

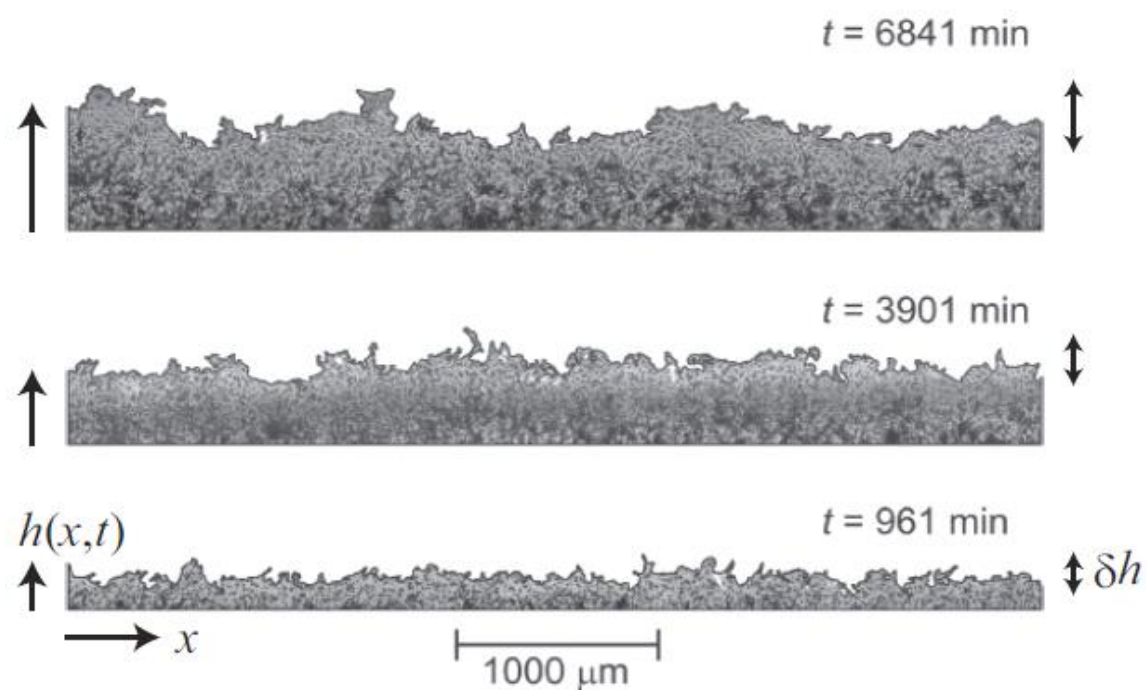
# *Stochastic six vertex model*

*Ivan Corwin (Columbia University)*

# Goals of first hour

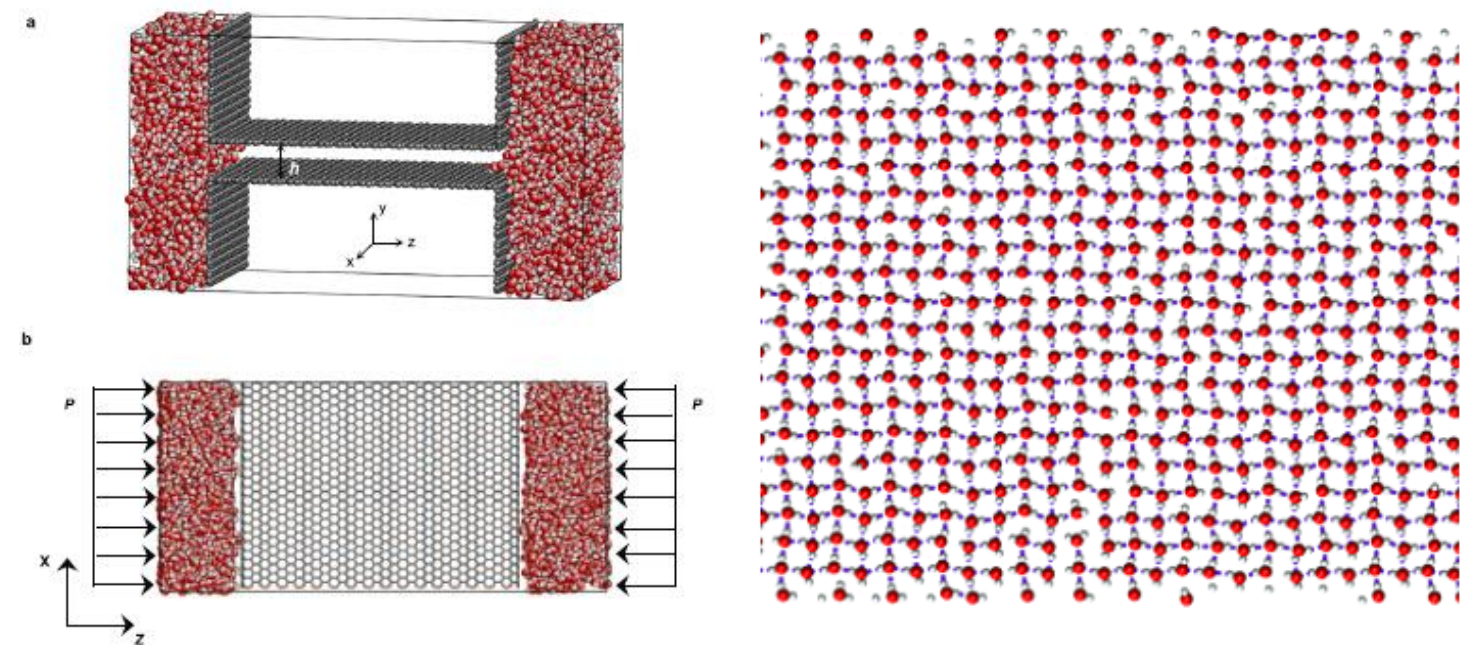
Physical goal: Uncover nonequilibrium **Kardar-Parisi-Zhang** (KPZ) universality class behavior in the equilibrium **six vertex model** (6V).

Growth dynamics of cancer cell colonies and their comparison with noncancerous cells. PRE, 2012, Huergo et. al.



*Kardar-Parisi-Zhang class*

Square ice in graphene nanocapillaries, Nature 2015, Algara-Siller et. al.

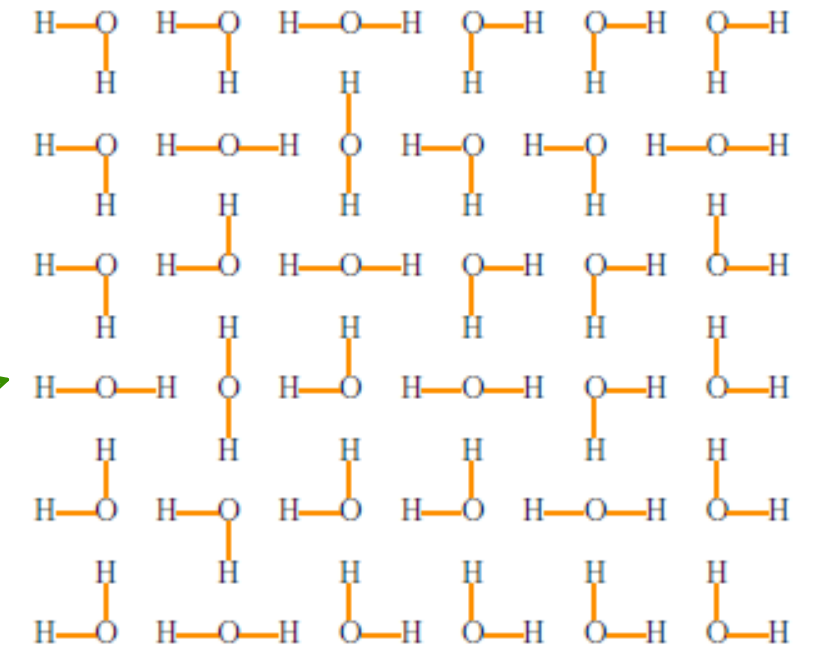


*Six vertex model*

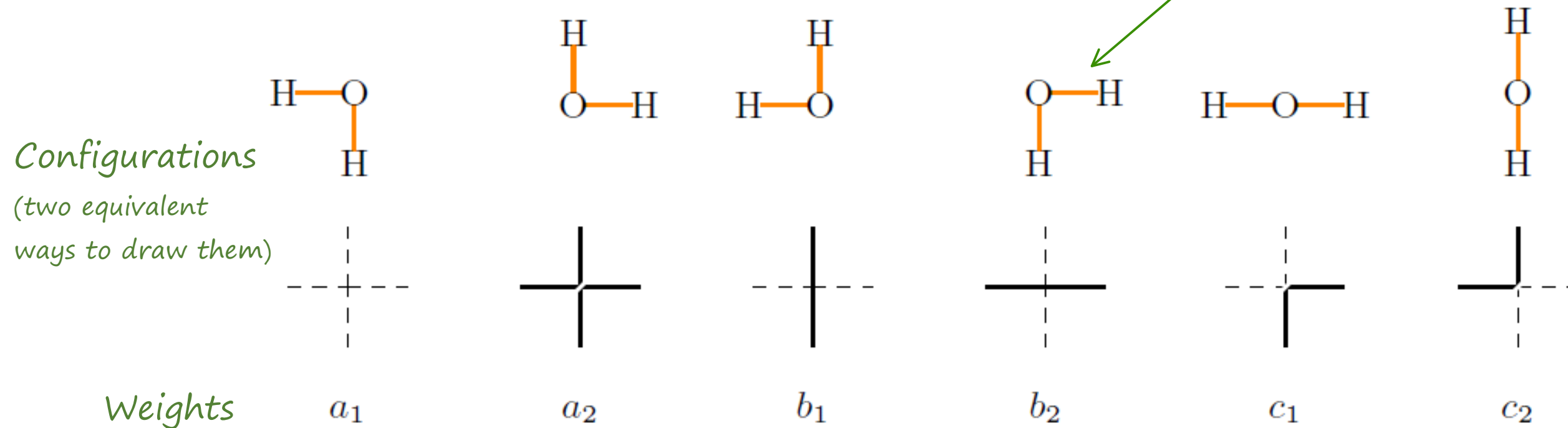
Mathematical goal: Describe how to analyze the stochastic six vertex model (S6V) via **Markov dualities** and **Bethe ansatz** methods.

# Six vertex model [Pauling '35], [Slater '41], [Lieb '67]

**Square-ice** model based on six orientations of  $H_2O$ . Other molecules (e.g.  $KH_2PO_4$ ) have unequal binding energy. Led [Slater '41] to the general six vertex model.



Probability proportional to product of weights.

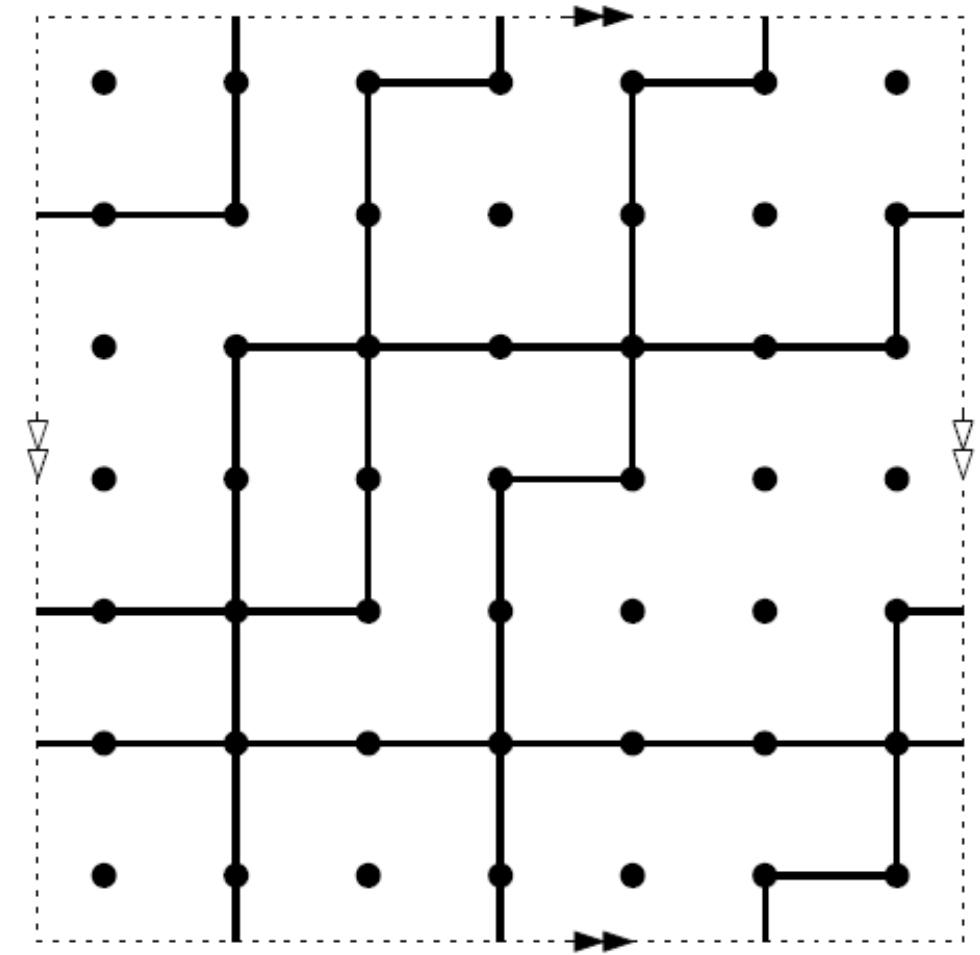


What happens in the large system limit? How do weights matter?

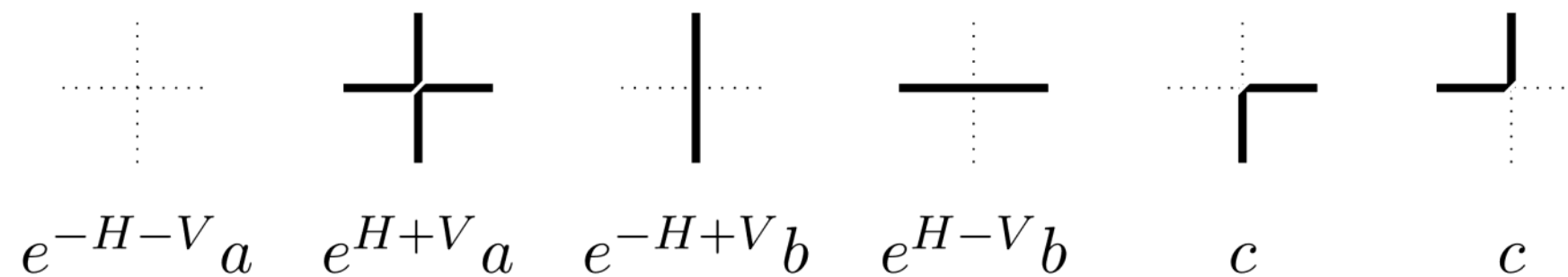
# Gibbs states

Infinite volume '**Gibbs states**' and free energies are key to answer these questions.

Given boundary of a subregion, probability of inside proportional to product of weights



Should be limits of the model on a torus.



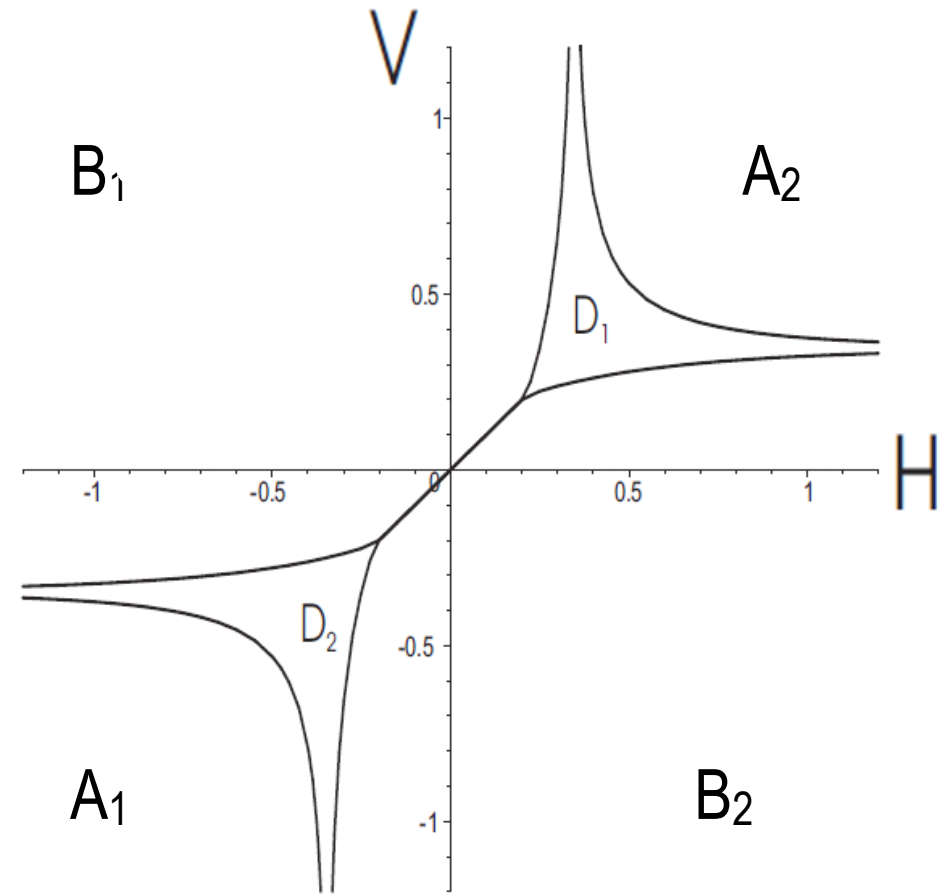
For  $(a,b,c)$  fixed, choosing different  $H$  and  $V$  **external fields** should lead to (possibly) different Gibbs states. The phase diagram of such Gibbs states is mostly conjectural and relies upon a key parameter

$$\Delta := \frac{a_1 a_2 + b_1 b_2 - c_1 c_2}{2\sqrt{a_1 a_2 b_1 b_2}} = \frac{a^2 + b^2 - c^2}{2ab}$$

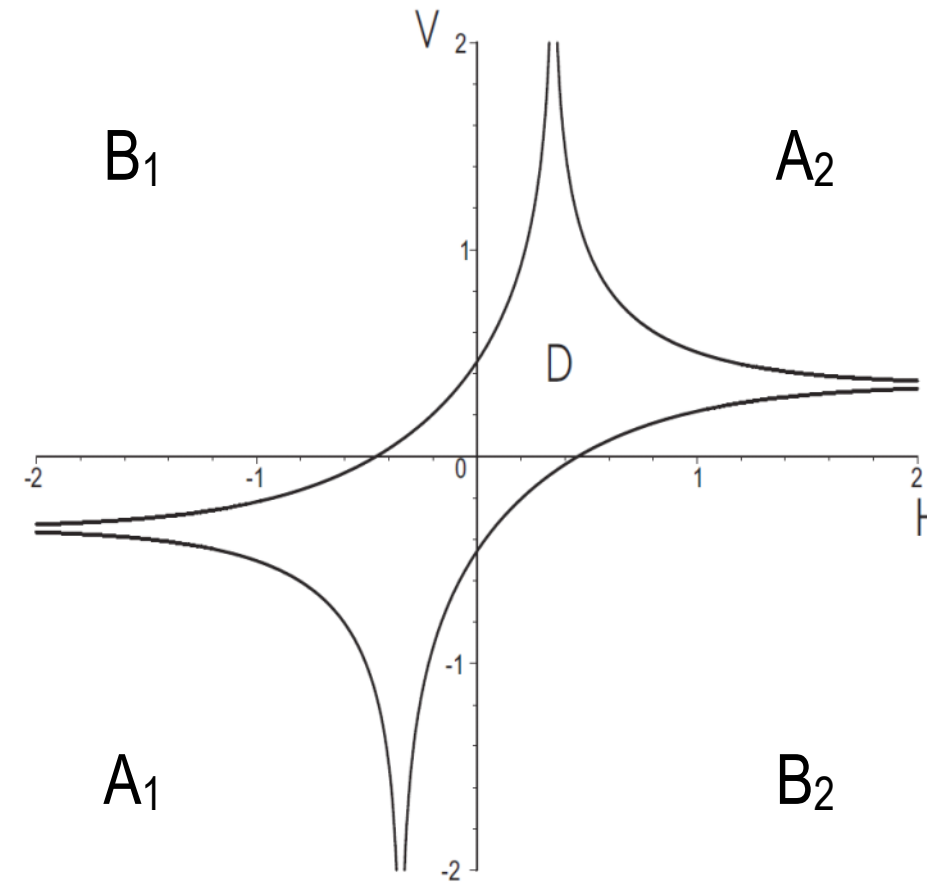
# Phase diagrams

From: Lectures on the integrability of the 6-vertex model, 2010, Reshetikhin.

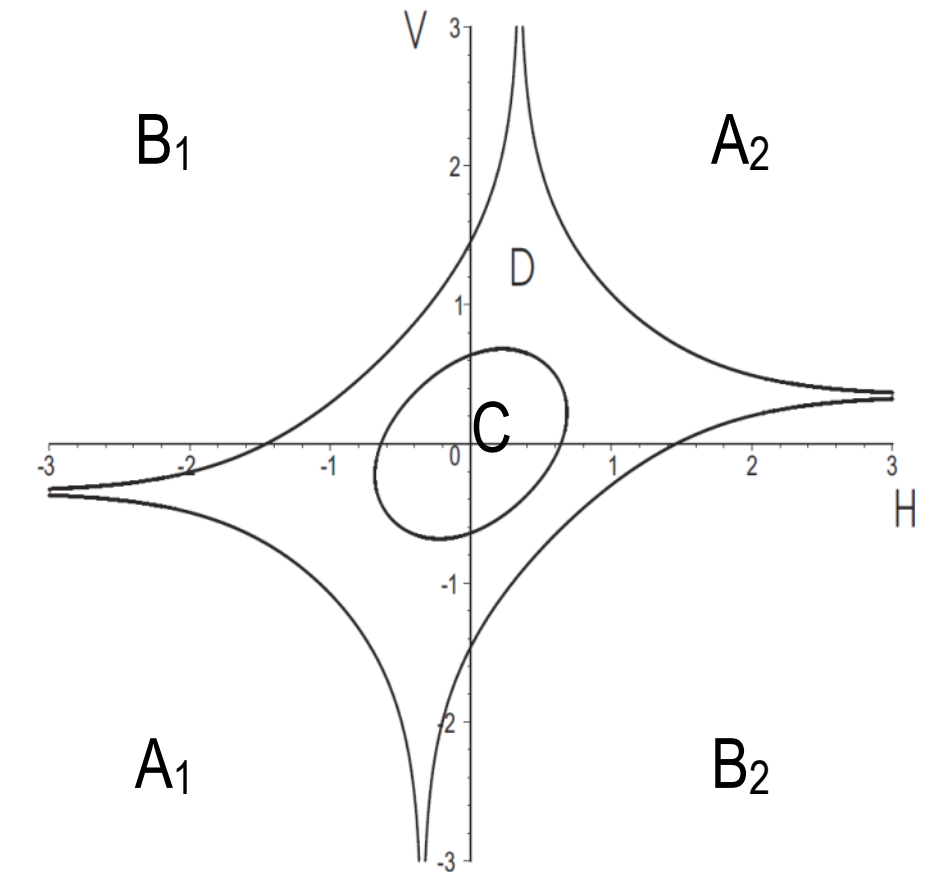
Ferroelectric:  $\Delta > 1$



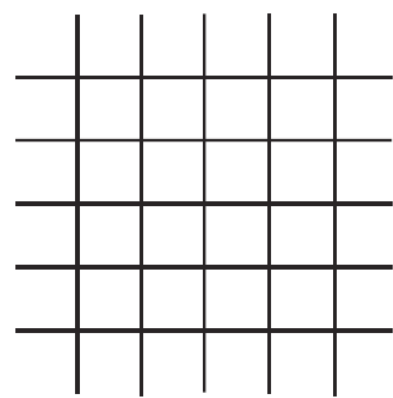
Disordered:  $|\Delta| < 1$



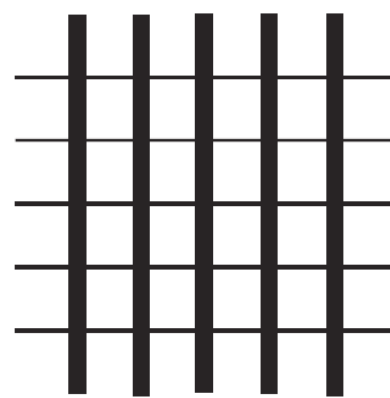
Antiferroelectric:  $\Delta < -1$



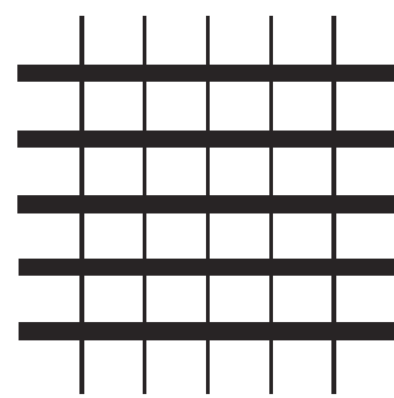
Ordered Gibbs states:



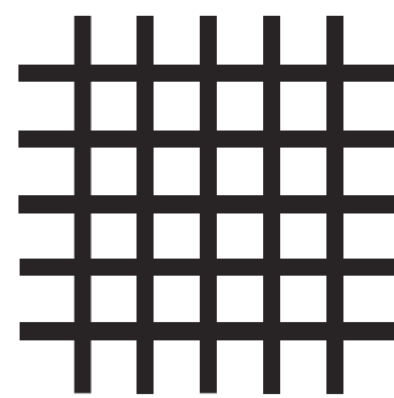
A<sub>1</sub>



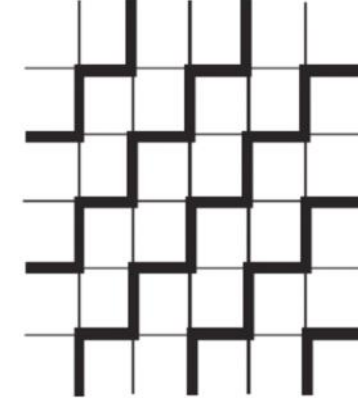
B<sub>1</sub>



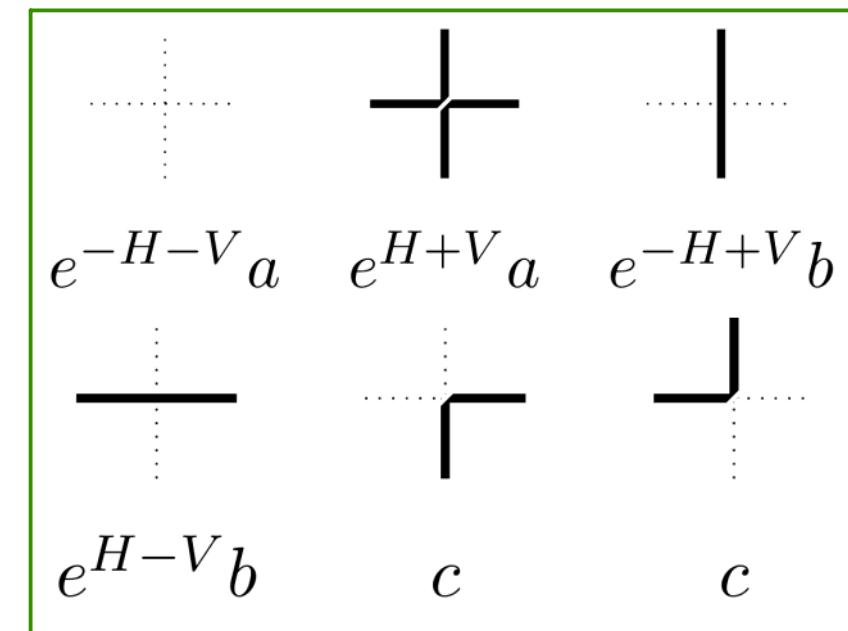
B<sub>2</sub>



A<sub>2</sub>



C



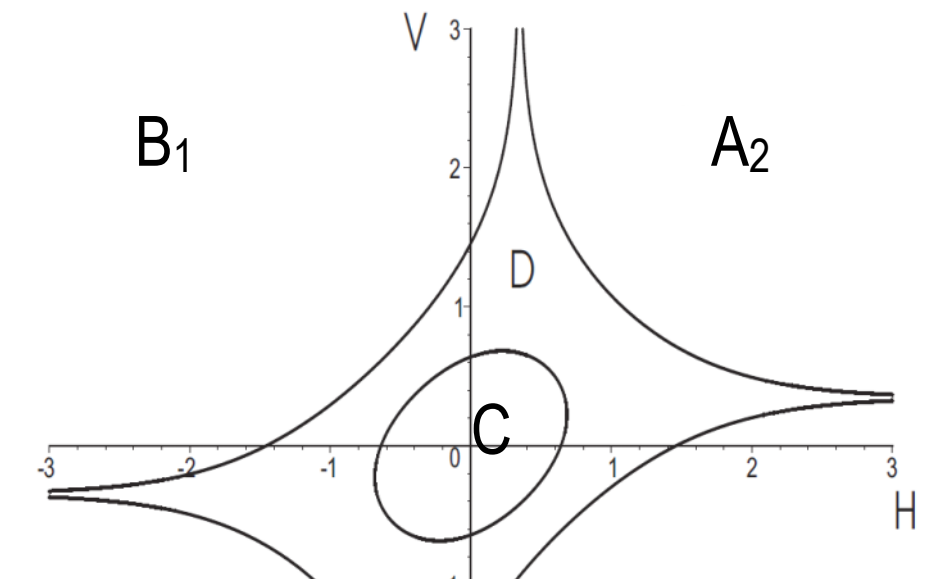
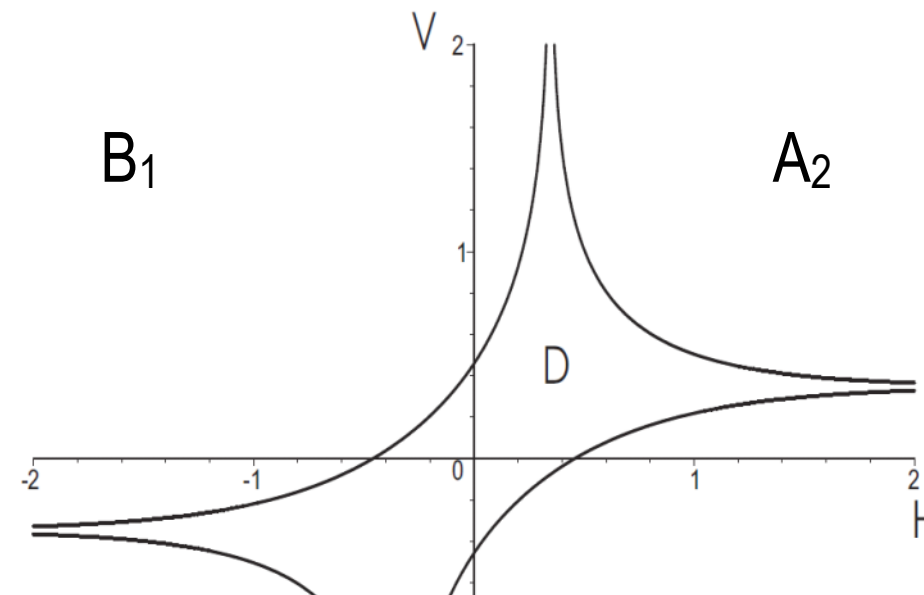
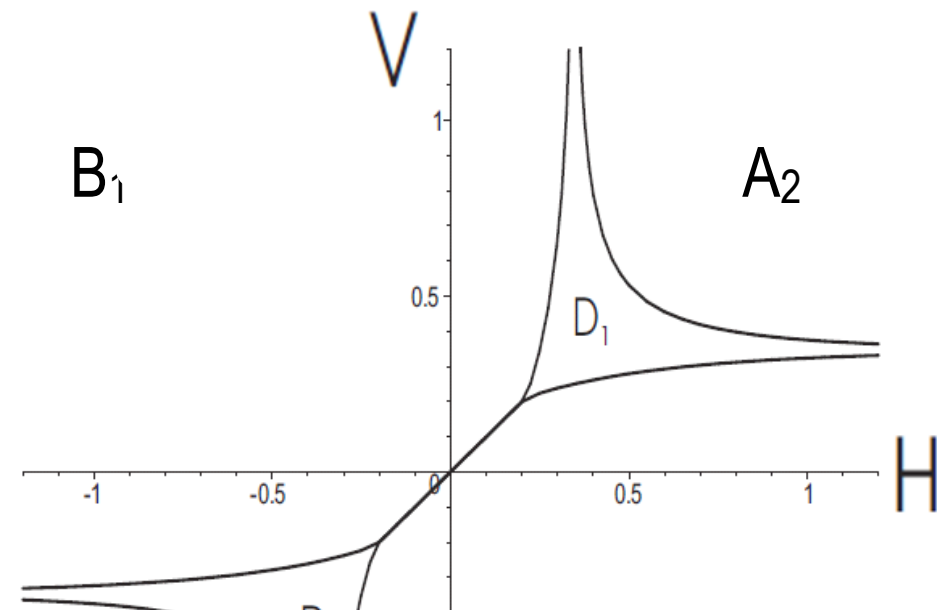
# Phase diagrams

From: Lectures on the integrability of the 6-vertex model, 2010, Reshetikhin.

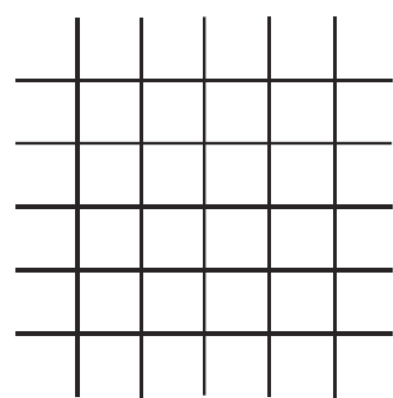
Ferromagnetic:  $\Delta > 1$

Disordered:  $|\Delta| < 1$

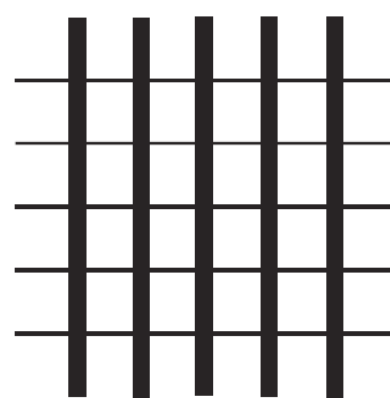
Antiferromagnetic:  $\Delta < -1$



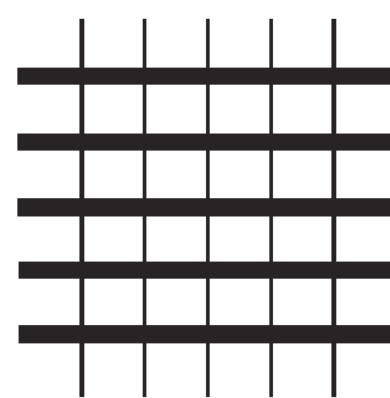
*What happens in the disordered phase, or at its boundary? How disordered is disordered?*



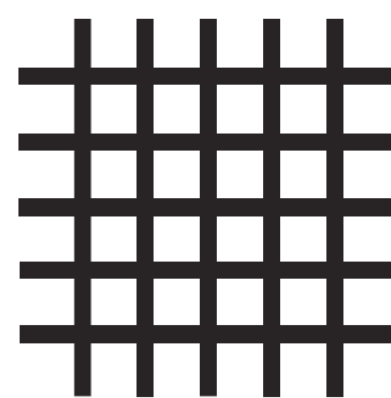
A<sub>1</sub>



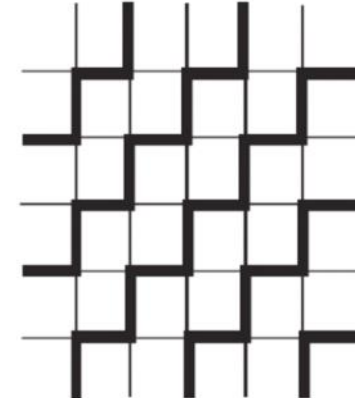
B<sub>1</sub>



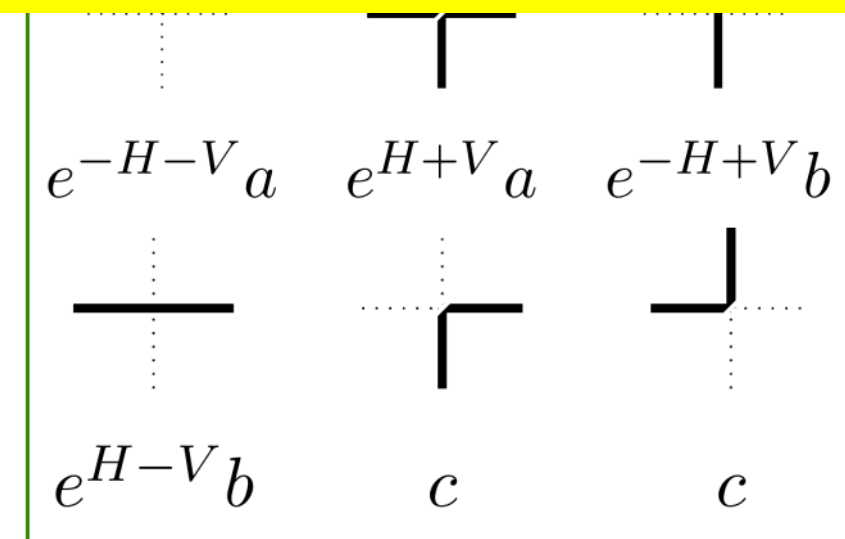
B<sub>2</sub>



A<sub>2</sub>



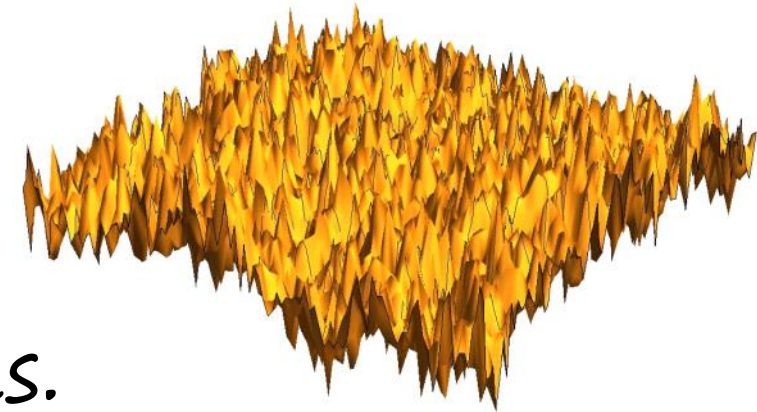
C



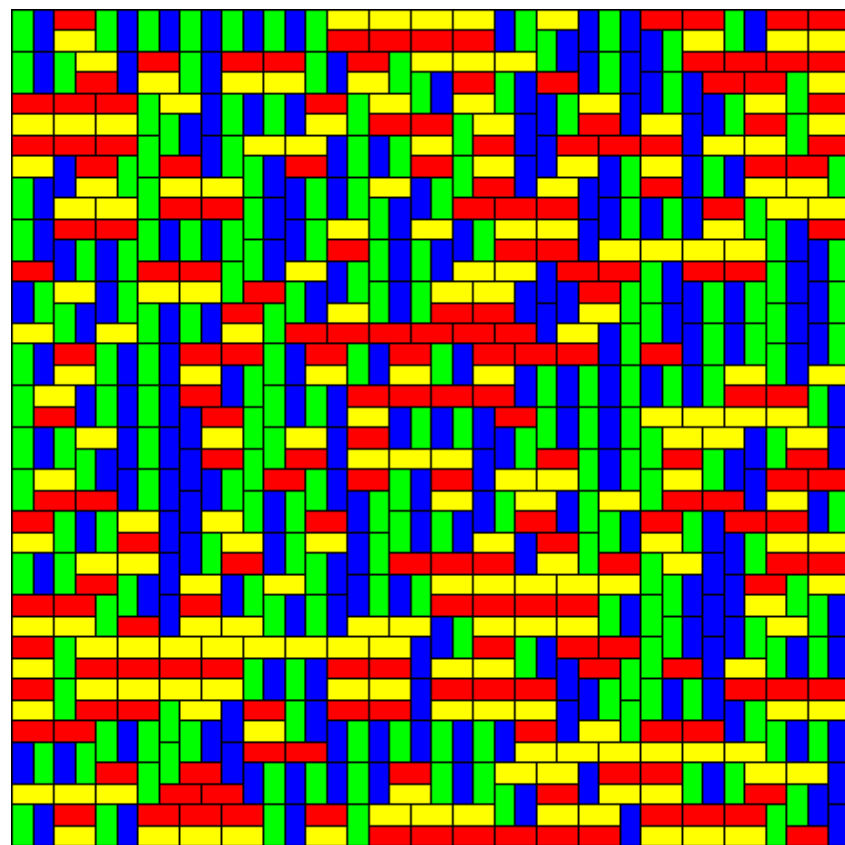
## Disordered phase (free fermion case)

Points in disordered phase lead to Gibbs states with various average horizontal and vertical line densities.

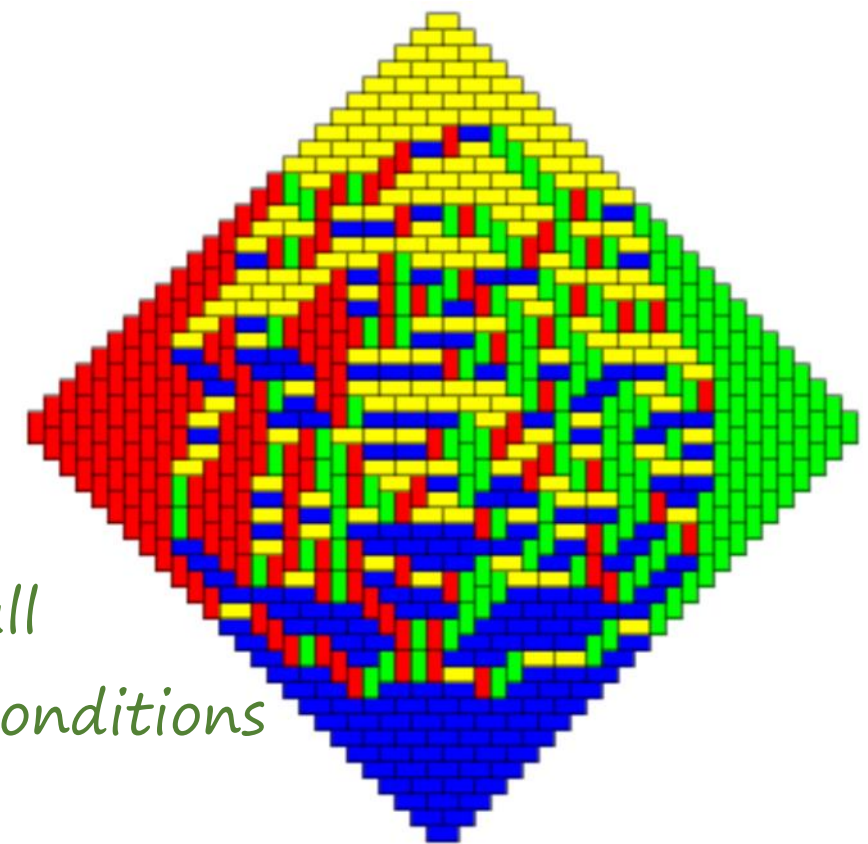
[Nienhuis '84] conjectured that disordered states have *Gaussian free field* height function fluctuations.



[Kenyon '01] proved results for *free fermion* case ( $\Delta = 0$ ).



Portion of a  
Gibbs state  
for the aztec  
tiling model

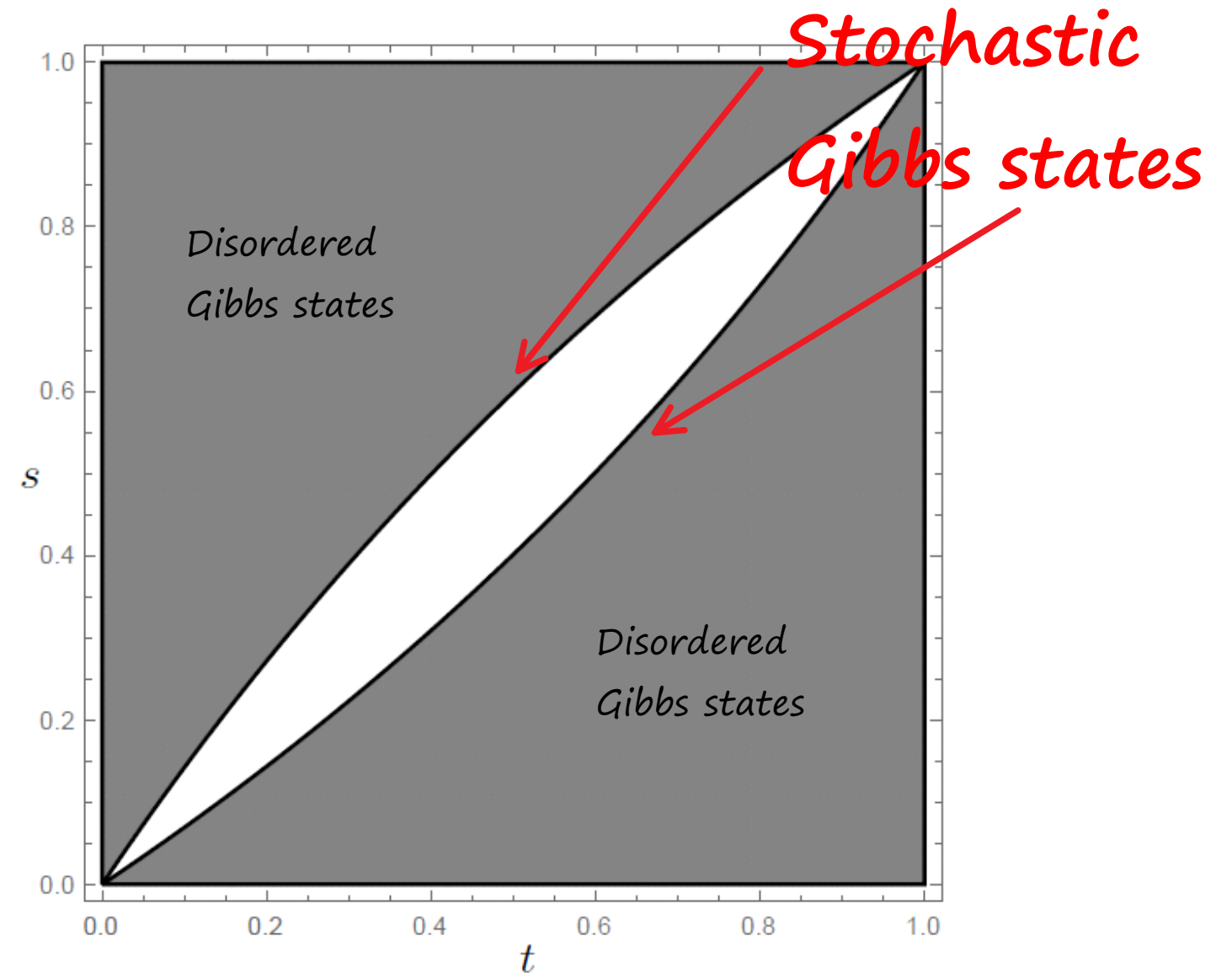
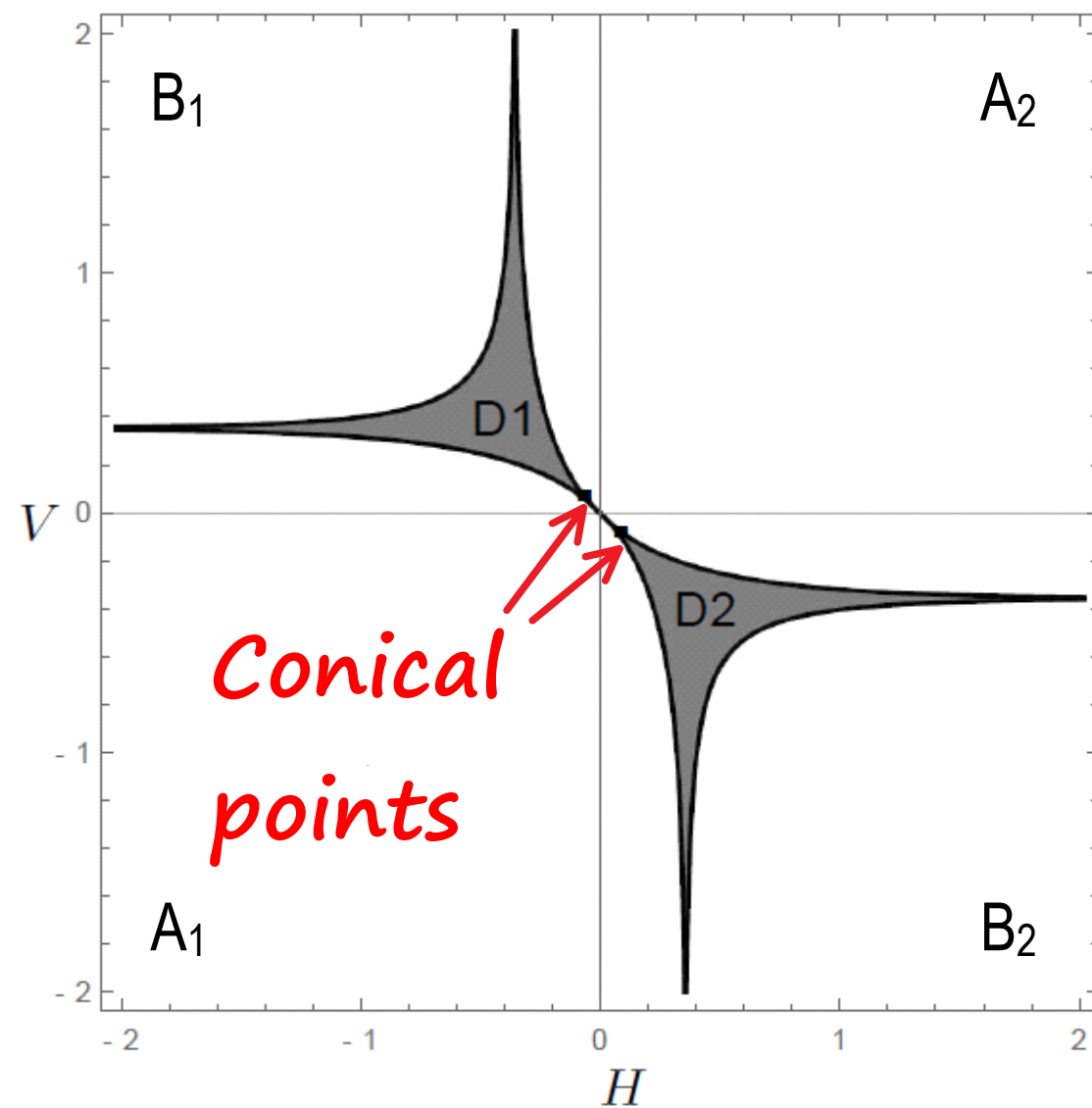


Domain wall  
boundary conditions

# Stochastic point

For  $\Delta > 1$  the 'conical points' in the phase diagram correspond with a one-parameter family of explicit 'stochastic' Gibbs states.

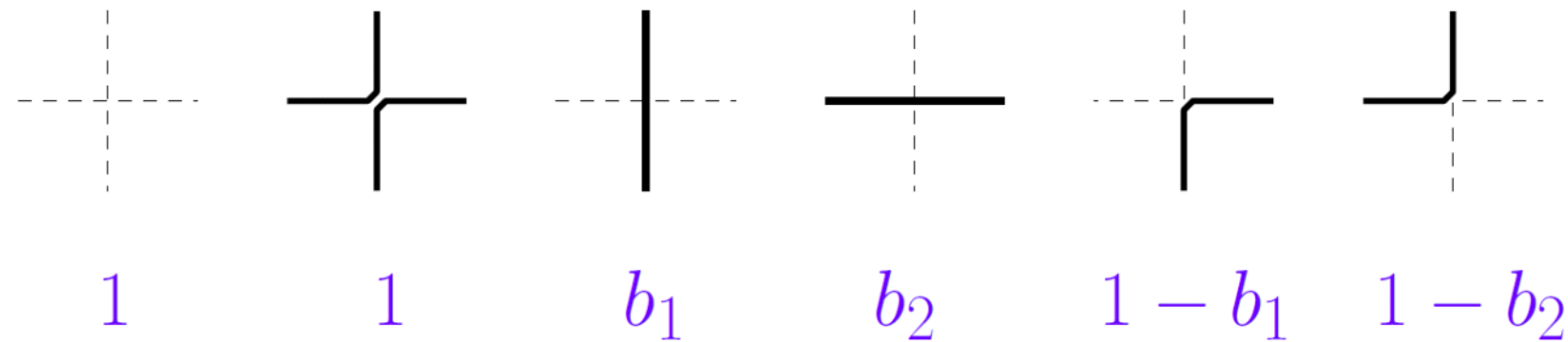
[Jayaprash-Sam '84] [Bukman-Shore '95] [Aggarwal '16]





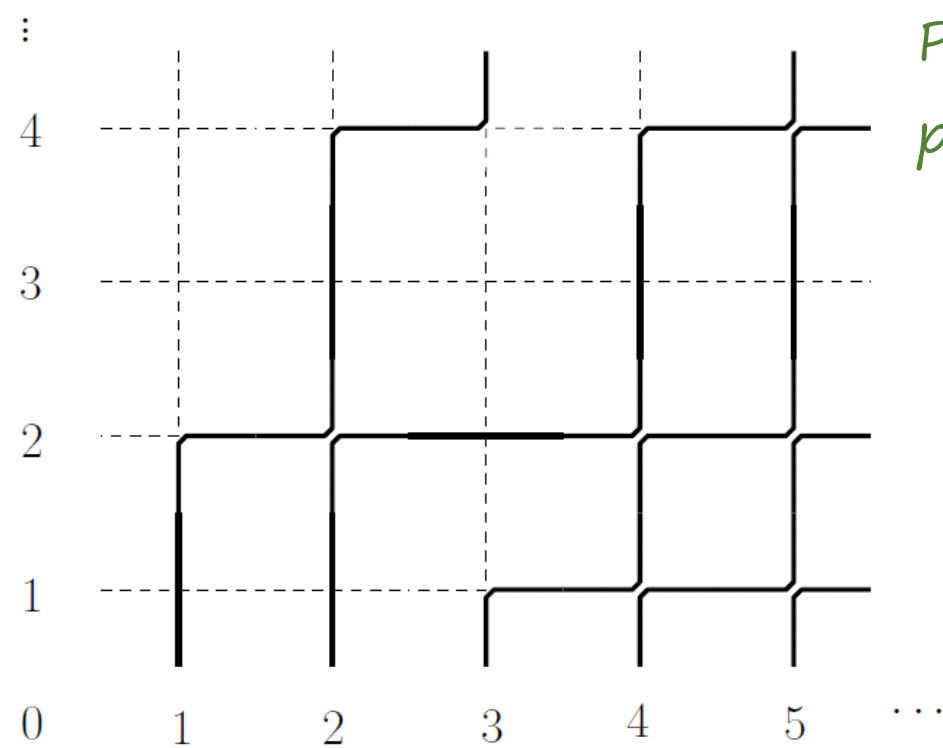
# Stochastic six vertex model

On first quadrant, for  $\Delta > 1$ , special choice of vertex weights



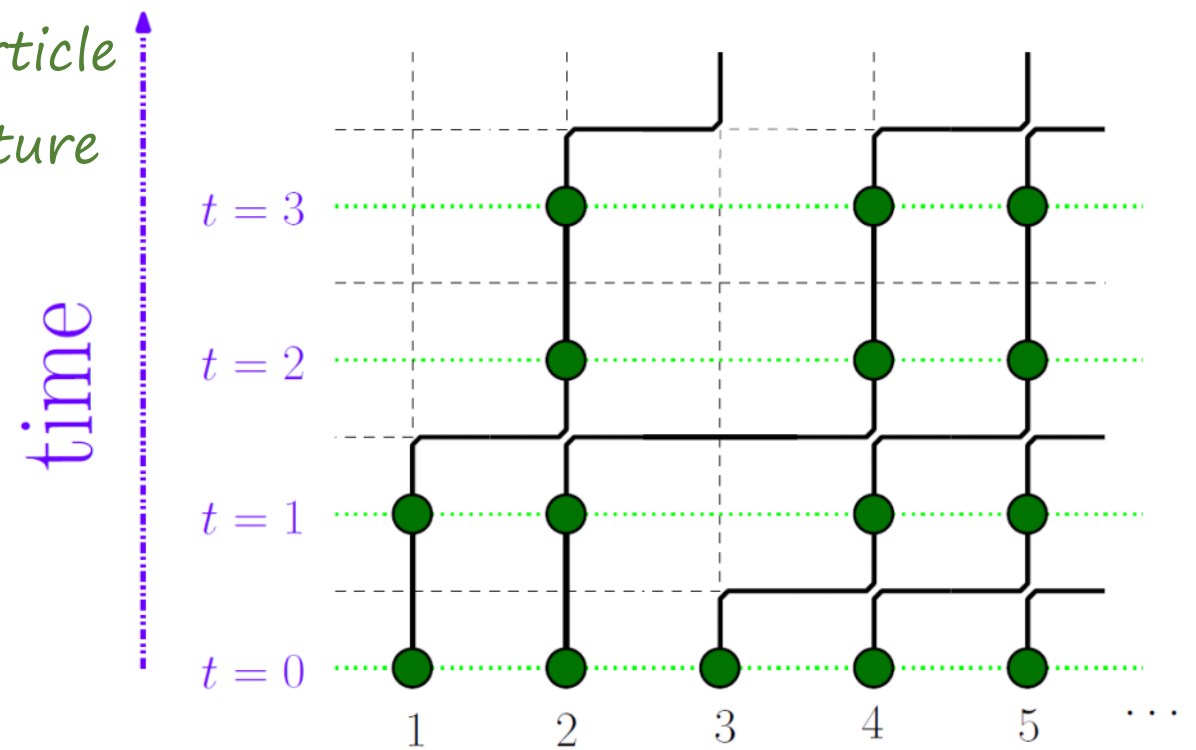
yields stochastic six vertex model (S6V) [Gwa-Spohn '92].

Markov update provides interacting particle system interpretation.



Path picture

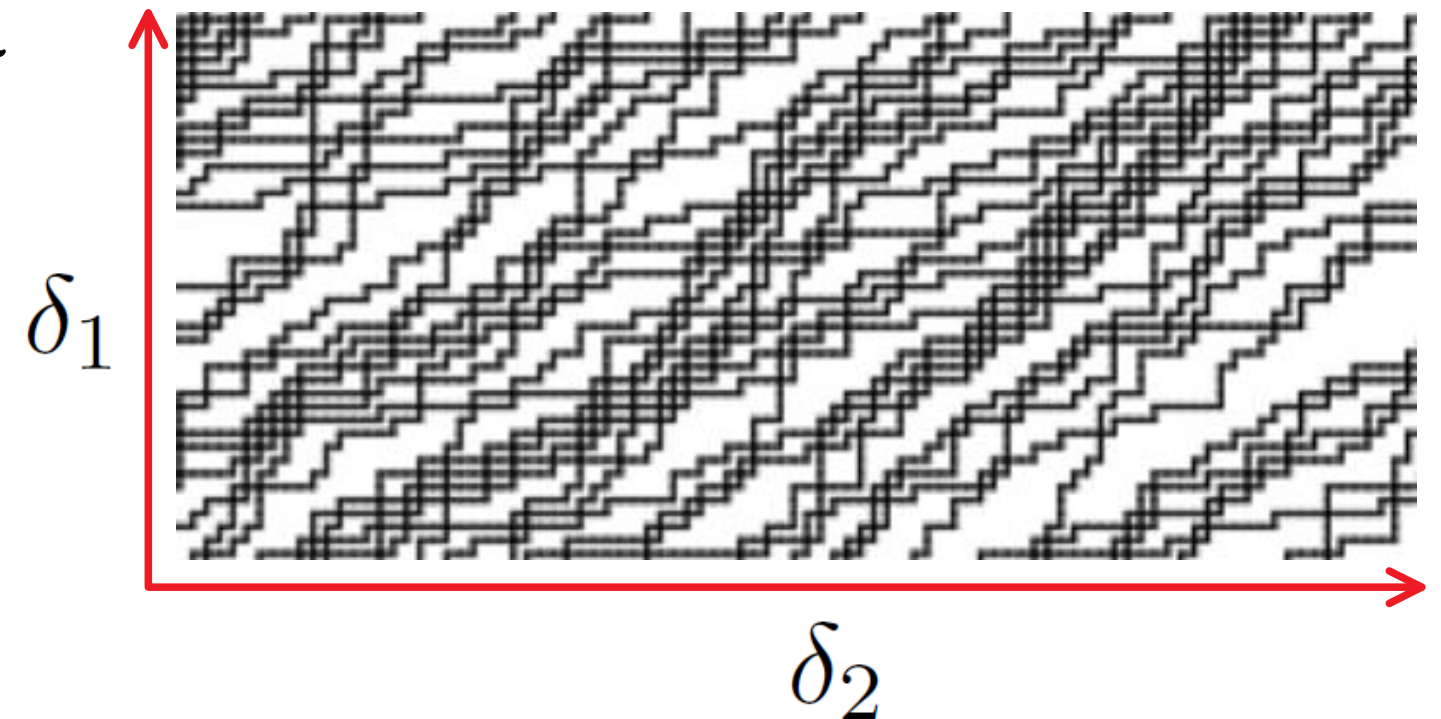
Particle picture



# Stochastic Gibbs states

Let  $\rho = \frac{1 - b_1}{1 - b_2}$  and  $(\delta_1, \delta_2)$  be solutions to  $\frac{\delta_1}{1 - \delta_1} = \rho \cdot \frac{\delta_2}{1 - \delta_2}$ .

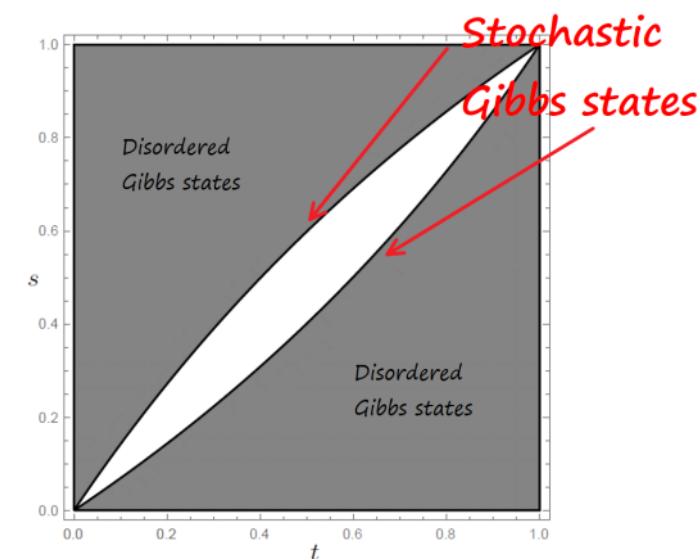
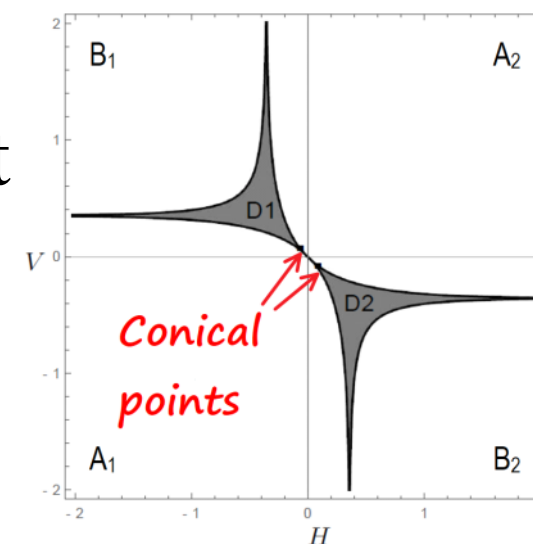
Bernoulli product measure  $(\delta_1$  on the  $y$ -axis and  $\delta_2$  on the  $x$ -axis) is stationary [Aggarwal '16] and hence produces an infinite volume 'stochastic Gibbs state'.



Theorem [Bukman-Shore '95][Aggarwal '16]:

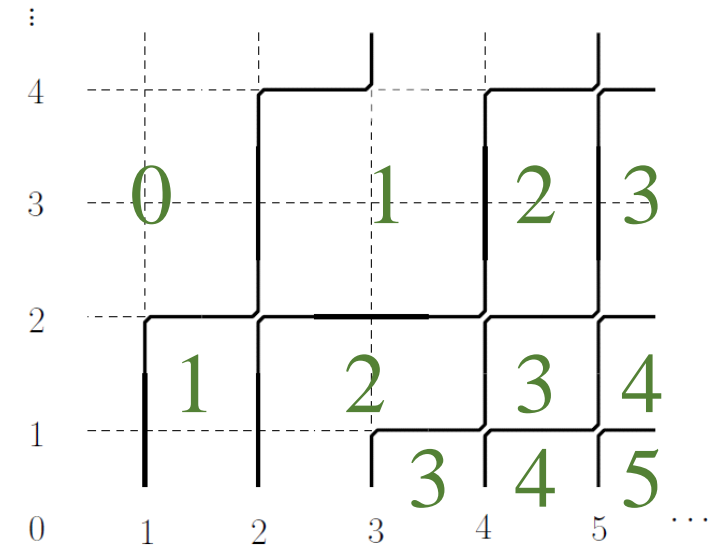
Stochastic Gibbs states (above) are conical point Gibbs states for the symmetric 6V model when

$$b_1 = \frac{b}{a}(\Delta + \sqrt{\Delta^2 - 1}), \quad b_2 = \frac{b}{a}(\Delta - \sqrt{\Delta^2 - 1}).$$

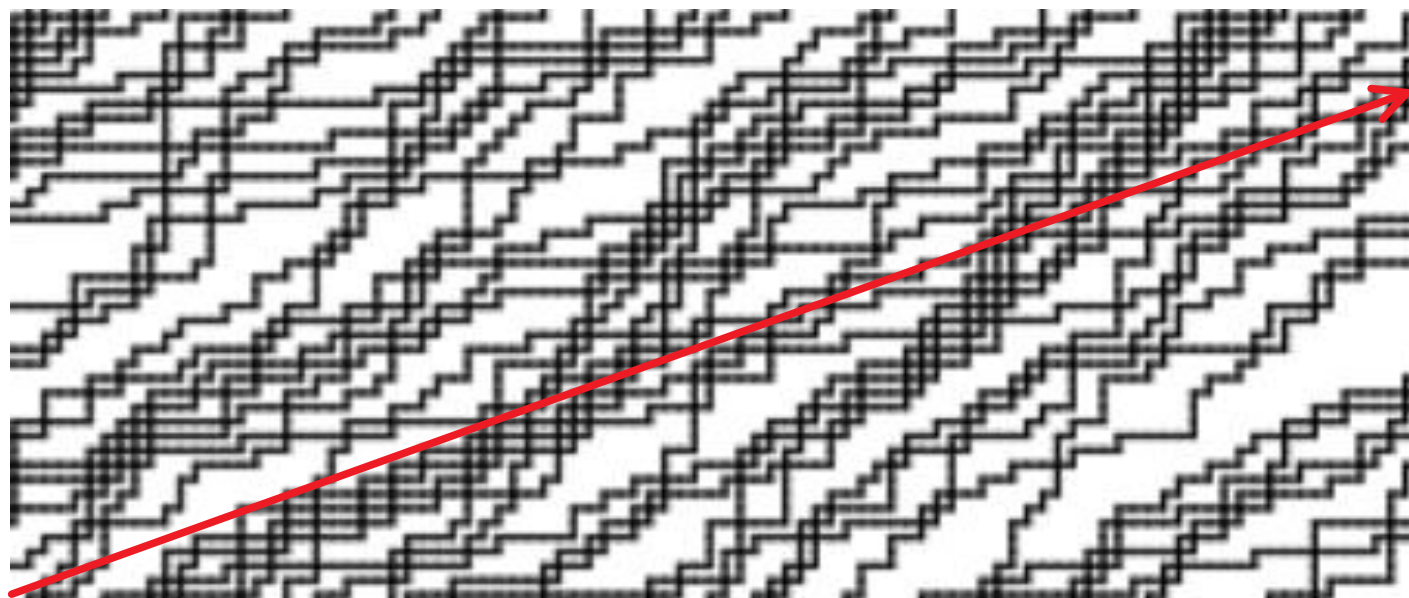


# Stationary $S6V$ height fluctuations

Define the *height function* (zero at the origin):



Theorem [Aggarwal '16]: For fixed  $b_1, b_2$  the stochastic Gibbs state height function fluctuates like distance<sup>1/2</sup> with Gaussian distribution, except along the '**characteristic direction**' where it's like **distance**<sup>1/3</sup> with stationary KPZ distribution [Baik-Rain '01].



*characteristic direction* comes from the Hamilton-Jacobi hydrodynamic limit flux (essentially  $\delta_1$  as a function of  $\delta_2$ ).

## Stationary S6V height fluctuations

Compare to conjectural Gaussian free field disordered phase behavior with logarithmic scale fluctuations.

Fluctuates like distance<sup>1/2</sup> with Gaussian distribution, except along the 'characteristic direction' where it's like distance<sup>1/3</sup> with stationary KPZ distribution [Baik-Rain '01].



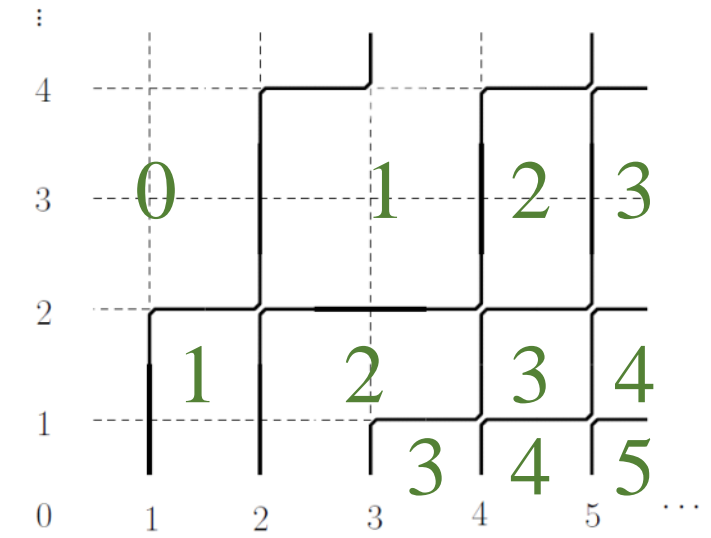
characteristic direction comes from the Hamilton-Jacobi hydrodynamic limit flux (essentially  $\delta_1$  as a function of  $\delta_2$ ).

# Step initial data S6V height fluctuations

**Theorem [Borodin-C-Gorin '14]:** For step initial data S6V

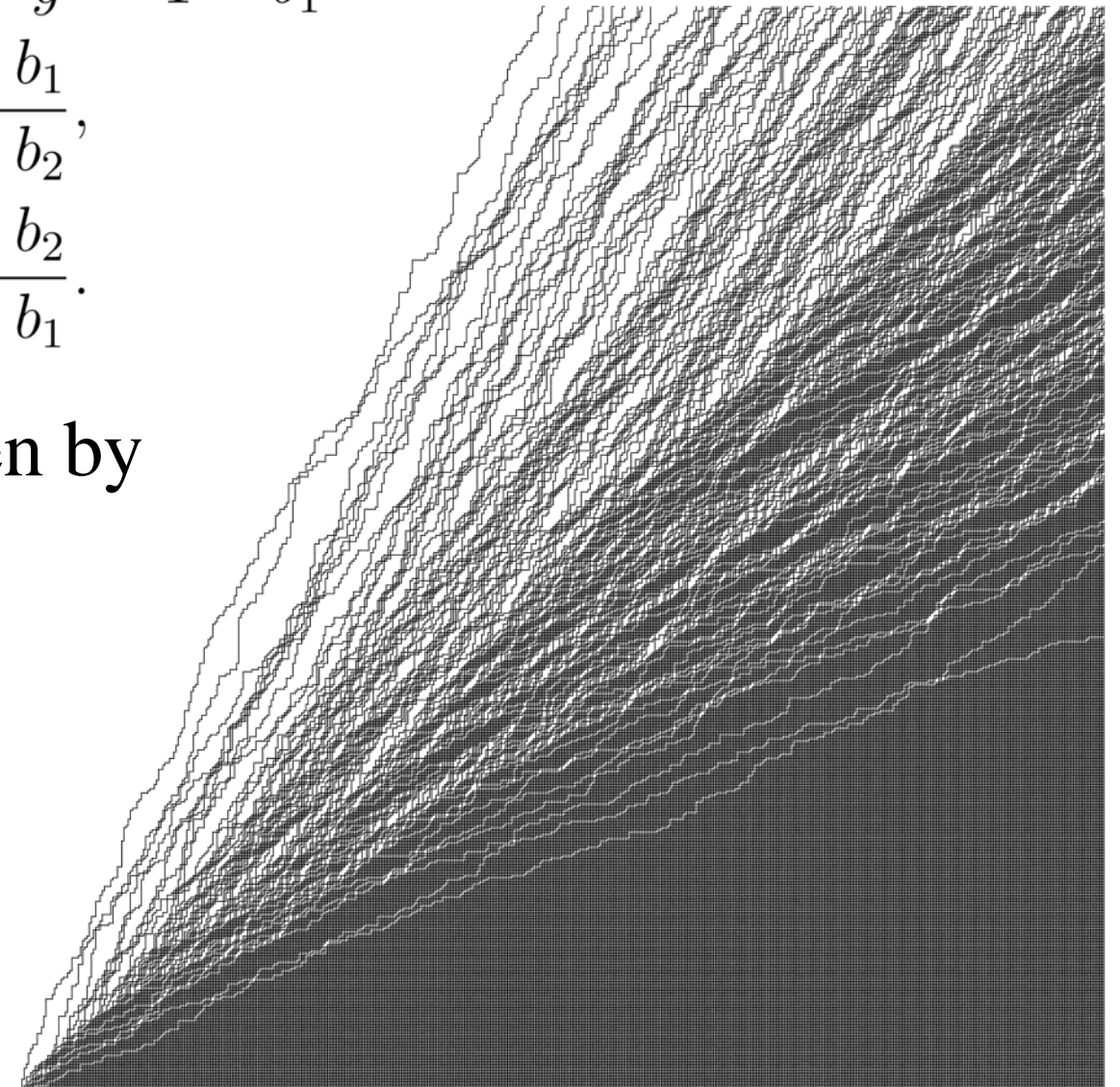
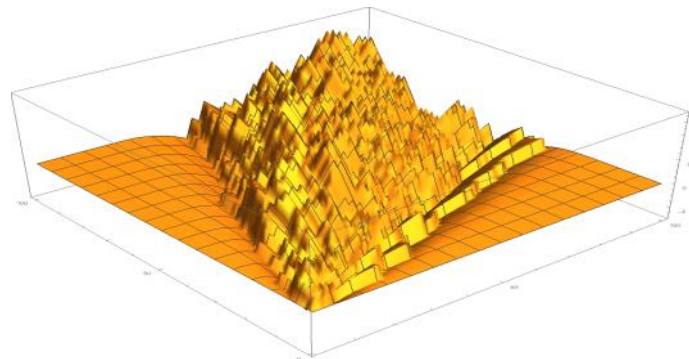
$$\lim_{L \rightarrow +\infty} \frac{H(Lx, Ly; \omega)}{L} = \mathcal{H}(x, y) \quad \text{where the limit shape is}$$

$$\mathcal{H}(x, y) = \begin{cases} \frac{\left(\sqrt{y(1-b_1)} - \sqrt{x(1-b_2)}\right)^2}{b_1 - b_2}, & \frac{1-b_1}{1-b_2} < \frac{x}{y} < \frac{1-b_2}{1-b_1}, \\ 0, & \frac{x}{y} \leq \frac{1-b_1}{1-b_2}, \\ x - y, & \frac{x}{y} \geq \frac{1-b_2}{1-b_1}. \end{cases}$$



The fluctuations around the limit shape are given by

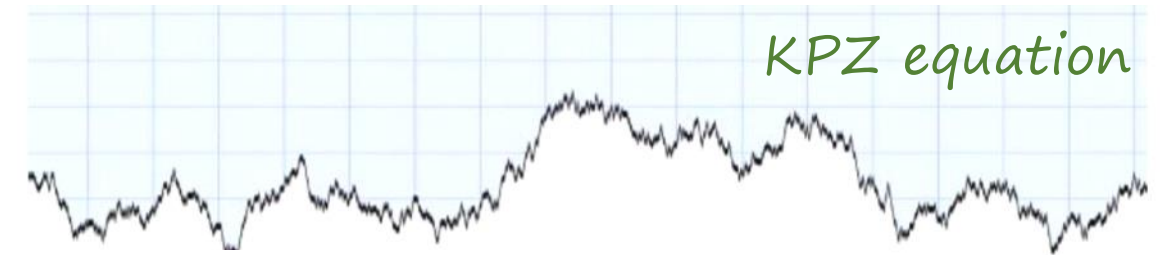
$$\lim_{L \rightarrow \infty} \mathbb{P} \left( \frac{\mathcal{H}(x, y)L - H(Lx, Ly; \omega)}{\sigma_{x,y} L^{1/3}} \leq s \right) = F_{\text{GUE}}(s)$$



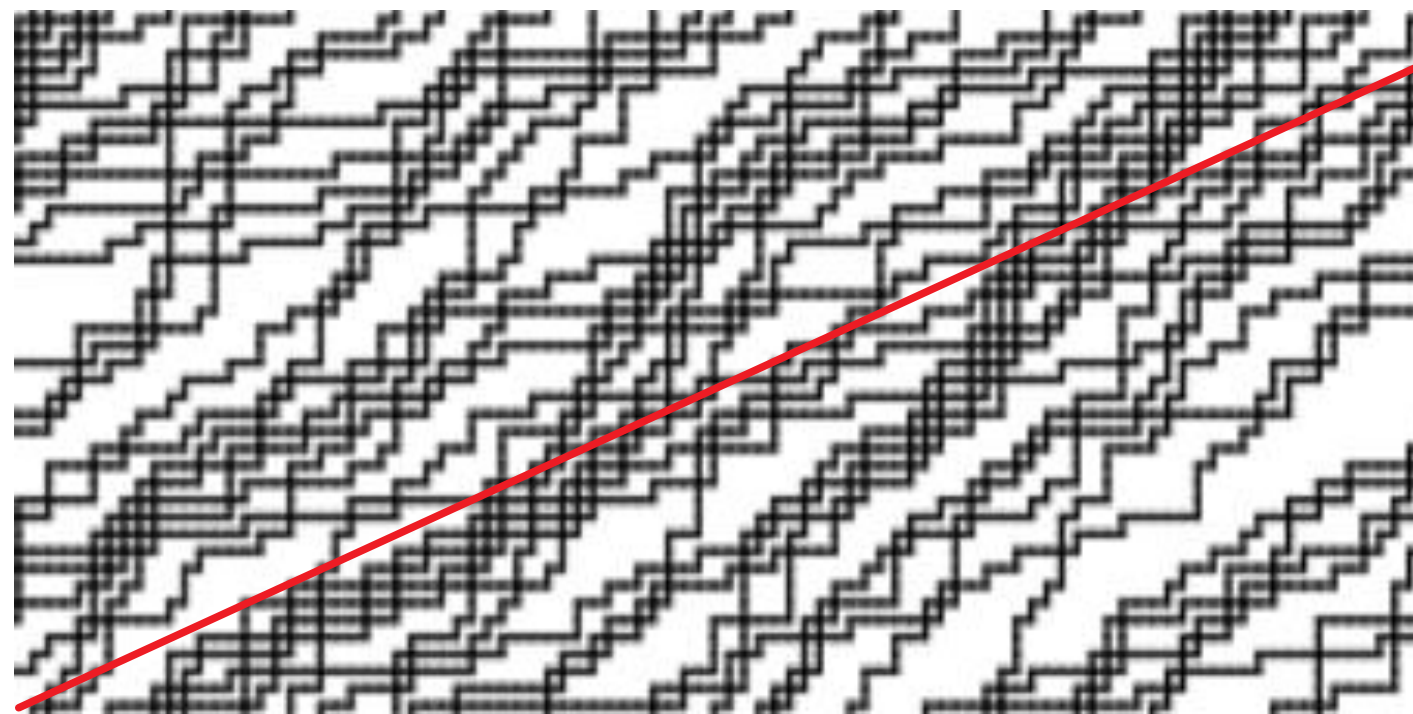
# SPDE limit of S6V

Theorem [C-Ghosal-Shen-Tsai '18]: Let  $b_1, b_2 \rightarrow b \in (0, 1)$  with  $\Delta = \frac{b_1 + b_2}{2\sqrt{b_1 b_2}} \approx 1 + \epsilon$ .

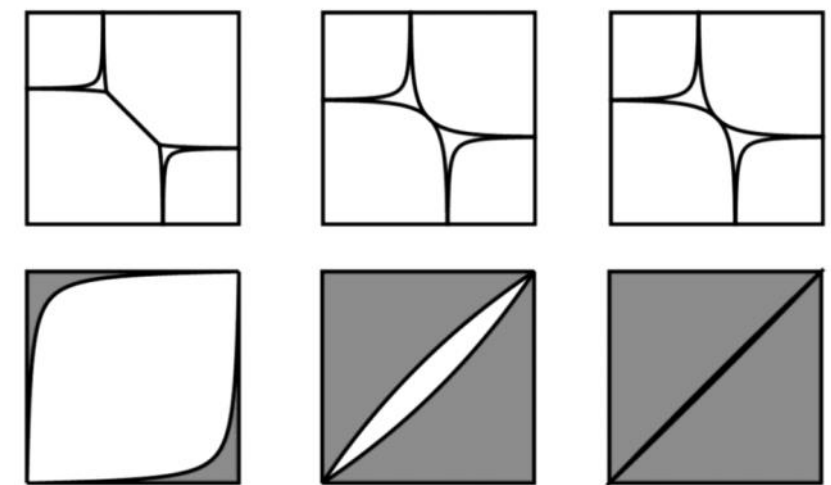
The stationary initial data S6V height function converges (after centering and scaling) along the characteristic directions to the stationary (Brownian initial data) solution to the



KPZ equation:  $\partial_t \mathcal{H}(t, x) = \frac{1}{2} \delta \Delta \mathcal{H}(t, x) + \frac{1}{2} \kappa (\partial_x \mathcal{H}(t, x))^2 + \sqrt{D} \dot{W}(t, x)$ .



$\epsilon^{-1}$   
 $\epsilon^{-2}$



Phase diagram as  $\Delta \rightarrow 1^+$

Stochastic Gibbs states converge to stationary solutions to the stochastic Burgers equation!

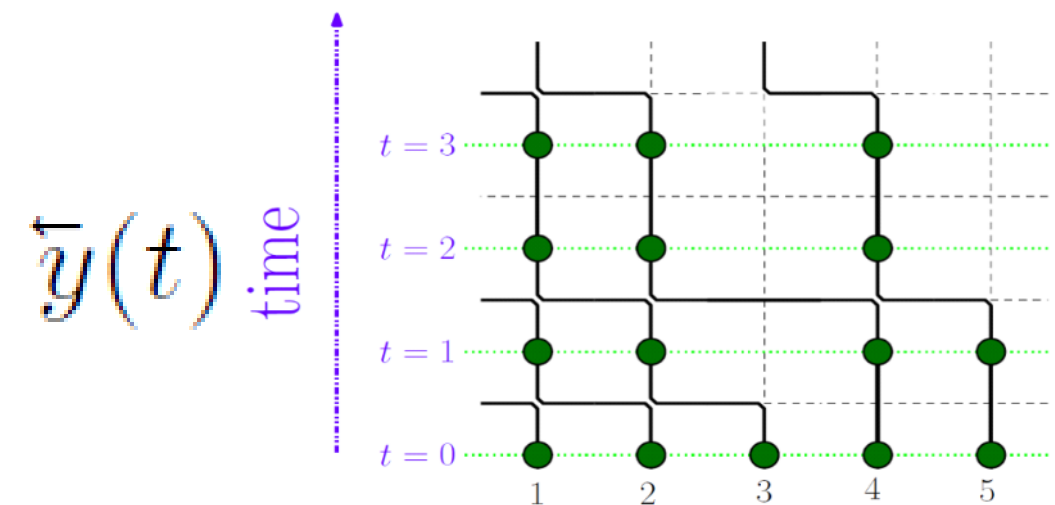
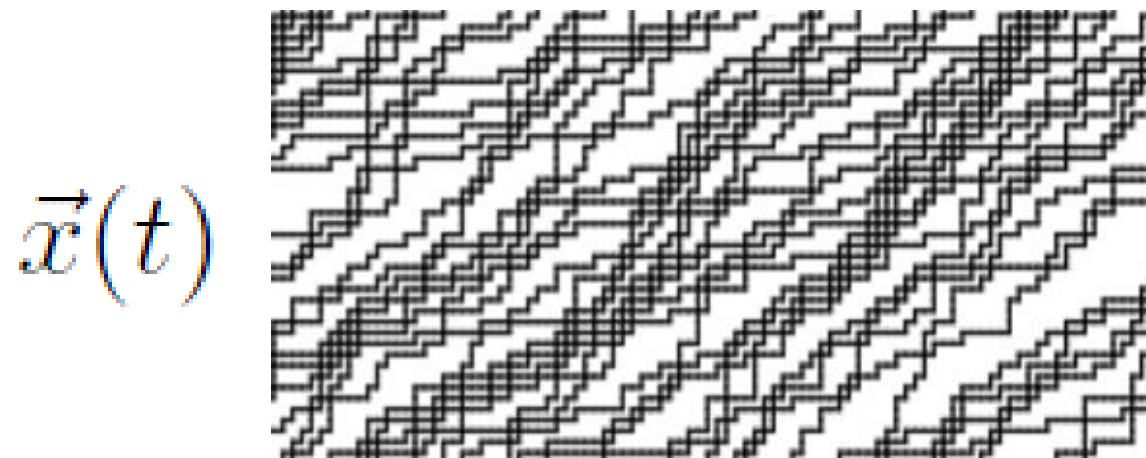
## Recap and what's next

- Gibbs states arise from  $6V$  on torus with external fields. Mapping between field strength and Gibbs state line densities is not simple.
- Disordered states should have GFF and log-correlated fluctuations.
- Stochastic Gibbs states arise at conical point. Fluctuations have  $1/3$  KPZ exponent in characteristic directions, and the entire field admits a limit when  $\Delta \rightarrow 1^+$  to the stationary KPZ equation.
- There are other KPZ class / equation convergence results.
- Rest of the talk will focus on two methods (**Markov duality** and **Bethe ansatz**) which play important roles in these type of results.

# Markov duality

Definition: Two Markov processes  $x(t) \in X$  and  $y(t) \in Y$  are dual with respect to

$$Q : X \times Y \rightarrow \mathbb{R} \text{ if for all } x \in X, y \in Y \text{ and } t \geq 0 : \mathbb{E}^x \left[ Q(x(t), y) \right] = \mathbb{E}^y \left[ Q(x, y(t)) \right].$$



Theorem [C-Petrov '15]: The S6V particle process  $\vec{x}(t)$  and the independent, space

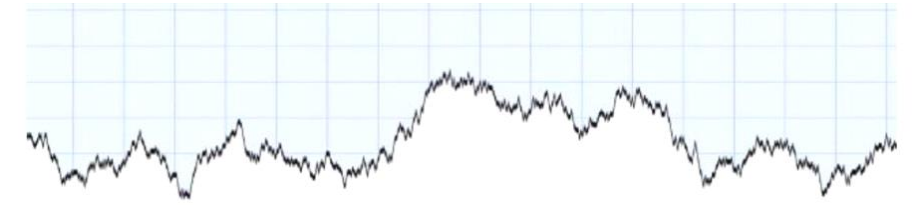
reversed S6V k-particle process  $\overleftarrow{y}(t)$  are dual with respect to  $Q(\vec{x}, \overleftarrow{y}) := \prod_{i=1}^k \tau^{N_{y_i}(\vec{x})}$ ,

where  $\tau = b_2/b_1 \in (0, 1)$  and  $N_y(\vec{x}) = \max \{n : x_n(t) \leq y\}$ .

Such dualities can be proved directly (as above or in [Borodin-C-Sasamoto '12]...), inductively ([Lin '19]) or based on quantum group symmetries ([Schutz '95], [Carinci-Giardina-Redig-Sasamoto '16], [Kuan '17]...)



# Microscopic stochastic heat equation



Definition: The Cole-Hopf solution to the KPZ equation

$$\partial_t \mathcal{H}(t, x) = \frac{1}{2} \delta \Delta \mathcal{H}(t, x) + \frac{1}{2} \kappa (\partial_x \mathcal{H}(t, x))^2 + \sqrt{D} \dot{W}(t, x)$$

is  $\mathcal{H}(t, x) := \log \mathcal{Z}(t, x)$ , where  $\mathcal{Z}$  solves the stochastic heat equation (SHE)

$$\partial_t \mathcal{Z}(t, x) = \frac{1}{2} \delta \Delta \mathcal{Z}(t, x) + \frac{\kappa}{\delta} \sqrt{D} \mathcal{Z}(t, x) \dot{W}(t, x).$$

S6V duality implies that  $\mathbb{E} \left[ \prod_{i=1}^k \tau^{N_{y_i}(\vec{x}(t))} \right] = \sum_{\vec{y}' \in \mathbb{Y}^k} \mathbb{P}_{S6V}(\vec{y} \rightarrow \vec{y}'; t) \prod_{i=1}^k \tau^{N_{y'_i}(\vec{x}(0))}$ .

-->  $\tau^{N_y(\vec{x}(t))}$  solves a discrete SHE with an explicit martingale whose quadratic variation involves the  $k=2$  duality function.

**Key challenge** in convergence to SHE is to control the martingale.

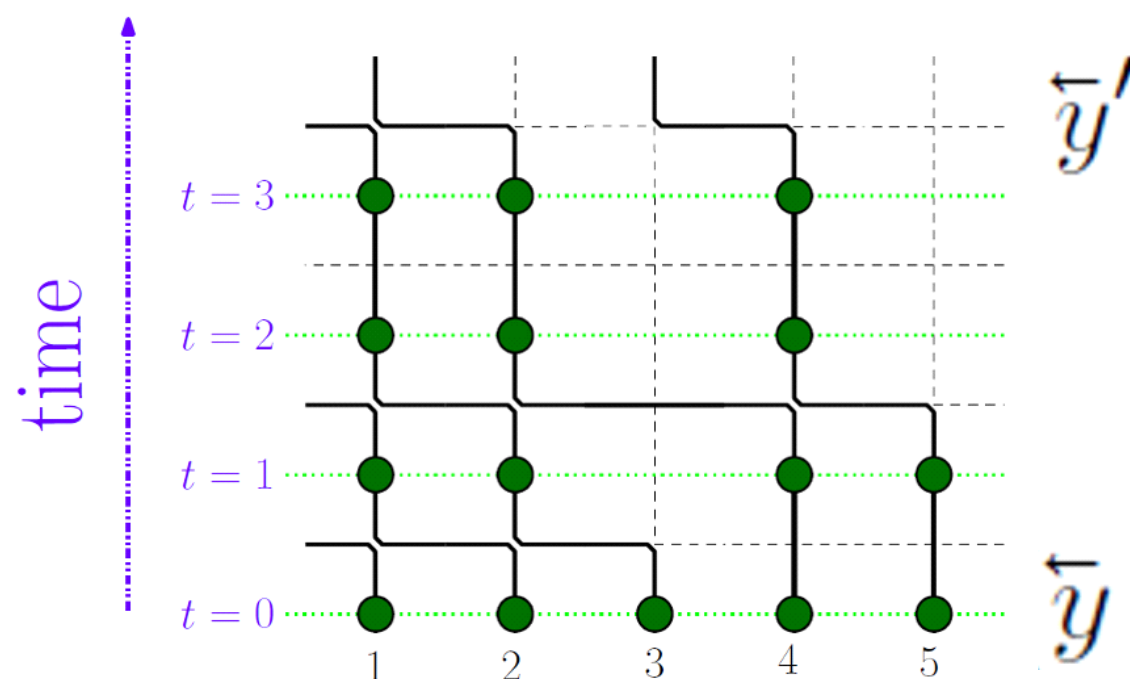
- [Bertini-Giacomin '95] does this for ASEP via complicated identity (doesn't work for S6V).
- [C-Ghosal-Shen-Tsai '17] uses 2-particle **duality** and **Bethe ansatz**.

# (Coordinate) Bethe ansatz

[Borodin-C-Gorin '14]: Explicit formulas for transition probabilities for  $k$ -particle S6V (in spirit of [Tracy-Widom '07] and [Lieb '67])

$$\mathbb{P}_{\text{S6V}}(\bar{y} \rightarrow \bar{y}'; t) = \frac{1}{(2\pi i)^k} \oint_{\mathcal{C}_R} \cdots \oint_{\mathcal{C}_R} \sum_{\sigma \in \mathfrak{S}_k} (-1)^\sigma \prod_{1 \leq i < j \leq k} \mathfrak{E}(z_i, z_j, \sigma) \prod_{i=1}^k z_{\sigma(i)}^{y_i - y'_{\sigma(i)} - 1} \mathfrak{D}(z_i, t) dz_i$$

where  $\mathfrak{E}(z_i, z_j, \sigma) := \frac{1 - (1 + \tau^{-1})z_{\sigma(i)} + \tau^{-1}z_{\sigma(i)}z_{\sigma(j)}}{1 - (1 + \tau^{-1})z_i + \tau^{-1}z_i z_j}$  and  $\mathfrak{D}(z, t) := \left( \frac{b_1 + (1 - b_1 - b_2)z^{-1}}{1 - b_2 z^{-1}} \right)^t$ .



Explicit formulas like these are also starting points for KPZ universality asymptotics

# Plancherel theory

Left/right eigenfunctions **diagonalize**  $k$ -particle  $S6V$  transition matrix:

For  $\vec{y} = (y_1 > \dots > y_k)$  and  $\vec{z} = (z_1, \dots, z_k) \in (\mathcal{C} \setminus \{1, \tau^{-1}\})^k$  define

$$\psi_{\vec{z}}^{\ell}(\vec{y}) = \sum_{\sigma \in S_k} \prod_{1 \leq b < a \leq k} \frac{z_{\sigma(a)} - \tau z_{\sigma(b)}}{z_{\sigma(a)} - z_{\sigma(b)}} \prod_{j=1}^k \left( \frac{1 - \tau^{-1} z_{\sigma(j)}}{1 - z_{\sigma(j)}} \right)^{y_j},$$

and

$$\psi_{\vec{z}}^r(\vec{y}) = \sum_{\sigma \in S_k} \prod_{1 \leq b < a \leq k} \frac{z_{\sigma(a)} - \tau^{-1} z_{\sigma(b)}}{z_{\sigma(a)} - z_{\sigma(b)}} \prod_{j=1}^k \left( \frac{1 - \tau^{-1} z_{\sigma(j)}}{1 - z_{\sigma(j)}} \right)^{-y_j}.$$

**Plancherel theory** [Borodin-C-Petrov-Sasamoto '14]:

The forward transform  $(\mathcal{F}f)(\vec{z}) := \sum_{y_1 > \dots > y_k} f(\vec{y}) \psi_{\vec{z}}^r(\vec{y})$  and the backward transform

$$(\mathcal{J}g)(\vec{y}) := \text{const} \cdot \int \cdots \int_{|w_j| = R \in (1, \tau^{-1}), 1 \leq j \leq k} \det \left[ \frac{1}{\tau w_i - w_j} \right]_{i,j=1}^k \prod_{j=1}^k \frac{w_j}{(1 - w_j)(1 - \tau^{-1} w_j)} \psi_{\vec{w}}^{\ell}(\vec{y}) g(\vec{w}) d\vec{w}$$

are mutual inverses so that  $\mathcal{F}\mathcal{J} = \text{Id}$  and  $\mathcal{J}\mathcal{F} = \text{Id}$ .

# Some extensions

- **Symmetric functions:**
  - [Borodin '14], [Borodin-Petrov '16], [Borodin-Wheeler '17] prove Cauchy identities, Pieri and branching rules for 'spin Hall-Littlewood' and 'spin  $q$ -Whittaker' functions.
- **Fusion:**
  - [C-Petrov '15] (building on [Kulish-Reshetikhin-Sklyanin '81] and [Mangazeev '14]) introduce **higher-spin** stochastic vertex models (Beta RWRE arises from this).
- **Elliptic:**
  - [Borodin '16] lifts to elliptic level 'dynamic  $S_6V$ ' and 'dynamic ASEP'. [Borodin-C '17] prove duality for dynamic ASEP and [C-Ghosal-Matetski '19] prove SPDE limit. [Aggarwal '16] gives higher-spin dynamic models.
- **High rank:**
  - [Kuniba-Mangazeev-Maruyama-Okado '16] introduce high rank models and [Kuan '17] proves their duality. [Borodin-Wheeler '18] develop their symmetric function theory.
- **And much more...**

# Summary

- Six vertex model has an interesting (mostly conjectural) phase diagram.
- Disordered Gibbs states are expected to have **GFF fluctuations**.
- Using the stochastic six vertex model, we can construct the one-parameter family of **stochastic Gibbs states** which arise at the conical point in the phase diagram.
- Stochastic Gibbs states show **KPZ universality class fluctuations** along their characteristic direction, and when  $\Delta \rightarrow 1^+$  they converge to the **stationary solutions to the KPZ equation**.
- **Duality** and **Bethe ansatz** are key tools in proving both results.

# *Stochastic six vertex model*

*Ivan Corwin (Columbia University)*

## Goals of second hour

- Prove a *stochastic heat equation (SHE) Laplace transform formula*:

$$\mathbb{E}_{\text{SHE}} \left[ e^{-u\mathcal{Z}(2\tau,0)e^{\tau/12}} \right] = \mathbb{E}_{\text{Airy}} \left[ \prod_{k=1}^{\infty} \frac{1}{1 + ue^{\tau^{1/3}} \mathbf{A}_k} \right]$$

where  $\mathcal{Z}(t, x)$  is the fundamental solution to the SHE:

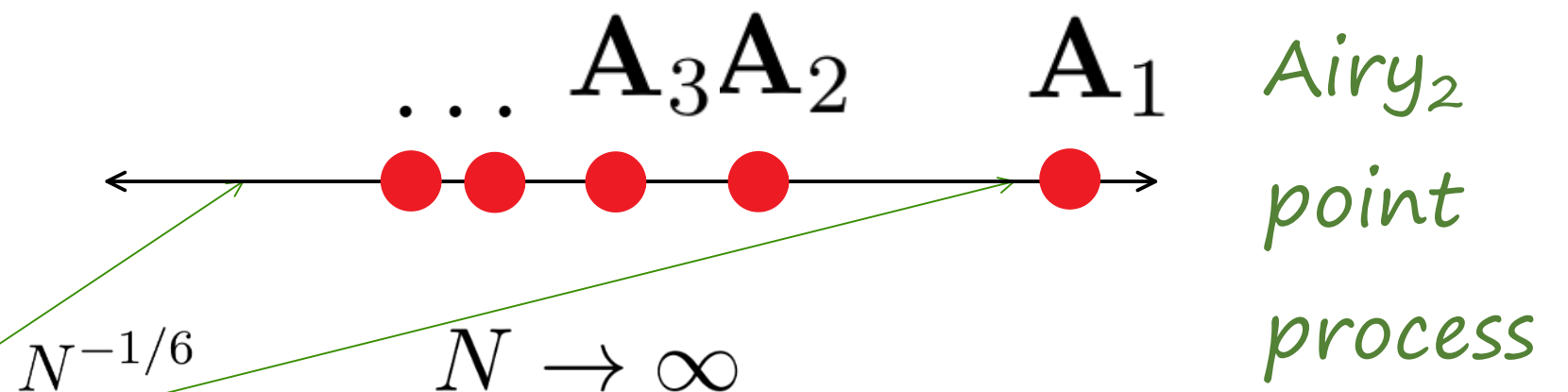
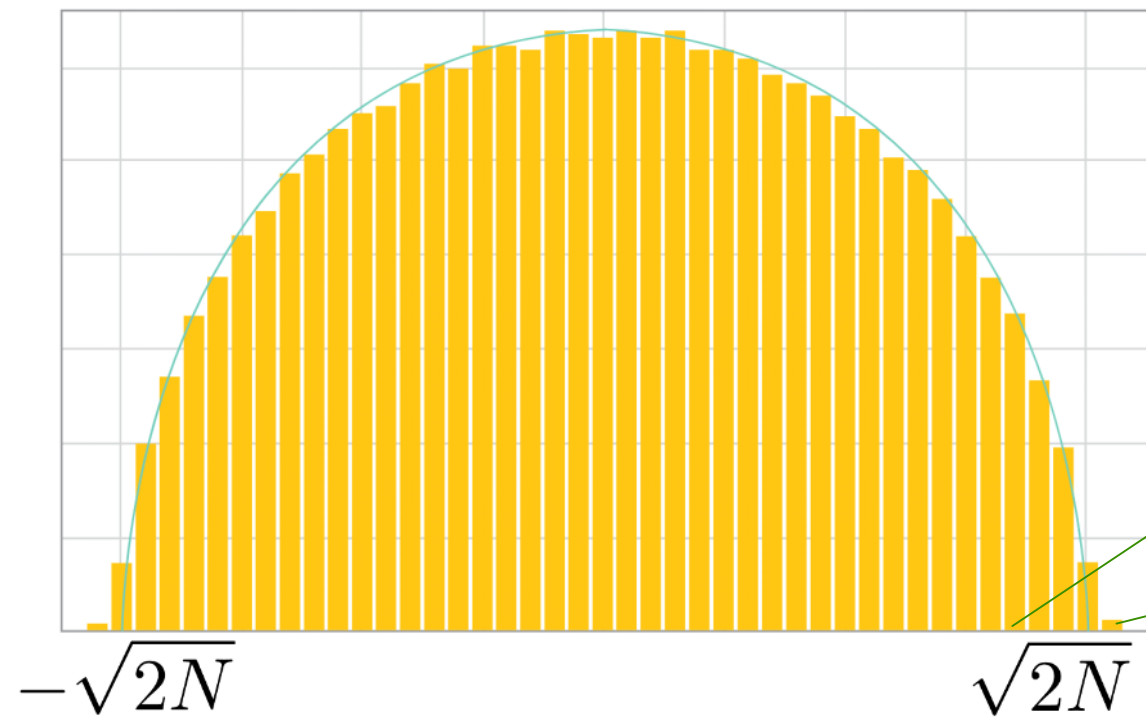
$$\partial_t \mathcal{Z}(t, x) = \frac{1}{2} \Delta \mathcal{Z}(t, x) + \mathcal{Z}(t, x) \dot{W}(t, x), \quad \mathcal{Z}(0, x) = \delta_{x=0}$$

and  $\mathbf{A}_1, \mathbf{A}_2, \mathbf{A}_3, \dots$ , is the *Airy<sub>2</sub> point process*.

- **Tomorrow:** Tsai will use this identity as the parting point to derive large deviations and tails for the KPZ equation.
- **Today:** We will derive this result using a combination of two tools:
  - *Yang-Baxter equation* and *Macdonald processes*

# Airy<sub>2</sub> point process

- Edge limit of GUE:  $\mathbb{P}_{\text{GUE}(N)}(\mathbf{A}_1, \dots, \mathbf{A}_N) \propto \prod_{1 \leq i < j \leq N} (\mathbf{A}_i - \mathbf{A}_j)^2 \prod_{j=1}^N e^{-\mathbf{A}_j^2} d\mathbf{A}_j$



- Airy<sub>2</sub> point process has other characterizations:
  - A **determinantal point process** with a simple explicit kernel.
  - The spectrum of the **stochastic Airy operator**.
  - The limit of many other point processes, e.g. the **Schur measure** that we will encounter later.



## How to prove the identity?

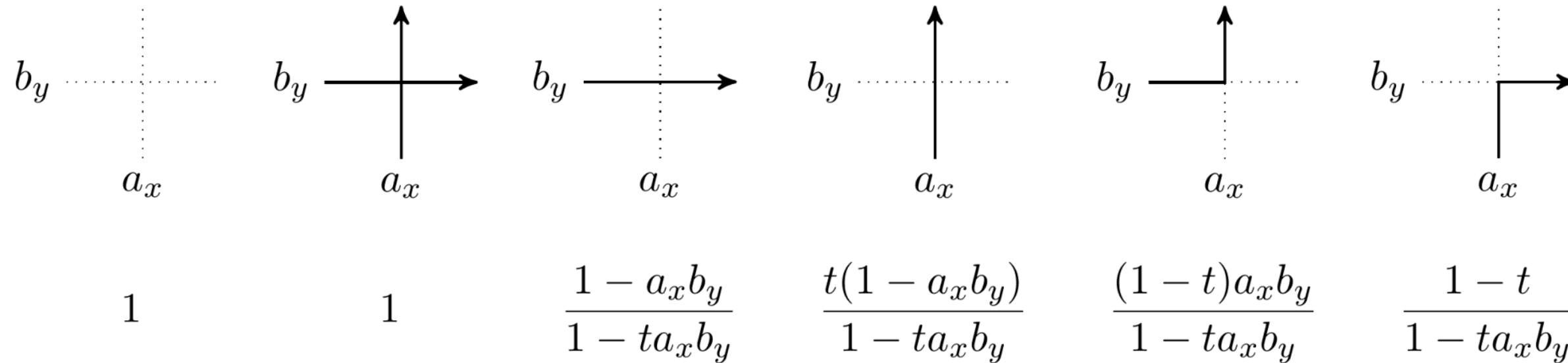
$$\mathbb{E}_{\text{SHE}} \left[ e^{-uZ(2\tau,0)} e^{\tau/12} \right] = \mathbb{E}_{\text{Airy}} \left[ \prod_{k=1}^{\infty} \frac{1}{1 + ue^{\tau^{1/3}} \mathbf{A}_k} \right]$$

- [Borodin-Gorin '16] proved this as an easy corollary of the SHE Fredholm determinant formula [Sasamoto-Spohn '10], [Calabrese-Le Doussal-Rosso '10], [Dotsenko '10], [Amir-C-Quastel '10].
- Where does such a formula come from? Can compute SHE moments via 'replica trick' (a version of Markov duality). **Moments DO NOT determine the distribution of the SHE**, so that route is not rigorous!
- Proof: We lift to a **discrete regularization**, the S6V model and use a non-trivial relationship between S6V and measures on partitions.

There are other approaches (including duality) I will not discuss...

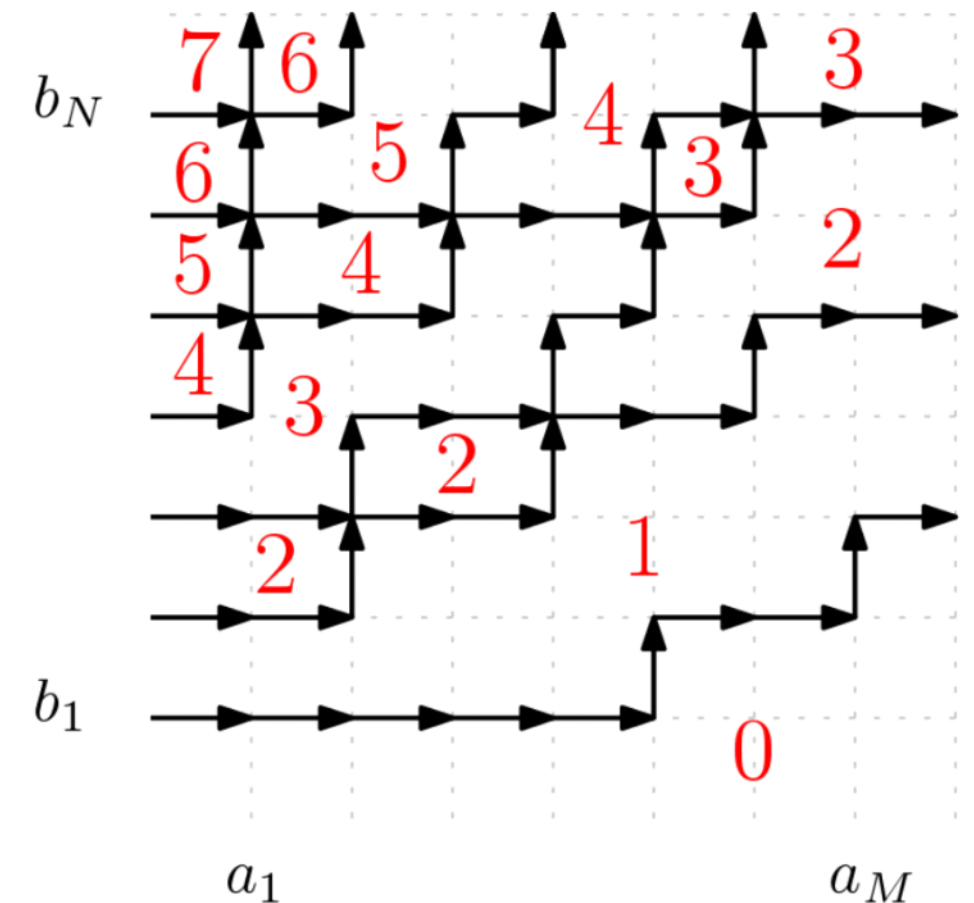
# Inhomogeneous stochastic six vertex model

- Consider an *inhomogeneous* version of S6V with weights



with  $a_x$  for column  $x$  and  $b_y$  is for row  $y$  such that  $a_x b_y$  and  $t$  in  $[0, 1)$ .

Fix *step initial data* and define a height function  $H(x, y)$  as shown (this picture is flipped versus earlier for convenience)



Recall  $(a; t)_n := \prod_{i=0}^{n-1} (1 - at^i)$

## A lifting of the identity

$$\begin{array}{ccc}
 \mathbb{E}_{S6V} \left[ \frac{1}{(-ut^{H(x,y)}; t)_\infty} \right] & \stackrel{\mathbf{2}}{=} \mathbb{E}_{HL} \left[ \frac{1}{(-ut^{\ell(\lambda)}; t)_\infty} \right] & \stackrel{\mathbf{3}}{=} \mathbb{E}_{Schur} \left[ \prod_{k \notin \{\lambda_i - i\}_{i \geq 1}} \frac{1}{1 + ut^{k+x}} \right] \\
 \downarrow \mathbf{1} & \text{Limit as } t \nearrow 1 \text{ and } x, y \nearrow \infty & \downarrow \mathbf{4} \\
 \mathbb{E}_{SHE} \left[ e^{-uZ(2\tau, 0)} e^{\tau/12} \right] & \xleftrightarrow{\text{The identity!}} & \mathbb{E}_{Airy} \left[ \prod_{k=1}^{\infty} \frac{1}{1 + ue^{\tau^{1/3} a_k}} \right]
 \end{array}$$

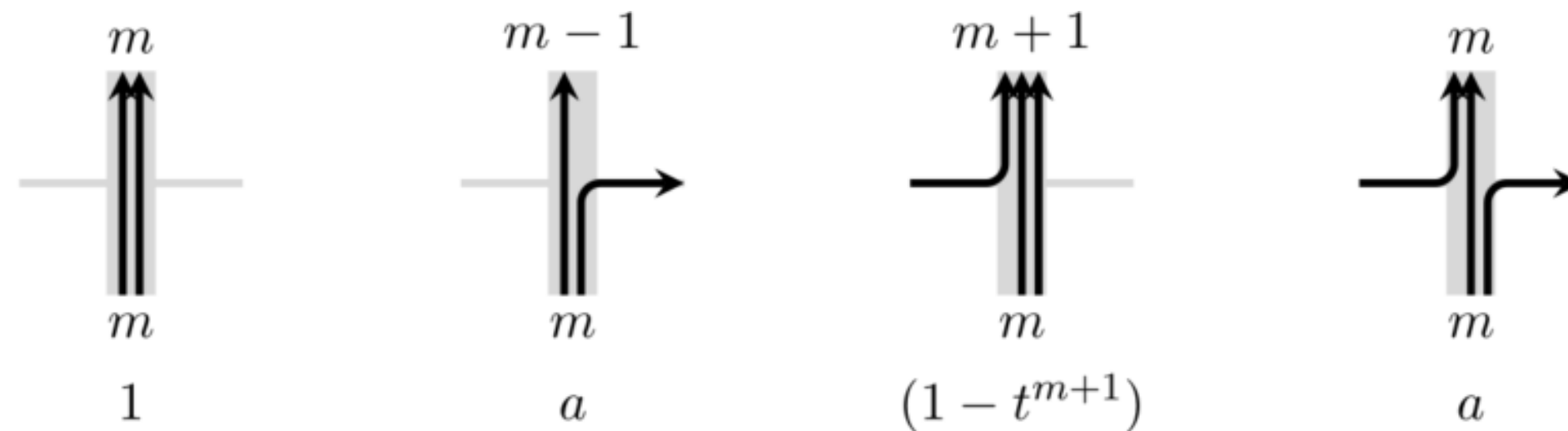
Limit 1: Convergence result from S6V to KPZ/SHE that we already saw along with limit of the Pochhammer symbol to an exponential.

Identities 2 and 3: *We will focus on these.* Identity 2 uses ideas from [Betea-Wheeler-Zinn-Justin 14] + [Borodin-Bufetov-Wheeler 16].

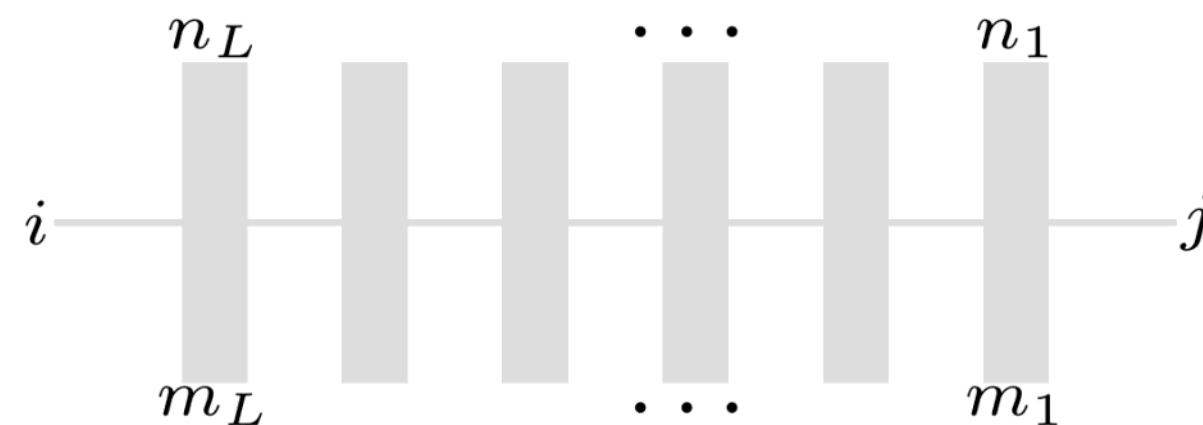
Limit 4: Follows from convergence of explicit determinantal kernels.

# $t$ -Boson vertex model

Consider the following vertex weights [Tsilevich '06] with arbitrary vertical lines and 0/1 horizontal lines, subject to line conservation:

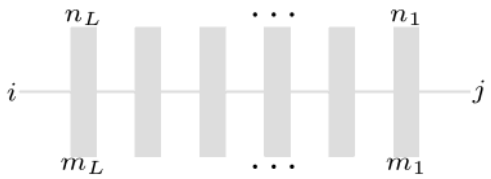


Here  $t$  and  $a$  or in  $[0,1)$ . We can put together weights like this



where the internal line configurations are summed over.

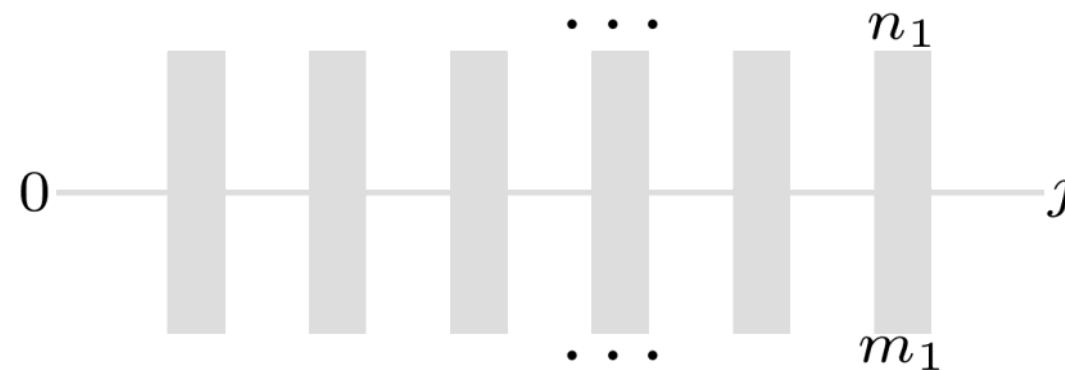
# Skew Hall-Littlewood polynomials

Taking  $L \rightarrow \infty$ , the weight of  is non-zero only when there is no incoming arrow on the right.

Partition notation:

$$\lambda = 1^{m_1} 2^{m_2} \dots$$

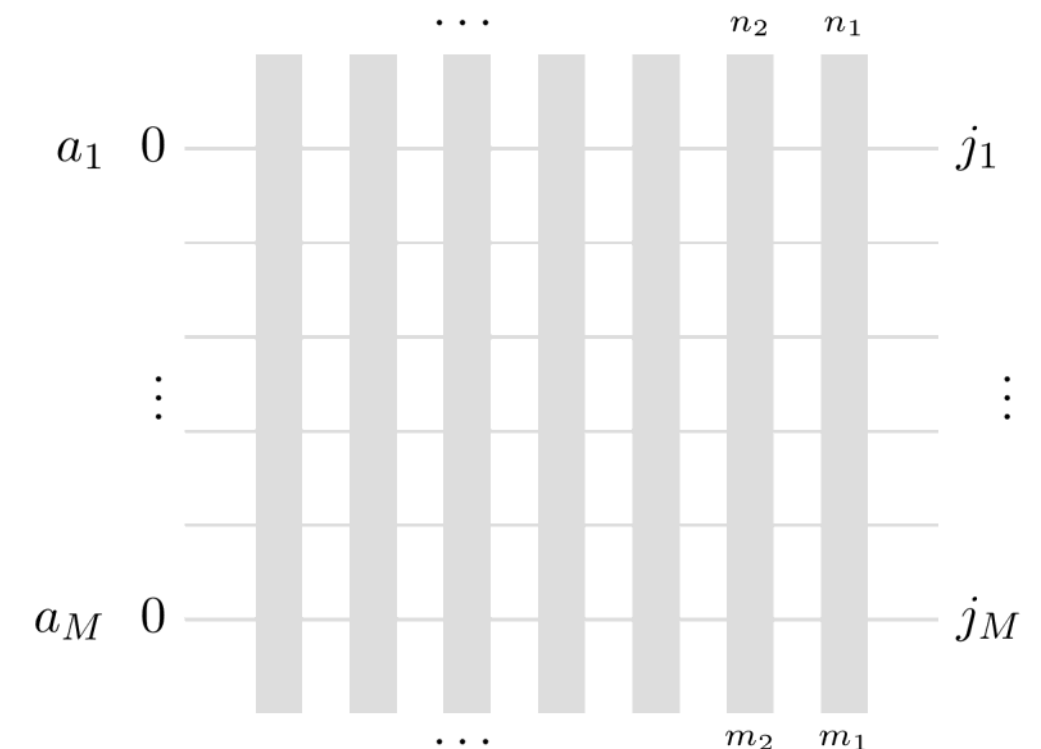
$$\mu = 1^{n_1} 2^{n_2} \dots$$



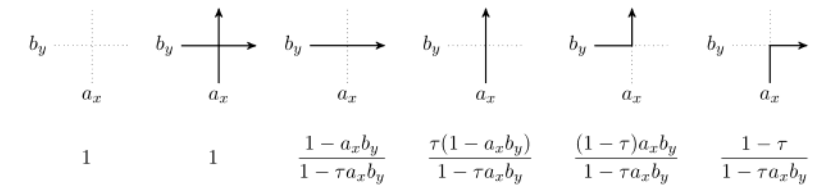
Line conservation implies partition lengths satisfy  $\ell(\lambda) = \ell(\mu) + 1$ .

**Skew Hall-Littlewood polynomial**  $P_{\lambda/\mu}(a) =$  weight from  $\lambda$  to  $\mu$ .

Multivariable skew Hall-Littlewood polynomial  $P_{\lambda/\mu}(a_1, \dots, a_M)$  involves stacking rows with variables  $a_1, \dots, a_M$  and summing weights over all possible internal configurations and  $j_1, \dots, j_M$ .



# Yang-Baxter equation



The relationship between Hall-Littlewood polynomials and the  $S6V$  model can be seen from the **Yang-Baxter equation**:

$$\sum_{0 \leq k_1, k_2 \leq 1} \sum_{p=0}^{\infty} \begin{matrix} n \\ \nearrow k_2 \\ \text{---} p \text{---} \\ \searrow k_1 \\ m \end{matrix} \begin{matrix} i_1 \\ \vdots \\ i_2 \end{matrix} \begin{matrix} j_2 \\ \vdots \\ j_1 \end{matrix} \begin{matrix} a \\ \nearrow \\ \searrow \\ b^{-1} \end{matrix} = \sum_{0 \leq k_1, k_2 \leq 1} \sum_{p=0}^{\infty} \begin{matrix} n \\ \nearrow k_1 \\ \text{---} p \text{---} \\ \searrow k_2 \\ m \end{matrix} \begin{matrix} i_1 \\ \vdots \\ i_2 \end{matrix} \begin{matrix} j_2 \\ \vdots \\ j_1 \end{matrix} \begin{matrix} b^{-1} \\ \nearrow \\ \searrow \\ a \end{matrix}$$

External lines are fixed and internal lines are summed over.

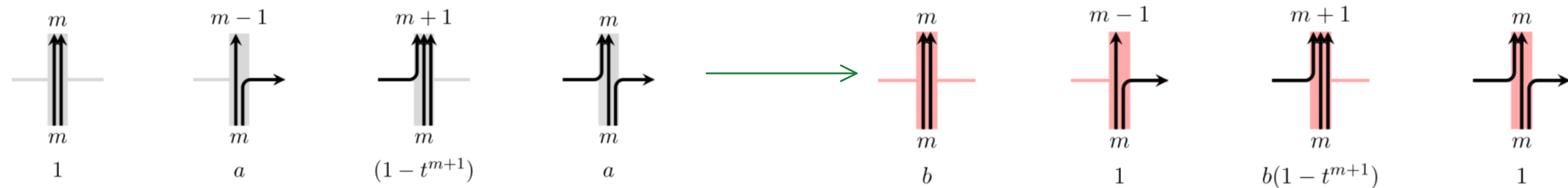
Spectral variables flip and the  $X$ 's are  **$S6V$  vertices rotated by  $45^\circ$** .

Iterating:

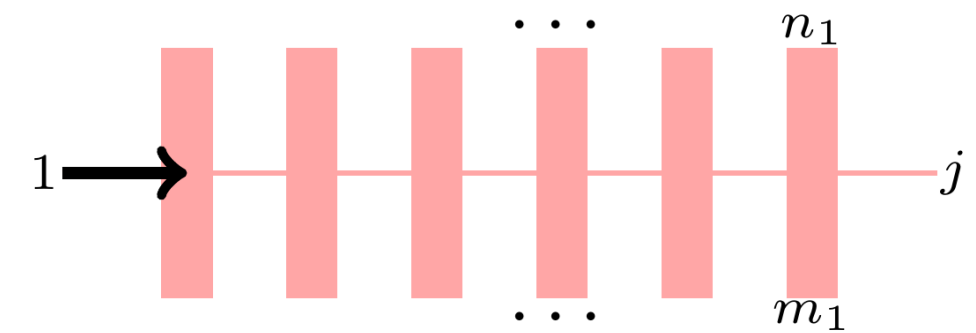
$$\sum_{0 \leq k_1, k_2 \leq 1} \sum_{p_1, \dots, p_L \geq 0} \begin{matrix} n_L & n_2 & n_1 \\ \nearrow k_2 & & \\ \text{---} p_L \text{---} & \dots & \text{---} p_2 \text{---} & \text{---} p_1 \text{---} \\ \searrow k_1 & & \\ m_L & m_2 & m_1 \end{matrix} \begin{matrix} i_1 \\ \vdots \\ i_2 \end{matrix} \begin{matrix} j_2 \\ \vdots \\ j_1 \end{matrix} \begin{matrix} a \\ \nearrow \\ \searrow \\ b^{-1} \end{matrix} = \sum_{0 \leq k_1, k_2 \leq 1} \sum_{p_1, \dots, p_L \geq 0} \begin{matrix} n_L & n_2 & n_1 \\ \nearrow k_1 & & \\ \text{---} p_L \text{---} & \dots & \text{---} p_2 \text{---} & \text{---} p_1 \text{---} \\ \searrow k_2 & & \\ m_L & m_2 & m_1 \end{matrix} \begin{matrix} i_1 \\ \vdots \\ i_2 \end{matrix} \begin{matrix} j_2 \\ \vdots \\ j_1 \end{matrix} \begin{matrix} b^{-1} \\ \nearrow \\ \searrow \\ a \end{matrix}$$

# Q polynomials and limiting Yang-Baxter equation

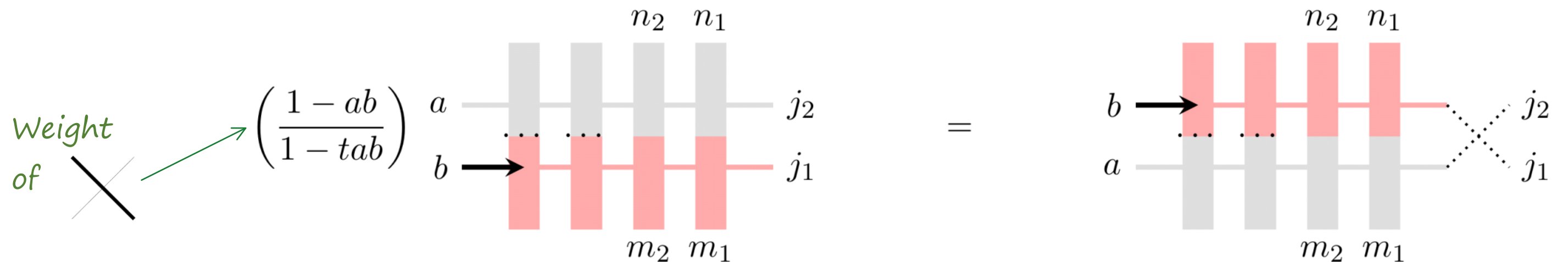
YBE involves  $a$  and  $b^{-1}$ . We introduce a dual set of  $t$ -Boson weights (in salmon) by replacing  $a$  by  $b^{-1}$  and multiplying through by  $b$ .



Composing and take  $L \rightarrow \infty$  only non-zero with right incoming line:



If we take the  $L \rightarrow \infty$  limit of the YBE, we arrive at:



# Hall-Littlewood process and Cauchy identity

Fixing parameters  $a_1, \dots, a_M$  and  $b_1, \dots, b_N$ , define the **Hall-Littlewood process** to be the probability measure on sequences of partitions:

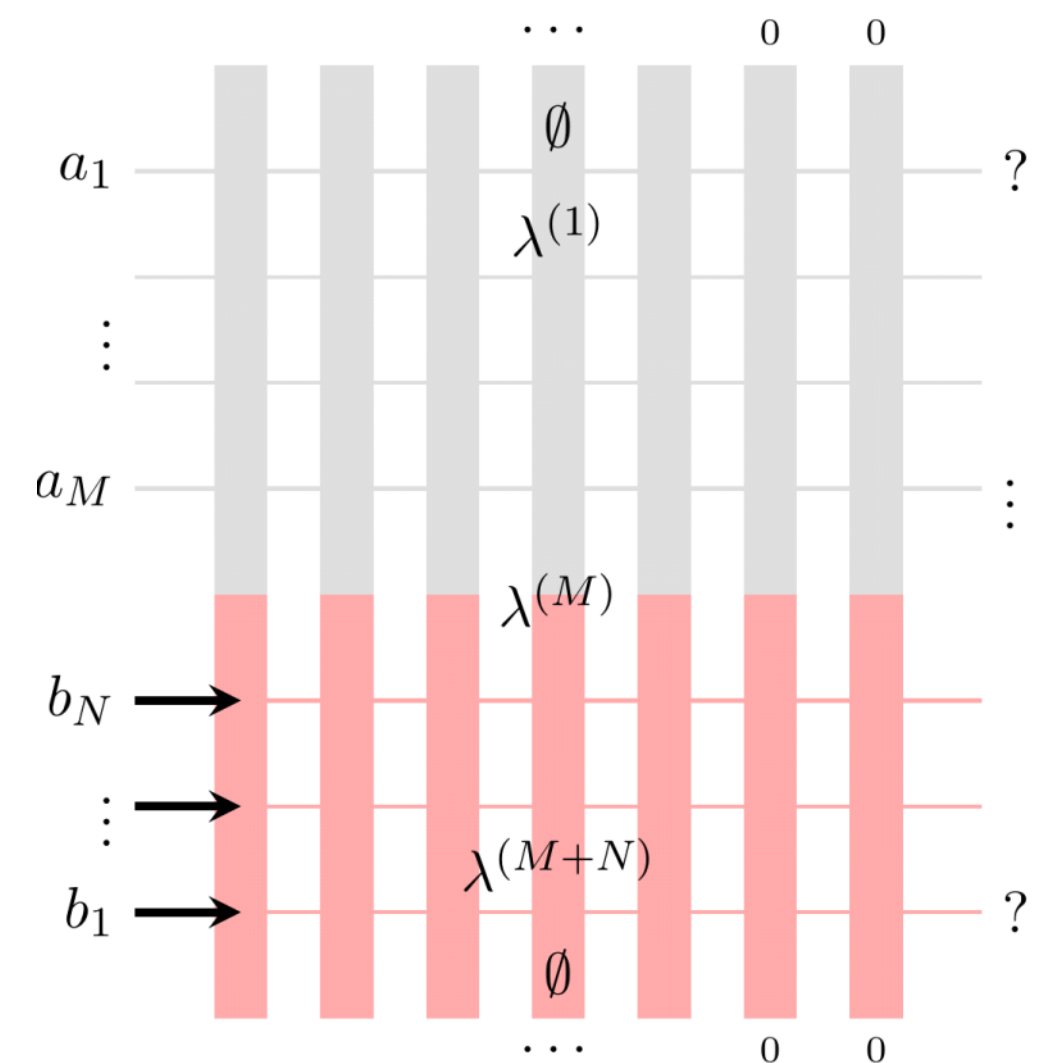
$$\mathbb{P}(\lambda^{(1)}, \dots, \lambda^{(M+N)}) = \frac{P_{\lambda^{(1)}}(a_1) P_{\lambda^{(2)}/\lambda^{(1)}}(a_2) \cdots P_{\lambda^{(M)}/\lambda^{(M-1)}}(a_M) Q_{\lambda^{(M)}/\lambda^{(M+1)}}(b_N) \cdots Q_{\lambda^{(M+N)}}(b_1)}{Z(a; b)}$$

Normalization is given by the Cauchy-Littlewood identity:

$$Z(a, b) = \prod_{i=1}^M \prod_{j=1}^N \frac{1 - ta_i b_j}{1 - a_i b_j}$$

In terms of the  $t$ -Boson vertex model, the Hall-Littlewood process is the weight of this configuration:

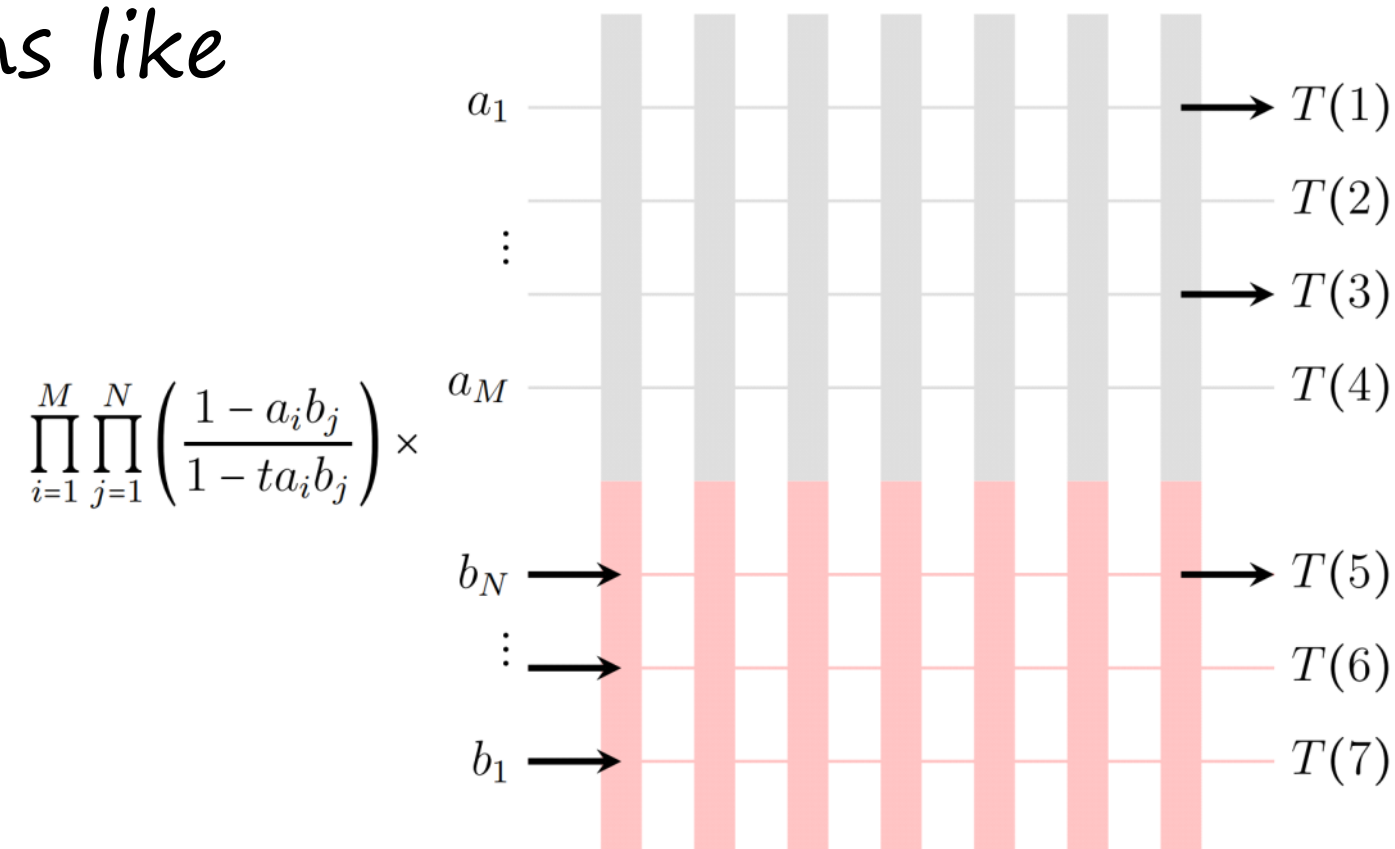
*'s relate to change in length of partitions.*



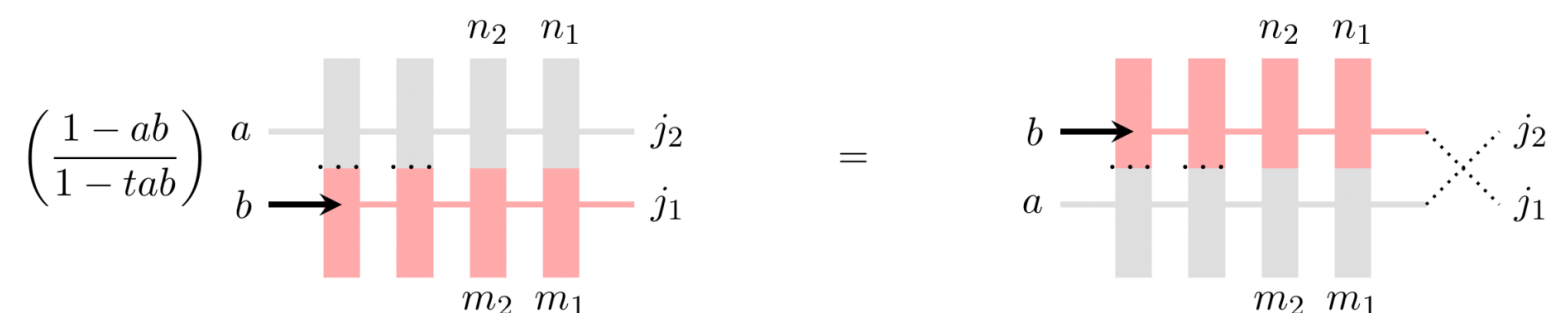


# Hall-Littlewood process lengths

Consider the probability of seeing lengths  $T(1), \dots, T(M+N)$  under the Hall-Littlewood process. This equals the sum of weights of all configurations like

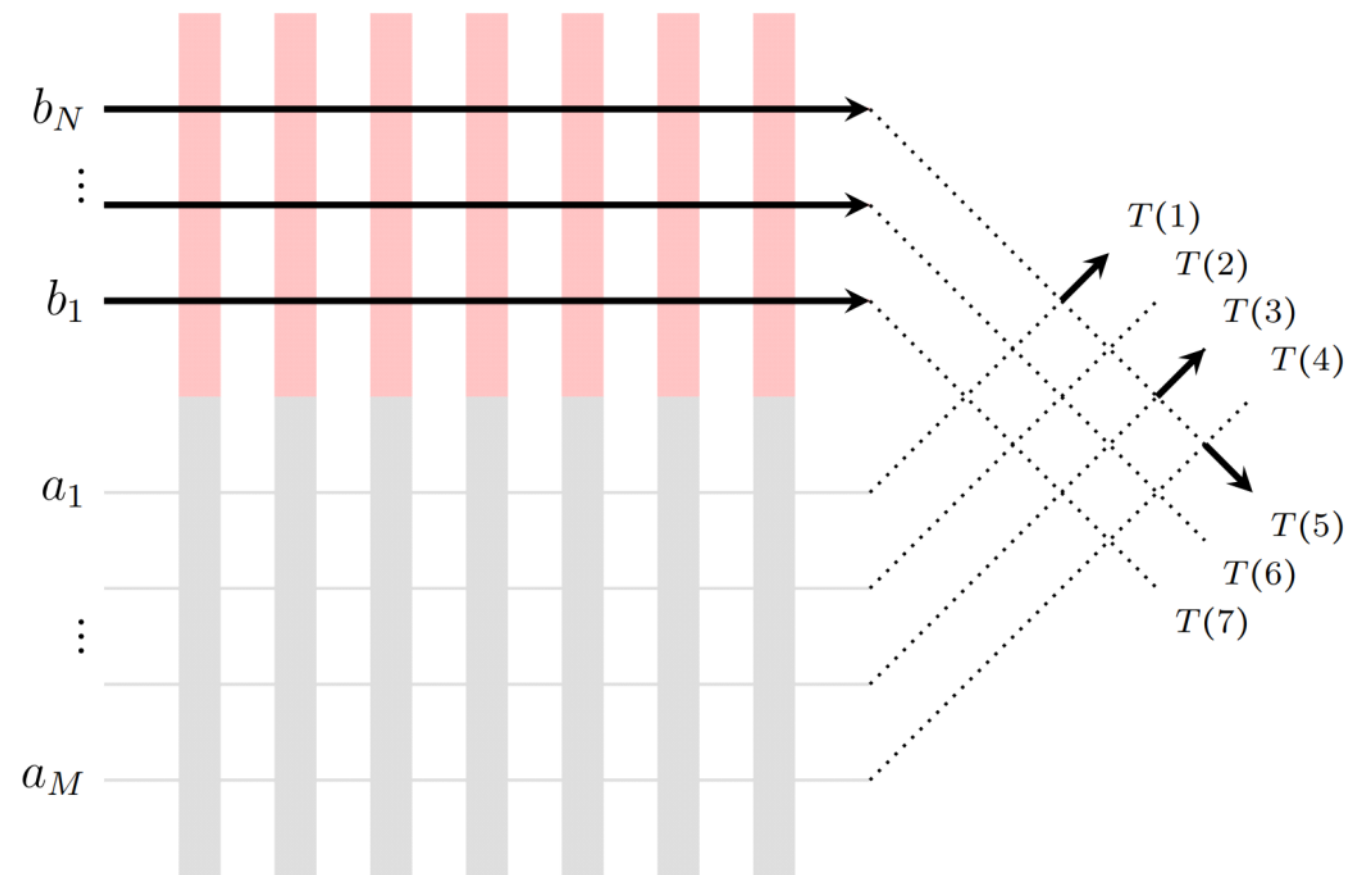


Now, use the YBE to **sequentially swap  $b$  and  $a$  rows**:



## Enter the $S6V$ height function

After swapping the order of all  $a$ 's and  $b$ 's, the normalizing constant has been fully absorbed and we are left equality to the weight of



The weight of the left side of the picture is 1 and the right side weight is precisely the probability of seeing the given sequence of output lines for the step initial data  $S6V$  model.

## We have proved identity 2

Recall identity 2:  $\mathbb{E}_{S6V} \left[ \frac{1}{(-ut^{H(x,y)}; t)_{\infty}} \right] \stackrel{\mathbf{2}}{=} \mathbb{E}_{HL} \left[ \frac{1}{(-ut^{\ell(\lambda)}; t)_{\infty}} \right]$

- We used the Yang-Baxter equation to match the distribution of the *Hall-Littlewood process lengths* to the distribution of the *output lines for the S6V model*. Identity 2 follows readily from this.
- The marginal of the Hall-Littlewood measure on a single intermediate partition is the *Hall-Littlewood measure* on a partition

$$\mathbb{P}(\lambda) = \frac{P_{\lambda}(a_1, \dots, a_M) Q_{\lambda}(b_1, \dots, b_N)}{Z(a; b)}$$

with Hall-Littlewood polynomials  $P$  and  $Q$ , and same  $Z$  as before.

### Identity 3: Enter the Macdonald measure

Replacing  $P$  and  $Q$  by Macdonald symmetric polynomials depending on  $q$  and  $t$  in  $[0,1)$ , [Borodin-C '11] define the **Macdonald measure**:

$$\mathbb{P}_{q,t}(\lambda) = \frac{P_\lambda(a_1, \dots, a_M; q, t) Q_\lambda(b_1, \dots, b_N; q, t)}{Z(a; b)}$$

The normalization is now given by the **Cauchy-Littlewood identity**:

$$\sum_{\lambda} P_\lambda(a_1, \dots, a_M) Q_\lambda(b_1, \dots, b_N) = Z_{q,t}(a, b) = \prod_{i=1}^M \prod_{j=1}^N \frac{(ta_i b_j; q)_\infty}{(a_i b_j; q)_\infty}$$

Macdonald polynomials have many degenerations, including

- **Hall-Littlewood polynomials** when  $q=0$ ,
- **Schur polynomials** when  $q=t$ .

## $q$ -independence

Define the **Macdonald difference operator** which act the  $a$ -variables

$$D_M^u = \sum_{I \subset \{1, \dots, M\}} t^{|I|(|I|-1)/2} \prod_{i \in I, j \notin I} \frac{ta_i - a_j}{a_i - a_j} \prod_{i \in I} T_{q,i} .$$

Here,  $u$  is arbitrary and  $T_{q,i}$  shifts  $a_i$  to  $qa_i$ . The **eigenrelation**

$$D_M^u P_\lambda(a; q, t) = e_\lambda(u, q, t) P_\lambda(a; q, t) \text{ with } e_\lambda(u, q, t) = \prod_{i=1}^M (1 + uq^{\lambda_i} t^{M-i})$$

not only defines the polynomials, but also enables us to calculate

$$\mathbb{E}_{q,t} [e_\lambda(u, q, t)] = \frac{D_M^u Z(a, b; q, t)}{Z(a, b; q, t)} .$$

It is easy to see that the right side is, in fact,  **$q$ -independent!**

Equating  **$q=0$  (Hall-Littlewood)** and  **$q=t$  (Schur)** yields identity 3.

## Recap

$$\begin{array}{ccc}
 \mathbb{E}_{S6V} \left[ \frac{1}{(-ut^{H(x,y)}; t)_\infty} \right] & \stackrel{\mathbf{2}}{=} \mathbb{E}_{HL} \left[ \frac{1}{(-ut^{\ell(\lambda)}; t)_\infty} \right] & \stackrel{\mathbf{3}}{=} \mathbb{E}_{Schur} \left[ \prod_{k \notin \{\lambda_i - i\}_{i \geq 1}} \frac{1}{1 + ut^{k+x}} \right] \\
 \downarrow \mathbf{1} & \text{Limit as } t \nearrow 1 \text{ and } x, y \nearrow \infty & \downarrow \mathbf{4} \\
 \mathbb{E}_{SHE} \left[ e^{-uZ(2\tau, 0)} e^{\tau/12} \right] & \xleftrightarrow{\text{The identity!}} & \mathbb{E}_{Airy} \left[ \prod_{k=1}^{\infty} \frac{1}{1 + ue^{\tau^{1/3} a_k}} \right]
 \end{array}$$

To prove an identity between SHE and Airy, we found an identity between two discrete regularizations, S6V and Schur measure.

- Identity 2 relates **S6V** to the **Hall-Littlewood process** using the Yang-Baxter equation for  $t$ -Bosons.
- Identity 3 relate the **Hall-Littlewood** and **Schur measure** using a further lifting to Macdonald measures.

## Some extensions

- Hall-Littlewood RSK:
  - [Bufetov-Matveev '17] provide a Markov chain on interlacing partitions which preserves that class of Hall-Littlewood processes and whose marginal on lengths is the S6V model.
- Yang-Baxter fields and bijectivization:
  - [Bufetov-Petrov '17] and [Bufetov-Mucciconi-Petrov '19] use Yang-Baxter equation to construct other Markov chains like above, also for higher-spin models.
- Half-space:
  - [Barraquand-Borodin-C-Wheeler '17] provide relation between half-space versions of the S6V and Hall-Littlewood process (special case of half-space Macdonald process [Barraquand-Borodin-C '18]) and prove half-space identity and asymptotics.
- Gibbsian line ensembles
  - [C-Dimitrov '17] interpret S6V and Hall-Littlewood relationship via Gibbsian line ensembles [C-Hammond '11] and prove predicted KPZ  $2/3$  transversal exponent.
- And much more...

# Summary of three lectures

- *What did I do*
  - *Lecture 1: Conjured and 'solved' the Beta RWRE out of thin air.*
  - *Lecture 2: 'Solved' S6V via Bethe ansatz diagonalization and Markov duality.*
  - *Lecture 3: Revealed a key source of solvability, the Yang-Baxter equation, and connected vertex models to symmetric function measures on partitions.*
- *What didn't I do?*
  - *Lots! For example, how does the Beta RWRE arise from S6V? From where does duality come? How does one actually perform asymptotics?...*
- *What will other people do at this program?*
  - *Imamura: Higher-spin models and another route to get Fredholm determinant formulas for measures in the Macdonald hierarchy.*
  - *Tsai: How to use the identity we discussed in this lecture to prove KPZ equation tail and large deviation results.*
  - *Basu: How inputs from integrable probability inform geometric problems in LPP.*