

STEP 3 Show that the "algebraically defined" Schur functions satisfy the condition

$$\sum_{\lambda} s_{\lambda}(x) s_{\lambda}(y) = \prod_{i,j} \frac{1}{1-x_i y_j}. \quad (*)$$

Cauchy formula

[Hence, the $\{s_{\lambda}\}$'s are a self-dual basis of A .]

Notice that it is enough to show that

$$\sum_{\substack{\lambda \text{ partitions in} \\ \text{at most } n \text{ parts}}} s_{\lambda}(x_1 \dots x_n) s_{\lambda}(y_1 \dots y_n) = \prod_{\substack{i=1 \dots n \\ j=1 \dots n}} \left(\frac{1}{1-x_i y_j} \right)$$

the identity $(*)$ will follow by letting $n \rightarrow \infty$.

Theorem [Cauchy formula] $\sum_{\substack{\lambda \text{ partition} \\ \text{in at most } n \\ \text{parts}}} s_{\lambda}(x_1 \dots x_n) s_{\lambda}(y_1 \dots y_n) =$

$$= \prod_{i,j=1 \dots n} \frac{1}{1-x_i y_j}.$$

proof - We show that

$$\sum_{\substack{\lambda \text{ partitions in at} \\ \text{most } n \text{ parts}}}$$

$$a_{\lambda+\delta}(x_1 \dots x_n) a_{\lambda+\delta}(y_1 \dots y_n) = \frac{a_{\delta}(x_1 \dots x_n) a_{\delta}(y_1 \dots y_n)}{\prod_{i,j=1}^n (1-x_i y_j)}$$

by showing that both sides of the equation
are equal To

$$\det \left(\frac{1}{1-x_i y_j} \right)_{i,j=1 \dots n}$$

So the plan is To evaluate This determinant in two
different ways...

CLAIM 1 $\det \left(\frac{1}{1-x_i y_j} \right)_{i,j=1 \dots n} = \frac{a_g(x_1 \dots x_n) a_g(y_1 \dots y_n)}{\prod_{i,j=1 \dots n} (1-x_i y_j)}$

for all $n \geq 1$.

Proof- By induction on n .

Base of induction: $n=1$, then $\det \left(\frac{1}{1-x_1 y_1} \right) = \frac{1}{1-x_1 y_1} =$
 $= \frac{a_g(x_1) a_g(y_1)}{1-x_1 y_1}$ ✓

Inductive hypothesis: True for $n-1$ variables.

Inductive step : Compute $\det \left(\frac{1}{1-x_i y_j} \right)_{i,j=1 \dots n}$.

We can write :

$$\det \left(\frac{1}{1-x_i y_j} \right)_{i,j=1 \dots n} = \det \left[\begin{array}{c|c|c|c|c} \frac{1}{1-x_1 y_1} & & & & \\ \hline & \frac{1}{1-x_2 y_1} & & & \\ \hline & & \ddots & & \\ \hline & & & \frac{1}{1-x_n y_1} & \\ \hline \end{array} \right]_{j=1 \dots n} =$$

$$\det \left[\begin{array}{c} \frac{1}{1-x_i y_j} \\ \vdots \\ \frac{y_j (x_i - x_1)}{(1-x_1 y_j)(1-x_i y_j)} \\ \vdots \\ \frac{1}{1-x_n y_j} \end{array} \right] =$$

$$\frac{1}{1-x_i y_j} - \frac{1}{1-x_1 y_j} = \frac{y_j (x_i - x_1)}{(1-x_1 y_j)(1-x_i y_j)}$$

$$\det \left[\begin{array}{c} 1 & 1 & \dots & 1 \\ \vdots & \vdots & & \vdots \\ \frac{(x_i - x_1) \det}{(1-x_1 y_j)} & \frac{y_j}{1-x_1 y_j} & \ddots & \frac{y_j}{1-x_n y_j} \end{array} \right] =$$

Take out a factor $\frac{1}{1-x_i y_j}$ from every j^{th} column
and a factor $(x_i - x_1)$ from every i^{th} row,
with $i = 2 \dots n$

$$\det \left[\begin{array}{c} 1 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ \frac{x_i - x_1}{1-x_1 y_j} \det & \frac{y_j - y_1}{(1-x_1 y_j)(1-x_i y_j)} & \ddots & \frac{y_j - y_1}{(1-x_1 y_j)(1-x_n y_j)} \end{array} \right] =$$

$$\frac{y_j}{1-x_i y_j} - \frac{y_1}{1-x_1 y_1} = \frac{y_j - y_1}{(1-x_1 y_j)(1-x_i y_1)}$$

$$\frac{1}{(1-x_1 y_1)} \prod_{i=2}^n \frac{(x_i - x_1)}{(1-x_i y_1)} \prod_{j=2}^n \frac{(y_j - y_1)}{(1-x_1 y_j)},$$

Take out a factor $y_j - y_1$ from the j^{th} column, $\forall j \geq 2$

and a factor $\frac{1}{1-x_i y_1}$ from the i^{th} row, $\forall i \geq 2$

$$\det \left[\frac{1}{1-x_i y_j} \right]_{i,j=2 \dots n} =$$

$$\det \left[\begin{array}{c} 1 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ \frac{1}{1-x_i y_j} & & & & \end{array} \right]_{i,j=2 \dots n}$$

now we can use induction

$$\begin{aligned}
 &= \frac{1}{1-x_1y_1} \prod_{i=2-n}^{\pi} \frac{x_i - x_1}{1-x_iy_1} \prod_{j=2-n}^{\pi} \frac{y_j - y_1}{1-x_1y_j} \frac{a_g(x_2 - x_n) a_g(y_2 - y_n)}{\prod_{i,j=2-n}^{\pi} (1-x_iy_j)} = \\
 &= \frac{(-1)^n \left[\prod_{i=2}^n (x_1 - x_i) \right] (-1)^n \left[\prod_{j=2-n}^n (y_1 - y_j) \right] \left[\prod_{2 < i < j < n} (x_i - x_j) \prod_{2 < i < j < n} (y_i - y_j) \right]}{\prod_{i,j=1-n}^{\pi} (1-x_iy_j)} = \\
 &= \frac{\prod_{1 < i < j < n} (x_i - x_j) \prod_{1 < i < j < n} (y_i - y_j)}{\prod_{i,j=1-n}^{\pi} (1-x_iy_j)} = \frac{a_g(x_1 - x_n) a_g(y_1 - y_n)}{\prod_{j=1}^n (1-x_jy_j)}
 \end{aligned}$$

OK

$$\text{CLAIM 2} \quad \det \left(\frac{1}{1-x_iy_j} \right)_{i,j=1-n} = \sum_{\substack{\lambda \text{ partitions in} \\ \text{at most } n \\ \text{parts}}} a_{\lambda+\delta}^{(x_1 - x_n)} a_{\lambda+\delta}^{(y_1 - y_n)}$$

Indeed we can write :

$$\det \left(\frac{1}{1-x_iy_j} \right)_{i,j=1-n} = \det \left(\sum_{k,j=0}^{\infty} (x_i y_j)^{d_{ij}} \right) \rightarrow$$

When you take the determinant, you never multiply in the same row. So you can use

The same index α_i for all the entries in the i^{th} row. || 6

$$\rightarrow = \det \left(\sum_{\alpha_i=0}^{\infty} (x_i y_j)^{\alpha_i} \right)_{\substack{i=1 \dots n \\ j=1 \dots n}} =$$

$$= \det \begin{vmatrix} \sum_{\alpha_1=0}^{\infty} x_1^{\alpha_1} y_j^{\alpha_1} \\ \sum_{\alpha_2=0}^{\infty} x_2^{\alpha_2} y_j^{\alpha_2} \\ \vdots \\ \sum_{\alpha_n=0}^{\infty} x_n^{\alpha_n} y_j^{\alpha_n} \end{vmatrix}_{j=1 \dots n} =$$

$$= \sum_{\alpha_1, \alpha_2, \dots, \alpha_n \geq 0} x_1^{\alpha_1} x_2^{\alpha_2} \dots x_n^{\alpha_n} \det \begin{bmatrix} y_j^{\alpha_1} \\ y_j^{\alpha_2} \\ \vdots \\ y_j^{\alpha_n} \end{bmatrix} =$$

↗ linearity
on rows

$$= \sum_{\alpha \in \mathbb{N}^n} x^\alpha \det \begin{bmatrix} y_j^{\alpha_i} \\ \vdots \\ y_j^{\alpha_n} \end{bmatrix}_{i=1 \dots n} =$$

compositions in
 n parts

$$= \sum_{\alpha \in \mathbb{N}^n} x^\alpha \det \begin{bmatrix} y_i^{\alpha_j} \\ \vdots \\ y_i^{\alpha_n} \end{bmatrix}_{i=1 \dots n} = \sum_{\alpha \in \mathbb{N}^n} x^\alpha \alpha_\alpha(y_1 \dots y_n) =$$

↗ The determinant
is invariant under
Transposition

$$= \sum_{\substack{\alpha \in \mathbb{N}^n \\ \text{distinct parts}}} x^\alpha \alpha_\alpha(y_1 \dots y_n)$$

$\alpha_\alpha = 0$ if α has repeated entries

any $\alpha \in \mathbb{N}^n$
with not equal
parts is a permu
tron of some $\lambda + \delta$,
for with $\lambda = \text{a part}$
(in at most n
parts)

$$= \sum_{\substack{\lambda: \text{partitions} \\ \text{in at most } n \text{ parts}}} \left(\sum_{\sigma \in S_n} \times \alpha_{\sigma(\lambda+\delta)}^{(y_1 \dots y_n)} \right) =$$

$$= \sum_{\substack{\lambda: \text{partitions in} \\ \text{at most } n \text{ parts}}} \left(\sum_{\sigma \in S_n} \operatorname{sgn}(\sigma) \times \alpha_{\sigma(\lambda+\delta)}^{(y_1 \dots y_n)} \right) =$$

$$\begin{array}{l} (\lambda+\delta) = \\ \text{and} \\ \text{add} \end{array}$$

$$= \sum_{\substack{\lambda: \text{partitions} \\ \text{in at most } n \text{ parts}}} \alpha_{\lambda+\delta}^{(x_1 \dots x_n)} \alpha_{\lambda+\delta}^{(y_1 \dots y_n)}.$$

Or

The Cauchy's formula follows. \square

Corollary - The set $\{S_\lambda\}_{\lambda \text{ partition}}$ is a self-dual basis of Λ .

Corollary - The bilinear form \langle , \rangle is symmetric and positive definite. So it's a scalar product on Λ .

To be postponed

Corollary - [Assume the equivalence of the two definitions of Schur functions] Then $\{h_\mu = \sum_{\lambda \vdash \mu} k_{\lambda \mu} S_\lambda\}$.

proof- We know that $s_\lambda = \sum_{g \leq \lambda} k_{\lambda g} m_g$.

Take the inner product of both sides with h_μ :

$$\langle s_\lambda, h_\mu \rangle = \sum_{g \leq \lambda} k_{\lambda g} \underbrace{\langle m_g, h_\mu \rangle}_{\delta_{g\mu}} = k_{\lambda\mu} \text{ if } \mu \leq \lambda \\ \text{and } = 0 \text{ otherwise}$$

Then, because $\{s_\lambda\}$ is an orthonormal basis of Λ , we can write

$$h_\mu = \sum_\lambda \langle s_\lambda, h_\mu \rangle s_\lambda = \sum_{\lambda \leq \mu} k_{\lambda\mu} s_\lambda$$

$$\Rightarrow \boxed{h_\mu = s_\mu + \sum_{\lambda > \mu} k_{\lambda\mu} s_\lambda}.$$

[Because $k_{\lambda\mu} = 0$ if $\mu > \lambda$, you can also write

$$h_\mu = \sum_\lambda k_{\lambda\mu} s_\lambda]$$