Ajeeth Gunna & Paul Zinn-Justin

Vertex models for Canonical Grothendieck polynomials and their duals


https://doi.org/10.5802/alco.235

© The author(s), 2023.

This article is licensed under the
Creative Commons Attribution 4.0 International License.
http://creativecommons.org/licenses/by/4.0/
Vertex models for Canonical Grothendieck polynomials and their duals

Ajeeth Gunna & Paul Zinn-Justin

Abstract
We study exactly solvable lattice models associated to canonical Grothendieck polynomials and their duals. We derive inversion relations and Cauchy identities.

1. Introduction
Grothendieck polynomials were introduced by Lascoux and Schützenberger in [10] as representatives of $K$-theoretic Schubert classes in flag varieties. Their connection to quantum integrability was noticed as early as [3], though it took some time to reformulate Grothendieck polynomials in the context of exactly solvable lattice models [19], where quantum integrability is most explicit. Recently, a large literature has developed around these ideas [2, 7, 13, 14, 15]. In this work we focus on symmetric Grothendieck polynomials (also called stable Grothendieck polynomials), i.e. the ones that are related to the $K$-theory of Grassmannians [1], though we expect many of our ideas to be applicable to more general (partial) flag varieties. We also consider their duals, in the sense of product/coproduct duality.\(^{(1)}\) We propose some new formulations of both Grothendieck and dual Grothendieck in terms of certain “bosonic” exactly solvable lattice models.

Let $\Lambda$ be the ring of symmetric functions. Even though the elements of $\Lambda$ are not polynomials, by abuse of language we shall refer to them as polynomials, identifying a symmetric function $F$ with the corresponding symmetric polynomial $F(x_1, \ldots, x_n)$. Schur polynomials $s_\lambda$ (where $\lambda$ runs over all partitions) form an orthonormal basis of $\Lambda$ under the Hall inner product. The involution map $\omega$, which sends $e_k$ (elementary symmetric polynomials) to $h_k$ (complete homogeneous symmetric polynomials), maps $s_\lambda$ to $s_{\lambda'}$ where $\lambda'$ is the transpose of $\lambda$. Let $\overline{\Lambda}$ be the completion of $\Lambda$, which is obtained by allowing infinite linear combinations of $s_\lambda$.

Grothendieck polynomials $G_\lambda$ are non homogeneous symmetric polynomials; with the appropriate choice of variables, $G_\lambda = s_\lambda + \text{higher order terms}$. When the number

\(^{(1)}\)These should not be confused with the dual Grothendieck polynomials that are e.g. considered in [15]. The latter are dual w.r.t. the natural scalar product of $K$-theory. In contrast, ours are dual w.r.t. the Hall inner product.
of variables grows, their degree grows, so they must be considered as elements of $\tilde{\Lambda}$. The structure constants $c_{\lambda,\mu}^{\nu}$ defined by

$$G_\lambda G_\mu = \sum_{\nu} c_{\lambda,\mu}^{\nu} G_\nu,$$

satisfy $c_{\lambda,\mu}^{\nu} = c_{\lambda,\mu}^{\nu}$ [4, Ex. 9.20]. However the image of $G_\lambda$ under $\omega$ is not $G_\lambda$. This implies that the family of polynomials $\omega(G_\lambda)$ has the same structure constants as Grothendieck polynomials. We shall not be dealing with structure constants in this paper, reserving them for subsequent work [5] and only mention them as motivation for what follows.

In [8], Lam and Pylyavskyy defined dual Grothendieck polynomials $(g_\lambda)$ as certain generating functions of reverse plane partitions. These polynomials are dual to $G_\lambda$ under the Hall inner product, and are of the form $g_\lambda = s_\lambda + \text{lower order terms}$. Similarly to Grothendieck polynomials, the image of $g_\lambda$ under the involution map is not $g_\lambda$.

In [16], Yeliussizov introduced a two parameter version of Grothendieck polynomials and their dual, which he called canonical Grothendieck polynomials and dual stable canonical Grothendieck polynomials. For more detailed combinatorial properties and definitions we refer the reader to [16]. Canonical Grothendieck polynomials and their dual satisfy the following relations:

$$\omega(G^{(\alpha,\beta)}_\lambda) = G^{(\beta,\alpha)}_{\lambda'}, \quad \omega(g^{(\alpha,\beta)}_\lambda) = g^{(\beta,\alpha)}_{\lambda'}.$$

In this paper, we shall study two types of vertex models, based on the way partitions are encoded, for both $G^{(\alpha,\beta)}_\lambda$ and $g^{(\alpha,\beta)}_\lambda$. We call a vertex model a row model (resp. column model) when the partitions are encoded by row (resp. column) multiplicities. Section 2 is devoted to the former, Section 3 to the latter. All the models studied in this paper appear to be new (see also [14]).

We then introduce (Section 4) generalised Grothendieck polynomials which are obtained by attaching additional variables to the vertical lines of the underlying lattice model. Along the process, we recover the generalised dual Grothendieck polynomials defined by Yeliussizov [18].

In Section 5, we show that the transfer matrices of these lattice models satisfy remarkable inversion relations. These show a deep connection between row and column lattice models, thus embodying the involution $\omega$ at the level of transfer matrices. This should be reminiscent of similar relations satisfied by the usual free fermionic vertex operators related to Schur functions (see e.g. [20], or [19] and references therein); indeed, our transfer matrices can be thought of as deformations of these vertex operators.

Finally, in Section 6, we show how “quantum integrability” in the form of RLL relations immediately implies the Cauchy identities

\begin{align}
(1) \quad & \sum_{\lambda} G^{(-\alpha,-\beta)}_\lambda(x_1,x_2,\ldots,x_m)g^{(\alpha,\beta)}_\lambda(y_1,y_2,\ldots,y_n) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - x_i y_j} \\
(2) \quad & \sum_{\lambda} G^{(-\beta,-\alpha)}_\lambda(x_1,x_2,\ldots,x_m)g^{(\alpha,\beta)}_\lambda(y_1,y_2,\ldots,y_n) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} (1 + x_i y_j)
\end{align}

for (generalised) Grothendieck polynomials and their duals. By specializing $\alpha = 0$ and $\beta = 1$, we recover the Cauchy identity for Grothendieck polynomials and its dual [9, 17]. Throughout this paper, $\alpha$ and $\beta$ are constants and our results hold for all values, with some intricacies for $\beta = 0$.

The appendix contains proofs of the RLL relations.
2. Row Vertex Models

2.1. Definition of Physical space. Let $V^r$ be an infinite dimensional vector space with basis indexed by collections of nonnegative integers $(m_i)_{i \in \mathbb{Z}_{>0}}$ such that only a finite number of $m_i$s are nonzero; we view it as a subspace of $\bigotimes_{i=1}^{\infty} V_i$ where each $V_i = \text{Span}([0], [1], \ldots)$ has a basis indexed by a single nonnegative integer:

$$V^r = \text{Span} \{ |m_1 \rangle \otimes |m_2 \rangle \otimes |m_3 \rangle \cdots \} \quad m_i \geq 0, \ i \geq 1.$$ (3)

We shall identify partitions with basis elements of $V^r$. Given a partition $\lambda$, which we view as a Young diagram, let $|\lambda\rangle$ be the basis vector with integers $m_i(\lambda) =$ number of rows of size $i$ of $\lambda$ (hence the superscript $r$). For example, we identify the partition $\lambda = (4, 4, 4, 3, 1)$ with the basis element $|1\rangle \otimes |0\rangle \otimes |1\rangle \otimes |3\rangle \otimes |0\rangle \ldots$ of $V^r$:

All the vertex models studied in this paper follow a general template. In order to not repeat ourselves, we shall study this model in detail and then skip the general arguments in other models.

2.2. Row vertex model for canonical Grothendieck polynomials.

2.2.1. Conventions. We use the standard diagrammatic formalism to interpret lattice models in terms of linear operators. We briefly review it here, and fix conventions.

All our lattice models are defined on some domain of the plane which consists of edges and vertices of valency 4. Edges traverse vertices to form lines, which are given a certain orientation: in all that follows, the domain is a (rectangular) region of the square lattice, so that lines can be either horizontal (also called “auxiliary” lines), in which case they are oriented left to right, or vertical (also called “physical” lines), in which case they are oriented bottom to top.

To each line is associated a vector space, and juxtaposition of lines corresponds to tensor product (the order of the factors is the order of the incoming external lines). These vector spaces come equipped with a basis labelled by the various states that edges of the lattice model carry. In our case, vertical lines are numbered $1, 2, \ldots$ from left to right, and vertical edges carry a nonnegative integer, so that to vertical line numbered $i$ we assign the vector space $V_i$ (and collectively they form the “physical space” $V^r$). Horizontal edges can carry either labels 0, 1, in which case we call the horizontal line fermionic and assign to it a space $F \cong \mathbb{C}^2$ (possibly adding a subscript to distinguish the various horizontal lines), or it can carry a nonnegative integer (bosonic line), in which case we call the corresponding vector space $W$. Graphically, when the auxiliary line is fermionic, we draw thin lines. When they are bosonic, we draw thick lines.
Finally, an important convention is that we transpose all linear operators in order to facilitate reading expressions from left to right; this means that if incoming lines at a vertex form $A \otimes B$ and outgoing lines form $C \otimes D$, then to the vertex is associated a linear operator from $C \otimes D$ to $A \otimes B$. We hope that this does not cause any confusion.

2.2.2. Definition of the $L$ matrix. In this subsection, the auxiliary line is fermionic. To every vertex we assign a (Boltzmann) weight that depends on the local configuration (i.e. states of the edges) around it. The weights are given as follows:

$$w_x \left( \begin{array}{c} d \\ a \\ c \\ b \end{array} \right) \equiv w_x(a, b, c, d) = \delta_{a+b, c+d} \begin{cases} x \frac{1-\alpha x}{1-\alpha x}, & \text{when } a = 1, \\ \frac{1+\beta x}{1-\alpha x}, & \text{when } a = 0 \text{ and } b \neq 0, \\ 1, & a, b, c, d = 0, \end{cases}$$

where $a, c \in \{0, 1\}$, and $b, d \in \mathbb{Z}_{\geq 0}$.

Let us now represent the vertices graphically with their Boltzmann weights written below them.

$$\begin{pmatrix} 0 & 0 & m & m-1 & m+1 & m \\ 0 & 0 & 0 & 1 & 0 & 1 \\ m & m & m & m & m & m \\ 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix}$$

The corresponding linear operator is the so-called $L$ matrix; it acts on $F_i \otimes V_j$, where $F_i = \text{span}\{|0\rangle, |1\rangle\}$. Let us first define annihilation and creation operators, $\phi_j$ and $\phi_j^\dagger$, acting on the $j$th factor $V_j$ of $V_r$:

$$\phi_j |m\rangle = |m-1\rangle \quad \phi_j^\dagger |m\rangle = |m+1\rangle \quad \phi_j |0\rangle = |0\rangle.$$

Then

$$L_{i,j}(x) = \frac{1}{1-\alpha x} \left( \delta_{0,m}(1-\alpha x) + (1-\delta_{0m})(1+\beta x)(1+\beta x) \phi_j^\dagger \right).$$

We shall now define dual $L$ matrices, $L^*$. We obtain $L^*$ by flipping the vertices upside down and replacing $0's$ with $1's$ and vice versa on the horizontal edges.

$$\begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & m & m & m+1 & m \\ 0 & 0 & 0 & 1 & 0 & 1 \\ m & m & m & m & m & m \\ 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix}$$

Then define $L^*$ acting on $F_i \otimes V_j$ as follows:

$$L_{i,j}^*(x) = \frac{1}{1-\alpha x} \left( \delta_{0,m}(1-\alpha x) + (1-\delta_{0m})(1+\beta x) \phi_j \right).$$

2.2.3. $R$-matrix and Yang–Baxter relations. Consider the vector spaces $F_i, F_j$ where $i < j$. Then we define a $R$-matrix which acts linearly on $F_i \otimes F_j$ as follows,

$$R_{i,j}(x_i, x_j) : |a\rangle \otimes |b\rangle \mapsto \sum_{c,d} R_{i,j}^{a\ldots d}(x_i, x_j) |c\rangle \otimes |d\rangle.$$
Graphically, we represent the entry $R_{a,c}^{b,d}$ as $a \rightarrow \leftarrow d$. We now give the $R$ matrix that underpins the integrability of the vertex model presented above. For convenience, let us represent $|0\rangle$ and $|1\rangle$ of $F$ as empty or occupied:

$$R_{ij}(x,y) = \begin{pmatrix}
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 \\
0 & 0 & 0 & 1
\end{pmatrix}_{ij} \in \text{End}(F_i \otimes F_j).$$

One recognizes this as the $R$-matrix of the five-vertex model [6] with spectral parameter $\frac{x}{1-\beta x}$. It can be obtained as a limit of the $R$ matrix of the stochastic six-vertex model where the quantum parameter is sent to 0.(2)

Together with the $L_{i,n}$ and $L_{j,n}$ matrices, $R_{ij}$ satisfies the RLL relation in End $(F_i \otimes F_j \otimes V_n)$:

$$R_{ij}(x,y)L_{i,n}(x)L_{j,n}(y) = L_{j,n}(y)L_{i,n}(x)R_{ij}(x,y) \begin{pmatrix} x & 1 & 0 \\ y & 0 & 0 \end{pmatrix} = \begin{pmatrix} x & 1 & 0 \\ y & 0 & 0 \end{pmatrix}.$$}

We skip the proof of the above equation; it is best checked by computer with symbolic calculation software.

2.2.4. Transfer matrices. We shall now build a vertex model based on the $L$-matrix above. It is convenient to depict a single row of the model as in the following picture:

$$w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) = *$$

where the $*$ on the left means that we are summing over all possible states. Even though we are considering an infinitely large row of vertices, the weight is uniquely defined. To see this, fix the labels on the top and bottom. Since there are only finitely many non zero labels on top and bottom, sufficiently far to the right the horizontal labels are constant, and we choose them to be 0s. Graphically, we show this by assigning 0 to the horizontal edge on the far right. Then, when the bottom and top labels are fixed, there is a unique configuration because of the local conservation around every vertex.

We now define the corresponding transfer matrix $T$ which acts linearly on $V^r$ as follows,

$$T(x) : |i_1\rangle \otimes |i_2\rangle \otimes \cdots \mapsto \sum_{k_1, k_2 \geq 0} w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) |k_1\rangle \otimes |k_2\rangle \otimes \cdots.$$
One can rewrite it in terms of the $L$-matrix as
\begin{equation}
T(x) = \lim_{n \to \infty} \langle \ast | L_{01}(x)L_{02}(x) \ldots L_{0n}(x) | 0 \rangle
\end{equation}
where the vector space attached to the horizontal line is labelled $0$ by the subscript, whereas the vertical lines are labelled $1, 2, \ldots$. Here $| 0 \rangle$ is the basis vector of the horizontal space, whereas $\langle \ast |$ is the sum of basis vectors of the dual of the horizontal space. The limit is entry-wise and is well-defined because of the aforementioned uniqueness of the configuration.

Similarly, we can define the dual transfer matrices $T^*$:
\begin{equation}
T^*(x) : | i_1 \rangle \otimes | i_2 \rangle \otimes \cdots \mapsto \sum_{k_1, k_2, \ldots \geq 0} w^* (\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) | k_1 \rangle \otimes | k_2 \rangle \otimes \cdots
\end{equation}
where the right boundary is fixed to be $1$:
\begin{equation}
w^* (\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) = \ast \begin{array}{cccc} k_1 & k_2 & k_3 & \cdots \\
1 & i_1 & i_2 & \ldots \end{array}.
\end{equation}
Equivalently,
\begin{equation}
T^*(x) = \lim_{n \to \infty} \langle \ast | L_{01}^*(x)L_{02}^*(x) \ldots L_{0n}^*(x) | 1 \rangle.
\end{equation}

Throughout this paper we use the same conventions to define transfer matrices.

### 2.2.5. Commutation relation of the transfer matrices

Observe that the sum of the entries in a column of the $R$ matrix is always 1. This means that the state which is the sum of all possible states is an eigenvector of the $R$ matrix with eigenvalue 1. This property can be reinterpreted as the fact that the partition function of a single vertex with fixed boundaries on the right and free boundary on the left is always 1:

\begin{equation}
\begin{array}{ccc}
x & \ast & x \\
y & \ast & = y & \ast
\end{array}
\end{equation}

Consider the product of two transfer matrices, $T(x)$ and $T(y)$. Graphically, taking the product amounts to stacking the two row to row transfer matrices one upon the other. Observe that the boundary on the left is free and for sufficiently large $n$, the boundary on the right is fixed. Recall that an edge with $\ast$ is a free boundary. Thus $T(x)T(y)$ is

\begin{equation}
y \ast x
\end{equation}

Now multiply $T(x)T(y)$ on the left by $R(x, y)$, and apply the RLL relation finitely many times:
2.2.6. Canonical Grothendieck polynomials. Given that the transfer matrices commute, the polynomials defined using them are invariant under permutation of the variables. It is also easy to see that $T(0) = 1$, so that these polynomials satisfy the stability property which makes them an element of $\tilde{\Lambda}$. We now prove that the polynomials defined using $T$ are canonical Grothendieck polynomials.

Before we prove it, let us recall the branching formula for $G(\alpha, \beta)_{\lambda}$ from [16, Proposition 8.8]. For a partition $\lambda = (\lambda_1, \lambda_2, \lambda_3, \ldots)$, denote $\bar{\lambda} = (\lambda_2, \lambda_3, \ldots)$. We have

$$G(\alpha, \beta)_{\lambda}(x_1, \ldots, x_n, x_{n+1}) = \sum_{\lambda/\mu \text{ hor. strip}} G(\alpha, \beta)_{\mu}(x_1, \ldots, x_n) G(\alpha, \beta)_{\lambda/\mu}(x_{n+1}), \quad (15)$$

and

$$G(\alpha, \beta)_{\lambda/\mu}(x) = \left( \frac{x}{1 - \alpha x} \right)^{|\lambda/\mu|} \left( \frac{1 + \beta x}{1 - \alpha x} \right)^{r(\lambda/\mu)}, \quad (16)$$

where $r(\lambda/\mu)$ is the number of non zero rows of $\lambda/\mu$. We shall take this branching formula as the definition of $G(\alpha, \beta)_{\lambda}$.

**Remark 2.1.** In order to dispel any confusion, we point out that $G_{\lambda/\mu}(x_1, \ldots, x_n)$ polynomials are not the same as skew Grothendieck polynomials $G_{\lambda/\mu}$. For a simple counter example, observe that for any partition $\lambda$, we have $G_{\lambda/\lambda}(x) = \left( \frac{1 + \beta x}{1 - \alpha x} \right)^{r(\lambda/\lambda)} \neq G_{\lambda/\lambda}(x) = 1$. **
Let us now look at an example to understand \( r(\mu/\bar{\lambda}) \). Consider the partitions \( \lambda = (4, 3, 2, 1) \) and \( \mu = (3, 2, 2, 1) \). Then \( \bar{\lambda} = (3, 2, 1) \) and \( r(\mu/\bar{\lambda}) = 2 \):

![Diagram showing partitions \( \lambda/\mu \) and \( \mu/\bar{\lambda} \)]

We can alternatively formulate \( r(\mu/\bar{\lambda}) \) as the number of removable boxes of \( \mu \) that do not lie in the same column as any box of \( \lambda/\mu \).

As a consequence of recording partitions with row multiplicities, every vertex with a non-zero label on the bottom edge corresponds to a removable box of \( \mu \). If a box is added to the \( i \)-th column of \( \mu \), then the removable box corresponding to that vertex at site \( i \) will be in the same column as the new box. So \( r(\mu/\bar{\lambda}) \) is precisely the number of vertices with zero label on the left edge and a non-zero label on the bottom edge.

**Theorem 2.2.** The canonical Grothendieck polynomials \( G_{\lambda}^{(\alpha, \beta)}(x) \) are given by

\[
\begin{align*}
G_{\lambda}^{(\alpha, \beta)}(x_1, \ldots, x_n) &= \langle 0 | T(x_1) \cdots T(x_n) | \lambda \rangle \\
G_{\lambda}^{(\alpha, \beta)}(x_1, \ldots, x_n) &= \langle \lambda | T^*(x_n) \cdots T^*(x_1) | 0 \rangle
\end{align*}
\]

where \( | \lambda \rangle = \bigotimes_{i=1}^{\infty} |m_i(\lambda)\rangle \), and similarly for the dual state \( \langle \lambda \rangle \).

**Proof.** We shall prove (17), and (18) follows immediately as a consequence of the way we defined the \( L^* \) matrix. Fix \( \lambda \), then we can just consider the finite transfer matrix of size \( \lambda_1 \). Inserting a complete set of states before the final transfer matrix, we have the following branching formula

\[
G_{\lambda}^{(\alpha, \beta)}(x_1, \ldots, x_n, x) = \sum_{\mu} \langle 0 | T(x_1) \cdots T(x_n) | \mu \rangle \langle \mu | T(x) | \lambda \rangle.
\]

On comparing the branching formula for \( G_{\lambda}^{(\alpha, \beta)} \) (eq. (15)), it is enough to show \( G_{\lambda/\mu}^{(\alpha, \beta)}(x) = \langle \mu | T(x) | \lambda \rangle \). Recall that for a horizontal strip \( \lambda/\mu \), we have

\[
G_{\lambda/\mu}^{(\alpha, \beta)}(x) = \left( \frac{x}{1 - \alpha x} \right)^{|\lambda/\mu|} \left( \frac{1 + \beta x}{1 - \alpha x} \right)^{r(\mu/\bar{\lambda})}.
\]

Based on the vertices that we used to define \( T \), one easily observes that \( \langle \mu | T(x) | \lambda \rangle \neq 0 \) if and only if \( \lambda/\mu \) is a horizontal strip. The label 1 on the left edge at site \( i \) amounts to adding a box in the \( i \)-th column from the left. For every such vertex, we get a factor of \( \frac{x}{1 - \alpha x} \). From our previous analysis, we see that \( r(\mu/\bar{\lambda}) \) is exactly the number of vertices with the label 0 on the left edge and a non-zero label on the bottom edge, where each such vertex has a weight of \( \frac{1 + \beta x}{1 - \alpha x} \). \( \square \)
Example 2.3. For partition $\lambda = (1,0)$ we have the following two possible configurations on the left, each with a unique configuration on the interior.

\[
\begin{array}{c|c|c|c|c|}
\hline
x_2 & 0 & 1 & 0 & 0 \\
\hline
x_1 & 0 & 1 & 0 & 0 \\
\hline
\end{array}
\quad
\begin{array}{c|c|c|c|c|}
\hline
1 & 0 & 0 & 0 & 0 \\
\hline
0 & 0 & 0 & 0 & 0 \\
\hline
\end{array}
\]

Therefore,

\[
G_\lambda^{(\alpha,\beta)}(x_1, x_2) = \left( \frac{x_1}{1 - \alpha x_1} \right) \left( \frac{1 + \beta x_2}{1 - \alpha x_2} \right) + \left( \frac{x_2}{1 - \alpha x_2} \right).
\]

Example 2.4. For partition $\lambda = (2,0)$, we have the following configurations.

\[
\begin{array}{c|c|c|c|c|}
\hline
x_2 & 0 & 0 & 1 & 0 \\
\hline
x_1 & 1 & 0 & 0 & 0 \\
\hline
\end{array}
\quad
\begin{array}{c|c|c|c|c|}
\hline
1 & 0 & 0 & 0 & 0 \\
\hline
0 & 0 & 0 & 0 & 0 \\
\hline
\end{array}
\quad
\begin{array}{c|c|c|c|c|}
\hline
1 & 0 & 0 & 0 & 0 \\
\hline
0 & 0 & 0 & 0 & 0 \\
\hline
\end{array}
\]

\[
G_\lambda^{(\alpha,\beta)}(x_1, x_2) = \left( \frac{x_1}{1 - \alpha x_1} \right)^2 \left( \frac{1 + \beta x_2}{1 - \alpha x_2} \right) + \left( \frac{x_1}{1 - \alpha x_1} \right) \left( \frac{x_2}{1 - \alpha x_2} \right) \left( \frac{1 + \beta x_2}{1 - \alpha x_2} \right) + \left( \frac{x_2}{1 - \alpha x_2} \right)^2
\]

Example 2.5. For the partition $\lambda = (1,1)$, there is a unique configuration, where the overall weight is the polynomial $G_\lambda^{(\alpha,\beta)}(x_1, x_2)$.

\[
\begin{array}{c|c|c|c|c|}
\hline
x_2 & 1 & 0 & 0 & 0 \\
\hline
x_1 & 0 & 1 & 0 & 0 \\
\hline
\end{array}
\]

\[
G_\lambda^{(\alpha,\beta)}(x_1, x_2) = \left( \frac{x_1}{1 - \alpha x_1} \right) \left( \frac{x_2}{1 - \alpha x_2} \right)
\]

2.3. Row vertex model for dual canonical Grothendieck polynomials. In this section, we consider a similar vertex model as the one introduced in Section 2.2, but with a bosonic auxiliary line. This means that that we shall associate an infinite dimensional vector space to the values a horizontal line can carry. The Boltzmann weights of the vertices are the following:

\[
(19) \quad w_x \left( \begin{array}{c}
d \\
b \\
c \\
a \\
\end{array} \right) \equiv w_x(a, b; c, d) = \delta_{a+b, c+d} \begin{cases} 
(\alpha + \beta)^{n-d-1}(x + \alpha)\beta^d & a > d, \\
\beta^{n-1}x & 0 < a \leq d, \\
1 & a = 0,
\end{cases}
\]

where $a, b, c, d \in \mathbb{Z}_{\geq 0}$.

Let $W = \text{Span}\{ |j\rangle \}_{j \in \mathbb{Z}_n}$ be an infinite dimensional vector space, and for $1 \leq i \leq n$, let $W_i$ be a copy of $W$. Then we define an $l$ matrix which acts linearly on $W_i \otimes V_j$ as
follows:

\[ l_{i,j}(x_i) : |a\rangle \otimes |b\rangle \mapsto \sum_{c,d \text{ where } a+b=c+d} w_{x_i}(a,b;c,d) |c\rangle \otimes |d\rangle . \]  

Let \( w(\{i_1, i_2, \ldots \}; \{k_1, k_2, \ldots \}) \) be the weight of single row of vertices.

\[ w(\{i_1, i_2, \ldots \}; \{k_1, k_2, \ldots \}) = \ast \begin{array}{ccc}
    k_1 & k_2 & k_3 \\
    t_1 & t_2 & t_3 \\
    \vdots & \ddots & \ddots \\
    \end{array} 0 . \]

We define the transfer matrix \( t \) which acts linearly on \( V^r \) as follows:

\[ t(x) : |i_1\rangle \otimes |i_2\rangle \otimes \cdots \mapsto \sum_{k_1, k_2, \ldots \geq 0} w(\{i_1, i_2, \ldots \}; \{k_1, k_2, \ldots \}) |k_1\rangle \otimes |k_2\rangle \otimes \cdots . \]

As the horizontal lines are bosonic, we represent the \( r \)-matrix as a cross of thick lines \( \begin{array}{c}
    x \\
    y \\
    \end{array} \). Consider the vector spaces \( W_i, W_j \) where \( i < j \). Define an \( r \)-matrix which acts linearly on \( W_i \otimes W_j \) as follows:

\[ r_{i,j}(x_i, x_j) : |a\rangle \otimes |b\rangle \mapsto \sum_{c,d \text{ where } a+b=c+d} r_{b,c}^{a,d}(x_i, x_j) |c\rangle \otimes |d\rangle . \]

where the entries of \( r \)-matrix here are the following:

\[ r_{b,c}^{a,d}(x,y) = \begin{cases}
    0 & b > c \\
    1 & b = c = 0 \\
    \frac{y}{x} & b = c > 0 \\
    \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-d-1} & b = 0, c \neq 0 \\
    \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-d-1} \left(\frac{y}{\beta}\right)^b & b > 0, b < c .
\end{cases} \]

Together with matrices \( l_{i,n} \) and \( l_{j,n} \), \( r_{i,j} \) satisfies the RLL relation in \( \text{End} (W_i \otimes W_j \otimes V_n) \) (see Appendix A.2):

\[ r_{i,j}(x,y)l_{i,n}(x)l_{j,n}(y) = l_{j,n}(y)l_{i,n}(x)r_{i,j}(x,y) \]

\[ \begin{array}{c}
    x \\
    y \\
    \end{array} \begin{array}{c}
    x \\
    y \\
    \end{array} = \begin{array}{c}
    x \\
    y \\
    \end{array} \begin{array}{c}
    x \\
    y \\
    \end{array} . \]

Remark 2.6. Observe that the \( r \) matrix is not defined at \( \beta = 0 \). However, the weights (19) for the \( l \)-matrix force all horizontal labels to be 0 or 1 as \( \beta \) is sent to zero, and such entries of the \( r \)-matrix are well defined. We shall also study a different model for \( g_{\lambda}^{(\alpha, \beta)} \) polynomials when \( \beta = 0 \) in Section 3.4.

2.3.1. Eigenvector of the \( r \)-matrix. We proceed as in the previous section, showing the state which is sum of all the possible states is an eigenvector of the \( r \)-matrix.

We show this by computing the partition function of a single vertex with fixed right
Vertex models for Canonical Grothendieck polynomials and their duals

boundary and free left boundary:

\[ Z(c, d) = \begin{array}{c}
\ast \\
\ast \\
c
\end{array} \begin{array}{c}
d \\
\ast \\
c
\end{array} \]

(where \( c, d \) are non negative integers) is constant and equal to 1.

We compute:

\[ Z(c, d) = \sum_{i=0}^{c} c + d - i \begin{array}{c}
d \\
i \\
c
\end{array} = (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-1} + \sum_{i=1}^{c-1} (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-i-1} \left( \frac{y}{\beta} \right)^i + \frac{y}{x} \]

\[ = (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-1} + \frac{y}{\beta} \left( \frac{1 - \frac{y}{x}}{1 - \frac{y}{\beta}} \right) \sum_{i=1}^{c-1} \left( 1 - \frac{y}{\beta} \right)^i + \frac{y}{x} \]

\[ = (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-1} + \frac{y}{\beta} \left( \frac{1 - \frac{y}{x}}{1 - \frac{y}{\beta}} \right) \frac{1 - (1 - \frac{y}{\beta})^{c-1}}{1 - (1 - \frac{y}{\beta})} + \frac{y}{x} \]

\[ = (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-1} + (1 - \frac{y}{x}) \left( 1 - (1 - \frac{y}{\beta})^{c-1} \right) + \frac{y}{x} \]

\[ = (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-1} + (1 - \frac{y}{x}) - (1 - \frac{y}{x}) (1 - \frac{y}{\beta})^{c-1} + \frac{y}{x} \]

\[ = 1. \]

By repeating the same argument as in Section 2.2.5, we get the commutation relation of the transfer matrices,

\[ t(x)t(y) = t(y)t(x). \]

Therefore, the polynomials defined using \( t \) are invariant under permutation of variables.

2.3.2. Canonical dual Grothendieck polynomials. In order to formulate the branching formula for \( g^{(\alpha, \beta)}_{\lambda} \), we need to establish some statistics on partitions.

For a skew-partition \( \lambda/\mu \), define

\[ r(\lambda/\mu) = \text{number of non zero rows}, \]
\[ c(\lambda/\mu) = \text{number of non zero columns}, \]
\[ b(\lambda/\mu) = \text{number of connected components}. \]
Let us now recall the branching formula of $g^{(\alpha,\beta)}_\lambda$ from [16, Theorem 8.6]. For $\lambda, \mu$, we have

$$g^{(\alpha,\beta)}_\lambda(x_1, \ldots, x_n, x_{n+1}) = \sum_{\mu \subseteq \lambda} g^{(\alpha,\beta)}_\mu(x_1, \ldots, x_n) g^{(\alpha,\beta)}_{\lambda/\mu}(x_{n+1}),$$

where

$$g^{(\alpha,\beta)}_{\lambda/\mu}(x) = \beta^{r(\lambda/\mu) - b(\lambda/\mu)} (\alpha + \beta)^{c(\lambda/\mu) - r(\lambda/\mu)} c(\lambda/\mu) b(\lambda/\mu) (x + x)^{b(\lambda/\mu) - b(\lambda/\mu)}$$

whenever $\mu \subseteq \lambda$ and 0 otherwise.

We shall use this branching formula as the definition of $g^{(\alpha,\beta)}_\lambda$. Let us compute some examples to understand the above statistics.

\begin{align*}
\lambda/\mu &= (4,3,2,1)/(2,2,2) & \lambda/\mu &= (4,3,3,2)/(2,2,2) & \lambda/\mu &= (4,3,3,3,2)/(2,2,2) \\
r(\lambda/\mu) &= 4 & r(\lambda/\mu) &= 5 & r(\lambda/\mu) &= 5 \\
c(\lambda/\mu) &= 4 & c(\lambda/\mu) &= 4 & c(\lambda/\mu) &= 4 \\
b(\lambda/\mu) &= 2 & b(\lambda/\mu) &= 2 & b(\lambda/\mu) &= 1 \\
\end{align*}

Let us unpack the information contained at a vertex. Consider a vertex \( \begin{pmatrix} a & b \\ c & d \end{pmatrix} \) at site \( i \). The label on the left edge \( a \) corresponds to adding \( a \) boxes to the \( i^{th} \) column of \( \mu \). The label \( d \) is the number rows of \( \lambda \) with size \( i \). We want to understand number of row of size \( i \) in \( \lambda/\mu \). There are three types of vertices, \( b < c \), \( b > d \), and \( b = c \). Let us look at the labels on the \( i^{th} \) column of \( \lambda/\mu \) in terms of the Young diagrams.

case: \( b < c \) case: \( b > c \) case: \( b = c \)
From the above pictures, it is evident that the number of non zero rows of size $i$ in $\lambda/\mu$ is $\min(a,d)$. Also observe that, in the last two cases, the skew diagram is disjoint. Therefore, the number of connected components is the number of vertices where $b > c$ or $b = c$. Finally, the $i^{th}$ column of $\lambda/\mu$ is non zero if and only if some boxes are added to it, i.e. when $a \neq 0$.

**Theorem 2.7.** The dual canonical Grothendieck polynomials $g^{(\alpha,\beta)}_\lambda(x)$ are given by

$$g^{(\alpha,\beta)}_\lambda(x_1,\ldots,x_n) = \langle 0| t(x_1) \ldots t(x_n) | \lambda \rangle$$

where $|\lambda\rangle = \bigotimes_{i=1}^{\infty} |m_i(\lambda)\rangle$.

**Proof.** On comparing with the branching formula (25), it enough to show that for $\mu \subseteq \lambda$,

$$g^{(\alpha,\beta)}_{\lambda/\mu}(x) = \langle \mu| t(x) | \lambda \rangle.$$ 

Recall that for $\mu \subseteq \lambda$, we have

$$g^{(\alpha,\beta)}_{\lambda/\mu}(x) = x^{r(\lambda/\mu) + b(\lambda/\mu)} \frac{(\alpha + \beta)^{\lambda/\mu - r(\lambda/\mu) - c(\lambda/\mu) + b(\lambda/\mu)}}{(\alpha + x)^{\lambda/\mu}}.$$ 

Let us study the exponent of $\beta$ in $\langle \mu| t(x) | \lambda \rangle$. Recall that the number of rows of size $i$ in $\lambda/\mu$ is $\min(a,d)$. The connected components are recorded by vertices where $b \geq c$. Therefore, by assigning the weight $\beta^d$ when $b < c$ and $\beta^{a-1}$ whenever $b \geq c$, we get the exponent of $\beta$ in the overall weight as $r(\lambda/\mu) - b(\lambda/\mu)$, which is precisely the exponent of $\beta$ in $g^{(\alpha,\beta)}_\lambda(x)$. Similarly, by doing the same for the other factors, one recovers the Boltzmann weights. □

**Example 2.8.** For the partition $\lambda = (1,0)$ we have the following two configurations corresponding to $g^{(\alpha,\beta)}_{1}(x_1,x_2) = x_1 + x_2$.

**Example 2.9.** For the partition $\lambda = (2,0)$, we have

$$g^{(\alpha,\beta)}_{2}(x_1,x_2) = (x_1 + \alpha)x_1 + x_1x_2 + (x_2 + \alpha)x_2 = x_1^2 + x_1x_2 + x_2^2 + \alpha(x_1 + x_2).$$

**Example 2.10.** For the partition $\lambda = (1,1)$, we have
\[ g^{(\alpha,\beta)}_{\lambda}(x_1, x_2) = \beta x_1 + x_1x_2 + \beta x_2 = x_1x_2 + \beta(x_1 + x_2). \]

**Example 2.11.** For the partition \(\lambda = (2, 1)\), we have

\[ g^{(\alpha,\beta)}_{\lambda}(x_1, x_2) = \beta x_1x_2 + x_1x_2^2 + (x_1 + \alpha)x_1x_2 + (x_2 + \alpha)\beta x_2 + (x_1 + \alpha)\beta x_1. \]

### 3. Column Vertex Models

#### 3.1. Definition of Physical space.
Recall that we identify partitions with basis elements of \(V^r\) by recording row multiplicities. In this section, for the physical space, we use the same vector space \((V^r)\) that we used in the earlier section. We shall denote it by \(V^c\). Even though \(V^c\) and \(V^r\) are identical, we distinguish them by the way we identify the partitions with the basis elements.

\[ V^c = \text{Span}\{ |m_1^c\rangle \otimes |m_2^c\rangle \otimes |m_3^c\rangle \cdots \} \quad m_i^c \geq 0, \ i \geq 1. \]

Given a partition, which we view as Young diagram, let \(|\lambda^c\rangle\) be the basis vector with integers \(m_i^c(\lambda) = \text{number of columns of size } i \text{ of } \lambda\).

For example, we identify the partition \(\lambda = (5, 4, 4, 3)\) with the basis element \(|1\rangle \otimes |0\rangle \otimes |1\rangle \otimes |3\rangle \otimes |0\rangle \cdots \) of \(V^c\):

\[ m_1^c(\lambda) = m_2^c(\lambda) = m_3^c(\lambda) = m_5^c(\lambda) = \cdots \]

It is useful to note that \(m_i^c(\lambda) = m_i^r(\lambda')\).

#### 3.2. Column vertex model for canonical Grothendieck polynomials.

**3.2.1. Definition of \(\tilde{L}\)-matrix and \(\tilde{R}\) matrix.** The main difference of the model considered in this section from the row model of \(G^{(\alpha,\beta)}_{\lambda}\) is that the horizontal line can now carry any non-negative integer. For every vertex, we assign the Boltzmann weights in the following way:

\[ w_x \begin{pmatrix} a & d \\ b & c \end{pmatrix} \equiv w_x(a, b; c, d) = \delta_{a+b, c+d} \begin{cases} \left(\frac{x}{1-\alpha x}\right)^a & b = c \\ \left(\frac{x}{1-\alpha x}\right)^a \left(\frac{1+\beta x}{1-\alpha x}\right)^b & b > c \\ 0 & b < c \end{cases} \]
where \(a, b, c, d\) are non-negative integers. Let \(W = \text{Span}\{j\}_{j \in \mathbb{Z}_{\geq 0}}\) be an infinite dimensional vector space, and for \(1 \leq i \leq n\), let \(W_i\) be a copy of \(W\). Let \(V_j \cong W\) be another copy. Then we define a \(\tilde{L}\) matrix which acts linearly on \(W_i \otimes V_j\) as follows:

\[
\tilde{L}_{i,j}(x_i) : \langle a \rangle \otimes \langle b \rangle \mapsto \sum_{c,d \text{ where } a+b=c+d} w_{x_i}(a, b; c, d) \langle c \rangle \otimes \langle d \rangle.
\]

Let \(w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\})\) be the weight of single row of vertices:

\[
w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) = \star \begin{array}{ccccccc}
k_1 & k_2 & k_3 & \ldots & \end{array} \begin{array}{ccccccc}
i_1 & i_2 & i_3 & \ldots & 0 \end{array}.
\]

We now define the transfer matrix \(\tilde{T}\) which acts linearly on \(V^n\) as follows,

\[
\tilde{T}(x) : \langle i_1 \rangle \otimes \langle i_2 \rangle \otimes \cdots \mapsto \sum_{k_1, k_2 \geq 0} w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) \langle k_1 \rangle \otimes \langle k_2 \rangle \otimes \cdots.
\]

Consider the vector spaces \(W_i, W_j\) where \(i < j\). Then we define an \(R\)-matrix which acts linearly on \(W_i \otimes W_j\) as follows,

\[
\tilde{R}_{i,j}(x_i, x_j) : \langle a \rangle \otimes \langle b \rangle \mapsto \sum_{c,d \text{ where } a+b=c+d} \tilde{R}_{i,j}^{ad}(x_i, x_j) \langle c \rangle \otimes \langle d \rangle,
\]

where the entries are:

\[
\tilde{R}_{i,j}^{ad}(x_i, x_j) = \begin{cases} a & \text{ when } b < c \\ b & \text{ when } b = c \\ c & \text{ otherwise.} \end{cases}
\]

Together with \(\tilde{L}_{i,n}\) and \(\tilde{L}_{j,n}\) matrices, \(\tilde{R}_{ij}\) satisfies the RLL relation in \(\text{End}(W_i \otimes W_j \otimes V_n)\) (see Appendix A.1):

\[
\tilde{R}_{ij}(x, y) \tilde{L}_{i,n}(x) \tilde{L}_{j,n}(y) = \tilde{L}_{j,n}(y) \tilde{L}_{i,n}(x) \tilde{R}_{ij}(x, y) \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 1-\alpha x \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}.
\]

3.2.2. Eigenvector of the \(\tilde{R}\) matrix. As in the previous section, we show that the partition function with single vertex and fixed right boundary and free boundary condition on the left

\[
Z(c, d) = \star \begin{array}{ccc}
d & \end{array} \begin{array}{ccc}
c & \end{array}.
\]
(where \(c, d\) are nonnegative integers) is constant and equal to 1. We compute:

\[
Z(c, d) = \sum_{i=0}^{d} c + d - i \sum_{i=0}^{d-1} \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)^{i} \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)^{d-i}
\]

\[
= \left( \frac{1}{1 - \alpha x} - \frac{x}{y(1 - \alpha x)} \right) \sum_{i=0}^{d-1} \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)^{i} \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)^{d-i}
\]

\[
= \left( \frac{y - x}{y(1 - \alpha x)} \right) \left( \frac{1 - \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)^{d-i}}{1 - \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)} \right) \left( \frac{x(1 - \alpha y)}{y(1 - \alpha x)} \right)^{d-i}
\]

\[
= 1.
\]

By repeating the same argument as in Section 2.2.5, we get the commutation relation of the transfer matrices:

\[\tilde{T}(x)\tilde{T}(y) = \tilde{T}(y)\tilde{T}(x)\]

Therefore, the polynomials defined using \(t\) are invariant under permutation of variables.

3.2.3. Canonical Grothendieck polynomials. Given that the transfer matrices commute, the polynomials defined using \(\tilde{T}\) are invariant under permutation of variables. We now prove that the polynomials defined using \(T\) are canonical Grothendieck polynomials.

**Theorem 3.1.** The canonical Grothendieck polynomials \(G^{(\alpha, \beta)}(x)\) are given by

\[
G^{(\alpha, \beta)}_{\lambda}(x_1, \ldots, x_n) = \langle 0 | \tilde{T}(x_1) \cdots \tilde{T}(x_n) | \lambda^c \rangle
\]

where \(|\lambda^c\rangle = \bigotimes_{i=1}^{\infty} |m^c_i(\lambda)\rangle\).

**Example 3.2.** Let us observe some examples to understand \(r(\mu/\bar{\lambda})\).

\[
\begin{array}{ccc}
\lambda/\mu = (5,3,1)/(4,2) & \lambda/\mu = (5,4,1)/(4,2) & \lambda/\mu = (5,4,2)/(4,2) \\
\begin{array}{c|c|c|c|c}
\mu_1 & \mu_2 & \mu_3 & \mu_4 \\
1 & 1 & 1 & 0 \\
2 & 2 & 0 & 0 \\
\end{array} & \begin{array}{c|c|c|c|c}
\lambda_1 & \lambda_2 & \lambda_3 & \lambda_4 \\
1 & 2 & 1 & 0 \\
1 & 2 & 0 & 0 \\
\end{array} & \begin{array}{c|c|c|c|c}
\lambda_1 & \lambda_2 & \lambda_3 & \lambda_4 \\
1 & 2 & 2 & 0 \\
1 & 2 & 0 & 0 \\
\end{array}
\end{array}
\]

\(r(\mu/\bar{\lambda}) = 2\) \hspace{1cm} \(r(\mu/\bar{\lambda}) = 1\) \hspace{1cm} \(r(\mu/\bar{\lambda}) = 0\)

**Proof.** Let us now understand the local configuration of vertices of this model. Consider a vertex \(a \longrightarrow b \longrightarrow c \) at site \(i\). The label \(a\) corresponds to adding \(a\) boxes to \(i^{th}\) row of \(\mu\). By recording the left nodes, we get \(\lambda/\mu\). Then, in-order to get a horizontal strip, the number of boxes that can be added to \(i^{th}\) row should be at most \(b\).
When $c < b$, we have a removable box in the $i^{th}$ row that is not in the same column with any box of $\lambda/\mu$. Therefore, $r(\mu/\lambda)$ is precisely the number of vertices where $c < b$.

Following the reasoning in Theorem 2.2, it is enough to show that \((u^c \bar{T}(x) | \lambda^c) = G_{\lambda \mu}^{(c,\beta)}\) for a horizontal strip $\lambda/\mu$. Recall that for a horizontal strip, we have

\[
G_{\lambda \mu}^{(c,\beta)}(x) = \left( \frac{x}{1 - \alpha x} \right)^{\lambda/\mu} \left( \frac{1 + \beta x}{1 - \alpha x} \right)^{r(\mu/\lambda)}.
\]

Observe that \((u^c \bar{T}(x) | \lambda^c) \neq 0\) if and only if $\lambda/\mu$ is a horizontal strip. From the above analysis and the way Boltzmann weights are defined, the proof is now immediate. □

**Example 3.3.** For the partition $\lambda = (2, 0)$, we have

\[
\begin{array}{ccc}
  x_2 & 0 & 1 \\
  x_1 & 1 & 0 \\
\end{array}
\]

\[
G_{\lambda}^{(c,\beta)}(x_1, x_2) = \left( \frac{x_1}{1 - \alpha x_1} \right)^2 \left( \frac{1 + \beta x_2}{1 - \alpha x_2} \right) = \left( \frac{x_1}{1 - \alpha x_1} \right) \left( \frac{x_2}{1 - \alpha x_2} \right).
\]

**Example 3.4.** For the partition $\lambda = (1, 1)$, we have

\[
\begin{array}{ccc}
  x_2 & 0 & 1 \\
  x_1 & 1 & 0 \\
\end{array}
\]

\[
G_{\lambda}^{(c,\beta)}(x_1, x_2) = \left( \frac{x_1}{1 - \alpha x_1} \right) \left( \frac{x_2}{1 - \alpha x_2} \right).
\]

### 3.3. Column vertex model dual canonical Grothendieck polynomials.

**3.3.1. Definition of $\tilde{I}$-matrix and $\tilde{T}$-matrix.** We consider the same vertex model as row model of $G_{\lambda}^{(c,\beta)}$, but with different Boltzmann weights. For every vertex, we assign the Boltzmann weights in the following way:

\[
w_x \left( \begin{array}{c} a \\ d \\ b \\ c \end{array} \right) \equiv w_x(a, b, c, d) = \delta_{a+b+c+d} \begin{cases} (\alpha + \beta)^{a-d-1} \beta(x + \alpha)^d & 0 < a > d \\ x(x + \alpha)^{a-1} & 0 < a \leq d \\ 1 & a = 0, \end{cases}
\]

where $a, b, c, d \in \mathbb{Z}_{\geq 0}$. Let $W = \text{Span}\{[j]\}_{j \in \mathbb{Z}_{\geq 0}}$ be an infinite dimensional vector space, and for $1 \leq i \leq n$, let $W_i$ be a copy of $W$. Let $V_j \cong W_i$ be a vector space. Then we define a $\tilde{I}$-matrix which acts linearly on $W_i \otimes V_j$ as follows,

\[
\tilde{I}_{i,j}(x_i) : [a] \otimes [b] \mapsto \sum_{c,d \text{ where } a+b+c+d} w_{x_i}(a, b, c, d) [c] \otimes [d].
\]
As usual, let \( w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) \) be the weight of single row of vertices.

\[
w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) = \begin{array}{c}
\k_1 \\
\k_2 \\
\vdots \\
0
\end{array} \begin{array}{c}
i_1 \\
i_2 \\
\vdots
\end{array}
\]

We now define the transfer matrix \( t \) which acts linearly on \( V^c \) as follows:

\[
\tilde{t}(x) : |i_1 \rangle \otimes |i_2 \rangle \otimes \cdots \mapsto \sum_{k_1, k_2, \ldots \geq 0} w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) |k_1 \rangle \otimes |k_2 \rangle \otimes \cdots.
\]

Consider the vector spaces \( W_i, W_j \) where \( i < j \). Then we define a \( \tilde{r} \)-matrix which acts linearly on \( W_i \otimes W_j \) as follows:

\[
\tilde{r}_{ij}(x_i, x_j) : |a \rangle \otimes |b \rangle \mapsto \sum_{c, d \text{ where } a + b = c + d} \tilde{r}_{b,c}^{a,d}(x_i, x_j) |c \rangle \otimes |d \rangle.
\]

Where the entries are the following:

\[
\tilde{r}_{i,j}^{k,l}(x, y) = \begin{cases}
0 & i < j \text{ and } k = l = 0 \\
1 & k = l > 0 \\
\left( \frac{x}{y} \right)^{1-k} & k = 0 \\
\left( \frac{x}{y} \right)^{-k} \left( \frac{y + \alpha}{x + \alpha} \right)^{-k} & k > 0
\end{cases}
\]

Together with matrices \( \tilde{l}_{i,n} \) and \( \tilde{l}_{j,n} \), \( \tilde{r}_{ij} \) satisfies the RLL relation in \( \text{End} (W_i \otimes W_j \otimes V_n) \) (see Appendix A.3).

\[
\tilde{r}_{ij}(x, y) \tilde{l}_{i,n}(x) \tilde{l}_{j,n}(y) = \tilde{l}_{j,n}(y) \tilde{l}_{i,n}(x) \tilde{r}_{ij}(x, y)
\]

3.3.2. Eigenvector of the \( \tilde{r} \) matrix. We proceed as in previous sections, computing the partition function of a single vertex with fixed right boundary with the Boltzmann weights of \( \tilde{r} \) matrix. We claim that for any non-negative integers \( c, d \),

\[
Z(c, d) = \begin{array}{c}
\bullet \\
\bullet \\
\bullet
\end{array} \begin{array}{c}
d \\
c
\end{array}
\]

the partition function is constant and is equal to 1. Let us first consider the case where \( d = 0 \). Then there is a unique vertex as the bottom left entry should be greater than or equal to \( c \) and also should satisfy the conservation. The weight of the unique configuration is 1.

\[
Z(c, 0) = \begin{array}{c}
0 \\
0
\end{array} \begin{array}{c}
c \\
c
\end{array}
\]
We now compute for the case where \( d > 0 \):

\[
Z(c, d) = \sum_{i=0}^{d} i \binom{c}{d+i} + \sum_{i=1}^{d-1} i \binom{c}{d-i} + \left(1 - \frac{x}{y}\right) + \frac{x}{y} \left(1 - \frac{x + \alpha}{y + \alpha}\right) + \frac{x}{y} \left(1 - \frac{x + \alpha}{y + \alpha}\right) + \frac{x}{y} \left(1 - \frac{x + \alpha}{y + \alpha}\right)
\]

\[
= 1.
\]

Using the argument in Section 2.2.5, we get the commutation relation of the transfer matrices,

\[
\tilde{t}(x)\tilde{t}(y) = \tilde{t}(y)\tilde{t}(x).
\]

Therefore, the polynomials defined using \( \tilde{t} \) are invariant under permutation of variables.

3.3.3. Dual canonical Grothendieck polynomials. Recall that for a skew-partition \( \lambda/\mu \), we have

\[
\begin{align*}
\quad r(\lambda/\mu) &= \text{number of non zero rows}, \\
\quad c(\lambda/\mu) &= \text{number of non zero columns}, \\
\quad b(\lambda/\mu) &= \text{number of connected components}.
\end{align*}
\]

We shall unpack the information contained at a vertex like we did in the case of row model of \( g^{(\alpha, \beta)}_{\lambda} \). Consider a vertex \( \begin{array}{c} \uparrow \downarrow \\
\alpha \\
\downarrow \uparrow \end{array} \) at site \( i \). The label on the left edge \( a \), corresponds to adding \( a \) boxes to the \( i^{th} \) row of \( \mu \). The label \( d \), is the number of columns of \( \lambda \) with size \( i \). We want to understand number of columns of size \( i \) in \( \lambda/\mu \). There are three types of vertices, \( b < c \), \( b > d \), and \( b = c \). Let us look at the labels on \( i^{th} \) row of \( \lambda/\mu \) in-terms of the Young diagram.

**case: \( b < c \)  

**case: \( b > c \)  

**case: \( b = c \)

It is evident from the pictures that the number of non zero columns of size \( i \) in \( \lambda/\mu \) is \( \min(a, d) \). Also, observe that the vertices where \( b \leq c \) detect the number of
connected components. The number of non empty rows in $\lambda/\mu$ is equal to the number of vertices where $a \neq 0$.

**Theorem 3.5.** The dual canonical Grothendieck polynomials $g_{\lambda}^{(\alpha,\beta)}(x)$ are given by
\begin{equation}
(41)
 g_{\lambda}^{(\alpha,\beta)}(x_{1},\ldots,x_{n}) = \langle 0 | \vec{l}(x_{1}) \cdots \vec{l}(x_{n}) | \lambda^{c} \rangle
\end{equation}
where $| \lambda^{c} \rangle = \bigotimes_{i=1}^{\infty} | m^{i}_{c}(\lambda) \rangle$.

**Proof.** Following the reasoning as in Theorem 2.2, it enough to show that for $\mu \subseteq \lambda$,
\[ g_{\lambda/\mu}^{(\alpha,\beta)}(x) = \langle \mu | t(x) | \lambda^{c} \rangle. \]
Recall that for $\mu \subseteq \lambda$, we have
\[ g_{\lambda/\mu}^{(\alpha,\beta)}(x) = \beta r(\lambda/\mu)-b(\lambda/\mu)(\alpha + \beta)^{\lambda/\mu -r(\lambda/\mu) -c(\lambda/\mu) +b(\lambda/\mu) \cdot x_{\lambda/\mu}}(\alpha + x)^{\lambda/\mu -b(\lambda/\mu)}. \]
Let us deal the $\beta$ factor in in $\langle \mu | \vec{l}(x) | \lambda \rangle$. Observe that the $\beta$ appears in a Boltzmann weight of a vertex only when $a \neq 0$ and $b < c$. From our previous analysis, we see that such vertices precisely count $r(\lambda/\mu) - b(\lambda/\mu)$. Similarly, we can check for all the other factors in $\langle \mu | \vec{l}(x) | \lambda \rangle$. \hfill \square

**Example 3.6.** For the partition $\lambda = (2,0)$, we have the following three configurations:
\[
\begin{array}{cccc}
  x_{2} & 2 & 0 & 0 \\
  x_{1} & 0 & 0 & 0 \\
  & 0 & 2 & 0 \\
  & 1 & 0 & 0 \\
  & 0 & 0 & 0 \\
  & 0 & 0 & 0 \\
  & 0 & 0 & 0 \\
\end{array}
\]
\[ g_{\lambda}^{(\alpha,\beta)}(x_{1},x_{2}) = x_{1}(x_{1} + \alpha) + x_{1}x_{2} + x_{2}(x_{2} + \alpha). \]

**Example 3.7.** For the partition $\lambda = (1,1)$, we have
\[
\begin{array}{cccc}
  x_{2} & 0 & 1 & 0 \\
  x_{1} & 0 & 1 & 0 \\
  & 0 & 1 & 1 \\
  & 0 & 0 & 0 \\
  & 1 & 1 & 0 \\
  & 0 & 0 & 0 \\
  & 0 & 0 & 0 \\
\end{array}
\]
\[ g_{\lambda}^{(\alpha,\beta)}(x_{1},x_{2}) = \beta x_{1} + x_{1}x_{2} + \beta x_{2}. \]

3.4. **Vertex model for $j$ polynomials.**

3.4.1. **Definition of $l$ matrix.** In this subsection, the auxiliary line is fermionic. Let
\begin{equation}
(42)
 l_{i,j}(x) = \begin{pmatrix}
 \phi_{j} \\
 x\phi_{j}^{\dagger} + \delta_{i,m}
\end{pmatrix}
\end{equation}
\begin{equation}
(43)
 be the $l$-matrix acting on $F_{i} \otimes V_{j}$. Below we represent the entries of $l$ graphically:
\[
\begin{array}{c}
 m \\
 m \\
 m-1 \\
 m+1 \\
 m \\
 m \\
 m \\
 m \\
 1 \\
 1 \\
 x \\
 x \\
 x+1
\end{array}
\]
Similarly, we have the dual \( l^* \) matrices,

\[
(44) \quad l^*_{i,j}(x) = \begin{pmatrix}
    x + \delta_{0,m} & x \phi_j \\
    \phi_j^* & 1
\end{pmatrix}
\]

The R matrix that makes this model integrable is same as (9) with \( \beta = 0 \). When \( \beta = 0 \) we denote this matrix as \( R(y/x) \).

\[
(45)
\]

\[
\begin{pmatrix}
    1 & 0 & 0 & 0 \\
    0 & 0 & \frac{y}{x} & 0 \\
    0 & 1 & 1 - \frac{y}{x} & 0 \\
    0 & 0 & 0 & 1
\end{pmatrix}
\]

The \( R \) matrix together with matrices \( l_i \) and \( l_j \), satisfies the RLL relation in \( \text{End}(F_i \otimes F_j \otimes V_n) \):

\[
(46)
\]

\[
R_{i,j}(y/x)l_{i,n}(x)l_{j,n}(y) = l_{j,n}(y)l_{i,n}(x)R_{i,j}(y/x)
\]

3.4.2. Row-row transfer matrices. We now define the transfer matrix \( t \) which acts linearly on \( V^r \) as follows:

\[
(47)
\]

\[
\begin{pmatrix}
    1 & 0 & 0 & 0 \\
    0 & 0 & \frac{y}{x} & 0 \\
    0 & 1 & 1 - \frac{y}{x} & 0 \\
    0 & 0 & 0 & 1
\end{pmatrix}
\]

\[
R_{i,j}(y/x)l_{i,n}(x)l_{j,n}(y) = l_{j,n}(y)l_{i,n}(x)R_{i,j}(y/x)
\]

\[
\begin{pmatrix}
    x & \quad \rightarrow \\
    y & \quad \rightarrow
\end{pmatrix}
\]

Remark 3.8. Observe that the transfer matrix \( t \) from row model of \( \mathcal{g}^{(1,0)}_\lambda \) and \( \mathcal{g}^{(1,0)}_\lambda \) are the same.

Similarly, we define the dual transfer matrix \( t^* \) which acts linearly on \( V^r \) as follows:

\[
(48) \quad t(x) : |i_1\rangle \otimes |i_2\rangle \otimes \cdots \mapsto \sum_{k_1, k_2, \ldots \geq 0} w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) |k_1\rangle \otimes |k_2\rangle \otimes \cdots,
\]

where \( w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) \) is the weight of the single row of vertices.

\[
w(\{i_1, i_2, \ldots\}; \{k_1, k_2, \ldots\}) = \frac{1}{k_1 k_2 \cdots}
\]

Remark 3.8. Observe that the transfer matrix \( t^* \) from row model of \( \mathcal{g}^{(1,0)}_\lambda \) and \( \mathcal{g}^{(1,0)}_\lambda \) are the same.
3.4.3. \( j \) polynomials. Recall that we denote dual Grothendieck polynomials by \( g_\lambda \), which is the \( \alpha = 0 \) and \( \beta = 1 \) specialization of \( g_\lambda^{(\alpha,\beta)} \). Then the \( \omega(g_\lambda) \) polynomials are called weak dual Grothendieck polynomials and we shall denote them by \( j_\lambda \):

\[
j_\lambda = \omega(g_\lambda) = \omega(g_\lambda^{(0,1)}) = g_\lambda^{(1,0)}.
\]

When \( \beta = 0 \), the branching formula of \( g_\lambda^{(\alpha,\beta)} \) reduces to the following [17]:

\[
j_\lambda(x_1, \ldots, x_n, x_{n+1}) = \sum_\mu j_{\lambda/\mu}(x_{n+1}) j_\mu(x_1, \ldots, x_n),
\]

where \( j_{\lambda/\mu}(x) \) is defined as follows,

\[
j_{\lambda/\mu}(x) = \begin{cases} x^{c(\lambda/\mu)}(1 + x)^{|\lambda/\mu| - c(\lambda/\mu)} & \lambda/\mu \text{ vert. strip}, \\ 0 & \text{otherwise}. \end{cases}
\]

**Theorem 3.9.** The dual weak Grothendieck polynomials \( j_\lambda(x) \) are given by

\[
j_\lambda(x_1, \ldots, x_n) = \langle 0 | t(x_1) \cdots t(x_n) | \lambda \rangle
\]

(50)

\[
j_\lambda(x_1, \ldots, x_n) = \langle \lambda' | t^*(x_n) \cdots t^*(x_1) | 0 \rangle
\]

(51)

where \( | \lambda' \rangle = \bigotimes_{i=1}^{\infty} | m_i^\prime(\lambda) \rangle \), and similarly for the dual state \( \langle \lambda' \rangle \).

**Proof.** Before we prove (50), let us observe an example to understand the vertices. For \( \mu = (5, 4, 4, 3) \) and \( \lambda = (5, 5, 3, 1, 1) \),

From the example above, observe that having 1 on the left horizontal edge at site \( i \) amounts to adding a box in row \( i \) of the Young diagram of \( \mu \). The number of such vertices amounts to the number of boxes added. The vertex \( 1 \) at site \( i \) can be
read as adding a box in two successive rows in the same column. So every such vertex amounts to $|\lambda/\mu| - c(\lambda/\mu)$. Using similar reasoning as in Theorem 2.2 and with the preceding analysis, the proof is immediate.

\[ \square \]

4. Generalised polynomials

In this section, we shall generalise the polynomials by introducing additional variables which are attached to the vertical lines of the underlying lattice model. In order to do that, we need the R matrix that underpins the integrability of a lattice model to satisfy the so-called difference property. Usually the difference property refers to entries of the R matrix being invariant under translation of the spectral variables. In this paper, we say that an R matrix satisfies the difference property when the entries are invariant under scaling of the spectral parameters i.e. the non constant entries are polynomials in ratio of the spectral variables.

4.1. Difference property of the R matrices. In this subsection, we study the difference property of the various R matrices studied in this paper.

4.1.1. R matrices of Row models. Consider the R matrix ((9)) for canonical Grothendieck polynomials. Observe that when $\beta = 0$, it satisfies the difference property. It also satisfies the property for general $\alpha$ and $\beta$ if we consider the spectral variables to be $\frac{x}{1-\beta x}$ instead of $x$.

\[ R(y/x) = R(x,y) = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & \frac{y}{x} & 0 \\ 0 & 1 & 1 - \frac{y}{x} & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \]

(52)

In the case of $g^{(\alpha,\beta)}_{\lambda}(\alpha,\beta)$, the r-matrix does not satisfy the difference property. It is also not defined for $\beta = 0$. Hence, we studied a different model for the case where $\beta = 0$. Observe that the R matrix for the vertex model of $j_{\lambda} = g^{(1,0)}_{\lambda'}$ is the same as that in (52).

4.1.2. R matrices of column models. The $\tilde{R}$ matrix of the column model of $G^{(\alpha,\beta)}_{\lambda}$,

\[ \tilde{R}_{b,c}^{a}(y,x) = \begin{pmatrix} x & \frac{1}{1-\alpha x} \\ \frac{y}{1-\alpha y} & 1 \end{pmatrix}^{a} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 - \frac{x}{(1-\alpha x)y} \end{pmatrix} \]

satisfies the difference property when $\alpha = 0$. It also satisfies the difference property in general, when we consider the spectral variables to be $\frac{x}{1+\alpha x}$ and $\frac{y}{1+\alpha y}$ instead of $x$ and $y$. 
The \( \tilde{r} \)-matrix of the column model of \( g^{(\alpha,\beta)}_\lambda \):

\[
\tilde{r}_{i,j}^{k,l}(x, y) = \begin{cases} 
0 & i < j \\
1 & k = l = 0 \\
\frac{x}{y} \frac{(y + \alpha)}{(x + \alpha)}^{1-k} & k = l > 0 \\
\frac{x}{y} \left( \frac{y + \alpha}{x + \alpha} - 1 \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{-k} & k > 0 
\end{cases} 
\]

(53)

satisfies the difference property when \( \alpha = 0 \).

4.2. Generalised polynomials. To summarize, in the case of row models we can only generalise \( G^{(\alpha,0)}(\lambda) \) and \( g^{(\alpha,0)}(\lambda) \). Similarly, in the case of column models we can generalise \( G^{(0,\beta)}(\lambda) \) and \( g^{(0,\beta)}(\lambda) \).

We generalise the polynomials by assigning variables to vertical lines. Let us assign the variable \( z_i \) to the \( i \)th vertical line from the left. In any model, the weight of a vertex formed with the intersection of \( i \)th horizontal line and \( j \)th vertical line is defined as follows:

\[
\tilde{w}_{(x_i, z_j)} \left( \begin{array} {c} a \\ c \\ \hline b \\ d \end{array} \right) = \tilde{w}(x_i z_j) (a, b; c, d).
\]

Let us name these polynomials.

(i) We call \( G^{\alpha}_\lambda = G^{(0,-\alpha)}_\lambda \) generalised Grothendieck polynomials.
(ii) We call \( g^{\alpha}_\lambda = g^{(0,\alpha)}_\lambda \) generalised dual Grothendieck polynomials.
(iii) We call \( J^{\alpha}_\lambda = G^{(-\alpha,0)}_\lambda \) generalised weak Grothendieck polynomials.
(iv) We call \( j^{\alpha}_\lambda = g^{(-\alpha,0)}_\lambda \) generalised weak dual Grothendieck polynomials.

When \( \alpha = 1 \), we shall drop the superscript.

Example 4.1. For the partition \( \lambda = (1) \), the generalised Grothendieck polynomial \( G^{(1)}(x_1, z_1) \) is \( \frac{x_1}{z_1} \). For comparison, the double Grothendieck polynomial is \( x_1 + y_1 (1-x_1) \) [12]. We observe that these two generalisations of Grothendieck polynomials are not the same.

Example 4.2. Let us look at a non trivial example. For the partition \( \lambda = (3,1) \) we have the following three configurations (edges with labels 0 suppressed):

\[
G^{(0, -1)}_{(3,1)}(x_1, x_2) = \\
\left( \frac{x_1^2}{z_1 z_2} \right) \left( 1 - \frac{x_1}{z_1} \right) \left( \frac{x_2^2}{z_1^2} \right) + \left( \frac{x_1^3}{z_1^3 z_2^2} \right) \left( \frac{x_2}{z_1} \right) + \left( 1 - \frac{x_1}{z_1} \right) \left( \frac{x_1}{z_2} \right) \left( \frac{x_2}{z_1} \right)^3.
\]
Remark 4.3. Observe that by setting both $\alpha$ and $\beta$ to 0, we get a generalised version of Schur polynomials. Generalised Schur polynomials from the row model and the column model are not the same. Let us denote the generalised Schur from the row (resp. column) model by $s^r$ (resp. $s^c$).

Example 4.4. For the partition $\lambda = (3, 1)$, we have
\[
\begin{align*}
s^r_{(3,1)}(x_1, x_2; z_1, z_2, \ldots) &= \left( \frac{x_1^3}{z_1 z_2 z_3} \right) \left( \frac{x_2}{z_1} \right) + \left( \frac{x_1^2}{z_1 z_2} \right) \left( \frac{x_2^2}{z_1 z_3} \right) + \left( \frac{x_1}{z_1} \right) \left( \frac{x_2^3}{z_1 z_2 z_3} \right) \\
\end{align*}
\]
\[
\begin{align*}
s^c_{(3,1)}(x_1, x_2; z_1, z_2, \ldots) &= \left( \frac{x_1^3}{z_1 z_2} \right) \left( \frac{x_2}{z_1} \right)^2 + \left( \frac{x_1^2}{z_1^2 z_2} \right) \left( \frac{x_2}{z_1} \right) + \left( \frac{x_1}{z_1} \right) \left( \frac{x_2^3}{z_1 z_2} \right).
\end{align*}
\]

$s^r_\lambda$ is a monomial multiple of $s_\lambda$, where the monomial is obtained by recording the columns of the Young diagram. Similarly, in the case of $s^c_\lambda$ the monomial is obtained by recording the rows of the Young diagram.

5. Duality between Column and Row models

In this section, we shall study a relation between the transfer matrix of the row and column model of $G^{(\alpha, \beta)}_\lambda$. Let us recall the necessary notation from various sections. The transfer matrices of the row model for $G^{(\alpha, \beta)}_\lambda$ are denoted by $T$, and those of the column model are denoted by $\tilde{T}$.

Proposition 5.1 (Inversion relation). The transfer matrices $\tilde{T}$ and $T$ satisfy the following identity:

\[
\begin{align*}
T(-x)\tilde{T}\left( \frac{x}{1 + (\alpha - \beta)x} \right) &= 1 \\
&= \begin{pmatrix}
& v_1 & u_2 & u_3 & u_4 & \cdots & \ast & 0 \\
& v_2 & v_3 & v_4 & \cdots & \ast & 0 & 0 \\
& 0 & 0 & \vdots & \vdots & \vdots & \vdots & \vdots \\
\end{pmatrix}
\end{align*}
\]

Proof. We shall prove the proposition for transfer matrices of size 1 and then apply induction to the size of the transfer matrices. Assign weights of $T(-x)$ for the vertex at the bottom and the weights of $\tilde{T}\left( \frac{x}{1 + (\alpha - \beta)x} \right)$ for the other vertex. Write $z$ for the vertical spectral parameter.

Observe that when $b = d$, there is a unique configuration with total weight 1. Now assume $b \neq d$
When \( b > 0 \),
\[
\begin{align*}
\left( \frac{1 - \beta \left( \frac{x}{z} \right)}{1 + \alpha \left( \frac{x}{z} \right)} \right) \left( \frac{\frac{x}{z}}{1 - \beta \left( \frac{x}{z} \right)} \right)^{d-b} & \left( \frac{1 + \alpha \left( \frac{x}{z} \right)}{1 - \beta \left( \frac{x}{z} \right)} \right) + \\
\left( \frac{-\frac{x}{z}}{1 + \alpha \left( \frac{x}{z} \right)} \right) \left( \frac{\frac{x}{z}}{1 - \beta \left( \frac{x}{z} \right)} \right)^{d-b-1} & \left( \frac{1 + \alpha \left( \frac{x}{z} \right)}{1 - \beta \left( \frac{x}{z} \right)} \right) \\
= 0.
\end{align*}
\]

When \( b = 0 \),
\[
\begin{align*}
\left( \frac{\frac{x}{z}}{1 - \beta \left( \frac{x}{z} \right)} \right)^{d} & + \left( \frac{-\frac{x}{z}}{1 + \alpha \left( \frac{x}{z} \right)} \right) \left( \frac{\frac{x}{z}}{1 - \beta \left( \frac{x}{z} \right)} \right)^{d-1} \left( \frac{1 + \alpha \left( \frac{x}{z} \right)}{1 - \beta \left( \frac{x}{z} \right)} \right) \\
= 0.
\end{align*}
\]

In order to apply the induction argument, we need to show that the transfer matrix of size \( n + 1 \) can be written as a multiple of the transfer matrix of size \( n \). Consider a transfer matrix of size \( n + 1 \) where top and bottom labels are fixed.

![Diagram](image)

Observe that when the left most boundary is fixed, then the contribution from the first site is fixed. Therefore, we can move the free boundary condition across the physical line at site 1. We then apply induction.

Define \( \widetilde{T}^\ast(x) \in \text{End}(V^c) \) as the adjoint of \( \widetilde{T}(x) \):
\[
\langle \lambda^c | \widetilde{T}^\ast(x) | \mu^c \rangle = \langle \mu^c | \widetilde{T}(x) | \lambda^c \rangle.
\]

Then the inversion relation between \( T^\ast(x) \) and \( \widetilde{T}^\ast(x) \) follows from the definition of \( \widetilde{T}^\ast(x) \) and the inversion relation of \( T(x) \) and \( \widetilde{T}(x) \):
\[
(55) \quad T^\ast(-x) \widetilde{T}^\ast \left( \frac{x}{1 + (\alpha - \beta)x} \right) = 1.
\]

In the case of \( g^{(\alpha,\beta)}_{\lambda} \), such an inversion relation does not exist for general \( \alpha \) and \( \beta \). But there is an inversion relation in the case where \( \alpha = 0 \) and \( \beta = 1 \). Since we are concerned with \( g^{(\alpha,\beta)}_{\lambda} \), it is convenient to specialize the Boltzmann weights from the column model of \( g^{(\alpha,\beta)}_{\lambda} \).

\[
w_{(x,z)} \left( \begin{array}{c} a & b & d \\ \end{array} \right) = w_{(\frac{x}{z})} \left( a, b; c, d \right) = \left( \frac{x}{z} \right)^{\min(a,d)}.
\]
Finally, we recall that the transfer matrix $t$ of $g^{(1,0)}$ from the row model and the transfer matrix $t$ of $j$ polynomials are the same. Therefore, graphically we represent $t$ with fermionic auxiliary line.

**Proposition 5.2.** The transfer matrices of $g^{(1,0)}$ from the row model and transfer matrices of $g^{(0,1)}$ from the column model satisfy the following relation:

$$t(-x) \tilde{t}(x) = 1$$

$$\begin{pmatrix}
  t(z) & \ast & u_1 & u_2 & u_3 & u_4 & \ldots & 0 \\
  \ast & \ast & v_1 & v_2 & v_3 & v_4 & \ldots & 0 \\
\end{pmatrix}$$

**Proof.** The proof is similar to Proposition 5.1. We shall prove the statement for transfer matrices of size 1. When $b = d$, there is a unique configuration with weight 1 and when $b \neq d$, we have two configurations which add up to 0.

$$\begin{pmatrix}
  \frac{x}{z} & d-b & 0 \\
  0 & b+1 & 0 \\
\end{pmatrix} + \begin{pmatrix}
  -\frac{x}{z} & d-1-b \\
  1 & 0 \\
\end{pmatrix} = 0.$$

Assume $b > 0$, then for any fixed $v, u, b, d$ the Boltzmann weights are fixed irrespective of the right boundary.

We can then simply slide the free boundary condition. Special care needs to be taken when $b = 0$. When $b = 0$, we have the following configurations:

$$\begin{pmatrix}
  \frac{x}{z} & d \\
  0 & 0 \\
\end{pmatrix} = \left(1 - \frac{x}{z}\right)^d (\frac{x}{z})^d,$$

Observe that in the case of the first configuration, there is a unique configuration suggesting that the right boundary is not free. But we can get away with it by adding the weight of the first configuration with the weight of the second configuration. Then we obtain $(1 - \frac{x}{z})(\frac{x}{z})^d \times$ times the transfer matrix of size $n$.

**6. Cauchy identities**

In this section, we shall prove Cauchy identities involving $G^{(-\alpha,-\beta)}$ and $g^{(\alpha,\beta)}$. We shall prove them by using the commutation relations between various combinations of the transfer matrices.
Proposition 6.1. The transfer matrix $T^*$ (eq. (13)) from the row model of $G_{\lambda}^{(-\alpha, -\beta)}(x)$, and the transfer matrix $t(y)(eq. (21))$ from the row model of $g_{\lambda}^{(\alpha, \beta)}(y)$ satisfy

$$t(y)T^*(x) = \frac{1}{1 - xy}T^*(x)t(y).$$

Proof. Firstly, note that the product of $T^*$ and $t$ is well defined only when we assume that the spectral variables satisfy $|x| < 1$ which ensures that the terms with unbounded degree are equal to 0.

The proof is similar to the way we proved that the transfer matrices commute. The $\mathcal{R}(x, y)$ matrix

$$(57)$$

$$\mathcal{R}(x, y) \in \text{End}(F \otimes W), \quad \mathcal{R}^{k, l}_{i, j} = \begin{cases} 1 - xy & j = k = 1, i = l = 0 \\ xy & k = l = 0, i = j = 1 \\ 1 - x\beta & k = 1 \\ x\beta & k = 0 \\ 1 & i = k = l = j = 0 \end{cases}$$

where $k, j \in \{0, 1\}$ and $i, l \in \mathbb{Z}_{\geq 0}$, together with the $L^*(x)$ matrix of $G_{\lambda}^{(-\alpha, -\beta)}(x)$ and the $l(y)$ matrix of $g_{\lambda}^{(\alpha, \beta)}(y)$ satisfies the RLL relation (see Appendix A.4):

$$(58)$$

$$\mathcal{R}(x, y)L^*(x)l(y) = l(y)L^*(x)\mathcal{R}(x, y) \in \text{End}(F \otimes W \otimes V)$$

Observe that the sum of all possible states is an eigenvector $\mathcal{R}$.

$$x \star \star y \star \star = \begin{array}{c} x \star \star \\ y \star \star \end{array}.$$

Then after multiplying the $\mathcal{R}$-matrix to $T^*(x)t(y)$ and repeatedly applying the RLL relation we get the following equation:

![Diagram](image)

Given below are the two possible configurations on the right hand side of the equation.

![Diagram](image)

The partition function of the first configuration has terms of unbounded degree and hence is 0. Therefore, there is a unique configuration on the right hand side and the
This implies the desired commutation relation:
\[ T^*(x)t(y) = (1 - xy)t(y)T^*(x). \]

**Theorem 6.2.** Canonical Grothendieck polynomials and their duals satisfy the following Cauchy identity:

\[ (59) \quad \sum_{\lambda} G^{(-\alpha,-\beta)}_{\lambda}(x_1, x_2, \ldots, x_m) g^{(\alpha, \beta)}_{\lambda}(y_1, y_2, \ldots, y_n) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - x_iy_j}. \]

**Proof.** Let
\[ G(x_1, x_2, \ldots, x_m, y_1, y_2, \ldots, y_n) = \langle 0 | t(y_1) t(y_2) \cdots T^*(x_2) T^*(x_1) | 0 \rangle. \]

Then by Theorems 2.2 and 2.7 we get that
\[ G(x_1, x_2, \ldots, x_m, y_1, y_2, \ldots, y_n) = \sum_{\lambda} G^{(-\alpha,-\beta)}_{\lambda}(x_1, x_2, \ldots, x_m) g^{(\alpha, \beta)}_{\lambda}(y_1, y_2, \ldots, y_n). \]

By repeatedly applying the commutation relation of Proposition 6.1, we obtain
\[ G(x_1, x_2, \ldots, x_m, y_1, y_2, \ldots, y_n) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - x_iy_j} \langle 0 | T^*(x_1) T^*(x_2) \cdots t(y_2) t(y_1) | 0 \rangle \\
= \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - x_iy_j}. \]

**Remark 6.3.** Recall that \( g^{(\alpha, \beta)}_{\lambda} \) and \( G^{(-\alpha,-\beta)}_{\lambda} \) were defined using the branching formulae (see (15) and (25)). The Cauchy identity then implies that these two families are dual with respect to the Hall inner product.

We can derive a skew version of the identity if we choose a different vector and covector. Let
\[ G(x_1, x_2, \ldots, x_m, y_1, y_2, \ldots, y_n) = \langle \mu | t(y_1) t(y_2) \cdots T^*(x_2) T^*(x_1) | \lambda \rangle, \]
then using the same reasoning as in the above theorem we get the following identity:
\[ \sum_{\nu} G_{\nu \neq \lambda}(x_1, x_2, \ldots, x_m) g_{\nu / \lambda}(y_1, y_2, \ldots, y_n) = \prod_{1 \leq i \leq n, 1 \leq j \leq m} \frac{1}{1 - x_iy_j} \sum_{\nu} G_{\mu \neq \nu}(x_1, x_2, \ldots, x_m) g_{\lambda / \nu}(y_1, y_2, \ldots, y_n). \]

We can do the same for all identities in this section but for simplicity we shall stick to the non-skew identities.

**Corollary 6.4.** Generalised weak Grothendieck polynomials and their duals satisfy the following identity:

\[ (60) \quad \sum_{\lambda} J_\lambda^\alpha(x_1, \ldots, x_m; z_1, z_2, \ldots, z_\lambda) J_\lambda^\beta(y_1, \ldots, y_n; \frac{1}{z_1}, \frac{1}{z_2}, \ldots) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - x_iy_j}. \]
Proof. Observe that when $\beta = 0$, the non constant entries of $R$ are $xy$ and $1 - xy$. If we introduce the inhomogeneities as $xz$ and $\frac{y}{x}$, then the $R$ matrix remains the same and thereby gives the same commutation relation. Then the proof of the identity is same as that of Theorem 6.2.

We now prove the following Cauchy identity:

\begin{equation}
\sum_{\lambda} G_{\lambda}^{(\beta, -\alpha)}(x_1, x_2, \ldots, x_m) G_{\lambda}^{(\alpha, \beta)}(y_1, y_2, \ldots, y_n) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} (1 + x_1 y_j).
\end{equation}

We can prove this identity by proving a commutation relation between $T^*$ and $\tilde{T}$. But we shall prove it using the inversion relation from (54).

**Theorem 6.5.** Canonical Grothendieck polynomials and their duals satisfy the following Cauchy identity:

\begin{equation}
\sum_{\lambda} G_{\lambda}^{(\beta, -\alpha)}(x_1, x_2, \ldots, x_m) G_{\lambda}^{(\alpha, \beta)}(y_1, y_2, \ldots, y_n) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} (1 + x_i y_j).
\end{equation}

**Proof.** By substituting $-x$ for $x$ in (56), we get the following relation.

\[ t(y) T^*(-x) = \frac{1}{1 + xy} T^*(-x) t(y). \]

By multiplying $\tilde{T}^* \left( \frac{x}{1 + x(\beta - \alpha)} \right)$ on both sides and applying the inversion relation (55), we get the following relation:

\begin{equation}
\tilde{T}^* \left( \frac{x}{1 + x(\beta - \alpha)} \right) t(y) = \frac{1}{1 + xy} t(y) \tilde{T}^* \left( \frac{x}{1 + x(\beta - \alpha)} \right).
\end{equation}

Let

\[ G(x_1, x_2, \ldots, x_n, y_1, y_2, \ldots, y_m) =
\]

\[ \langle 0 | t(y_1) t(y_2) \cdots \tilde{T}^* \left( \frac{x_1}{1 + (\beta - \alpha)x_1} \right) \cdots \tilde{T}^* \left( \frac{x_n}{1 + (\beta - \alpha)x_n} \right) | 0 \rangle. \]

Then by Theorem 2.7 and the definition of $\tilde{T}^*$ we obtain

\[ G(x_1, x_2, \ldots, x_n, y_1, y_2, \ldots, y_m) = \sum_{\lambda} G_{\lambda}^{(\beta, -\alpha)}(x_1, \ldots, x_n) G_{\lambda}^{(\alpha, \beta)}(y_1, \ldots, y_m). \]

On the other hand by repeatedly applying the commutation relation we get that

\[ G(x_1, x_2, \ldots, x_n, y_1, y_2, \ldots, y_m) \]

\[ = \prod_{1 \leq i \leq m} (1 + x_i y_j) \langle 0 | \tilde{T}^* \left( \frac{x_1}{1 + (\beta - \alpha)x_1} \right) \cdots \tilde{T}^* \left( \frac{x_n}{1 + (\beta - \alpha)x_n} \right) t(y_1) t \cdots (y_m) | 0 \rangle \]

\[ = \prod_{1 \leq i \leq m} (1 + x_i y_j). \]

\[ \Box \]

**Corollary 6.6.** Generalised Grothendieck polynomials and generalised weak dual Grothendieck polynomials satisfy the following identity:

\begin{equation}
\sum_{\lambda} G_{\lambda}(x_1, \ldots, x_m; z_1, z_2, \ldots) f_{\lambda}(y_1, \ldots, y_n; \frac{1}{z_1}, \frac{1}{z_2}, \ldots) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} (1 + x_i y_j).
\end{equation}

**Proof.** Plug in $\beta = 0$ and $\alpha = 1$ in Theorem 6.5 and use the inhomogeneous transfer matrices.\[ \Box \]
Theorem 6.7. Generalised Grothendieck polynomials and their duals satisfy the following identity:

\[ \sum_{\lambda} G_{\lambda}(x_1, \ldots, x_m; z_1, \ldots, z_m) g_{\lambda}(y_1, \ldots, y_n; \frac{1}{z_1}, \ldots, \frac{1}{z_m}) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - x_i y_j}. \]

Proof. Recall the commutation relation from Theorem 6.5:

\[ \tilde{T}^* \left( \frac{x}{1 + x(\beta - \alpha)} \right) t(y) = \frac{1}{1 + xy} t(y) \tilde{T}^* \left( \frac{x}{1 + x(\beta - \alpha)} \right). \]

In order to apply the inversion relation among the transfer matrices of the dual Grothendieck polynomials, we need to specialize the above commutation relation with \( \alpha = 1 \) and \( \beta = 0 \). We then get the following relation:

\[ \tilde{T}^* \left( \frac{x}{1 - x} \right) t(-y) = \frac{1}{1 - xy} t(-y) \tilde{T}^* \left( \frac{x}{1 - x} \right). \]

We now multiply the above equation by \( \tilde{t}(y) \) and apply the inversion relation (Proposition 5.2):

\[ \tilde{t}(y) \tilde{T}^* \left( \frac{x}{1 - x} \right) = \frac{1}{1 - xy} \tilde{T}^* \left( \frac{x}{1 - x} \right) \tilde{t}(y). \]

The result then follows immediately from the definition of \( \tilde{T}^* \), and Theorem 3.5. \( \square \)

Proposition 6.8. Generalised Grothendieck polynomials satisfy

\[ G_{\lambda}(z_1, \ldots, z_m; z_1, \ldots, z_m) = 1. \]

Proof. When \( \alpha = 0 \) and \( \beta = -1 \), \( \tilde{R} \) and \( \tilde{L} \) are same. The \( \tilde{R} \) matrix satisfies the unitary relation (for a proof refer to Appendix A.1.5):

\[ \begin{array}{c}
    x \\
    y
\end{array} = \begin{array}{c}
    x \\
    y
\end{array}. \]

Recall that there are two types of vertices (crossing and elbow) in the column model for Grothendieck polynomials:

\[ \begin{array}{c}
    c \\
    b
\end{array} \quad \begin{array}{c}
    c \\
    b
\end{array}. \]

Consider the model with \( m \) rows i.e. Grothendieck polynomial in \( m \) variables. We know that the number of variables restricts us to partitions with highest column being less than or equal to \( m \). So the only inhomogeneities are \( z_1, \ldots, z_m \).

Recall that the Boltzmann weight of a crossing has a \( \left( 1 - \frac{x}{z} \right) \) factor. So, when we set \( x_i = z_i \), the contribution of a configuration to the partition function is non zero only when the vertices along the diagonal are entirely elbows.
Then we get the desired result by repeatedly applying the unitary relation. □

As a consequence of the above proposition, we recover an identity for generalised dual Grothendieck polynomials, which is proved by Yeliussizov in [18].

**Corollary 6.9.** Dual Grothendieck polynomials satisfy the following identity:

\[
\sum_{l(\lambda) \leq m} g_{\lambda}(y_1, \ldots, y_n; \frac{1}{z_1}, \ldots, \frac{1}{z_m}) = \prod_{1 \leq i \leq m, 1 \leq j \leq n} \frac{1}{1 - z_i y_j}.
\]

**Proof.** Set \(x_i = z_i\) in (63). □

### Appendix A. RLL relations

**A.1. RLL for the column model of \(G_{\lambda}^{(a, b)}\).** For convenience, let us recall the Boltzmann weights of the column model of \(G_{\lambda}^{(a, b)}\) and the entries of the \(\tilde{R}\) matrix.

\[
\widetilde{R}_{a\, b\, c\, d} (y, x) = \begin{cases} 
0 & \text{when } b < c \\
1 & \text{when } b = c \\
\frac{1}{1 - \alpha x} - \frac{x}{(1 - \alpha x)y} & \text{otherwise.}
\end{cases}
\]

Let us try to understand the range of \(g\). First observe that whenever \(a' < g\), the summation is 0 because of the \(\tilde{R}\)-matrix. Based on the Boltzmann weights, the contribution of the top vertex is non zero if and only if \(g + b - c \geq c'\). Therefore, \(g\) on LHS can at most be \(a'\), and it has to be at least \(c' + c - b\). Similarly, on the RHS we have \(a + a' - d \leq c'\).
A.1.1. Assume $b > c$ and $d > a$. Let us now compute the LHS.

$$LHS = \left( \frac{y - x}{y(1 - ax)} \right) \left( \frac{x(1 - ay)}{x(1 - ax)} \right)^a \left( \frac{1}{1 - ay} \right)^{d-a} \left( \frac{x}{1 - ax} \right)^{c-b} \left( \frac{1 + \beta x}{1 - ax} \right) +$$

$$\left( \frac{y - x}{y(1 - ax)} \right) \left( \frac{x(1 - ay)}{x(1 - ax)} \right)^a \left( \frac{1 + \beta y}{1 - ay} \right) \left( \frac{y}{1 - ay} \right)^{a'}$$

$$\left( \sum_{g=c'+d+1}^{a'-1} \left( \frac{x(1 - ay)}{x(1 - ax)} \right)^g \right) +$$

$$\left( \frac{y - x}{y(1 - ax)} \right) \left( \frac{x(1 - ay)}{x(1 - ax)} \right)^a \left( \frac{1 + \beta x}{1 - ax} \right) \left( \frac{y}{1 - ay} \right)^{a'}$$

$$\left( \frac{1 + \beta x}{1 - ax} \right)^{(1 + \beta y)} \left( \frac{1 - \beta y}{1 - ax} \right) \left( \frac{1 - \beta y}{1 - ax} \right) \left( \frac{y}{1 - ay} \right)^{a'}$$

$$\left( \frac{x(1 - ay)}{y(1 - ax)} \right)^{c'} \left( \frac{x}{1 - ax} \right)^a \left( \frac{1 + \beta x}{1 - ax} \right) \left( \frac{y}{1 - ay} \right)^{a'} \left( \frac{1 + \beta y}{1 + \beta y} \right) \left( \frac{x}{1 - ax} \right)^{d-a}$$

$$= \left( \frac{1 + \beta x}{1 - ax} \right)^2 \left( \frac{x}{1 - ax} \right)^{a + c' + d - b} \left( \frac{y}{1 - ay} \right)^{d - a} \left( \frac{y}{1 - ay} \right) \left( \frac{y}{1 - ay} \right).$$

A.1.2. Assume $a < d$ and $b = c$. From the computation of the previous case, we can get the LHS by multiplying $\frac{1 - ax}{1 + \beta x}$ to the LHS of previous computation.

$$LHS = \left( \frac{1 + \beta x}{1 - ax} \right)^{a + c'} \left( \frac{y}{1 - ay} \right)^{d - a} \left( \frac{y}{1 - ay} \right).$$

On the right hand side, there is only one case because of the global condition, $a + a' + b = c + c' + d$.

$$RHS = \left( \frac{x}{1 - ax} \right)^{a + c'} \left( \frac{1 + \beta x}{1 - ax} \right) \left( \frac{y}{1 - ay} \right)^{a' - c'}$$

$$= \left( \frac{x}{1 - ax} \right)^{a + c'} \left( \frac{1 + \beta x}{1 - ax} \right) \left( \frac{y}{1 - ay} \right)^{d - a} \left( \frac{y}{1 - ay} \right).$$

A.1.3. Assume $a = d$ and $b > c$. RHS of the present case is a $\frac{1 - ax}{1 + \beta x}$ multiple of the RHS of the Appendix A.1.1.
\[ \text{RHS} = \left( \frac{1 + \beta x}{1 - \alpha x} \right)^a \left( \frac{y}{1 - \alpha y} \right)^a \left( \frac{x}{1 - \alpha x} \right)^{a' + c' - b} \left( \frac{y}{1 - \alpha y} \right)^d - a \]

For the LHS, there is just one valid configuration:

\[ \text{LHS} = \left( \frac{x}{1 - \alpha x} \right)^a \left( \frac{y}{1 - \alpha y} \right)^{-a} \left( \frac{y}{1 - \alpha y} \right)^a \left( \frac{x}{1 - \alpha x} \right)^{a'} \left( \frac{1 + \beta x}{1 - \alpha x} \right) \]

\[ = \left( \frac{x}{1 - \alpha x} \right)^{a + a'} \left( \frac{1 + \beta x}{1 - \alpha x} \right) \]

\[ = \left( \frac{x}{1 - \alpha x} \right)^{a + c' - b} \left( \frac{1 + \beta x}{1 - \alpha x} \right). \]

In the final step, we substitute \( a' = c + c' - b \), which follows from the global condition.

A.1.4. Assume \( a = d \) and \( b = c \). Recall from Appendix A.1.3 that when \( a = d \), there is a unique configuration on LHS. Similarly, recall from Appendix A.1.2 that there is a unique configuration on RHS when \( b = c \). We have already computed these two configurations and they are equal.

A.1.5. Unitary relation for the \( \tilde{R} \) matrix. The \( \tilde{R} \) satisfies the unitary relation:

\[ \begin{array}{cc}
\times & \\
\times & \\
y & y
\end{array} = \begin{array}{cc}
x & \\
x & \\
y & y
\end{array}. \]

Consider the following configuration:

\[ \begin{array}{cc}
\times & \\
\times & \\
a & a' \end{array} \quad \begin{array}{cc}
\times & \\
\times & \\
b & b'
\end{array}. \]

When \( a = b \) and \( a' = b' \), there is a unique configuration with weight 1. Now let us assume \( a \neq b \) and \( a' \neq b' \).

\[ = \left( \frac{x}{y} \right)^a \left( 1 - \frac{x}{y} \right) \left( \frac{y}{x} \right)^{a + a' - b'} + \sum_{g = b' + 1}^{a'} \left( \frac{x}{y} \right)^a \left( 1 - \frac{x}{y} \right) \left( \frac{y}{x} \right)^{a + a' - g} \left( 1 - \frac{y}{x} \right) + \left( \frac{x}{y} \right)^a \left( \frac{y}{x} \right)^a \left( 1 - \frac{y}{x} \right) \]

\[ = \left( 1 - \frac{x}{y} \right) \left( \frac{y}{x} \right)^{a' - b'} + \left( \frac{y}{x} \right)^{a' - b' - 1} \left( 1 - \frac{y}{x} \right) \left( 1 - \left( \frac{x}{y} \right)^{a' - b' - 1} \right) + \left( 1 - \frac{y}{x} \right) \]

\[ = 0. \]
A.2. RLL relation for row model of $g^{(\alpha, \beta)}_\lambda$. We recall the Boltzmann weights and $r$-matrix of the row model of $g^{(\alpha, \beta)}_\lambda$:

$$w_x \left( \begin{array}{c} a \\ b \\ c \\ d \end{array} \right) \equiv w_x(a, b; c, d) = \delta_{a+b+c+d} \begin{cases} (\alpha + \beta)^{a-d-1}x \beta^{d-1} & a > d \\ \beta^{a-1}x & 0 < a \leq d \\ 1 & a = 0 \end{cases}$$

where $a, b, c, d \in \mathbb{Z}_{\geq 0}$.

The entries of the $r$-matrix are the following:

$$r^{a,d}_{b,c}(x, y) = \begin{cases} 0 & b > c \\ 1 & b = c = 0 \\ \frac{x}{y} & b = c > 0 \\ \left(1 - \frac{x}{y}\right) & b = 0 \\ \left(1 - \frac{x}{y}\right) \left(1 - \frac{\beta}{y}\right)^{a-d-1} & b > 0 \end{cases}$$

Before we start proving the relation, let us analyze the cases we need to consider. Firstly, from the LHS, the weight of bottom vertex is fixed based on whether $b \leq c$ or $b > c$. Similarly, on the RHS, the weight of the top vertex is fixed based on the relation between $a$ and $d$.

We now look at the cases that arise from considering the entries of the $r$ matrix. Firstly, the entries of the $r$ matrix on LHS depends on whether $a' > 0$ or $a' = 0$. Similarly, the entry of $r$ matrix on RHS depends on whether $c > 0$ or $c = 0$.

In total, there are sixteen cases to consider. We shall divide these cases into four categories based on the conditions on $b, c$ and $a, d$. Then for each such case, we shall consider four sub-cases by the conditions on $a'$ and $c$.

A.2.1. Assume $b < c$ and $d < a$. Assume $a' > 0$ and $c > 0$.

To ease up the computation, we break up the summation into two parts, $0 \leq g \leq d$ and then $d < g \leq a$. 
Under the assumption that \( b < c \), we have

\[
\sum_{g=0}^{d} a_{g+a-g}' g \quad a_{d+c'-g} \quad c
\]

\[
+ \sum_{g=d+1}^{a} a_{g+a-g}' g \quad a_{d+c'-g} \quad c.
\]

\[
LHS = \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-1} \left(\frac{y}{\beta}\right) (\alpha + \beta)^{c-b-1}(x + \alpha)\beta^d+c' +
\]

\[
\sum_{g=1}^{d} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-g-1} \left(\frac{y}{\beta}\right) (\alpha + \beta)^{c-b-1}(x + \alpha)\beta^d+c'-g)
\]

\[
\sum_{g=d+1}^{a-1} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-g-1} \left(\frac{y}{\beta}\right) (\alpha + \beta)^{g-d-1}\beta^d(y + \alpha)
\]

\[
\left((\alpha + \beta)^{c-b-1}(x + \alpha)\beta^d+c'-g) + \left(\frac{y}{x}\right) (\alpha + \beta)^{a-d-1}\beta^d(y + \alpha)\right)
\]

\[
= (\alpha + \beta)^{c+a-b-d-2}\beta^d+c'-a(x + \alpha)\left(y + \alpha \frac{y}{x}\right).
\]

We compute the right hand side: Assume that \( b < c \).

\[
\sum_{g=0}^{c} a_{g+b-g}' \quad d \quad c_{a'+c'-g} \quad c_{b} \quad g \quad c_{c}'.
\]

\[
RHS = \frac{(\alpha + \beta)^{a-d-1}\beta^d(x + \alpha)}{(\alpha + \beta)^{a+b-1}(y + \alpha)\beta^a+b-c)
\]

\[
\sum_{g=1}^{b} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{c-g-1} \left(\frac{y}{\beta}\right) (\alpha + \beta)^{g-b-1}(y + \alpha)\beta^a+b-g) +
\]

\[
\sum_{g=b+1}^{c-1} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{c-g-1} \left(\frac{y}{\beta}\right) (\alpha + \beta)^{g-b-1}(y + \alpha)\beta^a+b-c) +
\]

\[
\frac{y}{x} ((\alpha + \beta)^{c-b-1}(y + \alpha)\beta^a+b-c)
\]

\[
= (1 - \frac{y}{x})(\alpha + \beta)^{c-b-1}\beta^a+b-c+y+
\]

\[
\frac{y}{x} ((\alpha + \beta)^{c-b-1}(y + \alpha)\beta^a+b-c)
\]

\[
RHS = (\alpha + \beta)^{a+b-2}\beta^d+a+b-c(y + \alpha)\left(x + \alpha\right)
\]

\[
= (\alpha + \beta)^{a+b-2}\beta^d+a+b-c(y + \alpha)\left(x + \alpha\right).
\]

Assume \( a' = 0 \) and \( c > 0 \).
A.2.2. Assume the product of the fixed Boltzmann weights. Algebraic Combinatorics states is an eigenvector of the $r$-matrix with fixed right boundary. Then from the fact that sum of all the possible entries of the $a < d$ matrix, we get that the overall sum of $a > d$, we do not need to consider the cases where $c = 0$.

**A.2.2. Assume $b \geq c$ and $d < a$. Assume $a' > 0$ and $c > 0$.**

\[
LHS = \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-1} (\alpha + \beta)^{c-b-1}(x + \alpha)\beta^{d+c} + \\
\sum_{g=1}^{d} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-g-1} (\beta^{g-1}y)^{(\alpha + \beta)^{c-b-1}(x + \alpha)\beta^{d+c-g} +} \\
\sum_{g=d+1}^{a+b-c} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-g-1} (\alpha + \beta)^{g-d-1}\beta^{d}(y + \alpha) \\
\left((\alpha + \beta)^{c-b-1}(x + \alpha)\beta^{d+c-g}\right) \\
= \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{c-b-1} (\alpha + \beta)^{a-d-1}(x + \alpha)\beta^{d}.
\]

We compute the RHS:

\[
RHS = (\alpha + \beta)^{a-d-1}\beta^{d}(x + \alpha) \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{c-1} + \\
\left(\sum_{g=1}^{b} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{c-g-1} \frac{y}{\beta}\right) \\
= (\alpha + \beta)^{a-d-1}\beta^{d}(x + \alpha) \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{c-b-1}.
\]

Since we assumed $b > c$, we do not need to consider the cases where $c = 0$.

Since $a < d$ and $b \geq c$, the Boltzmann weights from the vertices are fixed. When we factor out the contribution from the overall sum, we are left with entries of the $r$-matrix with fixed right boundary. Then from the fact that sum of all the possible states is an eigenvector of the $r$ matrix, we get that the overall sum of $RHS$ is just the product of the fixed Boltzmann weights.

\[
RHS = (\alpha + \beta)^{a-d-1}(x + \alpha)\beta^{d}\left(\beta^{a-1}y\right).
\]

**Assume $a' = 0$.**

\[
LHS = \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a-1} x\beta^{a-1} + 
\]
\[
\sum_{g=1}^{d} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a+1} \left(\beta^{a}y\right)\beta^{a+g-1} + \\
\sum_{g=d+1}^{a-1} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a+g-1} \left(\alpha + \beta\right)^{g-d-1} \beta^{d} \left(y + \alpha\right) x^\beta \beta^{a+\alpha'} \beta^{g-1} x^\beta \beta^{a+\alpha'} \beta^{g-1} + \\
\left((\alpha + \beta)^{a-d-1} \beta^{d} (y + \alpha)\right) \\
= (\alpha + \beta)^{a-d-1} \beta^{d} (x + \alpha).
\]

For \(\text{RHS}\), we again use the fact that the sum of all the possible states is an eigenvector of the \(r\)-matrix to conclude that \(\text{RHS} = \text{just the Boltzmann weight of the top vertex.}\)

Observe that the contribution from the bottom vertex is 1. So, the overall \(\text{RHS}\) is just the Boltzmann weight of the top vertex.

\(\text{RHS} = (\alpha + \beta)^{a-d-1} (x + \alpha) \beta^{d}.\)

**A.2.3. Assume** \(a \leq d\) \(\text{and} b < c\). Since we are assuming \(b < c\), its follows that \(c > 0\). Assume \(a' > 0\).

\(\text{LHS} = \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a} \left(\frac{y}{\beta}\right) \left(\alpha + \beta\right)^{c-b-1} (x + \alpha) \beta^{d+c'} + \\
\sum_{g=1}^{a-1} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{g-1} \left(\frac{y}{\beta}\right) \left(\beta^{g-1} y\right) \left(\alpha + \beta\right)^{b-c-1} (x + \alpha) \beta^{d+c' - g} + \\
\left(\frac{y}{x}\right) \left(\beta^{a-1} y\right) \left(\alpha + \beta\right)^{c-b-1} (x + \alpha) \beta^{d+c' - a} \\
= (\alpha + \beta)^{c-b-1} (x + \alpha) \beta^{d+c'} \left(\frac{y}{\beta}\right).
\]

Given that \(b < c\), we can get the \(\text{RHS}\) computation from Appendix A.2.1. The only difference being the Boltzmann weight corresponding to the top vertex.

\(\text{RHS} = x^\beta \beta^{a-1} (\alpha + \beta)^{c-b-1} \beta^{a'+b-c} \left(1 - \frac{y}{x}\right) y + \frac{y}{x} (y + \alpha) \\
= (\alpha + \beta)^{c-b-1} \beta^{a'+b-c+a} \left(\frac{y}{\beta}\right) (x + \alpha) \\
= (\alpha + \beta)^{c-b-1} \beta^{d+c'} \left(\frac{y}{\beta}\right) (x + \alpha).
\]

We do not need to consider the case where \(a' = 0\) as the global condition forces \(c'\) to negative.

\(c' = (a - d) + (b - c) < 0.\)

**A.2.4. Assume** \(a \leq d\) \(\text{and} b \geq c\). Assume \(a' > 0\).

\(\text{LHS} = \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{a} \left(\frac{y}{\beta}\right) \beta^{a'+1} x^\beta + \\
\sum_{g=1}^{a-1} \left(1 - \frac{y}{x}\right) \left(1 - \frac{y}{\beta}\right)^{g-1} \left(\frac{y}{\beta}\right) \beta^{g-1} y \beta^{a'+g-1} x^\beta + \\
\frac{y}{x} \beta^{a-1} y \beta^{a'-1} x^\beta \\
= \left(y^\beta \beta^{a'-1}\right) (x^\beta \beta^{a-1}).
\)
As a result of the assumptions, the Boltzmann weights are fixed. Using the fact that sum of all the possible states is an eigenvector, we get that \( \text{RHS} \) is just the product of the two fixed Boltzmann weights.

\[
\text{RHS} = (x^{\beta-1})(y^{\beta'-1}).
\]

Assume \( \alpha' = 0 \).

\[
\text{LHS} = \left(1 - \frac{y}{x}\right)\left(1 - \frac{y}{\beta}\right)^{a-1}x + \sum_{g=1}^{a-1} \left(1 - \frac{y}{x}\right)\left(1 - \frac{y}{\beta}\right)^{a-g-1}y^{\beta-a-g-1}x + \beta^{a-1}x.
\]

On the \( \text{RHS} \), it’s just the Boltzmann weight of the top vertex.

\[
\text{RHS} = \beta^{a-1}x.
\]

A.3. RLL for the column model of \( g^{(\alpha, \beta)}_\lambda \). We recall the Boltzmann weights, and the entries of \( \tilde{r} \)-matrix.

(67)

\[
w_x \left( \begin{array}{c} \begin{array}{c} a \ \ d \\ b \ \ c \end{array} \end{array} \right) \equiv w_x(a,b;c,d) = \delta_{\alpha,\beta} x^{(\alpha+\beta)}-d(x+\alpha)^d \begin{cases} 0 & 0 < a > d \\ 1 & 0 < a \leq d \\ a = 0 \end{cases} \]

(68)

\[
\tilde{r}^{k,l}_{i,j}(x,y) = \begin{array}{ll} k \leq i \leq l \end{array} = \begin{cases} 0 & i < j \quad k = l = 0 \\ 1 & k = l > 0 \\ x y \left( \begin{array}{c} x + \alpha \\ x + \alpha - 1 \end{array} \right)^{-k} (y + \alpha) (x + \alpha)^{-k} \end{cases}
\]

Firstly, on the \( \text{LHS} \) the Boltzmann weight of the bottom vertex is fixed based on the relation between \( b \) and \( c \). Similarly, from the \( \text{RHS} \), we see that the weight of the top vertex is determined by the relation between \( a \) and \( d \). So, we need to assume certain relations between \( b \) and \( c \), and \( a \) and \( d \). Furthermore, observe that entries of the \( \tilde{r} \) matrix depends on whether the top left label is equal to or greater than 0.
Therefore, in each subsection we assume some combination of relations between \( b \) and \( c \), and \( a \) and \( d \), and a condition on \( c' \).

A.3.1. Assume \( a > d \) and \( b > c \). As \( a > d \), we get that \( a > 0 \). Similarly, from the bottom label of the top vertex on LHS, we get that \( c' \neq 0 \). Therefore, we only need to consider the case where \( a > 0 \) and \( c' > 0 \).

\[
LHS = \left( \frac{x}{y} \right) \left( \frac{y + a}{x + a} \right)^{1-a} y(y + a)^{a'-1} + 
\sum_{g=a+1}^{a+a'-1} \left( \frac{x}{y} \right) \left( \frac{y - x}{y + a} \right) \left( \frac{y + a}{x + a} \right)^{1-a} y(y + a)^{g-d-1}(x + a)^{a'+g-1} \]
\[
= \left( \frac{x}{y} \right) \beta(a + \beta)^{a'-d-1}(y + a)^{d-a} x(x + a)^{a'+1} \left( \frac{y - \beta}{x - \beta} \right) \]

We compute the RHS. Observe that we need assume a condition on \( c' \). First, let us assume that \( c' > 0 \).

Observe that the label \( a' + b - g \) has to be positive. Based on our assumptions and the global condition, we conclude that

\[
a' + b - (c + c') = d - a < 0.
\]

Therefore, the range of \( g \) is from \( a \) to \( a' + b \).

\[
RHS = \beta(a + \beta)^{a'-d-1}(x + a)^{d-a}(y + a)^{a'-1} + \sum_{g=a+1}^{b} \left( \frac{x}{y} \right) \left( \frac{y + a}{x + a} \right)^{1+g-c'-c} \left( \frac{y - x}{y + a} \right) y(y + a)^{a'-1} + 
\sum_{g=b+1}^{a'+b} \left( \frac{x}{y} \right) \left( \frac{y + a}{x + a} \right)^{1+g-c'-c} \left( \frac{y - x}{y + a} \right) \beta(a + \beta)^{g-b-1}(y + a)^{a'+b-g}
\]
\[
= \left( \frac{x}{y} \right) \left( \frac{y - \beta}{x - \beta} \right) x(x + a)^{a'+a'-1}(y + a)^{d-a} \beta(a + \beta)^{a'-d-1} + 
\left( \frac{x}{y} \right) \left( \frac{y - x}{y + a} \right) \beta^2(x + a)^{d-a} x(x + a)^{a'-1}(y + a)^{a'+a'-1}.
\]

A.3.2. Assume \( a \leq d \) and \( b \geq c \). We now assume that \( a > 0 \) and \( c' > 0 \).

\[
LHS = \left( \frac{x}{y} \right) \left( \frac{y + a}{x + a} \right)^{1-a} y(y + a)^{a'-1} x(x + a)^{a'-1} + 
\sum_{g=a+1}^{d} \left( \frac{x}{y} \right) \left( \frac{y - x}{y + a} \right) \left( \frac{y + a}{x + a} \right)^{1-a} y(y + a)^{g-1} x(x + a)^{a'+a-g-1} + 
\sum_{g=d+1}^{a+a'-1} \left( \frac{x}{y} \right) \left( \frac{y - x}{y + a} \right) \left( \frac{y + a}{x + a} \right)^{1-a} \beta(a + \beta)^{g-d-1} y(y + a)^{a'+a-g-1} x(x + a)^{a'+a-g-1}
\]
\begin{align*} 
\left( \frac{x}{y} \right) & \left( y - x \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1-a} \beta(\alpha + \beta)^{a+\alpha'-d-1}(y + \alpha)^d \\
\left( \frac{x^2}{y} \right) & \left( \frac{y - \beta}{x - \beta} \right)(y + \alpha)^{d-a}(x + \alpha)^{2a + a' - d - 2 -} \\
\left( \frac{x}{y} \right) & \beta(\alpha + \beta)^{a+\alpha'-d-1}(y + \alpha)^{d-a}(x + \alpha)^{a-1}. 
\end{align*}

\[ \text{RHS} \frac{x(x + \alpha)^{a-1}}{x(x + \alpha)^{a-1}} = \left( \frac{x}{y} \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1-c'} y(y + \alpha)^{a'-1} + \\
\sum_{g=c+1}^{b} \left( \frac{x}{y} \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1+g-c'-c} \left( \frac{y - x}{y + \alpha} \right) y(y + \alpha)^{a'-1} + \\
\sum_{g=b+1}^{c+c'-1} \left( \frac{x}{y} \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1+g-c'-c} \left( \frac{y - x}{y + \alpha} \right) \beta(\alpha + \beta)^{g-b-1}(y + \alpha)^{a'+b-g} + \\
\left( 1 - \frac{x}{y} \right) \beta(\alpha + \beta)^{c+c'-b-1}(y + \alpha)^{a'+b-c-c'} \\
\text{RHS} \left( \frac{x^3}{y} \right) \left( \frac{y - \beta}{x - \beta} \right)(y + \alpha)^{d-a}(x + \alpha)^{2a + a' - d - 2} - \\
\left( \frac{x}{y} \right) \left( \frac{y - x}{x - \beta} \right) \beta(\alpha + \beta)^{a+\alpha'-d-1}(y + \alpha)^{d-a}(x + \alpha)^{a-1}. 
\] 

Assume \( a = 0 \) and \( c' > 0 \).

\[ \text{LHS} = x(x + \alpha)^{a-1} + \sum_{g=1}^{d} \left( 1 - \frac{x}{y} \right) y(y + \alpha)^{g-1} x(x + \alpha)^{a'-g-1} + \\
\sum_{g=d+1}^{a'-1} \left( 1 - \frac{x}{y} \right) \beta(\alpha + \beta)^{g-d-1}(y + \alpha)^{d} x(x + \alpha)^{a'-g-1} + \\
\left( 1 - \frac{x}{y} \right) \beta(\alpha + \beta)^{a'-d-1}(y + \alpha)^{d} \\
= x(x + \alpha)^{a'-d-1}(y + \alpha)^{d} \left( \frac{x}{y} \right) \left( \frac{y - \beta}{x - \beta} \right) - \\
\left( 1 - \frac{x}{y} \right) \beta(\alpha + \beta)^{d}(\alpha + \beta)^{a'-d-1} \left( \frac{\beta}{x - \beta} \right). 
\] 

Observe that, while computing RHS in the case where \( a \leq d \), it is only in the final step we multiply the Boltzmann weight of the top vertex. Here, the Boltzmann weight of the top vertex is 1. Therefore,

\[ \text{RHS} = \left( \frac{x}{y} \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1+b-c-c'} (y + \alpha)^{a'-1} \left( \frac{(y - \beta)x}{x - \beta} \right) - \\
\beta(\alpha + \beta)^{c+c'-b-1}(y + \alpha)^{d} \left( \frac{\beta(y - x)}{y(x - \beta)} \right) \\
= \left( \frac{x^2}{y} \right) \left( \frac{y - \beta}{x - \beta} \right)(y + \alpha)^{d}(x + \alpha)^{a'-d-1} - \\
\left( 1 - \frac{x}{y} \right) \left( \frac{\beta}{x - \beta} \right) \beta(\alpha + \beta)^{d}(\alpha + \beta)^{a'-d-1}. 
\]
In the above computation we have assumed $a' > d$. We now consider the case where $a' \leq d$.

$$LHS = x(x + \alpha)^{a'-1} + \sum_{g=1}^{a'-1} \left(1 - \frac{x}{y}\right) y(y + \alpha)^{g-1} x(x + \alpha)^{a'-g-1} + \left(1 - \frac{x}{y}\right) y(y + \alpha)^{a'-1} = y(y + \alpha)^{a'-1}.$$ 

On the RHS, the weights of the vertices are fixed for all $g$. Since the right boundary of the cross is fixed and the fact that sum of all possible states is an eigenvector, we get that $RHS = y(y + \alpha)^{a'-1}$.

A.3.3. Assume $a > d$ and $b < c$. As $a > d$, we will have $a > 0$. Also $c' > 0$, otherwise we get contradiction on the range of $g$.

$$LHS = \left(\frac{x}{y}\right) \left(\frac{y + \alpha}{x + \alpha}\right)^{1-a} \beta(\alpha + \beta)^{d-1}(y + \alpha)d(\alpha + \beta)^{a' + a'd'}(x + \alpha)^{d + c' - a} + \sum_{g=a+1}^{d+c'} \left(\frac{x}{y}\right) \left(\frac{y + \alpha}{x + \alpha}\right)^{1-a} \left(\frac{y - x}{y + \alpha}\right) \beta(\alpha + \beta)^{g-1}(y + \alpha)d(\alpha + \beta)^{a' + a'd'}(x + \alpha)^{d + c' - g} = \left(\frac{x}{y}\right) \beta^2(\alpha + \beta)^{a' + c' - b - d - 2}(y + \alpha)d(\alpha + \beta)^{a' + a'd'}(x + \alpha)^{d + c' - 1} + \left(\frac{x}{y}\right) \left(\frac{y - x}{y + \alpha}\right) \beta^2(\alpha + \beta)^{a' + b - c - d - 1}(y + \alpha)d(\alpha + \beta)^{a' + a'd'}(x + \alpha)^{d + c' - a + 1} \left(\frac{1 - \left(\frac{\alpha + \beta}{x + \alpha}\right)^{d + c' - a}}{x - \beta}\right).$$

We compute the RHS:

$$RHS = \left(\frac{x}{y}\right) \left(\frac{y + \alpha}{x + \alpha}\right)^{1-c'} \beta(\alpha + \beta)^{d-1}(y + \alpha)d(\alpha + \beta)^{a' + b - c}(x + \alpha)^{a' + b - c} + \sum_{g=c+1}^{a' + b} \left(\frac{x}{y}\right) \left(\frac{y - x}{y + \alpha}\right) \beta(\alpha + \beta)^{g-1}(y + \alpha)d^{a' + b - g} = \left(\frac{x}{y}\right) \beta^2(\alpha + \beta)^{a' + b - d - 2}(y + \alpha)d(\alpha + \beta)^{a' + b - d - 1}(x + \alpha)^{d + c' - a + 1} + \left(\frac{x}{y}\right) \left(\frac{y - x}{y + \alpha}\right) \beta^2(\alpha + \beta)^{a' + b - d - 1}(y + \alpha)d(\alpha + \beta)^{a' + b - d - 1}(x + \alpha)^{d + c' - 1}.$$
A.3.4. Assume \( a \leq d \) and \( b < c \). Assume \( a > 0 \) and \( c' > 0 \).

Recall that we have the global condition \( a + a' + b = c + c' + d \). Observe that, because of the assumptions, the range of \( g \) on the LHS is \( a + d + c' \).

On the RHS, we have \( a' + b - c - c' = d - a \geq 0 \). Therefore, the range of \( g \) is from \( c \) to \( c + c' \)

\[
\text{LHS} = \left( \frac{x}{y} \right) \frac{(y + \alpha)_{1-a}}{x + \alpha} g(y + \alpha)^{a-1} \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d+c'-a+} \\
\sum_{g=d+1}^{d+c'} \left( \frac{x}{y} \right) \frac{(y + \alpha)_{1-a}}{x + \alpha} g(y + \alpha)^{g-1} \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d+c'-g+} \\
\sum_{g=d+1}^{d+c'} \left( \frac{x}{y} \right) \frac{(y + \alpha)_{1-a}}{x + \alpha} \beta(\alpha + \beta)^{g-d-1} \\
(y + \alpha)^d \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d+c'-g} \\
= \left( \frac{x^2}{y} \right) \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d-a}(x + \alpha)^{a+c'-1} \left( \frac{y - \beta}{x - \beta} \right) - \\
\left( \frac{x}{y} \right) \beta^2 \left( \frac{y - x}{x - \beta} \right) (\alpha + \beta)^{c+c'-b-1}(y + \alpha)^{d-a}(x + \alpha)^{a-1}.
\]

\[
\text{RHS} = \frac{(x + \alpha)^{a-1}}{x(x + \alpha)^{a-1}} = \left( \frac{x}{y} \right) \frac{(y + \alpha)_{1-c'}}{x + \alpha} \beta(\alpha + \beta)^{c-b-1}(y + \alpha)^{a'+b-c+} \\
\sum_{g=c+1}^{c+c'-1} \left( \frac{x}{y} \right) \frac{(y + \alpha)_{1-g-c'}}{x + \alpha} \left( \frac{y - x}{y + \alpha} \right) \beta(\alpha + \beta)^{g-b-1}(y + \alpha)^{a'+b-c+g+} \\
\left( 1 - \frac{x}{y} \right) \beta(\alpha + \beta)^{c+c'-b-1}(y + \alpha)^{a'+b-c-c'} \\
\left( \frac{x^2}{y} \right) \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{a+c'-1} \left( \frac{y - \beta}{x - \beta} \right) - \\
\left( \frac{x}{y} \right) \beta^2 (\alpha + \beta)^{c+c'-b-1}(y + \alpha)^{a'+b-c-c'} (x + \alpha)^{a-1} \left( \frac{y - x}{x - \beta} \right).
\]

Assume \( a = 0 \) and \( c' > 0 \).

\[
\text{LHS} = \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d+c'} + \\
\sum_{g=1}^{d} \left( 1 - \frac{x}{y} \right) g(y + \alpha)^{g-1} \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d+c'-g+} \\
\sum_{g=d+1}^{d+c'} \left( 1 - \frac{x}{y} \right) \beta(\alpha + \beta)^{g-d-1}(y + \alpha)^d \beta(\alpha + \beta)^{c-b-1}(x + \alpha)^{d+c'-g} \\
= \left( \frac{x}{y} \right) \beta(\alpha + \beta)^{c-b-1}(y + \alpha)^d \left( \frac{y - \beta}{x - \beta} \right) -
\]
\[ (1 - \frac{x}{y}) \left( \frac{\beta^2}{x - \beta} \right) (\alpha + \beta)^{c + b - 1} (y + \alpha)^d. \]

Observe when \( a = 0 \), the only difference in the RHS from the earlier case is the weight of the Boltzmann weight of the top vertex.

\[
RHS = \left( \frac{x}{y} \right) \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^d (x + \alpha)^c \left( \frac{y - \beta}{x - \beta} \right) - \left( 1 - \frac{x}{y} \right) \left( \frac{\beta^2}{x - \beta} \right) (\alpha + \beta)^{c + c' - b - 1} (y + \alpha)^d.
\]

Assume \( a = 0 \) and \( c' = 0 \).

\[
LHS = \beta (\alpha + \beta)^{c - b - 1} (x + \alpha)^{d + c'} + \sum_{g=1}^{d} \left( \frac{1 - x}{y} \right) y (y + \alpha)^{g - 1} \beta (\alpha + \beta)^{c - b - 1} (x + \alpha)^{d - g} + \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^d.
\]

On the RHS, there is a unique configuration.

\[
RHS = \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^{a' + b - c} = \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^d \begin{pmatrix} d & 0 & 0 \\ 0 & 0 & 0 \\ a' & b & c \end{pmatrix}.
\]

Assume \( a > 0 \) and \( c' = 0 \).

\[
LHS = \left( \frac{x}{y} \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1-a} y (y + \alpha)^{a - 1} \beta (\alpha + \beta)^{c - b - 1} (x + \alpha)^{d + c' - a} + \sum_{g=a+1}^{d} \left( \frac{x}{y} \right) \left( \frac{y - x}{y + \alpha} \right) \left( \frac{y + \alpha}{x + \alpha} \right)^{1-a} y (y + \alpha)^{g - 1} \beta (\alpha + \beta)^{c - b - 1} (x + \alpha)^{d + c' - g} + \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^{d - a} (x + \alpha)^{a - 1}.
\]

Just like in the previous case, we have a unique configuration.

\[
RHS = x (x + \alpha)^{a - 1} \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^{a' + b - c} = x (x + \alpha)^{a - 1} \beta (\alpha + \beta)^{c - b - 1} (y + \alpha)^{d - a}.
\]

A.4. RLL FOR THE CAUCHY IDENTITY. We prove a relation between \( L^* \), the dual \( L \) matrix of row model of \( G_{\lambda}^{(\alpha - \beta)} \), and \( l \), from the row model of \( g_{\lambda}^{(\alpha, \beta)} \).

For convenience let us recall all the characters of the play:

\[
\begin{array}{cccccc}
0 & 1 & 0 & 1 & 0 & 0 \\
0 & m & 0 & m & 0 & m \\
1 & \frac{1 - \beta x}{1 + \alpha x} & 1 & \frac{1 - \beta x}{1 + \alpha x} & 1 & \frac{1 - \beta x}{1 + \alpha x} \\
\end{array}
\]

Algebraic Combinatorics, Vol. 6 #1 (2023) 152
Vertex models for Canonical Grothendieck polynomials and their duals

\[ w_x \begin{pmatrix} a & d \\ b & c \end{pmatrix} \equiv w_x(a, b; c, d) = \delta_{a+b, c+d} \begin{cases} (\alpha + \beta)^{a+b-1}(x + \alpha)\beta^d & a > d \\ \beta^{a-1}x & 0 < a \leq d \\ 1 & a = 0, \end{cases} \]

where \( a, b, c, d \in \mathbb{Z}_{\geq 0} \).

The \( R \) matrix \( \in \text{End}(F \otimes W) \):

\[ R_{i, j}^{k, l} = \begin{cases} 1 - xy & j = k = 1, i = l = 0 \\ xy & k = l = 0, i = j = 1 \\ 1 - x\beta & k = 1 \\ x\beta & k = 0 \\ 1 & i = k = l = j = 0 \end{cases} \]

where \( k, j \in \{0, 1\} \) and \( i, l \in \mathbb{Z}_{\geq 0} \).

Based on the entries of the \( R \)-matrix, we shall assume certain condition on \( a, a' \) and \( c, c' \). We shall divide the conditions of \( a, a' \) into three cases, \( a + a' > 1 \). We do the same with the conditions on \( c, c' \). Therefore, we shall have a total of 9 cases to consider.

A.4.1. Assume \( a + a' = 0 \) and \( c + c' = 0 \). Observe that, because of the global condition, \( b \) is equal to \( d \). So each side will have a unique configuration with identical weight.

A.4.2. Assume \( a + a' = 0 \). When \( a + a' = 0 \), there is a unique configuration on LHS. On the RHS we use the fact that sum of all the possibilities states is an eigenvector of the \( R \) matrix to conclude that the weight of RHS is just the product of the Boltzmann weights.

A.4.3. Assume \( a = 0 \) and \( a' = 1 \) and \( c + c' = 0 \).

\[ LHS = (x\beta) y \left( \frac{x}{1 + ax} \right) + (xy) \left( \frac{1 - \beta x}{1 + ax} \right) = \frac{xy}{1 + ax} \]

\[ \begin{pmatrix} b+1 & 0 \\ b & 1 \end{pmatrix} + \begin{pmatrix} b+1 & 0 \\ b & 1 \end{pmatrix} = \begin{pmatrix} b+1 & 0 \\ b & 1 \end{pmatrix}. \]
A.4.4. Assume $a = 0$ and $a' = 1$ and $c' = 0$ and $c = 1$. When $b > 0$:

\[
LHS = (x\beta) y \left( \frac{x}{1+\alpha x} \right) + xy \left( \frac{1-\beta x}{1+\alpha x} \right) \quad \begin{pmatrix}
0 & 1 & 0 & 0 \\
1 & 0 & 0 & 0 \\
b & b+1 & 0 & 0 \\
1 & 1 & 1 & 1
\end{pmatrix}.
\]

When $b = 0$, only the second configuration from the left is a valid configuration.

\[
LHS = xy
\]

When $b > 0$, we have:

\[
RHS = \begin{pmatrix}
0 & 1 & 0 & 0 \\
1 & 0 & 0 & 0 \\
b & b+1 & 0 & 0 \\
1 & 1 & 1 & 1
\end{pmatrix}.
\]

When $b = 0$,

\[
RHS = \begin{pmatrix}
0 & 1 & 0 & 0 \\
1 & 0 & 0 & 0 \\
b & b+1 & 0 & 0 \\
1 & 1 & 1 & 1
\end{pmatrix}.
\]

A.4.5. Assume $a = 0$ and $a' = 1$ and $c + c' \geq 1$ and $c' \neq 0$.

\[
LHS = (x\beta) y \left( \frac{x}{1+\alpha x} \right) + xy \left( \frac{1-\beta x}{1+\alpha x} \right) \quad \begin{pmatrix}
b-c'+1 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 \\
b & b+1 & 0 & 0 \\
1 & 1 & 1 & 1
\end{pmatrix}.
\]
Vertex models for Canonical Grothendieck polynomials and their duals

\[ \text{RHS} = \begin{pmatrix}
\begin{array}{ccc}
0 & 0 & 1 \\
1 & b & c + c' \\
b & 0 & 0
\end{array} & \begin{array}{ccc}
0 & 0 & 1 \\
1 & b & c + c' \\
b & 0 & 0
\end{array}
\end{pmatrix} + \begin{pmatrix}
\begin{array}{ccc}
b - c - c' + 1 & 0 & 1 \\
0 & b - c - c' + 1 & 0 \\
b & 0 & 0
\end{array}
\end{pmatrix}
\]

\[ = \left( \frac{x}{1 + \alpha x} \right) g(x \beta) + \left( \frac{x}{1 + \alpha x} \right) g(1 - x \beta)
\]

\[ = \frac{xy}{1 + \alpha x}.
\]

A.4.6. Assume \( a = 1 \) and \( a' = 0 \) and \( c + c' = 0 \).

\[ \text{LHS} = \begin{pmatrix}
\begin{array}{ccc}
0 & 0 & 1 \\
1 & b + 1 & 0 \\
b & 0 & 0
\end{array} & \begin{array}{ccc}
0 & 0 & 1 \\
1 & b + 1 & 0 \\
b & 0 & 0
\end{array}
\end{pmatrix}
\]

\[ = (1 - x \beta) g\left( \frac{x}{1 + \alpha x} \right) + (1 - xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]

\[ = \frac{1 - \beta x}{1 + \alpha x}.
\]

\[ \text{RHS} = \left( \frac{1 - \beta x}{1 + \alpha x} \right) \begin{pmatrix}
\begin{array}{ccc}
0 & 0 & 1 \\
1 & b + 1 & 0 \\
b & 0 & 0
\end{array}
\end{pmatrix}
\]

A.4.7. Assume \( a = 1 \) and \( a' = 0 \) and \( c = 1 \) and \( c' = 0 \). When \( b > 0 \), we have

\[ \text{LHS} = \begin{pmatrix}
\begin{array}{ccc}
0 & 0 & 1 \\
1 & b - 1 & 0 \\
b & 0 & 0
\end{array} + \begin{array}{ccc}
0 & 0 & 1 \\
1 & b & 0 \\
b & 0 & 0
\end{array}
\end{pmatrix}
\]

\[ = (1 - x \beta) g\left( \frac{x}{1 + \alpha x} \right) + (1 - xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]

\[ = \frac{1 - \beta x}{1 + \alpha x}.
\]

When \( b = 0 \), only the second configuration from the left is valid.

\[ \text{LHS} = (1 - xy).
\]

When \( b > 0 \), we have

\[ \text{RHS} = \left( xy \right) \left( \frac{1 - \beta x}{1 + \alpha x} \right) + \left( 1 - xy \right) \left( \frac{1 - \beta x}{1 + \alpha x} \right) \begin{pmatrix}
\begin{array}{ccc}
0 & 0 & 1 \\
1 & b & 0 \\
b & 0 & 0
\end{array} + \begin{array}{ccc}
0 & 0 & 1 \\
1 & b & 0 \\
b & 0 & 0
\end{array}
\end{pmatrix}
\]

\[ = \frac{1 - \beta x}{1 + \alpha x}.
\]
When $b = 0$, only the second configuration is valid.

\[ \text{RHS} = 1 - xy. \]

**A.4.8.** Assume $a = 1$ and $a' = 0$ and $c + c' \geq 1$ and $c' \neq 0$. When $b - c - c' + 1 \geq 1$, we have

\[
LHS = (1 - x\beta)(y + \alpha) \left( \frac{x}{1 + \alpha x} \right) + (1 - xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]

\[ = \frac{1 - \beta x}{1 + \alpha x} \left( \begin{array}{cc} b-c-c'+1 & b-c-c'+1 \\ 1 & 0 \\ 0 & 1 \\ 0 & b \end{array} \right) \]

When $b - c - c' + 1 = 0$, we have

\[
LHS = (1 - x\beta)(y + \alpha) \left( \frac{x}{1 + \alpha x} \right) + (1 - xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]

\[ = (1 - x\beta). \]

When $b - c - c' + 1 \geq 1$, we have

\[
RHS = (x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right) + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]

\[ = \left( \frac{1 - \beta x}{1 + \alpha x} \right) \left( \begin{array}{cc} b-c-c'+1 & b-c-c'+1 \\ 1 & 0 \\ 0 & 1 \\ 0 & b \end{array} \right) \]

When $b - c - c' + 1 = 0$, only the second configuration survives.

\[ \text{RHS} = (1 - x\beta). \]

**A.4.9.** Assume $a + a' > 1$ and $c + c' = 0$.

When $a = 0$,

\[
LHS = (x\beta) y \beta^{a'-1} \left( \frac{x}{1 + \alpha x} \right) + (x\beta) y \beta^{a'-2} \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]

\[ = (xy) \beta^{a'} \left( \frac{x}{1 + \alpha x} \right) + xy \beta^{a'-1} \left( \frac{1}{1 + \alpha x} \right) - (xy) \beta^{a'} \left( \frac{x}{1 + \alpha x} \right)
\]

\[ = \frac{xy \beta^{a'-1}}{1 + \alpha x}. \]
When $a = 1$,
\[
LHS = (1 - x\beta) y^{\beta a'} \left( \frac{x}{1 + \alpha x} \right) + (1 - x\beta) y^{\beta a' - 1} \left( \frac{1 - \beta x}{1 + \alpha x} \right)
\]
\[
= (1 - x\beta) y^{\beta a' - 1} \left( \frac{x\beta}{1 + \alpha x} + \frac{1 - \beta x}{1 + \alpha x} \right)
\]
\[
= (1 - x\beta) y^{\beta a' - 1} \left( \frac{1 + \beta x}{1 + \alpha x} \right).
\]

When $a = 0$,
\[
RHS = \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1}.
\]

When $a = 1$,
\[
RHS = \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1}.
\]

A.4.10. Assume $a + a' > 1$ and $c = 1$ and $c' = 0$.

When $a = 0$ and $b > 0$
\[
LHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1} + (x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 2}
\]
\[
= \frac{xy^{\beta a' - 1}}{1 + \alpha x}.
\]

When $a = 0$ and $b = 0$
\[
LHS = x\beta (y^{\beta a' - 2})
\]
\[
= x y^{\beta a' - 1}.
\]

When $a = 1$ and $b > 0$
\[
LHS = (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a'} + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1}
\]
\[
= \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1}.
\]

When $a = 1$ and $b = 0$
\[
LHS = (1 - x\beta) y^{\beta a' - 1}.
\]
When $a = 0$ and $b > 0$

$$RHS = (xy) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1} + (1 - xy) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1}$$

$$= \frac{xy y^{\beta a' - 1}}{1 + \alpha x}.$$

When $a = 0$ and $b = 0$

$$RHS = (xy) \left( \frac{x}{1 + \alpha x} \right) (y + a) y^{\beta a' - 1} + (1 - xy) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1}$$

$$= xy y^{\beta a' - 1}.$$

When $a = 1$ and $b > 0$

$$RHS = (xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1} + (1 - xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1}$$

$$= \frac{(1 - \beta x) y^{\beta a' - 1}}{1 + \alpha x}.$$

When $a = 1$ and $b = 0$

$$RHS = (xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right) (y + a) y^{\beta a' - 1} + (1 - xy) \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1}$$

$$= (1 - \beta x) y^{\beta a' - 1}.$$

A.4.11. Assume $a + a' > 1$ and $c + c' \geq 1$ and $c' \neq 0.$
When $a = 0$ and $b - c - c' \geq 0$
\[ LHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1} + (x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) y^{\beta a' - 2}, \]
\[ RHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1} + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right) y^{\beta a' - 1}. \]

When $a = 1$ and $b - c - c' \geq 0$,
\[ LHS = (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a'} + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) y^{\beta a' - 1}, \]
\[ RHS = (x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) y^{\beta a' - 1} + (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1}. \]

When $a = 0$ and $b - c - c' = -1$,
\[ LHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) (y + \alpha) \beta^{a' + b - c - c'} + (x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right) (y \beta^{a' - 2}), \]
\[ RHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) (y + \alpha) \beta^{a'} + (1 - \beta x) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1}. \]

When $a = 1$ and $b - c - c' = -1$,
\[ LHS = (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right) (y + \alpha) \beta^{a'} + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) (y \beta^{a' - 1}), \]
\[ RHS = (x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) (y + \alpha) \beta^{a' - 1} + (1 - \beta x) \left( \frac{x}{1 + \alpha x} \right) y^{\beta a' - 1}. \]

When $a = 0$ and $a' - 1 > a' + b - c - c'$,
\[ LHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) (\alpha + \beta)^{c' + c - b - 1} (y + \alpha) \beta^{a' + b - c - c'} + (x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) (\alpha + \beta)^{c' + c - b - 2} (y + \alpha) \beta^{a' + b - c - c'} , \]
\[ RHS = (x\beta) \left( \frac{x}{1 + \alpha x} \right) (\alpha + \beta)^{c' + c - b - 1} (y + \alpha) \beta^{a' + b - c - c'} + (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right) (\alpha + \beta)^{c' + c - b - 2} (y + \alpha) \beta^{a' + b - c - c' + 1} . \]

When $a = 1$ and $b - c - c' < -1$,
\[ LHS = (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right) (\alpha + \beta)^{c' + c - b - 1} (y + \alpha) \beta^{a' + b - c - c'} + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) (\alpha + \beta)^{c' + c - b - 2} (y + \alpha) \beta^{a' + b - c - c' + 1} , \]
\[ RHS = (x\beta) \left( \frac{1 - \beta x}{1 + \alpha} \right) (\alpha + \beta)^{c' + c - b - 1} (y + \alpha) \beta^{a' + b - c - c'} + (1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right) (\alpha + \beta)^{c' + c - b - 2} (y + \alpha) \beta^{a' + b - c - c' + 1} . \]
In all the cases, we have assumed that the vertex 1 does not appear. Let us now study the conditions on the nodes where such a vertex can occur.

Observe that it can appear in the second configuration of LHS when \(b = 0\) and \(c = 1\). Similarly, it can appear in the second configuration of RHS when \(a + a' + b - c - c' = 0\) and \(a' + b - c - c' + 1 = 0\), which reduces to the conditions \(a = 1\) and \(a' + b - c - c' = -1\).

When \(a = 0\), \(b = 0\) and \(c = 1\),

\[
\text{LHS} = (x \beta)(\alpha + \beta)c'^{-1}(y + \alpha)\beta^{a' - c'}
\]

\[
\begin{pmatrix}
0 & a' & c' \\
a' & 1 & 1 \\
0 & 1 & 0
\end{pmatrix}
\]

\[
\text{RHS} = \left(\frac{x}{1 + \alpha x}\right)(\alpha + \beta)c' (y + \alpha)\beta^{a' - c'}
\]

\[
+ (1 - x \beta) \left(\frac{x}{1 + \alpha x}\right)(\alpha + \beta)c'^{-1}(y + \alpha)\beta^{a' - c'}
\]

\[= x(\alpha + \beta)c'^{-1}(y + \alpha)\beta^{a' - c'}.
\]

When \(a = 1\) and \(b = 0\) and \(c = 1\),

\[
\text{LHS} = (1 - x \beta)(\alpha + \beta)c'^{-1}(y + \alpha)\beta^{a' - c'}
\]

\[
\begin{pmatrix}
a' & 1 & 0 \\
a' & 1 & 1 \\
a' & 0 & 1
\end{pmatrix}
\]
Vertex models for Canonical Grothendieck polynomials and their duals

\[ \text{RHS} = (x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right)(\alpha + \beta)^{c'}(y + \alpha)\beta^{a' - c'} + \\
(1 - \beta x) \left( \frac{1 - \beta x}{1 + \alpha x} \right)(\alpha + \beta)^{c-1}(y + \alpha)\beta^{a' - c'} \\
= (1 - \beta x)(\alpha + \beta)^{c-1}(y + \alpha)\beta^{a' - c'}. \]

When \( a = 1 \) and \( a' + b - c - c' = -1 \),

\[ \text{LHS} = \begin{pmatrix}
1 & a' + 1 \\
0 & a'
\end{pmatrix}
\begin{pmatrix}
y - c' \\
c
\end{pmatrix} + \\
\begin{pmatrix}
1 & a' \\
0 & b - c' + 1
\end{pmatrix}
\begin{pmatrix}
y + c' \\
c
\end{pmatrix}
\]

\[ = (1 - x\beta) \left( \frac{x}{1 + \alpha x} \right)(\alpha + \beta)^{a'}(y + \alpha) + \\
(1 - x\beta) \left( \frac{1 - \beta x}{1 + \alpha x} \right)(\alpha + \beta)^{a'-1}(y + \alpha) \\
= (1 - x\beta)(y + \alpha)(\alpha + \beta)^{a'-1}. \]

\[ \text{RHS} = (1 - x\beta)(\alpha + \beta)^{a'-1}(y + \alpha) \\
\begin{pmatrix}
0 & 1 \\
1 & c + c' - 1
\end{pmatrix}
\begin{pmatrix}
a' \\
c
\end{pmatrix}. \]

When we combine both the conditions, \( b = 0 \) and \( c = 1 \), and \( a = 1 \) and \( a' + b - c - c' = -1 \), we get \( a = 1 \) and \( a' = c' \) for which we have already computed \( \text{LHS} \) and \( \text{RHS} \).

\textbf{Acknowledgements.} The authors would like to thank J. Lamers for a careful reading of the manuscript, and T. Scrimshaw for discussions. The authors would also like to thank the two anonymous reviewers whose comments/suggestions helped improve this manuscript.

\textbf{References}


Ajeeth Gunna, School of Mathematics and Statistics, University of Melbourne, Parkville, Victoria 3010, Australia.

E-mail: agunna@student.unimelb.edu.au

Paul Zinn-Justin, School of Mathematics and Statistics, University of Melbourne, Parkville, Victoria 3010, Australia.

E-mail: pzinn@unimelb.edu.au