Spatio–spectral limiting on Boolean cubes Jubilee of Fourier Analysis and Applications, NWC at UMD, 2019 joint work with Jeff Hogan





- 1. Review: Time and band limiting: on $\mathbb{R},\,\mathbb{Z}$ and $\mathbb{Z}_{\textit{N}}$
- 2. Spatio-spectral limiting on graphs: definitions
- 3. Hypercube graphs
- 4. Results
- 5. Adjacency maps and invariant subspaces
- 6. Matrix reduction of spatio-spectral limiting operator
- 7. Numerical aspects
- 8. Potential extensions





Fourier transform:
$$\widehat{f}(\xi) = \int_{\mathbb{R}} f(t) \, \mathrm{e}^{-2\pi i t \xi} \, dt$$

Bandlimiting: $P_{\Omega}f(x) = (\widehat{f} \mathbb{1}_{[-\Omega/2, \Omega/2]})^{\vee}(x)$ Time limiting: $(Q_T f)(x) = \mathbb{1}_{[-T,T]}(x) f(x)$

- 1. What are the eigenfunctions of $P_{\Omega}Q_T$?
- 2. What is the distribution of eigenvalues of $P_{\Omega}Q_T$

Eigenvalue distribution:

Approximately $2\Omega T - O(\log(2\Omega T))$ eigenvalues close to one Plunge region of width proportional to $2\Omega T$ Exponential decay of remaining eigenvalues



Spatio-spectral limiting

$P_{\Omega}Q_T$ commutes with

(PDO)
$$(4T^2 - t^2) \frac{d^2}{dt^2} - 2t \frac{d}{dt} - \Omega^2 t^2$$

Eigenfunctions: Prolate Spheroidal Wave Functions (PSWFs) Methods to compute PSWFs based on PDO

¹S. Slepian, Some comments on Fourier analysis, uncertainty and modeling, SIAM Review, 25, 379–393 1983



Spatio-spectral limiting

Finite dimensional analogue: cycle



Discrete setting $\mathbb{Z} \leftrightarrow \mathbb{T}$: Slepian, (1978) DPSS Finite \mathbb{Z}_N setting: Grünbaum (1981), others Results analogous to continuous setting Zhu et al 2018: Non-asymptotic bound on plunge region² Many other developments in time and band limiting since 2000

²Z. Zhu, S. Karnik, M. A. Davenport, J. Romberg, and M. B. Wakin, The Eigenvalue Distribution of Discrete Periodic Time-Frequency Limiting Operators, IEEE Signal Process. Lett., **25**, 95–99, 2018.

Hypercubes: N = 5



Spatio-spectral limiting

Unnormalized Graph Laplacian and Fourier transform $\mathcal{G} = (\mathcal{V}, \mathcal{E})$

$$f: \mathcal{V} \to \mathbb{R}, Lf(v) = \sum_{w \sim v} f(v) - f(w)$$

 $L = D - A$

D: degree of each vertex A: adjacency map (undirected) Graph Fourier transform φ_n : eigenvectors of L.

$$\hat{f}(\lambda_{\ell}) = \langle f, \varphi_{\ell} \rangle$$

Analogue of Q_T : truncation to path neighborhood of a vertex Analogues of P_{Ω} : truncation to span { $\varphi_{\ell} : \lambda_{\ell}$ small} Motivation for GFT (e.g, Sardellitti Barbarossa Di Lorenzo [2016]): identify smooth clusters in vertex data that varies across clusters

Other time-frequency analysis on graphs: Shuman, Ricaud and Vandergheynst [e.g., ACHA 2016], Stanković, Daković and Sedjić [IEEE SP Magazine, 2017]

Our thesis: particular graphs admit concrete analytical expressions

Very particular graphs: Boolean hypercubes

$$\begin{array}{l} \mathcal{B}_{N} = \mathbb{Z}_{2}^{N} \\ \mathcal{B}_{N}: \text{ unweighted metric } \mathcal{C}ayley \text{ graph} \\ v = v_{S} = (\epsilon_{1}, \ldots, \epsilon_{N}), \ S \subset \{1, \ldots, N\}: \ i \in S \Leftrightarrow \epsilon_{i} = 1 \\ L = D - A \\ D = N I_{N} \\ \mathcal{A}_{RS} = 1 \text{ if } R\Delta S \text{ is a singleton} \end{array}$$



Spatio-spectral limiting



Figure: Adjacency matrix for N = 8 in dyadic lexicographic order.

Σ_r : Hamming sphere of radius r: vertices with r one-bits



Spatio-spectral limiting

Historical use: Sampling Known Fourier transform Non-Euclidean geometry

Our thesis: particular graphs admit concrete analytical expressions Accessible generalizations and restrictions: generalized hypercubes, partial cubes Spatio-spectral limiting: Tsitsvero, Barbarossa, Di Lorenzo [2016]: relate properties of compositions *QP* and *PQ* on graphs to (sub)-sampling strategies for recovery of sparse vertex functions.

Sampling of bandlimited vertex functions was developed in the setting of hypercubes by Mansour et al in early 1990s in context of learning (sparse) Boolean functions.

Lemma (Boolean Fourier transform)

Let $H_S(R) = (-1)^{|R \cap S|}$ and $L = L_{\mathcal{B}_N}$ as above. Then H_S is an eigenvector of L with eigenvalue 2|S|.



Figure: Hadamard (Fourier) matrix for N = 8 in dyadic lexicographic order.

Space-limiting matrix
$$Q = Q_K$$
: $Q_{R,S} = \begin{cases} 1, & R = S \& |S| \le K \\ 0, & \text{else} \end{cases}$
Spectrum-limiting matrix $P = P_K$ by $P = \bar{H}Q\bar{H}$

Results: identify eigenvectors of spatio–spectral limiting *PQP* Approach:

- Work in spectral domain: $QPQ = \overline{H}PQP\overline{H}$
- Identify salient invariant subspaces of QPQ
- These subspaces factor
- Reduce to small matrix problem on one of the factors
- Numerical computation via almost commuting operator and power method with a weight

Eigenspaces of spatio-spectral limiting on \mathcal{B}_N

A: adjacency matrix of \mathcal{B}_N (dyadic lexicographic order) $A = A_+ + A_-$: $A_- = A_+^T$; A_+ : lower triangular A_+ maps data on Σ_r to data on Σ_{r+1} : outer adjacency A_- maps data on Σ_r to data on Σ_{r-1} : inner adjacency



Figure: Highlighted: A_- , $\Sigma_3 \rightarrow \Sigma_2$

$$\ell^2(\Sigma_r) = A_+ \ell^2(\Sigma_{r-1}) \oplus \mathcal{W}_r$$

 \mathcal{W}_r : the orthogonal complement of $A_+\ell^2(\Sigma_{r-1})$ inside $\ell^2(\Sigma_r)$.

$$\ell^{2}(\Sigma_{r}) = A_{+}\ell^{2}(\Sigma_{r-1}) \oplus \mathcal{W}_{r} = \cdots = A_{+}^{r}\mathcal{W}_{0} \oplus A_{+}^{r-1}\mathcal{W}_{1} \oplus \cdots \oplus \mathcal{W}_{r}$$

Projection Matrix onto W_r : columns form a Parseval frame



Figure: Matrix of projection onto W_r , N = 8, r = 3.

Spatio-spectral limiting

Theorem Let $V \in W_r$ and k such that r + k < N. Then

$$A_{-}A_{+}^{k+1}V = [(N-2r) + \dots + (N-2(r+k))]A_{+}^{k}V$$

= $(k+1)(N-2r-k)A_{+}^{k}V$
= $m(r,k)A_{+}^{k}V$



Spatio-spectral limiting

$$C = [A_{-}, A_{+}] = A_{-}A_{+} - A_{+}A_{-}: \text{ commutator of } A_{+} \text{ and } A_{-}.$$

Proposition

C is diagonal with
$$C_{RR} = N - 2|R|$$
.

Theorem follows from induction on k

$$V \in \mathcal{V}_r$$
 if $V = \sum_{k=0}^{N-r} c_k A_+^k W$, $W \in \mathcal{W}_r$
Lemma
 A_+ and A_- map \mathcal{V}_r to itself.
Corollary
A maps \mathcal{V}_r to itself. Polynomials $p(A)$ preserve \mathcal{V}_r .

Proposition

The spectrum-limiting operator $P = P_K$ can be expressed as a polynomial p(A) of degree N.

Proof.

$$p_k = \prod_{j=0, j \neq k}^{N} \frac{x - (N - 2j)}{2(j - k)}; \qquad p(x) = \sum_{k=0}^{K} p_k$$

Then P = p(A) as verified on Hadamard basis.

Matrix of Spectral limiting P_K on \mathcal{V}_r

$$egin{aligned} M^P_{(N,K,r)} & ext{ of size } (N-r+1): ext{ represents } P_K ext{ on } \mathcal{V}_r \ P(A^k_+W) &= \sum_{\ell=0}^{N-r} M^P_{(N,K,r)}(k,\ell) A^\ell_+W, \quad (W\in\mathcal{W}_r) \end{aligned}$$

$$PV = \sum_{k=0}^{N-r} d_k A^k_+ W = \sum_{k=0}^{N-r} \sum_{\ell=0}^{N-r} M^P_{(N,K,r)}(k,\ell) c_\ell A^k_+ W \quad (W \in \mathcal{W}_r)$$

$$M^{QPQ}_{(N,K,r)}$$
: $(K - r + 1)$ -principal minor of $M^{P}_{(N,K,r)}$.

$$QPQV = \sum_{k=0}^{K-r} d_k A_+^k W = \sum_{k=0}^{K-r} \sum_{\ell=0}^{K-r} M_{(N,K,r)}^P(k,\ell) c_\ell A_+^k W, \quad (W \in \mathcal{W}_r)$$

Corollary (Coefficient eigenvectors of QPQ)

If $\mathbf{c} = [c_0, \dots c_{K-r}]^T$ is a λ -eigenvector of the principal minor $M_{(N,K,r)}^{QPQ}$ of size (K - r + 1) of the matrix $M_{(N,K,r)}^P$ then $V = \sum_{k=0}^{K-r} c_k A_+^k W$, any $W \in W_r$, is a λ -eigenvector of QPQ and $\overline{H}V$ is a λ -eigenvector of PQP.

Remark (Completeness)

Any eigenvector of QPQ is attached to one of the spaces \mathcal{V}_r

Matrix of $M_{(N,K,r)}^P$ of P by substituting M_A for A in P = p(A)



Figure: Matrices M_A and M^P , N = 9, K = 4, r = 1. (log scale)

Problem: large numbers Spatio-spectral limiting

Inner product on \mathcal{V}_r

$$\langle A_{+}^{k} W_{1}, A_{+}^{k} W_{2} \rangle = w(r, k) \langle W_{1}, W_{2} \rangle$$

$$w(r, k) = \prod_{j=0}^{k-1} m(r, j)$$

$$\left\langle \sum_{k=0}^{N-r} c_k A_+^k W_1, \sum_{k=0}^{N-r} d_k A_+^k W_2 \right\rangle = \left\langle W_1, W_2 \right\rangle \underbrace{\sum_{k=0}^{N-r} c_k d_k w(r,k)}_{\langle \boldsymbol{c}, \boldsymbol{d} \rangle_{W_r}}$$

Proposition

Coefficient eigenvectors of $M_{(N,K,r)}^{QPQ}$ are orthogonal wrt weight $[w(r,0), \ldots, w(r, K+1-r)]$

(BDO)
$$D(\alpha I - T^2)D + \beta T^2$$
.

T: diagonal, sqrts of eigenvalues of L

$$D = \overline{H}T\overline{H}$$
, $\overline{H} = 2^{-N/2}H$
 $D^2 = L$.

Proposition

If $\beta = 2\sqrt{K(K+1)}$ then BDO commutes with P_K . Equivalently, the conjugation of BDO by H commutes with Q_K .

BUT BDO *does not* commute with Q_K

$$M^{\text{HBDO}}(k,\ell) = \begin{cases} (2\sqrt{\ell(\ell-1)} - \beta)m(r,\ell-1-r); \ k = \ell - 1 \ge r \\ 2\ell(\alpha - N) + \beta N; k = \ell \ge r \\ 2\sqrt{\ell(\ell+1)} - \beta; k = \ell + 1; r \le \ell < N \\ 0, \text{ else }. \end{cases}$$

If $\alpha = \beta = 2\sqrt{K(K-1)}$:



Figure: Matrix M^{HBDO} , N = 9, K = 4

Theorem (\mathcal{V}_r is HBDO-invariant)

If $V \in \mathcal{V}_r$, $V = \sum_{k=0}^{N-r} c_k A_+^k W$, then $\text{HBDOV} = \sum_{k=0}^{N-r} d_k A_+^k W$ where $\mathbf{d} = M^{\text{HBDO}} \mathbf{c}$ where $\mathbf{c} = [c_0, \dots, c_{N-r}]^T$.

- ▶ Entries of $M^{\rm QPQ}$ can exceed maxint for moderate sized N
- M^{HBDO} is tridiagonal and eigendecomposition is fine
- ► HBDO and QPQ *almost commute*
- Eigenvectors of HBDO as seeds for weighted power method



Figure: Eigenvectors of PQP, N = 8, K = 3, r = 2. Dotted curves: two different elements W of W_r Dashed curves: corresponding eigenvectors V of QPQSolid curves: Eigenvector HV of PQP for eigenvector V of QPQ Algorithm 1 Adapted power method eigen-decomposition of QPQ

- 1: Inputs: $N, K \in \{0, ..., N\}, r \in \{0, ..., K\}$
- 2: Compute coefficient matrix $M_{(N,K,r)}^{\text{HBDO}}$ of $2^{-N}H\text{BDOH}$ on \mathcal{V}_r
- 3: Compute eigenvectors \boldsymbol{c} of $M_{(N,K,r)}^{\text{HBDO}}$
- 4: Sort eigenvectors $\boldsymbol{c}^k = [c_0^k, \ldots, c_{N-r}^k]$: $c_{K+1}^k = \cdots = c_{N-r}^k = 0$
- 5: Sub M_A for A: Compute $M_{(N,K,r)}^{QPQ}$: principal minor of $M_{(N,K,r)}^P$
- 6: **for** k = 0 to K r **do**
- 7: while stopping criteria = False do
- 8: Apply $M_{(N,K,r)}^{QPQ}$ factor-wise to d^k
- 9: Project output onto $(\operatorname{span}\{\boldsymbol{d}^0,\ldots,\boldsymbol{d}^{k-1}\})^{\perp}$ wrt $\langle\cdot,\cdot\rangle_w$
- 10: Update d^k = normalized projection (wrt $\|\cdot\|_w$)
- 11: end while
- 12: **end for**
- 13: Return: approximate coefficient eigenvectors $d^0, ..., d^{K-r}$ of M^{QPQ} , the matrix of QPQ acting on \mathcal{V}_r .



Figure: Eigenvalues of PQP with multiplicity (60460), N = 20, K = 6.

HAPPY BIRTHDAY JOHNNY!!



https://www.youtube.com/channel/ UCKChX5APWWHOLwu4CVestDA/featured?disable_polymer=1