# MULTIVARIATE CORRELATION INEQUALITIES FOR $P$-PARTITIONS 

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#### Abstract

Motivated by the Lam-Pylyavskyy inequalities for Schur functions, we give a far reaching multivariate generalization of Fishburn's correlation inequality for the number of linear extensions of posets. We then give a multivariate generalization of the Daykin-Daykin-Paterson inequality proving log-concavity of the order polynomial of a poset. We also prove a multivariate $P$-partition version of the cross-product inequality by Brightwell-Felsner-Trotter. The proofs are based on a multivariate generalization of the Ahlswede-Daykin inequality.


## 1. Introduction

Arguably, linear extensions play as much a central role in poset theory as standard Young tableaux in algebraic combinatorics. While the former combinatorial objects obviously generalize the latter, this connection is yet to be fully explored. In fact, the development in the two areas seem to move along parallel tracks as we explain below.

The story of this paper is an interplay between these two areas of combinatorics, which makes both the motivation and presentation of the results somewhat less accessible. To mitigate this, we include two separate (and almost completely non-overlapping) versions of the introduction addressing audiences with different background (see also §11.1).

The results themselves are postponed to later sections and assume fluency in both areas. While the reader may choose to read only the results that are closer to their interests, reading both sides of the story can enhance the experience. To help navigate between the areas, we include detailed notation and some background in Section 2.

Poset theoretic perspective. Our first result (Theorem 3.4) is a self-dual generalization of the remarkable Fishburn's correlation inequality (Theorem 3.1) for the numbers of linear extensions of poset order ideals. We further extend it to a correlation inequality for order polynomials, and then even further to their $q$-analogues and multivariate $\boldsymbol{q}$-analogues (Theorems 4.9 and 4.10). To understand the proofs it is worth examining the historical background and motivation behind earlier results.

Following up on the works by Harris (1960) and Kleitman (1966), Fortuin-Kasteleyn-Ginibre introduced the celebrated FKG inequality [FKG71]. This correlation inequality was further generalized in a series of papers, most notably by Ahlswede-Daykin [AD78], who proved a very general $A D$ inequality (Theorem 5.1), which is also called the four functions theorem [AS16, §6.1]. This result is so general that it has an elementary albeit somewhat involved proof by induction (ibid.). For the many followup investigations of correlation inequalities, see e.g. [AB08, §15], [Pak22, §5], and earlier overviews in [FS98, Gra83, Win86].

In a direct application to posets, Shepp [She80] was able to use the FKG inequality and a clever limit argument to prove the $X Y Z$ inequality (see e.g. [AS16, §6.4]), the most remarkable correlation inequality for linear extensions of posets, conjectured earlier by Rival and others. This brings us to Fishburn [Fis84], who established Fishburn's correlation inequality (Theorem 3.1)

[^0]as a tool in his proof of the strict version of the XYZ inequality. We note that Shepp's limit argument does not imply the strict version, so Fishburn's proof uses the AD inequality instead.

Motivated by enumerative applications and Fishburn's work, Björner [Bjö11] proved the $q-F K G$ inequality generalizing the FKG inequality. Christofides [Chr09] then found the $q-A D$ inequality, answering Björner's question. In a joint work with Panova [CPP22b], we employed Björner's $q$-FKG inequality to obtain $q$-analogues of inequalities for order polynomials of interest in enumerative combinatorics.

In our most recent paper [CP22], we find several correlation inequalities whose proof required the combinatorial atlas technique and does not have a natural $q$-analogue. Among other results, we proved a series of upper bounds on correlation inequalities (when they are written in the form of a ratio $\geq 1$ ), in some cases serving as a counterpart to the Fishburn's inequality.

The generality of our upper bounds in [CP22] and the self-dual nature of related results on Young tableaux naturally leads to our self-dual generalization of Fishburn's inequality. Just like the original proofs by Shepp and Fishburn, our proof is via the order polynomial, which naturally arises in this setting. Curiously, to prove our main theorem (Theorem 4.9), we use a multivariate generalization (Theorem 6.1) of Christofides's $q$-AD inequality.

At this point one would want to compare our results (notably Theorem 4.10), to those by Lam and Pylyavskyy [LP07], which are closely related and partly inspired this paper. They also prove a multivariate correlation inequality for order preserving maps on posets, which in some cases coincides with ours (cf. Corollary 4.5 and Remark 8.1). Unfortunately, their meet and join operations on order ideals are noncommutative and are therefore distinct from the more traditional definitions that we use. Thus, while the results in [LP07] might appear similar and even more general at a first glance (partially because they use the same notation), in full generality the similarity is misleading.

Now, Lam-Pylyavskyy's Cell Transfer Theorem [LP07, Thm 3.6] has a more general setting given by certain functions on poset's Hasse diagram. When it comes to skew Young diagrams, this allows the authors to recover the same reverse plane partitions results that we do, as well as semistandard Young tableaux results. We also recover their correlation inequality for Schur functions by making additional arguments (Section 8).

To summarize the comparison, neither result implies the other. Our meet and join notions are more standard, leading to a self-dual generalization of Fishburn's inequality. We are also using a more standard tool: the generalized AD inequality. On the other hand, the Lam-Pylyavskyy's ad hoc definitions allow them to recover the same Young tableaux results with an advantage of their proof giving an explicit combinatorial injection (cf. §11.2).

We give two applications of the multivariate AD inequality to poset inequalities. First, we prove a multivariate cross-product inequality for order preserving maps on posets (Theorem 10.1), giving a variation on the cross-product inequality by Brightwell-Felsner-Trotter [BFT95]. This result is new even for the usual (unweighted) setting. Note that the (original) cross-product inequality remains a conjecture in full generality (Remark 10.2).

Finally, we give a multivariate extension of the Daykin-Daykin-Paterson (DDP) inequality (Theorem 9.1), which was originally conjectured by Graham in [Gra83], and proved in [DDP84] by an ingenuous direct injection. ${ }^{1}$ In fact, Graham originally suggested that the DDP inequality could be proved by the AD inequality (see Remark 9.2 ). We provide such a proof in $\S 9.1$. Then, motivated by the structure of the multivariate AD inequality, we give a multivariate generalization of the DDP inequality (Theorem 9.3). We conclude with a multivariate log-concavity of the order polynomial (Corollary 9.5), generalizing our recent joint result with Panova [CPP22b].

[^1]Algebraic combinatorics perspective. Our main result is a generalization of the remarkable Lam-Pylyavskyy correlation inequality (Theorem 4.1) for Schur functions and reverse plane partitions to a self-dual (multivariate) correlation inequalities for general posets (Theorems 4.9 and 4.10). Specializations of our main result give correlation inequalities for $q$-analogues of the number standard Young tableaux for both straight and skew shapes, which generalize Björner's inequality (Corollary 3.2).

To understand the proofs it is worth examining the historical background and motivation behind earlier results. The study of inequalities for the symmetric functions goes back to Newton (1707), who proved the log-concavity $\mathbf{e}_{k}^{2} \geq \mathbf{e}_{k+1} \mathbf{e}_{k-1}$ of elementary symmetric polynomials $\mathbf{e}_{k}\left(x_{1}, \ldots, x_{n}\right)$, for all $x_{i} \in \mathbb{R}$. We refer to [Mac95, Sta99] for a thorough treatment of symmetric functions.

Over the past century, symmetric functions have received a great deal of attention due to their connections and applications in representation theory, as well as a host of other fields (enumerative algebraic geometry, integrable probability, etc.) With many identities came inequalities, which were often proved by tools from other areas. We refer to [Bre89, Bre94, Sta89] for somewhat dated surveys and to [Brä15, Huh18] for a more recent overviews of positivity results.

Some recent highlights include inequalities for values of Schur functions conjectured by Cuttler-Greene-Skandera [CGS11] and proved by Sra [Sra16], the log-concavity of normalized Schur polynomials by Huh-Matherne-Mészáros-St. Dizier [HMMS22], and the Schur positivity correlation inequality by Lam-Postnikov-Pylyavskyy [LPP07] (see Remark 4.2).

Building on the ideas which go back to MacMahon (1915), Stanley introduced in his thesis [Sta72] the P-partition theory, which is closely related to the study of the order polynomial of posets, and to the major index statistics on linear extensions [Sta99, §3.15]. Motivated by applications to plane partitions, the study of $P$-partitions became an important subject of its own. The order polynomial of a poset turned out to coincide with the Ehrhart polynomial of the order polytope (see e.g. [Sta99, §4.6.2]).

The Lam-Pylyavskyy paper [LP07] uses Stanley's $P$-partition theory to obtain inequalities for the numbers of $P$-partitions with multivariate weights. The authors presented an explicit combinatorial injection called the cell transfer, which proves inequalities in a very general setting. As the main application they succeeded in establishing the monomial positivity correlation inequality for Schur functions (Theorem 4.1), which was soon overshadowed by the stronger Schur positivity LPP correlation inequality mentioned above. Their approach also extends to monotonicity of quasisymmetric functions which arise from $P$-partitions [LP08].

In this paper, we take the core part of the Lam-Pylyavskyy general inequality and generalize it in the direction which is more natural from the poset theoretic point of view (Theorem 4.10). Since multivariate inequalities are uncommon in poset theory, we give a multivariate extension of the AD inequality, an important tool in the area. We then show that our multivariate extension is strong enough to also imply the above mentioned Lam-Pylyavskyy's monomial positivity.

Finally, we show that this multivariate approach can be used to prove new inequalities for general posets. Notably, we prove a new cross-product inequality (Theorem 10.1), and extend DDP and CPP log-concave inequalities for general posets (Theorem 9.3 and Corollary 9.5).

Paper structure. We start with a lengthy Section 2 with the background in both algebraic combinatorics and poset theory. We encourage the reader not to skip this section as we make some minor changes in definitions and standard notation to accommodate partly contradictory traditions in the two areas.

In the next two sections we present both known and new results in the order of increasing generality, pointing out the implications between results along the way. These implications tend to be quick and straightforward, and are included for clarity. In general, we opted for a complete
and detailed presentation of all corollaries and special cases as a way to fully explain connections between the results.

In a short Section 3, we present results only about linear extensions and standard Young tableaux. While the results are easy consequences of the $P$-partition results in Section 4, the idea is to make the linear extension's story completely self-contained. Our most general results (Theorems 4.9 and 4.10) are given at the end of Section 4.

We then proceed to the proofs. In Section 5, we give a self-contained simple proof of the generalized Fishburn's inequality (Theorem 3.4) deducing it from its order polynomial generalization (Theorem 4.8), which is proved via the AD inequality (Theorem 5.1). This proof is based on Fishburn's approach [Fis92], and is included here as a gentle introduction to our multivariate version.

In Section 6, we present the multivariate AD inequality (Theorem 6.1). This is the main tool of the paper, which we use to prove our main results in a short Section 7. In Section 8, we give a new proof of the Lam-Pylyavskyy inequality for Schur functions, also via the multivariate AD inequality.

In Section 9, we give a new proof and then a multivariate generalization (Theorem 9.3) of the DPP inequality. We follow this with the cross-product inequality for $P$-partitions (Theorem 10.1) in Section 10. We conclude with final remarks and open problems in Section 11.

## 2. Background, definitions and notation

2.1. Basic notations. We use $\mathbb{N}=\{0,1,2, \ldots\}, \mathbb{N}_{\geq 1}=\{1,2, \ldots\},[n]=\{1,2, \ldots, n\}$ and $\mathbb{R}_{+}=$ $\{x \geq 0\}$. To simplify the notation, for an element $a \in X$, we use $X-a$ to denote the subset $X \backslash\{a\}$. Similarly, for a subset $Y \subseteq X$, we write $X-Y$ in place of more general $X \backslash Y$.

For variables $\boldsymbol{q}=\left(q_{1}, \ldots, q_{n}\right)$ and a vector $\boldsymbol{a}=\left(a_{1}, \ldots, a_{n}\right) \in \mathbb{N}^{n}$, we write $\boldsymbol{q}^{\boldsymbol{a}}:=q_{1}^{a_{1}} \cdots q_{n}^{a_{n}}$. For a polynomial $F \in \mathbb{R}\left[z_{1}, \ldots, z_{n}\right]$, we write that $F \geq 0$ if $F\left(z_{1}, \ldots, z_{n}\right) \geq 0$ for all $\boldsymbol{z} \in \mathbb{R}^{n}$. For two polynomials $F, G \in \mathbb{R}\left[z_{1}, \ldots, z_{n}\right]$, we write $F \geq G$ if $F-G \geq 0$.

For polynomials $F, G \in \mathbb{R}[z]$, we write $F \geqslant_{z} G$ if $F-G \in \mathbb{R}_{+}[z]$ is a polynomial with nonnegative coefficients. For multivariate polynomials $F, G \in \mathbb{R}\left[z_{1}, \ldots, z_{n}\right]$, we define $F \geqslant_{z} G$ analogously. We drop the subscript in $\geqslant$ when the variables are clear. Obviously, $F \geqslant G$ implies $F \geq G$, but not vice versa, e.g. $x^{2}+y^{2} \geq 2 x y$ but $x^{2}+y^{2} \nsupseteq 2 x y$.
2.2. Posets. We refer to [Sta99, Ch. 3] and [Tro95] for standard definitions and notation. Let $\mathcal{P}=(X, \prec)$ be a partially ordered set on the ground set $X$ of size $|X|=n$, and with the partial order "々". A subposet is an induced poset $(Y, \prec)$ on the subset $Y \subseteq X$. For an element $x \subseteq X$, we denote by $\mathcal{P}-x$ the subposet of $\mathcal{P}$ on $X-x$.

For a poset $\mathcal{P}=(X, \prec)$, denote by $\mathcal{P}^{*}=\left(X, \prec^{*}\right)$ the dual poset with $x \prec^{*} y$ if and only if $y \prec x$, for all $x, y \in X$. For posets $\mathcal{P}=\left(X, \prec_{\mathcal{P}}\right)$ and $\mathcal{Q}=\left(Y, \prec_{\mathcal{Q}}\right)$, the parallel sum $\mathcal{P}+\mathcal{Q}=(Z, \prec)$ is the poset on the disjoint union $Z=X \sqcup Y$, where elements of $X$ retain the partial order of $\mathcal{P}$, elements of $Y$ retain the partial order of $\mathcal{Q}$, and elements $x \in X$ and $y \in Y$ are incomparable. Similarly, the linear sum $\mathcal{P} \oplus \mathcal{Q}=(Z, \prec)$, where $x \prec y$ for every two elements $x \in X$ and $y \in Y$ and other relations as in the parallel sum.

We use $\mathcal{C}_{n}$ and $\mathcal{A}_{n}$ to denote the $n$-element chain and antichain, respectively. Clearly, $\mathcal{C}_{n}=$ $\mathcal{C}_{1} \oplus \cdots \oplus \mathcal{C}_{1}$ ( $n$ times) and $\mathcal{A}_{n}=\mathcal{C}_{1}+\cdots+\mathcal{C}_{1}(n$ times $)$.

A lattice is a poset $\mathbb{L}=(\mathcal{L}, \prec)$ with meet $x \vee y$ (least upper bound) and join $x \wedge y$ (greatest lower bound) well defined, for all $x, y \in \mathcal{L}$. We also use ( $\mathcal{L}, \vee, \wedge$ ) to denote the lattice and the join and meet operations. The lattice $\mathbb{L}=(\mathcal{L}, \vee, \wedge)$ is distributive if it satisfied the distributive law: $x \wedge(y \vee z)=(x \wedge y) \vee(x \wedge z)$. Finally, for all $X, Y \subseteq \mathcal{L}$, we denote

$$
X \vee Y:=\{x \vee y: x \in X, y \in Y\} \quad \text { and } \quad X \wedge Y:=\{x \wedge y: x \in X, y \in Y\}
$$

2.3. Linear extensions and $P$-partitions. A linear extension of $\mathcal{P}$ is a bijection $L: X \rightarrow[n]$ that is order-preserving: $x \prec y$ implies $L(x)<L(y)$, for all $x, y \in X$. Denote by $\mathcal{E}(\mathcal{P})$ the set of linear extensions of $\mathcal{P}$, and let $e(\mathcal{P}):=|\mathcal{E}(\mathcal{P})|$ be the number of linear extensions. Observe that $e(\mathcal{P})=e\left(\mathcal{P}^{*}\right)$ and $e(\mathcal{P} \oplus \mathcal{Q})=e(\mathcal{P}) \cdot e(\mathcal{Q})$.

A subset $A \subseteq X$ is an upper ideal if $x \in A$ and $y \succ x$ implies $y \in A$. Similarly, a subset $A \subseteq X$ is a lower ideal if $x \in A$ and $y \prec x$ implies $y \in A$. We denote by $e(A)$ the number of linear extensions of the subposet $(A, \prec)$.

Let $\mathcal{P}=(X, \prec)$, where $X=\left\{x_{1}, \ldots, x_{n}\right\}$. We will always assume that $X$ has a natural labeling, i.e. $L: x_{i} \rightarrow i$ is a linear extension. A $\mathcal{P}$-partition is an order preserving map $A: X \rightarrow \mathbb{N}$, i.e. maps which satisfy $A(x) \leq A(y)$ for all $x \prec y$. Denote by $\operatorname{PP}(\mathcal{P})$ the set of $P$-partitions and let $\operatorname{PP}(\mathcal{P}, t)$ be the set of $P$-partitions with values at most $t .{ }^{2}$

Let $\Omega(\mathcal{P}, t):=|\operatorname{PP}(\mathcal{P}, t)|$ be the number of $\mathcal{P}$-partitions. This is the order polynomial corresponding to the poset $\mathcal{P} .{ }^{3}$ It is well-known and easy to see that

$$
\begin{equation*}
\Omega(\mathcal{P}, t) \sim \frac{e(\mathcal{P}) t^{n}}{n!} \quad \text { as } t \rightarrow \infty, \quad \text { where } \quad|X|=n \tag{2.1}
\end{equation*}
$$

Denote $|A|:=\sum_{x \in X} A(x)$ the sum of the entries in a $\mathcal{P}$-partition. Let

$$
\begin{equation*}
\Omega_{q}(\mathcal{P}, t):=\sum_{A \in \operatorname{PP}(\mathcal{P}, t)} q^{|A|} . \tag{2.2}
\end{equation*}
$$

Stanley showed, see [Sta99, Thm 3.15.7], that there is a statistics maj: $\mathcal{E}(\mathcal{P}) \rightarrow \mathbb{N}$, such that

$$
\begin{equation*}
\Omega_{q}(\mathcal{P}, \infty)=\frac{1}{(1-q)\left(1-q^{2}\right) \cdots\left(1-q^{n}\right)} \sum_{A \in \mathcal{E}(\mathcal{P})} q^{\operatorname{maj}(A)} \tag{2.3}
\end{equation*}
$$

More generally, let

$$
\begin{equation*}
\Omega_{q}(\mathcal{P}, t):=\sum_{A \in \operatorname{PP}(\mathcal{P}, t)} q_{1}^{A\left(x_{1}\right)} \cdots q_{n}^{A\left(x_{n}\right)} \tag{2.4}
\end{equation*}
$$

We call this GF the multivariate order polynomial. Note that Stanley gave a generalization of (2.3) for $\Omega_{\boldsymbol{q}}(\mathcal{P}, \infty)$ which we will not need, see [Sta99, Thm 3.15.5]. Finally, for $N \geq 0$, define

$$
\begin{equation*}
\mathrm{K}_{z}(\mathcal{P}, N):=\sum_{A \in \operatorname{PP}(\mathcal{P}, N)} z_{0}^{m_{0}(A)} \cdots z_{N}^{m_{N}(A)}, \tag{2.5}
\end{equation*}
$$

where $m_{i}(A):=\left|A^{-1}(i)\right|$ is the number of values $i$ in the $\mathcal{P}$-partition $A$.
2.4. Young diagrams and Young tableaux. We refer to [Mac95, Sag01] and [Sta99, Ch. 7] for standard definitions and notation. Let $\lambda=\left(\lambda_{1}, \ldots, \lambda_{\ell}\right)$ be an integer partition of $n$, write $\lambda \vdash n$, where $\lambda_{1} \geq \lambda_{2} \geq \ldots \geq \lambda_{\ell}>0$ and $\lambda_{1}+\ldots+\lambda_{\ell}=n$. Let $\ell(\lambda):=\ell$ denotes the number of parts. A conjugate partition $\lambda^{\prime}=\left(\lambda_{1}^{\prime}, \lambda_{2}^{\prime}, \ldots\right)$ is defined by $\lambda_{j}^{\prime}=\left|\left\{i: \lambda_{i} \geq j\right\}\right|$.

A Young diagram is the set of squares $\left\{(i, j) \in \mathbb{N}^{2}: 1 \leq j \leq \lambda_{i}, 1 \leq i \leq \ell\right\}$. In a mild abuse of notation, we use $\lambda$ to also denote the corresponding Young diagram, and refer to it as the straight shape. Let $\mu=\left(\mu_{1}, \mu_{2}, \ldots\right)$ be a partition such that $\mu_{i} \leq \lambda_{i}$ for all $0 \leq i \leq \ell$. The difference of Young diagrams is denoted by $\lambda / \mu$ and called the skew Young diagram of shape $\lambda / \mu$, or simply the skew shape $\lambda / \mu$. We use $|\lambda / \mu|$ for the size, i.e. the number of squares in $\lambda / \mu$.

A standard Young tableau of shape $\lambda / \mu$ is a bijection $A: \lambda / \mu \rightarrow[n]$ which increases in rows and columns: $A(i, j)<A(i+1, j)$ and $A(i, j)<A(i, j+1)$ whenever these are defined. Denote by $\operatorname{SYT}(\lambda / \mu)$ the set of standard Young tableaux of shape $\lambda / \mu$. We note that $|\operatorname{SYT}(\lambda)|$ can be

[^2]computed by the hook-length formula, see e.g. [Sta99, §7.21]. Similarly, the number $|\operatorname{SYT}(\lambda / \mu)|$ can be computed by the Aitken-Feit determinant formula, see e.g. [Sta99, §7.16].

Let poset $\mathcal{P}_{\lambda / \mu}=(\lambda / \mu, \prec)$ be defined by $(i, j) \preccurlyeq\left(i^{\prime}, j^{\prime}\right)$ if $i \leq i^{\prime}$ and $j \leq j^{\prime}$. For example, $P_{31 / 11} \simeq C_{2}$ and $\mathcal{P}_{321 / 21} \simeq A_{3}$. The set of linear extensions $\mathcal{E}\left(\mathcal{P}_{\lambda / \mu}\right)$ is in bijection with $\operatorname{SYT}(\lambda / \mu)$, so $e\left(\mathcal{P}_{\lambda / \mu}\right)=|\operatorname{SYT}(\lambda / \mu)|$.
2.5. Schur functions and reverse plane partitions. Let $A: \lambda / \mu \rightarrow \mathbb{N}$ be a function which increases in rows and columns. In this context, function $A$ is called a reverse plane partition. ${ }^{4}$ Let $\operatorname{RPP}(\lambda / \mu)$ denote the set of reverse plane partition of shape $\lambda / \mu$. We think of $A$ as a Young tableau with integers written in squares of $\lambda / \mu$. If $A \in \operatorname{RPP}(\lambda / \mu)$ is also increasing in columns and has all entries $\geq 1$, it is called a semistandard Young tableau. The set of such tableaux is denoted $\operatorname{SSYT}(\lambda / \mu)$. We use $\operatorname{RPP}(\lambda / \mu, t)$ and $\operatorname{SSYT}(\lambda / \mu, t)$ to denote reverse plane partitions and semistandard Young tableaux with entries $\leq t$.

Schur polynomial is a symmetric polynomial associated with the skew shape $\lambda / \mu$ and can be defined as

$$
\begin{equation*}
s_{\lambda / \mu}\left(z_{1}, \ldots, z_{N}\right)=\sum_{A \in \operatorname{SSYT}(\lambda / \mu, N)} z_{1}^{m_{1}(A)} \cdots z_{N}^{m_{N}(A)} \tag{2.6}
\end{equation*}
$$

where $m_{i}(A)=\left|A^{-1}(i)\right|$ is the number of $i$ 's in $A$. Schur functions are the stable limits of Schur polynomials as $n \rightarrow \infty$. They form a linear basis in the space of all symmetric functions.

For reverse plane partitions, observe the connection to the order polynomial:

$$
\begin{equation*}
\Omega(\lambda / \mu, t):=\Omega\left(\mathcal{P}_{\lambda / \mu}, t\right)=\sum_{A \in \operatorname{RPP}(\lambda / \mu, t)} t^{|A|} . \tag{2.7}
\end{equation*}
$$

In similar manner, consider the following multivariate GF for the reverse plane partitions:

$$
\mathcal{F}_{\lambda / \mu}\left(z_{0}, z_{1}, \ldots, z_{N}\right)=\sum_{A \in \operatorname{RPP}(\lambda / \mu, N)} z_{0}^{m_{0}(A)} z_{1}^{m_{1}(A)} \cdots z_{N}^{m_{N}(A)}
$$

Note the notation above, we have $\mathcal{F}_{\lambda / \mu}\left(z_{0}, z_{1}, \ldots, z_{N}\right)=\mathrm{K}_{z}\left(\mathcal{P}_{\lambda / \mu}, N\right)$.

## 3. Linear extensions

3.1. Fishburn's inequality. We start with the following fundamental inequality:

Theorem 3.1 (Fishburn's inequality [Fis84]). Let $\mathcal{P}=(X, \prec)$ be a finite poset, and let $A, B \subset X$ be lower ideals of $P$. Then:

$$
\begin{equation*}
\frac{e(A \cup B) \cdot e(A \cap B)}{e(A) \cdot e(B)} \geq \frac{|A \cup B|!\cdot|A \cap B|!}{|A|!\cdot|B|!} \tag{3.1}
\end{equation*}
$$

Using the notation

$$
f(P):=\frac{e(P)}{|X|!}
$$

Fishburn's inequality can be rewritten in a more concise form as a correlation inequality for probabilities:

$$
\begin{equation*}
f(A \cup B) \cdot f(A \cap B) \geq f(A) \cdot f(B) \tag{3.2}
\end{equation*}
$$

The original proof of Fishburn's inequality uses the $A D$ inequality. Note that it is tight for the antichain $\mathcal{P}=\mathcal{A}_{n}$.

[^3]3.2. Björner's inequality. For a skew Young diagram $|\lambda / \mu|=n$, we similarly denote
$$
f(\lambda / \mu):=f\left(\mathcal{P}_{\lambda / \mu}\right)=\frac{|\mathrm{SYT}(\lambda / \mu)|}{n!}
$$

Now (3.2) gives:
Corollary 3.2 (Björner's inequality [Bjö11, §6]). Let $\mu$ and $\nu$ be Young diagrams. Then:

$$
\begin{equation*}
f(\mu \vee \nu) \cdot f(\mu \wedge \nu) \geq f(\mu) \cdot f(\nu) \tag{3.3}
\end{equation*}
$$

where $\vee$ and $\wedge$ refer to the union and intersection of the Young diagrams.

Björner's proof used another Fishburn's result combined with the some calculations using the hook-length formula. The following result has an ambiguous status of being nominally new, yet it easily follows from the LP inequality (see $\S 4.1$ below).

Corollary 3.3 (generalized Björner's inequality). Let $\mu / \alpha$ and $\nu / \beta$ be skew Young diagrams. Then:

$$
\begin{equation*}
f(\mu / \alpha \vee \nu / \beta) \cdot f(\mu / \alpha \wedge \nu / \beta) \geq f(\mu / \alpha) \cdot f(\nu / \beta) \tag{3.4}
\end{equation*}
$$

where $\mu / \alpha \vee \nu / \beta:=(\mu \vee \nu) /(\alpha \vee \beta)$ and $\mu / \alpha \wedge \nu / \beta:=(\mu \wedge \nu) /(\alpha \wedge \beta)$.
In contrast with Björner's inequality, the generalized Björner inequality does not follow from Fishburn's inequality, at least not directly.
3.3. Generalized Fishburn's inequality. Our first new result is a common generalization of both the Fishburn's and the generalized Björner's inequalities.

Theorem 3.4. Let $\mathcal{P}=(X, \prec)$ be a finite poset. Let $A, B \subseteq X$ be lower ideals, and let $C, D \subseteq X$ be upper ideals of $\mathcal{P}$, such that $A \cap C=B \cap D=\varnothing$. Then:

$$
\begin{equation*}
f(X-V) \cdot f(X-W) \geq f(X-A-C) \cdot f(X-B-D) \tag{3.5}
\end{equation*}
$$

where $V:=(A \cap B) \cup(C \cup D)$ and $W:=(A \cup B) \cup(C \cap D)$.

Note that Fishburn's inequality (Theorem 3.1) is a special case $C=D=\varnothing$, and that Theorem 3.4 is self-dual. We prove the theorem using the AD inequality in Section 5.

Proof of [Theorem $3.4 \Longrightarrow$ Corollary 3.3]. Let $\mathcal{P}:=\mathcal{P}_{\lambda}$, where $\lambda:=\mu \vee \nu$. In the notation of Theorem 3.4, we have $X=\lambda$. Consider the following four subsets of the Young diagram $\lambda$ :

$$
\begin{equation*}
A:=\alpha, \quad B:=\beta, \quad C:=\lambda / \mu, \quad D:=\lambda / \nu \tag{3.6}
\end{equation*}
$$

Now observe that
$X-A-C=\mu / \alpha, \quad X-B-D=\nu / \beta, \quad X-V=(\mu \wedge \nu) /(\alpha \wedge \beta), \quad X-W=(\mu \vee \nu) /(\alpha \vee \beta)$.
Thus, (3.5) implies (3.4), as desired.

## 4. $P$-PARTITIONS

4.1. Schur functions. The following $L P$ inequality is the key result which inspired this paper.

Theorem 4.1 (Lam-Pylyavskyy inequality for Schur polynomials [LP07, Thm 4.5]). Let $\mu / \alpha$ and $\nu / \beta$ be skew Young diagrams, and let $\boldsymbol{z}=\left(z_{1}, \ldots, z_{N}\right)$, where $N \geq \ell(\mu), \ell(\nu)$. Then:

$$
\begin{equation*}
s_{\mu \vee \nu}(\boldsymbol{z}) \cdot s_{\mu \wedge \nu}(\boldsymbol{z}) \geqslant_{\boldsymbol{z}} s_{\mu}(\boldsymbol{z}) \cdot s_{\nu}(\boldsymbol{z}) . \tag{4.1}
\end{equation*}
$$

More generally, we have:

$$
\begin{equation*}
s_{\mu / \alpha \vee \nu / \beta}(\boldsymbol{z}) \cdot s_{\mu / \alpha \wedge \nu / \beta}(\boldsymbol{z}) \geqslant \geqslant_{z} s_{\mu / \alpha}(\boldsymbol{z}) \cdot s_{\nu / \beta}(\boldsymbol{z}), \tag{4.2}
\end{equation*}
$$

where $\mu / \alpha \vee \nu / \beta:=(\mu \vee \nu) /(\alpha \vee \beta)$ and $\mu / \alpha \wedge \nu / \beta:=(\mu \wedge \nu) /(\alpha \wedge \beta)$.
The original proof is completely combinatorial and uses an explicit injection. For completeness, we include a short argument showing how the LP inequality implies the Björner's and the generalized Björner's inequality.

Proof of $(4.2) \Longrightarrow$ (3.4). Recall the following analogue of (2.3) for skew Schur functions:

$$
\begin{equation*}
s_{\lambda / \tau}\left(1, q, q^{2}, \ldots\right)=\frac{1}{(1-q)\left(1-q^{2}\right) \cdots\left(1-q^{|\lambda / \tau|}\right)} \sum_{T \in \operatorname{SYT}(\lambda / \tau)} q^{\operatorname{maj}(T)}, \tag{4.3}
\end{equation*}
$$

where maj : $\operatorname{SYT}(\lambda / \tau) \rightarrow \mathbb{N}$ is the major index of a tableau, see e.g. [Sta99, Thm 7.19.11].
Let $n:=|\mu / \alpha|+|\nu / \beta|$. Substituting (4.3) into each of the four Schur functions in the LP inequality (4.2), multiplying both sides by $(1-q)^{n}$ and letting $q \rightarrow 1$, gives the generalized Björner's inequality (3.3).

Remark 4.2. The following truly remarkable Lam-Postnikov-Pylyavskyy inequality further extended (4.2) and resolved several open problems in the area:

$$
\begin{equation*}
s_{\mu / \alpha \vee \nu / \beta} \cdot s_{\mu / \alpha \wedge \nu / \beta} \geqslant \geqslant_{s} s_{\mu / \alpha} \cdot s_{\nu / \beta} . \tag{4.4}
\end{equation*}
$$

Here " $\geqslant_{s}$ " stands for Schur positivity, which is saying that the difference is a nonnegative sum of Schur functions. Although we will not need this extension, it does give a more conceptual proof of Björner's inequality.

In a different direction, Richards [Ric10] gave an analytic generalization of (4.1) for real $\lambda, \mu \in$ $\mathbb{R}^{\ell}$ and the determinant definition of Schur polynomials. It would be natural to conjecture that (4.4) also generalizes to this setting.

Proof of (4.4) $\Longrightarrow$ (3.3). Recall that for all $\mu \vdash k, \nu \vdash n-k$, we have:

$$
s_{\mu} \cdot s_{\nu}=\sum_{\lambda \vdash n} c_{\mu \nu}^{\lambda} s_{\lambda} \quad \text { and } \quad \chi^{\mu} \otimes \chi^{\nu} \uparrow_{S_{k} \times S_{n-k}}^{S_{n}}=\sum_{\lambda \vdash n} c_{\mu \nu}^{\lambda} \chi^{\lambda},
$$

where $c_{\mu \nu}^{\lambda}$ are the Littlewood-Richardson coefficients, see e.g. [Sag01, §4.9]. Equating dimensions in the second equality gives:

$$
f(\mu) \cdot f(\nu)=\sum_{\lambda \vdash n} c_{\mu \nu}^{\lambda} f(\lambda) .
$$

Thus $\varphi: s_{\lambda} \rightarrow f(\lambda)$ is a ring homomorphism from the ring of symmetric function to $\mathbb{Q}$ which maps Schur positive symmetric function to $\mathbb{Q}_{+}$. Applying $\varphi$ to the inequality (4.4) for $\alpha=\beta=\varnothing$ gives the desired inequality (3.3).
4.2. RPP variation. The following RPP variation is an easy corollary of the LP inequality (4.2):

Corollary 4.3. Let $\mu$ and $\nu$ be Young diagrams and let $t \geq 0$. Then:

$$
\begin{equation*}
\Omega(\mu \vee \nu, t) \cdot \Omega(\mu \wedge \nu, t) \geq \Omega(\mu, t) \cdot \Omega(\nu, t) \tag{4.5}
\end{equation*}
$$

Similarly, for the $q$-statistics we have:

$$
\begin{equation*}
\Omega_{q}(\mu \vee \nu, \infty) \cdot \Omega_{q}(\mu \wedge \nu, \infty) \geqslant_{q} \Omega_{q}(\mu, \infty) \cdot \Omega_{q}(\nu, \infty) \tag{4.6}
\end{equation*}
$$

More generally, we have:

$$
\begin{equation*}
\Omega_{q}(\mu \vee \nu, t) \cdot \Omega_{q}(\mu \wedge \nu, t) \geqslant_{q} \Omega_{q}(\mu, t) \cdot \Omega_{q}(\nu, t) . \tag{4.7}
\end{equation*}
$$

Proof of (4.1) $\Longrightarrow$ (4.6). Setting $N \leftarrow \infty$ and $\boldsymbol{z}=\left(z_{1}, z_{2}, \ldots\right) \leftarrow(q, q, \ldots)$, we get:

$$
\begin{equation*}
s_{\lambda}(q, q, \ldots)=\Omega_{q}(\lambda, \infty) \cdot q^{n(\lambda)}, \quad \text { where } \quad n(\lambda)=\sum_{(i, j) \in \lambda} i \tag{4.8}
\end{equation*}
$$

Note that $n(\mu \vee \nu)+n(\mu \wedge \nu)=n(\mu)+n(\nu)$. Substituting (4.8) into (4.1) and dividing both sides by $q^{n(\mu)+n(\nu)}$ gives (4.6).

Corollary 4.4. Let $\mathcal{P}=(X, \prec)$ be a finite poset, let $t \geq 0$, and let $A, B \subset X$ be lower ideals of $P$. Then:

$$
\begin{equation*}
\Omega(A \cup B, t) \cdot \Omega(A \cap B, t) \geq \Omega(A, t) \cdot \Omega(B, t) \tag{4.9}
\end{equation*}
$$

More generally, we have:

$$
\begin{equation*}
\Omega_{q}(A \cup B, t) \cdot \Omega_{q}(A \cap B, t) \geqslant_{q} \Omega_{q}(A, t) \cdot \Omega_{q}(B, t) \tag{4.10}
\end{equation*}
$$

Proof of (4.9) $\Longrightarrow$ (3.2). Let $t \rightarrow \infty$ and apply (2.1) to each term in (4.9).
Corollary 4.4 is a direct generalization of Corollary 4.3 , which follows by taking $A \leftarrow \mu$ and $B \leftarrow \nu$. Our next result is a multivariate generalization of Corollary 4.3.

Corollary 4.5. Let $\mu$ and $\nu$ be Young diagrams and let $N \geq 0$. Then:

$$
\begin{equation*}
\mathcal{F}_{\mu \vee \nu}(z) \cdot \mathcal{F}_{\mu \wedge \nu}(z) \geqslant_{z} \mathcal{F}_{\mu}(z) \cdot \mathcal{F}_{\nu}(z), \tag{4.11}
\end{equation*}
$$

where $\boldsymbol{z}=\left(z_{0}, z_{1}, \ldots, z_{N}\right)$.
Proof of (4.11) $\Longrightarrow$ (4.7). Let $N \leftarrow t$, and set $z_{i} \leftarrow q^{i}$ for all $0 \leq i \leq N$.
This result is implicit in [LP07] and follows from the following general theorem:
Theorem 4.6 (Lam-Pylyavskyy inequality for multivariate order polynomials [LP07, Prop. 3.7]). Let $\mathcal{P}=(X, \prec)$ be a finite poset, let $A, B \subset X$ be lower ideals of $P$, and let $N \geq 0$. Then:

$$
\begin{equation*}
\mathrm{K}_{z}(A \cup B, N) \cdot \mathrm{K}_{z}(A \cap B, N) \geqslant_{z} \mathrm{~K}_{z}(A, N) \cdot \mathrm{K}_{z}(B, N) \tag{4.12}
\end{equation*}
$$

This is the most general version of the LP inequality that we discuss in this paper. Note that (4.12) $\Longrightarrow$ (4.11) by taking $A \leftarrow \mu$ and $B \leftarrow \nu$.

Remark 4.7. As we mention in the introduction, the ultimate Lam-Pylyavskyy generalization uses the meet and join operations which are incompatible with those we employ in this paper. They are in fact, noncommutative and designed to allow the "cell transfer" direct injection.

Notably, (4.2) does not follow from (4.12), but from the proof of this ultimate Lam-Pylyavskyy generalization which happens to apply to skew shapes. We give a more streamlined derivation of (4.2) from our generalization below.
4.3. Main results. We begin with the order polynomial extension of the generalized Fishburn's inequality (Theorem 3.4) and the Lam-Pylyavskyy order polynomial inequality (Corollary 4.4).

Theorem 4.8. Let $\mathcal{P}=(X, \prec)$ be a finite poset. Let $A, B \subseteq X$ be lower ideals, and let $C, D \subseteq X$ be upper ideals of $\mathcal{P}$, such that $A \cap C=B \cap D=\varnothing$. Then:

$$
\begin{equation*}
\Omega(X-V, t) \cdot \Omega(X-W, t) \geq \Omega(X-A-C, t) \cdot \Omega(X-B-D, t), \tag{4.13}
\end{equation*}
$$

where $V:=(A \cap B) \cup(C \cup D)$ and $W:=(A \cup B) \cup(C \cap D)$. More generally, we have:

$$
\begin{equation*}
\Omega_{q}(X-V, t) \cdot \Omega_{q}(X-W, t) \geqslant_{q} \Omega_{q}(X-A-C, t) \cdot \Omega_{q}(X-B-D, t) . \tag{4.14}
\end{equation*}
$$

Corollary 4.4 is a special case of the theorem when $C=D=\varnothing$.
Proof of (4.13) $\Longrightarrow$ (3.5). Let $t \rightarrow \infty$ and apply (2.1) to each term in (4.13).
Here is our most general result in this direction, and the ultimate multivariate generalization of Fishburn's inequality (Theorem 3.1).

Theorem 4.9. Let $\mathcal{P}=(X, \prec)$ be a finite poset. Let $A, B \subseteq X$ be lower ideals, and let $C, D \subseteq X$ be upper ideals of $\mathcal{P}$, such that $A \cap C=B \cap D=\varnothing$. Then:

$$
\begin{equation*}
\Omega_{q}(X-V, t) \cdot \Omega_{q}(X-W, t) \geqslant_{q} \Omega_{q}(X-A-C, t) \cdot \Omega_{q}(X-B-D, t), \tag{4.15}
\end{equation*}
$$

where $V:=(A \cap B) \cup(C \cup D)$ and $W:=(A \cup B) \cup(C \cap D)$.

Proof of (4.15) $\Longrightarrow$ (4.14). Take $\boldsymbol{q} \leftarrow(q, \ldots, q)$.
Finally, we present another generalization of Theorem 4.8 for different choices of rank functions, and furthermore generalizes Lam-Pylyavskyy Theorem 4.6. We prove both theorems in Section 7.

Theorem 4.10. Let $\mathcal{P}=(X, \prec)$ be a finite poset. Let $A, B \subseteq X$ be lower ideals, and let $C, D \subseteq X$ be upper ideals of $\mathcal{P}$, such that $A \cap C=B \cap D=\varnothing$. Then:

$$
\begin{equation*}
\mathrm{K}_{z}(X-V, N) \cdot \mathrm{K}_{z}(X-W, N) \geqslant_{z} \mathrm{~K}_{z}(X-A-C, N) \cdot \mathrm{K}_{z}(X-B-D, N) \tag{4.16}
\end{equation*}
$$

where $V:=(A \cap B) \cup(C \cup D)$ and $W:=(A \cup B) \cup(C \cap D)$.
Proof of (4.16) $\Longrightarrow$ (4.14). Take $\boldsymbol{z} \leftarrow\left(1, q, q^{2}, \ldots, q^{N}\right)$.
In particular, these two theorems imply the following corollary for skew Young diagrams.
Corollary 4.11. Let $\mu / \alpha$ and $\nu / \beta$ be skew Young diagrams. Then:

$$
\begin{equation*}
\Omega_{\boldsymbol{q}}(\mu / \alpha \vee \nu / \beta, t) \cdot \Omega_{\boldsymbol{q}}(\mu / \alpha \wedge \nu / \beta, t) \geqslant_{\boldsymbol{q}} \Omega_{\boldsymbol{q}}(\mu / \alpha, t) \cdot \Omega_{\boldsymbol{q}}(\nu / \beta, t), \tag{4.17}
\end{equation*}
$$

and

$$
\begin{equation*}
\mathcal{F}_{\mu / \alpha \vee \nu / \beta}(\boldsymbol{z}) \cdot \mathcal{F}_{\mu / \alpha \wedge \nu / \beta}(\boldsymbol{z}) \geqslant_{z} \mathcal{F}_{\mu / \alpha}(\boldsymbol{z}) \cdot \mathcal{F}_{\nu / \beta}(\boldsymbol{z}), \tag{4.18}
\end{equation*}
$$

where $\boldsymbol{z}=\left(z_{0}, z_{1}, \ldots, z_{N}\right)$.
Proof. Let $\mathcal{P}, A, B, C, D$ be as in (3.6). By applying the same argument as in the proof of the [Theorem $3.4 \Longrightarrow$ Corollary 3.3] implication, the inequality (4.17) now follows from (4.15), while the inequality (4.18) follows from (4.16).

Remark 4.12. Although the inequalities (4.18) and (4.17) do not appear in [LP07], they follow from the approach in that paper.

## 5. The Ahlswede-Daykin inequality

In this section, we prove the first part of Theorem 4.8 by using the Ahlswede-Daykin (AD) inequality. Our approach is based on the proof in [Fis84]. For every $\rho: Z \rightarrow \mathbb{R}_{+}$and every $X \subseteq Z$, denote

$$
\begin{equation*}
\rho(X):=\sum_{x \in X} \rho(x) . \tag{5.1}
\end{equation*}
$$

Theorem 5.1 (Ahlswede-Daykin inequality $[\mathrm{AD} 78])$. Let $\mathbb{L}=(\mathcal{L}, \vee, \wedge)$ be a finite distributive lattice, and let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be nonnegative functions on $\mathcal{L}$. Suppose we have:

$$
\begin{equation*}
\alpha(x) \cdot \beta(y) \leq \gamma(x \vee y) \cdot \delta(x \wedge y) \quad \text { for every } \quad x, y \in \mathcal{L} . \tag{5.2}
\end{equation*}
$$

Then:

$$
\begin{equation*}
\alpha(X) \cdot \beta(Y) \leq \gamma(X \vee Y) \cdot \delta(X \wedge Y) \quad \text { for every } \quad X, Y \subseteq \mathcal{L} . \tag{5.3}
\end{equation*}
$$

Proof of the first part of Theorem 4.8. Let $\mathcal{P}=(X, \prec)$ be a poset, and let $t \geq 0$. We denote by $\mathbb{L}(\mathcal{P}, t)=(\mathcal{L}, \vee, \wedge)$ the distributive lattice on the set $\mathcal{L} \subseteq\{0, \ldots, t\}^{X}$ given by

$$
\begin{equation*}
\mathcal{L}:=\operatorname{PP}(\mathcal{P}, t)=\{T: X \rightarrow\{0, \ldots, t\}: T(x) \leq T(y) \quad \text { for all } x, y \in X \text { s.t. } x \prec y\}, \tag{5.4}
\end{equation*}
$$

with the join and meet operation given by

$$
[S \vee T](x)=\max \{S(x), T(x)\} \quad \text { and } \quad[S \wedge T](x)=\min \{S(x), T(x)\} \quad \text { for every } x \in X
$$

Recall that $\Omega(\mathcal{P}, t)=|\mathcal{L}|$. Let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be given by

$$
\begin{align*}
\alpha(T) & :=\mathbb{1}\{T(x)=0 \text { for all } x \in A, T(y)=t \text { for all } y \in C\}, \\
\beta(T) & :=\mathbb{1}\{T(x)=0 \text { for all } x \in B, T(y)=t \text { for all } y \in D\}, \\
\gamma(T) & :=\mathbb{1}\{T(x)=0 \text { for all } x \in A \cap B, T(y)=t \text { for all } y \in C \cup D\},  \tag{5.5}\\
\delta(T) & :=\mathbb{1}\{T(x)=0 \text { for all } x \in A \cup B, T(y)=t \text { for all } y \in C \cap D\} .
\end{align*}
$$

Note that

$$
\begin{aligned}
& \alpha(\mathcal{L})=\Omega(X-A-C, t), \quad \beta(\mathcal{L})=\Omega(X-B-D, t), \\
& \gamma(\mathcal{L})=\Omega(X-V, t), \quad \delta(\mathcal{L})=\Omega(X-W, t) .
\end{aligned}
$$

By the AD inequality (5.3), it thus suffices to verify (5.2), which in this case states:

$$
\begin{equation*}
\alpha(S) \cdot \beta(T) \leq \gamma(S \vee T) \cdot \delta(S \wedge T) \quad \text { for every } \quad S, T \in \mathcal{L} . \tag{5.6}
\end{equation*}
$$

Let $S, T \in \mathcal{L}$ be such that $\alpha(S)=\beta(T)=1$. Then:

$$
S(x)=0 \text { for } x \in A, \quad S(y)=t \text { for } y \in C, \quad T(x)=0 \text { for } x \in B, \quad T(y)=t \text { for } y \in D
$$

This gives:

$$
\begin{array}{ll}
\max \{S(x), T(x)\}=0 \text { for } x \in A \cap B, & \max \{S(y), T(y)\}=t \text { for } x \in C \cup D, \\
\min \{S(x), T(x)\}=0 \text { for } x \in A \cup B, & \min \{S(y), T(y)\}=t \text { for } x \in C \cap D .
\end{array}
$$

The first equation implies $\gamma(S \vee T)=1$, while the second equation implies $\delta(S \wedge T)=1$. This implies (5.6) and completes the proof of (4.13).

Remark 5.2. For the second (more general) part of Theorem 4.8, one can use the same approach with the AD inequality in Theorem 5.1 replaced with $q$-AD inequality by Christofides [Chr09]. Our proof of Theorem 4.9 given below, extends Theorem 4.8 using the multivariate $\boldsymbol{q}$-AD inequality.

## 6. Multivariate AD inequality

6.1. The statement. Let $\mathbb{L}:=(\mathcal{L}, \wedge, \vee)$ be a finite distributive lattice. Throughout this section, fix variables $q_{1}, \ldots, q_{\ell}$, and modular functions $r_{1}, \ldots, r_{\ell}: \mathcal{L} \rightarrow \mathbb{N}$ defined to satisfy

$$
r_{i}(x)+r_{i}(y)=r_{i}(x \vee y)+r_{i}(x \wedge y) \quad \text { for all } \quad x, y \in \mathcal{L} \text { and } \quad 1 \leq i \leq \ell .
$$

Write $\boldsymbol{q}:=\left(q_{1}, \ldots, q_{\ell}\right)$ and $\boldsymbol{r}:=\left(r_{1}, \ldots, r_{\ell}\right)$. For $x \in \mathcal{L}$, write

$$
\boldsymbol{r}(x):=\left(r_{1}(x), \ldots, r_{\ell}(x)\right) \quad \text { and } \quad \boldsymbol{q}^{r(x)}:=q_{1}^{r_{1}(x)} \ldots q_{\ell}^{r_{\ell}(x)} .
$$

For a function $\rho: \mathcal{L} \rightarrow \mathbb{R}_{+}$and subset $X \subseteq \mathcal{L}$, define

$$
\begin{equation*}
\rho_{\langle\boldsymbol{q}, r\rangle}(X):=\sum_{x \in X} \rho(x) \boldsymbol{q}^{r(x)} \in \mathbb{R}_{+}\left[q_{1}, \ldots, q_{\ell}\right] . \tag{6.1}
\end{equation*}
$$

Note that (6.1) is a multivariate $\boldsymbol{q}$-analogue of (5.1). We can now state the multivariate $\boldsymbol{q}$-analogue of the Ahlswede-Daykin inequality (Theorem 5.1).

Theorem 6.1 (multivariate $A D$ inequality). Let $\mathbb{L}=(\mathcal{L}, \wedge, \vee)$ be a finite distributive lattice, and let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be nonnegative functions on $\mathcal{L}$. Suppose we have

$$
\begin{equation*}
\alpha(x) \cdot \beta(y) \leq \gamma(x \vee y) \cdot \delta(x \wedge y) \quad \text { for every } \quad x, y \in \mathcal{L} . \tag{6.2}
\end{equation*}
$$

Then:

$$
\begin{equation*}
\alpha_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(X) \cdot \beta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(Y) \leqslant_{\boldsymbol{q}} \gamma_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(X \vee Y) \cdot \delta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(X \wedge Y) \quad \text { for every } \quad X, Y \subseteq \mathcal{L} . \tag{6.3}
\end{equation*}
$$

Our proof is strongly inspired by those of Björner [Bjö11] and Christofides [Chr09]. We closely follow the presentation from the former while incorporating some ideas from the latter paper.
6.2. The proof. We start by proving the following special case of Theorem 6.1 , which we use to obtain the theorem in the full generality.

Proposition 6.2. Let $\mathbb{L}=(\mathcal{L}, \wedge, \vee), \alpha, \beta, \gamma, \delta$ be as in Theorem 6.1. Then:

$$
\begin{equation*}
\alpha_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) \beta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) \leqslant_{q} \gamma_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) \delta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) . \tag{6.4}
\end{equation*}
$$

Proof of Proposition 6.2 $\Longrightarrow$ Theorem 6.1. Let $\alpha^{\prime}, \beta^{\prime}, \gamma^{\prime}, \delta^{\prime}: \mathcal{L} \rightarrow \mathbb{R}_{+}$be functions given by

$$
\alpha^{\prime}:=\alpha \circ \mathbf{1}_{X}, \quad \beta^{\prime}:=\beta \circ \mathbf{1}_{Y}, \quad \gamma^{\prime}:=\gamma \circ \mathbf{1}_{X \vee Y}, \quad \delta:=\delta^{\prime} \circ \mathbf{1}_{X \wedge Y} .
$$

Note that

$$
\begin{equation*}
\alpha^{\prime}(x) \cdot \beta^{\prime}(y) \leq \gamma^{\prime}(x \vee y) \cdot \delta^{\prime}(x \wedge y) \quad \text { for every } \quad x, y \in \mathcal{L} . \tag{6.5}
\end{equation*}
$$

Indeed, the LHS of (6.5) is equal to 0 if $x \notin A$ or $y \notin B$, so suppose that $x \in A, y \in B$. Then the inequality reduces to (6.2), which is part of the assumption. The inequality (6.3) then follows from (6.4) by noting that

$$
\begin{aligned}
& \alpha_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}^{\prime}(\mathcal{L})=\alpha_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(X), \quad \beta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}^{\prime}(\mathcal{L})=\beta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(Y), \\
& \gamma_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}^{\prime}(\mathcal{L})=\gamma_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(X \vee Y), \quad \delta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}^{\prime}(\mathcal{L})=\delta_{\langle\boldsymbol{q}, r\rangle}(X \wedge Y),
\end{aligned}
$$

as desired.

Proof of Proposition 6.2. Let

$$
\Phi(\boldsymbol{q}, \boldsymbol{r}):=\alpha_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L}) \cdot \beta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L})-\gamma_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L}) \cdot \delta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L}) .
$$

For $x, y \in \mathcal{L}$, we also define

$$
\phi(x, y):=\alpha(x) \cdot \beta(y)-\gamma(x) \cdot \delta(y) .
$$

A simple computation shows that

$$
\Phi(\boldsymbol{q}, \boldsymbol{r})=\sum_{(x, y) \in \mathcal{L}^{2}} \phi(x, y) \boldsymbol{q}^{\boldsymbol{r}(x)+\boldsymbol{r}(y)} .
$$

Let $\mathbf{d}:=\left(d_{1}, \ldots, d_{\ell}\right) \in \mathbb{N}^{\ell}$ be an arbitrary integer vector. Denote by

$$
\Phi_{\mathbf{d}}:=\left[q_{1}^{d_{1}} \cdots q_{\ell}^{d_{\ell}}\right] \Phi(q)
$$

the coefficient of the monomial $\boldsymbol{q}^{\mathbf{d}}$ in $\Phi(q)$. We then have:

$$
\Phi_{\mathbf{d}}=\sum_{\substack{(x, y) \in \mathcal{L}^{2}, r(x)+r(y)=\mathbf{d}}} \phi(x, y) .
$$

We now consider another, slightly coarser, grouping of terms. For $u, v \in \mathcal{L}$ satisfying $u \prec^{\diamond} v$, so in particular $u \neq v$, let $C(u, v)$ denote the set of (ordered) pairs ( $x, y$ ) in the interval $[u, v]$ such that $x \wedge y=v$ and $x \vee y=u$. Let

$$
\psi(u, v):=\sum_{(x, y) \in C(u, v)} \phi(x, y) .
$$

It follows from the modularity of $r_{1}, \ldots, r_{\ell}$ that

$$
\Phi_{\mathbf{d}}=\sum_{\substack{u \prec{ }^{\circ} v, r(u)+r(v)=\mathbf{d}}} \psi(u, v)+\sum_{\substack{u \in \mathcal{L}, 2 r(u)=\mathrm{d}}} \phi(u, u) .
$$

Since $\phi(u, u)=\alpha(u) \beta(u)-\gamma(u) \delta(u) \leq 0$ by (6.2), the proposition follows from Claim 6.3 below.

Claim 6.3. In notation above, for every $u, v \in \mathcal{L}$ such that $u \prec{ }^{\diamond} v$, we have $\psi(u, v) \leq 0$.
Proof of Claim 6.3. Note that $\psi(u, v)$ depends only on elements in the poset interval $[u, v]$, so by restricting to $[u, v]$ if necessary, we can without loss of generality assume that $u=\widehat{0}$ is the unique minimal element of $\mathcal{L}$, and $v=\widehat{1}$ is the unique maximal element of $\mathcal{L}$.

For $x \in \mathcal{L}$, a complement of $x$ is an element $y \in \mathcal{L}$ such that $x \wedge y=\widehat{0}$ and $x \vee y=\widehat{1}$. Note that in a finite distributive lattice every element has at most one complement (see e.g. [Bir67, Thm 10, p. 12]), and we denote this element by $x^{c}$ if it exists. Note that $\psi(\widehat{0}, \widehat{1})$ depends only on elements that have a complement in $\mathcal{L}$, and that the set of complemented elements in a finite distributive lattice form a sublattice of $\mathcal{L}$ (see e.g. [Bir67, p. 18]). By restricting to this sublattice if necessary, without loss of generality we can assume that every element $x \in \mathcal{L}$ has a unique complement $x^{c}$ (i.e., when $\mathcal{L}$ is a Boolean lattice).

Define four new functions $\alpha^{\prime}, \beta^{\prime}, \gamma^{\prime}, \delta^{\prime}: \mathcal{L} \rightarrow \mathbb{R}_{+}$as follows:

$$
\alpha^{\prime}(x):=\alpha(x) \beta\left(x^{c}\right), \quad \beta^{\prime}(x):=\alpha\left(x^{c}\right) \beta(x), \quad \gamma^{\prime}(x):=\gamma(x) \delta\left(x^{c}\right), \quad \delta^{\prime}(x):=\gamma\left(x^{c}\right) \delta(x)
$$

Note that

$$
\psi(\widehat{0}, \widehat{1})=\sum_{x \in \mathcal{L}} \phi\left(x, x^{c}\right)=\sum_{x \in \mathcal{L}} \alpha(x) \beta\left(x^{c}\right)-\gamma(x) \delta\left(x^{c}\right)=\alpha^{\prime}(\mathcal{L})-\gamma^{\prime}(\mathcal{L})
$$

It thus suffices to show that $\alpha^{\prime}(\mathcal{L}) \leq \gamma^{\prime}(\mathcal{L})$. Now observe that, for any $x, y \in \mathcal{L}$, we have:

$$
\begin{aligned}
\alpha^{\prime}(x) \beta^{\prime}(y) & =(\alpha(x) \beta(y))\left(\alpha\left(y^{c}\right) \beta\left(x^{c}\right)\right) \leq{ }_{(6.2)} \quad(\gamma(x \vee y) \delta(x \wedge y))\left(\gamma\left(y^{c} \vee x^{c}\right) \delta\left(y^{c} \wedge x^{c}\right)\right) \\
& \leq \gamma(x \vee y) \delta\left((y \vee x)^{c}\right) \gamma\left((y \wedge x)^{c}\right) \delta(x \wedge y)=\gamma^{\prime}(x \vee y) \delta^{\prime}(x \wedge y) .
\end{aligned}
$$

It then follows from the (usual) AD inequality (5.3), that

$$
\begin{equation*}
\alpha^{\prime}(\mathcal{L}) \beta^{\prime}(\mathcal{L}) \leq \gamma^{\prime}(\mathcal{L}) \delta^{\prime}(\mathcal{L}) . \tag{6.6}
\end{equation*}
$$

On the other hand, note that $\beta^{\prime}(\mathcal{L})=\alpha^{\prime}(\mathcal{L})$ and $\gamma^{\prime}(\mathcal{L})=\delta^{\prime}(\mathcal{L})$ by definition of the functions. Since the functions are nonnegative, (6.6) gives $\alpha^{\prime}(\mathcal{L}) \leq \gamma^{\prime}(\mathcal{L})$. This completes the proof.

## 7. Proof of main results

7.1. Proof of Theorem 4.9. Let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be as in (5.5). Note that these functions satisfy the assumption (5.6) of the multivariate AD inequality.

Let $\boldsymbol{q}:=\left(q_{1}, \ldots, q_{n}\right)$ be variables, with $n=|X|$. For any $i \in[n]$, let $r_{i}: \mathcal{L} \rightarrow \mathbb{R}_{+}$be the modular function given by $r_{i}(T):=T\left(x_{i}\right)$. For a subset $Y \subseteq X$, denote

$$
\boldsymbol{q}^{\mathbf{n}(Y)}:=\prod_{x_{i} \in Y}\left(q_{i}\right)^{t} .
$$

Then:

$$
\begin{aligned}
\alpha_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\Omega_{\boldsymbol{q}}(X-A-C, t) \cdot \boldsymbol{q}^{\mathbf{n}(C)}, \\
\beta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\Omega_{\boldsymbol{q}}(X-B-D, t) \cdot \boldsymbol{q}^{\mathbf{n}(D)}, \\
\gamma_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\Omega_{\boldsymbol{q}}(X-V, t) \cdot \boldsymbol{q}^{\mathbf{n}(C \cup D)}, \\
\delta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\Omega_{\boldsymbol{q}}(X-W, t) \cdot \boldsymbol{q}^{\mathbf{n}(C \cap D)} .
\end{aligned}
$$

The theorem now follows from the multivariate AD inequality (6.3).
7.2. Proof of Theorem 4.10. Let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be as in (5.5), with $t \leftarrow N$. Note that these functions satisfy the assumption of the multivariate AD inequality (see (5.6)). Let $\boldsymbol{q}:=\left(q_{0}, \ldots, q_{N}\right)$ be variables. For any $i \in\{0, \ldots, N\}$, let $r_{i}: \mathcal{L} \rightarrow \mathbb{R}_{+}$be the modular function where $r_{i}(T):=|\{x \in X: T(x)=i\}|$ is the number of $i$ 's in $T$. Then

$$
\begin{aligned}
\alpha_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\mathrm{K}_{z}(X-A-C, N) \cdot q_{0}^{|A|} q_{N}^{|C|}, \\
\beta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\mathrm{K}_{z}(X-B-D, M) \cdot q_{0}^{|B|} q_{N}^{|D|}, \\
\gamma_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\mathrm{K}_{z}(X-V, N) \cdot q_{0}^{|A \cap B|} q_{N}^{|C \cup D|}, \\
\delta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L}) & =\mathrm{K}_{z}(X-W, M) \cdot q_{0}^{|A \cup B|} q_{N}^{|C \cap D|} .
\end{aligned}
$$

The theorem now follows from the multivariate AD inequality (6.3).

## 8. Back to Schur polynomials

In this section we give a new proof of the Lam-Pylyavskyy inequality (4.2) for Schur polynomials via the multivariate AD inequality.

Proof of Theorem 4.1. Let $\mathcal{P}:=\mathcal{P}_{\lambda}$ be the poset of the Young diagram of shape $\lambda$, where $\lambda:=\mu \vee \nu$. Let $\mathbb{L}:=\left(\mathcal{L}^{\prime}, \wedge^{\prime}, \vee^{\prime}\right)$ be the distributive lattice given by $\mathcal{L}^{\prime}:=\operatorname{RPP}(\lambda, N)$, with the $\vee^{\prime}$ and $\wedge^{\prime}$ operation given by

$$
\left(S \vee^{\prime} T\right)(i, j):=\max \{S(i, j), T(i, j)\}, \quad\left(S \wedge^{\prime} T\right)(i, j):=\min \{S(i, j), T(i, j)\}
$$

For a skew Young diagram $\pi / \tau$ such that $\pi \subset \lambda$, let $\phi^{\pi / \tau}: \mathcal{L}^{\prime} \rightarrow \mathbb{R}_{+}$be the characteristic function of the reverse plane partition $T \in \operatorname{RPP}(\lambda, N)$ satisfying all these properties:

$$
\begin{aligned}
& T(i . j) \geq 1 \quad \text { for }(i, j) \in \lambda, \\
& T(i, j)=1 \quad \text { for }(i, j) \in \tau \quad \text { and } \quad T(i, j)=N \quad \text { for }(i, j) \in \lambda / \pi, \\
& T(i, j)<T(i+1, j) \quad \text { if }(i, j),(i+1, j) \in \pi / \tau .
\end{aligned}
$$

Note that these reverse plane partitions are in bijection with semistandard Young tableau of $\pi / \tau$ in $\operatorname{SSYT}(\pi / \tau, N)$.

We define functions $\zeta, \eta, \xi, \rho: \mathcal{L} \rightarrow \mathbb{R}_{+}$as follows:

$$
\zeta:=\phi^{\mu / \alpha}, \quad \eta:=\phi^{\nu / \beta}, \quad \xi:=\phi^{\mu / \alpha \wedge \nu / \beta}, \quad \rho:=\phi^{\mu / \alpha \vee \nu / \beta} .
$$

We now show that these functions satisfy the assumption of the multivariate AD inequality, i.e. for any $S, T \in \mathcal{L}$ :

$$
\zeta(S) \cdot \eta(T) \leq \xi(S \vee T) \cdot \rho(S \wedge T)
$$

The equation is vacuously true if $\zeta(S)=0$ or $\eta(T)=0$, so assume $\zeta(S)=\eta(T)=1$. We show only the proof that $\xi(S \vee T)=1$, as the proof of $\rho(S \wedge T)=1$ is similar. First, for $(i, j) \in \lambda$, we have:

$$
[S \vee T](i, j)=\max \{S(i, j), T(i, j)\} \geq 1
$$

Second, for $(i, j) \in \alpha \wedge \beta$,

$$
[S \vee T](i, j)=\max \{S(i, j), T(i, j)\}=1,
$$

Third, for $(i, j) \in \lambda /(\mu \wedge \nu)$,

$$
[S \vee T](i, j)=\max \{S(i, j), T(i, j)\}=N
$$

Fourth, let $(i, j),(i+1, j) \in(\mu \wedge \nu) /(\alpha \wedge \beta)$. We will need to show that

$$
\begin{equation*}
[S \vee T](i, j)<[S \vee T](i+1, j) \tag{8.1}
\end{equation*}
$$

Note that we must have either $(i, j) \in(\mu \wedge \nu) / \alpha$ or $(i, j) \in(\mu \wedge \nu) / \beta$. Without loss of generality, we assume the former holds. Then it follows that $(i+1, j) \in(\mu \wedge \nu) / \alpha$. Since $\zeta(S)=1$, this implies that

$$
S(i, j)<S(i+1, j) \leq \max \{S(i+1, j), T(i+1, j)\}=[S \vee T](i+1, j)
$$

Thus (8.1) follows if $T(i, j) \leq S(i, j)$, so suppose instead that $T(i, j)>S(i, j)$. This then implies $T(i, j)>1$. Since $\eta(T)=1$, this implies that $(i, j) \in(\mu \wedge \nu) / \beta$, which in turn implies that $(i+1, j) \in(\mu \wedge \nu) / \beta$. Thus we have:

$$
[S \vee T](i, j)=T(i, j)<T(i+1, j) \leq[S \vee T](i+1, j),
$$

which completes the proof of (8.1).
Let $\boldsymbol{z}:=\left(z_{1}, \ldots, z_{N}\right)$ be variables, and let $r_{i}: \mathcal{L} \rightarrow \mathbb{N}, i \in[N]$, be the modular function defined as follows: $r_{i}(T):=m_{i}(T)$ is the number of $i$ 's in $T$. It then follows that

$$
\begin{array}{ll}
A_{\langle z, r\rangle}=s_{\mu / \alpha} \cdot q_{1}^{|\alpha|} q_{N}^{|\lambda|-|\mu|}, & B_{\langle z, r\rangle}=s_{\nu / \beta} \cdot q_{1}^{|\beta|} q_{N}^{|\lambda|-|\nu|}, \\
C_{\langle z, r\rangle}=s_{\mu / \alpha \wedge \nu / \beta} \cdot q_{1}^{|\alpha \wedge \beta|} q_{N}^{|\lambda|-|\mu \wedge \nu|}, & D_{\langle z, r\rangle}=s_{\mu / \alpha \wedge \nu / \beta} \cdot q_{1}^{|\alpha \vee \beta|} q_{N}^{|\lambda|-|\mu \vee \nu|} .
\end{array}
$$

The theorem now follows from the multivariate AD inequality (6.3).
Remark 8.1. By the arguments analogous to the proofs in this and previous section, specifically the proof of (8.1) to account for strict comparisons, the multivariate AD inequality can be used to prove results analogous to Theorem 4.8 and Theorem 4.10 for both strict and non-strict ( $\mathcal{P}, \omega$ )partitions (see definitions in [Sta99, §3.15.1]). Similarly, we can extend out results to the more general $\mathbb{T}$-labelled $(\mathcal{P}, O)$ tableaux defined in [LP07]. We omit the details for brevity.

## 9. Multivariate Daykin-Daykin-Paterson inequality

9.1. The DDP inequality. Let $\mathcal{P}=(X, \prec)$ be a partially ordered set on $|X|=n$ elements. Fix $t \geq 0$ and an element $z \in X$. For integer $0 \leq k \leq t$, denote by $\operatorname{PP}(\mathcal{P}, t ; z, k)$ the set of $\mathcal{P}$-partitions $A \in \operatorname{PP}(\mathcal{P}, t)$ such that $A(z)=k$. Let $\Omega(\mathcal{P}, t ; z, k):=|\operatorname{PP}(\mathcal{P}, t ; z, k)|$ be the number of such $\mathcal{P}$-partitions. The following inequality was conjectured by Graham [Gra83] and proved by Daykin-Daykin-Paterson [DDP84].

Theorem 9.1 (Daykin-Daykin-Paterson inequality). Let $\mathcal{P}=(X, \prec)$ be a finite poset, let $t \in \mathbb{N}$, and let $z \in X$. Then, for every $0 \leq k \leq t$, we have:

$$
\begin{equation*}
\Omega(\mathcal{P}, t ; z, k)^{2} \geq \Omega(\mathcal{P}, t ; z, k-1) \cdot \Omega(\mathcal{P}, t ; z, k+1) \tag{9.1}
\end{equation*}
$$

More generally, for every positive integers $a, b \geq 1$,

$$
\begin{equation*}
\Omega(\mathcal{P}, t ; z, k+a) \cdot \Omega(\mathcal{P}, t ; z, k+b) \geq \Omega(\mathcal{P}, t ; z, k) \cdot \Omega(\mathcal{P}, t ; z, k+a+b) \tag{9.2}
\end{equation*}
$$

We give a new proof of Theorem 9.1 as an application of the AD inequality (5.3). The proof below sets the stage for the multivariate generalization of the theorem.

Proof of Theorem 9.1. We denote by $\mathbb{L}=(\mathcal{L}, \vee, \wedge)$ the distributive lattice on the set $\mathcal{L}$ given by

$$
\mathcal{L}:=\{T: X \rightarrow\{-b,-b+1, \ldots, t\}: T(x) \leq T(y) \forall x, y \in X \text { s.t. } x \prec y\},
$$

the set of order-preserving functions such that $-b \leq T(x) \leq t$ for every $x \in X$. The join and meet operation are given by

$$
[S \vee T](x):=\max \{S(x), T(x)\} \quad \text { and } \quad[S \wedge T](x):=\min \{S(x), T(x)\}
$$

for every $x \in X$. It is straightforward to verify that $\mathbb{L}$ is a distributive lattice.
Let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be characteristic function of subsets of $\mathcal{L}$ defined as follows:

$$
\begin{aligned}
\alpha & :=\mathbf{1}\{T(z)=k \text { and } T(x) \geq 0, \text { for all } x \in X\}, \\
\beta & :=\mathbf{1}\{T(z)=k+a \text { and } T(x) \leq t-b, \text { for all } x \in X\}, \\
\gamma & :=\mathbf{1}\{T(z)=k+a \text { and } T(x) \geq 0, \text { for all } x \in X\}, \\
\delta & :=\mathbf{1}\{T(z)=k \text { and } T(x) \leq t-b, \text { for all } x \in X\} .
\end{aligned}
$$

We will now verify the assumption of AD inequality, i.e. for every $S, T \in \mathcal{L}$, we have:

$$
\begin{equation*}
\alpha(S) \cdot \beta(T) \leq \gamma(S \vee T) \cdot \delta(S \wedge T) \tag{9.3}
\end{equation*}
$$

Without loss of generality we can assume that $\alpha(S)=\beta(T)=1$. Note that

$$
[S \vee T](z)=\max \{S(z), T(z)\}=\max \{k, k+a\}=k+a .
$$

Also note that, for every $x \in X$,

$$
[S \vee T](x)=\max \{S(x), T(x)\} \geq S(x) \geq 0
$$

This shows that $\gamma(S \vee T)=1$. Similarly, note that

$$
[S \wedge T](z)=\min \{S(z), T(z)\}=\min \{k, k+a\}=k
$$

Also note that, for every $x \in X$,

$$
[S \wedge T](x)=\min \{S(x), T(x)\} \leq T(x) \leq t-b
$$

This shows that $\delta(S \wedge T)=1$, and completes the proof of (9.3).
Now note that

$$
\begin{aligned}
& \alpha(\mathcal{L})=\mid\{T \in \mathcal{L}: T(z)=k \text { and } 0 \leq T(x) \leq t \forall x \in X\} \mid=\Omega(\mathcal{P}, t ; z, k), \quad \text { and } \\
& \gamma(\mathcal{L})=\mid\{T \in \mathcal{L}: T(z)=k+a \text { and } 0 \leq T(x) \leq t \forall x \in X\} \mid=\Omega(\mathcal{P}, t ; z, k+a) .
\end{aligned}
$$

Also note that

$$
\begin{align*}
\beta(\mathcal{L}) & =\mid\{T \in \mathcal{L}: T(z)=k+a \text { and }-b \leq T(x) \leq t-b \forall x \in X\} \mid \\
& =\mid\left\{T^{\prime} \in \mathcal{L}: T^{\prime}(z)=k+a+b \text { and } 0 \leq T^{\prime}(x) \leq t \forall x \in X\right\} \mid  \tag{9.4}\\
& =\Omega(\mathcal{P}, t ; z, k+a+b)
\end{align*}
$$

where the second equality is obtained through the substitution $T^{\prime}(x) \leftarrow T(x)+b$. Similarly, by the same substitution we have:

$$
\begin{align*}
\delta(\mathcal{L}) & =\mid\{T \in \mathcal{L}: T(z)=k \text { and }-b \leq T(x) \leq t-b \forall x \in X\} \mid \\
& =\mid\left\{T^{\prime} \in \mathcal{L}: T^{\prime}(z)=k+b \text { and } 0 \leq T^{\prime}(x) \leq t \forall x \in X\right\} \mid  \tag{9.5}\\
& =\Omega(\mathcal{P}, t ; z, k+b) .
\end{align*}
$$

Now (9.2) follows from the AD inequality (5.3).

Remark 9.2. The original proof of the DDP inequality was through an explicit injection [DDP84]. Curiously, Graham believed that there should exist a proof based on the FKG or AD inequalities. He lamented: "such a proof has up to now successfully eluded all attempts to find it" [Gra83, p. 15]. The proof above validates Graham's supposition.

We should also mention that if the order-preserving functions are replaced with linear extensions, the DPP inequality (9.1) becomes Stanley's inequality [Sta81], a major result in the area for which finding a direct combinatorial proof remans a challenging open problem. We refer to [Pak22, §6.3] for an extensive discussion and further references.
9.2. Multivariate DDP inequality. Let $\boldsymbol{q}:=\left(q_{1}, \ldots, q_{n}\right)$ be variables, and fix a natural labeling $X=\left\{x_{1}, \ldots, x_{n}\right\}$. Define

$$
\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k):=\sum_{A \in \operatorname{PP}(\mathcal{P}, t ; z, k)} q_{1}^{A\left(x_{1}\right)} \cdots q_{n}^{A\left(x_{n}\right)}
$$

We now present the multivariate version of DDP inequality (9.1), proved by the multivariate AD inequality (6.3).

Theorem 9.3 (multivariate $D D P$ inequality). Let $\mathcal{P}=(X, \prec)$ be a finite poset, let $t \in \mathbb{N}$, and let $z \in X$. Then, for every $0 \leq k \leq t$, we have:

$$
\begin{equation*}
\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k)^{2} \geqslant_{\boldsymbol{q}} \Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k-1) \cdot \Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+1) \tag{9.6}
\end{equation*}
$$

More generally, for every integer $a, b \geq 1$, we have:

$$
\begin{equation*}
\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+a) \cdot \Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+b) \geqslant_{\boldsymbol{q}} \Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k) \cdot \Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+a+b) \tag{9.7}
\end{equation*}
$$

Note that in contrast with the DPP inequality (9.1), the generalized log-concavity (9.7) does not follow from the (usual) log-concavity (9.6) via telescoping.

Proof. Let $\mathcal{L}, \alpha, \beta, \gamma, \delta$ be as in the proof of Theorem 9.1. Note that these functions satisfy the assumption (6.2) of the multivariate AD inequality (6.3). For all $1 \leq i \leq n$, let $r_{i}: \mathcal{L} \rightarrow \mathbb{R}_{+}$be the modular function given by $r_{i}(A):=A\left(x_{i}\right)$, where $A \in \mathcal{L}$. Then:

$$
\begin{array}{rlrl}
\alpha_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L}) & =\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k), & & \beta_{\langle\boldsymbol{q}, r\rangle}(\mathcal{L})=\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+a+b) \cdot\left(q_{1} \cdots q_{n}\right)^{-b} \\
\gamma_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L})=\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+a), & & \delta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L})=\Omega_{\boldsymbol{q}}(\mathcal{P}, t ; z, k+b) \cdot\left(q_{1} \cdots q_{n}\right)^{-b}
\end{array}
$$

The second part of the theorem now follows from the multivariate AD inequality (6.3), and thus also the first part (which is a special case).

Remark 9.4. In the context of Remark 8.1, Theorem 9.3 holds by the same argument if the order-preserving functions are replaced with the strict order-preserving functions. This approach can be extended to general $\mathbb{T}$-labelled $(\mathcal{P}, O)$ tableaux. However, the analogue of (9.6) does not hold if $\Omega_{q}$ is replaced with $\mathrm{K}_{z}$. This is because the weight functions for $\mathrm{K}_{z}$ is not invariant under the translation transformation used in the equations (9.4) and (9.5) in the proof of Theorem 9.1.
9.3. Log-concavity of the multivariate order polynomial. The following corollary follows immediately from Theorem 9.3, and can be viewed as a multivariate generalization of [CPP22b, Thm 4.7], and a poset generalization of the first formula in the proof of Lemma 6.13 in [LPR18, p. 550].

Corollary 9.5. Let $\mathcal{P}=(X, \prec)$ be a finite poset, and let $t \in \mathbb{N}_{\geq 1}$ be a positive integer. Then:

$$
\Omega_{\boldsymbol{q}}(\mathcal{P}, t)^{2} \geqslant_{q} \Omega_{\boldsymbol{q}}(\mathcal{P}, t-1) \cdot \Omega_{\boldsymbol{q}}(\mathcal{P}, t+1)
$$

More generally, for every integers $a, b \geq 1$, we have:

$$
\Omega_{q}(\mathcal{P}, t+a) \cdot \Omega_{q}(\mathcal{P}, t+b) \geqslant_{q} \Omega_{q}(\mathcal{P}, t) \cdot \Omega_{\boldsymbol{q}}(\mathcal{P}, t+a+b)
$$

Proof. Let $n:=|X|$. Let $\mathcal{P}^{\prime}:=\mathcal{P} \oplus z$ be the linear sum of $\mathcal{P}$ and an extra element $z$, which is the unique maximal element in $\mathcal{P}^{\prime}$. Since we use natural labeling, element $z$ corresponds to the variable $q_{n+1}$.

Note that for every $\ell, t \in \mathbb{N}$, we have:

$$
\begin{equation*}
\Omega_{q}\left(\mathcal{P}^{\prime}, t ; z, \ell\right)=\Omega_{\boldsymbol{q}}(\mathcal{P}, \ell) \cdot q_{n+1}^{\ell} \tag{9.8}
\end{equation*}
$$

On the other hand, it follows from applying Theorem 9.3 to $\mathcal{P}^{\prime}$ that

$$
\Omega_{\boldsymbol{q}}\left(\mathcal{P}^{\prime}, t ; z, k+a\right) \cdot \Omega_{\boldsymbol{q}}\left(\mathcal{P}^{\prime}, t ; z, k+b\right) \geqslant_{\boldsymbol{q}} \Omega_{\boldsymbol{q}}\left(\mathcal{P}^{\prime}, t ; z, k\right) \cdot \Omega_{\boldsymbol{q}}\left(\mathcal{P}^{\prime}, t ; z, k+a+b\right)
$$

The corollary now follows by applying (9.8) to the equation above.
Remark 9.6. Our proof of the $\boldsymbol{q}=\mathbf{1}$ version in [CPP22b, Thm 4.7] goes along similar lines, but uses the FKG rather than the AD inequality. Note that our [CPP22b, Thm 4.8] gives a strict log-concavity for order polynomials, with a substantially more involved proof.

## 10. Cross-Product inequality for $\mathcal{P}$-partitions

10.1. The statement. Let $\mathcal{P}=(X, \prec)$ be a poset on $|X|=n$ elements. Fix $t \geq 0$ and distinct elements $x, y, z \in X$. For integers $k, \ell \geq 0$, denote by

$$
\operatorname{SPP}(\mathcal{P}, t ; x, y, z ; k, \ell):=\{A \in \operatorname{PP}(\mathcal{P}, t): A(y)-A(x)=k \text { and } A(z)-A(y)=\ell\} .
$$

Denote

$$
\Lambda_{q}(k, \ell):=\sum_{A \in \operatorname{SPP}(\mathcal{P}, t ; x, y, z ; k, \ell)} q^{|A|}, \quad \Lambda_{\boldsymbol{q}}(k, \ell):=\sum_{A \in \operatorname{SPP}(\mathcal{P}, t ; x, y, z ; k, \ell)} q_{1}^{A\left(x_{1}\right)} \cdots q_{n}^{A\left(x_{n}\right)},
$$

and let $\mathrm{F}(k, \ell):=\Lambda_{1}(k, \ell)=|\operatorname{SPP}(\mathcal{P}, t ; x, y, z ; k, \ell)|$.
Theorem 10.1 (Cross-product inequality for $\mathcal{P}$-partitions). Let $\mathcal{P}=(X, \prec)$ be a finite poset, let $x, y, z \in \mathcal{P}$, and let $t \in \mathbb{N} \geq 1$ be a positive integer. Then, for every $k, \ell \geq 0$, we have:

$$
\begin{equation*}
\mathrm{F}(k, \ell+1) \cdot \mathrm{F}(k+1, \ell) \geq \mathrm{F}(k, \ell) \cdot \mathrm{F}(k+1, \ell+1) \tag{10.1}
\end{equation*}
$$

More generally:

$$
\begin{equation*}
\Lambda_{q}(k, \ell+1) \cdot \Lambda_{q}(k+1, \ell) \geqslant_{q} \Lambda_{q}(k, \ell) \cdot \Lambda_{q}(k+1, \ell+1) \tag{10.2}
\end{equation*}
$$

Even more generally:

$$
\begin{equation*}
\Lambda_{\boldsymbol{q}}(k, \ell+1) \cdot \Lambda_{\boldsymbol{q}}(k+1, \ell) \geqslant_{q} \Lambda_{\boldsymbol{q}}(k, \ell) \cdot \Lambda_{\boldsymbol{q}}(k+1, \ell+1) . \tag{10.3}
\end{equation*}
$$

Remark 10.2. Note that already the unweighted inequality (10.1) appears to be new. Note also that if the order-preserving functions are replaced with linear extensions, then a version of (10.1) is known as the cross-product conjecture [BFT95, Conj 3.1], a major open problem in the area. We refer to [CPP22a] for an extensive discussion and further references.
10.2. Proof of Theorem 10.1. We denote by $\mathbb{L}=(\mathcal{L}, \vee, \wedge)$ the distributive lattice on the set of order-preserving functions from $X$ to $\{0,1, \ldots, t\}$ :

$$
\mathcal{L}:=\{T: X \rightarrow\{0,1, \ldots, t\} \quad: T(v) \leq T(w) \forall v, w \in X \text { s.t. } v \prec w\} .
$$

The join and meet operation are given by

$$
\begin{aligned}
& {[S \vee T](w):=\max \{S(w)-S(y), T(w)-T(y)\}+\min \{S(y), T(y)\},} \\
& {[S \wedge T](w):=\min \{S(w)-S(y), T(w)-T(y)\}+\max \{S(y), T(y)\},}
\end{aligned}
$$

for every $w \in X$. This lattice was proved distributive by Shepp [She80, Eq. 2.4, 2.5], in his proof of the $X Y Z$ inequality (see also [AS16, §6.4]).

Let $\alpha, \beta, \gamma, \delta: \mathcal{L} \rightarrow \mathbb{R}_{+}$be characteristic function of subsets of $\mathcal{L}$ defined as follows:

$$
\begin{aligned}
\alpha & :=\mathbf{1}\{T(y)-T(x)=k \text { and } T(z)-T(y)=\ell\}, \\
\beta & :=\mathbf{1}\{T(y)-T(x)=k+1 \text { and } T(z)-T(y)=\ell+1\}, \\
\gamma & :=\mathbf{1}\{T(y)-T(x)=k \text { and } T(z)-T(y)=\ell+1\}, \\
\delta & :=\mathbf{1}\{T(y)-T(x)=k+1 \text { and } T(z)-T(y)=\ell\} .
\end{aligned}
$$

We will now verify the assumption (6.2) of the multivariate AD inequality:

$$
\begin{equation*}
\alpha(S) \cdot \beta(T) \leq \gamma(S \vee T) \cdot \delta(S \wedge T) \tag{10.4}
\end{equation*}
$$

for every $S, T \in \mathcal{L}$. Without loss of generality we can assume that $\alpha(S)=\beta(T)=1$. We have:

$$
\begin{aligned}
& {[S \vee T](x)-[S \vee T](y)=\max \{S(x)-S(y), T(x)-T(y)\}=\max \{-k,-k-1\}=-k,} \\
& {[S \vee T](z)-[S \vee T](y)=\max \{S(z)-S(y), T(z)-T(y)\}=\max \{\ell, \ell+1\}=\ell+1,} \\
& {[S \wedge T](x)-[S \wedge T](y)=\min \{S(x)-S(y), T(x)-T(y)\}=\min \{-k,-k-1\}=-k-1,} \\
& {[S \wedge T](z)-[S \wedge T](y)=\min \{S(z)-S(y), T(z)-T(y)\}=\min \{\ell, \ell+1\}=\ell .}
\end{aligned}
$$

This shows that $\gamma(S \vee T)=\delta(S \wedge T)=1$ and proves (10.4).
Finally, consider modular functions $r_{i}: \mathcal{L} \rightarrow \mathbb{R}_{+}$, for all $1 \leq i \leq n$, given by $r_{i}(T):=T\left(x_{i}\right)$. Then we have:

$$
\begin{aligned}
\alpha_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L}) & =\Lambda_{\boldsymbol{q}}(k, \ell), & & \beta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L})=\Lambda_{\boldsymbol{q}}(k+1, \ell+1), \\
\gamma_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L}) & =\Lambda_{\boldsymbol{q}}(k, \ell+1) & & \delta_{\langle\boldsymbol{q}, \boldsymbol{r}\rangle}(\mathcal{L})=\Lambda_{\boldsymbol{q}}(k+1, \ell) .
\end{aligned}
$$

The theorem now follows from the multivariate AD inequality (6.3).
Remark 10.3. Let us also mention that the proof in [CPP22a, $\S 3.1]$ shows that Theorem 10.1 implies a (multivariate) $\mathcal{P}$-partition version of the Kahn-Saks inequality [KS84, Thm 2.5]. On the other hand, while the KS inequality easily implies Stanley's inequality discussed earlier in Remark 9.2 (see e.g. [CPP23, §1.2]), the multivariate DPP inequality (Theorem 9.3) does not similarly follow from cross-product inequality for $\mathcal{P}$-partitions (Theorem 10.1). This is also demonstrated by the fact that different lattices are used in the proofs of the two theorems.

## 11. Final Remarks and open problems

11.1. This paper grew out of [CP22, §4.1] where we obtained superficially similar correlation inequalities which appear to have a very different nature and whose only known proof uses the combinatorial atlas technology. Our investigation was also partly motivated by the desire to bridge the gap between the two areas of combinatorics. Notably, we would like to emphasize the importance of the AD inequality to algebraic combinatorics, and the multivariate weighting to poset theory.

Note that there is a weighted version of $e(\mathcal{P})$ introduced in [CP21, §1.16]. While the results in [CP22] translate verbatim to the weighted setting, these weights seem incompatible with $\boldsymbol{q}$ weights in this paper. Similarly, the $q$-weight on $e(\mathcal{P})$ in [CPP22a] is also of different nature. On the other hand, the $q$-weighted order polynomial in [CPP22b] is exactly $\Omega_{q}(\mathcal{P}, t)$.
11.2. One distinguishing feature of poset inequalities is the difficulty of getting the equality conditions, see e.g. [CPP22b, $\S 9.9]$ for an overview. We are not aware of any equality conditions for the inequalities in this paper, proved or conjectured.

Another difficulty is finding a combinatorial interpretation for the difference of two sides. This was a major motivation for our investigation in [CP21]. We show in [IP22, §7.4] that the AD inequality (5.3) does not have a combinatorial interpretation in full generality, in a sense of being in \#P. Of course, the Lam-Postnikov-Pylyavskyy deep algebraic approach in [LPP07] (see Remark 4.2) is even less likely to give a combinatorial interpretation. We refer to [Pak22, §6] for an extensive survey.

Now, the Lam-Pylyavskyy's injective approach in [LP07] shows that the difference of coefficients on both sides in (4.12) has a combinatorial interpretation. By contrast, the limit arguments we use throughout this paper do not give a combinatorial interpretation for Fishburn's inequality (3.1). It would be interesting to see if (3.1) and the generalized Fishburn inequality (3.5) can be proved by a direct combinatorial argument giving a combinatorial interpretation.

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## References

[AB08] Rudolf Ahlswede and Vladimir Blinovsky, Lectures on advances in combinatorics, Springer, Berlin, 2008, 314 pp.
[AD78] Rudolf Ahlswede and David E. Daykin, An inequality for the weights of two families of sets, their unions and intersections, Z. Wahrsch. Verw. Gebiete 43 (1978), 183-185.
[AS16] Noga Alon and Joel H. Spencer, The probabilistic method (Fourth ed.), John Wiley, Hoboken, NJ, 2016.
[Bir67] Garrett Birkhoff, Lattice theory (third ed.), AMS, Providence, RI, 1967, 418 pp.
[Bjö11] Anders Björner, A $q$-analogue of the FKG inequality and some applications, Combinatorica 31 (2011), 151-164.
[Brä15] Petter Brändén, Unimodality, log-concavity, real-rootedness and beyond, in Handbook of enumerative combinatorics, CRC Press, Boca Raton, FL, 2015, 437-483. i
[Bre89] Francesco Brenti, Unimodal, log-concave and Pólya frequency sequences in combinatorics, Mem. AMS 81 (1989), no. 413, 106 pp.
[Bre94] Francesco Brenti, Log-concave and unimodal sequences in algebra, combinatorics, and geometry: an update, in Jerusalem combinatorics, AMS, Providence, RI, 1994, 71-89.
[BFT95] Graham R. Brightwell, Stefan Felsner and William T. Trotter, Balancing pairs and the cross product conjecture, Order 12 (1995), 327-349.
[CP21] Swee Hong Chan and Igor Pak, Log-concave poset inequalities, preprint (2021), 71 pp ; arXiv:2110. 10740.
[CP22] Swee Hong Chan and Igor Pak, Correlation inequalities for linear extensions, preprint (2022), 23 pp .; arXiv:2211.16637.
[CPP22a] Swee Hong Chan, Igor Pak and Greta Panova, The cross-product conjecture for width two posets, Trans. AMS 375 (2022), 5923-5961.
[CPP22b] Swee Hong Chan, Igor Pak, and Greta Panova, Effective poset inequalities, to appear in SIAM J. Discrete Math.; preprint (2022), 36 pp ; arXiv:2205. 02798.
[CPP23] Swee Hong Chan, Igor Pak and Greta Panova, Extensions of the Kahn-Saks inequality for posets of width two, Combinatorial Theory 3 (2023), no. 1, Paper No. 8, 35 pp.
[Chr09] Demetres Christofides, A $q$-analogue of the four functions theorem, preprint (2009), 6 pp .; arXiv: 0909.5137.
[CGS11] Allison Cuttler, Curtis Greene and Mark Skandera, Inequalities for symmetric means, European J. Combin. 32 (2011), 745-761.
[DDP84] David E. Daykin, Jacqueline W. Daykin, and Michael S. Paterson, On log concavity for order-preserving maps of partial orders, Discrete Math. 50 (1984), 221-226.
[Fis84] Peter C. Fishburn, A correlational inequality for linear extensions of a poset, Order 1 (1984), 127-137.
[Fis92] Peter C. Fishburn, Correlation in partially ordered sets, Discrete Appl. Math. 39 (1992), $173-191$.
[FS98] Peter C. Fishburn and Lawrence A. Shepp, The Ahlswede-Daykin theorem, in Numbers, information and complexity, Kluwer, Boston, MA, 2000, 501-516.
[FKG71] Cornelius M. Fortuin, Pieter W. Kasteleyn and Jean Ginibre, Correlation inequalities on some partially ordered sets, Comm. Math. Phys. 22 (1971), 89-103.
[Gra83] Ronald L. Graham, Applications of the FKG inequality and its relatives, in Mathematical programming: the state of the art, Springer, Berlin, 1983, 115-131.
[Huh18] June Huh, Combinatorial applications of the Hodge-Riemann relations, in Proc. ICM Rio de Janeiro, vol. IV, World Sci., Hackensack, NJ, 2018, 3093-3111.
[HMMS22] June Huh, Jacob P. Matherne, Karola Mészáros and Avery St. Dizier, Logarithmic concavity of Schur and related polynomials, Trans. AMS 375 (2022), 4411-4427.
[HSW22] June Huh, Benjamin Schröter and Botong Wang, Correlation bounds for fields and matroids, Jour. Eur. Math. Soc. 24 (2022), 1335-1351.
[IP22] Christian Ikenmeyer and Igor Pak, What is in \#P and what is not?, preprint (2022), 82 pp.; extended abstract in Proc. 63rd FOCS (2022), 860-871; arXiv:2204.13149.
[KS84] Jeff Kahn and Michael Saks, Balancing poset extensions, Order 1 (1984), 113-126.
[LPP07] Thomas Lam, Alexander Postnikov and Pavlo Pylyavskyy, Schur positivity and Schur log-concavity, Amer. J. Math. 129 (2007), 1611-1622.
[LP07] Thomas Lam and Pavlo Pylyavskyy, Cell transfer and monomial positivity, J. Algebraic Combin. 26 (2007), 209-224.
[LP08] Thomas Lam and Pavlo Pylyavskyy, $P$-partition products and fundamental quasi-symmetric function positivity, Adv. in Appl. Math. 40 (2008), 271-294.
[LPR18] Thomas Lam, Pavlo Pylyavskyy, and Reiho Sakamoto, Rigged configurations and cylindric loop Schur functions, Ann. Inst. Henri Poincaré D 5 (2018), 513-555.
[Mac95] Ian G. Macdonald, Symmetric functions and Hall polynomials (Second ed.), Oxford Univ. Press, New York, 1995, 475 pp .
[Pak22] Igor Pak, What is a combinatorial interpretation?, preprint (2022), 58 pp.; in Open Problems in Algebraic Combinatorics, to appear; arXiv:2209.06142.
[Ric10] Donald St. P. Richards, Log-convexity properties of Schur functions and generalized hypergeometric functions of matrix argument, Ramanujan J. 23 (2010), 397-407.
[Sag01] Bruce E. Sagan, The symmetric group, Springer, New York, 2001, 238 pp.
[She80] Lawrence A. Shepp, The FKG inequality and some monotonicity properties of partial orders, SIAM J. Algebraic Discrete Methods 1 (1980), 295-299.
[Sra16] Suvrit Sra, On inequalities for normalized Schur functions, European J. Combin. 51 (2016), 492-494.
[Sta72] Richard P. Stanley, Ordered structures and partitions, Memoirs of the AMS, AMS, Providence, RI, 1972, 104 pp.
[Sta81] Richard P. Stanley, Two combinatorial applications of the Aleksandrov-Fenchel inequalities, J. Combin. Theory, Ser. A 31 (1981), 56-65.
[Sta89] Richard P. Stanley, Log-concave and unimodal sequences in algebra, combinatorics, and geometry, in Graph theory and its applications, New York Acad. Sci., New York, 1989, 500-535.
[Sta99] Richard P. Stanley, Enumerative Combinatorics, vol. 1 (second edition) and vol. 2, Cambridge Univ. Press, 2012 and 1999.
[Tro95] Wiliam T. Trotter, Partially ordered sets, in Handbook of combinatorics, vol. 1, Elsevier, Amsterdam, 1995, 433-480.
[Win86] Peter M. Winkler, Correlation and order, in Combinatorics and ordered sets, AMS, Providence, RI, 1986, 151-174.


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[^1]:    ${ }^{1}$ This injection eluded us in the first version [CPP22b], when we were not aware of [DDP84] and proved an asymptotic version of the DDP inequality which we called Graham's conjecture.

[^2]:    ${ }^{2}$ In [Sta72, Sta99], Stanley uses $P$-partitions to denote order-reversing rather than order-preserving maps. We adopt this version for clarity and to unify the notation. Displeased readers can always think of dual posets.
    ${ }^{3}$ A standard definition for order polynomial is $\Omega(\mathcal{P}, t-1)$ as the values in the $\mathcal{P}$-partition are traditionally $\geq 1$. We adopt this version to simplify the notation and hope this does not lead to confusion.

[^3]:    ${ }^{4}$ Note that reverse plane partitions for $\lambda / \mu$ are actually $\mathcal{P}_{\lambda / \mu}$ - partitions. This is another notational compromise we make between the areas.

