

HIGH RESOLUTION SPECTRA ANALYSIS: ADVANCES AND APPLICATIONS

Tryphon Georgiou
University of Minnesota

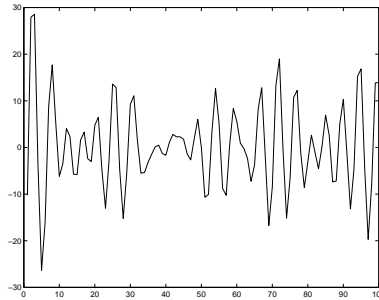
IN HONOR OF DR. MARC JACOBS

Collaboration: Chris Byrnes and Anders Lindquist

AFOSR Grant No. AF/F49620-00-1-0078

Given time-series data:

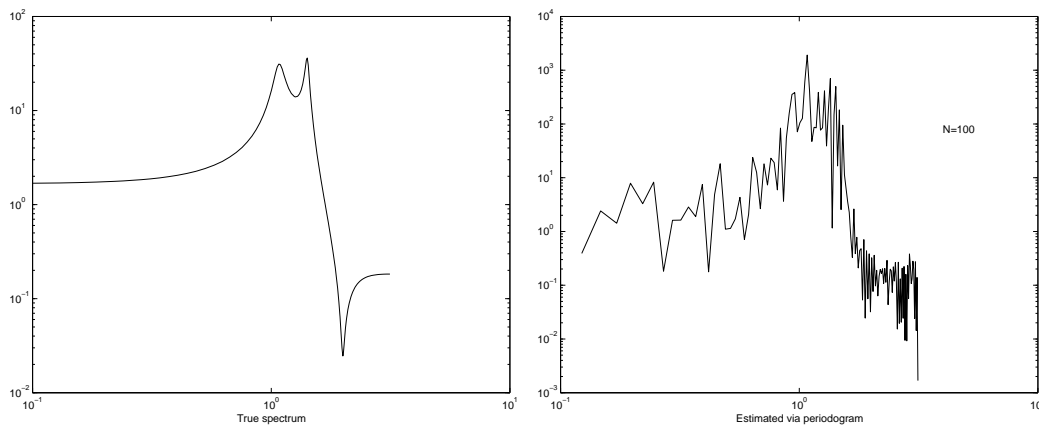
$$\{u_0, u_1, u_2, \dots, u_{N-1}\}$$



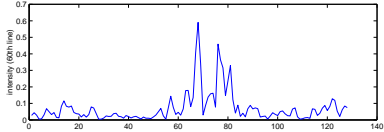
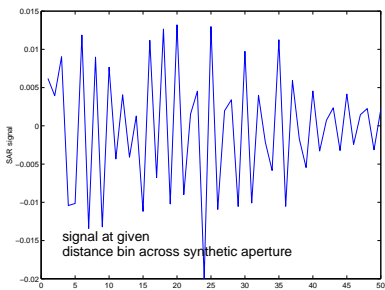
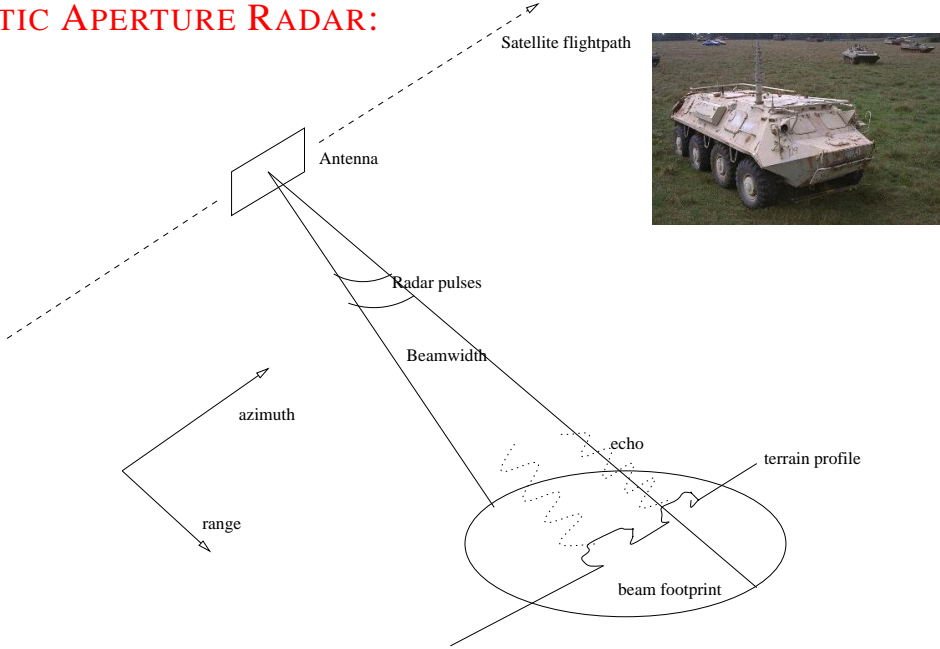
determine the **power spectrum of y** ,
i.e., periodicities and “color”

Methods:

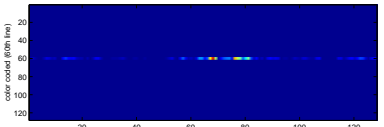
- Periodogram, FFT
- Model based (ARMA,...)
- Modern nonlinear (Maximum-entropy, maximum-likelihood,...)



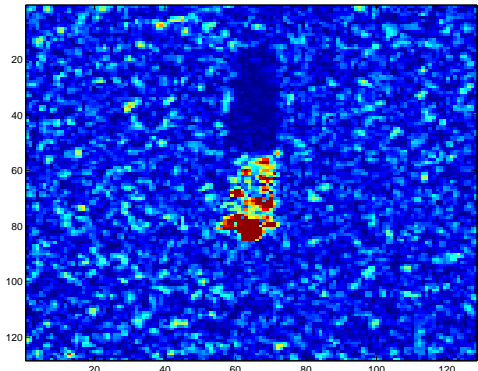
SYNTHETIC APERTURE RADAR:



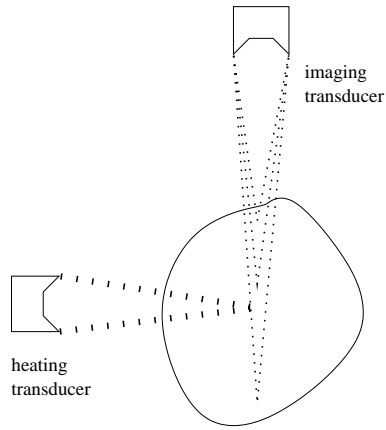
⇒



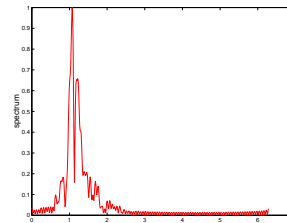
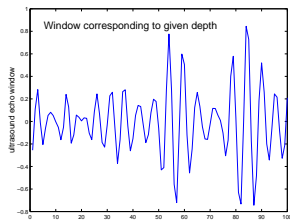
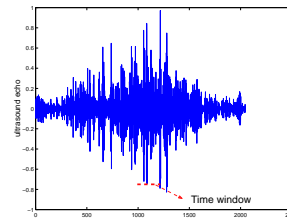
Line by line produces:



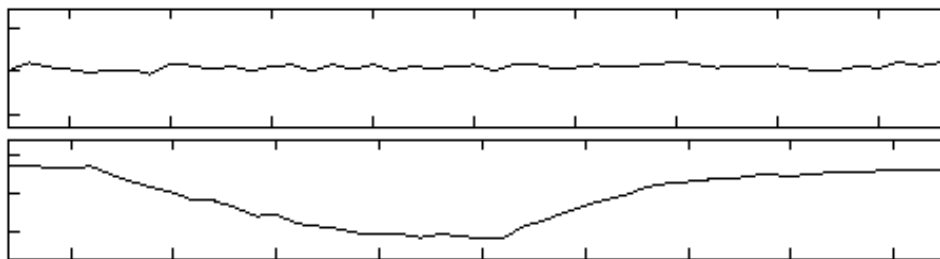
APPLICATIONS: NONINVASIVE TEMPERATURE SENSING



Ultrasound echo:



frequency vs. time \Rightarrow temperature profile



$$u_k = \int_{-\pi}^{\pi} e^{jk\theta} dv(\theta)$$

$\rho(\theta)d\theta \sim E\{dv(\theta)^2\}$ “energy density across frequencies”

Covariance statistics & spectral density

$$c_k = E\{u_t u_{t+k}\}$$



$$c_k = \frac{1}{2\pi} \int_{-\pi}^{\pi} e^{-jk\theta} \underbrace{\rho(\theta)d\theta}_{d\mu(\theta)}$$

c_0, c_1, c_2, \dots autocorrelation sequence



$F(z) = \frac{1}{2}c_0 + c_1z + c_2z^2 + \dots$ “positive-real” function



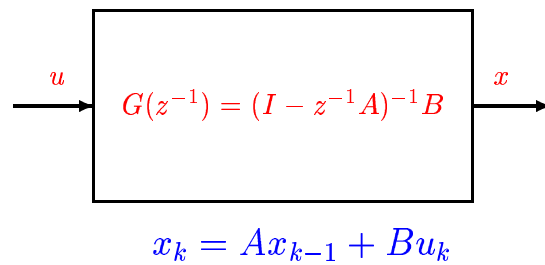
$$\begin{aligned}\rho(\theta) &= \operatorname{Re}(F(e^{j\theta})) \\ &= \dots c_2e^{-2j\theta} + c_1e^{-j\theta} + c_0 + c_1e^{j\theta} + c_2e^{2j\theta} + \dots \geq 0\end{aligned}$$

Given finite data c_0, \dots, c_N
all consistent spectra are given by:

$$\rho = \operatorname{Re} \left(\frac{A+BQ}{C+DQ} \right) \text{ with } Q \text{ a “free” parameter}$$

Spectral analysis \Leftrightarrow analytic interpolation

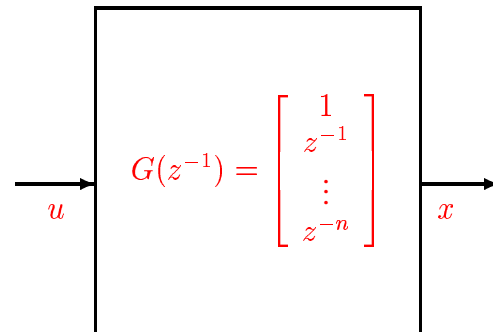
- **Theory** \Leftarrow
- **Academic examples** \Leftarrow
- **Applications** \Leftarrow



What is the structure of the state-covariance matrix $\Sigma := E\{xx^\}$?*

What are all spectra consistent with Σ ?

$$\Sigma = \int_{-\pi}^{\pi} \left(G(e^{j\theta}) \frac{d\mu(\theta)}{2\pi} G(e^{j\theta})^* \right)$$



$$x_k = \begin{bmatrix} u_k \\ u_{k-1} \\ \vdots \\ u_{k-n} \end{bmatrix}$$

$$\Sigma = \begin{bmatrix} c_0 & c_1 & \dots & c_n \\ c_1^* & c_0 & \dots & c_{n-1} \\ \vdots & \vdots & & \vdots \\ c_n^* & c_{n-1}^* & \dots & c_0 \end{bmatrix}$$

$$\begin{aligned}\Sigma &= \int_{-\pi}^{\pi} \left(G(e^{j\theta}) \frac{d\mu(\theta)}{2\pi} G(e^{j\theta})^* \right) \\ &\vdots \\ &= BH + H^* B^* + A\Sigma A^*\end{aligned}$$

THM: With (A, B) controllable pair, A stable, $\Sigma \geq 0$:

Σ is a covariance of $x_k = Ax_{k-1} + Bu_k$

\Leftrightarrow

$\Sigma = BH + H^* B^* + A\Sigma A^*$ has a solution H

\Leftrightarrow

$$\text{rank} \begin{bmatrix} \Sigma - A\Sigma A^* & B \\ B^* & 0 \end{bmatrix} = \text{rank} \begin{bmatrix} 0 & B \\ B^* & 0 \end{bmatrix}$$

THM: All input spectra consistent with Σ are

$$d\mu(\theta) \sim \lim_{r \rightarrow 1} \operatorname{Re} (F(re^{j\theta})) d\theta$$

where

$$F(\lambda) = F_0(\lambda) + Q(\lambda)V(\lambda) \text{ is positive-real}$$

Data A, B, Σ, H :

$$F_0(\lambda) = H(I - \lambda A)^{-1}B,$$

$$V(\lambda) = D + C\lambda(I - \lambda A)^{-1}B \text{ all-pass,}$$

$Q(\lambda)$ free parameter.

\Rightarrow LFT parametrization of all spectra consistent with Σ

Entropy:

$$\mathbb{I}(\mu) := \frac{1}{2\pi} \int_{-\pi}^{\pi} \log \det(\dot{\mu}(\theta)) d\theta$$

THM: A unique maximizing spectrum is

$$d\mu_o(\theta) := \left(\Phi(e^{j\theta})^{-1} \Omega^{-1} (\Phi(e^{j\theta})^{-1})^* \right) d\theta$$

Where:

$$\Phi(\lambda) := (B^* \Sigma^{-1} B)^{-1} B^* \Sigma^{-1} (I - \lambda A)^{-1} B,$$

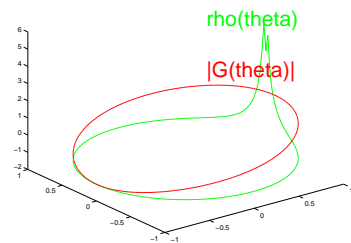
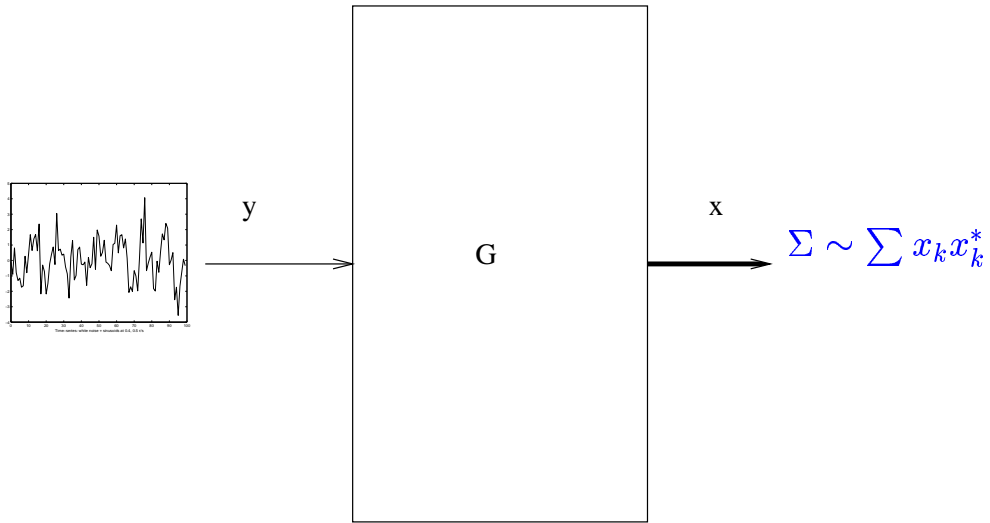
$$\Omega := (B^* \Sigma^{-1} B)^{-1}.$$

If $u_\ell = \text{“white-noise”} + \sum_{k=1}^m \sqrt{\rho_k} e^{j\ell\omega_k}$, then

$$\Sigma = \rho_0 I + \sum_{i=1}^m \rho_i G(e^{j\omega_i}) G(e^{j\omega_i})^*$$

THM: There exists a unique minimal decomposition of Σ corresponding to a sum of sinusoids

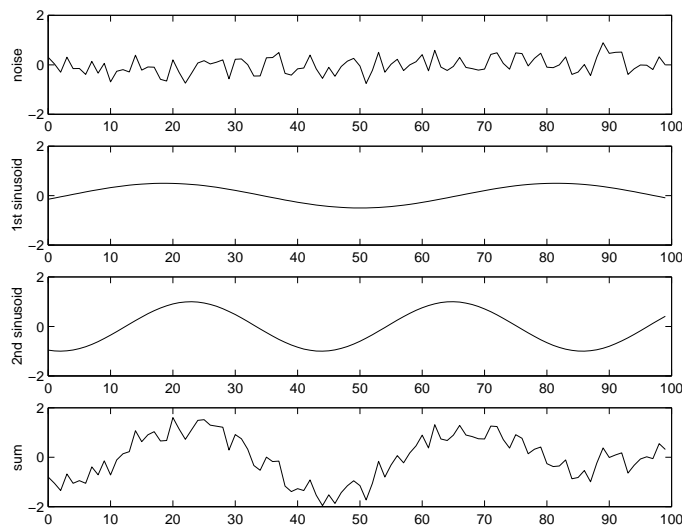
- minimal complexity spectra
- Kullback-Leibler distance & approximation of spectra



$$\Sigma = \frac{1}{2\pi} \int (G(e^{j\theta}) d\mu(\theta) G(e^{j\theta})^*)$$

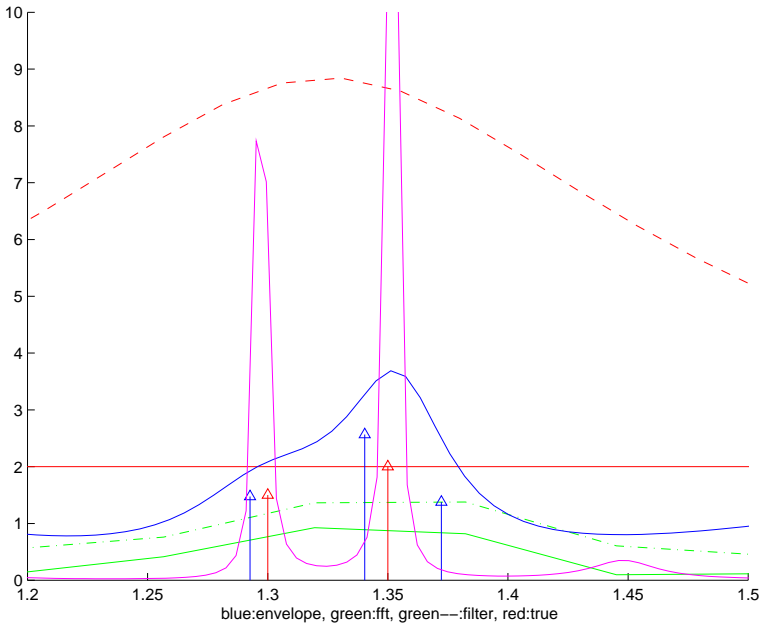
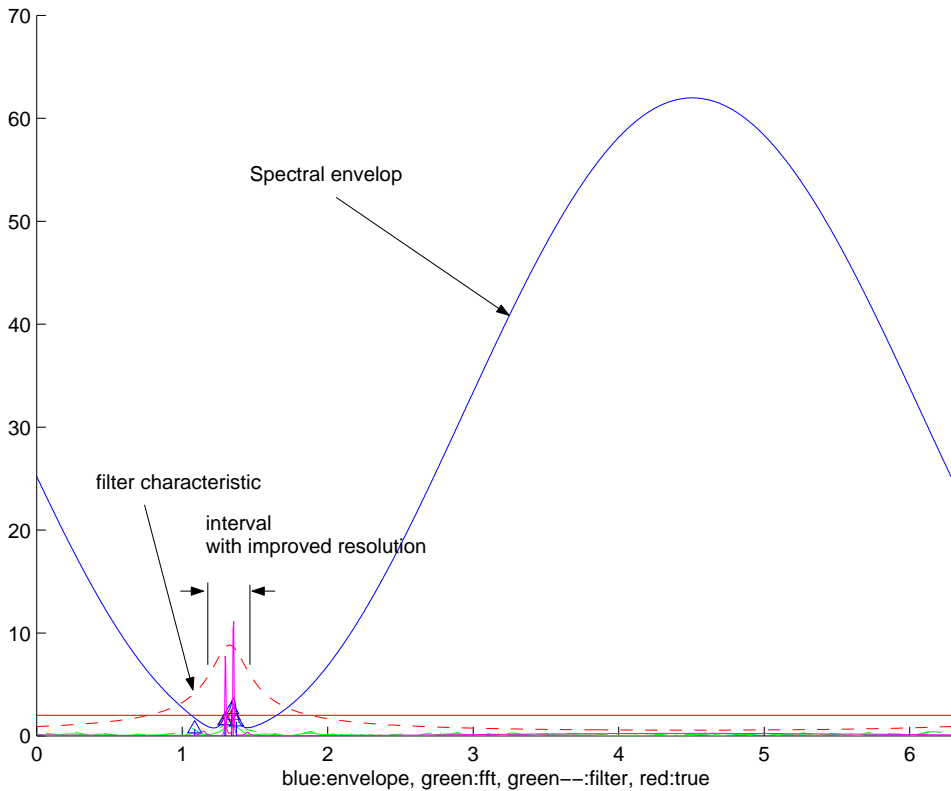
RESOLVING SINUSOIDS
BEATING FT-UNCERTAINTY BOUND

$$\mathbf{u}_k = \nu_k + A_1 \sin(\omega_1 k + \phi_1) + A_2 \sin(\omega_2 k + \phi_2), k = 1, \dots, n,$$

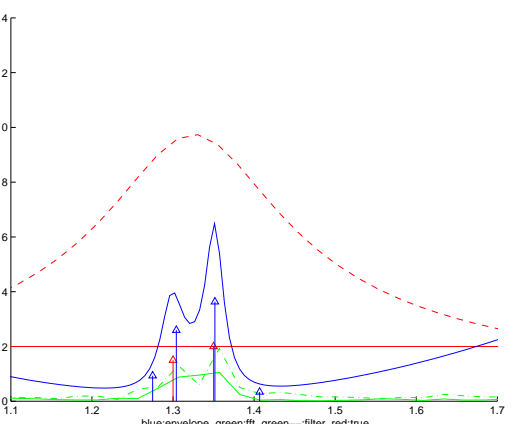
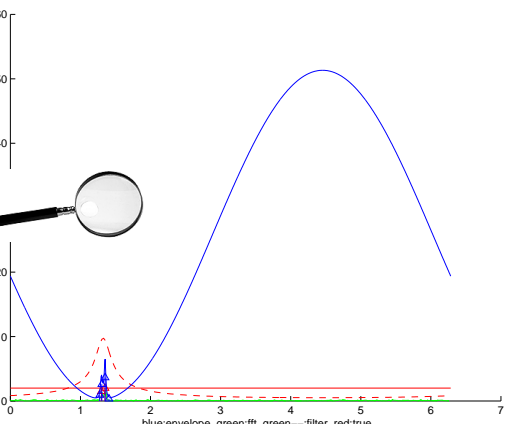
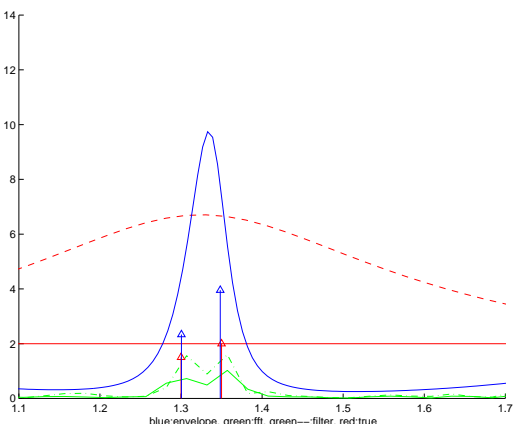
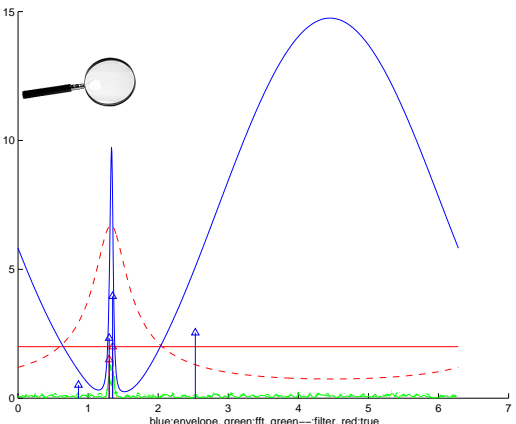
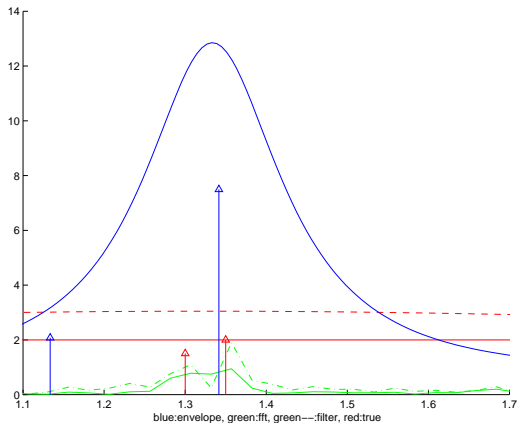
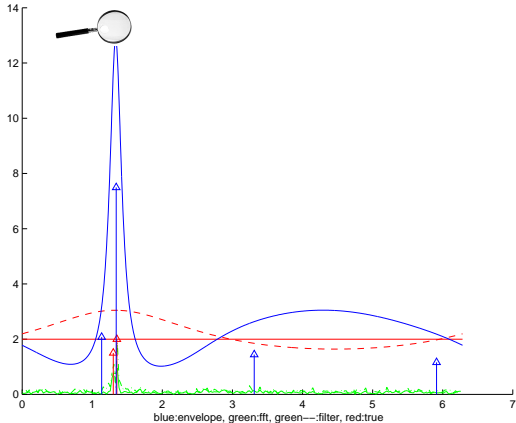


Noise, sinusoid 1, sinusoid 2, and their sum

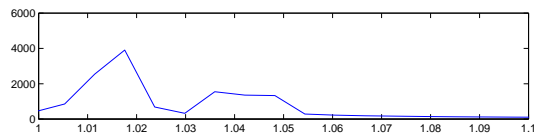
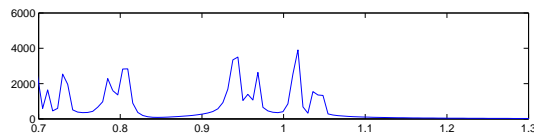
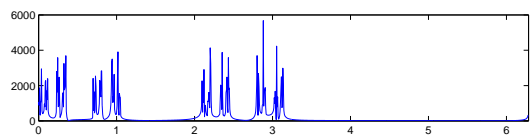
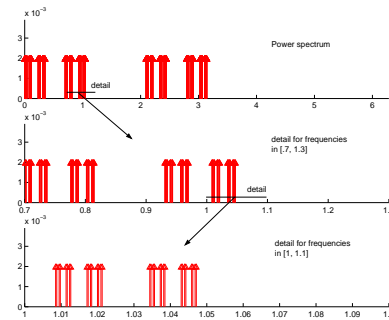
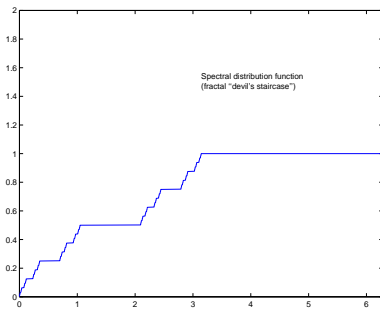
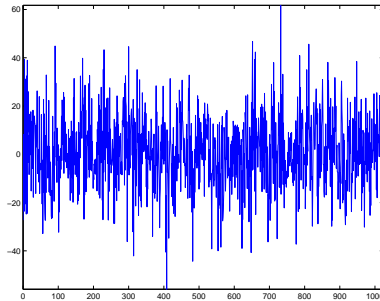
$$\omega_2 - \omega_1 < \frac{2\pi}{n} = \text{Fourier uncertainty bound}$$



ENVLPS & BNDRY INTRPLNTS

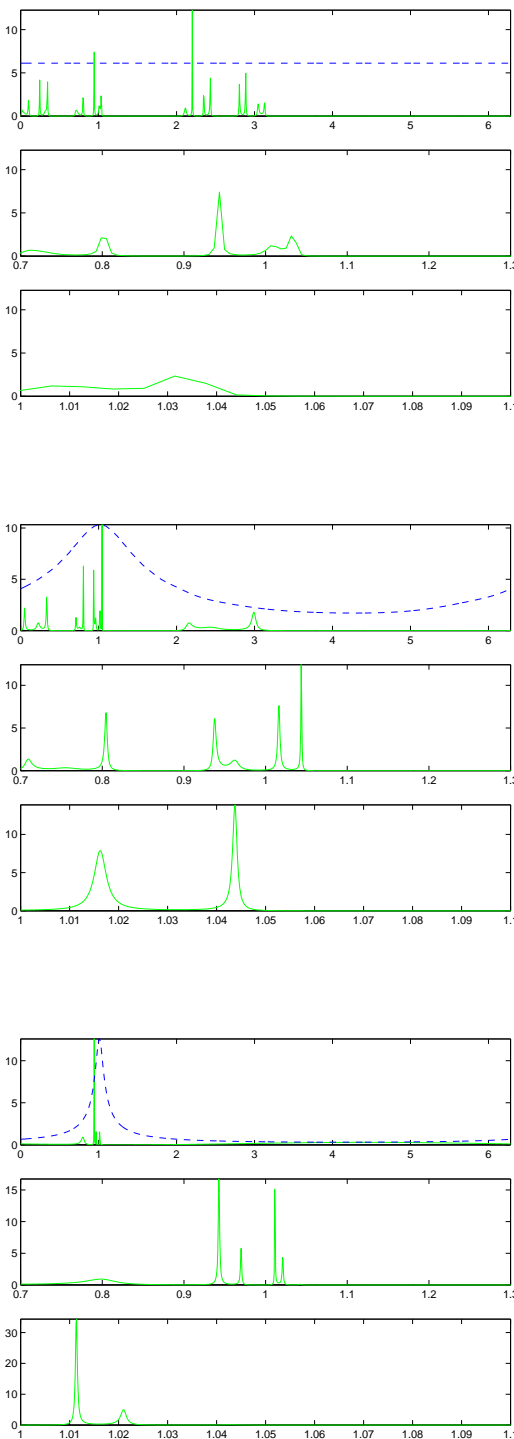


FRactal SPECTRUM LIMITS TO RESOLUTION?

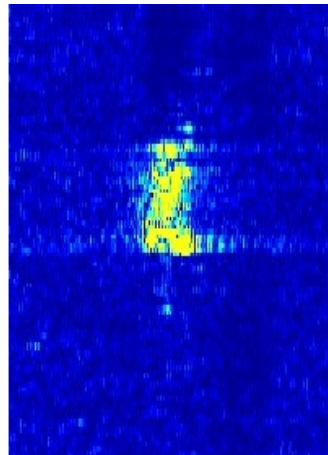
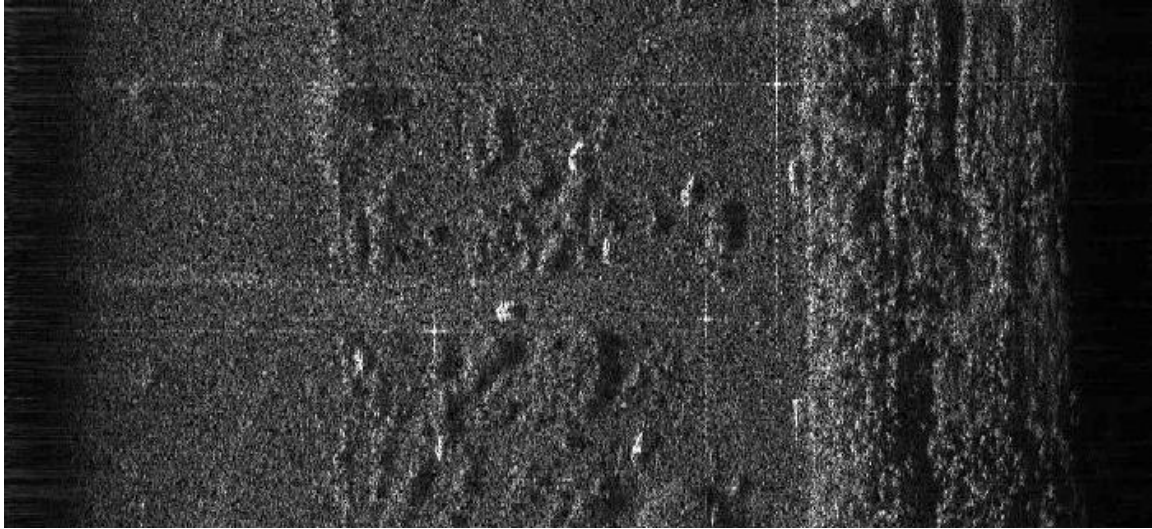


periodogram

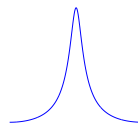
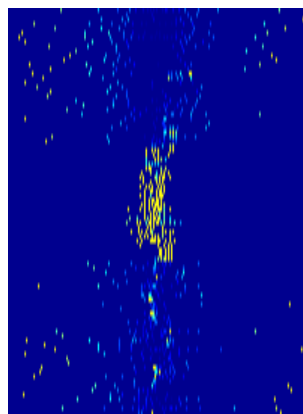
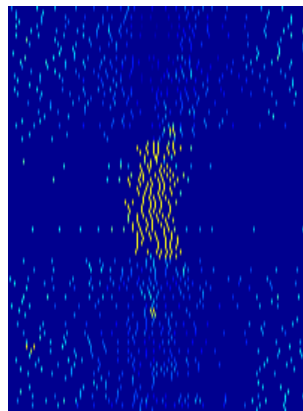
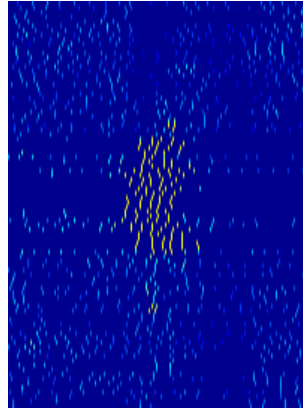
FRactal Spectrum:
with ME Interpolants

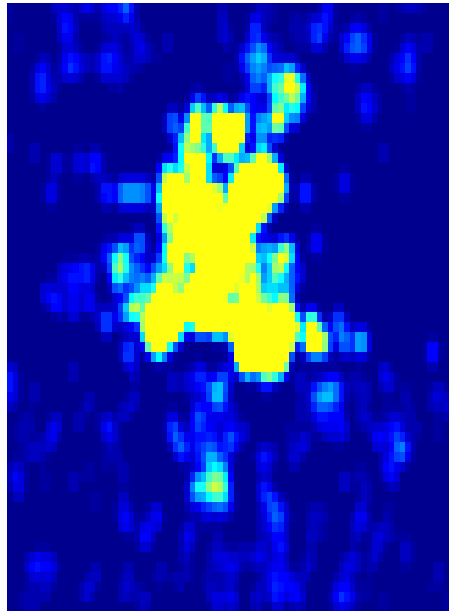
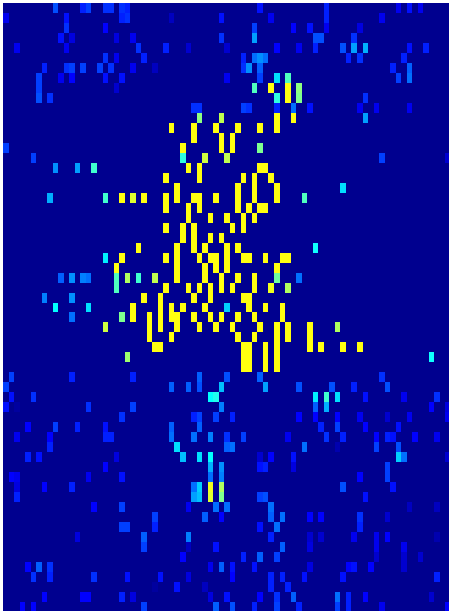


— Maximum entropy spectra
- - - Focusing filter

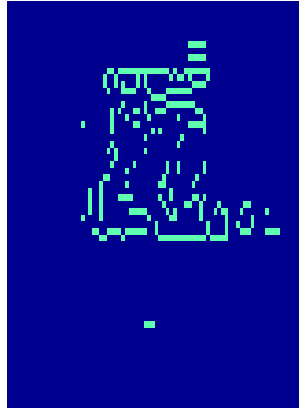
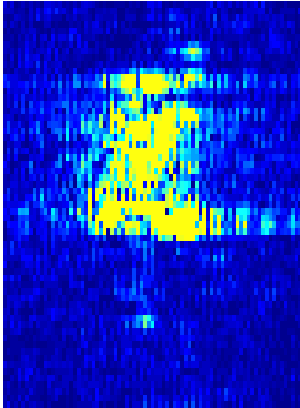


T-72

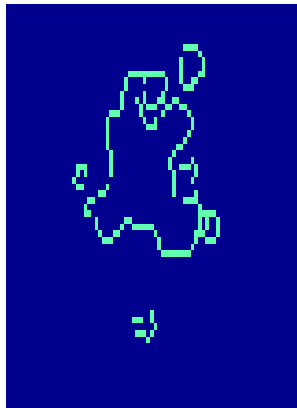
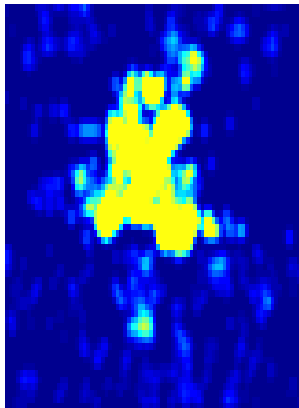


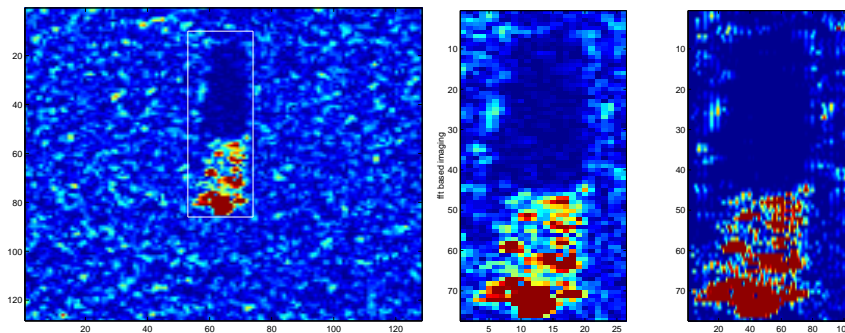


fft

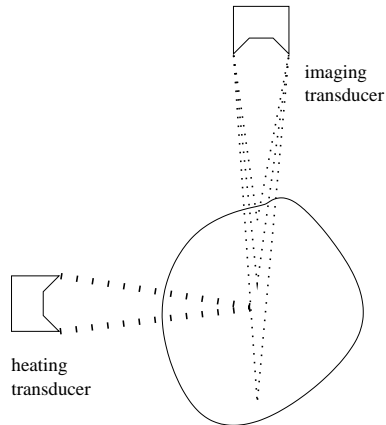


high resolution

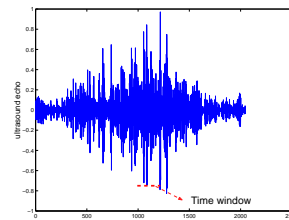




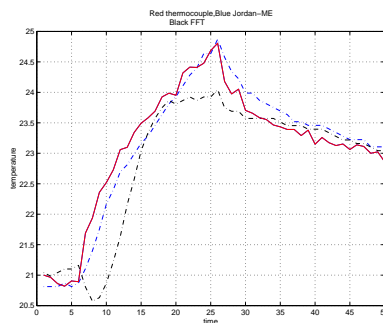
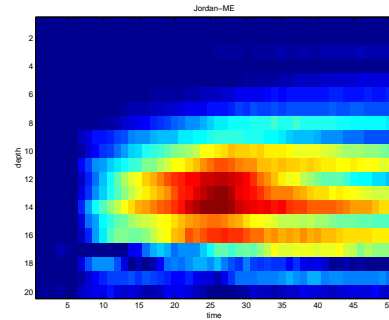
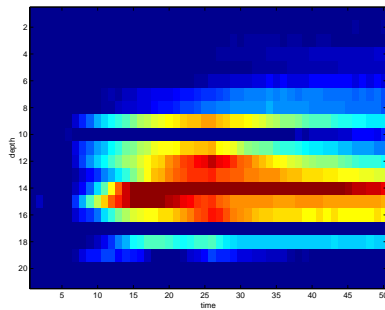
NON-INVASIVE ULTRASOUND TEMPERATURE SENSING



Ultrasound echo:



periodogram analysis vs. high resolution methods



Comparison with thermocouple (thermocouple, periodogram, high resolution)

- Collaboration: E. Ebbini

SUMMARY

- generalized statistics \sim analytic interpolation
- high resolution, applications

QUESTIONS AND ON-GOING RESEARCH PROGRAM:

- how can we quantify resolution?
- tradeoffs between variance and resolution
seeking an “ H_∞ -like paradigm”
- spacio-temporal dynamics and non-uniform arrays
- applications: SAR, medical imaging, polarimetry

Matlab code and references at:

<http://www.ece.umn.edu/users/georgiou>