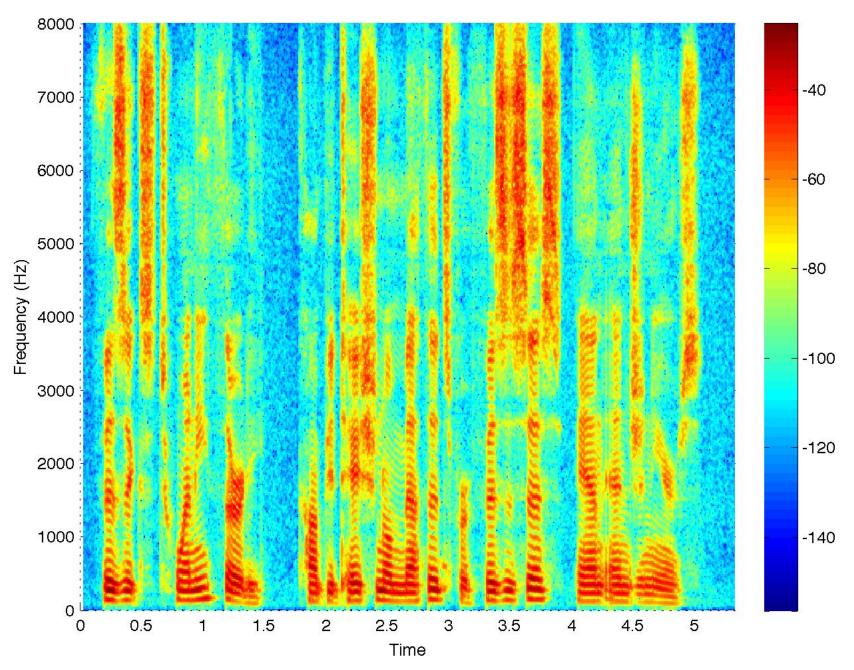
- Conceptual details (e.g., convolutions)
- Resonance in the inner ear?
- Nonlinear/active oscillators
- Other examples in biology

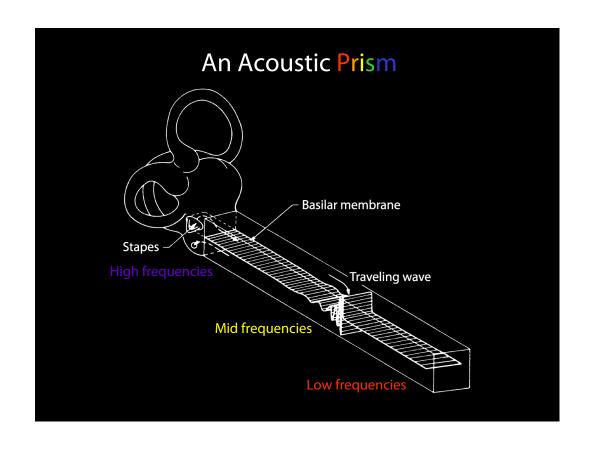




Tonotopy (re speech)







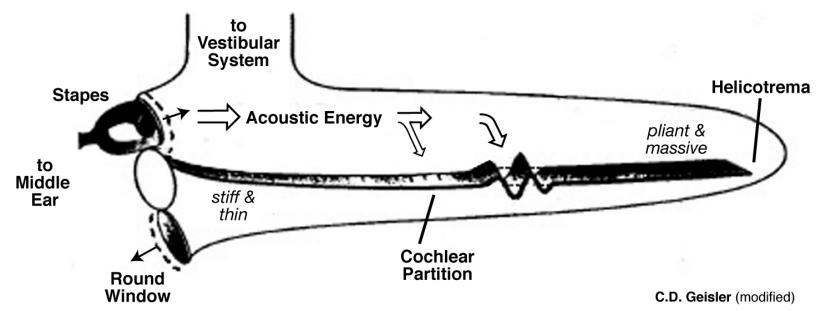
Theme/Question:

What is the (basic) physical basis for "tonotopy"?

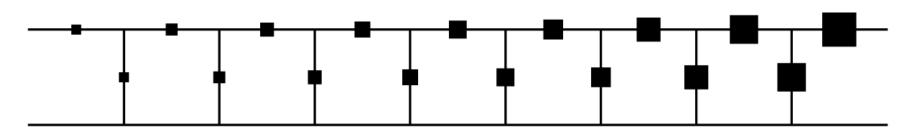
<u>Hint</u>: It ain't a traveling wave per se.... (though such provides a useful framework)

Tonotopy & Traveling waves

Mammalian Cochlea Uncoiled



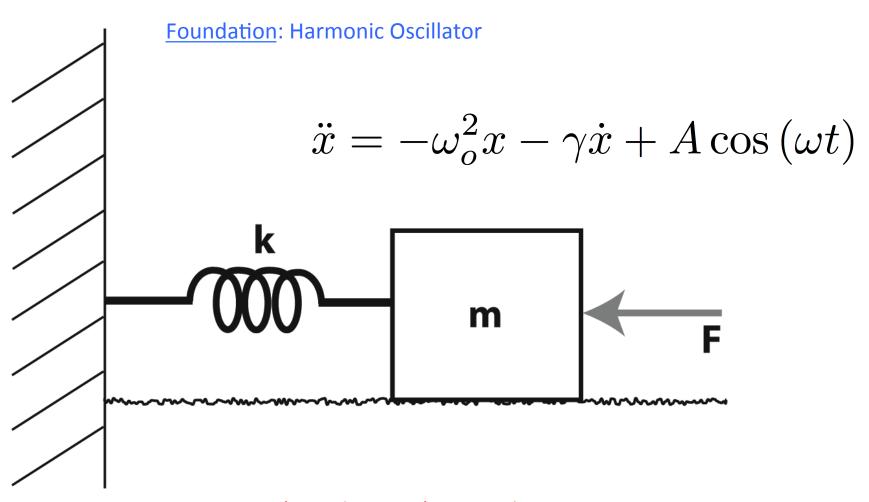
(one possible) Model: Non-uniform transmission line



 x_0

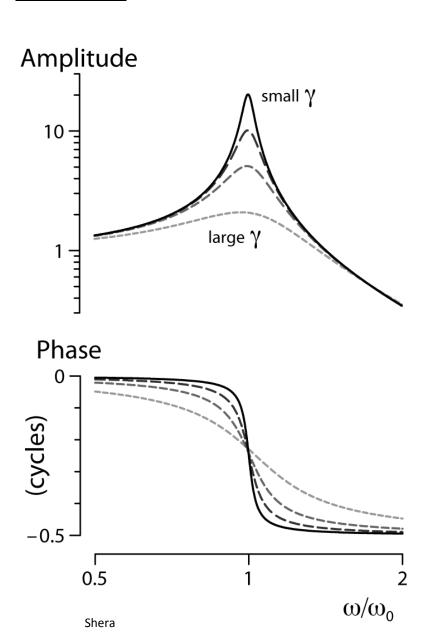
Big picture theme/question here:

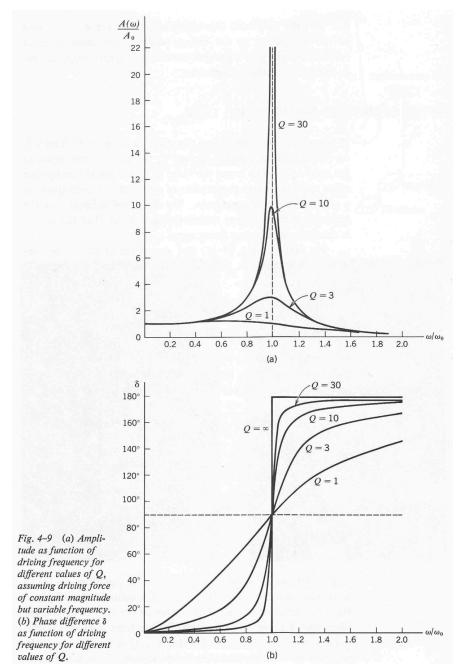
What is the (basic) physical basis for "tonotopy"?



Key (steady-state) principle: *Resonance*

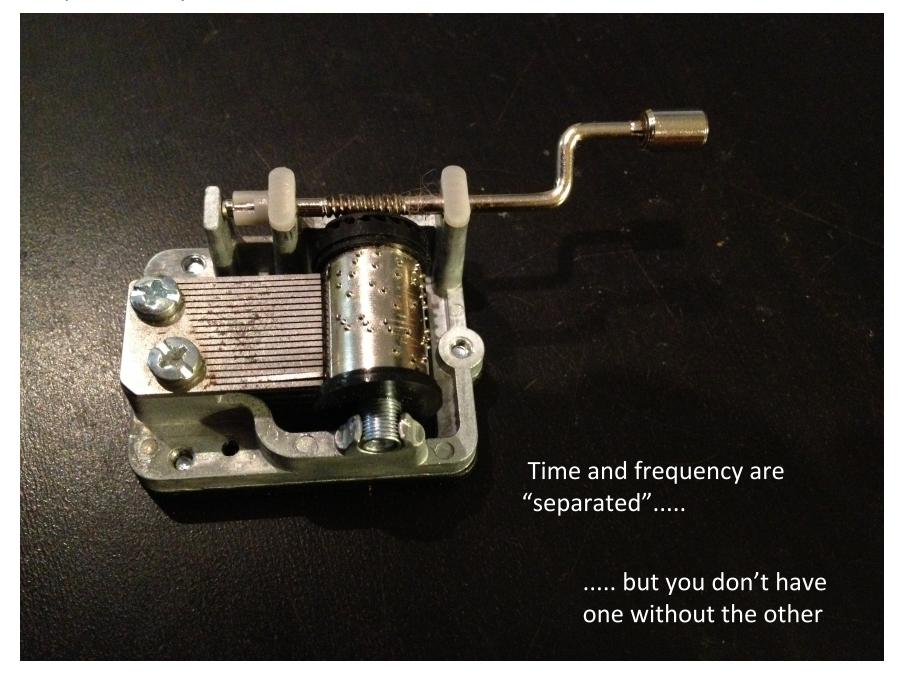
Foundation: Resonance

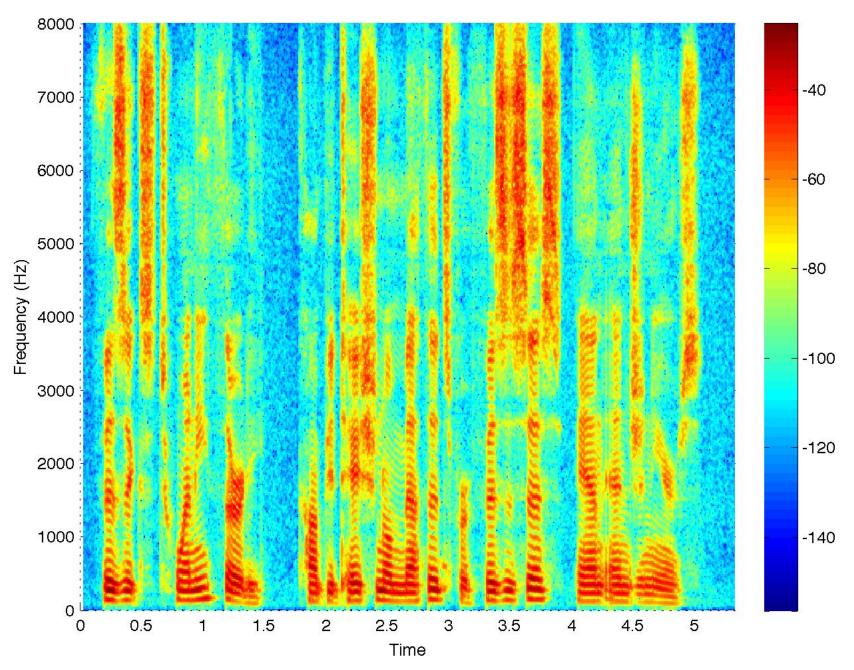




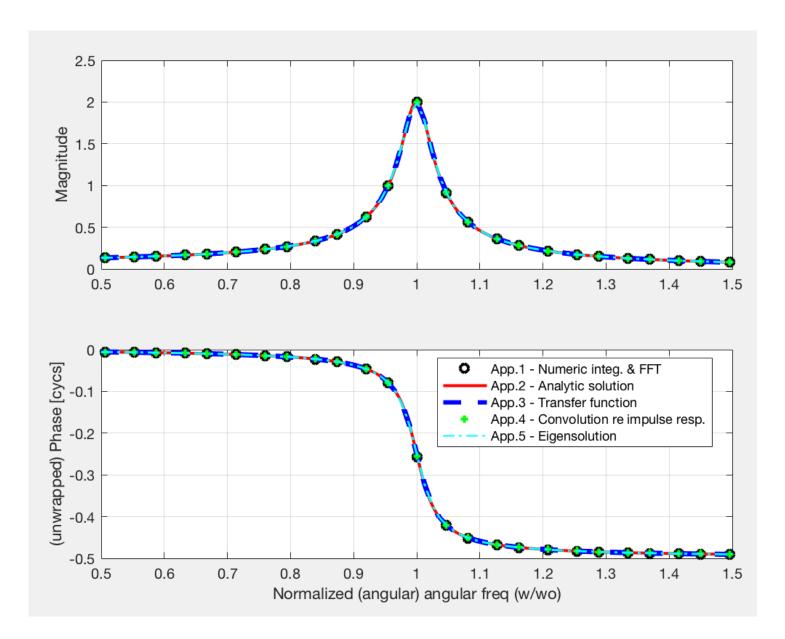
French (1971)

Aside: Spectral analysis



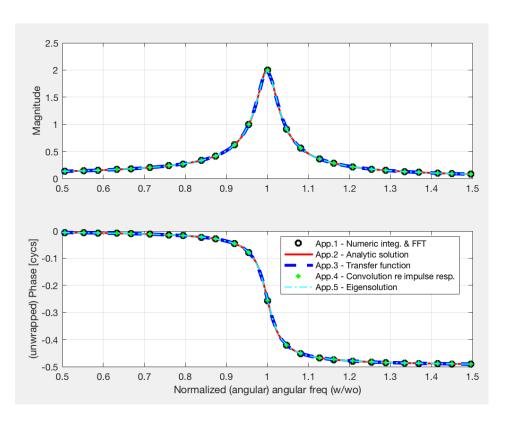


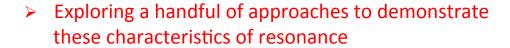
Focal Point

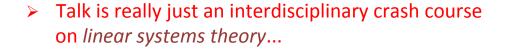


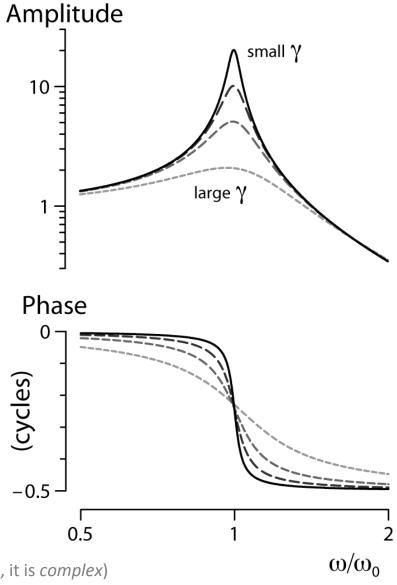
Relatively simple Matlab code...

Focal Point







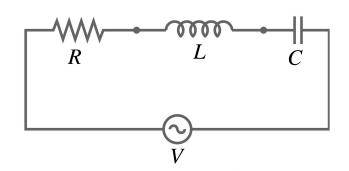


Nota bene: The inner ear is not really linear per se (i.e., it is *complex*)

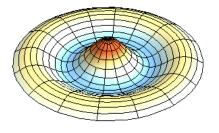
<u>Aside</u>

Resonance comes in a variety of "flavors", e.g.,:

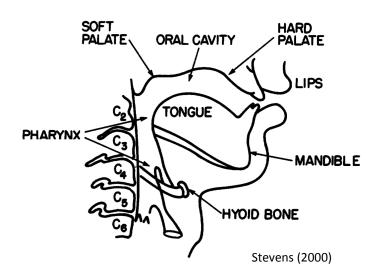
Externally forced 2nd order "systems" (i.e., energy is being input into them)



Standing waves

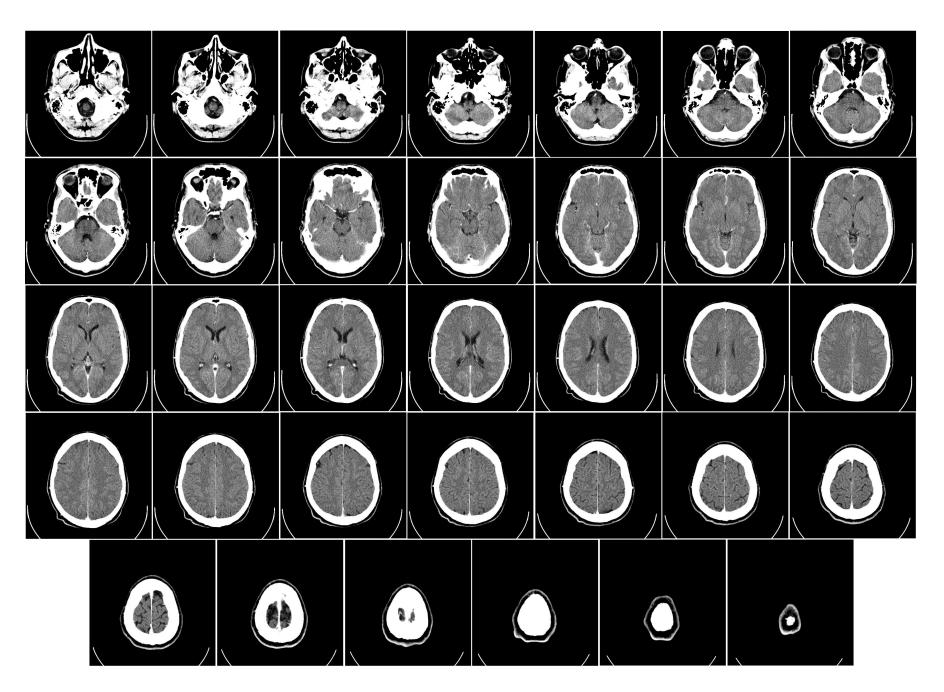


http://www.acs.psu.edu/drussell/demos/membranecircle/circle.html

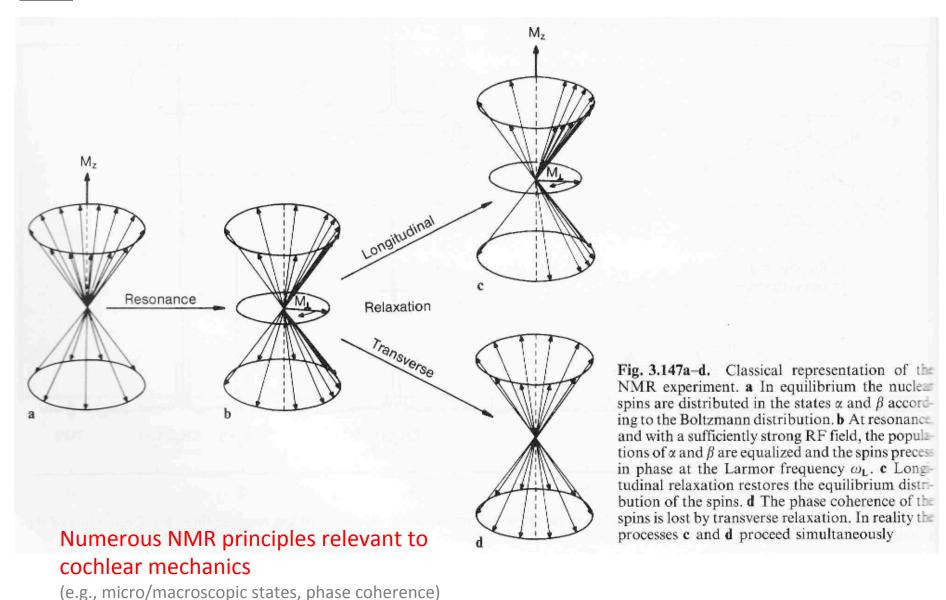




> NMR/MRI



Aside



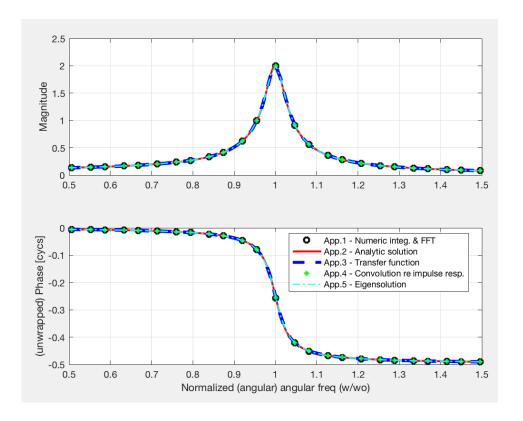
Hoppe et al (1983)

Resonance: The power to evoke enduring images, memories, and emotions — - Oxford dictionary, online



Sonia de Pasqua http://www.d-sho.com/

$$\ddot{x} = -\omega_o^2 x - \gamma \dot{x} + A\cos(\omega t)$$



<u>Note</u>: Via linear systems theory, these different "approaches" are not necessarily mutually exclusive (e.g., convolution theorem directly links #s 3 & 4)

Several basic approaches:

(all arriving at the same answer)

- 1. Numerically solve the ODEs and extract the relevant magnitudes and phases (via an FFT)
- 2. Analytic solution I (via Fourier transforms)
- 3. Impulse response I (and associated transfer function)
- **4.** Impulse response II (convolve in the time domain)
- 5. Analytic solution II (via eigensolutions)

Physics

- ODEs (e.g., Newton's 2nd, Hooke's Law)
- Resonance
- Notion of "steady-state"

Engineering

- Linear systems theory
- Convolutions
- Impedance/Admittance
- Impulse response
- Transfer functions

Mathematics

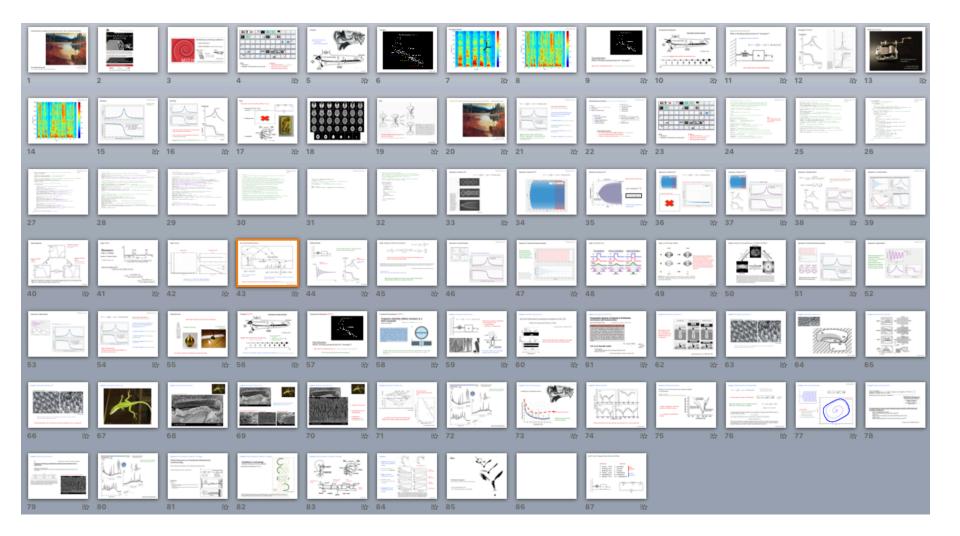
- Fourier transforms
- Complex #s
- Eigenvalues
- Phase space

Numerical

- Discrete Fourier transforms (FFT)
- Numerically solving ODEs (e.g., Euler, RK4, adaptive step-size and associated problems)
- Matlab syntax

Basic physical intuition:

- 2nd order system: two *reactive* elements (i.e., energy-storing)
- energy transferred back & forth between
- there is an optimal rate for such (i.e., resonant frequency)



To Do

- Resonance
- Highlight "interdisciplinary" approaches

Tangents

- Conceptual details (e.g., convolutions)
- Resonance in the inner ear?
- Nonlinear/active oscillators
- Other examples in biology

```
% ### EXhoResonance.m ###
                           2017.02.04 CB
% Code to solve the damped (sinusoidally-) driven harmonic oscillator (DDHO)
% for a variety of driving freqs. so to buildup the "resonance curve" via
% computation of the mag/phase of the Fourier transform of the steady-state
% response. Furthermore, the analytic solution for the DDHO as well as the
% transfer function are shown to be equivalent (Fig.1)
% Damped driven Harmonic Oscillator (DDHO)
d^2xdt^2 = -((P.wo)^2)*x - P.gamma*dx/dt + (P.A)*sin(P.w*t)
% Reqs:
% EXhoResonanceFunc.m (re ode45), rfft.m
clear;
% Oscillator params. and ICs
                                                                      Note:
P.p0 = 0.0; % Initial position {0}
                                                                      This is a slightly older version
P.v0 = 0.0; % Initial velocity {0}
                                                                      of the code (does not include
                   % resonant (angular) freq {10}
P.wo = 10;
P.gamma= 0.5; % damping coefficient {0.5}
                                                                      methods 4 & 5)
용 ___
% Sinusoidal driving term params.
P.A = 10; % Driving force amplitude {10}
P.wDrive= [5 15]; % start and end angular drive freqs. {[5 15]}
P.wDriveN= 25;
                      % # of drive freqs. to run {25}
% ---
P.tmax = 200; % Maximum time to solve [s; arb] \{200\}
             % sample rate for time step [Hz; arb] {150}
P.Npoints= 8192; % Number of points in time series for FFT, must be 2<sup>n</sup> {8192}
% ---
P.plotN= 1;
             % boolean re plotting the waveform and spectra for one driving freq.
{1}
P.plotNnum = round(P.wDriveN/2); % driving freq. index to plot {round(P.wDriveN/2)}
용 ___
P.solveType= 1; % 0-ode45, 1-hard-coded RK4 {1}
               % boolean re using a fixed step-size for ode45 {0}
P.stepF= 0;
```

EXhoResonance.m (cont)

```
용 ___
dt= 1/P.SR; % spacing of time steps
init0 = [P.p0 P.v0]'; % Column vector of initial conditions.
tspan = [0:dt:P.tmax]; % time interval for entire computation
tW=[0:1/P.SR:(P.Npoints-1)/P.SR]; % (shorter/later) time interval for FFT window
L = length(tspan); TW = L-(P.Npoints-1); % create offset point extracting FFT window
% ---
% create relevant freq. arrays (e.g., for FFT bin labeling)
freq= [0:P.Npoints/2]; % Note: these values are not angular (i.e., [freq]= 1/s, not
rads/s)
freq= P.SR*freq./P.Npoints;
df = P.SR/P.Npoints;
                             % freq. spacing between bins
wDT= linspace(P.wDrive(1), P.wDrive(2), 500); % create ang. freq. array for plotting
analytic solution
% ++++++++++++
% various relevant derived quantities (used post- main for loop)
                 % "quality factor" (Note: tau=1/P.qamma=Q/P.wo, where tau is time
Q= P.wo/P.gamma;
const. of build-up)
lambdaP= 0.5*(-P.gamma+ sgrt(P.gamma^2-4*P.wo^2)); % Eigenvalues, for x=0 (undriven)
lambdaM= 0.5*(-P.gamma- sgrt(P.gamma^2-4*P.wo^2));
% Note - Can also get eigenvalues via command: eig([0 1;-P.wo^2 -P.gamma])
Z= P.gamma+ i*(wDT- P.wo^2./wDT); % impedance (see notes above; assumes mass is
unity)
Y = 1./Z:
           % admittance (reciprocal of impedance)
용 ____
% grabbing driving freqs. from freq array
%%indx= find(freq>=P.fDrive(1) & freq<=P.fDrive(2)); % find relevant indicies</pre>
indx= find(freq>=P.wDrive(1)/(2*pi) & freq<=P.wDrive(2)/(2*pi)); % find relevant</pre>
indicies
subset
freqD= 2*pi*freq(indxB); % array of driving angular freqs
```

```
for mm=1:numel(freqD)
    P.w= freqD(mm); % extract driving freq.
     % *** Solve in one of two ways ***
    if P.solveType==0
        % use Matlab's ode45
        % ---
        % tell it to actually use the specified step-size
        if(P.stepF==1), options = odeset('MaxStep',1/P.SR); else options=[];
                                                                              end
       [t,y] = ode45(@EXhoResonanceFunc,tspan,init0,options,P);
    else
        % use 4th order Runge-Kutta code
        xPoints(1) = P.p0; vPoints(1) = P.v0; % initialize ICs into dummy arrays
        x= P.p0; v= P.v0; % kludge
        dt= 1/P.SR; % time step
        for nn=1:(length(tspan)-1)
            % ---
           t = tspan(nn); % Current time.
            % step1
           xk1=v:
           vk1 = -((P.wo)^2)*x - P.gamma*v + (P.A)*sin(P.w*t);
            % step 2
           xk2 = v + (dt/2)*vk1;
           vk2 = -((P.wo)^2)*(x + (dt/2)*xk1) - P.gamma*(v + (dt/2)*vk1)...
                + (P.A)*sin(P.w*(t+(dt/2)));
            % step 3
           xk3 = v + (dt/2)*vk2;
           vk3 = -((P.wo)^2)*(x + (dt/2)*xk2) - P.qamma*(v + (dt/2)*vk2)...
               + (P.A) * sin(P.w*(t+(dt/2)));
            % step 4
            xk4 = v + dt*vk3;
           vk4 = -((P.wo)^2)*(x + (dt)*xk3) - P.gamma*(v + dt*vk3)...
                + (P.A)*sin(P.w*(t+(dt/2)));
            % apply RK4 weighting
           x = x + (dt/6)*(xk1 + 2*xk2 + 2*xk3 + xk4);
           v = v + (dt/6)*(vk1 + 2*vk2 + 2*vk3 + vk4);
            % store away position and velocity
            xPoints(nn+1) = x; vPoints(nn+1) = v;
        end
       y(:,1)= xPoints'; y(:,2)= vPoints'; % repackage output
    end
```

```
읭 ___
                                                                                            (cont)
   ySPEC= y(TW:TW+P.Npoints-1,1); % steady-state portion of waveform for FFT
   sigSPEC= rfft(ySPEC);
    읭 ___
   wDrive(mm) = 2*pi*freq(indxB(mm)); % store away driving freqs.
   mag(mm) = abs(sigSPEC(indxB(mm))); % store away SS mag.
    용 ___
    % need to correct the phase re the duration of the window allowed for settling into steady-state
    tPhase= angle(sigSPEC(indxB(mm))); % extract the phase
   tPhase= angle(exp(i*(tPhase- wDrive(mm)*tspan(TW)))); % correct phase re onset
    phase(mm)= tPhase;
    용 ___
    % visualize relevant bits for one of the drive freqs.
   if mm==P.plotNnum
        읭 ___
        % integrated waveform and segment extracted for spectral analysis
        figure(2); clf;
        h1= plot(tspan,y(:,1)); hold on; grid on;
        xlabel('Time'); ylabel('Position');
        title('Time Waveform of integrated solution to damped driven HO equation')
       L = length(tspan); TW = L-(P.Npoints-1); % create offset point
        ySPEC= y(TW:TW+P.Npoints-1,1); % steady-state portion of waveform for FFT
        h2= plot(tspan(TW:TW+P.Npoints-1),ySPEC, 'r.', 'MarkerSize',3);
        legend([h1 h2], 'Entire waveform', 'Steady-state portion (used for FFT)')
        용 ___
        % phase space for waveform (entire and steady-state)
        figure(3); clf;
        hPS1= plot(y(:,1),y(:,2)); hold on; grid on;
        hPS2= plot(y(TW:TW+P.Npoints-1,1),y(TW:TW+P.Npoints-1,2),'r.-');
        xlabel('Position'); ylabel('Velocity'); title('Phase plane');
        legend([hPS1 hPS2], 'Entire waveform', 'Steady-state portion (used for FFT)')
        용 ____
        % plot spectra of steady-state waveform
        figure(4); clf;
        hS1= plot(2*pi*freq,db(sigSPEC)); hold on; grid on;
        xlabel('Freq [rads/s]'); ylabel('Spectral amplitude [dB]');
        hS2= plot(2*pi*freq(indxB(mm)),db(mag(mm)),'rs'); % indicate extracted freq.
        legend([hS1 hS2], 'Steady-state spectra', 'Driving freq.');
    end
    % ---
   disp([num2str(100*mm/numel(freqD)),'% done']);
end
```

```
% ++++++++++++
                                                                                   (cont)
% [Fig.1] ** Mags/phases extracted from the numeric steady-state responses **
figure(1); clf;
subplot(211); hh1= plot(wDrive/P.wo,mag,'ko','MarkerSize',6,'LineWidth',2); hold on; grid on;
ylabel('Magnitude');
subplot(212); hh2= plot(wDrive/P.wo,unwrap(phase)/(2*pi), 'ko', 'MarkerSize', 6, 'LineWidth', 2); hold on; grid on;
xlabel('Normalized (angular) angular freq (w/wo)'); ylabel('(unwrapped) Phase [cycs]');
% ++++++++++++
% [Fig.1] ** Analytic solution ** (see French, 1971; as noted above, these expressions are
% equivalent to using Fourier transforms, which implicitly assume sinusoidal steady-state, to
% solving the main ODE)
magT= P.A./sqrt((P.wo^2-wDT.^2).^2 + ((P.gamma*wDT).^2)); % mag (theory)
phaseT= atan((P.gamma*wDT)./(-P.wo^2+wDT.^2));
                                                  % phase (theory; note sign change in denom. re
convention)
figure(1);
subplot(211); hh3= plot(wDT/P.wo,magT,'r-','LineWidth',2);
subplot(212); hh4= plot(wDT/P.wo,unwrap(2*phaseT)/(4*pi),'r-','LineWidth',2); % kludge to get unwrapping
working
% ++++++++++++
% [Fig.1] ** "Transfer function" ** re linear systems theory (i.e., the Fourier transform of the
% impulse response of the DHO)
init0 = [0 \ 10]'; % set ICs such that there is an "impulse" at t=0
P.w= 0; % make sure to "turn off" drive
options= []; [t,yI] = ode45(@EXhoResonanceFunc,tspan,init0,options,P);
specI= rfft(yI(1:P.Npoints));
magI= abs(specI);
magI= magI* (max(mag)/max(magI));
                               % scale impulse mag. re max. value of driven case
phaseI= angle(specI);
figure(1);
subplot(211); hh5= plot(2*pi*freq/P.wo,magI, 'b--', 'LineWidth',2); xlim([wDT(1) wDT(end)]/P.wo);
subplot(212); hh6= plot(2*pi*freq/P.wo,unwrap(2*phaseI)/(4*pi), 'b--', 'LineWidth',2); % kludge to get unwrapping
working
xlim([wDT(1) wDT(end)]/P.wo);
```

```
EXhoResonance.m
% ++++++++++++
                                                                                           (cont)
% [Fig.1] Make a legend to put it all together (re Fig.1)
figure(1); subplot(211); legend([hh1 hh3 hh5], 'Numeric solution re steady-state FFT',...
    'Analytic solution', 'Transfer function');
% +++++++++++
% [Fig.5] Plot the impulse response and comparison to admittance
figure(5); clf;
subplot(221); plot(tW,yI(1:P.Npoints)); hold on; grid on; xlabel('Time [s]'); ylabel('x');
title('Impulse response (no drive; P.w=0, P.p0=0, P.v0=10'); xlim([0 tW(round(numel(tW)/3))]);
subplot(222); hI2= plot(freq,db(specI),'LineWidth',2); grid on; hold on; ylabel('Amplitude [dB]');
title('Transfer function (mag. of FFT of IR)'); xlim(P.wDrive/(2*pi));
subplot(224); hI3= plot(freq,angle(specI)/(2*pi), 'LineWidth',2); grid on; hold on;
xlabel('Frequency [Hz]'); ylabel('Phase [cycles]');
title('Transfer function (phase of FFT of IR)'); xlim(P.wDrive/(2*pi));
subplot(223);
hZa = plot(wDT, abs(Y), 'k-'); grid on; hOld on; hZb = plot(wDT, abs(Z), 'r.');
grid on; hold on; ylabel('Amplitude'); xlabel('Ang. requency [rad/s]'); legend([hZa
hZb], 'admittance', 'impedance');
용 ___
% for reference, also include (scaled) admittance to indicate (near?) equivalence
offset= max(db(Y)) - max(db(specI)); % scaling (in dB) to match up
%magY= fliplr(db(Y)- offset); % kludge
magY= (db(Y)- offset);
subplot(222); hI2b= plot(wDT/(2*pi), magY, r--);
legend([hI2 hI2b], 'Transf. func.', '(scaled) Admittance', 'Location', 'SouthWest');
angleY= angle(Y)/(2*pi) - angle(Y(1))/(2*pi); % there will be a slight vert. ofseet re angle(specI)/
(2*pi)
subplot(224); hI3b= plot(wDT/(2*pi),angleY, r--);
% ++++++++++++
% display some relevant #s to screen
disp(['Quality factor (P.wo/P.gamma)= ',num2str(Q)]);
```

disp(['Eigenvalues (for x=0, undriven case): ',num2str(lambdaP),' and ',num2str(lambdaM)]);

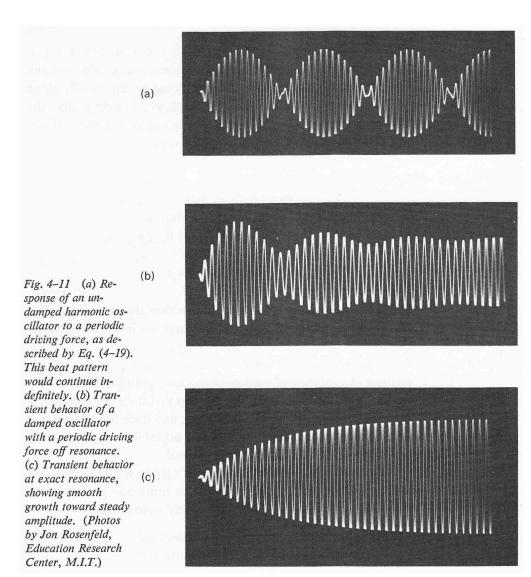
```
% Notes
% o To solve this numerically, need to turn 2nd order ODE into series of 1st order ODEs:
   dy/dt = -P.wo^2*x - P.gamma*y + (P.A)*sin(P.w*t)
% o For autonomous case (i.e., no drive), can rewrite in matrix form such that
% A= [0 1;-P.wo^2 -P.gamma]; straight-forward to find associated eigenvalues (see below)
% o via P.solveType, user can solve either via ode45 or a hard-coded RK4
% (both should yield the same solution!); Note that (surprisingly) ode45 actually seems
% slower than the RK4 (the slowest is ode45 w/ the fixed step-size), possibly due to
% the passing of the large-ish structure P; also note that the default ode45 routine
% (i.e., adaptive step-size) introduces harmonic disortions in the spectra
% due to its nonlinear nature
% o For the analytic solution (below manifest as magT and phaseT), the
% expression used below, as derived in French (1971) for the steady-state,
      exactly the same as if one simply put in the Fourier transform and
% solved for the resulting magnitude and phase [confirmed on the back of an
% envelope; let x(t) = X(w) \exp(i*w*t) and plug back in, solving for X(w); note then
% that magT=abs(X) and phaseT= angle(X)]
% o There are a few minor kludges below [e.g., vertical adjustment of the
% analytic solution so to match the (arbitrary?) ref. phase of the numeric
% solution)
% o Impedance (Z) for DDHO is (by definition) the complex ratio of the driving
% force and the (steady-state) velocity (see 4080W2016L10REF.pdf). Real part of Z (resistance)
% describes energy loss while imaginary part (reactance) describes energy storage
% o Comparison of the mags. for the transfer function and admittance
% (Fig.5, top right) are a bit kludgy (unsure why fliplr was needed) and
% off (worser overlap as you move away from wo)
```

```
function dy = EXhoResonanceFunc(t,y,P)
% Damped driven HO
% d^2xdt^2 = -((P.wo)^2)*x - P.gamma*dx/dt + (P.A)*sin(P.w*t)
% Note: y(1) = x, y(2) = dx/dt
dy = zeros(2,1);  % A column vector to be returned

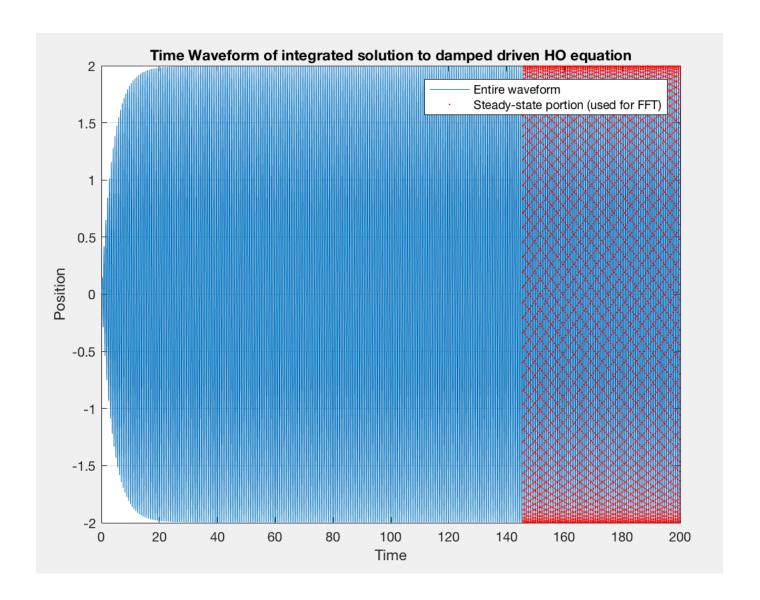
dy(1) = y(2);
dy(2) = -((P.wo)^2)*y(1) - P.gamma*y(2) + (P.A)*sin(P.w*t);
```

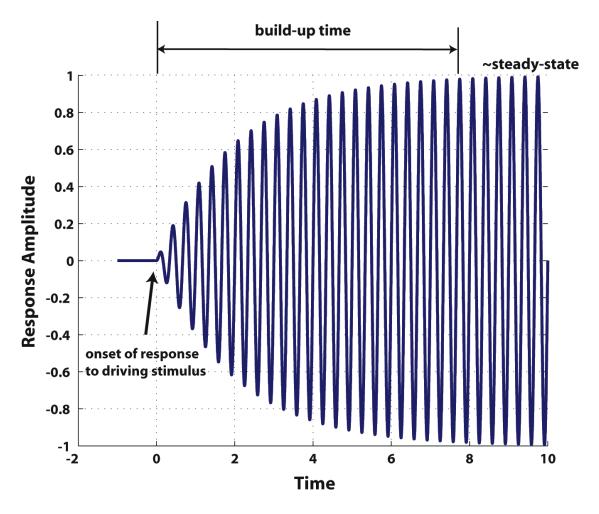
```
% RFFT: scaled real FFT, X=rfft(x)
% Returns the positive-frequency half of the transform X=FFT(x).
% The transform X is normalized so that if \{x\} is a sine wave of
% unit amplitude and frequency n*df, then X[n]=1.
% Usage:
          X=rfft(x)
% If x is N points long, NF=N/2+1 complex points are returned.
% See also IRFFT, FAST, FSST, FFT, IFFT,
function X=rfft(x)
  [m,n]=size(x);
  if (m==1 | n==1)
    % original...
   N=length(x)/2+1;
   xc=fft(x);
   X=xc(1:fix(N));
  else
    % do it column-wise...
   N=m/2+1;
   xc=fft(x);
   X=xc(1:fix(N),:);
  end
 X = X / (length(x)/2);
 return
```

$$\ddot{x} = -\omega_o^2 x - \gamma \dot{x} + A\cos(\omega t)$$



$$\ddot{x} = -\omega_o^2 x - \gamma \dot{x} + A \cos(\omega t)$$





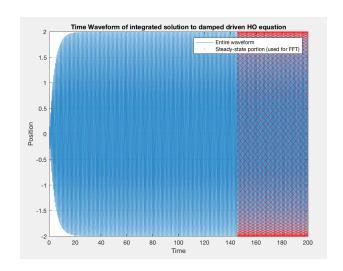
Note: Resonance takes time...

$$x(t) = A(\infty) \left[1 - e^{(-t/\tau)}\right]$$

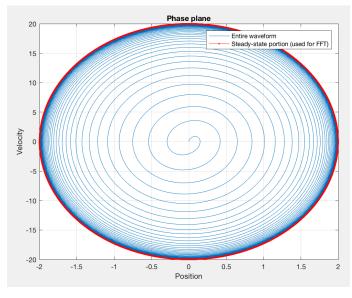
$$\tau = 1/\gamma = Q / \omega_o$$

→ Hence the importance of "steady-state"

Approach 1 – Numeric + FFT

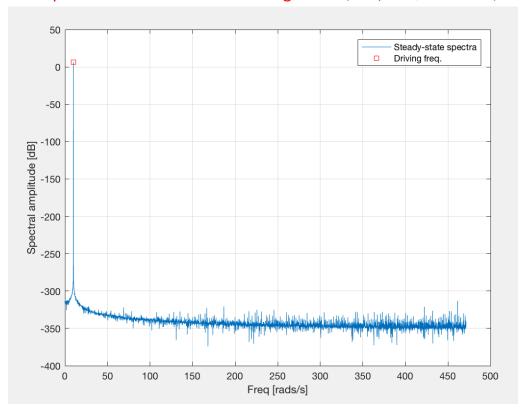


Can also plot in phase space...

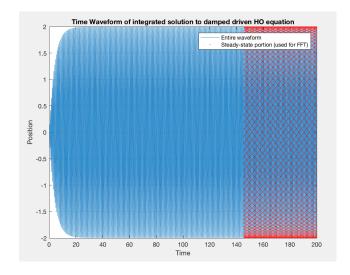


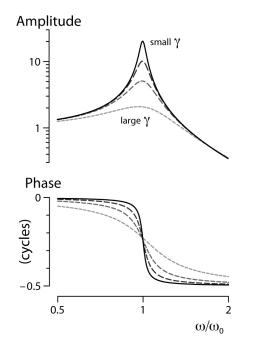
$$\ddot{x} = -\omega_o^2 x - \gamma \dot{x} + A \cos(\omega t)$$

Compute the FFT to extract the magnitude (and phase; not shown)

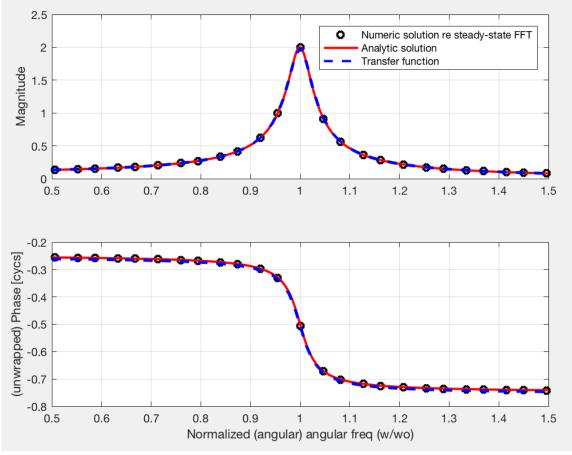


<u>Approach 1 – Numeric + FFT</u>





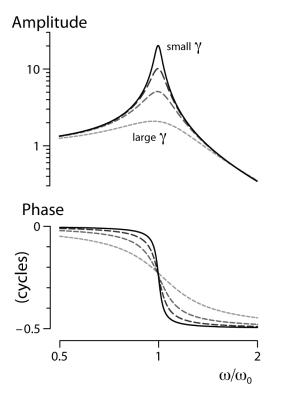
→ End up w/ the black circles....



Approach 2 – Analytic solution

$$A(\omega) = \frac{F_o/m}{[(\omega_o^2 - \omega^2)^2 + (\gamma \omega)^2]^{1/2}}$$

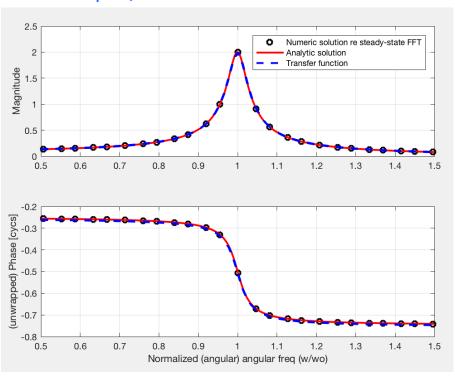
$$\delta(\omega) = \arctan\left(\frac{\gamma\omega}{\omega^2 - \omega_o^2}\right)$$



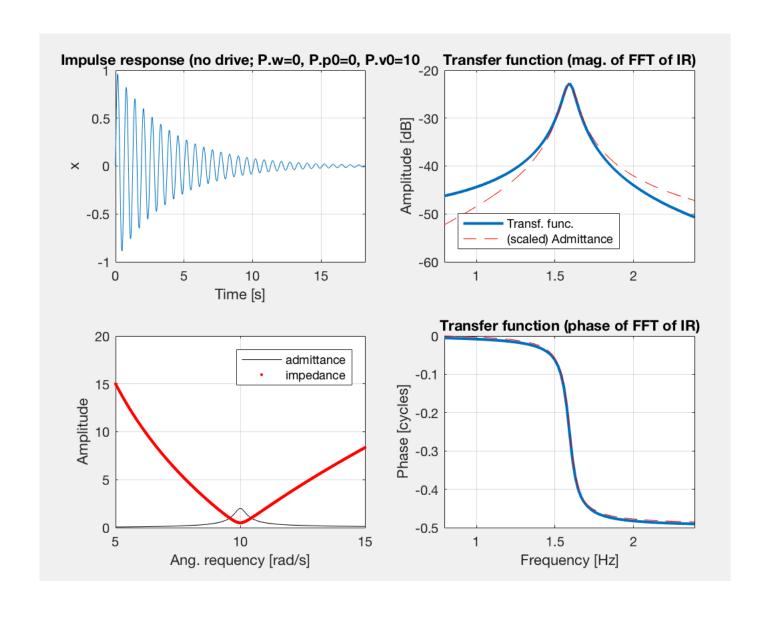
Can arrive here in a variety of ways

(including Fourier transforms; see notes at end)

→ End up w/ red line



Approach 3 – Transfer Function



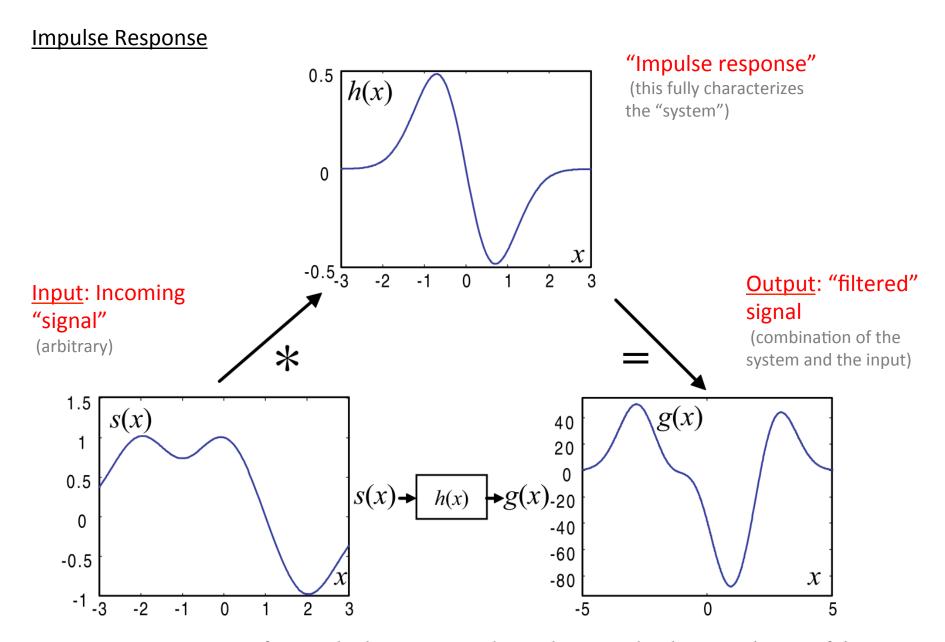


Fig. 4.11. Transmission of a signal. The transmitted signal is given by the convolution of the signal s(x) with the system's impulse response h(x)

Aside: Impulse

Resonance

CARL A. LUDEKE

JOURNAL OF APPLIED PHYSICS

VOLUME 13, JULY, 1942

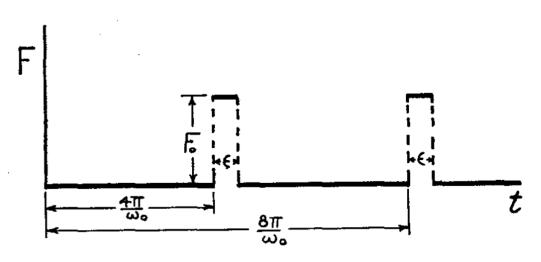


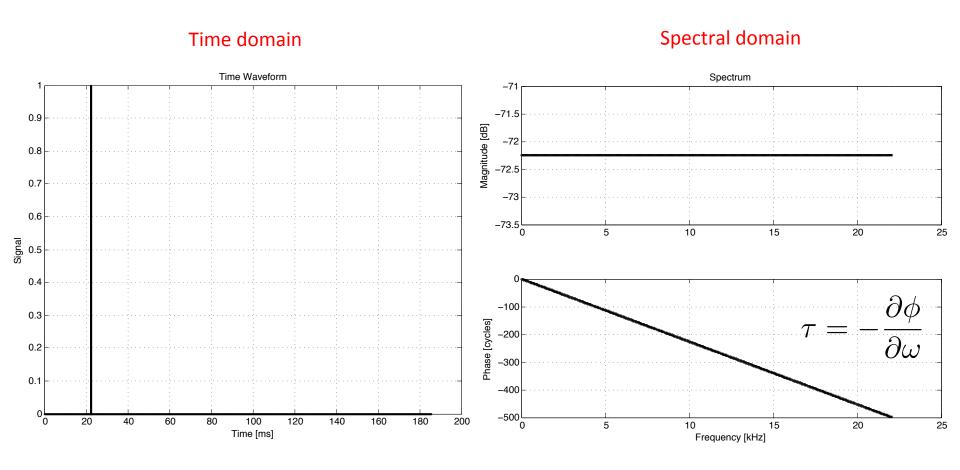
Fig. 4. The discontinuous force F supplied by the motor in Fig. 3, as a function of time t.

$$md^{2}x/dt^{2} + \beta dx/dt + kx$$

$$= \sum A_{n} \sin n\omega t + \sum B_{n} \cos n\omega t.$$

$$x = \sum \frac{A_n \sin (n\omega t - \varphi_n)}{\left[(k - mn^2 \omega^2)^2 + \beta^2 n^2 \omega^2 \right]^{\frac{1}{2}}} + \sum \frac{B_n \cos (n\omega t - \alpha_n)}{\left[(k - mn^2 \omega^2)^2 + \beta^2 n^2 \omega^2 \right]^{\frac{1}{2}}},$$

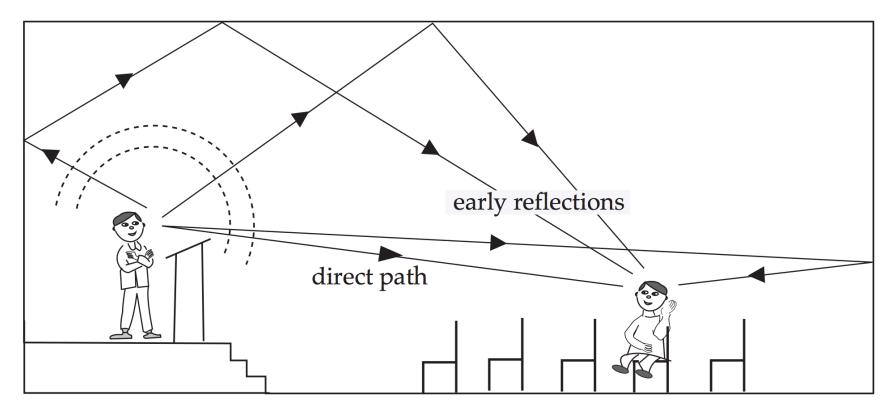
Aside: Impulse



Impulse (i.e., a "click") has a flat magnitude

(this is also a good place to mention the notion of a 'group delay')

Ex. Acoustic Impulse Response



$$g(x) = \mathcal{L}\{s(x)\}\$$

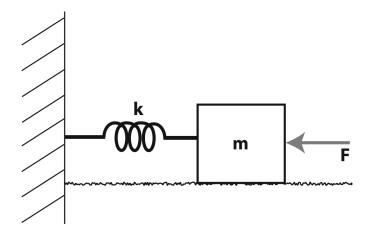
Room response (g) "filters" an input sound (s)

$$g(x) = s(x) * h(x) = \int_{-\infty}^{+\infty} s(\xi)h(x - \xi) d\xi$$

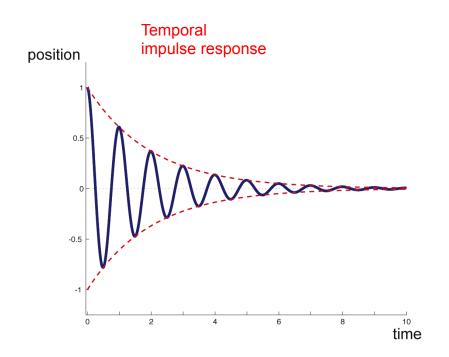
Room response (g) is just "convolution" between s and room's impulse response (h)

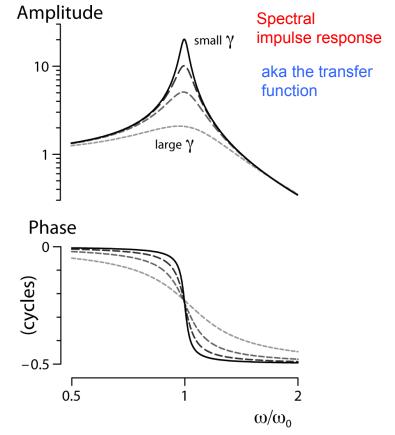
 \rightarrow All the relevant bits of the room's acoustics are contained in h (which we can easily measure!)

Transfer functions



→ The "transfer function" is simply the Fourier transform of the impulse response





Aside: Impedance & Pole/zero descriptions

$$\ddot{x} + \gamma \dot{x} + \omega_o^2 x = \frac{F_o}{m} e^{i\omega t}$$

$$Z \equiv \frac{F_{ext}}{\dot{x}} = \frac{x \cdot (-m\omega^2 + k + i\omega b)}{x \cdot i\omega} = b + i \left[m\omega - \frac{k}{\omega} \right]$$

Real part of Z (resistance) describes energy loss while imaginary part (reactance) describes energy storage

Poles and **Zeros** of a transfer function are the frequencies for which the value of the denominator and numerator of transfer function becomes zero respectively. The values of the poles and the zeros of a system determine whether the system is stable, and how well the system performs. Control systems, in the most simple sense, can be designed simply by assigning specific values to the poles and zeros of the system.

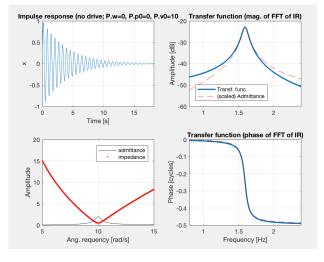
 $https://en.wikibooks.org/wiki/Control_Systems/Poles_and_Zeros$

Useful reference:

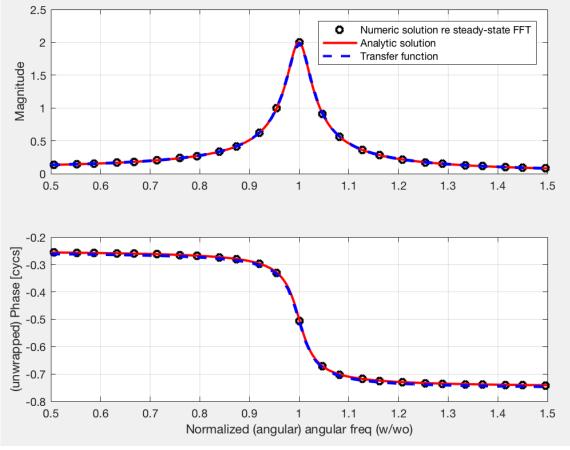
http://web.mit.edu/2.14/www/Handouts/PoleZero.pdf

<u>Note</u>: Electrical engineers commonly use complex frequency (s) representation, tied back to Laplace transforms

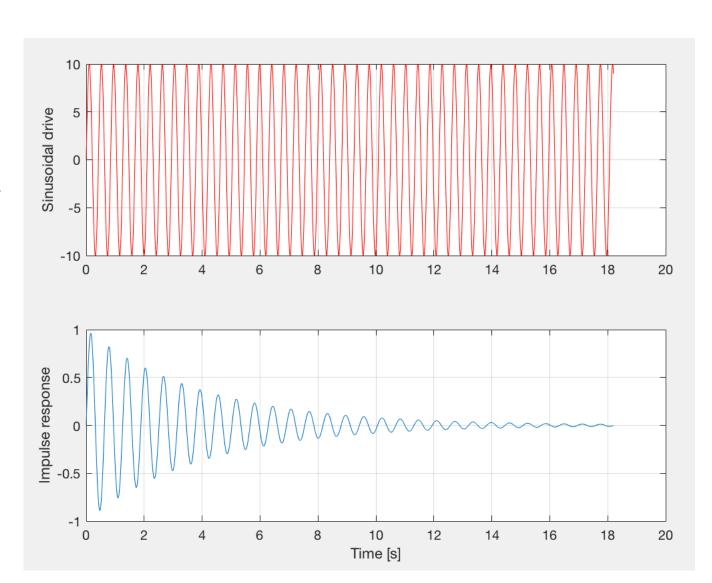
Approach 3 – Transfer Function



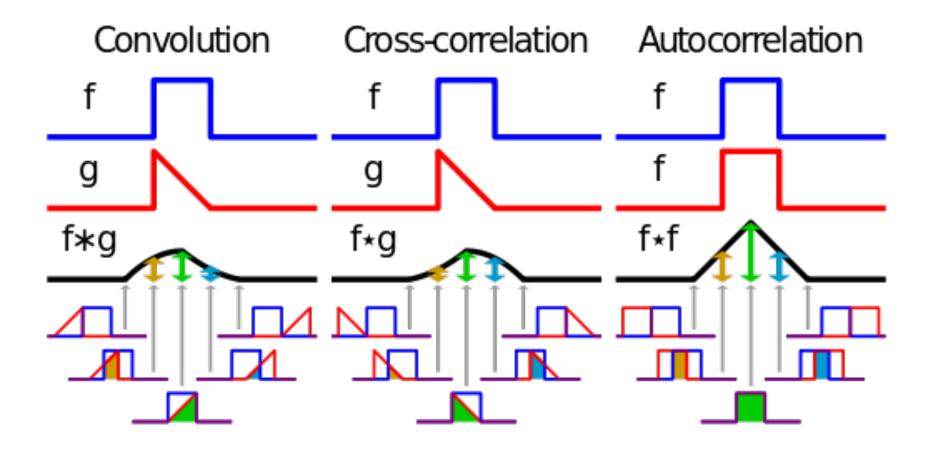
→ End up w/ dashed blue curve



Convolve (in time domain) input drive sinusoids w/ system's impulse response, then compute the FFT



Note: When convolving the impulse response and the drive, the initial transient is apparent, so we use the "long" time window and extract the "steady-state" portion of the convolved response



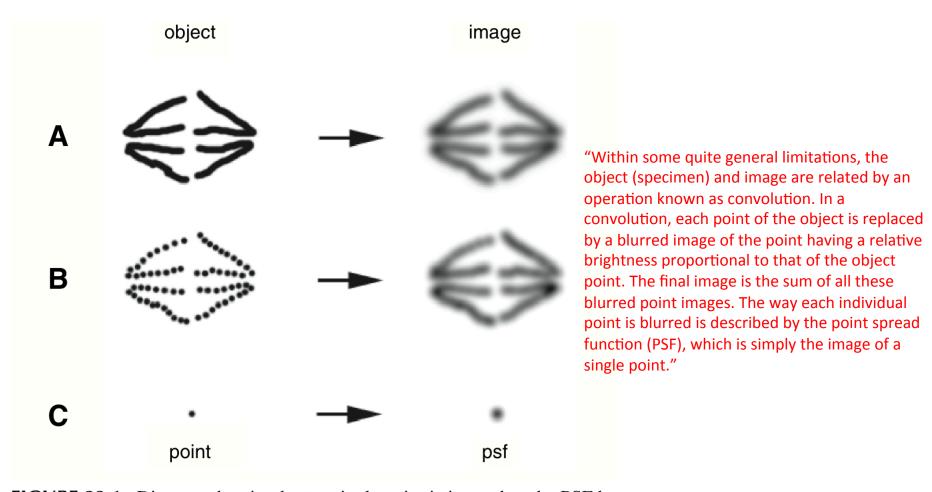


FIGURE 23.1. Diagram showing how a single point is imaged as the PSF by a microscope, and thus that the image of an extended object is the convolution of the object with the PSF.

Aside: Connections to tomography (e.g., CT) & Radon transforms

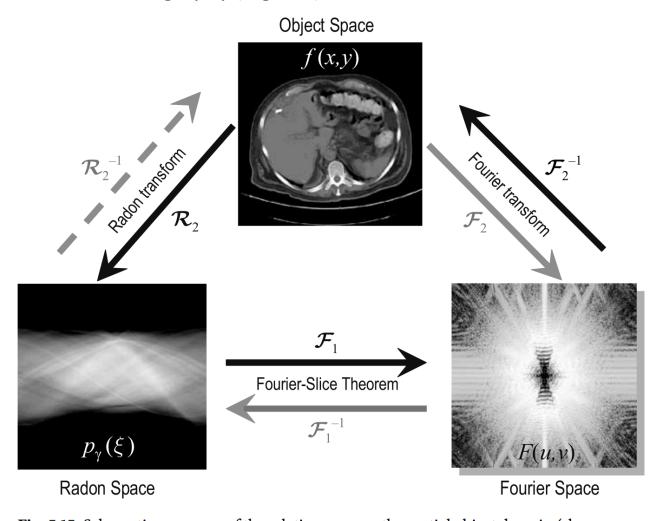
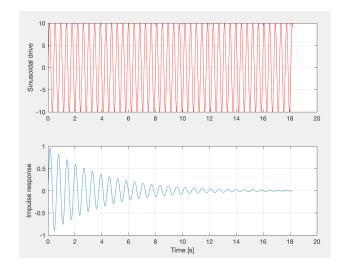
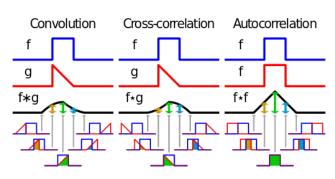


Fig. 5.15. Schematic summary of the relations among the spatial object domain (shown as an axial abdomen tomogram), the Radon space (given over an interval of 180° from the object), and the Fourier space (only absolute values are shown). The Fourier domain results directly from the spatial domain by a two-dimensional Fourier transform of the object, but can also be obtained by the Fourier slice theorem using a set of one-dimensional Fourier transforms of the projection profiles in the Radon space

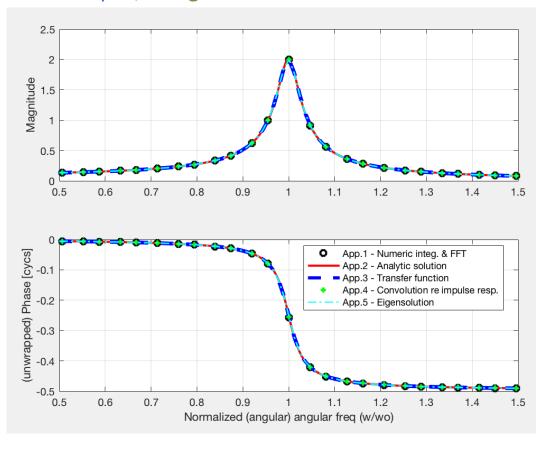
Approach 4 – Convolve the impulse response



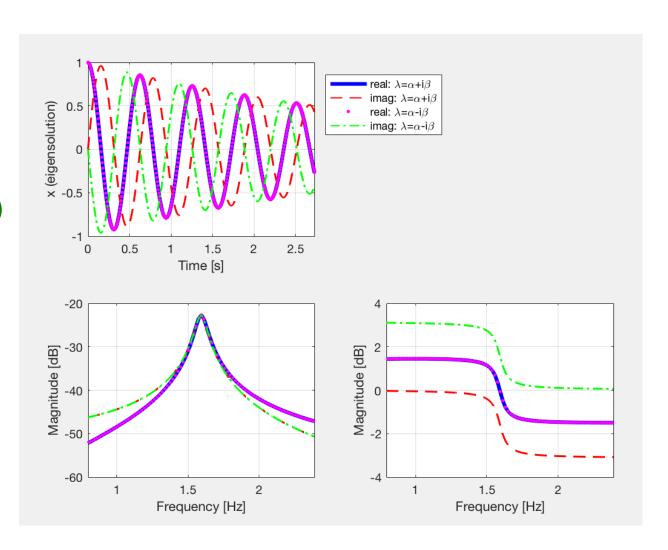


→ Now do this for a variety of different drive frequencies...

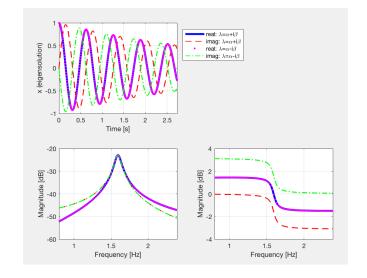
→ End up w/ the green +....



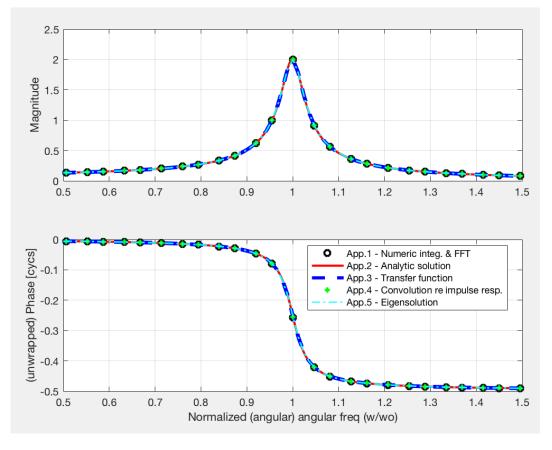
Determine eigenvalues for x=\dot{x}=0 (either numerically or analytically) as the resulting eigensolution, which is equivalent to the impulse response, then compute Fourier transform via FFT



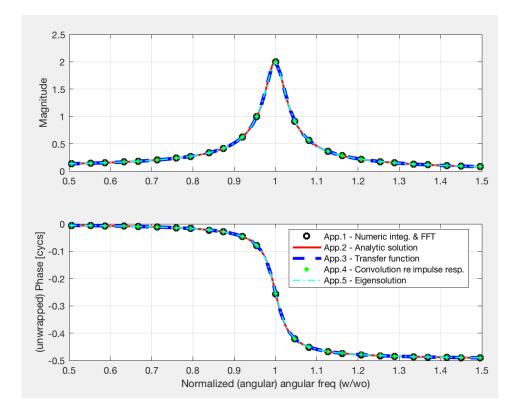
<u>Approach 5 – Eigensolutions</u>



→ End up w/ cyan dashed line....



$$\ddot{x} = -\omega_o^2 x - \gamma \dot{x} + A\cos(\omega t)$$



Note (re still to do #6) stochastic differential equation (SDE) for a purely noise-driven case

Several basic approaches:

(all arriving at the same answer)

- 1. Numerically solve the ODEs and extract the relevant magnitudes and phases (via an FFT)
- 2. Analytic solution I (via Fourier transforms)
- 3. Impulse response I (and associated transfer function)
- 4. Impulse response II (convolve in the time domain)
- **5.** Analytic solution II (via eigensolutions)

Noise-driven systems are very common physically....



Helmholtz resonator



in-situ earphone calibration



.... but harder to deal w/ analytically and computationally

Tonotopy REVISITED Mammalian Cochlea Uncoiled to Vestibular **System** Helicotrema **Stapes Acoustic Energy** pliant & massive to Middle stiff & Ear thin Cochlear

Model: Non-uniform transmission line

Round Window

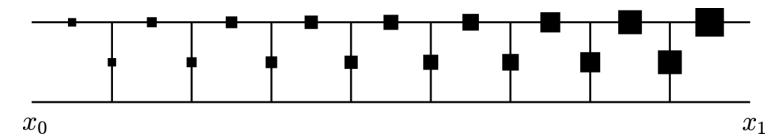
Ref: Zweig et al. 1976, Bergevin 2007

Several key ingredients:

- resonance
- longitudinal (e.g., fluid) coupling

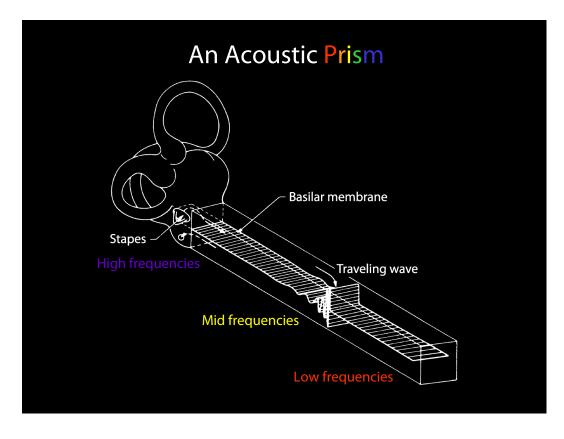
C.D. Geisler (modified)

WKB approximation



Partition

→ Now in much better shape to understand this model!



Theme/Question:

What is the (basic) physical basis for "tonotopy"?

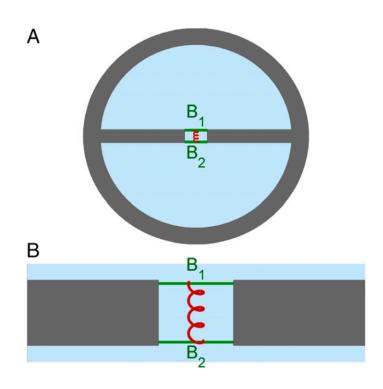
Hint: It ain't a traveling wave per se.... (though such provides a useful framework)

→ This picture may be a bit more complicated....

Frequency selectivity without resonance in a fluid waveguide

Marcel van der Heijden¹

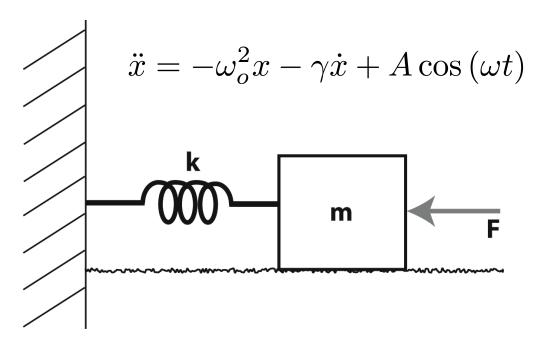
This work describes a simple waveguide that not only carries fluid waves, but also performs a spectral analysis. When driven by a complex input that contains several frequency components, it will spatially separate those components, in analogy to the separation of white light by a prism. The frequency tuning of the waveguide is not based on resonance, but on wave dispersion: Each wave has its own region in which it undergoes a steep deceleration, causing it to focus its energy and deliver it. This method of spectral analysis has not been described before. The waveguide bears a striking resemblance to the inner ear of mammals, both in terms of structure and behavior.

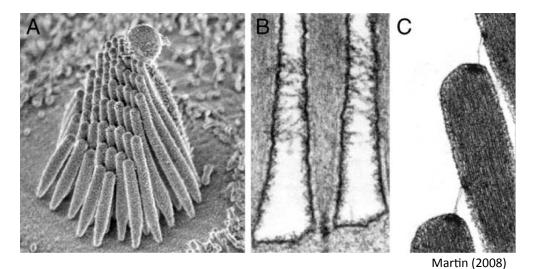


October 7, 2014 | vol. 111 | no. 40

→ Call it what you will (or will not), but all the basic ingredients for "resonance" are there (e.g., elements that trade energy back and forth on a cycle-by-cycle basis, "stiffness gradient", etc...)

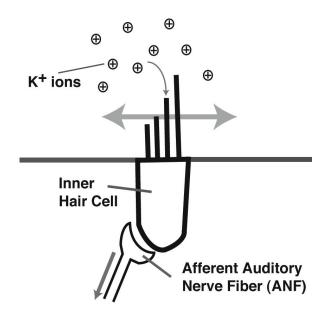
Tangent I: Resonance in the inner ear....





Resonance is chiefly a combination of two reactive forces:

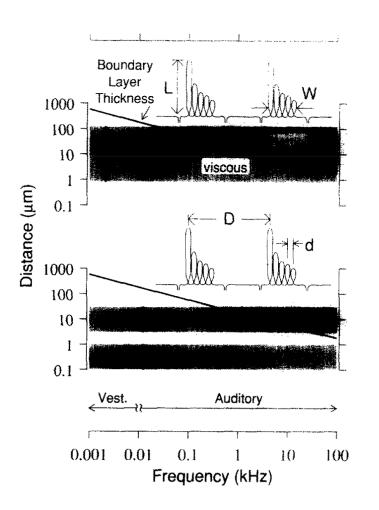
- spring/stiffness
- inertial (i.e., moving mass)



→ Can we model a hair cell bundle using this basic formulation?

The role of fluid inertia in mechanical stimulation of hair cells

Dennis M. Freeman and Thomas F. Weiss



Hearing Research, 35 (1988) 201-208

→ Hair cell bundles seem to operate in the region where viscous forces become relatively large....

Comparative Aspects of Hearing in Vertebrates and Insects with Antennal Ears

Joerg T. Albert¹ and Andrei S. Kozlov²

Current Biology 26, R1050-R1061, October 24, 2016

Box 2. How liquid in the inner ear has shaped the hair bundle.

A hair bundle operates at small Reynolds numbers on the order of 10^{-4} . The Reynolds number (Re) is defined by: Re = $uL\rho/\mu$, where u is the velocity, L is a linear dimension (e.g., a hair bundle's size), ρ is the density and μ is the dynamic viscosity of the fluid. The Reynolds number indicates the relative importance of inertia over viscous forces for a particular type of flow. For the hair bundle, a Reynolds number of much lower than 1 indicates the relative importance of viscous forces.

The Scallop Theorem

Life at low Reynolds number

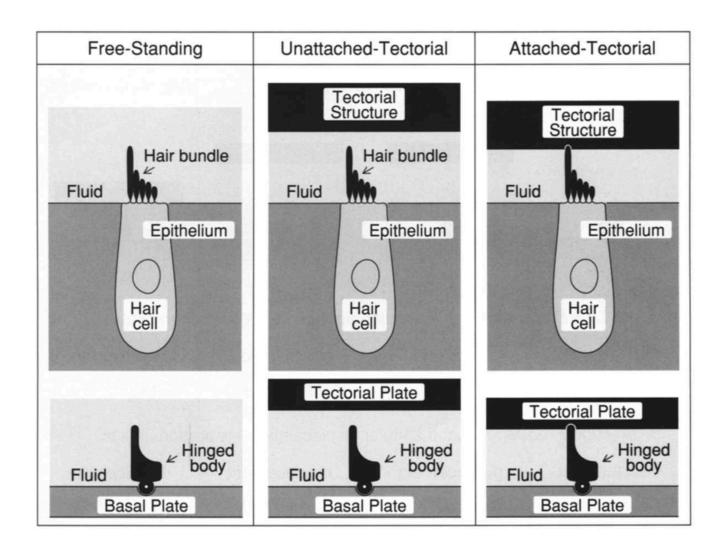
E. M. Purcell

Lyman Laboratory, Harvard University, Cambridge, Massachusetts 02138 (Received 12 June 1976)



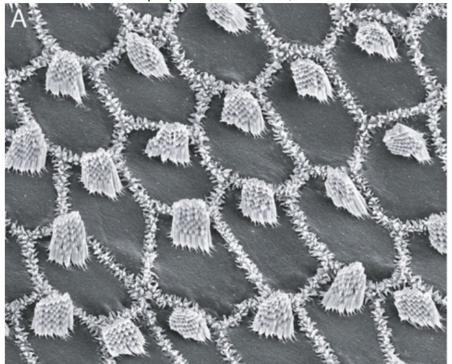
Tangent I: Resonance in the inner ear....

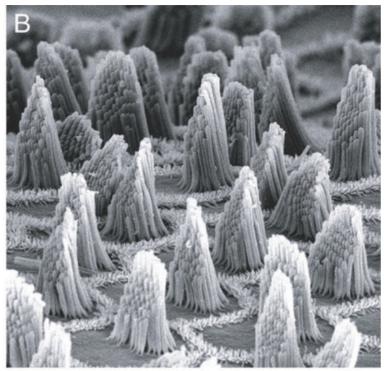
But things are a bit more complicated than just "what is the Reynold's # of a hair cell bundle?"....



<u>Tangent I</u>: Resonance in the inner ear....

Chicken basilar papilla (i.e., auditory hair cells)

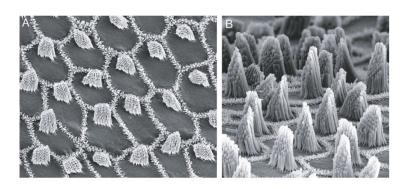


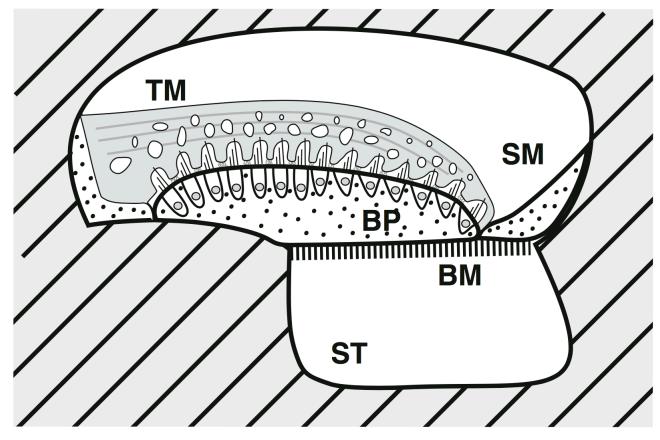


Hudspeth (2008)

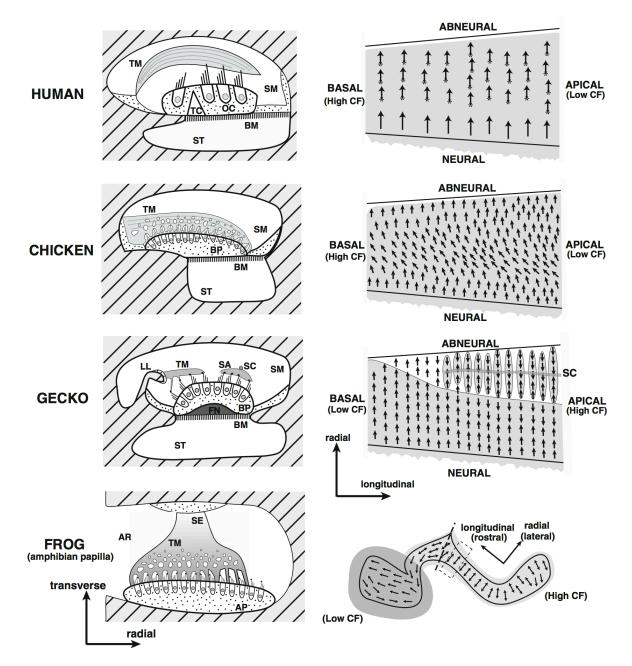
<u>Note</u>: Be careful! This picture can be misleading. In-vivo, there is a massive tectorial membrane (**TM**) overlying these hair cells....

<u>Tangent I</u>: Resonance in the inner ear....



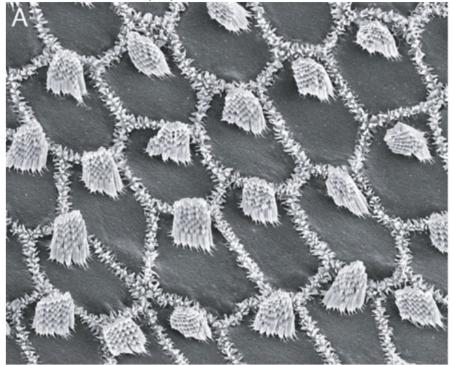


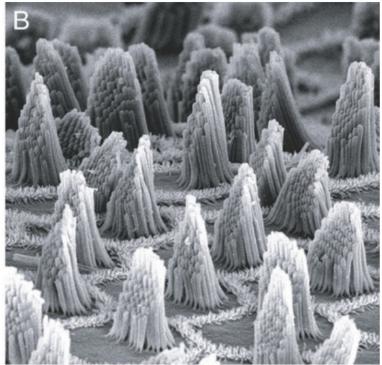
<u>Tangent I</u>: Resonance in the inner ear....



Tangent I: Resonance in the inner ear....

Chicken basilar papilla (i.e., auditory hair cells)





Hudspeth (2008)

<u>Note</u>: Be careful! This picture can be misleading. In-vivo, there is a massive tectorial membrane (**TM**) overlying these hair cells....

→ Coupling between hair cells affects mechanical properties (e.g., tuning) how?

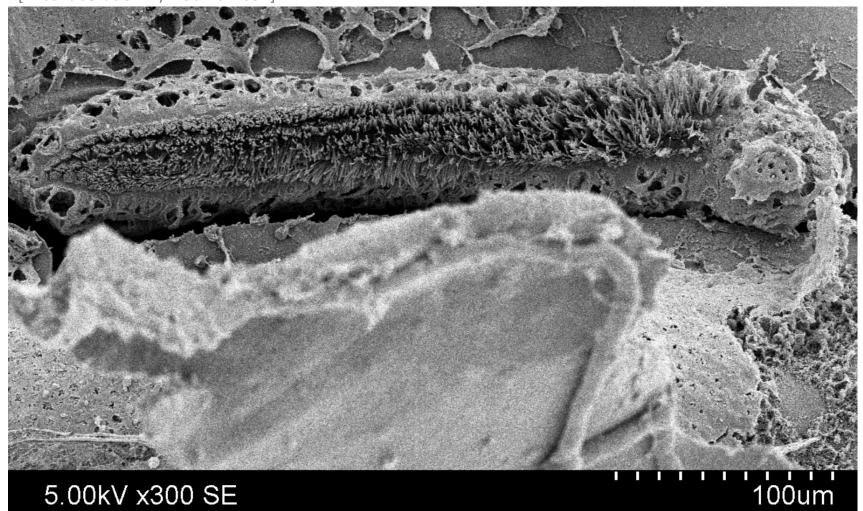
<u>Tangent I</u>: Resonance in the inner ear....



<u>Tangent I</u>: Resonance in the inner ear....

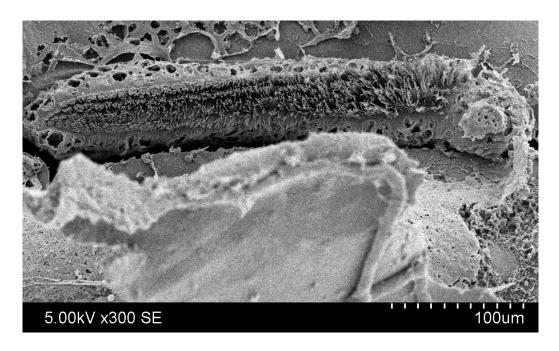


[in collaboration w/ Bob Harrison]



5.00kV x300 SE

<u>Tangent I</u>: Resonance in the inner ear....



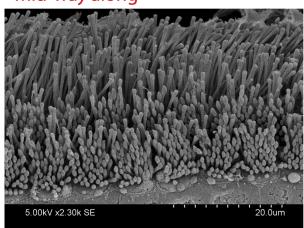


Note: These hair cells do not have an overlying TM (for the most part)

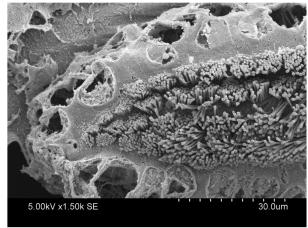
transition between free-standing (left) and TM (right) regions; note "tweenage" bundles



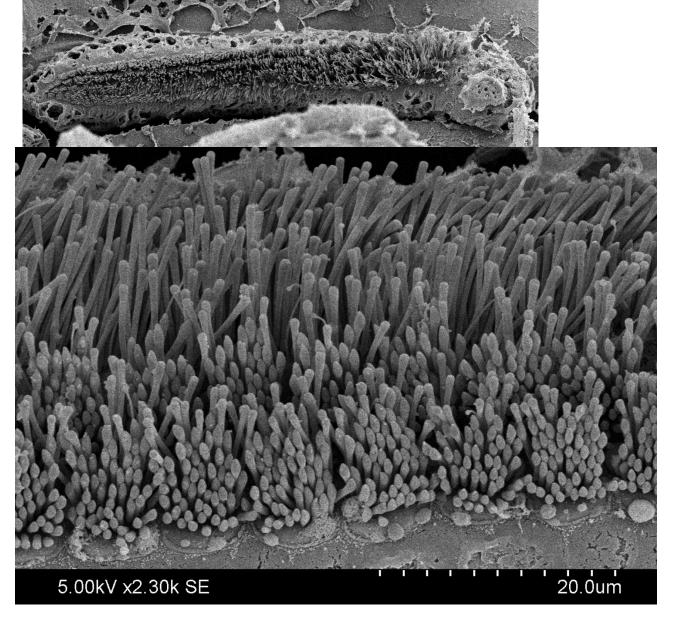
mid-way along



high-freq. (apical) end



Tangent I: Resonance in the inner ear....





- "phalanx" of hair cells
- Implications for intercell coupling?
- Longitudinal propagations? (e.g., traveling/standing waves)

Tangent I: Resonance in the inner ear....

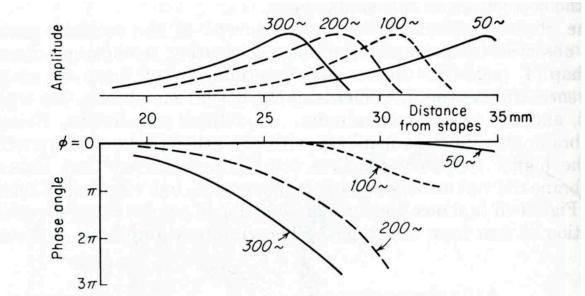
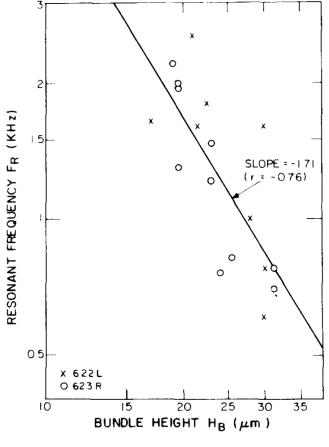


Fig. 11-58. Phase displacement and resonance curves for four low tones.

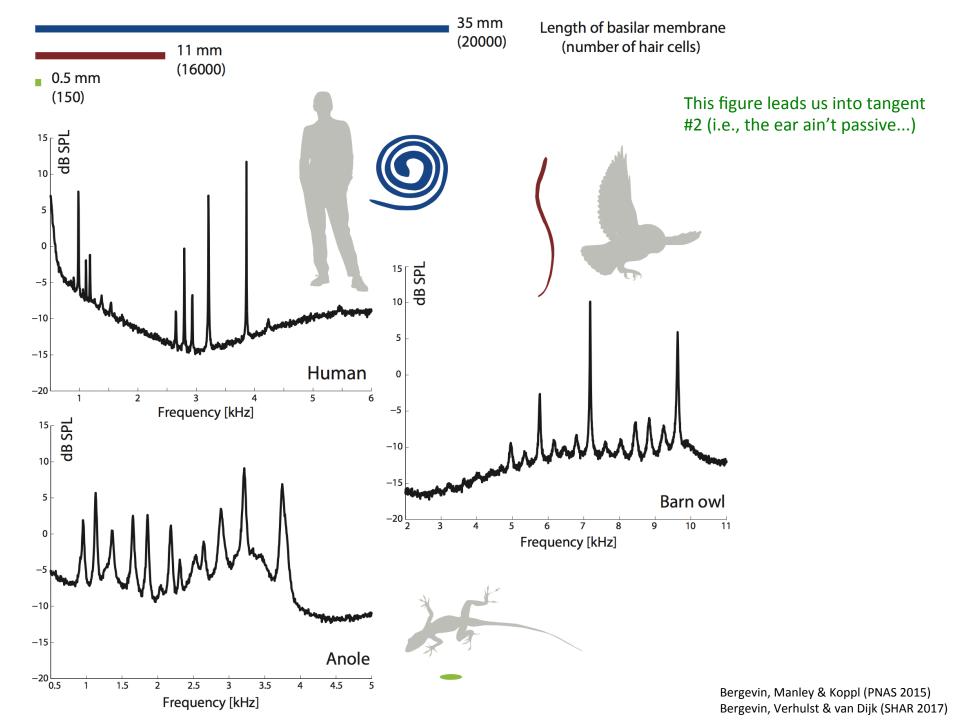
Bekesy (1960)

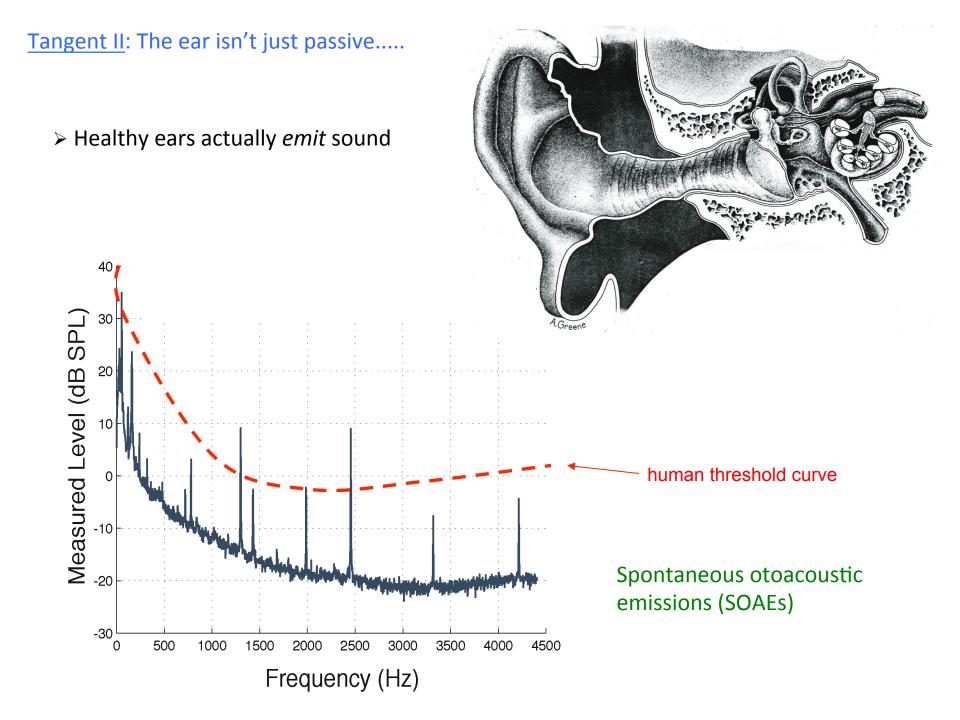
.... so clearly "tuned" responses can arise in the absence of an active process (see tangent #2)

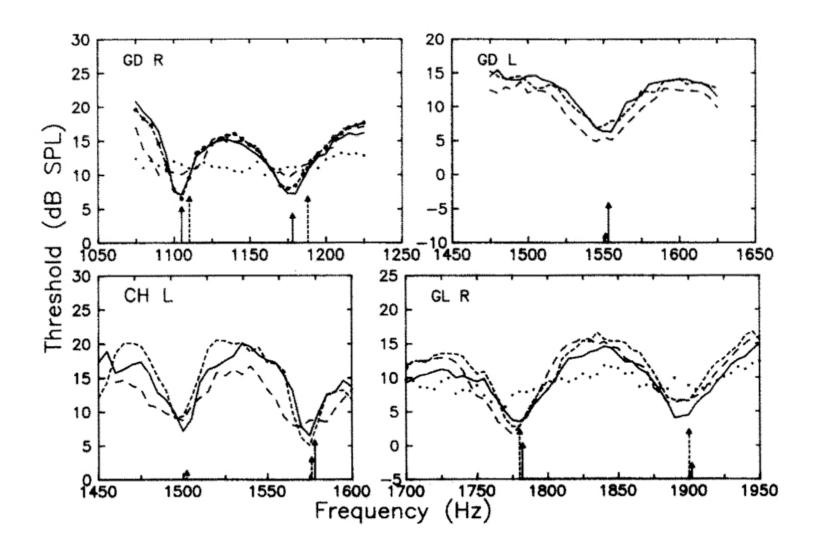
These are classic measurements from *dead* ears....



Frishkopf & DeRosier (1983)







> SOAEs directly tied to forward auditory transduction (i.e., neural responses)

Tangent II: SOAEs & ANF responses

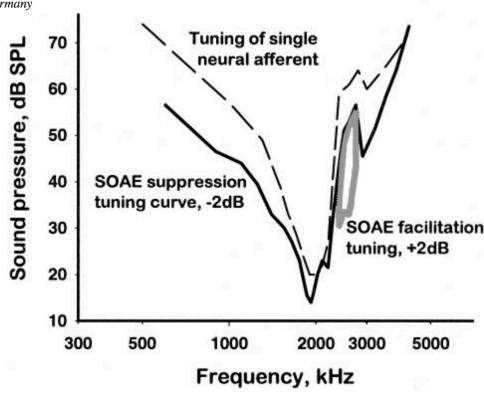
Evidence for an Active Process and a Cochlear Amplifier in Nonmammals

GEOFFREY A. MANLEY

Lehrstuhl für Zoologie, Technische Universität München, 85747 Garching, Germany

➤ SOAE "suppression" related to auditory nerve fiber tuning....

> ... probably in a complicated fashion



Tangent II: Simple heuristic for modeling SOAEs

$$\ddot{x} = -\omega_o^2 x - \gamma \dot{x} + A \cos(\omega t)$$

Passive, linear case doesn't do the trick....

Simple model to explain an SOAE peak:

$$\ddot{x} = -x - \varepsilon(x^2 - 1)\dot{x}$$

van der Pol oscillator

<u>Note</u>: This equation comes in different flavors/forms (e.g., "normal form", complex)

$$\dot{z} = -\mu z + i\omega_o z - |z|^2 z$$

where
$$z \in \mathbb{C}$$
, $\mu, \omega_o \in \mathfrak{R}$

² A (linear, undriven) harmonic oscillator can be described by a single, first–order ODE in terms of a complex variable z (e.g., [10]):

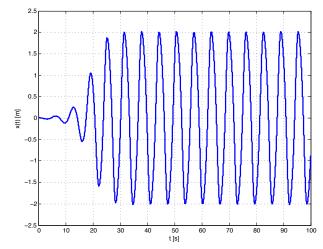
$$\dot{z} = -\mu z + i\omega_0 z$$
.

Via a change of variables, this can be re-expressed as a 2nd order (real-valued) ODE:

$$\ddot{x} + 2\mu\dot{x} + (\omega_o^2 + \mu^2)x = 0$$
.

Thus, the two notations are essentially equivalent. Note that in this case, no matter what the sign of ω_o is, the quantity $(\omega_o^2 + \mu^2)$ will always be positive. Thus the system will always have a positive stiffness, though the damping can be positive or negative (depending upon the sign of μ).

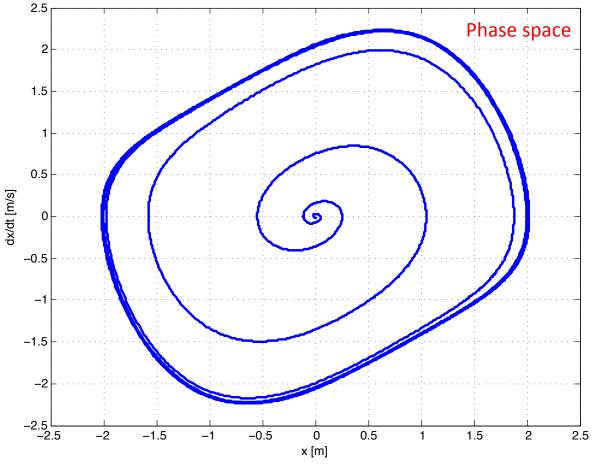
Tangent II: The ear isn't just passive.....



Limit cycles:

- Negative damping for small displacements injects energy into system
- Nonlinearity stabilizes
- Self-sustained oscillation!

$$\ddot{x} = -x - \varepsilon(x^2 - 1)\dot{x}$$



Tangent II: The ear isn't just passive.....

P.I.M. Johannesma

Workgroup Neurophysics, Lab. of Medical Physics and Biophysics, University of Nijmegen, Nijmegen, The Netherlands.

Narrow band filters and active resonators.

Psychophysical, Physiological and Behavioural Studies in Hearing

Proceedings of the 5th International Symposium on Hearing

Noordwijkerhout, The Netherlands April, 8-12, 1980

Are spontaneous otoacoustic emissions generated by self-sustained cochlear oscillators?

Carrick L. Talmadge and Arnold Tubis

Department of Physics, Purdue University, West Lafayette, Indiana 47907

Hero P. Wit

Institute of Audiology, University Hospital, P. O. Box 30.001, 9700 RB Groningen, The Netherlands

Glenis R. Long

Department of Audiology and Speech Science, Purdue University, West Lafayette, Indiana 47907

J. Acoust. Soc. Am. 89 (5), May 1991

4622

Biophysical Journal Volume 95 November 2008 4622-4630

Frequency Clustering in Spontaneous Otoacoustic Emissions from a Lizard's Ear

Andrej Vilfan* and Thomas Duke†

*J. Stefan Institute, Ljubljana, Slovenia; and †London Center for Nanotechnology and Department of Physics & Astronomy, London, United Kingdom

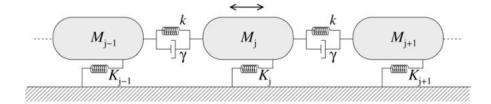
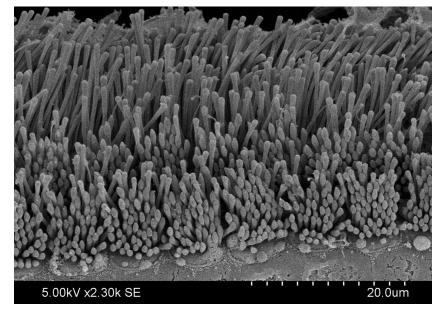
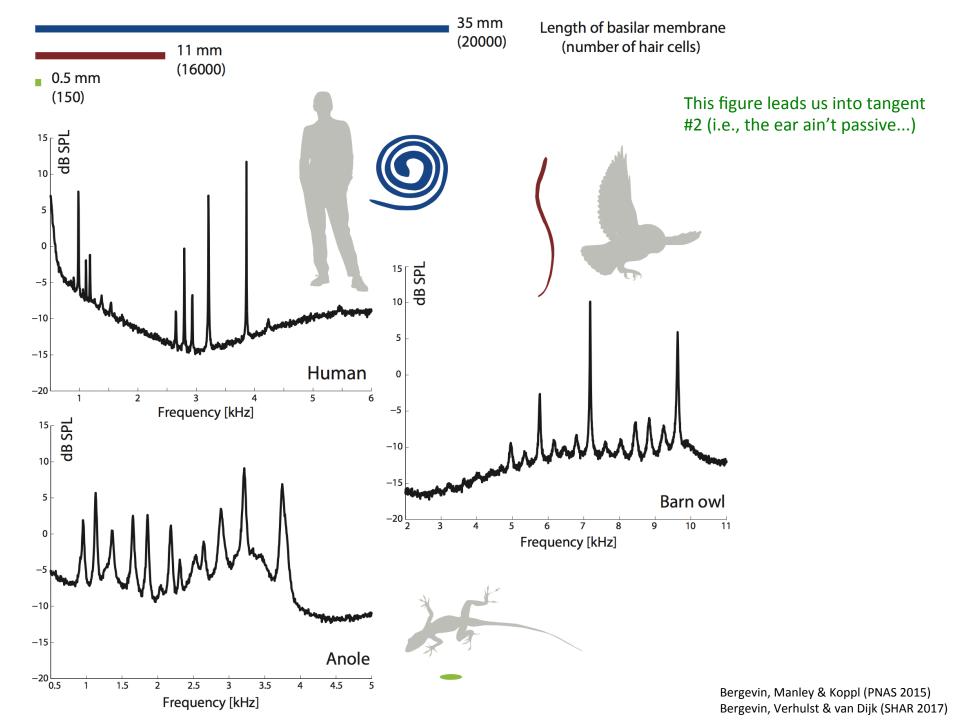


FIGURE 1 Mechanical equivalent of the model. Sallets are represented as inertial oscillators (mass M_j , spring K_j), coupled to their neighbors by elastic (constant k) and damping (constant γ) elements. In addition, there exists an active driving mechanism within each oscillator (not shown).

Unclear what the "correct" coupling should be...





Tangent III: Other examples of "oscillators" in biology...

Vibrissa Resonance as a Transduction Mechanism for Tactile Encoding

Maria A. Neimark,1* Mark L. Andermann,1,2* John J. Hopfield,3 and Christopher I. Moore4

The Journal of Neuroscience, July 23, 2003 • 23(16):6499 – 6509

vi·bris·sa

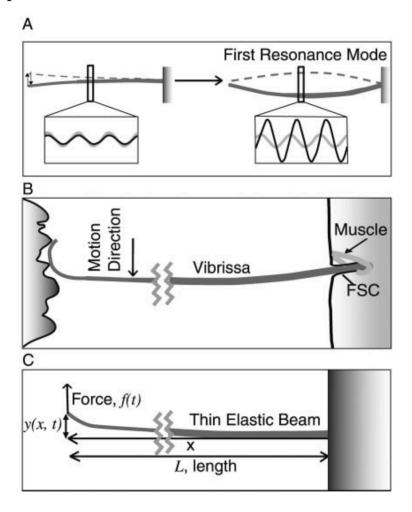
/vīˈbrisə/ •

noun zoology

any of the long stiff hairs growing around the mouth or elsewhere on the face of many mammals, used as organs of touch; whiskers.

ORNITHOLOGY

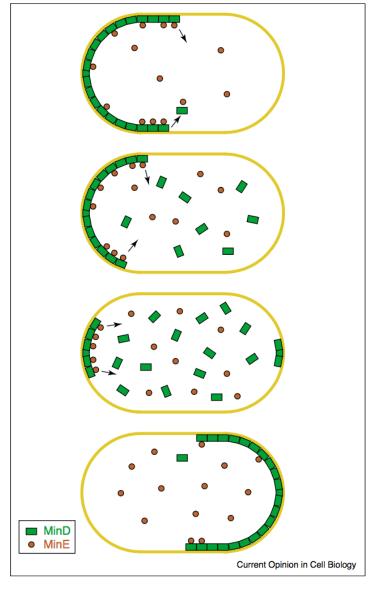
 each of the coarse bristlelike feathers growing around the gape of certain insectivorous birds that catch insects in flight.



Tangent III: Other examples of "oscillators" in biology...

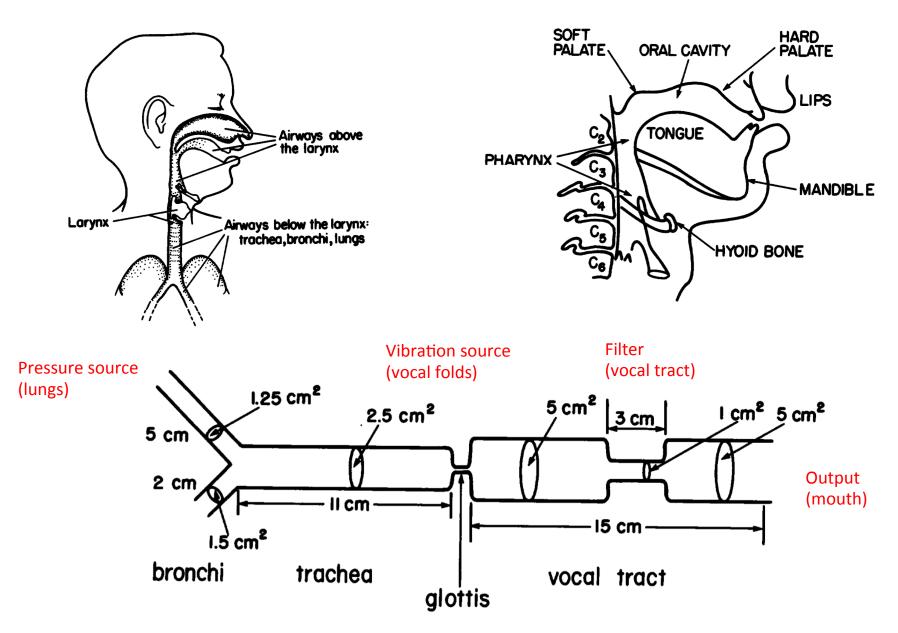
Oscillations in cell biology Karsten Kruse and Frank Jülicher

Current Opinion in Cell Biology 2005, 17:20–26



Schematic representation of Min oscillations in *E. coli*. MinD (green) is localized on the inner bacterial membrane (yellow) on one side of the cell, where it aggregates. MinE (red) induces disassembly of the MinD aggregates and detachment of MinD molecules into the cytoplasm. MinD then assembles on the membrane of the opposite side of the cell and the process is repeated.

Tangent III: Other examples of "oscillators" in biology...



Tangent III

- Source: Vibrating vocal folds make 'broadband' sound
- → Harmonics commonly referred to as "overtones"
- <u>Filter</u>: Vocal tract shapes source sound
- → Resulting 'shape' emphasizes features (e.g., formants)
- Formants: Filtered harmonics; basis for vowels

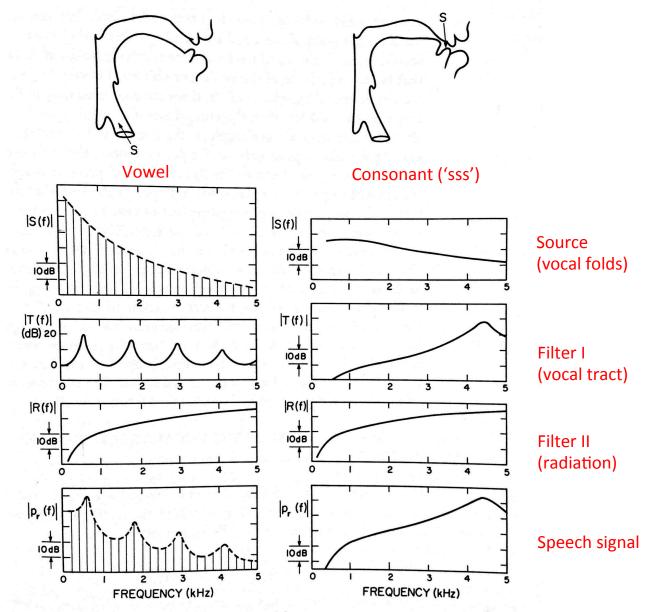


Figure 3.1 Sketches indicating components of the output spectrum $|p_r(f)|$ for a vowel and a fricative consonant. The output spectrum is the product of a source spectrum S(f), a transfer function T(f), and a radiation characteristic R(f). The source spectra are similar to those derived in figures 2.10 and 2.33 in chapter 2. For the periodic source, S(f) represents the amplitudes of spectral components; for the noise source, S(f) is amplitude in a specified bandwidth. See text.

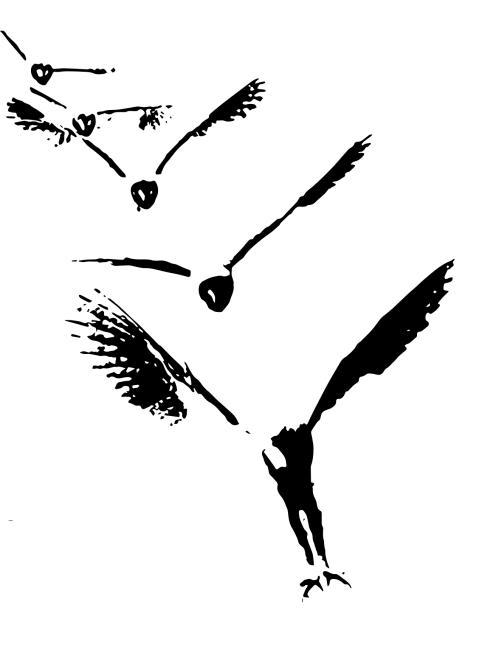
Fini



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Mechanical

Electrical

