



# **From classical sampling to average sampling in shift-invariant-like subspaces of Hilbert-Schmidt operators**

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# Outline

## (A) Classical sampling theory at a glance

- The classical WKS sampling theorem
- A hilbertian proof: some consequences
- The role of frames in sampling theory

## (B) Sampling in subspaces of Hilbert-Schmidt operators

- Two motivations
- The Weyl and Kohn-Nirenberg transforms for Hilbert-Schmidt operators on  $L^2(\mathbb{R}^d)$
- $\Lambda$ -shift-invariant subspaces  $V_S^2$  in  $\mathcal{HS}(\mathbb{R}^d)$
- The isomorphism  $\mathcal{T}_S$  between  $\ell_N^2(\Lambda)$  and  $V_S^2$
- An expression for the samples
- The sampling result in  $V_S^2$

# Classical sampling theory at a glance

## The WKS sampling theorem

Any function  $f$  in the *Paley-Wiener space*:

$$PW_\pi := \left\{ f \in L^2(\mathbb{R}) \cap C(\mathbb{R}), \quad \text{supp } \hat{f} \subseteq [-\pi, \pi] \right\}$$

i.e., *bandlimited* to the interval  $[-\pi, \pi]$ , can be expressed as

$$f(t) = \sum_{n=-\infty}^{\infty} f(n) \frac{\sin \pi(t-n)}{\pi(t-n)}, \quad t \in \mathbb{R}$$

The series converges in  $L^2(\mathbb{R})$ -sense and also absolutely y uniformly on  $\mathbb{R}$ .

WKS means Whittaker–Kotel'nikov–Shannon

- Any  $f \in PW_\pi$  can be expressed as

$$f(t) = \frac{1}{\sqrt{2\pi}} \int_{-\pi}^{\pi} \widehat{f}(w) e^{itw} dw = \left\langle \widehat{f}, \frac{e^{-itw}}{\sqrt{2\pi}} \right\rangle_{L^2[-\pi, \pi]}, \quad t \in \mathbb{R}$$

In particular,

$$f(n) = \left\langle \widehat{f}, \frac{e^{-inw}}{\sqrt{2\pi}} \right\rangle_{L^2[-\pi, \pi]}, \quad n \in \mathbb{Z}$$

- The sampling period is  $T_s = \frac{2\pi}{2\pi} = 1$
- Observe that it is a Lagrange-type interpolation formula

$$f(t) = \sum_{n=-\infty}^{\infty} f(n) \frac{G(t)}{G'(n)(t-n)}, \quad t \in \mathbb{R}, \quad \text{with } G(t) = \frac{\sin \pi t}{\pi}$$

# Shannon's original statement

**THEOREM 1:** *If a function  $f(t)$  contains no frequencies higher than  $W$  cps, it is completely determined by giving its ordinates at a series of points spaced  $1/2W$  seconds apart.*

$$f(t) = \frac{1}{\sqrt{2\pi}} \int_{-2\pi W}^{2\pi W} \hat{f}(w) e^{itw} dw, \quad t \in \mathbb{R}$$

Here the sampling period is  $T_s = \frac{2\pi}{4\pi W} = \frac{1}{2W}$ , and the sampling formula reads:

$$f(t) = \sum_{n=-\infty}^{\infty} f\left(\frac{n}{2W}\right) \frac{\sin \pi(2Wt - n)}{\pi(2Wt - n)}, \quad t \in \mathbb{R}$$

## A hilbertian easy proof:

1. We expand  $\hat{f}$  following the orthonormal basis  $\{e^{-inw}/\sqrt{2\pi}\}_{n\in\mathbb{Z}}$  for  $L^2[-\pi, \pi]$ :

$$\underbrace{\hat{f}} = \sum_{n=-\infty}^{\infty} \langle \hat{f}, \frac{e^{-inw}}{\sqrt{2\pi}} \rangle_{L^2[-\pi, \pi]} \frac{e^{-inw}}{\sqrt{2\pi}} = \sum_{n=-\infty}^{\infty} \underbrace{f(n)} \frac{e^{-inw}}{\sqrt{2\pi}}$$

2. We apply the inverse Fourier transform  $\mathcal{F}^{-1}$ :

$$f = \sum_{n=-\infty}^{\infty} f(n) \underbrace{\mathcal{F}^{-1}} \left( \frac{e^{-inw}}{\sqrt{2\pi}} \chi_{[-\pi, \pi]}(w) \right) (t) = \sum_{n=-\infty}^{\infty} f(n) \frac{\sin \pi(t-n)}{\pi(t-n)}$$

Convergence in the  $L^2$ -norm sense

Observe that  $\mathcal{F}^{-1} \left( \frac{1}{\sqrt{2\pi}} \chi_{[-\pi, \pi]}(w) \right) (t) = \frac{\sin \pi t}{\pi t}(t)$

## Some consequences and comments:

- Convergence in  $L^2$ -norm implies pointwise convergence, which is also uniform on  $\mathbb{R}$ : the Paley-Wiener space  $PW_\pi$  is a RKHS (reproducing kernel Hilbert space)

$$|f(t)| = \left| \left\langle \hat{f}, \frac{e^{-itw}}{\sqrt{2\pi}} \right\rangle_{L^2[-\pi, \pi]} \right| \leq \|f\|_{L^2(\mathbb{R})}, \quad \forall t \in \mathbb{R}$$

- The unconditional character of an orthonormal basis implies absolute convergence
- The sequence  $\left\{ \frac{\sin \pi(t-n)}{\pi(t-n)} \right\}_{n \in \mathbb{Z}}$  is an orthonormal basis for  $PW_\pi$
- The Parseval identity gives the *conservation of the signal energy*

$$\|f\|^2 = \sum_{n=-\infty}^{\infty} |f(n)|^2, \quad \forall f \in PW_\pi$$



- **Average sampling:** For a fixed function  $\psi \in L^2(\mathbb{R})$  the samples are taken from the LTI system  $\mathcal{A}_\psi$ , i.e.,

$$\mathcal{A}_\psi f(n) = \langle f, \psi(\cdot - n) \rangle = \langle f, T_n \psi \rangle = (f * \tilde{\psi})(n), \quad n \in \mathbb{Z},$$

where  $\tilde{\psi}(t) = \overline{\psi(-t)}$  is the **average function**

- **It includes also pointwise sampling:** For any  $f \in PW_\pi$ ,

$$f(n) = (f * \text{sinc})(n), \quad n \in \mathbb{Z}$$

- Due to the drawbacks of the sine cardinal function, one considers shift-invariant subspaces as

$$V_\varphi^2 = \left\{ f(t) = \sum_{n \in \mathbb{Z}} a_n \varphi(t - n) : \{a_n\} \in \ell^2(\mathbb{Z}) \right\} \subset L^2(\mathbb{R})$$

In particular,  $PW_\pi$  is a shift-invariant subspace with **generator** the **sine cardinal** function, i.e.,  $\varphi(t) = \text{sinc}(t) := \frac{\sin \pi t}{\pi t}$ ,  $t \in \mathbb{R}$

## Claude E. Shannon (1916–2001)



- ▶ B. S. Engineering Mathematics, University of Michigan, 1936  
Ph. D. Mathematics, MIT, 1940
- ▶ Research Mathematician, Bell Labs, 1941–1972; MIT Faculty Member, 1956–1978; Donner Professor of Science 1958
- ▶ Major publication: **A Mathematical Theory of Communication**, Bell System Technical Report, 1948
- ▶ Honorary Degree, Univ. of Michigan, 1961; The National Medal of Science, 1966; The Audio Engineering Society Gold Medal, 1985; The Kyoto Prize, 1985

*The american mathematician, computer scientist, communication engineer, and the founder of the field of **Information Theory**, whose work has laid the foundation for the telecommunication networks that lace the globe*

## Frames in a separable Hilbert space

- The sequence  $\{x_n\}_{n=1}^{\infty}$  is a frame for a separable Hilbert space  $\mathcal{H}$  if there exist two constants  $A, B > 0$  (frame bounds) such that:

$$A\|x\|^2 \leq \sum_{n=1}^{\infty} |\langle x, x_n \rangle|^2 \leq B\|x\|^2, \quad \forall x \in \mathcal{H}$$

- The preframe operator  $T$  is defined by:

$$\ell^2(\mathbb{Z}) \ni \{c_n\}_{n=1}^{\infty} \xrightarrow{T} \sum_{n=1}^{\infty} c_n x_n \in \mathcal{H}$$

It characterizes a frame:

$$\{x_n\}_{n=1}^{\infty} \text{ is a frame for } \mathcal{H} \iff T \text{ is bounded and surjective}$$

► Two important particular cases:

- $T$  is also **unitary**  $\iff \{x_n\}_{n=1}^\infty$  is an **orthonormal basis**
- $T$  is also **injective (isomorphism)**  $\iff \{x_n\}_{n=1}^\infty$  is a **Riesz basis**

► Given a frame  $\{x_n\}_{n=1}^\infty$  there exist frames  $\{y_n\}_{n=1}^\infty$  (**dual frames**) such that:

$$x = \sum_{n=1}^{\infty} \langle x, x_n \rangle y_n = \sum_{n=1}^{\infty} \langle x, y_n \rangle x_n \quad \forall x \in \mathcal{H}$$

► In case of bases, the dual frame is unique:

- For **orthonormal bases**  $\{y_n\}_{n=1}^\infty = \{x_n\}_{n=1}^\infty$
  - For **Riesz bases**  $\{y_n\}_{n=1}^\infty$  is its dual Riesz basis of  $\{x_n\}_{n=1}^\infty$
- For **overcomplete frames**, there exist infinite dual frames  $\{y_n\}_{n=1}^\infty$  satisfying the above equalities

## Summarizing the key points in the WKS theorem proof:

- (a)  $\mathcal{H}$  is a Hilbert space where we are sampling. The samples of any  $f \in \mathcal{H}$ , can be expressed as **frame coefficients**  $f(n) = \langle x, x_n \rangle_{\mathcal{K}}$  in a frame expansion

$$x = \sum_n \langle x, x_n \rangle y_n = \sum_n f(n) y_n \quad \text{in } \mathcal{K},$$

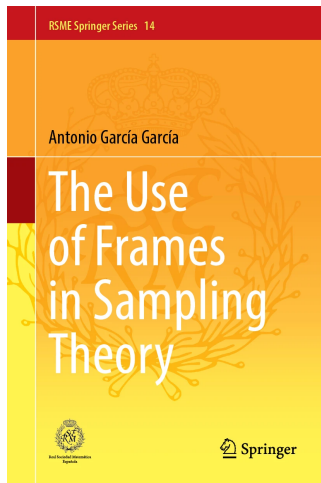
where  $\{x_n\}$  and  $\{y_n\}$  dual frames for an auxiliary Hilbert space  $\mathcal{K}$

- (b) There exists an isomorphism  $\mathcal{T} : \mathcal{K} \rightarrow \mathcal{H}$  relating both spaces, i.e.,  $\mathcal{T}x = f$
- (c) By means of this isomorphism  $\mathcal{T}$ , the **frame expansion in  $\mathcal{K}$**  in the first item yields a **sampling expansion in  $\mathcal{H}$** :

$$f = (\mathcal{T}x) = \sum_n f(n) \mathcal{T}(y_n) \quad \text{in } \mathcal{H}$$

The isomorphism  $\mathcal{T}$  should respect the structure in  $\mathcal{H}$

The above argument, borrowed from the hilbertian proof of the WSK sampling theorem, has been profusely used in my research and, as a consequence...



## Average sampling in shift-invariant subspace in $L^2(\mathbb{R}^d)$

- Consider a **classical shift-invariant subspaces** in  $L^2(\mathbb{R}^d)$  with a **stable set of generators**  $\Phi = \{\varphi_1, \varphi_2, \dots, \varphi_N\}$

$$V_\Phi^2 = \left\{ \sum_{n=1}^N \sum_{\alpha \in \mathbb{Z}^d} c_n(\alpha) \varphi_n(t - \alpha) : \{c_n(\alpha)\}_{\alpha \in \mathbb{Z}^d} \in \ell^2(\mathbb{Z}^d), n = 1, \dots, N \right\}$$

In other words, the sequence  $\{\varphi_n(t - \alpha)\}_{\alpha \in \mathbb{Z}^d; n=1,2,\dots,N}$  is a Riesz basis for  $V_\Phi^2$

- For any  $f \in V_\Phi^2$  we consider **average samples**

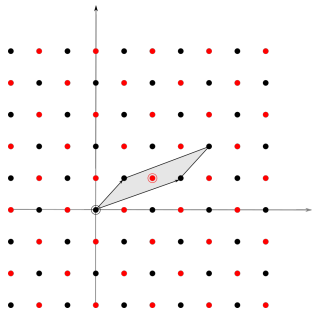
$$\{\langle f, \psi_m(\cdot - \alpha) \rangle\}_{\alpha \in P\mathbb{Z}^d}, \quad m = 1, 2, \dots, M,$$

$\psi_m$  are the **average functions** (not necessarily in  $V_\Phi^2$ ),  $m = 1, 2, \dots, M$

$P$  is a  $d \times d$  matrix with integer entries and  $\det P > 0$

The set  $\Lambda := P\mathbb{Z}^d$  is a **full rank lattice** in  $\mathbb{R}^d$

For instance, in  $\mathbb{Z}^2$ , the lattice  $\Lambda = P\mathbb{Z}^2$  where  $P = \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix}$  with  $\det P = 2$ , is depicted as the black points:



- $P[0, 1]^2$  is the fundamental parallelepiped of  $\Lambda$
- $\mathbb{Z}^2 \cap P[0, 1]^2 = \{(0, 0)^\top, (2, 1)^\top\}$
- $\mathbb{Z}^2 / \Lambda = \{\Lambda, (2, 1)^\top + \Lambda\}$



- Then, under appropriate hypotheses (similar to those in the second part of this talk) there exist  $M(\geq N \det P)$  sampling functions  $S_m \in V_\Phi^2$  such that, for each  $f \in V_\Phi^2$ :

$$f(t) = \sum_{m=1}^M \sum_{\alpha \in P\mathbb{Z}^d} \langle f, \psi_m(\cdot - \alpha) \rangle S_m(t - \alpha), \quad t \in \mathbb{R}^d$$

and the sequence  $\{S_m(t - \alpha)\}_{\alpha \in P\mathbb{Z}^d; m=1,2,\dots,M}$  is a frame for  $V_\Phi^2$

# Sampling in subspaces of Hilbert-Schmidt operators

## A mathematical motivation: Average sampling in shift-invariant-like subspaces of Hilbert-Schmidt operators on $L^2(\mathbb{R}^d)$

► A Hilbert-Schmidt operator  $H : L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d)$  is an integral operator

$$Hf(t) = \int_{\mathbb{R}^d} \kappa_H(t, s) f(s) ds$$

where the kernel  $\kappa_H \in L^2(\mathbb{R}^{2d})$  and  $\|H\|_{\mathcal{HS}} = \|\kappa_H\|_{L^2(\mathbb{R}^{2d})}$

► The *translation of an operator*  $S : L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d)$  by  $z = (x, \omega)$  in the phase space  $\mathbb{R}^d \times \widehat{\mathbb{R}}^d (\simeq \mathbb{R}^{2d})$  is defined by

$$\alpha_z(S) := \pi(z) S \pi(z)^*, \quad z \in \mathbb{R}^d \times \widehat{\mathbb{R}}^d$$

where  $\pi(z)$  denotes the *time-frequency shift* which acts on  $f \in L^2(\mathbb{R}^d)$  as

$$\pi(z)f(t) = e^{2\pi i \omega \cdot t} f(t - x), \quad t \in \mathbb{R}^d$$

► The set of translations  $\{\alpha_z\}_{z \in \mathbb{R}^{2d}}$  is a *unitary representation* of the additive group  $\mathbb{R}^{2d}$  on the Hilbert space  $(\mathcal{HS}(\mathbb{R}^d), \langle \cdot, \cdot \rangle_{\mathcal{HS}})$  of Hilbert-Schmidt operators on  $L^2(\mathbb{R}^d)$

► Let  $\Lambda$  be a **full rank lattice** in  $\mathbb{R}^{2d}$ , i.e.,  $\Lambda = A\mathbb{Z}^{2d}$  where  $A$  is a  $2d \times 2d$  real invertible matrix.

For  $S_n \in \mathcal{HS}(\mathbb{R}^d)$ ,  $n = 1, 2, \dots, N$ , we could consider the (closed) subspace of  $\mathcal{HS}(\mathbb{R}^d)$  given by

$$V_S^2 = \left\{ \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) \alpha_\lambda(S_n) : \{c_n(\lambda)\}_{\lambda \in \Lambda} \in \ell^2(\Lambda), n = 1, 2, \dots, N \right\}$$

in the case that  $\{\alpha_\lambda(S_n)\}_{\lambda \in \Lambda; n=1,2,\dots,N}$  is a **Riesz sequence** for  $\mathcal{HS}(\mathbb{R}^d)$ , i.e., a **Riesz basis** for  $V_S^2$

► We could define, for any  $T \in V_S^2$  its **average samples** at  $\Lambda$  by

$$\langle T, \alpha_\lambda(Q_m) \rangle_{\mathcal{HS}}, \quad \lambda \in \Lambda, \quad m = 1, 2, \dots, M$$

from  $M$  fixed operators  $Q_1, Q_2, \dots, Q_M$  in  $\mathcal{HS}(\mathbb{R}^d)$ , the **average operators** (not necessarily in  $V_S^2$ )

### A mathematical question

Under which hypotheses there exist  $M(\geq N)$  operators  $H_m \in V_S^2$  such that for each  $T \in V_S^2$

$$T = \sum_{m=1}^M \sum_{\lambda \in \Lambda} \langle T, \alpha_\lambda(Q_m) \rangle_{\mathcal{HS}} \alpha_\lambda(H_m) \quad \text{in } \mathcal{HS}\text{-norm}$$

where the sequence  $\{\alpha_\lambda(H_m)\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  is a **frame** for the Hilbert space  $V_S^2$  ?

- ▶ The *adjoint operator* is  $\pi(z)^* = e^{-2\pi i x \cdot \omega} \pi(-z)$  for  $z = (x, \omega) \in \mathbb{R}^{2d}$
- ▶ The *short-time Fourier transform (Gabor transform)*  $V_\psi \varphi$  of  $\varphi$  with window  $\psi$ , both in  $L^2(\mathbb{R}^d)$ , is defined by

$$V_\psi \varphi(z) = \langle \varphi, \pi(z)\psi \rangle_{L^2(\mathbb{R}^d)} = \int_{\mathbb{R}^d} \varphi(t) e^{-2\pi i t \cdot \omega} \overline{\psi(t-x)} dt, \quad z \in \mathbb{R}^{2d}$$

$V_\psi \varphi(z)$  gives time-frequency information on  $\varphi$  at  $z = (x, \omega)$

## A practical question

Why to sample Hilbert-Schmidt operators with precisely the average samples?

## A practical motivation: LTV versus LTI systems

### ► Linear time-invariant (LTI) system

$$y(t) = (Hx)(t) = \int_{-\infty}^{\infty} h(s) x(t-s) ds = \int_{-\infty}^{\infty} \widehat{h}(w) \widehat{x}(w) e^{2\pi i w t} dw$$

### ► Linear time-varying (LTV) system

$$y(t) = (Hx)(t) = \int_{-\infty}^{\infty} h(t,s) x(t-s) ds = \int_{-\infty}^{\infty} \sigma(t,w) \widehat{x}(w) e^{2\pi i w t} dw$$

where

$$\sigma = \mathcal{F}_2 h, \text{ i.e., } \sigma(t,w) = \int_{-\infty}^{\infty} h(t,s) e^{-2\pi i w s} ds$$

Thus, operator  $H$  is a pseudo-differential operator with symbol  $\sigma$

In particular, Hilbert-Schmidt operators model LTV systems:

$$Hf(t) = \int_{-\infty}^{\infty} \kappa_H(t, s) f(s) ds$$

where the integral kernel  $\kappa_H \in L^2(\mathbb{R}^2)$

Besides

$$Hf(t) = \int_{-\infty}^{\infty} \kappa_H(t, s) f(s) ds = \int_{-\infty}^{\infty} \kappa_H(t, t-s) f(t-s) ds$$

is a LTV system where  $h(t, s) = \kappa_H(t, t-s)$

Concerning the average samples:



In *Orthogonal frequency-division multiplexing* (OFDM) the digital information, i.e., a sequence of numbers  $\{c_\lambda\}$ ,  $\lambda$  in the lattice  $\Lambda = a\mathbb{Z}^d \times b\mathbb{Z}^d$  ( $a, b > 0$ ), is used as the coefficients of the input signal  $x(t) = \sum_{\mu \in \Lambda} c_\mu \pi(\mu)g(t)$  of a time-varying system  $H$  producing the output  $y(t) = Hx(t)$ . Then, it is considered the sequence of numbers

$$d_\lambda = \langle y, \pi(\lambda)\tilde{g} \rangle_{L^2(\mathbb{R}^d)} = \sum_{\mu \in \Lambda} c_\mu \langle H\pi(\mu)g, \pi(\lambda)\tilde{g} \rangle_{L^2(\mathbb{R}^d)}, \quad \lambda \in \Lambda,$$

The task: to recover the original data  $\{c_\lambda\}$  from the received data  $\{d_\lambda\}$

The matrix  $A = [a_{\lambda,\mu}]$ , where  $a_{\lambda,\mu} = \langle H\pi(\mu)g, \pi(\lambda)\tilde{g} \rangle_{L^2(\mathbb{R}^d)}$  is the so-called *channel matrix* associated with  $H$  and the functions (windows)  $g, \tilde{g}$  in  $L^2(\mathbb{R}^d)$

The **diagonal channel samples** of  $H$  with respect to  $g, \tilde{g}$  are

$$\langle H\pi(\lambda)g, \pi(\lambda)\tilde{g} \rangle_{L^2(\mathbb{R}^d)}, \quad \lambda \in \Lambda$$

They are also known as:

- ▶ The **lower symbol of the operator  $H$**  with respect  $g, \tilde{g} \in L^2(\mathbb{R}^d)$  and lattice  $\Lambda$  used in time-frequency analysis
- ▶ The samples of the **Berezin transform** of  $H$

$$\mathcal{B}^{g, \tilde{g}} H(z) := \langle H\pi(z)g, \pi(z)\tilde{g} \rangle_{L^2(\mathbb{R}^d)}, \quad z \in \mathbb{R}^{2d}$$

at the lattice  $\Lambda$  used in quantum physics

- ▶ Diagonal channel samples are a particular case of average samples

## Hilbert-Schmidt operators on $L^2(\mathbb{R}^d)$

► For a compact operator  $S$  on  $L^2(\mathbb{R}^d)$  there exist two orthonormal sequences  $\{x_n\}_{n \in \mathbb{N}}$  and  $\{y_n\}_{n \in \mathbb{N}}$  in  $L^2(\mathbb{R}^d)$  and a bounded sequence of positive numbers  $\{s_n(S)\}_{n \in \mathbb{N}}$  (*singular values of  $S$* ) such that

$$S = \sum_{n \in \mathbb{N}} s_n(S) x_n \otimes y_n \quad (\text{SVD})$$

with convergence of the series in the operator norm. The sequence  $\{s_n(S)\}$  consists of the eigenvalues of the operator  $|T| = (S^*S)^{1/2}$

Here,  $x_n \otimes y_n$  denotes the *rank-one operator*

$$(x_n \otimes y_n)(f) = \langle f, y_n \rangle_{L^2} x_n \quad \text{for } f \in L^2(\mathbb{R}^d)$$

► The class of *Hilbert-Schmidt operators* is  $\mathcal{HS}(\mathbb{R}^d) := \mathcal{T}^2$   
 $\mathcal{T}^2$  is the *Schatten 2-class*, i.e., singular values in  $\ell^2(\mathbb{N})$

The space  $\mathcal{HS}(\mathbb{R}^d)$  is a Hilbert space with the inner product

$$\langle S, T \rangle_{\mathcal{HS}} = \text{tr}(ST^*), \quad S, T \in \mathcal{HS}(\mathbb{R}^d)$$

Remind that the trace  $\text{tr}(S) = \sum_{n \in \mathbb{N}} \langle S e_n, e_n \rangle_{L^2}$  is a well-defined bounded linear functional on  $\mathcal{T}^1$ , and independent of the used orthonormal basis  $\{e_n\}_{n \in \mathbb{N}}$  in  $L^2(\mathbb{R}^d)$

► Concerning the norm of  $S$  in  $\mathcal{HS}(\mathbb{R}^d)$  we have

$$\|S\|_{\mathcal{HS}}^2 = \text{tr}(SS^*) = \sum_{n \in \mathbb{N}} \|S^*(e_n)\|_{L^2}^2 = \sum_{n \in \mathbb{N}} \|S(e_n)\|_{L^2}^2 = \sum_{n \in \mathbb{N}} s_n^2(S)$$

► A Hilbert-Schmidt operator  $S \in \mathcal{HS}(\mathbb{R}^d)$  can be seen also as an integral operator on  $L^2(\mathbb{R}^d)$  defined for each  $f \in L^2(\mathbb{R}^d)$  by

$$Sf(t) = \int_{\mathbb{R}^d} \kappa_S(t, x) f(x) dx \quad \text{a.e. } t \in \mathbb{R}^d$$

with **kernel**  $\kappa_S \in L^2(\mathbb{R}^{2d})$ . Besides,  $\langle S, T \rangle_{\mathcal{HS}} = \langle \kappa_S, \kappa_T \rangle_{L^2(\mathbb{R}^{2d})}$  for  $S, T \in \mathcal{HS}(\mathbb{R}^d)$

# The Weyl transform

The Weyl transform is a unitary operator

$$L^2(\mathbb{R}^{2d}) \ni f \longmapsto L_f \in \mathcal{HS}(\mathbb{R}^d)$$

where  $L_f : L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d)$  is the Hilbert-Schmidt operator defined in weak sense by

$$\langle L_f \phi, \psi \rangle_{L^2(\mathbb{R}^d)} = \langle f, W(\psi, \phi) \rangle_{L^2(\mathbb{R}^{2d})}, \quad \phi, \psi \in L^2(\mathbb{R}^d)$$

here

$$W(\psi, \phi)(x, \omega) = \int_{\mathbb{R}^d} \psi\left(x + \frac{t}{2}\right) \overline{\phi\left(x - \frac{t}{2}\right)} e^{-2\pi i \omega \cdot t} dt, \quad (x, \omega) \in \mathbb{R}^{2d},$$

is the *cross-Wigner distribution* of the functions  $\psi, \phi \in L^2(\mathbb{R}^d)$

For each  $S, T \in \mathcal{HS}(\mathbb{R}^d)$  with **Weyl symbols**  $a_S, a_T$  in  $L^2(\mathbb{R}^{2d})$  we have

$$\langle S, T \rangle_{\mathcal{HS}} = \langle a_S, a_T \rangle_{L^2(\mathbb{R}^{2d})}$$

# The Kohn-Nirenberg transform

The Kohn-Nirenberg transform is a unitary operator

$$L^2(\mathbb{R}^{2d}) \ni \sigma \longmapsto K_\sigma \in \mathcal{HS}(\mathbb{R}^d)$$

where  $K_\sigma : L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d)$  is the Hilbert-Schmidt operator defined in weak sense by

$$\langle K_\sigma \phi, \psi \rangle_{L^2(\mathbb{R}^d)} = \langle \sigma, R(\psi, \phi) \rangle_{L^2(\mathbb{R}^{2d})}, \quad \phi, \psi \in L^2(\mathbb{R}^d)$$

here

$$R(\psi, \phi)(x, \omega) = \psi(x) \overline{\widehat{\phi}(\omega)} e^{-2\pi i x \cdot \omega}, \quad (x, \omega) \in \mathbb{R}^{2d},$$

is the *Rihaczek distribution* of the functions  $\psi, \phi \in L^2(\mathbb{R}^d)$

For each  $S, T \in \mathcal{HS}(\mathbb{R}^d)$  with Kohn-Nirenberg symbols  $\sigma_S, \sigma_T$  in  $L^2(\mathbb{R}^{2d})$  we have

$$\langle S, T \rangle_{\mathcal{HS}} = \langle \sigma_S, \sigma_T \rangle_{L^2(\mathbb{R}^{2d})}$$

- There is a transition between Weyl and Kohn-Nirenberg calculus:

$$\sigma_s = Ua_s, \text{ where } \widehat{Ua_s}(\xi, u) = e^{\pi i u \cdot \xi} \widehat{a_s}(\xi, u), (\xi, u) \in \mathbb{R}^{2d}$$

- The Weyl and Kohn-Nirenberg transforms in  $\mathcal{HS}(\mathbb{R}^d)$  respect both the translations in the sense:

### A key property for both transforms

For  $f \in L^2(\mathbb{R}^{2d})$  and  $z \in \mathbb{R}^{2d}$  we have:

$$\underline{\mathcal{L}(T_z f) = \alpha_z(\mathcal{L}f)}$$

where  $\mathcal{L}$  denotes the Weyl or the Kohn-Nirenberg transform

- As a consequence:

Properties of  $V_s^2$  in  $\mathcal{HS}(\mathbb{R}^d) \longleftrightarrow$  Properties of  $V_{\sigma_s}^2$  (or  $V_{a_s}^2$ ) in  $L^2(\mathbb{R}^{2d})$

## $\Lambda$ -shift-invariant subspaces in $\mathcal{HS}(\mathbb{R}^d)$

Let  $\mathbf{S} = \{S_1, S_2, \dots, S_N\}$  be a fixed subset of  $\mathcal{HS}(\mathbb{R}^d)$  and let  $\Lambda$  be a lattice in  $\mathbb{R}^{2d}$ . We are searching for a necessary and sufficient condition such that  $\{\alpha_\lambda(S_n)\}_{\lambda \in \Lambda; n=1,2,\dots,N}$  is a **Riesz sequence** for  $\mathcal{HS}(\mathbb{R}^d)$ , i.e., a **Riesz basis** for the closed subspace

$$V_{\mathbf{S}}^2 := \overline{\text{span}}_{\mathcal{HS}} \{\alpha_\lambda(S_n)\}_{\lambda \in \Lambda; n=1,2,\dots,N} \subset \mathcal{HS}(\mathbb{R}^d)$$

In this case,  $\mathbf{S} = \{S_1, S_2, \dots, S_N\}$  is a *set of stable generators* for the  $\Lambda$ -shift-invariant subspace  $V_{\mathbf{S}}^2$  which can be described by

$$V_{\mathbf{S}}^2 = \left\{ \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) \alpha_\lambda(S_n) : \{c_n(\lambda)\}_{\lambda \in \Lambda} \in \ell^2(\Lambda), n = 1, 2, \dots, N \right\}$$



## Theorem

Let  $\Lambda$  be a lattice and  $S_n \in \mathcal{B}$ ,  $n = 1, 2, \dots, N$ . Then,  $\{\alpha_\lambda(S_n)\}_{\lambda \in \Lambda; n=1,2,\dots,N}$  is a Riesz sequence for  $\mathcal{HS}(\mathbb{R}^d)$  if and only if there exist two constants  $0 < m \leq M$  such that

$$m \mathbb{I}_N \leq G_S^W(z) \leq M \mathbb{I}_N \quad \text{for any } z \in \mathbb{R}^{2d},$$

where  $G_S^W(z)$  denotes the  $N \times N$  matrix-valued function

$$G_S^W(z) := \sum_{\lambda^\circ \in \Lambda^\circ} \mathcal{F}_W(\mathbf{S})(z + \lambda^\circ) \overline{\mathcal{F}_W(\mathbf{S})(z + \lambda^\circ)}^\top, \quad z \in \mathbb{R}^{2d}$$

and  $\mathcal{F}_W(\mathbf{S}) = (\mathcal{F}_W(S_1), \mathcal{F}_W(S_2), \dots, \mathcal{F}_W(S_N))^\top$

where:

- $\mathcal{S} \in \mathcal{B}$  is the Banach space of continuous operators on  $L^2(\mathbb{R}^d)$  with Weyl symbol  $a_s$  in the Feichtinger's algebra  $\mathcal{S}_0(\mathbb{R}^{2d})$ . In essence,  $\mathcal{B}$  consists of trace class operators on  $L^2(\mathbb{R}^d)$  with a norm-continuous inclusion  $\iota : \mathcal{B} \hookrightarrow \mathcal{T}^1$

Recall that  $\psi \in \mathcal{S}_0(\mathbb{R}^{2d})$  iff  $V_{\varphi_0}\psi \in L^1(\mathbb{R}^{2d})$  where

$\varphi_0(x) = 2^{d/4}e^{-\pi x \cdot x}$  is the normalized  $d$ -dimensional gaussian function

- $\Lambda^\circ$  is the adjoint lattice of the lattice  $\Lambda$ . Its associated matrix is  $A^{-\top}\Omega_d$  in case  $\Lambda = A\mathbb{Z}^{2d}$ , where

$$\Omega_d = \begin{pmatrix} O & I_d \\ -I_d & O \end{pmatrix}$$

- $\mathcal{F}_W(S)$  denotes the Fourier-Wigner transform of an operator  $S$  defined as the function

$$\mathcal{F}_W(S)(z) := e^{-\pi i x \cdot \omega} \operatorname{tr}[\pi(-z)S], \quad z = (x, \omega) \in \mathbb{R}^{2d}$$

- In our case,  $\mathcal{F}_W(S_n) = \mathcal{F}_s(a_{S_n})$  for  $n = 1, 2, \dots, N$ , where  $\mathcal{F}_s$  denotes the *symplectic Fourier transform* of  $a_{S_n}$  defined by

$$\mathcal{F}_s(a_{S_n})(z) := \int_{\mathbb{R}^{2d}} a_{S_n}(z') e^{-2\pi i \sigma(z, z')} dz', \quad z \in \mathbb{R}^{2d}$$

$\sigma(z, z') = \omega \cdot x' - \omega' \cdot x$  is the *standard symplectic form* in  $\mathbb{R}^{2d}$

Observe that  $\mathcal{F}_s f(x, w) = \mathcal{F}f(w, -x)$ ,  $(x, w) \in \mathbb{R}^{2d}$ , where  $\mathcal{F}$  denotes the classical  $2d$ -dimensional Fourier transform

- Condition  $m \mathbb{I}_N \leq G_S^W(z) \leq M \mathbb{I}_N$  means:

$$m \|x\|^2 \leq \langle G_S^W(z)x, x \rangle_{\mathbb{C}^N} \leq M \|x\|^2, \quad \forall x \in \mathbb{C}^N$$

# The isomorphism $\mathcal{T}_S$ (via the Kohn-Nirenberg transform)

The isomorphism  $\mathcal{T}_S$  between  $\ell_N^2(\Lambda)$  and  $V_S^2$

$$\begin{aligned} \mathcal{T}_S : \ell_N^2(\Lambda) &\longrightarrow V_{\sigma_S}^2 \subset L^2(\mathbb{R}^{2d}) \longrightarrow V_S^2 \subset \mathcal{HS}(\mathbb{R}^d) \\ (c_1, c_2, \dots, c_N)^\top &\longmapsto \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) T_\lambda \sigma_{S_n} \longmapsto \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) \alpha_\lambda(S_n) \end{aligned}$$

The isomorphism  $\mathcal{T}_S$  is the composition of the isomorphism  $\mathcal{T}_{\sigma_S} : \ell_N^2(\Lambda) \rightarrow V_{\sigma_S}^2$  which maps the standard orthonormal basis  $\{\delta_\lambda\}_{\lambda \in \Lambda}$  for  $\ell_N^2(\Lambda)$  onto the Riesz basis  $\{T_\lambda \sigma_{S_n}\}_{\lambda \in \Lambda; n=1,2,\dots,N}$  for  $V_{\sigma_S}^2$ , and the Kohn-Nirenberg (Weyl) transform between  $V_{\sigma_S}^2$  and  $V_S^2$

## An expression for the average samples

The **average samples** of any  $T = \sum_{n=1}^N \sum_{\mu \in \Lambda} c_n(\mu) \alpha_\mu(S_n)$  in  $V_S^2$  can be expressed as the **output of a discrete convolution system** in  $\ell_N^2(\Lambda)$ :

$$\langle T, \alpha_\lambda(Q_m) \rangle_{\mathcal{HS}} = \sum_{n=1}^N (a_{m,n} *_{\Lambda} c_n)(\lambda) = \langle \mathbf{c}, T_\lambda \mathbf{a}_m^* \rangle_{\ell_N^2(\Lambda)}, \quad \lambda \in \Lambda$$

where  $\mathbf{a}_m^* = (a_{m,1}^*, a_{m,2}^*, \dots, a_{m,N}^*)^\top$ ,  $a_{m,n}^*(\lambda) = \overline{a_{m,n}(-\lambda)}$ , and being

$$a_{m,n}(\mu) = \langle \sigma_{S_n}, T_\mu \sigma_{Q_m} \rangle_{L^2(\mathbb{R}^{2d})} = \langle S_n, \alpha_\mu(Q_m) \rangle_{\mathcal{HS}}, \quad \mu \in \Lambda$$

The **sampling condition** will depend on the  $M \times N$  matrix-valued function  $A(\lambda) = [a_{m,n}(\lambda)]$ ,  $\lambda \in \Lambda$ , whose entries are in  $\ell^2(\Lambda)$

## The diagonal channel samples revisited

For the diagonal channel samples of the operator  $T$

$$\langle T\pi(\lambda)g_m, \pi(\lambda)\tilde{g}_m \rangle_{L^2(\mathbb{R}^d)}, \quad \lambda \in \Lambda, \quad m = 1, 2, \dots, M$$

we have

Diagonal channel samples as average samples

$$\langle T\pi(\lambda)g_m, \pi(\lambda)\tilde{g}_m \rangle_{L^2(\mathbb{R}^d)} = \langle T, \alpha_\lambda(\tilde{g}_m \otimes g_m) \rangle_{\mathcal{HS}}, \quad \lambda \in \Lambda$$

In particular, for  $\varphi, \psi \in L^2(\mathbb{R}^d)$ ,

$$\alpha_z(\varphi \otimes \psi) = [\pi(z)\varphi] \otimes [\pi(z)\psi], \quad z \in \mathbb{R}^{2d}$$

# The symplectic Fourier transform in $\ell^2(\Lambda)$

- The dual group  $\widehat{\Lambda}$  is identified with  $\mathbb{R}^{2d}/\Lambda^\circ$ , where  $\Lambda^\circ$  is the **annihilator group** (adjoint lattice of  $\Lambda$ )

$$\Lambda^\circ = \{ \lambda^\circ \in \mathbb{R}^{2d} : e^{2\pi i \sigma(\lambda^\circ, \lambda)} = 1 \text{ for all } \lambda \in \Lambda \}$$

$\sigma(z, z') = \omega \cdot x' - \omega' \cdot x$  for  $z = (x, \omega)$  and  $z' = (x', \omega')$  in  $\mathbb{R}^{2d}$  is the *standard symplectic form*

- The symplectic Fourier transform of  $c \in \ell^1(\Lambda)$  is defined as

$$\mathcal{F}_s^\Lambda(c)(\dot{z}) := \sum_{\lambda \in \Lambda} c(\lambda) e^{2\pi i \sigma(\lambda, z)}, \quad \dot{z} \in \mathbb{R}^{2d}/\Lambda^\circ,$$

where  $\dot{z}$  denotes the image of  $z$  under the natural quotient map  $\mathbb{R}^{2d} \rightarrow \mathbb{R}^{2d}/\Lambda^\circ$

- Since  $\mathcal{F}_s^\Lambda$  is a Fourier transform it extends to a unitary mapping

## Symplectic Fourier transform in $\ell^2(\Lambda)$

$$\begin{aligned}\mathcal{F}_s^\Lambda : \ell^2(\Lambda) &\longrightarrow L^2(\widehat{\Lambda}) \\ c &\longmapsto \mathcal{F}_s^\Lambda(c)\end{aligned}$$

It satisfies:

- $\mathcal{F}_s^\Lambda(c *_\Lambda d) = \mathcal{F}_s^\Lambda(c) \mathcal{F}_s^\Lambda(d)$ , for  $c \in \ell^1(\Lambda)$  and  $d \in \ell^2(\Lambda)$
- If  $c, d \in \ell^2(\Lambda)$  and  $\mathcal{F}_s^\Lambda(c) \in L^\infty(\widehat{\Lambda}) \Rightarrow \mathcal{F}_s^\Lambda(c *_\Lambda d) = \mathcal{F}_s^\Lambda(c) \mathcal{F}_s^\Lambda(d)$

As usual, the convolution  $*_\Lambda$  of two sequences  $c, d$  is defined by

$$(c *_\Lambda d)(\lambda) = \sum_{\mu \in \Lambda} c(\mu) d(\lambda - \mu), \quad \lambda \in \Lambda$$



## Generalized stable sampling procedure in $V_S^2$

► It is a map  $\mathcal{S}_{\text{samp}} : V_S^2 \longrightarrow \ell_M^2(\Lambda)$  defined as

$$T = \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) \alpha_\lambda(S_n) \in V_S^2 \longmapsto \mathbf{s}_T := A *_\Lambda \mathbf{c} \in \ell_M^2(\Lambda)$$

where  $A = [a_{m,n}] \in \mathcal{M}_{M \times N}(\ell^2(\Lambda))$

That is:  $\mathbf{s}_T(\lambda) = (s_{T,1}(\lambda), s_{T,2}(\lambda), \dots, s_{T,M}(\lambda))^T$ ,  $\lambda \in \Lambda$ , where

$$s_{T,m}(\lambda) = \sum_{n=1}^N (a_{m,n} *_\Lambda c_n)(\lambda), \quad \lambda \in \Lambda, \quad m = 1, 2, \dots, M$$

► Besides, the involved matrix  $A = [a_{m,n}] \in \mathcal{M}_{M \times N}(\ell^2(\Lambda))$  satisfies:

$$0 < \alpha_A := \operatorname{ess\,inf}_{\xi \in \hat{\Lambda}} \lambda_{\min}[\hat{A}(\xi)^* \hat{A}(\xi)] \leq \beta_A := \operatorname{ess\,sup}_{\xi \in \hat{\Lambda}} \lambda_{\max}[\hat{A}(\xi)^* \hat{A}(\xi)] < \infty$$

► The matrix-valued function  $\widehat{A}(\xi) := [\mathcal{F}_s^\Lambda(a_{m,n})(\xi)]$ , a.e.  $\xi \in \widehat{\Lambda}$  is the **transfer matrix of  $A$**  where  $\mathcal{F}_s^\Lambda$  denotes the **symplectic Fourier transform in  $\ell^2(\Lambda)$**

► A crucial fact concerning the samples:

$$\sum_{n=1}^N (a_{m,n} *_\Lambda c_n)(\lambda) = \langle \mathbf{c}, T_\lambda \mathbf{a}_m^* \rangle_{\ell_N^2(\Lambda)}, \quad \lambda \in \Lambda$$

where  $\mathbf{c} = (c_1, c_2, \dots, c_N)^\top$  and  $\mathbf{a}_m^* = (a_{m,1}^*, a_{m,2}^*, \dots, a_{m,N}^*)^\top$

$\{T_\lambda \mathbf{a}_m^*\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  is a frame for  $\ell_N^2(\Lambda) \iff 0 < \alpha_A \leq \beta_A < \infty$

► For the **average sampling** the corresponding matrix  $A$  has entries

$$a_{m,n}(\lambda) = \langle \sigma_{S_n}, T_\lambda \sigma_{Q_m} \rangle_{L^2(\mathbb{R}^{2d})} = \langle S_n, \alpha_\lambda(Q_m) \rangle_{\mathcal{HS}}, \quad \lambda \in \Lambda,$$

i.e., the columns of  $A$  are the sequences of average samples of the generators  $S_n$  of  $V_S^2$

► Concerning its dual frames  $\{T_\lambda \mathbf{b}_m\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  with the same structure, they are obtained from the left-inverses  $\widehat{B} \in \mathcal{M}_{N \times M}(L^\infty(\widehat{\Lambda}))$  of the matrix  $\widehat{A}$ , i.e.,  $\widehat{B}(\xi)\widehat{A}(\xi) = \mathbb{I}_N$ , a.e.  $\xi \in \widehat{\Lambda}$

The  $\mathbf{b}_m$  are the columns of the matrix  $B \in \mathcal{M}_{N \times M}(\ell^2(\Lambda))$

► For instance, we can choose as  $\widehat{B}$  its Moore-Penrose pseudo-inverse  $\widehat{A}(\xi)^\dagger = [\widehat{A}(\xi)^* \widehat{A}(\xi)]^{-1} \widehat{A}(\xi)^*$

► All the suitable left-inverses  $\widehat{B}$  of the matrix  $\widehat{A}$  can be written from the Moore-Penrose pseudo-inverse as the  $N \times M$  matrices

$$\widehat{B}(\xi) = \widehat{A}(\xi)^\dagger + C(\xi) [\mathbb{I}_M - \widehat{A}(\xi)\widehat{A}(\xi)^\dagger], \quad \text{a.e. } \xi \in \widehat{\Lambda},$$

where  $C$  denotes any  $N \times M$  matrix with entries in  $L^\infty(\widehat{\Lambda})$

## Theorem (Sampling theorem in $V_S^2$ )

(1) Given a *sampling procedure*  $\mathcal{S}_{\text{samp}}$  in  $V_S^2$ , there exist  $M \geq N$  elements  $H_m \in V_S^2$ ,  $m = 1, 2, \dots, M$ , such that the sampling formula

$$T = \sum_{m=1}^M \sum_{\lambda \in \Lambda} s_{T,m}(\lambda) \alpha_\lambda(H_m) \quad \text{in } \mathcal{HS}\text{-norm}$$

holds for each  $T \in V_S^2$  where  $\{\alpha_\lambda(H_m)\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  is a frame for  $V_S^2$

(2) The convergence of the series is unconditional in  $\mathcal{HS}$ -norm, and besides,  $\|\mathbf{s}_T\|_{\ell_M^2} \asymp \|T\|_{\mathcal{HS}}$  in  $V_S^2$

(3) Reciprocally, if a sampling formula like above holds in  $V_S^2$  where

$$\mathbf{s}_T(\lambda) = (s_{T,1}(\lambda), s_{T,2}(\lambda), \dots, s_{T,M}(\lambda))^T := (A *_{\Lambda} \mathbf{c})(\lambda), \quad \lambda \in \Lambda,$$

where  $\beta_A < +\infty$ , and  $\{\alpha_\lambda(H_m)\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  is a frame for  $V_S^2$ , then  $\alpha_A > 0$

## Sketch of the proof

For any  $T = \sum_{n=1}^N \sum_{\mu \in \Lambda} c_n(\mu) \alpha_\mu(S_n)$  in  $V_S^2$  we have:

- For its samples  $s_{T,m}(\lambda) = \langle \mathbf{c}, T_\lambda \mathbf{a}_m^* \rangle_{\ell_N^2(\Lambda)}$ ,  $\lambda \in \Lambda$ , where the sequence  $\{T_\lambda \mathbf{a}_m^*\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  is a frame for  $\ell_N^2(\Lambda)$
- By using a dual frame  $\{T_\lambda \mathbf{b}_m\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  of the above frame  $\{T_\lambda \mathbf{a}_m^*\}_{\lambda \in \Lambda; m=1,2,\dots,M}$  we obtain

$$\mathbf{c} = \sum_{m=1}^M \sum_{\lambda \in \Lambda} \langle \mathbf{c}, T_\lambda \mathbf{a}_m^* \rangle_{\ell_N^2(\Lambda)} T_\lambda \mathbf{b}_m \quad \text{for each } \mathbf{c} \in \ell_N^2(\Lambda)$$

- Finally, applying the isomorphism  $\mathcal{T}_S$

$$T = \sum_{m=1}^M \sum_{\lambda \in \Lambda} s_{T,m}(\lambda) \mathcal{T}_S[T_\lambda \mathbf{b}_m] = \sum_{m=1}^M \sum_{\lambda \in \Lambda} s_{T,m}(\lambda) K_{T_\lambda(\mathcal{T}_{\sigma_S} \mathbf{b}_m)}$$

► That is

$$T = \sum_{m=1}^M \sum_{\lambda \in \Lambda} s_{T,m}(\lambda) \alpha_{\lambda}[K_{h_m}] = \sum_{m=1}^M \sum_{\lambda \in \Lambda} s_{T,m}(\lambda) \alpha_{\lambda}(H_m)$$

where  $H_m = K_{h_m}$  and  $h_m = \mathcal{T}_{\sigma_S}(\mathbf{b}_m)$

We have used that

$$\mathcal{T}_{\sigma_S}(T_{\lambda} \mathbf{b}_m) = T_{\lambda}(\mathcal{T}_{\sigma_S} \mathbf{b}_m) = T_{\lambda}(h_m)$$

Observe that  $\mathbf{b}_m = (b_{1,m}(\lambda), b_{2,m}(\lambda), \dots, b_{N,m}(\lambda))^{\top}$  is the  $m$ -th column of  $B$  (remind that  $\widehat{B} \widehat{A} = \mathbb{I}_N$ ). As a consequence,

$$H_m = \sum_{n=1}^N \sum_{\lambda \in \Lambda} b_{n,m}(\lambda) \alpha_{\lambda}(S_n), \quad m = 1, 2, \dots, M$$

- Under the above hypotheses, the **average sampling formula** reads

$$T = \sum_{m=1}^M \sum_{\lambda \in \Lambda} \langle T, \alpha_{\lambda}(Q_m) \rangle_{\mathcal{HS}} \alpha_{\lambda}(H_m) \quad \text{in } \mathcal{HS}\text{-norm}$$

- Under the same hypotheses, the case  $M = N$  implies that  $\{\alpha_{\lambda}(H_m)\}_{\lambda \in \Lambda; m=1,2,\dots,N}$  is a Riesz basis for  $V_S^2$
- Since convergence in  $\mathcal{HS}$ -norm implies convergence in operator norm, for each  $f \in L^2(\mathbb{R}^d)$  we get the **pointwise expansion**:

$$T(f) = \sum_{m=1}^M \sum_{\lambda \in \Lambda} \langle T, \alpha_{\lambda}(Q_m) \rangle_{\mathcal{HS}} [\alpha_{\lambda}(H_m)](f) \quad \text{in } L^2\text{-norm}$$

## An illustrative example

► Assume  $V_S^2$  with  $N$  stable generators of the form  $S_n = \varphi_n \otimes \tilde{\varphi}_n$  with  $\varphi_n, \tilde{\varphi}_n \in \mathcal{S}_0(\mathbb{R}^d)$ ,  $n = 1, 2, \dots, N$ . In this regard,

$$\mathcal{F}_W(\varphi_n \otimes \tilde{\varphi}_n)(z) = e^{\pi i x \cdot \omega} V_{\tilde{\varphi}_n} \varphi_n(z), \quad z = (x, \omega) \in \mathbb{R}^{2d}$$

► For each  $T \in V_S^2$  we consider the diagonal channel samples

$$\langle T\pi(\lambda)g_m, \pi(\lambda)\tilde{g}_m \rangle_{L^2(\mathbb{R}^d)}, \quad \lambda \in \Lambda$$

with  $g_m, \tilde{g}_m \in \mathcal{S}_0(\mathbb{R}^d)$ ,  $m = 1, 2, \dots, M$ . In this case,

$$\begin{aligned} a_{m,n}(\lambda) &= \langle (\varphi_n \otimes \tilde{\varphi}_n) \pi(\lambda) g_m, \pi(\lambda) \tilde{g}_m \rangle_{L^2(\mathbb{R}^d)} \\ &= \overline{V_{g_m} \tilde{\varphi}_n(\lambda)} V_{\tilde{g}_m} \varphi_n(\lambda), \quad \lambda \in \Lambda \end{aligned}$$

► It is known that the sequences  $\{a_{m,n}(\lambda)\}_{\lambda \in \Lambda}$  belong to  $\ell^1(\Lambda)$  and, as a consequence, the entries of  $\hat{A}$  are continuous functions on the compact  $\hat{\Lambda}$



► Thus, the **sampling conditions** in the definition of generalized stable sampling procedure reduce to

$$\det[\widehat{A}(\xi)^* \widehat{A}(\xi)] \neq 0 \quad \text{for all } \xi \in \widehat{\Lambda}$$

► Under the above circumstances:

Any  $T = \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) \alpha_\lambda(\varphi_n \otimes \widetilde{\varphi}_n) \in V_S^2$  can be recovered, in a stable way, from its diagonal channel samples  $\langle T\pi(\lambda)g_m, \pi(\lambda)\widetilde{g}_m \rangle_{L^2(\mathbb{R}^d)}$

Moreover, since

$$T(\eta) = \sum_{n=1}^N \sum_{\lambda \in \Lambda} c_n(\lambda) V_{\widetilde{\varphi}_n} \eta(\lambda) \pi(\lambda) \varphi_n = \sum_{n=1}^N \mathcal{G}_{\mathbf{c}_n}^{\widetilde{\varphi}_n, \varphi_n}(\eta), \quad \eta \in L^2(\mathbb{R}^d)$$

the operator  $T$  is nothing but a sum of **Gabor multipliers**  $\mathcal{G}_{\mathbf{c}_n}^{\widetilde{\varphi}_n, \varphi_n}$

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