

# Universality for optimal train fares (near the axis)

Sam McKeown  
University of Wisconsin-Madison

InterPlay between Integrable Probability and Interacting  
Particle systems  
*Université d'Angers*

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## Set-up

Consider a long train line where fares between stations are priced dynamically, varying by day. Let

$F_{k,t}$  = Fare for station <sub>$k$</sub>   $\rightarrow$  station <sub>$k+1$</sub>  on day  $t$ .

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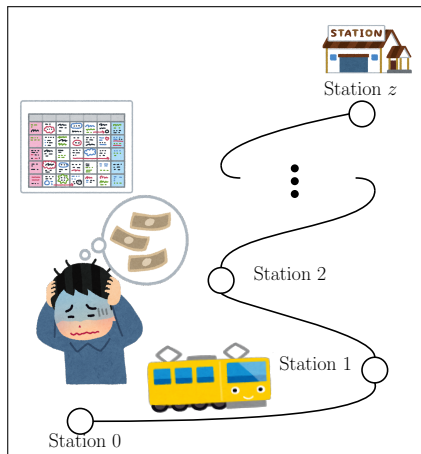
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Suppose we know all of the prices in advance. We're interested in the quantity

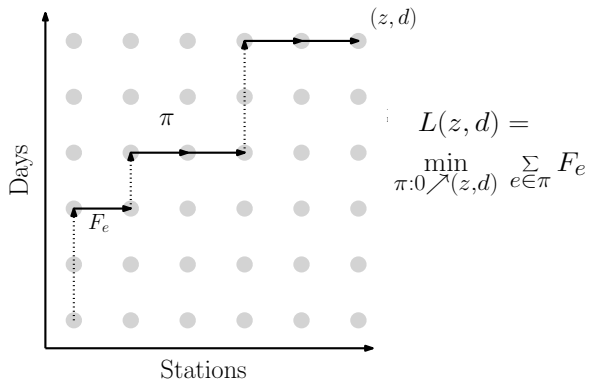
$$L(z, d) = \min_{0 \leq t_1 \leq \dots \leq t_s \leq d} \sum_{k=1}^s F_{k, t_k},$$

which is the optimal price to get to station  $z$  in at most  $d$  days.



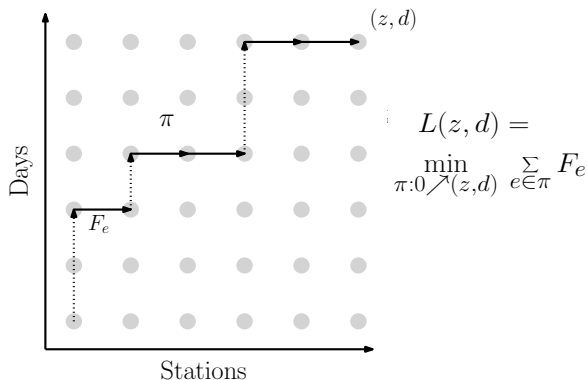
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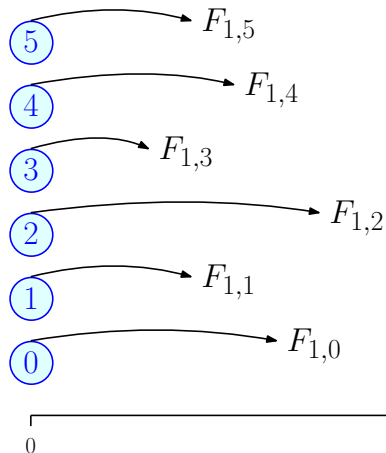


The existence of a deterministic *time constant*, how much we can expect to pay for each fare on the optimal route, is well known:

$$L(nz, nd) = n\ell(z, d) + o(n) \text{ as } n \rightarrow \infty.$$

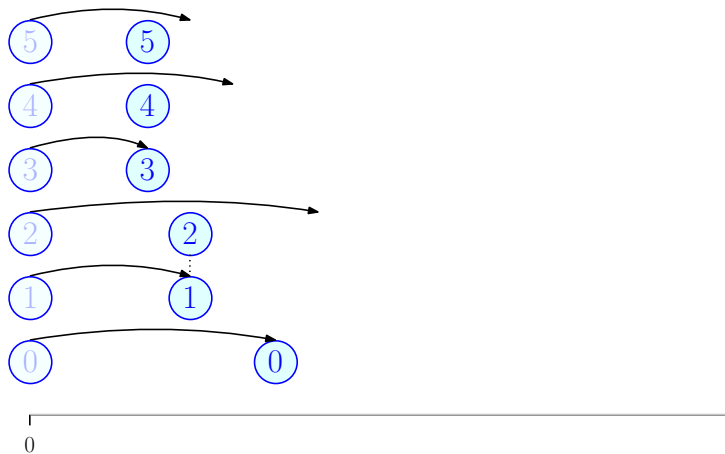
## As an IPS

There's also a mapping to the following system (which I won't explain here):



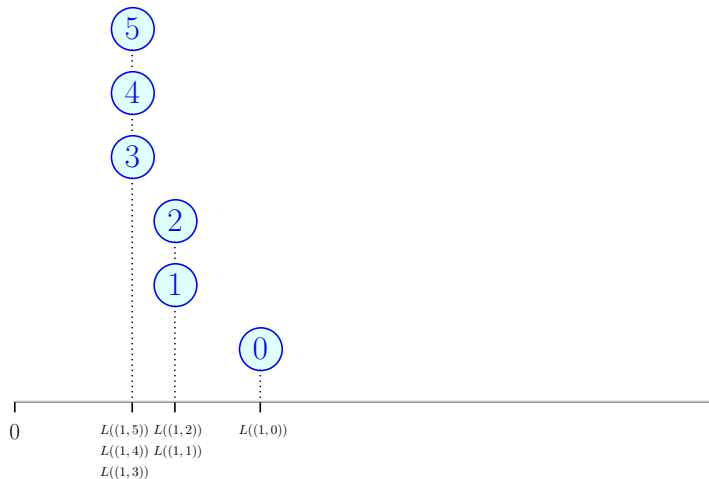
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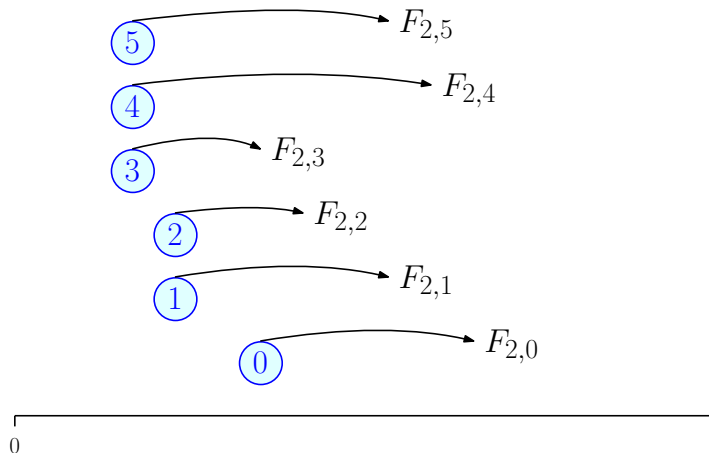
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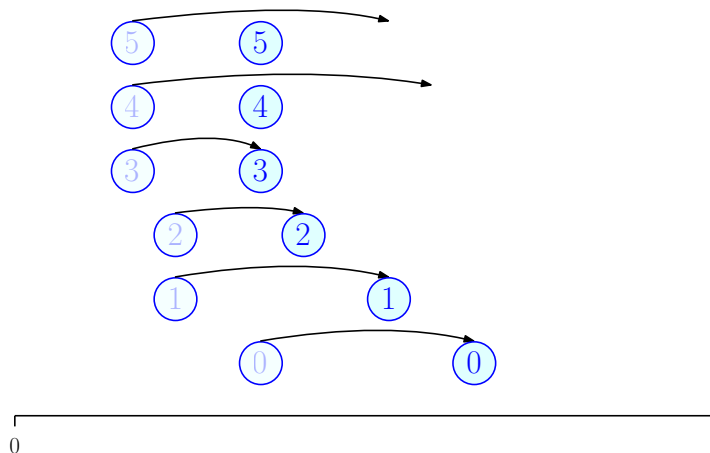
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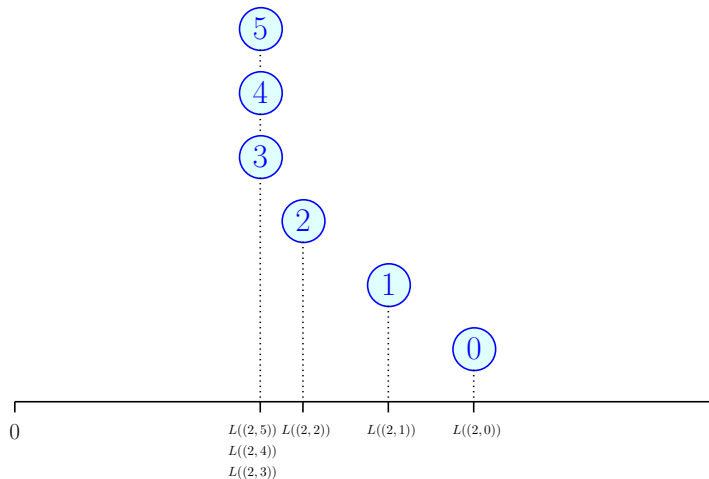
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## Many-station regime

In general we can't say much about  $\ell(z, d)$ , but two limiting regimes are tractable. Bodineau and Martin [Martin 2004; Bodineau and Martin 2005] found the leading terms of  $\ell$  and Tracy-Widom fluctuations in the *many-station* regime:

### Theorem

When  $d$  is small,

$$\ell(z, d) = z(\mu - 2\sigma\sqrt{d}) + o(\sqrt{d}),$$

and for  $0 < \alpha < 3/7$ ,

$$\frac{L(n, n^\alpha) - n\mu + 2\sigma n^{(1+\alpha)/2}}{\sigma n^{1/2-\alpha/6}} \Rightarrow TW_{GUE}.$$

Their proof of the latter goes through a Gaussian approximation and uses the KMT embedding.

## Many-day regime

The *many-days* regime is also interesting and depends on the shape of  $\lambda$  near its minimum. To state things more easily, suppose  $\lambda = \text{Unif}[0, 1]$ <sup>1</sup>

### Theorem

When  $z$  is small,

$$\ell(z, d) = \frac{z^2}{4d} + o(z^2),$$

and for  $0 \leq \alpha < 3/5$ ,

$$\frac{L(n^\alpha, n) - n^{2\alpha-1}/4}{cn^{4\alpha/3-1}} \Rightarrow TW_{GUE}.$$

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<sup>1</sup>Or any distribution with  $\min \text{supp } \lambda = 0$ , a density  $f$  near 0, with  $f(0) = 1$  and  $f$  Lipschitz on a small interval  $[0, \epsilon)$

## The integrable scaling limit

It's easy to check that

$$n \cdot \left( \min_{0 \leq t \leq n} F_{0,t} \right) \Rightarrow \text{Exp}(1).$$

This extends to a continuous time process level limit

$$\{nL(z, \lfloor nt \rfloor)\} \Rightarrow \{M(z, t)\},$$

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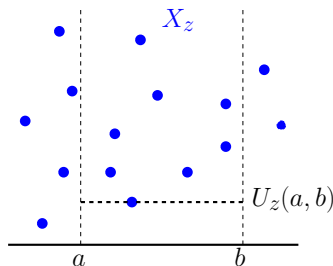
where  $M(z, t)$  is defined in terms of homogeneous Poisson processes.

Namely, for each  $z$  let  $X_z$  be an intensity 1 PPP on  $\mathbb{R} \times [0, \infty)$ , and let

$$U_z(a, b) = \text{Bottom}(X_z \cap [a, b] \times \mathbb{R}).$$

Then define

$$M(z, t) = \min_{0 \leq t_1 \leq \dots \leq t_z \leq t} \sum_{k=1}^z U_z(t_{k-1}, t_k).$$



## RMT inputs

This  $M(z, t)$  is scale invariant with  $M(z, at) \stackrel{d}{=} a^{-1} M(z, t)$  and has RMT marginals:

$$\mathbb{P}(M(z, 1) \geq x/4) = \det(I - K^{(z)} |_{L^2(0, x)}),$$

where  $K^{(z)}$  is the Bessel kernel with parameter  $z$ . This follows from the results of [Draief, Mairesse, and O'Connell 2005; Forrester 1993].



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where  $K^{(z)}$  is the Bessel kernel with parameter  $z$ . This follows from the results of [Draief, Mairesse, and O'Connell 2005; Forrester 1993]. Let  $m(z, t)$  be the time constant here. We can borrow RMT results to get:

### Proposition

For all  $z \geq 0$ ,

$$m(z, t) = \frac{z^2}{4t}$$

and

$$\frac{L(n, n) - n/4}{cn^{1/3}} \Rightarrow TW_{GUE}.$$

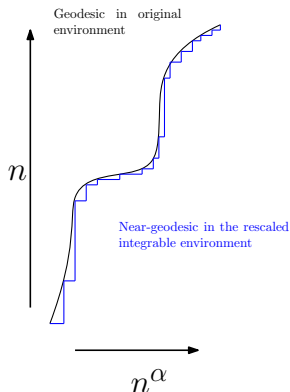
The line ensemble, Busemann process, etc. associated to  $\{M(z, t)\}$  have nice descriptions.

## Control of the geodesics

We prove the near-edge results for our train problem by producing a strong coupling to the  $M$  process which works well on thin rectangles  $[n^\alpha] \times [n]$ ,  $0 \leq \alpha < 3/5$ .

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Actually, if  $\alpha < 1/5$ , we get control of the optimal paths: with high probability they stay close to geodesics of the coupled  $M$ .

We should be able to leverage this to get the correct order of transversal fluctuations, and perhaps the Directed Landscape limit, once these are known for  $M$ .

Thanks for listening!

# References

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