

Sparse Interpolation

Annie Cuyt and Wen-shin Lee

Canazei Workshop 7–12/9/2014

Universiteit Antwerpen
Department of Mathematics
and Computer Science

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Motivation

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

interpolate

$$f(x) = \alpha_1 + \alpha_2 x^{100}$$

- ▶ Newton/Lagrange interpolation: 101 samples
- ▶ only 4 unknowns: $\alpha_1, \alpha_2, x^0, x^{100}$!
- ▶ how to solve it from 4 samples?

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

- ▶ exponential analysis
- ▶ generalized eigenvalue problems
- ▶ computer algebra
- ▶ orthogonal polynomials
- ▶ signal processing
- ▶ moment problems
- ▶ nonlinear approximation theory
- ▶ many applications . . .

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$x_s = s\Delta, \quad s = 0, 1, 2, \dots$$

$$\sum_{i=1}^n \alpha_i x_s^{k_i} = f_s, \quad n \ll \max(k_i), \quad k_i \in \mathbb{N}$$

$$\sum_{i=1}^{n_1} \alpha_{i,1} \cos(\phi_{i,1} x_s) + \sum_{i=1}^{n_2} \alpha_{i,2} \sin(\phi_{i,2} x_s) = f_s$$

$$\sum_{i=1}^n \alpha_i \exp(\phi_i x_s) = f_s, \quad \phi_i \in \mathbb{C}$$

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

1. Univariate exponential sparse interpolation
(Exercise)
2. Multivariate polynomial sparse interpolation
(Exercise)
3. Connection with rational approximation theory
(Exercise)
4. Applications unlimited

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Basics: exponential

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References



Figure: Gaspard Riche de Prony [1795]

interpolation problem:

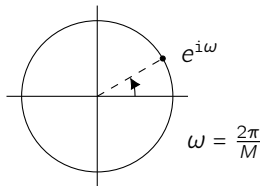
$$\sum_{i=1}^n \alpha_i \exp(\phi_i x_s) = f_s, \quad s = 0, \dots, 2n-1$$

$$x_s = s \frac{2\pi}{M}, \quad \omega = 2\pi/M$$

$$|\Im(\phi_i)| < M/2, \quad \Omega_i = \exp(\phi_i \omega),$$

$$f_s = \sum_{i=1}^n \alpha_i \Omega_i^s, \quad s = 0, \dots, 2n-1$$

$$\begin{cases} \alpha_1 + \dots + \alpha_n = f_0 \\ \alpha_1 \Omega_1 + \dots + \alpha_n \Omega_n = f_1 \\ \vdots \\ \alpha_1 \Omega_1^{2n-1} + \dots + \alpha_n \Omega_n^{2n-1} = f_{2n-1} \end{cases}$$



Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

finding Ω_i :

$$\prod_{i=1}^n (z - \Omega_i) = z^n + b_{n-1}z^{n-1} + \dots + b_1z + b_0$$

$$\begin{aligned} 0 &= \sum_{i=1}^n \alpha_i \Omega_i^s (\Omega_i^n + b_{n-1}\Omega_i^{n-1} + \dots + b_0) \\ &= \sum_{i=1}^n \alpha_i \Omega_i^{n+s} + \sum_{j=0}^{n-1} b_j \left(\sum_{i=1}^n \alpha_i \Omega_i^{j+s} \right) \\ &= f_{s+n} + \sum_{j=0}^{n-1} b_j f_{s+j} \end{aligned}$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$\begin{pmatrix} f_0 & \dots & f_{n-1} \\ \vdots & \ddots & \vdots \\ f_{n-1} & \dots & f_{2n-2} \end{pmatrix} \begin{pmatrix} b_0 \\ \vdots \\ b_{n-1} \end{pmatrix} = - \begin{pmatrix} f_n \\ \vdots \\ f_{2n-1} \end{pmatrix}$$

Hankel matrix:

$$H_n^{(r)} = \begin{pmatrix} f_r & \dots & f_{r+n-1} \\ \vdots & \ddots & \vdots \\ f_{r+n-1} & \dots & f_{r+2n-2} \end{pmatrix}$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

Hadamard polynomial:

$$H_n^{(0)}(z) = \begin{vmatrix} f_0 & \dots & f_{n-1} & f_n \\ \vdots & \ddots & \vdots & \vdots \\ f_{n-1} & \dots & f_{2n-2} & f_{2n-1} \\ 1 & \dots & z^{n-1} & z^n \end{vmatrix}$$

$$\begin{aligned} \prod_{i=1}^n (z - \Omega_i) &= \frac{H_n^{(0)}(z)}{|H_n^{(0)}|} \\ &= z^n + b_{n-1}z^{n-1} + \dots + b_1z + b_0 \end{aligned}$$

formally orthogonal polynomial:

$$\gamma : z^s \rightarrow f_s, \quad s = 0, 1, \dots$$

$$\gamma : \exp(\phi_i x_s) = \Omega_i^s \rightarrow \sum_{i=1}^n \alpha_i \Omega_i^s = f_s$$

$$\gamma : z^i \frac{H_n^{(0)}(z)}{|H_n^{(0)}|} \rightarrow 0, \quad i = 0, \dots, n-1$$

$$\frac{H_n^{(0)}(z)}{|H_n^{(0)}|} \perp_{\gamma} z^i, \quad i = 0, \dots, n-1$$

[Henrici, 1974]

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

roots of $\frac{H_n^{(0)}(z)}{|H_n^{(0)}|}$ from GEP:

$$H_n^{(0)} = \begin{pmatrix} 1 & \dots & & 1 \\ \Omega_1 & \Omega_2 & \dots & \Omega_n \\ \vdots & & & \vdots \\ \Omega_1^{n-1} & \dots & & \Omega_n^{n-1} \end{pmatrix} \begin{pmatrix} \alpha_1 & & & \\ & \ddots & & \\ & & \ddots & \\ & & & \alpha_n \end{pmatrix} \begin{pmatrix} 1 & \Omega_1 & \dots & \Omega_1^{n-1} \\ \vdots & \Omega_2 & \dots & \vdots \\ \vdots & \vdots & \dots & \vdots \\ 1 & \Omega_n & \dots & \Omega_n^{n-1} \end{pmatrix}$$

$$= V_n^T D_\alpha V_n$$

$$H_n^{(1)} = V_n^T D_\alpha \begin{pmatrix} \Omega_1 & & \\ & \ddots & \\ & & \Omega_n \end{pmatrix} V_n$$

Basics: exponential

$$\det\left(H_n^{(1)} - \lambda H_n^{(0)}\right) = \det\left(V_n^T D_\alpha \begin{pmatrix} \Omega_1 - \lambda & & \\ & \ddots & \\ & & \Omega_n - \lambda \end{pmatrix} V_n\right)$$
$$= 0 \text{ for } \lambda = \Omega_i, \quad i = 1, \dots, n$$

[Hua and Sarkar, 1990]

finding ϕ_j :

$$\exp(\phi_i) = \exp(\Re(\phi_i)) e^{i\Im(\phi_i)}$$

$$|\Im(\phi_i)| < \frac{M}{2} :$$

$$\begin{aligned} \arg(\Omega_i) &= \arg(\exp(\phi_i \omega)) \\ &= \Im(\phi_i) \frac{2\pi}{M} \in]-\pi, \pi[\end{aligned}$$

finding α_j :

$$\sum_{i=1}^n \alpha_i \Omega_i^{s+j} = f_{s+j}, \quad s = 0, \dots, n-1, \quad 0 \leq j \leq n$$

$$\begin{pmatrix} \Omega_1^j & \dots & \Omega_n^j \\ \vdots & & \vdots \\ \Omega_1^{j+n-1} & \dots & \Omega_n^{j+n-1} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} = \begin{pmatrix} f_j \\ \vdots \\ f_{j+n-1} \end{pmatrix}$$

remaining interpolation conditions are linearly dependent

finding n :

$$N < n : \left| H_N^{(r)} \right| \neq 0, \quad r = 0, 1, \dots$$

$$N = n : \left| H_N^{(r)} \right| \neq 0 \quad \text{if } \Omega_i \neq \Omega_j \text{ for } i \neq j \quad [\text{Kaltofen and Lee, 2003}]$$

$$N > n : \left| H_N^{(r)} \right| \equiv 0, \quad r = 0, 1, \dots$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$\phi(x) = \sum_{i=1}^4 \alpha_i \exp(\phi_i x)$$

$$\alpha_1 = 1$$

$$\alpha_2 = 2.4$$

$$\alpha_3 = -2.1$$

$$\alpha_4 = 0.2$$

$$\phi_1 = 0$$

$$\phi_2 = -5 + 19.97i$$

$$\phi_3 = 3 + 45i$$

$$\phi_4 = 5.3i$$

evaluate at $x_s = s \frac{2\pi}{100}$, $M = 100$, $|\Im(\phi_i)| < 50$

sequence f_0, \dots, f_7, \dots is linearly generated

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

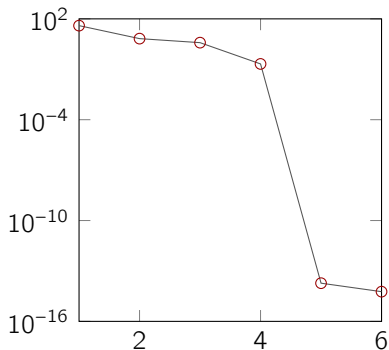


Figure: $H_N^{(0)}$ singular, $N = 6$

$$\frac{|\tilde{\Omega}_j - \Omega_j|}{|\Omega_j|} \leq 2 \times 10^{-12}, \quad \frac{|\tilde{\phi}_j - \phi_j|}{|\phi_j|} \leq 2 \times 10^{-12}$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$\phi(x) = \alpha_1 \exp(\phi_1 x) + \alpha_2 \exp(\phi_2 x)$$

4 unknowns $\phi_1, \phi_2, \alpha_1, \alpha_2$ identify $\phi(x)$ from

$$\phi(0) = \alpha_1 + \alpha_2$$

$$\phi'(0) = \alpha_1 \phi_1 + \alpha_2 \phi_2$$

$$\phi''(0) = \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2$$

$$\phi'''(0) = \alpha_1 \phi_1^3 + \alpha_2 \phi_2^3$$

solution:

$$\left| H_3^{(0)} \right| = \begin{vmatrix} \phi_0 & \phi_0' & \phi_0'' \\ \phi_0' & \phi_0'' & \phi_0''' \\ \phi_0'' & \phi_0''' & \phi_0^{IV} \end{vmatrix} = 0 \quad \text{symbolically}$$

$$n = 2$$

```

> #####
  ## Exercise: exponential ##
  #####
> restart:
> with(LinearAlgebra):
> Phi[0] := alpha[1] * exp(phi[1]*x) + alpha[2] * exp(phi[2]*x);

```

$$\Phi_0 := \alpha_1 e^{\phi_1 x} + \alpha_2 e^{\phi_2 x}$$

(1)

```

> for i from 1 to 4 do
  Phi[i] := diff(Phi[i-1], x);
od;

```

$$\Phi_1 := \alpha_1 \phi_1 e^{\phi_1 x} + \alpha_2 \phi_2 e^{\phi_2 x}$$

$$\Phi_2 := \alpha_1 \phi_1^2 e^{\phi_1 x} + \alpha_2 \phi_2^2 e^{\phi_2 x}$$

$$\Phi_3 := \alpha_1 \phi_1^3 e^{\phi_1 x} + \alpha_2 \phi_2^3 e^{\phi_2 x}$$

$$\Phi_4 := \alpha_1 \phi_1^4 e^{\phi_1 x} + \alpha_2 \phi_2^4 e^{\phi_2 x}$$

(2)

```

> for i from 0 to 4 do
  f[i] := simplify(subs(x=0, Phi[i]));
od;

```

$$f_0 := \alpha_1 + \alpha_2$$

$$f_1 := \alpha_1 \phi_1 + \alpha_2 \phi_2$$

$$f_2 := \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2$$

$$f_3 := \alpha_1 \phi_1^3 + \alpha_2 \phi_2^3$$

$$f_4 := \alpha_1 \phi_1^4 + \alpha_2 \phi_2^4$$

(3)

```

> H[3] := HankelMatrix([f[0], f[1], f[2], f[3], f[4]], 3);

```

..

$$H_3 := \begin{bmatrix} \alpha_1 + \alpha_2 & \alpha_1 \phi_1 + \alpha_2 \phi_2 & \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2 \\ \alpha_1 \phi_1 + \alpha_2 \phi_2 & \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2 & \alpha_1 \phi_1^3 + \alpha_2 \phi_2^3 \\ \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2 & \alpha_1 \phi_1^3 + \alpha_2 \phi_2^3 & \alpha_1 \phi_1^4 + \alpha_2 \phi_2^4 \end{bmatrix} \quad (4)$$

> Determinant(H[3]);

0

(5)

> H[2] := HankelMatrix([f[0],f[1],f[2]],2);

$$H_2 := \begin{bmatrix} \alpha_1 + \alpha_2 & \alpha_1 \phi_1 + \alpha_2 \phi_2 \\ \alpha_1 \phi_1 + \alpha_2 \phi_2 & \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2 \end{bmatrix}$$

(6)

> F2 := Matrix([[f[2]],[f[3]]]);

$$F2 := \begin{bmatrix} \alpha_1 \phi_1^2 + \alpha_2 \phi_2^2 \\ \alpha_1 \phi_1^3 + \alpha_2 \phi_2^3 \end{bmatrix}$$

(7)

> B:=LinearSolve(H[2],-F2);

$$B := \begin{bmatrix} \phi_1 \phi_2 \\ -\phi_1 - \phi_2 \end{bmatrix}$$

(8)

> B_poly := z^2 + B(2)*z + B(1);

$$B_poly := z^2 + (-\phi_1 - \phi_2)z + \phi_1 \phi_2$$

(9)

> z_root:=solve(B_poly=0, z);

$$z_root := \phi_2, \phi_1$$

(10)

> V := Transpose(VandermondeMatrix([z_root[1],z_root[2]],2,2));

$$V := \begin{bmatrix} 1 & 1 \\ \phi_2 & \phi_1 \end{bmatrix}$$

(11)

```
> F0 := Matrix([[f[0]],[f[1]]]);
```

$$F0 := \begin{bmatrix} \alpha_1 + \alpha_2 \\ \alpha_1 \phi_1 + \alpha_2 \phi_2 \end{bmatrix}$$

(12)

```
> A:=LinearSolve(V,F0);
```

$$A := \begin{bmatrix} \alpha_2 \\ \alpha_1 \end{bmatrix}$$

(13)

```
> alpha[1] := Pi;
alpha[2] := 5;
phi[1] := 2;
phi[2] := -1/7;
```

$$\begin{aligned} \alpha_1 &:= \pi \\ \alpha_2 &:= 5 \\ \phi_1 &:= 2 \\ \phi_2 &:= -\frac{1}{7} \end{aligned}$$

(14)

```
> for i from 0 to 4 do
  f[i] := simplify(subs(x=0, Phi[i]));
od;
```

$$\begin{aligned} f_0 &:= \pi + 5 \\ f_1 &:= 2\pi - \frac{5}{7} \\ f_2 &:= 4\pi + \frac{5}{49} \\ f_3 &:= 8\pi - \frac{5}{343} \\ f_4 &:= 16\pi + \frac{5}{2401} \end{aligned}$$

(15)


```
> H[3] := HankelMatrix([f[0],f[1],f[2],f[3],f[4]],3);
```

$$H_3 := \begin{bmatrix} \pi + 5 & 2\pi - \frac{5}{7} & 4\pi + \frac{5}{49} \\ 2\pi - \frac{5}{7} & 4\pi + \frac{5}{49} & 8\pi - \frac{5}{343} \\ 4\pi + \frac{5}{49} & 8\pi - \frac{5}{343} & 16\pi + \frac{5}{2401} \end{bmatrix}$$

(16)

```
> Determinant(H[3]);
```

0

(17)

```
> H[2] := HankelMatrix([f[0],f[1],f[2]],2);
```

$$H_2 := \begin{bmatrix} \pi + 5 & 2\pi - \frac{5}{7} \\ 2\pi - \frac{5}{7} & 4\pi + \frac{5}{49} \end{bmatrix}$$

(18)

```
> F2 := Matrix([[f[2]],[f[3]]]);
```

$$F2 := \begin{bmatrix} 4\pi + \frac{5}{49} \\ 8\pi - \frac{5}{343} \end{bmatrix}$$

(19)

```
> B:=LinearSolve(H[2],-F2);
```

$$B := \begin{bmatrix} -\frac{2}{7} \\ -\frac{13}{7} \end{bmatrix}$$

(20)

```
> B_poly := z^2 + B(2)*z + B(1);
```

$$B_poly := z^2 - \frac{13}{7}z - \frac{2}{7}$$

(21)

```
> z_root:=solve(B_poly=0,z);
```

$$z_root := 2, -\frac{1}{7} \quad (22)$$

```
> V := Transpose(VandermondeMatrix([z_root[1], z_root[2]], 2, 2));
```

$$V := \begin{bmatrix} 1 & 1 \\ 2 & -\frac{1}{7} \end{bmatrix} \quad (23)$$

```
> F0 := Matrix([[f[0]], [f[1]]]);
```

$$F0 := \begin{bmatrix} \pi + 5 \\ 2\pi - \frac{5}{7} \end{bmatrix} \quad (24)$$

```
> A:=LinearSolve(V,F0);
```

$$A := \begin{bmatrix} \pi \\ 5 \end{bmatrix} \quad (25)$$

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Basics: polynomial

Basics: polynomial



Figure: Michael Ben-Or and Prasoont Tiwari [1988]

interpolation problem:

$$\sum_{(k_1, \dots, k_d) \in K} \alpha_{k_1, \dots, k_d} x_1^{k_1} \cdots x_d^{k_d}, \quad \#K = n$$

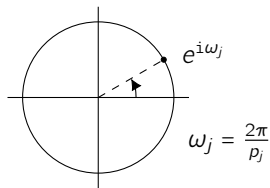
evaluate at

$$(x_1, \dots, x_d) = (\omega_1^s, \dots, \omega_d^s)$$

$$\omega_j = \exp(2\pi i/p_j), \quad p_j > \partial_j p(x_1, \dots, x_d), \quad p_j \text{ mutually prime}$$

$$p(\omega_1^s, \dots, \omega_d^s) = f_s, \quad 0 \leq s \leq 2n-1$$

$$\Omega_i = \omega_1^{k_1^{(i)}} \cdots \omega_d^{k_d^{(i)}}, \quad i = 1, \dots, n$$



[Giesbrecht, Labahn, and Lee, 2006]

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

finding Ω_i :

$$\prod_{i=1}^n (z - \Omega_i) = z^n + b_{n-1}z^{n-1} + \dots + b_1z + b_0$$

$$\begin{aligned} 0 &= \sum_{i=1}^n \alpha_i \Omega_i^s (\Omega_i^n + b_{n-1}\Omega_i^{n-1} + \dots + b_0) \\ &= \sum_{i=1}^n \alpha_i \Omega_i^{n+s} + \sum_{j=0}^{n-1} b_j \left(\sum_{i=1}^n \alpha_i \Omega_i^{j+s} \right) \\ &= f_{s+n} + \sum_{j=0}^{n-1} b_j f_{s+j} \end{aligned}$$

Basics: polynomial

$$\begin{pmatrix} f_0 & \cdots & f_{n-1} \\ \vdots & \ddots & \vdots \\ f_{n-1} & \cdots & f_{2n-2} \end{pmatrix} \begin{pmatrix} b_0 \\ \vdots \\ b_{n-1} \end{pmatrix} = - \begin{pmatrix} f_n \\ \vdots \\ f_{2n-1} \end{pmatrix}$$

Ω_i : zeros of formally orthogonal Hadamard polynomial

$$\prod_{i=1}^n (z - \Omega_i) = \frac{H_n^{(0)}(z)}{|H_n^{(0)}|}$$

finding $(k_1^{(i)}, \dots, k_d^{(i)})$:

reverse Chinese remainder theorem

$$m = p_1 \cdots p_d$$

$$\Omega_i = \omega^{k^{(i)}}, \quad \omega = \exp\left(\frac{2\pi i}{\prod_{j=1}^d p_j}\right),$$

$$k^{(i)} = k_1^{(i)} \frac{m}{p_1} + \cdots + k_d^{(i)} \frac{m}{p_d}$$

$$k_j^{(i)} \frac{m}{p_j} \bmod p_j = k^{(i)} \bmod p_j,$$

$$k_j^{(i)} < p_j, \quad \gcd(p_j, m/p_j) = 1, \quad j = 1, \dots, d$$

finding $\alpha_j = \alpha_{k_1^{(j)}, \dots, k_d^{(j)}}$:

$$\sum_{i=1}^n \alpha_i \Omega_i^{s+j} = f_{s+j}, \quad s = 0, \dots, n-1, \quad 0 \leq j \leq n$$

$$\begin{pmatrix} \Omega_1^j & \dots & \Omega_n^j \\ \vdots & & \vdots \\ \Omega_1^{j+n-1} & \dots & \Omega_n^{j+n-1} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} = \begin{pmatrix} f_j \\ \vdots \\ f_{j+n-1} \end{pmatrix}$$

remaining interpolation conditions are linearly dependent

finding n :

floating-point arithmetic:

$$\left| H_N^{(r)} \right| \equiv 0, \quad N > n, \quad r = 0, 1, \dots$$

exact arithmetic:

increase n till

$$\delta_s := f_s + b_{n-1}f_{s-1} + \dots + b_0f_{s-n}$$

equals 0, $s \geq 2n$

[Massey, 1969]

$$p(x, y) = x^5y + 2.2x^4y^4 - 0.5xy^{11} + 0.1xy^{12}$$

$$p_1 = 6, \quad p_2 = 13, \quad \omega_1 = \exp(2\pi i/6), \quad \omega_2 = \exp(2\pi i/13),$$

$$p(\omega_1^s, \omega_2^s), \quad s = 0, \dots, 7 \quad (\text{floating-point})$$

or

$$p(p_1^s, p_2^s), \quad s = 0, \dots, 7 \quad (\text{exact arithmetic})$$

sequence $f_0, f_1, \dots, f_7, \dots$ is linearly generated and $\delta_8 = 0$

Hadamard polynomial:

$$z^4 + (-3.67 + 0.0799i)z^3 + (5.35 - 0.216i)z^2 \\ (-3.67 + 0.216i)z + (0.997 - 0.0805i)$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$m = 78, \quad \omega = \exp(2\pi i/78),$$

$$\Omega_1 = \omega^{k(1)} = \omega^{71} = \omega^{5 \times 13 + 1 \times 6}$$

$$\Omega_2 = \omega^{k(2)} = \omega^{76} = \omega^{4 \times 13 + 4 \times 6}$$

$$\Omega_3 = \omega^{k(3)} = \omega^{79} = \omega^{1 \times 13 + 11 \times 6}$$

$$\Omega_4 = \omega^{k(4)} = \omega^{85} = \omega^{1 \times 13 + 12 \times 6}$$

$$K = \{(5, 1), (4, 4), (1, 11), (1, 12)\} \Rightarrow \text{terms } x^5y, x^4y^4, xy^{11}, xy^{12}$$

Vandermonde system

$$\sum_{i=1}^4 \alpha_i \Omega_i^{s+j} = f_{s+j}, \quad s = 0, \dots, n-1, \quad 0 \leq j \leq n$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$\begin{aligned} p(x, y) &= (x - 3)^5 (y + 5) + 2.2(x - 3)^4 (y + 5)^4 \\ &\quad - 0.5(x - 3)(y + 5)^{11} + 0.1(x - 3)(y + 5)^{12} \\ &= u^5 v + 2.2u^4 v^4 - 0.5uv^{11} + 0.1uv^{12}, \\ &\quad u = x - 3, v = y + 5 \end{aligned}$$

$$p_1 = 6,$$

$$\omega_1 = \exp(2\pi i/6),$$

$$u = \omega_1^s,$$

$$p_2 = 13,$$

$$\omega_2 = \exp(2\pi i/13),$$

$$v = \omega_2^s$$

$$p(\omega_1^s + 3, \omega_2^s - 5), \quad s = 0, \dots, 7$$

$$\begin{aligned}
 f_1 &:= 2.602841017834476 - 0.8741573457604577 \text{ I} \\
 f_2 &:= 2.064816858508122 - 1.589980617941464 \text{ I} \\
 f_3 &:= 1.329941999443174 - 2.035486797328322 \text{ I} \\
 f_4 &:= 0.5888159453592064 - 2.177111443869176 \text{ I} \\
 f_5 &:= 0.02061376974983473 - 2.067629883507751 \text{ I} \\
 f_6 &:= -0.2610215691867326 - 1.827541993129119 \text{ I} \\
 f_7 &:= -0.2414681965457515 - 1.605741666597298 \text{ I}
 \end{aligned}$$

(6)

> **HO[4] := HankelMatrix([f[0],f[1],f[2],f[3],f[4],f[5],f[6]],4);**

HO₄ := [[2.8 + 0. I, 2.602841017834476 - 0.8741573457604577 I, 2.064816858508122 - 1.589980617941464 I, 1.329941999443174 - 2.035486797328322 I],
 [2.602841017834476 - 0.8741573457604577 I, 2.064816858508122 - 1.589980617941464 I, 1.329941999443174 - 2.035486797328322 I, 0.5888159453592064 - 2.177111443869176 I],
 [2.064816858508122 - 1.589980617941464 I, 1.329941999443174 - 2.035486797328322 I, 0.5888159453592064 - 2.177111443869176 I, 0.02061376974983473 - 2.067629883507751 I],
 [1.329941999443174 - 2.035486797328322 I, 0.5888159453592064 - 2.177111443869176 I, 0.02061376974983473 - 2.067629883507751 I, -0.2610215691867326 - 1.827541993129119 I]]

(7)

> **H1[4] := HankelMatrix([f[1],f[2],f[3],f[4],f[5],f[6],f[7]],4);**

H1₄ := [[2.602841017834476 - 0.8741573457604577 I, 2.064816858508122 - 1.589980617941464 I, 1.329941999443174 - 2.035486797328322 I, 0.5888159453592064 - 2.177111443869176 I],
 [2.064816858508122 - 1.589980617941464 I, 1.329941999443174 - 2.035486797328322 I, 0.5888159453592064 - 2.177111443869176 I, 0.02061376974983473 - 2.067629883507751 I],
 [1.329941999443174 - 2.035486797328322 I, 0.5888159453592064 - 2.177111443869176 I, 0.02061376974983473 - 2.067629883507751 I, -0.2610215691867326 - 1.827541993129119 I],
 [0.5888159453592064 - 2.177111443869176 I, 0.02061376974983473 - 2.067629883507751 I, -0.2610215691867326 - 1.827541993129119 I, -0.2414681965457515 - 1.605741666597298 I]]

(8)

```

> eigenvalue_list := Eigenvalues(H1[4],H0[4], output=list);
eigenvalue_list := [0.8451900855580868 + 0.5344658261253400 I, 0.8451900855413997 - 0.5344658261265820 I, 0.9870502626364418
- 0.1604112808397337 I, 0.9967573080994497 + 0.08046656865700708 I]
(9)

> total_exp_list := [];
for i from 1 to 4 do
  total_exp_list := [op(total_exp_list), round(ExpLog(eigenvalue_list[i],6*13))];
od:
> total_exp_list;
[7, 71, 76, 1]
(10)

> exp_list:=[]:
for i from 1 to 4 do
  exp_list := [op(exp_list), [RevChRem(total_exp_list[i], 6, 6*13), RevChRem(total_exp_list[i], 13,
6*13)]];
od:
> exp_list;
[[1, 12], [5, 1], [4, 4], [1, 11]]
(11)

> V := Transpose(VandermondeMatrix(eigenvalue_list));
V := [[1. + 0. I, 1. + 0. I, 1. + 0. I, 1. + 0. I],
[0.8451900855580868 + 0.5344658261253400 I, 0.8451900855413997 - 0.5344658261265820 I, 0.9870502626364418
- 0.1604112808397337 I, 0.9967573080994497 + 0.08046656865700708 I],
[0.4286925614298439 + 0.9034504346214993 I, 0.4286925614003088 - 0.9034504346057614 I, 0.9485364419500248
- 0.3166679937654143 I, 0.9870502625782285 + 0.1604112807331159 I],
[-0.1205366802302719 + 0.9927088741336253 I, -0.1205366802550991 - 0.9927088740899947 I, 0.8854560256661491
- 0.4647231719910745 I, 0.9709418173518603 + 0.2393156640939936 I]]
(12)

> F := Matrix([[f[0]], [f[1]], [f[2]], [f[3]]]);
F :=
2.8 + 0. I
2.602841017834476 - 0.8741573457604577 I
2.064816858508122 - 1.589980617941464 I
1.329941999443174 - 2.035486797328322 I
(13)

> A := LinearSolve(V, F);

```


$$A := \begin{bmatrix} 0.1000000000041224 + 1.216000331768808 \cdot 10^{-11} I & \\ 1.000000000007150 + 2.565636441003167 \cdot 10^{-11} I & \\ 2.200000000261579 - 7.965585370092284 \cdot 10^{-11} I & \\ -0.5000000002728518 + 4.183975685423307 \cdot 10^{-11} I & \end{bmatrix} \quad (14)$$

```
> var_list := [u,v]:
```

```
p := 0:
```

```
for i from 1 to 4 do
  term := A[i,1]:
```

```
  for j from 1 to 2 do
    term := term*var_list[j]^exp_list[i,j]:
  od:
```

```
  p := p + term:
od:
```

```
> p;
```

$$(0.1000000000041224 + 1.216000331768808 \cdot 10^{-11} I) u v^{12} + (1.000000000007150 + 2.565636441003167 \cdot 10^{-11} I) u^5 v \\ + (2.200000000261579 - 7.965585370092284 \cdot 10^{-11} I) u^4 v^4 + (-0.5000000002728518 + 4.183975685423307 \cdot 10^{-11} I) u v^{11} \quad (15)$$

```
> poly_uv;
```

$$u^5 v + 2.2 u^4 v^4 - 0.5 u v^{11} + 0.1 u v^{12} \quad (16)$$

Motivation

Basics:
exponential

Basics:
polynomial

**Approximation
theory**

Applications

References

Approximation theory

Approximation theory

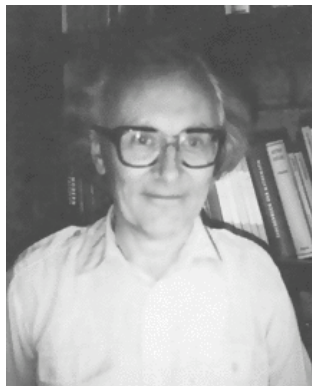


Figure: Henri Padé [1892] and Christian Pommerenke [1973]

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$f_s = \sum_{i=1}^n \alpha_i \exp(\phi_i x_s), \quad s = 0, 1, \dots, 2n-1$$

$$f(z) = \sum_{j=0}^{\infty} f_j z^j, \quad f_j = 0, \quad j < 0$$

$$p(z) = \sum_{i=0}^{\ell} a_i z^i,$$

$$q(z) = \sum_{i=0}^m b_i z^i$$

$$\left(\sum_{j=0}^{\infty} f_j z^j \right) q(z) - p(z) = \sum_{i \geq \ell+m+1} c_i z^i$$

Approximation theory

$$\begin{cases} f_0 b_0 = a_0 \\ f_1 b_0 + f_0 b_1 = a_1 \\ \vdots \\ f_\ell b_0 + \dots + f_{\ell-m} b_m = a_\ell \end{cases}$$

$$b_0 = 1$$

$$\begin{cases} f_{\ell+1} b_0 + \dots + f_{\ell-m+1} b_m = 0 \\ \vdots \\ f_{\ell+m} b_0 + \dots + f_\ell b_m = 0 \end{cases}$$

$$H_m^{(\ell+1-m)} \begin{pmatrix} b_m \\ \vdots \\ b_1 \end{pmatrix} = - \begin{pmatrix} f_{\ell+1} \\ \vdots \\ f_{\ell+m} \end{pmatrix}$$

$$[\ell/m](z) := p(z)/q(z)$$

$$\begin{aligned}f_s &= \sum_{i=1}^n \alpha_i \exp(\phi_i x_s) \\ &= \sum_{i=1}^n \alpha_i \Omega_i^s\end{aligned}$$

$$\begin{aligned}f(z) &= \sum_{j=0}^{\infty} f_j z^j \\ &= \sum_{i=1}^n \frac{\alpha_i}{1 - z\Omega_i} \\ &= \text{Laplace transform of } \sum_{i=1}^n \alpha_i \exp(\phi_i x)\end{aligned}$$

$$[n - 1/n](z) = p(z)/q(z)$$

$$\begin{aligned}q(z) &= \prod_{i=1}^n (1 - z\Omega_i) \\ &= z^n \frac{H_n^{(0)}(1/z)}{|H_n^{(0)}|} \\ &= b_0 z^n + b_1 z^{n-1} + \dots + b_{n-1} z + 1\end{aligned}$$

Motivation

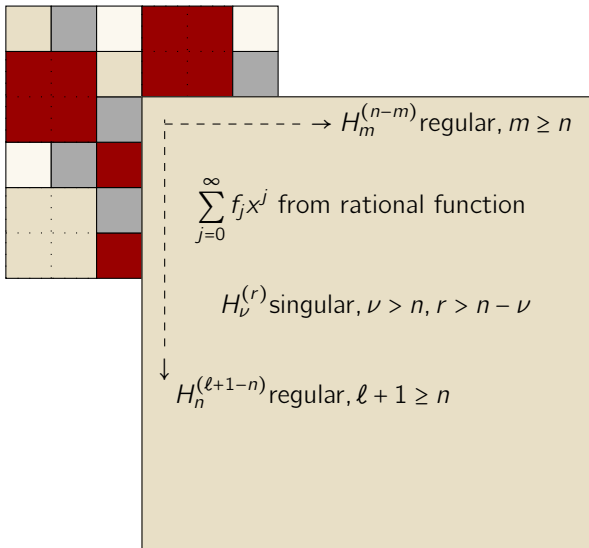
Basics:
exponential

Basics:
polynomial

**Approximation
theory**

Applications

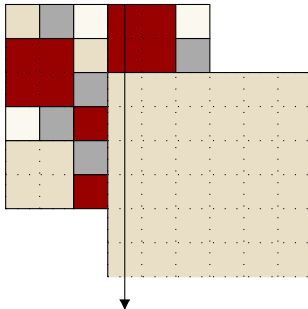
References



$f(z) + \varepsilon(z)$ meromorphic with poles in $0 \leq |z| < R$ of total multiplicity n [de Montessus de Ballore, 1905]



$[\nu/n](z) \rightarrow f(z) + \varepsilon(z)$ uniformly on compact sets excluding poles, with poles of $f(z) + \varepsilon(z)$ attracting poles of $[\nu/n](z)$ according to their multiplicity



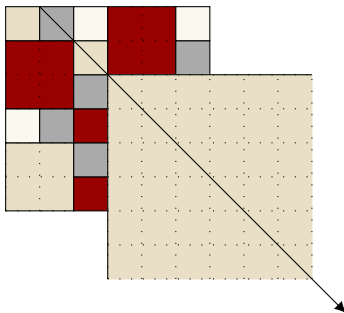
$f(z) + \varepsilon(z)$ analytic except for a countable number of poles

[Nuttall, 1970] and essential singularities [Pommerenke, 1973]



$[\nu - 1/\nu](z) \rightarrow f(z) + \varepsilon(z)$ in measure on compact sets, i.e.

$$\Lambda_2(\{z : |f(z) + \varepsilon(z) - [\nu - 1/\nu](z)| \geq \tau\}) \rightarrow 0$$



mathematical (noise free):

1. build $H_\nu^{(0)}$, $\nu = 0, 1, 2, \dots$
2. $H_\nu^{(0)} = U\Sigma V^T$ singular value decomposition
3. $\Sigma = \begin{pmatrix} \sigma_1 & & \\ & \ddots & \\ & & \sigma_\nu \end{pmatrix}$, $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n > \sigma_{n+1} = \dots = \sigma_\nu = 0$
4. find $\Omega_i, \phi_i, \alpha_i, i = 1, \dots, n$

numerical (with noise):

1. take ν large enough so that noise is clearly separated from n
2. solve $H_\nu^{(1)} v_i = \lambda_i H_\nu^{(0)} v_i, \quad i = 1, \dots, \nu, \quad \lambda_i = \Omega_i, \quad i = 1, \dots, n$
3. find ϕ_i
4. solve $\sum_{i=1}^n \alpha_i \exp(\phi_i x_j) = f_j, \quad 0 \leq j \leq 2\nu - 1$

Example: noise

$$\begin{aligned}\phi_1 &= 0, & \alpha_1 &= 1, \\ \phi_2 &= -0.2 + 39.5i, & \alpha_2 &= 2, \\ \phi_3 &= -0.5 + 40i, & \alpha_3 &= 4, \\ \phi_4 &= -1, & \alpha_4 &= 8,\end{aligned}$$
$$x_s = s \frac{2\pi}{100},$$
$$M = 100$$

$$\|\varepsilon(z)\|_\infty = 10^{-2}, \quad \text{uniform random noise}$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

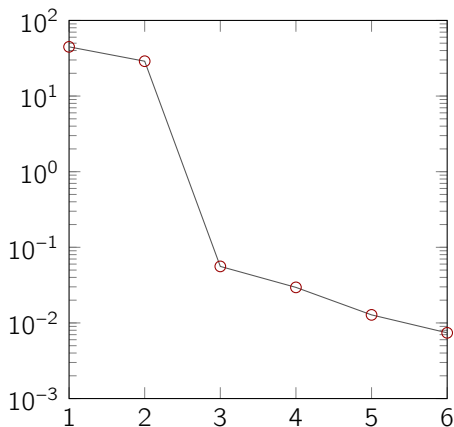


Figure: Singular values $H_\nu^{(0)}$ with $n = 4, \nu = 6$

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

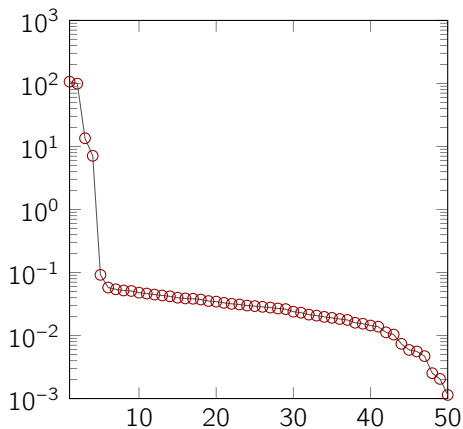


Figure: Singular values $H_{\nu}^{(0)}$ with $n = 4, \nu = 50$

Exercise: approximation

$$\phi_1 = 1.5i,$$

$$\alpha_1 = 10^{-3},$$

$$\phi_2 = 12.7i,$$

$$\alpha_2 = 2,$$

$$x_s = s \frac{2\pi}{100},$$

$$\phi_3 = -0.1 + 40i,$$

$$\alpha_3 = 4,$$

$$M = 100$$

$$\phi_4 = -0.3 + 25.2i,$$

$$\alpha_4 = 8,$$

$$\|\varepsilon(z)\|_\infty = 2 \times 10^{-3}, \quad \text{uniform random noise}$$


```

format long;

phi = [1.5*1i, 12.7*1i, -0.1+40*1i, -0.3+25.2*1i];
alpha = [10^(-3), 2, 4, 8];

eps = 2*10^(-3);

M = 100;

plot_signal

pause

plot_fft

pause

% synthesized input data with added noise
N = input('Enter the dimension for SVD: ');
% (10, 6, 3), (100, 100, 4)

randn('seed',0);

omega = 2*pi/M*(0:2*N-1);
f = syn_exp(alpha, phi, omega);

v = randn(size(f))+randn(size(f))*1i;
vv = v/norm(v,Inf);
f = f + eps*vv;

% form Hankel matrices H0 and H1 from y sequence
[H0,H1] = mat_ge(f);

plot_svd

pause

% reconstruct the parameters via generalized eigenvalues
n = input('Size of the model: ');

```

```
% compute the generalized eigenvalues and form the Vandermonde system
E = eig(H1(1:n,1:n),H0(1:n,1:n));
V = rot90(vander(E));

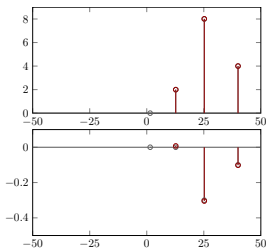
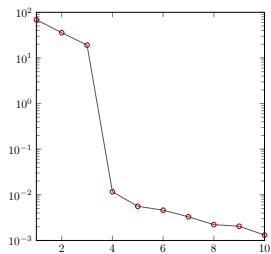
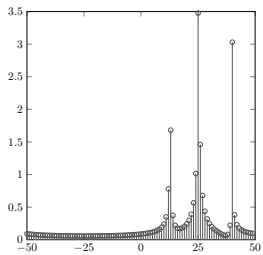
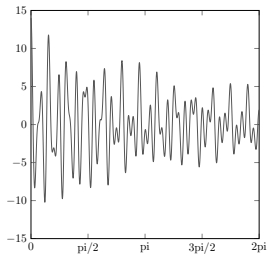
% amplitudes
A = V\f(1:n).';

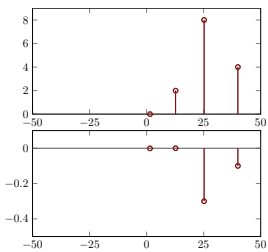
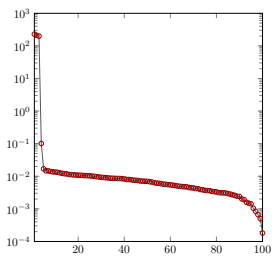
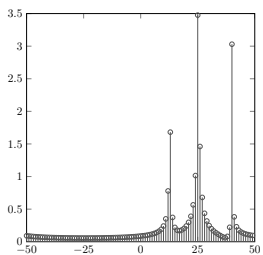
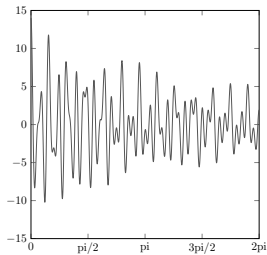
% frequencies and damping factors
alpha_rec = A;
phi_rec = log(E)*M/(2*pi);

pause

% extract the non-zero terms
extract

% plot computed parameters
plot_reconstructed_parameters
```





Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Applications

Exponential analysis in physical phenomena:

- ▶ power system transient detection
- ▶ motor fault diagnosis
- ▶ drug clearance / glucose tolerance
- ▶ magnetic resonance / infrared spectroscopy
- ▶ vibration analysis
- ▶ seismic data analysis
- ▶ music signal processing
- ▶ corrosion rate / crack initiation
- ▶ odour recognition with electronic nose
- ▶ typed keystroke recognition
- ▶ liquid (explosive) identification
- ▶ ...

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References



Figure: Transient detection and characterization

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

short lived high frequency signal:

- ▶ speech processing
- ▶ turbulent flow
- ▶ power lines
- ▶ ...

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

- ▶ model with $\phi_j = 120\pi i$,

$$\sum_{i=1}^n \alpha_i \cos(120\pi x + \gamma_i) \mathbf{1}_{[A_i, Z_i[}$$

- ▶ $n = 3, \alpha_i = 1, \gamma_{1,3} = -\pi/2, \gamma_2 = 3\pi/4$
- ▶ $[A_1, Z_1[= [0, 0.0308[$
 $[A_2, Z_2[= [0.0308, 0.0625[$
 $[A_3, Z_3[= [0.0625, 0.1058[$
- ▶ $M = 1200$
- ▶ uniformly distributed noise in $[-0.05, 0.05]$

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

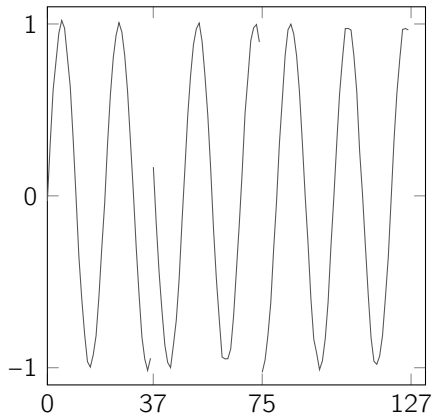


Figure: Given transient signal

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

- ▶ at each instance: 2 exponential terms
- ▶ characteristics of terms change
- ▶ inspect rank of

$$H_4^{(1)} = \begin{pmatrix} f_1 & f_2 & f_3 & f_4 \\ f_2 & f_3 & f_4 & f_5 \\ f_3 & f_4 & f_5 & f_6 \\ f_4 & f_5 & f_6 & f_7 \end{pmatrix}$$

$$[A_1, Z_1] = [0/M, 37/M[,$$

$$[A_2, Z_2] = [37/M, 75/M[,$$

$$[A_3, Z_3] = [75/M, 127/M[$$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

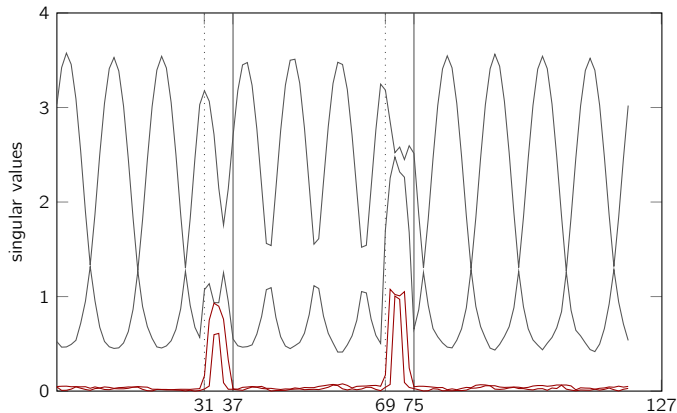


Figure: Numerical rank of $H_4^{(r)}$ evolving over time x

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References



Figure: Reconstructing undersampled audio signals

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

song containing 29 notes of 0.25 seconds each:

- ▶ $M = 44100$ (Hz)
- ▶ 11025 samples per note, 319725 in total
- ▶ $16.35 \leq \phi_i \leq 4978.03$, $i = 1, \dots, 100$
- ▶ complex exponential model

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

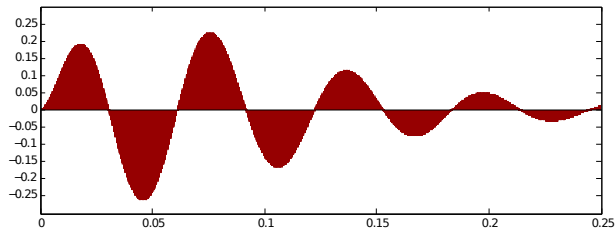


Figure: Sampled signal produced by 1 note

compressive sensing (optimisation, probabilistic)

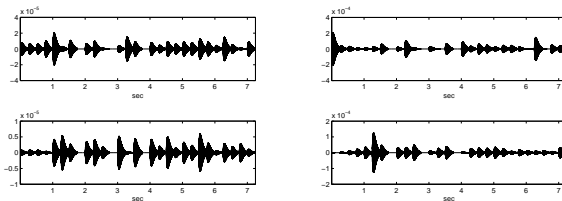


Figure: 4 runs with 1229 samples

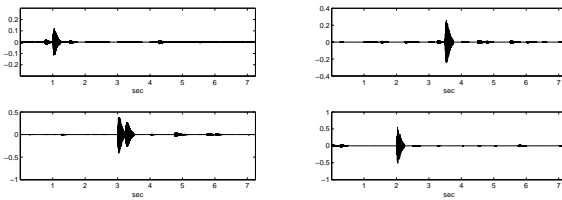


Figure: 4 runs with 456 samples

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

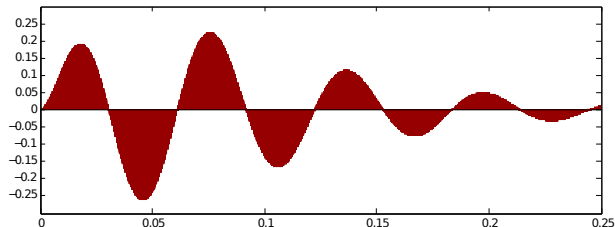


Figure: Full set of samples per note

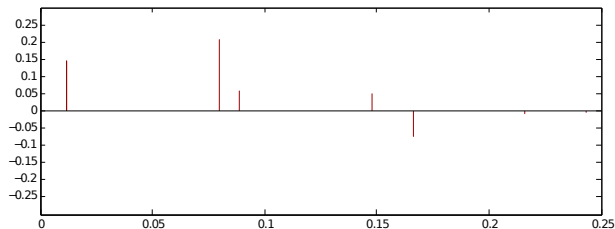


Figure: Sparse interpolation with 7 samples per note

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References



Figure: Preventive diagnosis of a broken rotor bar

3-phase induction motors:

- ▶ consume 40 – 50% of all electricity in industrialized countries
- ▶ rotor made up metal bars
- ▶ stator current signal analysed
- ▶ broken bar(s) characterized by sideband frequencies
- ▶ difficult to diagnose under low or no load



Figure: Stator and rotor

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

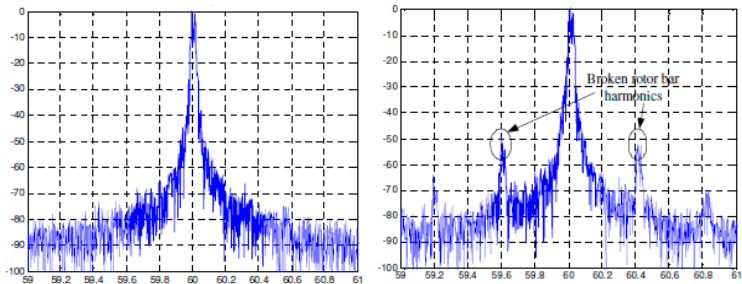


Figure: Stator current FFT spectra: healthy and with 1 broken bar

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

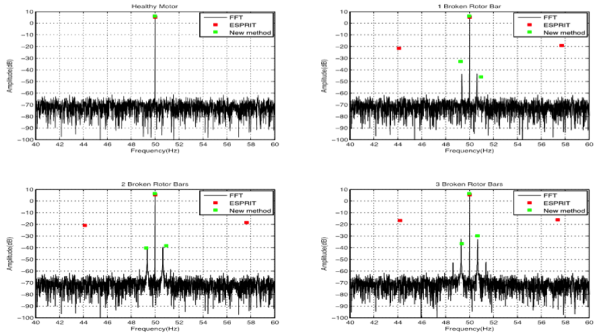


Figure: 10% load, 16dB noise, $\nu = 400$

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

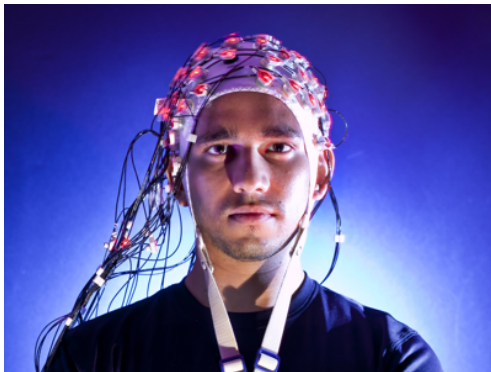


Figure: Sparse EEG approximation

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Bio-electrical signals:

- ▶ electrical activity of cells and tissues
- ▶ clinical studies of health status
- ▶ ECG, EEG, EMG, EOG, ...
- ▶ sparse model ($n = 8$) is approximate

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory**Applications**

References

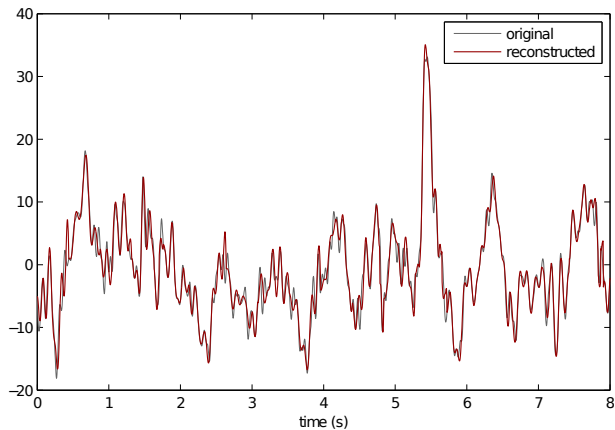


Figure: Reconstruction of 8 second [1 – 20] Hz bandpass filtered EEG

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References



Figure: Sparse EOG approximation

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Polysomnogram:

- ▶ 12 channels
- ▶ 22 wire attachments to patient
- ▶ heart rate, leg measurement, airflow (chest, abdomen), chin muscle, EEG, EOG, ...

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

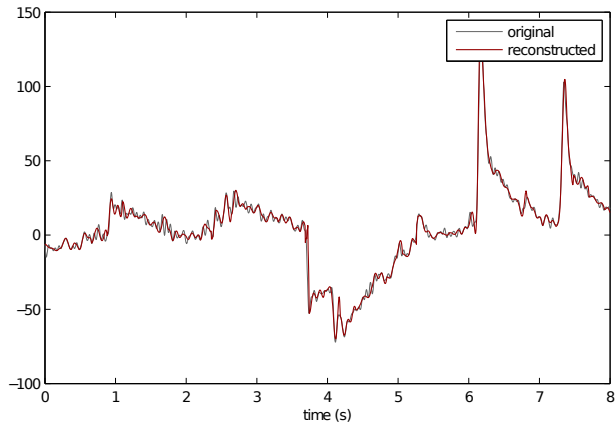


Figure: Reconstruction of 8 second ECG (CC = 99.2%)

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

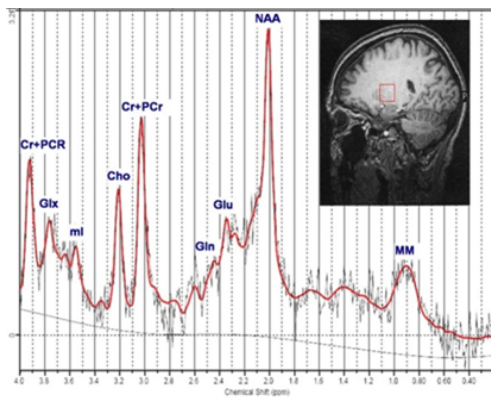


Figure: Spectral analysis of FID

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

Magnetic resonance spectroscopy:

- ▶ physical and chemical properties of molecules
- ▶ a.o. concentration of metabolites in the brain
- ▶ frequencies clustered → high frequency resolution
- ▶ free induction decay → time constraint
- ▶ Fourier methods need additional tools

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

$$\phi(x) = 5 \times 10^{-2} + 2e^{(-0.97+i79.94\pi)x} + 4e^{(-1+i80\pi)x} + 8e^{-1.1x} + \varepsilon(x)$$
$$\|\varepsilon(x)\|_{\infty} = 10^{-3}, \quad \text{circular Gaussian noise}$$

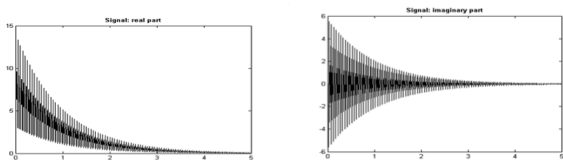


Figure: The real (left) and imaginary (right) part of $\phi(x)$

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory**Applications**

References

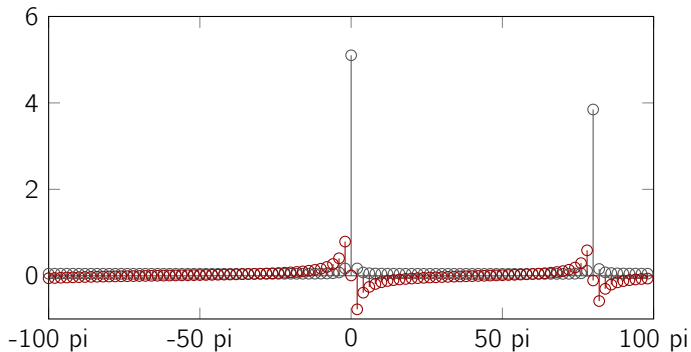


Figure: Real (black) and imaginary (red) parts of FFT

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

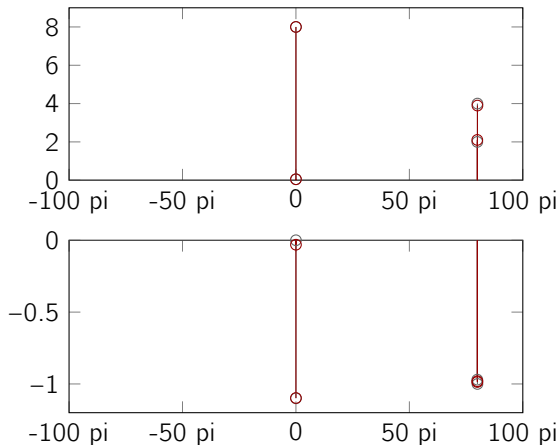


Figure: Amplitudes (top) and damping factors (bottom) of $\phi(x)$

Motivation

Basics:
exponential

Basics:
polynomial

Approximation
theory

Applications

References

References

Motivation

Basics:
exponentialBasics:
polynomialApproximation
theory

Applications

References

- M. Ben-Or and P. Tiwari. A deterministic algorithm for sparse multivariate polynomial interpolation. In *STOC '88: Proceedings of the twentieth annual ACM symposium on Theory of computing*, pages 301–309, New York, NY, USA, 1988. ACM.
<http://dx.doi.org/10.1145/62212.62241>.
- M. de Montessus de Ballore. Sur les fractions continues algébriques. *Rend. Circ. Mat. Palermo*, 19:185–257, 1905. <http://dx.doi.org/10.1007/BF03014011>.
- R. de Prony. Essai expérimental et analytique sur les lois de la dilatabilité des fluides élastiques et sur celles de la force expansive de la vapeur de l'eau et de la vapeur de l'alkool, à différentes températures. *J. Ec. Poly.*, 1:24–76, 1795.
- M. Giesbrecht, G. Labahn, and W. Lee. Symbolic-numeric sparse interpolation of multivariate polynomials. In *ISSAC'06*, 2006.
<http://dx.doi.org/10.1145/1145768.1145792>. 9–12 July.
- P. Henrici. *Applied and computational complex analysis I*. John Wiley & Sons, New York, 1974.
- Y. Hua and T. K. Sarkar. Matrix pencil method for estimating parameters of exponentially damped/undamped sinusoids in noise. *IEEE Trans. Acoust. Speech Signal Process.*, 38: 814–824, 1990. <http://dx.doi.org/10.1109/29.56027>.
- E. Kaltofen and W.-s. Lee. Early termination in sparse interpolation algorithms. *J. Symbolic Comput.*, 36(3-4):365–400, 2003. [http://dx.doi.org/10.1016/S0747-7171\(03\)00088-9](http://dx.doi.org/10.1016/S0747-7171(03)00088-9). International Symposium on Symbolic and Algebraic Computation (ISSAC'2002) (Lille).
- J. L. Massey. Shift-register synthesis and BCH decoding. *IEEE Trans. Inform. Theory*, 15 (1):122–127, 1969. <http://dx.doi.org/10.1109/TIT.1969.1054260>.
- J. Nuttall. The convergence of Padé approximants of meromorphic functions. *J. Math. Anal. Appl.*, 31:147–153, 1970. [http://dx.doi.org/10.1016/0022-247X\(70\)90126-5](http://dx.doi.org/10.1016/0022-247X(70)90126-5).
- H. Padé. *Sur la représentation approchée d'une fonction par des fractions rationnelles*. PhD thesis, Faculté des sciences de Paris, 1892.
- C. Pommerenke. Padé approximants and convergence in capacity. *J. Math. Anal. Appl.*, 41: 775–780, 1973. [http://dx.doi.org/10.1016/0022-247X\(73\)90248-5](http://dx.doi.org/10.1016/0022-247X(73)90248-5).