

# Lecture Series 3

October 5, 2024

In this chapter, we will use  $\mathcal{H} = \{z : \text{Im}(z) > 0\}$  to denote the open upper half plane, and  $\mathbf{z}$  to be the vector  $\{z_1, \dots, z_n\}$ .

## 1 Stable Polynomials

We have previously defined the notion of real stable polynomials and introduced some Lemmas and Theorems that was used to prove Theorem 2.13 in Chapter 2.

**Definition 1.1.** A polynomial  $f(\mathbf{z})$  is real stable (resp. stable) if and only if for every  $\mathbf{e} \in \mathbb{R}_{>0}^n$  and  $\mathbf{x} \in \mathbb{R}^n$ , the univariate restriction

$$t \mapsto f(t\mathbf{e} + \mathbf{x})$$

is real-rooted (resp. stable).

**Lemma 1.2.** For positive semidefinite matrices  $A_1, \dots, A_n \succeq 0$  and Hermitian  $B$ , the determinantal polynomial

$$\det \left( \sum_{i=1}^n z_i A_i + B \right)$$

is real stable.

*Proof sketch.* Assume that  $A_i$  are positive definite and consider a univariate restriction

$$t \mapsto \det \left( t \sum_{i=1}^n e_i A_i + \left( \sum_{i=1}^n x_i A_i + B \right) \right)$$

Since the  $e_i$  are positive,  $M := \sum_{i=1}^n e_i A_i \succ 0$  has a negative square root  $M^{-1/2}$  and we may write the above as

$$t \mapsto \det \left( M^{-1/2} \right) \det \left( tI + M^{1/2} \left( \sum_{i=1}^n x_i A_i + B \right) M^{1/2} \right) \det \left( M^{-1/2} \right)$$

Since this is a multiple of a characteristic polynomial of a Hermitian matrix, it must be real-rooted.

The positive semidefinite case could be handled by taking a limit of positive definite matrices. Recall that the limit along each univariate restriction must be real-rooted or zero.  $\square$

**Theorem 1.3.** [2] *If  $q(x, y)$  is a real stable polynomial of degree  $d$ , then there are real symmetric  $d \times d$  positive semidefinite matrices  $A, B$  and symmetric matrix  $C$  such that*

$$q(x, y) = \pm \det(xA + yB + C).$$

This theorem is known to be false for more than 2 variables.

**Example 1.4.** *Let  $G = (V, E)$  be a connected undirected graph. Then the spanning tree polynomial*

$$P_G(\mathbf{z}) = \sum_{\text{spanning tree } T} \prod_{e \in T} z_e$$

*is real stable.*

The following Theorem states the closure properties of some linear transformations.

**Theorem 1.5.** *The following linear transformations on  $\mathbb{C}[\mathbf{z}]$  maps every stable polynomial to another stable polynomial or to zero.*

1. *Permutation.*  $f(z_1, z_2, \dots, z_n) \mapsto f(z_{\sigma(1)}, \dots, z_{\sigma(n)})$  for some permutation  $\sigma : [n] \rightarrow [n]$ .
2. *Scaling.*  $f(z_1, \dots, z_n) \mapsto f(az_1, \dots, z_n)$  where  $a > 0$ .
3. *Diagonalization.*  $f(z_1, z_2, \dots, z_n) \mapsto f(z_2, z_2, z_3, \dots, z_n) \in \mathbb{C}[z_2, \dots, z_n]$ .
4. *Inversion.*  $f(z_1, z_n) \mapsto z_1^d f(-1/z_1, \dots, z_n)$  where  $d = \deg_1(f)$  is the degree of  $z_1$  in  $f$ .
5. *Specialization.*  $f \mapsto f(a, z_2, \dots, z_n) \in \mathbb{C}[z_2, \dots, z_n]$  where  $a \in \mathcal{H} \cup \mathbb{R}$ .
6. *Differentiation.*  $f \mapsto \frac{\partial}{\partial z_1} f$ .

*Proof.* (1) to (3) follows from definition.

(4) follows because  $z \mapsto -1/z$  preserves the upper half plane.

(6) is a consequence of the Gauss-Lucas Theorem below.  $\square$

**Theorem 1.6** (Gauss-Lucas). *If  $f \in \mathbb{C}[z]$ , the roots of  $f'(z)$  lie in the convex hull of the roots of  $f(z)$ .*

*Proof.* Let  $\lambda_1, \dots, \lambda_n \in \mathbb{C}$  be the roots of  $f$  and assume WLOG that  $f$  and  $f'$  have no common roots. If  $f' = 0$ , then we have

$$0 = \frac{f'(z)}{f(z)} = \sum_{i=1}^n \frac{1}{z - \lambda_i} = \sum_{i=1}^n \frac{\overline{z - \lambda_i}}{|z - \lambda_i|^2}$$

Rearranging, we obtain

$$z = \sum_{i=1}^n \frac{|z - \lambda_i|^{-2}}{\sum_{i=1}^n |z - \lambda_i|^{-2}} \lambda_i$$

which is a convex combination. □

## 1.1 A probabilistic application

### 1.1.1 Poisson Binomial Distribution

The distribution of a sum of independent Bernoulli random variables is called a Poisson Binomial Distribution, that is

$$X = \sum_{i=1}^n X_i$$

where  $X_i$  are independent Bernoullis with  $\mathbb{E}(X_i) = b_i \in (0, 1)$ , and taking  $p_k = \mathbb{P}[X = k]$ . We are interested in knowing if such a distribution is unimodal, that is whether there is some  $m$  such that  $p_0 \leq p_1 \leq \dots \leq p_m \geq \dots \geq p_n$ .

Consider the following generating function of the distribution

$$q(x) \stackrel{\text{def}}{=} \sum_{k=0}^n p_k x^k = \prod_{i=1}^n (b_i x + (1 - b_i))$$

where the independence of  $X_i$  yields a factorization of  $q(x)$  into linear terms. This factorization implies that  $q(x)$  is real-rooted with strictly negative roots  $\lambda_i := -\frac{1-b_i}{b_i} < 0$ . Using the following Newton's Inequalities, which states

**Theorem 1.7** (Newton Inequalities). *If  $\sum_{k=0}^n a_k x^k$  is real-rooted, then*

$$\left( \frac{a_k}{\binom{n}{k}} \right)^2 \geq \frac{a_{k-1}}{\binom{n}{k-1}} \frac{a_{k+1}}{\binom{n}{k+1}}$$

for  $k = 1, \dots, n - 1$ .

After cancellation of the factorials, it reduces to

$$a_k^2 \geq \left(1 + \frac{1}{k}\right) \left(1 + \frac{1}{n-k}\right) a_{k-1} a_{k+1},$$

which is strictly stronger than

$$a_k^2 \geq a_{k_1} a_{k+1} \quad (\text{log-concavity})$$

This implies unimodality, where the probabilities  $p_k$  must be unimodal. Now we introduce the next proposition, which will be used in proving the next Theorem.

**Proposition 1.8.** *Suppose  $p(x) = \sum_{k=0}^n a_k x^k$  is a real-rooted polynomial with nonnegative coefficients and  $a_0 \neq 0, p(1) = 1$ . Then there are independent Bernoulli random variables  $X_1, \dots, X_n$  such that*

$$a_k = \mathbb{P} \left[ \sum_{i=1}^n X_i = k \right]$$

*Proof.* Factor  $p(x)$  as  $C \prod_{i=1}^n (x + \lambda_i)$  for some  $\lambda_i > 0$ . Since  $p(1) = 1$ , we must have

$$C = \frac{1}{\prod_{i=1}^n (1 + \lambda_i)}$$

Then we have

$$p(x) = \prod_{i=1}^n (b_i x + (1 - b_i))$$

for  $b_i = \frac{1}{1 + \lambda_i} \in (0, 1)$ . Taking  $X_i$  with  $\mathbb{E}(X_i) = b_i$  proves the claim.  $\square$

### 1.1.2 Application

Suppose  $G = (V, E)$  is a graph,  $F \subset E$  is a cut, and  $T$  is a uniformly random spanning tree of  $G$ . The distribution of the random variable  $|F \cap T|$  is a Poisson Binomial Distribution.

**Theorem 1.9.** *The distribution of  $|F \cap T|$  is a Poisson Binomial Distribution.*

*Proof.* The generating polynomial of a random variable  $T \cap F$  is obtained by setting all the variables  $\mathbf{z}_e, e \notin F$ :

$$Q_G(\mathbf{z}|_F) = P_G(\mathbf{z}_F, 1, \dots, 1),$$

where we observe that the coefficient of the monomial  $z^S := \prod_{e \in S} z_e$  in  $Q_G$  is equal to the number of spanning trees  $T$  for which  $T \cap F = S$ . As setting the variables to real numbers preserve stability,  $Q_G$  is real stable. Thus its diagonal restriction

$$Q_G(x, x, \dots, x) = \sum_{k=0}^{|F|} x^k \mathbb{P}[|T \cap F| = k]$$

must be real-rooted. Normalizing by  $Q_G(1, 1, \dots, 1)$  and applying the previous proposition finishes the proof.  $\square$

## 1.2 Characterization of Stability Preserving Operators

Let  $\mathbb{C}_k[z_1, \dots, z_n]$  be the vector space of complex polynomials in  $z_1, \dots, z_n$  in which each variable has degree of at most  $k$ . We call a linear transformation nondegenerate if its range has dimension at least 2.

**Theorem 1.10.** *A nondegenerate linear operator  $T : \mathbb{C}_k[z_1, \dots, z_n] \rightarrow \mathbb{C}[z_1, \dots, z_n]$  preserves stability iff the  $2n$ -variate polynomial*

$$G_T(z_1, \dots, z_n, w_1, \dots, w_n) := T \left[ (z_1 + w_1)^k \dots (z_n + w_n)^k \right]$$

is stable, where the operator  $T$  only acts on the  $z$  variables.

This theorem says that there is a single  $2n$ -variate polynomial whose stability guarantees the stability of all of the  $n$ -variate images  $T(f)$ .

### 1.2.1 Heilmann-Lieb Theorem

Given a graph with nonnegative edge weights  $w_e \geq 0, e \in E$ , we define

$$m_k := \sum_{\text{matching } M, |M|=k} \prod_{e \in M} w_e$$

The relevant generating function is the matching polynomial

$$\mu_G(x) := \sum_{k=0}^{n/2} x^{n-2k} (-1)^k m_k.$$

We state the following theorem:

**Theorem 1.11** (Heilmann-Lieb). *For every weighted graph  $G$  with nonnegative edge weights,  $\mu_G(x)$  is real-rooted.*

*Proof.* Given a graph  $G$  with positive edge weights  $w_{uv} > 0, uv \in E$ , consider the multivariate polynomial

$$Q_G(\mathbf{z}) = \prod_{uv \in E} (1 - w_{uv} z_u z_v),$$

where the variables  $z_v$  are indexed by  $v \in V$ . As  $Q_G$  is a product of real stable polynomials, it is real stable. Consider the multiaffine part operator

$$\text{MAP} : \mathbb{C}[\mathbf{z}] \rightarrow \mathbb{C}_1[\mathbf{z}]$$

defined on monomials of degree at most  $m := |E|$  in each variable by

$$\text{MAP} \left( \prod_{e \in S} z_e^{d_e} \right) = \begin{cases} \prod_{e \in S} z_e & \text{if } d_e \leq 1 \text{ for all } e \\ 0 & \text{otherwise} \end{cases}$$

The symbol of this operator is given by

$$G_{\text{MAP}}(\mathbf{z}, \mathbf{w}) = \text{MAP} \left( \prod_{v \in V} (z_v + w_v)^m \right) = \prod_{v \in V} (w_v^m + m z_v w_v^{m-1}) = \prod_{v \in V} w_v^{m-1} (w_v + m z_v),$$

which is real stable. Since MAP is nondegenerate, Theorem 1.10 states that it preserves stability. Thus,

$$\text{MAP}(Q_G) = \sum_{\text{matching } M} (-1)^{|M|} \prod_{\text{edge } uv \in M} w_{uv} \prod_{\text{vertex } v \in M} z_v$$

is real stable, and its univariate diagonal restriction  $z_v \leftarrow x, v \in V$ :

$$\sum_{\text{matching } M} (-1)^{|M|} x^{2|M|} \prod_{uv \in M} w_{uv}$$

is real-rooted. But this is just the reversal of  $\mu_G(x)$  □

## 2 Multiaffine Real Stable Polynomials

A multiaffine polynomial is a multivariate polynomial in which each variable has degree at most one. We use  $\mathbb{R}_1[z_1, \dots, z_n]$  or  $\mathbb{R}_{MA}[z_1, \dots, z_n]$  to denote vector spaces of multiaffine polynomials.

**Definition 2.1.**  $f \in \mathbb{R}[z_1, \dots, z_n]$  is *Strongly Rayleigh* if for every  $i \neq j$ ,

$$\partial_{z_i} f(\mathbf{x}) \cdot \partial_{z_j} f(\mathbf{x}) \geq \partial_{z_i z_j} f(\mathbf{x}) \cdot f(\mathbf{x}) \quad \forall \mathbf{x} \in \mathbb{R}^n.$$

**Theorem 2.2.** [1] *A real multiaffine polynomial  $f \in \mathbb{R}_1[z_1, \dots, z_n]$  is stable if and only if it is Strongly Rayleigh.*

*Proof.* ( $\Rightarrow$ ) Suppose  $f$  is real stable. Fix  $\mathbf{x} \in \mathbb{R}^n$ . Consider the bivariate restriction

$$g(s, t) := f(\mathbf{x} + s e_i + t e_j)$$

which is a multiaffine bivariate polynomial

$$g(s, t) = a + bs + ct + dst$$

with real coefficients

$$a = f(\mathbf{x}) \quad b = \partial_{z_i} f(\mathbf{x}) \quad c = \partial_{z_j} f(\mathbf{x}) \quad d = \partial_{z_i z_j} f(\mathbf{x})$$

Since every univariate restriction of  $g$  along a direction in the positive orthant  $\mathbb{R}_{>0}^2$  is a restriction of a specialization of  $f$  (by fixing all the variables other than  $z_i, z_j$ ), all such restrictions are real-rooted and  $g$  is itself real stable. Observe by applying the closure properties that for every  $\lambda > 0$ , the polynomial  $g(\lambda r, r) = a + (\lambda b + c)r + d\lambda r^2$  must be real-rooted, hence  $(\lambda b + c)^2 \geq 4ad\lambda$  for

all  $\lambda > 0$ . If  $b$  and  $c$  are nonzero of the same sign then setting  $\lambda = c/b$  yields the inequality. If they have opposite signs or if one of them is zero then we can see that  $g$  cannot be real stable unless it is zero.

( $\Leftarrow$ ) We prove this by induction. Suppose  $f(\mathbf{z}, z_{n+1}) = g(\mathbf{z}) + z_{n+1}h(\mathbf{z})$  is Strongly Rayleigh, with  $g, h \in \mathbb{R}_1[z_1, \dots, z_n]$ . Note that both  $g(\mathbf{z}), h(\mathbf{z})$  are Strongly Rayleigh by closure properties. Let  $z_{n+1} = \alpha \in \mathbb{R}$ . Observe that  $g(\mathbf{z}) + \alpha h(\mathbf{z}) \in \mathbb{R}_1[\mathbf{z}]$  is Strongly Rayleigh. By induction, it must be stable for every  $\alpha$ . If it is identically zero for some  $\alpha$ , then  $g(\mathbf{z}) \equiv -\alpha h(\mathbf{z})$  and we may factor  $f$  as  $f(\mathbf{z}, z_{n+1}) = (z_{n+1} - \alpha)h(\mathbf{z})$ , which is stable and we are done.

Otherwise,  $g(\mathbf{z}) + \alpha h(\mathbf{z}) \neq 0$  for all  $\alpha$  and for all  $\mathbf{z} \in \mathcal{H}^n$ . This means that

$$\Phi(\mathbf{z}) := \frac{g(\mathbf{z})}{h(\mathbf{z})} \notin \mathbb{R} \quad \forall \mathbf{z} \in \mathcal{H}^n$$

Since  $\Phi$  is continuous on  $\mathcal{H}^n$  we must have either

$$\begin{aligned} \operatorname{Im}(\Phi(\mathbf{z})) > 0 \quad \forall \mathbf{z} \in \mathcal{H}^n \quad \text{or} \\ \operatorname{Im}(\Phi(\mathbf{z})) < 0 \quad \forall \mathbf{z} \in \mathcal{H}^n \end{aligned}$$

In the latter case it is immediate that  $f(\mathbf{z}, z_{n+1})$  is stable. In the former, we find that by changing the sign of  $z_{n+1}$ ,  $f(\mathbf{z}, -z_{n+1})$  must be stable. By the forward direction of the theorem proven earlier, this means that it is Strongly Rayleigh. Applying the definition of Strongly Rayleigh to the pairs  $i, n+1$ , we obtain the reversed inequalities:

$$\partial_{z_i} f(\mathbf{x}) \cdot \partial_{z_{n+1}} f(\mathbf{x}) \leq \partial_{z_i z_{n+1}} f(\mathbf{x}) \cdot f(\mathbf{x}) \quad \forall \mathbf{x} \in \mathbb{R}^n$$

Since  $f$  is also Strongly Rayleigh these must be equalities. We could check that this is only possible when  $g = h$ .  $\square$

## 2.1 Multiaffine Stability Preservers

In this subsection, we derive a sufficient condition for establishing that a linear transformation on  $\mathbb{C}_1[z_1, \dots, z_n]$  preserves stability. The main purpose of this subsection is to show that a transformation  $T$  preserves stability of  $n$ -variate polynomials. It suffices to show that the stability of a single  $2n$ -variate generating polynomial derived from it.

**Lemma 2.3** (Lieb-Sokal). *Suppose  $f(\mathbf{z}) + wg(\mathbf{z}) \in \mathbb{C}[z, w]$  is stable and the degree of  $z_1$  in  $g$  is at most 1. Then*

$$f(\mathbf{z}) - \partial_{z_1} g(\mathbf{z}) \in \mathbb{C}[\mathbf{z}]$$

*is stable.*

*Proof.* By closure properties,  $f$  and  $g$  must be stable. Let  $\alpha \in \mathcal{H}$ . Observe that  $-\alpha^{-1} \in \mathcal{H}$ . Hence

$$h(\mathbf{z}, \alpha) := \alpha g(z_1 - \alpha^{-1}, z_2, \dots, z_n) \neq 0$$

for all  $\mathbf{z} \in \mathcal{H}^n$ . Since  $g$  is affine in  $z_1$ , then  $h$  is a multiaffine stable polynomial. Expanding the first variable, we find that

$$h(\mathbf{z}, \alpha) = \alpha g(\mathbf{z}) - (\partial_{z_1} g)(\mathbf{z})$$

is stable. Rearranging, we have

$$\operatorname{Im} \left( \frac{-\partial_{z_1} g(\mathbf{z})}{g(\mathbf{z})} \right) \geq 0$$

for all  $\mathbf{z} \in \mathcal{H}^n$ . Similarly, since  $f + wg$  is stable, we know that  $\operatorname{Im}(f(\mathbf{z})/g(\mathbf{z})) \geq 0$  for all  $\mathbf{z} \in \mathcal{H}^n$ . Hence, summing up we have

$$\operatorname{Im} \left( \frac{-\partial_{z_1} g(\mathbf{z}) + f(\mathbf{z})}{g(\mathbf{z})} \right) \geq 0 \quad \forall \mathbf{z} \in \mathcal{H}^n,$$

so  $\partial_{z_1} g(\mathbf{z}) + f(\mathbf{z}) + vg(\mathbf{z}) \in \mathbb{R}[\mathbf{z}, v]$  is stable. Specializing  $v$  to 0 gives the result.  $\square$

This Lemma is used in the proof of the following Borcea-Branden Theorem, which follows because every linear operator on a space of bounded degree polynomials can be written as an appropriate sum of differential operators.

**Theorem 2.4** (Borcea-Branden). *Suppose  $T : \mathbb{C}_1[\mathbf{z}] \rightarrow \mathbb{C}[\mathbf{z}]$  is a linear operator. If the algebraic symbol*

$$G_T(z_1, \dots, z_n, w_1, \dots, w_n) := T \left( \prod_{j=1}^n (z_j + w_j) \right) = \sum_{S \subset [n]} T(z^S) w^{[n] \setminus S}$$

*is stable then  $T$  is stability-preserving.*

*Proof.* Suppose  $f(\mathbf{z}) = \sum_{S \subset [n]} a_S z^S \in \mathbb{C}_1[\mathbf{z}]$  is stable. Since  $w_j \mapsto -1/w_j$  preserves  $\mathcal{H}$ , the hypothesis implies that

$$w_1 \dots w_n G(z_1, \dots, z_n, -1/w_1, \dots, -1/w_n) = \sum_{S \subset [n]} T(z^S) (-1)^{n-|S|} w^S$$

is stable. Multiplying by  $f(v_1, \dots, v_n) \in \mathbb{C}_1[v_1, \dots, v_n]$ , we find that

$$\sum_{S \subset [n]} T(z^S) (-1)^{n-|S|} w^S f(v_1, \dots, v_n)$$

is stable. Since  $f$  is multiaffine, we can use Lieb-Sokal Lemma to replace each  $w_i$  by  $-\partial_{v_i}$ , which shows

$$\sum_{S \subset [n]} T(z^S) (-1)^{n-|S|} (-1)^{|S|} (\partial_v^S) f(v_1, \dots, v_n)$$



is stable, where  $\partial_v^S := \prod_{j \in S} \partial_{v_j}$ . Setting  $v_1 = \dots = v_n = 0$  preserves stability. So

$$\sum_{S \subset [n]} T(z^S) (-1)^n (\partial_v^S) f(0, \dots, 0)$$

is stable. Since  $(\partial_v^S f = a_S$ , the expression is equal to

$$(-1)^n \sum_{S \subset [n]} T(z^S) a_S = (-1)^n T(f).$$

Thus  $T$  preserves stability as we desire.  $\square$

## 2.2 Polarization and High Degree Polynomials

Polarization is a tool that allows one to transfer results about multiaffine polynomials to polynomials of higher degree.

**Definition 2.5.** *Given a polynomial  $f \in \mathbb{C}_k[z_1, \dots, z_n]$ , the polarization of  $f$  is the unique polynomial*

$$F \in \mathbb{C}_1[z_{11}, \dots, z_{1k}, \dots, z_{n1}, \dots, z_{nk}]$$

such that

1. The restriction  $z_{ji} \leftarrow z_j, j = 1, \dots, n$  is equal to the original polynomial:

$$F(z_1, \dots, z_1, \dots, z_n, \dots, z_n) = f(z_1, \dots, z_n)$$

2. For every  $j = 1, \dots, n$ ,  $F$  is symmetric in  $z_{j1}, \dots, z_{jn}$ .

The polarization operation is denoted by  $F = \Pi_k^\uparrow(f)$ . The inverse is called projection and denoted by  $f = \Pi_k^\downarrow(F)$ . It is trivial to see that if  $F$  is stable then  $f = \Pi_k^\downarrow(F)$  is also stable. The converse is also true, which is the following theorem.

**Theorem 2.6.** *If  $f \in \mathbb{C}_k[z_1, \dots, z_n]$  is stable then  $\Pi_k^\uparrow(f)$  is also stable.*

The proof is a result of the following Lemma:

**Lemma 2.7.** *If  $f \in \mathbb{C}_1[z_1, \dots, z_n]$  is stable, then for every  $\theta \in [0, 1]$  :*

$$(1 - \theta)f(z_1, z_2, \dots, z_n) + \theta f(z_2, z_1, \dots, z_n)$$

is stable.

*Proof.* Setting all the variables other than  $z_1, z_2$  to values in  $\mathcal{H}$ , it is sufficient to prove the claim for bivariate polynomials. We notice that

$$T : g(z_1, z_2) \mapsto (1 - \theta)g(z_1, z_2) + \theta g(z_2, z_1)$$

is a linear operator on  $\mathbb{C}_1[z_1, z_2]$ . Its symbol is

$$\begin{aligned} G_T(z_1, z_2, w_1, w_2) &= T((z_1 + w_1)(z_2 + w_2)) \\ &= z_1 z_2 + w_1((1 - \theta)z_2 + \theta z_1) + w_2((1 - \theta)z_1 + \theta z_2) + w_1 w_2 \end{aligned}$$

We will show that  $G_T$  is Strongly Rayleigh, which also implies that it is stable, which finishes the proof by Theorem 2.4. By symmetry, we only need to check the following two inequalities:

$$\partial_{z_1} G_T \cdot \partial_{z_2} G_T - \partial_{z_1 z_2} G_T \cdot G_T \geq 0$$

and

$$\partial_{z_1} G_T \cdot \partial_{w_1} G_T - \partial_{z_1 w_1} G_T \cdot G_T \geq 0$$

where the polynomials are evaluated at real points. We could simplify the expressions by computation to  $\theta(1 - \theta)(w_1 - w_2)^2$  and  $\theta(z_1 - w_1)^2$  respectively. Hence they must be nonnegative.  $\square$

*Proof sketch of Theorem 2.6.* Let  $T_{ij,\theta}$  be the partial symmetrization operator which swaps indices  $i$  and  $j$  with probability  $\theta$ . It can be shown by induction that for every  $n$  there is a finite sequence of pairs  $i_1 j_1, \dots, i_N j_N$  and numbers  $\theta_1, \dots, \theta_N$  so that for every polynomial  $f(z_1, \dots, z_n)$ :

$$T_{i_N j_N, \theta_N} \dots T_{i_1 j_1, \theta_1} f = \mathbb{E}_\sigma (f(z_{\sigma(1)}, \dots, z_{\sigma(n)})) =: \text{Sym}(f)$$

where the expectation is taken over a random permutation  $\sigma$  of  $[n]$  - i.e., we can generate a uniformly random permutation by performing appropriately biased swaps on a predetermined sequence of pairs. Hence the symmetrization operator  $\text{Sym}(f)$  preserves stability.

Let  $\Pi_{k,j}^\uparrow$  where  $j = 1, \dots, n$  be the operator which polarizes the variable  $z_j$  only, and note that  $\Pi_k^\uparrow = \Pi_{k,n}^\uparrow \circ \dots \circ \Pi_{k,1}^\uparrow$ . Thus it is sufficient to show that  $\Pi_{k,1}^\uparrow$  preserves stability. By setting every  $z_j, j \neq 1$  to a number in  $\mathcal{H}$ , it suffices to handle the univariate case. Thus we let

$$g(z) = C \prod_{i=1}^k (z - \alpha_i)$$

be a univariate stable polynomial. As each  $\alpha_i \notin \mathcal{H}$ , each the polynomials  $z_i - \alpha_i$  are stable, hence the product

$$G(z_1, \dots, z_k) = C \prod_{i=1}^k (z_i - \alpha_i)$$

is stable. By the previous paragraph,  $\text{Sym}(G)$  must be stable; but as  $\text{Sym}(G)$  is symmetric in  $z_1, \dots, z_k$  and projects to  $g$ , it is equal to the polarization of  $g$ .  $\square$

Hence a polynomial is stable iff its polarization is stable.

**Theorem 2.8.** *Suppose  $T : \mathbb{C}_k[z_1, \dots, z_n] \rightarrow \mathbb{C}[z_1, \dots, z_n]$  is a linear transformation and*

$$G_T(z_1, \dots, z_n, w_1, \dots, w_n) := T[(z_1 + w_1)^k \dots (z_n + w_n)^k]$$

*is stable. Then  $T$  preserves stability.*

*Proof.* Define an operator  $\Pi_k^\uparrow(T) : \mathbb{C}_1[z_{11}, \dots, z_{nk}] \rightarrow \mathbb{C}_1[z_{11}, \dots, z_{nk}]$  by

$$\Pi_k^\uparrow(T)(f) = \Pi_k^\uparrow \circ T \circ \Pi_k^\downarrow(f)$$

It is simple to check that  $G_{\Pi_k^\uparrow(T)} = \Pi_k^\uparrow(G_T)$ . Since  $G_T$  is stable, Theorem 2.6 implies that  $G_{\Pi_k^\uparrow(T)} = \Pi_k^\uparrow(G_T)$  is stable. Hence by Theorem 2.4,  $\Pi_k^\uparrow(T)$  preserves stability. But  $T = \Pi_k^\downarrow \circ \Pi_k^\uparrow(T) \circ \Pi_k^\uparrow$  which by Theorem 2.6 preserves stability.  $\square$

## References

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