

Parisian ruin for fluid flow risk processes



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Fluid flow model.

Consider a fluid flow model with Brownian noise (started at level $u \in \mathbb{R}$) on the form

$$V_t = u + \int_0^t r_{J_s} ds + \int_0^t \sigma_{J_s} dB_s, \quad (t \geq 0), \quad (1)$$

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- for every $i \in E$, $r_i \in \mathbb{R} \setminus \{0\}$ and $\sigma_i \geq 0$.

Suppose that $V_t \rightarrow +\infty$ as $t \rightarrow \infty$ a.s..

Notation for the fluid flow model.

- E is partitioned and ordered into

$$E^\sigma := \{i \in E : \sigma_i > 0\},$$

$$E^+ := \{i \in E : \sigma_i = 0, r_i > 0\}, \text{ and}$$

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- The infinitesimal generator of $\{J_t\}_{t \geq 0}$ is written as

$$\Lambda = \begin{pmatrix} \Lambda^{\sigma\sigma} & \Lambda^{\sigma+} & \Lambda^{\sigma-} \\ \Lambda^{+\sigma} & \Lambda^{++} & \Lambda^{+-} \\ \Lambda^{-\sigma} & \Lambda^{-+} & \Lambda^{--} \end{pmatrix}, \quad (2)$$

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- Define the row vectors

$$\mathbf{r}_\sigma := \{r_i : i \in E^\sigma\}, \quad \mathbf{r}_+ := \{r_i : i \in E^+\},$$

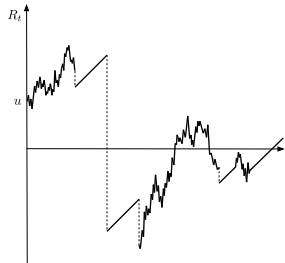
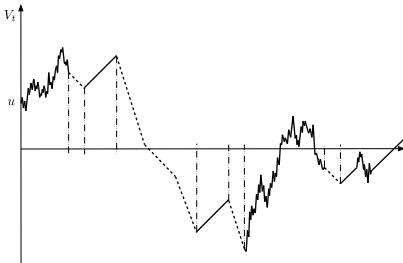
$$\mathbf{r}_- := \{r_i : i \in E^-\}, \quad \boldsymbol{\sigma} := \{\sigma_i : i \in E^\sigma\}.$$

Which is the connection with risk theory?

We define the **fluid flow risk process** $\{R_t\}_{t \geq 0}$ by regarding the linear downward segments of $\{V_t\}_{t \geq 0}$ as downward jumps of the same height.

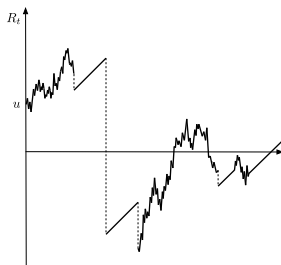
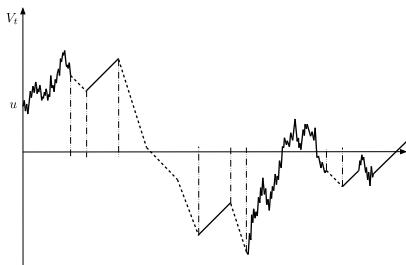
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Classic task: Compute $\psi(u) := \mathbb{P}(\inf_{s \geq 0} R_s < 0 | R_0 = u)$, the **classic probability of ruin**.

Classic risk processes as fluid flow risk processes.

Example (Cramér-Lundberg process)

The classic **Cramér-Lundberg process** with linear drift $p > 0$, Poisson arrival rate β , and $\text{PH}(\alpha, \mathbf{S})$ -distributed claims can be represented as a fluid flow risk process $\{R_t\}_{t \geq 0}$ with characteristics $E^\sigma = \emptyset$, $E^+ = \{1\}$, $E^- = \{2, 3, \dots, m+1\}$, $\mathbf{r}_+ = (p)$, $\mathbf{r}_- = (-1, \dots, -1)$ and

$$\Lambda = \begin{pmatrix} -\beta & \beta\alpha \\ -\mathbf{S}\mathbf{e} & \mathbf{S} \end{pmatrix}.$$

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$$\mathbf{\Lambda} = \begin{pmatrix} -\beta & \beta\alpha \\ -\mathbf{S}\mathbf{e} & \mathbf{S} \end{pmatrix}.$$

Remark

*Instead of having upward linear segments, we can opt to have a Brownian component, so that **Lévy risk processes with phase-type jumps** are an example of a fluid flow risk process.*

Classic risk processes as fluid flow risk processes.

Example (Sparre-Andersen process)

The Sparre-Andersen process with $\text{PH}(\alpha, \mathbf{S})$ -distributed claims and $\text{PH}(\pi, \mathbf{T})$ -distributed interarrival times can be represented as a fluid flow risk process $\{R_t\}_{t \geq 0}$ with characteristics $E^\sigma = \emptyset$, $E^+ = \{1, \dots, n\}$, $E^- = \{n+1, \dots, n+m\}$, $\mathbf{r}_+ = (1, \dots, 1)$, $\mathbf{r}_- = (-1, \dots, -1)$ and

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Remark

*We can represent risk processes with MAP arrivals and phase type jumps as fluid flow risk processes. These processes are basically **Markov additive risk processes with phase-type jumps**.*

Parisian ruin (when $E_\sigma = \emptyset$).

Definition

Suppose that $E_\sigma = \emptyset$, let $\{L_i\}_{i \geq 1}$ be i.i.d. clocks and associate each L_i to the (possible) i -th excursion below zero of $\{R_t\}_{t \geq 0}$.

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$$\psi_p(u) = \mathbb{P}(\text{Parisian ruin happens} \mid R_0 = u).$$

In other words, each time the reserve from an insurance company gets below 0, the company is given a (random) time window in order to recover.

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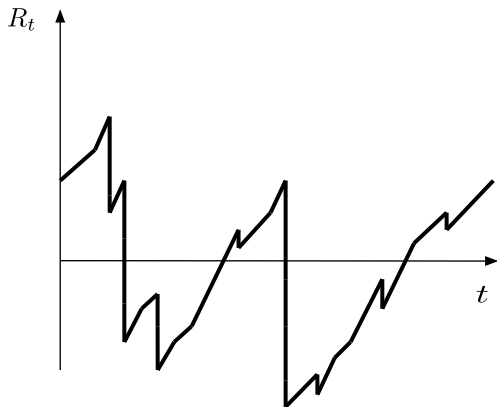
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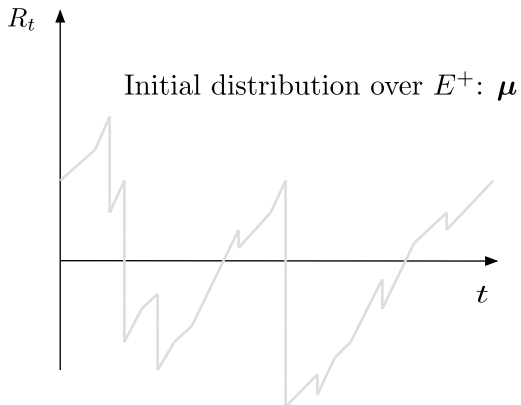
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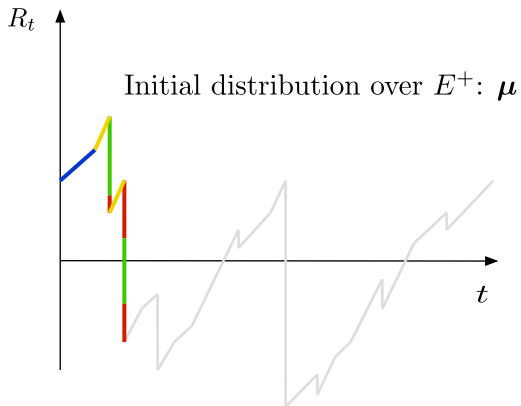
Erlangization arguments can be used to approximate the deterministic clocks case!

Fluid flow risk processes.
Parisian ruin.

Definition.
The associated Markov chain.
Main result.
Main result.
Numerical example.

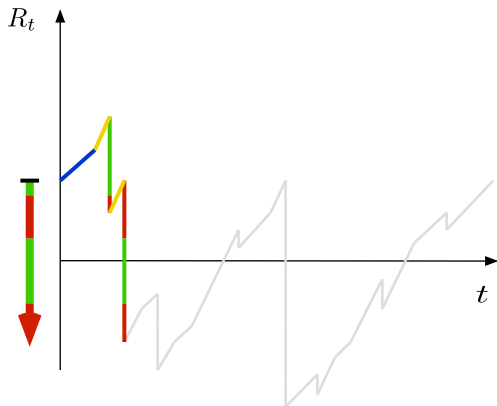


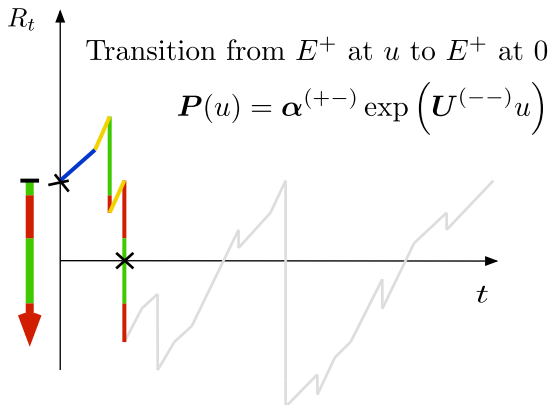




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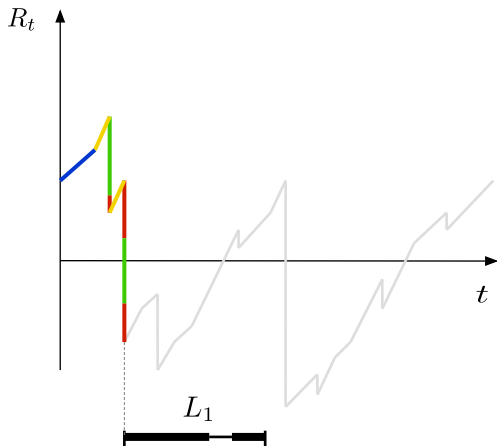
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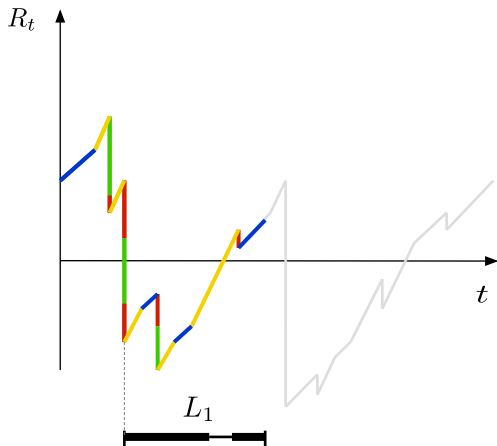
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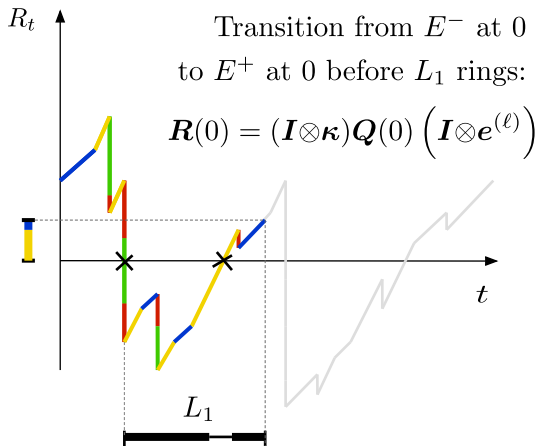
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$Q(\cdot)$ is of the form $Q(s) = \beta^{(-+)} \exp\left(V^{(++)}s\right)$

where $\beta^{(-+)}$ and $V^{(++)}$ are the upcrossing probabilities and intensity matrix for the FFM with intensity matrix

$$\begin{pmatrix} \Lambda^{++} \oplus K & \Lambda^{+-} \otimes I \\ \Lambda^{-+} \otimes I & \Lambda^{--} \otimes I^{(\ell)} \end{pmatrix},$$

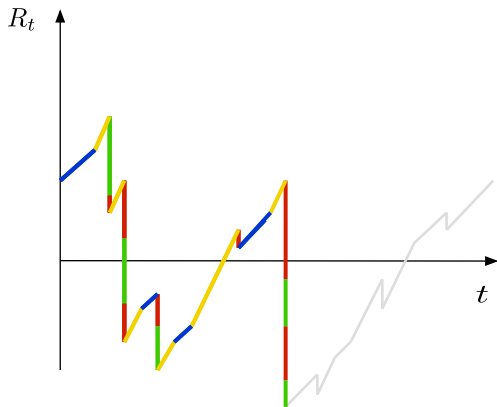
and drifts

$$r_+ \otimes e^T,$$

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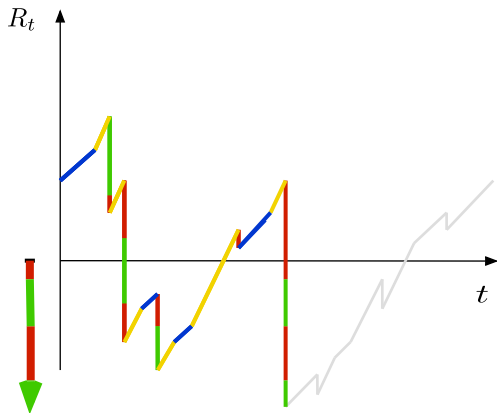
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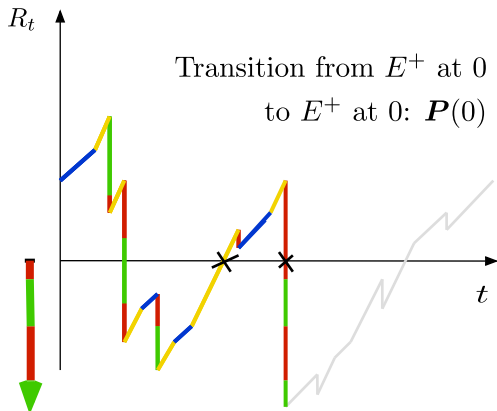
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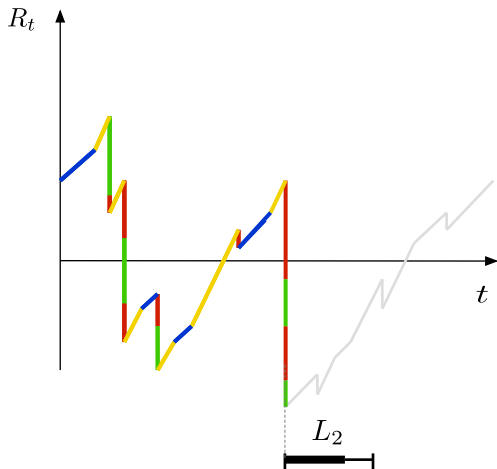
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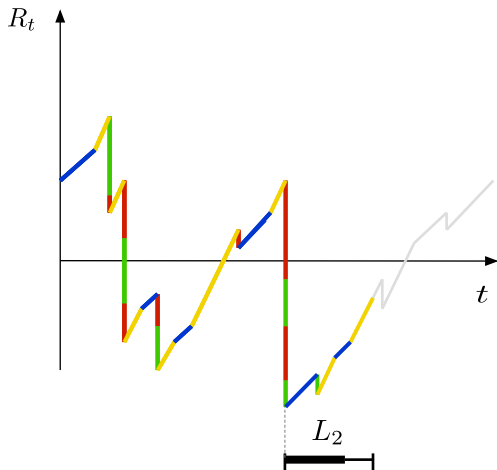
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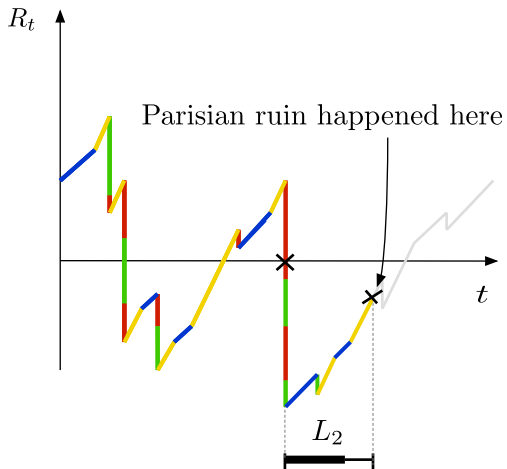
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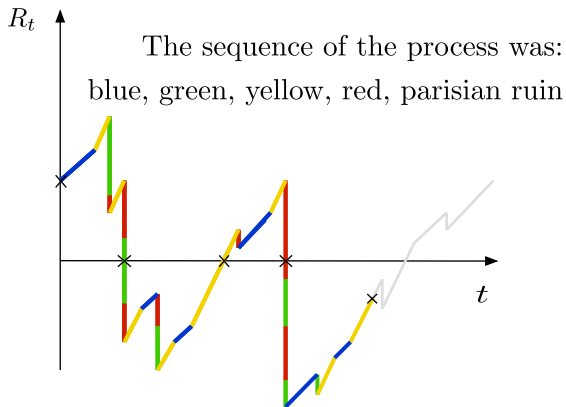


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The transition probability matrix.

We construct a Markov chain whose state space is partitioned into E^+ , E^- , ∂_N and ∂_P . Its initial distribution is

$$(\mathbf{0}, \mu \mathbf{P}(u), 1 - \mu \mathbf{P}(u) \mathbf{e}, 0),$$

and its transition matrix is given by

$$\begin{pmatrix} \mathbf{0} & \mathbf{P}(0) & \mathbf{e} - \mathbf{P}(0) \mathbf{e} & \mathbf{0} \\ \mathbf{R}(0) & \mathbf{0} & \mathbf{0} & \mathbf{e} - \mathbf{R}(0) \mathbf{e} \\ \mathbf{0} & \mathbf{0} & 1 & 0 \\ \mathbf{0} & \mathbf{0} & 0 & 1 \end{pmatrix}$$

Main result.

Theorem (Probability of Parisian ruin for the fluid flow risk process (if $E^\sigma = \emptyset$)).

If $E^\sigma = \emptyset$,

$$\psi_p(u) = \mu \mathbf{P}(u) (\mathbf{I} - \mathbf{R}(0) \mathbf{P}(0))^{-1} (\mathbf{e} - \mathbf{R}(0) \mathbf{e}) \quad (3)$$

or alternatively,

$$\psi_p(u) = \mu \mathbf{P}(u) \mathbf{v}_p,$$

where

$$\mathbf{v}_p = (\mathbf{I} - \mathbf{R}(0) \mathbf{P}(0))^{-1} (\mathbf{e} - \mathbf{R}(0) \mathbf{e}).$$

- The interpretation of the vector \mathbf{v}_p is

$$\mathbf{v}_p(i) = \mathbb{P} \left(\begin{array}{l} \{R_t\}_{t \geq 0} \text{ gets ruined} \\ \text{in a parisian way} \end{array} \mid \begin{array}{l} \text{The first downcrossing of 0} \\ \text{occured while in state } i \end{array} \right).$$

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- Notice that \mathbf{v}_p is independent of the initial reserve.
- Since classic ruin for $\{R_t\}_{t \geq 0}$ is given by $\mu \mathbf{P}(u) \mathbf{e}$, in order to compare parisian ruin with classical ruin, it is insightful to compare \mathbf{v}_p and \mathbf{e} .

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Theorem (Probability of parisian ruin for the fluid flow risk process (if $E^\sigma \neq \emptyset$)).

If $E^\sigma \neq \emptyset$, then

$$\psi_p(u) = \mu \mathbf{P}(u) \mathbf{v}_p,$$

where

$$\mathbf{v}_p = \lim_{\epsilon \downarrow 0} (\mathbf{I} - \mathbf{R}(\epsilon) \mathbf{P}(\epsilon))^{-1} (\mathbf{e} - \mathbf{R}(\epsilon) \mathbf{e}).$$

A Fluid flow risk process example

Take

$$\mathbf{r} = (0.15, 1.2, -1, -1, -1, -1), \quad \boldsymbol{\sigma} = (0.2, 0.4, 0, 0, 0, 0), \quad \text{and}$$

$$\mathbf{\Lambda} = \left(\begin{array}{cc|cccc} -0.5 & 0 & 0.5 & 0 & 0 & 0 \\ 0 & -0.5 & 0.5 & 0 & 0 & 0 \\ \hline 5 & 0 & -6 & 1 & 0 & 0 \\ 3 & 1 & 0 & -6 & 2 & 0 \\ 1 & 2 & 0 & 0 & -6 & 3 \\ 0 & 1 & 0 & 0 & 0 & -1 \end{array} \right)$$

In this model, the length of its jumps “dictates” in which environmental state we will end up after such jump ends.

Numerical example.

The values of ν_p over $\{1, 2, 3, 4, 5, 6\}$ for the 20-stage Erlang-distributed parisian clock case of mean 1, 5, 10 and 20 are shown below.

	$\mathbb{E}(L) = 1$	$\mathbb{E}(L) = 5$	$\mathbb{E}(L) = 10$	$\mathbb{E}(L) = 20$
σ_1	0.6677	0.4667	0.3915	0.2359
σ_2	0.5532	0.3922	0.3243	0.2022
-1	0.7985	0.5512	0.448	0.2749
-2	0.7737	0.5577	0.4585	0.2914
-3	0.7694	0.5926	0.4964	0.3316
-4	0.8191	0.6654	0.5647	0.391

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- Finite time horizon parisian ruin.