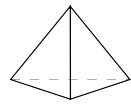
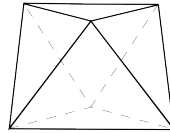


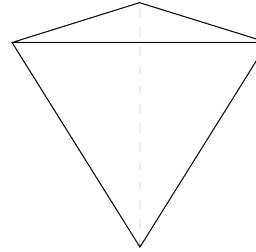
Extensions of parametric inference



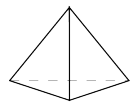
Q_1



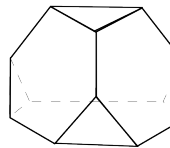
Q_2



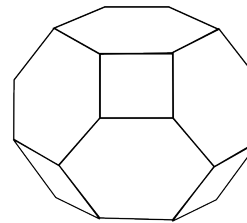
Q_3



P_1



P_2



P_3

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Statistical inference under a log-linear likelihood

- $S \subset \mathbb{Z}_+^d$ set of lattice pts (possible explanations for observed data)
- For $\mathbf{m} \in S$, **log-likelihood** is $\ell(\mathbf{m}) \propto c \cdot \mathbf{m}$ (linear in \mathbf{m})
- c is given vector of parameters.
- **Inference**: Find most probable explanation, $\operatorname{argmax}_{\mathbf{m} \in S} \ell(\mathbf{m})$.
- **Algebraic derivation of Inference**: Given polynomial $f \in \mathbb{Z}_+[x_1, \dots, x_d]$, find leading term of f under term order c .

Recursively defined polynomials over \mathbb{Z}_+

- A polynomial $f \in \mathbb{Z}_+[x_1, \dots, x_d]$ is **recursively presented** over \mathbb{Z}_+ if $f = F_N$, where
 - $F_0 := 1$
 - $F_k := G_k(F_0, \dots, F_{k-1})$ where $G_k \in \mathbb{Z}_+[x][y_0, \dots, y_{k-1}]$,
 $k = 1, \dots, N$.
- For inference applications, definition of G_k usually depends on given observed data
- **Evaluation over semirings**: The entire set F_0, F_1, \dots, F_N can be efficiently evaluated in sequential order, using the recursive presentations. (Dynamic programming)
 - **Example**: Finding leading terms. (Inference)
 - **Linear complexity**: Number of semiring operations = number of operations in the recursive presentations G_1, \dots, G_N .

Parametric inference

- **Inference**: Given $f \in \mathbb{Z}_+[x_1, \dots, x_d]$, and term order vector c , compute the leading term of f .
- **Parametric inference** Find all possible leading terms of f , as term order varies.
- **Algebraic formulation**: Compute the Newton polytope $NP(f)$.
 - Recall: $NP(f) = \text{conv}\{(m_1, \dots, m_d) \mid x_1^{m_1} \cdots x_d^{m_d} \in f\}$.
- Normal cone of $(m_1, \dots, m_d) \in NP(f)$ gives all term orders c for which $x_1^{m_1} \cdots x_d^{m_d}$ is leading term of f

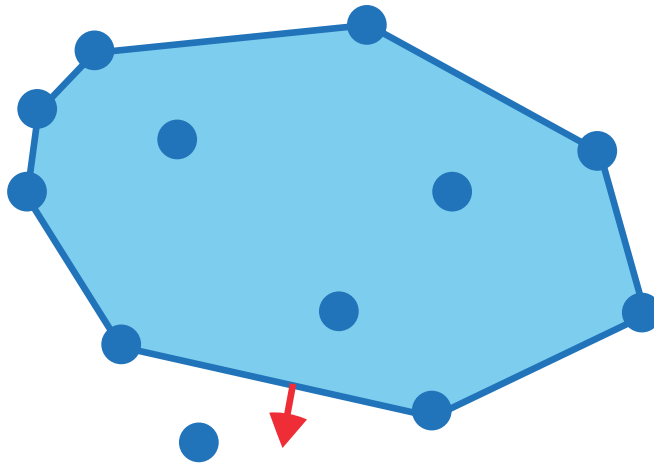
Computing $NP(f)$ for recursively presented $f \in \mathbb{Z}_+[x]$

- **Polytope semiring method** (Pachter/Sturmfels): Evaluate recursive presentation of f , over the **polytope semiring**:
 - $P \odot Q := \text{Minkowski sum } P + Q$
 - $P \oplus Q := \text{conv}(P \cup Q)$
- Theorem: $f(e_1, \dots, e_d) = NP(f)$, where $\{e_i\}$ is standard basis.

Incremental method to compute $NP(f)$

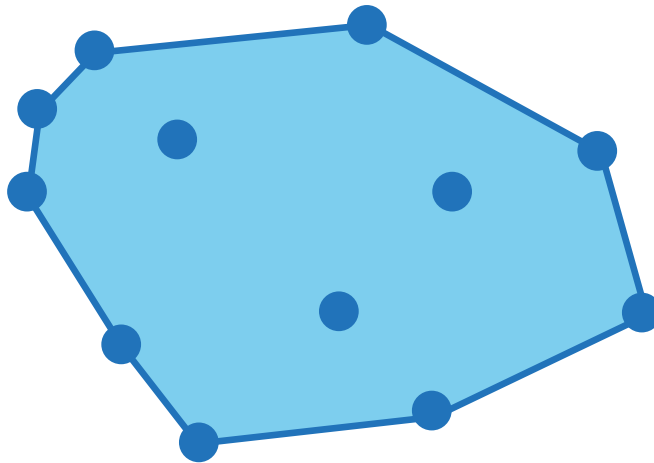
- Find new vertices by repeatedly computing leading terms of f wrt different term orders.
- Software [iB4e](#) incrementally builds polytope P , given subroutine `Optimize(c)` which solves LP $\max c \cdot x$ subject to $x \in P$.

How `iB4e` computes a polytope



To find new vertices, query `Optimize(c)` with c orthogonal to **facets** of current polytope.

How `iB4e` computes a polytope



To find new vertices, query `Optimize(c)` with c orthogonal to **facets** of current polytope.

Parametric k -best inference, and k -set polytopes

Parametric k -best inference: motivation

- Correct explanation may be near-optimal likelihood, instead of optimal
- What if we want to find near-optimal explanations as parameters vary?

Parametric k -best inference

- Given $f \in \mathbb{Z}_+[x_1, \dots, x_d]$,
- Let $S = \{(m_1, \dots, m_d) \mid x_1^{m_1} \cdots x_d^{m_d} \in f\}$ be set of exponent vectors of all terms in f .
- **Problem:** Compute all k -tuples $(\mathbf{m}^1, \mathbf{m}^2, \dots, \mathbf{m}^k) \subset S$ which can become the top k points in S under some linear functional.
- In other words, compute all taylor expansions of f that have k terms

k -sets and ordered k -sets

- Given $S = \{s_1, \dots, s_n\} \subset \mathbb{R}^d$,
- Suppose $T \in \binom{S}{k}$, and $c \cdot t > c \cdot s$ for all $t \in T$ and $s \in S \setminus T$.
Then T is called the **k -set** induced by c .
- Suppose $c \cdot s_{\sigma(1)} > c \cdot s_{\sigma(2)} > \dots > c \cdot s_{\sigma(k)} > \dots$. Then
 $(s_{\sigma(1)}, \dots, s_{\sigma(k)})$ is the **ordered k -set** induced by c .

(Unordered) k -set polytopes

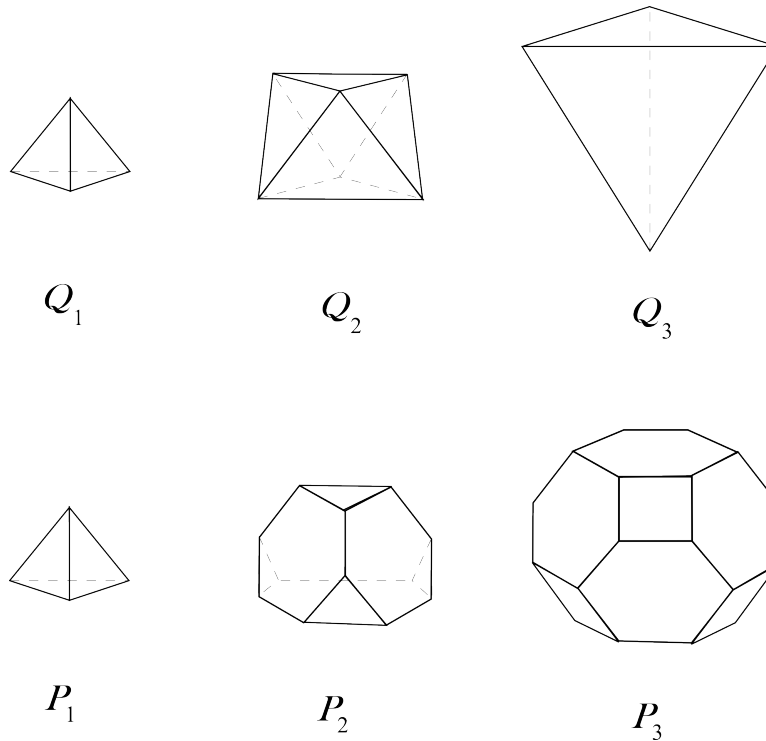
- The k -set polytope $Q_k(S)$ is

$$Q_k(S) = \text{conv} \left\{ \sum_{s \in T} s \mid T \in \binom{S}{k} \right\}$$

- Studied by Edelsbrunner, Fukuda and colleagues.
- **Theorem:**
 - $\sum_{s \in T} s$ is a vertex of $Q_k(S)$ iff $T \in \binom{S}{k}$ is a k -set,
 - The normal cone $N(\sum_{s \in T} s)$ is set of c which induce k -set T .

Ordered k -set polytopes

- Note the equivalence:
 - (s_1, \dots, s_k) is the ordered k -set induced by c , iff
 - $\{s_1, \dots, s_j\}$ is the (unordered) j -set induced by c , for each $j = 1, \dots, k$.
- Thus vertices of $Q_1(S) + \dots + Q_k(S)$ correspond to ordered k -sets. Normal fan characterizes how ordered k -sets depend on c .
- We call $P_k(S) := Q_1(S) + \dots + Q_k(S)$ the **ordered k -set polytope** for S .



Example k -set polytopes $Q_k(S)$ and $P_k(S)$, for $S =$ vertices of a tetrahedron. Note the relation $P_3(S) = Q_1(S) + Q_2(S) + Q_3(S)$.

Parametric k -best inference

- $S =$ exponent vectors of terms in $f \in \mathbb{Z}_+[x_1, \dots, x_d]$.
- Vertices of $P_k(S)$ give all k -tuples of top k leading terms that arise in f as term order varies.
- Normal fan of $P_k(S)$ gives the term orders which yield each choice of top k terms.

Building $P_k(S)$ for parametric k -best inference

- S = exponent vectors of terms in $f \in \mathbb{Z}_+[x_1, \dots, x_d]$.
- Vertex-finding oracle $\text{Optimize}(c)$ for $P_k(S)$:
 - Find the leading k terms x^{z_1}, \dots, x^{z_k} for f , under term order vector c .
 - Return vertex $\sum_{i=1}^k (k - i + 1) z_i$
- Thus incremental polytope construction in B4e can construct $P_k(S)$.

Finding top k leading terms

- Let $LT^k(f)$ denote the set of top k terms of f for given term order c .
- For $f, g \in \mathbf{Z}_+[x_1, \dots, x_d]$ we have:
 - $LT^k(f + g) = LT^k(LT^k(f) + LT^k(g))$
 - $LT^k(fg) = LT^k(LT^k(f) \cdot LT^k(g))$
- Thus $LT^k(\)$ can be computed efficiently for recursively presented polynomials.

Complexity of parametric k -best inference

- For $k = 1$, best known bounds for parametric inference come from general argument of Pachter/Sturmfels.
- Let \mathbb{V} be the volume of lattice polytope $P \subset \mathbb{R}^d$, d fixed.
- **Andrews theorem:** If $\mathbb{V} > 0$, P has $O(\mathbb{V}^{(d-1)/(d+1)})$ vertices.
- **Corollary:** If $f \in \mathbb{Z}_+[x_1, \dots, x_d]$ has degree $n_i > 0$ in variable x_i , then $NP(f)$ has $O(\prod_{i=1}^d n_i^{(d-1)/(d+1)})$ vertices.

Complexity of parametric k -best inference

- S = exponent vectors of terms in $f \in \mathbb{Z}_+[x_1, \dots, x_d]$. Let $\mathbb{V} = \text{vol}(NP(f))$.
- Observation: $Q_k \subset k \cdot NP(f)$, and $\dim Q_k(S) = \dim(NP(f))$.
- (Andrews theorem): If $\mathbb{V} > 0$, then $Q_k(S)$ has $O((k^d \mathbb{V})^{(d-1)/(d+1)})$ vertices, and $P_k(S)$ has $O((k^{2d} \mathbb{V})^{(d-1)/(d+1)})$ vertices.
- Polynomial in k
- For fixed k , same as best known bounds for $k = 1$ case.

Constrained parametric inference

“Reasonable” vertices of polytopes

- **General problem:** For a polytope P , we might only want vertices whose normal cones intersect a prescribed cone C .
- Often the case in parametric inference
- **Example:** [Bayesian networks](#): logs of transition probabilities should be non-positive. Also some types of transitions might be prescribed to be more likely than others.

Envelopes of polytopes

- Given polytope $P \subset \mathbb{R}^d$, and a cone $C \subset \mathbb{R}^d$ with apex 0, the C -envelope of P is the set of faces $\{F \subset P \mid N(F) \cap C \neq \{0\}\}$.
- Graph of C -envelope is **connected subgraph** of graph of P .
- For parametric inference, we will assume C is full-dimensional and pointed.

Incremental construction of C -envelopes

- Can we use `iB4e` to compute C -envelope of polytope P , given oracle `Optimize()` for P ?
 - Can we **avoid finding vertices outside the C -envelope**, to speed up the computation?
- **Yes:**
 - Let u_1, \dots, u_m be generators of the dual C' .
 - Initialize $P' = \text{conv}(\{Nu_i\})$ where N is a large number.
 - Then finish running `iB4e` to add vertices to P' , using `Optimize()` oracle for P .
- Output P' is desired envelope. (Proof: Farkas Lemma)

Thank you

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