

Lecture 4: More on additive martingales

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II.2) Martingales

For $\lambda \in \mathbb{R}$, $t \geq 0$, we define $W_t^\lambda := \sum_{v \in \mathcal{N}(t)} e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t}$.

Remark: $W_t^0 = e^{-mt} N_t$.

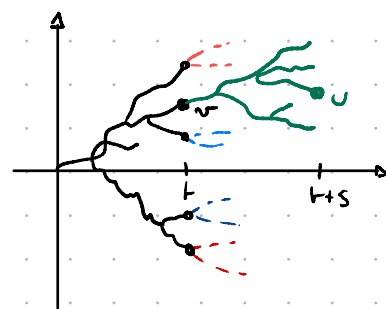
Proposition: $(W_t^\lambda)_{t \geq 0}$ is an $(\mathcal{F}_t)_{t \geq 0}$ -martingale, called additive martingale.

Proof: First note that, for $t \geq 0$,

$$\mathbb{E}[W_t^\lambda] \stackrel{\text{many-to-one}}{=} e^{-mt} \mathbb{E}\left[e^{\lambda B_t - \frac{\lambda^2}{2}t}\right] = 1 \quad \text{because } \mathbb{E}[e^{\lambda B_t}] = e^{\frac{\lambda^2}{2}t}.$$

Now let $t, s \geq 0$.

$$\begin{aligned} W_{t+s}^\lambda &= \sum_{v \in \mathcal{N}(t+s)} \sum_{\substack{u \in \mathcal{N}(t+s) \\ \text{s.t. } u \geq v}} e^{\lambda X_u(t+s) - (\frac{\lambda^2}{2} + m)(t+s)} \\ &= \sum_{v \in \mathcal{N}(t)} e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t} \underbrace{\sum_{\substack{u \in \mathcal{N}(t+s) \\ \text{s.t. } u \geq v}} e^{\lambda(X_u(t+s) - X_v(t)) - (\frac{\lambda^2}{2} + m)s}}_{\text{by the branching property, conditionally on } \mathcal{F}_t, \text{ these are independent random variables with the same law as } W_s^\lambda} \end{aligned}$$



by the branching property, conditionally on \mathcal{F}_t , these are independent random variables with the same law as W_s^λ .

$$\text{So } \mathbb{E}[W_{t+s}^\lambda | \mathcal{F}_t] = \sum_{v \in \mathcal{N}(t)} e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t} \underbrace{\mathbb{E}[W_s^\lambda]}_{=1} = W_t^\lambda. \quad \square$$

Consequence: $W_t^\lambda \xrightarrow{\text{a.s.}} W_\infty^\lambda$ (positive martingale!)

Proposition: Let $q = P(\forall t, N_t \geq 1)$ be the probability of survival.

For any $\lambda \in \mathbb{R}$, $P(W_\infty^\lambda > 0) \in \{0, q\}$.

\Rightarrow either $W_\infty^\lambda > 0$ a.s. on survival, or $W_\infty^\lambda = 0$ a.s.

Proof (similar to the case $\lambda = 0$!)

At the first branching time T_1 , there are L children each one starts an independent BBM, write $W_\infty^{\lambda,1}, \dots, W_\infty^{\lambda,L}$ their associated limit of additive martingales. Then, $W_\infty^\lambda = 0 \Leftrightarrow \forall i \in \{1, \dots, L\}, W_\infty^{\lambda,i} = 0$.

So $P(W_\infty^\lambda = 0) = \mathbb{E}[P(W_\infty^\lambda = 0)^L]$ hence it is a fixed point of $f(s) = \mathbb{E}[s^L]$.

The two fixed points of f are $1-q$ and 1 .



Last time we saw a theorem saying $\forall |\lambda| < \sqrt{2m}$, (W_t^λ) is uniformly integrable. The proof was based on a 2nd moment calculation for a truncated version of W_t^λ .
 → we will see today a stronger result using a different proof:

Theorem (Neveu 1988): Assume $E[L^2] < \infty$.

- If $|\lambda| < \sqrt{2m}$, then $W_t^\lambda \rightarrow W_\infty^\lambda$ in L^p for any $p \in (1, 2] \cap (1, \frac{2m}{\lambda^2})$ and in particular $W_\infty^\lambda > 0$ a.s. on survival.
- If $|\lambda| \geq \sqrt{2m}$, then $W_\infty^\lambda = 0$ a.s.

Lemma (von Bahr - Esseen 1965): Let $n \geq 1$, $p \in [1, 2]$. Let $X_1, \dots, X_n \in L^p$ be independent centered r.v. Then $E[|\sum_{k=1}^n X_k|^p] \leq 2 \sum_{k=1}^n E[|X_k|^p]$.

Proof: see Exercise 3 at the end of the file if you are curious.

Remark: This is clearly true for $p=1$ and $p=2$ (with constant 1 instead of 2).

Proof of the theorem:

- Case $|\lambda| < \sqrt{2m}$: Recall $W_{t+s}^\lambda = \sum_{v \in \mathcal{N}(t)} e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t} \sum_{\substack{u \in \mathcal{N}(t+s) \\ \text{s.t. } u \geq v}} e^{\lambda(X_u(t+s) - X_v(t)) - (\frac{\lambda^2}{2} + m)s}$
 and, by the branching property, conditionally on \mathcal{F}_t ,
 the $W_s^\lambda(v, t)$ for $v \in \mathcal{N}(t)$ are independent and have the same law as W_s^λ .
(the additive martingale for the BBT starting from v at time t)

$$\text{Then } W_{t+s}^\lambda - W_t^\lambda = \sum_{v \in \mathcal{N}(t)} \underbrace{e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t}}_{\mathcal{F}_t\text{-measurable}} \underbrace{(W_s^\lambda(v, t) - 1)}_{\text{indep. and centered given } \mathcal{F}_t}$$

So by the lemma:

$$\begin{aligned} E[|W_{t+s}^\lambda - W_t^\lambda|^p | \mathcal{F}_t] &\leq \sum_{v \in \mathcal{N}(t)} E\left[\left(e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t} |W_s^\lambda(v, t) - 1|\right)^p | \mathcal{F}_t\right] \\ &= \left(\sum_{v \in \mathcal{N}(t)} e^{\lambda p X_v(t) - (\frac{\lambda^2}{2} + m)pt}\right) E[|W_s^\lambda - 1|^p] \\ &= e^{(\frac{\lambda^2 p}{2} + mp)t - (\frac{\lambda^2}{2} + m)pt} \underbrace{W_t^{\lambda p}}_{\text{has expectation 1}} E[|W_s^\lambda - 1|^p] \end{aligned}$$

Taking the expectation: $\mathbb{E}[|W_{t+s}^\lambda - W_t^\lambda|^p] \leq e^{(p-1)(\frac{\lambda^2}{2} - m)t} \mathbb{E}[|W_s^\lambda - 1|^p]$
 Now take $t \in \mathbb{N}$ and $s=1$. < 0 because $p < \frac{2m}{\lambda^2}$

Note that $\mathbb{E}[|W_1^\lambda - 1|^p] \stackrel{\text{Jensen}}{\leq} \mathbb{E}[(W_1^\lambda - 1)^2]^{p/2} \leq C(\lambda, p)$ (check it with the many-to-two).
 so $\|W_{t+1}^\lambda - W_t^\lambda\|_p$ is summable and $(W_t^\lambda)_{t \in \mathbb{N}}$ is bounded in L^p .

Then, $(W_t^\lambda)_{t \geq 0}$ is bounded in L^p :

$$\mathbb{E}[|W_t^\lambda|^p] \stackrel{\text{martingale}}{=} \mathbb{E}[|\mathbb{E}[W_{t+1}^\lambda | \mathcal{F}_t]|^p] \stackrel{\text{Jensen}}{\leq} \mathbb{E}[\mathbb{E}[|W_{t+1}^\lambda|^p | \mathcal{F}_t]] = \mathbb{E}[|W_{t+1}^\lambda|^p] \text{ bounded.}$$

So $(W_t^\lambda)_{t \geq 0}$ converges in L^p .

• Case $|\lambda| > \sqrt{2m}$: Taking $s \rightarrow \infty$ in the decomposition of W_{t+s}^λ above, we get

$$W_\infty^\lambda = \sum_{v \in \mathcal{V}(t)} e^{\lambda X_v(t) - (\frac{\lambda^2}{2} + m)t} W_\infty^\lambda(v, t) \quad \text{where } W_\infty^\lambda(v, t) = \lim_{s \rightarrow \infty} W_s^\lambda(v, t) \text{ (a.s. limit)}$$

Let $p \in (0, 1)$. Note that $\mathbb{E}[(W_\infty^\lambda)^p] < \infty$ because $W_\infty^\lambda \in L^1$ by Fatou ($\Rightarrow \mathbb{E}[W_\infty^\lambda] \leq 1$).

By subadditivity of $x \mapsto x^p$,

$$(W_\infty^\lambda)^p \leq \sum_{v \in \mathcal{V}(t)} e^{\lambda p X_v(t) - (\frac{\lambda^2}{2} + m)pt} (W_\infty^\lambda(v, t))^p \quad \text{independent and same law as } W_\infty^\lambda$$

$$\begin{aligned} \mathbb{E}[(W_\infty^\lambda)^p] &\leq \mathbb{E}\left[\sum_{v \in \mathcal{V}(t)} e^{\lambda p X_v(t) - (\frac{\lambda^2}{2} + m)pt}\right] \times \mathbb{E}[(W_\infty^\lambda)^p] \\ &= e^{(\frac{\lambda p^2}{2} + m)t - (\frac{\lambda^2}{2} + m)pt} \mathbb{E}[W_t^{\lambda p}] = e^{(p-1)(\frac{\lambda^2}{2}p - m)t} < 1 \quad \text{if } p \in \left(\frac{2m}{\lambda^2}, 1\right). \end{aligned}$$

So $\mathbb{E}[(W_\infty^\lambda)^p] = 0$ and $W_\infty^\lambda = 0$ a.s.

• Case $|\lambda| = \sqrt{2m}$: see exercise at the end of the file. ▣

II.3) Back to the number of particles

For $x \in \mathbb{R}$ and $t \geq 0$, let $N_t(x) = \sum_{v \in \mathcal{V}(t)} \mathbb{1}_{X_v(t) \geq x}$.

Reminder: We have seen that $\mathbb{E}[N_t(at)] \sim \frac{1}{\sqrt{2\pi t}} e^{(m - \frac{a^2}{2})t}$ for $a > 0$

and that $\forall a \in (0, \sqrt{2m})$, $\mathbb{P}(N_t(at) \geq 1) \xrightarrow{t \rightarrow \infty} 1$.

(or even in exercise $\mathbb{P}(N_t(at) \geq e^{(m - \frac{a^2}{2} - \varepsilon)t}) \xrightarrow{t \rightarrow \infty} 1$.)

Theorem: For any $a \in (0, \sqrt{2m})$, $\frac{1}{\sqrt{2\pi t}} e^{-(m - \frac{a^2}{2})t} N_t(at) \xrightarrow[t \rightarrow \infty]{P} W_\infty^a$.

Remark: The convergence actually holds a.s. but this is harder to prove.
 See Louis Chabaignier's master thesis for details.

Lemma: Let $(x_t)_{t \geq 0}$ be such that $\frac{x_t}{\sqrt{t}} \rightarrow \infty$ as $t \rightarrow \infty$.

Then $E[N_t(x)] \sim \sqrt{\frac{t}{2\pi}} \frac{1}{x} e^{-mt - \frac{x^2}{2t}}$ as $t \rightarrow \infty$, uniformly in $x \geq x_t$.

Proof: By the many-to-one $E[N_t(x)] = e^{mt} P(B_t \geq x) = e^{mt} P(B_1 \geq \frac{x}{\sqrt{t}})$
and $P(B_1 \geq y) \sim \frac{e^{-y^2/2}}{\sqrt{2\pi} y}$ as $y \rightarrow \infty$. ■

Proof of the theorem: The strategy is the following: consider some time $s = s(t) \leq t$ such that $s \rightarrow \infty$ as $t \rightarrow \infty$. Then we will prove

① $a\sqrt{2\pi t} e^{-(m - \frac{a^2}{2})t} E[N_t(at) | \mathcal{F}_s] \xrightarrow{P} W_\infty^a$ if $s = o(\sqrt{t})$.

② $\sqrt{t} e^{-(m - \frac{a^2}{2})t} (N_t(at) - E[N_t(at) | \mathcal{F}_s]) \xrightarrow{L^p} 0$ for $p \in (1, \frac{2m}{a^2})$ if $\frac{s}{\log t} \rightarrow \infty$.

This implies the result by choosing for example $s = t^{1/4}$.

Step ①: $N_t(at) = \sum_{v \in \mathcal{N}(s)} \sum_{\substack{u \in \mathcal{U}(t) \\ u \geq v}} \mathbb{1}_{X_u(t) - X_v(s) \geq at - X_v(s)}$.

Writing $\varphi(t, x) = E[N_t(x)]$ and using the branching property

$$E[N_t(at) | \mathcal{F}_s] = \sum_{v \in \mathcal{U}(s)} \varphi(t-s, at - X_v(s))$$

to estimate this we restrict ourselves to an event where $|X_v(s)|$ is bounded deterministically

Let $E_s = \{ \forall v \in \mathcal{N}(s), |X_v(s)| \leq \sqrt{2ms} \}$. Recall that $P(E_s) \rightarrow 1$ as $s \rightarrow \infty$.

On E_s , uniformly in $v \in \mathcal{N}(s)$, (here the error term can be bounded deterministically)

$$\varphi(t-s, at - X_v(s)) = \sqrt{\frac{t-s}{2\pi}} \times \frac{(1+o(1))}{(at - X_v(s))} \times \exp\left(m(t-s) - \frac{(at - X_v(s))^2}{2(t-s)}\right) \quad \text{by the lemma above}$$

$$= \sqrt{\frac{t}{2\pi}} \frac{(1+o(1))}{at} \exp\left(m(t-s) - \frac{a^2}{2} \left(\frac{t^2}{t-s}\right) + \frac{a}{t-s} X_v(s) - \frac{X_v(s)^2}{2(t-s)}\right)$$

$$t \times \left(1 - \frac{s}{t}\right)^{-1} = t \left(1 + \frac{s}{t} + O\left(\frac{s^2}{t^2}\right)\right) \stackrel{(\ominus)}{=} t + s + o(1)$$

$$= aX_v(s) + O\left(\frac{s^2}{t}\right) \stackrel{(\ominus)}{=} o(1)$$

$$\stackrel{(\ominus)}{=} aX_v(s) + o(1)$$

using $s = o(\sqrt{t})$

$$= \frac{1+o(1)}{a\sqrt{2\pi t}} \exp\left(m(t-s) - \frac{a^2}{2}(t+s) + aX_v(s)\right)$$

Therefore, on E_s , $E[N_t(at) | \mathcal{F}_s] = \frac{1+o(1)}{a\sqrt{2\pi t}} e^{(m - \frac{a^2}{2})t} \sum_{v \in \mathcal{U}(s)} e^{aX_v(s) - (m + \frac{a^2}{2})s}$

Since $P(E_s) \rightarrow 1$, this shows ①.

$$= W_s^a \xrightarrow{t \rightarrow \infty} W_\infty^a$$

Step ②: Recall $N_t(at) = \sum_{v \in \mathcal{N}(s)} \Gamma_v$ with $\Gamma_v = \sum_{\substack{u \in \mathcal{N}(t) \\ u \geq v}} \mathbb{1}_{X_u(t) \geq at}$

$$\text{So } N_t(at) - \mathbb{E}[N_t(at) | \mathcal{F}_s] = \sum_{v \in \mathcal{N}(s)} (\underbrace{\Gamma_v - \mathbb{E}[\Gamma_v | \mathcal{F}_s]}_{\text{given } \mathcal{F}_s, \text{ indep. and centered r.v.}})$$

By von Bahr-Esseen inequality (applied given \mathcal{F}_s): for any $p \in [1, 2]$,

$$\mathbb{E}[|N_t(at) - \mathbb{E}[N_t(at) | \mathcal{F}_s]|^p | \mathcal{F}_s] \leq \sum_{v \in \mathcal{N}(s)} \mathbb{E}[|\Gamma_v - \mathbb{E}[\Gamma_v | \mathcal{F}_s]|^p | \mathcal{F}_s]$$

$$\begin{aligned} |X - \mathbb{E}[X]|^p &\leq 2(|X|^p + \mathbb{E}[|X|^p]) \\ \text{so } \mathbb{E}[|X - \mathbb{E}[X]|^p] &\leq 2\mathbb{E}[|X|^p] \quad (\leq) \quad 2 \sum_{v \in \mathcal{N}(s)} \mathbb{E}[|\Gamma_v|^p | \mathcal{F}_s] \end{aligned}$$

$$\text{But } \forall x \in \mathbb{R}, \mathbb{1}_{x \geq at} \leq e^{a(x-at)}$$

$$\text{so } \Gamma_v \leq \sum_{\substack{u \in \mathcal{N}(t) \\ u \geq v}} e^{aX_u(t) - at} = e^{aX_v(s) + (m - \frac{a^2}{2})t - (m + \frac{a^2}{2})s} \underbrace{\sum_{\substack{u \in \mathcal{N}(t) \\ u \geq v}} e^{a(X_u(t) - X_v(s)) - (m + \frac{a^2}{2})(t-s)}}_{= W_s^a(v, t) \stackrel{(d)}{=} W_s^a}$$

Choose $p < \frac{2m}{a^2}$, then $(W_s^a)_{s \geq 0}$ is bounded in L^p ,

$$\text{so } \mathbb{E}[|\Gamma_v|^p | \mathcal{F}_s] \leq C e^{apX_v(s) + (m - \frac{a^2}{2})pt - (m + \frac{a^2}{2})ps}$$

$$\mathbb{E}[|N_t(at) - \mathbb{E}[N_t(at) | \mathcal{F}_s]|^p | \mathcal{F}_s] \leq 2C e^{(m - \frac{a^2}{2})pt} e^{(m + \frac{ap}{2})s - (m + \frac{a^2}{2})ps} W_s^{ap}$$

Taking the expectation:

$$\mathbb{E}\left[\sqrt{t} e^{-(m - \frac{a^2}{2})t} (N_t(at) - \mathbb{E}[N_t(at) | \mathcal{F}_s])^p\right] \leq 2C t^{p/2} e^{\overbrace{(p-1)(\frac{a^2}{2}p - m)s}^{< 0}}$$

$$\xrightarrow{t \rightarrow \infty} 0 \quad \text{because } \frac{s}{\log t} \rightarrow \infty. \quad \square$$

Exercise 1: Back to Neveu's theorem

We prove here the case $|\lambda| = \sqrt{2m}$ (and give a new proof of the case $|\lambda| > \sqrt{2m}$).

1. Prove that $W_\infty^\lambda = e^{\lambda X_\phi(\tau_\phi) - (\frac{\lambda^2}{2} + m)\tau_\phi} (W_\infty^{\lambda,1} + \dots + W_\infty^{\lambda,L_\phi})$, where $(W_\infty^{\lambda,i})_{i \geq 1}$ are iid with the same law as W_∞^λ and independent of $(\tau_\phi, X_\phi(\tau_\phi), L_\phi)$.

2. For $p \in (0, 1)$, show that $\mathbb{E}[(W_\infty^{\lambda,1} + \dots + W_\infty^{\lambda,L_\phi})^p] = \left(1 + p\left(\frac{\lambda^2}{2} + m\right) - \frac{(\lambda p)^2}{2}\right) \mathbb{E}[(W_\infty^\lambda)^p]$.

3. For $|\lambda| > \sqrt{2m}$, deduce that $W_\infty^\lambda = 0$ a.s.

4. Show that $\mathbb{E}[(W_\infty^{\lambda,1} + \dots + W_\infty^{\lambda,L_\phi})^p] = (m+1) \mathbb{E}[W_\infty^{\lambda,1} (W_\infty^{\lambda,1} + \dots + W_\infty^{\lambda,L_\phi})^{p-1}]$

5. Combining 2. and 4. and looking at the derivative at $p = 1^-$, prove that $(m+1) \mathbb{E}\left[W_\infty^{\lambda,1} \log\left(\frac{W_\infty^{\lambda,1} + \dots + W_\infty^{\lambda,L_\phi}}{W_\infty^{\lambda,1}}\right)\right] = \left(m - \frac{\lambda^2}{2}\right) \mathbb{E}[W_\infty^\lambda]$

6. For $|\lambda| = \sqrt{2m}$, deduce that $W_\infty^\lambda = 0$ a.s.

Exercise 2: Optimality of the p in Neveu's theorem for $|\lambda| < \sqrt{2m}$

Let $|\lambda| < \sqrt{2m}$. Using question 2 of exercise 1, prove that, for $p > \frac{\lambda^2}{2m}$, $\mathbb{E}[(W_\infty^\lambda)^p] = +\infty$.

Exercise 3: von Bahr - Esseen inequality

Let $p \in [1, 2]$.

1.a. For $x \geq 0$, prove that $2(|x|^p + 1) - |x-1|^p - |x+1|^p \geq 0$.

1.b. For $x, y \in \mathbb{R}$, prove that $|x+y|^p + |x-y|^p \leq 2(|x|^p + |y|^p)$

1.c. For $X, Y \in L^p$ independent with Y symmetric, prove that $\mathbb{E}[|X+Y|^p] \leq \mathbb{E}[|X|^p] + \mathbb{E}[|Y|^p]$.

2. For $X, Y \in L^p$ such that $\mathbb{E}[Y|X] = 0$, prove that $\mathbb{E}[|X|^p] \leq \mathbb{E}[|X+Y|^p]$.

3. Prove von Bahr - Esseen inequality.

Hint: Use 3.a. to get $\mathbb{E}[|\sum_{k=1}^n X_k|^p] \leq \mathbb{E}[|\sum_{k=1}^{n-1} X_k + X_n - X'_n|^p]$ where X'_n is independent of (X_1, \dots, X_n) and has the same law as X_n .