Math 249 Lecture 13 Notes

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1 Schur Functions

1.1 Definition and connection to antisymmetric functions

Last lecture, we had the antisymmetric functions

$$a_{\lambda+\rho} = \sum_{\sigma \in S_n} \varepsilon(\sigma)\sigma(x^{\lambda+\rho}) = \det \begin{bmatrix} x_1^{(\lambda+\rho)_1} & x_1^{(\lambda+\rho)_2} & \cdots & x_1^{(\lambda+\rho)_n} \\ x_2^{(\lambda+\rho)_1} & x_2^{(\lambda+\rho)_2} & \cdots & x_2^{(\lambda+\rho)_n} \\ \vdots & \vdots & & \vdots \\ x_n^{(\lambda+\rho)_1} & x_n^{(\lambda+\rho)_2} & \cdots & x_2^{(\lambda+\rho)_n} \end{bmatrix}$$

and the Vandermonde determinant

$$a_{\rho} = \det \begin{bmatrix} x_1^{n-1} & x_1^{n-2} & \cdots & 1 \\ x_2^{n-1} & x_2^{n-2} & \cdots & 1 \\ \vdots & \vdots & & \vdots \\ x_n^{n-1} & x_n^{n-2} & \cdots & 1 \end{bmatrix} = \prod_{i < j} (x_i - x_j).$$

We also had that a_{ρ} divides $a_{\lambda+\rho}$ for all λ .

Definition 1.1. The Schur functions are functions of the form

$$s_{\lambda}(x_1,\ldots,x_n) = \frac{a_{\lambda+\rho}}{a_{\rho}}.$$

We have an isomorphism of abelian groups $\Lambda(x_1,\ldots,x_n)\to \mathbb{Z}[x_1,\ldots,x_n]^{\varepsilon}$ given by $f\mapsto a_{\rho}f$. So the $\{s_{\lambda}\}$ form a \mathbb{Z} -basis of $\Lambda_{\mathbb{R}}(x_1,\ldots,x_n)$. If $|\lambda|=d$, then $\deg(s_{\lambda})=d$.

1.2 Independence of the number of variables

Remarkably, the Schur functions do not depend on the number of variables n.

Lemma 1.1. Let $\ell(\lambda)$ be the number of parts in the partition λ . Then

$$s_{\lambda}(x_1, \dots, x_{n-1}, 0) = \begin{cases} s_{\lambda}(x_1, \dots, x_{n-1}) & \ell(\lambda) < n \\ 0 & \ell(\lambda) = n \end{cases}$$

Proof. If $\ell(\lambda) = n$, then $x^{\lambda+\rho}$ is a multiple of $x_1 \cdots x_n$. So $a_{\lambda+\rho}$ is divisible by $x_1 \cdots x_n$, which makes $a_{\lambda+\rho}(x_1, \ldots, x_{n-1}, 0) = 0$.

If
$$\ell(\lambda) < n$$
,

$$a_{\lambda+\rho}(x_1,\dots,x_{n-1},0) = \det \begin{bmatrix} x_1^{\lambda_1+n-1} & x_1^{\lambda_2+n-2} & \cdots & x_1^{\lambda_n} \\ x_2^{\lambda_1+n-1} & x_2^{\lambda_2+n-2} & \cdots & x_2^{\lambda_n} \\ \vdots & \vdots & & \vdots \\ x_{n-1}^{\lambda_1+n-1} & x_{n-1}^{\lambda_2+n-2} & \cdots & x_{n-1}^{\lambda_n} \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$
$$= a_{\lambda+\rho_{n-1}+(1,\dots,1)}(x_1,\dots,x_{n-1})$$
$$= x_1 \cdots x_{n-1} a_{\lambda+\rho_{n-1}}(x_1,\dots,x_{n-1}).$$

Now note that

$$a_{\rho}(x_1, \dots, x_{n-1}, 0) = \left(\prod_{i < j \le n-1} (x_i - x_j)\right) x_1 x_2 \cdots x_{n-1} = x_1 x_2 \cdots x_{n-1} a_{\rho_{n-1}},$$

so we get

$$s_{\lambda}(x_{1}, \dots, x_{n-1}, 0) = \frac{a_{\lambda+\rho}(x_{1}, \dots, x_{n-1}, 0)}{a_{\rho}(x_{1}, \dots, x_{n-1}, 0)}$$

$$= \frac{x_{1} \dots x_{n-1} a_{\lambda+\rho_{n-1}}(x_{1}, \dots, x_{n-1})}{x_{1} x_{2} \dots x_{n-1} a_{\rho_{n-1}}}$$

$$= \frac{a_{\lambda+\rho_{n-1}}(x_{1}, \dots, x_{n-1})}{a_{\rho_{n-1}}}$$

$$= s_{\lambda}(x_{1}, \dots, x_{n-1}).$$

Corollary 1.1. The coefficient of m_{μ} in s_{λ} , $K_{\lambda,\mu} = \langle m_{\mu} \rangle s_{\lambda}$, is independent of the number of variables n (provided that $n \geq \ell(\lambda), \ell(\mu)$).

Proof. Let $s_{\lambda}(x_1, x_2, \dots) = \sum_{\mu} K_{\lambda, \mu} m_{\mu}(x_1, x_2, \dots)$. Then

$$s_{\lambda}(x_1, x_2, \dots, x_n, 0, 0, \dots) = \begin{cases} s_{\lambda}(x_1, \dots, x_n) & \ell(\lambda) \le n \\ 0 & \ell(\lambda) > n. \end{cases}$$

1.3 Relationship between e_k and s_{λ}

If f is a symmetric function, $\langle s_{\lambda} \rangle f = \langle x^{\rho+\lambda} \rangle f a_{\rho}$. What is the coefficient of s_{λ} in e_k ?

$$e_k a_\rho = e_k \sum_{\sigma \in S_n} \varepsilon(\sigma) \sigma(x_1^{n-1} x_2^{n-2} \cdots x_n^0)$$

$$= \sum_{\sigma \in S_n} \varepsilon(\sigma) \sigma(e_k x_1^{n-1} x_2^{n-2} \cdots x_n^0)$$

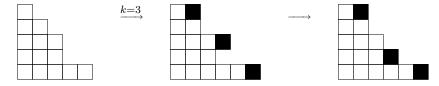
$$= a_{(1^k) + \rho}$$

because this multiplication adds 1 to each exponent of k variables in each monomial of the right term. However, if we make any two variables have the same exponent, then the coefficient of that term will be 0 because the product of a symmetric and an antisymmetric function must be antisymmetric; so we can only get the term where we added 1 to the exponents of x_1, \ldots, x_k . Since $e_k a_\rho = a_{(1^k)+\rho}$, we have that $e_k = s_{(1^k)}$.

If we want to express $e_k s_\lambda$ in terms of a sum of s_μ terms, we can show that $e_k a_{\lambda+\rho} = \sum_{\mu} c_\mu a_{\mu+\rho}$ and then divide by a_ρ on both sides. We have

$$e_k a_{\lambda+\rho} = e_k \sum_{\sigma \in S_n} \varepsilon(\sigma) \sigma(x_1^{\lambda_1+n-1} \cdots x_n^{\lambda_n}) = \sum_{\sigma \in S_n} \varepsilon(\sigma) \sigma(e_k x_1^{\lambda_1+n-1} \cdots x_n^{\lambda_n}) = \sum_{\sigma \in S_n} a_{\mu+\rho},$$

where $\mu - \lambda$ is a sum of k distinct unit vectors. We call this a *vertical* k-strip. To get one of these μ , we take the Young diagram of λ and add k boxes, at most 1 to each row.



The only terms that survive are terms where we add boxes consecutively from the bottom, i.e. terms such that μ is still a partition. You can imagine gravity pulling the black boxes down. So we get that

$$e_k s_{\lambda} = \sum_{\mu} s_{\mu},$$

where the sum is over paritions μ such that μ/λ is a vertical k-strip.

Example 1.1.

$$e_2 = s_{(1,1)}$$

$$e_{(1,1)} = e_1 e_1 = e_1 s_1 = s_{(2)} + s_{(1,1)}$$

1.4 Relationship to continuous characters

Let $\rho: \mathrm{GL}_n(\mathbb{C}) \to \mathrm{GL}_N(\mathbb{C})$ be a continuous representation; that is, $\mathrm{GL}_n(\mathbb{C}) \circlearrowleft \mathbb{C}^N$. Then the character χ is continuous because the trace map is continuous. Diagonalizable matrices are dense in $\mathrm{GL}_n(\mathbb{C})$, so the values of χ are determined by its values on diagonal matrices of eigenvalues

$$D = \begin{bmatrix} x_1 & & & \\ & x_2 & & \\ & & \ddots & \\ & & & x_n \end{bmatrix}.$$

The value of χ on these matrices is invariant under permutations of the diagonal entries, so $\chi \in \Lambda(x_1, \ldots, x_n)$. We will not prove this here, but the theory of Lie groups gives us that

$$\chi(D) = \sum_{\sigma \in S_n} \sigma\left(\frac{x^{\lambda}}{\prod_{i < j} (1 - x_j / x_i)}\right) = \sum_{\sigma \in S_n} \sigma\left(\frac{x^{\lambda + \rho}}{\prod_{i < j} (x_i - x_j)}\right) = \frac{1}{a_\rho} \sum_{\sigma \in S_n} \varepsilon(\sigma) \sigma(x^{\lambda + \rho})$$
$$= s_{\lambda}.$$