

The SVD of the linearized EIT problem on a disk

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The SVD

The **Singular Value Decomposition**(SVD)

- ▶ helps us to understand the **ill-conditioning** of a linear inverse imaging problem
- ▶ characterizes **consistent data**
- ▶ gives a **basis** for images in order of **how easy their components are to recover**

Some important inverse imaging problems an explicit analytic form of the SVD is known. For example, for the **Radon transform on a disk** the singular functions

- ▶ on the **image side** are **Zernike** disk functions
- ▶ on the **data side** they are **Fourier basis functions** in two angular coordinates [3, Ch7].

The quest for SVD of linear 2D EIT

- ▶ WL was advised in 1985 'The first thing to do in EIT is calculate the SVD' by Jennifer Scott at Oxford
- ▶ First calculated numerically in [2] and presented at *3rd EU Workshop on EIT Copenhagen 1990* as keynote talk 'What can you see with EIT'.
- ▶ No one has diagonalised linearised EIT analytically yet
- ▶ **New** The spectrum is **not discrete**
- ▶ **New** For the discrete part of the spectrum right singular functions look a bit like Zernike disk functions.
- ▶ **New** The non-discrete part can cause us problems with practical EIT reconstruction



Figure: The conference outing for the Copenhagen meeting was a trip on the schooner Halmø. Photo credit Brian Brown

SVD of Radon I

Let R be Radon transform integral along lines, R^* backprojection operator then

$$R^*R = (-\Delta)^{-1/2}$$

where $\Delta = \nabla^2$ the Laplacian. On the plane \mathbb{R}^2 we can diagonalize using the Fourier transform

$$g(x) = R^*R[f](x) \quad \hat{g}(\omega) = |\omega|^{-1}\hat{f}(\omega)$$

we say this has a *continuous spectrum* as any non-negative real number is an eigenvalue. On the unit disk $D = \{x : ||x|| \leq 1\}$

$$R^*RZ_{n,k} = \frac{4\pi}{n+1}Z_{n,k}$$

so the reciprocal integers for a discrete spectrum and the Zernike disk functions (polynomial in r and Fourier in θ) are the eigenfunctions.

SVD of Radon II

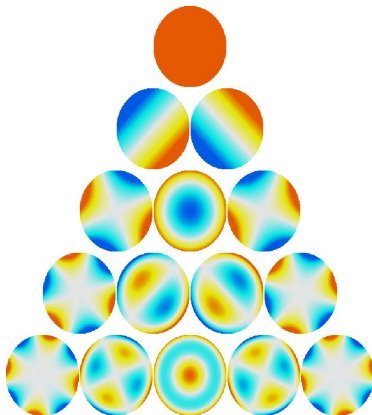


Figure: Zernike disk functions [3]. $Z_{n,r}(r, \theta) = p_{n,k}(r) \begin{cases} \sin \\ \cos \end{cases} k\theta$ where $p_{n,k}$ is a polynomial of order n . They are widely used in optics to describe lenses where they have cutesy names like 'astigmatism and defocus'.

The Hilbert transform I



The Hilbert transform on the real line \mathbb{R}

$$H[f](y) = \text{P.V.} \int_{-\infty}^{\infty} \frac{1}{y-x} f(x) dx = \frac{1}{x} * f(x),$$

$$\widehat{H[f]}(\omega) = -i \operatorname{sgn}(\omega) \hat{f}(\omega)$$

Restricted Hilbert transform

$$H[f](y) = \text{P.V.} \int_a^b \frac{1}{y-x} f(x) dx, \quad y \in [c, d]$$

- ▶ If () $a = c, b = d$
→ spectrum is discrete.
- ▶ If () $c = b$ (intervals abut)
→ we have a continuous spectrum [1].

Linearized EIT on a disk with trig patterns I

On a unit disk

- ▶ Potential for uniform conductivity $\phi_k^{\text{trig}} = r^k \text{trig} k\theta$ where trig is sin or cos.
- ▶ Fréchet derivative of forward problem at unit conductivity with perturbation in conductivity $\eta(r, \theta)$

$$K_{km}^{\text{cos}}[\eta] = \int_D \eta(r, \theta) \nabla \phi_k^{\text{cos}} \cdot \nabla \phi_m^{\text{cos}} r dr d\theta$$

- ▶ to get the normal operator $N = K^*K$ we sum over k, m this can be explicitly summed to give a kernel function for N as a function of r, θ, r', θ'
- ▶ for the case $\eta(r)$ independent of theta this reduces to

$$\frac{4 (rr')^2 ((rr')^2 + 1)}{(1 - (rr')^2)^3}$$

Linearized EIT on a disk with trig patterns II

- ▶ With some change of variables this can be written in terms of a restricted Hilbert transform from $[0, 1]$ to $[1, \infty)$ so we expect spectrum of N to have a **continuous part**.
- ▶ The continuous part of the spectrum means that the numerical SVD will **depend on the discretization** of the disk.

Numerical results

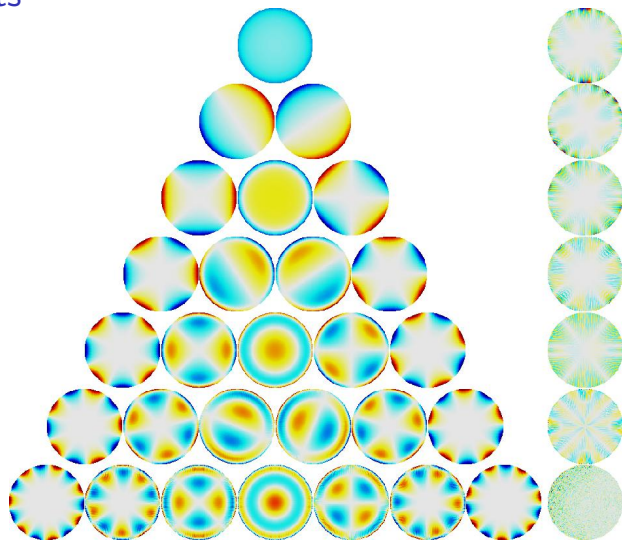


Figure: Singular functions of EIT Jacobian of the unit disk low-order sinusoidal patterns *Left:* Zernike-like *Right:* mesh-dependent

Numerical results + Zernike

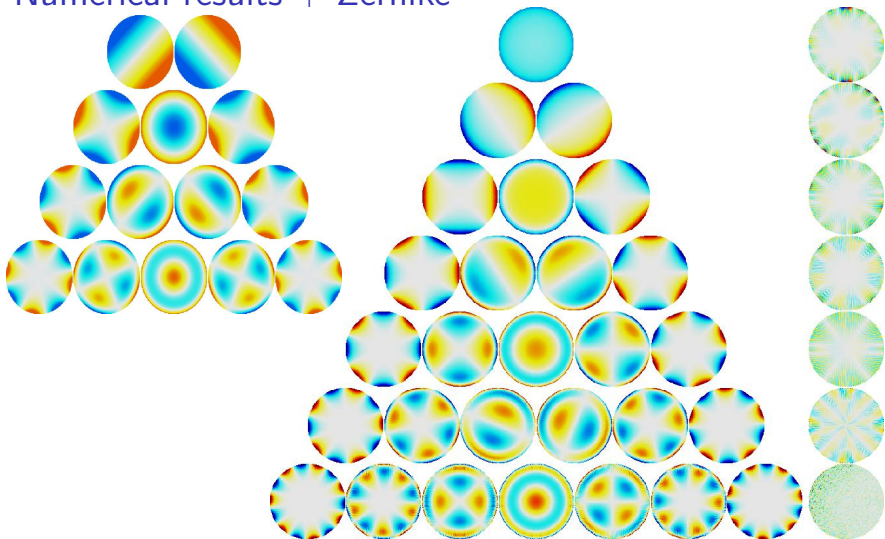


Figure: Singular functions of EIT Jacobian of the unit disk low-order sinusoidal patterns *Left:* Zernike-like *Right:* mesh-dependent

Conclusions I

- ▶ There are right singular functions of linearized EIT on the disk that are like Zernike functions
- ▶ There are also some mesh dependent singular functions concentrated near the boundary that we think reflect a non discrete part of the spectrum
- ▶ With a regularized linear solution your images are likely to be mesh dependent
- ▶ More details in forthcoming preprint...

References I

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