

# Diffusion scaling of a limit-order book model

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nearly complete work with  
Christopher Almost  
John Lehoczky  
Xiaofeng Yu

Thera Stochastics  
In Honor of Ioannis Karatzas  
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Build a “zero-intelligence” Poisson model of the limit-order book and determine its diffusion limit.

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- ▶ Diffusion limit — Evolution of the limiting limit-order book is described in terms of Brownian motions.
- ▶ Computation of statistics – Use the limiting limit-order model to compute statistics of model dynamics. □

## Partial history

- ▶ CONT, R., STOIKOV, S. & TALREJA, R. (2010) “A stochastic model for order book dynamics,” *Operations Research* **58**, 549–563.

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Always a one tick spread and orders queue only at the best bid and best ask prices. If one of these is depleted, both move one tick and the book reinitializes. Derive the diffusion-scaled limit.

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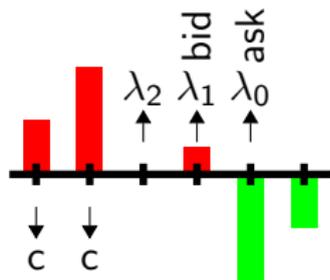
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- ▶ The pre-limit model, which the limiting model approximates, has more realistic behavior. For the parameters considered here, there is a one-tick spread 76% of the time.

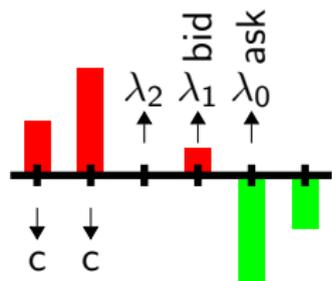
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- ▶ We present some computations in the limiting model. □

# Arrivals and cancellations of **buy** orders

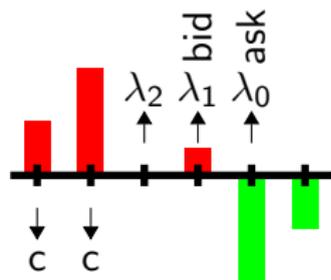


## Arrivals and cancellations of buy orders



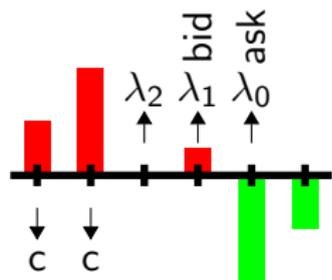
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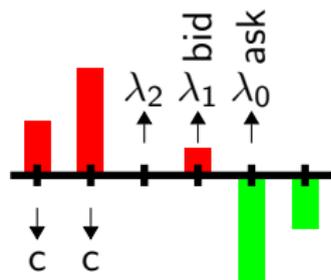
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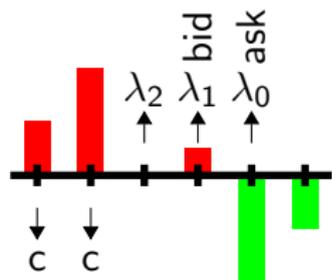
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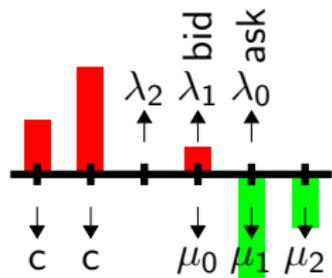
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- ▶ All processes are independent of one another. □

## Arrivals and cancellations of sell orders



- ▶ All orders are of size 1.
- ▶ Poisson arrivals of **market sells** at rate  $\mu_0$ . These execute at the best bid price.
- ▶ Poisson arrivals of **limit sells** at one and two ticks above the best bid price at rates  $\mu_1$  and  $\mu_2$ , respectively.
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## Constraints on parameters

In order to have a diffusion limit, among the **six** parameters  $\lambda_0$ ,  $\lambda_1$ ,  $\lambda_2$ ,  $\mu_0$ ,  $\mu_1$ , and  $\mu_2$ , there are **three** degrees of freedom. Let  $a$  and  $b$  be positive constants satisfying  $a + b > ab$ . Then

$$\lambda_1 = (a - 1)\lambda_0,$$

$$\lambda_2 = (a + b - ab)\lambda_0,$$

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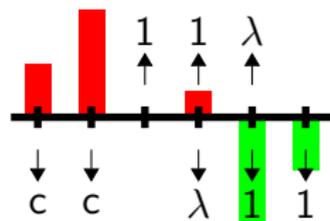
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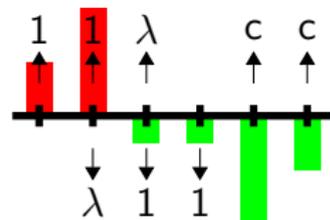
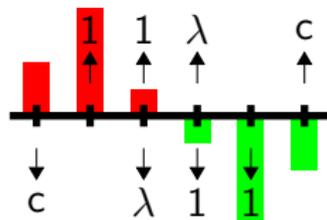
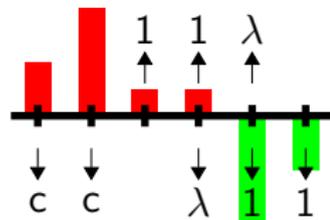
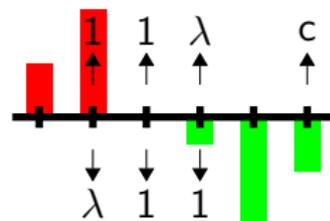
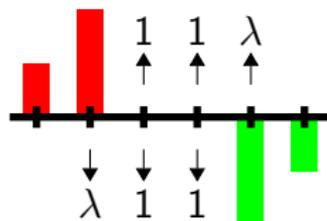
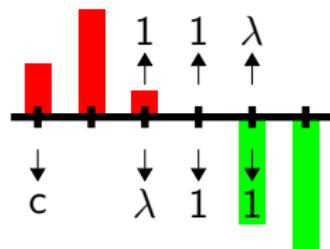
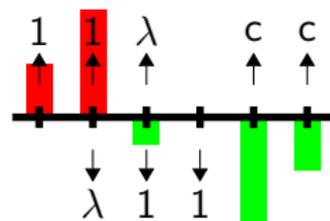
To simplify the presentation, we set

$$\begin{aligned}\lambda_1 &= \lambda_2 = \mu_1 = \mu_2 = 1, \\ \lambda_0 &= \mu_0 = \lambda := (1 + \sqrt{5})/2.\end{aligned}$$

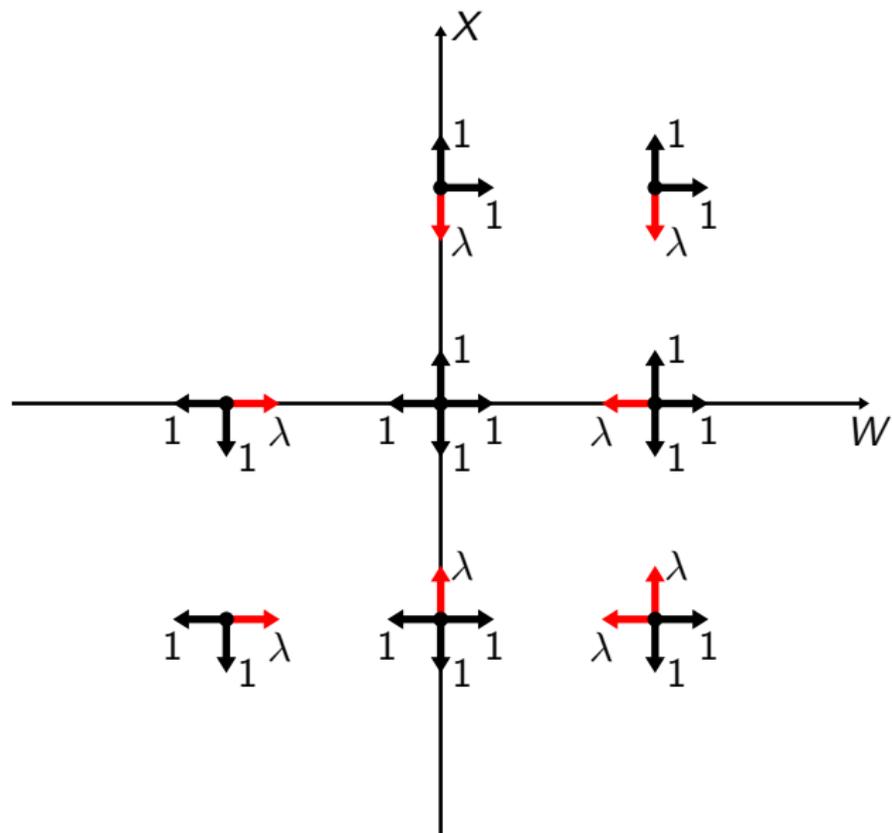
# Limit-order book arrivals and departures



*U V W X Y Z*



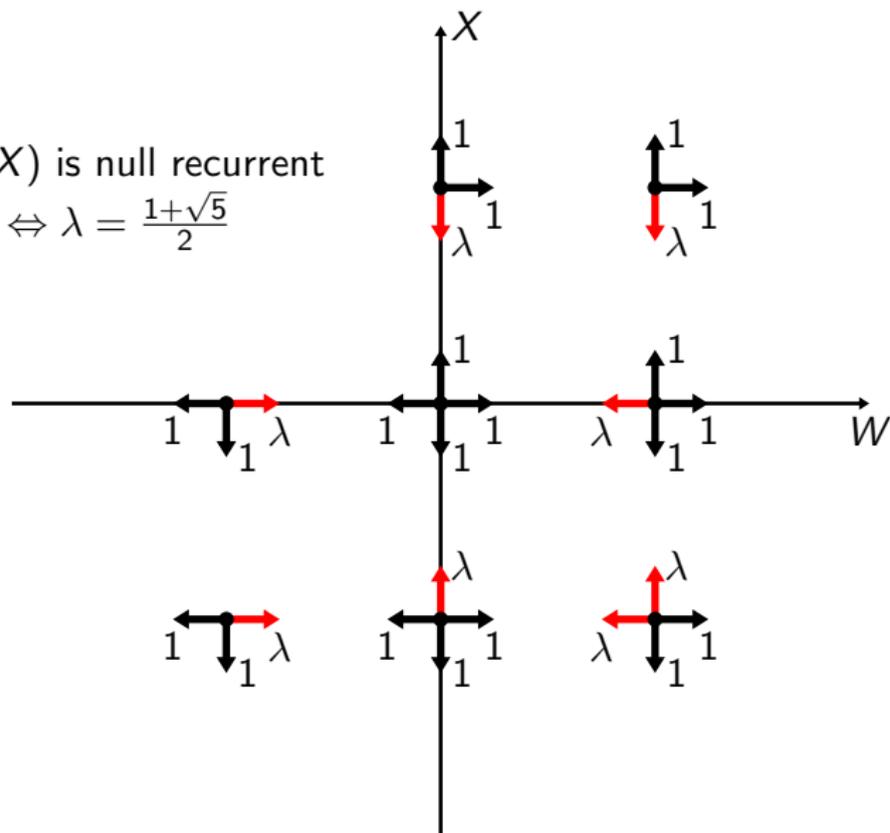
## Transitions of $(W, X)$



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$(W, X)$  is null recurrent

$$\Leftrightarrow \lambda = \frac{1+\sqrt{5}}{2}$$



# Split Brownian motion

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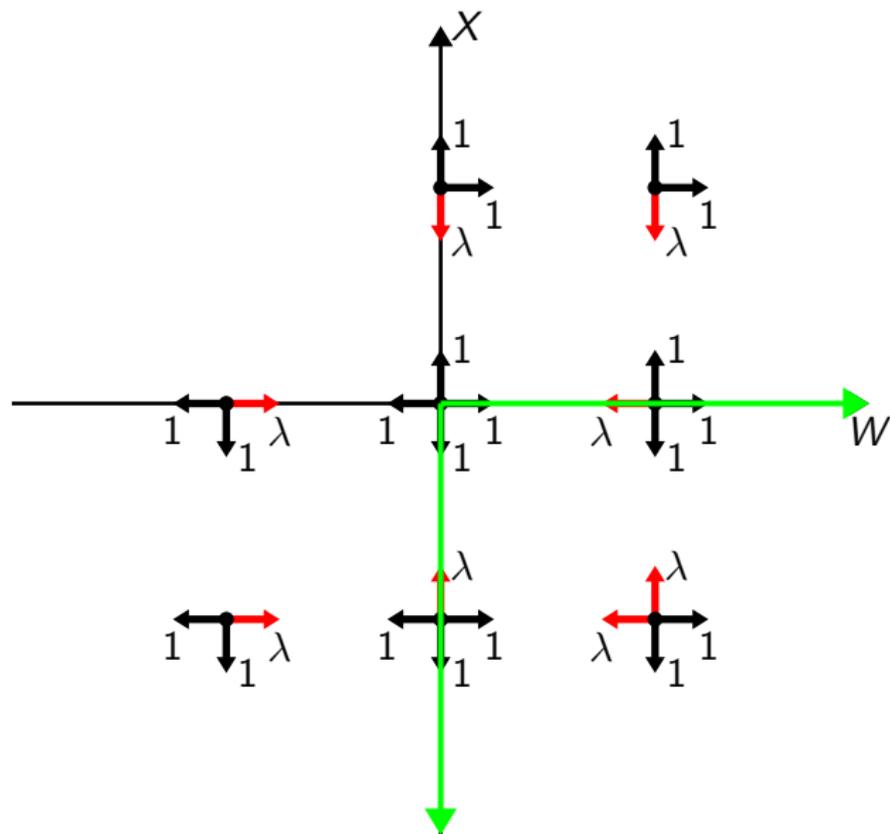
## Theorem

*Conditional on the bracketing processes  $V$  and  $Y$  remaining nonzero,  $(\widehat{W}^n, \widehat{X}^n)$  converges in distribution to a **split Brownian motion***

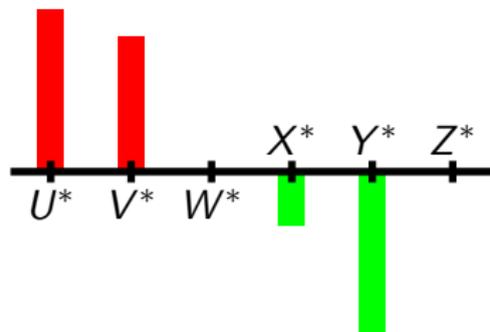
$$(W^*, X^*) = (\max\{G^*, 0\}, \min\{G^*, 0\}),$$

*where  $G^*$  is a one-dimensional Brownian motion with variance  $4\lambda$  per unit time.* □

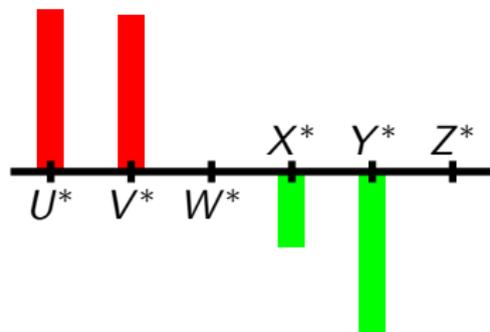
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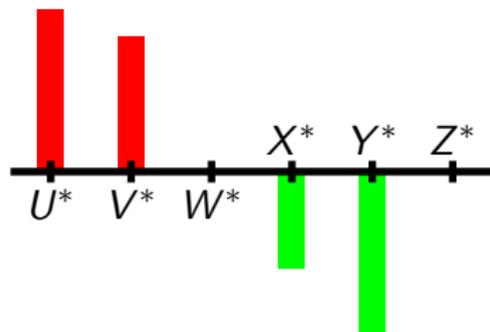
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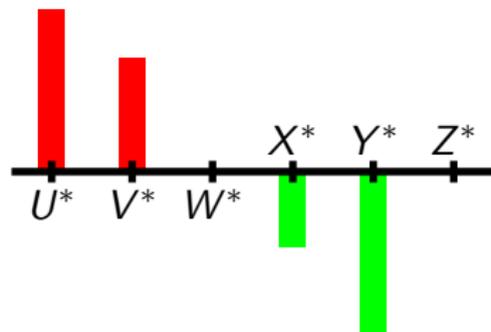
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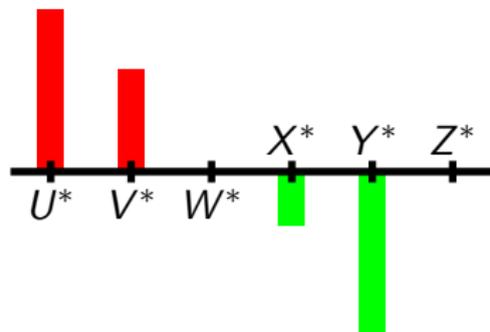
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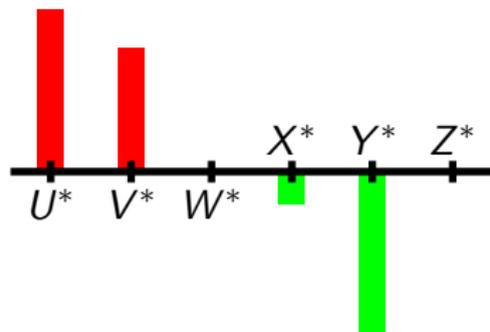
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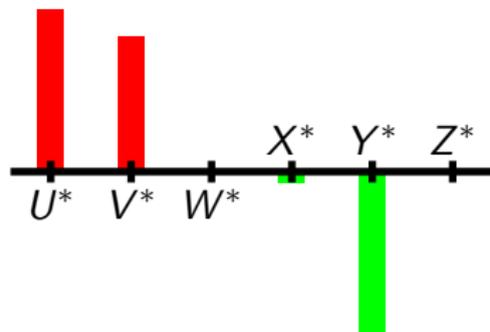
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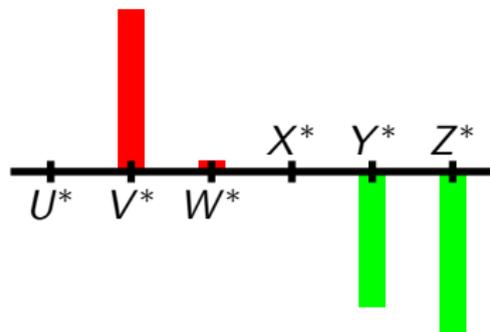
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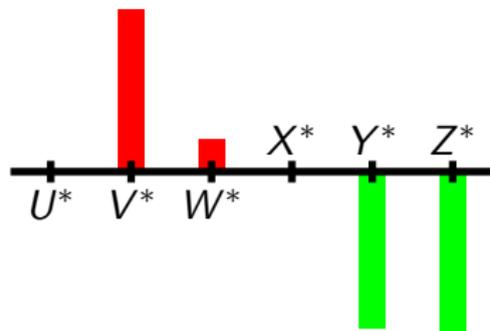
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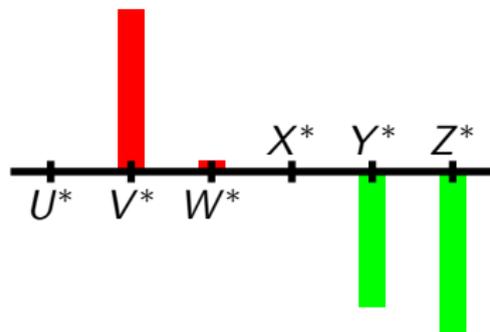
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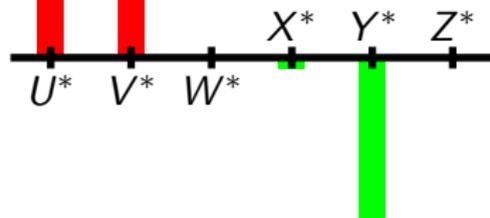
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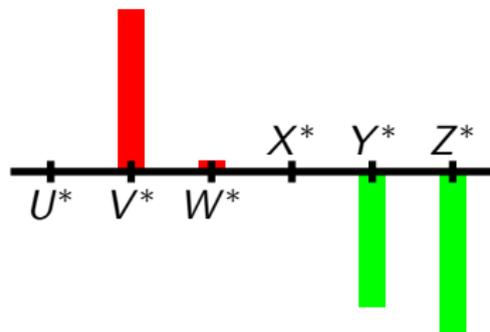
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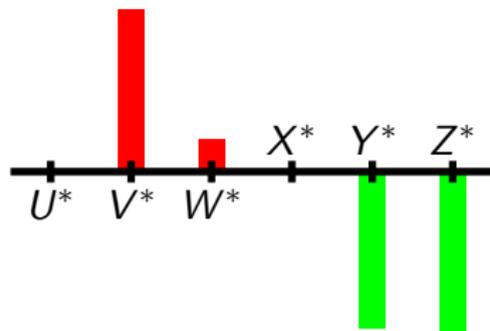
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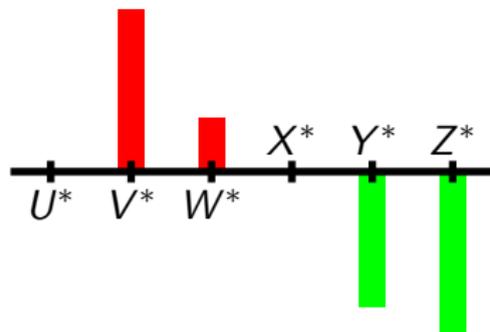
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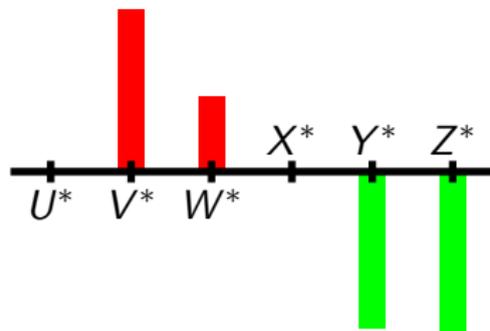
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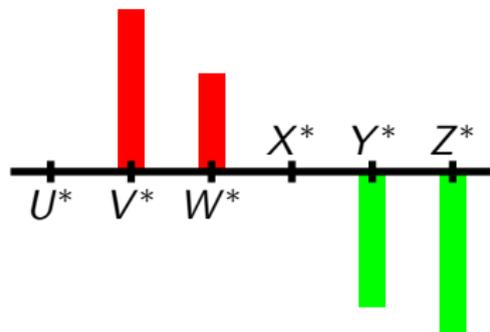
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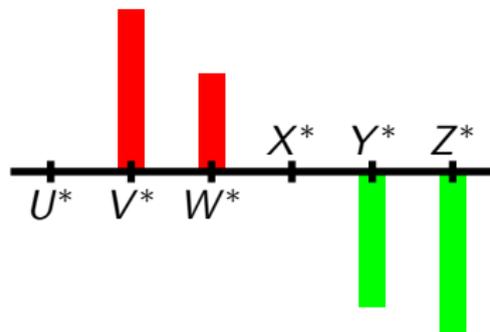
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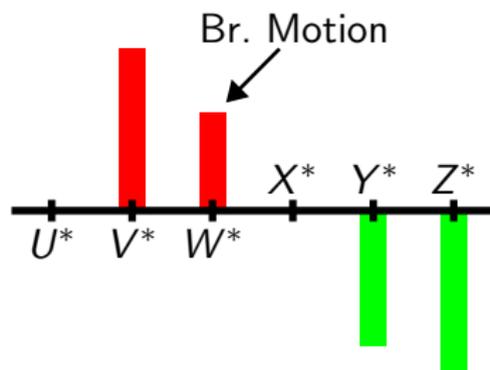
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# The other queues

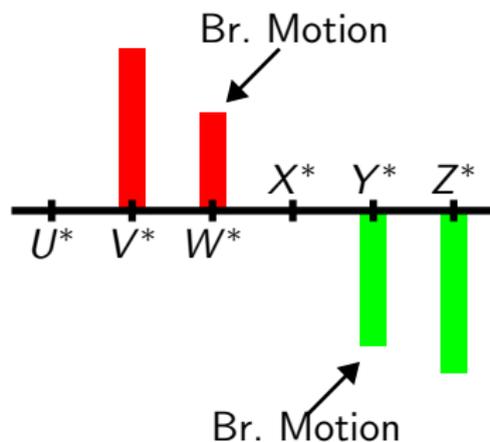


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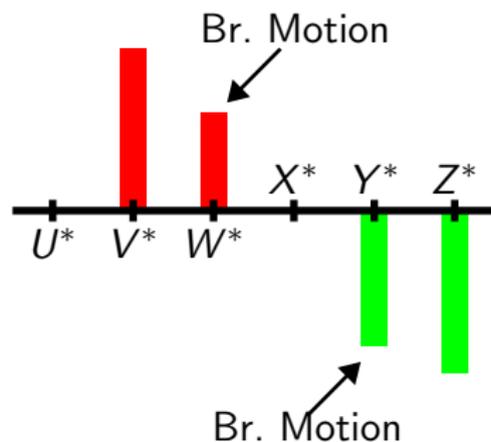
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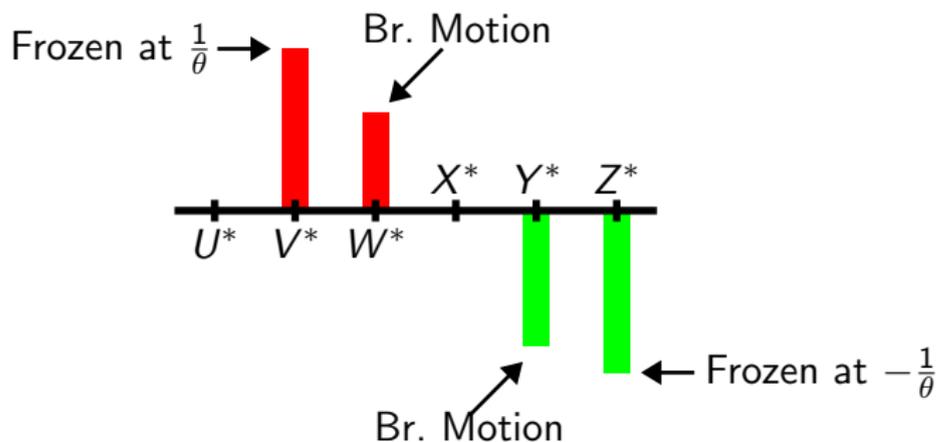
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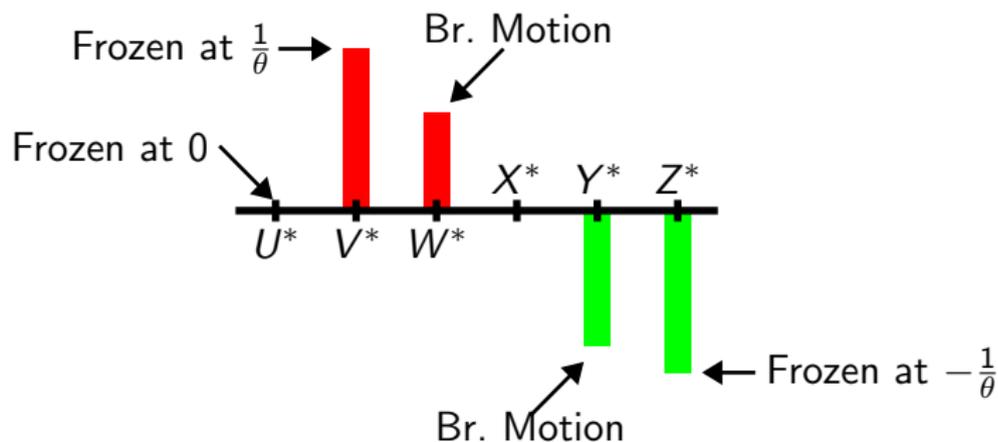
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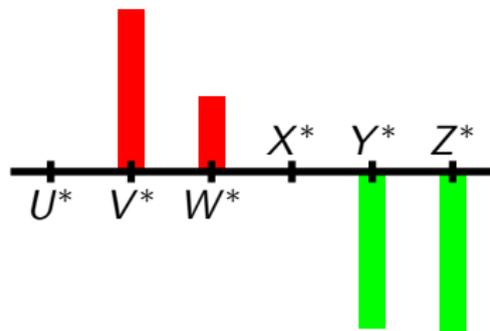
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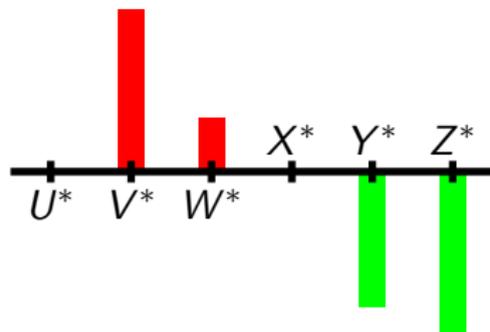
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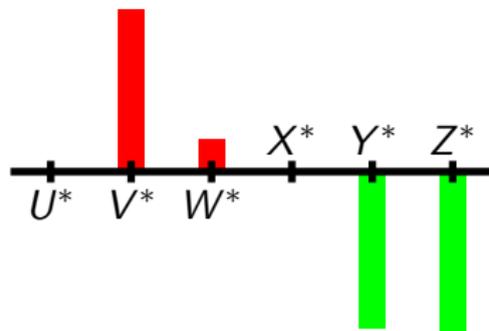
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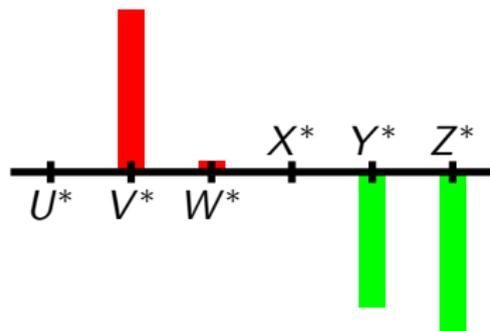
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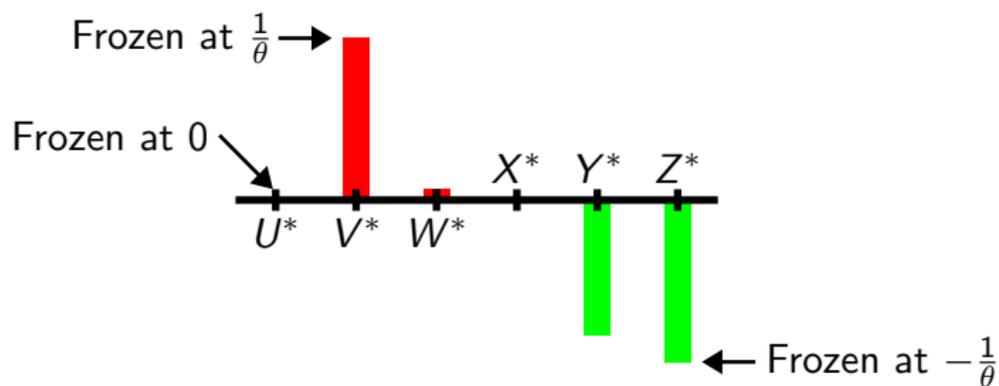
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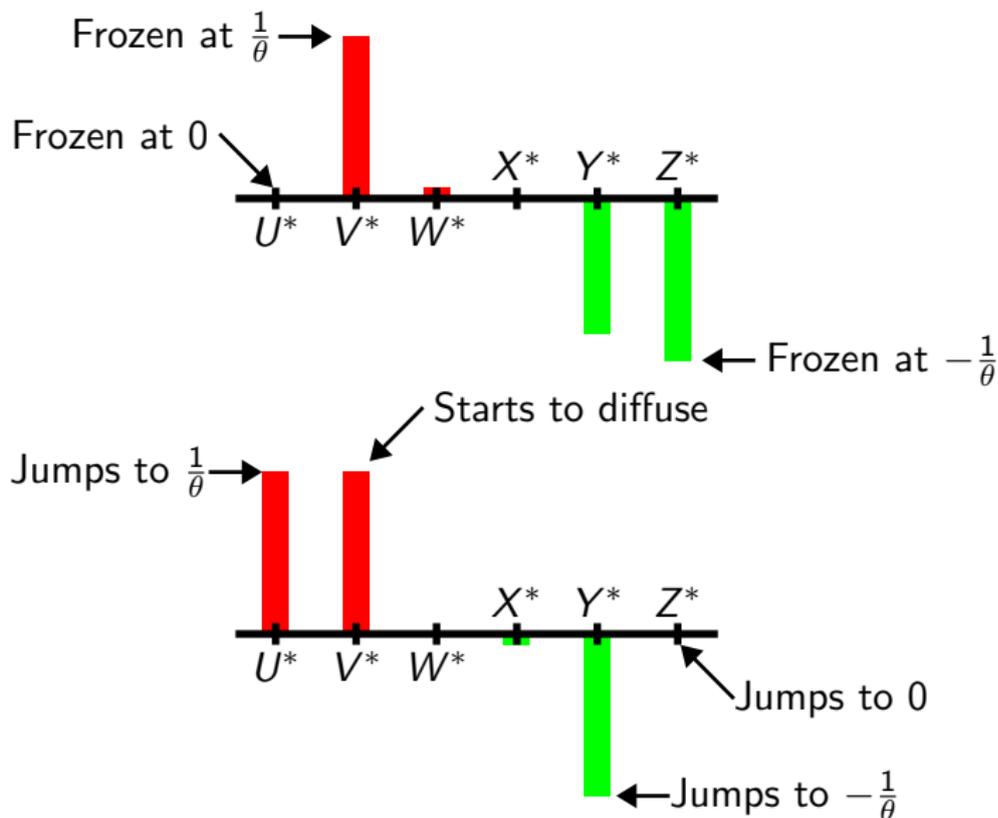
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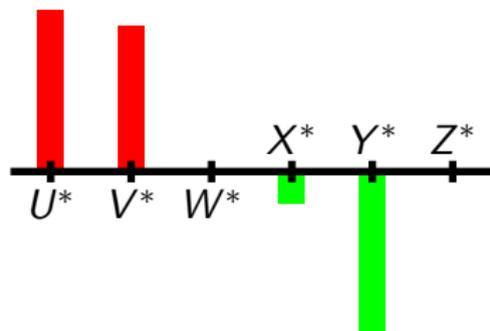
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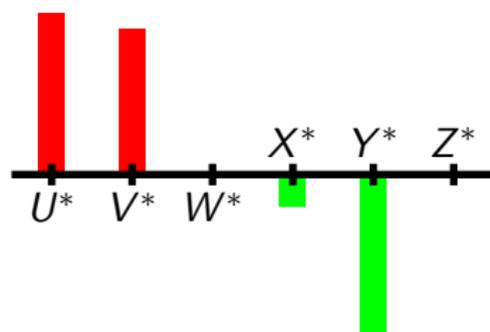


## The other queues



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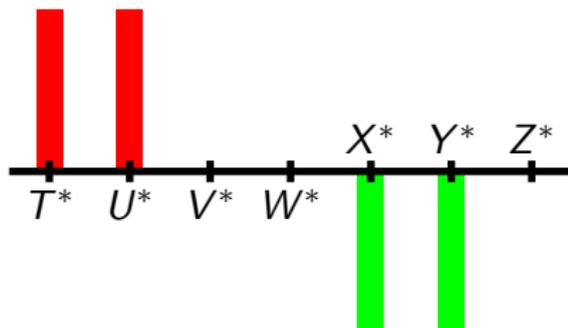
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A. METZLER, *Stat. & Probab. Letters*, 2010: “On the first passage problem for correlated Brownian motion.”



# The other queues

Suppose  $V^*$  wins.



- ▶ Reset the “bracketing processes” to be  $U^*$  and  $X^*$ .
- ▶  $(V^*, W^*)$  begins executing a split Brownian motion. □

# Snapped Brownian motion

Let's consider the  $V^*$  process in more detail.

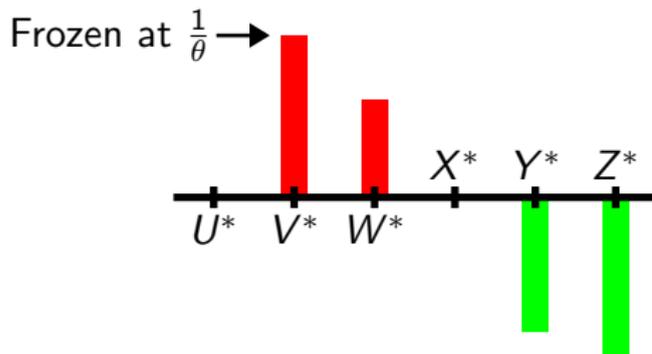
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Let's consider the  $V^*$  process in more detail.

As long as the “bracketing processes”  $V^*$  and  $Y^*$  remain nonzero,  $(W^*, X^*)$  executes a split Brownian motion:

$$(W^*, X^*) = (\max\{G^*, 0\}, \min\{G^*, 0\}),$$

where  $G^*$  is a one-dimensional Brownian motion with variance  $4\lambda$  per unit time.

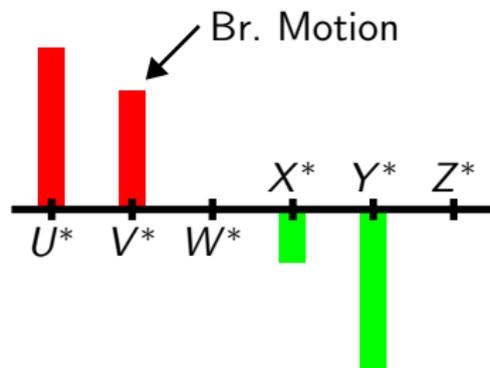


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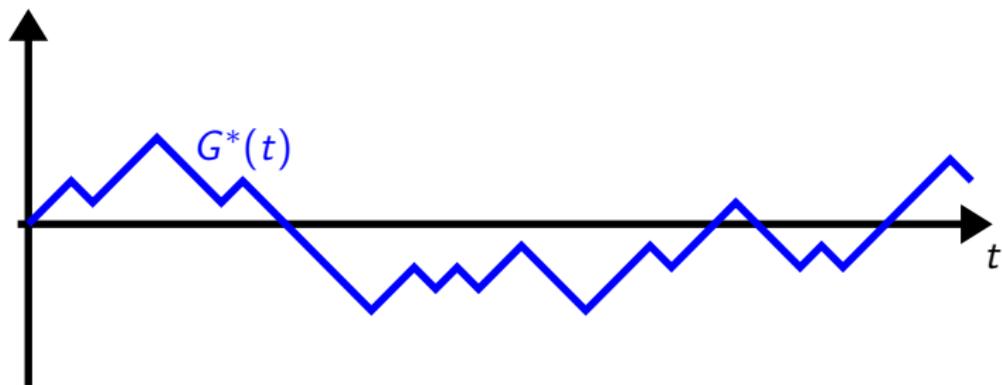
Still have the split Brownian motion,

$$(W^*, X^*) = (\max\{G^*, 0\}, \min\{G^*, 0\}),$$

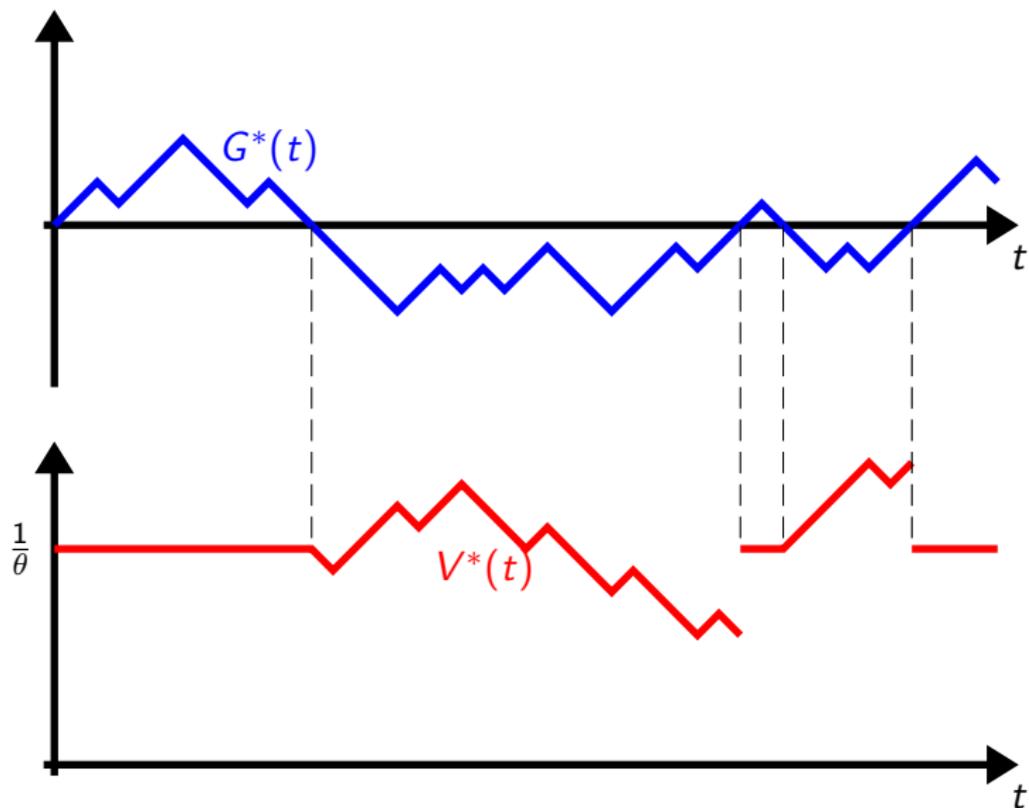
but now  $V^*$  is diffusing.



# Snapped Brownian motion



# Snapped Brownian motion



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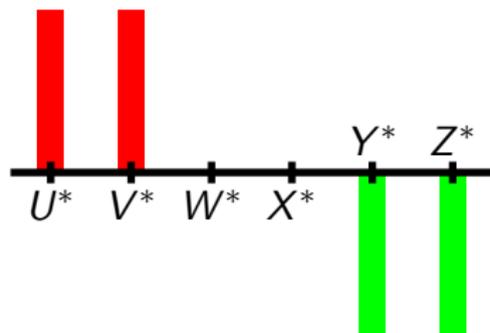
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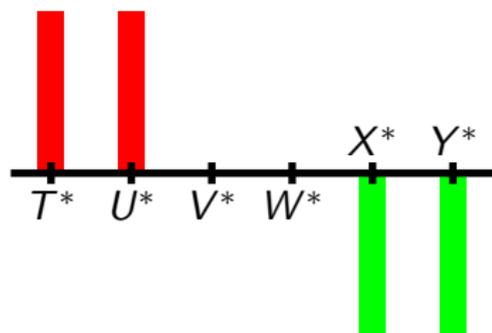
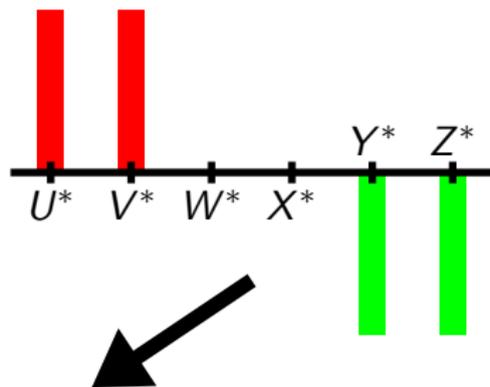
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- ▶ We transition through the three-tick spread using the concept of a snapped Brownian motion. □

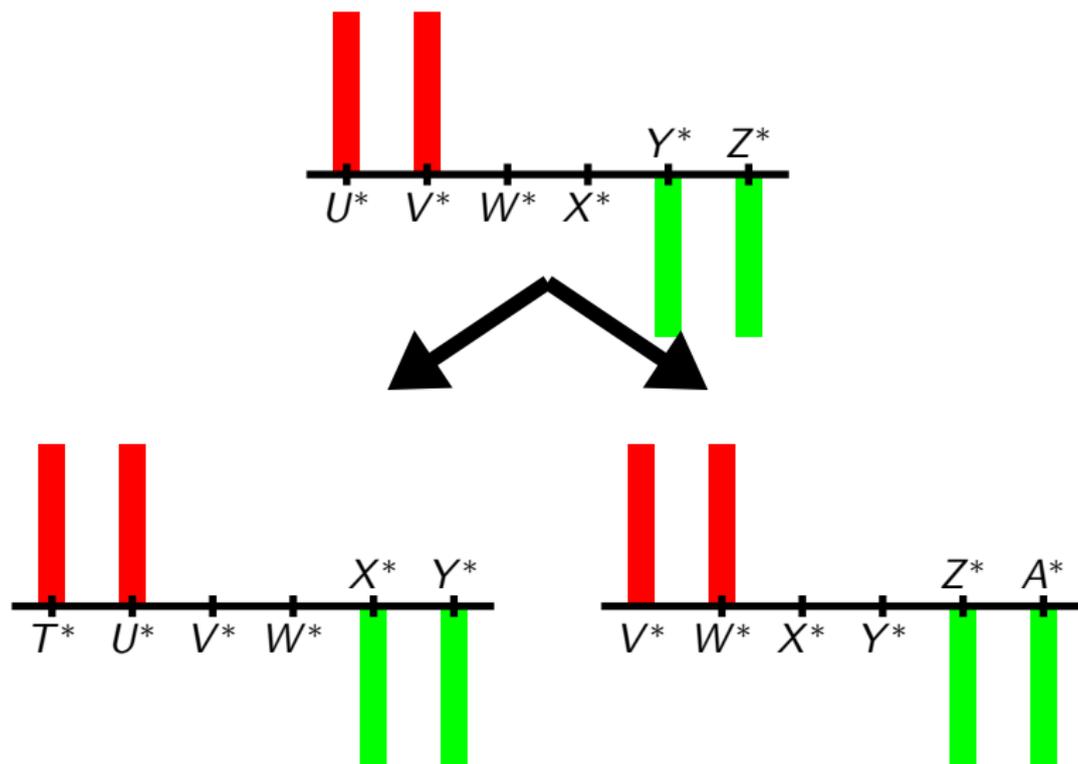
## Renewal states



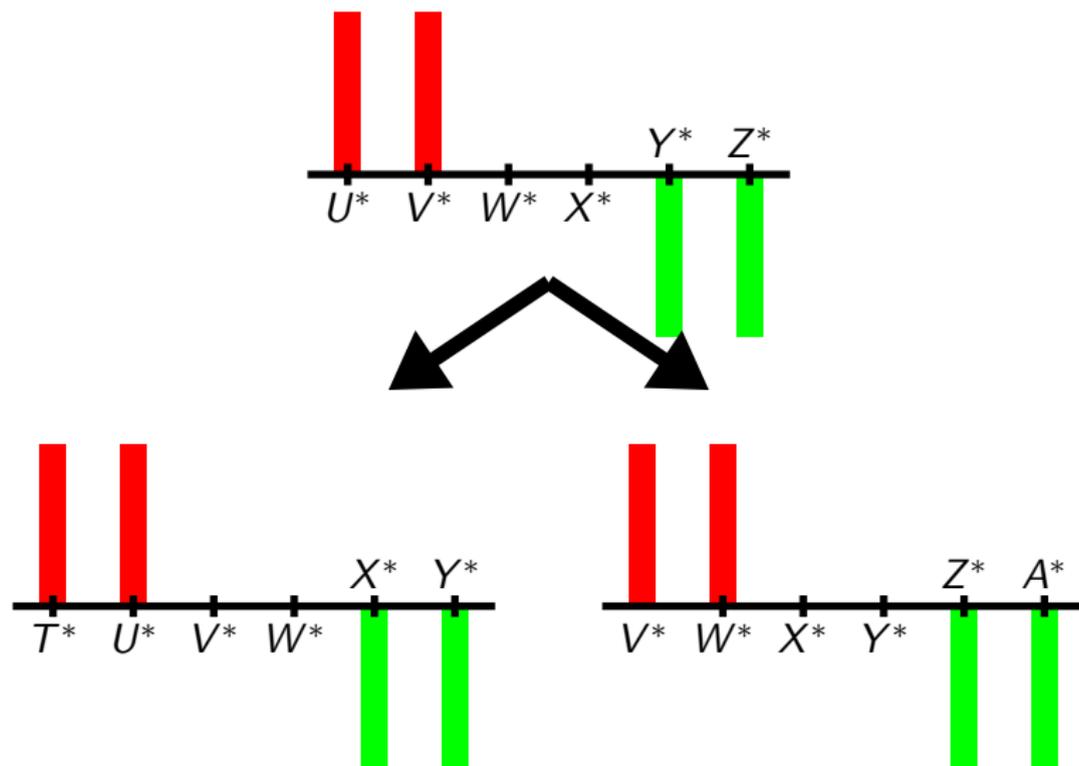
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Which way? How long?

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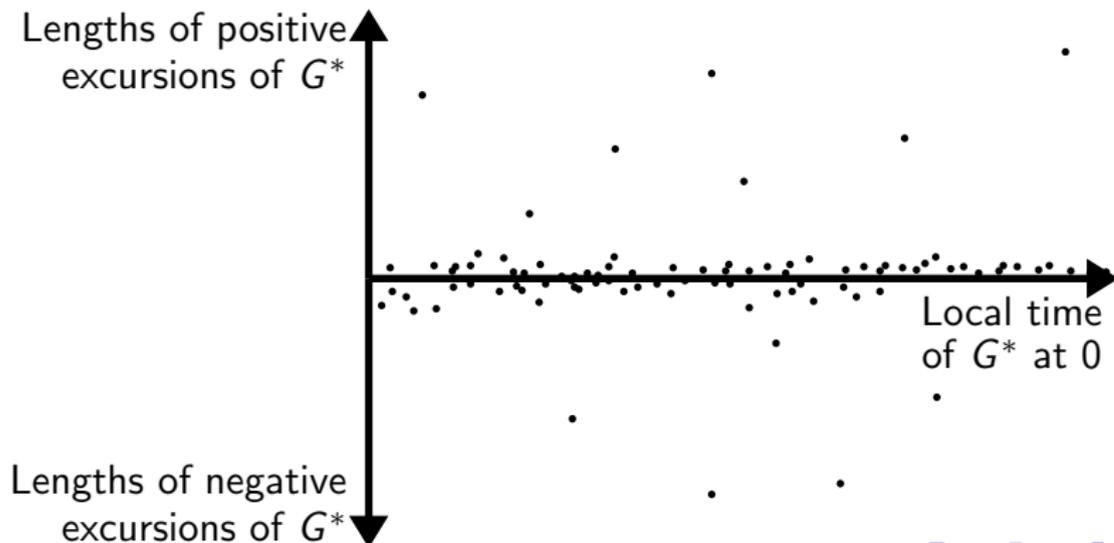
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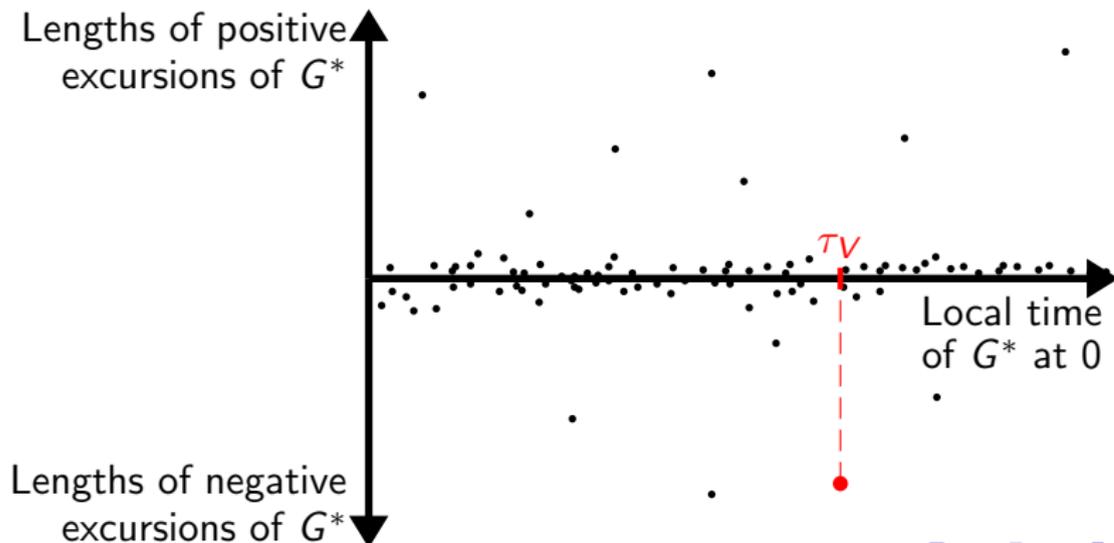
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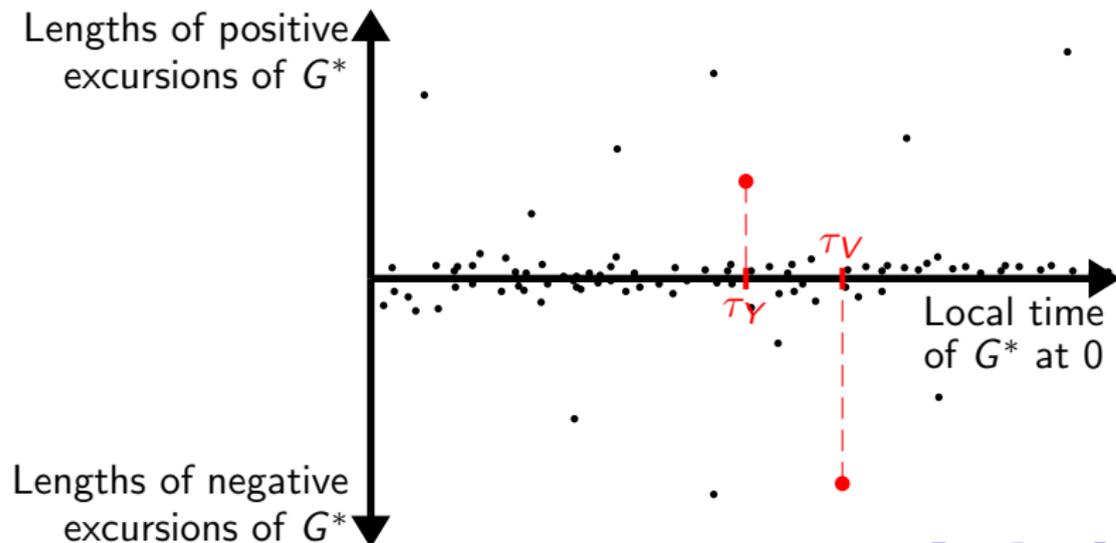
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- ▶  $\nu_\times^\pm(dt d\ell)$  – Lengths of positive (negative) excursions of  $G^*$  during which  $Y^*$  ( $V^*$ ) does not reach zero. Lévy measure is

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- ▶ For (ii), we observe that the distribution of the length of the “last excursion” is  $\mu_0(d\ell)/\mu_0((0, \infty))$ .
- ▶ Adapt Metzler again to compute the distribution of the **elapsed time**  $T_2$  in the “last excursion,” conditioned on its length.

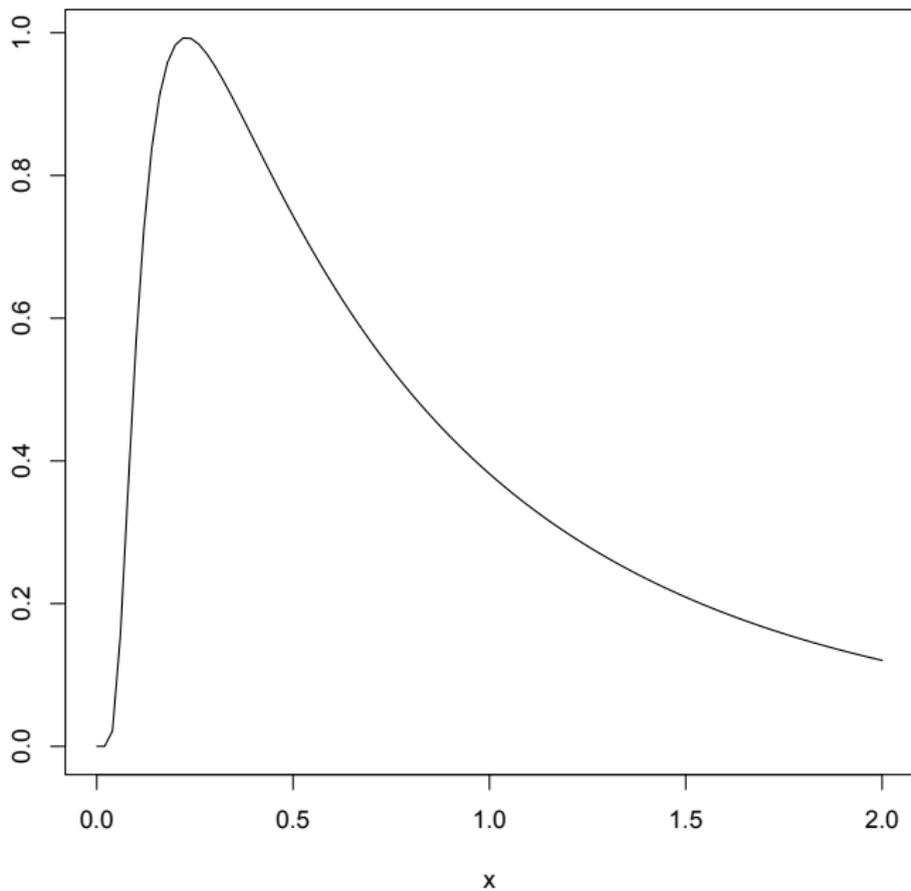
## Calculation of renewal time distribution

The moment-generating function of  $T_1 + T_2$  is

$$\begin{aligned} & \mathbb{E} \left[ e^{-\alpha(T_1+T_2)} \right] \\ &= \int_0^\infty \int_0^\ell e^{-\alpha s} \frac{p(s, \ell)}{\sqrt{2\pi\ell^3}} ds d\ell \bigg/ \left( \sqrt{\frac{\alpha}{2}} + \int_0^\infty e^{-\alpha\ell} \frac{p(\ell) d\ell}{2\sqrt{2\pi\ell^3}} \right), \end{aligned}$$

where  $p(s, \ell)$  is the conditional density in  $s$  of the elapsed time  $T_2$  given that the “last excursion” has length  $\ell$ .

# Probability Density Function



# Concluding remarks

Stochastic Processes and their Applications 11 (1981) 215-260  
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**MARTINGALES AND STOCHASTIC INTEGRALS  
IN THE THEORY OF CONTINUOUS TRADING**

J. Michael HARRISON

*Graduate School of Business, Stanford University, Stanford, CA 94305, U.S.A.*

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Stochastic Processes and their Applications 15 (1983) 313-316  
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**SHORT COMMUNICATION**

**A STOCHASTIC CALCULUS MODEL OF CONTINUOUS  
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### 1.6. Outline of the paper

This paper is aimed at readers with a good command of probability and stochastic processes, but no particular knowledge of economics. On the former dimension, we assume familiarity with the Strasbourg theory of martingales and stochastic integration, as developed in the definitive treatise by Meyer [32]. This assumption is perhaps unrealistic, but we cannot provide a systematic tutorial on stochastic integrals and an adequate treatment of our nominal subject matter in a reasonable amount of space.

(Also, the former task is best left to others. We are working dangerously close to the boundaries of our knowledge as things stand.)

Most of this paper will be accessible to those who know about stochastic integrals with respect to Brownian motion, and the rest should come into focus after a little study of the relevant foundational material. (On first reading, specialize general results to the case where  $S$  is an Ito process.) To facilitate such study, we consistently refer to Meyer [32] by page number for basic definitions and standard results, and his notation and terminology are used wherever possible. For a nice overview of the Strasbourg approach to stochastic integration, plus some new results and illuminating commentary, see the recent survey by Dellacherie [9] in this journal. A comprehensive treatment of stochastic calculus is given by Jacod [18], and it appears that the second volume of Williams [38] will be another good coursebook on martingales and stochastic integrals in the Strasbourg

# Φρέδος Παπαγγέλου

Fredos Papangelou  
Emeritus Professor  
School of Mathematics  
University of Manchester



# Vorlesungen über Maßtheorie von F. Papangelou

*“Sei  $I_n^*$  das nach beiden Seiten um seine zweifache Länge erweiterte Intervall  $I_n$ .“*

# Vorlesungen über Maßtheorie von F. Papangelou

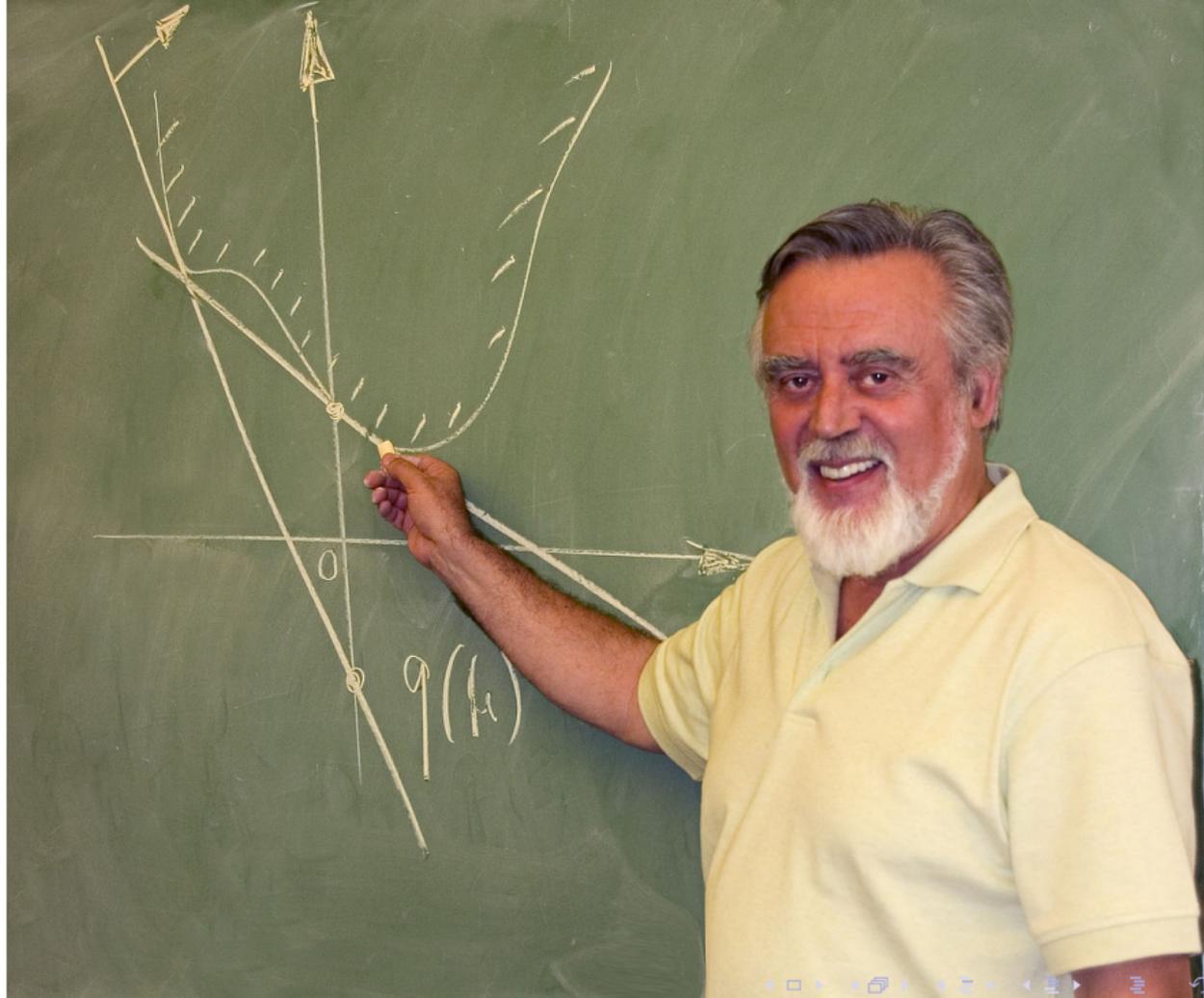
*“Sei  $I_n^*$  das nach beiden Seiten um seine zweifache Länge erweiterte Intervall  $I_n$ .”*

English translation:

*“Let  $I_n^*$  be the toward-both-sides-for-its-doubled-length-extended interval  $I_n$ .”*

# Δημήτρης Μπερτσέκας

Dimitri Bertsekas  
Jerry McAfee Professor  
Electrical and Computer Engineering  
Massachusetts Institute of Technology





# STOCHASTIC OPTIMAL CONTROL: THE DISCRETE- TIME CASE

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*“Martingales sprang fully armed from the forehead of Joseph Doob.” — Karatzas*

**If you encounter Greeks bearing gifts....**

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welcome them with open arms.**