

Back to the roots: a spectrum of what was realized

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Outline

- 1 Spectra
- 2 Realization and spectra
- 3 Kronecker
- 4 1D LTI regular
- 5 Applications
- 6 mD SI models
- 7 Applications
- 8 Misfit
- 9 Many more...
- 10 Valedictum

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Eigenvalues and vectors: For matrix $A \in \mathbb{R}^{n \times n}$:

$$Ax = x\lambda, x \in \mathbb{C}^n, \lambda \in \mathbb{C}, x \neq 0$$

Characteristic equation - fundamental theorem of algebra:

$$p(\lambda) = \det(\lambda I_n - A) = \lambda^n + \alpha_1 \lambda^{n-1} + \dots + \alpha_{n-1} \lambda + \alpha_n = 0$$

Since Galois, for $n \geq 5$: no solution in radicals !

Numerical linear algebra = iterative algorithms + finite precision machines

Cayley-Hamilton:

$$A^n + \alpha_1 A^{n-1} + \dots + \alpha_{n-1} A + \alpha_n I_n = 0$$

Eigenvalue decomposition - Jordan Canonical Form (JCF):

$$A = X J X^{-1}$$

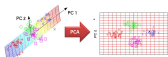
Eigen-objects: Operator (object) = object \times scalar

Continuous spectrum:

$$d(e^{(\alpha \pm j\beta t)})/dt = (e^{(\alpha \pm j\beta t)})(\alpha \pm j\beta t), (d./dt + \int . dt)e^{\alpha t} = e^{\alpha t} \left(\frac{\alpha^2 + 1}{\alpha} \right)$$

Discrete spectrum: e.g. standing waves

Dimensionality Reduction Principal Component Analysis



PCA

Let Y_1 and Y_2 be two orthogonal matrices of size D by m , and let $u \in \text{span}(Y_1)$ and $v \in \text{span}(Y_2)$ be unit vectors.



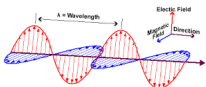
The first principal angle/canonical corr between $\text{span}(Y_1)$ and $\text{span}(Y_2)$ is

$$\cos \theta_1 = \max_{u \in \text{span}(Y_1), v \in \text{span}(Y_2)} |u^T v|, \text{ subject to } \|u\| = \|v\| = 1.$$

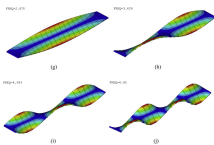
Can. Corr./Principal Angles

$$\frac{\partial^2 y}{\partial t^2} = \frac{1}{v_w^2} \frac{\partial^2 y}{\partial t^2}$$

Wave equation



Maxwell's laws



Modal shapes

1. $\nabla \cdot \mathbf{D} = \rho_V$
2. $\nabla \cdot \mathbf{B} = 0$
3. $\nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t}$
4. $\nabla \times \mathbf{H} = \frac{\partial \mathbf{D}}{\partial t} + \mathbf{J}$

Maxwell's field equations

The PageRank problem



The PageRank random surfer
1. With probability beta, follow a random-walk step

2. With probability (1-beta), jump randomly - dist. v

Goal find the stationary dist. x

$$\mathbf{x} = \beta \mathbf{A} \mathbf{D}^{-1} \mathbf{x} + (1 - \beta) \mathbf{v}$$

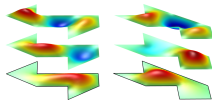
Alg Solve the linear system $(\mathbf{I} - \beta \mathbf{A} \mathbf{D}^{-1}) \mathbf{x} = (1 - \beta) \mathbf{v}$

Solution Jump vector

Symmetric adjacency matrix Diagonal degree matrix

David Gleich - Purdue CS256

Graph spectral analysis



Hear the shape of a drum?

Answer: No !

Series resonant condition:

$$Z = R \quad \omega = \frac{1}{\sqrt{LC}} \quad \text{Phase} = \phi = 0$$

Imaginary

Phase diagram

$Z = R + jX_c - jX_l$

$X_c = -\frac{1}{\omega C}$

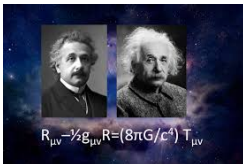
$X_l = \omega L$

$Z = \sqrt{R^2 + (X_c - X_l)^2}$

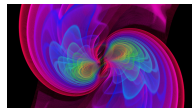
$\text{Phase} = \phi = \tan^{-1} \left[\frac{X_c - X_l}{R} \right]$

$$H(t)|\psi(t)\rangle = i\hbar \frac{\partial}{\partial t} |\psi(t)\rangle$$

Schrödinger equation



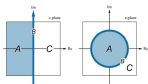
Matter curves spacetime moves matter



Gravitational waves

Mapping between the s plane and the z plane

- Primary strip and Complementary strips (cont.)



Mapping regions of the s -plane onto the z -plane

Stability

Kalman Decomposition Theorem

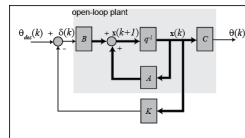
An equivalence transformation exists to transform any state-space equation into the following canonical form:

$$\begin{bmatrix} \dot{x}_{cc} \\ \dot{x}_{co} \\ \dot{x}_{oc} \\ \dot{x}_{oo} \end{bmatrix} = \begin{bmatrix} A_{cc} & 0 & A_{12} & 0 \\ A_{21} & A_{22} & A_{23} & A_{24} \\ 0 & 0 & A_{32} & 0 \\ 0 & 0 & A_{42} & A_{44} \end{bmatrix} \begin{bmatrix} x_{cc} \\ x_{co} \\ x_{oc} \\ x_{oo} \end{bmatrix} + \begin{bmatrix} B_{cc} \\ B_{co} \\ 0 \\ 0 \end{bmatrix} u(t)$$

$$y = \begin{bmatrix} C_{cc} & 0 & C_{oc} & 0 \end{bmatrix} \begin{bmatrix} x_{cc} \\ x_{co} \\ x_{oc} \\ x_{oo} \end{bmatrix} + D u(t)$$

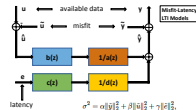
where subscript co indicates the controllable and observable, and the bar over the subscript indicates *not*.

Controllability/observability



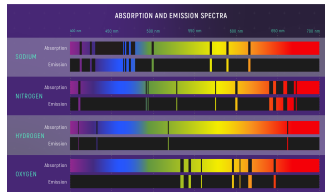
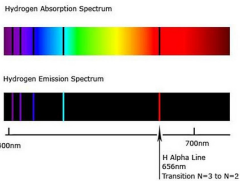
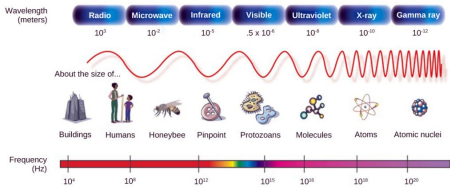
Pole placement

Observers	Kalman Filter Riccati Hamil. EVP	H_{∞} -filter Riccati Sympl. EVP
Control	LQR Riccati Hamil. EVP	H_{∞} -control Riccati Sympl. EVP



LS LTI System ID = EVP !

If you want to find the secrets of the universe, think in terms of energy, frequency and vibration. Nikola Tesla



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From Kepler to Newton: realization from data to (internal) state



Kepler (1571-1630)

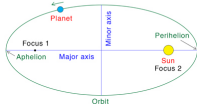


Newton (1642-1726)

Kepler's Laws

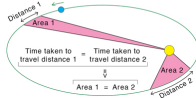
First Law

All planets move around the Sun in elliptical orbits with the Sun at one of the foci



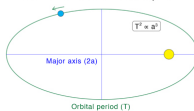
Second Law

A planet sweeps out equal areas in equal intervals of time



Third Law

The square of the orbital period of a planet is proportional to the cube of the orbit's semi-major axis

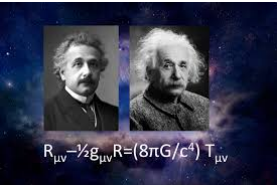


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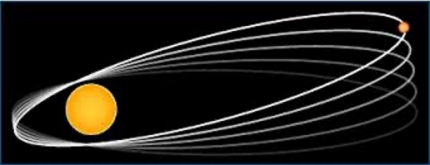
$$\mathbf{F} = m \mathbf{a}$$

$$F = G m M / r^2$$

Einstein and Popper



Einstein (1879-1955)



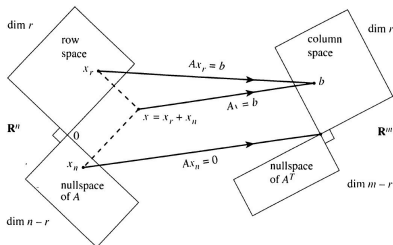
Popper (1902-1994)

Popper's demarcation criterion :
 A model/theory is scientific when refutable
 Models forbid more than they allow

In this talk, eigen-(multi-)spectra to...:

- characterize (multi-)shift invariant subspaces
- realize regular linear time-invariant state space models from 1D model-compliant data
- realize mD shift-invariant state space models from mD model-compliant data
- realize 1D state space models from data that are not model-compliant via mD realization

Fundamental Theorem of linear algebra: 4 matrix subspaces and their dimension/rank (SVD)



Fundamental Theorem of algebra: n roots

$$p(\lambda) = \det(\lambda I_n - A) = \lambda^n + \alpha_1 \lambda^{n-1} + \dots + \alpha_{n-1} \lambda + \alpha_n = 0$$

The Theorems meet in Cayley - Hamilton

$$A^n + \alpha_1 A^{n-1} + \dots + \alpha_{n-1} A + \alpha_n I_n = 0$$

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Taylor / McLaurin series expansion

$$\begin{aligned}
 f(z) &= f(0) + \frac{f'(0)}{1!}z + \frac{f''(0)}{2!}z^2 + \dots + \frac{f^{(k)}(0)}{k!}z^k + \dots \\
 &= \gamma_0 + \gamma_1 z + \gamma_2 z^2 + \dots + \gamma_k z^k + \dots
 \end{aligned}$$

When rational in z^{-1} ?

$$\begin{aligned}
 f(z^{-1}) &= \frac{\beta_0 z^n + \beta_1 z^{n-1} + \dots + \beta_{n-1} z + \beta_n}{\alpha_0 z^n + \alpha_1 z^{n-1} + \dots + \alpha_{n-1} z + \alpha_n} \\
 &= \frac{\sum_{i=0}^n \beta_i z^{n-i}}{\sum_{i=0}^n \alpha_i z^{n-i}} \\
 &= \gamma_0 + \gamma_1 z^{-1} + \gamma_2 z^{-2} + \dots = \sum_{i=0}^{\infty} \gamma_i z^{-i} \\
 \Rightarrow \sum_{i=0}^n \beta_i z^{n-i} &= \left(\sum_{i=0}^n \alpha_i z^{n-i} \right) \left(\sum_{i=0}^{\infty} \gamma_i z^{-i} \right)
 \end{aligned}$$



Kronecker (1823-1891)

Example: $n = 2$:

$$\beta_0 z^2 + \beta_1 z + \beta_2 = (\alpha_0 z^2 + \alpha_1 z + \alpha_2)(\gamma_0 + \gamma_1 z^{-1} + \gamma_2 z^{-2} + \gamma_3 z^{-3} + \dots)$$

Equate likewise powers of z :

$$\begin{aligned} z^2 : \quad \beta_0 &= \alpha_0 \gamma_0 \\ z^1 : \quad \beta_1 &= \alpha_0 \gamma_1 + \alpha_1 \gamma_0 \\ z^0 : \quad \beta_2 &= \alpha_0 \gamma_2 + \alpha_1 \gamma_1 + \alpha_2 \gamma_0 \\ z^{-1} : \quad 0 &= \alpha_0 \gamma_3 + \alpha_1 \gamma_2 + \alpha_2 \gamma_1 \\ z^{-2} : \quad 0 &= \alpha_0 \gamma_4 + \alpha_1 \gamma_3 + \alpha_2 \gamma_2 \\ &\vdots \quad \quad \quad \vdots \\ z^{-k} : \quad 0 &= \alpha_0 \gamma_{k+2} + \alpha_1 \gamma_{k+1} + \alpha_2 \gamma_k \end{aligned}$$

Coefficients $\gamma_k, k \geq 2$ satisfy 3-term linear recurrence

$$\alpha_0 \gamma_{k+2} + \alpha_1 \gamma_{k+1} + \alpha_2 \gamma_k = 0, k \geq 2,$$

Kronecker and Hankel

$$\begin{pmatrix} \alpha_2 & \alpha_1 & \alpha_0 & 0 & 0 & 0 & \dots \\ 0 & \alpha_2 & \alpha_1 & \alpha_0 & 0 & \ddots & \ddots \\ 0 & 0 & \alpha_2 & \alpha_1 & \alpha_0 & 0 & \ddots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix} \begin{pmatrix} \gamma_2 & \gamma_3 & \gamma_4 & \gamma_5 & \dots \\ \gamma_3 & \gamma_4 & \gamma_5 & \ddots & \vdots \\ \gamma_4 & \gamma_5 & \gamma_6 & \ddots & \vdots \\ \gamma_5 & \gamma_6 & \gamma_7 & \ddots & \vdots \\ \ddots & \ddots & \ddots & \ddots & \vdots \end{pmatrix} = 0$$

Rational function series expansion \iff Hankel matrix rank deficient

- Banded Toeplitz \times rank deficient Hankel = 0
- Rank Hankel = degree of rational function
- Recurrence relation coefficients = denominator



Hankel (1839-1873)

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State $x_k \in \mathbb{R}^n$, output $y_k \in \mathbb{R}^l$:

$$\begin{aligned}
 x_{k+1} &= Ax_k & X(z) &= (zI_n - A)^{-1}x_0 \\
 & & &= (I_n + Az^{-1} + Az^{-2} + \dots)x_0 \\
 \\
 y_k &= Cx_k & Y(z) &= C(zI_n - A)^{-1}x_0 \\
 &= CA^k x_0 & &= Cx_0 + (CAx_0)z^{-1} + (CA^2x_0)z^{-2} + \dots \\
 & & &= y_0 + y_1 z^{-1} + y_2 z^{-2} + \dots
 \end{aligned}$$

Resolvent is rational:

$$(zI_n - A)^{-1} = \text{adj}(A) / \det(zI_n - A)$$

Kronecker/Hankel: Factorize (SVD) to go from data to state space model

$$\begin{pmatrix}
 y_1 & y_2 & y_3 & y_4 & \dots \\
 y_2 & y_3 & y_4 & y_5 & \vdots \\
 y_3 & y_4 & y_5 & y_6 & \vdots \\
 \dots & \dots & \dots & \dots & \vdots
 \end{pmatrix} = \begin{pmatrix}
 C \\
 CA \\
 CA^2 \\
 CA^3 \\
 \vdots
 \end{pmatrix} \begin{pmatrix}
 x_0 & Ax_0 & A^2x_0 & \dots
 \end{pmatrix}$$

With characteristic equation and Cayley-Hamilton

$$p(\lambda) = \det(\lambda I_n - A) = \lambda^n + \alpha_1 \lambda^{n-1} + \dots + \alpha_{n-1} \lambda + \alpha_n = 0$$

$$\begin{pmatrix} \alpha_n & \alpha_{n-1} & \alpha_{n-2} & \dots & \alpha_1 & \alpha_0 & 0 & 0 & \dots \\ 0 & \alpha_n & \alpha_{n-1} & \ddots & \alpha_2 & \alpha_1 & \alpha_0 & 0 & \ddots \\ 0 & 0 & \alpha_n & \ddots & \ddots & \alpha_2 & \alpha_1 & \alpha_0 & \ddots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix} \begin{pmatrix} C \\ CA \\ CA^2 \\ CA^3 \\ \vdots \end{pmatrix} = T \cdot \Gamma = 0$$

Left null space (Γ) = banded Toeplitz T .

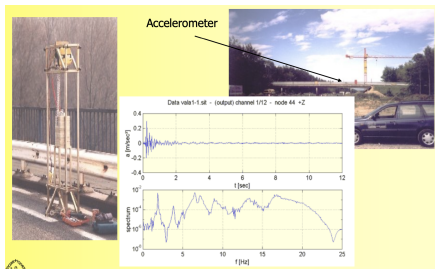
Right null space (T) = shift-invariant:

$$\underline{\Gamma} A = \bar{\Gamma} \iff \text{rank}(\underline{\Gamma} \bar{\Gamma}) = n = \text{rank}(\underline{\Gamma}) \text{ (PRC)} \implies A = \underline{\Gamma}^\dagger \bar{\Gamma}$$

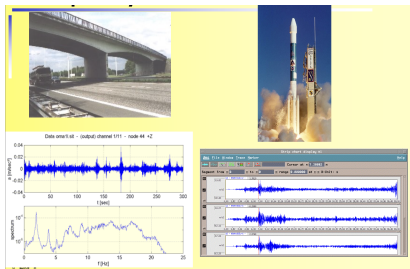
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Application: Impulse response and stochastic realization

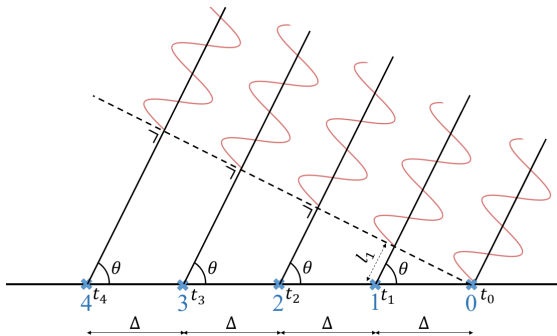


Impulse response



Stochastic realization

Application: Direction of Arrival: Uniform linear array, narrow band sources, far field

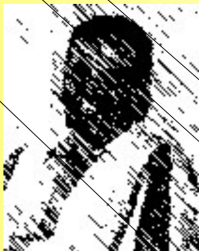


$$\begin{aligned}
 y_i(t) &= \sin\left(\omega t + \frac{\omega(i\Delta) \cos \theta}{c}\right) = \sin(\omega t + \varphi_i) \\
 &= \sin(\omega t) \cos \varphi_i + \cos(\omega t) \sin \varphi_i = \begin{pmatrix} \cos \varphi_i & \sin \varphi_i \end{pmatrix} \begin{pmatrix} \sin \omega t \\ \cos \omega t \end{pmatrix}
 \end{aligned}$$

Application: Shape from moments

Calculate moments of 'pdf' and show that

$$\int_{-T}^T p_f(t, \theta) t^k dt = \sum_{j=0}^k \binom{k}{j} \cos^{k-j}(\theta) \sin^j(\theta) \mu_{k-j, j}$$



$$\mu_{pq} \rightarrow \tau_k$$

$$\tau_k = \sum_{j=1}^n a_j z_j^k$$

→ Realization theory !

Application: Cepstrum realization

Power cepstrum = power spectrum of log of power spectrum


$$\log \Phi(z) = \sum_{k=-\infty}^{+\infty} c_k \cdot z^{-k}$$

Cepstral coefficients $c_k = c_{-k}, \forall k;$

$$c_0 = 2 \log \rho$$

$$k c_k = \sum \alpha_i^k - \sum \beta_i^k$$

**i-th cepstral coefficient
= sum of i-th powers
of poles and zeros**



spectrum	cepstrum
frequency	quefrequency
phase	saphe
magnitude	gamnitude
filtering	liftering
harmonic	rahmonic
period	repiod

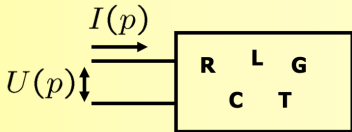
→ Realization theory !

Application: Electrical circuit power spectrum by R, L, C, T, G

A transfer function $Z(p)$ is **realizable** as a passive electrical circuit

\Leftrightarrow there exists an interconnection of a finite number of R's, L's, C's, T's and G's such that

$$Z(p) = \frac{U(p)}{I(p)}$$



$\Leftrightarrow Z(p)$ is positive real

$\Leftrightarrow p \in \mathbb{C}_+ \Rightarrow Z(p) \in \mathbb{C}_+$

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mD shift invariant systems ($m = 2$)

$x_{k,l} \in \mathbb{R}^n, y_{k,l} \in \mathbb{R}$:

$$\begin{aligned} x_{k+1,l} &= A_1 x_{k,l} \\ x_{k,l+1} &= A_2 x_{k,l} & A_1 A_2 &= A_2 A_1 \\ y_{k,l} &= C x_{k,l} \end{aligned}$$

$$Y = \begin{pmatrix} y_{00} & y_{10} & y_{01} & y_{20} & y_{11} & y_{02} & y_{30} & \dots \\ y_{10} & y_{20} & y_{11} & y_{30} & y_{21} & y_{12} & y_{40} & \dots \\ y_{01} & y_{11} & y_{02} & y_{21} & y_{12} & y_{03} & y_{31} & \dots \\ y_{20} & y_{30} & y_{21} & y_{40} & y_{31} & y_{22} & y_{50} & \dots \\ y_{11} & y_{21} & y_{12} & y_{31} & y_{22} & y_{13} & y_{41} & \dots \\ y_{02} & y_{12} & y_{13} & y_{22} & y_{13} & y_{04} & y_{32} & \dots \\ y_{30} & y_{40} & y_{31} & y_{50} & \dots & \dots & \dots & \dots \\ y_{21} & y_{31} & y_{22} & y_{41} & \dots & \dots & \dots & \dots \\ y_{12} & y_{22} & y_{13} & y_{32} & \dots & \dots & \dots & \dots \\ y_{03} & y_{13} & y_{04} & y_{23} & \dots & \dots & \dots & \dots \\ y_{40} & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} \quad \text{rank}(Y) = n$$

$$= \Gamma \Delta = \begin{pmatrix} C \\ \hline CA_1 \\ CA_2 \\ \hline CA_1^2 \\ CA_1 A_2 \\ CA_2^2 \\ \hline \vdots \end{pmatrix} \left(x_0 \mid A_1 x_0 \mid A_2 x_0 \mid A_1^2 x_0 \mid \dots \right)$$

The column space of Γ is a **multi-shift-invariant subspace**:

$$\underline{\Gamma} A_1 = S_1 \Gamma = \begin{pmatrix} C \\ \hline CA_1 \\ CA_2 \\ \hline CA_1^2 \\ CA_1 A_2 \\ CA_2^2 \\ \hline \vdots \\ \hline CA_1^{p-2} \\ CA_1^{p-3} A_2 \\ \hline \vdots \\ CA_2^{p-2} \end{pmatrix} A_1 = \begin{pmatrix} CA_1 \\ \hline CA_1^2 \\ CA_1 A_2 \\ \hline CA_1^3 \\ CA_1^2 A_2 \\ CA_1 A_2^2 \\ \hline \vdots \\ \hline CA_1^{p-1} \\ CA_1^{p-2} A_2 \\ \hline \vdots \\ CA_1 A_2^{p-2} \end{pmatrix} \quad \text{and} \quad \underline{\Gamma} A_2 = S_2 \Gamma$$

- Selector matrix S_1 selects the block rows (2, 4, 5, 7, 8, 9, ...).
- Selector matrix S_2 selects the block rows (3, 5, 6, 8, 9, 10, ...).
- Find A_1, A_2 by solving set of linear equations (PRC: $\text{rank}(\underline{\Gamma}) = n$)

$$A_1 = \underline{\Gamma}^\dagger S_1 \Gamma \quad \text{and} \quad A_2 = \underline{\Gamma}^\dagger S_2 \Gamma .$$

- A multi-shift invariant subspace is determined by the eigenvalues of its shifts A_1 and A_2

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- All mD generalizations of DOA, shape-from-moments, power spectra, etc.
- Bilinear system identification
- *Rooting multivariable polynomials*
- *Multi-parameter eigenvalue problems*
- *Global optimum of prediction-error-methods*
- ...

Application: Two polynomials in two variables

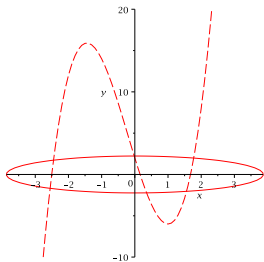
- Consider

$$\begin{cases} p(x, y) = x^2 + 3y^2 - 15 = 0 \\ q(x, y) = y - 3x^3 - 2x^2 + 13x - 2 = 0 \end{cases}$$

- Fix a monomial order, e.g., $1 < x < y < x^2 < xy < y^2 < x^3 < x^2y < \dots$

- Construct quasi-Toeplitz Macaulay matrix M :

$$\begin{matrix} p(x, y) \\ q(x, y) \\ x \cdot p(x, y) \\ y \cdot p(x, y) \end{matrix} \begin{bmatrix} 1 & x & y & x^2 & xy & y^2 & x^3 & x^2y & xy^2 & y^3 \\ -15 & & & 1 & & 3 & & & & \\ -2 & 13 & 1 & -2 & & & -3 & & & \\ -15 & & & & & & 1 & & 3 & \\ & -15 & & & & & & 1 & & 3 \end{bmatrix} \begin{pmatrix} 1 \\ x \\ y \\ x^2 \\ xy \\ \vdots \\ xy^2 \\ y^3 \end{pmatrix} = 0$$



$$\begin{cases} p(x, y) = x^2 + 3y^2 - 15 = 0 \\ q(x, y) = y - 3x^3 - 2x^2 + 13x - 2 = 0 \end{cases}$$

Continue to enlarge M :

it #	form	1	x	y	x ²	xy	y ²	x ³	x ² y	xy ²	y ³	x ⁴	x ³ y	x ² y ²	xy ³	y ⁴	x ⁵	x ⁴ y	y ³ x ²	x ³ xy ²	y ⁴	
d = 3	p	-15			1																	
	xp		-15					1														
	yp			-15						1												
	qp																					
d = 4	x ² p				-15							1										
	xy ² p					-15							1									
	xq		-2																			
	yp			-2																		
d = 5	x ³ p							-15														
	x ² yp								-15													
	xy ² p									-15												
	y ³ p										-15											
	x ² q					-2																
	xyq						-2															
y ² q																						

- # rows grows faster than # cols \Rightarrow overdetermined system
- If solution exists: rank deficient by construction!

nD realization in the null space

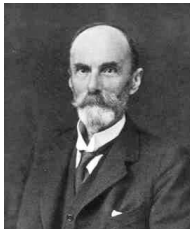
- Macaulay matrix M :

$$M = \begin{bmatrix} \times & \times & \times & \times & 0 & 0 & 0 \\ 0 & \times & \times & \times & \times & 0 & 0 \\ 0 & 0 & \times & \times & \times & \times & 0 \\ 0 & 0 & 0 & \times & \times & \times & \times \end{bmatrix}$$

- Solutions generate vectors in kernel of M :

$$MK = 0$$

- Number of solutions s follows from rank decisions 'mind-the-gap':



Francis Sowerby Macaulay

Vandermonde nullspace K
built from s solutions (x_i, y_i) :

1	1	...	1
x_1	x_2	...	x_s
y_1	y_2	...	y_s
x_1^2	x_2^2	...	x_s^2
$x_1 y_1$	$x_2 y_2$...	$x_s y_s$
y_1^2	y_2^2	...	y_s^2
x_1^3	x_2^3	...	x_s^3
$x_1^2 y_1$	$x_2^2 y_2$...	$x_s^2 y_s$
$x_1 y_1^2$	$x_2 y_2^2$...	$x_s y_s^2$
y_1^3	y_2^3	...	y_s^3
x_1^4	x_2^4	...	x_s^4
$x_1^3 y_1$	$x_2^3 y_2$...	$x_s^3 y_s$
$x_1^2 y_1^2$	$x_2^2 y_2^2$...	$x_s^2 y_s^2$
$x_1 y_1^3$	$x_2 y_2^3$...	$x_s y_s^3$
y_1^4	y_2^4	...	y_s^4
\vdots	\vdots	\vdots	\vdots

Setting up an eigenvalue problem in x

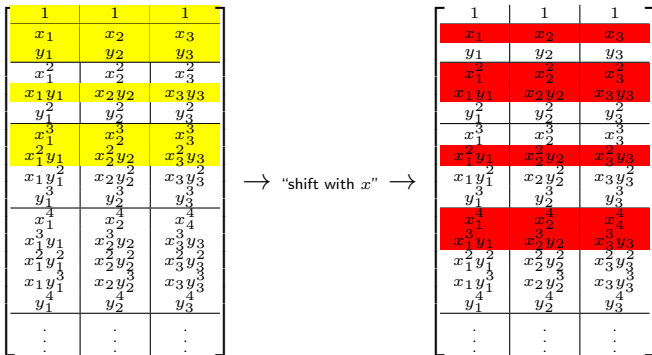
- Choose s linear independent rows in K

$$S_1 K$$

- This corresponds to finding linear dependent columns in M

1	1	...	1
x_1	x_2	...	x_s
y_1	y_2	...	y_s
x_1^2	x_2^2	...	x_s^2
$x_1 y_1$	$x_2 y_2$...	$x_s y_s$
y_1^2	y_2^2	...	y_s^2
x_1^3	x_2^3	...	x_s^3
$x_1^2 y_1$	$x_2^2 y_2$...	$x_s^2 y_s$
$x_1 y_1^2$	$x_2 y_2^2$...	$x_s y_s^2$
y_1^3	y_2^3	...	y_s^3
x_1^4	x_2^4	...	x_s^4
$x_1^3 y_1$	$x_2^3 y_2$...	$x_s^3 y_s$
$x_1^2 y_1^2$	$x_2^2 y_2^2$...	$x_s^2 y_s^2$
$x_1 y_1^3$	$x_2 y_2^3$...	$x_s y_s^3$
y_1^4	y_2^4	...	y_s^4
⋮	⋮	⋮	⋮
⋮	⋮	⋮	⋮

Shifting the selected rows gives (shown for 3 columns)



simplified:

1	1	1
x_1	x_2	x_3
y_1	y_2	y_3
$x_1 y_1$	$x_2 y_2$	$x_3 y_3$
x_1^3	x_2^3	x_3^3
$x_1^2 y_1$	$x_2^2 y_2$	$x_3^2 y_3$

$$\begin{bmatrix} x_1 & & \\ & x_2 & \\ & & x_3 \end{bmatrix} =$$

x_1	x_2	x_3
x_1^2	x_2^2	x_3^2
$x_1 y_1$	$x_2 y_2$	$x_3 y_3$
$x_1^2 y_1$	$x_2^2 y_2$	$x_3^2 y_3$
x_1^4	x_2^4	x_3^4
$x_1^3 y_1$	$x_2^3 y_2$	$x_3^3 y_3$

Finding the x -roots

Let $D_x = \text{diag}(x_1, x_2, \dots, x_s)$, then

$$S_1 KD_x = S_x K,$$

where S_1 and S_x select rows from K w.r.t. shift property We have

$$S_1 KD_x = S_x K$$

Generalized Vandermonde K is not known as such, instead a null space basis Z is calculated, which is a linear transformation of K :

$$ZV = K$$

which leads to

$$(S_x Z)V = (S_1 Z)VD_x$$

Here, V is the matrix with eigenvectors, D_x contains the roots x as eigenvalues.

Setting up an eigenvalue problem in y

It is possible to shift with y as well. . .

We find

$$S_1 K D_y = S_y K$$

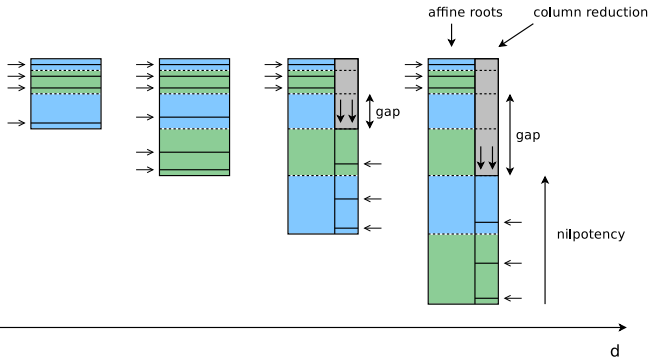
with D_y diagonal matrix of y -components of roots, leading to

$$(S_y Z) V = (S_1 Z) V D_y$$

Some interesting observations:

- same eigenvectors V !
- $(S_x Z)^{-1}(S_1 Z)$ and $(S_y Z)^{-1}(S_1 Z)$ commute
 \implies 'commutative algebra'

'Mind the Gap' with roots at infinity !



Rank, nullity and null space: SVD-ize the Macaulay matrix

Basic Algorithm outline

Find a basis for the nullspace of M using an SVD:

$$M = \begin{bmatrix} \times & \times & \times & \times & 0 & 0 & 0 \\ 0 & \times & \times & \times & \times & 0 & 0 \\ 0 & 0 & \times & \times & \times & \times & 0 \\ 0 & 0 & 0 & \times & \times & \times & \times \end{bmatrix} = [X \quad Y] \begin{bmatrix} \Sigma_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} W^T \\ Z^T \end{bmatrix}$$

Hence,

$$MZ = 0$$

Deflate roots at ∞ by detecting 'mind-the-gap' and column compression:

$$Z^T = \begin{pmatrix} Z_{11} & 0 \\ Z_{21} & Z_{22} \end{pmatrix}$$

We have

$$S_1 K D = S_{\text{shift}} K$$

with K generalized Vandermonde, not known as such. Instead a basis Z_{11} is computed as

$$Z_{11} V = K$$

which leads to

$$(S_{\text{shift}} Z_{11}) V = (S_1 Z_{11}) V D$$

S_1 selects linear independent rows.

S_{shift} selects rows 'hit' by the shift.

Application: Multiparameter Eigenvalue Problem (MEVP)

Given $A_0, \dots, A_m \in \mathbb{R}^{p \times q}$ with $p \geq q$, find $\lambda_i \in \mathbb{C}, i = 1, \dots, m$ and $x \neq 0 \in \mathbb{C}^q$ so that

$$(A_0 + A_1 \lambda_1 + \dots + A_m \lambda_m) x = 0$$

Special cases:

- Ordinary EVP: $A_0 \in \mathbb{R}^{n \times n}, A_1 = -I_n, A_i = 0, i \geq 2$
- 'Generalized' EVP: $A_0, A_1 \in \mathbb{R}^{n \times n}, A_i = 0, i \geq 2$

Basic idea to solve an MEVP (illustrated for $m = 2$)

$$(A_0 + A_1\lambda_1 + A_2\lambda_2) x = 0$$

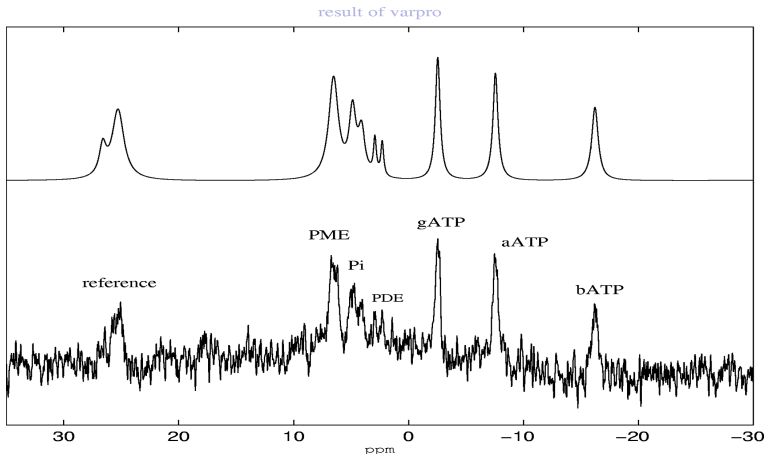
$$\begin{matrix} \times 1 \\ \times \lambda_1 \\ \times \lambda_2 \\ \times \lambda_1^2 \\ \vdots \end{matrix} \begin{pmatrix} A_0 & A_1 & A_2 & 0 & 0 & 0 & 0 & \dots \\ 0 & A_0 & 0 & A_1 & A_2 & 0 & 0 & \dots \\ 0 & 0 & A_0 & 0 & A_1 & A_2 & 0 & \dots \\ 0 & 0 & 0 & A_0 & 0 & 0 & A_1 & \dots \\ \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix} \begin{pmatrix} x \\ \frac{x\lambda_1}{x\lambda_2} \\ \frac{x\lambda_1^2}{x\lambda_1\lambda_2} \\ \frac{x\lambda_1\lambda_2}{x\lambda_2^2} \\ \frac{x\lambda_1^3}{x\lambda_1^3} \\ \vdots \end{pmatrix} = 0$$

Block 'quasi'-Toeplitz structure + 'generalized' Vandermonde structure

Outline

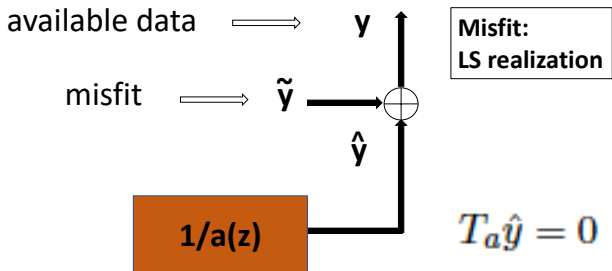
- 1 Spectra
- 2 Realization and spectra
- 3 Kronecker
- 4 1D LTI regular
- 5 Applications
- 6 mD SI models
- 7 Applications
- 8 Misfit**
- 9 Many more...
- 10 Valedictum

Errors using inadequate data are much less than those using no data at all.
Charles Babbage.



Data not model-compliant

Misfit case: Least squares realization (n_a)



$$\sigma^2 = \|\tilde{\mathbf{y}}\|_2^2$$

Misfit case: Least squares realization (ref: Kailath 80 !)

Data : $y \in \mathbb{R}^N$. **Model:** Data = model-compliant data + misfit:

$$y = \hat{y} + \tilde{y}$$

Model-compliance (Popper: models forbid more than allow) :

Image model:

$$\hat{y} = \Gamma \hat{x}_0$$

Kernel model

$$\begin{aligned} \hat{Y} a &= T_{N-n}^a \hat{y} \\ &= \begin{pmatrix} \alpha_n & \alpha_{n-1} & \dots & \alpha_0 & 0 & \dots & 0 \\ 0 & \alpha_n & \alpha_{n-1} & \ddots & \alpha_0 & \ddots & \ddots \\ \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \end{pmatrix} \begin{pmatrix} \hat{y}_0 \\ \hat{y}_1 \\ \vdots \\ \hat{y}_{N-1} \end{pmatrix} = 0 \end{aligned}$$

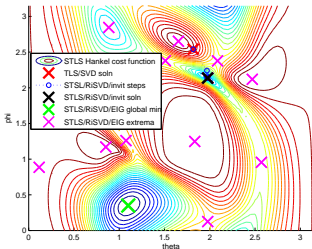
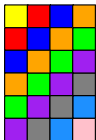
Least squares minimization:

$$\min \|\tilde{y}\|_2^2 \text{ subject to model - compliancy}$$

Multi-parameter EVP

$$\left(T_{N-n}^a y \quad T_{N-n}^a (T_{N-n}^a)^T (T_{N-2n}^a)^T \right) \begin{pmatrix} -1 \\ g \end{pmatrix} = 0$$

$$\begin{aligned} \min_v \quad & \tau^2 = v^T M^T D_v^{-1} M v \\ \text{s. t.} \quad & v^T v = 1. \end{aligned}$$



method	TLS/SVD	STLS inv. it.	STLS eig
v_1	.8003	.4922	.8372
v_2	-.5479	-.7757	.3053
v_3	.2434	.3948	.4535
τ^2	4.8438	3.0518	2.3822
global solution?	no	no	yes

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- Volterra equation of the 2nd kind, affine EVP
 $Ax = x\lambda + b$
- PdE separation of variables (diffusion and wave equations,...)
- All multivariate optimization problems are MEVPs
- All Prediction Error Methods solve a MEVP (heuristically)
- H_2 model reduction is a MEVP
- Real roots only = realization theory + nonnegative definiteness of Hankel
- etc... etc...

Outline

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What do we have in common ?



Lennart Ljung



Thomas Kailath



Kailath and Obama



European Research Council
Established by the European Commission

European
Innovation
Council



What do we have in common ?

- Topics: System identification, power spectrum estimation, DOA, subspace fitting, array signal processing
- Research grants and awards
- Lots of (old) common friends and colleagues: Boyd, Swindlehurst, Moonen, Viberg, van der Veen, Wahlberg, Slock, ...
- 35 years
- Wine





Kära Björn,

tack för vetenskapen och vänskapen!

(Dear Bjorn, thanks for the science and the friendship !)