

Modélisation stochastique des avalanches par des processus de fragmentation-branchement

Madalina Deaconu

Inria - Institut Élie Cartan de Lorraine, Nancy

avec L. Beznea et O. Lupaşcu-Stamate

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- **Model introduction**

- Motivation: fragmentation examples
- Fragmentation equation

- **Stochastic approach**

- SDE with jumps
- Branching processes and fragmentation

- **Application - modelisation**

- Fragmentation, branching and avalanche
- Numerical algorithm

Introduction

Motivation: fragmentation examples

- **Fragmentation models**

- in astrophysics: stellar fragmentation, meteorits
- in crystallography: crystals fragmentation
- in nuclear physics: atoms fission
- in geophysics: rupture phenomena like avalanches, earthquakes, etc.

Introduction

Fragmentation equation

- **Continuous mass model**

- Consider the evolution of an infinite particle system
- The particles are completely characterized by their mass (continuous)
- Equation which describes the evolution of the concentration of particles $c(t, x)$ in the system

- **Fragmentation equation (EF)**

$$\begin{cases} \frac{\partial}{\partial t} c(t, x) = \int_x^1 F(x, y-x) c(t, y) dy \\ \quad - \frac{1}{2} c(t, x) \int_0^x F(y, x-y) dy, \forall t \geq 0, \forall x \in [0, 1], \\ c(0, x) = c_0(x), \forall x \in [0, 1]. \end{cases} \quad (EF)$$

- **Properties / Remarks**

- Binary fragmentation: a particle of mass $x + y$ splits in two particles of masses x and y
- $F : [0, 1] \times [0, 1] \rightarrow \mathbb{R}_+$ fragm. kernel, symmetric $F(x, y) = F(y, x)$
- **Mass conservation**
- Complex linear integro-differential \rightarrow difficult to solve

Probabilistic approach

Fragmentation equation

- **Key property**

- Mass conservation

$$\int_0^1 xc(t, x)dx = \int_0^1 xc_0(x)dx = 1, \forall t \geq 0$$

- Probability $Q_t(dx) = xc(t, x)dx, \forall t \geq 0$

- **Objectives and steps**

- Construct a stochastic process with jumps $(X_t)_{t \geq 0}$ whose law is $Q_t(dx)$
- Weak form of the equation (EF) in order to introduce the stochastic approach and the infinitesimal generator
- Construct a Markov process with jumps (stochastic differential equation)
- Get properties for the solution of (EF) via this approach

Probabilistic approach

Fragmentation and branching process

- **Hypothesis on F**

(H) F is continuous from $[0, 1]^2$ to $\mathbb{R}_+ \cup \{+\infty\}$.

The rate of loss of mass for the particle of mass x is:

$$\psi(x) = \begin{cases} \frac{1}{x} \int_0^x y(x-y)F(y, x-y)dy & \text{if } x > 0, \\ 0 & \text{if } x = 0. \end{cases}$$

F is s.t. ψ is continuous on $[0, 1]$.

- **Remark: the particle mass (size) $x \in [0, 1]$** - connexion with the fragmentation introduced by Bertoin
- **Example:** $F(x, y) := x + y$

Probabilistic approach

Weak solution for the fragmentation equation

- **Definition** The family $(Q_t)_{t \geq 0}$ of probabilities on $[0, 1]$ is a **weak solution** of (EF) with initial data Q_0 if:

$$\langle Q_t, \phi \rangle = \langle Q_0, \phi \rangle + \int_0^t \langle Q_s, \mathcal{F}\phi \rangle ds, \quad \forall \phi \in \mathcal{C}^1([0, 1]), t \geq 0, \quad (\text{EF-weak})$$

where $\langle Q_t, \phi \rangle = \int_0^1 \phi(y) Q_t(dy)$ and for all $x \in [0, 1]$:

$$\mathcal{F}\phi(x) = \int_0^x [\phi(x-y) - \phi(x)] \frac{x-y}{x} F(y, x-y) dy.$$

- **Aim:** Construct a process $(X_t)_{t \geq 0}$ whose law is $(Q_t)_{t \geq 0}$

- **Definition of the solution of the SDE (SDE – F)**

Let $(\Omega, \mathcal{G}, (\mathcal{G}_t)_{t \geq 0}, \mathbb{P})$ be a probability space and Q_0 a probability on $[0, 1]$.

X is a solution of (SDE-F) if:

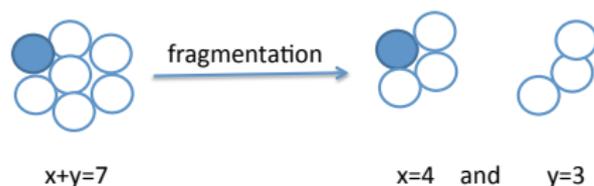
- $X = (X_t)_{t \geq 0}$ is an adapted process on $(\Omega, \mathcal{G}, (\mathcal{G}_t)_{t \geq 0}, \mathbb{P})$ having trajectories in $\mathbb{D}([0, +\infty), [0, 1])$.
- The law of $X_0 = Q_0$.
- There exists a Poisson measure $N(ds, dy, du)$ adapted to $(\mathcal{G}_t)_{t \geq 0}$ on $[0, +\infty) \times [0, 1) \times [0, 1)$ with intensity $ds dy du$ s.t.:

$$X_t = X_0 - \int_0^t \int_0^1 \int_0^1 y \mathbf{1}_{\{y \in (0, X_{s-})\}} \mathbf{1}_{\{u \leq \frac{X_{s-} - y}{X_{s-}} F(y, X_{s-} - y)\}} N(ds, dy, du)$$

Probabilistic approach

Interpretation

- **Illustration**



- **Interpretation**

At random Poissonian times the particle splits into two particles with smaller masses

- **Mass** the new particle is $X_{s-} - y$, $y \in (0, X_{s-})$
- **Time** at a rate $F(y, X_{s-} - y) \frac{X_{s-} - y}{X_{s-}}$

Probabilistic approach

Link (EF), (EF-weak) and (SDE-F)

Theorem

Under (H) there exists a solution $X = (X_t)_{t \geq 0}$ of the SDE of fragmentation (SDE - F). Let Q_t be the law of (X_t) , $t \geq 0$. Then, the family $\{Q_t\}_{t \geq 0}$ is a solution of (EF - weak).

- **How to come back to the initial equation (EF) ?**

Suppose that for all $t \geq 0$, Q_t is s.t. $Q_t(dx) = q(t, x)dx$. Then $c(t, x) := \frac{q(t, x)}{x}$ is a solution, in the weak sense, of (EF): for all $t \geq 0$ and all test function φ s.t. $\phi(x) = \frac{\varphi(x)}{x} \in C^1([0, 1])$ we have:

$$\int_0^1 \varphi(x)c(t, x)dx = \int_0^1 \varphi(x)c_0(x)dx$$
$$+ \frac{1}{2} \int_0^t \int_0^1 \int_0^x [\varphi(x-y) + \varphi(y) - \varphi(x)]F(y, x-y)c(s, x)dydxds.$$

Probabilistic approach

Branching process

E a measurable space (for us $E = [0, 1]$)

$$\hat{E} := \left\{ \mu \text{ non-negative measure on } E : \mu = \sum_{k=1}^m \delta_{x_k}, x_1, \dots, x_m \in E \right\} \cup \{\mathbf{0}\}.$$

Definition

A Markov process X on \hat{E} is a **branching process** iff for all measures $\mu_1, \mu_2 \in \hat{E}$:

$$X^{\mu_1 + \mu_2} \stackrel{(d)}{=} X^{\mu_1} + X^{\mu_2}.$$

- **Probabilistic interpretation** At the initial time the particle starts in a point from E and evolves according with a basis process X , up to a random time, when it gives birth to a number m of independent particles having same law as the mother particle (same law as X).
- **Fragmentation** The particles split independently one to each other so we can associate a branching property

Probabilistic approach

Fragmentation and branching

- **Steps**
 - Markov processes and fragmentation equation
 - Branching process associated to a fragmentation kernel
 - Fragmentation as a limit of the branching process

Probabilistic approach

Markov process and fragmentation equation

Theorem

Let F be a **bounded fragmentation kernel** and $Q_0 = \delta_x$. Then:

- $(EF - \text{weak})$ has a unique solution $(Q_{t,x})_{t \geq 0}$.
- The family of kernels $(Q_t)_{t \geq 0}$ on $[0, 1]$:

$$Q_t f(x) := \langle Q_{t,x}, f \rangle, \forall f \in \mathcal{P}\mathcal{C}_b([0, 1]), x \in [0, 1],$$

is also the transition function of the Markov process with jumps $X^0 = (\Omega, \mathcal{G}, \mathcal{G}_t, X_t^0, \mathbb{P}^x)$, with values in $[0, 1]$ (Ethier, Kurtz).

- For every function $\phi \in \mathcal{C}([0, 1])$ and $\nu \in \mathcal{P}([0, 1])$, $(\phi(X_t^0) - \int_0^t \mathcal{F}\phi(X_s^0) ds, t \geq 0)$ is a $(\mathcal{G}_t)_{t \geq 0}$ -martingale under \mathbb{P}^ν .
- The SDE (SDE - F), with initial condition δ_x , has a solution and this solution has same law as (X^0, \mathbb{P}^x) .

Probabilistic approach

Markov processes for truncated fragmentation kernel

Tools

- **Construct a sequence of thresholds for the particle size** fix a sequence $(d_n)_{n \geq 1} \subseteq (0, 1)$, strictly non-increasing converging to 0.

- **Truncated kernel (bounded)**

Let F be a fragmentation kernel. For $n \geq 1$ define

$$F_n(x, y) := \mathbb{1}_{(d_n, 1]}(x \wedge y) F(x, y), \quad x, y \in E = [0, 1].$$

- **Markov process for the kernel F_n** (size greater than d_n), by using the theorem.
- **Markov process truncated by the size d_n**
- **Result** Construct a branching process associated with. We need to:
 - Define the **Markov process**: $X = (\Omega, \mathcal{F}, \mathcal{F}_t, X_t, \theta_t, P^x)$, on E
 - A **Markovian kernel**: $B : bp\mathcal{B}(\widehat{E}) \rightarrow bp\mathcal{B}(E)$

Probabilistic approach

Branching process associated to the kernel

Theorem

(Beznea, Deaconu, Lupaşcu, SPA, 2015)

There exists a branching process on \widehat{E} , induced by the process X with state space E , and by the Markovian kernel B .

- **Branching process induced by the solution of the fragmentation equation**

By the Theorem, for all $n \geq 1$, there exists a branching process on \widehat{E}_n , induced by the process X^n on E_n , and by the kernel B^n , associated to the fragmentation kernel F .

Probabilistic approach

Branching process associated with the fragmentation

- **Idea of proof**

For $n \geq 1$ we note $(\widehat{P}_t^n)_{t \geq 0}$ the transition function of the branching process on \widehat{E}_n , induced by the basis process X^n and by the Markovian kernel B^n .

We can prove that there exists a projective limit of the sequence $(\widehat{P}_t)_{t \geq 0}$ on S^\downarrow , which is a transition function

- S^\downarrow link with the fragmentation model introduced by Bertoin

Probabilistic approach

Avalanches and fragmentation phenomenon

- A simplified physical model (fragmentation kernel) for the avalanche
- Stochastic differential equation of fragmentation
- Simulation



Probabilistic approach

Physical model for the avalanche kernel

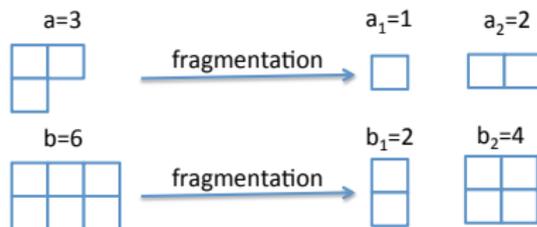
- Describe the avalanche via the fragmentation model with a fragmentation kernel which capture the physical properties.
- Fragmentation kernel for the avalanche

$$x + y \rightarrow x, y$$

such that

$$\frac{\min(x, y)}{\max(x, y)} = cst.$$

- **Example** If $cst. = 1/2$



- **Fragmentation kernel for the avalanche model**

There exists a function $\Phi : (0, \infty) \rightarrow (0, \infty)$ such that

$$F(x, y) = \Phi\left(\frac{x}{y}\right), \quad \forall x, y > 0, \quad \text{and} \quad \Phi(x) = \Phi\left(\frac{1}{x}\right), \quad \forall x > 0.$$

- **Example:** For $r > 0$, define the function

$$\Phi^r(x) := \frac{1}{2}\delta_r(x) + \frac{1}{2}\delta_{1/r}(x), \quad \forall x > 0.$$

In this situation $F^r(x, y) = \frac{1}{2}(\delta_r(\frac{x}{y}) + \delta_{1/r}(\frac{y}{x}))$.

- **Difficulty:** Φ^r is not continuous but we can construct a sequence of functions which approximates it and use the results before.

Probabilistic approach

Stochastic differential equation

- **Branching kernel associated with F^r :**

$$N_x^{F^r} := \lambda_0(\beta x \delta_{\beta x} + (1 - \beta)x \delta_{(1-\beta)x}),$$

where $\lambda_0 := \frac{\beta^2 + (1-\beta)^2}{4}$ with $\beta := \frac{r}{1+r}$

- **Warning:** N^{F^r} is no more Markovian and has no density w.r.t. the Lebesgue measure.

Take $d_1 < \beta \leq 1/2$ and $d_{n+1}/d_n < \beta$ for all $n \geq 1$. For n fixed let

$$E_n = \bigcup_{k=1}^n E'_{k-1}.$$

Define the kernel N_n^r on E_n as

$$N_n^r f := \sum_{k=1}^n \mathbb{1}_{E'_{k-1}} N^{F^r}(f \mathbb{1}_{E'_{k-1}}), \forall f \in bp\mathcal{B}(E_n).$$

- First order integral operator \widetilde{N}_n^r ,

$$\mathcal{F}_n^r f(x) := \widetilde{N}_n^r f(x) = \int_{E_n} [f(y) - f(x)] (N_n^r)_x(dy), \forall f \in bp\mathcal{B}(E_n) \text{ and } x \in E_n$$

\mathcal{F}_n^r is the generator of a (continuous time) jump Markov process $X^{r,n} = (X_t^{r,n})_{t \geq 0}$. Its transition function is $P_t^{r,n} := e^{\mathcal{F}_n^r t}$, $t \geq 0$.

Probabilistic approach

Stochastic differential equation for the avalanche - discontinuous kernel

$$E_{\beta,x} := \{ \beta^i (1-\beta)^j x : i, j \in \mathbb{N} \} \cup \{0\}, \quad E_{\beta,x,n} := E_{\beta,x} \cap E_n$$

Theorem

(Beznea, Deaconu, Lupaşcu-Stamate, MATCOM, 2018)

If $n \geq 1$ then E_n is absorbing w.r.t. the Markov process $X^{r,n}$ (in E_n) and

- (i) For every $\phi \in \text{bpB}(E_n)$ and each probability ν on E_n , $(\phi(X_t^{r,n}) - \int_0^t \mathcal{F}_s^r \phi(X_s^{r,n}) ds, t \geq 0)$ is a martingale under \mathbb{P}^ν , w.r.t. the natural filtration of $X^{r,n}$.
- (ii) If $x \in E_n$, $n \geq 1$, then the following SDE of fragmentation for avalanches, with initial distribution δ_x , has a solution which is equal in law with $(X^{r,n}, \mathbb{P}^x)$:

$$X_t = X_0 - \int_0^t \int_0^\infty p(d\alpha, ds) X_{\alpha-} \sum_{k=1}^n (1-\beta) \mathbb{1}_{\left[\frac{d_k}{\beta} \leq X_{\alpha-} < d_{k-1}, \frac{s}{\lambda_0 \beta} < X_{\alpha-}\right]} \\ + \beta \mathbb{1}_{\left[\frac{d_k}{1-\beta} \leq X_{\alpha-} < \frac{d_k}{\beta}, \frac{s}{\lambda_0(1-\beta)} < X_{\alpha-} \leq \frac{s}{\lambda_0 \beta}\right] \cup \left[\frac{d_k}{\beta} \leq X_{\alpha-} < d_{k-1}, \frac{s}{\lambda_0} < X_{\alpha-} \leq \frac{s}{\lambda_0 \beta}\right]},$$

where $p(d\alpha, ds)$ is a Poisson measure with intensity $q := d\alpha ds$.

- (iii) If $x \in E_n$ then \mathbb{P}^x -a.s. $X_t^{r,n} \in E_{\beta,x,n}$ for all $t \geq 0$.

Probabilistic approach

Algorithm

- **Initialisation:** Sample the initial particle $X_0 \sim Q_0$
- **Step p:**
 - Sample a random variable $S_p \sim \text{Exp}(\lambda_0)$
 - Let $T_p = T_{p-1} + S_p$
 - Set $X_t = X_{p-1}$ for all $t \in [T_{p-1}, T_p[$
 - Define

$$X_p = \begin{cases} \beta X_{p-1} & \text{with probability } \beta X_{p-1}, \\ (1 - \beta) X_{p-1} & \text{with probability } (1 - \beta) X_{p-1}, \\ X_{p-1} & \text{with probability } 1 - X_{p-1} \end{cases} \quad (0.1)$$

- **Stop:** When $T_p > T$
- **Outcome:** The approximation of the mass of the particle at time T , X_{p-1}

Numerical results

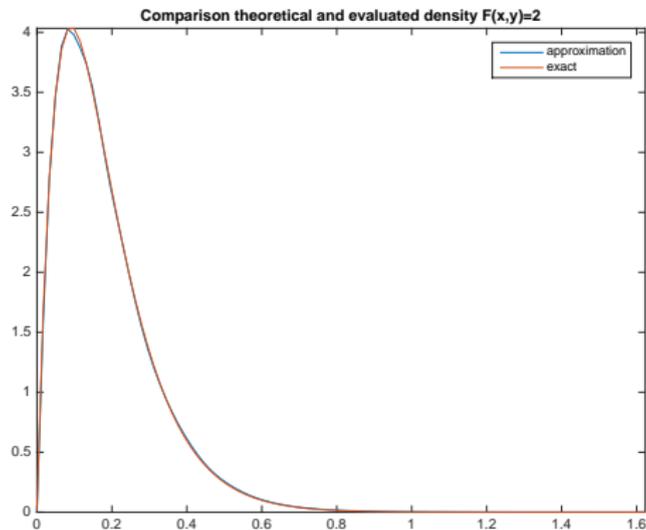


Figure : Comparison between the exact solution and the algorithm in the case $F(x, y) = 2$.

Numerical results

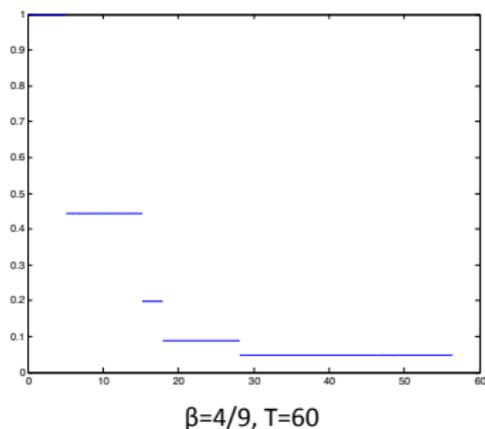
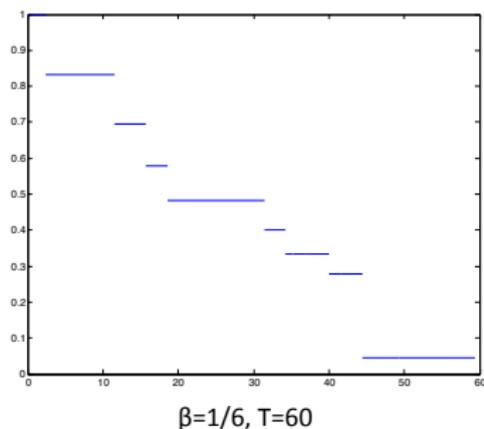


Figure : Path of the fragmentation process with the discontinuous kernel F^r and the size of the initial particle 1

Monte Carlo with 10^5 simulations

β	Mean \hat{T}_M	Confidence interval
$\frac{1}{6}$	0.1566	0.0020
$\frac{1}{3}$	0.1356	0.0017
$\frac{4}{9}$	0.1350	0.0016

Numerical results

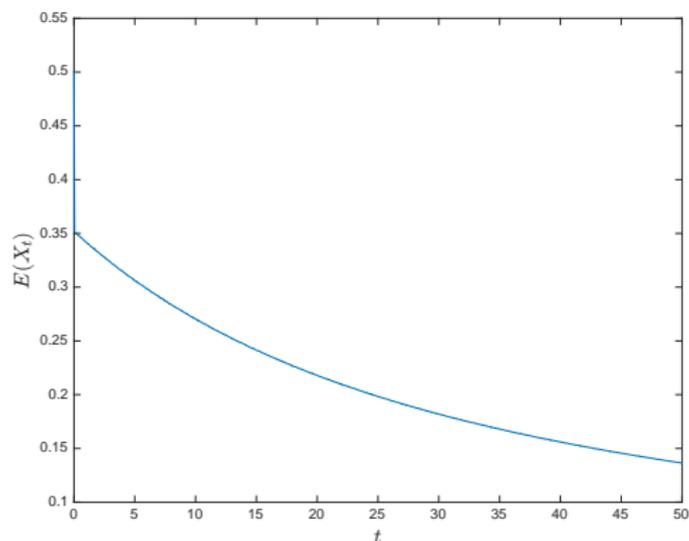


Figure : Time evolution of the empirical mean $t \mapsto \hat{I}_M(t)$ for $r = 0.2$, Monte Carlo parameter $M = 10^6$, and $t \in [0, 50]$.

Conclusion and perspectives

- **Go further**

Construct a complex model of coagulation / fragmentation depending on the position and with physical kernels

- Phase of snow accumulation (coagulation) before the avalanche begins
- The rupture phase (fragmentation)
- The phase of accumulation at the end of the avalanche (coagulation)

- Evaluate and control the risk connected to the avalanche

- **Collaboration**

- Geophysicians - Irstea Grenoble

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