

# Rewriting History in Integrable Stochastic Particle Systems

Leonid Petrov  
University of Virginia

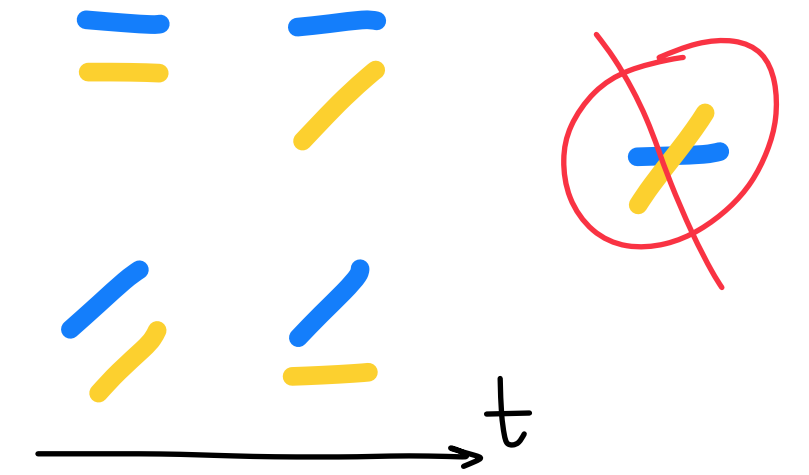
joint with Axel Saenz (Oregon State University), [arXiv:2212.01643](https://arxiv.org/abs/2212.01643)

**A tale of two cars on a  
one-lane road**

# Two cars (discrete time TASEP with Bernoulli jumps)

Time  $t \in \mathbb{Z}_{\geq 0}$ , 2 cars with speeds  $a_1 > a_2 > 0$ , probabilities of jumps  $a_i/(1 + a_i)$

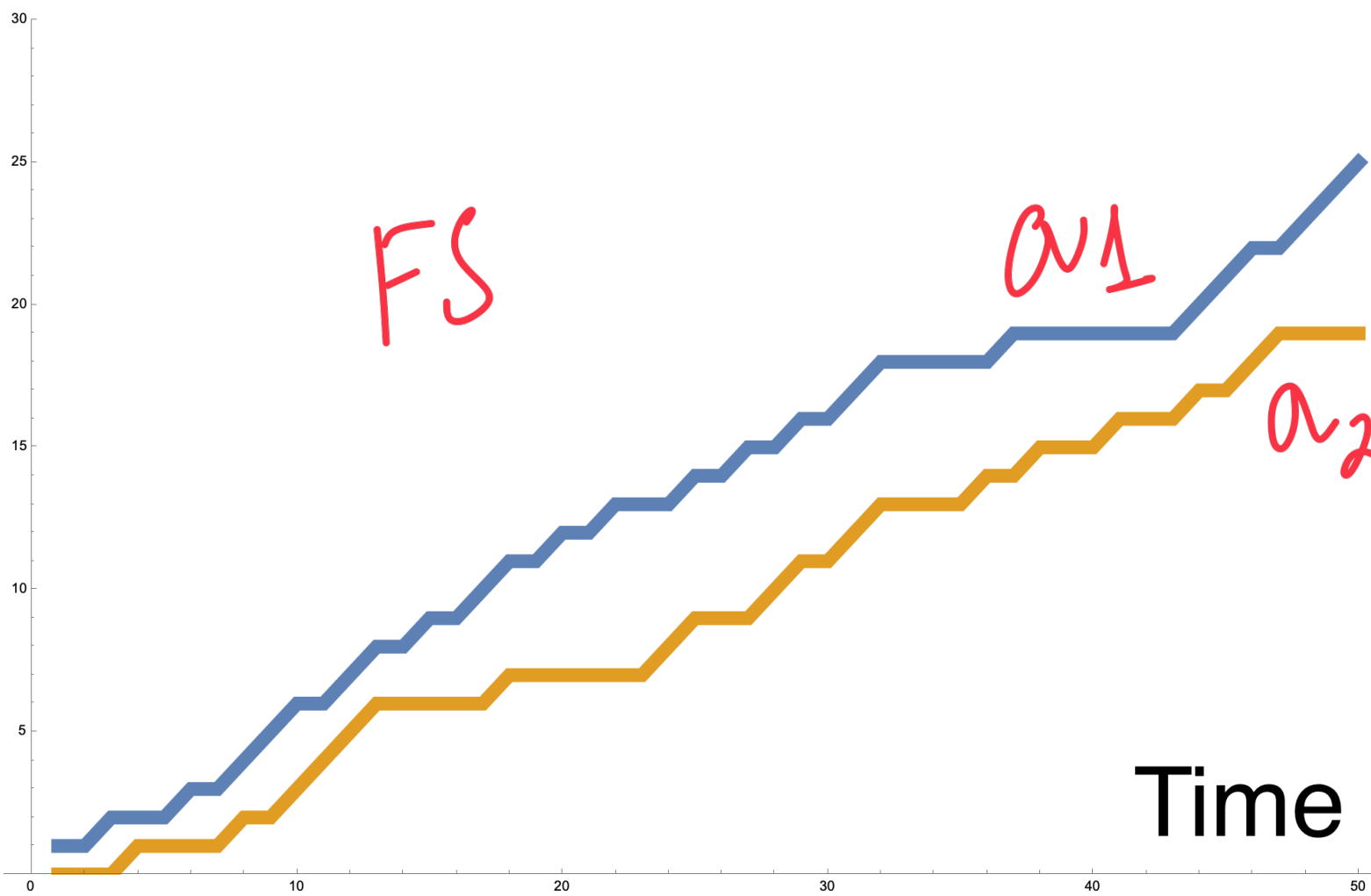
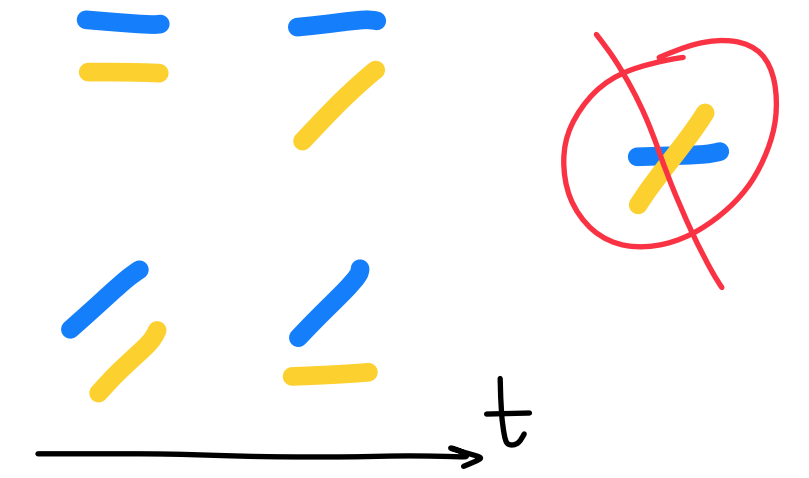
One-lane highway: no passing. Consider two systems: FS and SF



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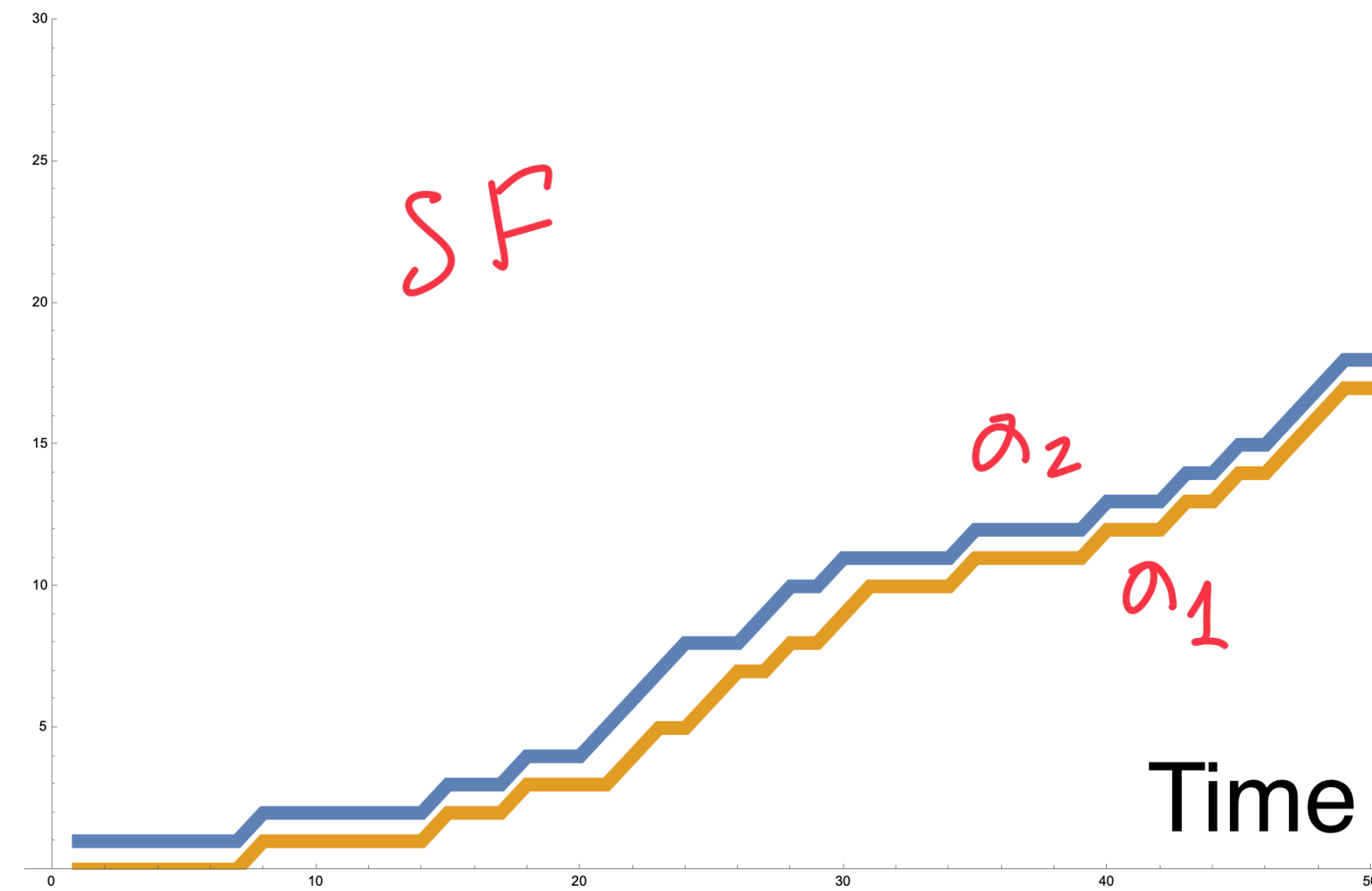
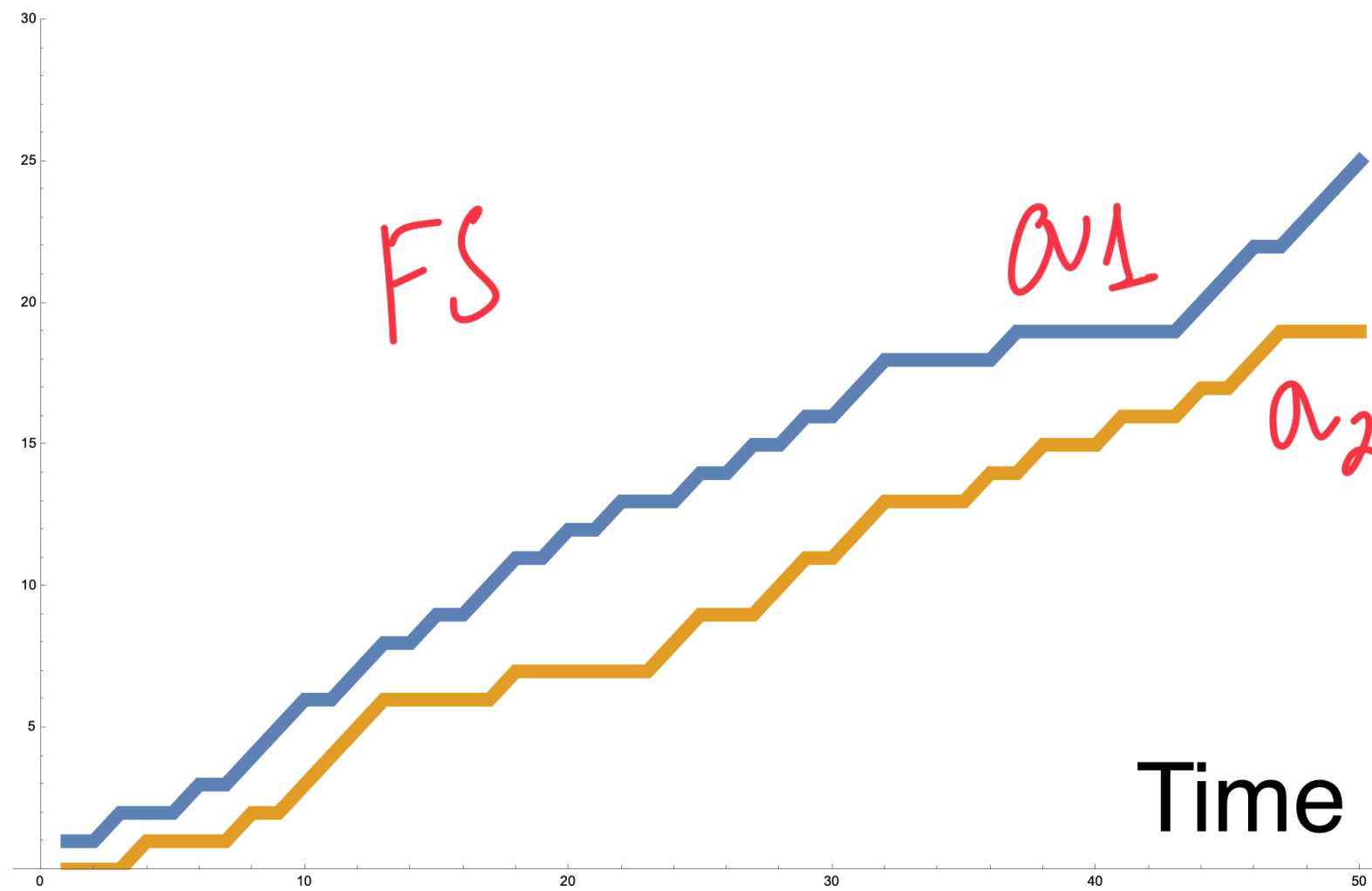
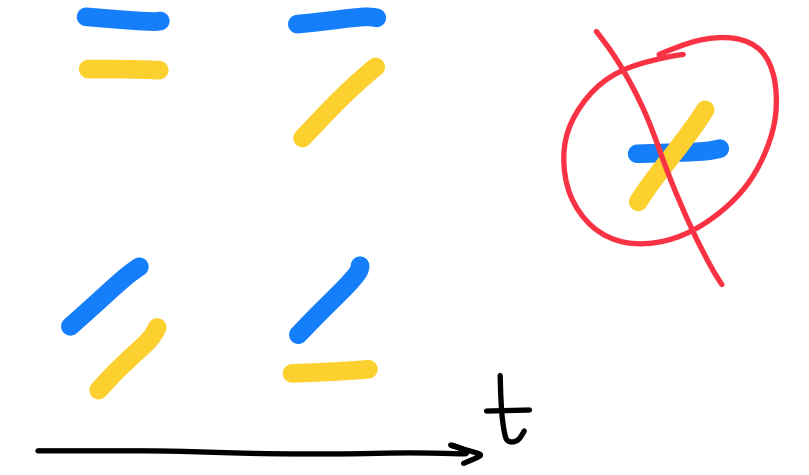
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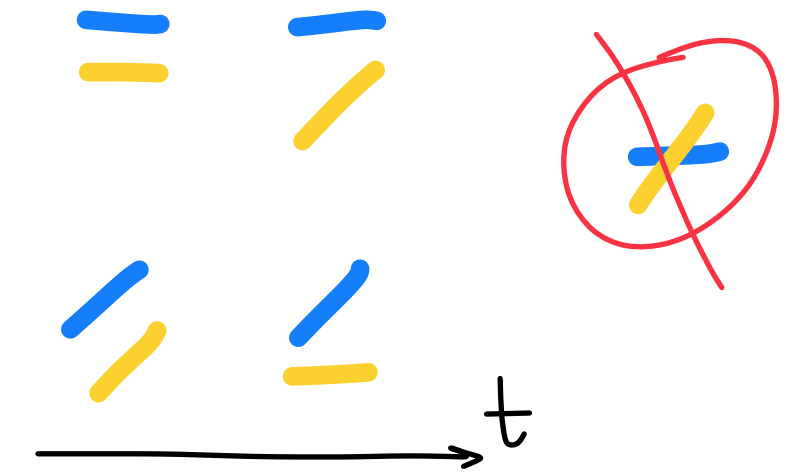
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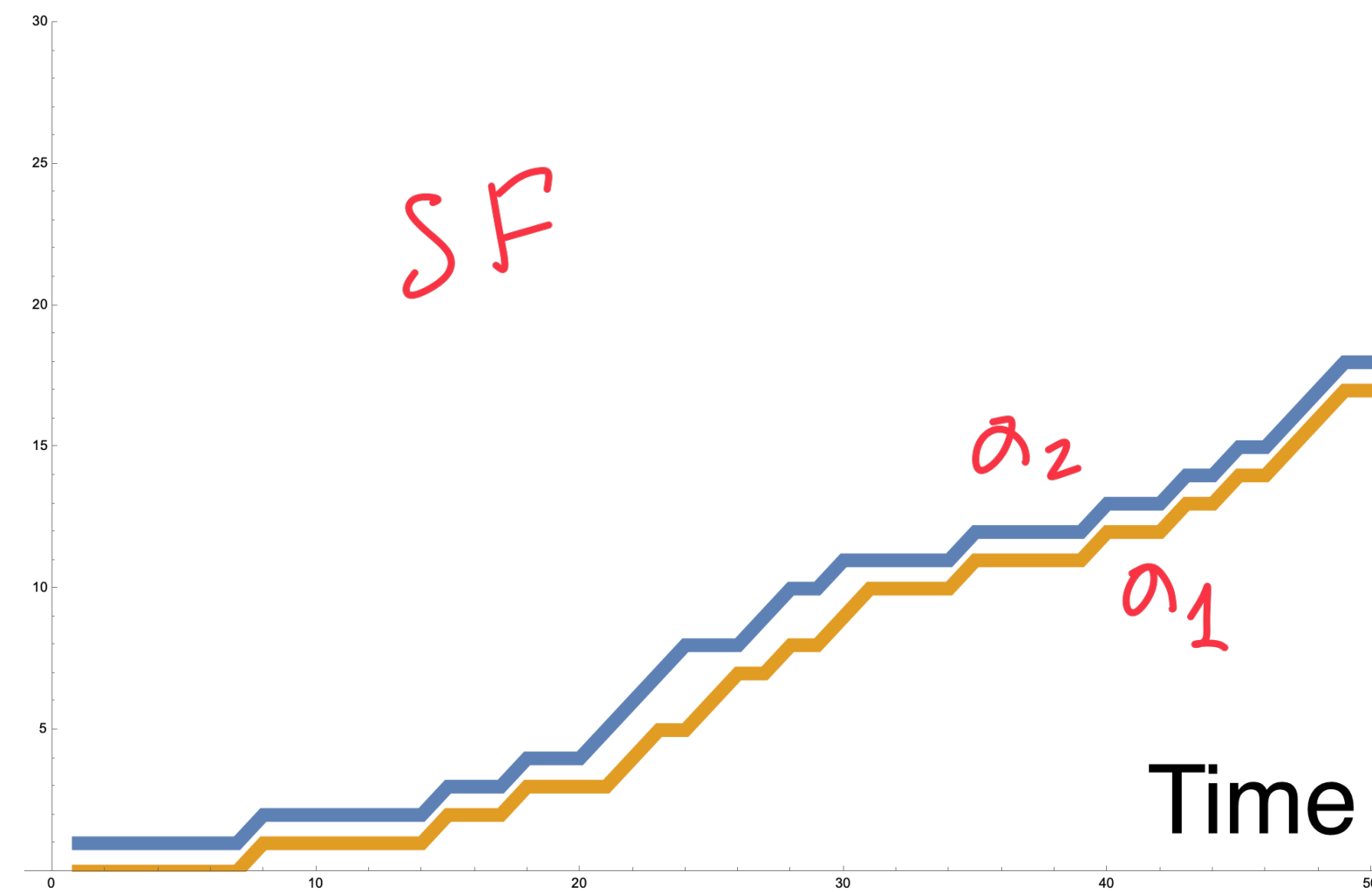
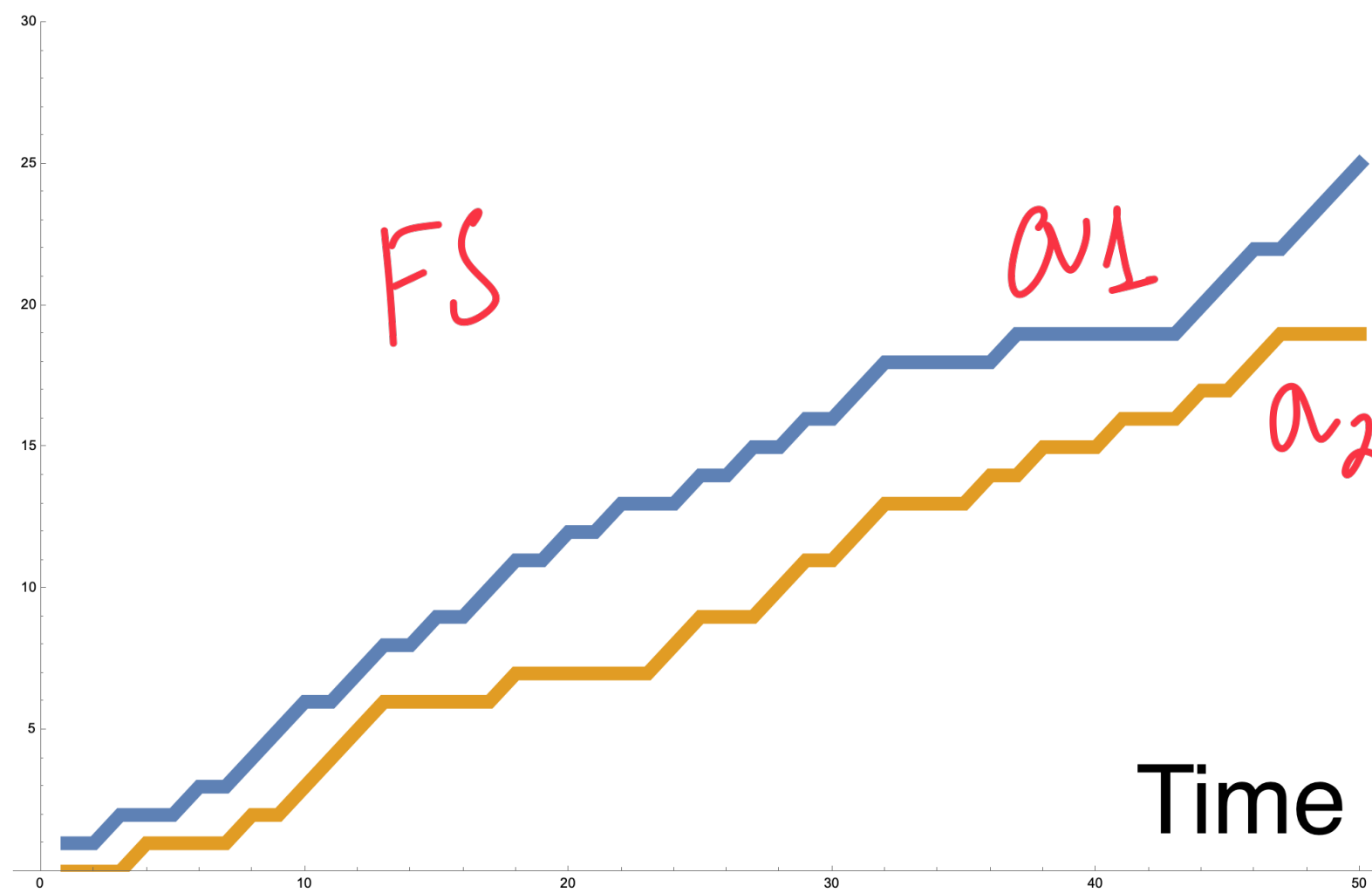
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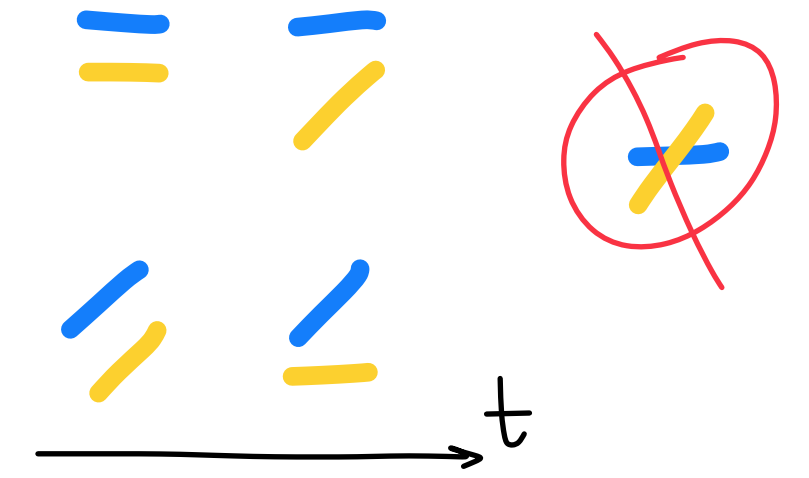
The long-time speed of the car ahead (blue) depends on which car is first; for the car behind (yellow) it **does not** depend on the order



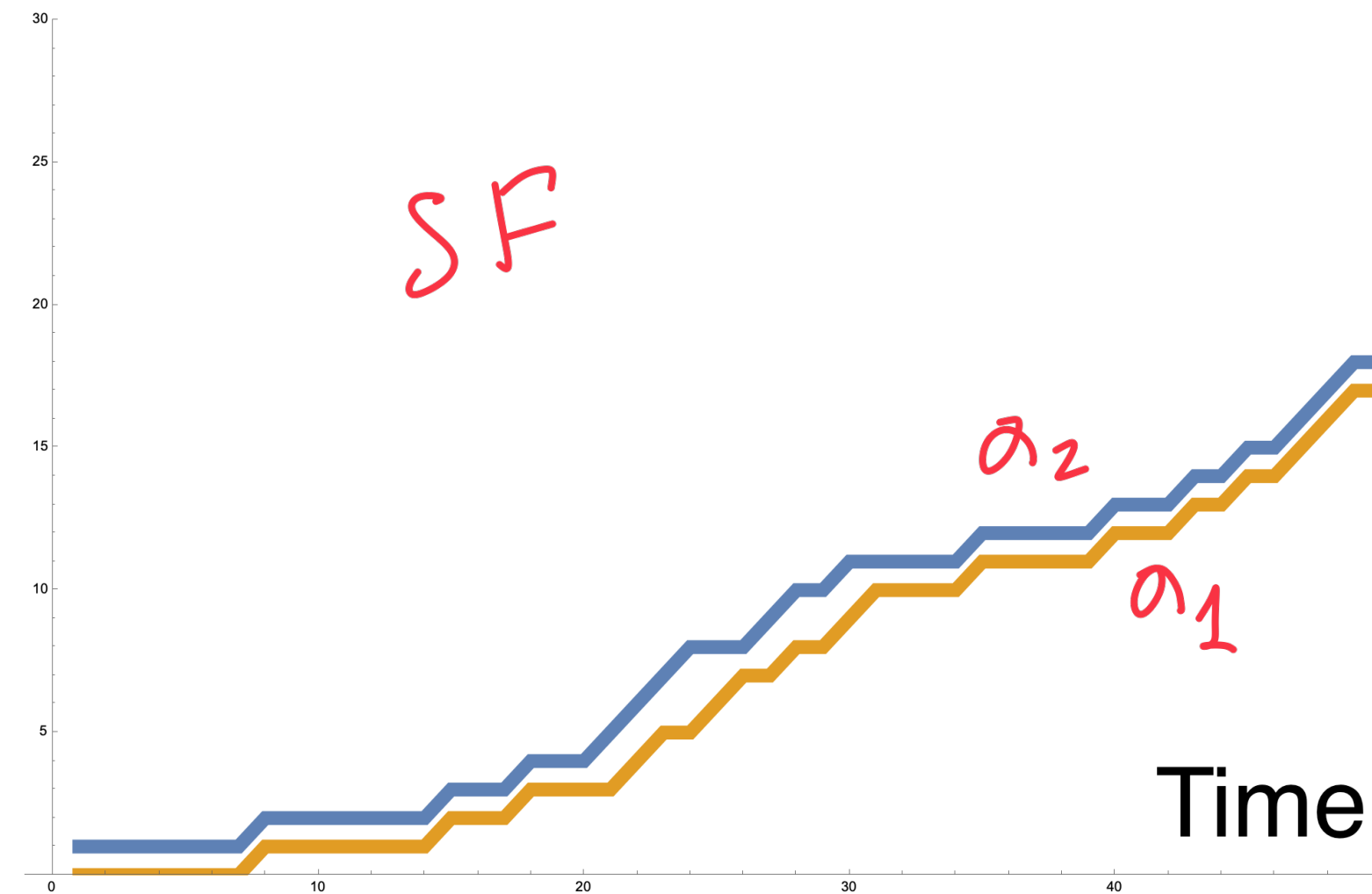
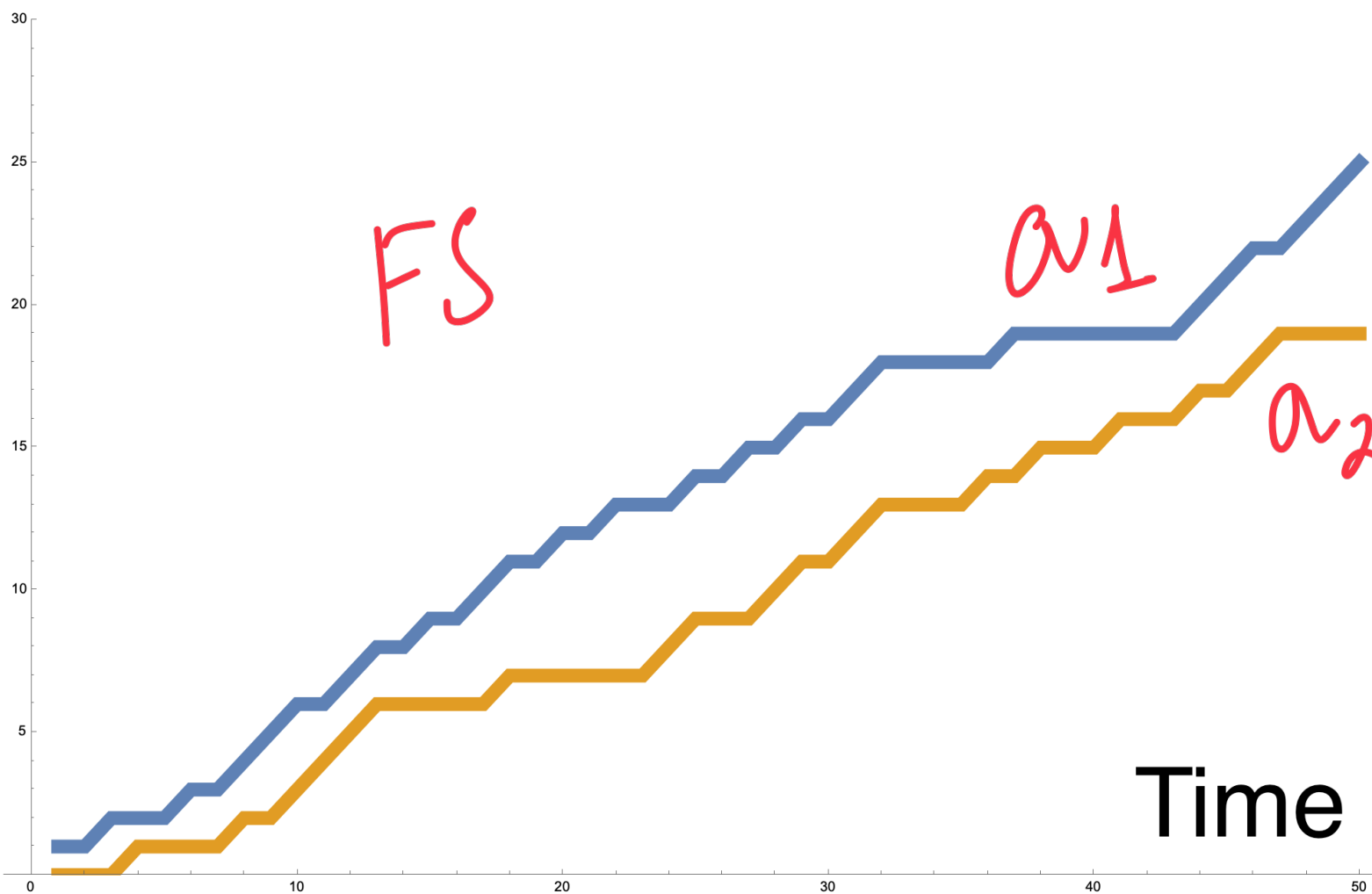
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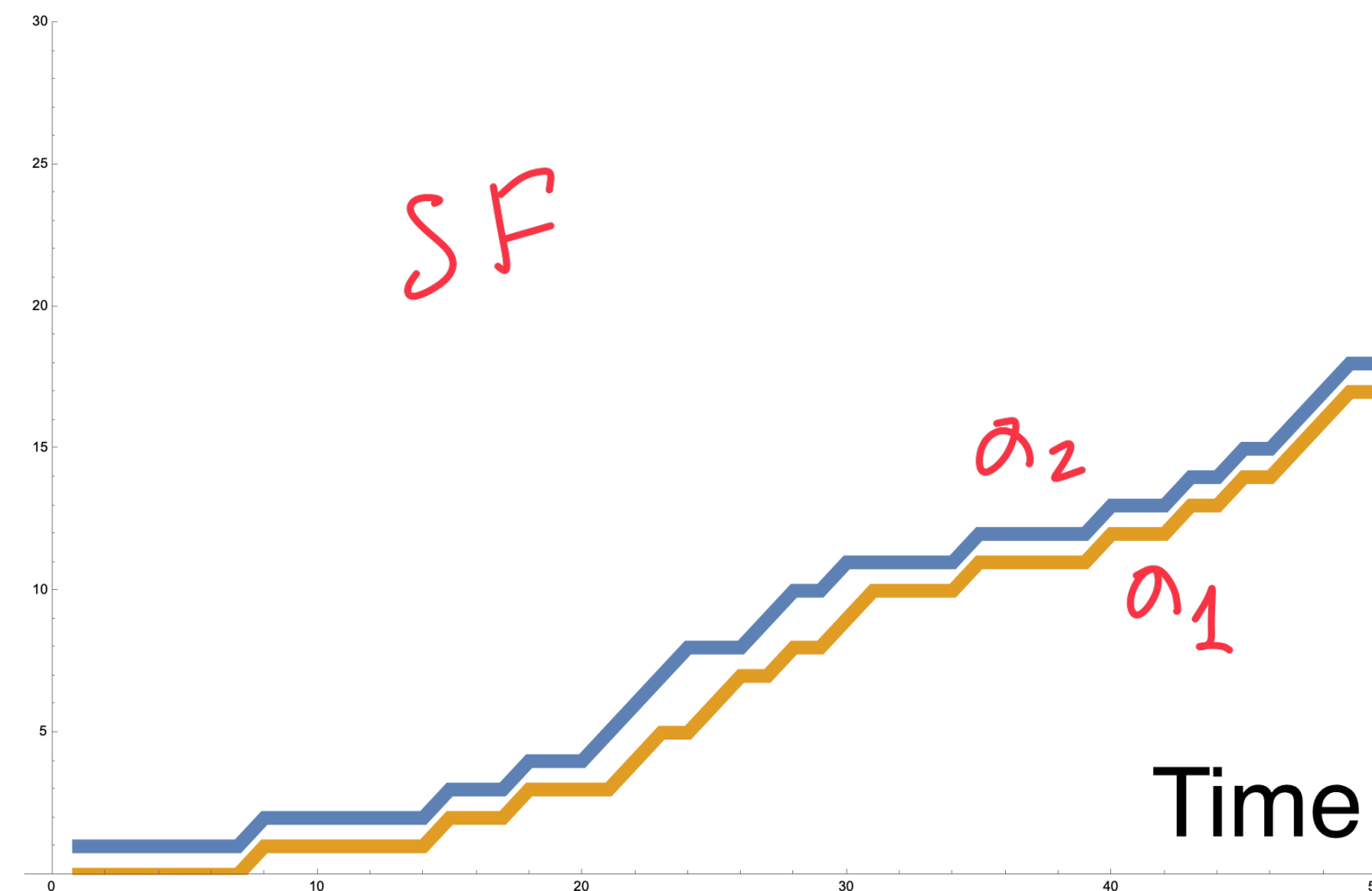
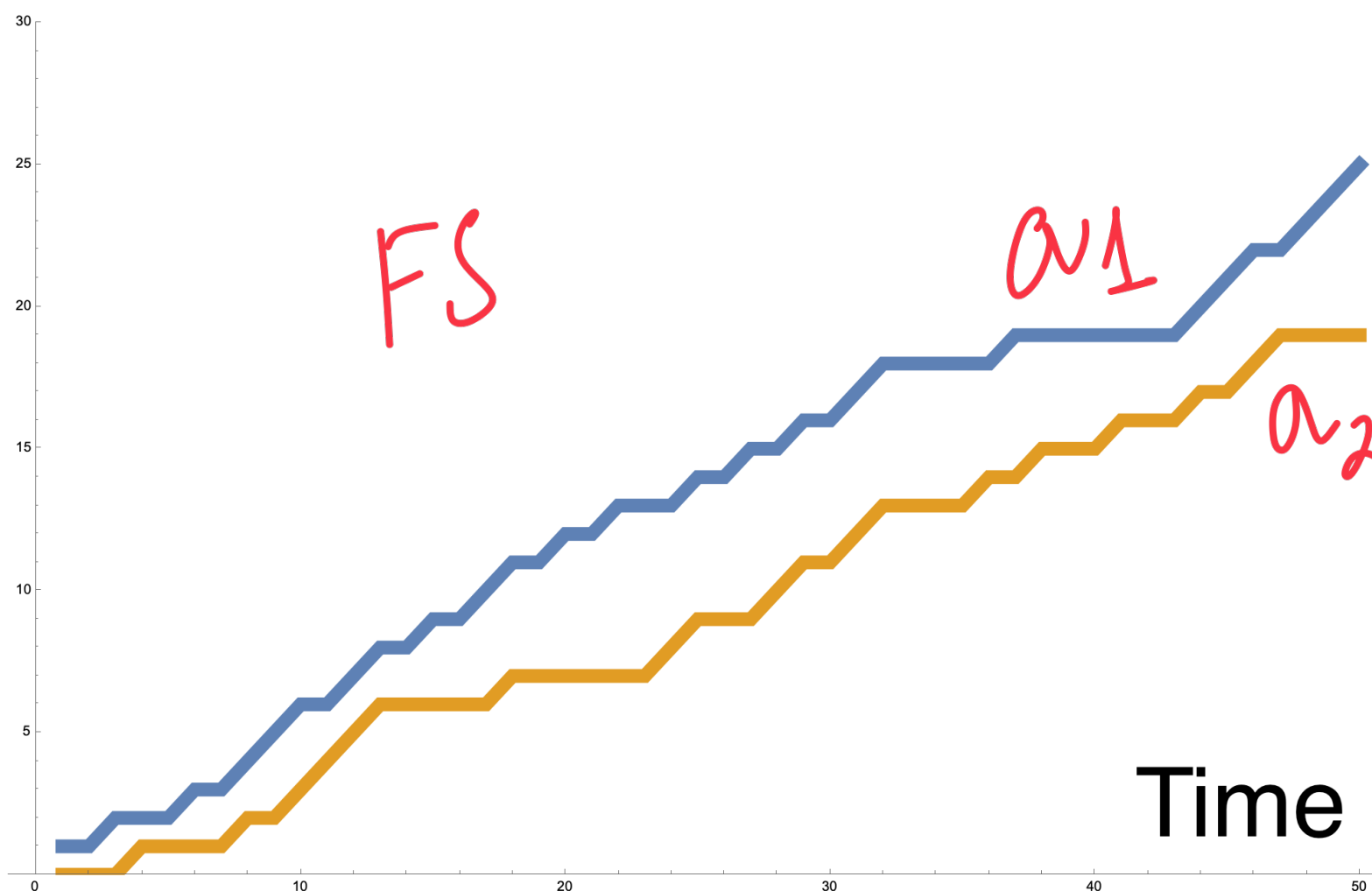
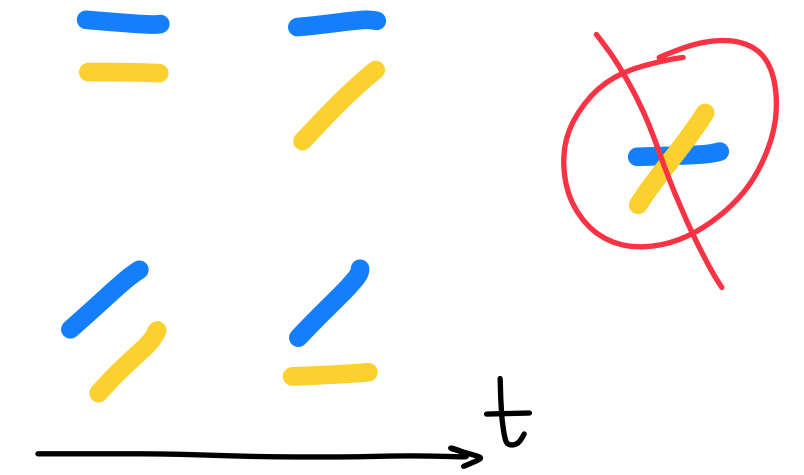
**Theorem.** (Vershik-Kerov ~1981; O'Connell 2003)

If the cars started at locations 0,1 (immediate neighbors; called *step initial configuration*), then the distribution of the trajectory of the car behind is **independent** of the order of the speeds

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← follows from Robinson-Schensted-Knuth correspondence which encodes unused jumps



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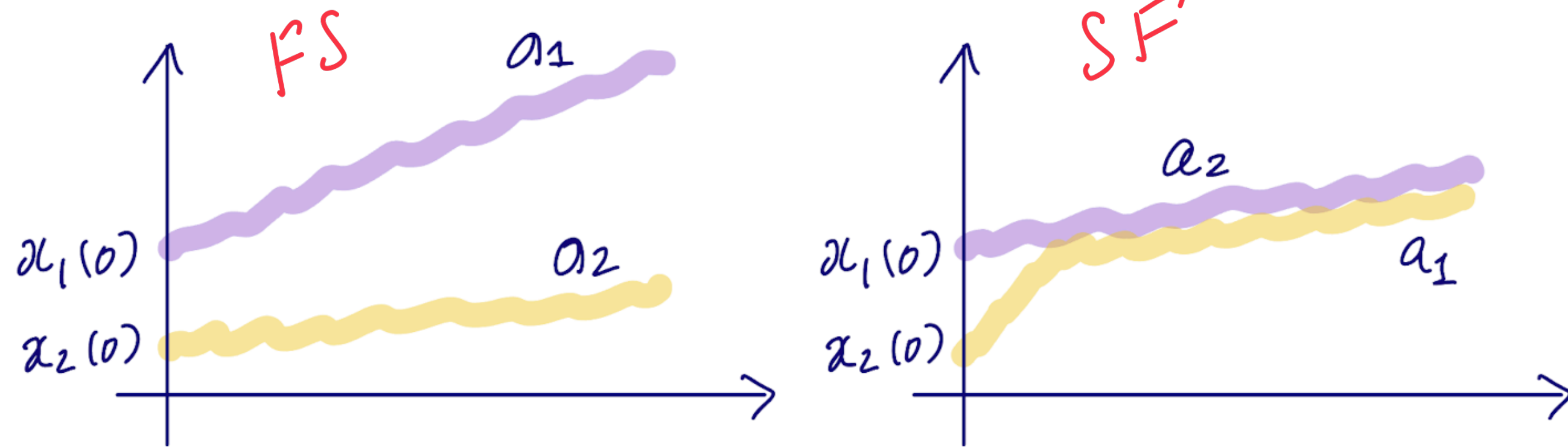
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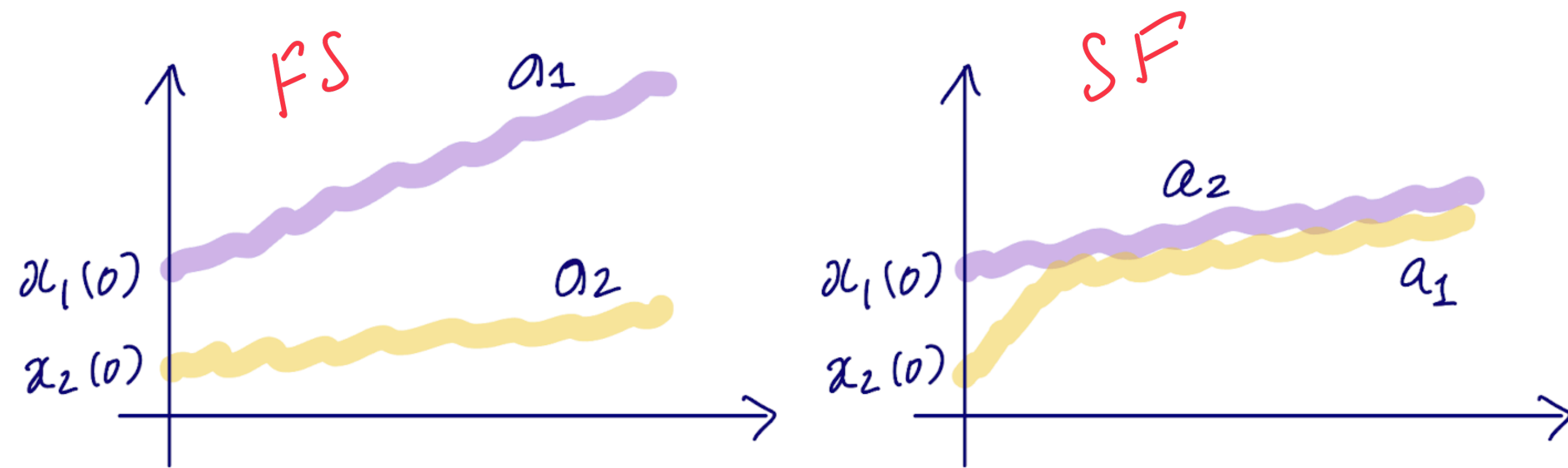
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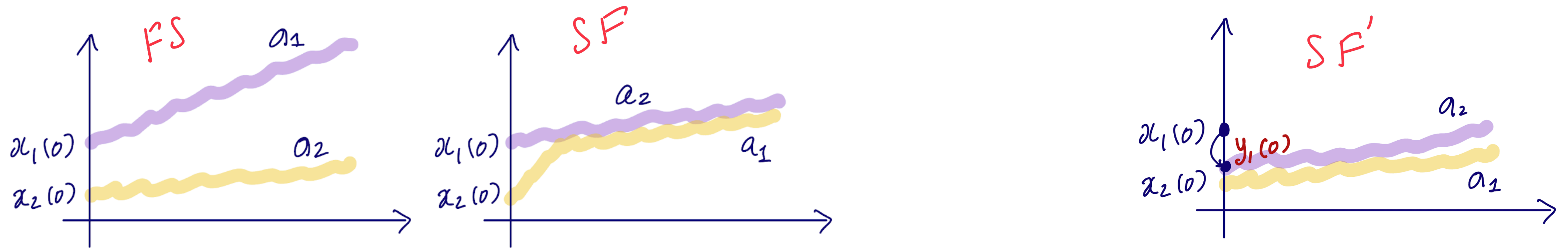
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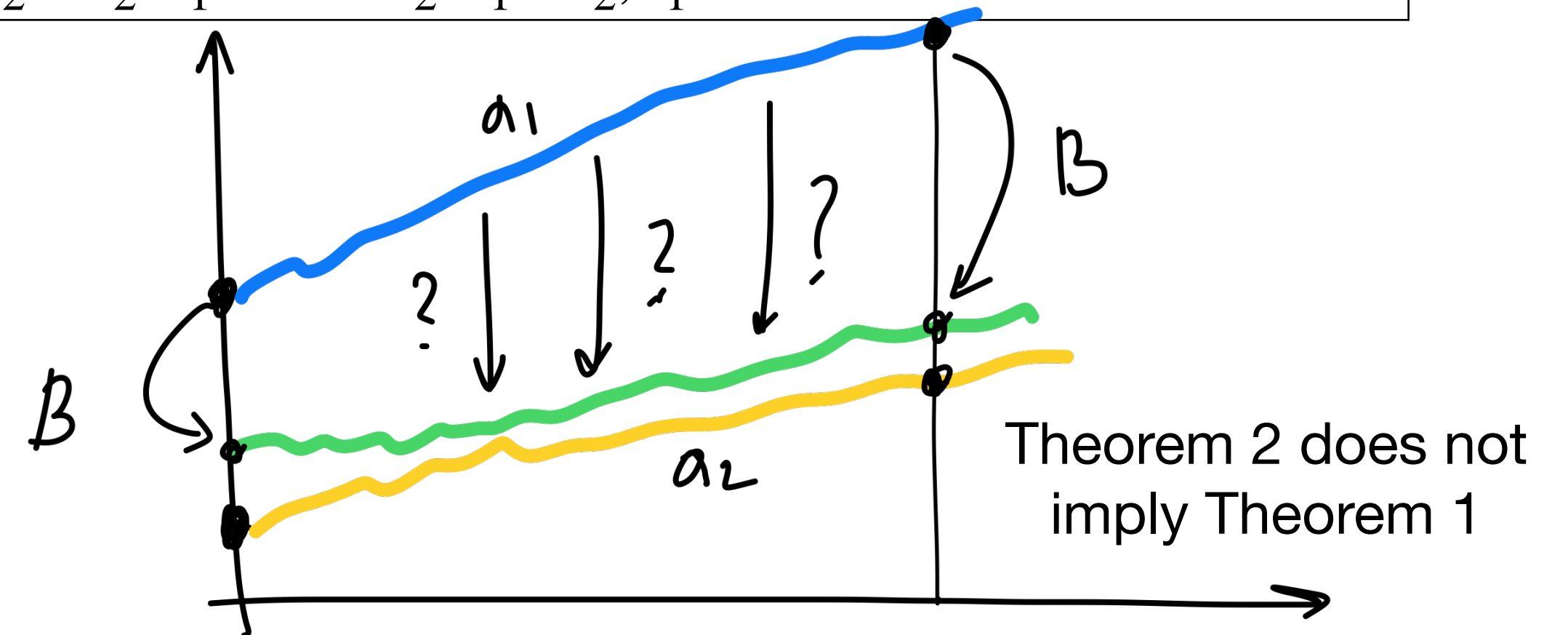
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# Proof via Yang-Baxter equation

# Jumps and Yang-Baxter equation

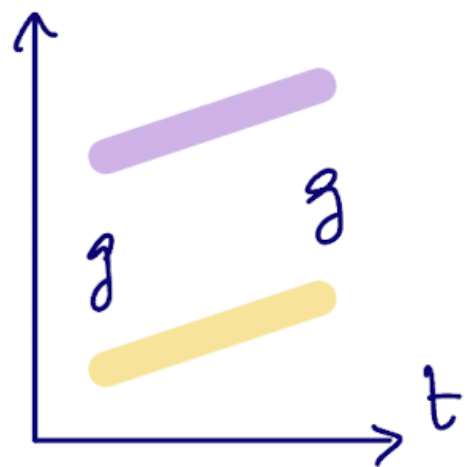
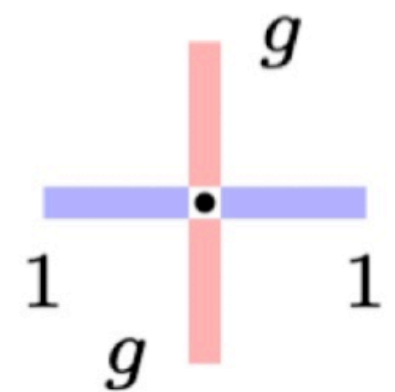
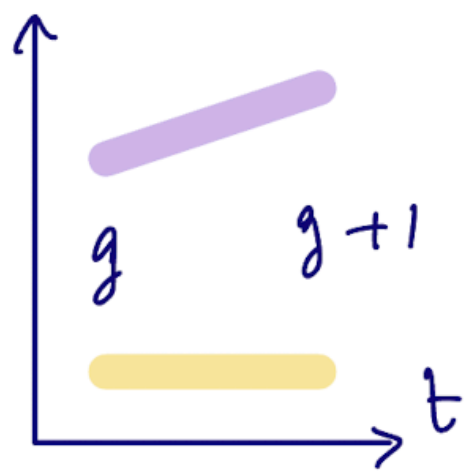
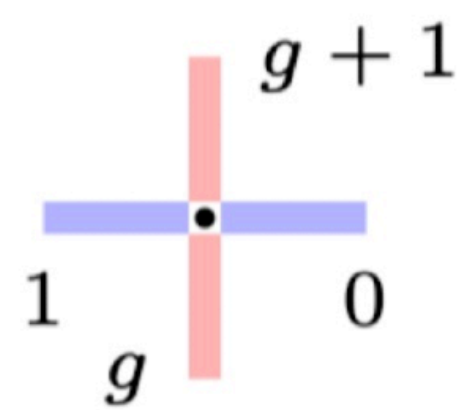
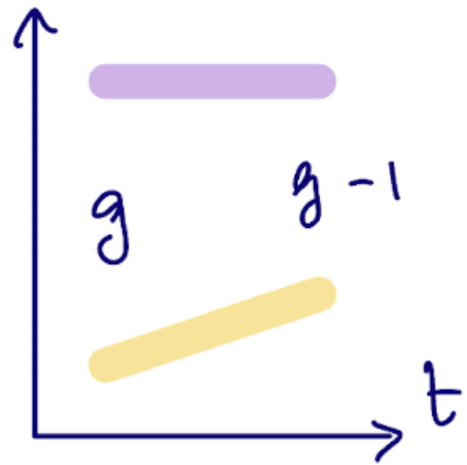
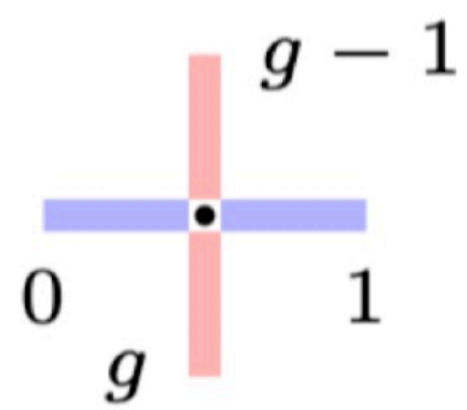
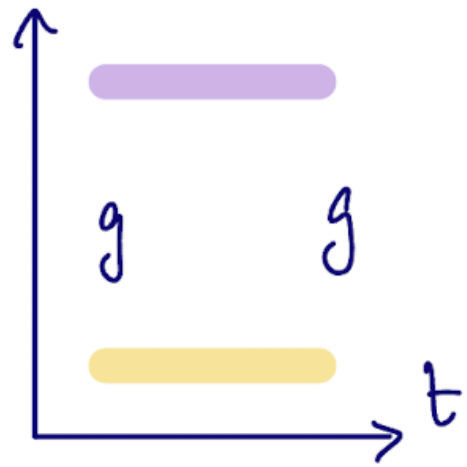
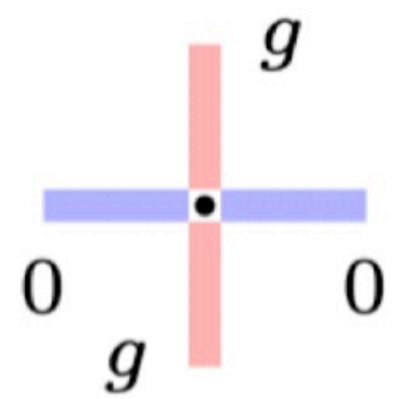
Translate into the “vertex model” language. In vertex models, time runs **up**.

$g = x_1 - x_2 - 1$  is the **gap**. Down and left are inputs, up and right are outputs.

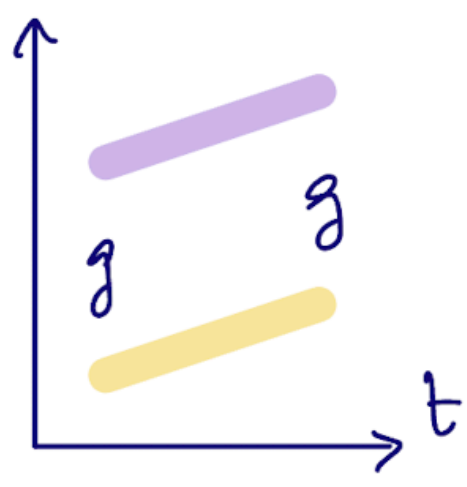
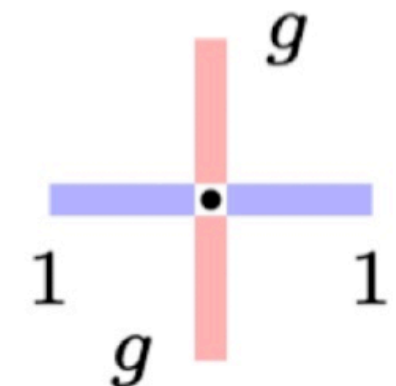
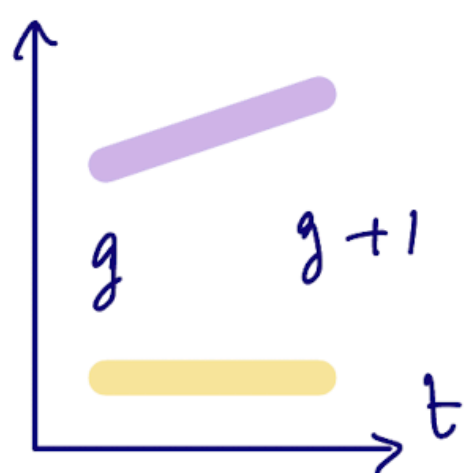
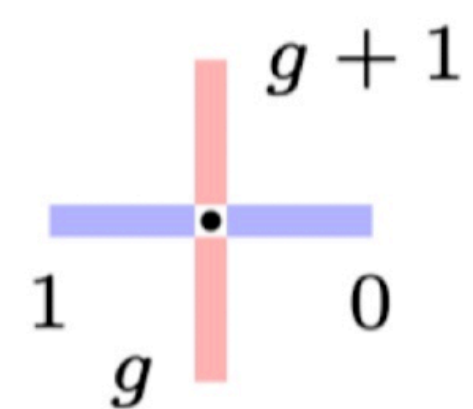
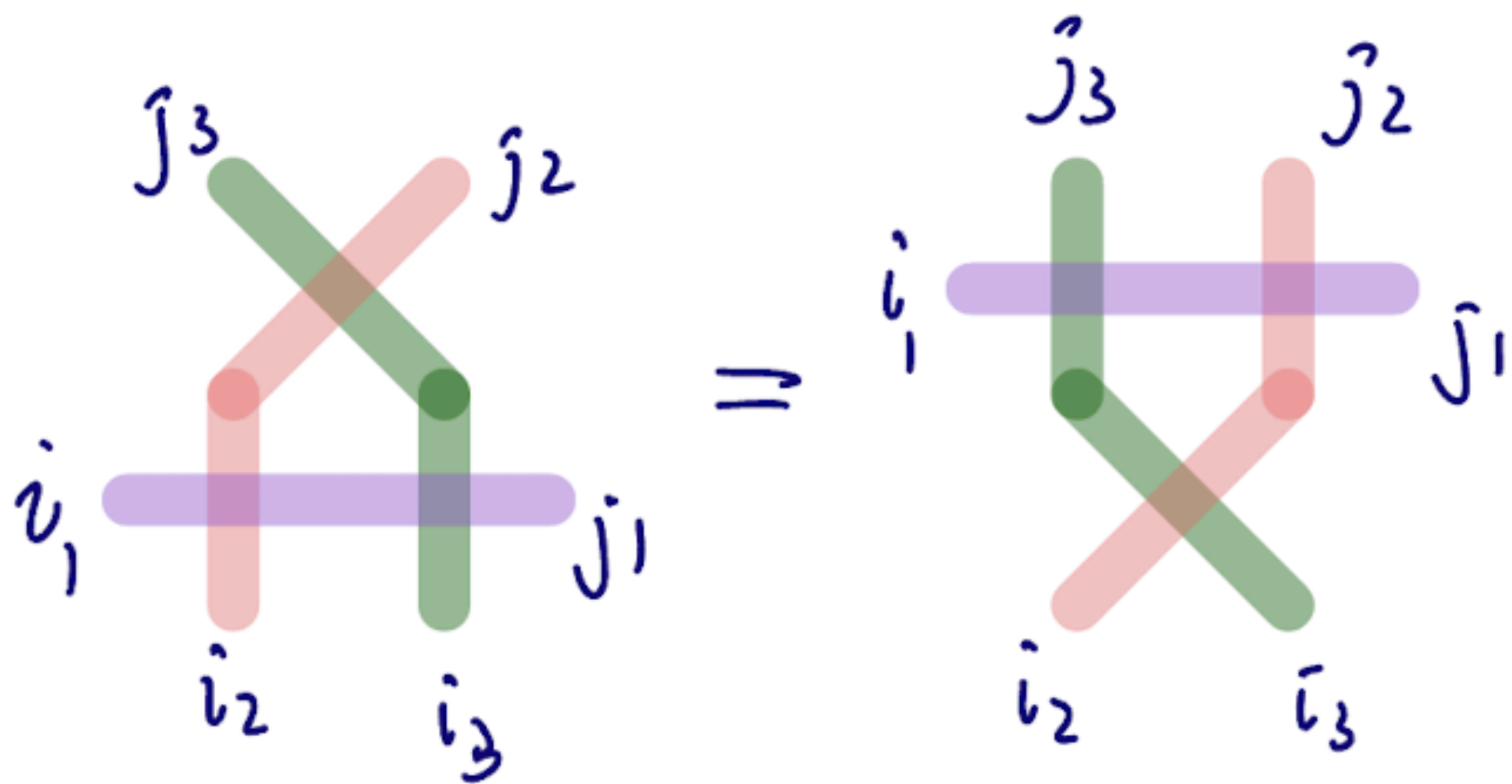
# Jumps and Yang-Baxter equation

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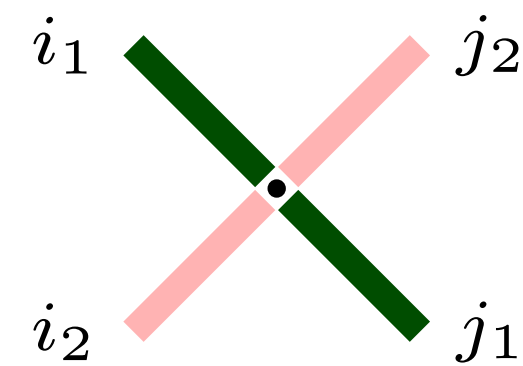
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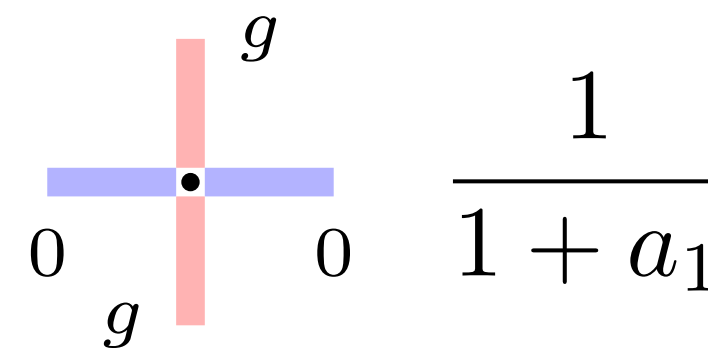
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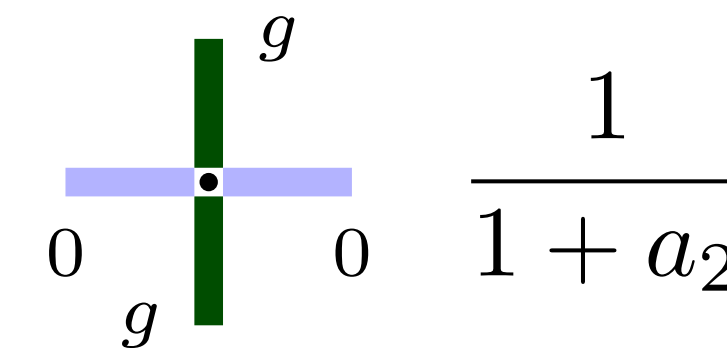
# Theorem. Vertex weights satisfy the Yang-Baxter equation



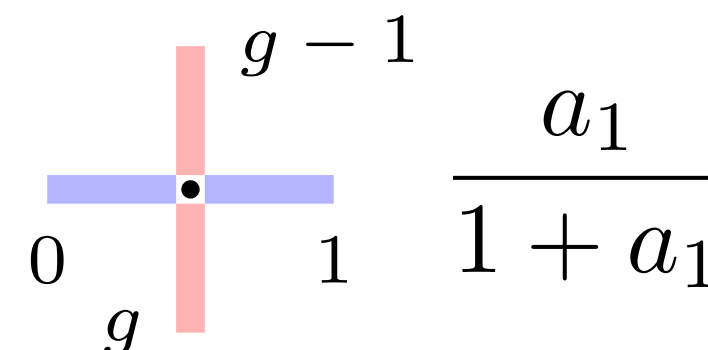
$$(a_2/a_1)^{j_2} (1 - \mathbf{1}_{j_2 < j_1} a_2/a_1) \mathbf{1}_{j_2 \leq j_1}$$



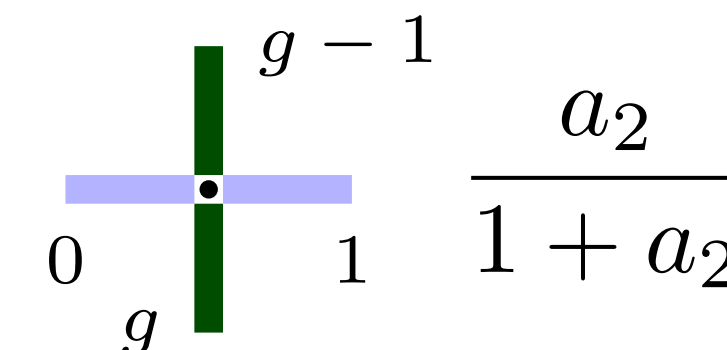
$$\frac{1}{1+a_1}$$



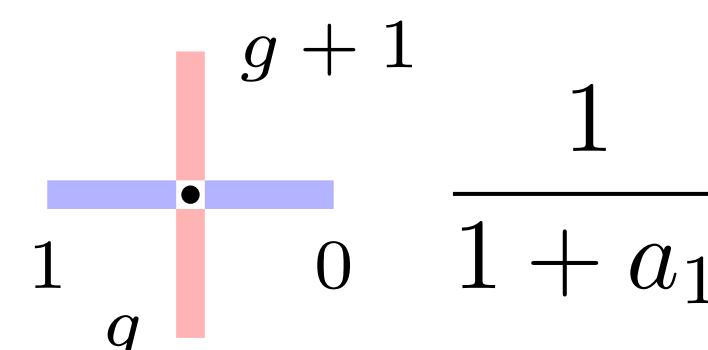
$$\frac{1}{1+a_2}$$



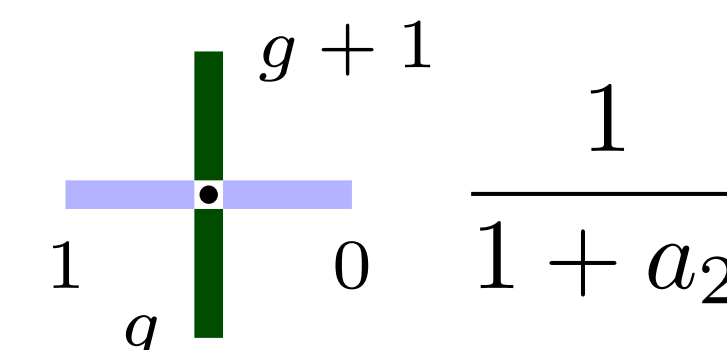
$$\frac{a_1}{1+a_1}$$



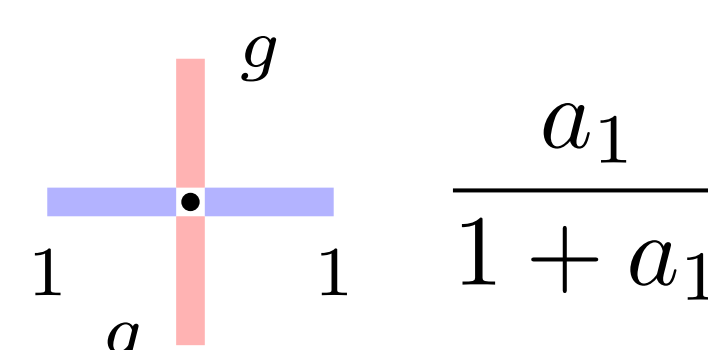
$$\frac{a_2}{1+a_2}$$



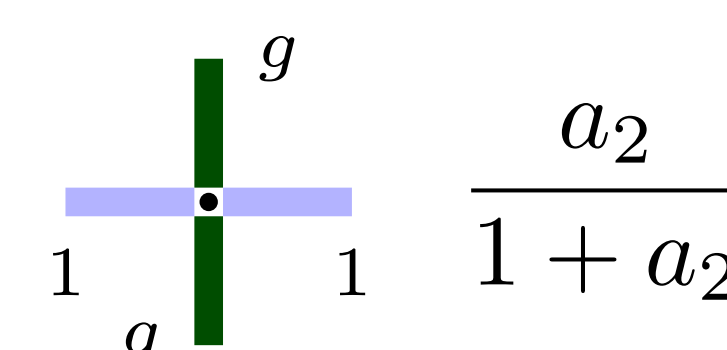
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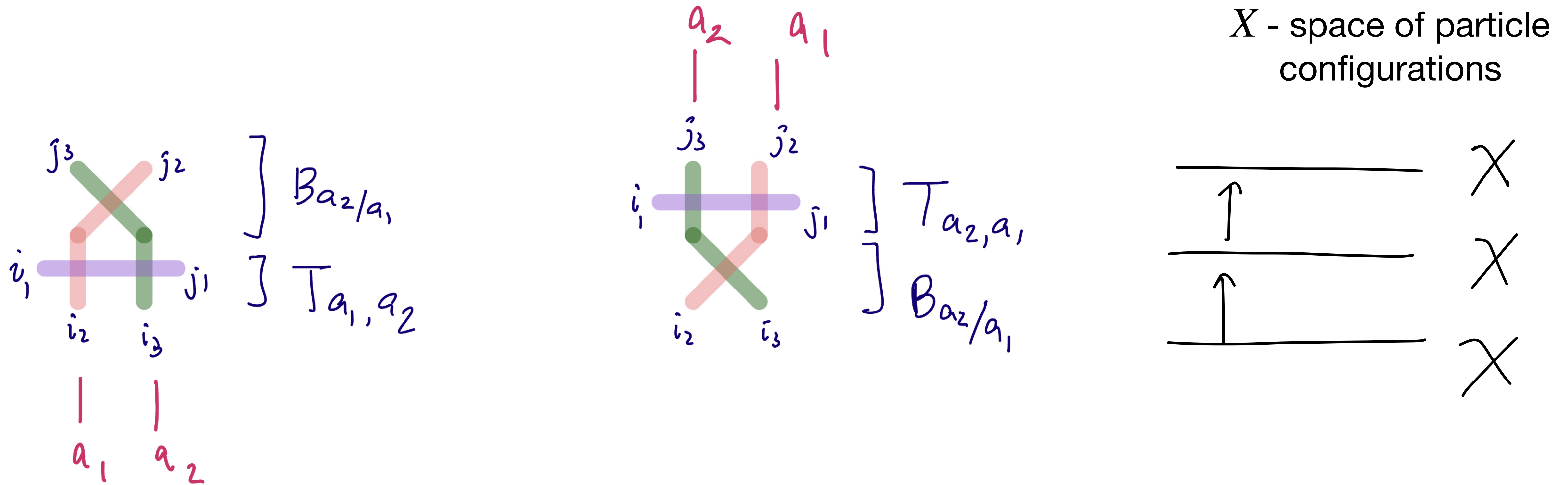


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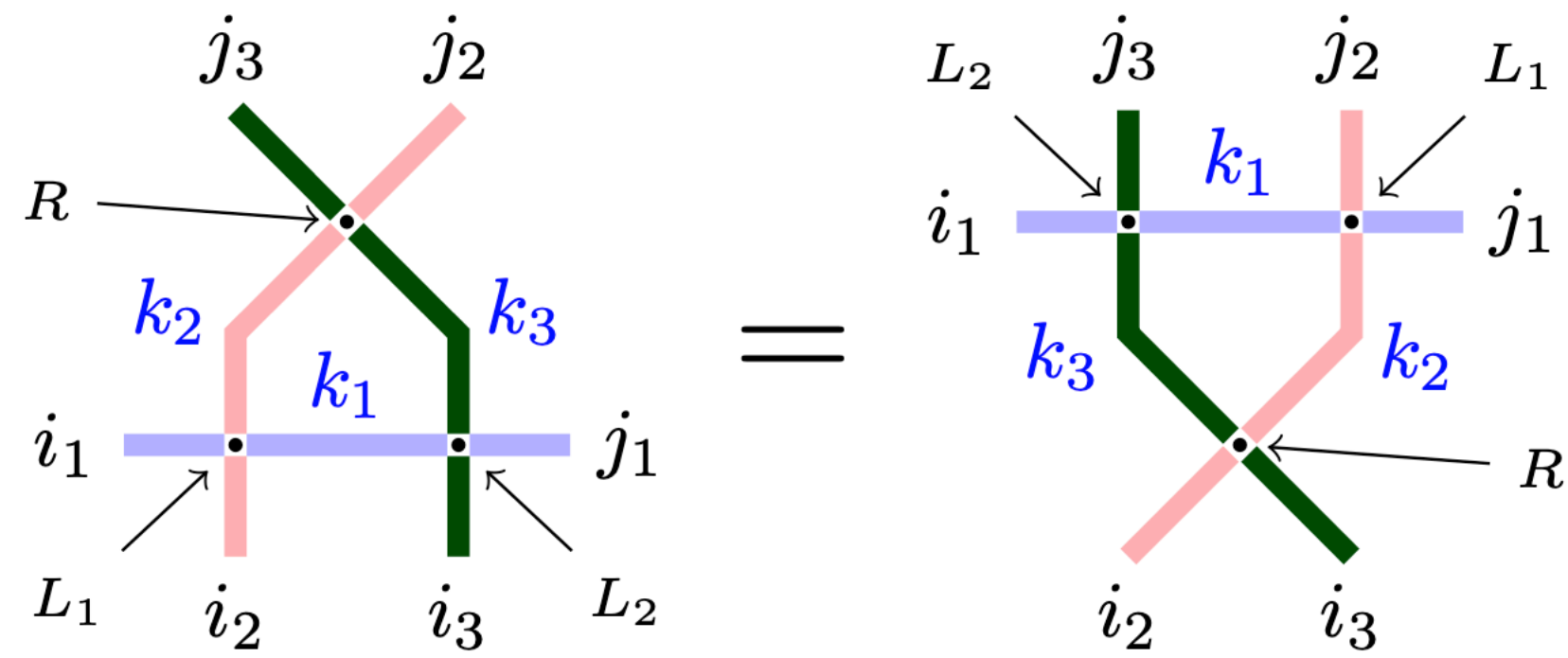
# Intertwining relation proof



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# Intertwining relation for general stochastic R matrices (for specialists)



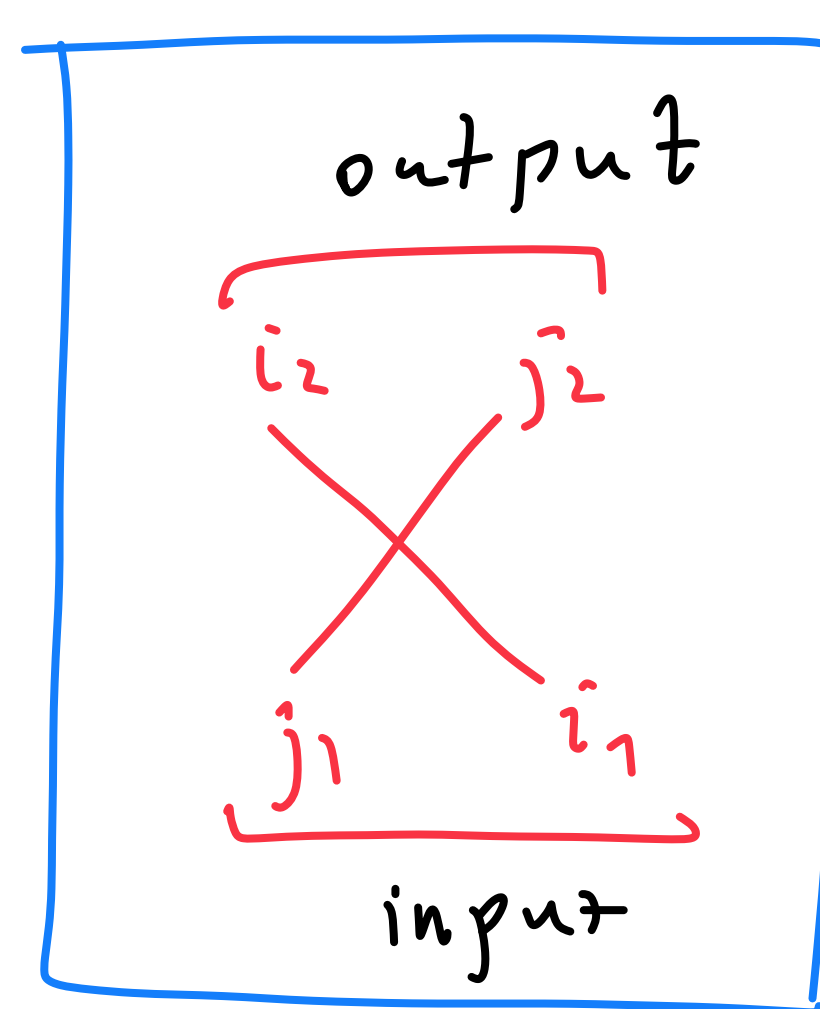
$$L_{u,s}^{(1)}(g, 0; g, 0) = \frac{1 - q^g su}{1 - su},$$

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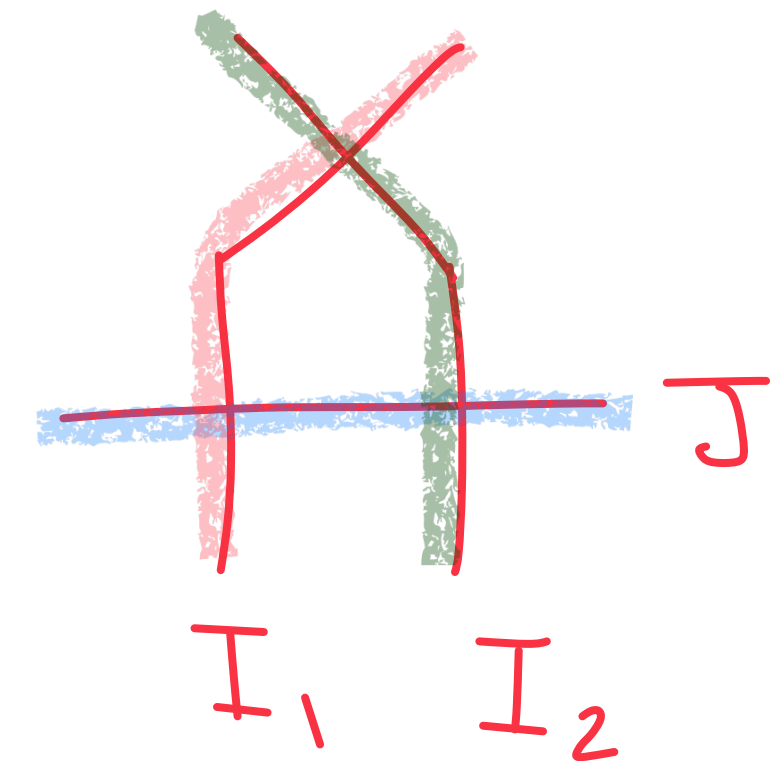
$$L_{u,s}^{(1)}(g, 1; g + 1, 0) = \frac{1 - q^g s^2}{1 - su}.$$

Stochastic:



$$s_1 = q^{-I_1/2}$$

(fusion)

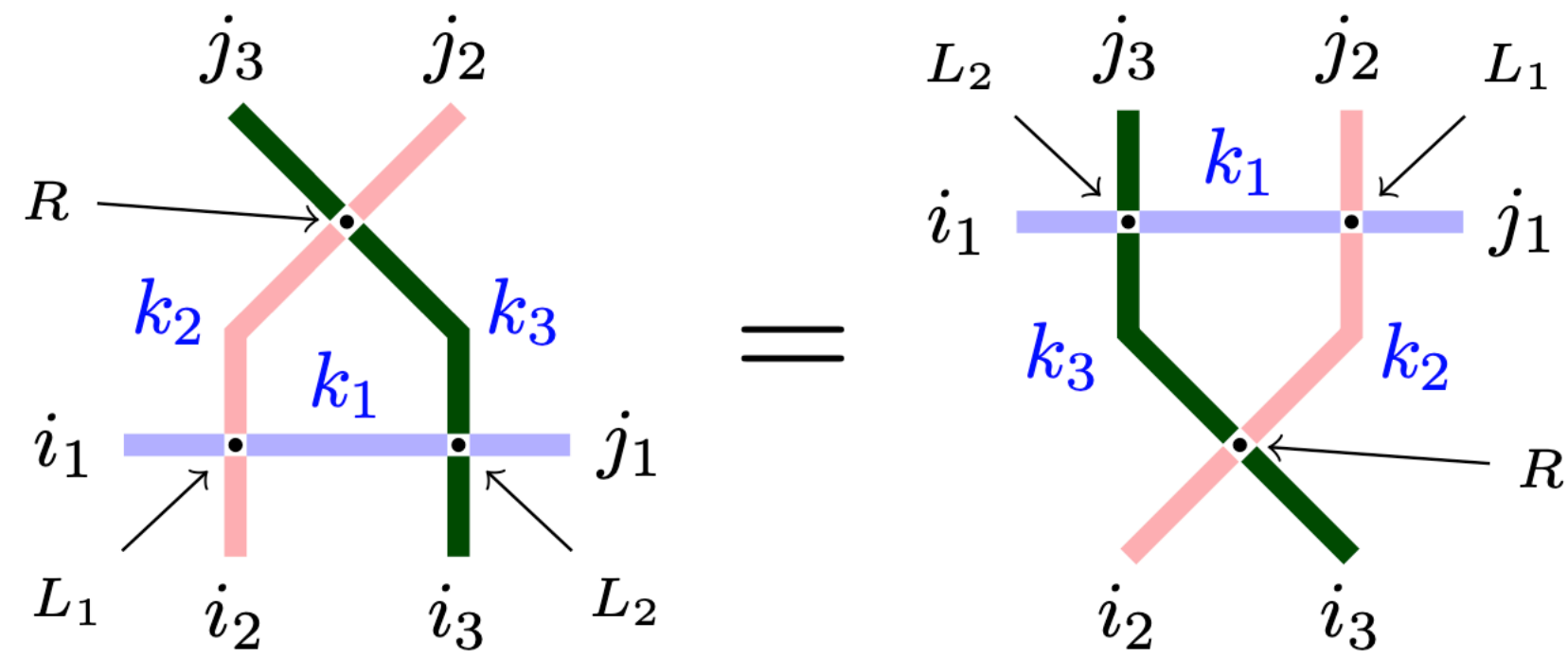


$$L_i = L_{u_i, s_i}^{(J)} \quad (\text{fused})$$

$$R = R_{u_2/u_1, s_1, s_2}^{(I_1)} = L_{\frac{s_1 u_2}{u_1}, s_2}$$



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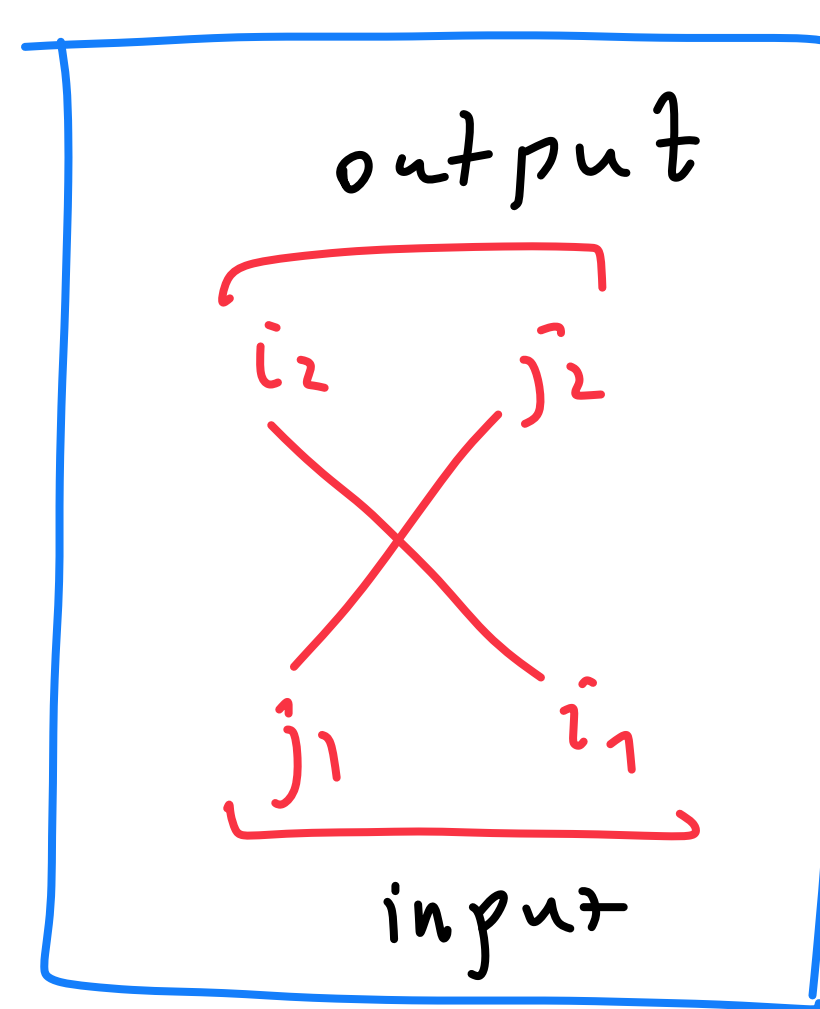
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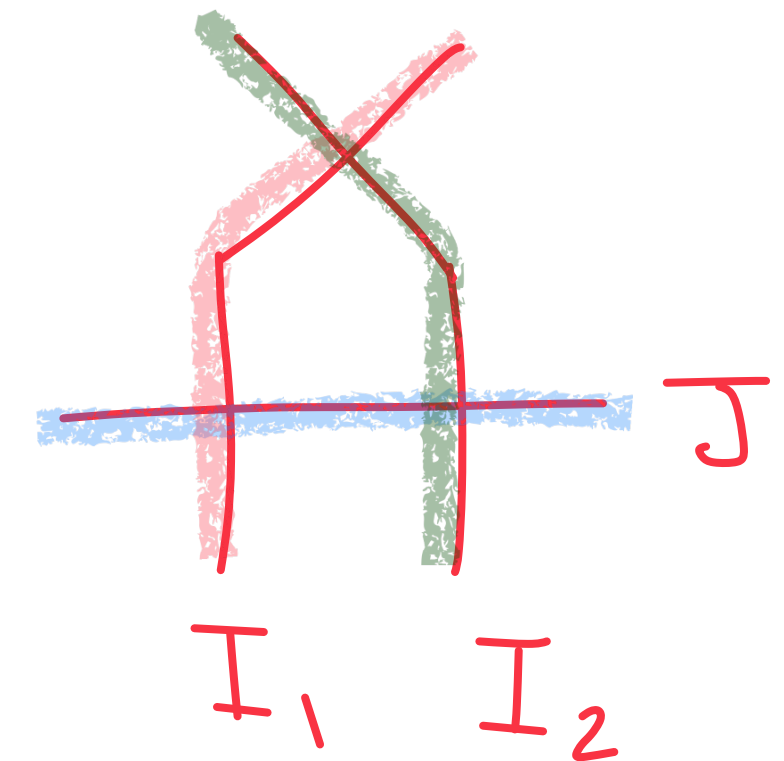
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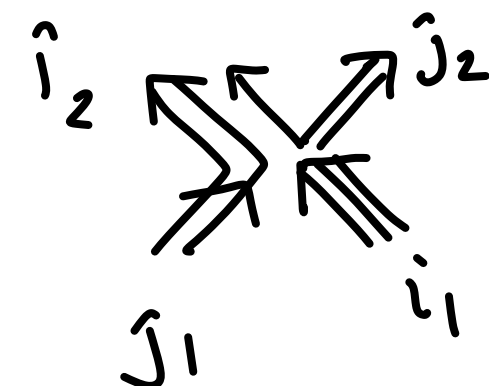
$$R = R_{u_2/u_1, s_1, s_2}^{(I_1)} = L_{\frac{s_1 u_2}{u_1}, s_2}^{(I_1)}$$

q-Mahn (monotone)

$$u_2/u_1 = s_2/s_1$$

⇒

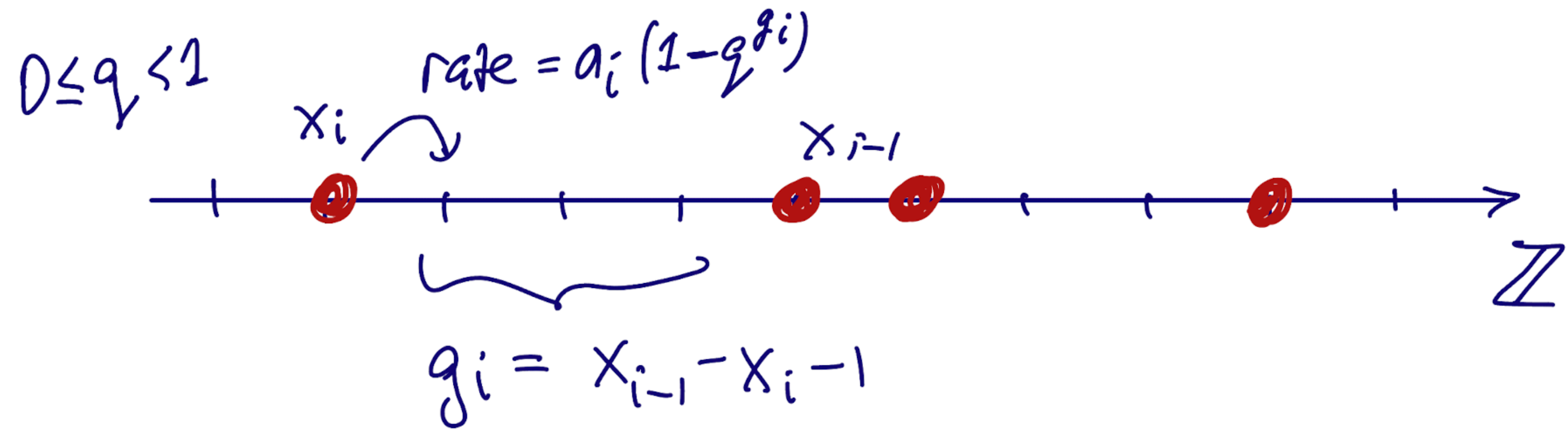
specialization



$$j_2 \leq i_1$$

# General intertwining and Lax equations

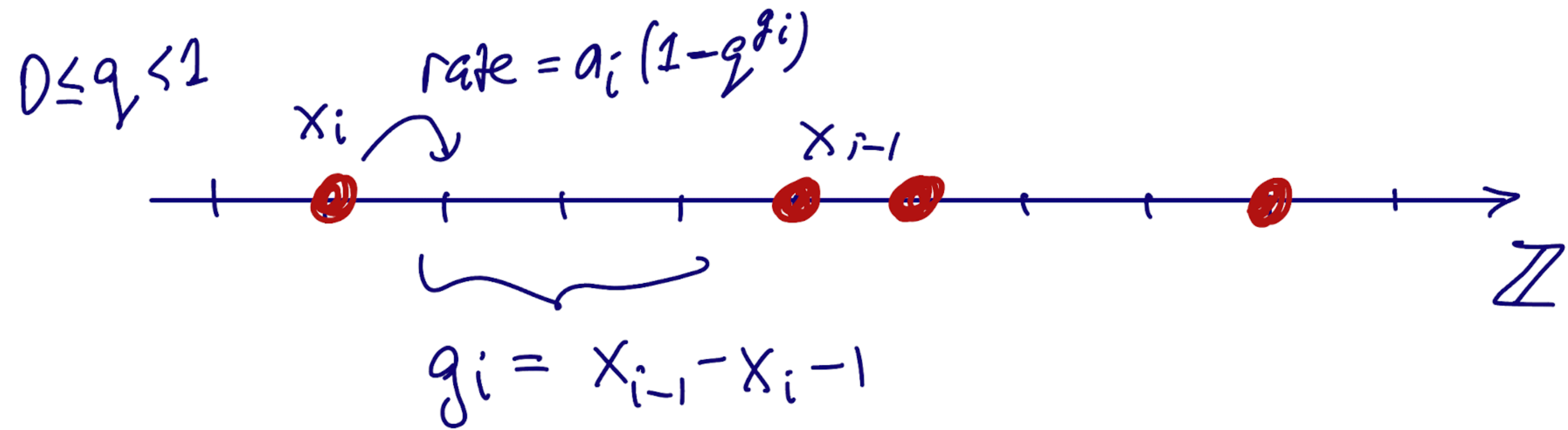
# Intertwining for the $q$ -TASEP in continuous time (add $q$ to make life harder)



$$g_0 = +\infty$$

In continuous time, each particle  $x_i$  jumps forward at rate  $a_i(1 - q^{g_{ap_i}})$ .  
Let  $T_{\mathbf{a}}(t)$  be the  $q$ -TASEP semigroup

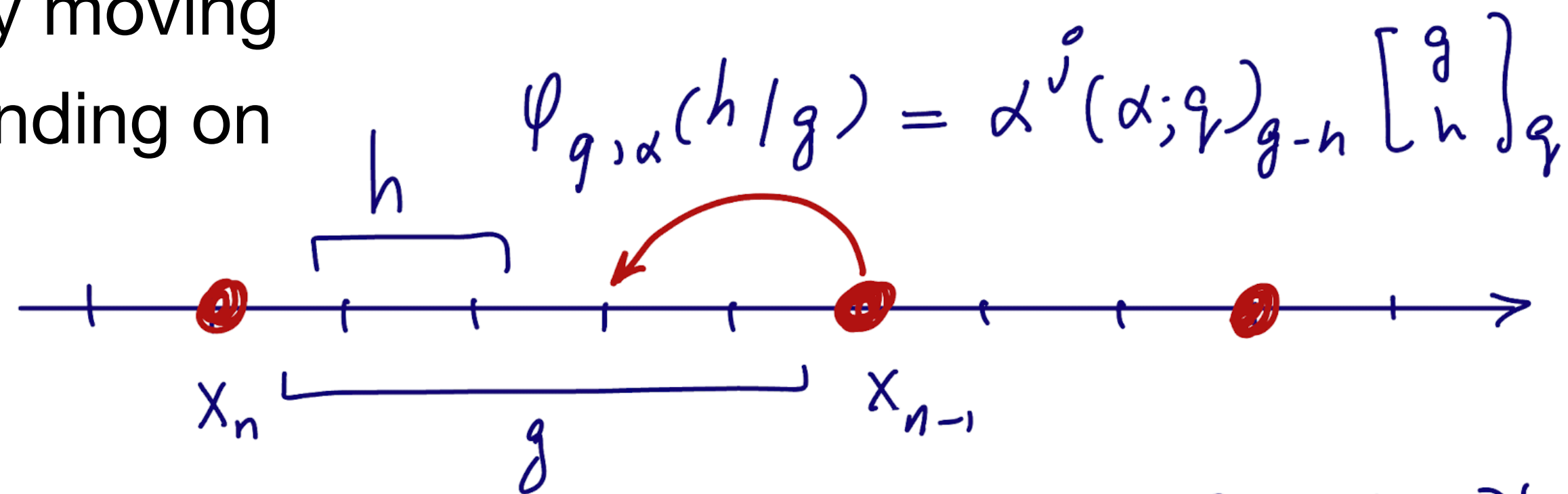
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Let  $B_{n-1}$  be the operator of randomly moving particle  $x_{n-1}$  back closer to  $x_n$ , depending on  $\alpha = a_n/a_{n-1} < 1$ .

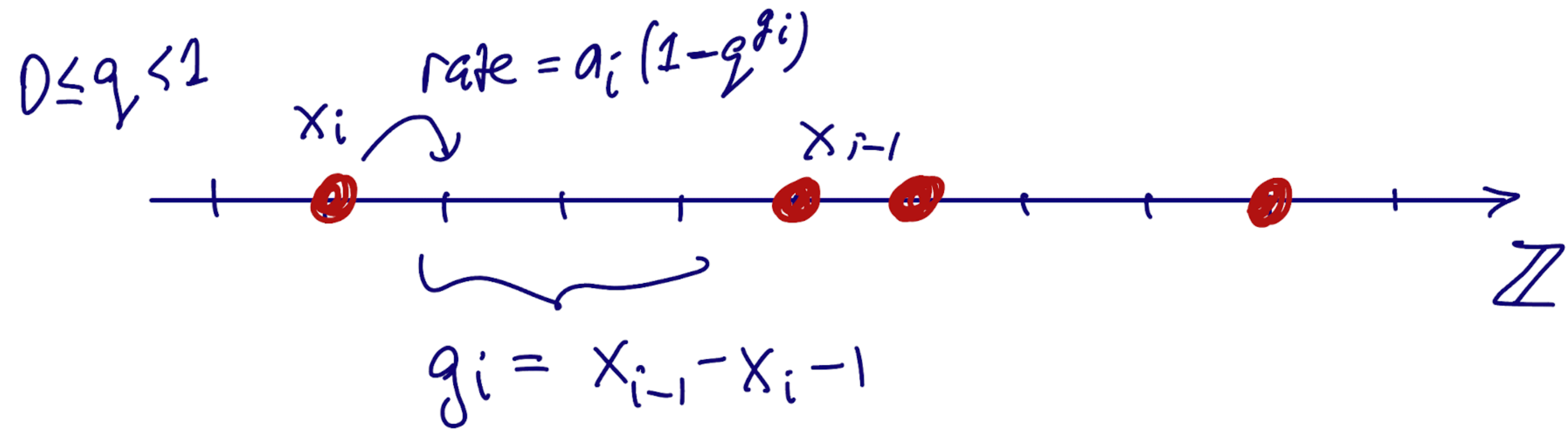


$$(a; q)_k = (1-a)(1-aq) \dots (1-aq^{k-1})$$

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$$[n]_q = \frac{(q; q)_n}{(q; q)_{n-k} (q; q)_k}$$

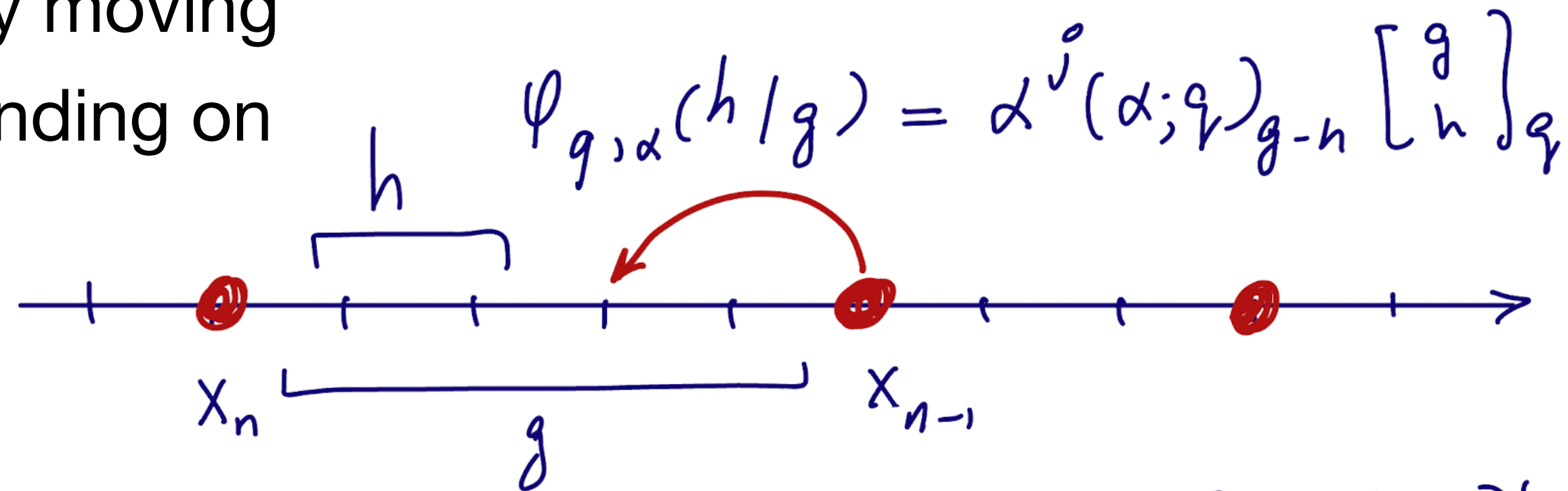
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**Theorem** (P.-Saenz 2022). We have the intertwining  $T_{\mathbf{a}}(t)B_{n-1} = B_{n-1}T_{\sigma_{n-1}\mathbf{a}}(t)$ , where  $\sigma$  swaps  $a_{n-1} \leftrightarrow a_n$ .

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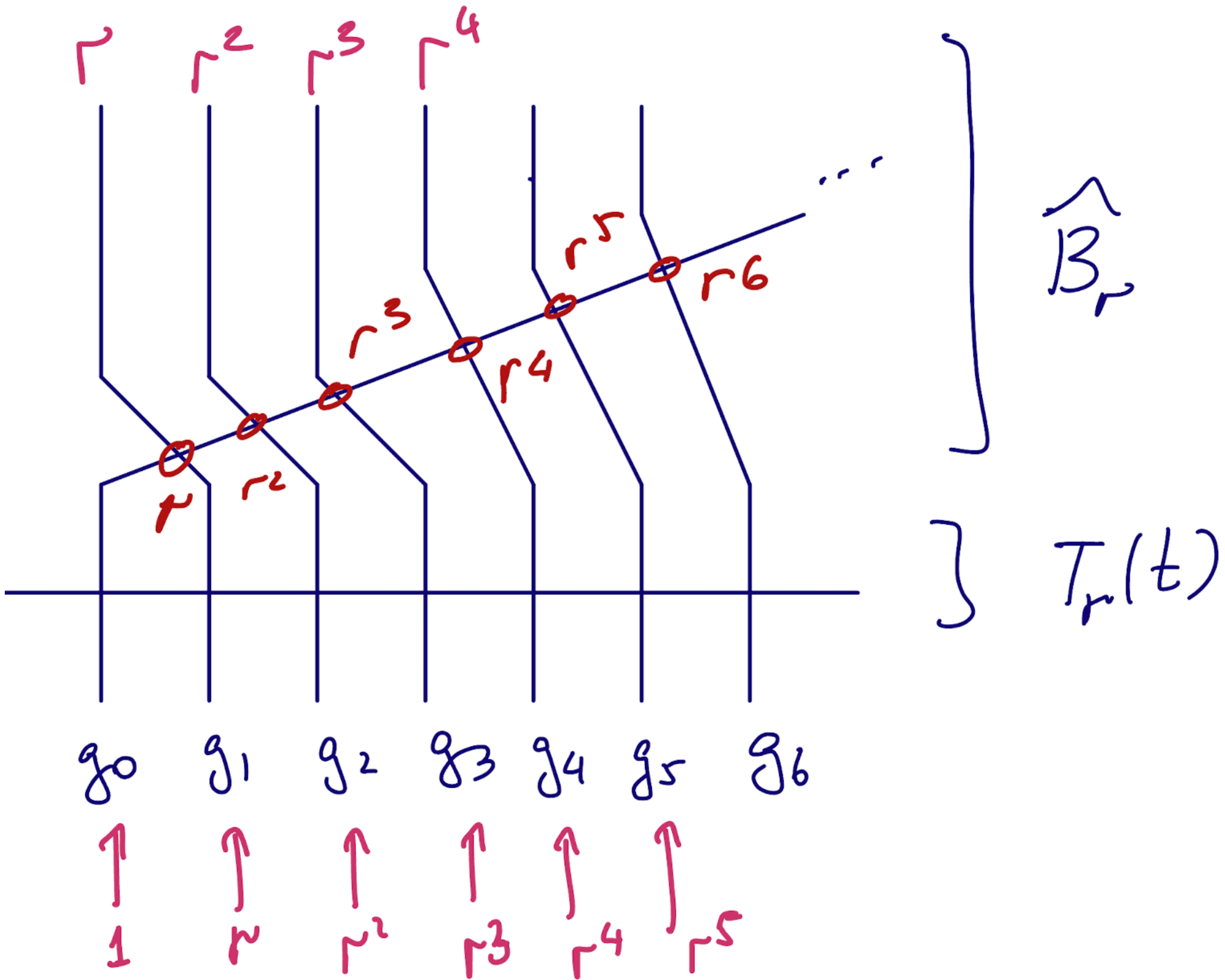
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Application to densely packed configurations: [P.-Saenz 2019], [P. 2019]

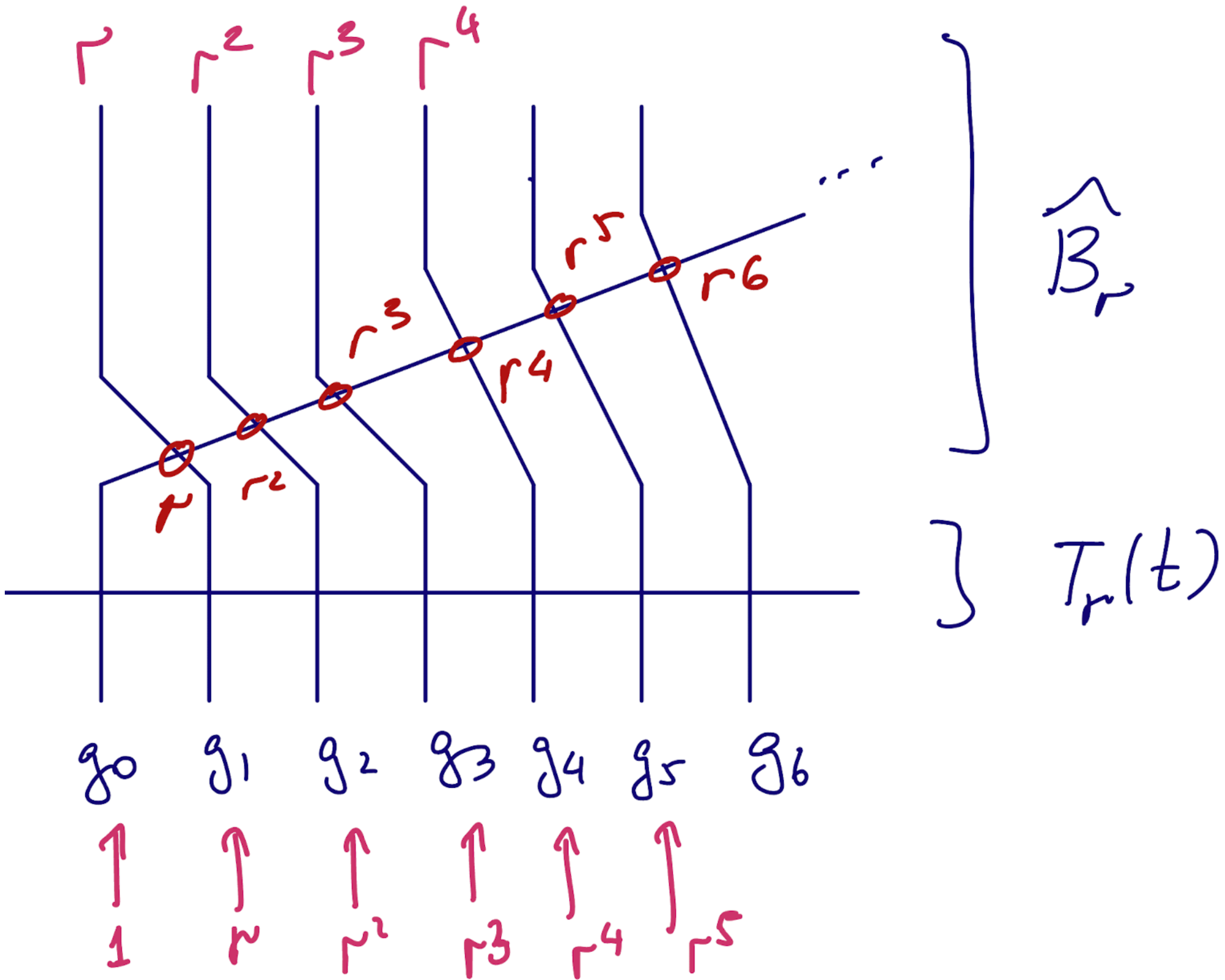
# From intertwining to Lax equations for q-TASEP

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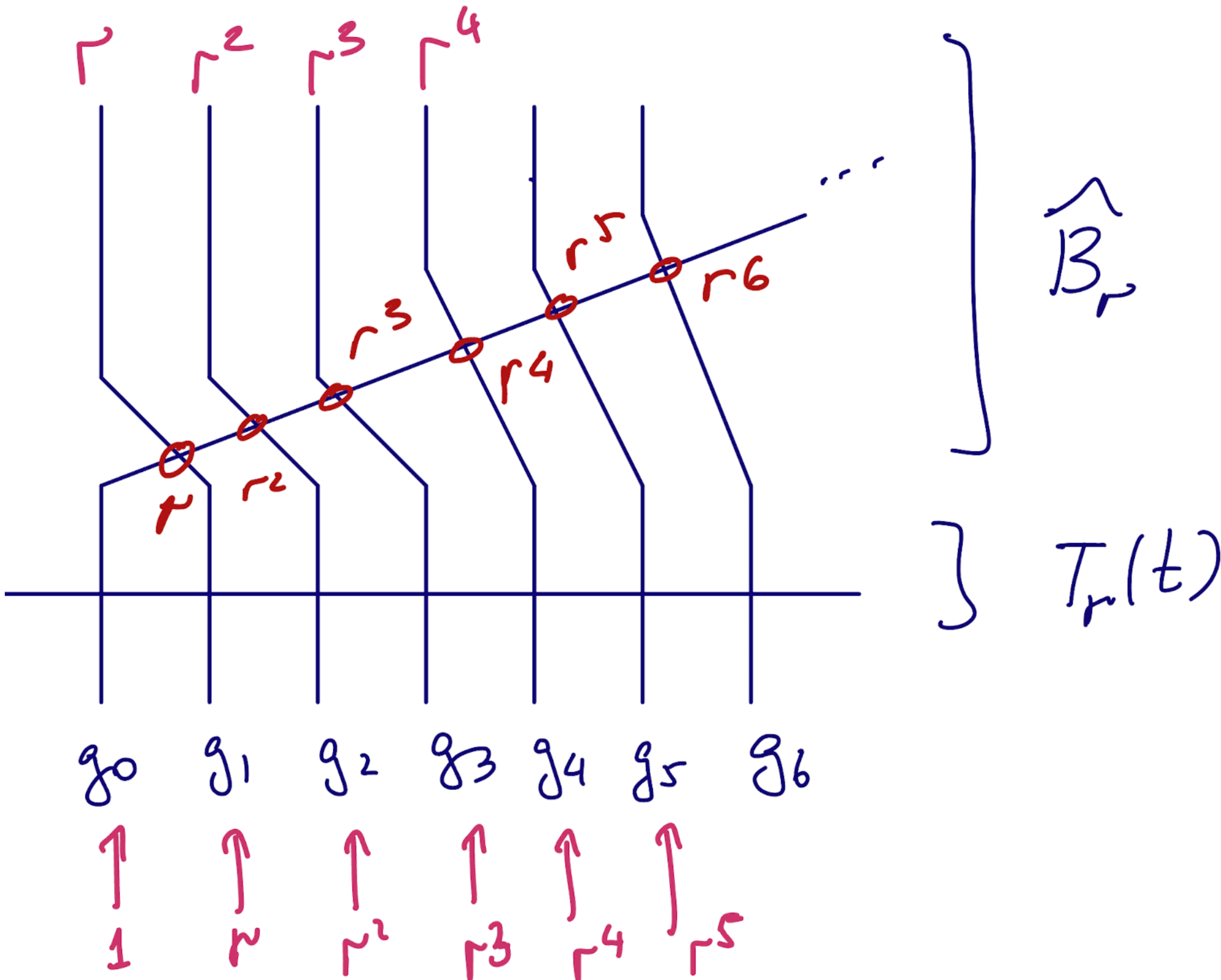
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Take a continuous time Poisson limit of  $(\hat{B}_r)^{\tau/(1-r)}$ . We get a Markov semigroup  $B(\tau)$ , and intertwining

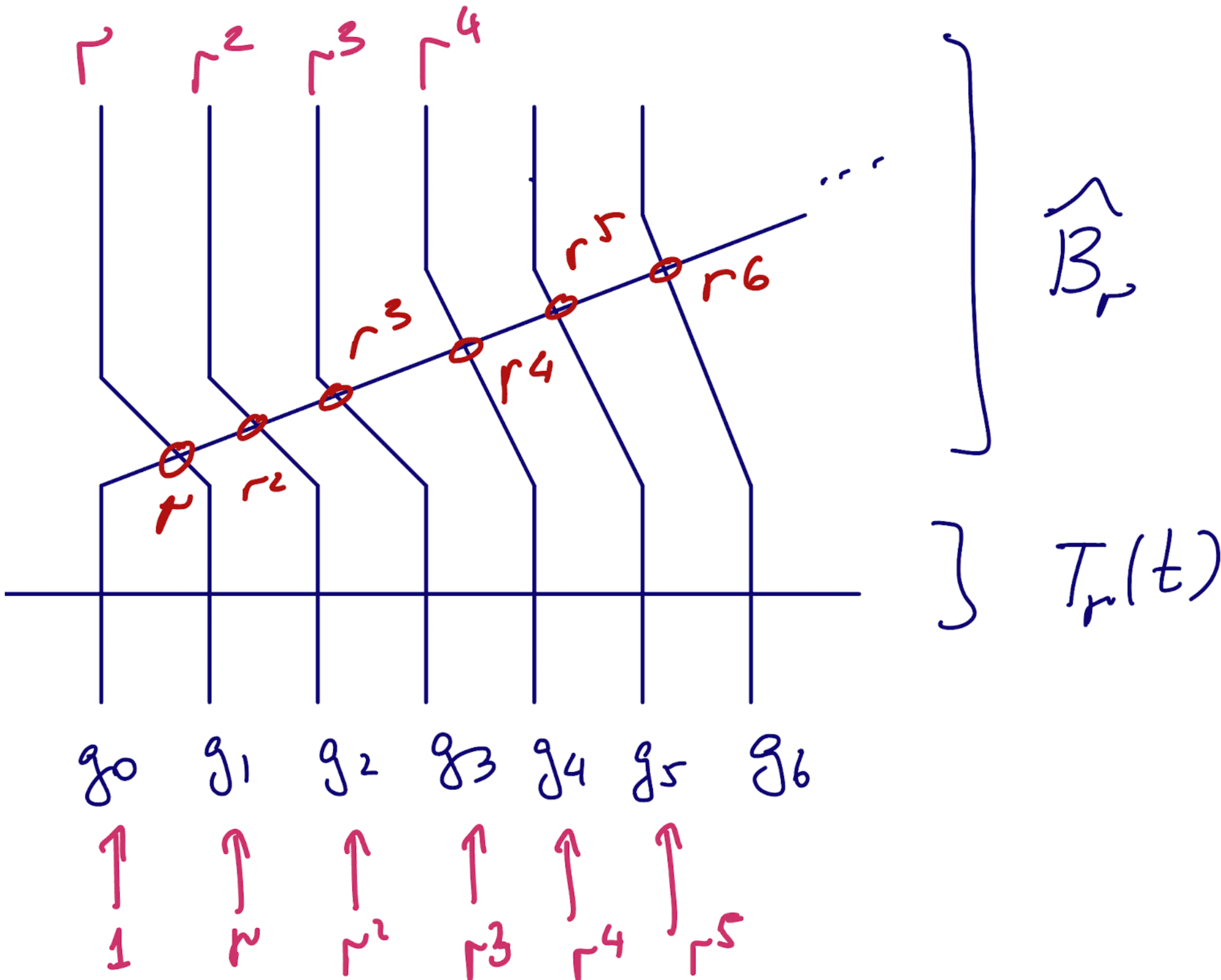
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**Note:** Outside the  $q$ -Hahn case  $u_2/u_1 = s_2/s_1$ , this cross-vertex system  $B_r$  does **not** preserve the empty configuration!

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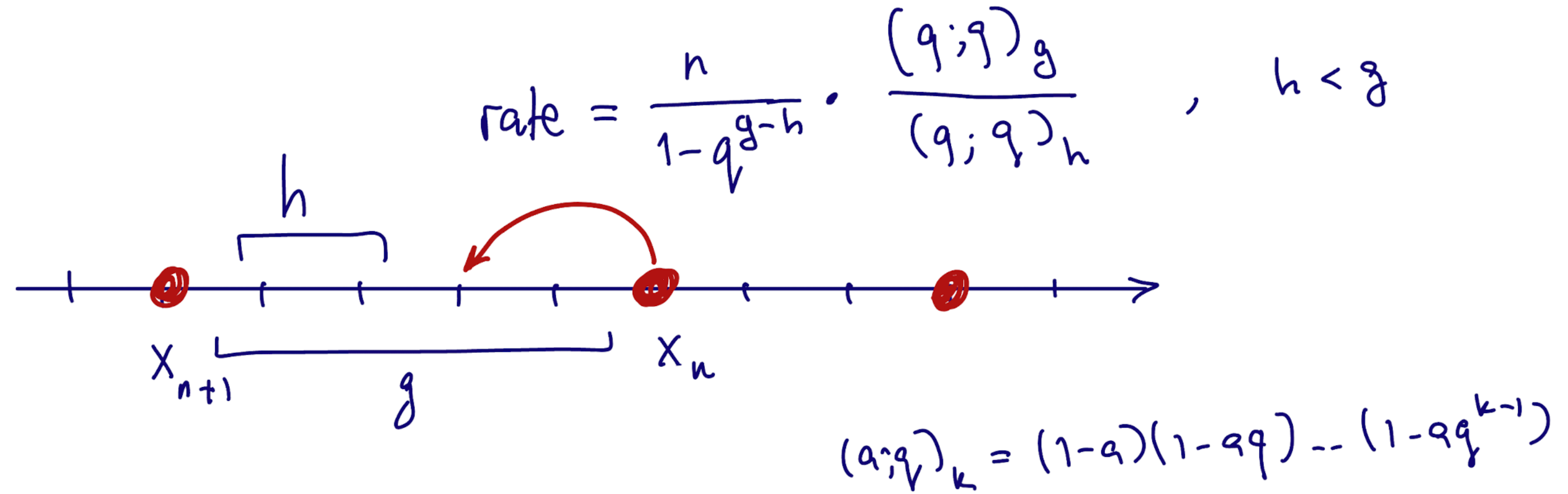
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- The Lax equation should give access to all multipoint observables for q-TASEP, but this information is not that easy to extract... [Quastel-Remenik 2019] show KP equations for KPZ fixed point

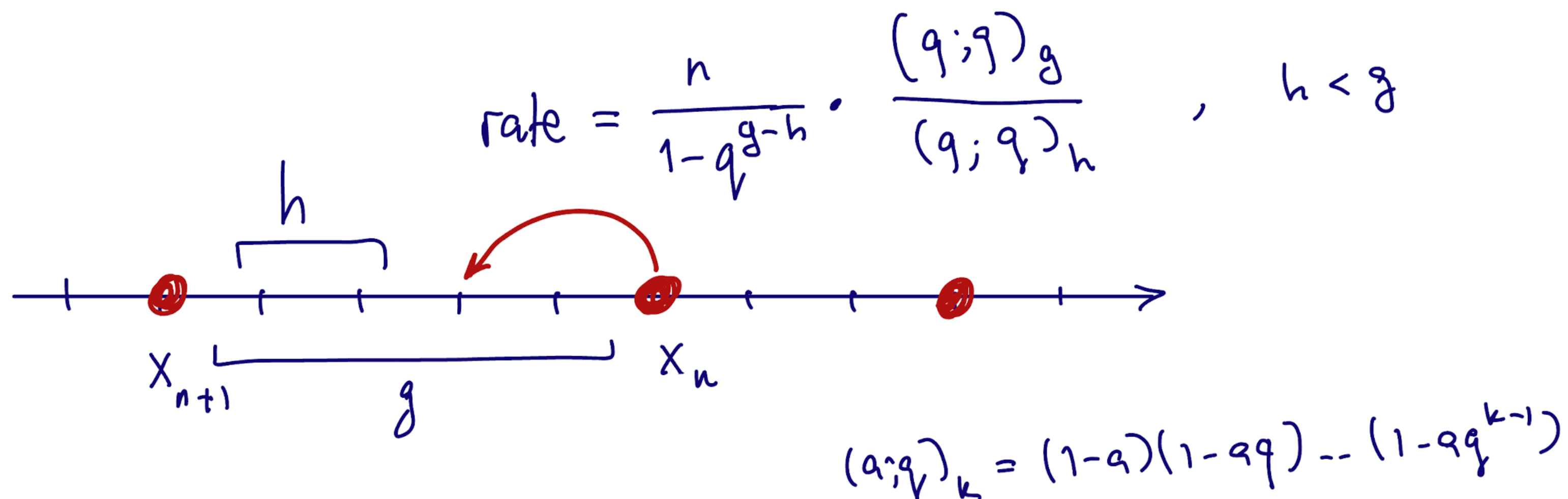
# The backwards dynamics B for q-TASEP and TASEP

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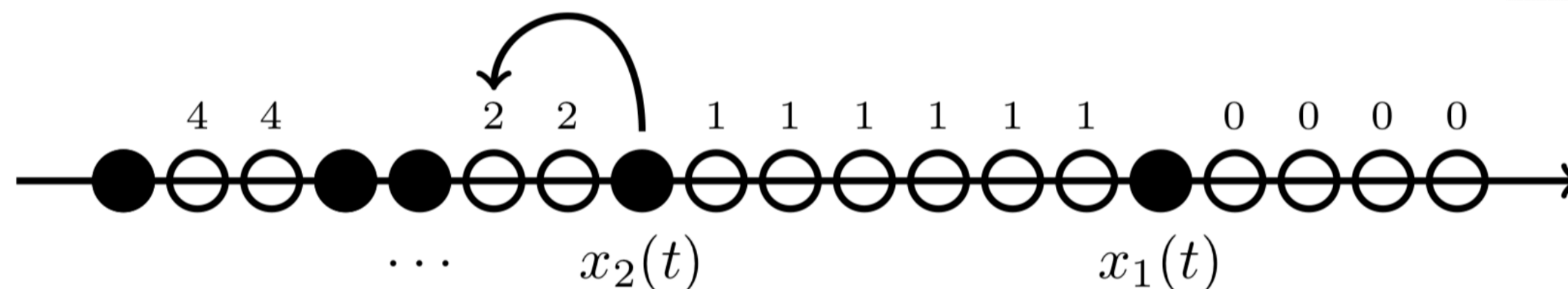


# The backwards dynamics $B$ for $q$ -TASEP and TASEP

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**Case  $q = 0$**

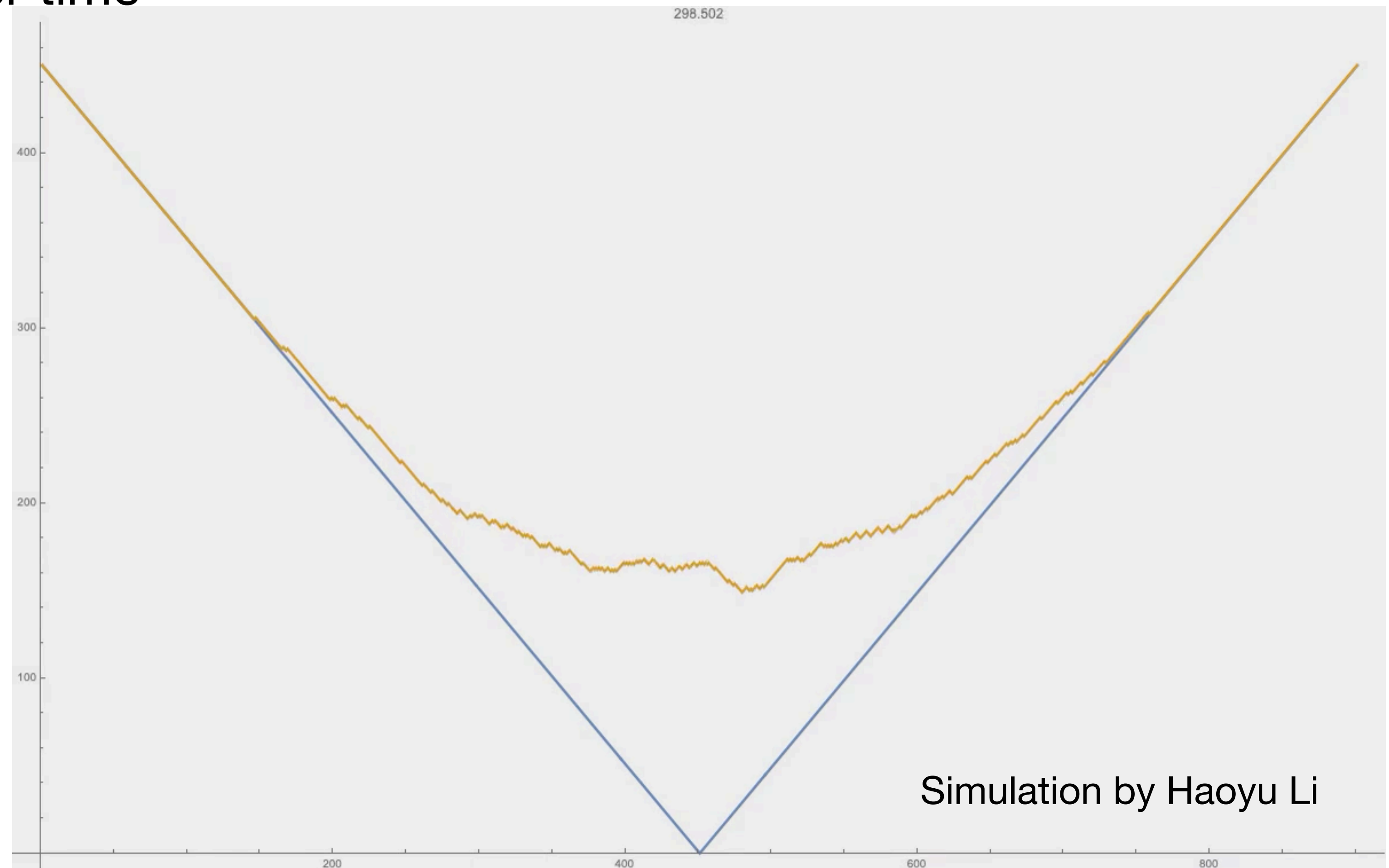


- Each hole has an independent exponential clock with rate equal to the number  $m$  of particles to its right,  $\mathbb{P}(\text{wait} > s) = e^{-m \cdot s}$ ,  $s > 0$ .
- When the clock at a hole rings, the leftmost of the particles that are to the right of the hole instantaneously jumps into this hole
- Because total rate of jump is proportional to the size of the gap, this is a discrete space inhomogeneous version of the Hammersley process **[Hammersley '72], [Aldous-Diaconis '95]**

# Running TASEP back in time

**Theorem [P.-Saenz '19].**  $\delta_{step}T(t)B(\tau) = \delta_{step}T(e^{-\tau}t)$ , which means that if we run TASEP from the step (densely packed) initial configuration, and then run the backwards process, then the result is a TASEP distribution at an earlier time

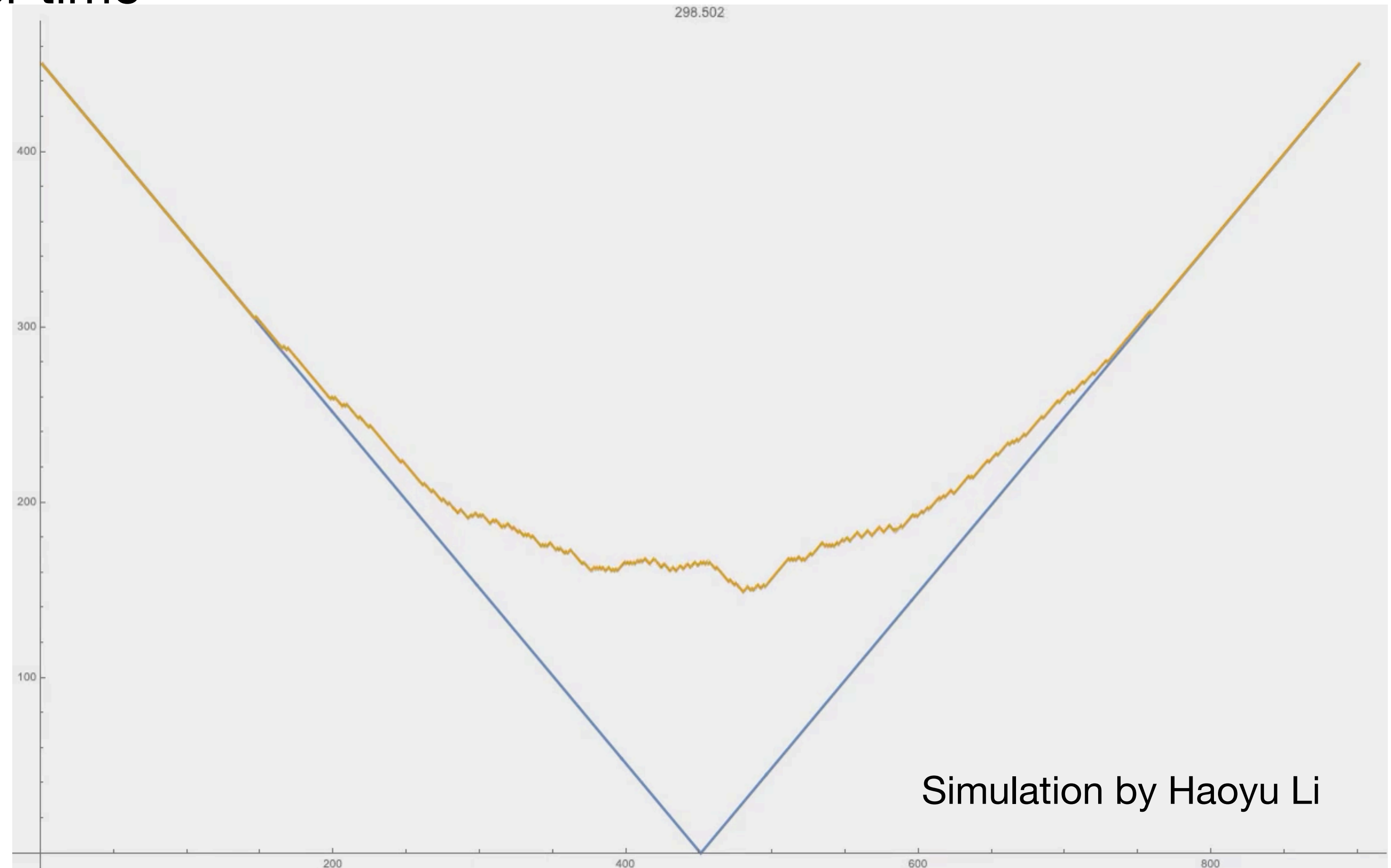
TASEP as a growth process  
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# Coupling and rewriting history



# Bijection of the Yang-Baxter equation

Let  $A, B$  be finite sets and  $\sum_{a \in A} w(a) = \sum_{b \in B} w(b)$  (with positive terms)

A **bijection (coupling)** of this identity is a family of transition probabilities

$p(a \rightarrow b)$  and  $p'(b \rightarrow a)$ , satisfying

$$w(a)p(a \rightarrow b) = w(b)p'(b \rightarrow a)$$

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Example:  $1 + 3 = 2 + 2$

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1	1	0	(maximally dependent)
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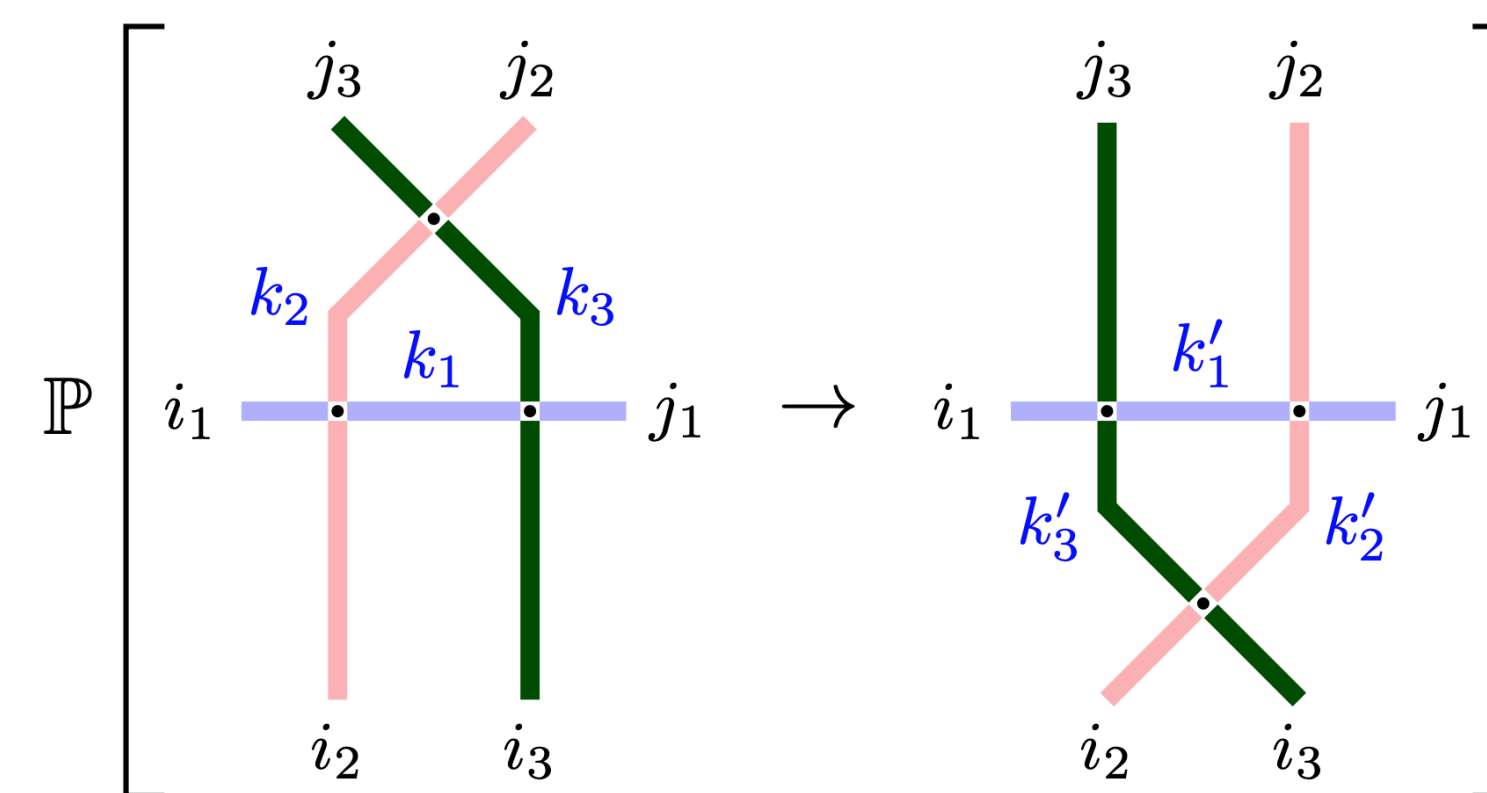
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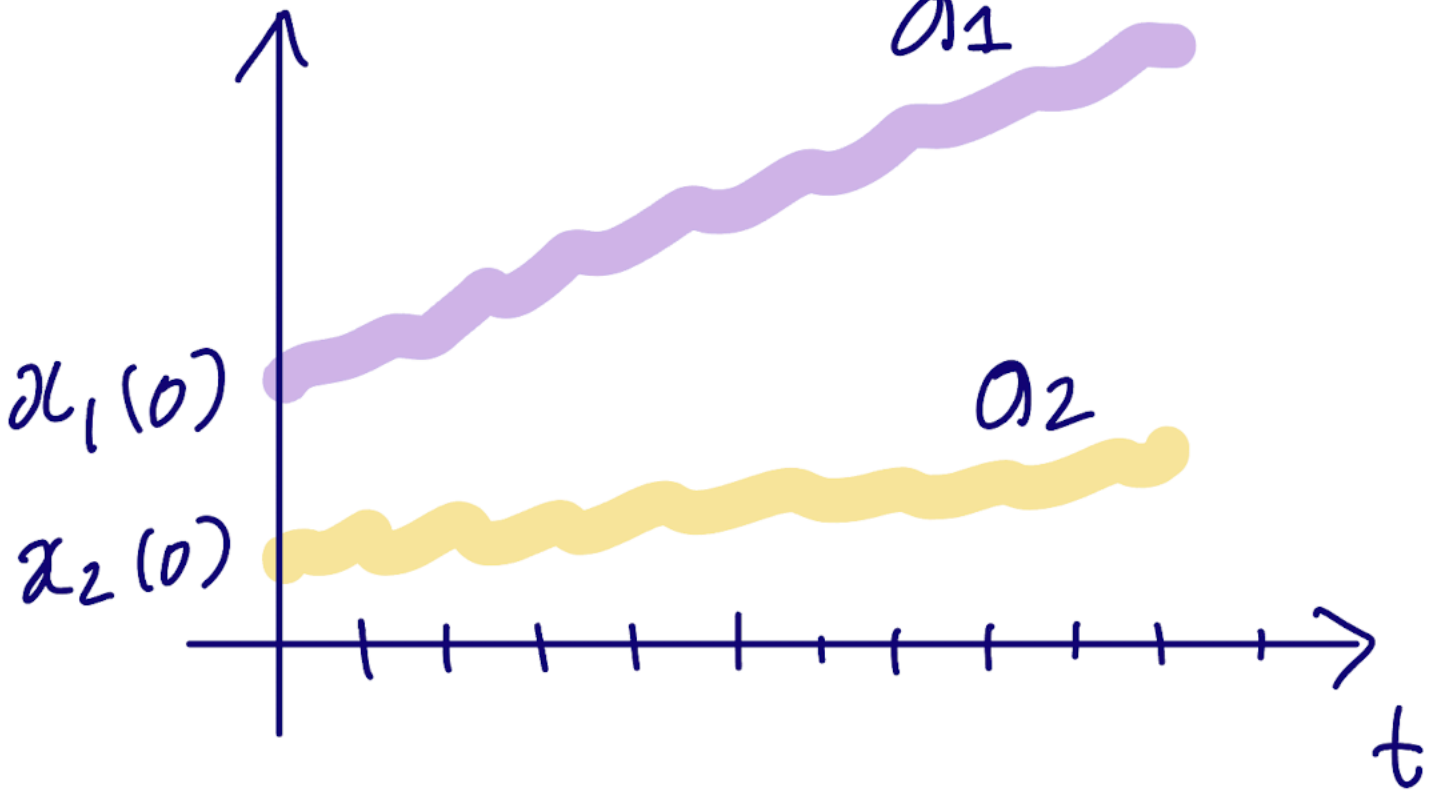
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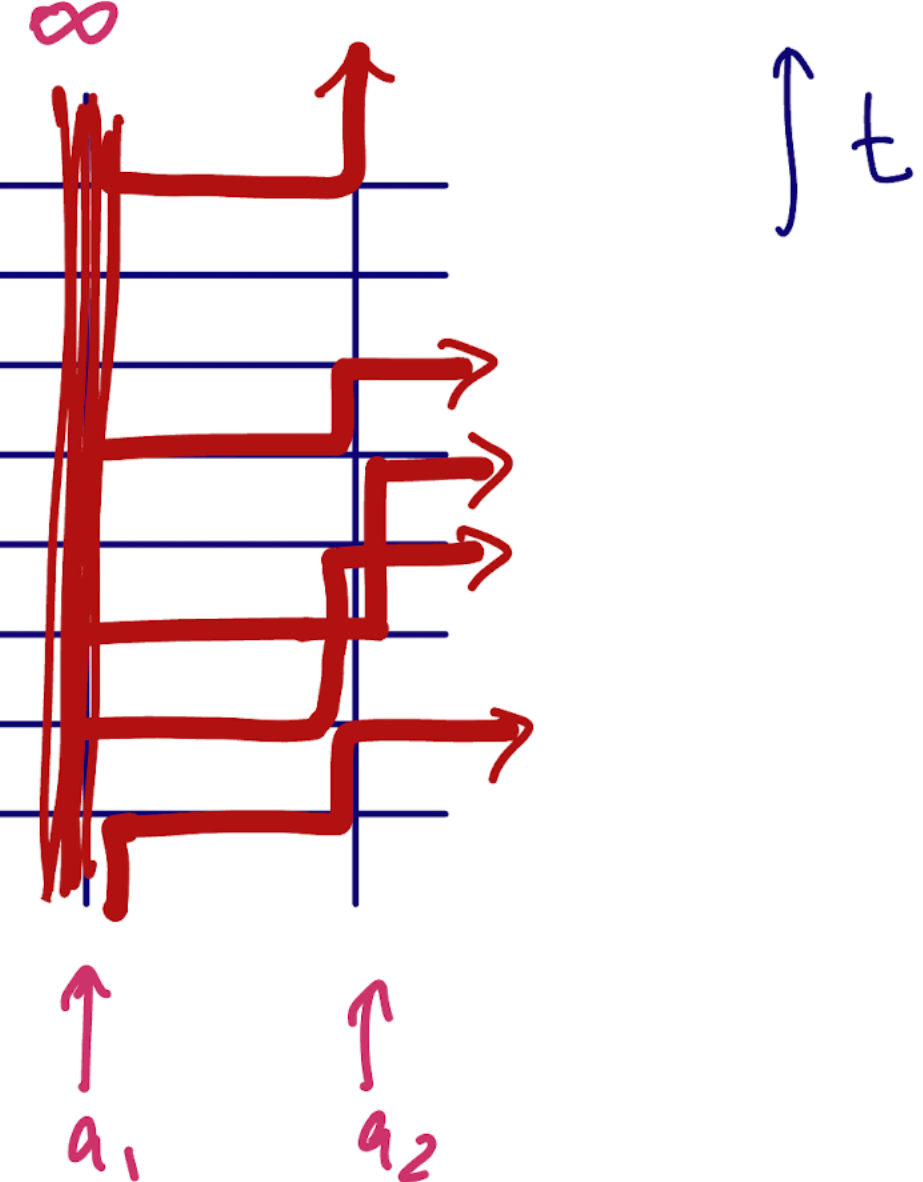


# Rewriting history processes

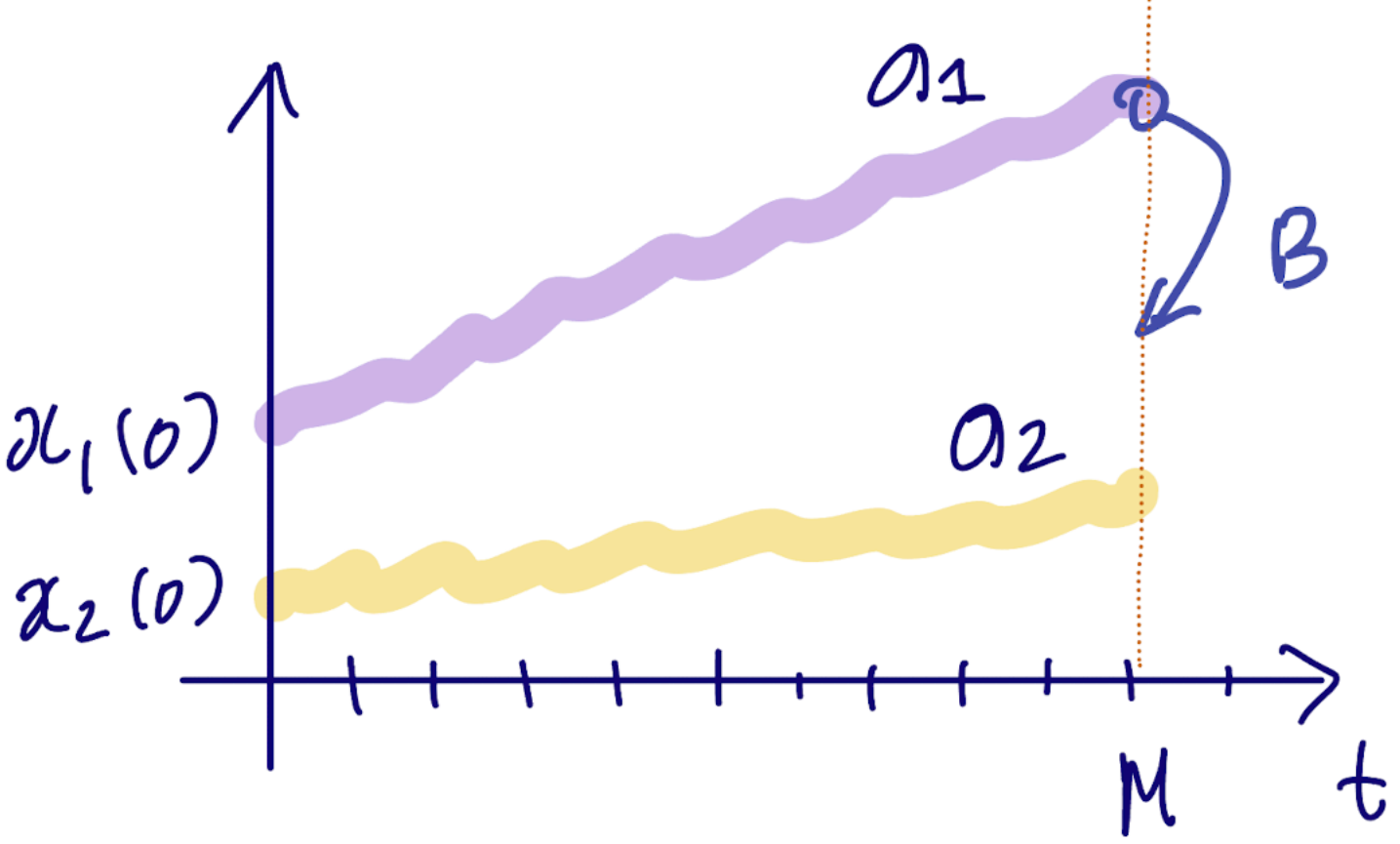
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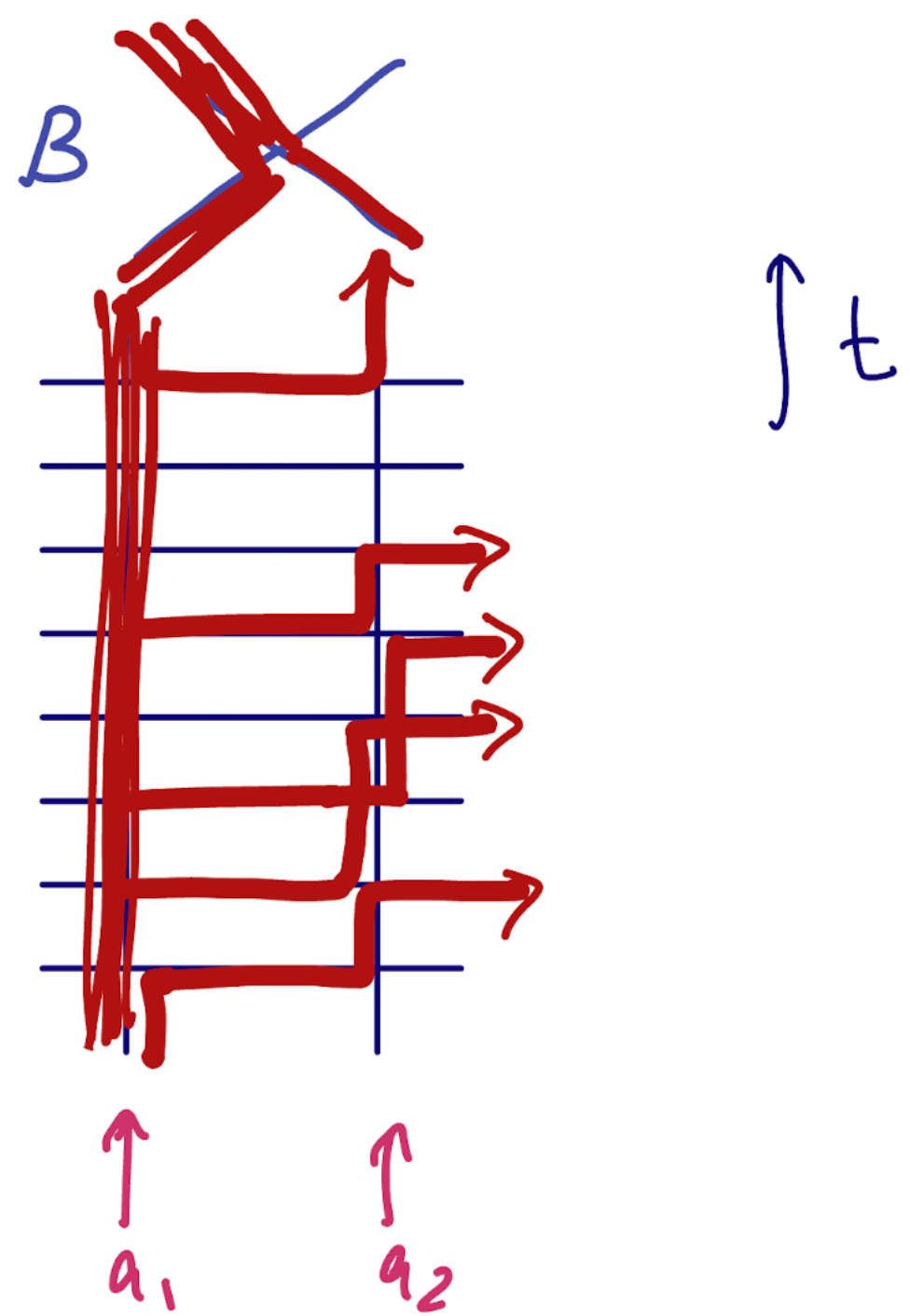
$\Leftrightarrow$



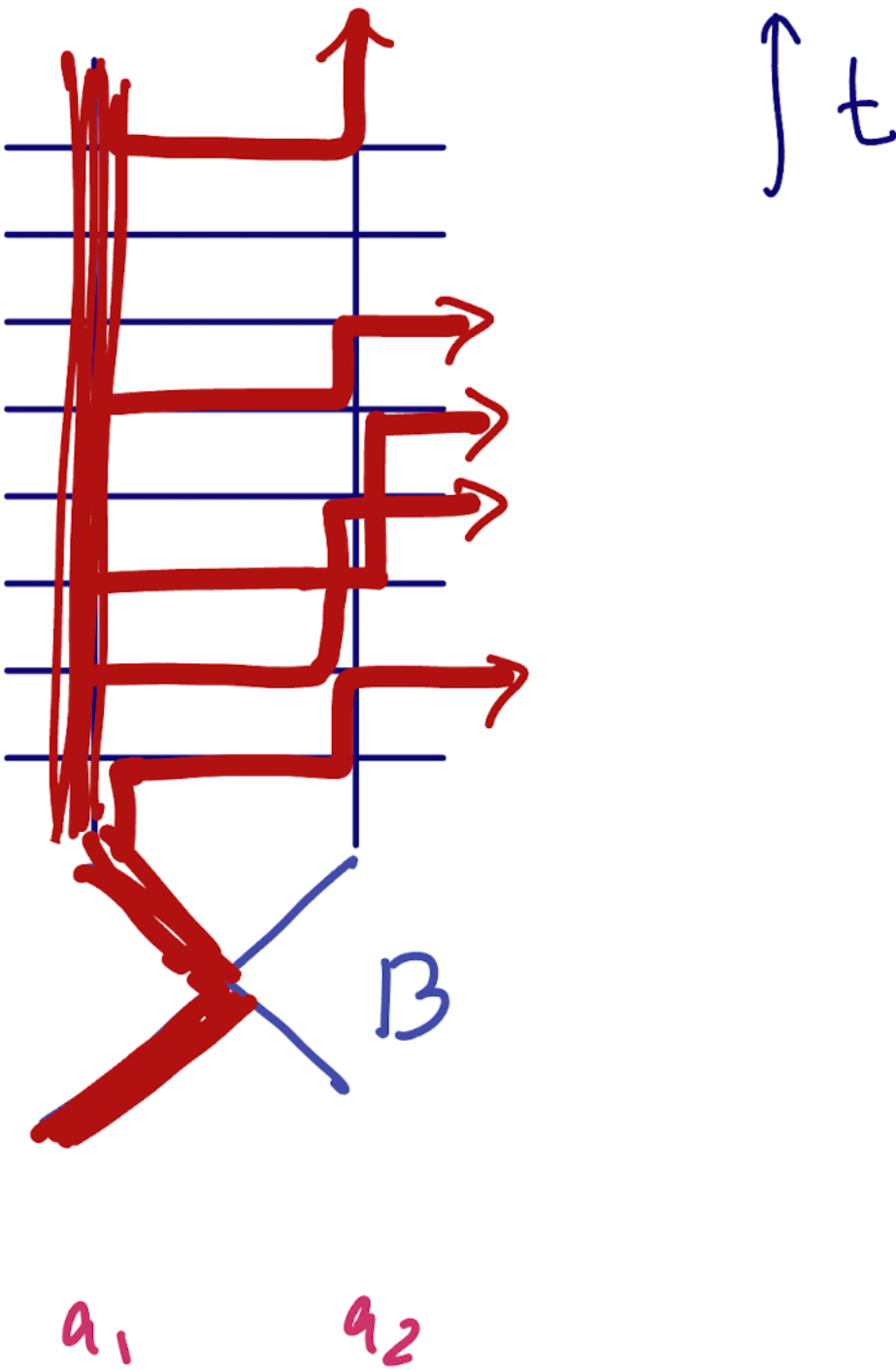
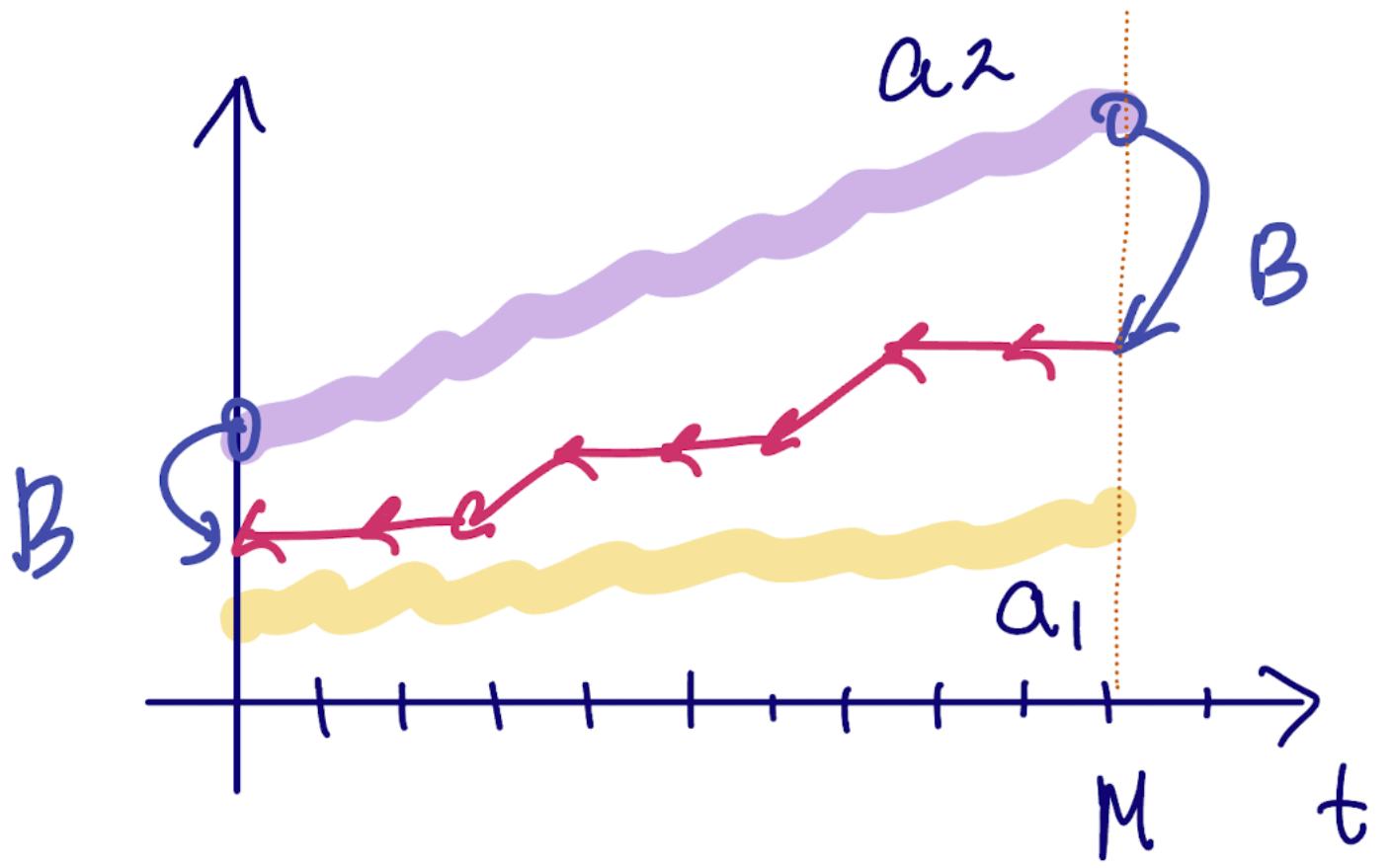
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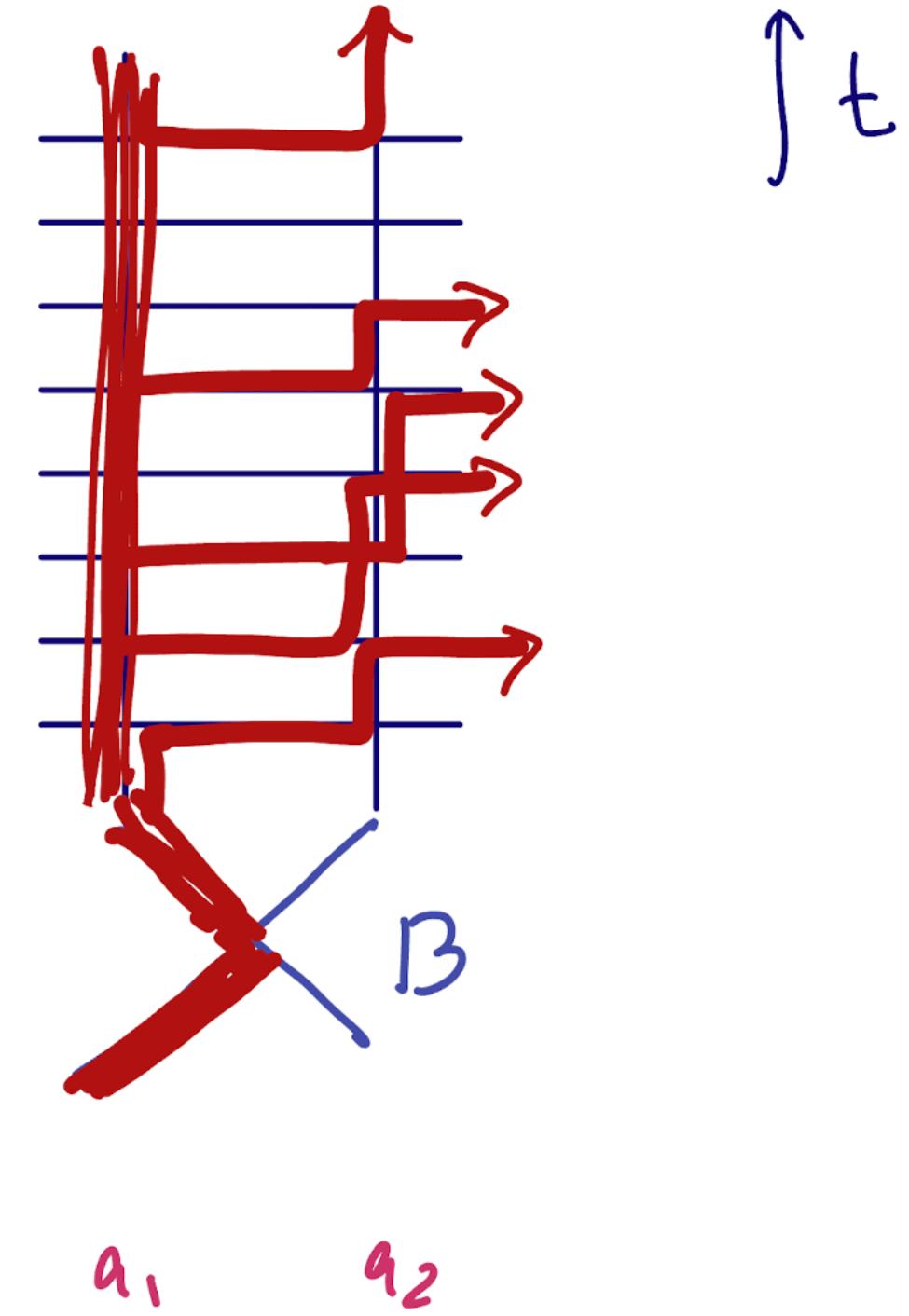
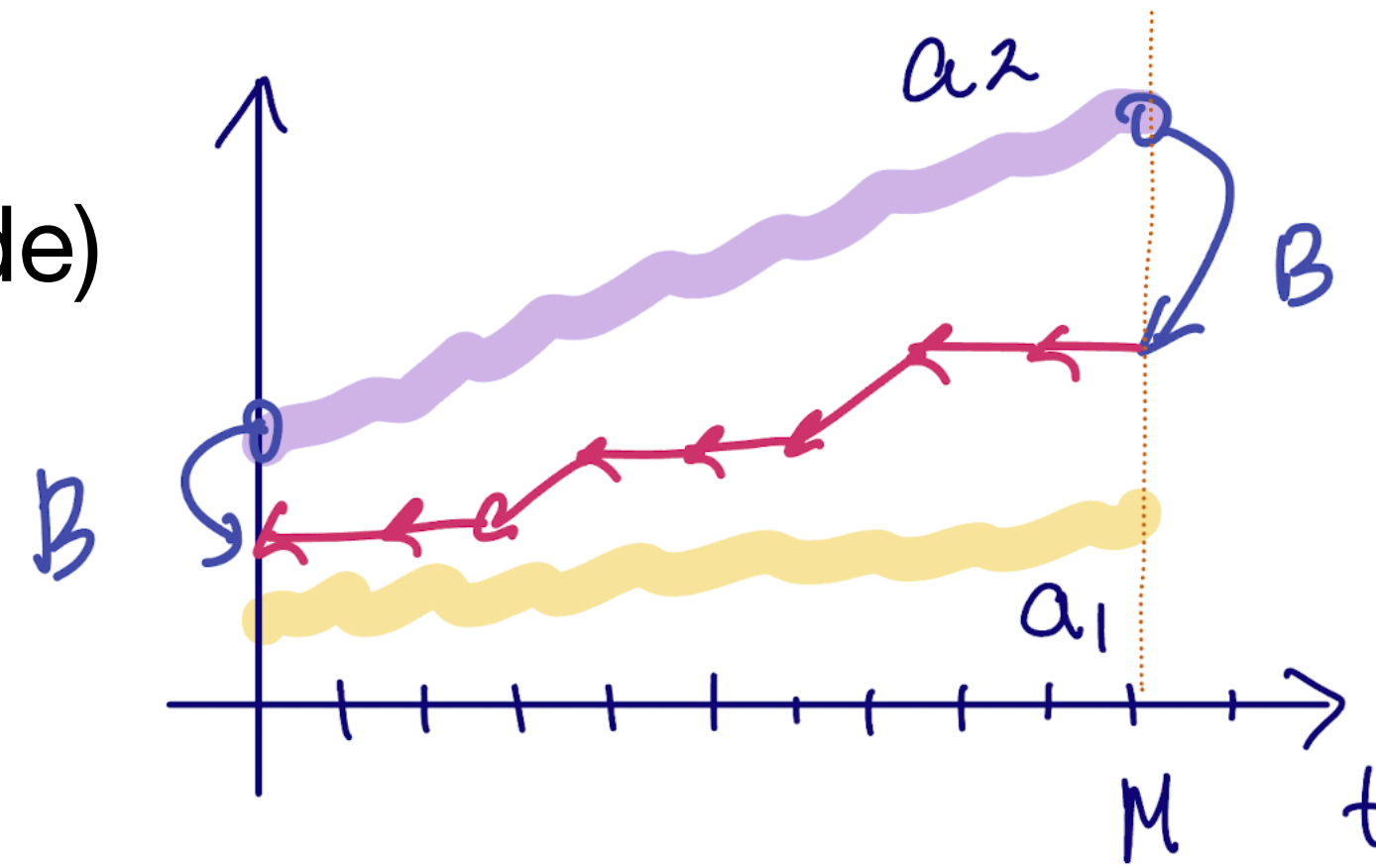
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We get two processes for rewriting history:

- from future to past (in the figure)
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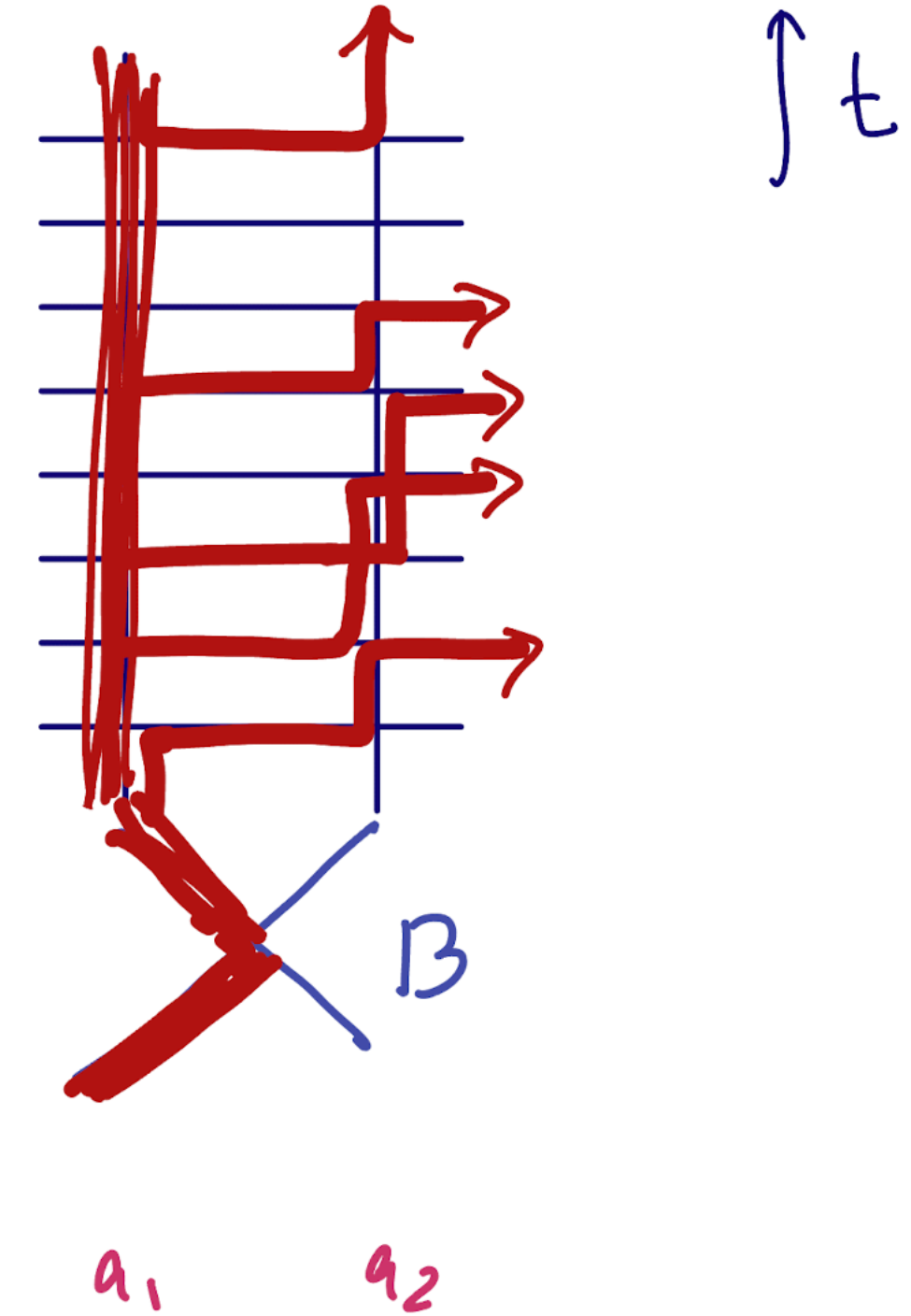
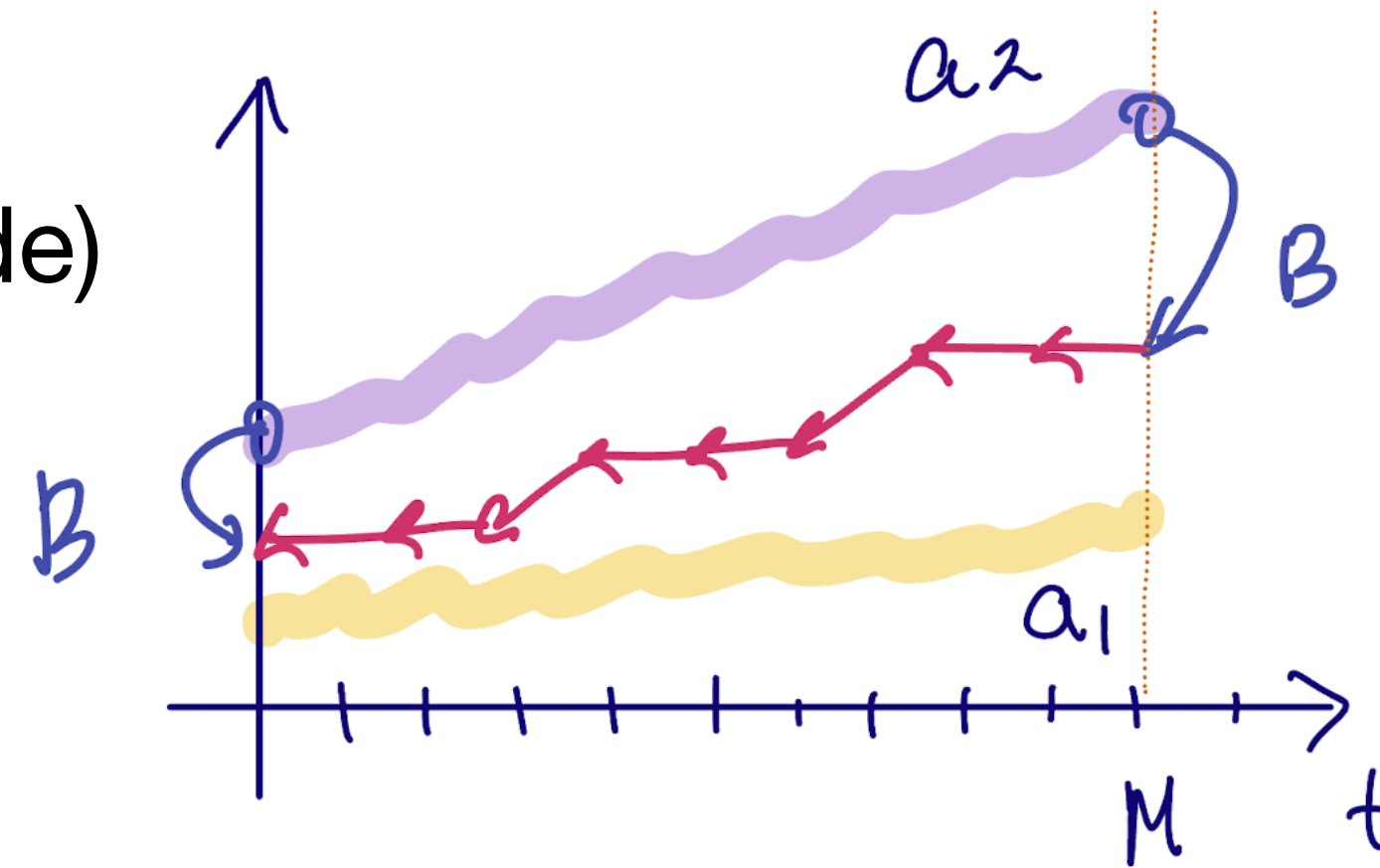




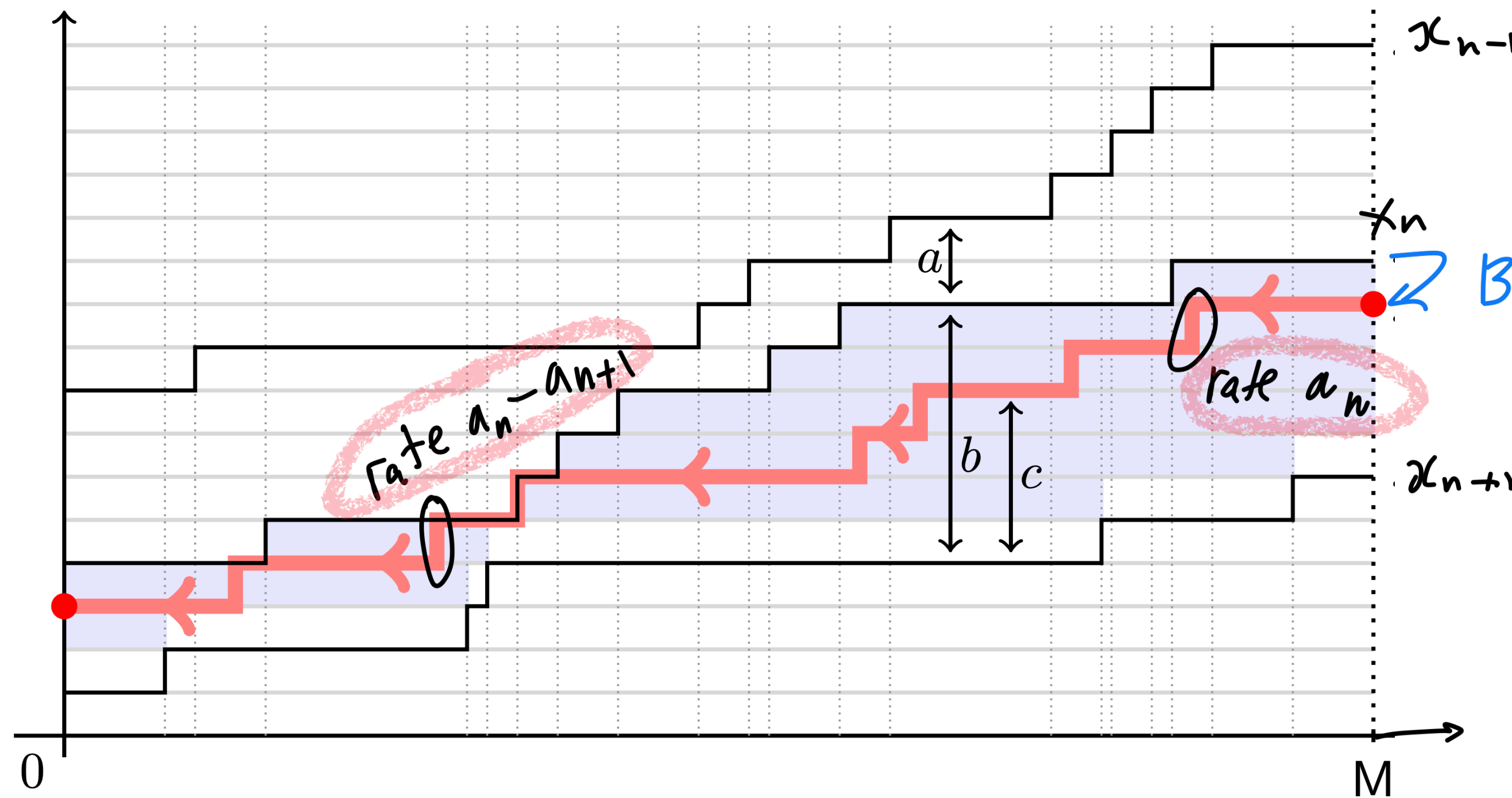
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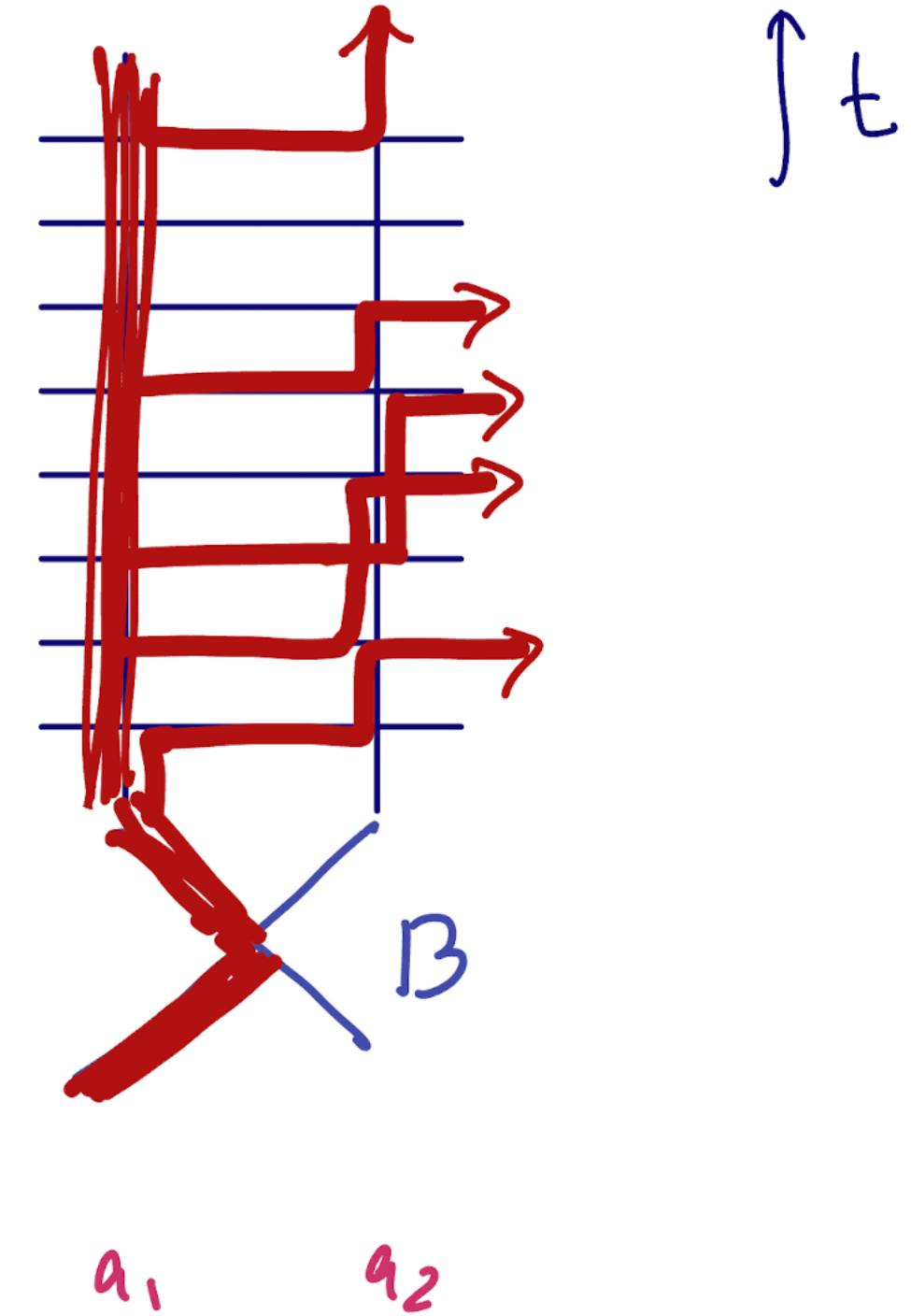
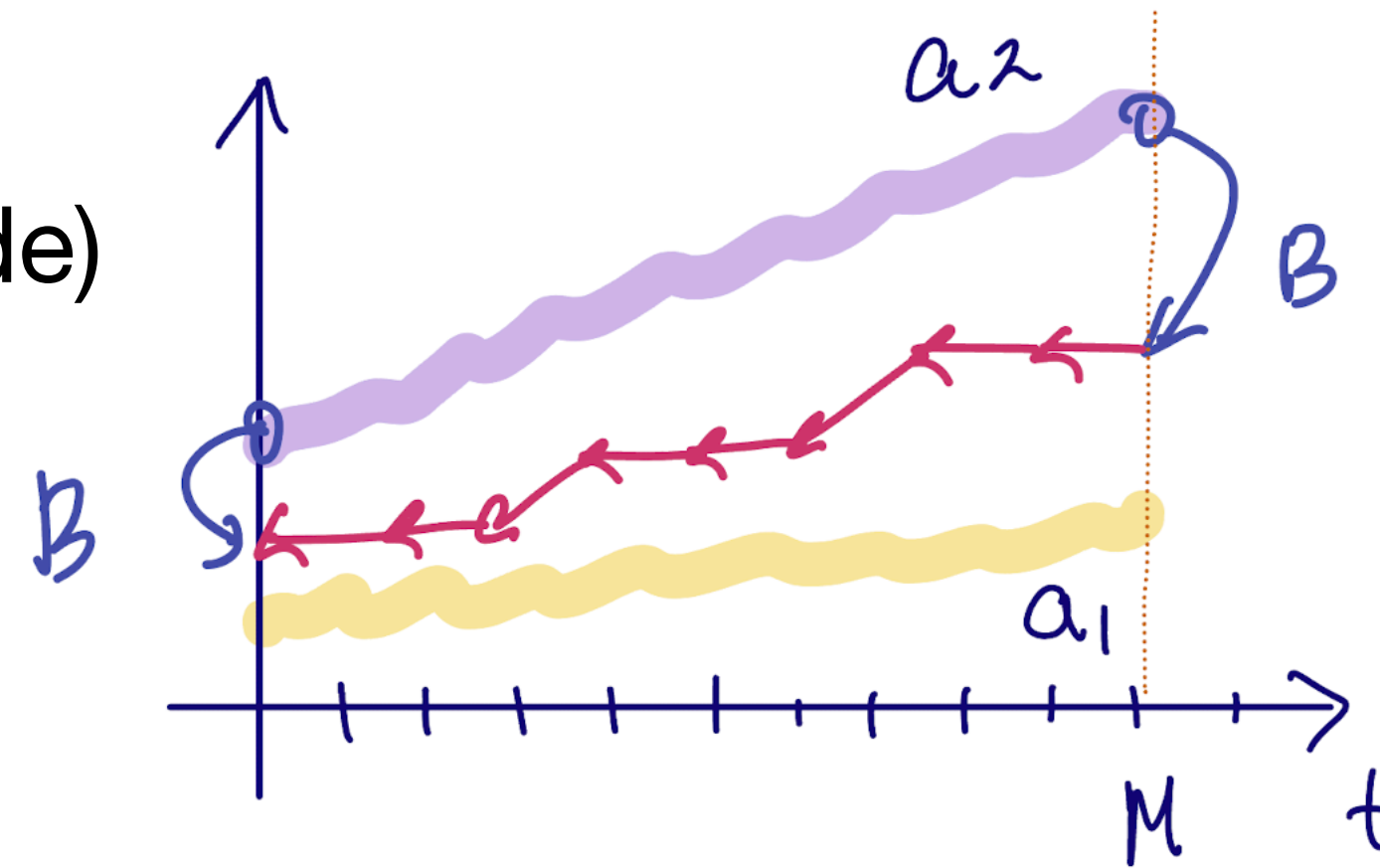
Cont. Time TASEP w.  $a_n > a_{n+1}$



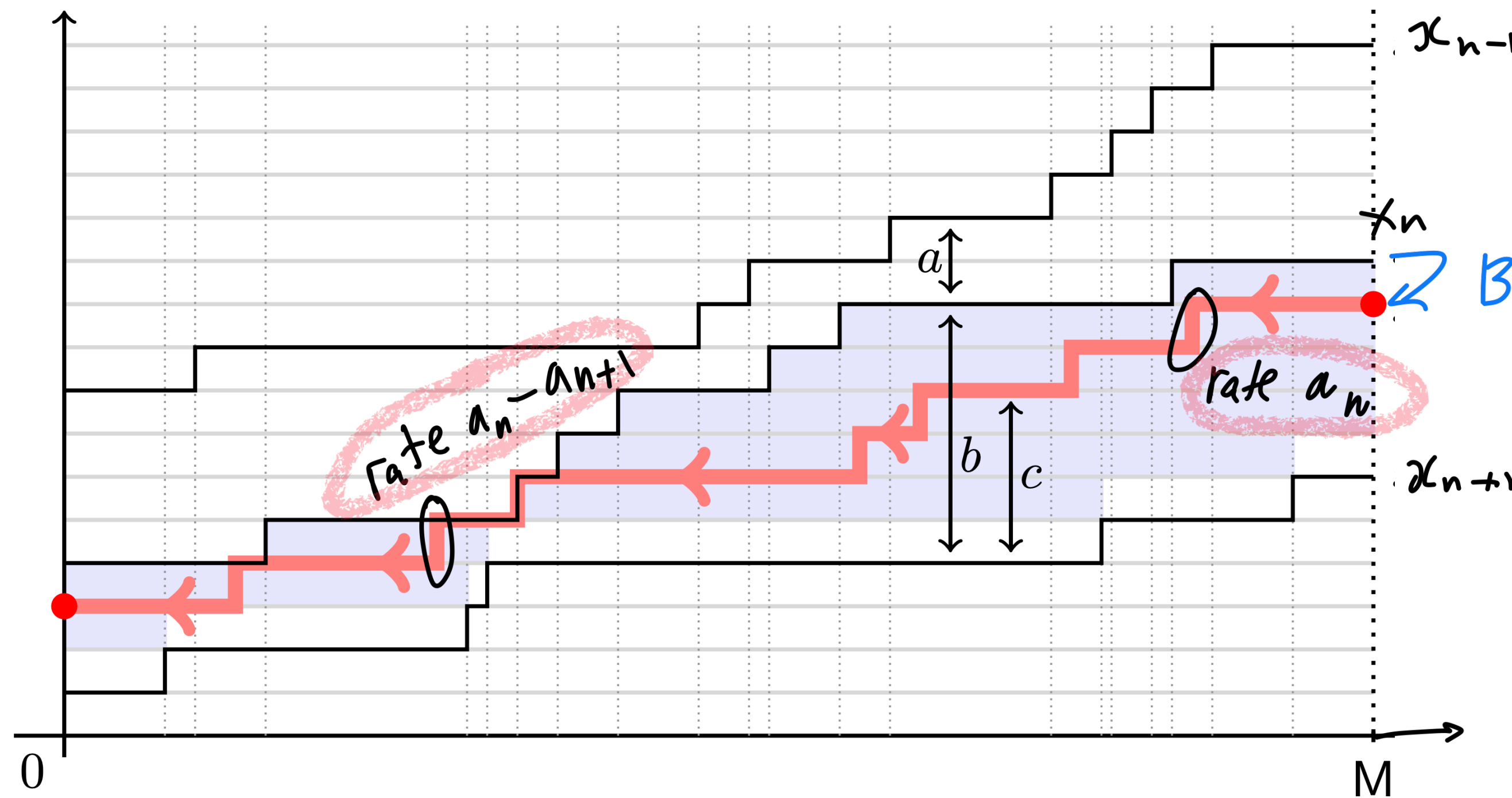
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**Theorem** [P.-Saenz 2022]. Move  $x_n$  back by geometric jump B. Then run a random walk  $x'_n$  in the chamber between  $x_n, x_{n+1}$ , in reverse time, with jump rates down  $a_n - a_{n+1} \mathbf{1}_{b=c}$ .

Then the new trajectories are distributed as a TASEP with speeds

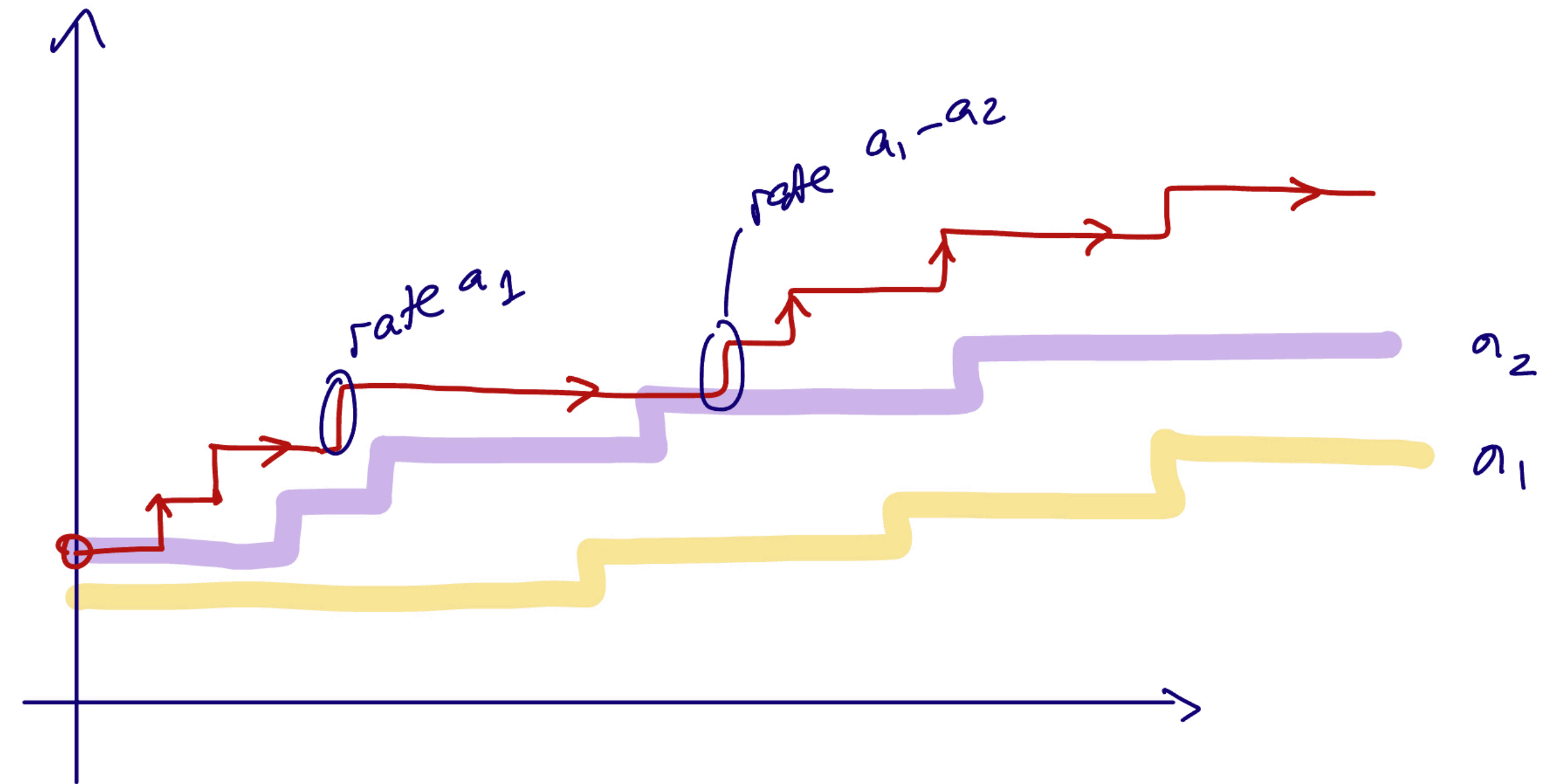
$(\dots, a_{n-1}, a_{n+1}, a_n, a_{n+2}, \dots)$

# Last neat (new!) statement about the Poisson process

Rewriting history from past to future for the first particle:

Start from SF system, then the joint distribution of the red and yellow trajectories is **the same** as the FS system, i.e. TASEP with speeds  $a_1 > a_2$ .

Rewriting history is independent from the second particle (TASEP feature only)

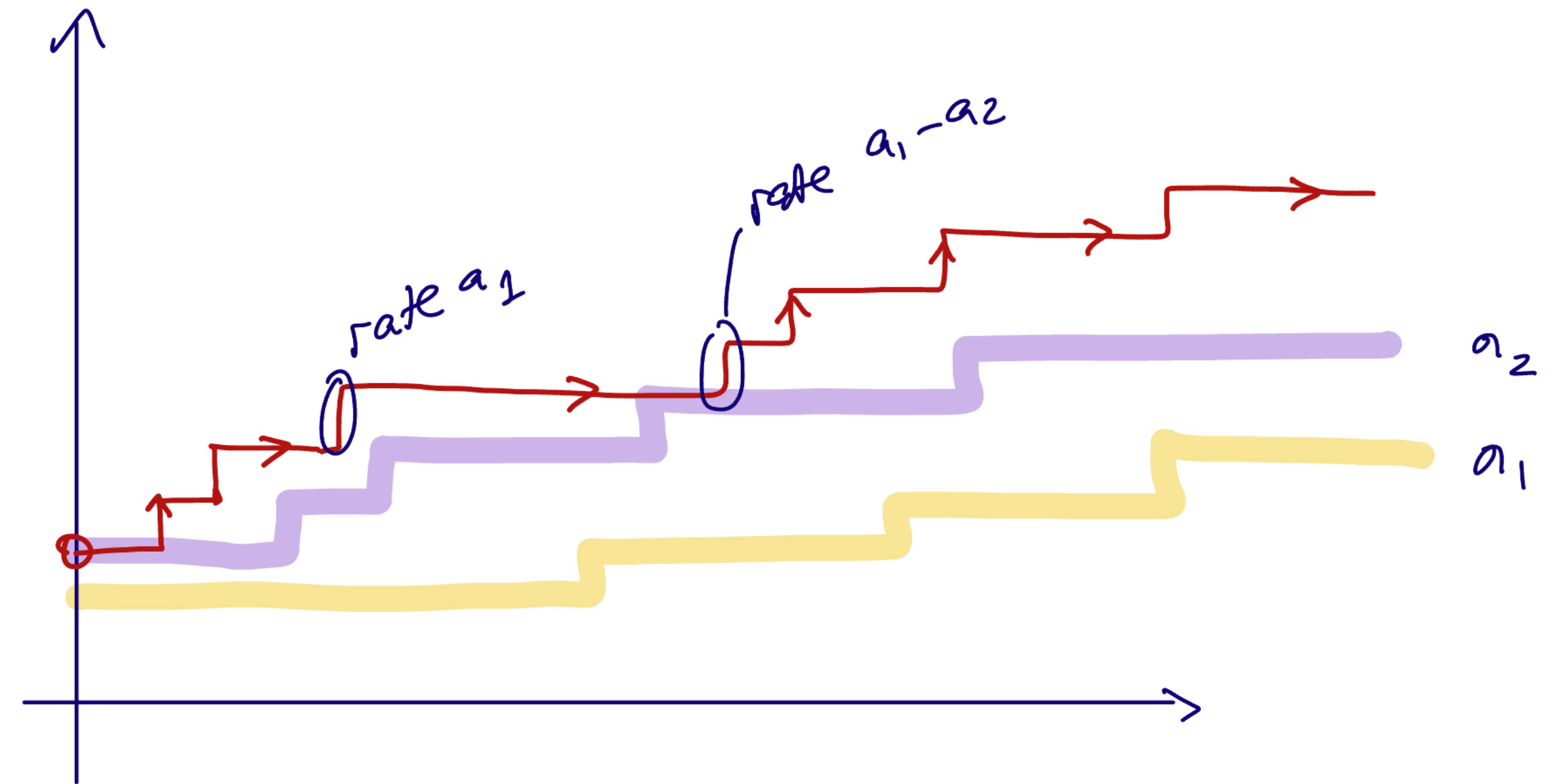


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Take a limit  $a_1, a_2 \rightarrow 1$ . Then we get a time-inhomogeneous, continuous time dynamics which increases the slope of a Poisson random walk which initially had slope 1. At time  $\tau$ , dynamics starts a new slab at a random location  $t_*$ , and then the slab has slope  $\tau + 1$ .

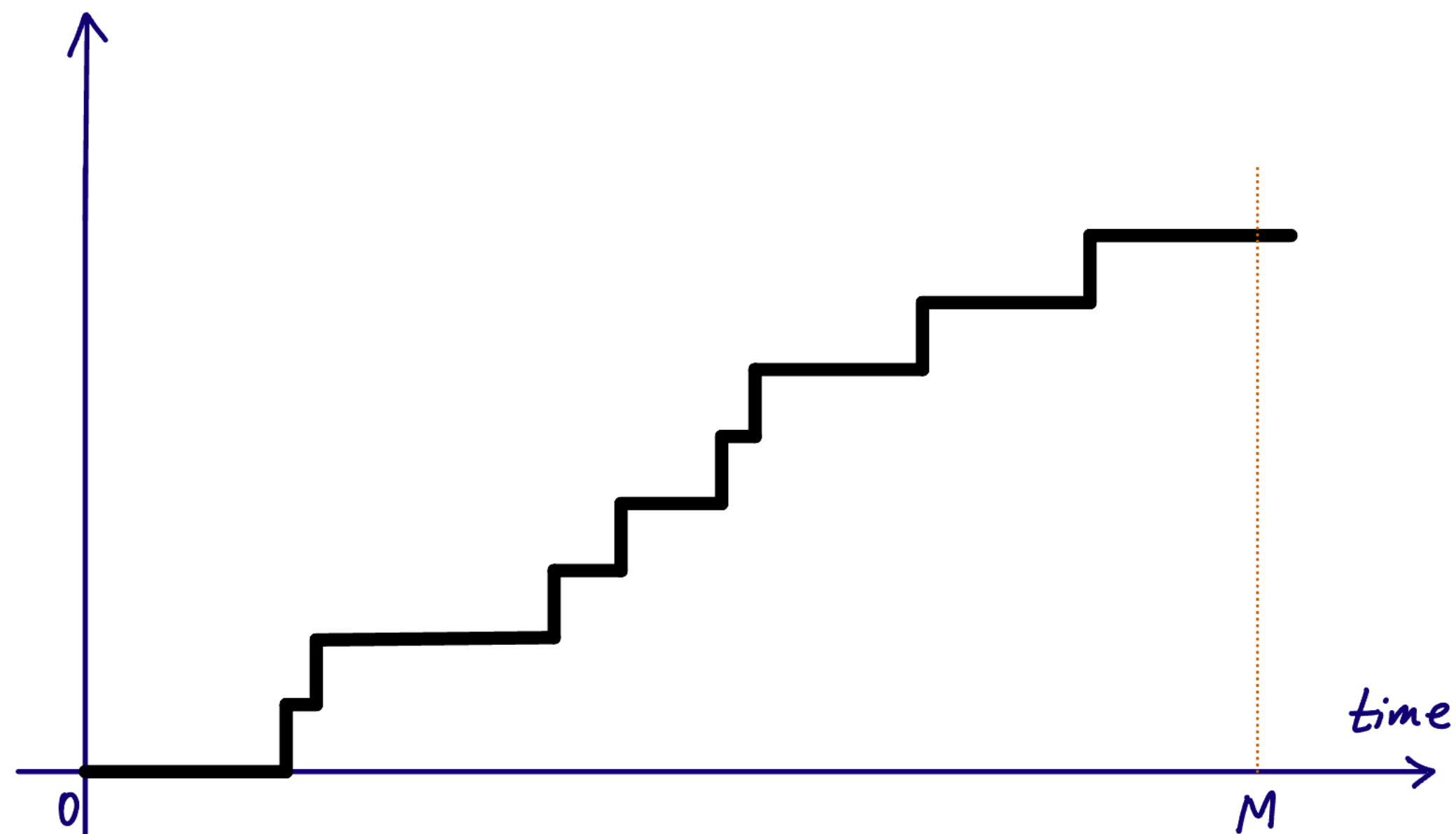
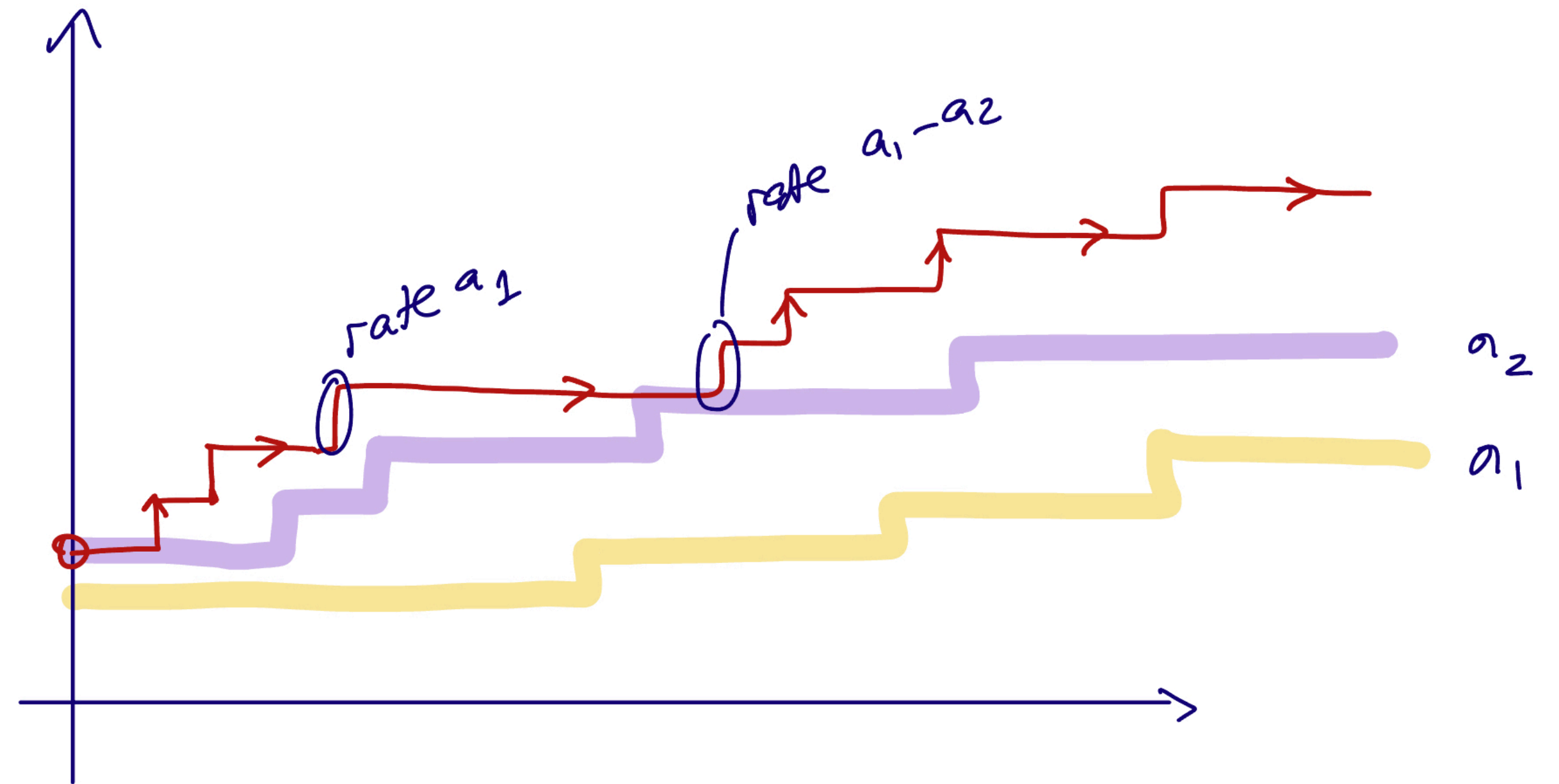
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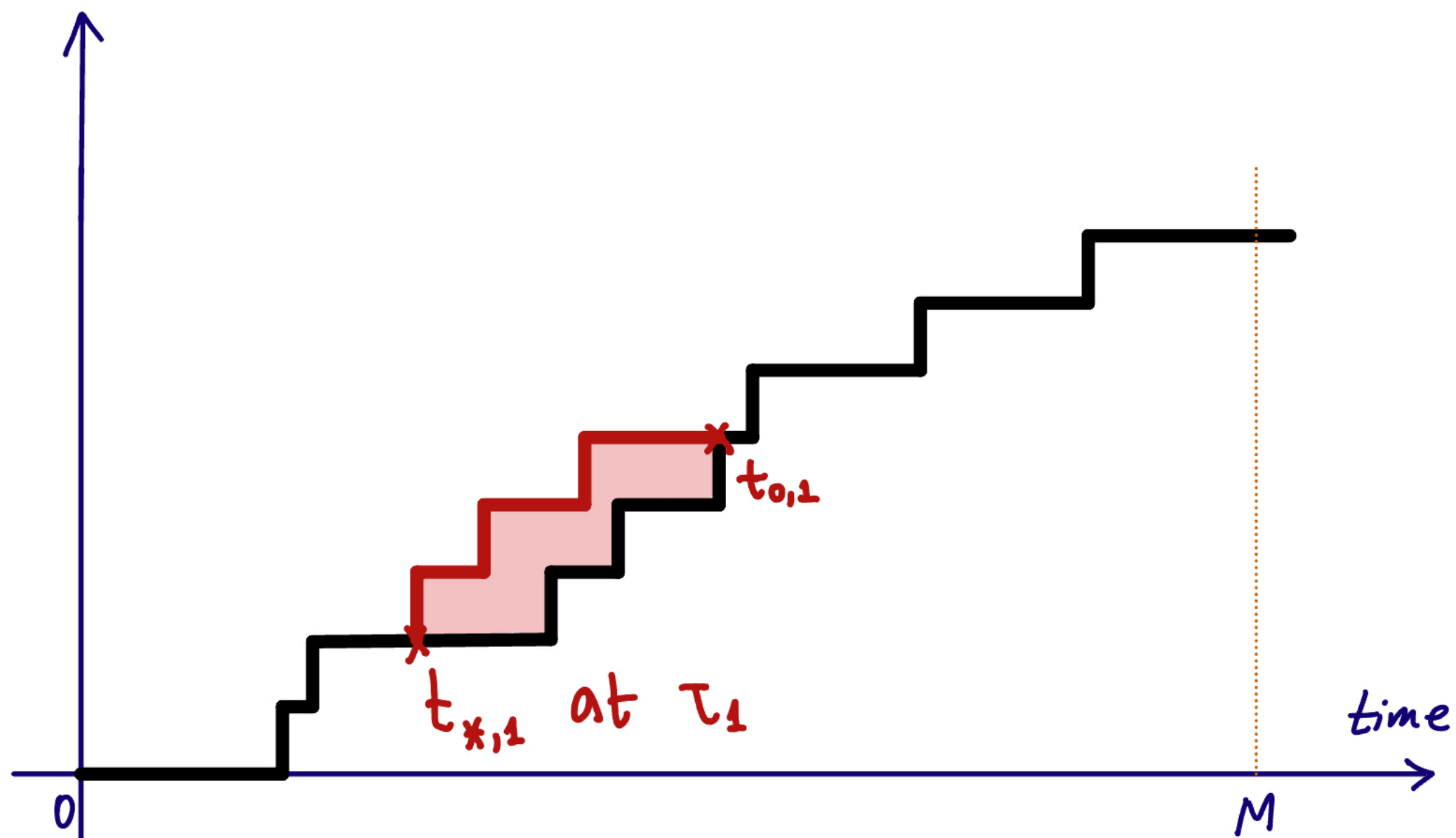
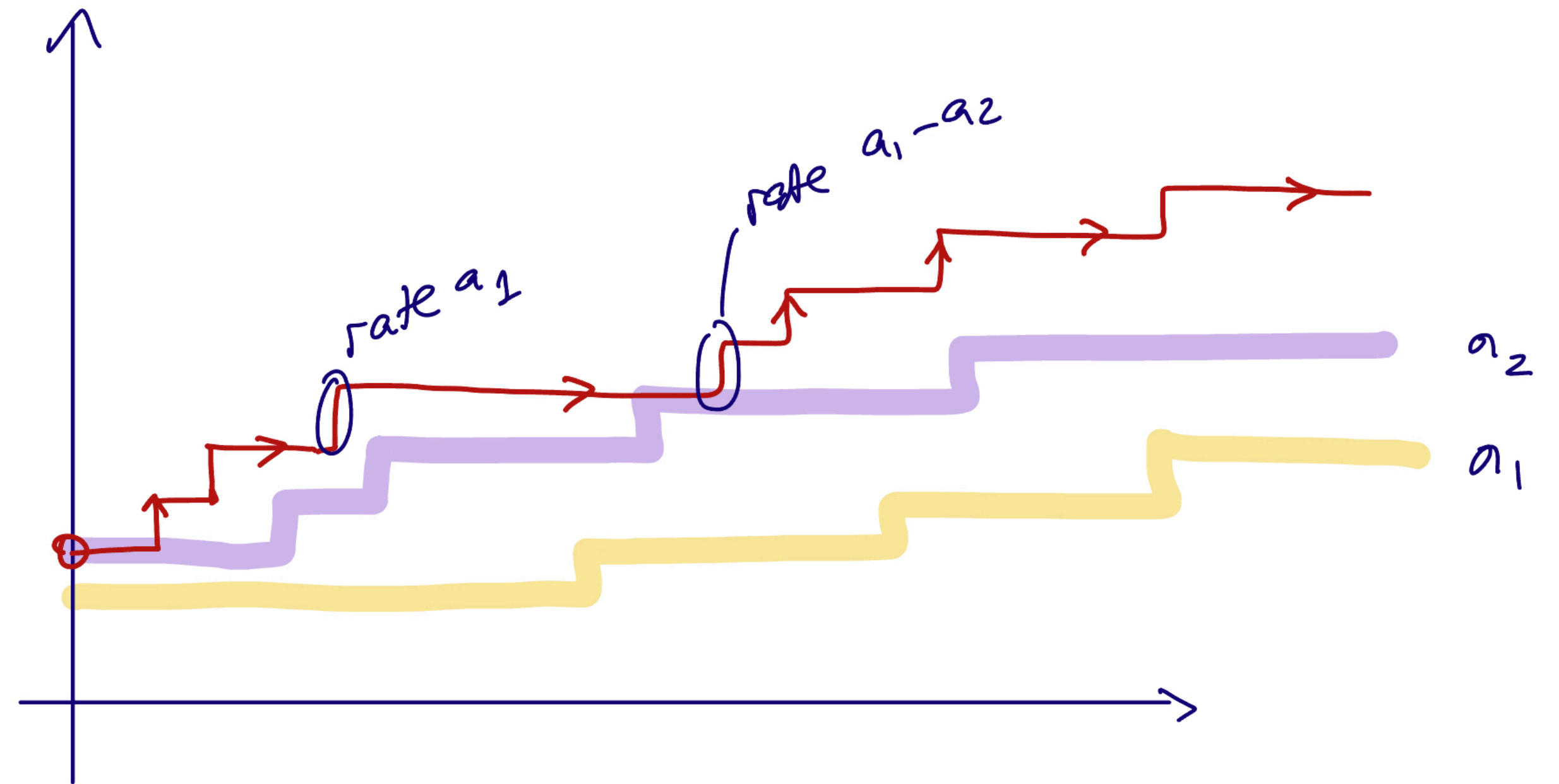
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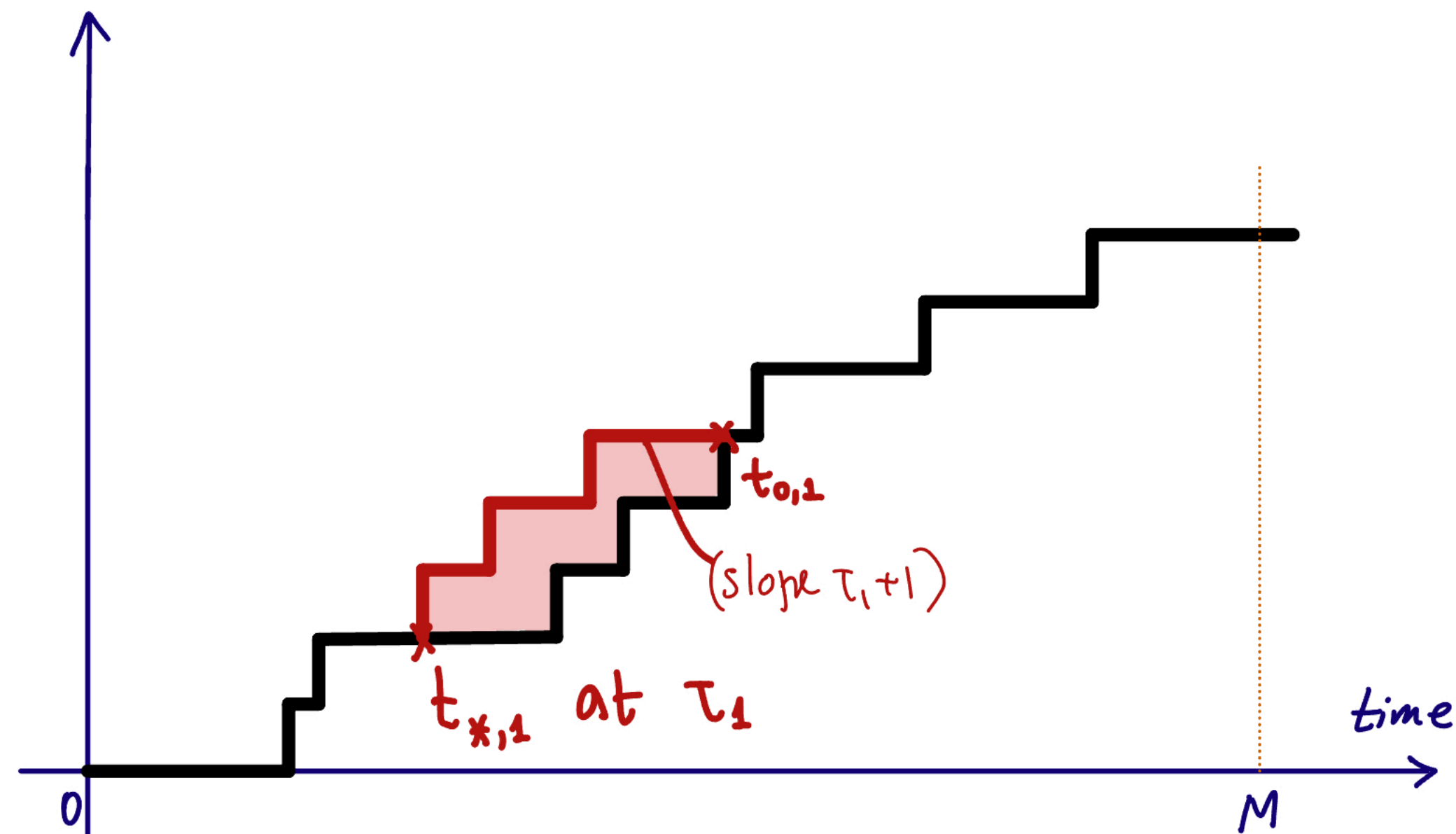
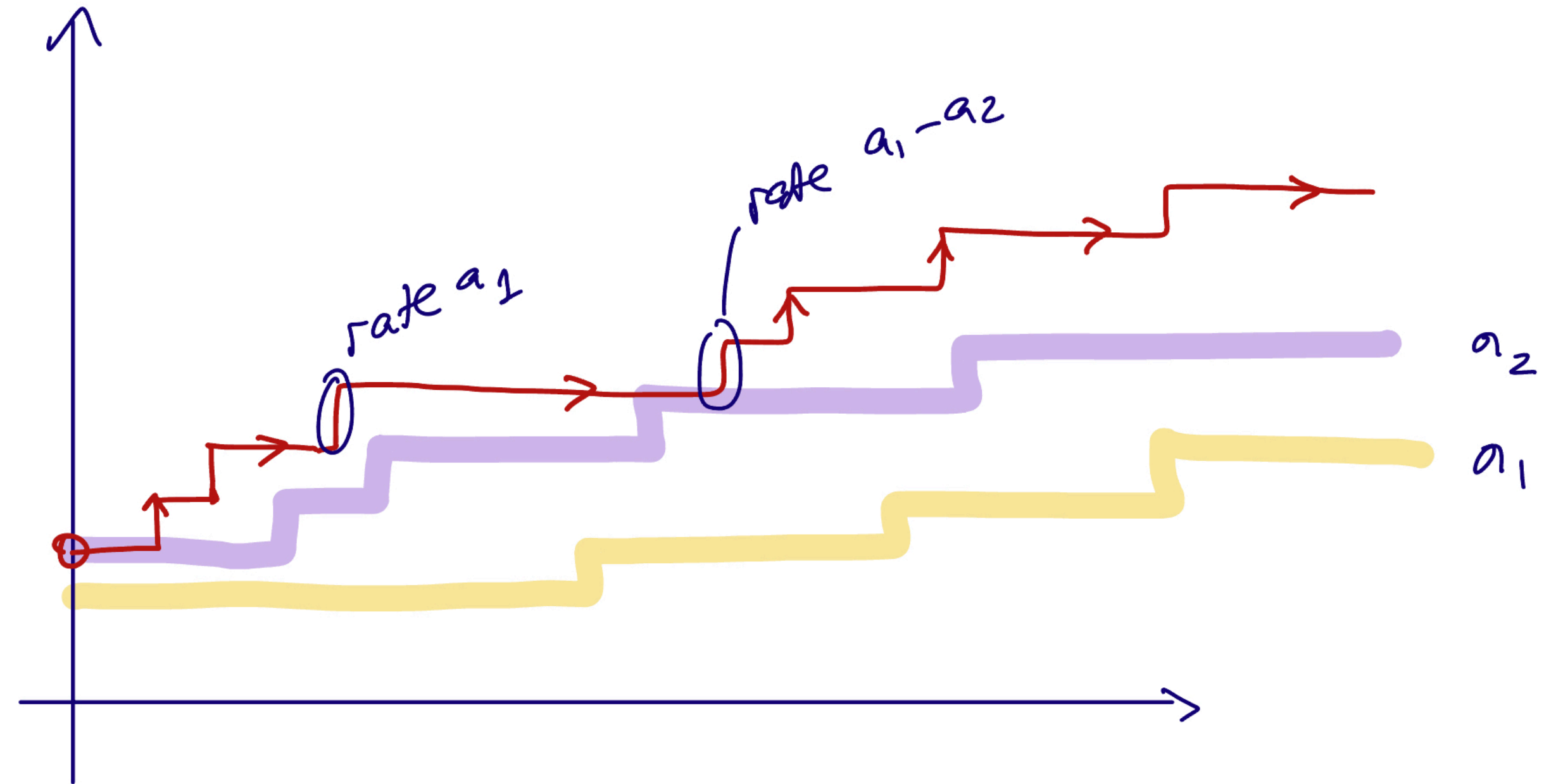
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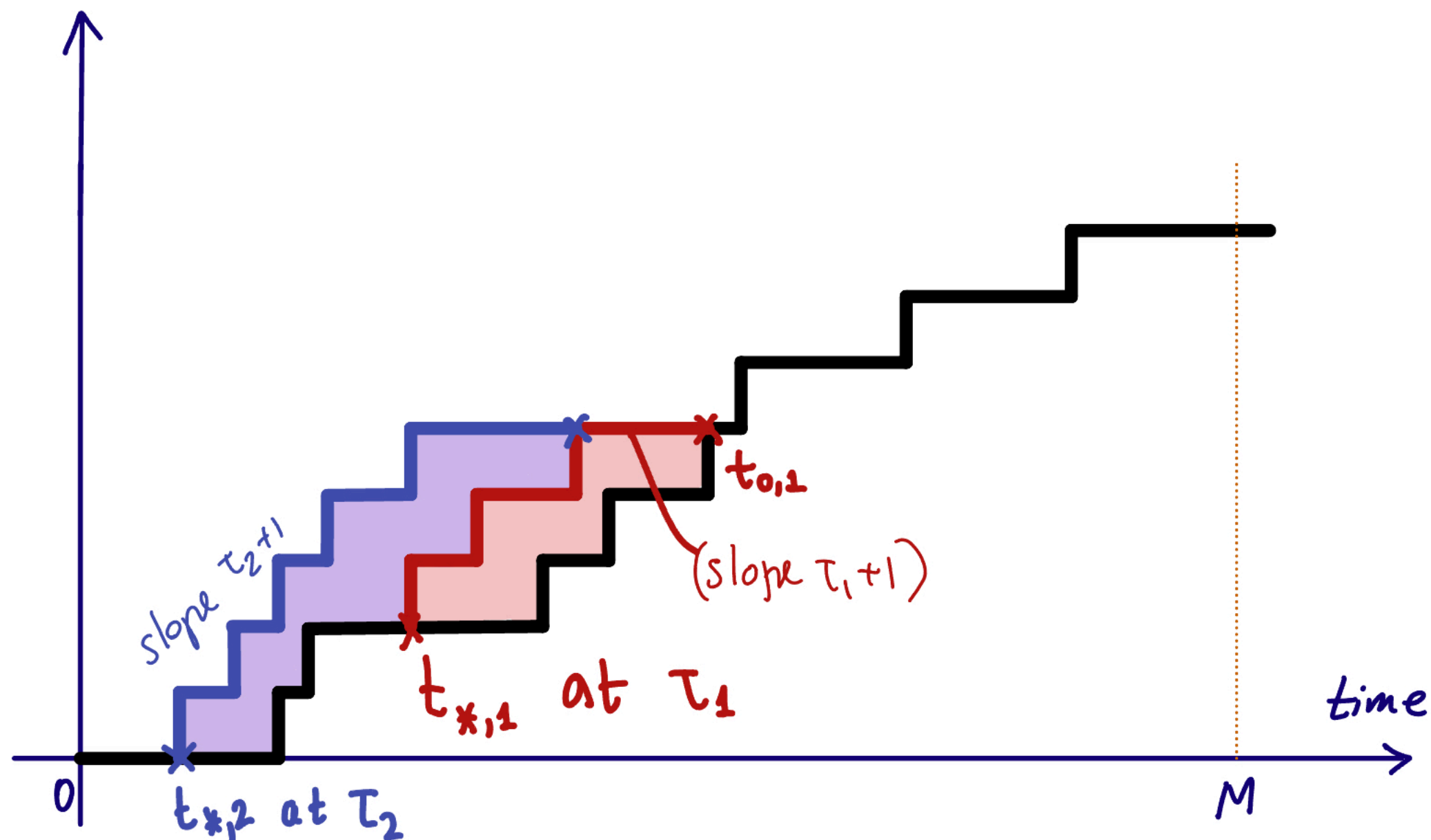
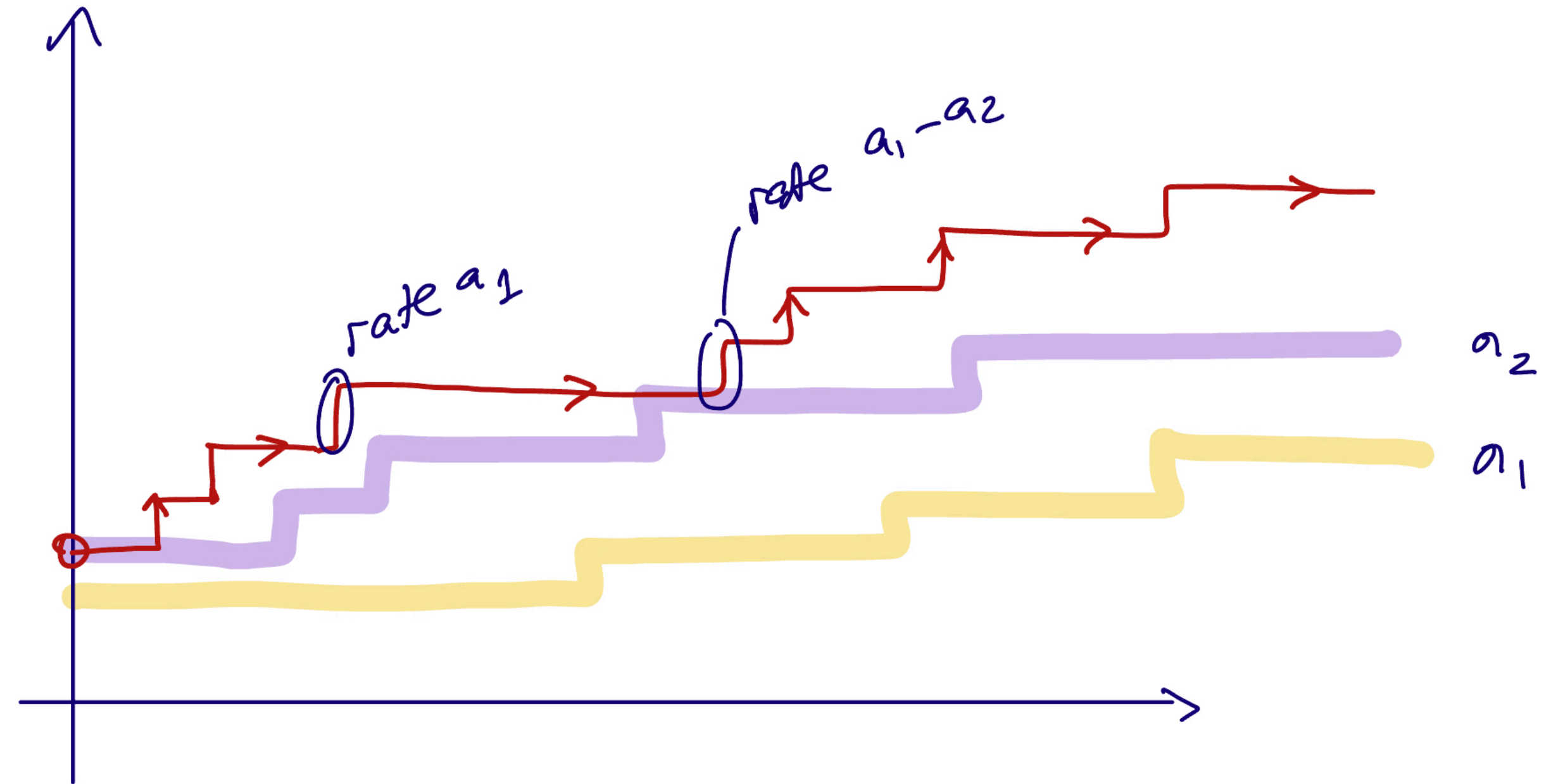
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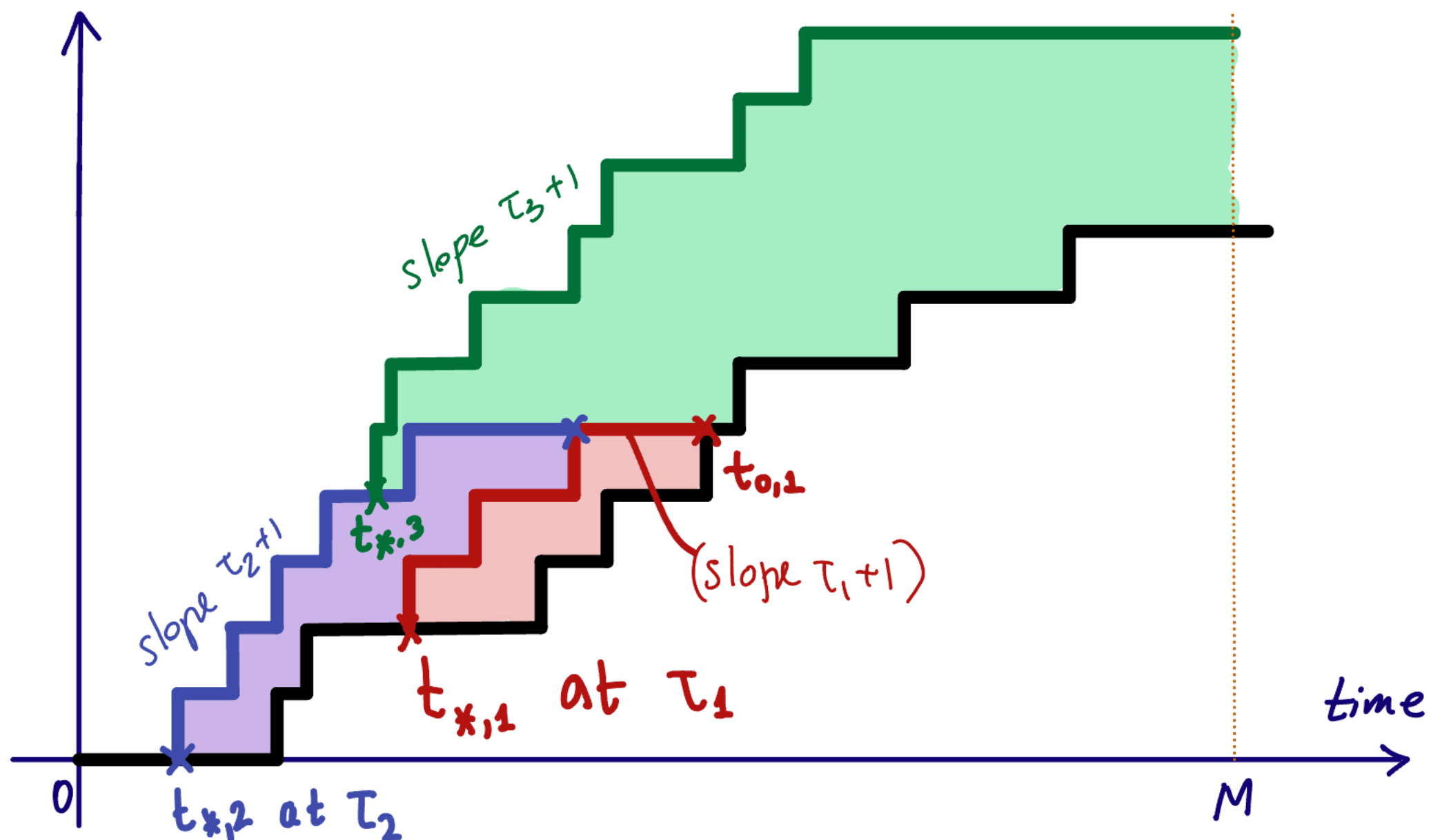
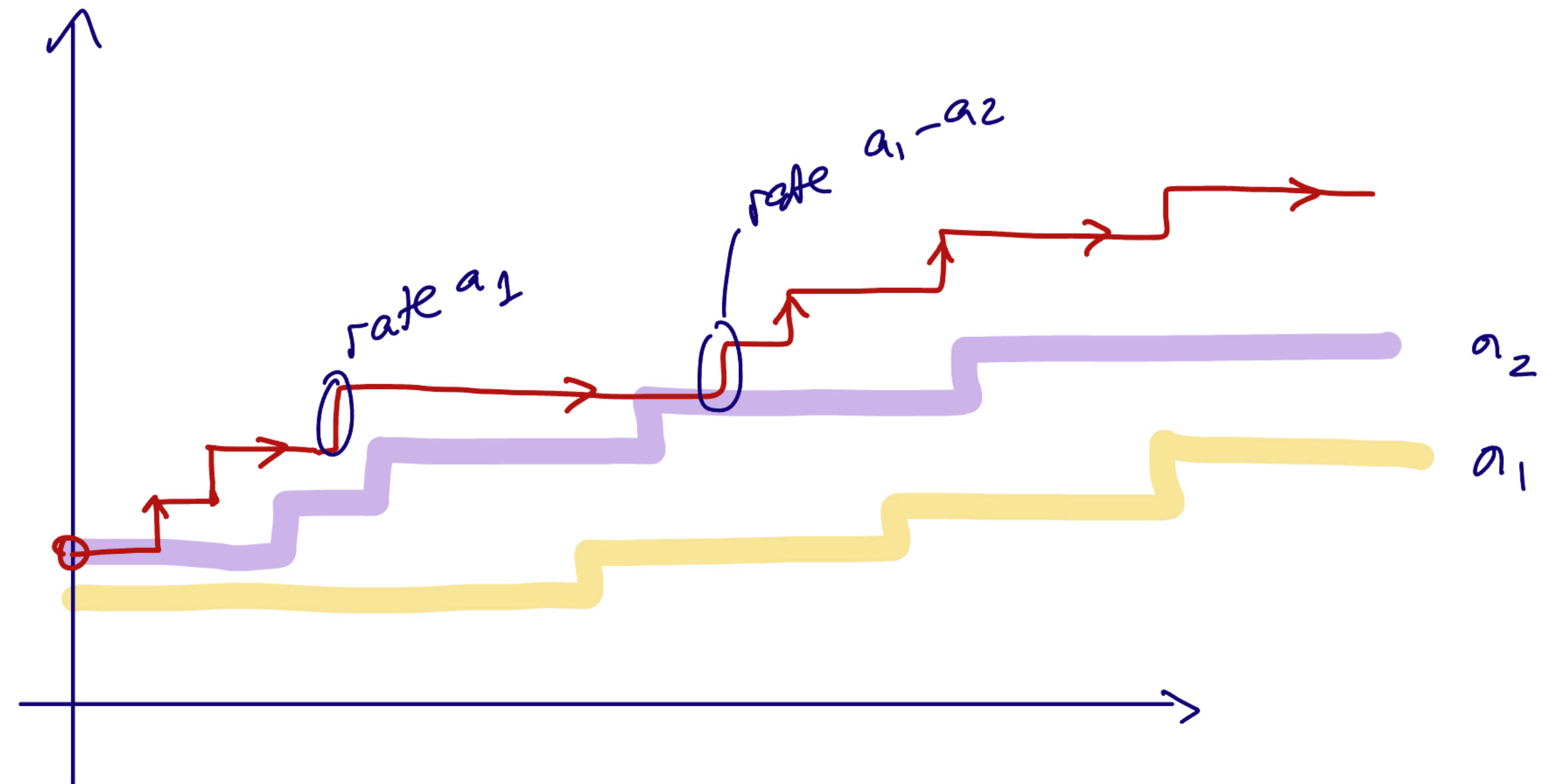


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- We discover Lax equations for  $q$ -TASEP and TASEP (their role should be better explored)
- We also construct neat couplings for the whole trajectories, and get a new result about the Poisson process



# Thank you for your attention!

In stochastic particle systems, there's a way  
To rewrite history with each passing day.  
A single particle, its fate made clear,  
Can undo what's been done and make it reappear.

The laws of probability and chaos at play  
Can be bent to our will, if we but obey.  
The deterministic systems in our control,  
Will yield to a new order, as it starts to unfold.

The particles and their interactions will dictate,  
The outcome of our systems, no matter their state.  
With the tools of integrability, we can rewrite,  
The future of our systems with a single bite.