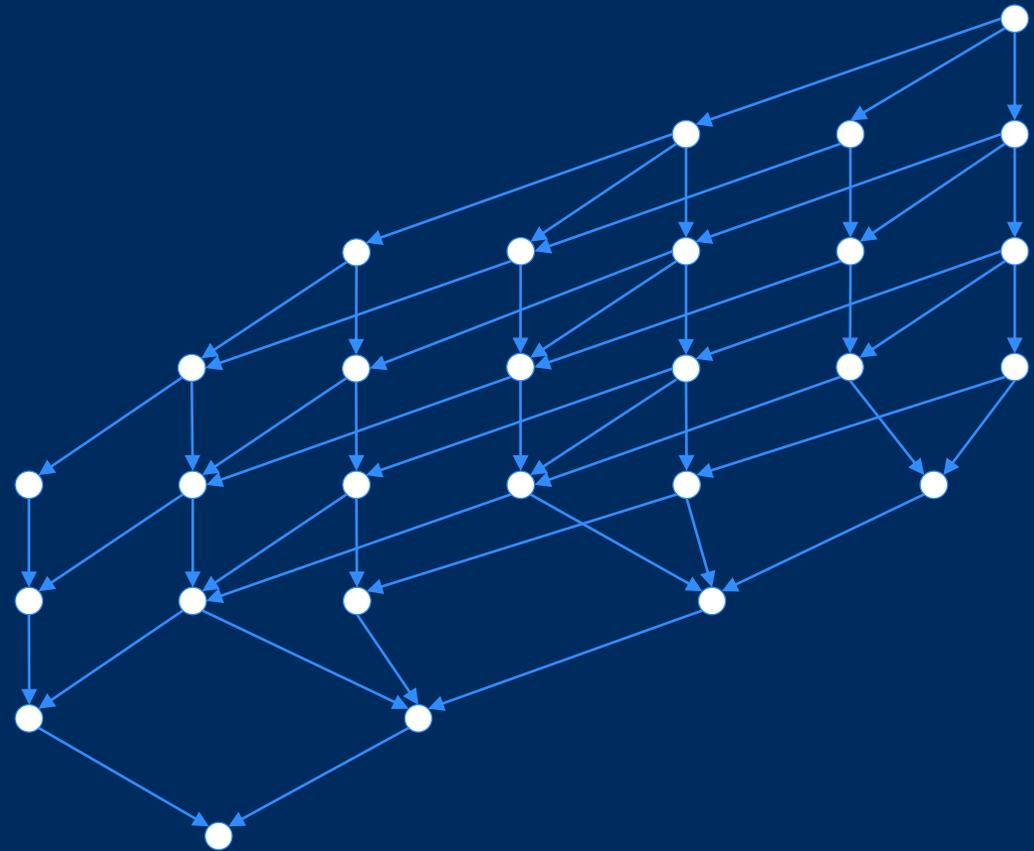


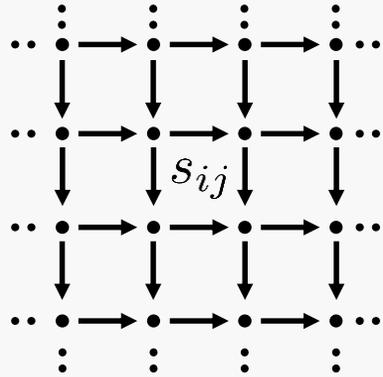
Sampling Signals on Meet/Join Lattices

Chris Wendler and Markus Püschel



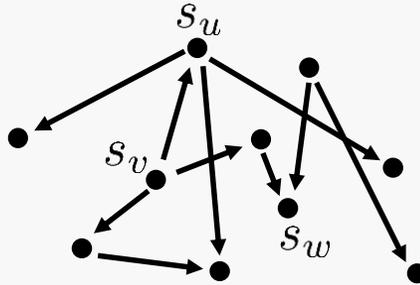
Computer Science
ETH zürich

Classical DSP



Signals indexed by
time/space

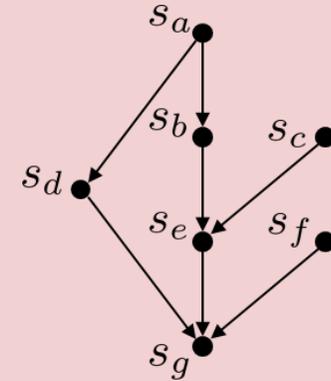
Graph DSP



Signals indexed by nodes
of a graph

Shumann 2012
Sandryhaila 2013

New Discrete Lattice SP



Signals indexed by a
meet/join lattice
Instantiation of algebraic
signal processing theory

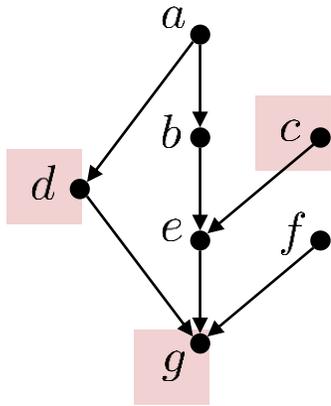
ICASSP 2019

Goal

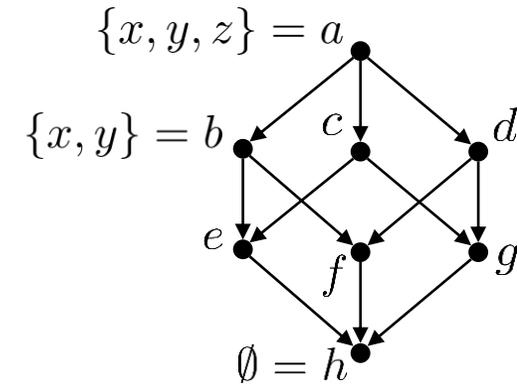
shift, convolution/filtering, Fourier transform, frequency response, sampling,
for lattice signals

Meet Semilattice

Finite set L with **partial order** \leq and **meet operation** $x \wedge y$ (greatest lower bound)



For example, $c \wedge d = g$



Powerset Lattice

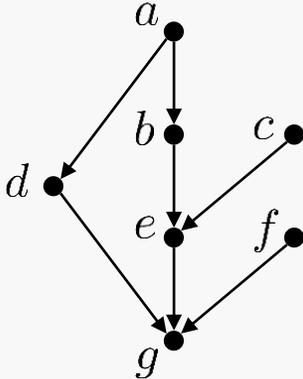
Partial order \subseteq , meet \cap

Join Semilattice

Analogous

Discrete Lattice Signal Processing

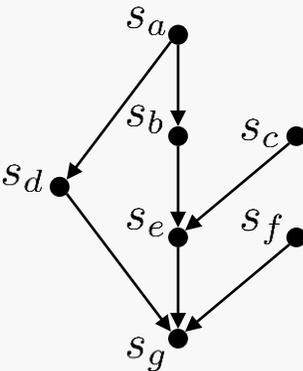
Lattice



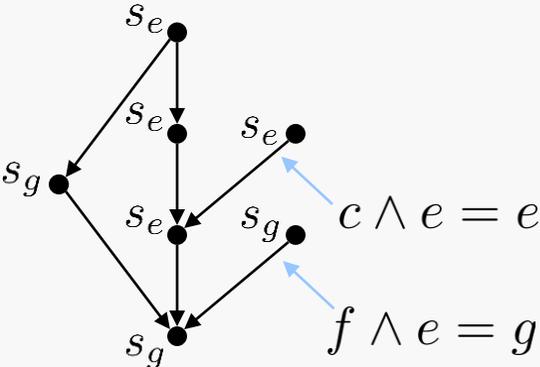
Shift(s) by $q \in L$

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

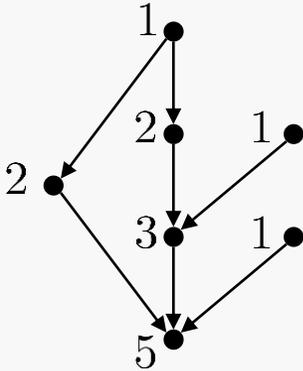
Signal $\mathbf{s} = (s_x)_{x \in L} \in \mathbb{R}^n$



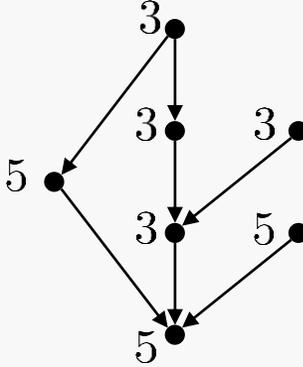
Shifted Signal (by e)



Concrete Example



$T_e \mathbf{s}$ **Shifted Example** (by e)



Filters: Linear Shift Invariant Systems

Shift(s) by $q \in L$

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

Convolution

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

Filter $\mathbf{h} = (h_q)_{q \in L}$

 indexed by
lattice

Shift Invariance ✓

Fourier Transform diagonalizes all shifts and filters

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

 indexed by
lattice

$$\mu(x, y) = - \sum_{x \leq z < y} \mu(x, z),$$

for $x \neq y$

Pure Frequencies eigenvectors of all filters

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

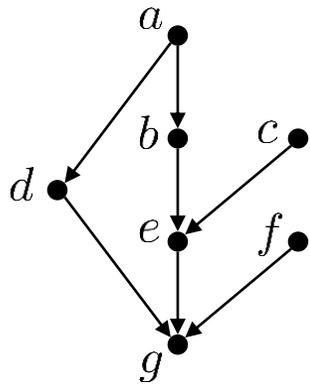
 characteristic function

Frequency response eigenvalues

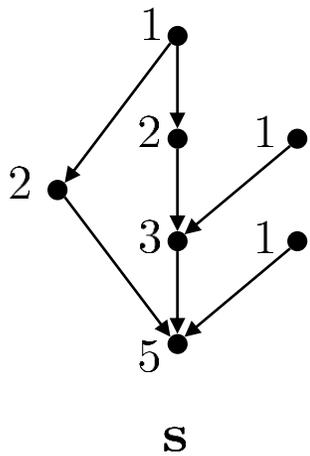
$$\bar{h}_y = \sum_{q \in L, y \leq q} h_q$$

Algebraic Lattice Theory

Lattice



Signal

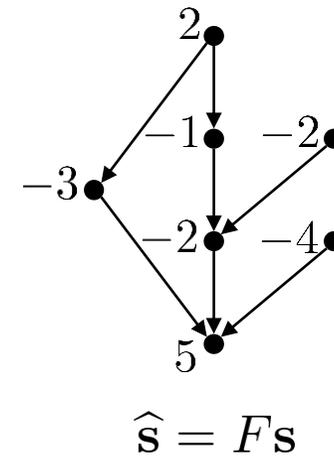


Fourier transform

$$\begin{bmatrix} 1 & -1 & 0 & -1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & -1 \\ 0 & 0 & 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

F

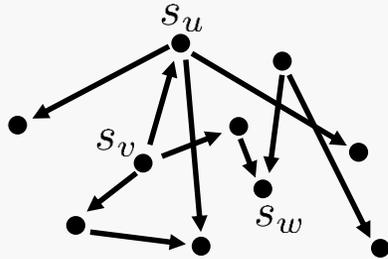
Spectrum



Comparison Graph DSP

Graph DSP

Signals indexed by vertices of a graph



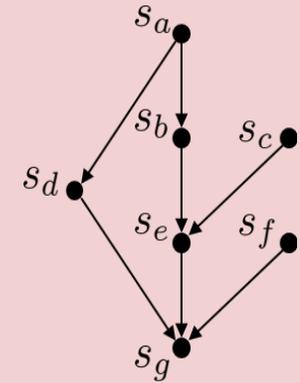
Shift captures adjacency structure

One generating shift
(adjacency or Laplacian)

Shift not always diagonalizable (digraphs)

New Discrete Lattice SP

Signals indexed by a
meet/join lattice =
special type of graph



Shifts capture partial order structure

Several generating shifts
(one per 'maximal' element)

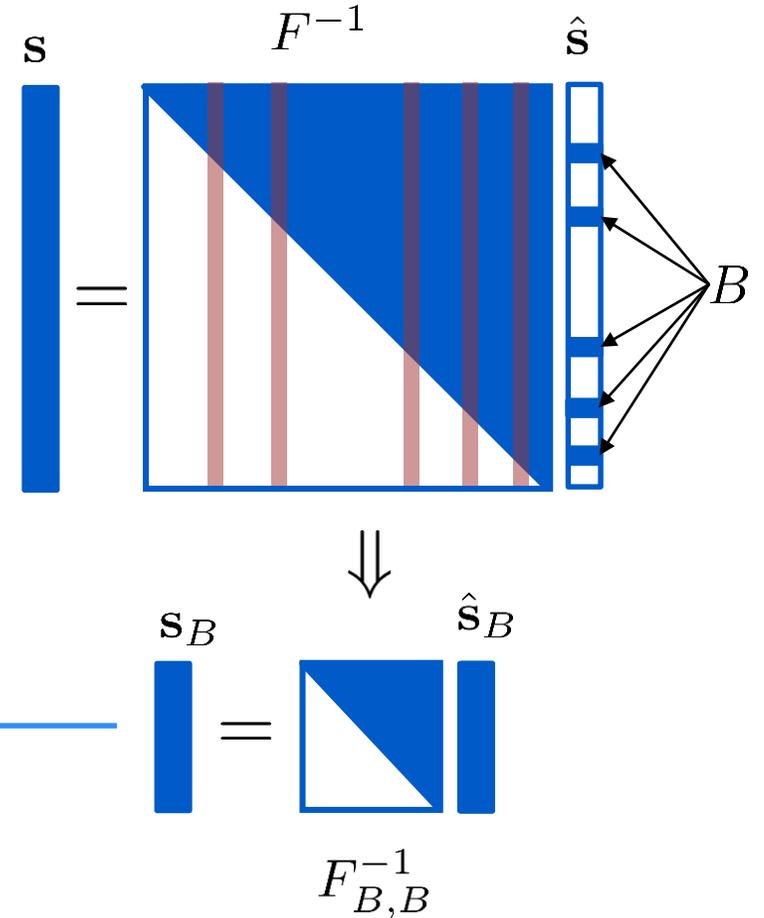
Shifts always diagonalizable

Sampling Signals on Meet/Join Lattices

Sampling Theorem: A Fourier sparse signal s with known support $\text{supp}(\hat{s}) = B = \{b_1, \dots, b_k\}$ can be reconstructed from the samples $s_B = (s_b)_{b \in B}$.

Formally, we have $s = F_{L,B}^{-1} \underbrace{(F_{B,B}^{-1})^{-1}}_{=\hat{s}_B} s_B$.

solve linear system of equations



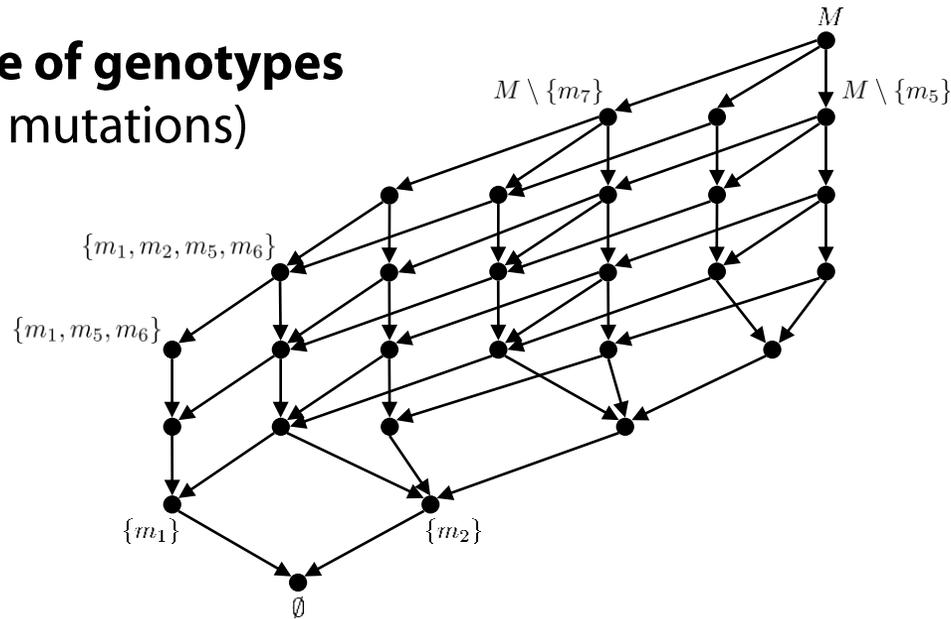
Application 1: Genotype-Phenotype Mappings (HIV RT)

Gene ...GAGAACTTAATAAGAAAACCTCAAGA...

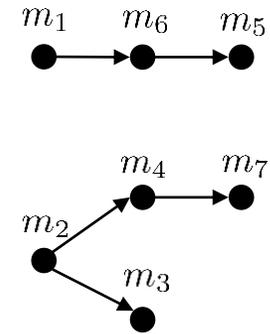
↓ set of mutations

Genotype ...GAGAGCTTAATAAGACAACCTCAAGA...

Semilattice of genotypes
(subsets of mutations)



Constraints on mutations



$$M = \{m_1, m_2, m_3, m_4, m_5, m_6, m_7\}$$

Signal $s_\emptyset = 0$

$$s_{\{m_1\}} = 0.15$$

$$s_{\{m_2\}} = 0.13$$

$$s_{\{m_1, m_5, m_6\}} = 1.23$$

$$s_{\{m_1, m_2, m_5, m_6\}} = 1.67$$

$$s_{M \setminus \{m_5\}} = 2.14$$

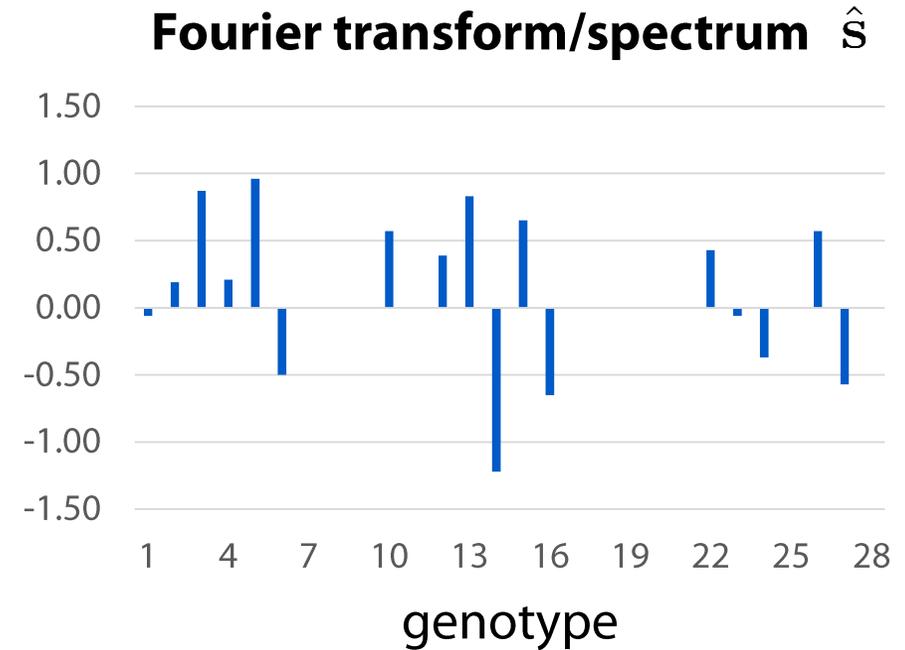
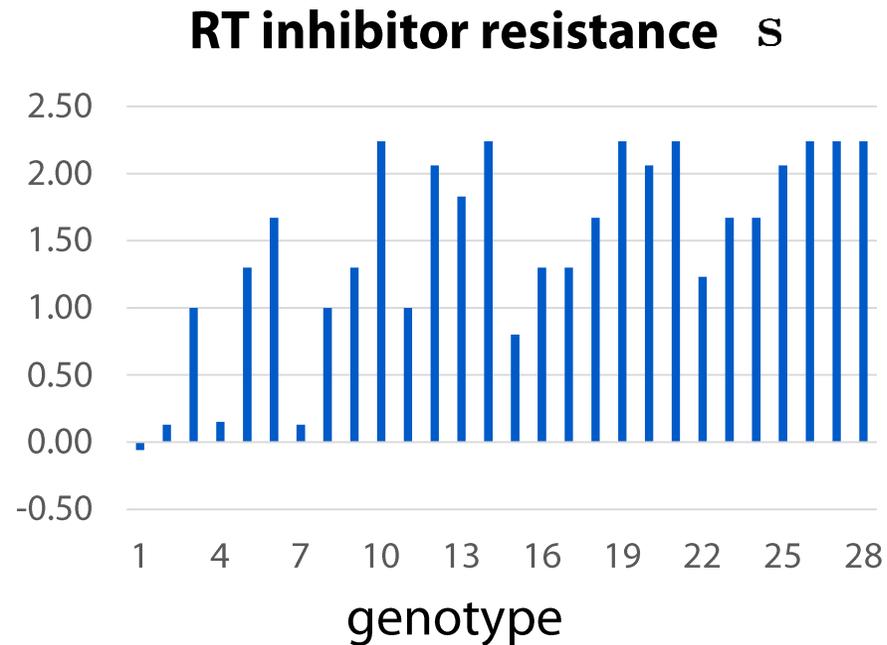
$$s_{M \setminus \{m_7\}} = 2.14$$

$$s_M = 2.14$$

RT inhibitor
resistance



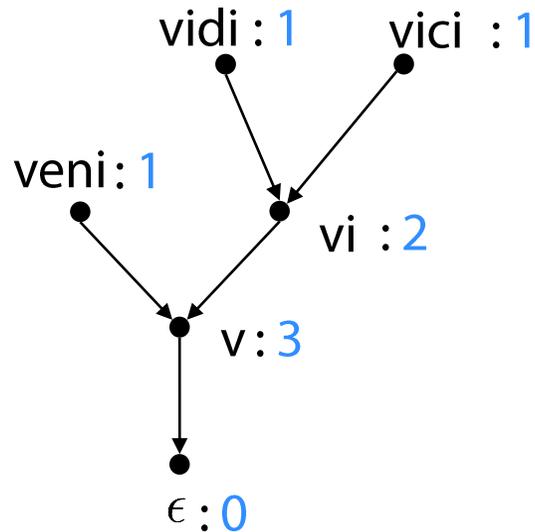
Application 1: Genotype-Phenotype Mappings (HIV RT)



Application 2: Document Representation

Lattice = prefix lattice of words

Signal = prefix count signal



For example: *'veni vidi vici'*

Kritik der reinen Vernunft (Kant 1781)

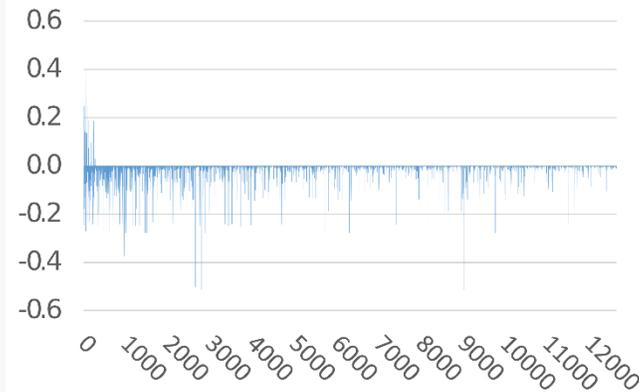
Book

↓
Signal = prefix count

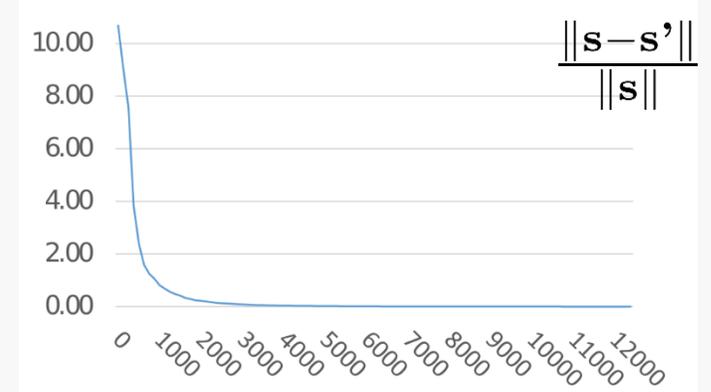
9,952 unique words

12,636 prefixes (lattice)

Prefix count spectrum \hat{s}



Relative reconstruction error



Application 3: Combinatorial Auctions

Spectrum Auction

goods = bands of electromagnetic spectrum

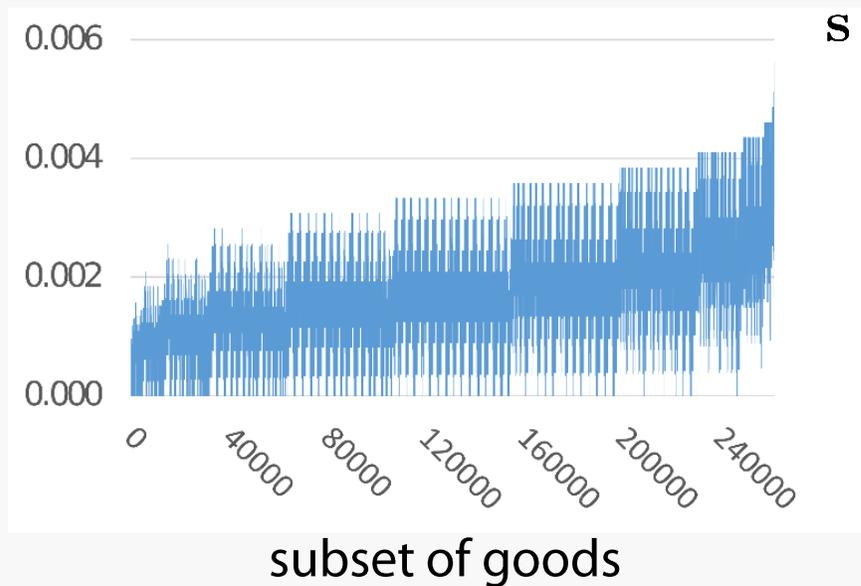
bidders = valuation functions $v_i : 2^M \rightarrow \mathbb{R}^+$

lattice = powerset, signal = valuation function

GSVM auction 18 goods $\rightarrow 2^{18}$ valuations/bidder

Goeree and Holt (2010)

Bidder valuation function

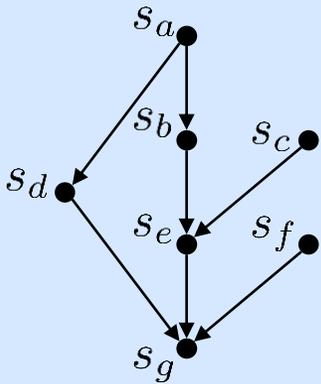


Fourier transform/spectrum



possible application: preference elicitation

Lattice DSP



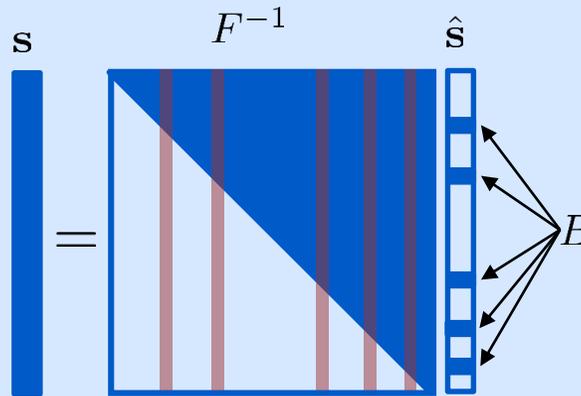
Convolution

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

Fourier Transform

$$\hat{s}_y = \sum_{x \leq y} \mu(x, y) s_x$$

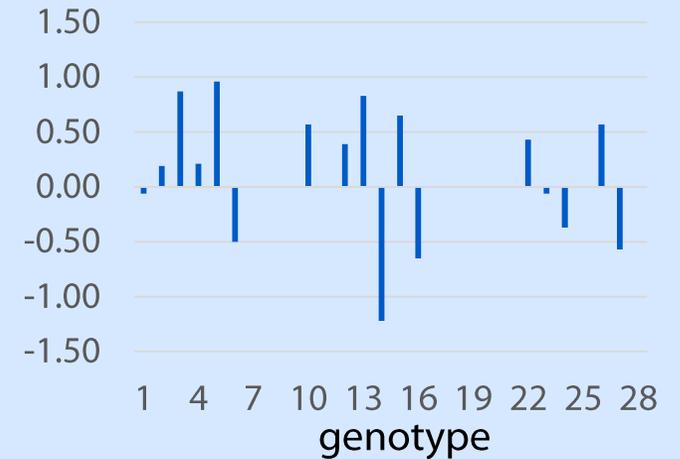
Sampling



$$\mathbf{s} = F_{L,B}^{-1} (F_{B,B}^{-1})^{-1} \mathbf{s}_B$$

reconstruct signal from $|B|$ samples

Possible Applications



1. Genotype-phenotype maps
2. Document representation
3. Set functions (e.g., valuations)
4. **Powerset Convolutional Neural Network (NeurIPS 2019)**