

# Frame theory from signal processing and back again – a sampling

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# Outline

- 1 The narrow-band ambiguity function (with Robert L. Benedetto and Joseph Woodworth)
  - NWC Applications: Waveform design and radar
- 2 Ambiguity functions for vector-valued data (with Travis D. Andrews and Jeffrey J. Donatelli)
  - NWC Applications: Multi-sensor environments and MIMO
  - Set-up and problem
  - Group frame multiplications
- 3 Graph uncertainty principles (with Paul Koprowski)
  - NWC Applications: Non-linear spectral methods for dimension reduction and classification

## Outline continued

- 4 Balayage and STFT frame inequalities (with E. Au-Yeung)
  - NWC Applications: Non-uniform sampling and super-resolution imaging
- 5 Quantum detection (with Andrew Kebo)
  - NWC Applications: Finite frames and probabilistic frames in terms of quantum detection and POVMs
- 6 Reactive sensing (with Michael Dellomo)
  - NWC Applications: Engine diagnosis with disabled sensors by mean of multiplicative frames, whose factors account for parameter intensity and sensor sensitivity

- Let  $H$  be a separable Hilbert space, e.g.,  $H = L^2(\mathbb{R}^d)$ ,  $\mathbb{R}^d$ , or  $\mathbb{C}^d$ .
- $F = \{x_n\} \subseteq H$  is a *frame* for  $H$  if

$$\exists A, B > 0 \text{ such that } \forall x \in H, \quad A\|x\|^2 \leq \sum |\langle x, x_n \rangle|^2 \leq B\|x\|^2.$$

## Theorem

If  $F = \{x_n\} \subseteq H$  is a frame for  $H$  then

$$\forall x \in H, \quad x = \sum \langle x, S^{-1}x_n \rangle x_n = \sum \langle x, x_n \rangle S^{-1}x_n,$$

where  $S : H \rightarrow H$ ,  $x \mapsto \sum \langle x, x_n \rangle x_n$  is well-defined.

- Frames are a natural tool for dealing with numerical stability, overcompleteness, noise reduction, and robust representation problems.

- THE NARROW BAND AMBIGUITY FUNCTION

# Ambiguity function and STFT

- Woodward's (1953) *narrow band cross-correlation ambiguity function* of  $v, w$  defined on  $\mathbb{R}^d$  :

$$A(v, w)(t, \gamma) = \int v(s+t) \overline{w(s)} e^{-2\pi i s \cdot \gamma} ds.$$

- The *STFT* of  $v$  :  $V_w v(t, \gamma) = \int v(x) \overline{w(x-t)} e^{-2\pi i x \cdot \gamma} dx$ .
- $A(v, w)(t, \gamma) = e^{2\pi i t \cdot \gamma} V_w v(t, \gamma)$ .
- The *narrow band ambiguity function*  $A(v)$  of  $v$  :

$$A(v)(t, \gamma) = A(v, v)(t, \gamma) = \int v(s+t) \overline{v(s)} e^{-2\pi i s \cdot \gamma} ds$$

# The discrete periodic ambiguity function

- Given  $u : \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}$ .
- The *discrete periodic ambiguity function*,

$$A(u) : \mathbb{Z}/N\mathbb{Z} \times \mathbb{Z}/N\mathbb{Z} \longrightarrow \mathbb{C},$$

of  $u$  is

$$A(u)(m, n) = \frac{1}{N} \sum_{k=0}^{N-1} u[m+k] \overline{u[k]} e^{-2\pi i kn/N}.$$

# CAZAC sequences

- $u : \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}$  is  
*Constant Amplitude Zero Autocorrelation (CAZAC)* if

$$\forall m \in \mathbb{Z}/N\mathbb{Z}, \quad |u[m]| = 1, \quad (\text{CA})$$

and

$$\forall m \in \mathbb{Z}/N\mathbb{Z} \setminus \{0\}, \quad A(u)(m, 0) = 0. \quad (\text{ZAC})$$

- Are there only finitely many non-equivalent CAZAC sequences?
  - "Yes" for  $N$  prime and "No" for  $N = MK^2$ ,
  - Generally unknown for  $N$  square free and not prime.

# Björck CAZAC sequences

Let  $p$  be a prime number, and  $\left(\frac{k}{p}\right)$  the *Legendre symbol*.

A *Björck CAZAC sequence* of length  $p$  is the function  $b_p : \mathbb{Z}/p\mathbb{Z} \rightarrow \mathbb{C}$  defined as

$$b_p[k] = e^{i\theta_p(k)}, \quad k = 0, 1, \dots, p-1,$$

where, for  $p = 1 \pmod{4}$ ,

$$\theta_p(k) = \arccos\left(\frac{1}{1 + \sqrt{p}}\right) \left(\frac{k}{p}\right),$$

and, for  $p = 3 \pmod{4}$ ,

$$\theta_p(k) = \frac{1}{2} \arccos\left(\frac{1-p}{1+p}\right) [(1 - \delta_k) \left(\frac{k}{p}\right) + \delta_k].$$

$\delta_k$  is the Kronecker delta symbol.

# Björck CAZAC discrete periodic ambiguity function

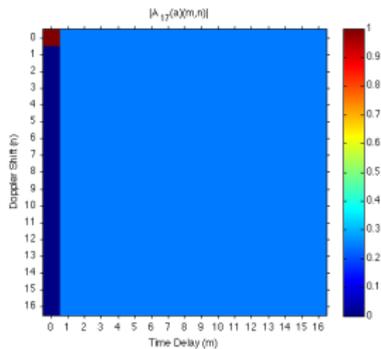
Let  $A(b_p)$  be the Björck CAZAC discrete periodic ambiguity function defined on  $\mathbb{Z}/p\mathbb{Z} \times \mathbb{Z}/p\mathbb{Z}$ .

Theorem (J. and R. Benedetto and J. Woodworth)

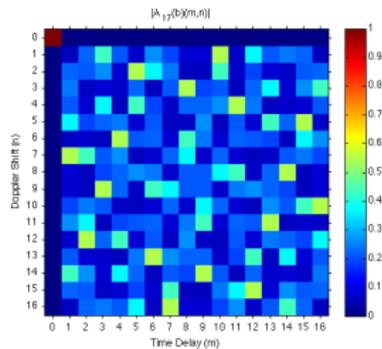
$$|A(b_p)(m, n)| \leq \frac{2}{\sqrt{p}} + \frac{4}{p}$$

for all  $(m, n) \in \mathbb{Z}/p\mathbb{Z} \times \mathbb{Z}/p\mathbb{Z} \setminus (0, 0)$ .

- The proof is at the level of Weil's proof of the Riemann hypothesis for finite fields and depends on Weil's exponential sum bound.
- Elementary construction/coding and intricate combinatorial/geometrical patterns.



(a)



(b)

**Figure :** Absolute value of the ambiguity functions of the Alltop and Björck sequences with  $N = 17$ .

# Problems and remarks

- For given CAZACs  $u_p$  of prime length  $p$ , estimate minimal local behavior  $|A(u_p)|$ . For example, with  $b_p$  we know that the lower bounds of  $|A(b_p)|$  can be much smaller than  $1/\sqrt{p}$ , making them more useful in a host of mathematical problems, cf. Welch bound.
- Even more, construct all CAZACs of prime length  $p$ .
- Optimally small coherence of  $b_p$  allows for computing sparse solutions of Gabor matrix equations by greedy algorithms such as OMP.

- AMBIGUITY FUNCTIONS FOR VECTOR-VALUED DATA

# Modeling for multi-sensor environments

- Multi-sensor environments and vector sensor and MIMO capabilities and modeling.
- Vector-valued DFTs
- Discrete time data vector  $u(k)$  for a  $d$ -element array,

$$k \mapsto u(k) = (u_0(k), \dots, u_{d-1}(k)) \in \mathbb{C}^d.$$

We can have  $\mathbb{R}^N \rightarrow GL(d, \mathbb{C})$ , or even more general.

# Ambiguity functions for vector-valued data

- Given  $u : \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}^d$ .
- For  $d = 1$ ,  $A(u) : \mathbb{Z}/N\mathbb{Z} \times \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}$  is

$$A(u)(m, n) = \frac{1}{N} \sum_{k=0}^{N-1} u(m+k) \overline{u(k)} e^{-2\pi i kn/N}.$$

## Goal

Define the following in a meaningful, computable way:

- Generalized  $\mathbb{C}$ -valued periodic ambiguity function  $A^1(u) : \mathbb{Z}/N\mathbb{Z} \times \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}$
- $\mathbb{C}^d$ -valued periodic ambiguity function  $A^d(u)$ .

The STFT is the *guide* and the *theory of frames* is the technology to obtain the goal.

# Preliminary multiplication problem

- Given  $u : \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}^d$ .
- If  $d = 1$  and  $e_n = e^{2\pi in/N}$ , then

$$A(u)(m, n) = \frac{1}{N} \sum_{k=0}^{N-1} \langle u(m+k), u(k)e_{nk} \rangle.$$

## Preliminary multiplication problem

To characterize sequences  $\{\varphi_k\} \subseteq \mathbb{C}^d$  and compatible multiplications  $*$  and  $\bullet$  so that

$$A^1(u)(m, n) = \frac{1}{N} \sum_{k=0}^{N-1} \langle u(m+k), u(k) * \varphi_{n \bullet k} \rangle \in \mathbb{C}$$

is a meaningful and well-defined *ambiguity function*. This formula is clearly motivated by the STFT.

# $A^1(u)$ for DFT frames

- Given  $u : \mathbb{Z}/N\mathbb{Z} \rightarrow \mathbb{C}^d$ ,  $d \leq N$ .
- Let  $\{\varphi_k\}_{k=0}^{N-1}$  be a DFT frame for  $\mathbb{C}^d$ , let  $*$  be componentwise multiplication in  $\mathbb{C}^d$  with a factor of  $\sqrt{d}$ , and let  $\bullet = +$  in  $\mathbb{Z}/N\mathbb{Z}$ .

In this case  $A^1(u)$  is well-defined by

$$\begin{aligned} A^1(u)(m, n) &= \frac{1}{N} \sum_{k=0}^{N-1} \langle u(m+k), u(k) * \varphi_{n \bullet k} \rangle \\ &= \frac{d}{N^2} \sum_{k=0}^{N-1} \sum_{j=0}^{N-1} \langle \varphi_j, u(k) \rangle \langle u(m+k), \varphi_{j+nk} \rangle. \end{aligned}$$

# $A^1(u)$ for cross product frames

- Take  $* : \mathbb{C}^3 \times \mathbb{C}^3 \rightarrow \mathbb{C}^3$  to be the cross product on  $\mathbb{C}^3$  and let  $\{i, j, k\}$  be the standard basis.
- $i * j = k, j * i = -k, k * i = j, i * k = -j, j * k = i, k * j = -i,$   
 $i * i = j * j = k * k = 0.$   $\{0, i, j, k, -i, -j, -k, \}$  is a tight frame for  $\mathbb{C}^3$  with frame constant 2. Let

$$\varphi_0 = 0, \varphi_1 = i, \varphi_2 = j, \varphi_3 = k, \varphi_4 = -i, \varphi_5 = -j, \varphi_6 = -k.$$

- The index operation corresponding to the frame multiplication is the non-abelian operation  $\bullet : \mathbb{Z}_7 \times \mathbb{Z}_7 \rightarrow \mathbb{Z}_7$ , where  
 $1 \bullet 2 = 3, 2 \bullet 1 = 6, 3 \bullet 1 = 2, 1 \bullet 3 = 5, 2 \bullet 3 = 1, 3 \bullet 2 = 4$ , etc.
- We can write the cross product as

$$u \times v = u * v = \frac{1}{2^2} \sum_{s=1}^6 \sum_{t=1}^6 \langle u, \varphi_s \rangle \langle v, \varphi_t \rangle \varphi_{s \bullet t}.$$

- Consequently,  $A^1(u)$  is well-defined.

Generalize to quaternion groups, order 8 and beyond.

# Frame multiplication

## Definition (Frame multiplication)

Let  $\mathcal{H}$  be a finite dimensional Hilbert space over  $\mathbb{C}$ , and let  $\Phi = \{\varphi_j\}_{j \in J}$  be a frame for  $\mathcal{H}$ . Assume  $\bullet : J \times J \rightarrow J$  is a binary operation. The mapping  $\bullet$  is a *frame multiplication* for  $\Phi$  if it extends to a bilinear product  $*$  on all of  $\mathcal{H}$ .

- The mapping  $\bullet$  is a frame multiplication for  $\Phi$  if and only if there exists a bilinear product  $*$  :  $\mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}$  such that

$$\forall j, k \in J, \quad \varphi_j * \varphi_k = \varphi_{j \bullet k}.$$

- There are frames with no frame multiplications.

- Slepian (1968) - *group codes*.
- Forney (1991) - *geometrically uniform* signal space codes.
- Bölcskei and Eldar (2003) - *geometrically uniform* frames.
- Han and Larson (2000) - *frame bases and group representations*.
- Zimmermann (1999), Pfander (1999), Casazza and Kovacević (2003), Strohmer and Heath (2003), Vale and Waldron (2005), Hirn (2010), Chien and Waldron (2011) - *harmonic frames*.
- Han (2007), Vale and Waldron (2010) - *group frames, symmetry groups*.

# Harmonic frames

- $(\mathcal{G}, \bullet) = \{g_1, \dots, g_N\}$  abelian group with  $\widehat{\mathcal{G}} = \{\gamma_1, \dots, \gamma_N\}$ .
- $N \times N$  matrix with  $(j, k)$  entry  $\gamma_k(g_j)$  is *character table* of  $\mathcal{G}$ .
- $K \subseteq \{1, \dots, N\}$ ,  $|K| = d \leq N$ , and columns  $k_1, \dots, k_d$ .

## Definition

Given  $U \in \mathcal{U}(\mathbb{C}^d)$ . The *harmonic frame*  $\Phi = \Phi_{\mathcal{G}, K, U}$  for  $\mathbb{C}^d$  is

$$\Phi = \{U((\gamma_{k_1}(g_j), \dots, \gamma_{k_d}(g_j))) : j = 1, \dots, N\}.$$

Given  $\mathcal{G}$ ,  $K$ , and  $U = I$ .  $\Phi$  is the *DFT – FUNTF* on  $\mathcal{G}$  for  $\mathbb{C}^d$ . Take  $\mathcal{G} = \mathbb{Z}/N\mathbb{Z}$  for usual *DFT – FUNTF* for  $\mathbb{C}^d$ .

## Definition

Let  $(\mathcal{G}, \bullet)$  be a finite group, and let  $\mathcal{H}$  be a finite dimensional Hilbert space. A finite tight frame  $\Phi = \{\varphi_g\}_{g \in \mathcal{G}}$  for  $\mathcal{H}$  is a *group frame* if there exists

$$\pi : \mathcal{G} \rightarrow \mathcal{U}(\mathcal{H}),$$

a unitary representation of  $\mathcal{G}$ , such that

$$\forall g, h \in \mathcal{G}, \quad \pi(g)\varphi_h = \varphi_{g \bullet h}.$$

Harmonic frames are group frames.

## Theorem (Abelian frame multiplications – 1)

Let  $(\mathcal{G}, \bullet)$  be a finite abelian group, and let  $\Phi = \{\varphi_g\}_{g \in \mathcal{G}}$  be a tight frame for  $\mathcal{H}$ . Then  $\bullet$  defines a frame multiplication for  $\Phi$  if and only if  $\Phi$  is a group frame.

## Theorem (Abelian frame multiplications – 2)

Let  $(\mathcal{G}, \bullet)$  be a finite abelian group, and let  $\Phi = \{\varphi_g\}_{g \in \mathcal{G}}$  be a tight frame for  $\mathbb{C}^d$ . If  $\bullet$  defines a frame multiplication for  $\Phi$ , then  $\Phi$  is unitarily equivalent to a harmonic frame and there exists  $U \in \mathcal{U}(\mathbb{C}^d)$  and  $c > 0$  such that

$$cU(\varphi_g * \varphi_h) = cU(\varphi_g) cU(\varphi_h),$$

where the product on the right is vector pointwise multiplication and  $*$  is defined by  $(\mathcal{G}, \bullet)$ , i.e.,  $\varphi_g * \varphi_h := \varphi_{g \bullet h}$ .

- There is an analogous characterization of frame multiplication for non-abelian groups (T. Andrews).
- Consequently, vector-valued ambiguity functions  $A^d(u)$  exist for functions  $u$  on a finite dimensional Hilbert space  $\mathcal{H}$  if frame multiplication is well-defined for a given tight frame for  $\mathcal{H}$  and a given finite group  $\mathcal{G}$ .
- It remains to extend the theory to infinite Hilbert spaces and groups.
- It also remains to extend the theory to the non-group case, e.g., our cross product example.

- GRAPH UNCERTAINTY PRINCIPLES

# Uncertainty principles – 1

The Heisenberg uncertainty principle inequality is

$$\forall f \in L^2(\mathbb{R}), \quad \|f\|_{L^2(\mathbb{R})}^2 \leq 4\pi \|t f(t)\|_{L^2(\mathbb{R})} \left\| \gamma \hat{f}(\gamma) \right\|_{L^2(\hat{\mathbb{R}})}.$$

Additively, we have

$$\forall f \in L^2(\mathbb{R}), \quad \|f\|_{L^2(\mathbb{R})}^2 \leq 2\pi \left( \|t f(t)\|_{L^2(\mathbb{R})}^2 + \left\| \gamma \hat{f}(\gamma) \right\|_{L^2(\hat{\mathbb{R}})}^2 \right).$$

Equivalently, for  $f \in \mathcal{S}(\mathbb{R})$ ,

$$\|f\|_{L^2(\mathbb{R})}^2 \leq \left\| \hat{f}' \right\|_{L^2(\hat{\mathbb{R}})}^2 + \|f'\|_{L^2(\mathbb{R})}^2.$$

We shall extend this inequality to graphs.

## Uncertainty principles – 2

- In signal processing, uncertainty principles dictate the trade off between high spectral and high temporal accuracy, establishing limits on the extent to which the “instantaneous frequency” of a signal can be measured (Gabor, 1946)
- Weighted, Euclidean, LCAG, non- $L^2$  uncertainty principles, proved by Fourier weighted norm inequalities, e.g., Plancherel, generalizations of Hardy’s inequality, e.g., integration by parts, and Hölder (alas).
- DFT: Chebatorov, Grünbaum, Donoho and Stark, Tao.
- Generalize the latter to graphs.

# Graph theory – background

- Problem: propose, prove, and understand uncertainty principle inequalities for graphs, see A. Agaskar and Y. M. Lu on *A spectral graph uncertainty principle*
- Generally: There is no obvious solution because of the loss on general graphs of the cyclic structure associated with the DFT.
- Locally: Radar/Lidar data analysis at NWC uses non-linear spectral kernel methods, with *essential* graph theoretic components for dimension reduction and remote sensing.

## Definition

A *graph* is  $G = \{V, \mathbf{E} \subseteq V \times V, w\}$  consisting of a set  $V$  called vertices, a set  $\mathbf{E}$  called edges, and a weight function

$$w : V \times V \longrightarrow [0, \infty).$$

Write  $V = \{v_j\}_{j=0}^{N-1}$  and keep the ordering fixed, but arbitrary.

# Graph theory – assumptions

- For any  $(v_i, v_j) \in V \times V$  we have

$$w(v_i, v_j) = \begin{cases} 0 & \text{if } (v_i, v_j) \in \mathbf{E}^c \\ c > 0 & \text{if } (v_i, v_j) \in \mathbf{E}. \end{cases}$$

- $G$  is undirected, i.e.,  $w(v_i, v_j) = w(v_j, v_i)$ .
- $w(v_i, v_i) = 0$ , i.e.,  $G$  has no loops.
- $G$  is connected, i.e., given any  $v_i$  and  $v_j$ , there exists at most one edge between them, and there exists a sequence of vertices  $\{v_k\}$ ,  $k = 0, \dots, d \leq |V| = N$ , such that

$$(v_i, v_0), (v_0, v_1), \dots, (v_d, v_j) \in \mathbf{E}.$$

- $G$  is *unit weighted* if  $w$  takes only the values 0 and 1.

# Graph Laplacian

- $N \times N$  symmetric *adjacency matrix*,  $A$ , for  $G$  :

$$A = (A_{ij}) = (w(v_i, v_j)).$$

- The *degree matrix*,  $D$ , is the  $N \times N$  diagonal matrix,

$$D = \text{diag} \left( \sum_{j=0}^{N-1} A_{0j}, \sum_{j=0}^{N-1} A_{1j}, \dots, \sum_{j=0}^{N-1} A_{(N-1)j} \right).$$

- The *graph Laplacian*,

$$L = D - A,$$

is the  $N \times N$  symmetric, positive semi-definite matrix, with real ordered eigenvalues  $0 = \lambda_0 \leq \dots \leq \lambda_{N-1}$  and orthonormal eigenbasis,  $\{\chi_j\}_{j=0}^{N-1}$ , for  $\mathbb{R}^N$ .

# Graph Fourier transform

- Formally, the Fourier transform  $\hat{f}$  at  $\gamma$  of  $f$  defined on  $\mathbb{R}$  is the inner product of  $f$  with the complex exponentials, that are the eigenfunctions of the Laplacian operator  $\frac{d^2}{dt^2}$  on  $\mathbb{R}$ .
- Thus, define the *graph Fourier transform*,  $\hat{f}$ , of  $f \in \ell^2(G)$  in the graph Laplacian eigenbasis:

$$\hat{f}[j] = \langle \chi_j, f \rangle, \quad j = 0, \dots, N-1.$$

If

$$\chi = [\chi_0, \chi_1, \dots, \chi_{N-1}],$$

then  $\hat{f} = \chi^* f$ , and, since  $\chi$  is unitary, we have the *inversion formula*:

$$f = \chi \chi^* f = \chi \hat{f}.$$

# Difference operator for graphs

The *difference operator*,

$$D_r : \ell^2(G) \longrightarrow \mathbb{R}^{|\mathbf{E}|},$$

with coordinate values representing the change in  $f$  over each edge, is defined by

$$(D_r f)[k] = (f[j] - f[i]) (w(e_k))^{1/2},$$

where  $e_k = (v_j, v_i)$  and  $j < i$ .

- $D_r$  can be defined by the *incidence matrix* of  $G$ .
- If  $G$  is a unit weighted circulant graph, then  $D_r$  is the intuitive difference operator of Lammers and Maeser.

# Difference uncertainty principle for graphs

## Theorem

Let  $G$  be a connected and undirected graph. Then,

$$\forall f \in \ell^2(G), \quad 0 < \tilde{\lambda}_0 \|f\|^2 \leq \|D_r f\|^2 + \left\| D_r \hat{f} \right\|^2 \leq \tilde{\lambda}_{N-1} \|f\|^2,$$

where

$$\Delta = \text{diag}\{\lambda_0, \dots, \lambda_{N-1}\}$$

and where  $0 < \tilde{\lambda}_0 \leq \tilde{\lambda}_1 \leq \dots \leq \tilde{\lambda}_{N-1}$  are the eigenvalues of  $L + \Delta$ .  
The bounds are sharp.

# Frame difference uncertainty principle for graphs

$\{e_j\}_{j=0}^{N-1} \subseteq \mathbb{C}^d$  is a *frame* for  $\mathbb{C}^d$  if

$$\exists 0 < A \leq B \text{ such that } \forall f \in \mathbb{C}^d, \quad 0 < A \|f\|^2 \leq \sum_{j=0}^{N-1} |\langle f, e_j \rangle|^2 \leq B \|f\|^2.$$

- If  $A = B = 1$  then the frame is a *Parseval frame*.
- Define the  $d \times N$  matrix  $E = [e_0, e_1, \dots, e_{N-1}]$ , where  $\{e_j\}_{j=0}^{N-1}$  is a Parseval frame for  $\mathbb{C}^d$ . Then  $EE^* = I_{d \times d}$ .

## Theorem

Let  $G$  be a connected and undirected graph. Then, for every  $d \times N$  Parseval frame  $E$ ,

$$\sum_{j=0}^{d-1} \tilde{\lambda}_j \leq \|D_r \chi^* E^*\|_{fr}^2 + \|D_r E^*\|_{fr}^2 \leq \sum_{j=N-d}^{N-1} \tilde{\lambda}_j.$$

The bounds are sharp.

# Feasibility region

The difference operator *feasibility region*  $FR$  is

$$FR = \{(x, y) : \exists f \in \ell^2(G), \|f\| = 1, \text{ such that } \|D_r f\|^2 = x \text{ and } \|D_r \hat{f}\|^2 = y\}.$$

## Theorem

- $FR$  is a closed subset subset of  $[0, \lambda_{N-1}] \times [0, \lambda_{N-1}]$ , where  $\lambda_{N-1}$  is the maximum eigenvalue of the Laplacian  $L$ .
- $(\frac{1}{N} \sum_{j=0}^{N-1} \lambda_j, 0)$  and  $(0, L_{0,0})$  are the only points of  $FR$  on the axes.
- $FR$  is in the half plane defined by  $x + y \geq \tilde{\lambda}_0 > 0$  with equality if and only if  $\hat{f}$  is in the eigenspace associated with  $\tilde{\lambda}_0$ .
- If  $N \geq 3$ , then  $FR$  is a convex region.

# Complete graph

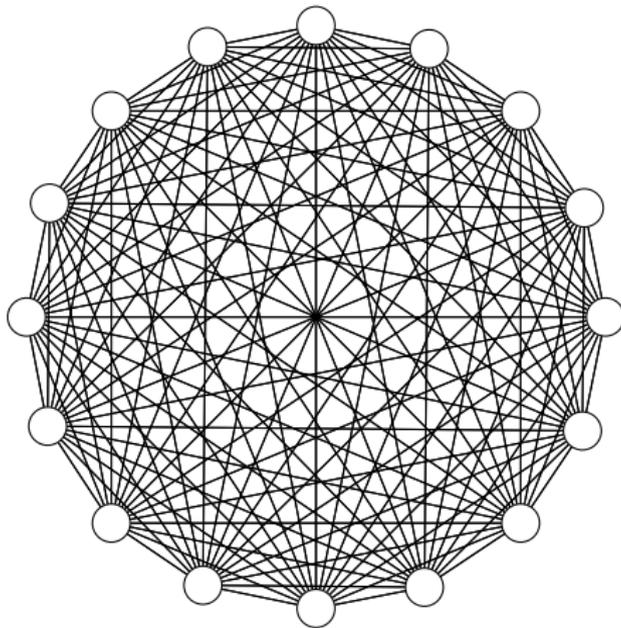
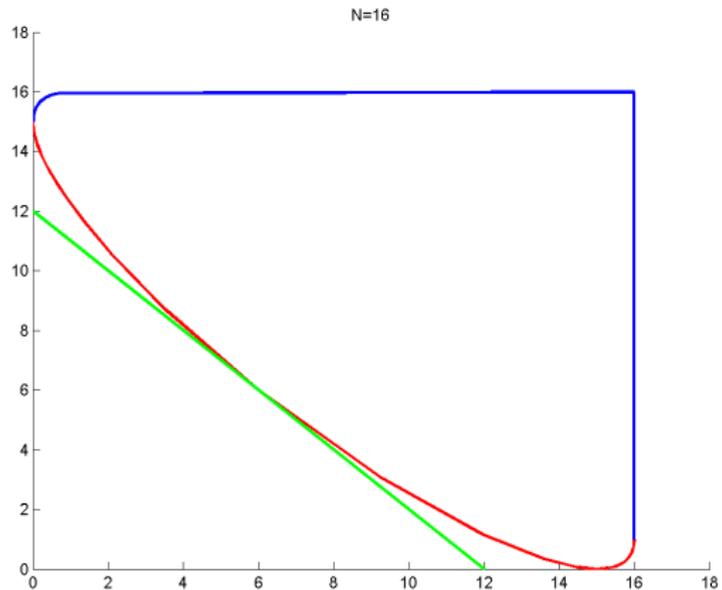


Figure : A unit weighted complete graph with 16 vertices.

# Feasibility region



# Difference uncertainty curve

The *difference uncertainty curve*  $\omega$  is the lower boundary of *FR* defined as

$$\forall x \in [0, \lambda_{N-1}], \quad \omega(x) = \inf_{g \in \ell^2(G)} \langle g, Lg \rangle$$

$$\text{subject to } \langle g, \Delta g \rangle = x.$$

Given  $x \in [0, \lambda_{N-1}]$ .  $g_x \in \ell^2(G)$  *attains the difference uncertainty curve* at  $x$  if, for all  $g$  for which  $\langle g, \Delta g \rangle = x$ , we have

$$\langle g_x, Lg_x \rangle \leq \langle g, Lg \rangle.$$

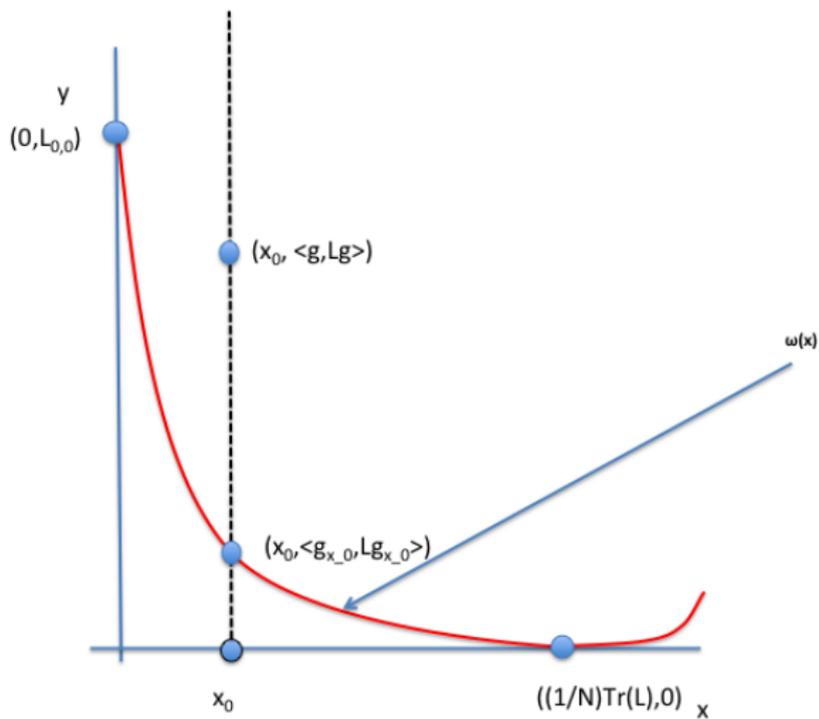


Figure : The difference uncertainty curve (red) for a connected graph  $G$

# Uncertainty curve theorem

## Theorem

A unit normed function  $f \in \ell^2(G)$ , with  $\|D_r f\|^2 = x \in (0, \lambda_{N-1})$ , achieves the uncertainty curve at  $x$  if and only if  $\hat{f}$  is a nonzero eigenfunction for  $K(\alpha) = L - \alpha\Delta$  associated with the minimal eigenvalue of  $K(\alpha)$ , where  $\alpha \in (-\infty, \infty)$ .

# Uncertainty principle problem and comparison

- Lammers and Maeser, Grünbaum, Agaskar and Lu.
- The Agaskar and Lu problem.
- Critical comparison between the graph theoretical feasibility region and the comparable Bell Labs uncertainty principle region.

- BALAYAGE AND STFT FRAME INEQUALITIES

# Balayage and spectral synthesis

## Definition

(Balayage – Beurling) Let  $E \subseteq \mathbb{R}^d$  and  $\Lambda \subseteq \widehat{\mathbb{R}}^d$  be closed sets. *Balayage* is possible for  $(E, \Lambda) \subseteq \mathbb{R}^d \times \widehat{\mathbb{R}}^d$  if

$$\forall \mu \in M_b(\mathbb{R}^d), \exists \nu \in M_b(E) \text{ such that } \widehat{\mu} = \widehat{\nu} \text{ on } \Lambda.$$

Define

$$\mathcal{C}(\Lambda) = \{f \in C_b(\mathbb{R}^d) : \text{supp}(\widehat{f}) \subseteq \Lambda\}.$$

## Definition

(Spectral synthesis) A closed set  $\Lambda \subseteq \widehat{\mathbb{R}}^d$  is a set of *spectral synthesis* (*S-set*) if

$$\forall f \in \mathcal{C}(\Lambda) \text{ and } \forall \mu \in M_b(\mathbb{R}^d), \widehat{\mu} = 0 \text{ on } \Lambda \Rightarrow \int f d\mu = 0.$$

# Balayage and a non-uniform Gabor frame theorem

Let  $\mathcal{S}_0(\mathbb{R}^d)$  be the *Feichtinger algebra*.

## Theorem

Let  $E = \{(s_n, \sigma_n)\} \subseteq \mathbb{R}^d \times \widehat{\mathbb{R}}^d$  be a separated sequence; and let  $\Lambda \subseteq \widehat{\mathbb{R}}^d \times \mathbb{R}^d$  be an S-set of strict multiplicity that is compact, convex, and symmetric about  $0 \in \widehat{\mathbb{R}}^d \times \mathbb{R}^d$ . Assume balayage is possible for  $(E, \Lambda)$ . Given  $g \in L^2(\mathbb{R}^d)$ , such that  $\|g\|_2 = 1$ . Then

$\exists A, B > 0$ , such that  $\forall f \in \mathcal{S}_0(\mathbb{R}^d)$ , for which  $\text{supp}(\widehat{V_g f}) \subseteq \Lambda$ ,

$$A \|f\|_2^2 \leq \sum_{n=1}^{\infty} |V_g f(s_n, \sigma_n)|^2 \leq B \|f\|_2^2.$$

# Balayage and a non-uniform Gabor frame theorem (continued)

## Theorem

Consequently, the frame operator,  $S = S_{g,E}$ , is invertible in  $L^2(\mathbb{R}^d)$ -norm on the subspace of  $\mathcal{S}_0(\mathbb{R}^d)$ , whose elements  $f$  have the property,  $\text{supp}(\widehat{V_g f}) \subseteq \Lambda$ . Further, if  $f \in \mathcal{S}_0(\mathbb{R}^d)$  and  $\text{supp}(\widehat{V_g f}) \subseteq \Lambda$ , then

$$f = \sum_{n=1}^{\infty} \langle f, \tau_{s_n} e_{\sigma_n} g \rangle S_{g,E}^{-1}(\tau_{s_n} e_{\sigma_n} g),$$

where the series converges unconditionally in  $L^2(\mathbb{R}^d)$ .

$E$  does not depend on  $g$ .

- There is a formulation of the non-uniform Gabor frame theorem in terms of the Woodward ambiguity function.
- The theory is also developed for pseudo-differential operators.
- Elementary examples satisfy hypotheses of non-uniform Gabor frame theorem.
- Analogous results, with give and take of hypotheses and conclusions:
  - Gröchenig's theorem involving an analysis of convolution operators on the Heisenberg group;
  - Meyer - Matei theory involving quasi-crystals.

*That's all folks!*