

# A link between discrete convolution systems and sampling via frame theory

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# Outline

- ▶ Statement of the problem
- ▶ A brief on discrete frames
- ▶ Discrete convolution systems and frames
- ▶ The resulting sampling theory
- ▶ Some sampling results as particular cases

# Statement of the problem

- ▶  $(G, +)$  is a countable discrete abelian group
- ▶  $G \ni g \mapsto U(g) \in \mathcal{U}(\mathcal{H})$  a unitary representation of  $G$  on a separable Hilbert space  $\mathcal{H}$
- ▶ Given a set  $\Phi = \{\varphi_1, \varphi_2, \dots, \varphi_N\}$  in  $\mathcal{H}$ , consider the  $U$ -invariant subspace of  $\mathcal{H}$

$$\mathcal{V}_\Phi = \left\{ \sum_{n=1}^N \sum_{g \in G} x_n(g) U(g) \varphi_n : x_n \in \ell^2(G), n = 1, 2, \dots, N \right\}$$

where the sequence  $\{U(g)\varphi_n\}_{g \in G; n=1,2,\dots,N}$  is a **Riesz sequence** for  $\mathcal{H}$

- ▶ We define an **average sampling** in  $\mathcal{V}_\Phi$  as follows:

Given  $M$  elements  $\psi_m \in \mathcal{H}$ ,  $m = 1, 2, \dots, M$ , which do not belong necessarily to  $\mathcal{V}_\Phi$ , for any  $f \in \mathcal{V}_\Phi$  we define for  $m = 1, 2, \dots, M$

$$\mathcal{L}_m f(g) := \langle f, U(g)\psi_m \rangle_{\mathcal{H}}, \quad g \in G$$

## Motivating fact

These samples can be expressed as the output of a convolution system defined in  $\ell_N^2(G)$ .

Indeed, for any  $f = \sum_{n=1}^N \sum_{h \in G} x_n(h) U(h)\varphi_n$  in  $\mathcal{V}_\Phi$  one gets

$$\mathcal{L}_m f(g) = \sum_{n=1}^N (a_{m,n} * x_n)(g), \quad g \in G$$

where  $a_{m,n}(g) = \langle \varphi_n, U(g)\psi_m \rangle_{\mathcal{H}}$ ,  $g \in G$ , belongs to  $\ell^2(G)$

In general, we can describe a **sampling procedure** in  $\mathcal{V}_\Phi$  as a **convolution system**

$$\begin{aligned} \mathcal{A} : \ell_N^2(G) &\longrightarrow \ell_M^2(G) \\ \mathbf{x} = (x_1, x_2, \dots, x_N)^\top &\longmapsto \mathcal{A}(\mathbf{x}) = A * \mathbf{x} = (\mathcal{L}_1 f, \mathcal{L}_2 f, \dots, \mathcal{L}_M f)^\top \end{aligned}$$

with an associated matrix  $A = [a_{m,n}] \in \mathcal{M}_{M \times N}(\ell^2(G))$

### Definition

The above **sampling procedure**  $\mathcal{V}_\Phi \ni f \longmapsto \mathcal{L}f := \mathcal{A}(x) \in \ell_M^2(G)$  is **stable** if there exist two positive constants  $0 < \alpha \leq \beta$  such that

$$\alpha \|f\|_{\mathcal{H}}^2 \leq \|\mathcal{L}f\|_{\ell_M^2(G)}^2 \leq \beta \|f\|_{\mathcal{H}}^2 \quad \text{for all } f \in \mathcal{V}_\Phi$$

## Goal

- ▶ **To characterize** the stable sampling procedures in  $\mathcal{V}_\Phi$  obtained by means of a convolution system  $\mathcal{A}$
- ▶ **To obtain** the reconstruction formulas in  $\mathcal{V}_\Phi$  associated with the sampling procedure  $\mathcal{A}$ , i.e., sampling formulas having the form

$$f = \sum_{m=1}^M \sum_{g \in G} \mathcal{L}_m f(g) U(g) S_m \quad \text{in } \mathcal{H}$$

for some elements  $S_1, S_2, \dots, S_M \in \mathcal{V}_\Phi$  such that the sequence  $\{U(g)S_m\}_{g \in G; m=1,2,\dots,M}$  is a frame for  $\mathcal{V}_\Phi$

This is done by linking the convolution system  $A * x$  defined in  $\ell_N^2(G)$  with a frame of translates  $\{T_g \mathbf{c}_m\}_{g \in G; m=1,2,\dots,M}$  in  $\ell_N^2(G)$ , where  $T_g \mathbf{c}_m(h) = \mathbf{c}_m(h - g)$ ,  $h \in G$ , for  $\mathbf{c}_m \in \ell_N^2(G)$

# A brief on discrete frames



A sequence  $\{x_n\}_{n=1}^{\infty}$  is a **frame** for a separable Hilbert space  $\mathcal{H}$  if there exist constants  $0 < A \leq B$  (frame bounds) such that

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

**Preframe operator  $T$  (synthesis)** and its **adjoint  $T^*$  (analysis)**

$$\begin{array}{ccc} T : \ell^2(\mathbb{N}) & \longrightarrow & \mathcal{H} \\ \{c_n\}_{n=1}^{\infty} & \longmapsto & \sum_{n=1}^{\infty} c_n x_n \end{array} \quad \rightsquigarrow \quad \begin{array}{ccc} T^* : \mathcal{H} & \longrightarrow & \ell^2(\mathbb{N}) \\ x & \longmapsto & \{\langle x, x_n \rangle\}_{n=1}^{\infty} \end{array}$$

**Frame operator  $S$**

$$S : \mathcal{H} \longrightarrow \mathcal{H}, \quad Sx := TT^*x = \sum_{n=1}^{\infty} \langle x, x_n \rangle x_n, \quad x \in \mathcal{H}$$

## Some properties

- ▶  $T$  is **bounded** if and only if  $\{x_n\}_{n=1}^{\infty}$  is a **Bessel sequence**
- ▶  $S$  is a **bounded, invertible, positive, self-adjoint operator** and the sequence  $\{S^{-1}x_n\}_{n=1}^{\infty}$  is a frame for  $\mathcal{H}$  (**canonical dual frame**)
- ▶ For each  $x \in \mathcal{H}$  we have,

$$x = \sum_{n=1}^{\infty} \langle x, x_n \rangle S^{-1}x_n = \sum_{n=1}^{\infty} \langle x, S^{-1}x_n \rangle x_n$$

- ▶  $\{x_n\}_{n=1}^{\infty}$  is a frame  $\iff T$  is bounded and onto  $\iff T^*$  is injective with a closed range  $\iff S$  is bounded and invertible
- ▶ The sequences  $\{x_n\}_{n=1}^{\infty}$  and  $\{y_n\}_{n=1}^{\infty}$  are **dual frames** if for each  $x \in \mathcal{H}$  we have

$$x = \sum_{n=1}^{\infty} \langle x, x_n \rangle y_n = \sum_{n=1}^{\infty} \langle x, y_n \rangle x_n$$

## Some properties

- ▶  $\{x_n\}_{n=1}^{\infty}$  and  $\{y_n\}_{n=1}^{\infty}$  Bessel sequences and  $T_y T_x^* = T_x T_y^* = I_{\mathcal{H}}$   
 $\iff \{x_n\}_{n=1}^{\infty}$  and  $\{y_n\}_{n=1}^{\infty}$  dual frames
- ▶  $\{x_n\}_{n=1}^{\infty}$  is an exact frame  $\iff \{x_n\}_{n=1}^{\infty}$  is a Riesz basis (a unique dual)

Overcomplete (inexact) frames have an infinity of duals

- ▶ In finite dimension: frame  $\iff$  spanning set

The set  $\{x_1, x_2, \dots, x_M\}$  in  $\mathbb{C}^N$  is a frame  $\iff \text{rank } \mathbf{M} = N$ ,  
where  $\mathbf{M} := (x_1, x_2, \dots, x_M)$

Its frame operator is  $S = \mathbf{M}\mathbf{M}^*$  and  $A_{opt} = \sigma_1^2$ ;  $B_{opt} = \sigma_N^2$

( $\sigma_1$  and  $\sigma_N$  are the smallest and largest singular value of  $\mathbf{M}$  respectively)

## Some applications of frames

1. In **Sampling theory** by identifying the data samples as frame coefficients  $\{\langle x, x_n \rangle\}$  and looking for appropriate duals  $\{y_n\}$   
This is the aim of this talk
2. In **Signal processing** as a **noise suppressing** method:

$$\sum_{n=1}^{\infty} (\langle x, x_n \rangle + \epsilon_n) S^{-1} x_n = x + S^{-1} \left( \sum_{n=1}^{\infty} \epsilon_n x_n \right)$$

In case of Riesz or orthonormal bases, **error**  $\propto \|\{\epsilon_n\}\|_2 > 0$   
For frames, it can be suppressed whenever  $\{\epsilon_n\} \in \ker T$

### 3. In **Quantum mechanics**: Gleason functions for Parseval frames

#### Definition

A Gleason function  $g$  of weight  $W$  is a function  $g : B_d \subset \mathcal{H}_d \rightarrow K$  such that  $\sum_{j \in J} g(x_j) = W$  for all  $\{x_j\}_{j \in J}$  Parseval frame for  $\mathcal{H}_d$

- For instance,  $g(x) = \langle Ax, x \rangle$ ,  $x \in B_d$ , with  $A$  self-adjoint is a Gleason function of weight  $W = \text{tr } A$
- **Positive operator valued measurements**:  $\{E_n\} \subset \mathcal{E}(\mathcal{H})$  such that  $I = \sum_n E_n \iff$  Parseval frames for  $\mathcal{H}$   
 $E \in \mathcal{E}(\mathcal{H})$  if  $E$  is a self-adjoint operator with  $0 \leq E \leq I$
- To obtain **generalized probability measures**  $\nu : \mathcal{E}(\mathcal{H}) \rightarrow \mathbb{R} \Rightarrow \nu(E) = \text{tr}(\rho E)$ ,  $E \in \mathcal{E}(\mathcal{H})$ , for a density operator  $\rho$  (Busch/Gleason theorem)

4. In **Numerical analysis** to **computing the best approximation**

- ▶ Let  $\Omega \subset (-1, 1)^d$  be a compact domain; the orthonormal basis  $\{2^{-d/2} e^{i\pi n \cdot t}\}_{n \in \mathbb{Z}^d}$  for  $L^2(-1, 1)^d$  restricted to  $\Omega$  is a **tight, linearly-independent frame with bounds  $A = B = 1$  for  $L^2(\Omega)$**
- ▶ The augmented Fourier basis  $\{ \frac{1}{\sqrt{2}} e^{in\pi t} \}_{n \in \mathbb{Z}} \cup \{ \sqrt{k+1/2} P_k \}_{k=1}^K$  with  $K$  Legendre polynomials is a **tight, linearly-independent frame with bounds  $A = 1$  and  $B = 2$  for  $L^2(-1, 1)$**
- ▶ A basis of (Legendre, Chebyshev) polynomials plus modified polynomials to approximate functions  $f(t) = w(t)g(t) + h(t)$ ,  $t \in [-1, 1]$ , where  $g, h$  are smooth and  $w \in L^\infty(-1, 1)$  may be singular, oscillatory, etc. Indeed,  $\{\varphi_n(t)\}_{n \in \mathbb{N}} \cup \{w(t)\varphi_n(t)\}_{n \in \mathbb{N}}$  is a **frame for  $L^2(-1, 1)$  with bounds**

$$A = 1 + \operatorname{ess\,inf}_{t \in (-1, 1)} |w(t)|^2 \quad \text{and} \quad B = 1 + \operatorname{ess\,sup}_{t \in (-1, 1)} |w(t)|^2$$

- ▶ To compute the best approximation of  $f$  from  $\mathcal{H}_N := \text{span}\{\phi_n\}_{n \in I_N}$

$$P_N f = \sum_{n \in I_N} x_n \phi_n \Leftrightarrow \mathbf{G}_N \mathbf{x} = \mathbf{y}, \quad \langle P_N f, \phi_n \rangle = \langle f, \phi_n \rangle, \quad n \in I_N$$

$$\mathbf{G}_N = \{\langle \phi_m, \phi_n \rangle\} \in \mathbb{C}^{N \times N}; \quad \mathbf{x} = \{x_n\} \in \mathbb{C}^N; \quad \mathbf{y} = \{\langle f, \phi_n \rangle\} \in \mathbb{C}^N$$

- ▶ Ill-conditioned problem  $\Rightarrow$  To regularize by using the SVD of  $\mathbf{G}_N = \mathbf{V} \Sigma \mathbf{V}^*$  to construct a truncated SVD projection

$$P_N^\epsilon f = \sum_{\sigma_n > \epsilon} \frac{\langle f, T_N \mathbf{v}_n \rangle}{\sigma_n} T_N \mathbf{v}_n \quad (T_N^* T_N \mathbf{v}_n = \mathbf{G}_N \mathbf{v}_n = \sigma_n \mathbf{v}_n)$$

$$\{T_N \mathbf{v}_n\} \text{ orthogonal basis for } \mathcal{H}_N: \quad \langle T_N \mathbf{v}_n, T_N \mathbf{v}_m \rangle = \sigma_n \delta_{n,m}$$

- ▶ Finally, to give an estimation for  $\|f - P_N^\epsilon f\|$

$$\|f - P_N^\epsilon f\| \leq \|f - T_N \mathbf{c}\| + \sqrt{\epsilon} \|\mathbf{c}\|$$

- **Ejemplo 1.** Let  $\Omega \subset (-1, 1)^d$  be a compact Lipschitz domain. If  $f \in H^{kd}(-1, 1)^d$ , there exists a constant  $C > 0$ , independent of  $f$  and  $N$ , such that

$$\|f - P_N^\epsilon f\|_{L^2(\Omega)} \leq C(N^{-k} + \sqrt{\epsilon}) \|f\|_{H^{kd}(-1,1)^d}$$

- **Ejemplo 2.** Let  $K \in \mathbb{N}$  be fixed. If  $f \in H^k(-1, 1)$  for  $0 \leq k \leq K$ , there exists a constant  $C > 0$ , independent of  $f$  and  $N$ , such that

$$\|f - P_N^\epsilon f\|_{L^2(-1,1)} \leq C(N^{-k} + \sqrt{\epsilon}) \|f\|_{H^k(-1,1)}$$



# Discrete convolution systems and frames

## Discrete convolution systems: some needed properties

Let

$$\begin{aligned} \mathcal{A} : \ell_N^2(G) &\longrightarrow \ell_M^2(G) \\ \mathbf{x} &\longmapsto \mathcal{A}(\mathbf{x}) = A * \mathbf{x} \end{aligned}$$

be a convolution system with matrix  $A = [a_{m,n}] \in \mathcal{M}_{M \times N}(\ell^2(G))$

- ▶  $\mathcal{A}$  is a **well defined bounded operator** if and only if  $\widehat{A} \in \mathcal{M}_{M \times N}(L^\infty(\widehat{G}))$ , where  $\widehat{A}(\xi) := [\widehat{a}_{m,n}(\xi)]$  (transfer matrix)
- ▶ Its **adjoint operator**  $\mathcal{A}^* : \ell_M^2(G) \rightarrow \ell_N^2(G)$  is also a (bounded) convolution operator with matrix  $A^* = [a_{m,n}^*]^\top \in \mathcal{M}_{N \times M}(\ell^2(G))$  where  $a_{m,n}^*(g) := \overline{a_{m,n}(-g)}$ ,  $g \in G$ . Its transfer matrix is  $\widehat{A}^*(\xi)$  is the transpose conjugate of  $\widehat{A}(\xi)$ , i.e.,  $\widehat{A}(\xi)^*$ , a.e.  $\xi \in \widehat{G}$
- ▶  $\mathcal{A}$  (bounded) is **injective with closed range**  $\iff \mathcal{A}^* \mathcal{A}$  invertible  $\iff \delta_A := \operatorname{ess\,inf}_{\xi \in \widehat{G}} \det[\widehat{A}(\xi)^* \widehat{A}(\xi)] > 0$

## When discrete convolution systems meets frames

- ▶ Let  $\mathbf{a}_m^*$  denote the  $m$ -th column of the matrix  $A^*$ , then

$$[\mathcal{A}(\mathbf{x})]_m(g) = [A * \mathbf{x}]_m(g) = \langle \mathbf{x}, T_g \mathbf{a}_m^* \rangle_{\ell_N^2(G)}$$

In other words, the operator  $\mathcal{A}$  is the analysis operator of the sequence  $\{T_g \mathbf{a}_m^*\}_{g \in G; m=1,2,\dots,M}$  in  $\ell_N^2(G)$ . Consequently:

- ▶  $\{T_g \mathbf{a}_m^*\}_{g \in G; m=1,2,\dots,M}$  is a frame for  $\ell_N^2(G) \iff \delta_A > 0$
- ▶ Assume that  $\widehat{A} \in \mathcal{M}_{M \times N}(L^\infty(\widehat{G}))$  and  $\widehat{B} \in \mathcal{M}_{N \times M}(L^\infty(\widehat{G}))$ , and let  $\mathbf{b}_m$  denote the  $m$ -th column of the matrix  $B$ . Then the sequences  $\{T_g \mathbf{a}_m^*\}_{g \in G; m=1,2,\dots,M}$  and  $\{T_g \mathbf{b}_m\}_{g \in G; m=1,2,\dots,M}$  form a pair of dual frames for  $\ell_N^2(G)$  if and only if

$$\widehat{B}(\xi) \widehat{A}(\xi) = I_N, \text{ a.e. } \xi \in \widehat{G} \quad (\iff \mathcal{B}\mathcal{A} = \mathcal{I}_{\ell_N^2(G)})$$

## When discrete convolution systems meets frames and sampling

►  $\{T_g \mathbf{a}_m^*\}$  frame  $\iff \delta_A > 0$ , and there exist a dual frame  $\{T_g \mathbf{b}_m\}$  via the Moore-Penrose pseudo inverse  $\widehat{A}(\xi)^\dagger = [\widehat{A}(\xi)^* \widehat{A}(\xi)]^{-1} \widehat{A}(\xi)^*$

► Thus, for each  $\mathbf{x} \in \ell_N^2(G)$  we have

$$\mathbf{x} = \sum_{m=1}^M \sum_{g \in G} \langle \mathbf{x}, T_g \mathbf{a}_m^* \rangle_{\ell_N^2(G)} T_g \mathbf{b}_m = \sum_{m=1}^M \sum_{g \in G} \mathcal{L}_m f(g) T_g \mathbf{b}_m$$

► Applying the natural isomorphism  $\mathcal{T}_{U,\Phi} : \ell_N^2(G) \rightarrow \mathcal{V}_\Phi$  which maps the standard orthonormal basis  $\{\delta_{g,n}\}_{g \in G; n=1,2,\dots,N}$  for  $\ell_N^2(G)$  onto the Riesz basis  $\{U(g)\varphi_n\}_{g \in G; n=1,2,\dots,N}$  for  $\mathcal{V}_\Phi$ , for each  $f = \mathcal{T}_{U,\Phi} \mathbf{x} \in \mathcal{V}_\Phi$  we get

$$f = \sum_{m=1}^M \sum_{g \in G} \mathcal{L}_m f(g) \underbrace{\mathcal{T}_{U,\Phi}(T_g \mathbf{b}_m)}_{U(g)(\mathcal{T}_{U,\Phi} \mathbf{b}_m)} = \sum_{m=1}^M \sum_{g \in G} \mathcal{L}_m f(g) U(g) \underbrace{(\mathcal{T}_{U,\Phi} \mathbf{b}_m)}_{S_m}$$

# The resulting sampling theory

## Theorem

Let  $A = [a_{m,n}] \in \mathcal{M}_{M \times N}(\ell^2(G))$  be the matrix defining the sampling vector  $\mathcal{L}f$  for each  $f \in \mathcal{V}_\Phi$ , and assume that  $\hat{A} \in \mathcal{M}_{M \times N}(L^\infty(\hat{G}))$ .

The following statements are equivalent:

(a) The constant  $\delta_A := \operatorname{ess\,inf}_{\xi \in \hat{G}} \det[\hat{A}(\xi)^* \hat{A}(\xi)] > 0$

(b) There exist constants  $0 < \alpha \leq \beta$  such that

$$\alpha \|f\|^2 \leq \sum_{m=1}^M \sum_{g \in G} |\mathcal{L}_m f(g)|^2 \leq \beta \|f\|^2, \quad f \in \mathcal{V}_\Phi$$

(c) There exists  $\hat{B} \in \mathcal{M}_{N \times M}(L^\infty(\hat{G}))$  such that  $\hat{B}(\xi) \hat{A}(\xi) = I_N$ , a.e.  $\xi \in \hat{G}$

- (d) *There exists a matrix  $B \in \mathcal{M}_{N \times M}(\ell^2(G))$  with  $\widehat{B} \in \mathcal{M}_{N \times M}(L^\infty(\widehat{G}))$ , such that*

$$\mathbf{x} = B * \mathcal{L}f \quad \text{and} \quad f = \sum_{n=1}^N \sum_{g \in G} x_n(g) U(g) \varphi_n, \quad f \in \mathcal{V}_\Phi$$

*That is, the bounded convolution system  $\mathcal{B} : \ell_M^2(G) \rightarrow \ell_N^2(G)$  satisfies  $\mathcal{B} \mathcal{A} = \mathcal{I}_{\ell_N^2(G)}$*

- (e) *There exist  $M$  elements  $S_m \in \mathcal{V}_\Phi$  such that the sequence  $\{U(g)S_m\}_{g \in G; m=1,2,\dots,M}$  is a frame for  $\mathcal{V}_\Phi$ , and the formula*

$$f = \sum_{m=1}^M \sum_{g \in G} \mathcal{L}_m f(g) U(g) S_m \quad \text{in } \mathcal{H}$$

*holds in  $\mathcal{V}_\Phi$*

(f) *There exists a frame  $\{S_{g,m}\}_{g \in G; m=1,2,\dots,M}$  for  $\mathcal{V}_\Phi$  such that for each  $f \in \mathcal{V}_\Phi$  we have the expansion*

$$f = \sum_{m=1}^M \sum_{g \in G} \mathcal{L}_m f(g) S_{g,m} \quad \text{in } \mathcal{H}$$

The reconstruction elements  $\{S_m\}_{m=1,2,\dots,M}$  in  $\mathcal{V}_\Phi$  are necessarily obtained from the columns  $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_M$  of a matrix  $B$  satisfying (c)

$$S_m = \mathcal{T}_{U,\Phi} \mathbf{b}_m = \sum_{n=1}^N \sum_{g \in G} b_{n,m} U(g) \varphi_n, \quad m = 1, 2, \dots, M$$

If  $M = N$  and  $\operatorname{ess\,inf}_{\xi \in \widehat{G}} |\det[\widehat{A}(\xi)]| > 0$  we are in the **Riesz bases** setting



# Sampling results as particular cases

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

$\mathcal{H} := L^2(\mathbb{R}^d)$ ,  $G := \mathbb{Z}^d$  and  $[U(p)f](t) := f(t - p)$ ,  $t \in \mathbb{R}^d$  and  $p \in \mathbb{Z}^d$

The average samples

$$\mathcal{L}_m f(p) := \langle f, \psi_m(\cdot - p) \rangle_{L^2(\mathbb{R}^d)}, \quad p \in \mathbb{Z}^d$$

have the matrix  $A = [a_{m,n}]$ , with  $a_{m,n}(p) = \langle \varphi_n, \psi_m(\cdot - p) \rangle_{L^2(\mathbb{R}^d)}$ . Under the hypotheses in the sampling theorem, for the classical shift-invariant subspace  $V_{\Phi}^2$  of  $L^2(\mathbb{R}^d)$  described as

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

we obtain

$$f(t) = \sum_{m=1}^M \sum_{p \in \mathbb{Z}^d} \langle f, \psi_m(\cdot - p) \rangle_{L^2(\mathbb{R}^d)} S_m(t - p) \quad \text{in } L^2(\mathbb{R}^d)$$

As  $V_{\Phi}^2$  is a RKHS, we have also pointwise convergence

## Sampling associated with a crystallographic group

$P$  is a non-singular  $d \times d$  matrix,  $\Gamma = \{\gamma_1 = I, \gamma_2, \dots, \gamma_N\}$  a finite subgroup of  $O(d)$  of order  $N$  such that  $\gamma(P\mathbb{Z}^d) = P\mathbb{Z}^d$  for each  $\gamma \in \Gamma$ .

Consider the **crystallographic group** is  $\mathcal{C}_{P,\Gamma} := P\mathbb{Z}^d \rtimes_{\sigma} \Gamma$  (Recall that  $(x, \gamma) \cdot (x', \gamma') = (x + \gamma x', \gamma \gamma')$ ) and its **quasi regular representation** on  $L^2(\mathbb{R}^d)$

$$U(p, \gamma)f(t) = f[\gamma^{\top}(t - p)], \quad p \in P\mathbb{Z}^d, \gamma \in \Gamma \text{ and } f \in L^2(\mathbb{R}^d)$$

Consider the  $U$ -invariant subspace in  $L^2(\mathbb{R}^d)$

$$\begin{aligned} \mathcal{V}_{\varphi} &= \left\{ \sum_{(p, \gamma) \in \mathcal{C}_{P, \Gamma}} x(p, \gamma) \underbrace{\varphi[\gamma^{\top}(t - p)]}_{U(p, \gamma)\varphi(t)} : \{x(p, \gamma)\} \in \ell^2(\mathcal{C}_{P, \Gamma}) \right\} \\ &= \left\{ \sum_{n=1}^N \sum_{p \in P\mathbb{Z}^d} x_n(p) \underbrace{U(p, I) \underbrace{U(0, \gamma_n)\varphi}_{\varphi_n}}_{\varphi_n(t-p)} : x_n \in \ell^2(P\mathbb{Z}^d), n = 1, 2, \dots, N \right\} \end{aligned}$$

Choosing  $M$  functions  $\psi_m \in L^2(\mathbb{R}^d)$ ,  $m = 1, 2, \dots, M$ , we consider the **average sampling** in  $\mathcal{V}_\varphi$

$$\mathcal{L}_m f(p) = \langle f, U(p, I)\psi_m \rangle = \langle f, \psi_m(\cdot - p) \rangle, \quad p \in P\mathbb{Z}^d.$$

Here, the matrix  $A = [a_{m,n}]$  has entries

$$a_{m,n}(p) = \langle \varphi(t), \psi_m(\gamma_n t - p) \rangle_{L^2(\mathbb{R}^d)}$$

Under the hypotheses in the sampling theorem, there exist  $M \geq N$  sampling functions  $S_m \in \mathcal{V}_\varphi$  for  $m = 1, 2, \dots, M$ , such that the sequence  $\{S_m(\cdot - p)\}_{p \in P\mathbb{Z}^d; m=1,2,\dots,M}$  is a frame for  $\mathcal{V}_\varphi$  and, for each  $f \in \mathcal{V}_\varphi$ , we have

$$f(t) = \sum_{m=1}^M \sum_{p \in P\mathbb{Z}^d} \langle f, \psi_m(\cdot - p) \rangle_{L^2(\mathbb{R}^d)} S_m(t - p) \quad \text{in } L^2(\mathbb{R}^d)$$

## The case of pointwise sampling

$G$  is a countable discrete subgroup of a LCA group  $\tilde{G}$

$t \in \tilde{G} \mapsto U(t) \in \mathcal{U}(L^2(\tilde{G}))$  a unitary representation of  $\tilde{G}$  on  $\mathcal{H} = L^2(\tilde{G})$

Let  $\mathcal{V}_\Phi$  be the corresponding  $U$ -invariant subspace of  $L^2(\tilde{G})$ ; for any  $f \in \mathcal{V}_\Phi$  we define, from  $M$  fixed points  $t_m \in \tilde{G}$ ,  $m = 1, 2, \dots, M$ , its samples as

$$\mathcal{L}_m f(g) := [U(-g)f](t_m), \quad g \in G, \quad m = 1, 2, \dots, M$$

The associated matrix  $A$  has entries  $a_{m,n}(g) = [U(-g)\varphi_n](t_m)$ ,  $g \in G$ . If the functions  $[U(g)\varphi_n](t)$ ,  $g \in G$  and  $n = 1, 2, \dots, N$ , are **continuous** on  $\tilde{G}$ , and the condition

$$\sup_{t \in \tilde{G}} \sum_{g \in G} |[U(g)\varphi_n](t)|^2 < +\infty, \quad n = 1, 2, \dots, N$$

holds, then the subspace  $\mathcal{V}_\Phi$  is a reproducing kernel Hilbert space of **continuous bounded functions** in  $L^2(\tilde{G})$

► Whenever  $\mathcal{H} = L^2(\mathbb{R}^d)$  and  $[U(p)f](t) := f(t - p)$ ,  $t \in \mathbb{R}^d$ ,  $p \in \mathbb{Z}^d$ , the samples read

$$\mathcal{L}_m f(p) = [U(-p)f](t_m) = f(p + t_m), \quad p \in \mathbb{Z}^d \text{ and } m = 1, 2, \dots, M$$

Under hypotheses in the sampling theorem on the matrix  $A = [a_{m,n}]$ , where  $a_{m,n}(p) = \varphi_n(t_m + p)$ , we obtain in the shift-invariant subspace  $V_{\Phi}^2$  of  $L^2(\mathbb{R}^d)$  the sampling formula

$$f(t) = \sum_{m=1}^M \sum_{p \in \mathbb{Z}^d} f(p + t_m) S_m(t - p), \quad t \in \mathbb{R}^d$$

holds, for some functions  $S_m \in V_{\Phi}^2$ ,  $m = 1, 2, \dots, M$ . The convergence of the series in  $L^2(\mathbb{R}^d)$ -norm implies pointwise convergence which is absolute and uniform on  $\mathbb{R}^d$

► In the example of the quasi regular representation of the crystallographic group  $\mathcal{C}_{P,\Gamma} = P\mathbb{Z}^d \rtimes_{\sigma} \Gamma$ , for each  $f \in \mathcal{V}_{\varphi}$  the defined samples read

$$\mathcal{L}_m f(p) = [U(-p, I)f](t_m) = f(p + t_m), \quad p \in P\mathbb{Z}^d \text{ and } m = 1, 2, \dots, M$$

Under hypotheses in the sampling theorem on the matrix  $A = [a_{m,n}]$ , where  $a_{m,n}(p) = \varphi[\gamma_n^{\top}(t_m - p)]$ , there exist  $M$  functions  $S_m \in \mathcal{V}_{\varphi}$ ,  $m = 1, 2, \dots, M$ , such that, for each  $f \in \mathcal{V}_{\varphi}$ , the sampling formula

$$f(t) = \sum_{m=1}^M \sum_{p \in P\mathbb{Z}^d} f(p + t_m) S_m(t - p), \quad t \in \mathbb{R}^d$$

holds. The convergence of the series in the  $L^2(\mathbb{R}^d)$ -norm sense implies pointwise convergence which is absolute and uniform on  $\mathbb{R}^d$ .

## Sampling in a subgroup $H$ of $G$

Let  $H$  be a subgroup of  $G$  with finite index  $L$ . Then,

$$G = (g_1 + H) \cup (g_2 + H) \cup \cdots \cup (g_L + H)$$

with  $(g_l + H) \cap (g_{l'} + H) = \emptyset$ ,  $l \neq l'$ , for a fix set  $\{g_0, g_1, \dots, g_L\}$  of representatives of the cosets of  $H$ . For  $\mathcal{V}_\Phi$  we have

$$\begin{aligned}\mathcal{V}_\Phi &= \left\{ \sum_{n=1}^N \sum_{g \in G} x_n(g) U(g) \varphi_n : x_n \in \ell^2(G) \right\} \\ &= \left\{ \sum_{n=1}^N \sum_{l=1}^L \sum_{h \in H} \underbrace{x_n(g_l + h)}_{x_{nl}(h)} \underbrace{U(g_l + h) \varphi_n}_{U(h) \underbrace{U(g_l) \varphi_n}_{\varphi_{nl}}} \right\}\end{aligned}$$

$$(x_{11}(h), \dots, x_{1L}(h), x_{21}(h), \dots, x_{2L}(h), \dots, x_{N1}(h), \dots, x_{NL}(h))^T \in \ell^2_{NL}(H)$$

(here we consider a new index  $nl$ , from 11 to  $NL$ )



► In  $\mathcal{H} := L^2(\mathbb{R}^d)$ ,  $G := \mathbb{Z}^d$  and  $[U(p)f](t) := f(t - p)$ ,  $t \in \mathbb{R}^d$  and  $p \in \mathbb{Z}^d$ , consider  $H := P\mathbb{Z}^d$  a sublattice in  $\mathbb{Z}^d$  where  $P$  denotes a  $d \times d$  matrix of integer entries with positive determinant  $L := \det P$ .

Average sampling in the shift-invariant subspace  $V_{\Phi}^2$  has the matrix  $A = [a_{m,nl}]$ , where  $a_{m,nl}(p) = \langle \varphi_n(t), \psi_m(t - p + gl) \rangle_{L^2(\mathbb{R}^d)}$ ,  $p \in P\mathbb{Z}^d$

The corresponding sampling formula reads

$$f(t) = \sum_{m=1}^M \sum_{p \in P\mathbb{Z}^d} \langle f, \psi_m(\cdot - p) \rangle_{L^2(\mathbb{R}^d)} S_m(t - p), \quad t \in \mathbb{R}^d$$

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

Convergence in the  $L^2(\mathbb{R}^d)$ -norm sense implies pointwise convergence which is absolute and uniform on  $\mathbb{R}^d$

► Whenever  $\mathcal{H} = L^2(\mathbb{R}^d)$   $G := \mathbb{Z}^d$  and  $[U(p)f](t) := f(t - p)$ ,  $t \in \mathbb{R}^d$  and  $p \in \mathbb{Z}^d$ , for  $M$  fixed points  $t_m \in \mathbb{R}^d$ , we consider in the shift-invariant subspace  $V_{\Phi}^2$  the samples

$$\mathcal{L}_m f(p) := [U(-p)f](t_m) = f(p + t_m), \quad p \in P\mathbb{Z}^d, \quad m = 1, 2, \dots, M$$

This pointwise sampling has matrix  $A = [a_{m,nl}]$ , where  $a_{m,nl}(p) = \varphi_n(t_m + p - gl)$ ,  $p \in P\mathbb{Z}^d$ , for  $m = 1, 2, \dots, M$ ,  $n = 1, 2, \dots, N$  and  $l = 1, 2, \dots, L$ .

The corresponding sampling formula reads

$$f(t) = \sum_{m=1}^M \sum_{p \in P\mathbb{Z}^d} f(p + t_m) S_m(t - p), \quad t \in \mathbb{R}^d.$$

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

Convergence of the series in the  $L^2(\mathbb{R}^d)$ -norm sense implies pointwise convergence which is absolute and uniform on  $\mathbb{R}^d$

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