

# BIG MATHS

## A PLAY IN THREE ACTS

---

Julien Fageot  
Virginie Uhlmann

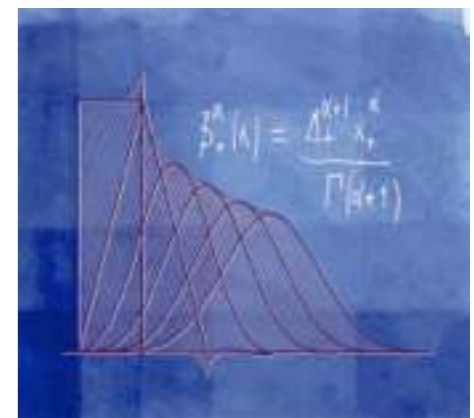


Twenty Years of Biomedical Imaging and Splines (TYBIS)

(and Virginie's Birthday)



March 23, 2018



# PREAMBLE

- BIG simply loves maths.
  - 20 years in 20 minutes: 1/500 000
- The BIG question: How to model signals?
  - We do *applied* and even *applicable* mathematics.
- “We are a continuous lab.” - M. Unser



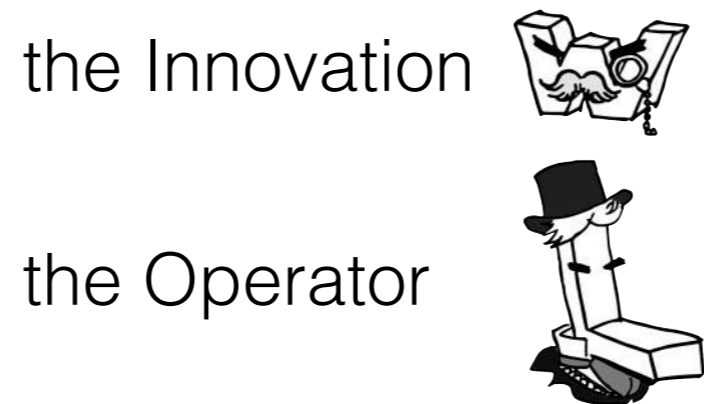
A Play in  
Three Acts

Michael Unser's

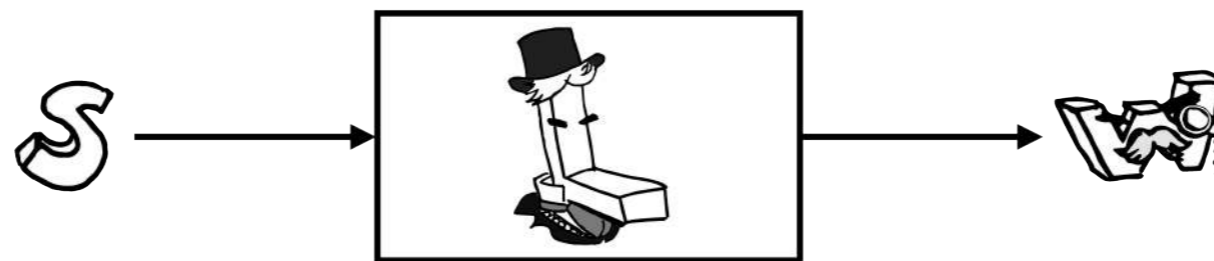
**BIG MATHS**

# THE GAME OF THE INNOVATION AND THE OPERATOR

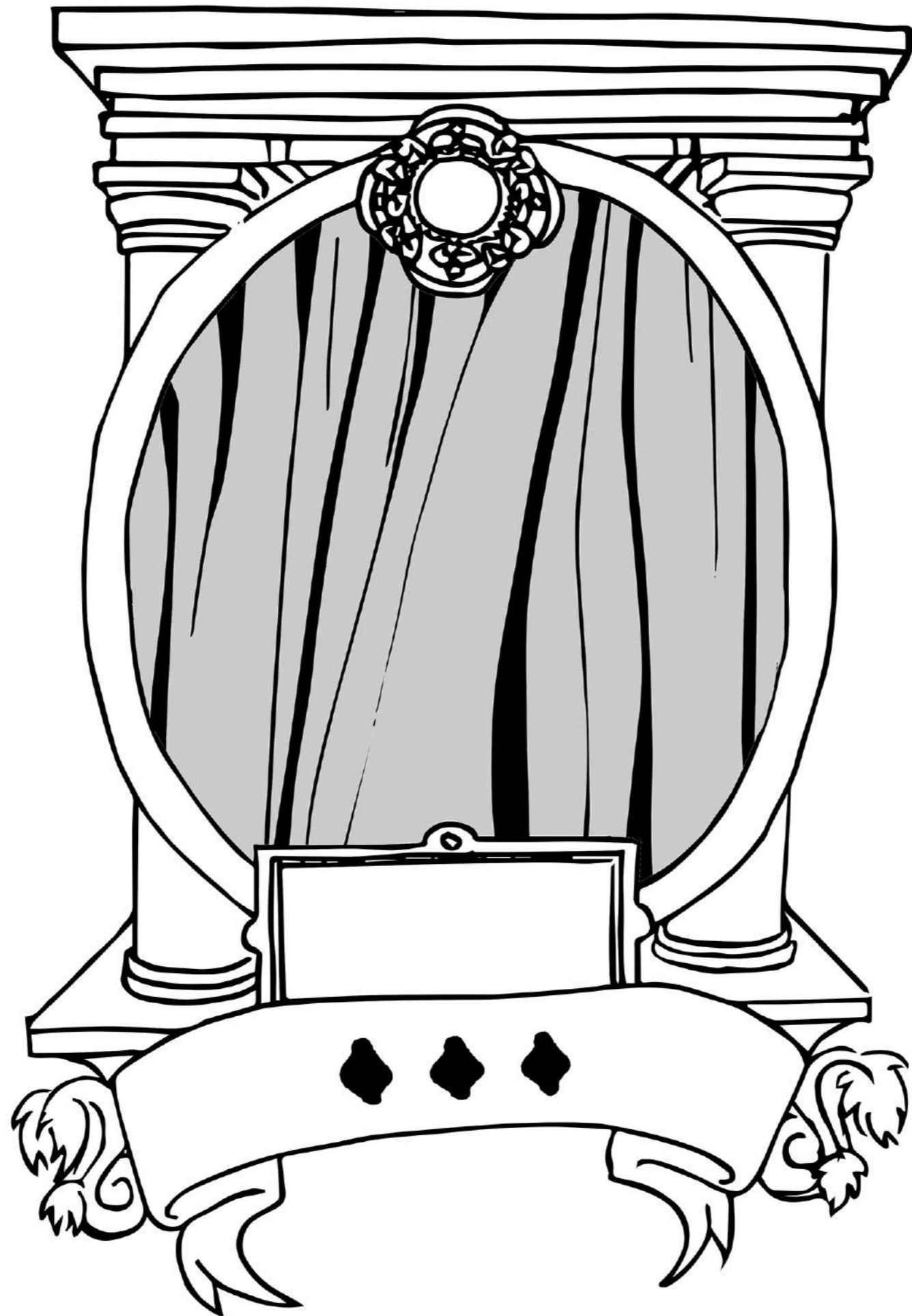
The characters:

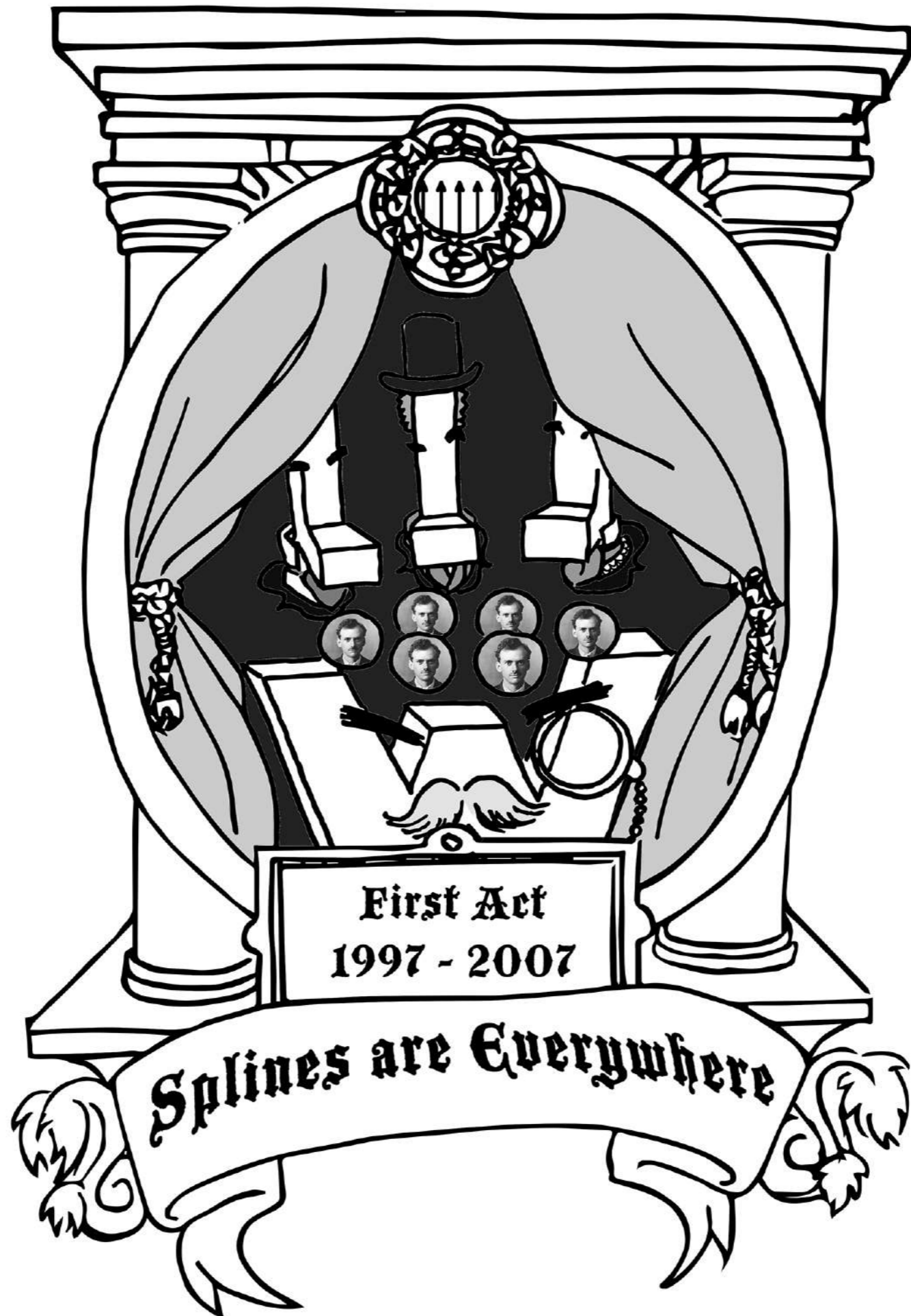


The story of how they combine to model signals



$$Ls = w$$





First Act

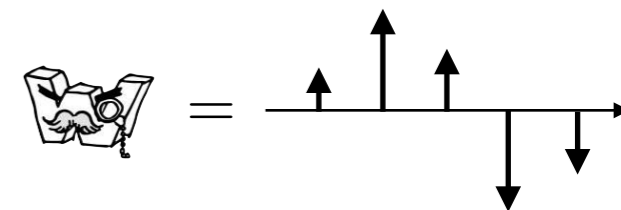
1997 - 2007

Splines are Everywhere

# FROM SCHOENBERG TO UNSER

A function  $s$  is a uniform  $L$ -spline if

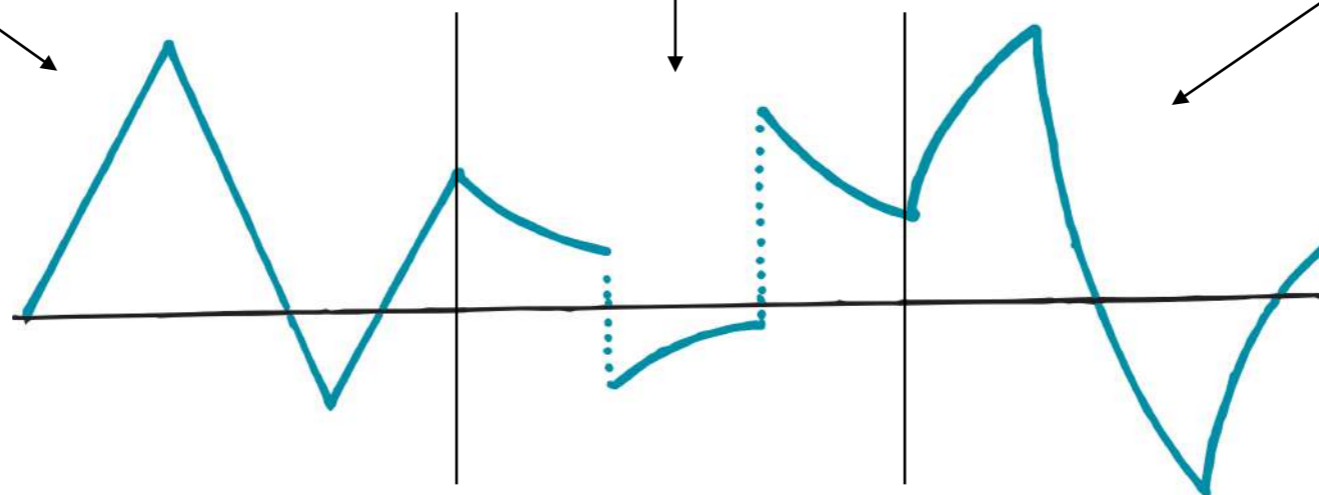
$$Ls = \sum_{k \in \mathbb{Z}} a_k \delta(\cdot - k)$$



$L = D^N$   
piecewise polynomial  
[S'64,U'99,BTU'01]

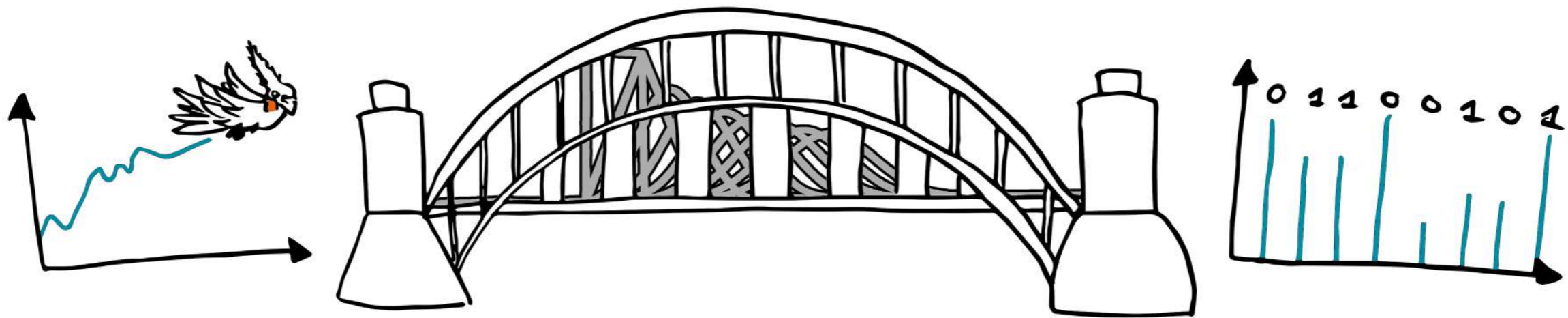
$L = D - \alpha I$   
Exponential splines  
[UB'05,U'05]

$L = D^\gamma$   
Fractional splines  
[BU'00,BU'07,UB'07]



$L$  general  
"spline-admissible"  
[UB'05]

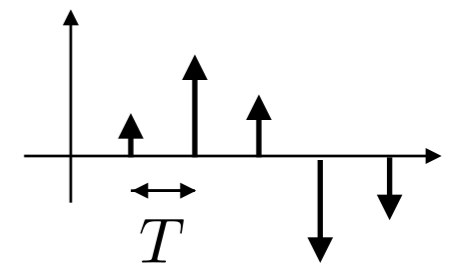
# “AND HERE, YOU SEE! SPLINES!” - AN ENTHUSIASTIC RESEARCHER



- Localization: L-splines =  $\left\{ s(t) = \sum_{k \in \mathbb{Z}} b_k \beta_L(t - k) \right\}$

- Reproduction properties: polynomials, exponentials

- Approximation power:  $\|s - \mathcal{P}_T^{\text{spline}}\{s\}\|_{L_2}$  when  $T \rightarrow 0$



- Minimum curvature:  $\min \|D^2 s\|_{L_2}$  such that  $s(k) = y_k$

- Statistical optimality: MMSE estimator of sampled Gaussian processes

- Minimal support properties: “The shorter, the better.” - a spline aficionado



# THE BLU-UNSER COLLABORATION

A textbook case: How well can you approximate with splines?

- First order question: The order of approximation?

$$\|s - \mathcal{P}_T^{\text{spline order } N}\{s\}\|_{L_2} \leq CT^N \|s^{(N)}\|_{L_2} \quad [\text{BU}'99\text{a, BU}'99\text{b}]$$

- Second order question: What is the constant?

$$C = \frac{1}{(2\pi)^N} \sqrt{2 \sum_{k=1}^{\infty} \frac{1}{k^{2N}}} \quad [\text{BU}'99\text{b, UB}'05]$$

- Comparison: splines vs. Daubechies wavelets

$$\pi = \lim_{N \rightarrow \infty} \frac{T_{\text{spline order } N}}{T_{\text{Daubechies order } N}} \quad [\text{BU}'99\text{c}]$$

# ARE YOU SPLINE-ADMISSIBLE?

- An operator is spline-admissible if you can do splines with it.
- More precisely?

*Definition 1:* L is a spline-admissible operator of order  $r > 1/2$  if and only if the following conditions are met.  
 1) L is a linear, shift-invariant operator with a frequency response  $\hat{L}(\omega)$  such that

**DEFINITION 6.2 (Spline admissibility)** The Fourier-multiplier operator  $L: \mathcal{X}(\mathbb{R}^d) \rightarrow \mathcal{S}'(\mathbb{R}^d)$  with frequency response  $\hat{L}(\omega)$  is called *spline admissible* if

$$\rho_L(r) = \mathcal{F}^{-1} \left\{ \frac{1}{\hat{L}(\omega)} \right\} (r) < +\infty$$

*Definition 2:* L is spline-admissible of order  $r > 1/2$  if and only if we have

- 1) L is a linear, shift-invariant operator of order  $r > 1/2$ .
- 2) L has a well-defined inverse  $L^{-1}$  and its frequency response  $\rho(t)$  is a function of slow growth (i.e.,  $\rho \in \mathcal{S}'$ ). Thus, L admits  $\rho(t)$  as Green's function:  $L\{\rho(t)\} = \delta(t)$ .

A linear and continuous operator L from  $\mathcal{S}(\mathbb{R}^d)$  to  $\mathcal{S}'(\mathbb{R}^d)$  is *spline-admissible* if

- it is shift-invariant, meaning that
 
$$L\{\varphi(\cdot - x_0)\} = L\{\varphi\}(\cdot - x_0)$$
 for every  $\varphi \in \mathcal{S}(\mathbb{R}^d)$  and  $x_0 \in \mathbb{R}^d$ ; and
- there exists a measurable function of slow growth (bounded by a polynomial)  $\rho_L$  such that
 
$$L\{\rho_L\} = \delta$$
 with  $\delta$  the Dirac delta function. The function  $\rho_L$  is a *Green's function* of L. [FUU'17]

- 3) **Definition 1 (Spline-admissible operator).** A linear operator  $L: \mathcal{M}_L(\mathbb{R}^d) \rightarrow \mathcal{S}'(\mathbb{R}^d)$  is called spline-admissible if
  1. it is shift-invariant;
  2. there exists a function  $\rho_L: \mathbb{R}^d \rightarrow \mathbb{R}$  of slow growth such that  $L\{\rho_L\} = \delta$ , where  $\delta$  is the Dirac impulse.
- 4)  $n_0 = \inf\{n \in \mathbb{N} : \rho_L \in L_{\infty, n}(\mathbb{R}^d)\}$ .
- 5) the (growth-restricted) null space of L,
 
$$\mathcal{N}_L = \{q \in L_{\infty, n_0}(\mathbb{R}^d) : L\{q\} = 0\},$$
 has the finite dimension  $N_0 \geq 0$ .  
 The native space of L,  $\mathcal{M}_L(\mathbb{R}^d)$ , is then specified as
 
$$\mathcal{M}_L(\mathbb{R}^d) = \{f \in L_{\infty, n_0}(\mathbb{R}^d) : \|Lf\|_{\mathcal{M}} < \infty\}.$$

functions  $\{\beta(t - k)\}_{k \in \mathbb{Z}}$  form an  $L_p$ -stable Riesz basis. Specifically, the following two conditions must be satisfied for all  $1 \leq p \leq \infty$ :

$$\inf_{\|c\|_{\ell_p} = 1} \left\| \sum_{k \in \mathbb{Z}} c[k] \beta(t - k) \right\|_{L_p} > 0$$

$$\sup_{\|c\|_{\ell_p} = 1} \left\| \sum_{k \in \mathbb{Z}} c[k] \beta(t - k) \right\|_{L_p} < \infty.$$

[UFW'17]

[UB'07]

# ARE YOU SPLINE-ADMISSIBLE?

- An operator is spline-admissible if you can do splines with it.
- More precisely?
- The answer:  
A. Amini, M. Unser, *A universal formula for generalized cardinal B-splines*, Applied and Computational Harmonic Analysis, in press.

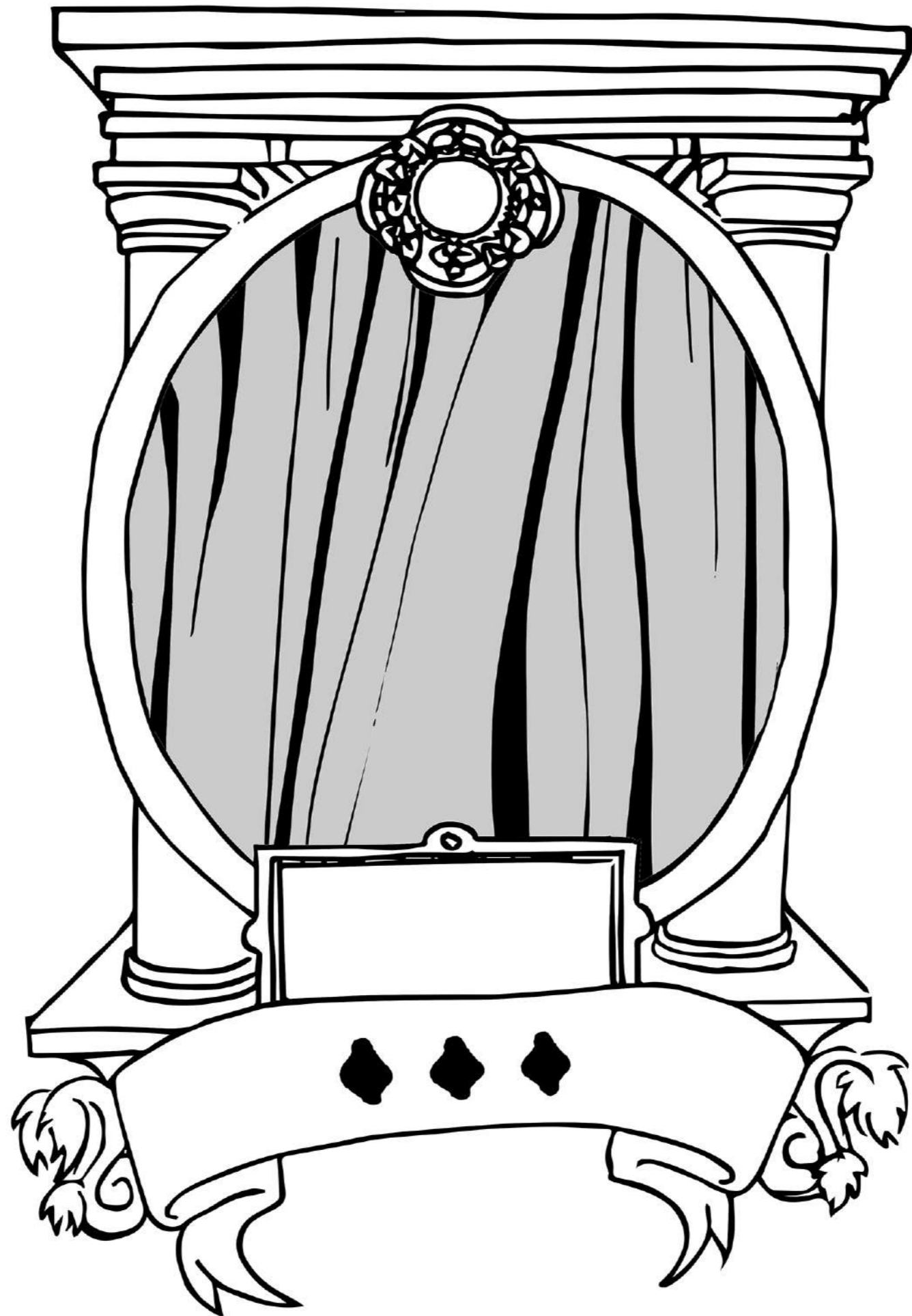
**Definition 2.3.** Let the continuous function  $\widehat{L}(\omega) = L(i\omega)$  be the Fourier multiplier associated with the operator L. We call L “*spline-admissible*” if there exists  $0 \leq \Omega \in \mathbb{R}$  such that

- (i)  $\widehat{L}(\omega)$  has a finite number of (distinct) zeros, all included in  $[-\Omega, \Omega]$ , and the spacing between no two of them is an integer multiple of  $2\pi$ ;
- (ii)  $\widehat{L}(\omega)$  is twice differentiable at all  $|\omega| > \Omega$  and there exist constants  $0 < c_1, \epsilon_1 \in \mathbb{R}$  such that

$$\forall \omega, |\omega| > \Omega : \left| \frac{d^2}{d\omega^2} \log \widehat{L}(\omega) \right| \leq \frac{c_1}{|\omega|^{1+\epsilon_1}};$$

- (iii) there exist constants  $0 < c_2, \epsilon_2 \in \mathbb{R}$  such that

$$\forall \omega, |\omega| > \Omega : \left| \widehat{L}(\omega) \right| \geq c_2 |\omega|^{\frac{1}{2} + \epsilon_2}.$$





Second Act  
2007 - 2015

Sparse and Random

# THE RISE AND FALL OF GAUSSIAN MODELS

-  = Gaussian white noise



- The Fourier-Gauss-Hilbert realm

Fourier transforms optimally represent Gaussian processes [U'84]

Quadratic optimization equivalent to optimal Gauss-based estimation

- But... sparsity!

Gaussian is not sparse

New mathematical methods: wavelets, FRI, compressed sensing, ....

Stochastic models for sparse signals?

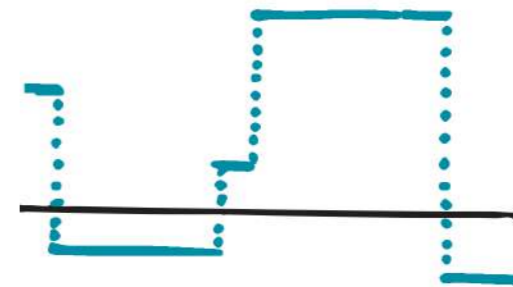
# FROM PIECEWISE-SMOOTH TO SPARSE PROCESSES

- Innovation: random, independent at every point, stationary



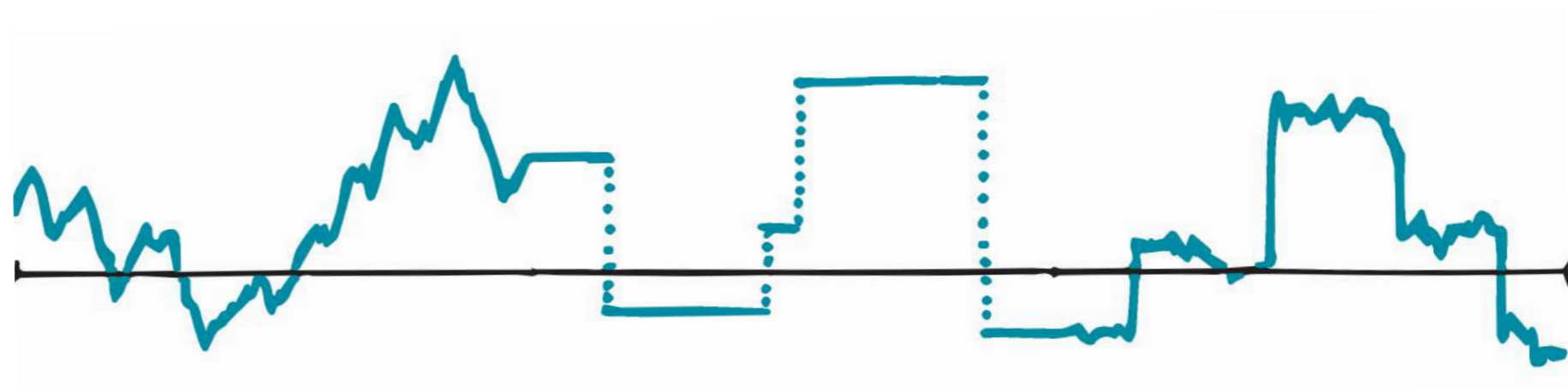
- = Poisson white noise

When random processes are splines [UT'11]



- **Definition:** A *sparse stochastic process* satisfies  $Ls = w$

$L = D$



$w$

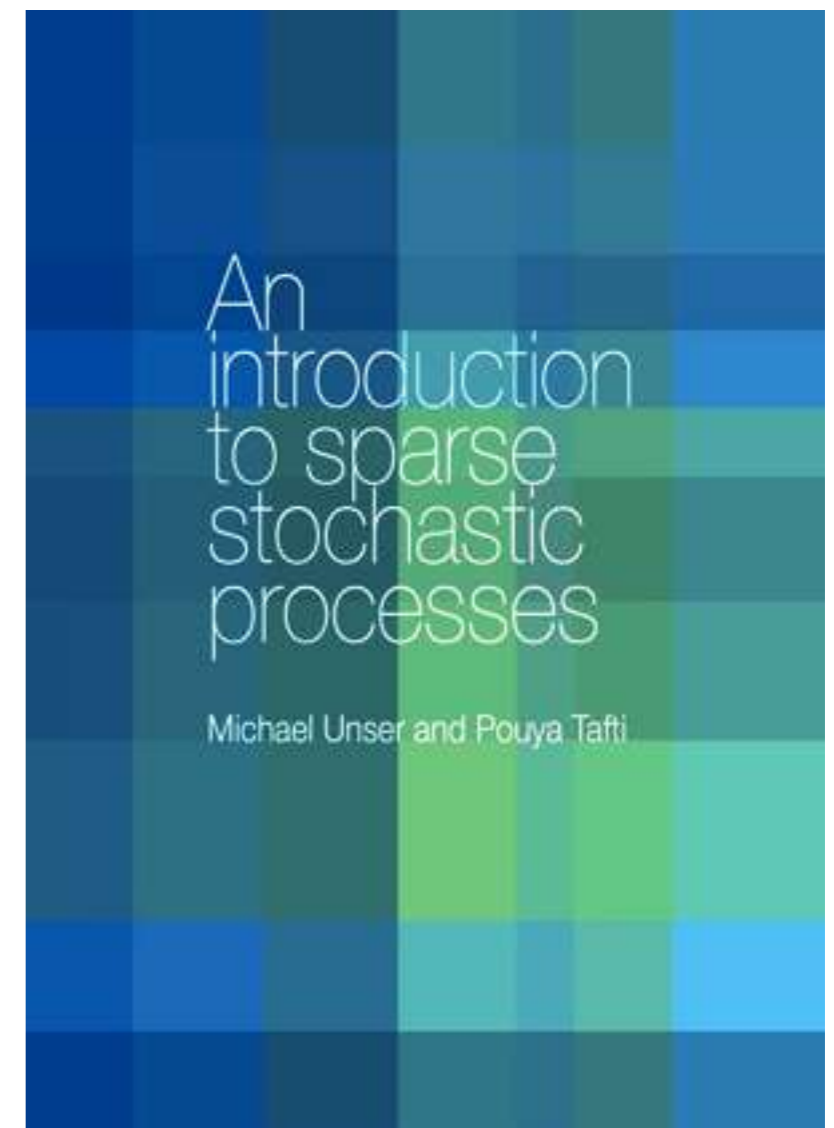
Gauss

Poisson

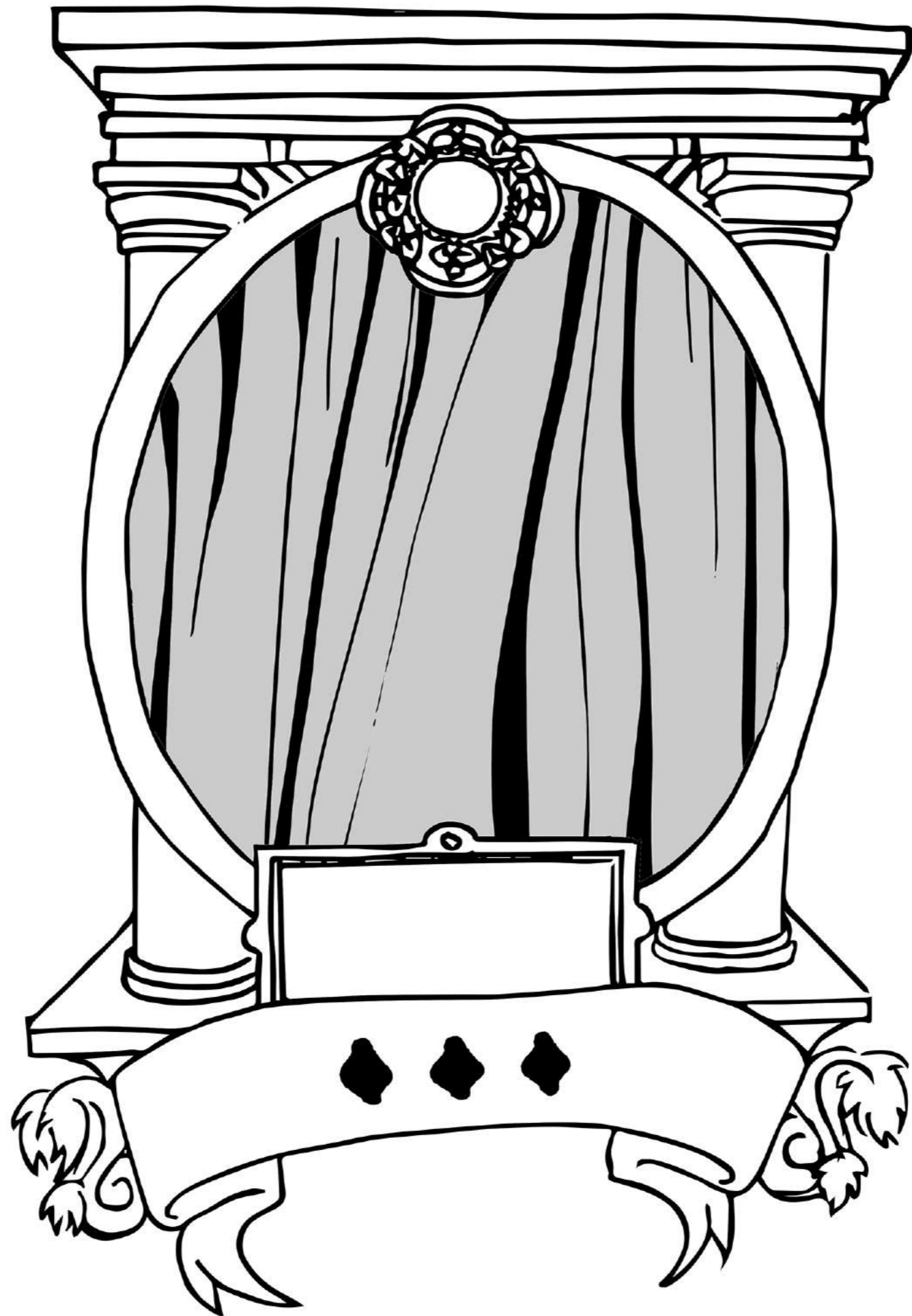
Cauchy

# MATHEMATICS OF SPARSE STOCHASTIC PROCESSES

- *Signal processing*: probabilistic model for sparse signals
- *Mathematics*: analysis of stochastic differential equations driven by a Lévy white noise
- The foundations: construct your mathematics [UT'14,FAU'14]
- Wavelets are optimal to compress sparse processes [PU'15]
- Estimation methods based on sparse priors [BKNU'13,KPAM'13]
- The mathematical meaning of sparsity [AFU'17,FUW'17]





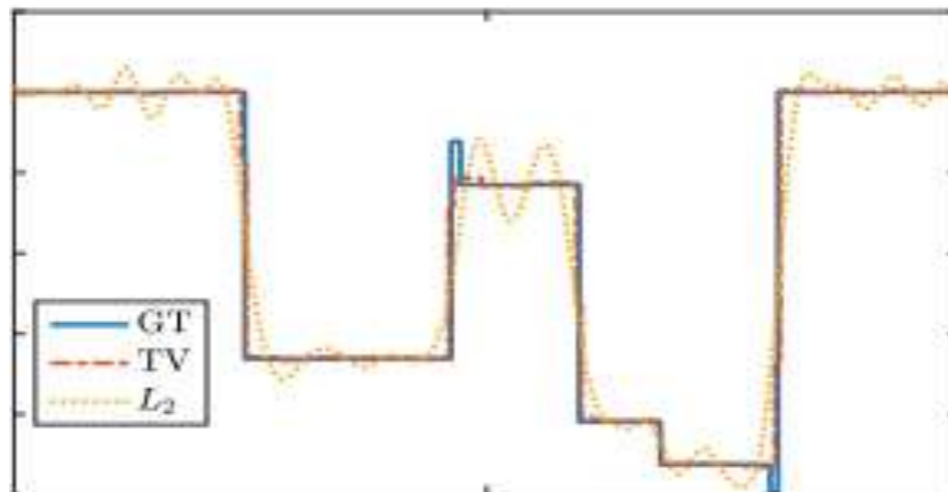




Third Act  
2015 -

The Return of the Splines

# OPTIMIZATION METHODS FOR SPARSE SIGNALS



[GFU'18]

## Representer Theorem in $L_2$

$$\min \sum_{m=1}^M (y_m - s(t_m))^2 + \lambda \|Ls\|_{L_2}^2$$

$$L^*L\{s_{\text{optimal}}\} = \sum_{m=1}^M b_m \delta(\cdot - t_m)$$

## Representer Theorem in $\mathcal{M}$ (the real $L_1$ )

$$\min \sum_{m=1}^M (y_m - s(t_m))^2 + \lambda \|Ls\|_{\mathcal{M}}$$

$$L\{s_{\text{optimal}}\} = \sum_{k=1}^K c_k \delta(\cdot - x_k) \quad K \leq M$$

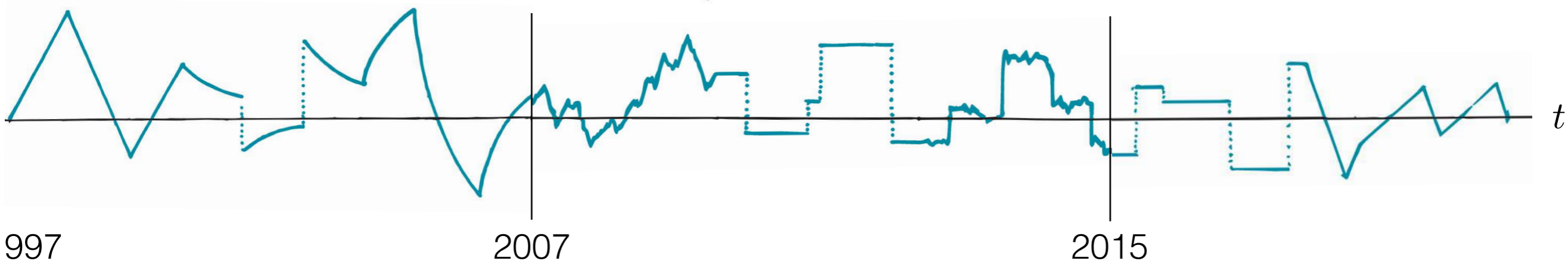
First complete mathematical framework to analyse sparse promoting regularization in continuous-domain [UFW'17,U&co'19]

New algorithms to reconstruct sparse signals

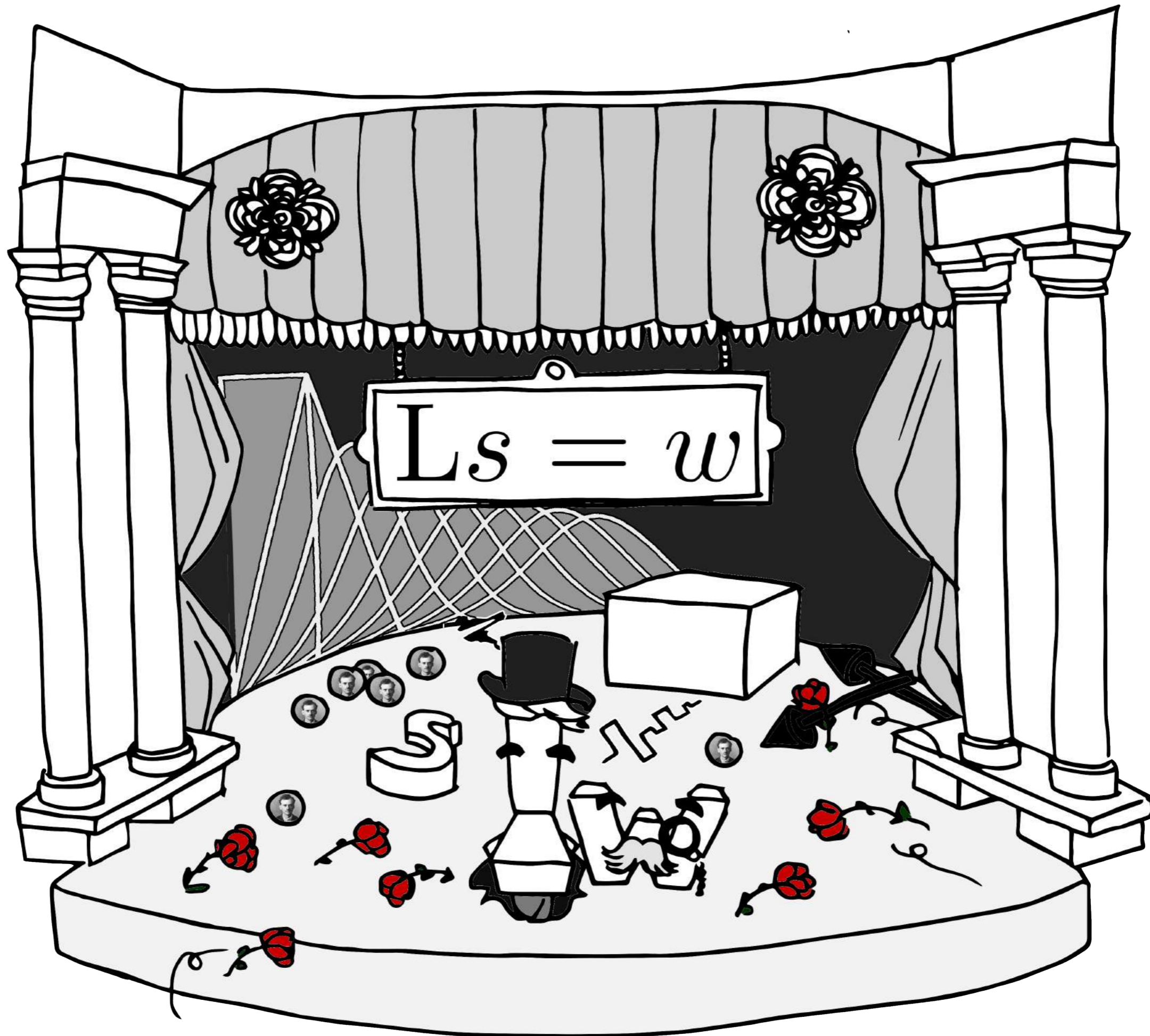
Harshit Gupta, Shayan Aziznejad, Thomas Debarre



- From the particular to the general, and back
- Original contributions to mathematics:  
approximation theory, stochastic processes, functional analysis
- Facing nowadays challenges in signal processing



A unified masterpiece



$Ls = w$