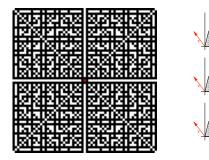
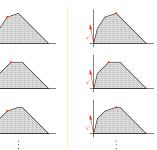
Primitive Sets and Inference Functions: Pure and Applied Combinatorics

Kevin Woods, Oberlin College (joint work with Sergi Elizalde, Dartmouth)





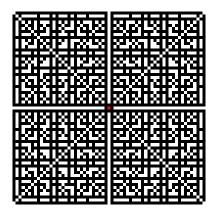
#### Two Stories

#### Pure Story: Geometry of Numbers

Applied Story: Inference for Bayesian networks

# The Pure Story

Question: What proportion of  $(a, b) \in \mathbb{Z}^2$  are visible from the origin?



i.e., a and b relatively prime i.e., (a, b) is a basis for the lattice span<sub> $\mathbb{R}$ </sub> $(a, b) \cap \mathbb{Z}^2$ .

Probability = 
$$\left(1 - \frac{1}{4}\right) \left(1 - \frac{1}{9}\right) \left(1 - \frac{1}{25}\right) \cdots$$
  
=  $\frac{1}{\prod_{p \text{ prime}} 1/(1 - p^{-2})}$   
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## Generalizing

The probability that a point in  $\mathbb{Z}^d$  is visible from the origin is  $1/\zeta(d)$ . [Nymann, 1974]

$$S=\{s_1,s_2,\ldots,s_m\}\subseteq \mathbb{Z}^d$$
 is primitive if it is a basis for $ext{span}_\mathbb{R}(S)\cap \mathbb{Z}^d$ 

The probability that S is primitive is

$$\frac{1}{\zeta(d)\zeta(d-1)\cdots\zeta(d-m+1)}.$$

[Elizalde, W]

Write S as rows of a matrix:

Column operations (over  $\mathbb{Z}$ ) don't change primitivity.

$$egin{bmatrix} \gcd(a,b,c) & 0 & 0 \ d' & e' & f' \end{bmatrix}$$

Must have gcd(a, b, c) = 1 (probability  $1/\zeta(3)$ ).

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Difficult, but interesting

- Triangulations
- Volumes of cross sections of *d*-cubes [Ball, 1989]
- Prime number theorem

# Applied Story

This has it backwards. The applied story came first.

It uses combinatorial tools, but also inspired the previous combinatorial result.

#### Given: Genomes of parent strains:

#### AAAAAA

#### CCCCCC

Observed: Child strain

ATACCC

Inference: Explanation of what recombination happened.

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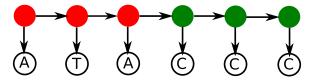
Given R and M, the costs of a recombination event or a mutation.

Minimize  $R \cdot r + M \cdot m$  over all possible explanations (*r*=number of recombinations, *m*=number of mutations).

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This is one example of inference in a Bayesian network / graphical model.



## Inference Functions

Given R and M and a length n,

Inference Function is a map Input: Length *n* DNA sequence (the child) Output: Best possible explanation

Different R and M may give different inference functions. There seem to be

 $(2^n)^{4^n}$ 

possible functions.

## No Worries

Actually, there are only  $O(n^2)$  inference functions.

In general, this is  $O(n^{d(d-1)})$ , where *d* is the number of parameters. [Elizalde, W]

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Example: For binary HMM's of length 5, there are

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Only

5266

are actually inference functions. [Weibel]

## Relation to Combinatorics

Translate to statement about Minkowski sums of polytopes:

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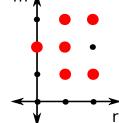
In proving that the  $O(n^{d(d-1)})$  bound is tight, needed to know that a positive fraction of choices of

$$\{s_1,\cdots,s_m\}\subseteq\mathbb{Z}^d$$

are primitive.

Parents:

AAA m CCC Given Child: TAC



8 possible explanations.

Graph (r, m) for each explanation.

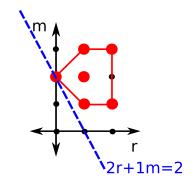
Example: R = 2, M = 1

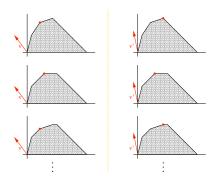
Minimize

2r + 1m

over all points.

Linear Programming!





Two different inference functions (for different R, M).

Inference Function = Vertex of Minkowski Sum

#### Theorem (Gritzmann, Sturmfels)

Let  $P_1, P_2, \ldots, P_k$  be polytopes in  $\mathbb{R}^d$ , and let m denote the number of non-parallel edges of  $P_1, \ldots, P_k$ . Then the number of vertices of  $P_1 + \cdots + P_k$  is at most

$$2\sum_{j=0}^{d-1}\binom{m-1}{j}.$$

Note that this bound is independent of the number k of polytopes.