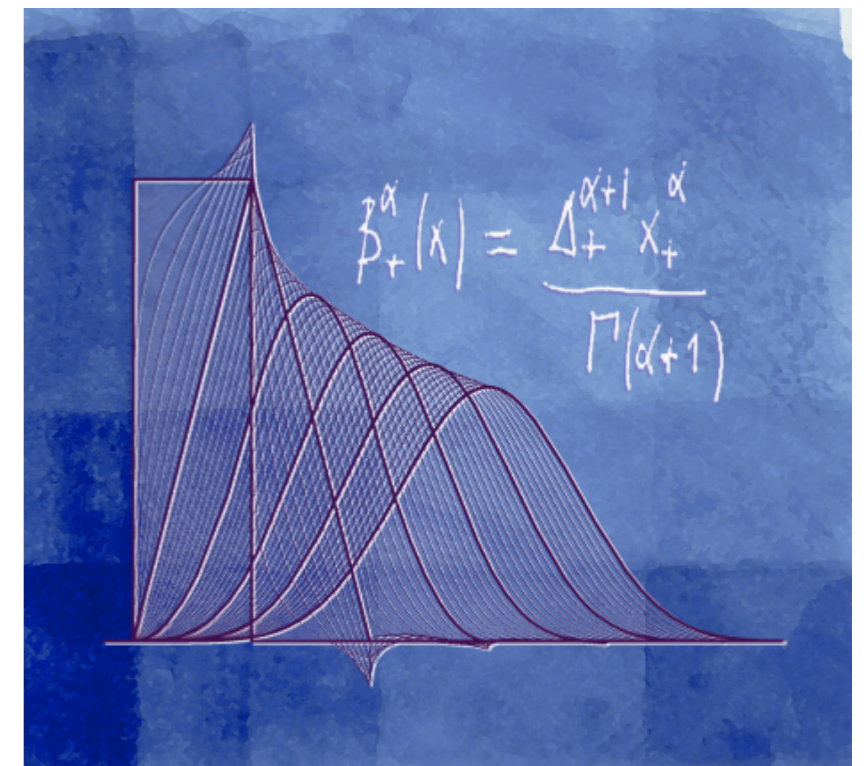


Vers une théorie unificatrice pour le traitement numérique/analogique des signaux

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Is continuous-time signal processing dead ?

- Arguments in favor of its suppression:
 - The modern world is discrete (CDs, DVDs, WEB, etc...)
 - Modern SP courses concentrate on digital signal processing
 - Most processing is discrete (DSPs, PCs, etc...)
 - Students don't like the Laplace transform...
- However...
 - Real-world signals are continuous
 - Often, the end product is analog: control systems, sound reproduction systems, etc.
 - Don't forget the interface: A-to-D and D-to-A
 - Some discrete algorithms require continuous-time thinking

Revival of continuous-time thinking

- Recent trends in SP
 - Wavelet theory, multiresolution analysis
 - Self-similarity, fractals, analysis of singularities
 - Partial differential equations
 - Spline-based signal processing
- Continuous/discrete formulation
 - **“Think analog, act digital”**
 - Applications:
 - Fractional delays, sampling rate conversion
 - Discretization of differential operators
 - Interpolation
 - ...

OUTLINE

- In search of the missing link
- E-splines
- B-spline calculus
- Application: hybrid signal processing

IN SEARCH OF THE MISSING LINK

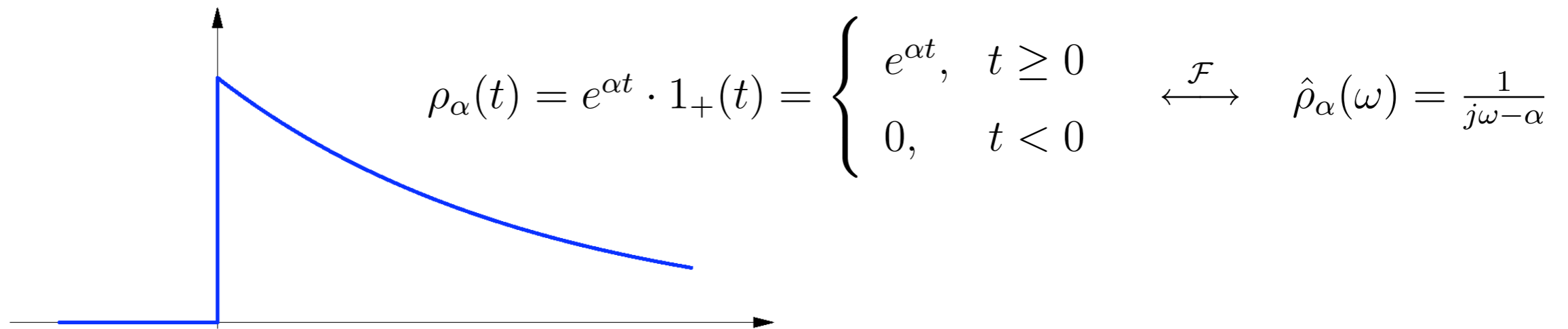
Knowing splines is an advantage
[Schoenberg, 1946]

Teach “Signals and Systems” ...

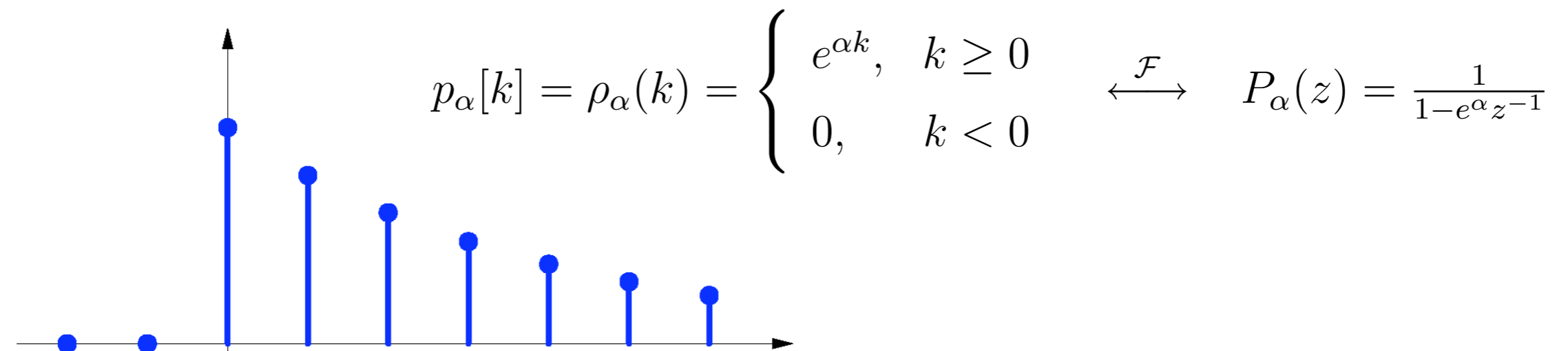
Continuous vs discrete: example

- Causal exponential

- Continuous-time version



- Discrete-time version



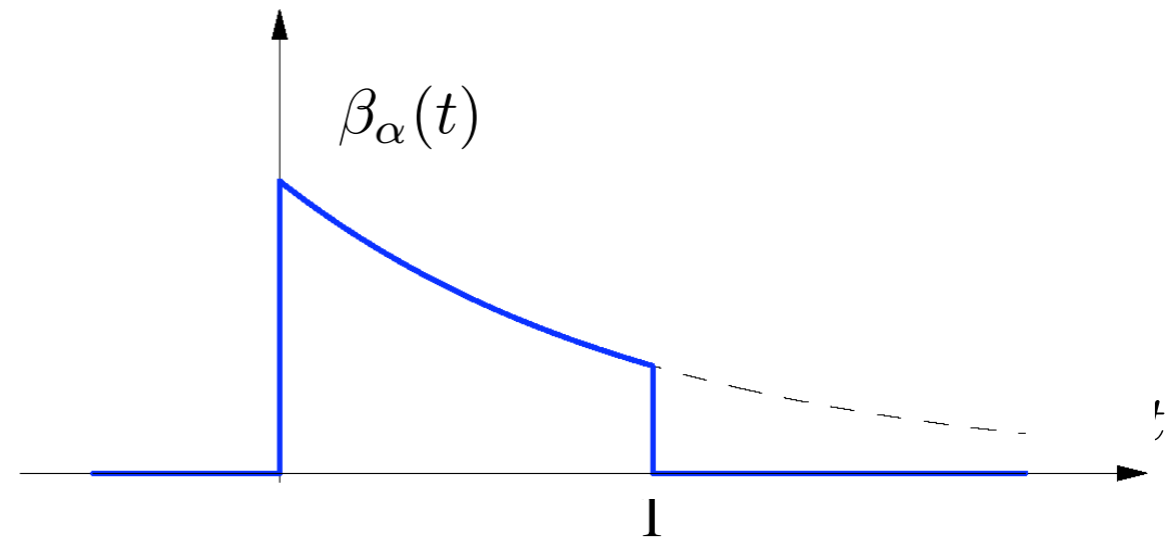
What is the link ?

- Answer: ratio of Fourier transforms

$$\hat{\beta}_\alpha(\omega) = \frac{\hat{\rho}_\alpha(\omega)}{P_\alpha(e^{j\omega})} = \frac{1 - e^\alpha e^{-j\omega}}{j\omega - \alpha}$$

\mathcal{F}^{-1}

$$\beta_\alpha(t) = \rho_\alpha(t) - e^\alpha \cdot \rho_\alpha(t - 1)$$



- Reproduction formula

$$\rho_\alpha(t) = 1_+(t) \cdot e^{\alpha t} = \sum_{k=0}^{+\infty} e^{\alpha k} \beta_\alpha(t - k) = \sum_{k \in \mathbb{Z}} p_\alpha[k] \beta_\alpha(t - k)$$

Continuous-time signal

Compactly-supported basis functions

Discrete signal

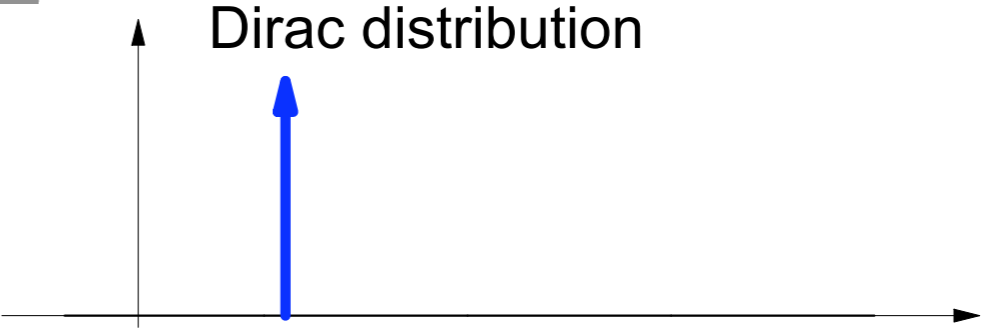
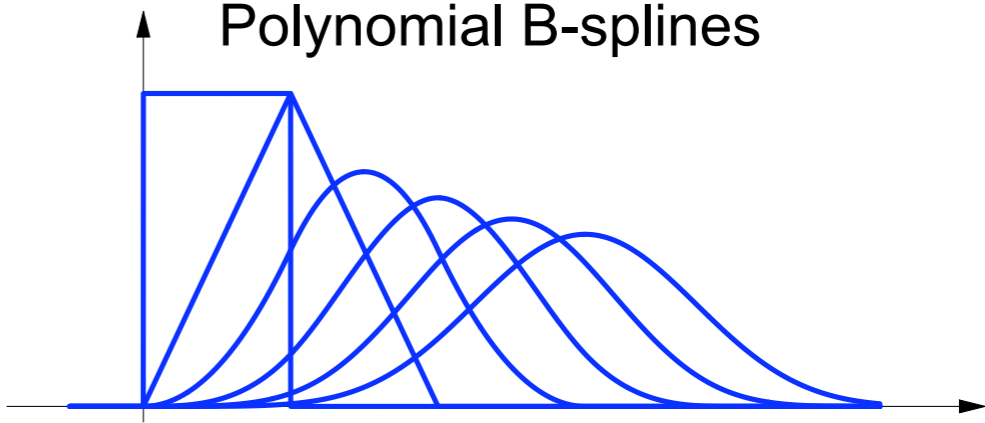
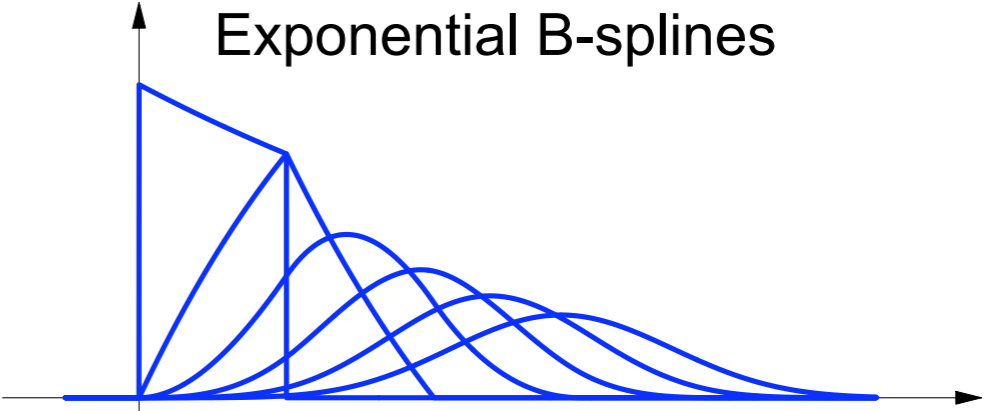
Basic continuous-time convolution operators

Operator	Notation	Impulse response	Frequency response
Identity	$I\{ \}$	$\delta(t)$	1
Shift	$S_\tau\{f\} = f(t - \tau)$	$\delta(t - \tau)$	$e^{-j\omega\tau}$
Integral	$D^{-1}\{ \} = \int_{-\infty}^t dt$	$1_+(t)$	$\pi\delta(\omega) + \frac{1}{j\omega}$
Multiple integral	$D^{-n}\{ \}$	$\frac{t_+^{n-1}}{(n-1)!}$	$\frac{j^{n-1}\pi\delta^{(n-1)}(\omega)}{(n-1)!} + \frac{1}{(j\omega)^n}$
Simple differential system	$(D - \alpha I)^{-1}\{ \}$	$1_+(t) \cdot e^{\alpha t}$	$\frac{1}{j\omega - \alpha}$ $\text{Re}\{\alpha\} < 0$
Iterated differential system	$(D - \alpha I)^{-n}\{ \}$	$\frac{t_+^{n-1} e^{\alpha t}}{(n-1)!}$	$\frac{1}{(j\omega - \alpha)^n}$ $\text{Re}\{\alpha\} < 0$

... and their discrete counterparts

Name	Discrete time specification	z-transform
Unit impulse	$\delta[k]$	1
Shift	$\delta[k - k_0]$	z^{-k_0}
Unit step	$p_0[k] = \begin{cases} 0, & k < 0 \\ 1, & k \geq 0 \end{cases}$	$\frac{1}{1 - z^{-1}}$
Discrete monomial	$p_0^{[n-1]}[k] = \begin{cases} 0, & k < 0 \\ \prod_{m=1}^{n-1} (k + m), & k \geq 0 \end{cases}$	$\frac{1}{(1 - z^{-1})^n}$
Causal exponential	$p_\alpha[k] = \begin{cases} 0, & k < 0 \\ e^{\alpha k}, & k \geq 0 \end{cases}$	$\frac{1}{1 - e^\alpha z^{-1}}$
Discrete exponential monomial	$p_\alpha^{[n-1]}[k] = \begin{cases} 0, & k < 0 \\ e^{\alpha k} \prod_{m=1}^{n-1} (k + m), & k \geq 0 \end{cases}$	$\frac{1}{(1 - e^\alpha z^{-1})^n}$

D-to-A translating B-splines

B-spline	Operator L	Order N	Frequency response	
$\delta(t)$	$I\{ \}$	0	1	 <p>Dirac distribution</p>
$\delta(t - \tau)$	$S_\tau\{ \}$	0	$e^{-j\omega\tau}$	
$\beta_{(0)}(t)$	$D\{ \} = \frac{d}{dt}$	1	$\frac{1 - e^{-j\omega}}{j\omega}$	 <p>Polynomial B-splines</p>
$\beta_{(0,\dots,0)}(t)$	$D^n\{ \}$	n	$\left(\frac{1 - e^{-j\omega}}{j\omega}\right)^n$	
$\beta_\alpha(t)$	$(D - \alpha I)\{ \}$	1	$\frac{1 - e^{\alpha - j\omega}}{j\omega - \alpha}$	 <p>Exponential B-splines</p>
$\beta_{(\alpha,\dots,\alpha)}(t)$	$(D - \alpha I)^n\{ \}$	n	$\left(\frac{1 - e^{\alpha - j\omega}}{j\omega - \alpha}\right)^n$	

E-SPLINES

- Generalized splines
- Exponential B-splines
- B-spline properties
- B-spline representation

General concept of an *L*-spline

$L\{\cdot\}$: differential operator (shift-invariant)

$\delta(t)$: Dirac distribution

Definition A: The continuous-time function $s(t)$ is an ***L-spline*** with knots $\{t_k\}_{k \in \mathbb{Z}}$ iff:

$$L\{s(t)\} = \sum_{k \in \mathbb{Z}} a_k \delta(t - t_k)$$

Definition B: The continuous-time function $s(t)$ is a ***cardinal L-spline*** iff:

$$L\{s(t)\} = \sum_{k \in \mathbb{Z}} a[k] \delta(t - k)$$

Exponential spline defining operator

- General differential system

$$(D^N + a_1 D^{N-1} + \cdots + a_N I) \{y(t)\} = (D^M + \cdots + b_M I) \{x(t)\}$$

$$\iff L_{\vec{\alpha}} \{y(t)\} = x(t)$$

- Rational transfer function

$$L_{\vec{\alpha}}(\omega) = \frac{\prod_{n=1}^N (j\omega - \alpha_n)}{\prod_{m=1}^M (j\omega - \gamma_m)}$$

- Exponential spline parameters

$$\vec{\alpha} = \underbrace{(\alpha_1, \cdots, \alpha_N)}_{\text{Poles}}; \underbrace{(\gamma_1, \cdots, \gamma_M)}_{\text{Zeros (optional)}} \text{ with } M < N$$

Poles

Zeros (optional)

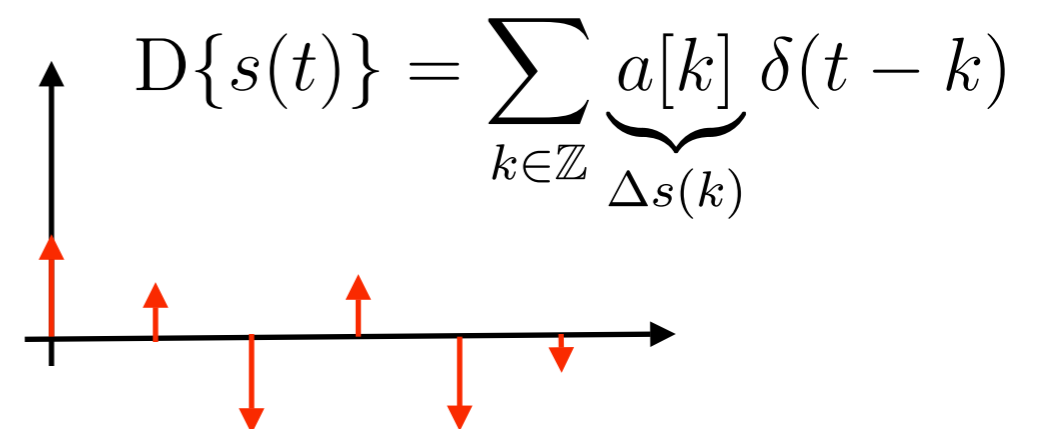
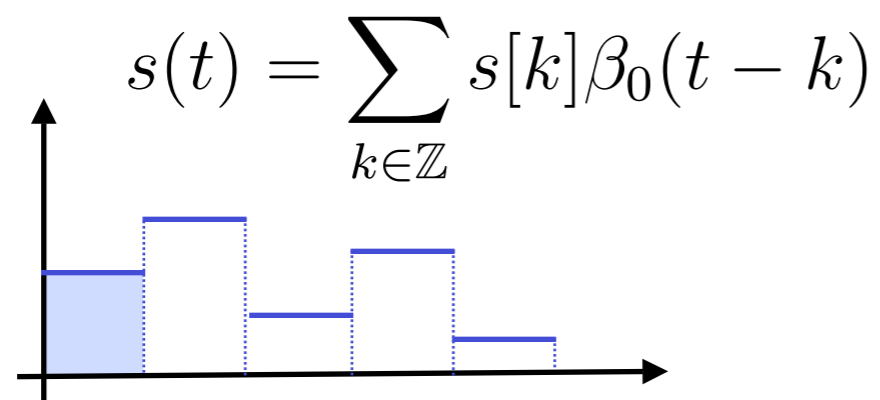
Example: piecewise-constant splines

- Spline-defining operators

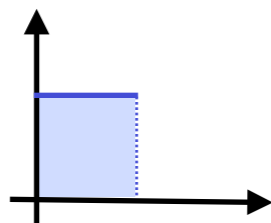
Continuous-time derivative: $D = \mathcal{L}_0\{\cdot\} \longleftrightarrow j\omega$

Discrete-time derivative: $\Delta\{\cdot\} \longleftrightarrow 1 - e^{-j\omega}$

- Piecewise constant or D-spline



- B-spline function:



$$\beta_0(t) = \Delta\{1_+(t)\} \longleftrightarrow \frac{1 - e^{-j\omega}}{j\omega}$$

Exponential B-splines

- Localization operator (weighted finite differences)

$$\Delta_{\vec{\alpha}}(z) = \prod_{n=1}^N (1 - e^{\alpha_n} z^{-1}) \quad \text{Mapping: } z = e^s$$

- Fourier domain formula

$$\hat{\beta}_{\vec{\alpha}}(\omega) = \frac{\Delta_{\vec{\alpha}}(e^{j\omega})}{L_{\vec{\alpha}}(\omega)}$$

Discrete-time approximation

Continuous-time operator

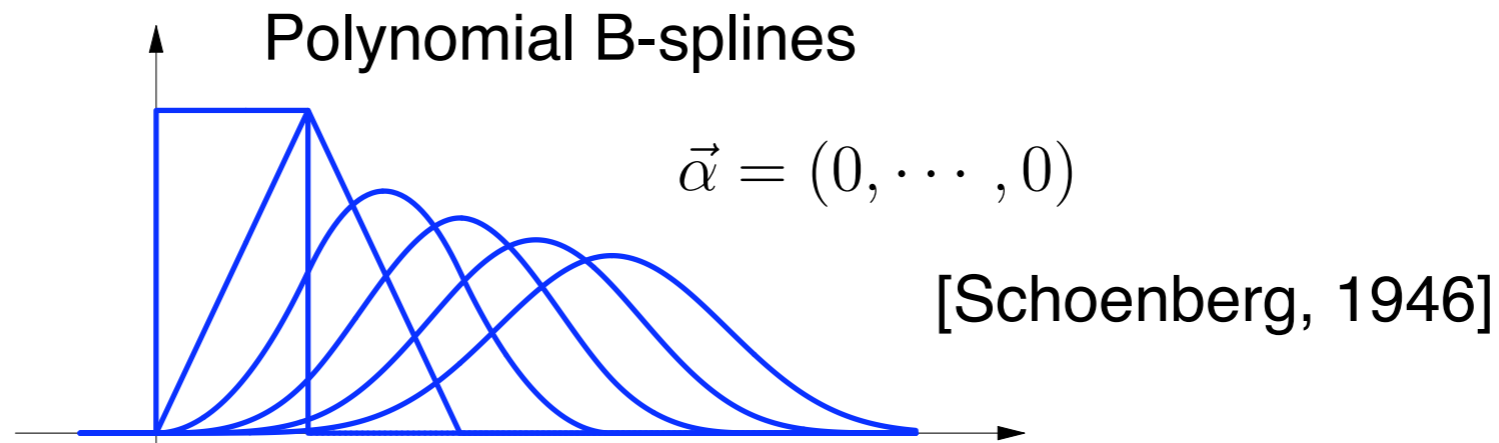
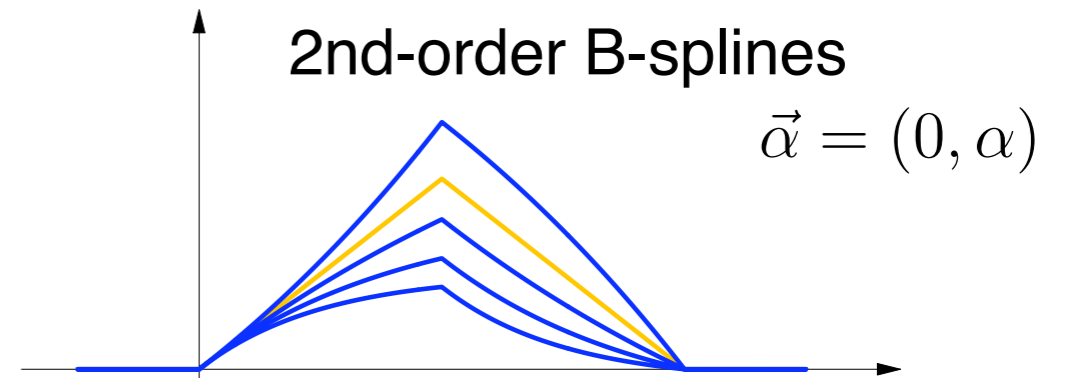
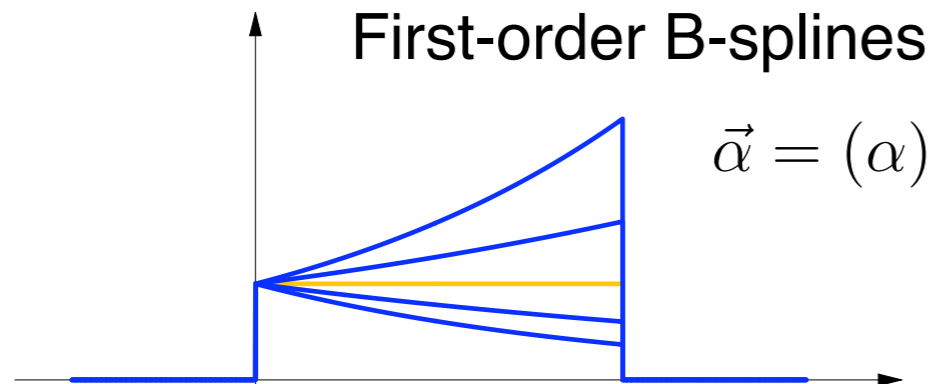
- Time-domain formula (inverse Laplace transform)

$$\beta_{\vec{\alpha}}(t) = \mathcal{L}^{-1} \left\{ \left(\prod_{n=1}^N \frac{1 - e^{\alpha_n - s}}{s - \alpha_n} \right) \cdot \prod_{m=1}^M (s - \gamma_m) \right\}$$

poles

zeros

Exponential B-splines (Cont'd)



■ Properties

- Piecewise exponential/polynomial (E-spline)
- Compact support: size N
- Continuity: Hölder of order $n = N - M - 1$

B-spline convolution property

- Convolution property

$$(\beta_{\vec{\alpha}_1} * \beta_{\vec{\alpha}_2})(t) = \beta_{(\vec{\alpha}_1 : \vec{\alpha}_2)}(t)$$

$$(\vec{\alpha}_1 : \vec{\alpha}_2) =$$

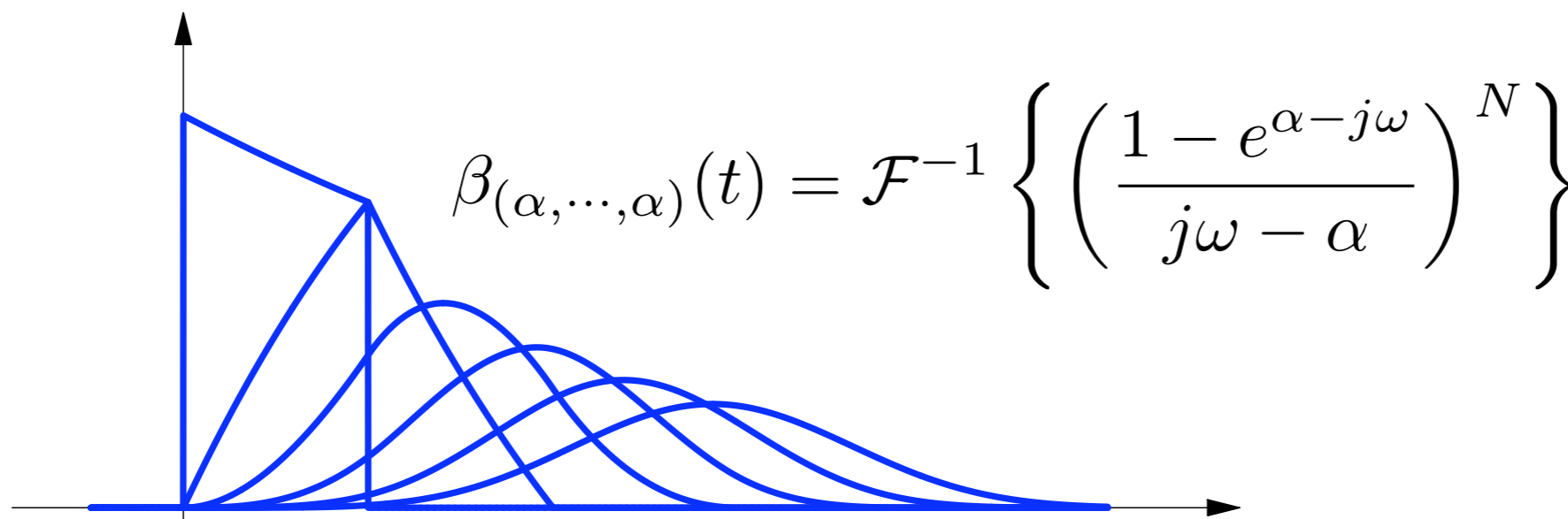
$$\left(\underbrace{\alpha_{1,1}, \dots, \alpha_{1,N_1}, \alpha_{2,1}, \dots, \alpha_{2,N_2}}_{\text{concatenation of poles}}; \underbrace{\gamma_{1,1}, \dots, \gamma_{1,M_1}, \gamma_{2,1}, \dots, \gamma_{2,M_2}}_{\text{concatenation of zeros}} \right)$$

concatenation of poles

concatenation of zeros

- Example: g-splines

[Panda et al., 1996]



E-splines: B-spline representation

- Space of cardinal E-splines

$$V_{\vec{\alpha}} = \left\{ s(t) : L_{\vec{\alpha}}\{s(t)\} = \sum_{k \in \mathbb{Z}} a[k] \delta(t - k) \right\} \cap L_2$$

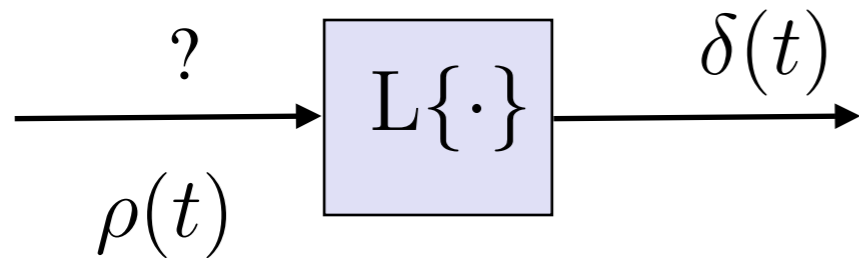
- B-spline representation

Theorem: The set of functions $\{\beta_{\vec{\alpha}}(t - k)\}_{k \in \mathbb{Z}}$ provides a Riesz basis of $V_{\vec{\alpha}}$ if and only if $\alpha_n - \alpha_m \neq j2\pi k, k \in \mathbb{Z}$ for all pairs of distinct, purely imaginary poles.

$$V_{\vec{\alpha}} = \left\{ \underbrace{s(t)}_{\text{continuous-time signal}} = \sum_{k \in \mathbb{Z}} \underbrace{c[k]}_{\text{discrete-time signal (B-spline coefficients)}} \beta_{\vec{\alpha}}(t - k) : c \in \ell_2 \right\}$$

Green function reproduction

- Green function



$\rho(t)$: Green function of $L\{\cdot\}$

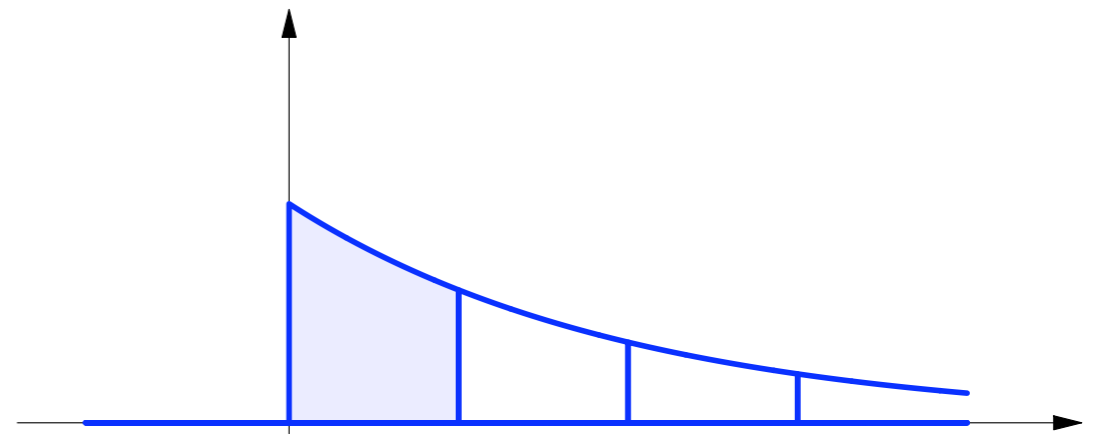
\Updownarrow

$$L\{\rho(t)\} = \delta(t)$$

- Green function reproduction = A-to-D translation

$$\rho_{\vec{\alpha}}(t) = \sum_{k \in \mathbb{Z}} p_{\vec{\alpha}}[k] \beta_{\vec{\alpha}}(t - k)$$

with
$$P_{\vec{\alpha}}(z) = \prod_{n=1}^N \frac{1}{1 - e^{\alpha_n} z^{-1}}$$



B-SPLINE CALCULUS

- Interpolation
- Convolution
- Modulation
- Differential operators
- ...

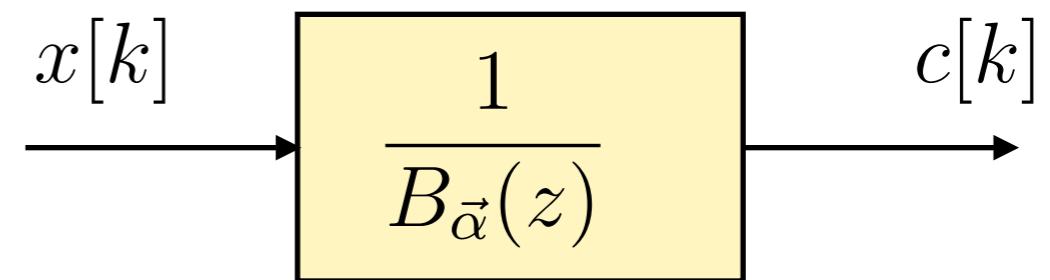
Interpolation

- Interpolation condition

$$x[k] = \sum_{n \in \mathbb{Z}} c[n] \beta_{\vec{\alpha}}(t - n) \Big|_{t=k} = (b_{\vec{\alpha}} * c)[k]$$

- B-spline kernel: $B_{\vec{\alpha}}(z) = \sum_{k=0}^{N-1} \beta_{\vec{\alpha}}(k) z^{-k}$

- Digital filtering algorithm



Recursive IIR filter

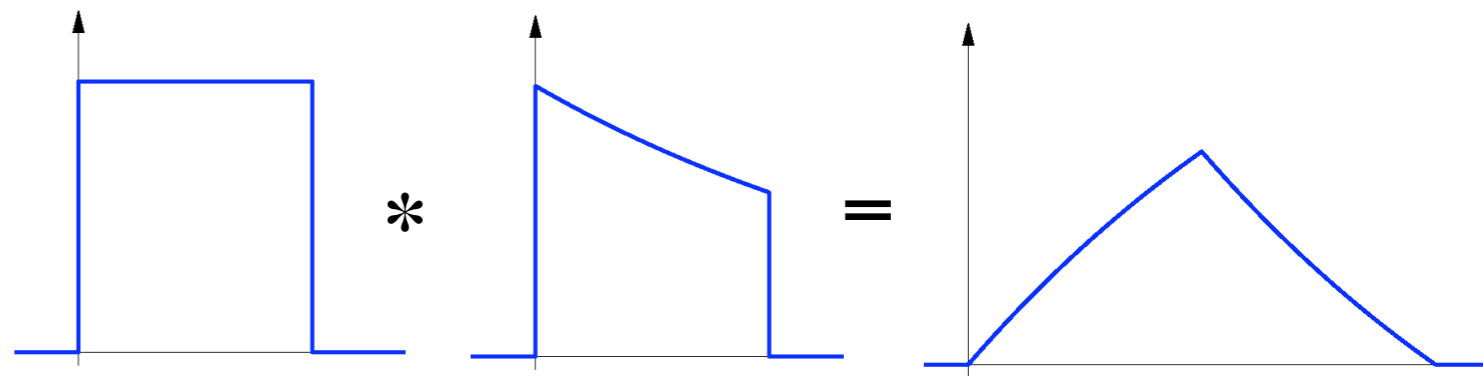
Convolution

- Input signals

$$s_1(t) = \sum_{k \in \mathbb{Z}} c_1[k] \beta_{\vec{\alpha}_1}(t - k) \quad s_2(t) = \sum_{k \in \mathbb{Z}} c_2[k] \beta_{\vec{\alpha}_2}(t - k)$$

- B-spline convolution property

$$(\beta_{\vec{\alpha}_1} * \beta_{\vec{\alpha}_2})(t) = \beta_{(\vec{\alpha}_1 : \vec{\alpha}_2)}(t)$$



- Continuous-time convolution

$$(s_1 * s_2)(t) = \sum_{k \in \mathbb{Z}} (c_1 * c_2)[k] \beta_{(\vec{\alpha}_1 : \vec{\alpha}_2)}(t - k)$$

Discrete-time convolution

Augmented order B-spline

Modulation

- Input signal $s(t) = \sum_{k \in \mathbb{Z}} c[k] \beta_{\vec{\alpha}}(t - k)$

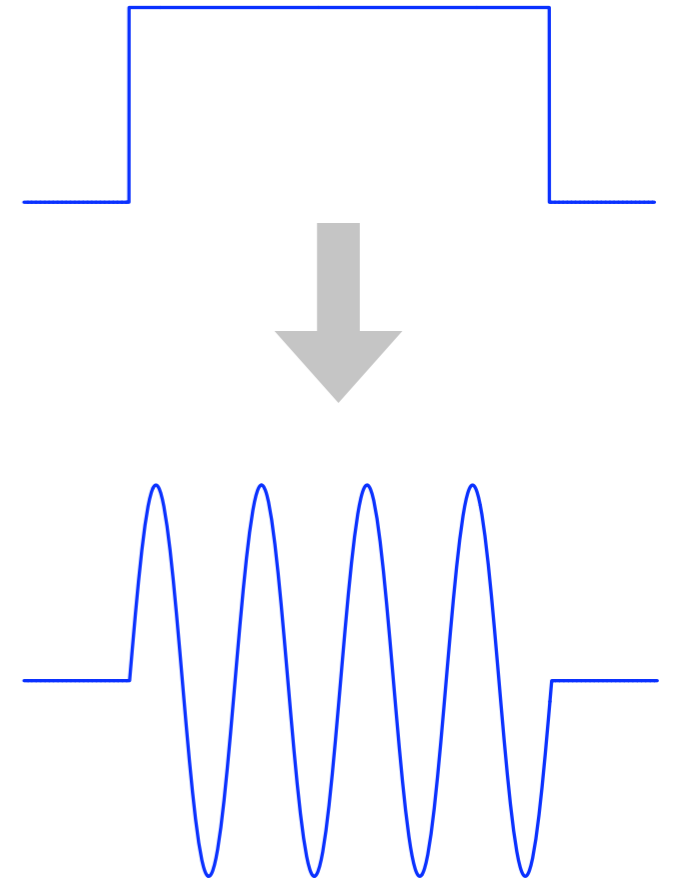
- B-spline modulation property

$$\beta_{\vec{\alpha}}(t) \cdot e^{j\omega_0 t} = \beta_{\vec{\alpha} + \vec{j}\omega_0}(t)$$

- Continuous-time modulation

$$s(t) \cdot e^{j\omega_0 t} = \sum_{k \in \mathbb{Z}} (c[k] \cdot e^{j\omega_0 k}) \beta_{\vec{\alpha} + \vec{j}\omega_0}(t - k)$$

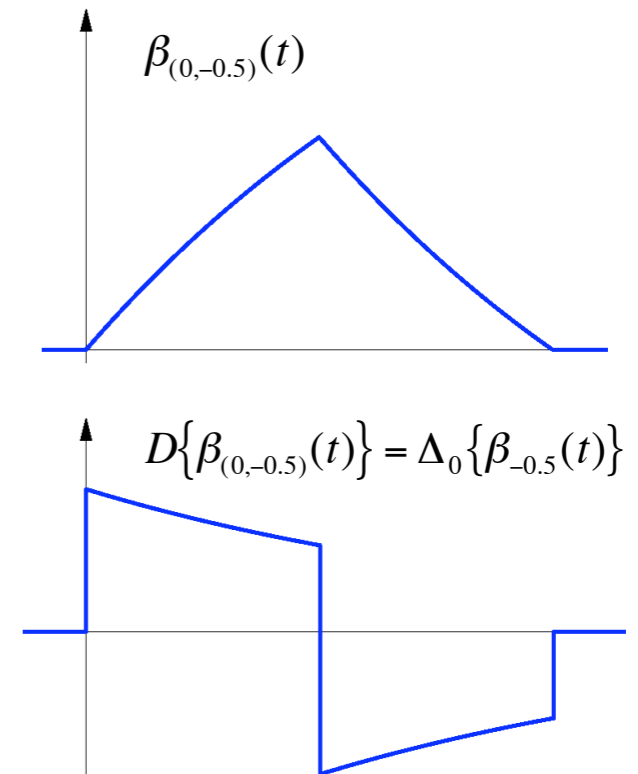
Discrete-time modulation



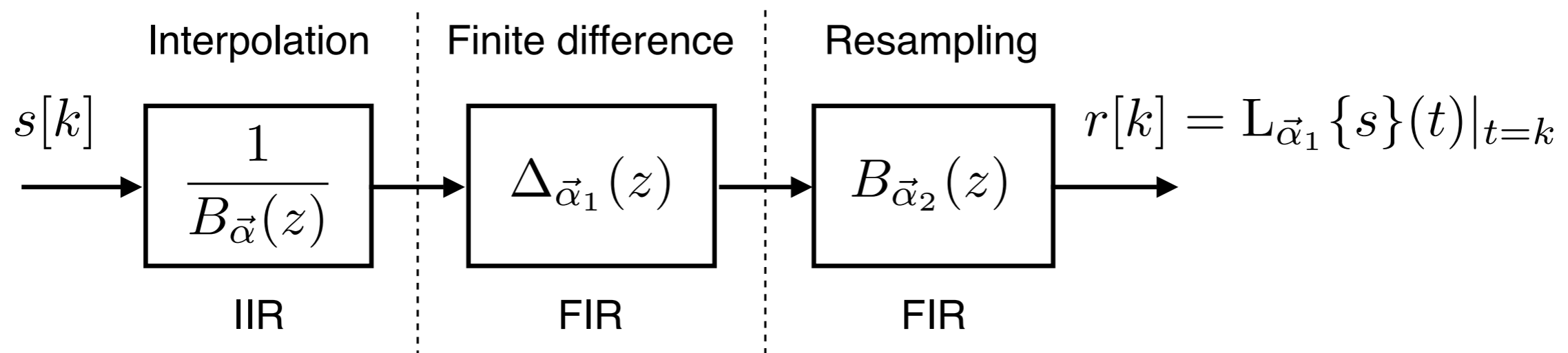
Differential operators

- B-spline differentials

$$L_{\vec{\alpha}_1} \{ \beta_{(\vec{\alpha}_1:\vec{\alpha}_2)}(t) \} = \Delta_{\vec{\alpha}_1} \{ \beta_{\vec{\alpha}_2}(t) \}$$



- Implementation of differential operator



APPLICATION: HYBRID SIGNAL PROCESSING

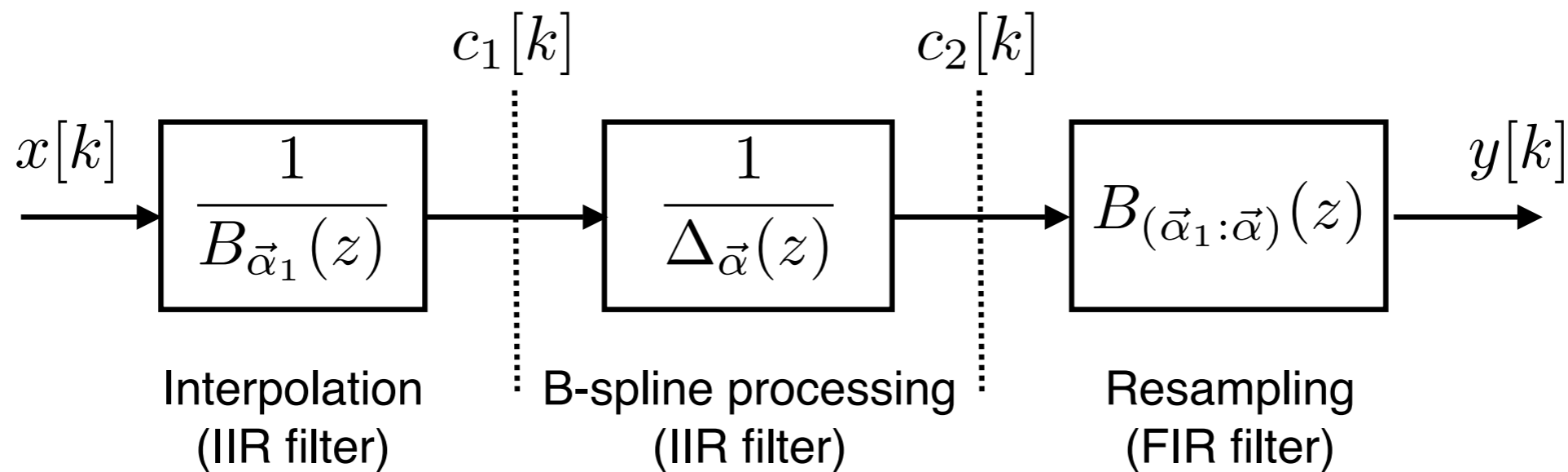
- Analog filtering in the B-spline domain
- Consistent sampling
- Digitally-compensated D-to-A conversion

Analog filtering in the B-spline domain

Analog filter: $h(t) = \sum_{k \in \mathbb{Z}} p[k] \beta_{\vec{\alpha}}(t - k)$

Input signal: $x(t) = \sum_{k \in \mathbb{Z}} c_1[k] \beta_{\vec{\alpha}_1}(t - k)$

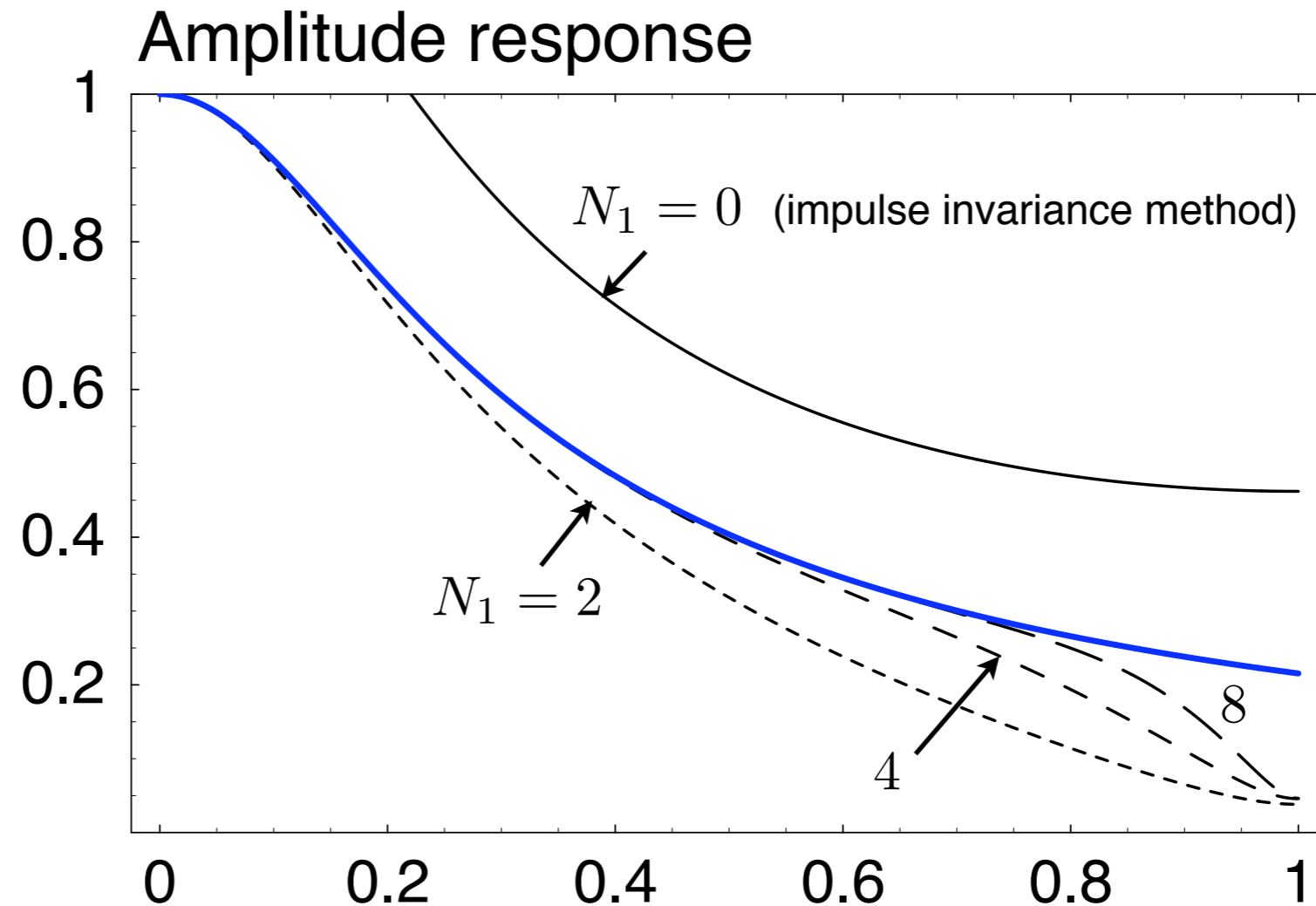
Output signal: $y(t) = \sum_{k \in \mathbb{Z}} (p * c_1)[k] \beta_{(\vec{\alpha}_1: \vec{\alpha})}(t - k)$



$$R_2(z) = \frac{B_{(\vec{\alpha}_1: \vec{\alpha})}(z)}{\Delta_{\vec{\alpha}}(z)} = P(z) \cdot B_{(\vec{\alpha}_1: \vec{\alpha})}(z)$$

Example: first order butterworth

Filter to design: $H(s) = \frac{-\alpha}{s - \alpha}$



Input model: polynomial spline of order N_1

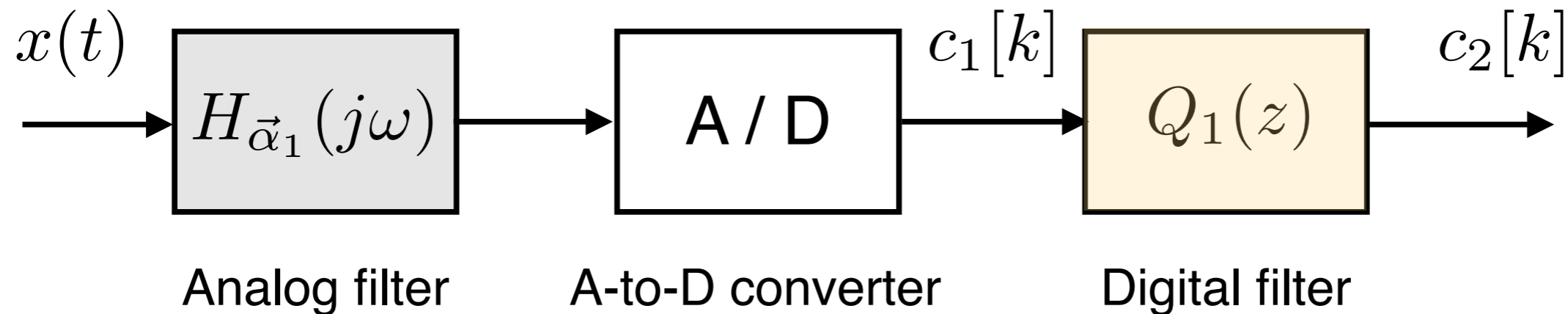
Design example: $\vec{\alpha}_1 = (0, 0) \implies R_{12}(z) = \frac{0.2786 + 0.2213z^{-1}}{1 - 0.5z^{-1}}$

Consistent sampling system

Reconstructed signal: $y(t) = \sum_{k \in \mathbb{Z}} c_2[k] \varphi_2(t - k)$

Consistency requirement:

$$\forall k \in \mathbb{Z}, \langle x(t), \varphi_1(t - k) \rangle = \langle y(t), \varphi_1(t - k) \rangle$$



Digital reconstruction filter: $Q_1(z) = \frac{\Delta_{\vec{\alpha}_1}(z)}{\sum_{k=0}^{N_1+N_2} \beta_{(\vec{\alpha}_1:\vec{\alpha}_2)}(k) z^{-k}}$

Digitally-compensated D-to-A conversion

Reconstructed signal:

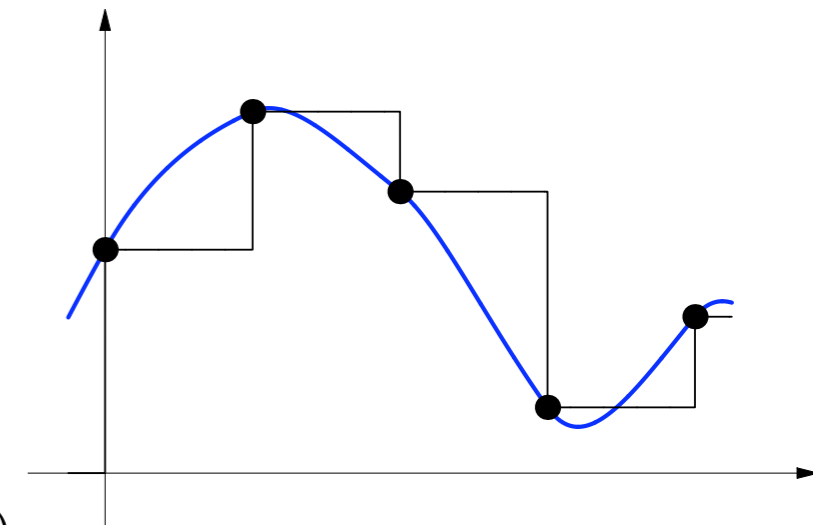
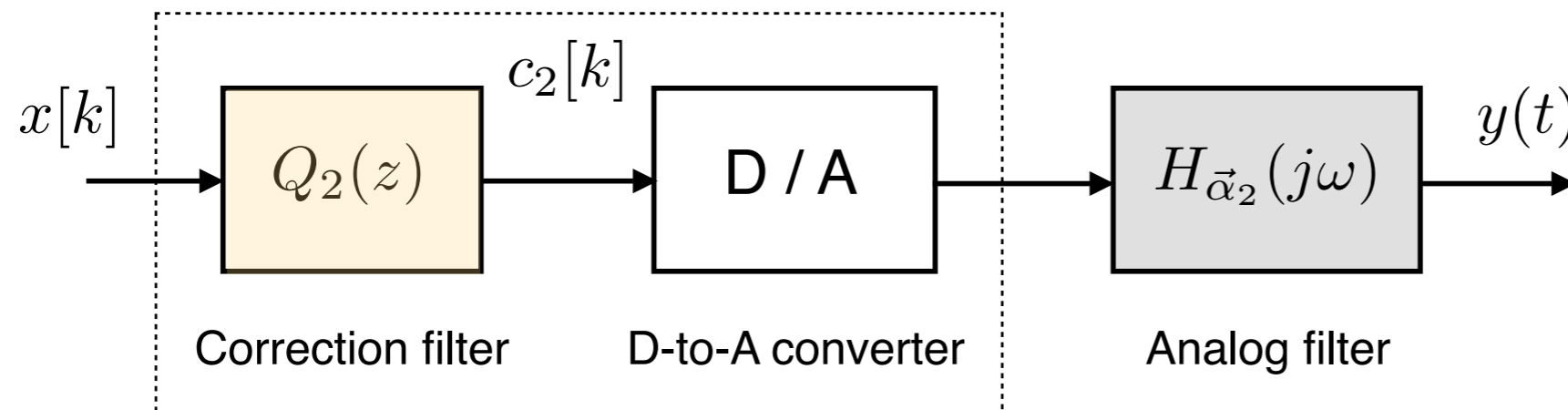
$$y(t) = \sum_{k \in \mathbb{Z}} c_2[k] \varphi_2(t - k)$$

Interpolation condition:

$$y(t)|_{t=k} = x[k]$$

Equivalent synthesis function:

$$\varphi_2(t) = (\beta_{(0)} * \rho_{\vec{\alpha}_2})(t)$$



Digital correction filter:
$$Q_2(z) = \frac{\Delta_{\vec{\alpha}_2}(z)}{\sum_{k=0}^{N_2+1} \beta_{(0:\vec{\alpha}_2)}(k) z^{-k}}$$

CONCLUSION

- Cardinal E-splines: numerous attractive properties
 - B-spline representation = discrete signal
 - Family closed with respect to primary continuous-time signal processing operators (e.g., convolution, modulation, differential operators)
 - Easy to manipulate (e.g., recursive filtering algorithms, explicit formulas)
 - Generality: include all known brands of splines (polynomial, trigonometric, hyperbolic) and many more

The end: Thank you!

- The key collaborator: **Thierry Blu**
- For more info:
 - M. Unser, "Splines: A Perfect Fit for Signal and Image Processing," *IEEE Signal Processing Magazine*, vol. 16, no. 6, pp. 22-38, November 1999.
 - M. Unser, T. Blu, "Cardinal Exponential Splines: Part I—Theory and Filtering Algorithms," *IEEE Trans. Signal Processing*, vol. 53, no. 4, pp. 1425-1438, April 2005.
 - M. Unser, "Cardinal Exponential Splines: Part II—Think Analog, Act Digital," *IEEE Trans. Signal Processing*, vol. 53, no. 4, pp. 1425-1438, April 2005.
- Preprints and demos: <http://bigwww.epfl.ch/>

More to come ...

- Variational properties: “Tikhonov” splines
- Unified formulation of stochastic signal processing
 - Hybrid Wiener filter
 - Fractals
- New type of exponential-preserving wavelets and multiresolution analysis
- Multi-dimensional extensions: polyharmonic splines, vector-splines

Think analog, act digital

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