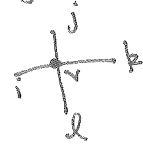


Last time, we were trying to understand "partition function" ①

$$Z = \sum_S \prod_{x \in S} R_{ij}^{kl}(x)$$

admissible states S
states x : vertex

R : Boltzmann weight
for



Motivation - Infer macroscopic thermodynamics from R - probability of state according to its energy.

but we can just think of $R_{ij}^{kl}(x)$ as abstract function.

(produce interesting class of Z 's: special functions from careful choices of function R)

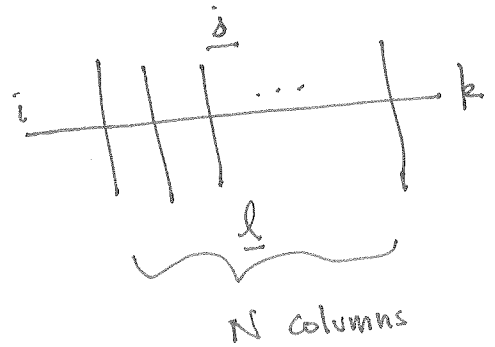
We had transformed this into algebraic

context: $V = \langle v_1, \dots, v_m \rangle$ where m : # of possible decorations on edges

define matrix $R \in \text{End}(V \otimes V)$ whose coeffs. are R_{ij}^{kl}

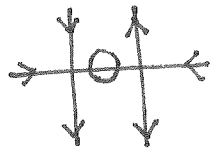
Similarly, matrix $T \in \text{End}(V \otimes (V^{\otimes N}))$ whose coeffs are

$T_{i, \underline{j}}^{k, \underline{l}}$: the partition function of the system with 1 row, N columns



In 6-vertex model, might want algebraic manifestation

of



system with one internal edge.

in our notation: T $\leftarrow, \uparrow\uparrow$
 $\rightarrow, \uparrow\uparrow$

And every other choice of boundary arrows

would give a different matrix coeff. in T .

matrix coeff. in $T \in \text{End}(V \otimes (V^{\otimes 2}))$

Claim: $T = R_{01} R_{02} \dots R_{0N}$ for any R, T corresponding to lattice model.

in our example, $T = R_{01} R_{02}$

where $R_{01} \in \text{End}(V \otimes (V^{\otimes 2}))$ means $R \otimes \text{id}$:
 $V \otimes (V \otimes V)$ with R acting on $V_{(0)} \otimes V_{(1)}$ and

similarly R_{02} means act by id. on $V_{(1)}$ and id. on $V_{(2)}$.
 by R on $V_{(0)} \otimes V_{(2)}$.

So if $R = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & \lambda & 1-g\lambda & 0 \\ 0 & 1-g^{-1}\lambda & \lambda & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$, try to describe partition function on previous page.

(see example in prior notes)

Then $Z = \sum_{i_1, \dots, i_N} \sum_{k_1, \dots, k_M} \prod_{j=1}^N T_{i_j, j}^{k_j, r^{(j)}} = \sum_{r^{(1)}} \dots \sum_{r^{(M-1)}} T_{i_1, j}^{k_1, r^{(1)}} \cdot T_{i_2, r^{(1)}}^{k_2, r^{(2)}} \dots T_{i_M, r^{(M-1)}}^{k_M, r^{(M)}}$

messy in general to explain/simplify this calculation.

If we use toroidal boundary conditions (view boundary edge as internal edge wrapped around)

In one-row partition function, $Z = \text{trace}_V(T)^{\frac{L}{d}}$
 sum of diagonal terms in map $V \otimes V^{\otimes N} \rightarrow V \otimes V^{\otimes N}$
 so $\text{trace}_V(T) = V^{\otimes N} \rightarrow V^{\otimes N}$
 and now the above really is just matrix multiplication so in multi-row Z we get...

so this is still $(\dim V)^N \times (\dim V)^N$ matrix

Z in (*) is equal to
 (with horizontal edges having toroidal
 boundary conditions)

$$\left(\left(\text{trace}_V(T) \right)^M \right)_{j, l}$$

(3)

If we also make vertical boundary edges toroidal (so edges in l, j internal)
 then we take another trace on $V^{\otimes N}$, get

$$Z = \text{trace}_{V^{\otimes N}} \left(\left(\text{trace}_V(T) \right)^M \right) \cdot \left(\text{Proposition 7.5.1. in Chari-Pressley} \right)$$

really should emphasize:
 "toroidal boundary conds."

Note: Can choose any other boundary conditions,
 just get less familiar functionals that pick
 off a particular matrix coeff. in $\text{End}(V \otimes V^{\otimes N})$.

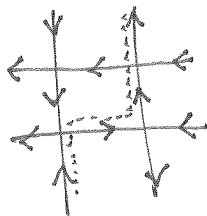
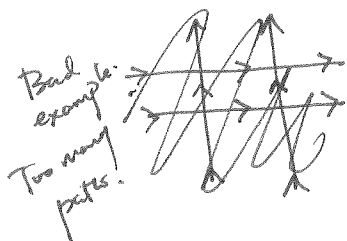
In particular as $M \rightarrow \infty$, $Z \sim K_N^M$ where K_N : largest e-value
 of $\text{trace}_V(T)$
 in $\text{End}(V^{\otimes N})$

How can we find this e-value?

Model dependent. In the context of 6-vertex model
 with toroidal boundary conditions, then

admissible decorations of lattice \leftrightarrow 2 in/2 out arrows
 at each vertex.

\leftrightarrow also assign path to, say,
 up and ~~right~~ arrows.
 left



then admissible states correspond
 to non-crossing paths moving
 through lattice.

(so biggest
 among 2^N
 e-values
 assuming they
 are distinct)

toroidal boundary conditions ensure that paths never end - old arcade / space invaders geometry - if path exits out right, then comes back on left.

Want evales of $\text{trace}_V(T) = \text{End}(V^{\otimes N})$ ←

if $n=0$, boring. ^{piece of T} is a 1×1 matrix with e-value $\lambda = \text{wt}(SE)^N + \text{wt}(SW)^N$ and on and on. (see Ch. 8 of Baxter's book)

formulas get ever more complicated. Guess shape of eigenfunctions, find all e-values. (Bethe Ansatz)

Lieb / Sutherland in '1967 to "solve" 6-vertex model with toroidal bdy. conditions.

if not toroidal, then what?

6 weights → symmetric on reversing arrows "field-free" case. 3 weights.

In nice cases: ~~B~~ We can diagonalize our transfer matrix T:

$T = P T_{\text{diag}} P^{-1}$
matrix of eigenvectors (distinct) e-values

"transfer matrices commute"

and find that P is independent of some parameters. call it x_i for T_i = in row i

Annoying to write $\text{trace}(T)$ and understand it as collapsed T. cases based on $n := \uparrow$ number of up edges ^{just write "T"} along bottom of one row system ^{for collapsed version?} (coeff. of T).

So then positions (column index) of up arrows recorded

$x_1 < x_2 < \dots < x_n$
!!
X λ

if we had e-value for T and e-vector g.

$g(x) \cdot \lambda = \sum_Y T(x, Y) g(Y)$

Because of toroidal bdy. conditions, ensures break T up into blocks.

But ~~transfer matrices~~ ^{diag. matrices} commute, so

$T(x_1) T(x_2) = T(x_2) T(x_1)$