

# HIGH RESOLUTION SPECTRAL ANALYSIS

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*Spectral Analysis ~ Analytic Interpolation*

Motivation

Theoretical advances

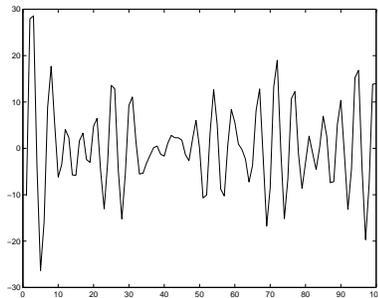
Tools for high resolution analysis

Applications: SAR, ultrasound

- Signal analysis/filtering

Given time-series data:

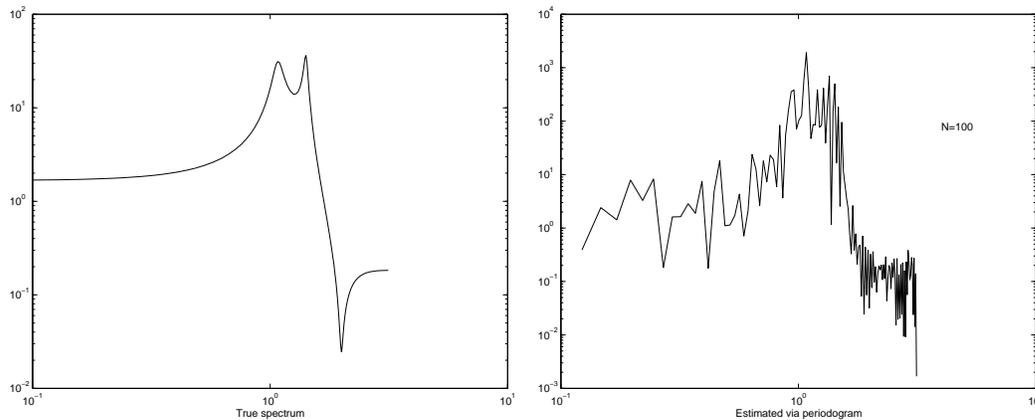
$$\{u_0, u_1, u_2, \dots, u_{N-1}\}$$



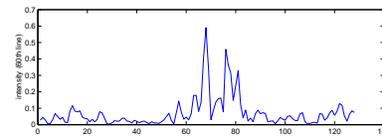
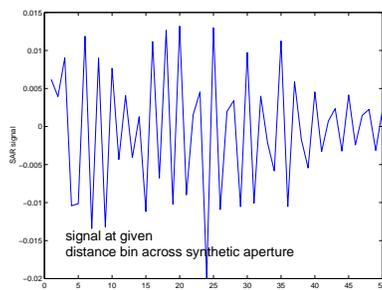
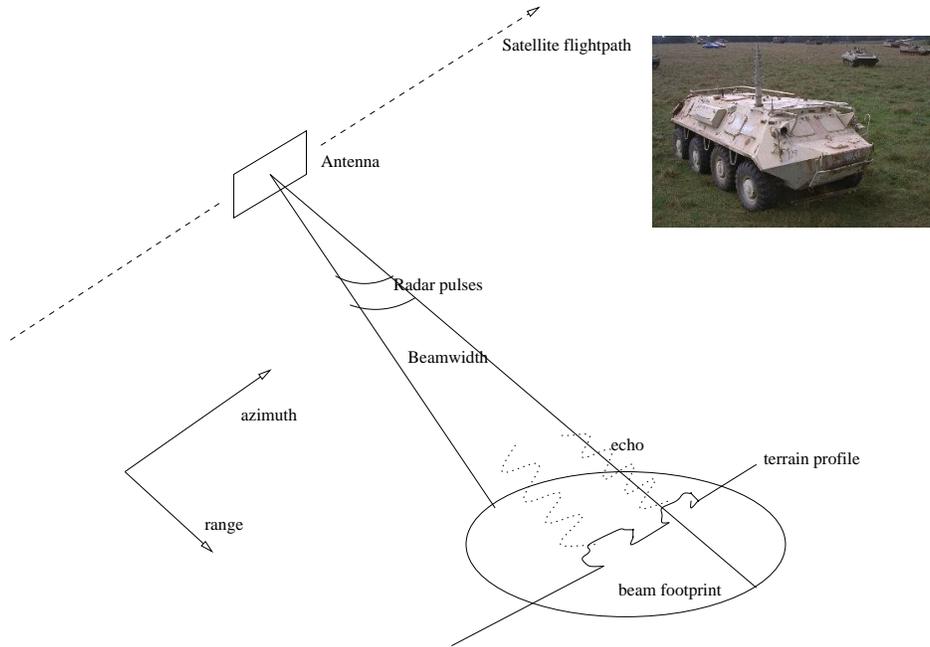
determine the **power spectrum of  $y$** ,  
i.e., periodicities and “color”

**Methods:**

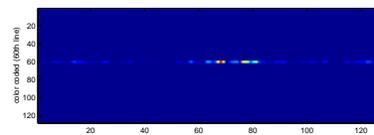
- Periodogram, FFT
- Model based (ARMA,...)
- Modern nonlinear (Maximum-entropy, maximum-likelihood,...)



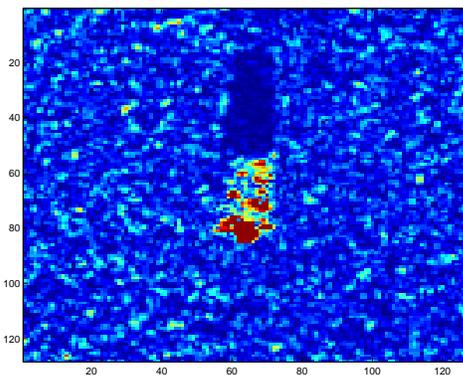
SAR:



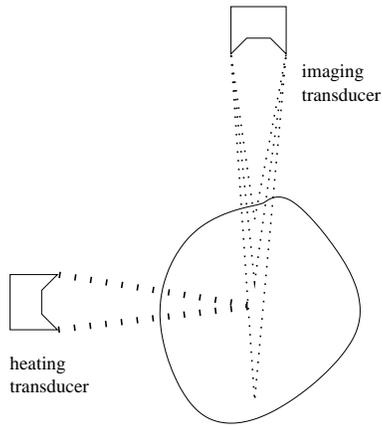
⇒



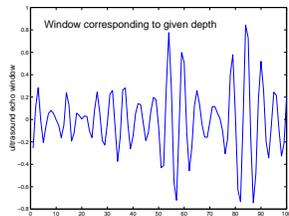
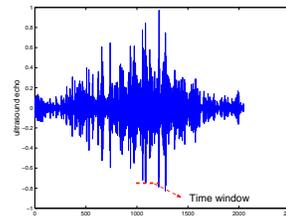
Line by line produces:



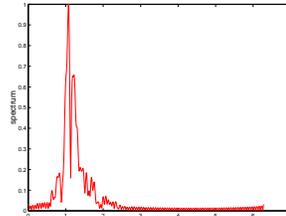
APPLICATIONS: SPEECH, CODING/COMMUNICATIONS, MR IMAGING, ULTRASOUND, SAR ETC.  
 ULTRASOUND – Noninvasive temperature sensing CONT.



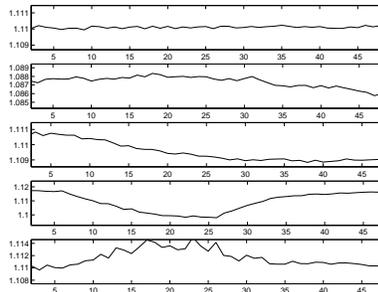
Ultrasound echo:



⇒



harmonic shift ⇒ temperature profile



- Reference/collaboration: E. Ebbini

**Moment problem**

Find “mass density”  $\rho(x)$ :

$$\int_I x^n \rho(x) dx = c_n, \text{ for } n = 0, 1, \dots, N.$$

**Analytic interpolation**

Find  $f$  analytic, having  $\operatorname{Re}(f) > 0$ , and

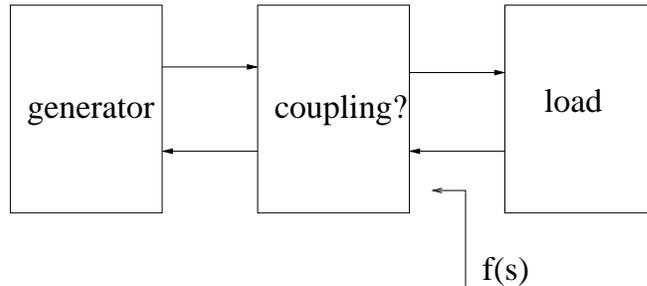
$$f(z) \sim c_0 + c_1 z + c_2 z^2 + \dots + c_N z^N + \dots$$

- $\rho \sim \operatorname{Re}(f)$
- Analogous problems for  $|f| < 1$ , and  $f(z_i) = w_i$ .

Circuits:        positivity  $\Leftrightarrow$  passivity  
Control:    contractiveness  $\Leftrightarrow$  signal attenuation  
Signals:        positivity  $\Leftrightarrow$  admissible probability structure

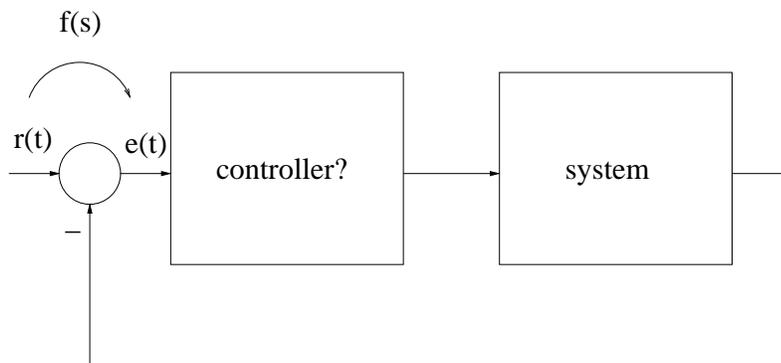
**Circuit theory** (Fano, Youla-Saito, Helton, ...)

*Maximal power transfer*



**Control** (Zames, Tannenbaum, Doyle, Francis, Helton, ...)

*Robust performance, etc.*



**Signal analysis** (Levinson, Burg, Pisarenko, ...)

*Modeling from covariance statistics,...*

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

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

### Covariance statistics & spectral density

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

$\Updownarrow$

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

$c_0, c_1, c_2, \dots$  autocorrelation sequence

$\Updownarrow$

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

$\Updownarrow$

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

**RESULTS: Given finite data  $c_0, \dots, c_N$ :**

- **PARAMETRIZATION OF SOLUTIONS:**

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

- SPECIFIC CHOICES OF  $Q$  LEAD TO DIFFERENT “METHODS”:

**Maximum Entropy/central solution**

maximizing  $\int \log(\operatorname{Re} f(e^{j\theta}))d\theta$

**Pisarenko/MUSIC/ESPRIT**

...

- **Generalized covariance/statistics**

*Collaboration with Chris Byrnes and Anders Lindquist*

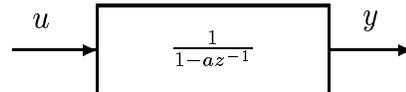
**Local improvement of resolution**

**Spectral analysis  $\sim$  problem in generalized interpolation**

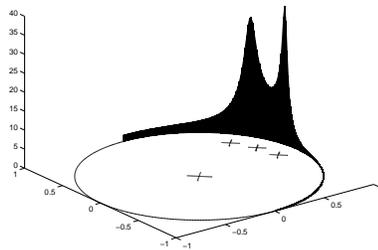
- Multivariable theory  $\Leftarrow$
- Special canonical interpolants  $\Leftarrow$
- Applications  $\Leftarrow$

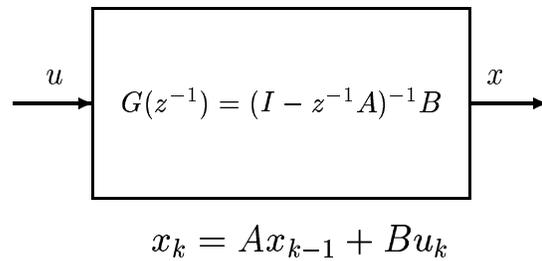
SPECTRAL ANALYSIS OF TIME SERIES  
GENERALIZED ANALYTIC INTERPOLATION

$$\begin{aligned}
 E\{y(k)^2\} &= E\{(u(k) + au(k-1) + a^2u(k-2) + \dots)^2\} \\
 &= c_0(1 + a^2 + a^4 + \dots) \\
 &\quad + 2c_1a(1 + a^2 + a^4 + \dots) \\
 &\quad + 2c_2a^2(1 + a^2 + a^4 + \dots) + \dots \\
 &= \frac{2}{1-a^2}F(a)
 \end{aligned}$$



$$F(a) = \frac{1-a^2}{2}E\{y(k)^2\}$$

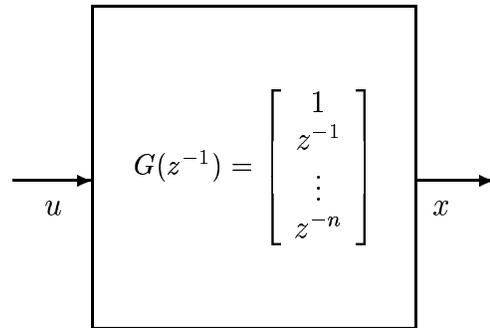




*What is the structure of the state-covariance matrix  $\Sigma := E\{xx^*\}$ ?*

*What are all spectra consistent with  $\Sigma$ ?*

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



- $G(z^{-1})$  “steering vector” of a uniform array

$$\mathbf{x}_k = \begin{bmatrix} u_k \\ u_{k-1} \\ \vdots \\ u_{k-n} \end{bmatrix}$$

and

$$\Sigma = \begin{bmatrix} c_0 & c_1 & \dots & c_n \\ c_1^* & c_0 & \dots & c_{n-1} \\ \vdots & \vdots & & \vdots \\ c_n^* & c_{n-1}^* & \dots & c_0 \end{bmatrix} \text{ is (block) Toeplitz}$$

- “Steering vector”  $G(z^{-1}) = (I - z^{-1}A)^{-1}B$  with nontrivial dynamics

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

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

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

$\Leftrightarrow$

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

$\Leftrightarrow$

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

Let  $\Sigma$  state-covariance and  $H$  as before:

All input spectra consistent with  $\Sigma$  are

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

where

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

**Data:**

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

and

$$V = \left[ \begin{array}{c|c} A & B \\ \hline C & D \end{array} \right] \text{ is inner,}$$

$$\lambda := z^{-1}, \text{ w.l.o.g. } AA^* + BB^* = I$$

- $\Sigma$  is the relevant "Pick" matrix
- LFT parametrization of all  $F$ 's
- for scalar input  $\Sigma = W + W^*$  with  $W$  commuting with  $A$

Define the entropy functional

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

Then the unique spectrum maximizing  $\mathbb{I}$  is

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

where

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

and

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

- If  $\Phi(\lambda) = I - \lambda P_1 - \lambda^2 P_2 + \dots$

$$\hat{u}_k = P_1 u_{k-1} + P_2 u_{k-2} + \dots,$$

has minimal variance  $\Omega$ , in fact it is a **min-max optimal predictor** (i.e., minimizes uniformly over spectra consistent with  $\Sigma$ ).

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

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

**Given ANY state-cov.  $\Sigma$ ,**  
 **$\exists$  a unique decomposition as above**  
**for a minimal value of  $m$**

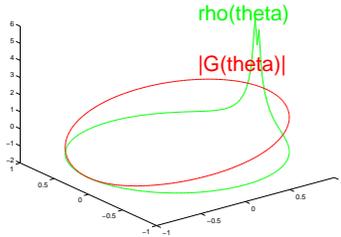
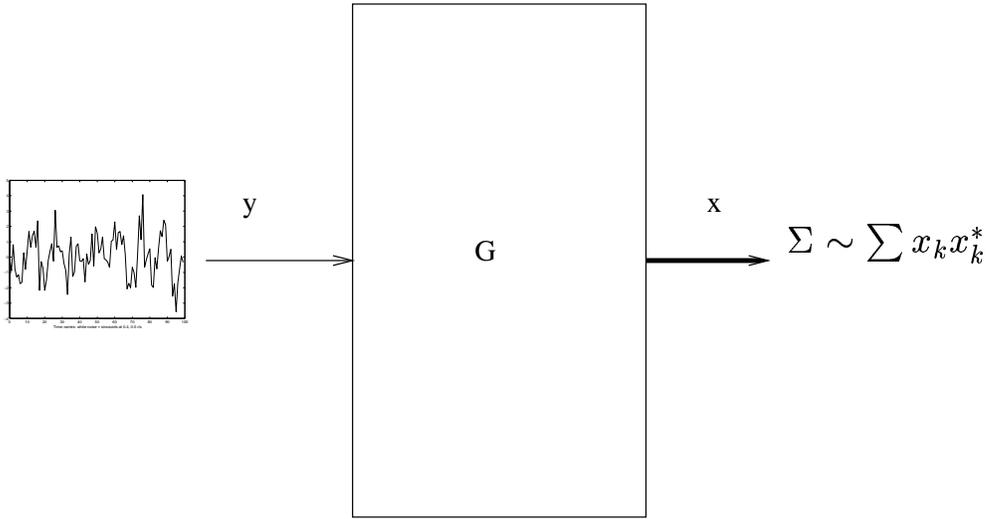
- $\rho_0 = \sigma_{\min}(\Sigma)$ ,  $\rho$ 's and  $\omega$ 's obtained via **SVD** of  $\Sigma$
- If  $G(z^{-1}) = (1, z^{-1}, \dots, z^{-n}) \Rightarrow$   
*Carathèodory-Fèjer/Pisarenko - decomposition*  
Methods of MUSIC, ESPRIT

If  $\Sigma, \Sigma_0$  are both state covariances for  $G(z^{-1})$   
and  $\Sigma_1 := \Sigma - \Sigma_0 \geq 0$ , then  $\Sigma_1$  is also a state covariance

**Pf:** ...  $\Sigma_1 = A\Sigma_1A^* + BH_1 + H_1^*B^*$  and  $\geq 0$ .  $\square$

$\Sigma_0 = G(e^{j\theta})G(e^{j\theta})^* \sim$  complex sinusoidal mass

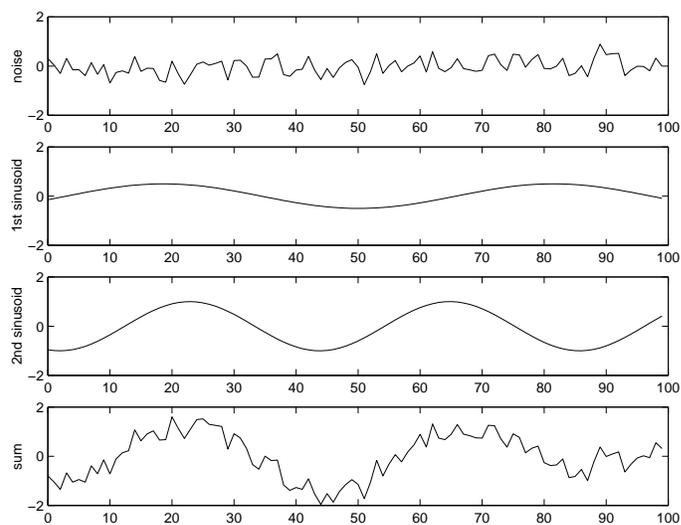
$\rho(\theta) = \max \{ \rho : \Sigma - \rho G(e^{j\theta})G(e^{j\theta})^* \geq 0 \} \Rightarrow$  envelope



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

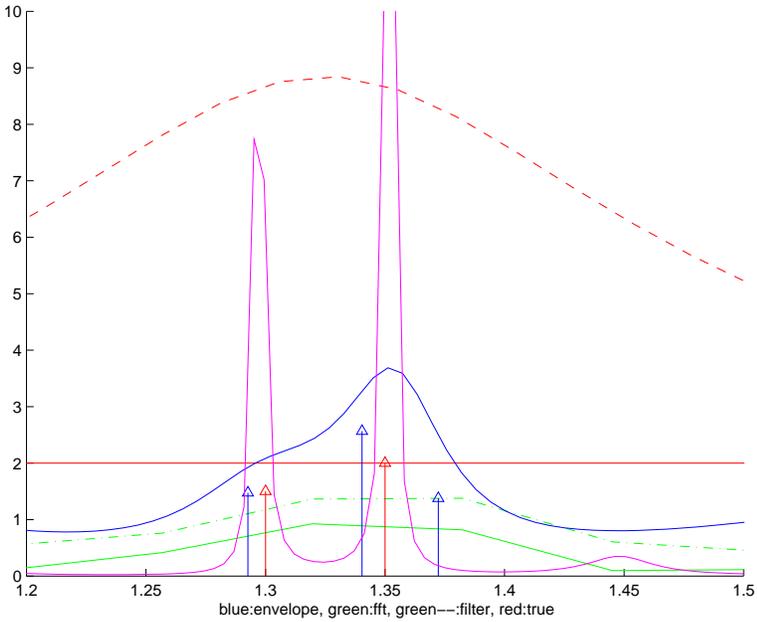
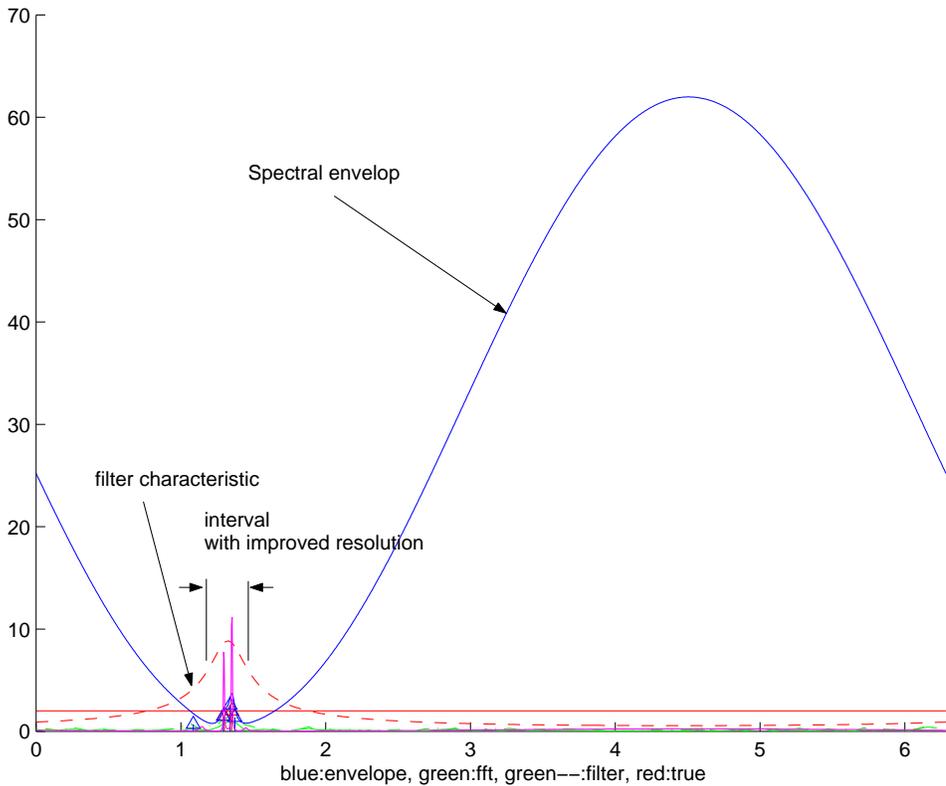
RESOLVING SINUSOIDS  
BEATING FT-UNCERTAINTY BOUND

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

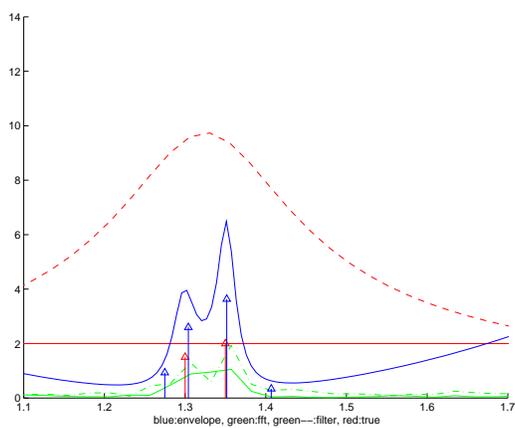
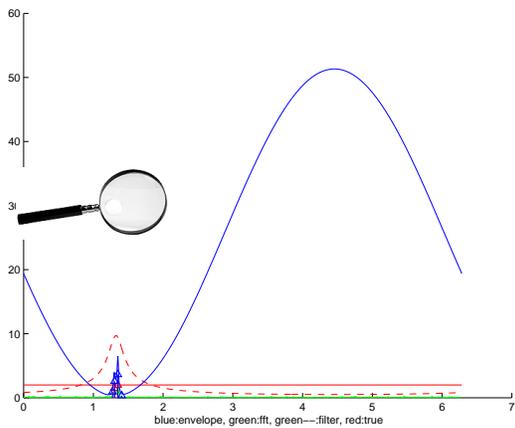
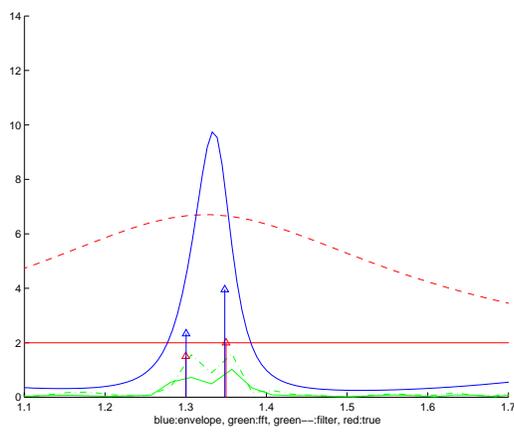
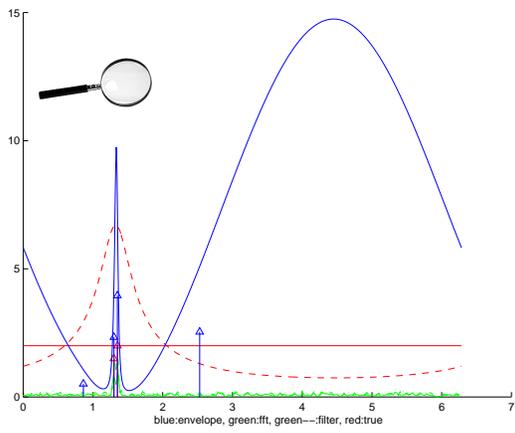
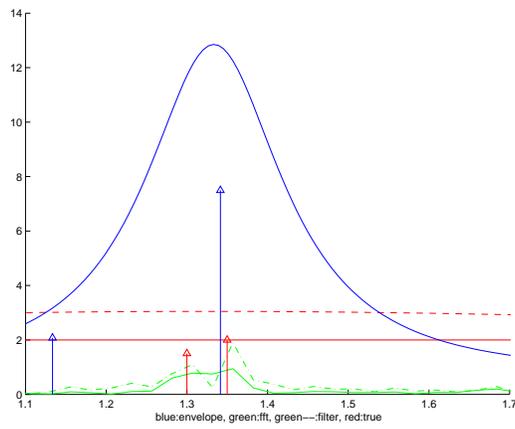
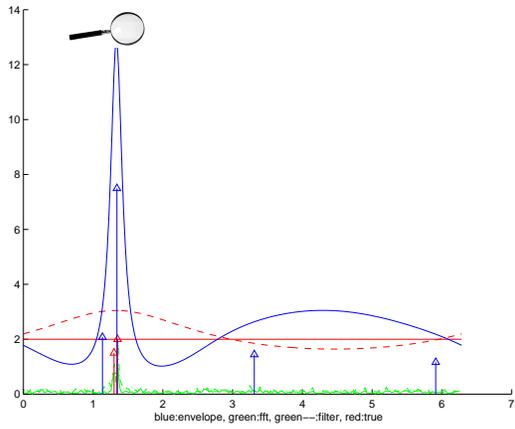


Noise, sinusoid 1, sinusoid 2, and their sum

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



# ENVLPS & BNDRY INTRPLNTS



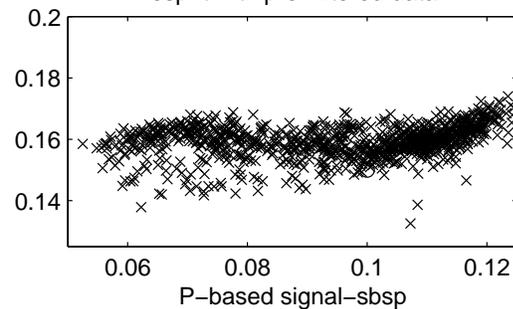
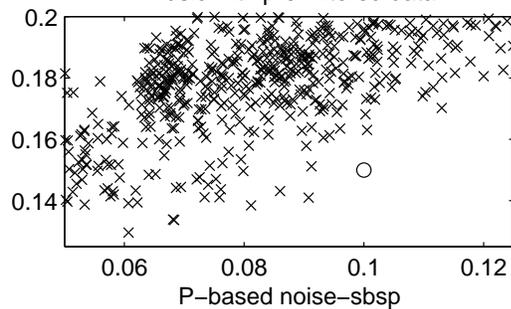
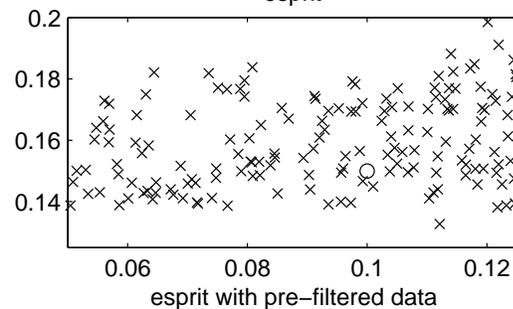
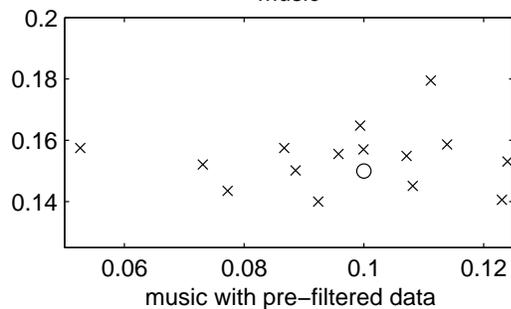
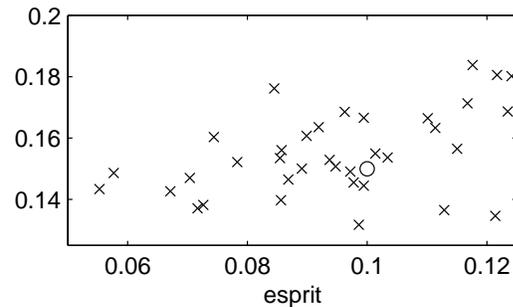
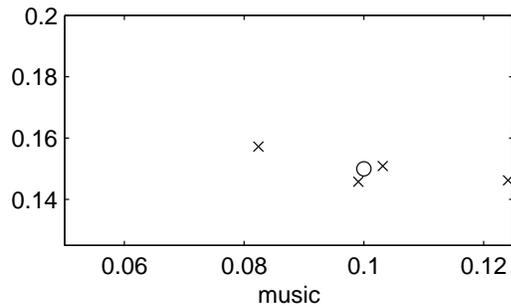


□ Darts (1000 runs) thrown by:

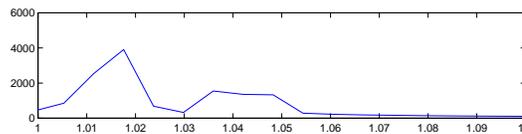
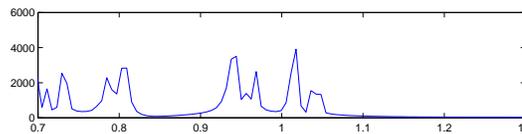
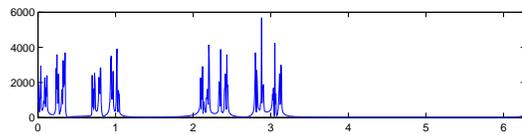
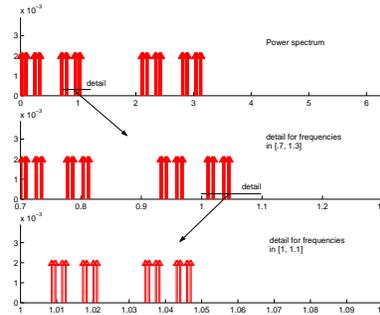
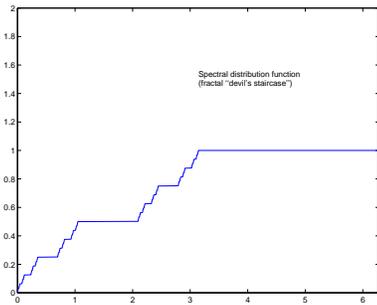
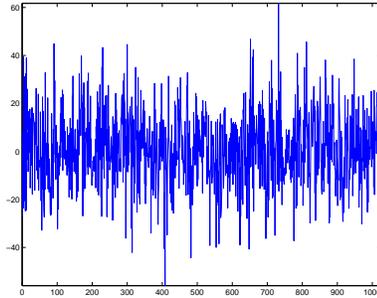
row 1: MUSIC, ESPRIT

row 2: MUSIC, ESPRIT on “prefiltered data”

row 3: Boundary interpolant of state-cov data

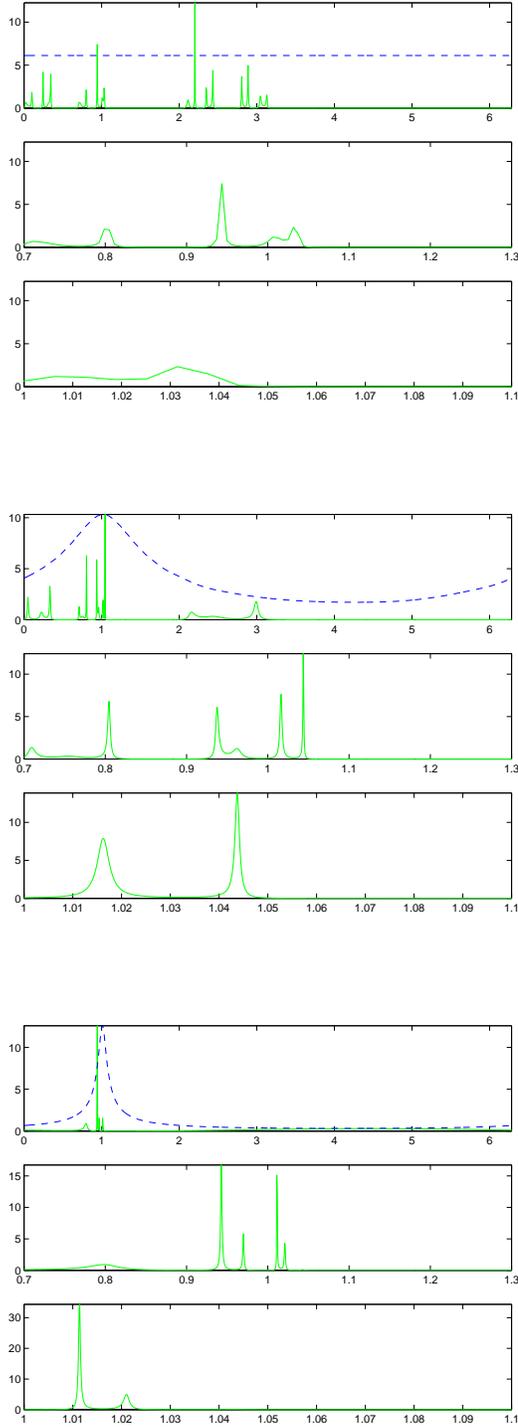


# FRactal SPECTRUM LIMITS TO RESOLUTION?

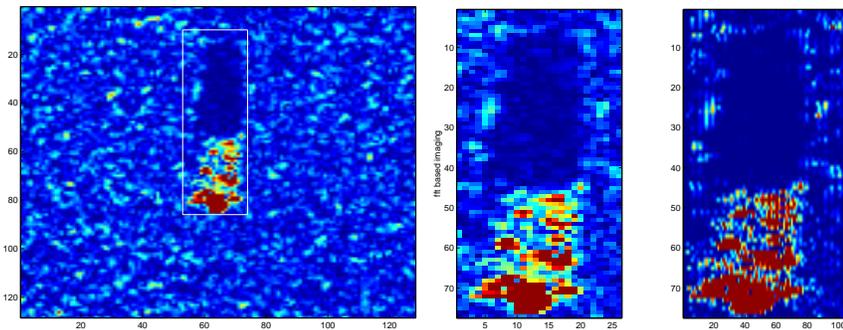


periodogram

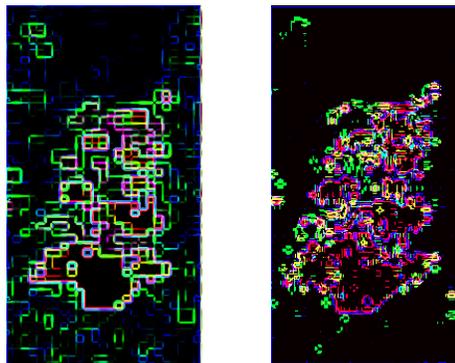
FRactal Spectrum:  
FOCUSING WITH ME INTERPOLANTS



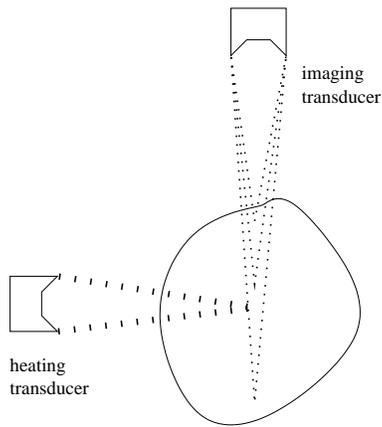
APC image  
MSTAR image of APC  
and detail vs. “high resolution” reconstruction



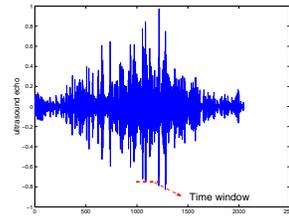
- Using edge detection (MSTAR-image vs. high resolution)



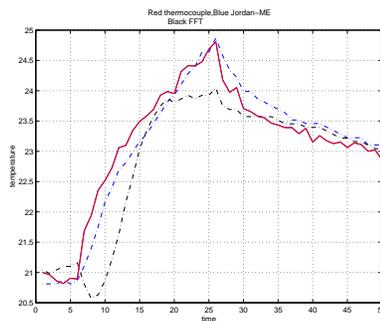
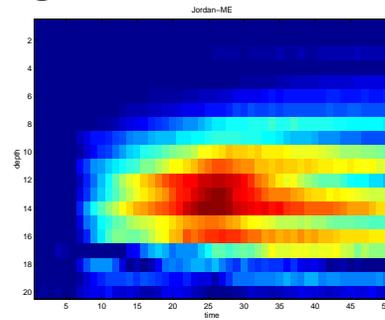
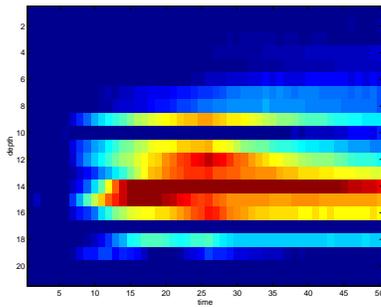
## NON-INVASIVE ULTRASOUND TEMPERATURE SENSING



### Ultrasound echo:



### periodogram analysis vs. high resolution methods



**Comparison with thermocouple**  
(thermocouple-red, periodogram-black, high resolution-blue)

- Reference/collaboration: E. Ebbini

## RESULTS

- covariance statistics  $\sim$  analytic interpolation
- high resolution methods, applications

## WORK IN PROGRESS

- spacio-temporal dynamics and non-uniform arrays  
(Laurent Baratchard-INRIA)
- temperature sensing  
(Emad Ebbini, and A. Nasiri-Amini)
- SAR imaging & target recognition  
(Allen Tannenbaum)
- polarimetric SAR  
(Firooz Sadjadi-Lockheed)

## FUNDAMENTAL QUESTIONS

- ¿**how can we quantify resolution?**  
¿fundamental limits beyond Fourier uncertainty?
- tradeoffs between variance and resolution  
seeking an “ $H_\infty$ -like paradigm”

Matlab code and references at:

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