On Exits of Oscillating Random Walks Under Delayed Observation

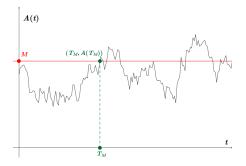
Ryan T. White

Florida Institute of Technology

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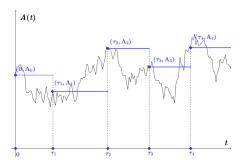
A Standard Problem

- Let A(t) be a Levy process on a filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t), \mathbb{P})$
- We consider the **first exit** of A(t) from $(-\infty, M)$, i.e. a level crossing of M
- The first exit time is $T_M = \inf\{t : A(t) > M\}$



Delayed Observation

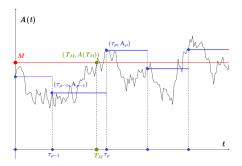
- A(t), the real-time process, is assumed to be inaccessible to the observer the position of A can be observed only upon a renewal process $\{\tau_i\}$
- The **observed process** is the piecewise-constant subsequence $\{A(\tau_i)\} := \{A_i\}$.



Realtime Process with Observed Process

Complications of Delayed Observation

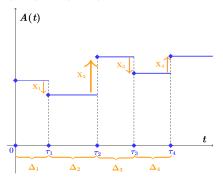
- The first exit time $T_M \notin \{\tau_j\}$ a.s., so the value of $A(T_M)$ is a.s. inaccessible to the observer
- Consider the post-exit time τ_{ρ} for $\rho = \inf\{n : A(\tau_n) > M\}$ and pre-exit time $\tau_{\rho-1}$



Realtime Process with Observed Process, Threshold, and First Exit Time

Increments of the Observed Process

• Denote $(\Delta_i, X_i) = (\tau_i - \tau_{i-1}, A_i - A_{i-1})$



The Observed Process and Increments

- Assume (Δ_j, X_j) are **jointly** *i.i.d.* random vectors with **dependent** components
 - \bullet $|X_j|$ and Δ_j are likely to be positively correlated in applications

Applications of Continuous Time Random Walks (CTRWs)

- Modeling anomalous diffusion in physics [Stanislavsky and Weron, 2010, Metzler and Klafter, 2007]
- Reliability of large-scale networks under attack and/or benign node failure [Dshalalow and White, 2013, 2014, White, 2015]
- Stochastic games [Dshalalow and Treerattrakoon, 2010]
- Geophysical motion of particles on planetary surfaces [Schumer et al., 2009]
- Financial forensics and modeling [Scalas, 2004, Jurlewicz et al., 2009, Dshalalow and Liew, 2006]

Known Results for Monotone CTRW Exiting $(-\infty, M]$

Joint Laplace-Stieltjes transforms:

$$\mathbb{E}\left[e^{-\alpha_0 A_{\rho-1} - \alpha A_{\rho} - h_0 \tau_{\rho-1} - h \tau_{\rho}}\right]$$

- ullet Marginal Laplace-Stieltjes transforms, e.g. $\mathbb{E}\left[e^{-lpha A_{ au
 ho}}
 ight]$
- Marginal distributions and moments (tractable in special cases)
- **Joint** distributions, e.g. $\mathbb{P}\{A_{\rho} < s, \tau_{\rho} < t\}$
- Similar results in 2-3 spatial dimension CTRW exiting rectangles
- Similar results for monotone "streaks" within oscillating CTRW

An Operational Calculus Approach to Monotone CTRW

① Consider the family $\{\rho(q) = \inf\{n : A_n > q\}$ and partition the sample space

$$\mathbb{E}\left[e^{-\alpha_0 A_{\rho(q)-1} - \alpha A_{\rho(q)} - h_0 \tau_{\rho(q)-1} - h \tau_{\rho(q)}}\right]$$

$$= \sum_{j=0}^{\infty} \mathbb{E}\left[e^{-\alpha_0 A_{j-1} - \alpha A_j - h_0 \tau_{j-1} - h \tau_j} \underbrace{\mathbf{1}_{\{\rho(q)=j\}}}_{\text{The only q-dependent term}}\right]$$

$$\sum_{j=0}^{\infty} \mathbb{E}\left[e^{-\alpha_0 A_{j-1} - \alpha A_j - h_0 \tau_{j-1} - h \tau_j} \left(e^{-x A_{j-1}} - e^{-x A_j}\right)\right]$$

Exploit independent and stationary increments

$$\sum_{j=0}^{\infty} \underbrace{\mathbb{E}\left[e^{-(\alpha_0+\alpha+x)X_1-(h_0+h)\Delta_1}\right]^{j-1}}_{\text{corresponds to }[0,\tau_{\rho-1}]} \underbrace{\left(\mathbb{E}\left[e^{-\alpha X_1-h\Delta_1}\right]-\mathbb{E}\left[e^{-(\alpha+x)X_1-h\Delta_1}\right]\right)}_{\text{corresponds to }(\tau_{\rho-1},\tau_{\rho}]}$$

- Sum as a geometric series
- Invert the Laplace-Carson under special cases with $q \to M$

Monotone CTRW Result

It is known

$$\Phi(\alpha_0, \alpha, h_0, h) = \mathbb{E}\left[e^{-\alpha_0 A_{\rho-1} - \alpha A_{\rho} - h_0 \tau_{\rho-1} - h \tau_{\rho}}\right]$$
$$= \mathcal{LC}_x^{-1} \left(\frac{\gamma(\alpha, h) - \gamma(\alpha + x, h)}{1 - \gamma(\alpha_0 + \alpha + x, h_0 + h)}\right) (M),$$

where

$$\gamma(\alpha,h) = \underbrace{\mathbb{E}\left[e^{-\alpha X_1 - h\Delta_1}\right]}_{\text{joint LST of an increment}}$$

- Inversions for many special cases are analytically or numerically tractable
- Similar approaches work in more dimensions

The Monotone Approach Fails

- The approach above does not work with general CTRW (convergence may fail)
- Consider the monotone increasing "streaks" broken up by "drops" in the CTRW, e.g.

$X_1 \ge 0$	$X_2 < 0$	$X_3 \ge 0$	$X_4 \ge 0$	$X_5 < 0$	$X_6 \ge 0$		$X_{\rho} \geq 0$
7	¥	7	7	¥	7		7
monotone ↑	DROP	monotone ↑		DROP	monotone ↑		

• The idea: exploit the monotone approach on the increasing "streaks"

Partitioning the Space

- Summing over every possible sequence of increasing "streaks" is unrealistic
- Can we partition the sample space usefully?
 - Restrict interest to the event $\mathcal{E} = \{A_k \text{ eventually exits}\}$
 - ullet Partition into the events $\mathcal{E}_n=\{A_k \text{ exits between the } (n-1) \text{th and } n \text{th drop} \}$

$$\begin{split} \Phi_{\mathcal{E}}(\alpha_0,\alpha,h_0,h) :&= \mathbb{E}\left[e^{i\alpha_0A_{\rho-1} + i\alpha A_{\rho} + ih_0\tau_{\rho-1} + ih\tau_{\rho}}\mathbf{1}_{\mathcal{E}}\right] \\ &= \sum_{n=0}^{\infty} \mathbb{E}\left[e^{i\alpha_0A_{\rho-1} + i\alpha A_{\rho} + ih_0\tau_{\rho-1} + ih\tau_{\rho}}\mathbf{1}_{\mathcal{E}_n}\right] \\ &=: \sum_{k=0}^{\infty} \Phi_{\mathcal{E}_n}(\alpha_0,\alpha,h_0,h) \end{split}$$

Calculating $\Phi_{\mathcal{E}_n}$ (I)

"Drop" indices

$$\mu_1=\inf\{m:X_m<0\}$$
 (the first "drop")
$$\mu_n=\inf\{m:X_m<0,m>\mu_{n-1}\}$$
 (the n th "drop")

• For n > 2, sum for all possible $\mu_{n-1} < \rho < \mu_n$

$$\begin{split} & \Phi_{\mathcal{E}_n}(\alpha_0, \alpha, h_0, h) \\ & = \mathbb{E}\left[e^{i\alpha_0 A_{\rho-1} + i\alpha A_{\rho} + ih_0 \tau_{\rho-1} + ih\tau_{\rho}} \mathbf{1}_{\{\mu_{n-1} < \rho < \mu_n\}}\right] \\ & = \sum^{\infty} \sum^{\infty} \sum^{\infty} \mathbb{E}\left[e^{i\alpha_0 A_{k-1} + i\alpha A_k + ih_0 \tau_{k-1} + ih\tau_k} \mathbf{1}_{\{\mu_{n-1} = j, \rho = k, \mu_n = l\}}\right] \end{split}$$

Side-Stepping a Combinatorial Nightmare

- The indicator $\mathbf{1}_{\{\mu_{n-1}=j,\rho=k,\mu_n=l\}}$ allows for n-2 "drop" indices below j
 - i.e. $\mu_1 < \mu_2 < \dots < \mu_{n-2} < j$
 - ullet Each has the same **number** of "drops" and positive jumps, i.e. $\binom{j-1}{n-2}$ **equal** terms
- Apply Fourier-Carson transform and exploit independent and stationary increments:

$$\begin{split} &\mathbb{E}\left[e^{i\alpha_{0}A_{k-1}+i\alpha A_{k}+ih_{0}\tau_{k-1}+ih\tau_{k}}\mathbf{1}_{\{\mu_{n-1}=j,\rho=k,\mu_{n}=l\}}\right]^{*}\\ &=\binom{j-1}{n-2}\left(\gamma^{+}\right)^{j-(n-1)}\left(\gamma^{-}\right)^{n-1}\left(\sigma^{1}-\sigma\right)\left(\gamma^{+}\right)^{k-j-1}P\{X_{1}\geq0\}^{l-k-1}P\{X_{1}<0\}\\ &\bullet \delta^{+}(\alpha,h)=\boxed{\mathbb{E}\left[e^{i\alpha X_{1}+ih\Delta_{1}}\mathbf{1}_{\{X_{1}\geq0\}}\right]} \end{split}$$

joint CF of nonnegative increment

$$\bullet \ \delta^{-}(\alpha,h) = \underbrace{\mathbb{E}\left[e^{i\alpha X_{1}+ih\Delta_{1}}\mathbf{1}_{\{X_{1}<0\}}\right]}_{\text{the ST}}$$

joint CF of negative increment

•
$$\gamma^+ = \delta^+(\alpha_0 + \alpha + x, h_0 + h)$$

$$\gamma^{-} = \delta^{-}(\alpha_0 + \alpha + x, h_0 + h)$$

• $\sigma^1 = \delta^+(\alpha, h)$ $\sigma = \delta^+(\alpha + x, h)$

Calculating $\Phi_{\mathcal{E}_n}$ (II)

Simple calculations show

$$\Phi_{\mathcal{E}_n}^*(\alpha_0,\alpha,h_0,h)$$

$$= p^{-} \left(\sigma^{1} - \sigma\right) \left(\gamma^{-}\right)^{n-1} \sum_{j=n-1}^{\infty} {j-1 \choose n-2} \left(\gamma^{+}\right)^{j-(n-1)} \sum_{k=j+1}^{\infty} {\gamma^{+}}^{k-j-1} \sum_{l=k+1}^{\infty} {(p^{+})}^{l-k-1}$$

$$= \frac{\sigma^1 - \sigma}{1 - \gamma^+} \left(\frac{\gamma^-}{1 - \gamma^+} \right)^{n-1}$$

Calculating $\Phi_{\mathcal{E}}$

Simply use the geometric series,

$$\Phi_{\mathcal{E}}^*(\alpha_0, \alpha, h_0, h) = \sum_{k=0}^{\infty} \Phi_{\mathcal{E}_n}^*(\alpha_0, \alpha, h_0, h)$$
$$= \frac{\sigma^1 - \sigma}{1 - \gamma^+ - \gamma^-}$$
$$= \frac{\sigma^1 - \sigma}{1 - \gamma}$$

where $\gamma=\mathbb{E}\left[e^{i(\alpha_0+\alpha+x)X_1+i(h_0+h)\Delta_1}\right]$ (joint CF of increment)

• Invert the Fourier-Carson transform

$$\Phi_{\mathcal{E}}(\alpha_0, \alpha, h_0, h) = \mathcal{F}\mathcal{C}_x^{-1} \left(\frac{\sigma^1 - \sigma}{1 - \gamma} \right) (M)$$

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