

On Shocks in the TASEP

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This talk is about a type of
wave motion
based on the article arXiv:1801.06193.

Recall the wave equation

$$\partial_t^2 u - \partial_x^2 u = 0.$$

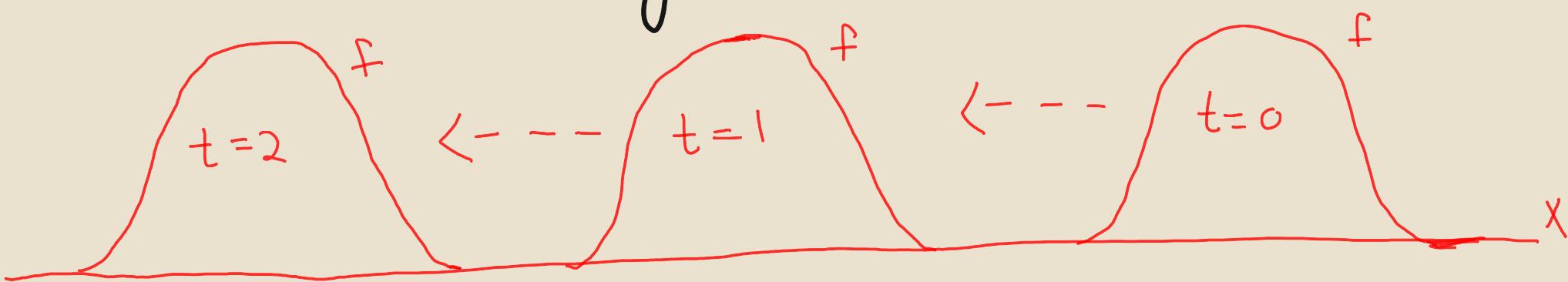
An even simpler one is

$$\partial_t u - \partial_x u = 0.$$

This is because the solution of the latter is

$$u(t, x) = f(t+x), \quad f = u(0, \cdot),$$

which is a travelling front:



The solution of the wave equation is a superposition of travelling fronts:

$$u(t, x) = f(t+x) + g(t-x).$$

But waves can easily form shocks:



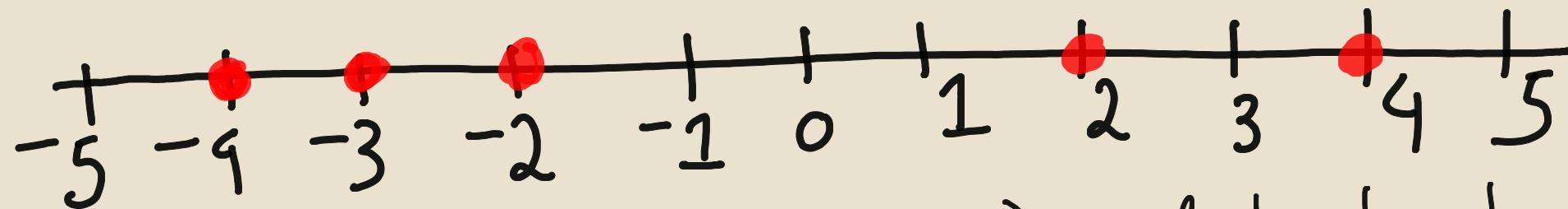
They are modelled by introducing nonlinearities in the wave equation. A prototype is Burgers' equation

$$\partial_t u - u \partial_x u = 0.$$

Shocks form when the initial wave profile decreases in the direction of flow ($\partial_x u_0 < 0$) somewhere.

There are microscopic models to study Burgers' equation, certain interacting particle systems, which discretize it.

The most basic is the TASEP. totally asymmetric simple exclusion process



- Each ● jumps right ($x \mapsto x+1$) randomly at rate 1.
- All jumps are independent.
- But subject to exclusion rule: ● cannot jump on top another ●.

Suppose particles are initially placed such that there is a macroscopic density profile u_0 :

$$\int_a^b u_0(x) dx = \lim_{L \rightarrow +\infty} \frac{1}{L} \cdot \# \left\{ \begin{array}{l} \text{Particles in } [aL, bL] \\ \text{at time 0.} \end{array} \right\}.$$

Then there is a macroscopic density profile $u(t, x)$ for further times:

$$\int_a^b u(t, x) dx = \lim_{L \rightarrow +\infty} \frac{1}{L} \cdot \# \left\{ \begin{array}{l} \text{Particles in } [aL, bL] \\ \text{at time } tL. \end{array} \right\}.$$

Furthermore, the density $u(t, x)$ satisfies
Burgers' equation in the form



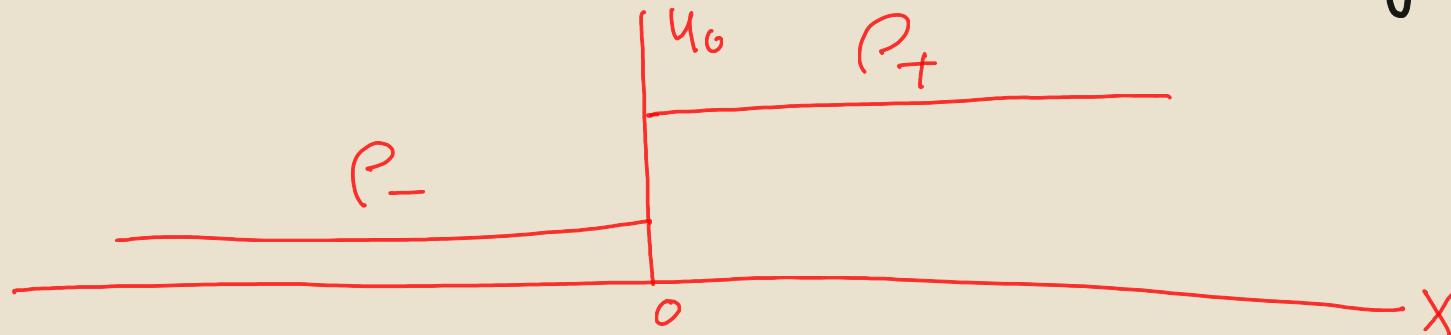
$$\partial_t u - \partial_x (u(1-u)) = 0,$$

$$u(0, x) = u_0(x).$$

It is the entropic solution (the physical one).

General theorems like this in the context of hydrodynamic limits of interacting particles.

The most basic case of the shock is the increasing step:



The (entropic) solution of $\textcircled{*}$ is the travelling front

$$u(t, x) = u_0(x - \nu t), \text{ where}$$

$\nu = 1 - P_- - P_+$ is the front speed.

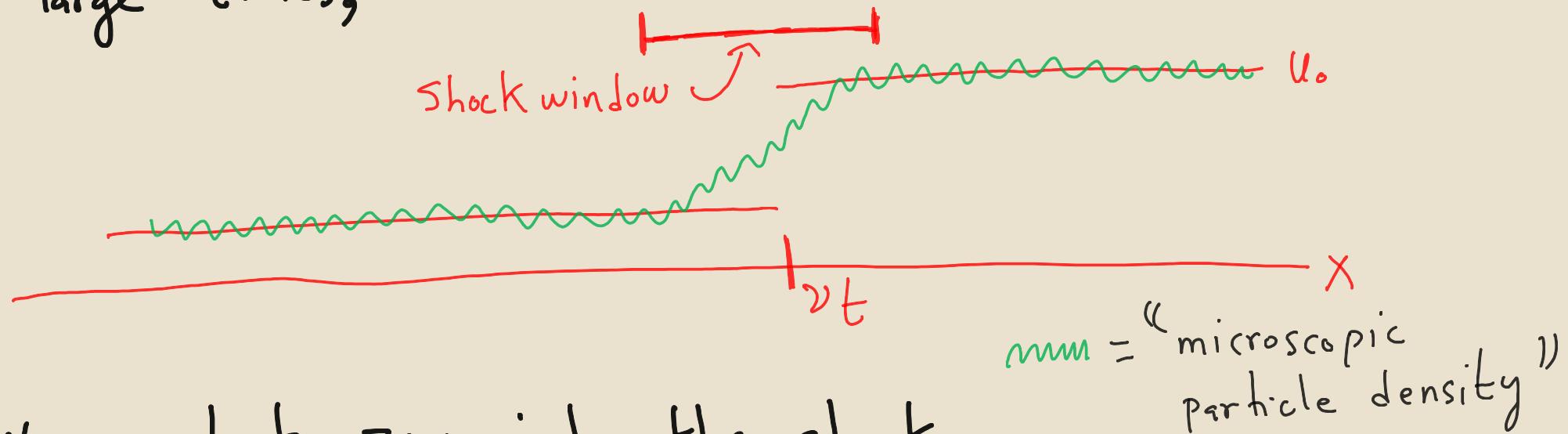
In TASEP terms:

- Particles left of origin move at speed $1 - P_-$ and those to the right move at speed $1 - P_+$.
- Since $1 - P_- > 1 - P_+$, they collide to create a jam, which moves at speed ν .

Model u_0 by placing particles on \mathbb{Z} in a deterministic way such that

$$\rho_{\pm} = \lim_{L \rightarrow \infty} \frac{1}{L} \cdot \#\{\text{particles in } [0, \pm L]\}.$$

For large times, the TASEP looks like so



$m_m = \text{"microscopic particle density"}$

We want to zoom into the shock window and observe the microscopic particle fluctuations.

We will do so by introducing a parameter to soften the shock.

For $\beta \in \mathbb{R}$ and $t \geq 2|\beta|^3$, set

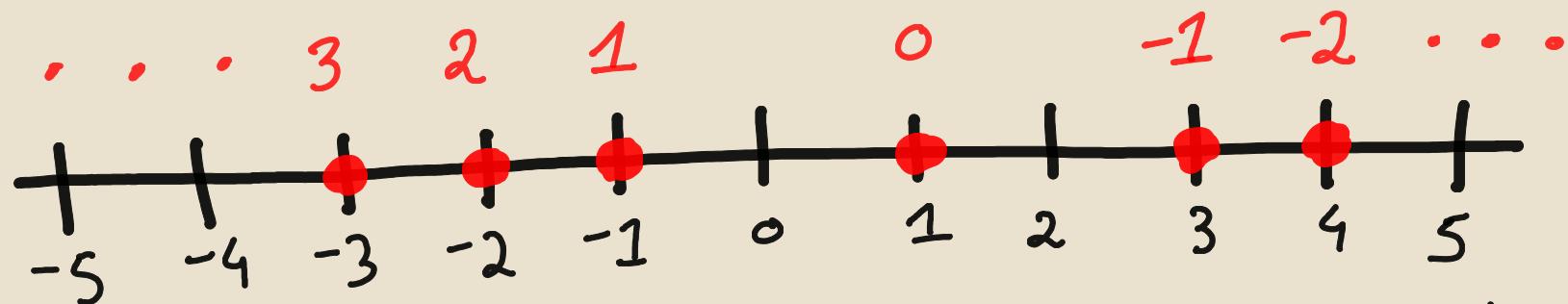
$$\rho_{\pm} = \frac{1 \pm \beta (t/2)^{-1/3}}{2}.$$

- Run TASEP to time t , t as above, and observe shock window.
- Shock fluctuations are derived in the double limit $t \rightarrow \infty$ followed by $\beta \rightarrow \infty$.
- This mysterious yet judicious rescaling of densities is motivated by KPZ theory . . .

which allows for :

- ① Exact calculations of relevant statistics inside the shock window for large times.
- ② Analysis of these statistics in the large β limit, which are given by Fredholm determinants.
- ③ A process level description of the microscopic particle fluctuations inside the shock window by way of ① and ②.

Tag the particles in TASEP like so



$X_t(n)$ = position of particle n at time t .

The particle that is macroscopically at $2t$ at time t

has number $n_t^{\text{shk}} = (\rho_- \cdot \rho_+) t$.

- Particles within distance $t^{2/3} |\beta|$ of n_t^{shk} are correlated with it.
- Their positions fluctuate to the order $t^{1/3}$.

So consider $n(t, x) = n_t^{shk} + x \left(\frac{t^{2/3}}{2\beta} \right)$, $x \in \mathbb{R}$.

Define

$$\Psi_t^\beta(x) = \frac{X_t(n(t, x)) - (x/\beta)(t/2)^{2/3}}{-(t/2)^{1/3}}$$

macroscopic
position of
 $n(t, x)$.

Theorem. In the limit $t \rightarrow \infty$ followed by $\beta \rightarrow \infty$,
 Ψ_t^β converges in the sense of finite dimensional laws to
 $\Psi(x) = \max \left\{ 2^{-2/3} X_{TW_1} - x, 2^{-2/3} X'_{TW_1} + x \right\}$.

X_{TW_1} and X'_{TW_1} are independent GOE Tracy-Widom random variables.

Corollary: $\lim_{\beta \rightarrow \infty} \lim_{t \rightarrow \infty} \Pr(X_t(n_t^{shk}) \geq -at^{\frac{1}{3}}) = F_1(2a)^2$.

\downarrow
CDF of
GOF TW.

The theorem should be interpreted as follows.

$$\max \left\{ \bar{2}^{-\frac{2}{3}} X_{TW_1} - x, \bar{2}^{-\frac{2}{3}} X'_{TW_1} + x \right\} = |x - D| + S,$$

$$D = (X_{TW_1} - X'_{TW_1}) / 2^{\frac{5}{3}} \text{ and } S = (X_{TW_1} + X'_{TW_1}) / 2^{\frac{5}{3}}.$$

So ψ_t^β converges to a \checkmark that is centered at D
 and shifted by S .

The microscopic density () is

$\frac{d}{dx} V =$ an increasing step with jump at D .

The microscopic position of the shock is at D .

This is simply because

$X_t = \text{particle position} = S_{\text{particle density}}$.

More remarks.

- The soft shock setting was introduced by [Ferrari-Nejjar, 14]
They conjectured the corollary.
- Sometimes the microscopic position of the shock
is identified with the position of the 2nd class
particle. It is indeed proved by [Ferrari-Ghosal-Nejjar, 17]
that the rescaled position of the 2nd class
particle converges to D in the setting of the
increasing step.

- One can also model the increasing step with random initial data: place a particle at site x with probability ρ_- (if $x < 0$) or ρ_+ (if $x > 0$).
 In this case the shock fluctuations are of the order $t^{1/2}$. Here the randomness of the initial data overtakes the randomness inherent in TASEP, and shock fluctuations become Gaussian. A series of works by (Pablo) Ferrari, Kipnis, Liggett and many others.

I alluded to an auxillary result : large time limit
of Ψ_t^β .

Theorem. For every $\beta \in \mathbb{R}$, Ψ_t^β converges as $t \rightarrow \infty$ in
the sense of finite dimensional laws to

$$\Psi^\beta(x) = h^\beta(1, x/2\beta) - \beta^2.$$

$h^\beta(1, x)$ is a random function. It is
given by KPZ theory or rather the KPZ fixed point.

It is a manifestation of the integrability
of TASEP, and we have ...

$$\Pr(h^\beta(1, x) \leq a) = \det(I - e^{-x\delta^2} K_\beta e^{x\delta^2})_{L^2(a, \infty)}$$

for some trace class operator $e^{-x\delta^2} K_\beta e^{x\delta^2}$.

More generally, given $x_1 < x_2 < \dots < x_K$ and a_1, a_2, \dots, a_K ,

$$\Pr(h^\beta(1, x_i) \leq a_i \text{ for } i=1, \dots, K) =$$

$$\det(I - e^{-x_K \delta^2} K_\beta e^{x_K \delta^2} (I - e^{(x_K - x_1)\delta^2} e^{(x_2 - x_1)\delta^2} \dots e^{(x_K - x_{K-1})\delta^2}))_{L^2(\mathbb{R})}$$

with $X_a = \text{proj onto } L^2(-\infty, a)$.

The key object is K_β .

The operator K_β .

Introduce operators $S_x = e^{\frac{1}{3}d^3 + xd^2}$ $\in L^2(a, \infty)$ for $x \in \mathbb{R}$.

Kernel: $S_x(u, v) = e^{\frac{2}{3}x^3 + x(v-u)} A_i(v-u+x^2)$.

$B(t)$ is Brownian motion with diffusivity 2.

$\gamma(u) = \inf \{t_{\geq 0} : B(t) \leq 2\beta t - u\}$ with $u \in \mathbb{R}$.

Define operator $\mathcal{S}^{\text{hypo}}$ by its Kernel
 $\mathcal{S}^{\text{hypo}}(u, v) = \mathbb{E} S_{-\gamma(u)}(B(\gamma(u)), v) \cdot \mathbf{1}_{\{\gamma(u) < \infty\}}$.

$K_\beta = S_0^* \mathcal{S}^{\text{hypo}} + \mathcal{S}^{\text{hypo}*} S_0 - \mathcal{S}^{\text{hypo}*} \mathcal{S}^{\text{hypo}}$.

- $\mathfrak{f}^{\text{hyp}^0}$ is introduced in generality in [Quastel-Remenik, 16].
 - It is a fundamental object that describes statistics of the KPZ fixed point introduced in [Matetski- Quastel- Remenik, 17]
- The latter describes the law of growth interfaces in the KPZ universality class in a continuum limit.
- $\mathfrak{f}^{\text{hyp}^0}$ is obtained as a (singular) limit
- $$\lim_{L \rightarrow \infty} P_L^{\text{hit}} \cdot e^{-L\partial^2} \cdot K_A; \quad \xrightarrow{\text{Airy Kernel}}$$

$$P_L^{\text{hit}}(u,v) = \Pr \left(\begin{array}{l} \text{A Brownian bridge from } BB(0)=u \text{ to } BB(l)=v \\ \text{hits the hypograph of the curve } 2\beta t \end{array} \right).$$

A few remarks on $h^\beta(1, x)$.

- A one-parameter family of random functions that recover many of the Airy process. E.g.,

$\beta = -\infty$: The process $\text{Airy}_2(x) - x^2$.

$\beta = 0$: The process $\text{Airy}_1(x)$.

As $\beta \rightarrow -\infty$: $h^\beta(1, x+\beta) - \beta^2 + 2\beta(x+\beta) \rightarrow \text{Airy}_{2 \rightarrow 1}(x)$.

- The soft-shock scaling is similar to the Baik-Ben Arous-Péché Scaling from random matrices within the context of particle systems. The functions $h^\beta(1, x)$ are the corresponding distributional limits.

An idea of the proof.

Consider the probability that $\psi^\beta(0) \leq a$, which is

$$\Pr(\psi^\beta(1,0) \leq \beta^2 + a) = \det(I - K\beta)^{L^2[a+\beta^2, \infty)}$$

$$= \det(I - e^{\beta d} K \beta e^{-\beta d})^{L^2[a, \infty)}$$

(\hookrightarrow translation)

There are two key steps:

① Calculate $\mathcal{S}^{\text{hypo}}$.

② Factorize $I - e^{\beta d} K \beta e^{-\beta d} = (I - A)(I - B)$.

Lemma 1. Define operators M_β and R by

$$M_\beta f(x) = e^{\beta x} f(x) \quad \text{and} \quad Rf(x) = f(-x).$$

$$g^{\text{hyp}} = x_0 S_0 + (1-x_0) M_\beta R M_\beta^{-1} S_0.$$

Lemma 2. $I - e^{\beta^2 \partial} K_\beta e^{-\beta^2 \partial} = (I-Q)^* (I-Q),$

$$Q = M_\beta (A + E_\beta) M_\beta^{-1}, \quad A = S_0^* R S_0, \quad E_\beta = \text{error}.$$

$$A(u,v) = 2^{-\frac{1}{3}} A_i(2^{-\frac{1}{3}}(u+v)) \quad \text{and} \quad \det(I-A)_{L^2_{[0,\infty)}} = \Pr(X_{TW_i} \leq 2^{\frac{2}{3}}a).$$

Decompose $(I-Q)^*(I-Q) = \underbrace{(I-Q)^*X_a(I-Q)}_X + \underbrace{(I-Q)^*(I-X_a)(I-Q)}_Y.$

Since $I-X_a = \text{proj. onto } L^2[a, \infty),$

$$\det(Y)_{L^2[a, \infty)} = \det(I-Q)_{L^2[a, \infty)}^2.$$

The trace norms of X and E_β on $L^2[a, \infty)$ vanish as $\beta \rightarrow \infty.$

$$\begin{aligned} \text{So } \Pr(\Psi^\beta(0) \leq a) &= \det((I-Q)^*(I-Q))_{L^2[a, \infty)} \\ &\approx \det(Y)_{L^2[a, \infty)} \\ &\approx \det(I-A)_{L^2[a, \infty)}^2 \end{aligned}$$

□

How does one calculate $\mathcal{S}^{\text{hypo}}$?

- Consider $\beta=0$. For general β , use Cameron-Martin Theorem to tilt space so that the line $2\beta t$ becomes flat. This results in the conjugation by M_β in Lemma 1.

- Only need to consider u_{γ_0} in $\mathcal{S}^{\text{hypo}}(u, v)$.
 $\tilde{\gamma}$ = hitting time of B to origin started from u .

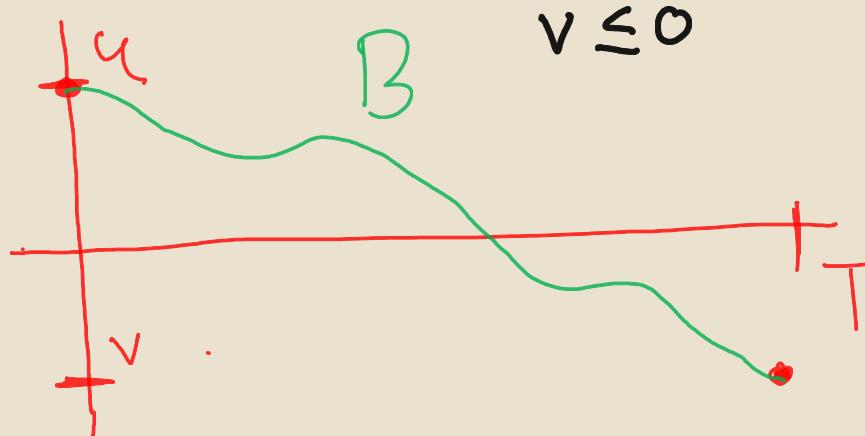
$$\mathcal{S}^{\text{hypo}}(u, v) = \mathbb{E}\left[e^{\frac{1}{3}\delta^3 - \gamma\delta^2}_{(0, v)}\right] = \lim_{T \rightarrow \infty} \mathbb{E}\left[e^{\frac{1}{3}\delta^3 - \gamma\delta^2}_{(0, v)} \cdot 1_{\{\gamma \leq T\}}\right].$$

- All the randomness enters from the left, so
- $$\mathbb{E} \left[e^{\frac{1}{3}\partial^3 - \gamma \partial^2} (0, v) \mathbf{1}_{\{\gamma \leq T\}} \right] = \mathbb{E} \left[e^{(\frac{1}{3}\partial^3 - \gamma \partial^2)}_{(0, \cdot)} \mathbf{1}_{\{\gamma \leq T\}} \right] \cdot e^{\frac{1}{3}\partial^3 - \gamma \partial^2}_{(\cdot, v)}.$$
- multiplication*

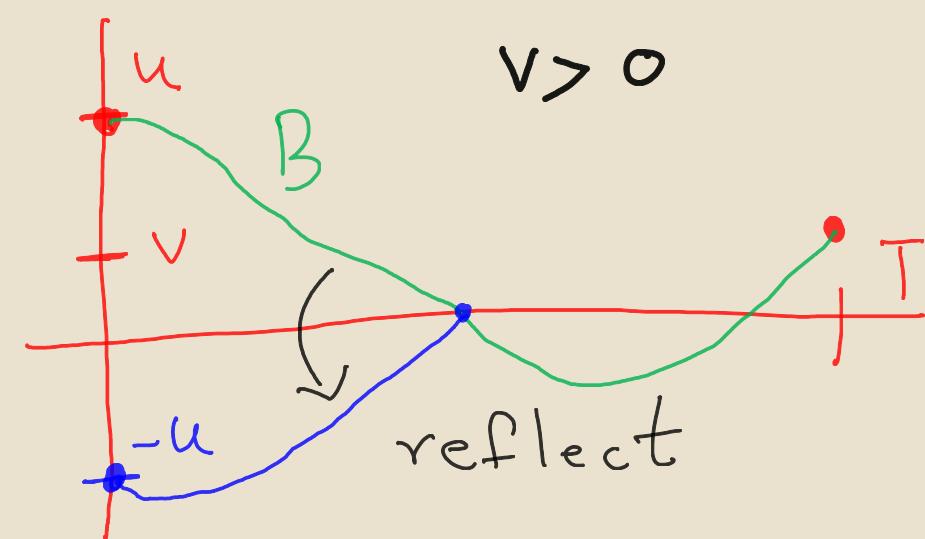
- By strong Markov property,

$$\mathbb{E}[\dots] = \Pr \left(B \text{ goes from } B(0)=u \text{ to } B(T)=v \text{ while hitting origin} \right).$$

- There are 2 cases:



$$v \leq 0$$



$$v > 0$$

So by the reflection principle,

$$\mathbb{E}[\dots] = \begin{cases} e^{T\partial^2}(u,v) & \text{if } v \leq 0, \\ e^{T\partial^2}(-u,v) & \text{if } v > 0. \end{cases}$$

$\hookrightarrow Re^{T\partial^2}(u,v)$

$$\begin{aligned} S_{\text{hypo}}(u,v) &= \lim_{T \rightarrow \infty} e^{T\partial^2} x_0 \cdot S_T(u,v) + Re^{T\partial^2} (1-x_0) \cdot S_T(u,v) \\ &= RS_0(u,v) + \lim_{T \rightarrow \infty} (I-R) e^{T\partial^2} x_0 \cdot S_T(u,v). \end{aligned}$$

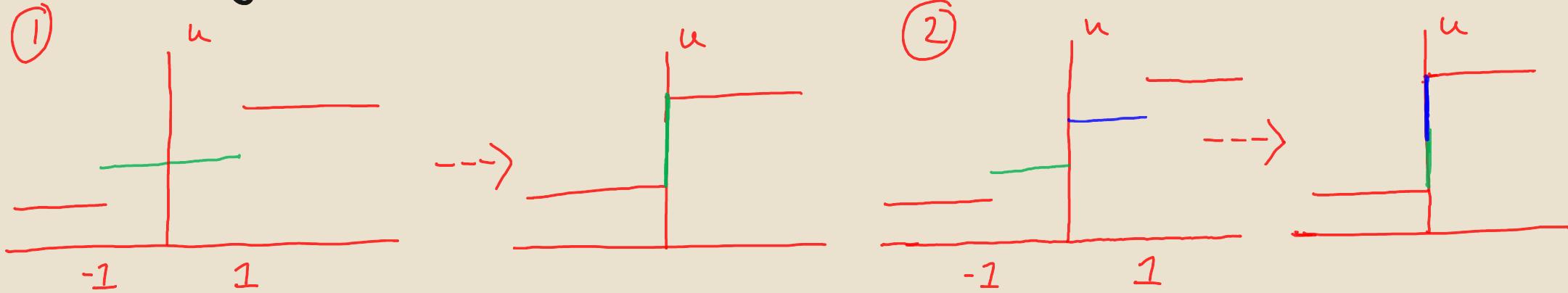
↳ $v \leq 0$ ↳ $v > 0$

The limit turns out to vanish to give

$$S_{\text{hypo}} = x_0 S_0 + (1-x_0) RS_0.$$

↳ $u \leq 0$ ↳ $u > 0$

Merger of multiple shocks



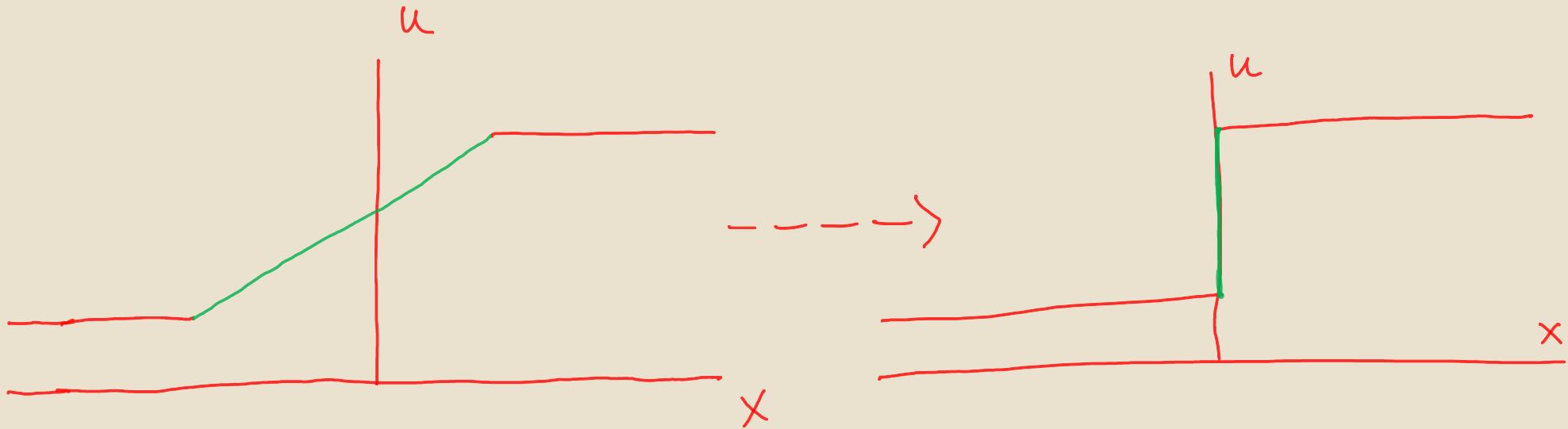
A similar approach may be used to show that at the merger of $M-1$ shocks the fluctuations are given by the maximum of M GOE T-Ws. The process level description is

$$X \mapsto \max \left\{ X_{TW_1}^{(0)} - x, X_{TW_1}^{(1)} - \left(1 - \frac{2 \cdot 1}{M-1}\right)x, X_{TW_1}^{(2)} - \left(1 - \frac{2 \cdot 2}{M-1}\right)x, \dots, X_{TW_1}^{(M-1)} + x \right\}.$$

independent ↑

Questions remain

The emerging shock:



Fluctuations at the shock?

Thank you

