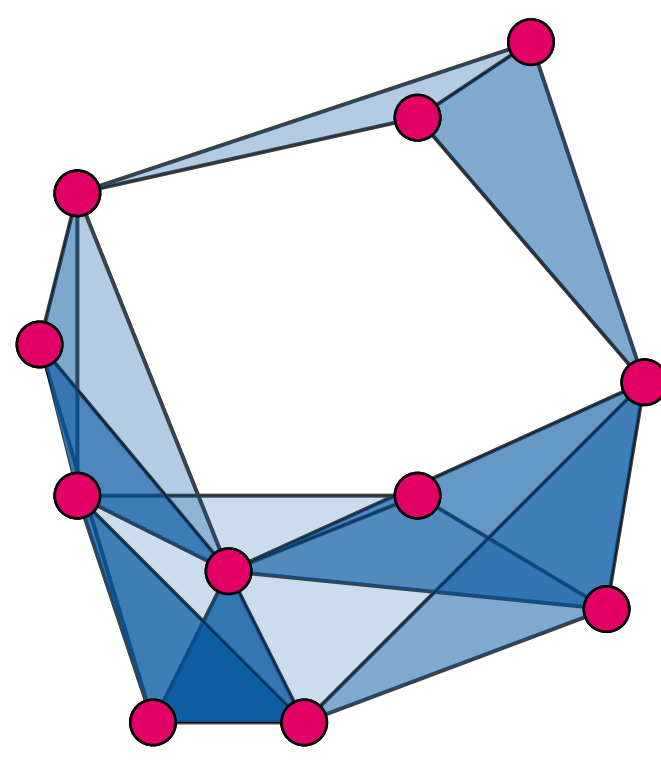


SPARSIFICATION OF SIMPLICIAL COMPLEXES VIA NETWORK DENSITY OF STATES

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Simplicial complexes



Boundary maps (simplex $\rightarrow$ boundary):

$$B_1: \mathcal{V}_1(\mathcal{K}) \rightarrow \mathcal{V}_0(\mathcal{K}), \quad B_2: \mathcal{V}_2(\mathcal{K}) \rightarrow \mathcal{V}_1(\mathcal{K})$$

edges nodes triangles edges

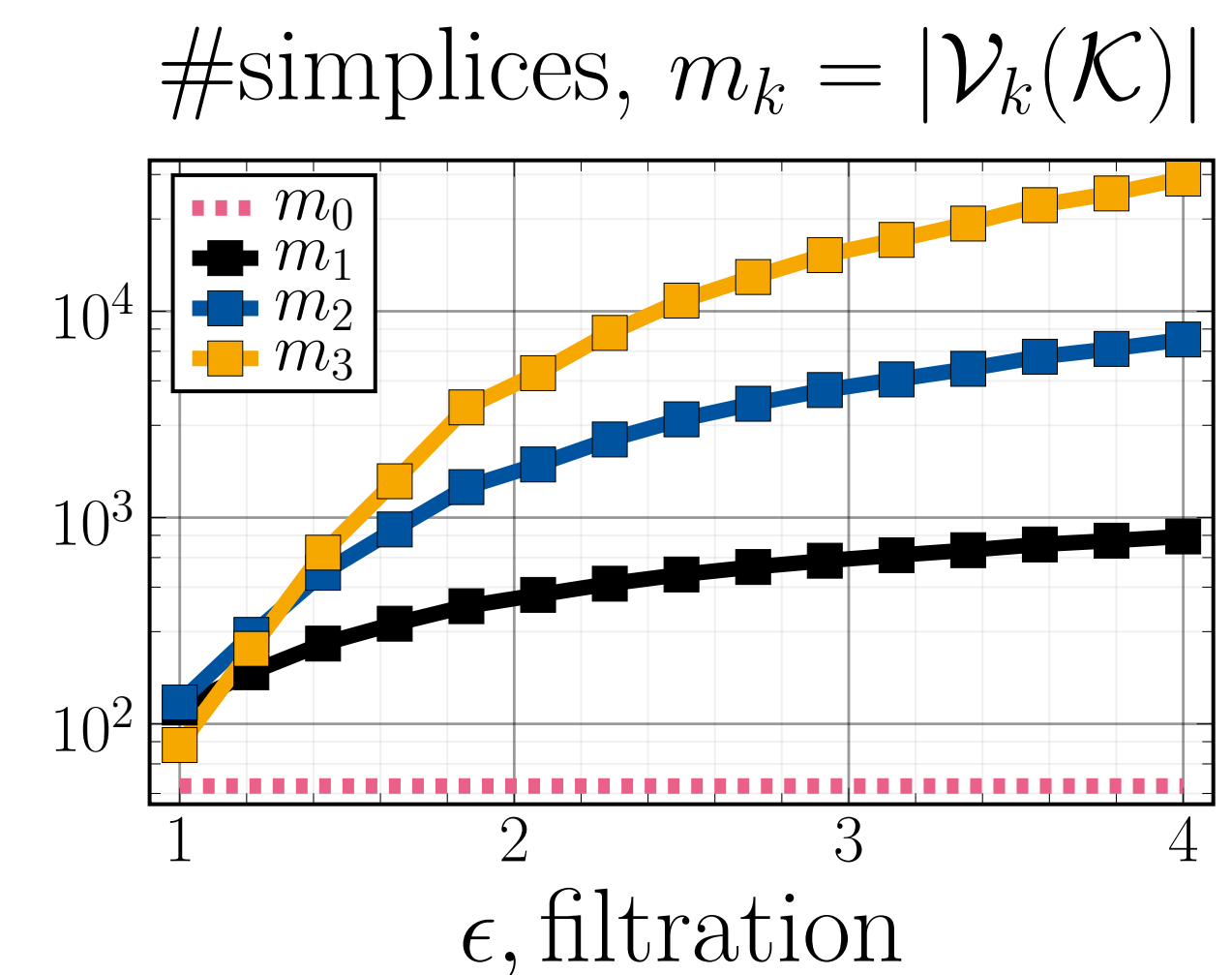
Hodge Laplacians:

$$L_k = \underbrace{B_k^\top B_k}_{\text{down-adjacent } k\text{-simplices}} + \underbrace{B_{k+1} B_{k+1}^\top}_{\text{up-adjacent } k\text{-simplices}} = L_k^\downarrow + L_k^\uparrow$$

Def. collection of simplices closed for face inclusion

$$\mathcal{K} = \left\{ \underbrace{\mathcal{V}_0(\mathcal{K})}_{\text{nodes}}, \underbrace{\mathcal{V}_1(\mathcal{K})}_{\text{edges}}, \underbrace{\mathcal{V}_2(\mathcal{K})}_{\text{triangles}}, \dots \right\}$$

- shift operator for simplicial complexes
- encodes **topological information**
- used in **topological data analysis**, **GNNs**, **random walks**, etc.

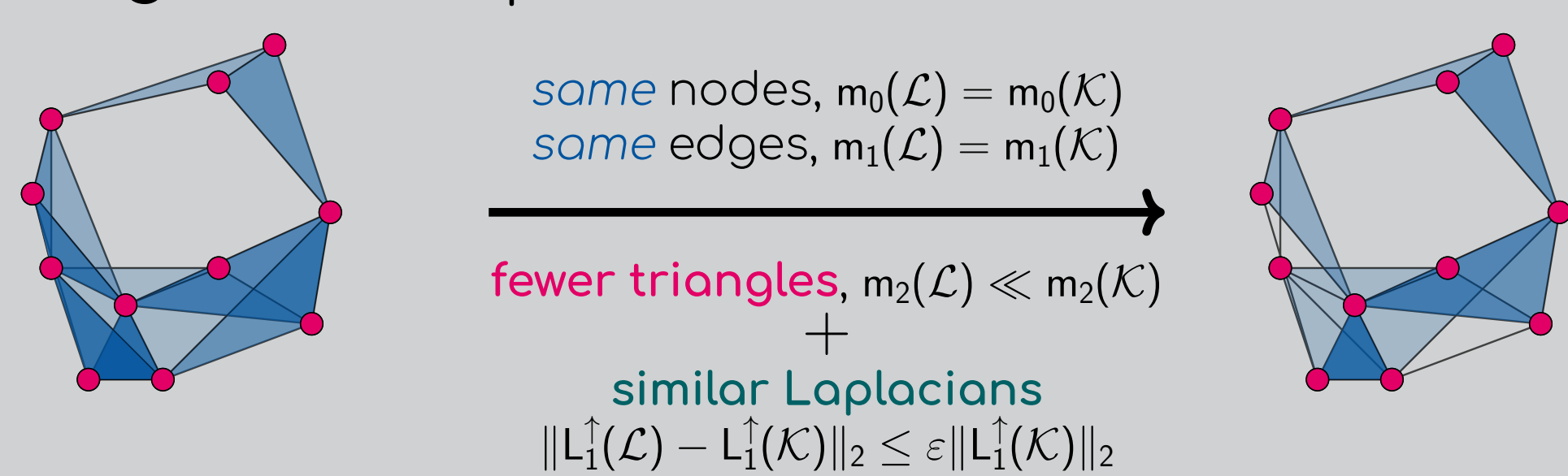


Problem

- number of simplices m_k explodes
- expensive inference for GNNs
- expensive solution $L^\dagger$

Sparsification

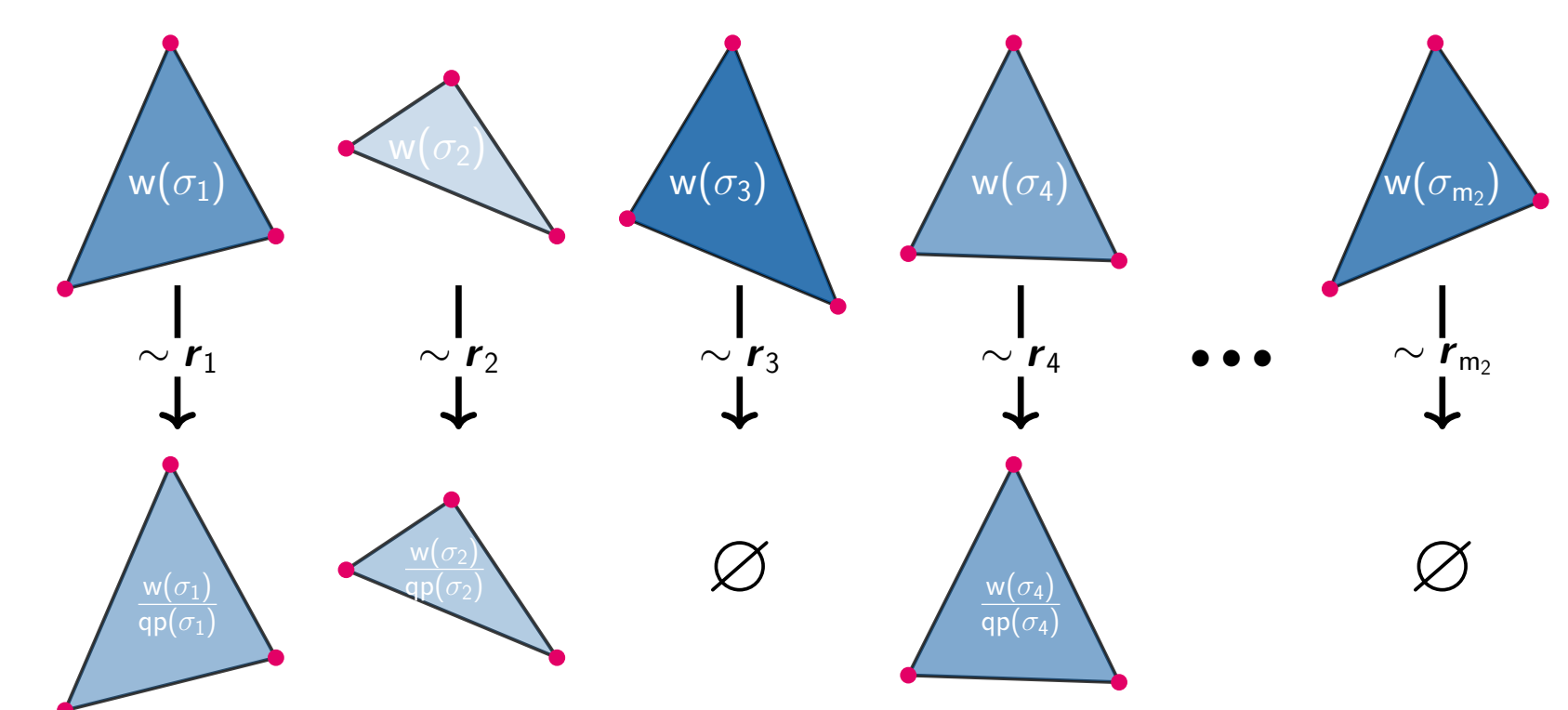
Original complex $\mathcal{K}$ is **too dense**!



Spielman Spectral Sparsification

Sample $q(m_k) = \mathcal{O}(m_k \log m_k / \epsilon)$ simplices according to GER

$$r = \text{diag} \left(B_k^\top \underbrace{(L_k^\uparrow)^\dagger}_{\text{expensive}} B_k \right)$$

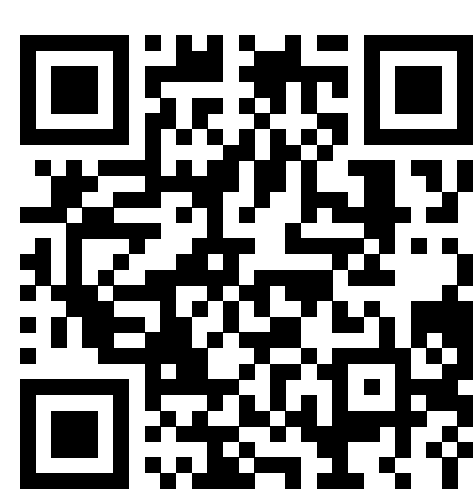


Problems with Direct Approach

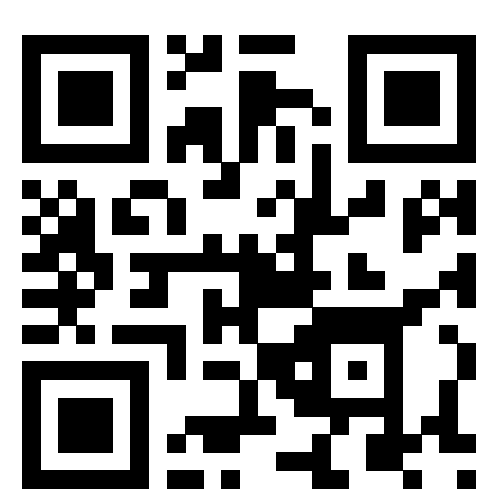
- lack of efficient solvers for $(L_k^\uparrow)^\dagger$ for dense simplicial complexes ($m_{k+1} \gg m_k \ln m_k / \epsilon$);
- poorly conditioned matrix $L_k^\uparrow$;
- computational cost $\mathcal{O}(m_k^3)$ and up to $\mathcal{O}(m_0^{3(k+1)})$.

TL;DR Results

- you can sparsify **any** complex at **any** level with $\mathcal{O}(m_k^{1+\frac{4}{k+1}})$ complexity;
- arbitrary controlled approximation error;
- limited to sparse **matrix multiplications** $\rightarrow$ **small memory consumption**.



arXiv



code

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Novel approximation method

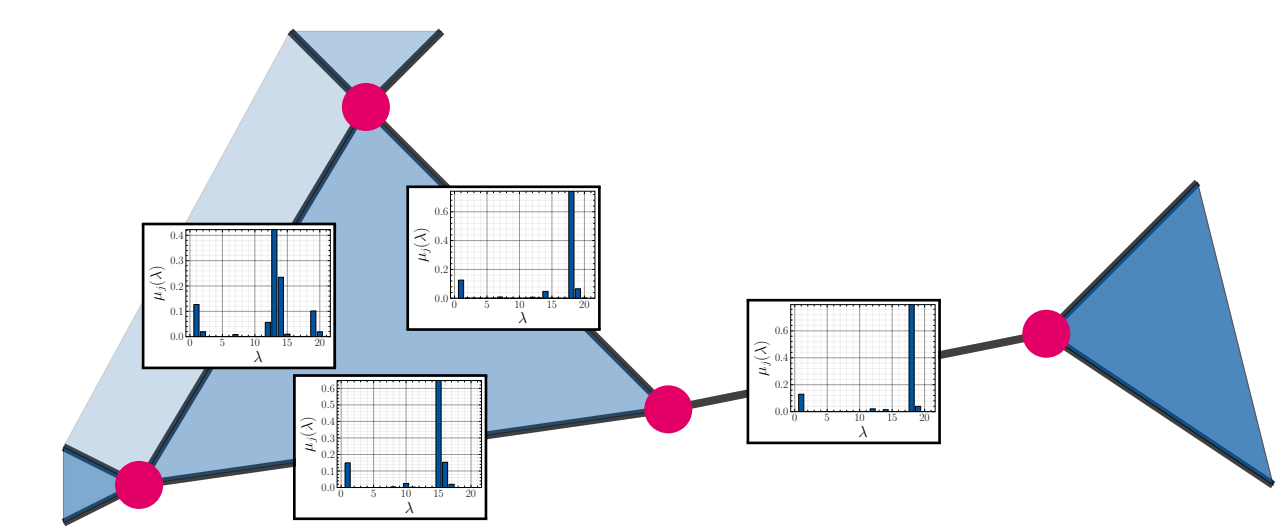
Let $\{\lambda_i, q_i\}$ be eigenpairs of $L_k^\uparrow$.

- **local density of state** is

$$\mu_j(\lambda | L_k^\uparrow) = \sum_i |q_{ij}|^2 \delta(\lambda - \lambda_k)$$

- encodes spectral contribution of each edge/simplex;
- provides a functional description of the spectrum;
- still requires $\mathcal{O}(m_k^3)$ computation!

Can one avoid computing the eigenpairs?

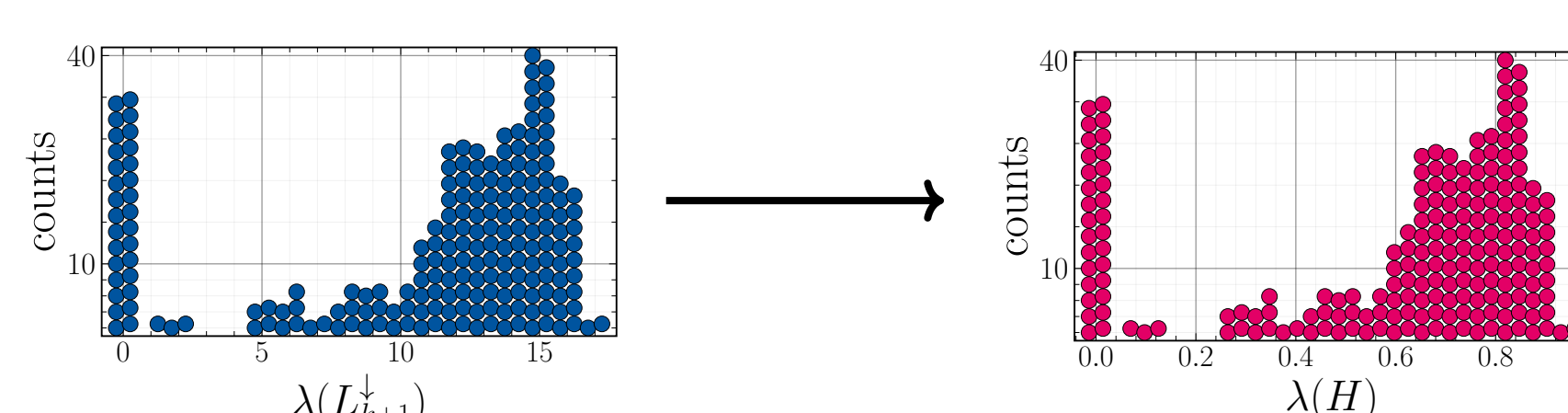


Thm

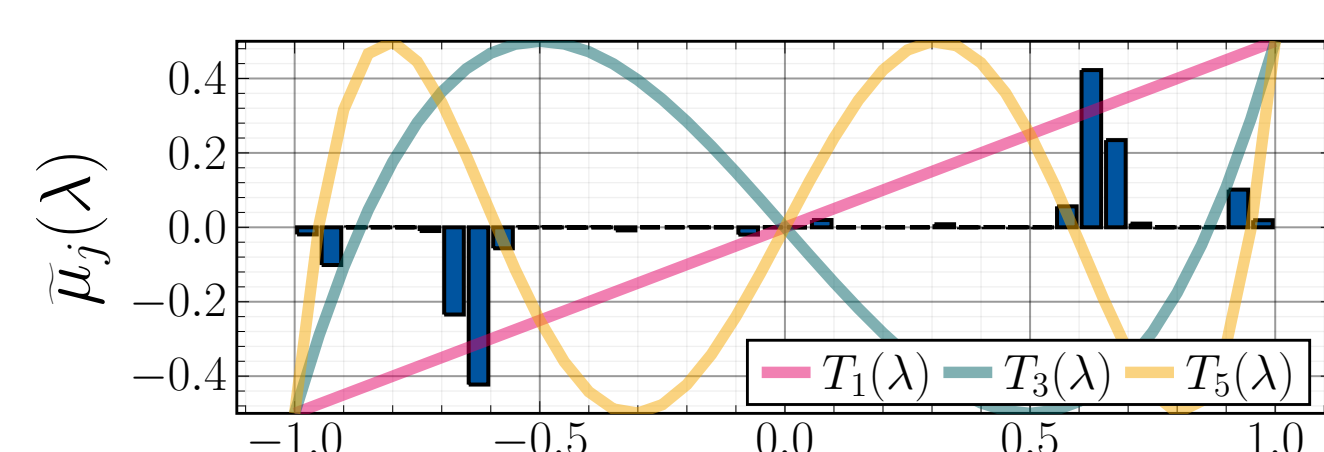
GER r for $L_k^\uparrow$ is defined in terms of the local spectral information:

$$r_i = \int_{\mathbb{R} \setminus \{0\}} \mu_i(\lambda | L_{k+1}^\downarrow) d\lambda$$

Kernel Ignoring Decomposition



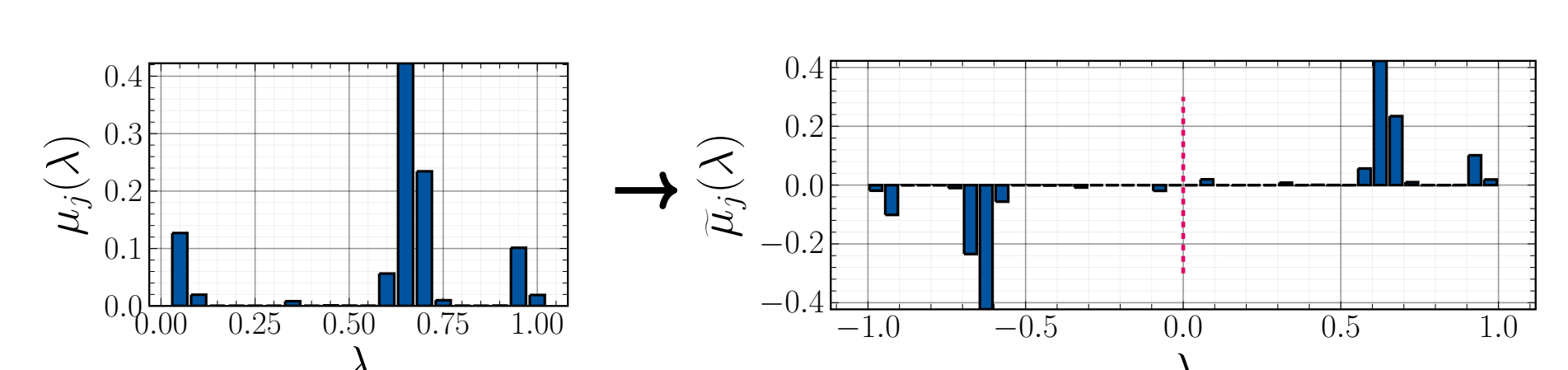
1. scale $L_{k+1}^\downarrow \rightarrow H$ so $\sigma(H) \in [0, 1]$



3. decompose using **odd Chebyshev polynomials**

$$\tilde{\mu}_j(\lambda | H) = \sum_{m=1}^{M/2} d_{mj} T_{2m+1}(\lambda)$$

Idea: approximate LDoS $\mu_j(\lambda | L_k^\uparrow)$ at **every point except $\lambda = 0$** .

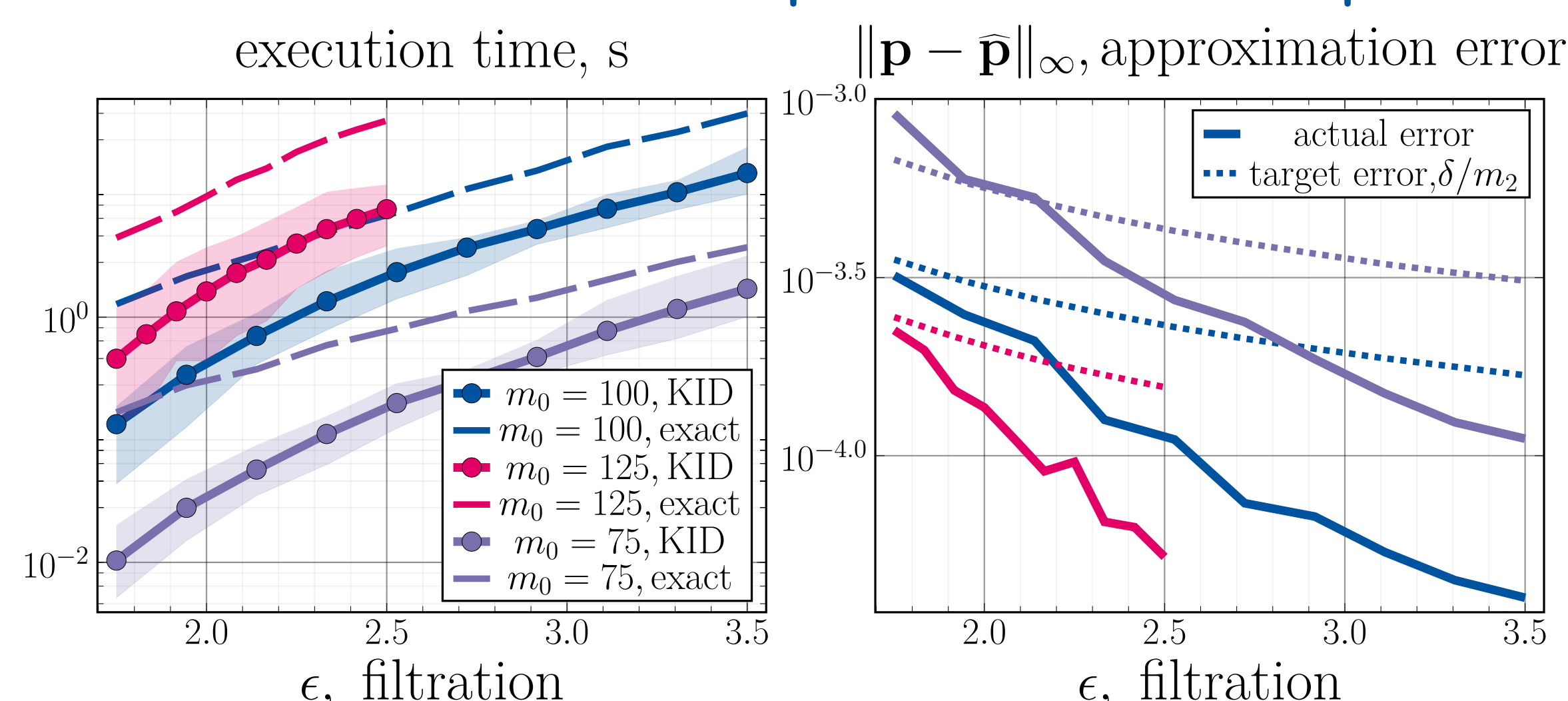


2. exclude $\lambda = 0$ and symmetrize $\mu_j(\lambda | L_k^\uparrow)$ to make an **odd** $\tilde{\mu}_j(\lambda | H)$

$$\mathbb{E} \left[\mathcal{N}(0, 1) \odot T_m(H) \times \mathcal{N}(0, 1) \right] = \text{diag} \left[T_m(H) \right]$$

4. use MonteCarlo-estimate for d_{mj} , $d_{m\bullet} = \text{diag} T_m(H)$

Benchmark: Vietoris-Rips Filtration Complexes



Real-life datasets

dataset	GER time	KID time
email-enron	2.9s	0.31s
contact-hs	59.81s	0.0013s
NDC-classes	7.53s	0.246s
vegas-bars	NA	0.1179s
algebra	2.221s	0.433s