

# Foundation of Digital Signal Processing: Signal Spaces, System Representation, and Quantization Effects

Holger Boche

Lehrstuhl für Theoretische Informationstechnik  
Technische Universität München

Strobl11

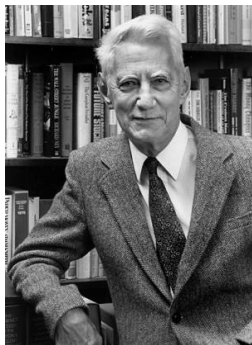
June 16, 2011

# Outline

- ➊ Introduction
- ➋ General Sampling Series
- ➌ Quantization and Thresholding
- ➍ Realizations of Band-Pass Type Systems for  $\mathcal{B}_{\pi}^{\infty}$
- ➎ Conclusion

# Claude E. Shannon (1916-2001)

## Shannon – Founder of the Information Age



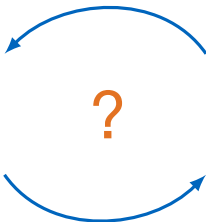
From: [IEEE Information Theory Society]

Today's “digital world” is based on his theoretical work!

# Analog World Versus Digital World



Analog world



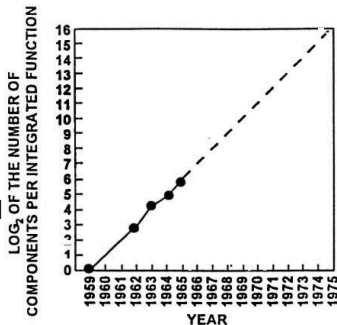
Digital world

# Moore's Law

## Moore's Law

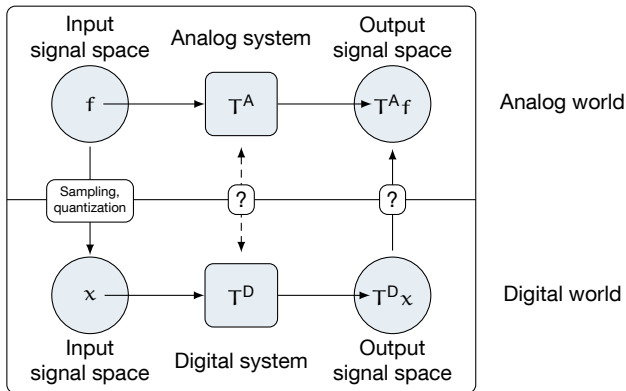
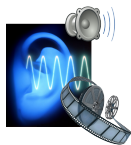
The complexity of electronic circuits, measured in the number of transistors on a chip, doubles approximately every two years.

- It was postulated in 1965 by Gordon Moore.
- Based on the development between 1959 and 1965, he originally stated a doubling every year. (Corrected in 1975 to the current two year statement.)
- This remarkable progress in technology could be observed over the last 50 years.
- The miniaturization cannot be continued forever due to physical limitations.



From: G. Moore, "Cramming More Components onto Integrated Circuits," Electronics Magazine, 38, 1965.

# Analog World Versus Digital World



# Richard Feynman

“Richard Feynman was one of the 20th century’s most influential physicists . . . ” [The Observer, Sunday 15 May 2011]



He discusses the problem of transmitting a function of time and writes in this context:

“Consideration of such a problem will bring us on to consider the famous *Sampling Theorem*, another baby of Claude Shannon.”

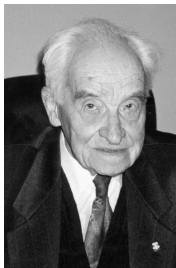


R. P. Feynman, *Feynman Lectures on Computation*, A. J. Hey and R. W. Allen, Eds. Penguin Books, 1999.

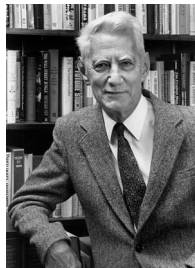
# Whittaker-Kotel'nikov-Shannon Sampling Series



Edmund T. Whittaker



Vladimir A. Kotel'nikov



Claude E. Shannon



Herbert Raabe

$$\sum_{k=-\infty}^{\infty} f(k) \frac{\sin(\pi(t-k))}{\pi(t-k)}$$



Kinnosuke Ogura



# Outline

- 1 Introduction
- 2 General Sampling Series  
Signal Reconstruction  
System Approximation
- 3 Quantization and Thresholding
- 4 Realizations of Band-Pass Type Systems for  $\mathcal{B}_{\pi}^{\infty}$
- 5 Conclusion

# Signal Reconstruction

We analyze the local and global convergence behavior of the sampling series

$$\sum_{k=-\infty}^{\infty} f(t_k) \phi_k(t).$$

for the **Paley-Wiener space**  $\mathcal{PW}_{\pi}^1$ .

- $\phi_k, k \in \mathbb{Z}$ , are certain reconstruction functions
- $\{t_k\}_{k \in \mathbb{Z}}$  is the sequence of real sampling points

We assume that  $t_0 = 0$ , and

$$\dots < t_{-N} < \dots < t_{-1} < t_0 < t_1 < \dots < t_N < \dots$$

- The sampling points  $\{t_k\}_{k \in \mathbb{Z}}$  are the zeros of **sine-type functions**.

# Signal Spaces

- $\mathcal{B}_\sigma$  is the set of all entire functions  $f$  with the property that for all  $\epsilon > 0$  there exists a constant  $C(\epsilon)$  with  $|f(z)| \leq C(\epsilon) \exp((\sigma + \epsilon)|z|)$  for all  $z \in \mathbb{C}$ .

## Definition (Bernstein Space)

The Bernstein space  $\mathcal{B}_\sigma^p$  consists of all signals in  $\mathcal{B}_\sigma$ , whose restriction to the real line is in  $L^p(\mathbb{R})$ ,  $1 \leq p \leq \infty$ .

$\mathcal{B}_{\sigma,0}^\infty$  denotes the set of all signals in  $f \in \mathcal{B}_\sigma^\infty$  that satisfy  $\lim_{|t| \rightarrow \infty} f(t) = 0$ .

# Signal Spaces

## Definition (Paley-Wiener Space)

For  $1 \leq p \leq \infty$  we denote by  $\mathcal{PW}_\sigma^p$  the Paley-Wiener space of functions  $f$  with a representation  $f(z) = \frac{1}{2\pi} \int_{-\sigma}^{\sigma} g(\omega) e^{iz\omega} d\omega$ ,  $z \in \mathbb{C}$ , for some  $g \in L^p[-\sigma, \sigma]$ .

The norm for  $\mathcal{PW}_\sigma^p$  is given by  $\|f\|_{\mathcal{PW}_\sigma^p} = \left( \frac{1}{2\pi} \int_{-\sigma}^{\sigma} |\hat{f}(\omega)|^p d\omega \right)^{1/p}$ .

- We focus on  $\mathcal{PW}_\pi^1$ .
- $\mathcal{PW}_\pi^p \supset \mathcal{PW}_\pi^s$  for  $1 \leq p < s \leq \infty$ ;  $\|f\|_\infty \leq \|f\|_{\mathcal{PW}_\pi^1}$ .
- $\mathcal{PW}_\pi^2$  is the space of bandlimited function with finite energy.

# Operational Meaning of Signal Spaces

Motivated by Feichtinger's axiomatic approach to characterize function spaces with desirable properties.

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Motivated by Feichtinger's axiomatic approach to characterize function spaces with desirable properties.

$\mathcal{PW}_\pi^1$  and the convergence of wide-sense stationary (WSS) stochastic processes  $X$

- Mean-square error:  $\mathbb{E} \left| X(t) - \sum_{k=-N}^N X(t_k) \phi_k(t) \right|^2$

Let  $T > 0$ . We have

$$\lim_{N \rightarrow \infty} \max_{t \in [-T, T]} \left| f(t) - \sum_{k=-N}^N f(t_k) \phi_k(t) \right| = 0$$

for all  $f \in \mathcal{PW}_\pi^1$ , if and only if

$$\lim_{N \rightarrow \infty} \max_{t \in [-T, T]} \mathbb{E} \left| X(t) - \sum_{k=-N}^N X(t_k) \phi_k(t) \right|^2 = 0$$

for an important subclass bandlimited wide-sense stationary processes (l-processes).

# Sine-Type Functions

## Definition

An entire function  $f$  of exponential type  $\pi$  is said to be of sine type if

- (i) the zeros of  $f$  are separated, and
- (ii) there exist positive constants  $A$ ,  $B$ , and  $H$  such that

$$A e^{\pi|y|} \leq |f(x + iy)| \leq B e^{\pi|y|}$$

whenever  $x$  and  $y$  are real and  $|y| \geq H$ .

## Example

$\sin(\pi z)$  is a function of sine type and its zeros are  $t_k = k$ ,  $k \in \mathbb{Z}$ .

# Zeros of Sine Type Functions and Complete Interpolating Sequences

## Definition (Complete Interpolating Sequence)

We say that  $\{t_k\}_{k \in \mathbb{Z}}$  is a complete interpolating sequence for  $\mathcal{PW}_\pi^2$  and coefficient space  $\ell^2$  if the interpolation problem  $f(t_k) = c_k$ ,  $k \in \mathbb{Z}$ , has exactly one solution  $f \in \mathcal{PW}_\pi^2$  for every sequence  $\{c_k\}_{k \in \mathbb{Z}} \in \ell^2$ .

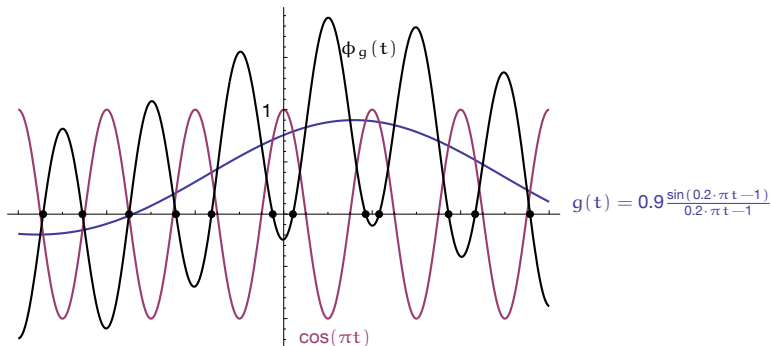
## Lemma

*If  $\{t_k\}_{k \in \mathbb{Z}}$  is the set of zeros of a function of sine type, then the system  $\{e^{i\omega t_k}\}_{k \in \mathbb{Z}}$  is a Riesz basis for  $L^2[-\pi, \pi]$ , and  $\{t_k\}_{k \in \mathbb{Z}}$  is a complete interpolating sequence for  $\mathcal{PW}_\pi^2$ .*



# Construction of Sampling Patterns

- real-valued signal  $g \in \mathcal{PW}_\pi^1$ ,  $\|g\|_{\mathcal{PW}_\pi^1} < 1$   
 $\Rightarrow \phi_g(t) = g(t) - \cos(\pi t)$  is a function of sine type.
- The zeros  $\{t_k\}_{k \in \mathbb{Z}}$  of  $\phi_g$  are all real, because we assumed that  $g$  is real-valued and  $\|g\|_{\mathcal{PW}_\pi^1} < 1$ .  
 $\rightarrow$  method to construct arbitrarily many sampling patterns  $\{t_k\}_{k \in \mathbb{Z}}$ .



# Reconstruction Functions

If  $\{t_k\}_{k \in \mathbb{Z}}$  are the zeros of a function of sine type, then the product

$$\phi(z) = z \lim_{N \rightarrow \infty} \prod_{\substack{|k| \leq N \\ k \neq 0}} (1 - z/t_k)$$

converges uniformly on  $|z| \leq R$  for all  $R < \infty$  and  $\phi$  is an entire function of exponential type  $\pi$ .

$$\phi_k(t) = \frac{\phi(t)}{\phi'(t_k)(t - t_k)}$$

is the unique function in  $\mathcal{PW}_\pi^2$  that solves the interpolation problem

$$\phi_k(t_l) = \begin{cases} 1 & l = k \\ 0 & l \neq k \end{cases}$$

and  $\{\phi_k\}_{k \in \mathbb{Z}}$  is a Riesz basis for  $\mathcal{PW}_\pi^2$ .

# Local Uniform Convergence

## Theorem

Let  $\phi$  be a function of sine type, whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then we have

$$\lim_{N \rightarrow \infty} \max_{t \in [-T, T]} \left| f(t) - \sum_{k=-N}^N f(t_k) \phi_k(t) \right| = 0$$

for all  $T > 0$  and all  $f \in \mathcal{PW}_{\pi}^1$ .

- The sampling series is locally uniformly convergent.

## Remark

- The same result holds even for  $f \in \mathcal{B}_{\pi, 0}^{\infty}$ .

# Brown's Theorem

## Theorem (Brown)

*For all  $f \in \mathcal{PW}_\pi^1$  and  $T > 0$  fixed we have*

$$\lim_{N \rightarrow \infty} \max_{t \in [-T, T]} \left| f(t) - \sum_{k=-N}^N f(k) \frac{\sin(\pi(t-k))}{\pi(t-k)} \right| = 0.$$

The Shannon sampling series is uniformly convergent on all compact subsets of  $\mathbb{R}$  for the space  $\mathcal{PW}_\pi^1$ .

# Global Uniform Convergence

## Theorem

Let  $\phi$  be a function of sine type, whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then, for all  $0 < \beta < 1$  and all  $f \in \mathcal{PW}_{\beta\pi}^1$ , we have

$$\lim_{N \rightarrow \infty} \max_{t \in \mathbb{R}} \left| f(t) - \sum_{k=-N}^N f(t_k) \phi_k(t) \right| = 0.$$

- If oversampling is used the sampling series is globally uniformly convergent.

# Local Convergence with Oversampling

- For  $\mathcal{B}_{\beta\pi}^\infty$ ,  $0 < \beta < 1$ , we have local uniform convergence.

## Theorem

*Let  $\phi$  be a function of sine type, whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then, for all  $T > 0$ ,  $0 < \beta < 1$ , and all  $f \in \mathcal{B}_{\beta\pi}^\infty$ , we have*

$$\lim_{N \rightarrow \infty} \max_{t \in [-T, T]} \left| f(t) - \sum_{k=-N}^N f(t_k) \phi_k(t) \right| = 0.$$

# Conjecture 1.1

## Conjecture (Local Convergence Behavior for Complete Interpolating Sequences)

*There exist a complete interpolating sequence  $\{t_k^1\}_{k \in \mathbb{Z}}$ ,  $f_1 \in \mathcal{PW}_\pi^1$ , and  $t_1 \in \mathbb{R}$ , such that*

$$\limsup_{N \rightarrow \infty} \left| f_1(t_1) - \sum_{k=-N}^N f_1(t_k) \phi_k^1(t_1) \right| = \infty.$$

## Remark

If this conjecture is true, it shows that the zero sequences of sine-type functions have very nice properties.

# Zeros of Sine Type Functions and Complete Interpolating Sequences

## Theorem (Avdonin and Joó)

*If  $\{t_k\}_{k \in \mathbb{Z}} \subset \mathbb{R}$  is a complete interpolating sequence then there exists  $d \in (0, 1/4)$  and a sine-type function with zeros  $\{\mu_k\}_{k \in \mathbb{Z}}$  such that*

$$d(t_{k-1} - t_k) \leq \mu_k - t_k \leq d(t_{k+1} - t_k)$$

*for all  $k \in \mathbb{Z}$ .*



# Global Convergence Behavior without Oversampling

- Two positive results for  $\mathcal{PW}_\pi^1$ :
  - 1) local uniform convergence when no oversampling is used,
  - 2) global uniform convergence when oversampling is used.
- Global convergence behavior without oversampling?
- Recent result: For  $\mathcal{PW}_\pi^1$  and a large class of reconstruction processes a globally bounded signal reconstruction is impossible if the samples are taken equidistantly at Nyquist rate.
- Non-equidistant sampling  $\rightarrow$  additional degree of freedom, which may help to improve the convergence behavior.

# A Subclass of the Functions of Sine Type

## Definition

By  $\mathcal{S}$  we denote the set of all entire functions  $\phi$  with separated real zeros  $\{t_k\}_{k \in \mathbb{Z}}$  that have a representation as Fourier-Stieltjes integral in the form

$$\phi(t) = \frac{1}{2\pi} \int_{-\pi}^{\pi} e^{i\omega t} d\mu(\omega),$$

where  $\mu(\omega)$  is a real function of bounded variation on the interval  $[-\pi, \pi]$  and has a jump discontinuity at each endpoint.

- $\mathcal{S}$  is a subclass of the functions of sine type.

# Global Convergence Behavior without Oversampling

## Theorem

*Let  $\phi$  be a function of sine type in  $\mathcal{S}$ , whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then there exists a signal  $f_1 \in \mathcal{PW}_\pi^1$  such that*

$$\limsup_{N \rightarrow \infty} \max_{t \in \mathbb{R}} \left| f_1(t) - \sum_{k=-N}^N f_1(t_k) \phi_k(t) \right| = \infty.$$

# Global Convergence Behavior of the Shannon Sampling Series without Oversampling

A special case of the previous theorem concerns the global convergence behavior of the Shannon sampling series:

## Corollary

*There exists a signal  $f_1 \in \mathcal{PW}_\pi^1$  such that*

$$\limsup_{N \rightarrow \infty} \max_{t \in \mathbb{R}} \left| f_1(t) - \sum_{k=-N}^N f_1(k) \frac{\sin(\pi(t-k))}{\pi(t-k)} \right| = \infty.$$

## Conjecture 1.2

### Conjecture (Global Convergence Behavior for Complete Interpolating Sequences)

*For each complete interpolating sequences  $\{t_k\}_{k \in \mathbb{Z}}$  there exists a  $f_1 \in \mathcal{PW}_\pi^1$  such that*

$$\limsup_{N \rightarrow \infty} \left( \sup_{t \in \mathbb{R}} \left| f_1(t) - \sum_{k=-N}^N f_1(t_k) \phi_k(t) \right| \right) = \infty.$$

## Definition

We call a bandlimited wide-sense stationary process  $X$  **I-process** if its correlation function  $R_X$  has the representation

$$R_X(\tau) = \frac{1}{2\pi} \int_{-\pi}^{\pi} S_X(\omega) e^{i\omega\tau} d\omega,$$

for some non-negative  $S_X \in L^1[-\pi, \pi]$ .

# Stochastic Processes: Local Convergence Behavior

## Theorem

*Let  $\phi$  be a function of sine type, whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then, for all I-processes  $X$  and all  $T > 0$ , we have*

$$\lim_{N \rightarrow \infty} \max_{t \in [-T, T]} \mathbb{E} \left| X(t) - \sum_{k=-N}^N X(t_k) \phi_k(t) \right|^2 = 0.$$

- We have a good local convergence behavior for I-processes.

# Stochastic Processes:

## Global Convergence Behavior without Oversampling

- There exist I-processes such that the global mean-square approximation error increases unboundedly.

### Theorem

*Let  $\phi$  be a function of sine type in  $\mathcal{S}$ , whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then there exists an I-process  $X_1$  such that*

$$\limsup_{N \rightarrow \infty} \sup_{t \in \mathbb{R}} \mathbb{E} \left| X_1(t) - \sum_{k=-N}^N X_1(t_k) \phi_k(t) \right|^2 = \infty.$$



# Stochastic Processes:

## Global Convergence Behavior with Oversampling

- Similar to the deterministic case, oversampling improves the global convergence behavior of the series for I-processes.

### Theorem

*Let  $\phi$  be a function of sine type, whose zeros  $\{t_k\}_{k \in \mathbb{Z}}$  are all real. Then, for all  $0 < \beta < 1$  and all I-processes  $X$ , whose power spectral density  $S_X(\omega)$  is supported in  $[-\beta\pi, \beta\pi]$ , we have*

$$\sup_{N \in \mathbb{N}} \sup_{t \in \mathbb{R}} \mathbb{E} \left| \sum_{k=-N}^N X(t_k) \phi_k(t) \right|^2 < \infty.$$

# Idea of Sampling-Based Signal Processing

- In many applications the task is to reconstruct some transformation  $Tf$  of  $f \in \mathcal{PW}_\pi^1$  and not  $f$  itself.

Key idea of sampling-based signal processing:

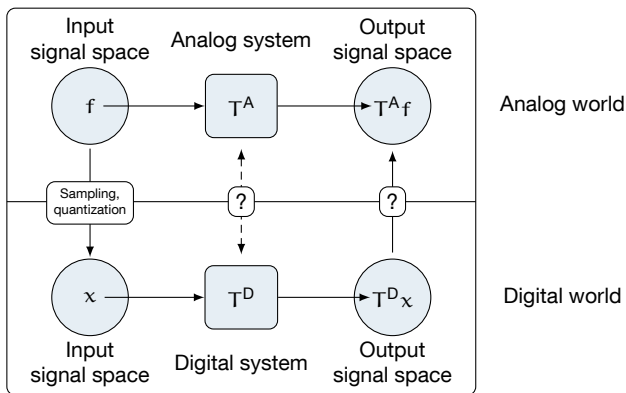
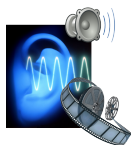
- Not the whole signal is used to calculate some transformation of the signal, but only the samples of the signal.  
→ Calculate  $Tf$  from the samples of  $f$
- Corresponds to the natural situation in digital signal processing, where only the samples of the signal are available.

The question:

- Is it always possible to calculate  $Tf$  from the samples of  $f$ ?

Sampling-based signal processing should be potentially possible because  $f$ , as a bandlimited signal, is uniquely determined by its samples.

# Analog World Versus Digital World



# Stable Linear Time Invariant Systems

A linear system  $T : \mathcal{PW}_\pi^1 \rightarrow \mathcal{PW}_\pi^1$  is called **stable linear time invariant (LTI)** system if:

- $T$  is **bounded**, i.e.,  $\|T\| = \sup_{\|f\|_{\mathcal{PW}_\pi^1} \leq 1} \|Tf\|_{\mathcal{PW}_\pi^1} < \infty$  and
- $T$  is **time invariant**, i.e.,  $(Tf(\cdot - \alpha))(t) = (Tf)(t - \alpha)$  for all  $f \in \mathcal{PW}_\pi^1$  and  $t, \alpha \in \mathbb{R}$ .

The **Hilbert transform**  $H$  and the **low-pass filter** are stable LTI systems.

## Example (Hilbert transform)

The Hilbert transform  $\tilde{f}$  of a signal  $f \in \mathcal{PW}_\pi^1$  is defined by

$$\tilde{f}(t) = (Hf)(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} -i \operatorname{sgn}(\omega) \hat{f}(\omega) e^{i\omega t} d\omega,$$

where  $\operatorname{sgn}$  denotes the signum function.

# Representation of Stable LTI Systems

- For every stable LTI system  $T : \mathcal{PW}_\pi^1 \rightarrow \mathcal{PW}_\pi^1$  there is exactly one function  $\hat{h}_T \in L^\infty[-\pi, \pi]$  such that

$$(Tf)(t) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \hat{h}_T(\omega) \hat{f}(\omega) e^{i\omega t} d\omega$$

for all  $f \in \mathcal{PW}_\pi^1$ , and the integral is absolutely convergent.

- Every  $\hat{h}_T \in L^\infty[-\pi, \pi]$  defines a stable LTI system  $T : \mathcal{PW}_\pi^1 \rightarrow \mathcal{PW}_\pi^1$ .

The operator norm  $\|T\| := \sup_{\|f\|_{\mathcal{PW}_\pi^1} \leq 1} \|Tf\|_{\mathcal{PW}_\pi^1}$  is given by  $\|T\| = \|\hat{h}_T\|_\infty$ .

# System Approximation

System approximation process:

$$\sum_{k=-N}^N f(t_k)(T\phi_k)(t)$$

- $T : \mathcal{PW}_{\pi}^1 \rightarrow \mathcal{PW}_{\pi}^1$  is a stable LTI system.
- $\phi_k \in \mathcal{PW}_{\pi}^2$ ,  $k \in \mathbb{Z}$ , are reconstruction functions.
- $f$  is a signal in  $\mathcal{PW}_{\pi}^1$ .

# System Approximation

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- $\phi_k \in \mathcal{PW}_{\pi}^2$ ,  $k \in \mathbb{Z}$ , are reconstruction functions.
- $f$  is a signal in  $\mathcal{PW}_{\pi}^1$ .

## Theorem

*Let  $\{t_k\}_{k \in \mathbb{Z}} \subset \mathbb{R}$  be a complete interpolating sequence for  $\mathcal{PW}_{\pi}^2$  and  $\phi_k$ ,  $k \in \mathbb{Z}$ , the corresponding reconstruction functions. Then, for all  $t \in \mathbb{R}$  there exists a stable LTI system  $T_1$  with continuous  $\hat{h}_{T_1}$  and a signal  $f_1 \in \mathcal{PW}_{\pi}^1$  such that*

$$\limsup_{N \rightarrow \infty} \left| (T_1 f_1)(t) - \sum_{k=-N}^N f_1(t_k)(T_1 \phi_k)(t) \right| = \infty.$$

# Conjecture 2.1

## Conjecture

*The divergence remains even if oversampling is applied:*

*Let  $\{t_k\}_{k \in \mathbb{Z}} \subset \mathbb{R}$  be a complete interpolating sequence and  $0 < \beta < 1$ . Then, for all  $t \in \mathbb{R}$  there exists a stable LTI system  $T_1$  and a signal  $f_1 \in \mathcal{PW}_{\beta\pi}^1$  such that*

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If this conjecture is true, it implies a no go result for sampling based signal processing.

# General Sampling Functionals

- Sampling of the signal  $f$  corresponds to a point evaluation of  $f$  at the sampling points  $\{t_k\}_{k \in \mathbb{Z}}$ .
- It is also possible to consider **more general linear functionals**  $c_k : \mathcal{PW}_\pi^1 \rightarrow \mathbb{C}$ ,  $k \in \mathbb{Z}$ .
- For example, functionals that also take the signal values in the proximity of the sampling points into account.

New approximation process:

$$\sum_{k=-N}^N c_k(f) (T\phi_k)(t).$$

- In the classical sampling approach the functionals are given by  $c_k(f) = f(t_k)$ ,  $k \in \mathbb{Z}$ .

## Conjecture 2.2

For approximation processes that use the general evaluation functionals we have the following conjecture.

### Conjecture

*Let  $\sigma < \pi$ . There exists a sequence of continuous linear functionals  $c_k$ ,  $k \in \mathbb{Z}$ , on  $\mathcal{PW}_\pi^1$  such that for all stable LTI systems  $T$  and all  $f \in \mathcal{PW}_\sigma^1$  we have*

$$\lim_{N \rightarrow \infty} \sup_{t \in \mathbb{R}} \left| (Tf)(t) - \sum_{k=-N}^N c_k(f) (T\phi_k)(t) \right| = 0.$$

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If this conjecture is true, it shows that for all stable LTI systems a digital implementation is possible using general sampling functionals.

- It would be interesting to find suitable functionals.

# Outline

## ① Introduction

## ② General Sampling Series

## ③ Quantization and Thresholding

Motivation

Threshold and Quantization Operator

Signal Reconstruction

System Approximation

## ④ Realizations of Band-Pass Type Systems for $\mathcal{B}_{\pi}^{\infty}$

## ⑤ Conclusion

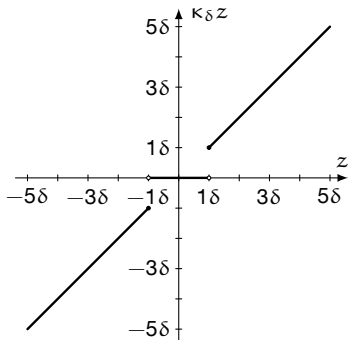
# Signal Processing Under Quantization and Thresholding

- The principle of digital signal processing relies on the fact that certain **bandlimited signals** can be **perfectly reconstructed from their samples**.
- **Reconstruction** of the signal:  $\{f(k)\}_{k \in \mathbb{Z}} \rightarrow f$
- **Approximation** of a transformation:  $\{f(k)\}_{k \in \mathbb{Z}} \rightarrow Tf$
- Perfect reconstruction only possible if the sample values are known exactly.
- Not given in practical applications, because **samples are disturbed** (quantizers with limited resolution, thresholding effects).

# The Threshold Operator $\Theta_\delta$

- The **threshold operator**  $\Theta_\delta$  sets all signal values, whose absolute value is smaller than some **threshold**  $\delta > 0$  to zero.
- For continuous functions  $f : \mathbb{R} \rightarrow \mathbb{C}$ :

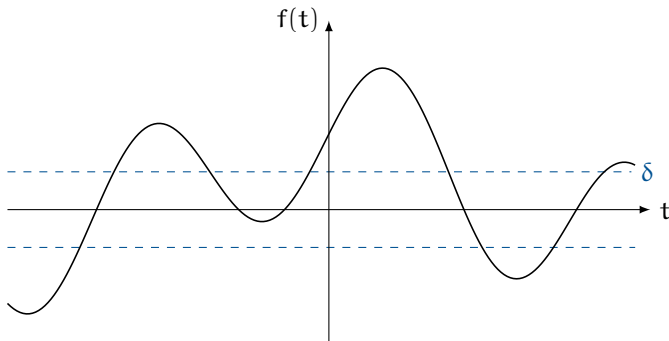
$$(\Theta_\delta f)(t) = \kappa_\delta f(t), \quad t \in \mathbb{R}, \quad \text{where} \quad \kappa_\delta z = \begin{cases} z & |z| \geq \delta \\ 0 & |z| < \delta \end{cases}$$



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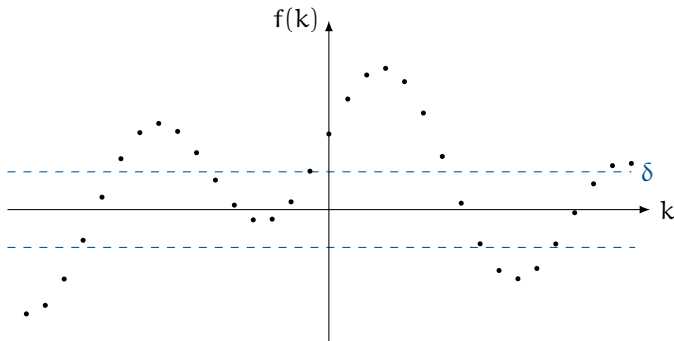




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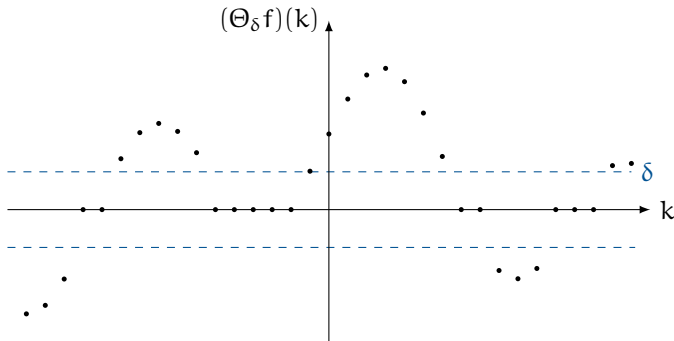
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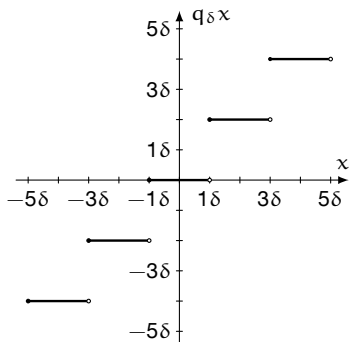
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# The Quantization Operator $\Upsilon_\delta$

- $2\delta$  is the quantization step size
- For continuous functions  $f : \mathbb{R} \rightarrow \mathbb{C}$ :

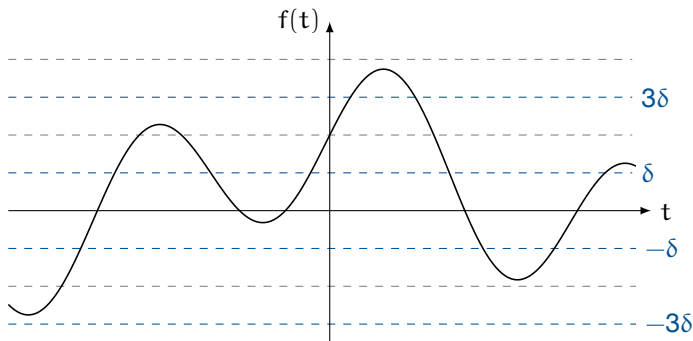
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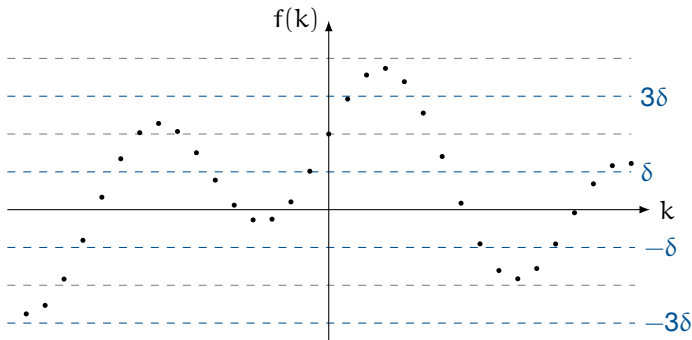
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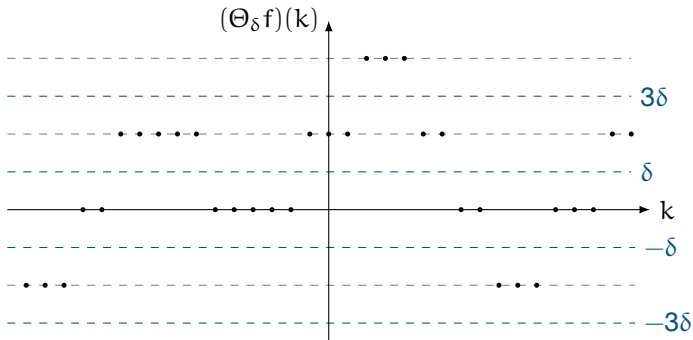
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# The Reconstruction Process $A_\delta$

- The threshold operator is applied on the samples  $\{f(k)\}_{k \in \mathbb{Z}}$  of signals  $f \in \mathcal{PW}_\pi^1$ .
- The resulting samples  $\{(\Theta_\delta f)(k)\}_{k \in \mathbb{Z}}$  are used to build an approximation

$$(A_\delta f)(t) := \sum_{k=-\infty}^{\infty} (\Theta_\delta f)(k) \frac{\sin(\pi(t-k))}{\pi(t-k)} = \sum_{\substack{k=-\infty \\ |f(k)| \geq \delta}}^{\infty} f(k) \frac{\sin(\pi(t-k))}{\pi(t-k)}$$

of the original signal  $f$ .

- We have  $\lim_{|t| \rightarrow \infty} f(t) = 0$  (Riemann-Lebesgue lemma)  
 $\Rightarrow$  the series has only finitely many summands  
 $\Rightarrow A_\delta f \in \mathcal{PW}_\pi^2 \subset \mathcal{PW}_\pi^1$ .

Since the series uses all “important” samples of the signal, one could expect  $A_\delta f$  to have an approximation behavior similar to the Shannon sampling series.

# Properties of the Reconstruction Process $A_\delta$

- ❶ For every  $\delta > 0$ ,  $A_\delta$  is a non-linear operator.
- ❷ For every  $\delta > 0$ , the operator  $A_\delta : (\mathcal{PW}_\pi^1, \|\cdot\|_{\mathcal{PW}_\pi^1}) \rightarrow (\mathcal{PW}_\pi^2, \|\cdot\|_{\mathcal{PW}_\pi^2})$  is discontinuous.
- ❸ For some  $f \in \mathcal{PW}_\pi^1$ , the operator  $A_\delta$  is also discontinuous with respect to  $\delta$ .

The non-linearity of the threshold operator makes the analysis difficult.



# Approximation of Stable LTI Systems

- In many applications the task is to reconstruct some transformation  $Tf$  of  $f \in \mathcal{PW}_\pi^1$  and not  $f$  itself.
- The goal is to approximate the desired transformation  $Tf$  of a signal  $f$  by an approximation process, which uses only the samples of the signal that are disturbed by the threshold operator.

# System Approximation under Thresholding

- If the samples  $\{f(k)\}_{k \in \mathbb{Z}}$  are known perfectly we can use

$$\sum_{k=-N}^N f(k) T(\text{sinc}(\cdot - k))(t) = \sum_{k=-N}^N f(k) h_T(t - k)$$

to obtain an approximation of  $Tf$ .

- Here: samples are disturbed.  $\rightarrow$  Approximate  $Tf$  by

$$(T_\delta f)(t) := (TA_\delta f)(t) = \sum_{k=-\infty}^{\infty} (\Theta_\delta f)(k) h_T(t - k)$$

- Goal: small approximation error  
Since

$$|(T_\delta f)(t) - (Tf)(t)| \leq |(T_\delta f)(t)| + \|T\| \|f\|_{\mathcal{PW}_\pi^1}$$

it is interesting how large  $\sup_{\|f\|_{\mathcal{PW}_\pi^1} \leq 1} |(T_\delta f)(t)|$  can get.

# Pointwise Stability

- The following theorem gives a necessary and sufficient condition for  $\sup_{\|f\|_{\mathcal{PW}_\pi^1} \leq 1} |(T_\delta f)(t)|$  to be finite.

## Theorem

*Let  $T$  be a stable LTI system,  $0 < \delta < 1/3$ , and  $t \in \mathbb{R}$ . Then we have*

$$\sup_{\|f\|_{\mathcal{PW}_\pi^1} \leq 1} |(T_\delta f)(t)| < \infty$$

*if and only if*

$$\sum_{k=-\infty}^{\infty} |h_T(t-k)| < \infty. \quad (*)$$

- Note that (\*) is nothing else than the BIBO stability condition for discrete-time systems.

# Pointwise Convergence

## Corollary

Let  $T$  be a stable LTI system,  $0 < \delta < 1/3$ , and  $t \in \mathbb{R}$ . If

$$\sum_{k=-\infty}^{\infty} |h_T(t-k)| < \infty \quad (*)$$

then we have

$$\lim_{\delta \rightarrow 0} \sup_{f \in \mathcal{PW}_{\pi}^1} |(Tf)(t) - (T_{\delta}f)(t)| = 0.$$

- If (\*) is fulfilled, then we have a good pointwise approximation behavior because the approximation error converges to zero as the threshold  $\delta$  goes to zero.

## Example: Ideal Low-Pass Filter

Even for common stable LTI systems like the ideal low-pass filter there are problems because (\*) is not fulfilled.

### Example

$T_L$ : ideal low-pass filter,  $h_{T_L}(t) = \sin(\pi t)/(\pi t)$

$\rightarrow \sum_{k=-\infty}^{\infty} |h_{T_L}(t - k)| = \infty$  for all  $t \in \mathbb{R} \setminus \mathbb{Z}$

For  $t \in \mathbb{R} \setminus \mathbb{Z}$  and  $0 < \delta < 1/3$ ,

$$\sup_{\|f\|_{\mathcal{PW}_{\pi}^1} \leq 1} |(T_{L,\delta} f)(t)| = \sup_{\|f\|_{\mathcal{PW}_{\pi}^1} \leq 1} \left| \sum_{\substack{k=-\infty \\ |f(k)| \geq \delta}}^{\infty} f(k) \frac{\sin(\pi(t-k))}{\pi(t-k)} \right| = \infty.$$

# Global Stability

We can also give a necessary and sufficient condition for the uniform boundedness on the whole real axis.

## Theorem

*Let  $T$  be a stable LTI system and  $0 < \delta < 1/3$ . We have*

$$\sup_{\|f\|_{\mathcal{P}W^1_\pi} \leq 1} \|T_\delta f\|_\infty < \infty$$

*if and only if*

$$\sup_{0 \leq t \leq 1} \sum_{k=-\infty}^{\infty} |h_T(t-k)| < \infty$$

*if and only if*

$$\int_{-\infty}^{\infty} |h_T(\tau)| d\tau < \infty. \quad (**)$$

- Note that (\*\*) is nothing else than the BIBO stability condition for continuous-time systems.

# Global Uniform Convergence

## Corollary

*Let  $T$  be a stable LTI system and  $0 < \delta < 1/3$ . If*

$$\int_{-\infty}^{\infty} |h_T(\tau)| \, d\tau < \infty. \quad (**)$$

*then we have*

$$\lim_{\delta \rightarrow \infty} \sup_{f \in \mathcal{PW}_{\pi}^1} \|Tf - T_{\delta}f\|_{\infty} = 0.$$

- This shows the good global approximation behavior of  $T_{\delta}f$  if  $(**)$  is fulfilled.

# Threshold Tending to Zero

## Theorem

There exists a signal  $f_1 \in \mathcal{PW}_\pi^1$  such that

$$\limsup_{\delta \rightarrow 0} \left| \sum_{\substack{k=-\infty \\ |f_1(k)| \geq \delta}}^{\infty} f_1(k) \frac{\sin(\pi(t-k))}{\pi(t-k)} \right| = \infty$$

for all  $t \in \mathbb{R} \setminus \mathbb{Z}$ .

## Remark

Much more difficult behavior compared to the Shannon sampling series

$$\sum_{k=-N}^N f(k) \frac{\sin(\pi(t-k))}{\pi(t-k)}$$

→ local uniform convergence (Brown's theorem).



# Conjecture 3.1

## Conjecture

*Let  $\{t_k\}_{k \in \mathbb{Z}} \subset \mathbb{R}$  be a complete interpolating sequence. Then, there exists a signal  $f_1 \in \mathcal{PW}_\pi^1$  such that for all  $t \in \mathbb{R} \setminus \{t_k\}_{k \in \mathbb{Z}}$  we have*

$$\limsup_{\delta \rightarrow 0} \left| \sum_{\substack{k=-\infty \\ |f_1(t_k)| \geq \delta}}^{\infty} f_1(t_k) \phi_k(t) \right| = \infty.$$

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## Open Problems:

- Is a stable system implementation under quantization possible if oversampling is applied?
- If Conjecture 2.2 (general sampling functionals) is true, what are the consequences for quantization?

# Outline

- ① Introduction
- ② General Sampling Series
- ③ Quantization and Thresholding
- ④ Realizations of Band-Pass Type Systems for  $\mathcal{B}_{\pi}^{\infty}$ 
  - Introduction
  - Linear and Non-Linear Realizations of Band-Pass Type Systems
  - Frequency Splitting
- ⑤ Conclusion

# Introduction

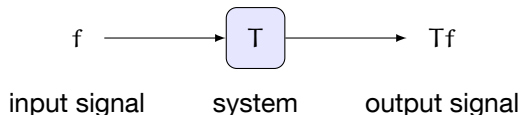
- Filters are a widely used tool in signal processing and system theory.
- Descriptive when the signals are treated in the frequency domain.
- Filters can be categorized according to their passband:
  - low-pass type, high-pass type, **band-pass type**, band-stop type
- All signals that have only frequencies within the passband are not disturbed by the filter.

We use the term **system** instead of filter because filters are often assumed to be linear and time-invariant, and we do not want to restrict our analysis a priori to systems with those properties.

# The Analyzed Systems

- We analyze band-pass type systems operating on bounded bandlimited signals.
- Important in all applications where the peak value of the signal has to be controlled.
- In wireless communication systems the peak value of the transmitted signals has to be bounded by some constant in order that the power amplifier does not overload (clipping of the signal).

# Efficient Band-Pass Type System



Band-pass type systems should be **efficient** in the sense that

**P1)**  $\text{range}(T) \subseteq \mathcal{B}_{[\omega_1, \omega_2]}^\infty$ ,

**P2)**  $Tf = f$  for all  $f \in \mathcal{B}_{[\omega_1, \omega_2]}^\infty$ ,

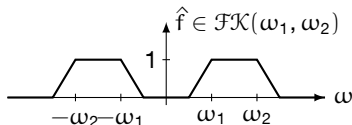
**P3)**  $T : \mathcal{B}_\pi^\infty \rightarrow \mathcal{B}_{[\omega_1, \omega_2]}^\infty$  is bounded.

# An Alternative Definition of Band-Pass Signals

## Definition

For  $0 \leq \omega_1 < \omega_2 < \infty$  let

$$\mathcal{K}(\omega_1, \omega_2) = \left\{ f \in L^1(\mathbb{R}) : \hat{f}(\omega) = 1 \text{ for } |\omega| \in [\omega_1, \omega_2] \right\}$$



## Definition

The space  $\mathcal{B}_{[\omega_1, \omega_2]}^\infty$  consists of all signals  $f \in L^\infty(\mathbb{R})$  that fulfill  $f(t) = \int_{-\infty}^{\infty} f(\tau) K(t - \tau) d\tau$  for all  $t \in \mathbb{R}$  and all  $K \in \mathcal{K}(\omega_1, \omega_2)$ .

Note that we have  $\mathcal{B}_{\omega_2}^\infty = \mathcal{B}_{[0, \omega_2]}^\infty$ , according to this definition.

# No Linear Realization of Efficient Band-Pass Type Systems

## Theorem

*Let  $0 \leq \omega_1 < \omega_2 \leq \pi$  with  $\omega_2 - \omega_1 < \pi$ . There exists no linear operator  $T$  defined on  $\mathcal{B}_\pi^\infty$  with the properties*

- ①  $\text{range}(T) \subseteq \mathcal{B}_{[\omega_1, \omega_2]}^\infty$  *and*
- ②  $Tf = f$  *for all*  $f \in \mathcal{B}_{[\omega_1, \omega_2]}^\infty$
- ③  $T : \mathcal{B}_\pi^\infty \rightarrow \mathcal{B}_{[\omega_1, \omega_2]}^\infty$  *is bounded*

Consequently, a linear realization of efficient band-pass type systems for the signal space  $\mathcal{B}_\pi^\infty$  cannot exist.

- The result is very general, because there are many conceivable realizations.
- For example we do not restrict the systems to be time-invariant.



# Non-Linear Realization of Efficient Band-Pass Type Systems

- Now we drop the requirement that the system is linear.
- The following theorem shows that a **non-linear realization** of efficient band-pass type systems is possible for the space  $\mathcal{B}_\pi^\infty$ .

## Theorem

*Let  $0 \leq \omega_1 < \omega_2 \leq \pi$ . There exists an operator  $T$  defined on  $\mathcal{B}_\pi^\infty$  with the properties*

- ①  $\text{range}(T) \subseteq \mathcal{B}_{[\omega_1, \omega_2]}^\infty$
- ②  $Tf = f$  for all  $f \in \mathcal{B}_{[\omega_1, \omega_2]}^\infty$ , and
- ③  $\|Tf\|_\infty \leq 2\|f\|_\infty$  for all  $f \in \mathcal{B}_\pi^\infty$ .

# Comparing Signals in the Frequency Domain

## Definition

We say that  $f \in \mathcal{B}_\pi^\infty$  and  $g \in \mathcal{B}_\pi^\infty$  agree on the open frequency interval  $(\omega_1, \omega_2)$ ,  $-\infty < \omega_1 < \omega_2 < \infty$ , if

$$\int_{-\infty}^{\infty} f(\tau)h(\tau) \, d\tau = \int_{-\infty}^{\infty} g(\tau)h(\tau) \, d\tau$$

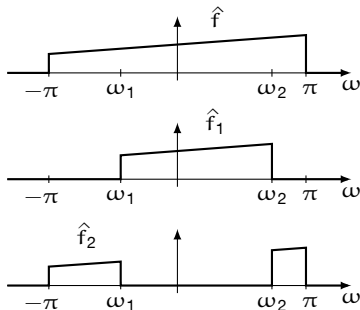
for all  $h \in L^1(\mathbb{R})$  with  $\hat{h}(\omega) = 0$  for all  $\omega \in \mathbb{R} \setminus (\omega_1, \omega_2)$ .

- For  $f \in \mathcal{B}_\pi^2$  this definition is equivalent to the definition that uses the Fourier transform.
- Makes only a statement about what it means that two signals agree on **open sets** of frequencies.

# Frequency Splitting for $\mathcal{B}_\pi^2$

- For  $\mathcal{B}_\pi^2$  it is possible to **split** a signal with respect to its **frequency content**.
- The signal  $f_1$  is given by

$$f_1(t) = \frac{1}{2\pi} \int_{\omega_1}^{\omega_2} \hat{f}(\omega) e^{i\omega t} d\omega.$$



For every signal  $f \in \mathcal{B}_\pi^2$  and every frequency interval  $[\omega_1, \omega_2]$  it is possible to split  $f$  into two signals  $f_1 \in \mathcal{B}_\pi^2$  and  $f_2 \in \mathcal{B}_\pi^2$  such that  $f$  agrees with  $f_1$  on the frequency interval  $[\omega_1, \omega_2]$  and with  $f_2$  on the frequency interval  $[-\pi, \pi] \setminus [\omega_1, \omega_2]$ .

# Frequency Splitting for $\mathcal{B}_\pi^\infty$ ?

## Question

Given  $f \in \mathcal{B}_\pi^\infty$ . Can we find a decomposition  $f = f_1 + f_2$  with  $f_1 \in \mathcal{B}_{\omega_1}^\infty$ ,  $0 < \omega_1 < \pi$ , and  $f_2 \in \mathcal{B}_\pi^\infty$ , such that  $f$  and  $f_1$  agree on the frequency interval  $(-\omega_1, \omega_1)$ ?

If “yes”:

- It would immediately follow that  $f_2$  agrees with the zero function on the frequency interval  $(-\omega_1, \omega_1)$  and that  $f_2 \in \mathcal{B}_{[\omega_1, \pi]}^\infty$ .
- $f_1$  would be the low-pass part of  $f$ , which agrees with  $f$  on the open frequency interval  $(-\omega_1, \omega_1)$ .
- $f_2$  would be the band-pass part of  $f$ , which agrees with  $f$  on the open set of frequencies  $(-\pi, -\omega_1) \cup (\omega_1, \pi)$ .

# No Frequency Splitting for $\mathcal{B}_\pi^\infty$

## Theorem

*Let  $0 < \omega_1 < \pi$ . There exists a signal  $f \in \mathcal{B}_{\pi,0}^\infty$  such that there exists no signal  $f_1 \in \mathcal{B}_{\omega_1}^\infty$  such that*

$$\int_{-\infty}^{\infty} f(\tau)h(\tau) \, d\tau = \int_{-\infty}^{\infty} f_1(\tau)h(\tau) \, d\tau$$

*for all  $h \in \mathcal{B}_{\omega_1}^1$ .*

- A frequency splitting is not possible for signals in  $\mathcal{B}_\pi^\infty$ .
- This signal theoretic result implies that there exists **no filter**—regardless of how complicated the realization is made—that can perform this task.
- Result is also true for the band-pass case.

# Approximate Frequency Splitting

For  $0 < \omega_1 < \infty$ ,  $\delta > 0$ , and  $1 < \kappa < \infty$  let  $\overline{\mathcal{K}}(\omega_1, \delta, \kappa)$  denote the set of all functions  $K \in L^1(\mathbb{R})$ , whose Fourier transform fulfills  $\|\hat{K}\|_\infty \leq \kappa$ ,  $\hat{K}(\omega) = 1$  for  $|\omega| \leq \omega_1$ , and  $\hat{K}(\omega) = 0$  for  $|\omega| > \omega_1 + \delta$ .

For  $K \in \overline{\mathcal{K}}(\omega_1, \delta, \kappa)$  we define the system  $\Psi_K : \mathcal{B}_{\pi,0}^\infty \rightarrow \mathcal{B}_{\omega_1+\delta,0}^\infty$  by

$$(\Psi_K f)(t) = \int_{-\infty}^{\infty} f(\tau) K(t - \tau) \, d\tau.$$

- For signals in  $\mathcal{B}_{\pi}^2$ ,  $\hat{K}$  has the meaning of a transfer function.
- Relaxation of P1).

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- For signals in  $\mathcal{B}_{\pi}^2$ ,  $\hat{K}$  has the meaning of a transfer function.
- Relaxation of P1).

## Theorem

*For all  $0 < \omega_1 < \pi$  and  $1 < \kappa < \infty$  we have*

$$\liminf_{\delta \rightarrow 0} \inf_{K \in \overline{\mathcal{K}}(\omega_1, \delta, \kappa)} \|\Psi_K\| = \infty.$$

# Conclusion

- We analyzed the convergence behavior of sampling series for **signal reconstruction** and **system approximation**.
- It was shown that **quantization** and **thresholding** impair the convergence of the sampling series.
- For the space  $\mathcal{B}_{\pi}^{\infty}$ , a **linear realization of efficient band-pass type systems** and **frequency splitting** are not possible



Happy birthday Hans!