

# Integrable Probability

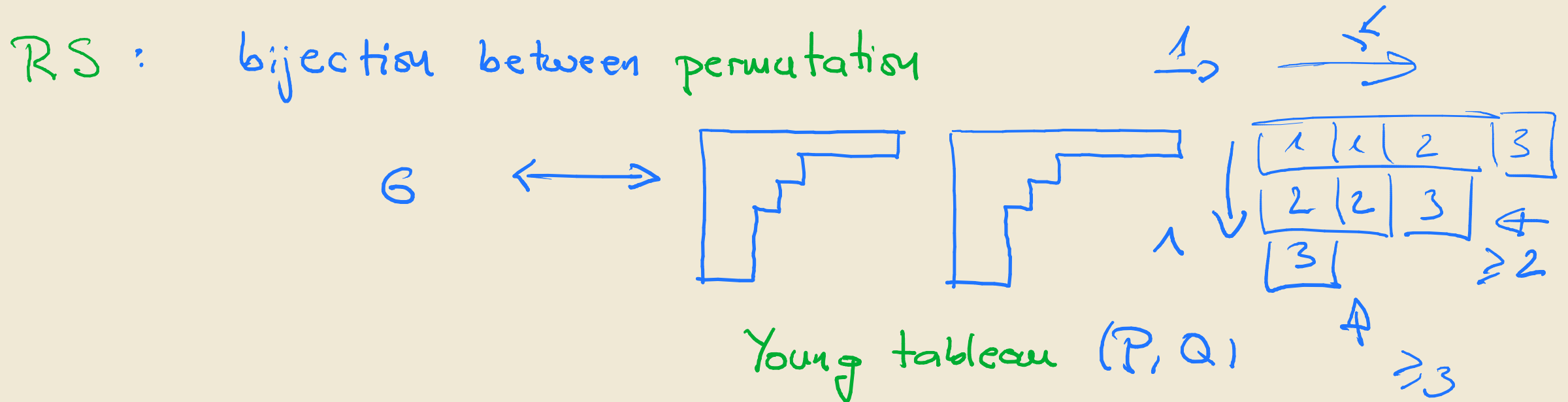
## Lecture 3

Review of RSK. 2

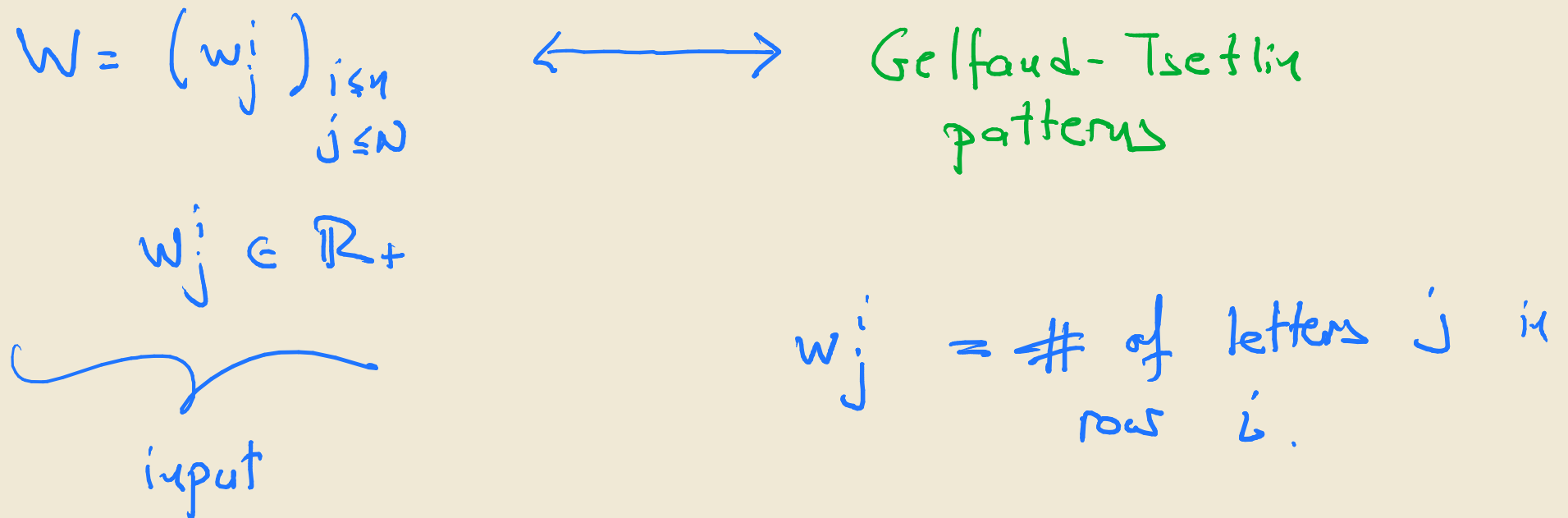
a first example of Integrable Probability =

a solvable Last Passage Percolation

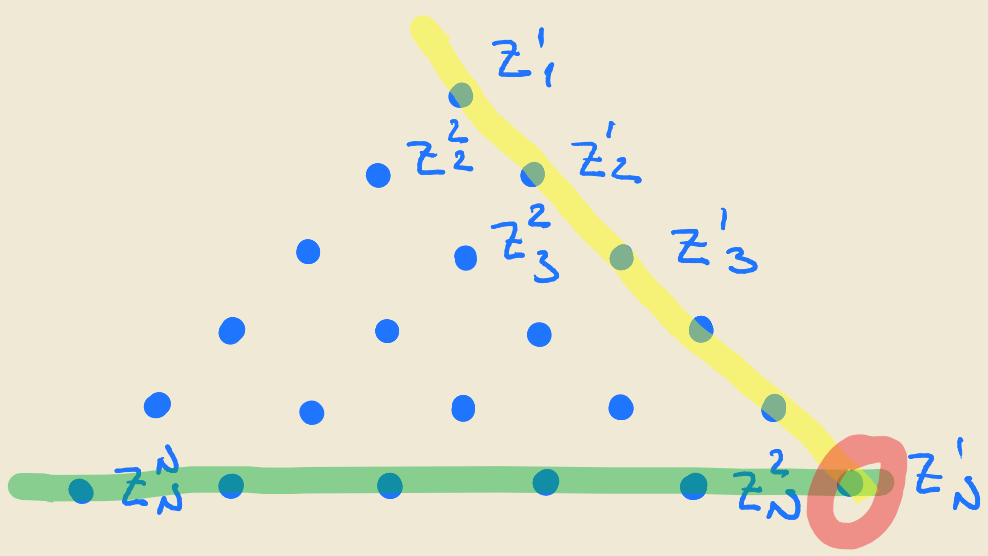
# Review of RSK.



RSK : bijection between matrices



# Definition: Gelfand-Tsetlin patterns



Interlacing triangular patterns

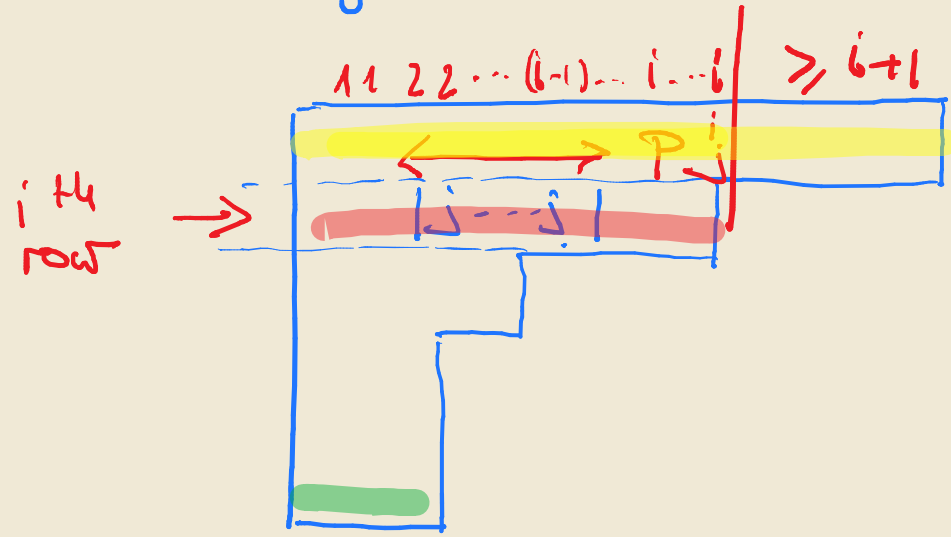
$$z_{j+1}^{i+1} \leq z_j^i \leq z_j^{i+1}$$

left particle

right particle

Correspondence to Young tableaux

$P_j^i = \#$  of  $j$ 's in row  $i$  of Young tableau



In GT-variables

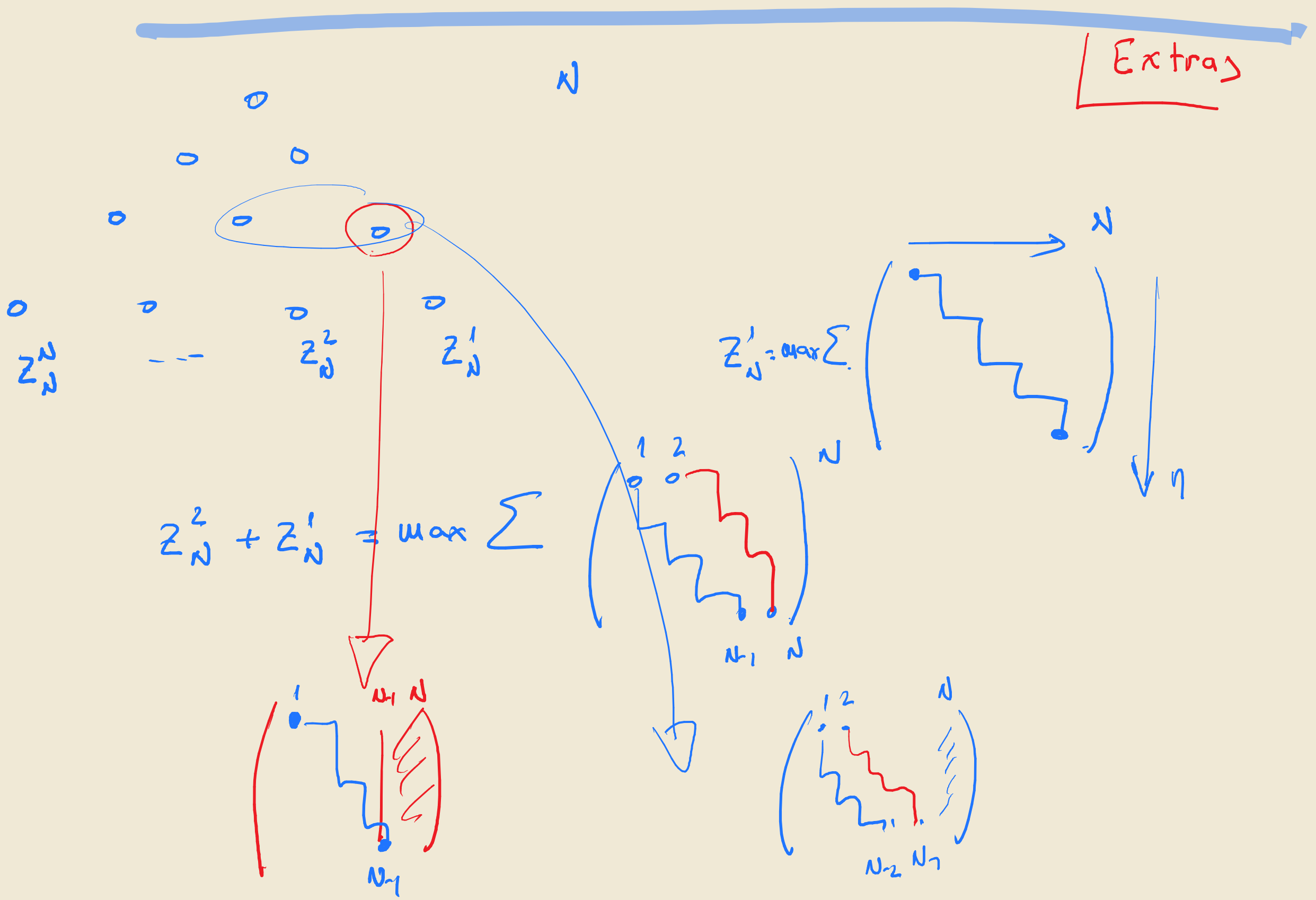
$$z_j^i = P_i^i + P_{i+1}^i + \dots + P_j^i \quad j \geq i$$

= total # entries up to  $j$  in row  $i$

Bottom row of GT = shape of Young tableau

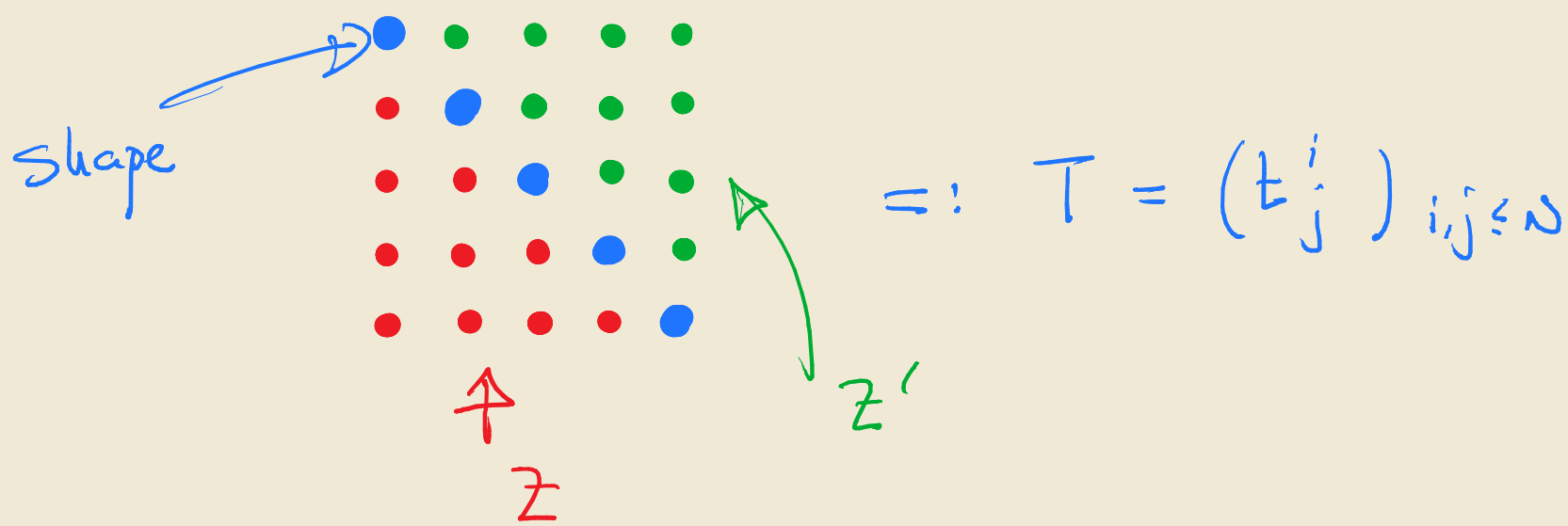
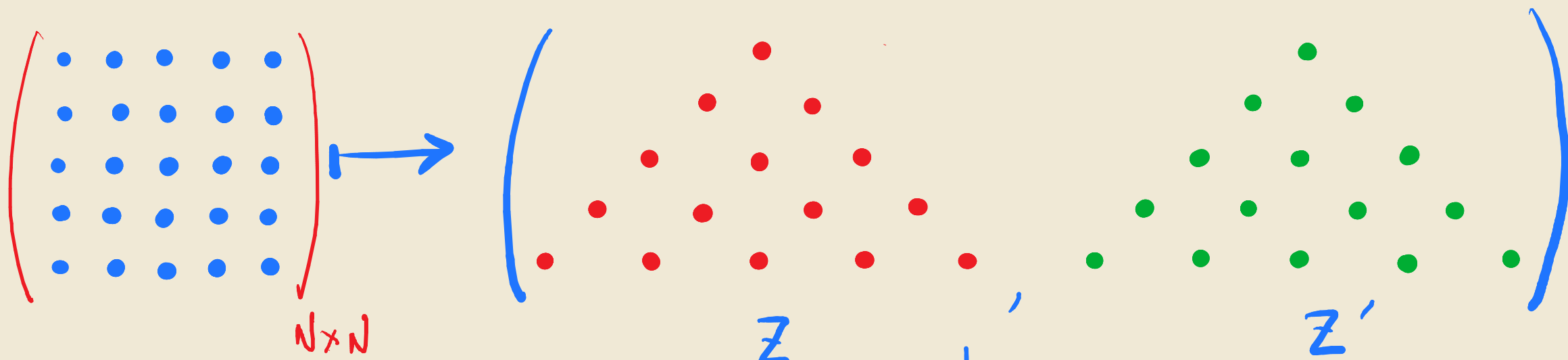
$$z_N^i = P_i^i + \dots + P_N^i$$

= length of  $i$ -th row



# RSK as bijection between matrices

$$W = (w_j^i)_{1 \leq i, j \leq N} \longleftrightarrow (Z, Z') \cong (P, Q)$$

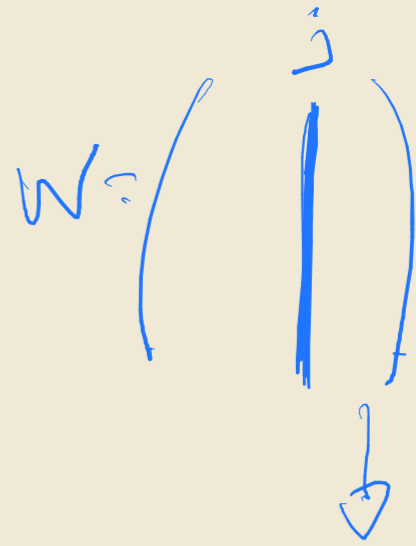


Extras

$$\begin{aligned} \text{RSK}(W) &= (Z, Z') \cong (P, Q) \\ \text{RSK}(W^t) &= (Z', Z) \cong (Q, P) \end{aligned} \quad \left. \vphantom{\begin{aligned} \text{RSK}(W) \\ \text{RSK}(W^t) \end{aligned}} \right\}$$

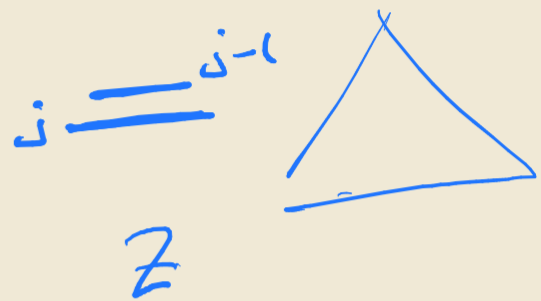
if  $W = W^t$  or  $\sigma = \sigma^{-1}$  then  $P = Q$ .

Towards the integrable model:  
a fundamental property of RSK



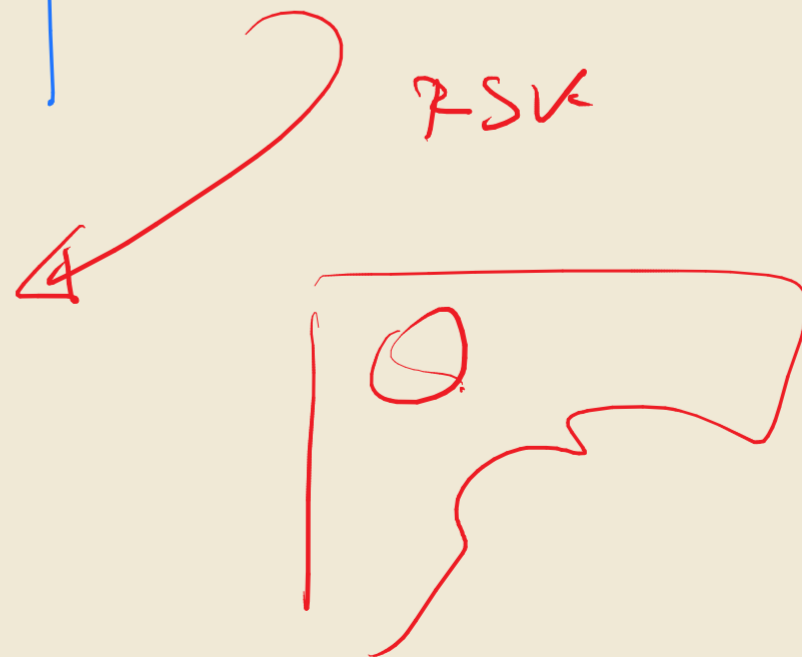
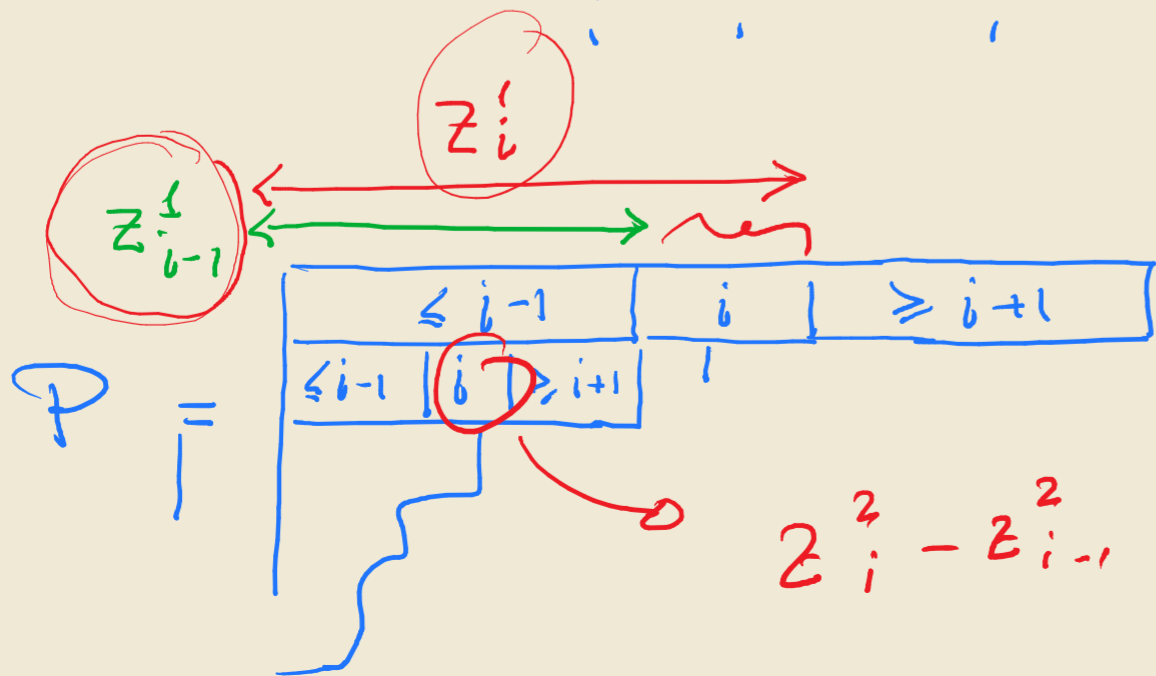
Let  $(Z, Z') = \text{RSK}(W)$  then

$$\sum_{1 \leq i \leq n} w_j^i = \sum_{i=1}^j z_j^i - \sum_{i=1}^{j-1} z_{j-1}^i$$



Explanation 1: In the combinatorial setting  
both sides represent the total  
#  $i$ 's inserted.

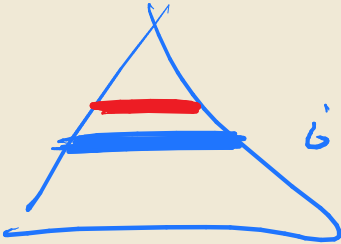
$$(w_j^i) \approx \begin{matrix} 1^{w_1^1} & 2^{w_2^1} & \dots & i^{w_i^1} & \dots \\ 1^{w_1^2} & 2^{w_2^2} & \dots & i^{w_i^2} & \dots \\ \vdots & \vdots & & \vdots & \vdots \end{matrix}$$



# Explanation 2 : Path representation

$$\max \sum \left( \begin{array}{c} 1 \dots j \\ \text{[Diagram of paths from } i \text{ to } i-j+1, i, i+1 \dots j \text{]} \\ i-j+1 \quad i \end{array} \right) = z_i^1 + z_i^2 + \dots + z_i^j$$

if  $j=i$



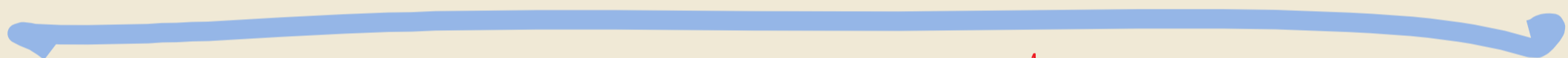
$$\max_{\pi} \sum \left( \begin{array}{c} \text{[Diagram of vertical paths from } i \text{ to } 1 \text{]} \\ 1 \quad \dots \quad i \end{array} \right) = \sum_{\substack{1 \leq k \leq n \\ 1 \leq l \leq i}} w_{e}^k$$

Also,

$$i \left( \begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{array} \right) = (z')^i_i + \dots + (z')^i_i - (z')^i_{i-1} - \dots - z^i_{i-1}$$

Explanation:  $RSK(w^t) = (z', z)$

& apply the previous property



Extras

\*  $RSK(w) = (z, z')$

\*  $RSK(w^t) = (z', z)$

\* bottom row  $z =$  bottom  $z' =$  shape

\*  $\left( \begin{array}{c} \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \\ \curvearrowright \end{array} \right) = z_{n1} = z'_{n1}$

&  $\left( \begin{array}{c} \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \\ \uparrow \end{array} \right) \approx \begin{array}{c} z \\ \triangle \\ \text{---} \\ \triangle \\ j \quad \text{---} \quad j-1 \end{array} \quad \& \quad \left( \begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{array} \right)_i = \begin{array}{c} z' \\ \triangle \\ \text{---} \\ \triangle \\ i-1 \quad \text{---} \quad i \end{array}$

# Integrable Model:

Last Passage Percolation with  
geometric random variables

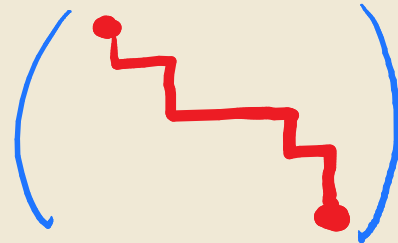
Then let  $W = (W_{ij})_{i \leq j}$  with  $\{W_{ij}\}$  independent

$$\mathbb{P}(W_{ij} = w_{ij}) = (1 - p_i q_j) (p_i q_j)^{w_{ij}} \sim \text{Geom}(p_i q_j)$$

then

$$\mathbb{P}(\tau_N \leq x) = \Gamma_{p,q} \sum_{\lambda: \lambda_1 \leq x} S_\lambda(p) S_\lambda(q)$$

with  $\tau_N = \max_{\pi} \sum \left( \text{diagram} \right)$



$$\Gamma_{p,q} = \prod_{i,j} (1 - p_i q_j)$$

$$S_\lambda(p) = S_\lambda(p_1, \dots, p_N) \text{ Schur function}$$

$$\lambda = \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N \text{ "partition"}$$

Extras

$$\sum_{\substack{(\lambda_1, \lambda_2, \dots, \lambda_N) \\ x \geq \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N}} S_{(\lambda_1, \lambda_2, \dots, \lambda_N)}(p_1, p_2, \dots, p_N) \cdot S_\lambda(q)$$

$S_\lambda(p)$



Proof

$$\begin{aligned}
 & P(\tau_N \leq x) \\
 &= \sum_{\mathbf{w}} P(\mathbf{w}) \mathbb{1}_{\tau_N(\mathbf{w}) \leq x} \\
 &= \prod_{i,j} (1-p_i q_j) \sum_{\mathbf{w}} \prod_{i,j} p_i^{w_{ij}^i} q_j^{w_{ij}^j} \mathbb{1}_{\tau_N(\mathbf{w}) \leq x} \\
 &= \prod_{i,j} (1-p_i q_j) \sum_{\mathbf{w}} \prod_{i,j} p_i^{w_{ij}^i} q_j^{w_{ij}^j} \mathbb{1}_{\tau_N(\mathbf{w}) \leq x} \\
 &= \prod_{i,j} (1-p_i q_j) \sum_{\mathbf{z}, \mathbf{z}'} \prod_{i,j} p_i^{z_i} q_j^{z'_j} \mathbb{1}_{\tau_N(\mathbf{z}, \mathbf{z}') \leq x} \\
 &= \prod_{i,j} (1-p_i q_j) \sum_{\mathbf{z}, \mathbf{z}'} \prod_{i,j} p_i^{z_i} q_j^{z'_j} \mathbb{1}_{\tau_N(\mathbf{z}, \mathbf{z}') \leq x} \\
 &= \prod_{i,j} (1-p_i q_j) \sum_{\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N} \sum_{\mathbf{z}: \text{sk}(\mathbf{z}) = \lambda} \prod_{i,j} p_i^{z_i} q_j^{z'_j} \\
 &= \prod_{i,j} (1-p_i q_j) \sum_{\lambda: \lambda_1 \leq x} S_\lambda(p) S_\lambda(q) \\
 & \quad \lambda_1 \geq \dots \geq \lambda_N
 \end{aligned}$$

change of variables

by definition

by Schur's definition

by one of its definitions

Questions?

$$\text{Exp}(\alpha_i + \beta_j)$$

$$P(X > u) = e^{-\sum_j (\alpha_i + \beta_j) u}$$

$$\prod_{i,j} e^{-x_i} e^{-y_j}$$

$$\prod_j e^{-\sum_i x_i} \prod_i e^{-\sum_j y_j}$$

Extras

for polymer models similar lines lead

$$\mathbb{E} e^{-u Z_N} \propto \int_{\mathbb{R}_{>0}^n} \varphi_\alpha(x) \varphi_\beta(x) \prod \frac{dx_i}{x_i}$$

$$P(w_{ij} \in dx) = \frac{1}{P(\alpha_i + \beta_j)} w_{ij}^{-\alpha_i - \beta_j} e^{-\frac{1}{w_{ij}}} \frac{dw_{ij}}{w_{ij}}$$

# Corollary (Cauchy identity)

$$s_{\lambda, \lambda, \lambda, \dots}(\alpha_1, \alpha_2, \dots)$$

$$\sum_{\lambda} s_{\lambda}(P) s_{\lambda}(Q) = \prod_{i,j} \frac{1}{1 - P_i Q_j}$$

## Combinatorial form of Schur

$$s_{\lambda}(P) = \sum_{T: \text{sh}(T) = \lambda} \prod_{i \text{ in } T} P_i = \text{generating series of } \gamma, T.$$
$$= \sum_{\substack{Z: \text{GT} \\ \text{bottom row} = \lambda}} \prod P_i^{|\mathbb{Z}^i| - |\mathbb{Z}^{i-1}|}$$

## Determinantal formula

$$s_{\lambda}(P) = \frac{\det (P_i^{\lambda_j + N - j})_{1 \leq i, j \leq N}}{\det (P_i^{N - j})_{1 \leq i, j \leq N}}$$

# Determinantal & Asymptotic analysis

## I. Basic tools

- Fredholm determinants
- Cauchy - Binet identity
- Sylvester identity



# Fredholm determinants

$$\iint \det \begin{pmatrix} K(x_1, x_2) & K(x_1, x_n) \\ K(x_2, x_1) & K(x_n, x_1) \end{pmatrix} dx_1 dx_n$$

Let  $K: L^2(X, \mu) \rightarrow L^2(X, \mu)$

$$(Kf)(x) = \int_X K(x, \gamma) f(\gamma) \mu(d\gamma)$$

trace class i.e.  $\int_X |K(x, x)| \mu(dx) < \infty$

the Fredholm determinant

$$\det(I+K)_{L^2(X, \mu)} = 1 + \sum_{n=1}^{\infty} \frac{1}{n!} \int_{X^n} \det(K(x_i, x_j)) \prod_{i=1}^n \mu(dx_i)$$

Intuition: Let  $K$  a finite matrix. Then

$$\det(I+K) = \prod_{i=1}^n (1 + \lambda_i) =$$

$$= 1 + \sum_{k=1}^n \sum_{1 \leq i_1 < i_2 < \dots < i_k \leq n} \lambda_{i_1} \lambda_{i_2} \dots \lambda_{i_k}$$

1<sup>st</sup>  $\sum_{i=1}^n \lambda_i = \text{Tr}(K) = \sum_X K(x, x)$  (Extract)

2<sup>nd</sup>  $\sum_{1 \leq i \neq j \leq n} \lambda_i \lambda_j = \underbrace{(\lambda_1 + \dots + \lambda_n)^2}_{\text{Tr}(K)^2} - \sum_{i=1}^n \lambda_i^2$

$\parallel$   $\sum_{x, \gamma} K(x, x) K(\gamma, \gamma) - \sum_X K(x, x)^2$

$$\sum_{x, \gamma} \det \begin{vmatrix} K(x, x) & K(x, \gamma) \\ K(\gamma, x) & K(\gamma, \gamma) \end{vmatrix}$$

$$\sum_{x, \gamma} K(x, \gamma) K(\gamma, x)$$

$$\det \left( \kappa(x_i, x_j) \right)_{i, j \leq n} \xrightarrow{n} \infty$$



$$\det \left( I + \kappa \right)_{\substack{\uparrow \\ \mathcal{L}^2(X)}}$$