

# INTRODUCTION TO TIME-FREQUENCY DISTRIBUTIONS

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(Preliminary version)

## OUTLINE OF PRESENTATION

1. Introduction
  - a. What are we trying to do?
  - b. Desired properties of time-frequency dists
2. Basics
  - a. Densities
  - b. Time-bandwidth product
  - c. Analytic signals
3. Continuous (Morlet) wavelet transform
  - a. Basic properties
  - b. Localization properties
4. Short-Time Fourier Transform (STFT)
  - a. Basic properties
  - b. Time-frequency plane tilings
  - c. Spectrogram properties
5. Wigner-Ville distribution
  - a. Basics
  - b. Properties
  - c. More properties
6. Choi-Williams (RID) distribution
  - a. Basic idea
  - b. Examples
7. Cohen's class of TFDs using kernels
  - a. Basic idea
  - b. Examples
8. Applications of TFDs
  - a. Reconstruction of dispersive layered media
  - b. Tracking non-constant blood flow in MRI

## INTRODUCTION

Q: What are we trying to do?

A: To obtain a "time-local" Fourier transform which can track time-varying spectral properties.

Fourier transform can't track time variations of:

EX #1:  $x(t) = \begin{cases} \sin(3t), & \text{if } t < 4; \\ \sin(5t), & \text{if } t > 4. \end{cases}$

EX #2: Chirp  $x(t) = e^{j(at)t} = e^{jat^2}$ :  
frequency increases linearly with time.

EX #3: Speech, whale sounds, waves through lossy media.

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## DESIRED PROPERTIES

$\mathcal{W}_{\tau,\omega}\{x(t)\}$  is spectral density of  $x(t)$  for  $t \approx \tau$ .

We would like the following to hold:

1.  $\mathcal{W}_{\tau,\omega}\{x(t)\} \geq 0$  (since it is a density)
2.  $\int \int \mathcal{W}_{\tau,\omega}\{x(t)\} d\tau d\omega = \int |x(t)|^2 dt$  (total energy conserved)
3. Marginals (more specific energy conservation):
  - a.  $\int \mathcal{W}_{\tau,\omega}\{x(t)\} d\omega = |x(t)|^2$  (energy at time  $t$ )
  - b.  $\int \mathcal{W}_{\tau,\omega}\{x(t)\} dt = |X(\omega)|^2$  (energy at freq  $\omega$ )
4. Shifts in time and frequency properly represented:
  - a.  $\mathcal{W}_{\tau,\omega}\{x(t - t_0)\} = \mathcal{W}_{\tau - t_0, \omega}\{x(t)\}$
  - b.  $\mathcal{W}_{\tau,\omega}\{x(t)e^{j\omega_0 t}\} = \mathcal{W}_{\tau, \omega - \omega_0}\{x(t)\}$

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## PROBLEMS

1. Shorter time window  $\rightarrow$  poor freq resolution.
2. Longer window  $\rightarrow$  can't track nonstationarities.
3. Must trade off time and frequency resolution.

## BASICS

1. **Densities:** A density is something you *integrate* to get the quantity of interest:
  - a. Energy spectral density  $|X(\omega)|^2$   
→ energy in band  $[\omega_0, \omega_0 + \delta]$  is  $\frac{2}{2\pi}|X(\omega_0)|^2\delta$ .
  - b. We really want "time-freq *density*," not "distribution."
2. **Time-bandwidth product:**

Regard  $|x(t)|^2 \geq 0$  as a prob. density func. for  $t$  and also regard  $|X(\omega)|^2 \geq 0$  as a pdf for  $\omega$ .

  - a.  $\Delta t = \text{duration} = \sigma_t$
  - b.  $\Delta \omega = \text{bandwidth} = \sigma_\omega$
  - c. Then  $(\Delta t)(\Delta \omega) \geq \frac{1}{2}$ ; equality for Gaussian.
3.  $x(t) = A(t)e^{j\phi(t)} \rightarrow A(t) = \text{amplitude}; \phi(t) = \text{phase}$ .
  - a.  $\frac{d\phi}{dt} = \text{instantaneous frequency}$ .
  - b. Can't define these from  $Re[x(t)]$  only. Leads to:
4. **Analytic signal** of  $x(t)$  is  $x_a(t) = x(t) - j\mathcal{H}\{x(t)\}$ .
  - a. *Hilbert xform*  $\mathcal{H}\{x(t)\} = x(t) * \frac{-1}{\pi t}$   
Transfer function:  $jSGN[\omega] \leftrightarrow \frac{\pi}{2}$  phase shift.
  - b.  $X_a(\omega) = \begin{cases} 2X(\omega), & \text{if } \omega > 0; \\ 0, & \text{if } \omega < 0. \end{cases}$
  - c. EX:  $x(t) = \cos(at) \rightarrow x_a(t) = e^{jat}$  (quadrature)
  - d.  $x(t)$  causal  $\rightarrow Im[X(\omega)] = \mathcal{H}\{Re[X(\omega)]\}$ .

# THE CONTINUOUS (MORLET) WAVELET TRANSFORM

**Def'n:** Continuous or *Morlet* wavelet xform:

$$\mathcal{W}_b^a\{x(t)\} = \int x(t) \frac{1}{\sqrt{a}} \psi^*\left(\frac{t-b}{a}\right) dt, a > 0$$

1. Large  $a \gg 0 \rightarrow$  long time scale, coarse resolution.  
Small  $0 < a \ll 1 \rightarrow$  short time scale, fine resolution.

2. *Requirements* on the *mother wavelet*  $\psi(t)$ :

- a.  $\Psi(0) = 0 \rightarrow$  bandpass (important requirement).

- b.  $C = \int_0^\infty \frac{|\Psi(\omega)|^2}{|\omega|} d\omega < \infty$ .

This is satisfied in practice if  $\Psi(0) = 0$

and if  $\lim_{|\omega| \rightarrow \infty} \Psi(\omega) = 0 \rightarrow$  no impulse in  $\psi(t)$ .

3. *Reconstruct*  $x(t)$  from  $\mathcal{W}_b^a\{x(t)\}$  using the formula

$$x(t) = \frac{1}{C} \int_0^\infty \int \mathcal{W}_b^a\{x(t)\} \frac{1}{\sqrt{a}} \psi\left(\frac{t-b}{a}\right) \frac{db da}{a^2}$$

Proof: See V&K p.302-3.

4. **Properties:**

- a. Shift-invariant:  $\mathcal{W}_b^a\{x(t - t_0)\} = \mathcal{W}_{b-t_0}^a\{x(t)\}$ .

- b. Scale:  $\mathcal{W}_b^a\{x(t/c)\} = \sqrt{|c|} \mathcal{W}_{b/c}^{a/c}\{x(t)\}$ .

- c. Parseval:  $\int |x(t)|^2 dt = \frac{1}{C} \int \int |\mathcal{W}_b^a\{x(t)\}|^2 \frac{da db}{a^2}$ .

Proof: See V&K p.306-7.

5. Morlet wavelet:  $\psi(t) = \frac{1}{2\pi} e^{-t^2/2} e^{-j\omega_0 t}$ ;  $\Psi(\omega) = e^{-(\omega - \omega_0)^2/2}$ .

$\omega_0 = 5.336 \rightarrow$  1st peak of  $Re[\psi(t)] =$  half  $t = 0$  value.

6. **Sample:**  $a = 2^m, b = n2^m \rightarrow$  dyadic wavelet sampling.

This xform is heavily overdetermined and redundant.

## Localization Properties of Morlet Wavelet

Let the mother wavelet have duration  $\delta t$   
and bandpass spectrum  $[\omega_0 - \frac{\delta\omega}{2}, \omega_0 + \frac{\delta\omega}{2}]$ :

1.  $a \ll 1 \rightarrow$  only "see"  $x(t)$  for  $b - \frac{a\delta t}{2} < t < b + \frac{a\delta t}{2}$ .
  - a. Narrow interval  $\rightarrow$  can track nonstationarity (vary  $b$ ).
  - b. Good time localization; "see" fine details of  $x(t)$ .
  - c. "Zoom-in" on details of  $x(t)$ .
2.  $a \ll 1 \rightarrow$  "see"  $X(\omega)$  for  $\frac{\omega_0}{a} - \frac{\delta\omega}{2a} < \omega < \frac{\omega_0}{a} + \frac{\delta\omega}{2a}$ .
  - a. High and broad frequency band.
  - b. Octave-band filtering (constant  $\frac{\text{width}}{\text{center}}$ ).

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3.  $a \gg 1 \rightarrow$  "see"  $x(t)$  for  $b - \frac{a\delta t}{2} < t < b + \frac{a\delta t}{2}$ .
  - a. Broad interval  $\rightarrow$  can't track nonstationarity.
  - b. Little time localization; "see" coarse details of  $x(t)$ .
  - c. "Zoom-out" on  $x(t)$ ; see "big picture."
4.  $a \gg 1 \rightarrow$  only "see"  $X(\omega)$  for  $\frac{\omega_0}{a} - \frac{\delta\omega}{2a} < \omega < \frac{\omega_0}{a} + \frac{\delta\omega}{2a}$ .
  - a. Low and narrow frequency band.
  - b. Octave-band filtering.

# THE SHORT-TIME FOURIER TRANSFORM (STFT)

**Def'n:** The STFT is defined as

$$STFT_{\omega,\tau}\{x(t)\} = \int w(t - \tau)x(t)e^{-j\omega t} dt.$$

1. Fourier xform of *windowed* (by  $w(t - \tau)$ )  $x(t)$ .  
As  $\tau$  changes, pick off  $x(t)$  at different times.
2. *Requirements* on the window  $w(t)$ : None.  
Usually normalize  $\int |w(t)|^2 dt = 1$ .  
Should be localized in time and frequency to be useful.
3. *Reconstruct*  $x(t)$  from  $STFT_{\omega,\tau}\{x(t)\}$ :  
$$x(t) = \frac{1}{2\pi} \int \int STFT_{\omega,\tau}\{x(t)\} w(t - \tau) e^{j\omega t} d\omega d\tau.$$
4. **Properties:**
  - a. STFT time-frequency tilings all same size.  
Wavelet: t-f tilings have different sizes.
  - b. Parseval:  $\int |x(t)|^2 dt = \frac{1}{2\pi} \int \int |STFT_{\omega,\tau}\{x(t)\}|^2 d\omega d\tau$ .  
Proof: V&K p.313-4.
  - c. Spectrogram:  $|STFT_{\omega,\tau}\{x(t)\}|^2$  is local psd.
5. Gabor function:  $w(t) = be^{-at^2}$ ;  $W(\omega) = b\sqrt{\frac{\pi}{a}}e^{-\omega^2/4a}$ .  
Best localization in time and frequency.
6. **Sample:**  $\omega = m\omega_0, \tau = n\tau_0$   
since time-frequency tilings all have same size.  
STFT is heavily overdetermined and redundant.

## TILINGS OF T-F PLANE: STFT AND MORLET

For both sampled STFT and Morlet wavelet xforms:  
Express  $x(t)$  using basis functions localized in time and freq.

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### STFT:

1. Basis funcs= $\{w(t - n\tau_0)e^{jm\omega_0t}, n, m \in integers\}$ .
  2.  $w(t)$ =window localized in time and frequency  $\rightarrow$   
 $w(t - n\tau_0)e^{jm\omega_0t}$  centered at  $(t, \omega) = (n\tau_0, m\omega_0)$ .
  3. No scaling of t-f plane tiles.
  4.  $STFT_{\omega, \tau}\{x(t)\}$  projects  $x(t)$  onto t-f tiles.
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### Continuous (Morlet) Wavelet Transform

1. Basis funcs= $\{2^{-m/2}\psi(2^{-m}t - n), n, m \in integers\}$ .
  2.  $\psi(t)$ =basis localized in time and frequency  $\rightarrow$   
 $\psi(2^{-m}t - n)$  centered at  $(t, \omega) = (2^m n, 2^{-m}\omega_0)$ .
  3. Now t-f plane tiles are scaled by  $2^{\pm m}$ .
  4. Wavelet xform projects  $x(t)$  onto t-f tiles.
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See next page for figures of t-f tilings.



## SPECTROGRAM PROPERTIES

**DEF:** The *spectrogram* is  $|STFT_{\omega,\tau}\{x(t)\}|^2$ .

1.  $Spectrogram = |\int w(t - \tau)x(t)e^{-j\omega t} dt|^2$
  2.  $= \frac{1}{2\pi} |\int W(\omega - \omega')X(\omega')e^{j\omega'\tau} d\omega'|^2$   
using (generalized) Parseval's theorem  
and also  $\mathcal{F}\{w(t - \tau)\} = W(\omega)e^{-j\tau\omega}$ .
  3. Can view as time-varying local energy spectral density.
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### Properties

1. Tradeoff between time and freq resolutions:
  - a. Narrow window  $\rightarrow$  good time localization, poor freq resolution (based on too little data).
  - b. Spectral peaks convolved with  $W(\omega) \rightarrow$  blurred.
  - c. Also, short time support  $\rightarrow$  large bandwidth.
  - d. Wide window  $\rightarrow$  good frequency resolution, poor time localization (signal varies *within* window).
  - e. Can see these explicitly in above formulae:  
tradeoff between compactness of  $w(t)$  and  $W(\omega)$  (time-bandwidth product).
2. Simple and actually does work fairly well.
3. Can use signal-dependent window  $w(t)$ .
4. Does NOT satisfy marginals

## THE WIGNER-VILLE T-F DISTRIBUTION

**Def'n:** Wigner-Ville distribution is defined as

$$W_{\omega,\tau}\{x(t)\} = \frac{1}{2\pi} \int x(\tau + t/2)x^*(\tau - t/2)e^{-j\omega t} dt.$$

1. This is 2-D Fourier *dual of ambiguity function*.
2. Dates back to 1932; used in quantum mechanics!
3.  $x(\tau + t/2)x^*(\tau - t/2) = \text{instantaneous autocorrelation}$  of  $x(t)$  having lag  $t$  at time  $\tau$ .

So this is instantaneous version of  $S_x(\omega) = \mathcal{F}\{R_x(t)\}$ .

4. *Reconstruct  $x(t)$  from  $W_{\omega,\tau}\{x(t)\}$ :*

$$x(t)x^*(0) = \int \int W_{\omega,\tau=t/2}\{x(t)\}e^{j\omega t} d\omega d\tau.$$

### 5. Nice Properties:

- a. *Marginals are preserved:* If  $\mathcal{F}\{x(t)\} = X(\omega)$ ,  
 $\int W_{\omega,\tau}\{x(t)\}d\tau = |X(\omega)|^2$ ;  $\int W_{\omega,\tau}\{x(t)\}d\omega = |x(\tau)|^2$ .

Compare to spectrogram: Total energy conserved:

$$\int \int |STFT_{\omega,\tau}\{x(t)\}|^2 d\omega d\tau = \int |x(t)|^2 dt$$

but marginals *not* conserved separately.

- b. *Time and frequency shift-invariant:*

$$W_{\omega,\tau}\{x(t - t_0)e^{j\omega_0 t}\} = W_{\omega - \omega_0, \tau - t_0}\{x(t)\}$$

- c. Sharp resolution: For chirp  $x(t) = e^{j(at)t} = e^{jat^2}$   
(this is linearly increasing frequency  $\omega = at$ )

$$W_{\omega,\tau}\{x(t)\} = \delta(\omega - at) \text{ (try it!)}$$

### 6. Bad Properties:

- a. *Not non-negative:* Can have  $W_{\omega,\tau}\{x(t)\} < 0$ !

*Very bad for a purported energy spectral density!*

- b. *Cross terms:* Since nonlinear, no superposition!

Solution: Use RID (reduced interference distribution) of Choi-Williams (insert kernel which windows in t-f).

## More Properties of Wigner Distribution

1.  $x(t) = A(t)e^{j\phi(t)}$

→ average freq at  $t$  is  $\frac{\int \omega W_{\omega,\tau}\{x(t)\}d\omega}{\int W_{\omega,\tau}\{x(t)\}d\omega} = \frac{d\phi}{dt}$

=instantaneous frequency, as it should!

2. Product:  $W_{\omega,\tau}\{x_1(t)x_2(t)\} = \int W_{\omega',\tau}\{x_1(t)\}W_{\omega-\omega',\tau}\{x_2(t)\}d\omega'$

Product in time → convolution in freq.

3. Convolution:  $W_{\omega,\tau}\{x_1(t)*x_2(t)\} = \int W_{\omega,\tau'}\{x_1(t)\}W_{\omega,\tau-\tau'}\{x_2(t)\}d\tau'$

Product in freq → convolution in time.

4. Example of cross terms:  $x(t) = e^{j\omega_1 t} + e^{j\omega_2 t} \rightarrow$

$$W_{\omega,\tau}\{x(t)\} = \delta(\omega-\omega_1) + \delta(\omega-\omega_2) + 2\delta(\omega - \frac{\omega_1+\omega_2}{2}) \cos(\omega_2-\omega_1)t$$

5. Can compute in frequency domain as

$$W_{\omega,\tau}\{x(t)\} = \int x(\tau + t/2)x^*(\tau - t/2)e^{-j\omega t} dt$$

$$= \frac{1}{2\pi} \int X(\omega - \theta/2)X^*(\omega + \theta/2)e^{-j\tau\theta} d\theta$$

using (generalized) Parseval's theorem on  $\mathcal{F}_{t \rightarrow \theta}$

and also  $\mathcal{F}\{x(t - \tau)\} = X(\theta)e^{-j\tau\theta}$ .

## CHOI-WILLIAMS (RID) T-F DISTRIBUTION

**Def'n:** The Choi-Williams t-f distribution,  
a type of RID (Reduced Interference Distribution):

$$CW_{\omega,\tau}\{x(t)\} = \int \int \frac{1}{|t|} k\left(\frac{u-\tau}{t}\right) x(u+t/2) x^*(u-t/2) e^{-j\omega t} du dt.$$

where the *kernel*  $k(t) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-t^2/(2\sigma^2)}$

### 1. Advantages:

- a. Kernel  $k(t)$  tends to suppress cross terms, at the price of reducing resolution.
  - b. Marginals still satisfied for any  $\sigma$ .
- ### 2. Can trade these off by varying $\sigma$ :
- a.  $\sigma \rightarrow 0$ :  $k(t) \rightarrow \delta(t) \rightarrow$ Wigner $\rightarrow$ cross terms.
  - b.  $\sigma \rightarrow \infty$ :  $k(t) \rightarrow k \rightarrow$ no cross terms, no resolution!
- ### 3. Other choices of kernel $k(t)$ can be made.

## EXAMPLES: SUPPRESSION OF CROSS TERMS

**Example 1:**  $x(t) = e^{j1t} + e^{j9t}$

(a): Wigner (b): Choi-Williams  $\sigma = 0.001$  (c):  $\sigma = 0.003$

**Example 2:**  $x(t) = e^{(a_1+jb_1)t^2/2+j\omega_1t} + e^{(a_2+jb_2)t^2/2+j\omega_2t}$

(sum of 2 chirps; instantaneous freqs  $\omega = b_i t + \omega_i, i = 1, 2$ )

(a): Wigner (b): Choi-Williams

## COHEN'S CLASS OF TFDS USING KERNELS

**IDEA** (L. Cohen): Define time-freq dist. (TFD) as:

1. Define *ambiguity function*  $A_{\theta,\tau}\{x(t)\}$  as

$$A_{\theta,\tau}\{x(t)\} = \int x(u + \tau/2)x^*(u - \tau/2)e^{j\theta u} du.$$

Note  $4\pi^2 W_{\omega,t}\{x(t)\} = \mathcal{F}\mathcal{F}\{A_{\theta,\tau}\{x(t)\}\}$   
 $= \int \int A_{\theta,\tau}\{x(t)\} e^{-j\theta t} e^{-j\omega\tau} d\theta d\tau.$

2. Define  $W'_{\omega,t}\{x(t)\}$  as

$$W'_{\omega,t}\{x(t)\} = \frac{1}{4\pi^2} \int \int k(\theta, \tau) A_{\theta,\tau}\{x(t)\} e^{-j\theta t} e^{-j\omega\tau} d\theta d\tau$$

for some *kernel*  $k(\theta, \tau)$ .

Note this convolves  $W_{\omega,\tau}\{x(t)\}$  with  $\mathcal{F}^{-1}\mathcal{F}^{-1}\{k(\theta, \tau)\}$ .

### Examples:

1.  $k(\theta, \tau) = 1 \rightarrow$ Wigner-Ville (easy).
2.  $k(\theta, \tau) = e^{-\theta^2\tau^2/\sigma} \rightarrow$ Choi-Williams using
 
$$\int e^{j\theta(u-t)} e^{-\theta^2\tau^2/\sigma} d\theta = \frac{\sqrt{\sigma\pi}}{\tau} e^{-\frac{\sigma}{4\tau^2}(u-t)^2}$$
3. Other choices possible (see below).
4. To satisfy marginals, need  $k(0, \tau) = k(\theta, 0) = 1$ .  
 To satisfy energy conservation, need  $k(0, 0) = 1$ .
5. Kernel can be made signal-dependent (Baraniuk-Jones).
6. Recover signal from  $W'_{\omega,t}\{x(t)\}$  using

$$2\pi s(t)s^*(0) = \int \int \int \frac{W'_{\omega,u}\{x(t)\}}{k(\theta,\tau)} e^{j\omega t} e^{j\theta u} e^{-j\theta t/2} du d\omega d\theta$$

For spectrogram, how can we recover signal from  $|\mathcal{F}|$ ?  
 Is magnitude of STFT, not  $\mathcal{F}$ !

# RECONSTRUCTION OF DISPERSIVE LAYERED MEDIA

(Chien and Yagle, 1989)

## Given:

1. Dispersive (frequency-dependent absorption) layered (absorption, density vary with depth) medium.
2. Reflection response  $R(t)$  of medium to an impulse (deconvolved from actual source explosion signal)

## Goal:

To reconstruct density  $\rho_i$  and absorption  $Q_i$  in  $i^{th}$  layer of acoustic medium for all  $i$ .

## Assumptions:

1. Absorption factor =  $e^{-\frac{|\omega|}{2cQ_i}z}$   
 $z$ =distance in  $i^{th}$  layer,  $c$ =wave speed.
2. Reflection coefficient  $r_i = \frac{\rho_i c_i - \rho_{i+1} c_{i+1}}{\rho_i c_i + \rho_{i+1} c_{i+1}}$ .
3. Neglect multiple reflections.
4. Layers equally thick (use very thin layers).

## Approach:

1. Compute time-freq distribution of  $R(t)$ .
2. Window tfd in time to separate layer reflections.
3. Intervals between layer reflections  $\rightarrow c_i$ .
4. Fit exponential curve to freq variation at each time.  
"Time constant" =  $2 \sum_i Q_i c_i$  (actually fit line to log).
5. Amplitude of tfd (intercept)  $\rightarrow r_i \rightarrow \rho_i$ .

# TRACKING NON-CONSTANT BLOOD FLOW IN MRI

(Sahiner and Yagle, 1989)

**Given:**

1. Moving object (e.g., blood) undergoing MRI.  
Motion blurs MRI signal; must deconvolve it.
2. MRI signal of moving object.  
Both motion and object are unknown.

**Goal:**

To reconstruct both the unknown velocity and the unknown (nuclear) spin density

**Assumptions:**

1. Everything occurs in one direction (projection).
2. Velocity affine function of location
  - a. Reasonable for thin MRI slice excitation of steady-state flow in blood vessel.
  - b. Use different functions in each slice → trapezoidal approx. to velocity vs. location.
3.  $\hat{\rho}(\frac{k}{v_1}(e^{v_1 t} - 1)) \approx \text{constant}$ .

**Approach:**

1. Compute time-freq distribution of MRI signal

$$x(t) = e^{j2\pi k \frac{v_0}{v_1^2}(e^{v_1 t} - v_1 t - 1)} \hat{\rho}(\frac{k}{v_1}(e^{v_1 t} - 1))$$

velocity  $v(x) = v_0 + v_1 x$ ,  $\hat{\rho}(k) = \mathcal{F}\{\text{spin density}(x)\}$ .

2. Instant freq =  $\frac{d\phi}{dt} = \frac{kv_0}{v_1}(e^{v_1 t} - 1)$
3. Trajectory of tfd maxima →  $v_0, v_1$ .