

I Diffusive versus driven systems

Examples:

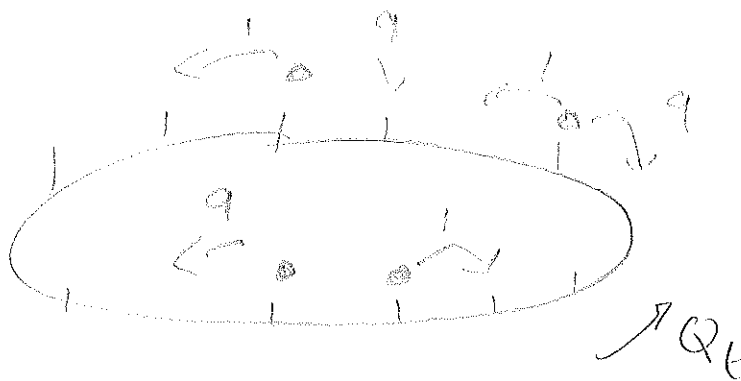
SSEP WASEP

ASEP TASEP

diffusive

driven

N particles on a ring of L sites



$q = 1$ SSEP

symmetric simple
exclusion
process

$q = 1 - \frac{\lambda}{L}$

WASEP

↑ weakly δ

$q < 1$

ASEP

$q = 0$

TASEP

(E2)

Steady state of all these systems =

all configurations are equally likely

N particles on L sites

(Note that for $q \neq 1$, they don't

satisfy detailed balance)

But:

$$\frac{dP(\mathcal{C})}{dt} = \sum_{\mathcal{C}'} \Gamma(\mathcal{C}, \mathcal{C}') P(\mathcal{C}') - \Gamma(\mathcal{C}', \mathcal{C}) P(\mathcal{C})$$

if all $P(\mathcal{C})$ are equal = P

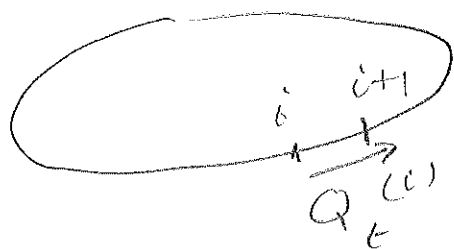
$$\sum_{\mathcal{C}'} \Gamma(\mathcal{C}, \mathcal{C}') = (1+q) n(\mathcal{C})$$

number of clusters in $\mathcal{C}$

$$\sum_{\mathcal{C}'} \Gamma(\mathcal{C}, \mathcal{C}') = (1+q) n(\mathcal{C})$$

So: $P(\mathcal{C})$ indep of $\mathcal{C}$ is

the steady state



$$J \equiv \frac{d Q_t^{(i)}}{dt} = \langle n_i (1 - n_{i+1}) \rangle - q \langle n_i n_{i+1} (1 - n_i) \rangle$$

$$= \frac{N(L-N)}{L(L-1)} (1-q)$$

$$\approx p(1-p)(1-q) \quad \text{Steady state}$$

For $q = 1 - \frac{d}{L} \Rightarrow J = p(1-p) \frac{d}{L}$

diffusive

TASEP $q = 0$

$$J = p(1-p)$$

driven

~~Hydrodynamic equation~~

~~$$\frac{dp}{dt} = - \frac{dJ}{dx}$$~~

q indep of L

$$J = p(1-p)(1-q)$$

(E4)

II Hydrodynamic equation

Suppose a situation close to equilibrium where the density varies on a large scale L

$$\langle n_i \rangle = \rho \left(\frac{i}{L} \right)$$

$$\frac{i}{L} = x$$

Diffusive system

microscopic current $J_i = \frac{1}{L} j(x)$

evolution

$$\frac{dp}{dt} = J_0 - J_L = -\frac{1}{L^2} \frac{dj}{dx}$$

$$\text{time} \sim L^2$$

Driven system

$$J = j(x)$$

$$\frac{dp}{dt} = -\frac{1}{L} \frac{dj}{dx}$$

$$\sim L$$

$$J = \rho(1-\rho)$$

TASEP

$$\frac{dp}{dt} = - \frac{d}{dx} [p(1-p)]$$

Burgers equation

deterministic equation

$$p(x, 0) = p_0(x)$$

can be solved in a parametric

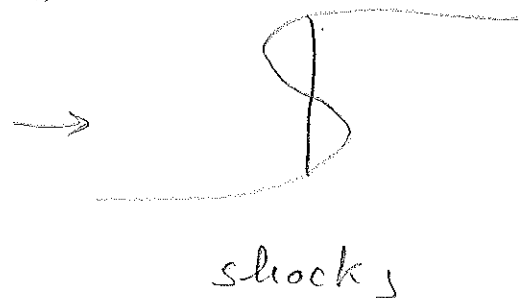
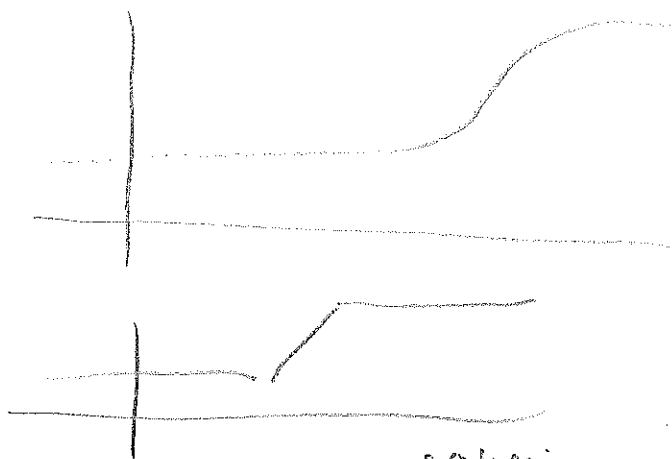
form : one varies y

$$p(x, t) = p_0(y)$$

$$x = y + (1 - 2p_0(y))t$$

Exercise : show that this is a solution
of Burgers equation

generalise it to $\frac{dp}{dt} = - \frac{d f(p)}{dx}$



$$\frac{dp}{dt} = - \frac{d}{dx} (p(1-p))$$

$$p = A(t) + B(t)x$$



$$\frac{dB}{dt} = \gamma B^2$$

$$\frac{dA}{dt} = \gamma AB - B$$

$$B(t) = \frac{B(0)}{1 - \gamma B(0)t}$$

$$B(0) < 0$$

fan

$$B(t) \rightarrow 0$$

$$B(0) > 0$$

stock

$$B(t) \rightarrow 0 \text{ at } t = \frac{1}{\gamma B(0)}$$

$$t = \frac{1}{\gamma B(0)}$$

III

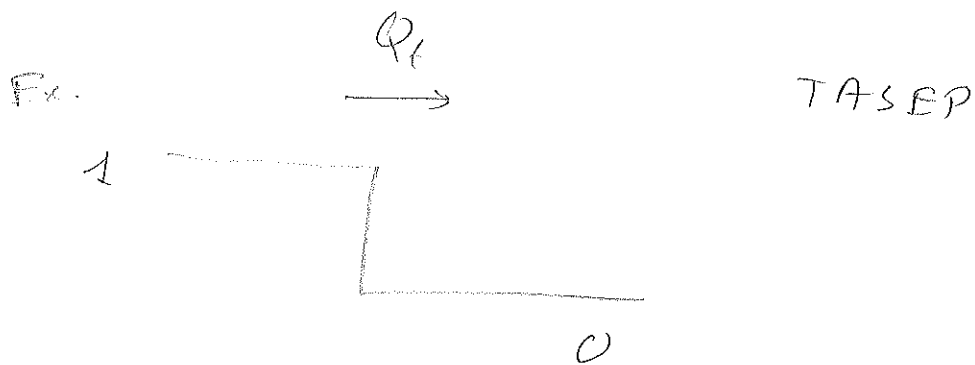
Fluctuations of the curve

(E7)

Burgers + Noise $\Rightarrow$

Kardar - Parisi Zang 1986

KPZ



$$Q_t = \frac{t}{4} + \text{fluctuations}$$

$\downarrow$

(diffusive π)

$t^{1/3} \times \text{random variable}$

$\downarrow$

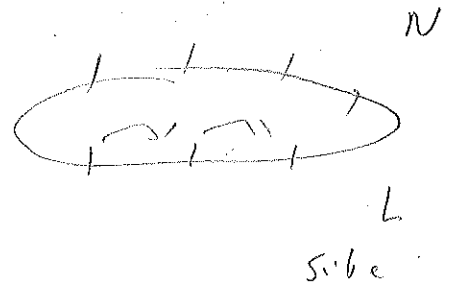
Tracy Widom
distribution

(random matrix)

~ 2008

On the ring

exact



$$P(Q) \sim \exp - t \underbrace{F_{N,L}(q)}_{\text{explicitly}} \quad \text{for all } N \text{ and } L$$

$N, L \rightarrow \infty$

Universal scaling function

$$\frac{N}{L} = p$$

$$q > q_{\text{hyp}}$$

F much larger
the

$$q < q_{\text{hyp}}$$

F slower

slow down

F slower much q , F faster

Bethe ansatz calculation
 q large

$$\boxed{q \rightarrow 0}$$

$q \rightarrow \infty$

$$e^{-t}$$

$$e^{-t(N q \log q)}$$

TASEP

D. Lebowitz 38

D. Appert 39

E7b

$$\langle e^{iQ(t)} \rangle \sim e^{\mu(\lambda)t}$$

$$\mu(\lambda) = \rho(1-\rho)\lambda$$

$$+ \sqrt{\frac{\rho(1-\rho)}{2\pi N^2}} G\left(\lambda \sqrt{N(\rho(1-\rho))}\right)$$

G universal function

$$G(\beta) = \sum_{n \geq 1} (-1)^{n+1} \frac{2^n}{n^{5/2}}$$

$$\beta = \sum_{n \geq 1} (-1)^{n+1} \frac{2^n}{n^{3/2}}$$

G universal

IV.

Matrix method for the

open system

in the steady state

Goal

$$\rightarrow P_{ss}(\{p_k\}) \sim e^{-L F(\{p_k\})}$$

Matrix method

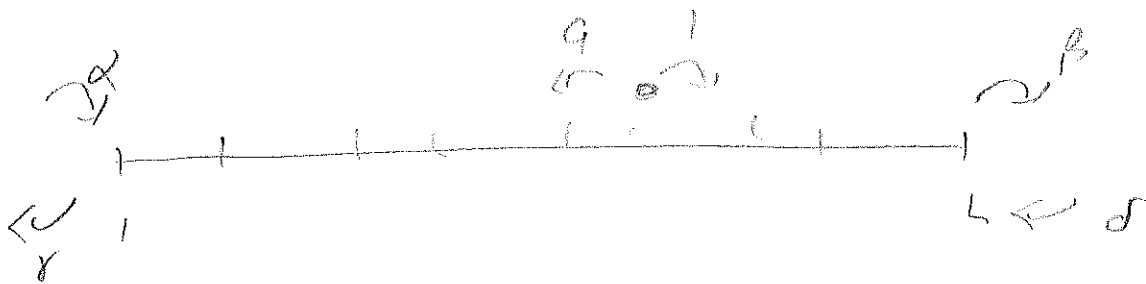
(initially used for

quantum systems)

Exact description of the steady state

for the SSEP, ASEP --

Evans, Hakim, Pasquier 1992



$P_{\text{steady state}} (n_1, \dots, n_L)$ can be determined exactly

$\langle w | = (w_1, w_2, \dots, w_K) \leftarrow \text{vectors}$

$$|v\rangle = \begin{pmatrix} v_1 \\ \vdots \\ v_K \end{pmatrix}$$

D, E matrices
or
operators

(E9)

$$P(\tau_1, \dots, \tau_L) = \frac{1}{Z_L} \langle W | \prod_{i=1}^L (\tau_i D + (-\tau_i) E) | V \rangle$$

$$DE - q E D = D + E$$

$$\langle W | (\alpha E - \gamma D) = \langle W |$$

$$(p D - \delta E) | V \rangle = | V \rangle$$

decreases the degree

(Note that any product

$$D^{n_1} E^{m_1} D^{n_2} E^{m_2} \dots$$

can be reduced to sums of

$$E^m D^n$$

(2^L numbers reduce to L^2 numbers)

E9a

Example

TASEP

$$\gamma = \delta = 0$$

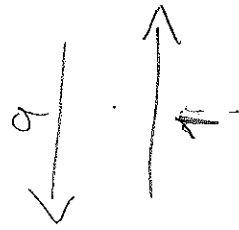
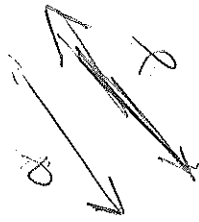


$$\langle W | E^n D^m | v \rangle = \frac{1}{d^n} \frac{1}{\beta^m}$$

Check



$$\langle W | D^3 E^4 \rangle$$

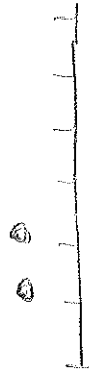


- gain

Loss

$$\begin{aligned} & + \alpha \langle W | E D^2 E^4 | V \rangle \\ & + q \langle W | D^2 E D E^3 | V \rangle \\ & + \beta \langle W | D^3 E^3 D | V \rangle \end{aligned}$$

$$\begin{aligned} & - \alpha \langle W | D^3 E^4 | V \rangle = \langle W | D^2 E^4 | V \rangle \\ & - \langle W | D^3 E^4 | V \rangle = - \langle W | D^2 (D+E) E^3 | V \rangle \\ & - \delta \langle W | D^3 E^4 | V \rangle = \langle W | D^3 E^3 | V \rangle \end{aligned}$$



(E11)

Can one find such matrices

Yes

Example

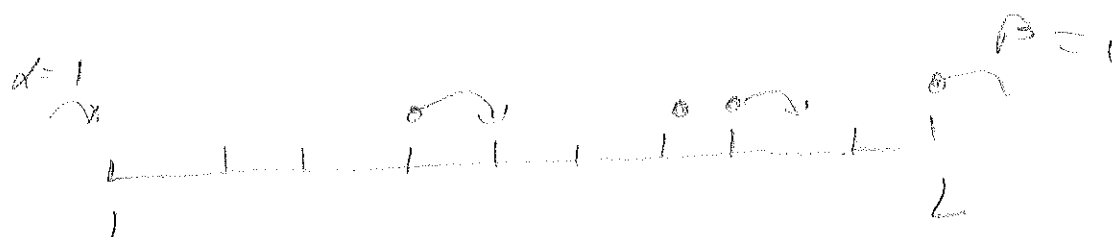
$$\alpha = 1$$

$$\beta = 1$$

$$\gamma = 0$$

$$\delta = 0$$

TASEP



$$D = \begin{pmatrix} 1 & 1 & & & 0 \\ & \diagdown & & & \\ & 0 & \diagdown & & \\ & & & \diagdown & \\ & & & & 1 \end{pmatrix}$$

$$E = \begin{pmatrix} 1 & & & & \\ & \diagdown & & & \\ & 1 & \diagdown & & \\ & & & \diagdown & \\ & & & & 1 \end{pmatrix}$$

$$\langle V | = (1, 0, \dots)$$

$$|V\rangle = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

(E1.2)

Is it necessary : no

the algebraic rules allow to
calculate all the weights

V Steady state properties

(1) Profile

$$\langle n_i \rangle = \frac{\langle W | (D+E)^{i-1} D (D+E)^{L-i} | V \rangle}{\langle W | (D+E)^L | V \rangle}$$

(2) Correlations

$$\langle n_i n_j \rangle = \frac{\langle W | (D+E)^{i-1} D (D+E)^{j-i-1} D (D+E)^{L-j} | V \rangle}{\langle W | (D+E)^L | V \rangle}$$

(3) Current in the steady state

$$\begin{aligned} \langle n_i (1 - n_{i+1}) \rangle &= q \langle n_{i+1} (1 - n_i) \rangle \\ &= \frac{\langle W | (D+E)^{i-1} (DE - qED) D^{L-i-1} | V \rangle}{\langle W | (D+E)^L | V \rangle} \end{aligned}$$

The curvat is

(E13)

$$\frac{\langle W | (D+E)^{L-1} | V \rangle}{\langle W | (D+E)^L | V \rangle}$$

VI Non gaussian fluctuations

Prob on large scale of a density profile

TASEP

$d=1$, $\beta=1$

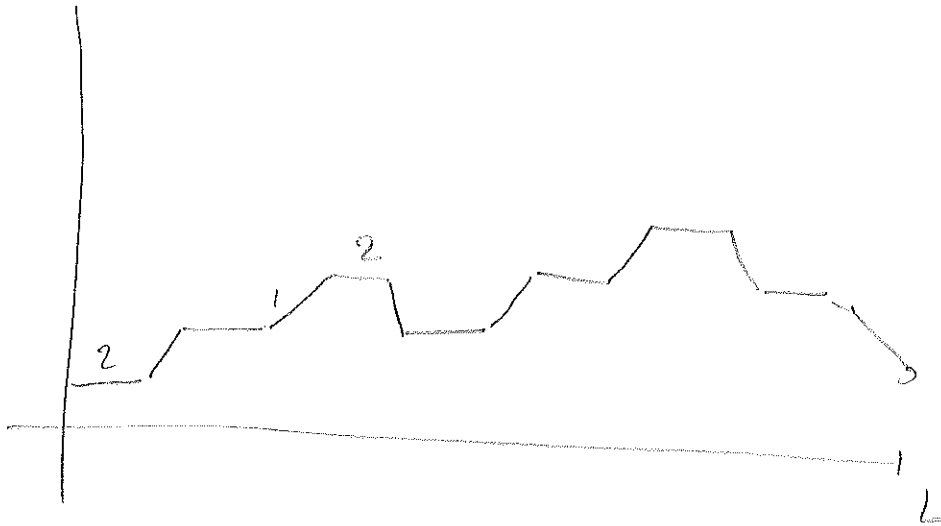
$$D+E = \begin{pmatrix} 2 & 1 & & \\ 1 & 2 & 1 & \\ & 1 & 2 & \\ & & \ddots & \ddots \end{pmatrix}$$

$$\langle W | = (1, 0, \dots)$$

$$|V\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

$$\langle W | (D + E)^L | V \rangle$$

E 14



Paths which remain positive

$$\langle W | (D + E)^L | V \rangle = \frac{(2L+2)!}{(L+2)! (L+1)!} \quad L \geq 2$$

Catalan number

$\Rightarrow$

$$\rho(x) - \frac{1}{2} \approx \frac{1}{\sqrt{L}} \left(\overset{\uparrow}{B(x)} + \overset{\uparrow}{Y(x)} \right)$$

$$x = \frac{1}{L}$$

Brownian

~~extra~~ motion

Brownian

excursion

Non gaussian fluctuations

VII How to compute large deviations

$$P(\{p(x)\}) \sim e^{-L \mathcal{F}(p(x))}$$

1 way

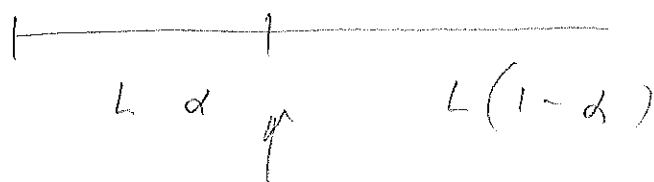
$$\text{Ex: } P(n_1, \dots, n_L) = \langle w | \prod (n_i D + (1-n_i) E) | V \rangle$$

$$\left\langle \frac{\sum d_i n_i}{L} \right\rangle = \langle w | (D e^{d_i} + E) | V \rangle \sim e^{L \mathcal{F}(d_i/n_i)}$$

$$d_i = L \left(\frac{i}{L} \right)$$

in general G is non local.

One ...

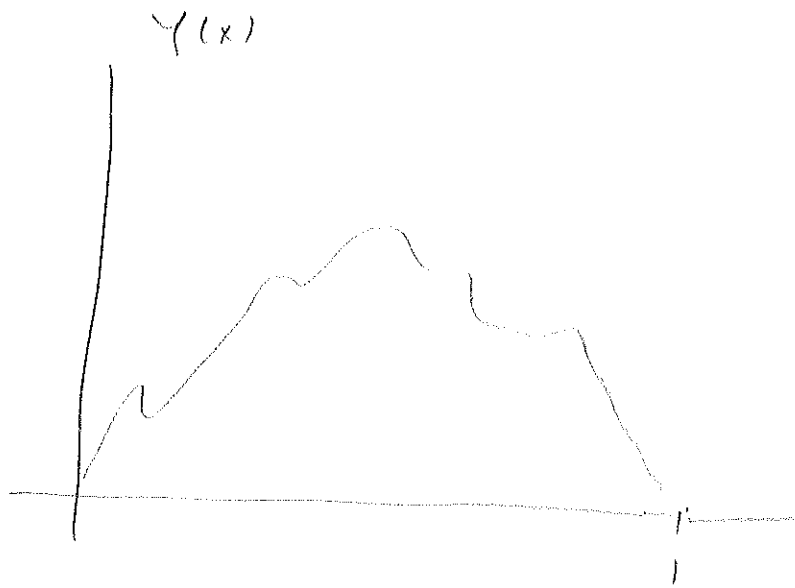


insert a basis of vectors

(2)

Another way

(E16)

 $Y(x)$ Brownian excursion

$$\int_0^1 \phi'(x) dx$$

$$\Rightarrow \left(\rho(x) - \frac{1}{2} \right) = \frac{1}{\sqrt{L}} \left(B^{\circ}(x) + Y^{\circ}(x) \right)$$

↑

Brownian
motion

↑

Brownian
excursion

↑

Enaud - Leboritz
2004

Non gaussian

