

# Asymptotic behavior of a one-dimensional avalanche model through a particular stochastic process

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**Aim:** to investigate the invariant behaviour of the stochastic differential equations of coagulation-fragmentation

**Motivation:** Application to avalanches, a real problem



# An avalanche model

An avalanche model as **an interacting particle system** with values in  $\{0, 1\}^{\mathbb{Z}}$  (cf. [X. Bressaud, N. Fournier, *Annals of Probab.*, 2009]) :

- Let  $\Gamma = ((\Gamma_t(i))_{t \geq 0})_{i \in \mathbb{Z}}$  be an independent family of Poisson processes with rate 1.
- We assume that on each site  $i \in \mathbb{Z}$ , the flakes are falling following  $(\Gamma_t(i))_{t \geq 0}$  and the birth flakes follows Poisson processes with rate 1.
  - If a flake falls on a vacant site  $i$  of  $\mathbb{Z}$ , i.e.  $i = 0$ , this site becomes occupied,  $i = 1$ .
  - If a flake falls on an occupied site  $i \in \mathbb{Z}$ , **an avalanche starts**: the whole connected component of occupied sites around  $i$  becomes vacant.

The **avalanche process**  $((\mathbf{y}_t(i))_{t \geq 0})_{i \in \mathbb{Z}}$  is defined for  $t \geq 0$  and  $i \in \mathbb{Z}$ , as:  
 $\mathbf{y}_t(i) = 1$  if the site  $i$  is occupied at the moment  $t$ , or  
 $\mathbf{y}_t(i) = 0$  if the site  $i$  is vacant at time  $t$ .

**The state space** is

$$E := \{\mathbf{y} \in \{0, 1\}^{\mathbb{Z}} \text{ such that } \liminf_{i \rightarrow -\infty} \mathbf{y}(i) = \liminf_{i \rightarrow \infty} \mathbf{y}(i) = 0\}.$$

# A related binary coagulation-fragmentation model for the avalanche model

- The fragmentation / coagulation phenomenon for an infinite particles system.
- Each particle is characterised by its size and, at some random time, it can coagulate or fragment into two particles.
- $c(i, t) \geq 0$ : the density of  $i$  clusters at the time  $t$  in the system, solution of the **coagulation-fragmentation equation (C-FE)**:

$$\left\{ \begin{array}{l} \frac{\partial}{\partial t} c(i, t) = \frac{1}{2} \sum_{j=1}^{i-1} (K(i-j, j)c(i-j, t)c(j, t) - F(i-j, j)c(i, t)) \\ \quad - \sum_{j \in \mathbb{N}^*} (K(i, j)c(j, t)c(i, t) - F(i, j)c(i+j, t)) \\ c(i, 0) = c_0(i), \text{ for } (i, t) \in \mathbb{N}^* \times \mathbb{R}_+. \end{array} \right.$$

$K(i, j)$  : the rate of coagulation of two particles with masses  $i$  and  $j$  to form a cluster with mass  $i + j$ ,

$F(i, j)$  : the fragmentation rate of a cluster  $i + j$  breaks up into two clusters with masses  $i$  and  $j$ ,

# Probabilistic model for the avalanche

## Properties

- At some random time a particle can split into two particles or it can coagulate with another one

- *The total mass will be preserved:*

$$\sum_{i \geq 1} ic(i, t) = \sum_{i \geq 1} ic_0(i) \text{ for all } t \geq 0, \text{ and } \sum_{i \geq 1} ic_0(i) < \infty.$$

Therefore

$$Q_t(dx) = \sum_{i=1}^{+\infty} ic(i, t) \delta_i(dx) \text{ is a probability measure on } \mathbb{N} \text{ for each } t.$$

## Aim:

1. To construct a jump Markov process  $(X_t)_{t \geq 0}$  such that

$$\mathbb{P}(X_t \in dx) = Q_t(dx).$$

*This process will describe the evolution of the size of a typical particle in the coagulation-fragmentation system.*

2. A numerical approximation for the steady state of the process and a convergence result

# Hypothesis

- The **coagulation kernel**  $K : G \times G \mapsto \mathbb{R}_+$  is a continuous symmetric map. There exists a constant  $C$  such that for all  $x, y \in G$

$$K(x, y) \leq C(1 + x + y).$$

There exists a continuous nonnegative function  $\phi : G \rightarrow [1, \infty)$  such that  $x \rightarrow \phi(x)/x$  is nonincreasing on  $G$ , and for all  $x, y \in G$ ,

$$0 \leq K(y, x) = K(x, y) \leq \phi(x)\phi(y).$$

- The **fragmentation kernel**  $F : G \times G \mapsto \mathbb{R}_+$  is a continuous symmetric map. The function  $\Psi : G \mapsto \mathbb{R}_+$  defined by

$$\Psi(0) = 0, \quad \Psi(x) = \frac{1}{x} \int_0^x y(x-y)F(y, x-y)dy, \quad \text{for } x > 0,$$

is continuous.

# A related coagulation-fragmentation equation to the avalanche

- We consider the coagulation and fragmentation kernels: for all  $i, j \in \mathbb{N}^*$  we take  $K(i, j) = 2$  and

$$F(i, j) = \begin{cases} i + j - 1, & \text{if } i = 1 \text{ or } j = 1, \\ 0, & \text{if } i \neq 1 \text{ and } j \neq 1. \end{cases}$$

The particularity of the fragmentation kernel is that a particle  $i$  could be only split into two particles one of mass 1 and the other one of mass  $i - 1$ .

The coagulation-fragmentation equation related to the avalanche model is :

$$\begin{cases} \frac{\partial}{\partial t} c(i, t) = \frac{1}{m_0(t)} \sum_{j=1}^{i-1} c(i-j, t)c(j, t) - (i+1)c(i, t) + ic(i+1, t) \\ c(i, 0) = c_0(i), \end{cases} \quad \text{with } m_0(t) = \sum_{i \geq 1} c(i, t).$$



# The interpretation of the avalanche in relation with coagulation-fragmentation model

- We say that two neighbour edges  $(i-1, i)$  and  $(i, i+1)$  belong to the same particle if  $\mathbf{y}_t(i) = 1$ .
- We assume that each edge has a mass equal to 1.
- A particle of mass  $i$  contains  $i$  edges and  $i-1$  occupied sites.
- $(i, j)$  belongs to a particle with mass 1 if and only if  $\mathbf{y}(i) = \mathbf{y}(j) = 0$ .
- For a configuration  $\mathbf{y} \in E$  and for  $i \in \mathbb{N}$ , we assume that there exists the average number of particles with mass  $i$  per unit of length

$$c(i, \mathbf{y}_t) := \lim_{n \rightarrow \infty} \frac{\text{number of particles with mass } i \text{ in } [-n, n]}{2n+1}.$$

- Based on the invariance by translation of the model, we admit

$$c(i, t) := c(i, \mathbf{y}_t) = \frac{1}{i} \mathbb{P}[\text{the edge } (0, 1) \text{ belongs to a particle with mass } i \text{ in } \mathbf{y}_t].$$

- The family  $(c(\cdot, t))_{t \geq 0} = (c(i, t))_{i \geq 0, t \geq 0}$  would also satisfy  $\sum_{i \geq 0} ic(i, t) = 1$  for all  $t \geq 0$ , and

$$\left\{ \begin{array}{l} \frac{\partial}{\partial t} c(1, t) = -2c(1, t) + \sum_{i \geq 1} ic(i+1, t), \\ \frac{\partial}{\partial t} c(i, t) = -2c(i, t) - (i-1)c(i, t) + \\ \quad \frac{1}{m_0(t)} \sum_{j=1}^{i-1} c(i-j, t)c(j, t) + ic(i+1, t) \quad \text{for all } i \geq 2. \end{array} \right. \quad (0.1)$$

# The interpretation of the specific coagulation-fragmentation eq. in relation with the avalanche

The first equation counts the particles with mass 1 (the isolated edge).

$$\frac{\partial}{\partial t} c(1, t) = -2c(1, t) + \sum_{i \geq 1} ic(i+1, t)$$

- a particle with mass 1 **disappears** when a snowflake falls on an isolated edge and it coagulates at a constant rate 1 with its two neighbours,  $K \equiv 2$ .
- a particle with mass 1 **appears** if a flake falls on a particle of mass  $i+1$  and an avalanche starts: it breaks into two particles of masses 1 and  $i$ ,  $F(i, 1)$ .

In the second equation, for all  $i \geq 2$ ,

$$\frac{\partial}{\partial t} c(i, t) = -2c(i, t) - (i-1)c(i, t) + \frac{1}{m_0(t)} \sum_{j=1}^{i-1} c(i-j, t)c(j, t) + ic(i+1, t)$$

- the particles of mass  $i \geq 2$  which **disappear** after becoming larger with rate 2, when a flake falls on one of its extremities.
- the particles of mass  $i$  which **disappear** when a flake falls on a particle with mass  $i$  and an avalanche occurs with the rate  $F(i-j, j)$
- the particles of mass  $i$  which **appear** when a flake falls on one extremity of a particle of mass  $j$ , which is the extremity of a particle of mass  $i-j$  and they coagulate with the constant rate 2.
- the particles of mass  $i$  which **result** after the splitting of a larger particle  $i+1$  into two smaller particles, according to  $F(i, 1)$ .

Study the **equilibrium properties of the avalanche process**  $\mathbf{y}$  by using the steady state of the system (0.1) and an adapted stochastic model

### Proposition

The system of equations (0.1) admits a unique steady state denoted by  $c = (c(i))_{i \geq 1}$ , that is:

(i)  $c(i) \geq 0$  for each  $i \geq 1$

(ii)  $\sum_{i \geq 1} ic(i) = 1$

(iii)  $c = (c(i))_{i \geq 1}$  satisfies

$$\begin{aligned} 2c(1) &= \sum_{i \geq 1} ic(i+1), \\ (i+1)c(i) &= ic(i+1) + \frac{1}{M_c} \sum_{j=1}^{i-1} c(j)c(i-j) \quad \text{for all } i \geq 1 \end{aligned} \quad (0.2)$$

where  $M_c := \sum_{i \geq 1} c(i)$  is supposed uniquely defined and finite.

The steady state is given implicitly by (0.2). We can approximate numerically  $M_c \simeq 0,459134$  and then the steady state of the system.

## Definition

A stochastic process  $(X_t)_{t \geq 0}$  is a **solution of the stochastic differential equation of coagulation-fragmentation (SDECF)** if there exists a filtered probability space  $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$  such that

- $X_0$  is a  $x_{C_0}(dx)$  distributed random variable,
- $(X_t)_{t \geq 0}$  is a càdlàg  $G$ -valued  $(\mathcal{F}_t)_{t \geq 0}$  adapted process,
- There exist two independent Poisson measures  $N(ds, dy, dz)$  and  $M(ds, dy, dz)$  adapted to  $(\mathcal{F}_t)_{t \geq 0}$  on  $[0, +\infty) \times G \times [0, +\infty)$  with intensity measures  $dsQ_s(dy)dz$ , and  $dsdydz$  respectively, where for  $s \geq 0$ ,  $Q_s$  is the distribution of  $X_s$ , such that a.s. for all  $t \geq 0$

$$X_t = X_0 + \int_0^t \int_0^{+\infty} \int_0^{+\infty} y \mathbf{1}_{\{z \leq \frac{K(X_{s-}, y)}{y}\}} \mathbf{1}_{\{y < \infty\}} N(ds, dy, dz) - \int_0^t \int_0^{+\infty} \int_0^{+\infty} y \mathbf{1}_{\{y \in (0, X_{s-})\}} \mathbf{1}_{\{z \leq \frac{X_{s-} - y}{X_{s-}} F(y, X_{s-} - y)\}} M(ds, dy, dz).$$

The mass of a typical particle in the system is obtained by adding, at some random Poissonian times, the mass of another typical particle, driven by the  $K$ ; either by splitting, driven by the  $F$ , this mass into two smaller masses.

# Numerical approximation for the stochastic differential equation of the coagulation-fragmentation equation

- Under the Hypothesis on  $F$  and  $K$ , there exists a unique weak solution to the (SDECF), cf. [Fournier, Giet, *J. Stat. Phys.* 2003], in the discrete case [Jourdain, *Markov Processes and Related Fields*, 2003].

## Aim

- to approximate **the steady state of the avalanche** by using a numerical approach for the solution of the stochastic differential equation of the coagulation-fragmentation equation (SDECF);
- to give a **convergence result**

# Simulation of the coagulation-fragmentation process as a solution of (SDECf)

**Step 0:** Sampling  $X_0^1, \dots, X_0^n$  i.i.d.r.v.,  $\sim Q_0(dx) = xc_0(dx)$ .

**Step p.1:** Sampling a r. v.  $y^i \sim \mathcal{U}([0, X_{T_{p-1}}^i])$  for all  $i \in \{1, \dots, n\}$ .

Compute  $m_{p,f}^i = \frac{X_{T_{p-1}}^i - y^i}{X_{T_{p-1}}^i} F(y^i, X_{T_{p-1}}^i - y^i)$  for all  $i \in \{1, \dots, n\}$ .

Sampling a random variable  $S_{p,f}^i \sim \mathcal{E}(m_{p,f}^i)$  for all  $i \in \{1, \dots, n\}$ .

$$S_{p,f} = \min_i \{S_{p,f}^i\}, \quad i_f = i \{S_{p,f}^i\}.$$

**Step p.2:** Compute  $m_{p,c} = \sup_{i,j} \frac{K(X_{T_{p-1}}^i, X_{T_{p-1}}^j)}{X_{T_{p-1}}^j}$  for all  $i, j \in \{1, \dots, n\}$ .

Set  $i_c = i, j_c = j$  the couple for which the sup is realised.

Sampling a random variable  $S_{p,c} \sim \mathcal{E}(nm_{p-1,c})$ .

**Step p.3:** Set  $S_p = \min\{S_{p,f}, S_{p,c}\}$ .

**Step p.3f:** If  $S_p = S_{p,f}$  then  $X_{T_{p-1}}^{i_f}$  could fragment.

Sampling a random variable  $u \sim \mathcal{U}([0, 1])$ ; If  $u \leq m_{p,f}^{i_f}$ , then fragmentation occurs and set

$$X_{T_p}^{i_f} = X_{T_{p-1}}^{i_f} - y^{i_f}; \quad X_{T_p}^k = X_{T_{p-1}}^k \quad \text{for all } k \neq i_f.$$

Else set  $X_{T_p}^k = X_{T_{p-1}}^k$ , for all  $k \in \{1, \dots, n\}$ .



**Step p.3c:** If  $S_p = S_{p,c}$  then a coagulation can occur. At time  $S_p$  the pair  $(i_c, j_c)$  could coagulate and sampling  $u \sim \mathcal{U}([0, m_{p,c}])$ .

If  $u \leq \frac{K(X_{T_{p-1}}^{i_c}, X_{T_{p-1}}^{j_c})}{X_{T_{p-1}}^{j_c}}$ , then the coagulation occurs and set

$$X_{T_p}^{i_c} = X_{T_{p-1}}^{i_c} + X_{T_{p-1}}^{j_c}; X_{T_p}^k = X_{T_{p-1}}^k \text{ for all } k \neq i_c.$$

Else set  $X_{T_p}^k = X_{T_{p-1}}^k$  for all  $k \in \{1, \dots, n\}$ .

**Step p.4:** Set  $T_p = T_{p-1} + S_p$  and for all  $k \in \{1, \dots, n\}$  and  $t \in [T_{p-1}, T_p)$ ,

$$X_t^k = X_{T_{p-1}}^k.$$

**Stop:** When  $T_p > T$ , set, for all  $t \in [T_{p-1}, T]$ ,  $X_t^k = X_{T_{p-1}}^k$  for all  $k \in \{1, \dots, n\}$ .

**Outcome:** The approximated particles mass at time  $T$ ,  $X_{T_{p-1}}^k$  for all  $k \in \{1, \dots, n\}$ .

## Numerical results.

In Table 1 we give the values of  $c_i, i = 1, \dots, 5$ , **the concentration of the particles of mass  $i$**  in the system  $N = 15, T = 100$  and Monte Carlo parameter  $M = 10^4$ .

**Table:** Numerical approximations of concentration of particles  $c_i, i = 1, \dots, 5$

$c_i$	Algorithm A1	Value given by Proposition 0.1
$c_1$	0.2707	0.270433
$c_2$	0.1003	0.081732
$c_3$	0.0209	0.042954
$c_4$	0.0050	0.025178
$c_5$	0.0013	0.015186

$c_6 \simeq 0.009232, c_7 \simeq 0.005625, c_8 \simeq 0.003430, c_9 \simeq 0.002092, c_{10} \simeq 0.001275, \dots$

So, we can compute  $\sum_{i=1}^{50} c_i = 0.459036$ , which is a good approximation of  $M_c$ .

# The convergence of the algorithm

## Recursive Algorithm for the (SDECF)

Sampling  $X_0^1, \dots, X_0^n$  i.i.d.r.v., according to  $Q_0(dx) = x c_0(dx)$ .

Set the final time  $T$  and the initial  $t = 0$ ;

Set  $x = X_0 = (X_0^1, \dots, X_0^n)$ ;

While ( $t < T$ ), do

{  
Compute the fragmentation or coagulation time  $S_p = \min\{S_{p,f}, S_{p,c}\}$  and the  $i_f, i_c, j_c$  as in Step p.1 and Step p.2 from the Algorithm A1.

Set  $X_s = x$  for all  $s \in [t, (t + S_p) \wedge T]$

Set  $t = t + S_p$

If  $t \leq T$

{

Choose  $w$  uniformly in  $[0, 1]$

Set  $x_m = \text{mass}(t, w, x)$

Set  $x(i_f) = x_m(i_f) - y(i_f)$ , for the fragmentation case

or set  $x(i_c) = x_m(i_c) + x_m(j_c)$ , for the coagulation case

}

}

## Théorème

Let  $T < \infty$ . We assume that  $K, F$  satisfies the conditions from the Hypothesis and suppose that for all  $x, x', y \in \mathbb{N}^*$  if  $x \leq x'$  then  $K(x, y) \leq K(x', y)$ . Denote by  $C_T$  the total number of times that Recursive Algorithm takes before ending.

Then

$$\mathbb{E}[C_T] < \infty$$

and the Recursive Algorithm ends a.s. We denote by  $(X_t)_{t \in [0, T]}$  be the process constructed by the algorithm. Then  $(X_t)_{t \in [0, T]}$  satisfies the stochastic differential equation of coagulation-fragmentation (SDECF).

[O. L-S, M. Deaconu, *Asymptotic behavior of a one-dimensional avalanche model through a particular stochastic process*, preprint, 2023]

# Sketch of the proof

- (i) For any  $r \geq 0$ , any  $v \in (0, 1)$ , and any  $z \in \mathbb{N}^*$ , we denote by  $P_{r,v,z}(dx, dc)$  the law of a couple of random variables  $(X_{r,v,z}, C_{r,v,z})$ :
- $C_{r,v,z}$  the (possibly infinite) number of times that the execution of the mass  $(r, v, z)$
  - $X_{r,v,z}$  is the result of the function  $\text{mass}(r, v, z)$ , i.e. gives the possibility to have a fragmentation or a coagulation procedure.

(ii) For each  $r \geq 0$ , we denote by

- $C_r$  the (possibly infinite) total number of times that Recursive Algorithm calls the function  $\text{mass}$  to obtain  $X_r$ .
- Then  $C_r$  is a nondecreasing  $\mathbb{N} \cup \{\infty\}$ -valued process. and since  $X_{r,v,z}$  is simulated essentially in the same way as  $X_r$ , in law, we have

$$(X_{r,v,z}, C_{r,v,z}) \stackrel{(d)}{=} (X_r(\mathbf{1}_{\{v \leq [K(z, X_r)/X_r]/\lambda(z)\}} \mathbf{1}_{\{X_r < \infty\}} \mathbf{1}_{\text{coag.}} - \mathbf{1}_{\{X_r \in (0, z)\}} \mathbf{1}_{\{u \leq \frac{z-X_r}{z} F(X_r, z-X_r)\}} \mathbf{1}_{\text{frag.}}, C_r), (0.3)$$

For each  $r \in [0, T]$  we denote by  $Q_r$  the law of the  $\mathbb{N}^* \cup \{\infty\}$ -valued r. v.  $X_r$ .  $(X_r)_{r \in [0, T]}$  is now well-defined as a càdlàg,  $\mathbb{N}^* \cup \{\infty\}$ -valued process, and  $X_0$  has the distribution  $xQ_0(dx)$ .

Let  $D$  be the set  $\mathbb{N} \times (\mathbb{N} \cup \{\infty\})$ . For each  $r \in [0, T]$  we have that

$$\bullet \quad X_r = X_0 + \int_0^r \int_0^1 \int_D x M^c(ds, dv, d(x, c)) - \int_0^r \int_0^1 \int_D x M^f(ds, du, d(x, c)), \quad (0.4)$$

where  $M^c(ds, dv, d(x, c))$  and  $M^f(ds, du, d(x, c))$  are random integer-valued measures (see [JacSh87] on  $[0, T] \times (0, 1) \times D$ , with compensator  $\lambda(X_{s-}) ds dv P_{s,v,X_{s-}}(dx, dc)$ , respectively  $ds du P_{s,u,X_{s-}}(dx, dc)$ ).

$$\bullet \quad C_r = \int_0^r \int_0^1 \int_D (1+c) M^c(ds, dv, d(x, c)) + \int_0^r \int_0^1 \int_D (1+c) M^f(ds, dv, d(x, c)) \quad (0.5)$$

- $(X_r)_{r \in [0, T]}$  satisfies (SDECF).

Let us now prove that  $\mathbb{E}[C_T] < \infty$ .

- For  $r \in [0, T]$  let  $n_r$  be the distribution of the random variable  $C_r$  which is  $\mathbb{N} \cup \{+\infty\}$  valued. By using formula (21) we can express (0.5):

$$C_r = \int_0^r \int_{\mathbb{N}} (1+c) \nu(ds, dc), \quad (0.6)$$

where  $\nu$  is a random integer-valued measure of the form  $\nu = \nu^c + \nu^f$ , and  $\nu^c$  and  $\nu^f$  denote also random integer-valued measures with compensators  $\lambda(X_{r-})dn_r(dc)$  and  $drn_r(dc)$  respectively.

- Let  $A \in (0, +\infty)$ , we have:

$$C_r \wedge A \leq \int_0^r ds ((1+c) \wedge A) \nu(ds, dc). \quad (0.7)$$

By taking expectation and denoting  $m_T = \sup_{t \in [0, T]} \mathbb{E}(\lambda(X_t) + X_t)$  we get

$$\mathbb{E}(C_r \wedge A) \leq m_T + m_T \int_0^r \mathbb{E}(C_T \wedge A) ds. \quad (0.8)$$

- By Gronwall Lemma we deduce that  $\mathbb{E}(C_T \wedge A) \leq \gamma_T$ , where  $\gamma_T$  is a constant not depending on  $A$ , and we take the limit as  $A$  goes to infinity.

## Probabilistic approaches

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