

Recap.

- $f \in \mathbb{R}[x_1, \dots, x_n]$ is (real-) stable if $f(x_1, \dots, x_n) \neq 0$ whenever $\lambda_j(x_j) > 0 \ \forall j$

Equivalently if $t \rightarrow f(te - x)$ is real-rooted for all $e \in \mathbb{R}_{>0}^n$, $x \in \mathbb{R}^n$

- f stable $\implies D_v f = \sum_{i=1}^n v_i \frac{\partial f}{\partial x_i}$ stable for all $v \in \mathbb{R}_{\geq 0}^n$.

- If A_0 is hermitian, and A_1, \dots, A_n are PSD, then

$$f(x) = \det(A_0 + x_1 A_1 + \dots + x_n A_n)$$

is stable.

Example Multivariate Eulerian polynomials

For $\sigma = \sigma_1 \sigma_2 \dots \sigma_n \in \mathfrak{S}_n$, let

$$DT(\sigma) = \{ \sigma_i : \sigma_i > \sigma_{i+1} \}, \text{ where } \sigma_0 = \sigma_{n+1} = 0$$

$$AT(\sigma) = \{ \sigma_i : \sigma_{i-1} < \sigma_i \}$$

$$A_n(x, y) = \sum_{\sigma \in \mathfrak{S}_n} x^{DT(\sigma)} y^{AT(\sigma)}, \quad x^S = \prod_{i \in S} x_i$$

$$A_1 \quad 010$$

$$\downarrow$$

$$x, y$$

Insert $n+1$ in a permutation $\sigma \in \mathfrak{S}_n$

$$\begin{array}{ccc} \sigma_i > \sigma_{i+1} & \cdot & \text{Effect} \\ \uparrow & & \downarrow \\ n+1 & & x^{DT(\sigma)} y^{AT(\sigma)} \\ & & \downarrow \\ x_{n+1} y_{n+1} \frac{\partial}{\partial x_{\sigma_i}} x^{DT(\sigma)} y^{AT(\sigma)} \end{array}$$

Recursion

$$A_{n+1} = x_{n+1} y_{n+1} \left(\sum_{i=1}^n \left(\frac{\partial}{\partial x_i} + \frac{\partial}{\partial y_i} \right) \right) A_n$$

\rightarrow preserves stability

\uparrow preserves stability

By induction A_n is stable.

Example Plücker-polynomial

$$A = [v_1, v_2, \dots, v_n] \in \mathbb{C}^{r \times n}, \text{ rank} = r$$

$$f = \det(x_1 v_1 v_1^* + x_2 v_2 v_2^* + \dots + x_n v_n v_n^*)$$

is stable

$$f = \det \left(A \begin{bmatrix} x_1 & & 0 \\ & \ddots & \\ 0 & & x_n \end{bmatrix} A^* \right)$$

Cauchy-Binet

$$= \sum_{\substack{S \subseteq [n] \\ |S|=r}} \det(A(S)) \overline{\det(A(S))} \prod_{i \in S} x_i$$

\uparrow $r \times r$
submatrix using
in rows S

support of $f = \{S : \text{coeff. of } S \text{ is nonzero}\}$

= set of bases of the linear
matroid defined by v_1, \dots, v_n .

Characterization of homogeneous stable quadratics

Sylvester's law of inertia: Let A be a real symmetric $n \times n$ matrix. Then $P^T A P$ has the same signature as A for all $P \in GL_n(\mathbb{R})$.

Cauchy interlacing: The eigenvalues of any principal submatrix of a real symmetric matrix A interlaces the eigenvalues of A .

$$\lambda_n \leq \mu_{n-1} \leq \lambda_{n-1} \leq \mu_{n-2} \leq \dots \leq \lambda_2 \leq \mu_1 \leq \lambda_1$$

eigenvalues of A

eigenvalues of submatrix

Any hom. quadratic may be written

$$f = \sum_{i,j} a_{ij} x_i x_j = x^T A x$$

$$A = \frac{1}{2} \left(\partial_i \partial_j f \right)_{i,j=1}^n = \frac{1}{2} \nabla^2 f$$

Prop. Let $f = x^T A x$, where A is a symmetric $n \times n$ matrix with nonnegative coeff. TFAE

(1). A has exactly one positive eigenvalue.

(2). A is stable (and $\neq 0$)

(3). $(e^T A x)^2 \geq (e^T A e)(x^T A x)$

for all $e \in \mathbb{R}_{>0}^n$ and $x \in \mathbb{R}^n$

(4). —||—

for some $e \in \mathbb{R}^n$ s.t. $e^T A e > 0$
and all $x \in \mathbb{R}^n$

Proof: $e \in \mathbb{R}_{>0}^n$

$$\begin{aligned} f(t) &= (te - x)^T A (te - x) \\ &= t^2 (e^T A e) - 2t (e^T A x) + (x^T A x) \end{aligned}$$

$$\text{stable} \Leftrightarrow (e^T A x)^2 \geq (e^T A e)(x^T A x)$$

which proves (2) \Leftrightarrow (3).

(1) \Rightarrow (2): Take $e \in \mathbb{R}_{>0}^n$ and $x \in \mathbb{R}^n$.

If $e \parallel x$, then (3) is obvious.

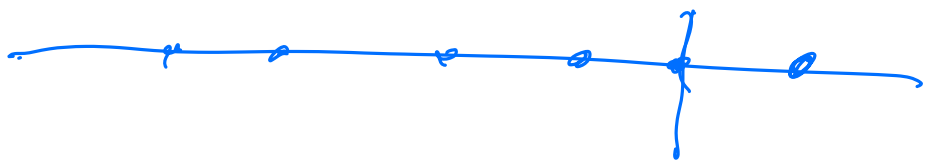
Assume $e \not\parallel x$. Extend to a basis

$$P = [e, x, v_3, \dots, v_n] \in GL_n(\mathbb{R})$$

Sylvester implies

$$P^T A P = \begin{bmatrix} e^T A e & e^T A x & \dots \\ e^T A x & x^T A x & \dots \\ \vdots & \vdots & \ddots \end{bmatrix}$$

has exactly one pos. eigenvalue



Cauchy interlacing says that

$B = \begin{bmatrix} e^T A e & e^T A x \\ e^T A x & x^T A x \end{bmatrix}$ has at most one pos. eigenvalue.

Because $e^T A e > 0$, $e_i^T B e_i > 0$, so B is not NSD, so it has exactly one pos. eigenvalue.

standard basis vector

$\therefore \det(B) \leq 0$

$\llcorner (e^T A e)(x^T A x) - (e^T A x)^2$

(3) \Rightarrow (4) trivially

(4) \Rightarrow (1). Assume (4). ^{Then} $e^T A e > 0$

$(e^T A x)^2 \geq (e^T A e) \cdot (x^T A x), \quad \forall x \in \mathbb{R}^n$

Let

$H = \{ x \in \mathbb{R}^n : e^T A x = 0 \}$ hyperplane

Then A (restricted to H) is NSD.

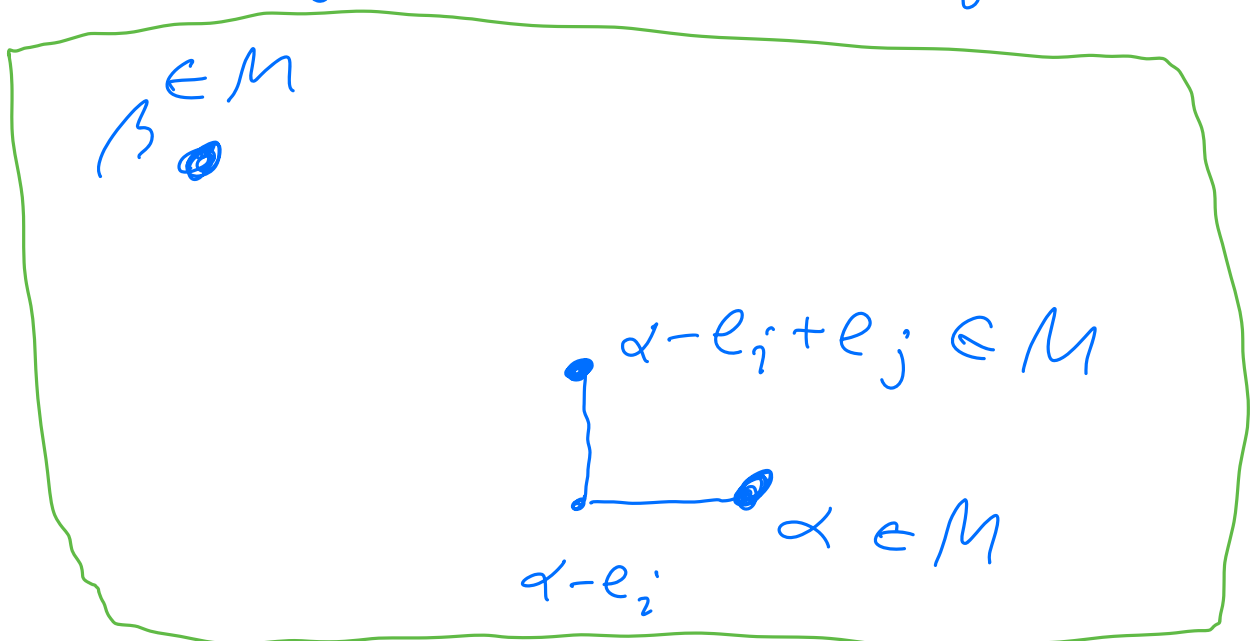
By Cauchy and Sylvester we see that A has at most one positive eigenvalue. Since $e^T A e > 0$ it has a positive eigenvalue. \square

M-convex sets (aka
 polymatroids,
 integer points of
 generalized permutahedra)

Def. M finite subset \mathbb{N}^n .
 M is called M-convex if

(EA) $\alpha, \beta \in M, \alpha_i > \beta_i \implies$

$\exists j \beta_j > \alpha_j$ and $\alpha - e_i + e_j \in M$



Example: $M \subseteq \{0, 1\}^n$, then M

is M -convex iff M is the set of bases of a matroid.

Def. If $f = \sum_{\alpha \in \mathbb{N}^n} a(\alpha) x^\alpha$ ($x^\alpha = x_1^{\alpha_1} \dots x_n^{\alpha_n}$)
 The support of f is

$$\text{supp}(f) = \{ \alpha \in \mathbb{N}^n : a(\alpha) \neq 0 \}$$

$$i \in [n] : \partial_i M = \{ \alpha - e_i : \alpha \in M, \alpha_i > 0 \}$$

(contraction)

$$\tau M = \bigcup_{i \in [n]} \partial_i M \quad (\text{truncation})$$

$$a(\alpha) \geq 0 : \text{supp}(\partial_i f) = \partial_i \text{supp}(f)$$

$$v \in \mathbb{R}_{>0}^n : \text{supp}(D_v f) = \tau \text{supp}(f).$$

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

Lemma Suppose $M \subseteq \mathbb{N}^n$, $\sum_{i=1}^n \alpha_i > 2$,
 for some $\alpha \in M$.

If $\partial_i M$ is M -convex for all i
 τM is M -convex,

then M is M -convex,