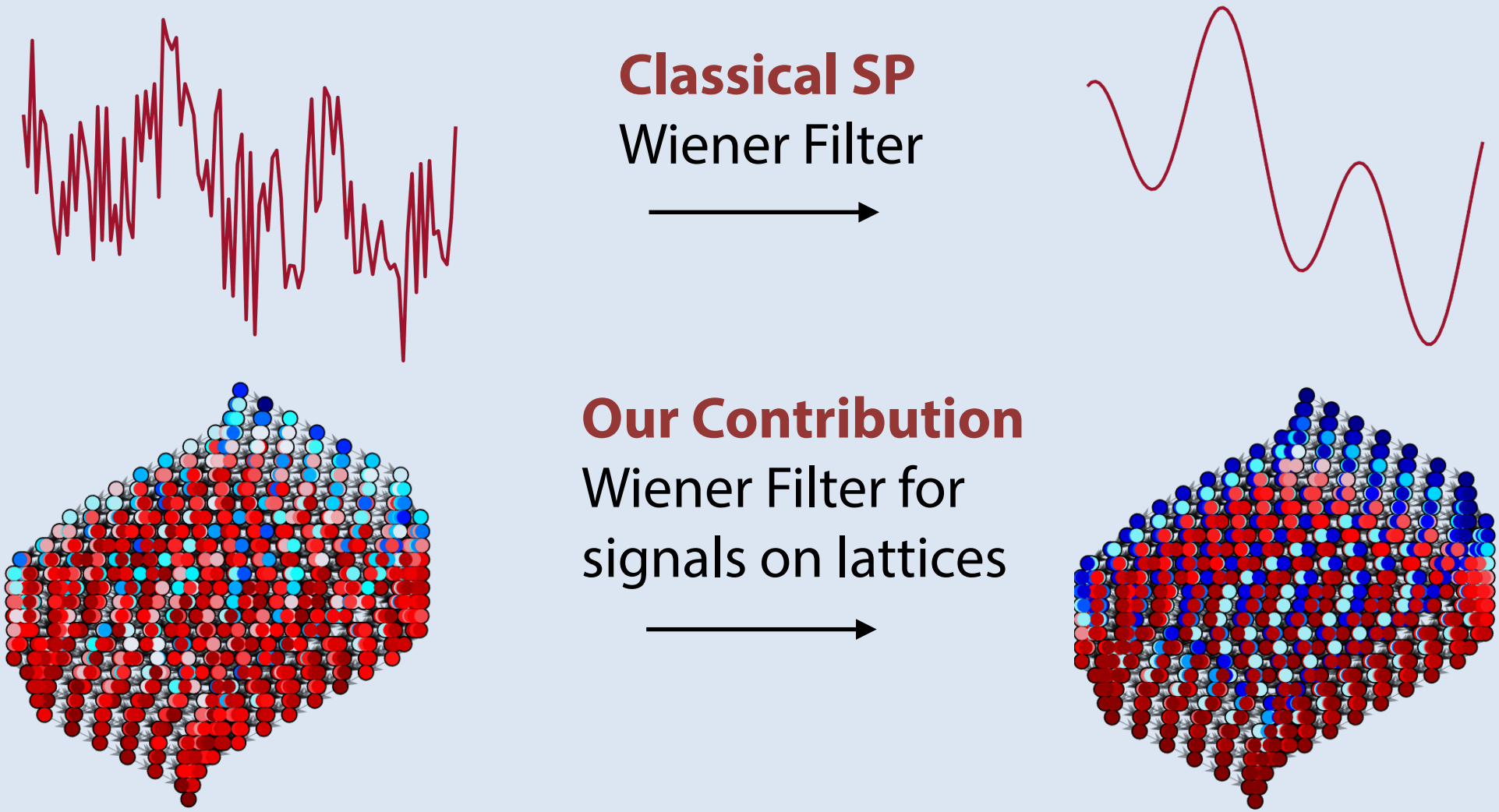


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Goal

Classical SP
Wiener Filter

Our Contribution
Wiener Filter for
signals on lattices



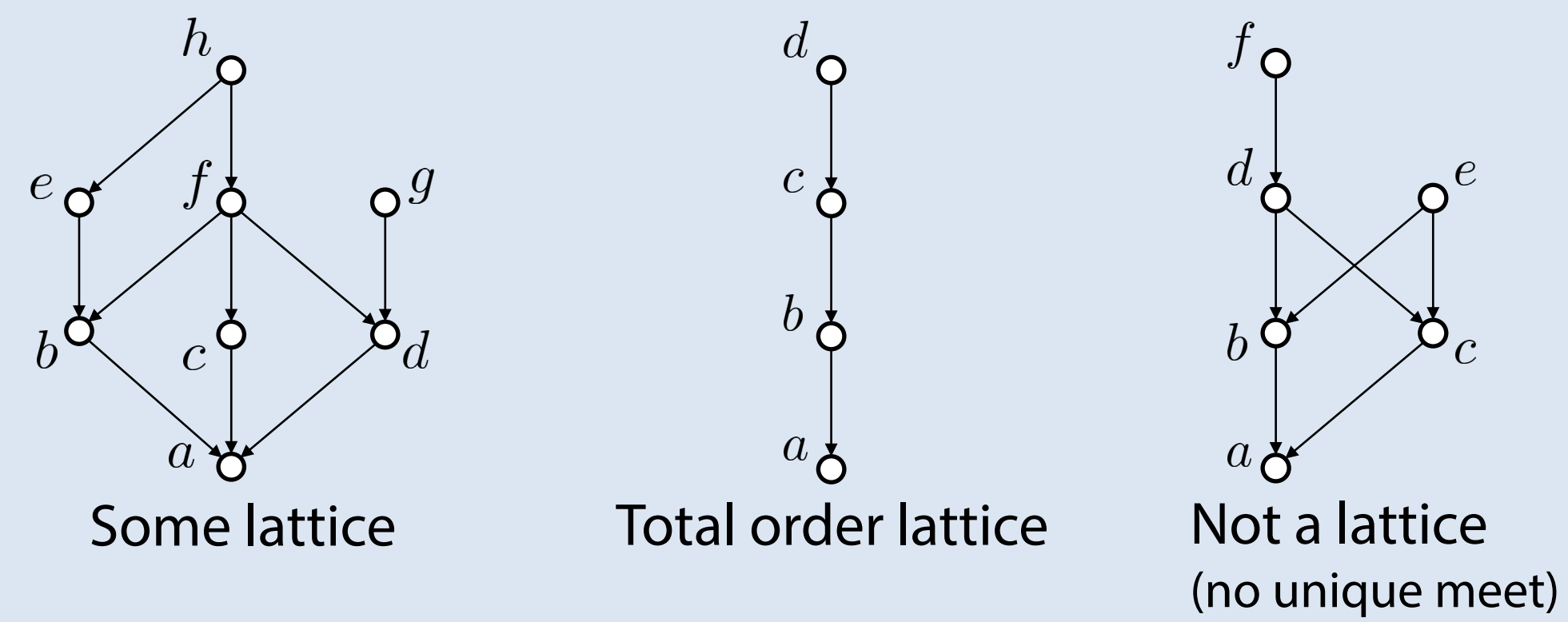
Lattices

Meet Semilattice

- Finite set L with **partial order** \leq
- Meet operation:** $a \wedge b$ (greatest lower bound of a and b)
- b **covers** a : $a < b$, no x : $a < x < b$

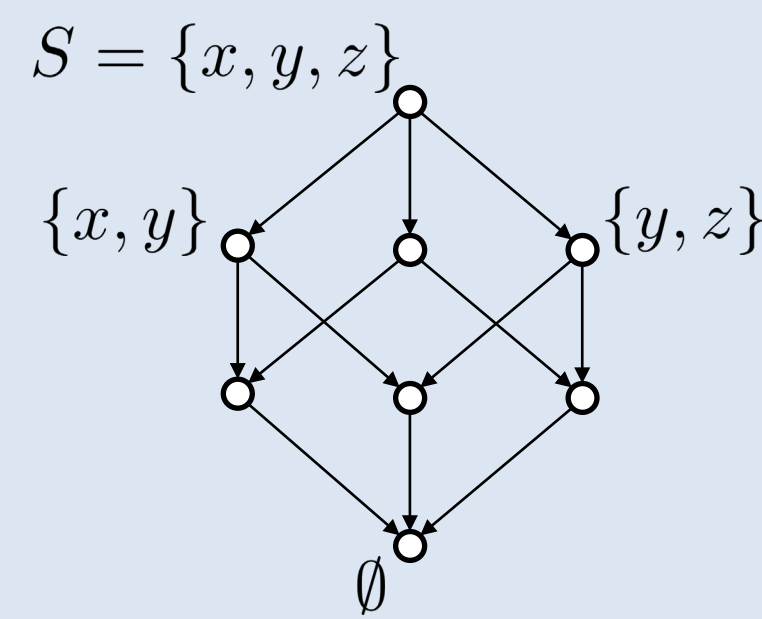
Cover Graph Representation of Lattices

- Nodes: L
- Edges: $(b, a) \Leftrightarrow b$ covers a



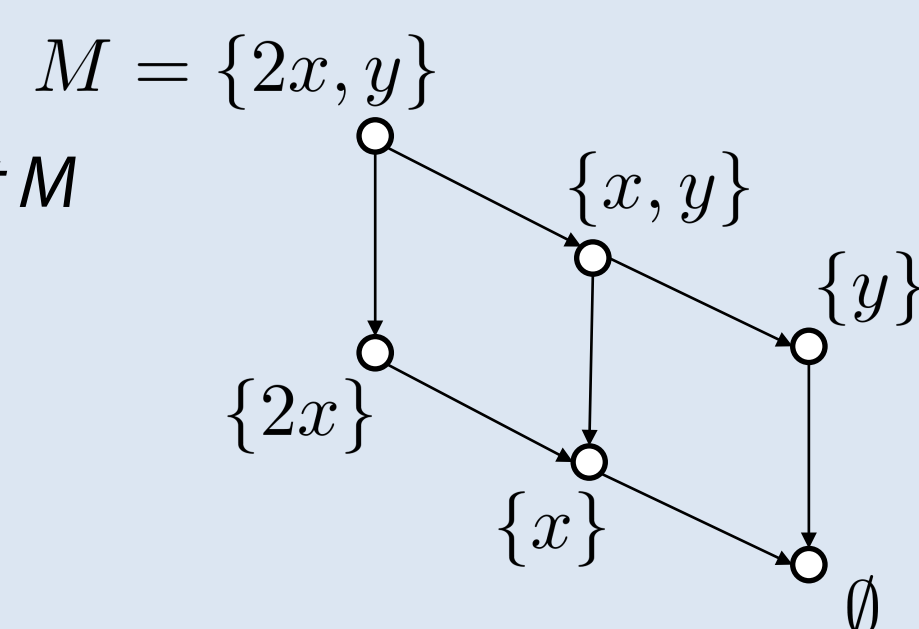
Powerset Lattice

Elements: *Subsets of given set S*
 Partial Order: \subseteq
 Meet: \cap



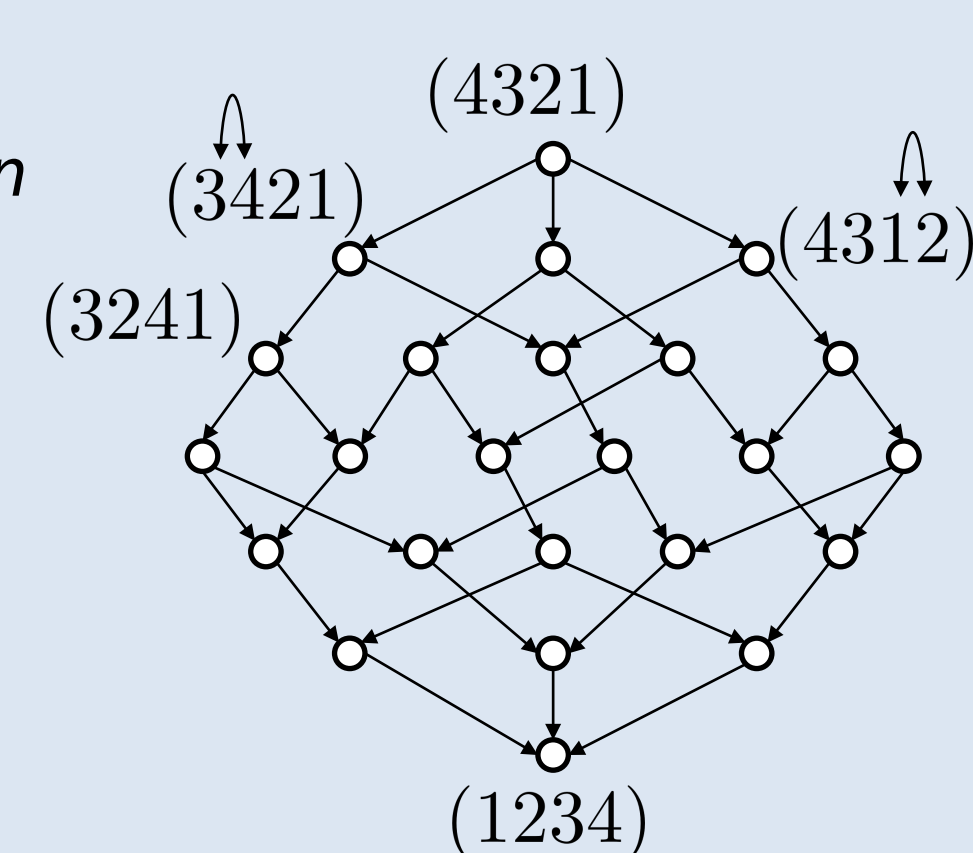
Multiset Lattice

Elements: *Submultisets of multiset M*
 Partial Order: \subseteq
 Meet: \cap



Permutation Lattice

Elements: *Permutations of length n*
 b covers a :
 $(b_1, \dots, b_n) = (a_1, \dots, a_{i+1}, a_i, \dots, a_n)$
 Partial Order and Meet: *Derived from cover graph*

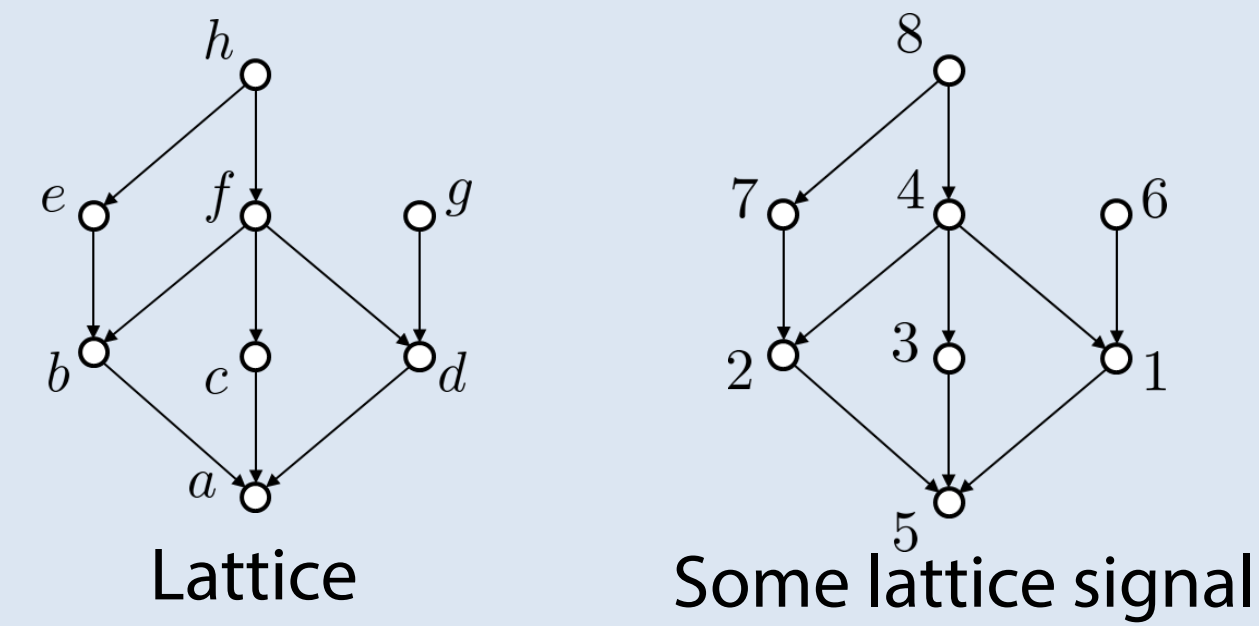


Grätzer, *Lattice Theory: Foundation*, 2011

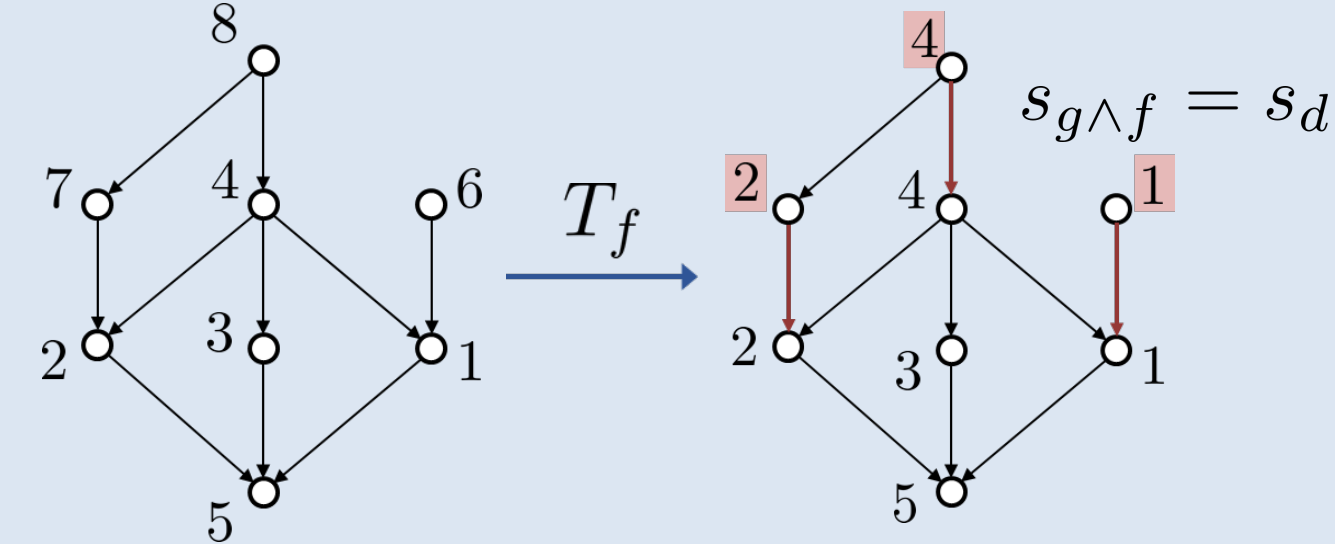
Lattice Signal Processing

Derivation: Algebraic Signal Processing
 Shift \rightarrow Convolution \rightarrow Fourier Transform

Lattice Signal
 $s = (s_x)_{x \in L} \in \mathbb{R}^n$



Shift(s) by $q \in L$
 $T_q s = (s_{x \wedge q})_{x \in L}$



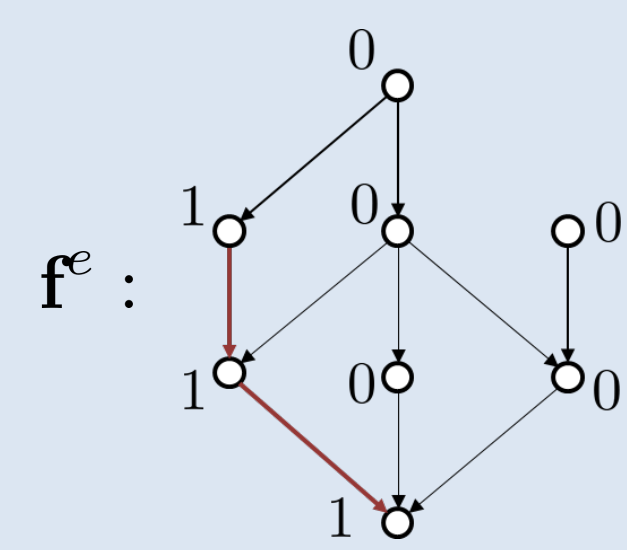
Convolution is shift-invariant

$$h * s = \left(\sum_{q \in L} h_q s_{x \wedge q} \right)_{x \in L}$$

Fourier Basis Vectors

$$f^y = (\mu_{y \leq x})_{x \in L}, \quad y \in L$$

characteristic function



Fourier Transform diagonalizes all shifts and filters

$$\hat{s}_y = \sum_{x \leq y} \mu(x, y) s_x \quad \mu(x, x) = 1, \text{ for } x \in L$$

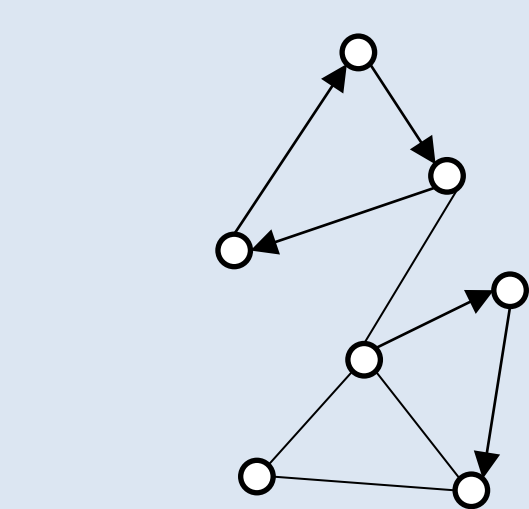
$$s_x = \sum_{y \leq x} \hat{s}_y \quad \mu(x, y) = - \sum_{x \leq z < y} \mu(x, z), \text{ for } x \neq y$$

Fourier transform matrix for above lattice

$$DLT = [\mu(x, y)]_{y, x \in L} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ -1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 1 & 0 & 0 & 0 \\ 2 & -1 & -1 & -1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & -1 & -1 & 0 & 1 \end{bmatrix}$$

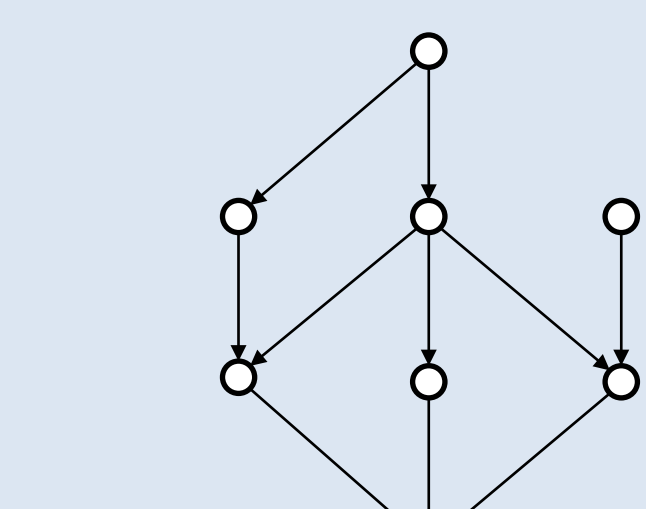
A Discrete Signal Processing Framework for Meet/Join Lattices, ICASSP 2019
 Discrete Signal Processing on Meet/Join Lattices, arXiv, submitted for publication

Lattice DSP vs. Graph DSP



Graph DSP:

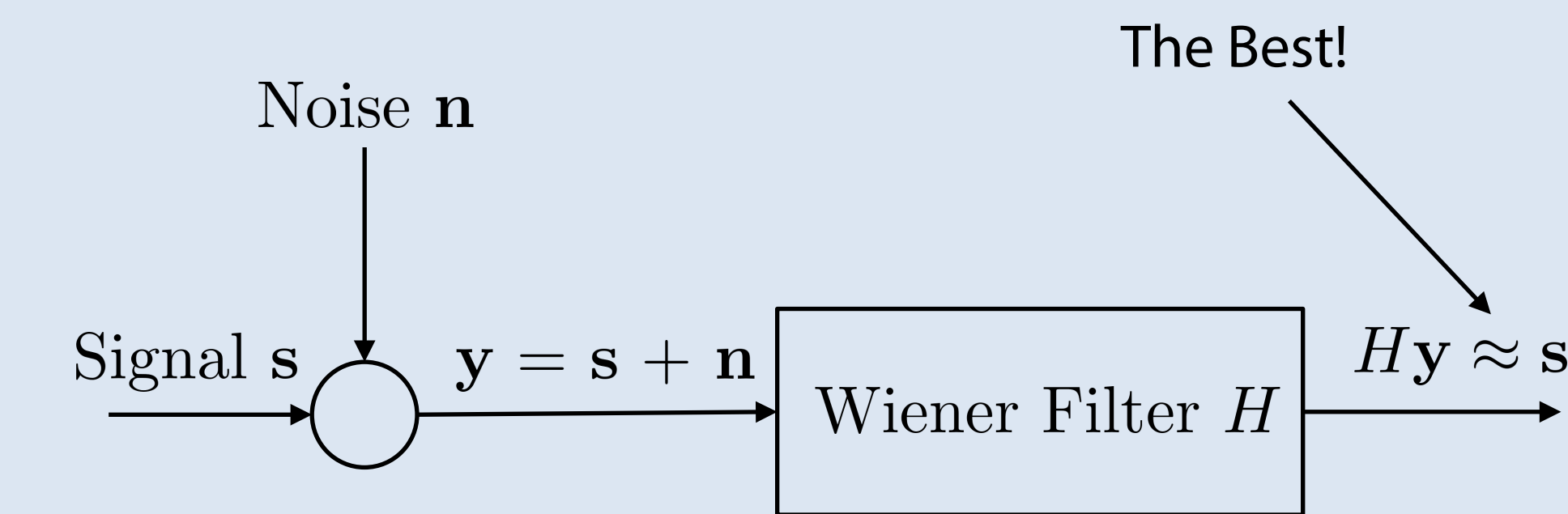
Shift captures adjacency structure
 One generating shift (adjacency or Laplacian)
 Shift not always diagonalizable (digraphs)



Lattice DSP:

Shifts capture partial order structure
 Several generating shifts (one per "maximal" element)
 Shifts always diagonalizable

Wiener Filter on Lattices



Key idea: (For Graphs: Gavili, Zhang *Trans. Signal Proc.* 2017)

Replace Shifts with Energy-Preserving Shift

$$T_{ep} = DLT^{-1} \cdot \Lambda_{ep} \cdot DLT$$

with $\Lambda_{ep} = \text{diag}(\exp(2\pi i k/|L|) \mid k = 0, \dots, |L| - 1)$

Preserves energy

$$\|T_{ep} s\|_2 = \|s\|_2$$

Every other shift representable

$$T_q = p_q(T_{ep})$$

Wiener Filter

$$H = \sum_{k=0}^N h_k T_{ep}^k$$

Filter Coefficients $h = (h_0, \dots, h_N)$

$$\min_h \|Hy - s\|_2^2$$

Closed-form solution

$$B^H B h = B^H s$$

with $B = [y \ T_{ep} y \ \dots \ T_{ep}^N y]$

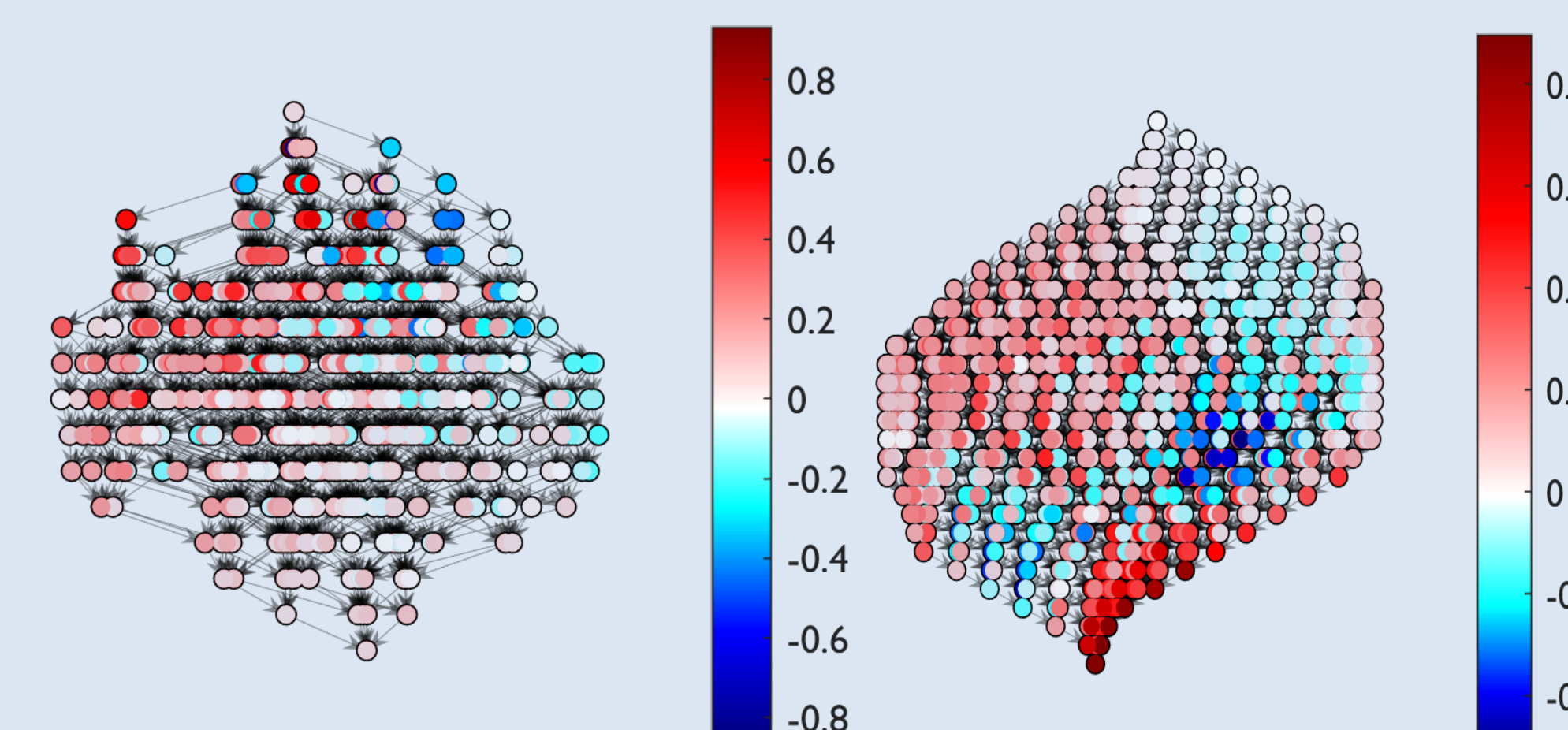
Lattice White Noise

$$n = DLT^{-1} \hat{n}$$

\hat{n} — Gaussian noise vector

Q: Why?

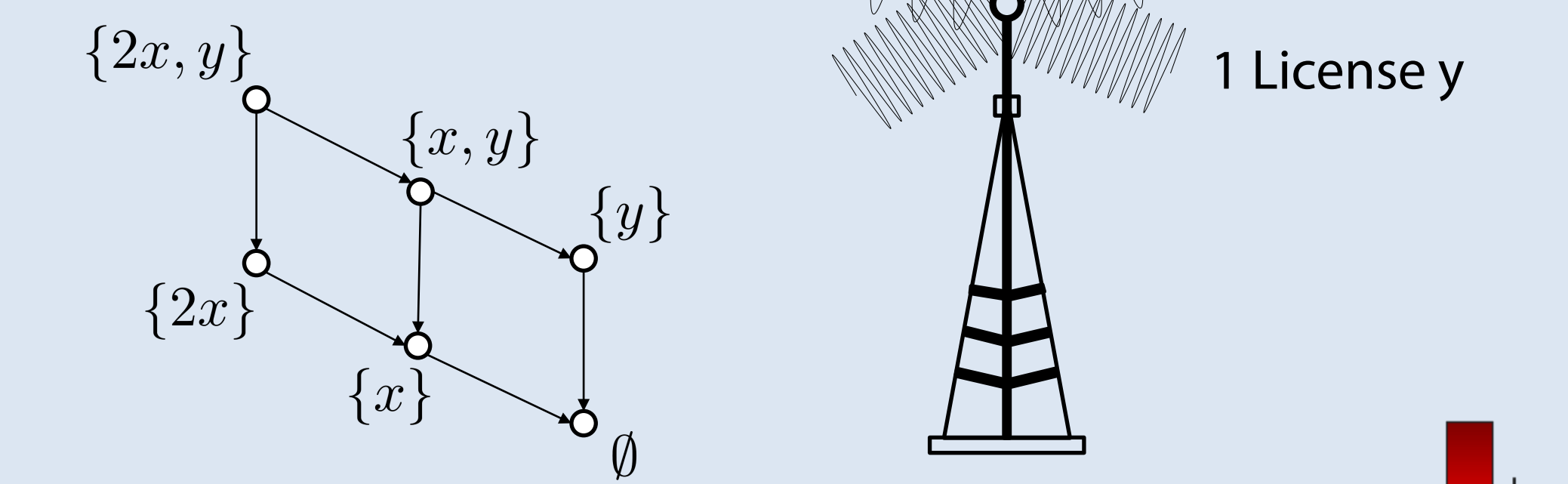
A: Lattice Fourier transform is **not** orthogonal. Thus n is not Gaussian.



Two examples of lattice white noise on different lattices

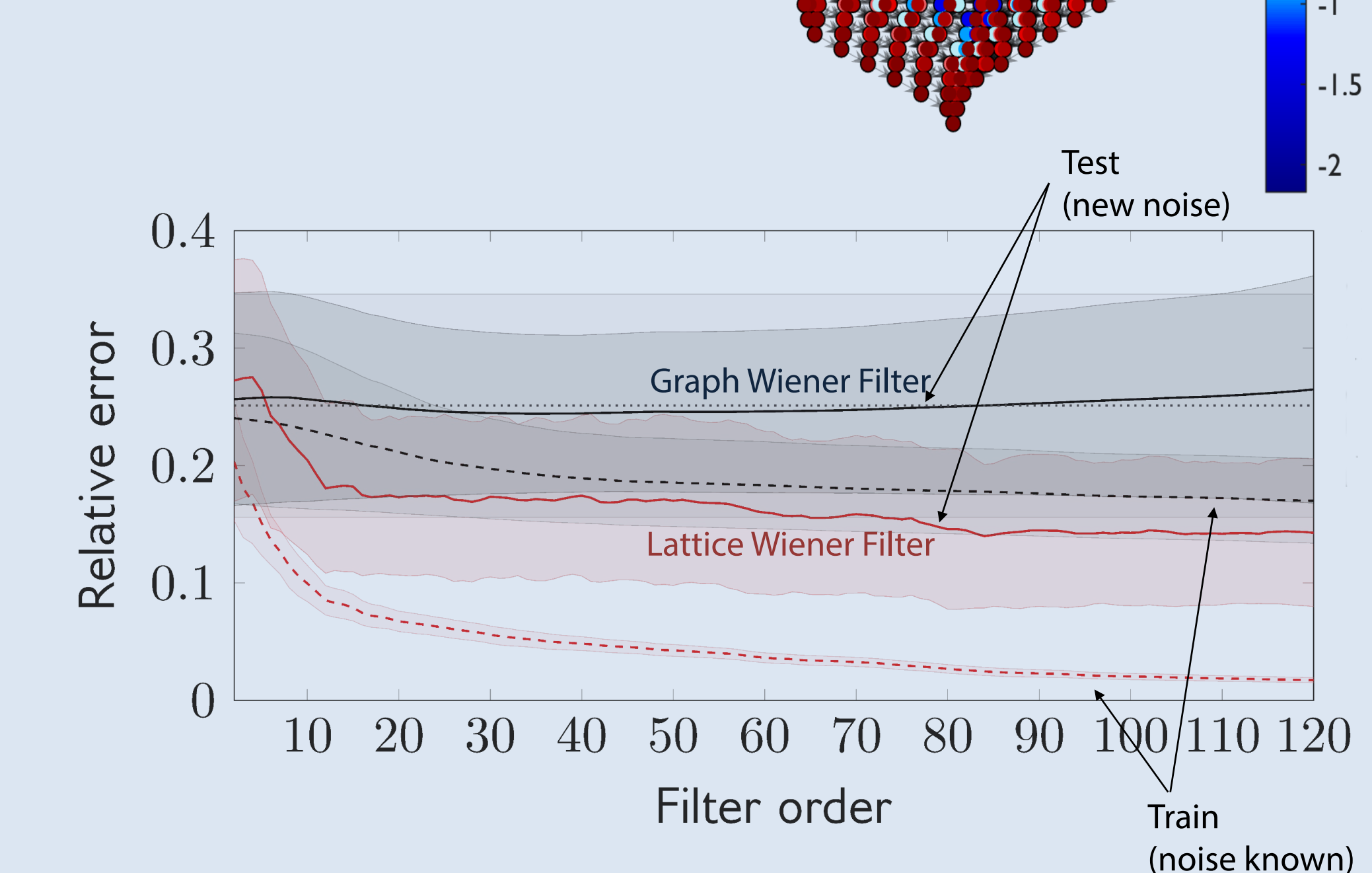
Applications

Spectrum Auctions

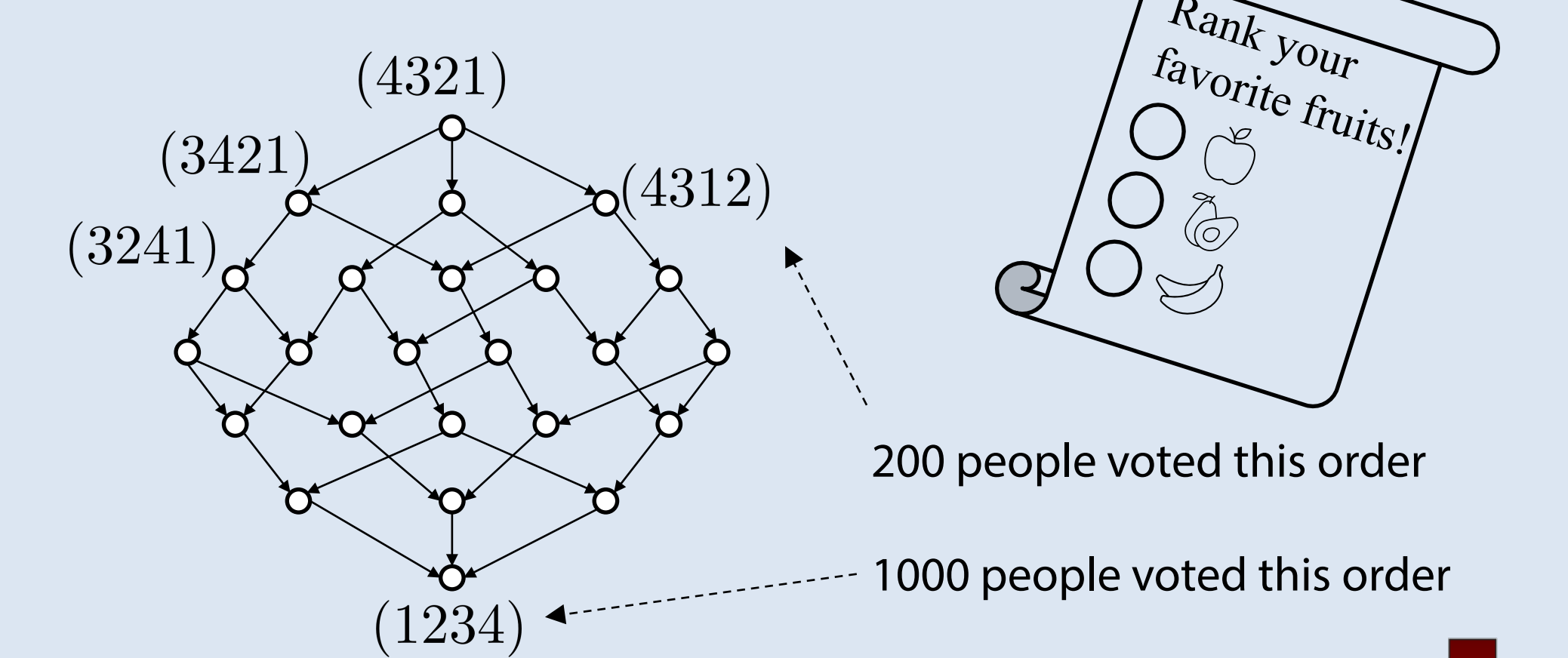


Multiset of licenses = lattice

Bidder = signal of values for each submultiset of licenses



Rankings



Rankings = lattice of permutations of choices

Poll = signal of how many people voted for each permutation

