The Schur-Positivity Poset

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Slides and papers available from www.facstaff.bucknell.edu/pm040/

- Introduction to the Schur-positivity poset
- Some known properties
- Some unknown properties
- Focus on necessary conditions for $A \leq_s B$.

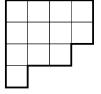
Preview

n = 4 , P F F ₽ ₽ ΗP \square

Schur functions

• Partition
$$\lambda = (\lambda_1, \lambda_2, \dots, \lambda_\ell)$$

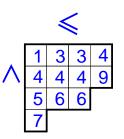
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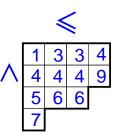
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The Schur function s_{λ} in the variables $x = (x_1, x_2, ...)$ is then defined by

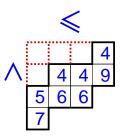
$$s_{\lambda} = \sum_{\text{SSYT } T} x_1^{\#1\text{'s in } T} x_2^{\#2\text{'s in } T} \cdots$$

Example.

$$s_{4431} = x_1 x_3^2 x_4^4 x_5 x_6^2 x_7 x_9 + \cdots$$

Skew Schur functions

- Partition $\lambda = (\lambda_1, \lambda_2, \dots, \lambda_\ell)$
- μ fits inside λ .
- Young diagram. Example: λ/µ = (4, 4, 3, 1)/(3, 1)
- Semistandard Young tableau (SSYT)



The skew Schur function $s_{\lambda/\mu}$ in the variables $x = (x_1, x_2, ...)$ is then defined by

$$s_{\lambda/\mu} = \sum_{\text{SSYT } T} x_1^{\#1\text{'s in } T} x_2^{\#2\text{'s in } T} \cdots$$

Example.

 $s_{4431/31} = x_4^3 x_5 x_6^2 x_7 x_9 + \cdots$

- Skew Schur functions are symmetric in the variables $x = (x_1, x_2, ...)$.
- The Schur functions form a basis for the algebra of symmetric functions (over Q, say).
- ► Connections to algebraic geometry (Schubert calculus), representation theory (S_n, GL(n, C)).

Littlewood-Richardson Rule

$$m{s}_{\lambda/\mu} = \sum_{
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Littlewood-Richardson rule [Littlewood-Richardson 1934, Schützenberger 1977, Thomas 1974].

 $c_{\mu\nu}^{\lambda}$ is the number of SSYT of shape λ/μ and content ν whose reverse reading word is a ballot sequence.

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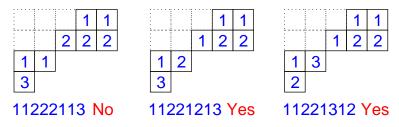
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Example.

When
$$\lambda = (5,5,2,1), \mu = (3,2),
u = (4,3,1),$$
 we get $c_{\mu
u}^{\lambda} = 2$



Key point: $c_{\mu\nu}^{\lambda} \geq 0$.

 $s_{\lambda/\mu}$ is Schur-positive, i.e. coefficients in Schur expansion are all non-negative.

Natural connections between Schur-positivity and representation theory.

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When is $m{s}_{\lambda/\mu} - m{s}_{\sigma/ au}$ Schur-positive?

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Definition.

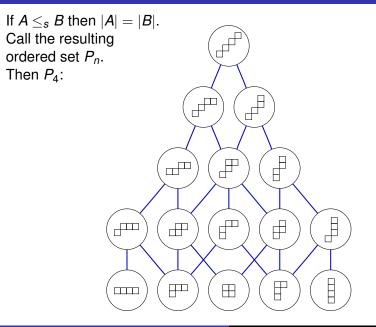
Let A, B be skew shapes. We say that

$$A \leq_s B$$
 if $s_B - s_A$

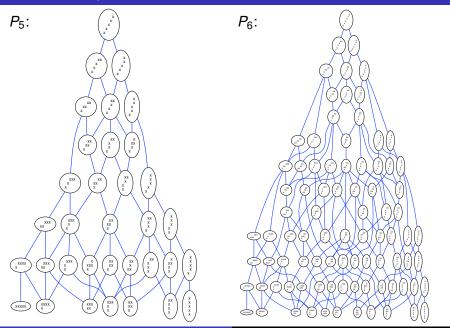
is Schur-positive.

Goal: Characterize the Schur-positivity order \leq_s in terms of skew shapes.

Example of a Schur-positivity poset

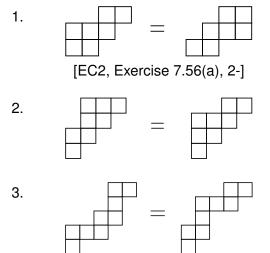


More examples



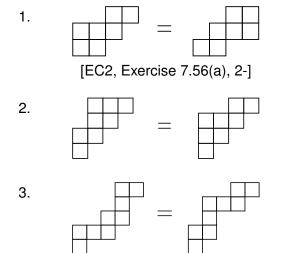
Known properties: first things first

 \leq_s is not yet anti-symmetric. So identify skew shapes such as



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Definition.

A ribbon is a connected skew shape containing no 2×2 rectangle.

Question: When is $s_A = s_B$?

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Complete classification of equality of ribbon Schur functions

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Complete classification of equality of ribbon Schur functions

- ► Vic Reiner, Kristin Shaw, Stephanie van Willigenburg (2006)
- McN., Stephanie van Willigenburg (2006)

Enough for our purposes: we can consider P_n to be a poset.

Open Problem: Find necessary and sufficient conditions on *A* and *B* for $s_A = s_B$.

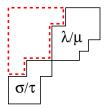
Known properties: Sufficient conditions

Sufficient conditions for $A \leq_s B$:

- Alain Lascoux, Bernard Leclerc, Jean-Yves Thibon (1997)
- Andrei Okounkov (1997)
- Sergey Fomin, William Fulton, Chi-Kwong Li, Yiu-Tung Poon (2003)
- Anatol N. Kirillov (2004)
- Thomas Lam, Alex Postnikov, Pavlo Pylyavskyy (2005)
- François Bergeron, Riccardo Biagioli, Mercedes Rosas (2006)

Þ ...

Note: $s_{\lambda/\mu}s_{\sigma/\tau}$ is a special case of s_A .

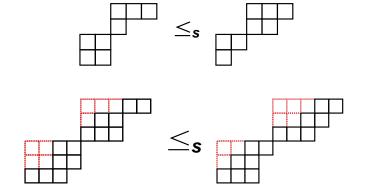


Lam, Postnikov and Pylyavskyy's result

Theorem [LPP]. For skew shapes λ/μ and σ/τ ,

$$\mathbf{s}_{\lambda/\mu}\mathbf{s}_{\sigma/ au} \leq_{\mathbf{s}} \mathbf{s}_{\lambda\cup\sigma/\mu\cup au}\mathbf{s}_{\lambda\cap\sigma/\mu\cap au}$$

Examples.



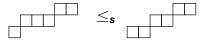
Known properties: special classes of skew shapes

Notation. Write $\lambda \preccurlyeq \mu$ if λ is less than or equal to μ in dominance order, i.e.

$$\lambda_1 + \cdots + \lambda_i \leq \mu_1 + \cdots + \mu_i$$
 for all *i*.

► Macdonald's "Symmetric functions and Hall polynomials": For horizontal strips, A ≤_s B if and only if

row lengths of $A \succ$ row lengths of B



 P_n restricted to horizontal strips: (dual of the) dominance lattice.

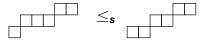
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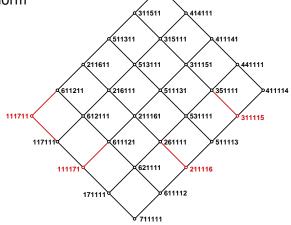
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Ron King, Trevor Welsh, Stephanie van Willigenburg (2007): For ribbons with decreasing row lengths and equal numbers of rows, same is true.



Known properties: special classes of skew shapes

 McN., Stephanie van Willigenburg (2007): For a given number of boxes and rows, the poset of multiplicity-free ribbons is always of the form

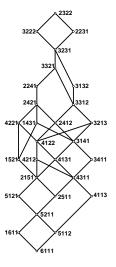


Unknown properties: general ribbons

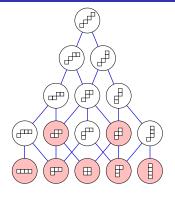
Open Problem: Explain the Schur-positivity order for general ribbons.

Suffices to fix #boxes and #rows.

Ribbons with 9 boxes and 4 rows:

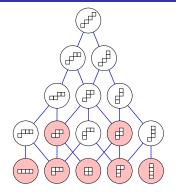


Unknown properties: maximal connected skew shapes



Question: What are the maximal elements of P_n among the connected skew shapes?

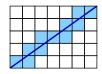
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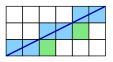


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Conjecture [McN., Pylyavskyy]. For each r = 1, ..., n, there is a unique maximal connected element with r rows, namely the ribbon marked out by the diagonal of an r-by-(n - r + 1) box.

Examples.





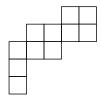
Question: Suppose $A \leq_s B$ (i.e. $s_B - s_A$ is Schur-positive). Then what can we say about the shapes *A* and *B*?

Such necessary conditions for $A \leq_s B$ give us a way to show that $C \not\leq_s D$.

Example. If $A \leq_s B$, then |A| = |B|.

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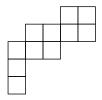
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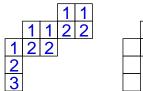
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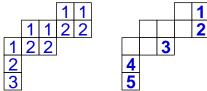
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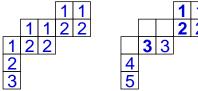
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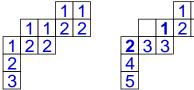
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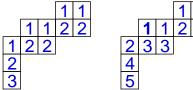
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Proof:

 $A \leq_{s} B$

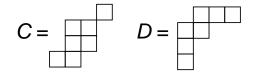
- \Leftrightarrow $s_B s_A$ is Schur-positive
- \Rightarrow support(*A*) \subseteq support(*B*)
- \Rightarrow rows(A) \succ rows(B) and $(cols(A))^t \preccurlyeq (cols(B))^t$
- \Leftrightarrow rows(A) \succ rows(B) and cols(A) \succ cols(B).

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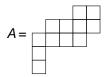
 $rows(C) = 2221 \prec 3211 = rows(D).$ Thus $C \not\leq_s D$.

Definitions [Reiner, Shaw, van Willigenburg]. For a skew shape *A*, let $\operatorname{overlap}_{k}(i)$ be the number of columns occupied in common by rows $i, i + 1, \ldots, i + k - 1$.

Then $rows_k(A)$ is the weakly decreasing rearrangment of

 $(\text{overlap}_k(1), \text{overlap}_k(2), \ldots).$

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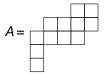
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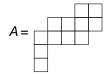
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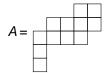
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- overlap₂(1) = 2, overlap₂(2) = 3, overlap₂(3) = 1, overlap₂(4) = 1, so rows₂(A) = 3211.

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$$(\text{overlap}_k(1), \text{overlap}_k(2), \ldots).$$

Define $cols_k(A)$ similarly.



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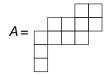
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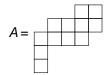
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- rows₃(A) = 11, rows_k(A) = \emptyset for k > 3.
- ▶ $cols_1(A) = cols(A) = 33222$, $cols_2(A) = 2221$, $cols_3(A) = 211$, $cols_4(A) = 11$, $cols_k(A) = \emptyset$ for k > 4.

Theorem [RSvW]. Let *A* and *B* be skew shapes. If $s_A = s_B$, then

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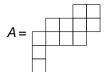
In fact, it suffices to assume that support(A) \subseteq support(B).

Corollary. Let A and B be skew shapes. If support(A) = support(B), then

$$rows_k(A) = rows_k(B)$$
 for all k.

Relating rows $_k(A)$ and cols $_k(A)$

Let $\operatorname{rects}_{k,\ell}(A)$ denote the number of $k \times \ell$ rectangular subdiagrams contained inside *A*.



$$rects_{3,1}(A) = 2$$
, $rects_{2,2}(A) = 3$, etc.

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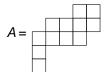
• $rows_k(A) = rows_k(B)$ for all k;

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$$\operatorname{cols}_{\ell}(A) = \operatorname{cols}_{\ell}(B)$$
 for all ℓ ;

• rects<sub>k,
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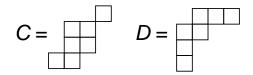
- rows_k(A) \succ rows_k(B) for all k;
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Summary result

Theorem [McN]. Let *A* and *B* be skew shapes. If $A \leq_s B$, i.e. $s_B - s_A$ is Schur-positive, or if *A* and *B* satisfy the weaker condition that support(*A*) \subseteq support(*B*), then the following three equivalent sets of conditions are true:

- rows_k(A) \succeq rows_k(B) for all k;
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Example.



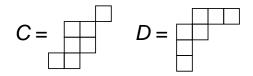
 $rows(C) = 2221 \prec 3211 = rows(D)$. Thus $C \not\leq_s D$.

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- $rows_k(A) \succcurlyeq rows_k(B)$ for all k;
- $\operatorname{cols}_{\ell}(A) \succcurlyeq \operatorname{cols}_{\ell}(B)$ for all ℓ ;
- $\operatorname{rects}_{k,\ell}(A) \ge \operatorname{rects}_{k,\ell}(B)$ for all k, ℓ .

Example.



 $rows(C) = 2221 \prec 3211 = rows(D)$. Thus $C \not\leq_s D$. $rows_2(C) = 21 \succ 111 = rows_2(D)$. Thus $D \not\leq_s C$.

Outlook

- Instead of looking at the Schur-positivity poset, could look at the support containment poset; it seems to have more structure.
- Almost nothing is known about the covering relations in P_n.
- Why restrict to skew Schur functions? Could try:
 - Stanley symmetric functions
 - Hall-Littlewood polynomials
 - LLT-polynomials
 - Cylindric Schur functions
 - Skew Grothendieck polynomials
 - Poset quasisymmetric functions
 - Wave Schur functions
 - ▶ ...