

## Recap

•  $M \subseteq \mathbb{N}^n$  is  $M$ -convex if

$$(EA) \quad \alpha, \beta \in M, \alpha_i > \beta_i \Rightarrow \exists j \beta_j > \alpha_j, \alpha - e_i + e_j \in M$$

$$\text{Contraction: } \partial_i M = \{ \alpha - e_i : \alpha \in M, \alpha_i > 0 \}$$

$$\text{Truncation: } \tau M = \bigcup_{i=1}^n \partial_i M$$

• If  $M$  is  $M$ -convex, then so are  $\partial_i M$  and  $\tau M$

Converse:

Lemma R. Suppose  $M \subseteq \{ \alpha \in \mathbb{N}^n : \sum_{i=1}^n \alpha_i = r \}$ ,  $r \geq 3$ .

If  $\tau M$  and  $\partial_i M$  are  $M$ -convex for all  $i$ , then

$M$  is  $M$ -convex.  $f = \sum_{\alpha} a_{\alpha} x^{\alpha}$   
 $\text{supp}(f) = \{ \alpha \in \mathbb{N}^n : a_{\alpha} \neq 0 \}$

Remark: If  $f \in \mathbb{R}_{\geq 0}[x_1, \dots, x_n]$  is homogeneous, then  $\text{supp}(\partial_i f) = \partial_i \text{supp}(f)$  and  $\text{supp}(D_v f) = \tau \text{supp}(f)$  for all  $v \in \mathbb{R}_{> 0}^n$ .

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Theorem (Choe-Oxley-Sokal-Wagner, 2004).

If  $f \in \mathbb{R}_{\geq 0}[x_1, \dots, x_n]$  is homogeneous and stable, then  $\text{supp}(f)$  is  $M$ -convex.

Proof. Since stable polynomials are closed under  $\partial_i$  and  $D_v$ , the supports of such polynomials are closed under  $\partial_i$  and  $\tau$ . It remains to consider  $d=2$ .

The support of quadratic stable pol.,  
are M-convex:

Proof. Previous lecture we may  
assume  $f(x) = x^T A x$ ,  $A$ 's symmetric  
 $A = (a_{ij})_{i,j=1}^n$ ,  $a_{ij} \geq 0$

$$(\star) \quad (x^T A y)^2 \geq (x^T A x)(y^T A y), \quad \begin{matrix} x \in \mathbb{R}_{\geq 0}^n \\ y \in \mathbb{R}^n \end{matrix}$$

proof by contradiction. Assume

$$\alpha = e_i + e_j \in M \quad \text{violate (EA).}$$

$$\beta = e_k + e_l \in M$$

$$\text{Then } \{i, j\} \cap \{k, l\} = \emptyset$$

$$\neg(\text{EA}): \quad \begin{aligned} \alpha - e_i + e_k &= e_j + e_k \notin M \\ \alpha - e_i + e_l &= e_j + e_l \notin M \end{aligned}$$

$$x = e_i + t e_j, \quad t > 0$$

$$y = e_k + e_l$$

$$x^T A y = a_{ik} + a_{il} \quad (a_{jk} = 0)$$

$$x^T A x = a_{ii} + t^2 a_{jj} + 2t a_{ij}$$

$$y^T A y = a_{kk} + a_{ll} + 2a_{kl}$$

$t \xrightarrow{\text{if}} \infty$  then the RHS of  $(\star)$  dominates.  $\square$

Question: Which matroids are supports of stable polynomials?

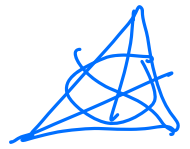
Yes:

- Representable over  $\mathbb{C}$ .
- Vámos's matroid (Wagner, Wei)

(Used to construct counterexamples to generalized Lax conjecture)

No: (B., González D'León)

- No projective geometry
- No binary-non-regular.



# C-Lorentzian polynomials

Minkowski volume polynomials:

$K_1, \dots, K_n$  convex bodies in  $\mathbb{R}^d$ :

$$\text{Vol}(x_1 K_1 + \dots + x_n K_n) = \sum_{i_1, \dots, i_d} V(K_{i_1}, K_{i_2}, \dots, K_{i_d}) x_{i_1} x_{i_2} \dots x_{i_d}$$

↑  
Minkowski sum

↑  
Mixed volumes  $\geq 0$

Alexandrov-Fenchel inequalities:

$$V(K_1, K_2, K_3, \dots, K_d)^2 \geq V(K_1, K_1, K_3, \dots, K_d) \cdot V(K_2, K_2, K_3, \dots, K_d)$$

Let  $f \in \mathbb{R}[x_1, \dots, x_n]$  be homogeneous of deg.  $d$   
and  $v_1, \dots, v_m \in \mathbb{R}^n$ . Then

$$f(t_1 v_1 + \dots + t_m v_m) = \frac{1}{d!} \sum_{i_1, \dots, i_d} (D_{v_{i_1}} D_{v_{i_2}} \dots D_{v_{i_d}} f) \cdot t_{i_1} t_{i_2} \dots t_{i_d}$$

↑  
mixed derivatives

• Let  $C \subseteq \mathbb{R}^n$  be a convex open cone.

Def:  $f$  is C-Lorentzian if for all

$$v_1, v_2, \dots, v_d \in C,$$

$$(P) \quad D_{v_1} D_{v_2} \dots D_{v_d} f > 0$$

$$(AF) \quad (D_{v_1} \dots D_{v_d} f)^2 \geq (D_{v_1} D_{v_1} D_{v_3} \dots D_{v_d} f) \cdot (D_{v_2} D_{v_2} D_{v_3} \dots D_{v_d} f).$$

we can replace (A $\neq$ ) with an equivalent condition:

(L) The Hessian of  $Dv_1 Dv_2 \dots Dv_{d-2} f$  has exactly one pos. eigenvalue.

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• If  $C = \mathbb{R}_{>0}^n$ , then we say that  $f$  is Lorentzian.

In this case

(L)  $\Leftrightarrow Dv_1 \dots Dv_{d-2} f$  is stable.

Remark: Since stable<sup>pol.</sup> are closed under  $Dv f$ ,  $v \in \mathbb{R}_{>0}^n$ , and if  $f \neq 0$  has nonnegative coefficients, then

(P)  $Dv_1 \dots Dv_d f > 0$

(L)  $Dv_1 \dots Dv_{d-2} f$  is stable.

it follows that homogenous stable polynomials are Lorentzian.

## Examples

- Stable homogeneous polynomials are Lorentzian.
- The polynomial  $\det(X)$  is  $C$ -Lorentzian  
 $C = \{ \text{positive definite symmetric matrices} \}$
- Minkowski volume polynomials are Lorentzian.
- Normalized Schur and Schubert polynomials are Lorentzian.
- Various Matroid polynomials are Lorentzian.

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•  $C$ -Lorentzian is a closed property.

• If  $v_1, v_2 \in \overline{C}$ , then

$$f(sv_1 + tv_2) = \frac{1}{d!} \sum_{k=0}^d \binom{d}{k} \binom{d-k}{v_1} \binom{d-k}{v_2} f s^k t^{d-k}$$

$= a_k$

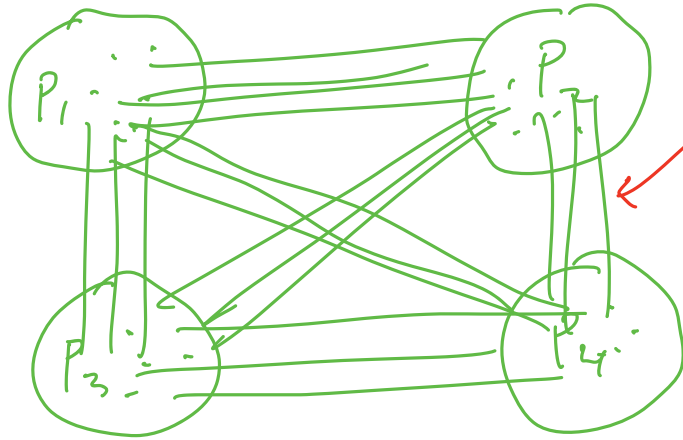
$$(AF) \Rightarrow a_k^2 \geq a_{k-1} a_{k+1}, \quad 0 < k < d.$$

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Non-example  $x^3 + y^3$ . (See later)

- How does a rank 2  $M$ -convex set look like?

- If  $M \subseteq \{0, 1\}^n$ . Let  $L \subseteq E$  be the loops, and let  $P_1, \dots, P_m$  be the parallel classes.



edge means  $e_i + e_j \in M$

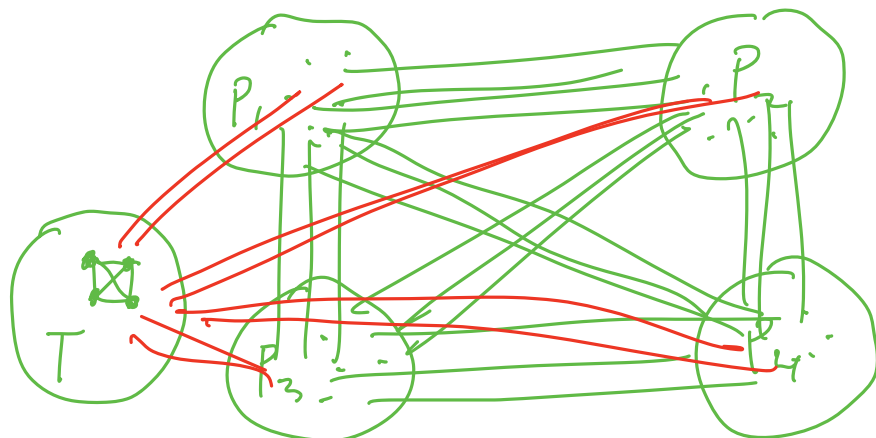
complete graph.

- If  $M \subseteq \{0, 1, 2\}^n$ .

Suppose  $\underset{\alpha}{2} e_k \in M$  and  $e_i + e_j \in M$

Then  $\alpha - e_i + e_k = e_j + e_k \in M$  and  
 $\alpha - e_j + e_k = e_i + e_k \in M$

Let  $T = \{k : 2e_k \in M\}$



## Perron-Frobenius:

We say that a matrix  $A = (a_{ij})_{i,j=1}^n$  is regular if:

(P) All off-diagonal entries are nonnegative

(C) For every distinct  $i, j \in [n]$ , there is a sequence  $i = i_0, i_1, \dots, i_k = j$  s.t.

$$a_{i_0 i_1} \cdot a_{i_1 i_2} \cdot \dots \cdot a_{i_{k-1} i_k} \neq 0$$

Perron-Frobenius: A regular matrix  $A$  has a unique eigenvector whose entries are all positive. The corresponding eigenvalue is simple and is the largest eigenvalue of  $A$ .

Lemma (Bochner method). Let  $f \in \mathbb{R}[x_1, \dots, x_n]$  be a homogeneous polynomial of degree  $d \geq 3$ , and let  $x \in \mathbb{R}_{>0}^n$ . If

(1). The Hessian of  $f$  at  $x$  is regular, and for all  $i$ ,

(2).  $\partial_i f(x) > 0$ , and

(3). The Hessian of  $\partial_i f$  at  $x$  has exactly one positive eigenvalue,

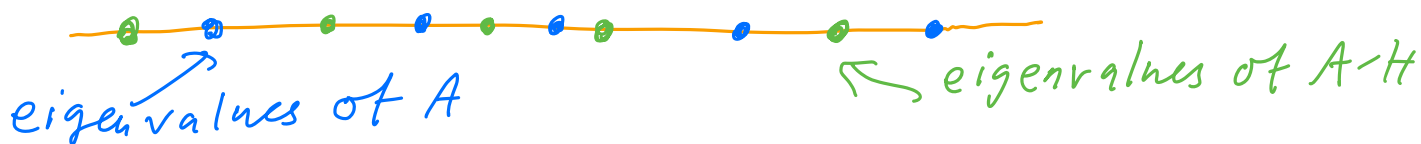
then the Hessian of  $f$  at  $x$  has exactly one positive eigenvalue.



# Proof of "Bochner method lemma":

A version of Cauchy interlacing (rank one perturbations) is

Lemma C: If  $A$  is a symmetric matrix and  $H$  is a rank one PSD matrix, then the eigenvalues of  $A-H$  interlaces  $\underbrace{\hspace{1cm}}_{\text{left to right}}$  the eigenvalues of  $A$ :



Notation:  $\nabla f = (\partial_1 f, \partial_2 f, \dots, \partial_n f)$ ,  $\nabla^2 f = (\partial_i \partial_j f)_{i,j=1}^n$

Lemma D: Suppose  $g \in \mathbb{R}[x_1, \dots, x_n]$  is homogeneous of degree  $d$ , and  $g(x) > 0$ . If  $\nabla^2 g(x)$  has exactly one positive eigenvalue, then

$d \cdot g \cdot \nabla^2 g - (d-1) \nabla g (\nabla g)^T$  (evaluated at  $x$ ) is negative semidefinite.

Proof: Recall Euler's formula for homogeneous functions:  $d \cdot g = \sum_{i=1}^n x_i \partial_i g$ . This gives

$$\begin{aligned} (a) \quad x^T \nabla^2 g x &= \sum_{i,j} x_i x_j \partial_i \partial_j g = \sum_i x_i \sum_j x_j \partial_j (\partial_i g) \stackrel{\text{E.F.}}{=} \\ &= \sum_i x_i (d-1) \partial_i g = d(d-1)g \end{aligned}$$

$$(b) \quad x^T \nabla g (\nabla g)^T x = \sum_{i,j} x_i x_j \partial_i g \partial_j g = d^2 g^2$$

If  $\nabla^2 g(x)$  has exactly one positive eigenvalue and  $g(x) > 0$ , then the matrix

$$B_\mu = d \cdot g \cdot \nabla^2 g - \mu \nabla g (\nabla g)^T, \quad \mu > 0$$

has at most one pos. eigenvalue by Lemma C (and exactly one pos. eigenvalue for  $\mu > 0$  sufficiently small).

By (a) and (b),  $x^T B_\mu x = d^2(d-1)g^2 - \mu d^2 g^2$ .

Hence when  $\mu \leq d-1$ ,  $B_\mu$  is NSD.  $\square$

• Recall the Löwner order on symmetric matrices:

$$A \leq B \quad \text{if} \quad B - A \text{ is PSD.}$$

Proof (of Bochner method): For ease of notation let  $f_i = \partial_i f$ . (All polynomials below are evaluated at  $x$ ).

By Lemma D:

$$(d-1) f_i \nabla^2 f_i \leq (d-2) \nabla f_i (\nabla f_i)^T \quad (\star)$$

Then by Euler's formula,

$$(d-2) \nabla^2 f = \left( (d-2) \partial_k \partial_l f \right)_{k,l} = \left( \sum_i x_i \partial_i \partial_k \partial_l f \right)_{k,l}$$

$$= \sum_i x_i \nabla^2 f_i \stackrel{(*)}{\leq} \sum_i \frac{x_i}{f_i} \frac{(d-2)}{(d-1)} \nabla f_i (\nabla f_i)^T$$

We may rewrite this inequality as

$$(d-1) \nabla^2 f \leq (\nabla^2 f) D (\nabla^2 f) \quad (*)$$

where  $D$  is the diagonal matrix

$$D = \text{diag}(x_1/f_1, x_2/f_2, \dots, x_n/f_n).$$

If  $A = D^{1/2} (\nabla^2 f) D^{1/2}$ , then  $(*)$  translates as

$$A^2 - (d-1)A \geq 0.$$

Hence if  $\lambda$  is an eigenvalue of  $A$ , then

$$0 \leq \lambda^2 - (d-1)\lambda = \lambda(\lambda - (d-1)),$$

i.e., no eigenvalue of  $A$  lies in the open interval  $(0, d-1)$ .

Consider the vector  $v = D^{-1/2} x = (x_i^{1/2} f_i^{1/2})_i$ .

Then

$$Av = D^{1/2} (\nabla^2 f) D^{1/2} D^{-1/2} x = D^{1/2} (\nabla^2 f) x$$

$$= D^{1/2} \left( \sum_j x_j d_{j\bar{j}} f \right)_i = (d-1) D^{1/2} (f_i)_i$$

$$= (d-1) \left( \frac{x_i^{1/2}}{f_i^{1/2}} f_i \right)_i = (d-1) v$$

Since  $\nabla^2 f$  is regular, so is  $A$ .

Hence by Perron-Frobenius,  $v$  is the unique positive eigenvector corresponding to the maximum eigenvalue (which is  $d-1$ ) of  $A$ . Since  $A$  has no eigenvalue in  $(0, d-1)$ , we conclude that the matrix

$$A = D^{\frac{1}{2}} (D^c f) D^{\frac{1}{2}}$$

has exactly one positive eigenvalue.

By Sylvester, so does  $D^c f$ .

□

Theorem (B., Huh). Let  $f \in \mathbb{R}_{\geq 0}[x_1, \dots, x_n]$  be homog. of degree  $d$ . Then  $f$  is Lorentzian iff

(M).  $\text{supp}(f)$  is  $M$ -convex, and

(L).  $\delta_{i_1} \dots \delta_{i_{d-2}} f$  has at most one positive eigenvalue.

## Recall:

Lemma (Bochner method). Let  $f \in \mathbb{R}[x_1, \dots, x_n]$  be a homogeneous polynomial of degree  $d \geq 3$ , and let  $x \in \mathbb{R}_{>0}^n$ . If

(1). The Hessian of  $f$  at  $x$  is regular, and for all  $i$ ,

(2).  $\partial_i f(x) > 0$ , and

(3). The Hessian of  $\partial_i f$  at  $x$  has exactly one positive eigenvalue,

then the Hessian of  $f$  at  $x$  has exactly one positive eigenvalue.

## Proof of thm:

$\Rightarrow$ ). Assume  $f$  is Lorentzian. Then  $Dv_1 \dots Dv_{d-2} f$  has Hessian with exactly one pos. eigenvalue. But then by continuity  $\partial_{i_1} \partial_{i_2} \dots \partial_{i_{d-2}} f$  is either  $\equiv 0$ , or has at most one positive eigenvalue. This gives (L).

(M). Induction over  $d \geq 2$ . If  $d=2$ , then  $f$  is stable, so  $\text{supp}(f)$  is  $M$ -convex. By definition Lorentzian polynomials are closed under  $Dv f$ ,  $v \in \mathbb{R}_{>0}^n$ , but the Lorentzian property is closed

so  $\partial_i f$  is either identically zero  
or Lorentzian if  $f$  is Lorentzian.  
Hence the supports of Lorentzian pol.  
are closed under  $\partial_i M$  and  $\tau M$ .

(M) follows from Lemma R (by induction<sup>n</sup>)

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Other direction next time.