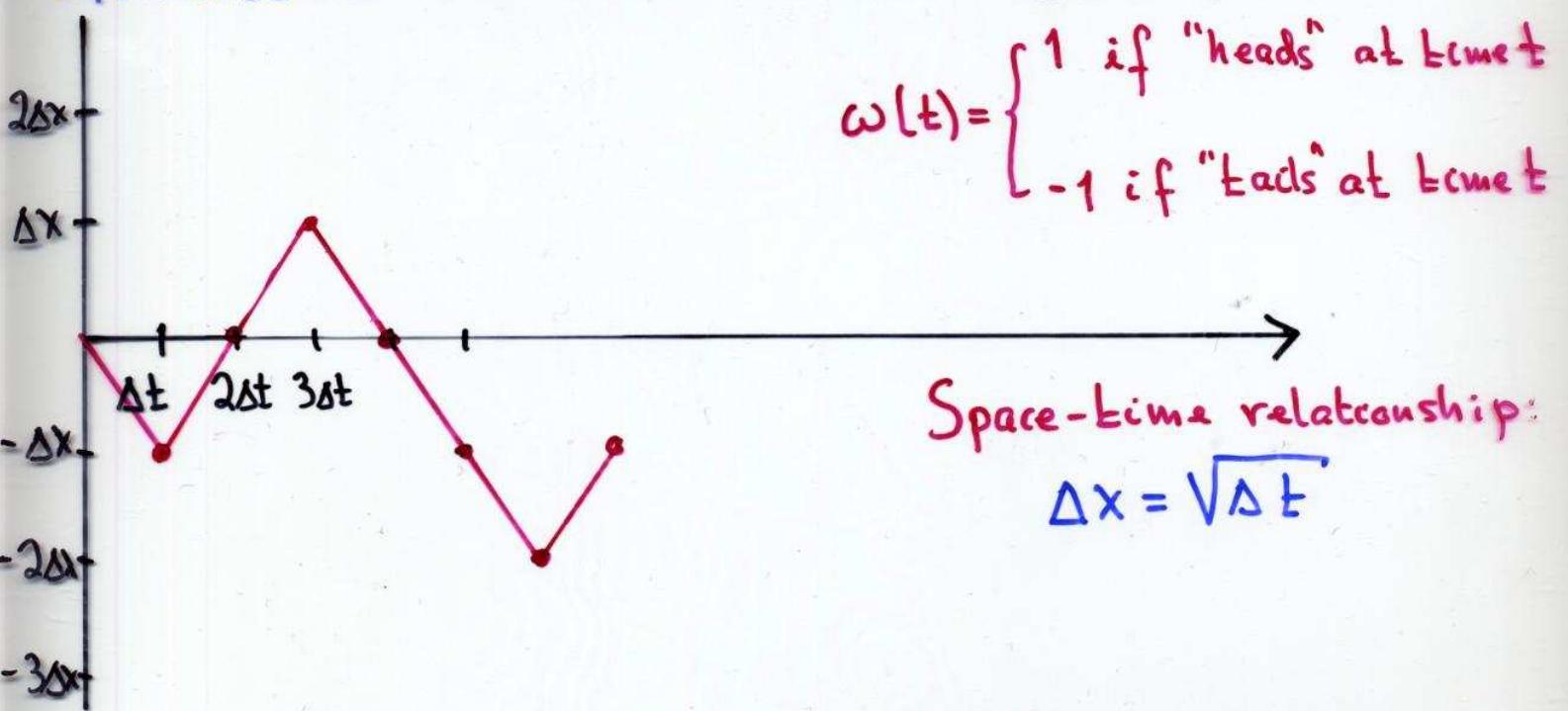


NONLINEAR STOCHASTIC INTEGRALS FOR HYPERFINITE LÉVY PROCESSES

(<http://folk.uio.no/Lindstro>)

1. Stochastic integration

Anderson's Brownian motion: $\Delta t \approx 0, \Delta x \approx 0$



Formal definition: $B: \Omega \times T \rightarrow {}^* \mathbb{R}$ given by

$$B(\omega, t) = \sum_{s=0}^{t-\Delta t} \omega(s) \sqrt{\Delta t}$$

If $\underline{X}(\omega, t)$ is another stochastic process, the stochastic integral $\int \underline{X} dB$ is defined by:

$$\int_0^t \underline{X} dB = \sum_{\delta < t} \underline{X}(\omega, \delta) \Delta B(\omega, \delta)$$

In order to get a decent theory, one "has to" require that \underline{X} is nonanticipating, i.e.

$$\underline{X}(\omega, \delta) = \underline{X}(\omega', \delta)$$

whenever $\omega(r) = \omega'(r)$ for all $r \leq \delta$ (\underline{X} can't look into the future)

Extension to higher dimensions:

$$B(\omega, t) = \begin{pmatrix} B_1(\omega, t) \\ B_2(\omega, t) \\ \vdots \\ B_d(\omega, t) \end{pmatrix}$$

independent versions of Anderson's process

$$\int_0^t \underline{X} dB = \sum_{\delta=0}^t \underbrace{\underline{X}(\omega, \delta)}_{\text{values in } (m \times d) \text{-matrices over } \mathbb{R}} \underbrace{\Delta B(\omega, \delta)}_{\text{values in } \mathbb{R}^d}$$

Hence the increments of the integral process

$$I(\omega, t) = \int_0^t \bar{X} dB = \sum_{\Delta t} \bar{X}(\omega, \Delta t) \Delta B(\omega, \Delta t)$$

Are produced by letting matrices $\bar{X}(\omega, \Delta t)$ act on the increments $\Delta B(\omega, \Delta t)$ of the original process, i.e. by applying linear functions to the increments $\Delta B(\omega, \Delta t)$.

QUESTION: What happens if we let nonlinear functions act on the increments?

But first we need to extend the class of integrator processes.

2. Hyperfinite Lévy processes

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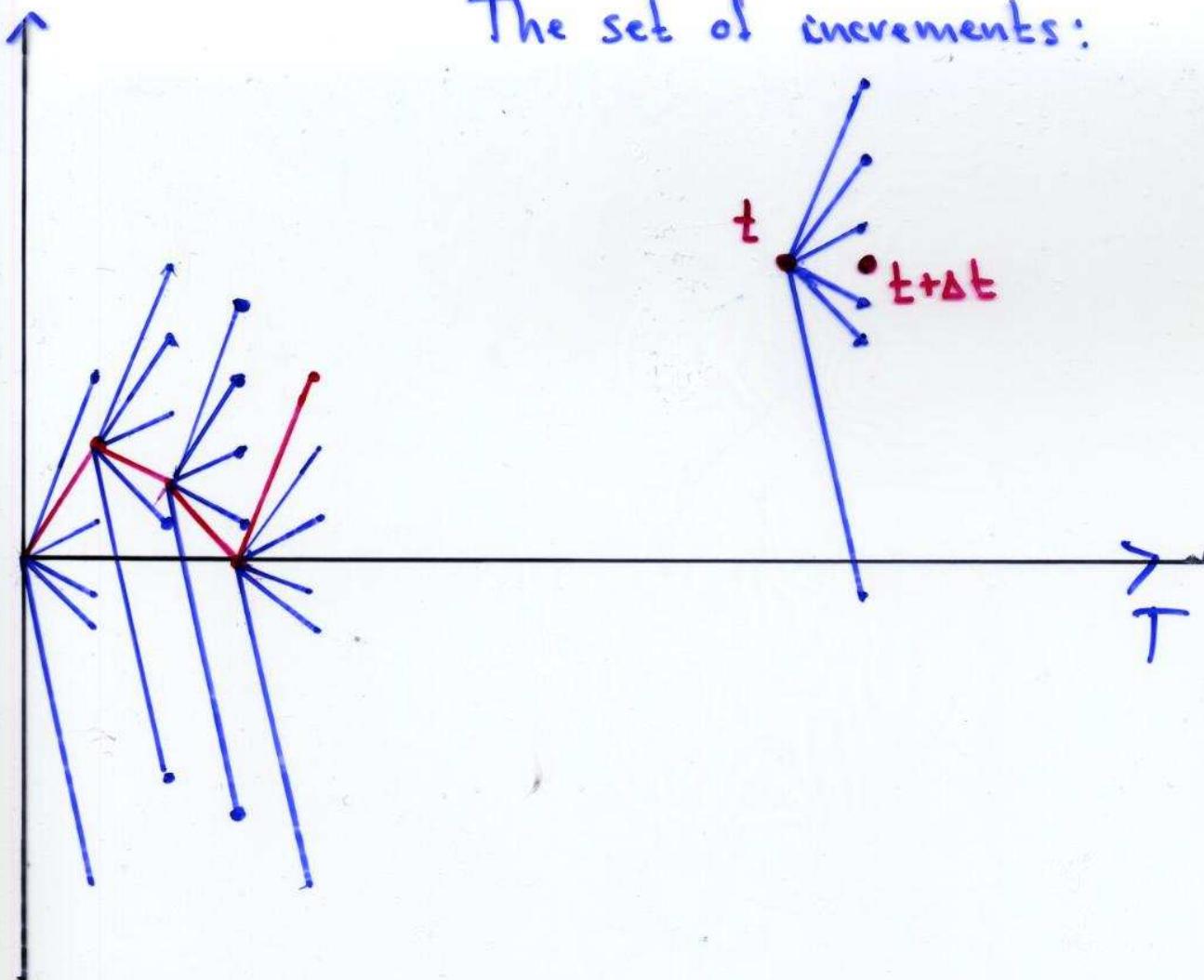
In Anderson's case, the process has only two options at each time t : it may go up $\sqrt{\Delta t}$ with probability $1/2$ or down $\sqrt{\Delta t}$ with probability $1/2$.

Let us extend the possibilities:

Fix a hyperfinite set $A \subseteq {}^*\mathbb{R}^d$ of increments and a corresponding set $\{p_a\}_{a \in A}$ of transition probabilities such that $\sum_{a \in A} p_a = 1$ (and $p_a \geq 0$ for all $a \in A$)

The idea is that at each instant $t \in T$, the process will choose its next increment to be a with probability p_a (independent of its past)

The set of increments:



Formally: An internal process $L: \Omega \times T \rightarrow^* \mathbb{R}^d$ is a hyperfinite random walk with increments A and transition probabilities $\{p_a\}_{a \in A}$ if:

- (i) $L(0) = 0$
- (ii) The increments $\Delta L(0), \Delta L(\Delta t), \Delta L(2\Delta t), \dots$ are $*$ -independent
- (iii) $P[\Delta L(\omega, t) = a] = p_a$ for all $t \in T$ and all $a \in A$.

So far we have no size restrictions, and a hyperfinite random walk may jump around in the infinite far. We want the process to stay finite for all finite times in the following sense:

Definition: A hyperfinite random walk L is called a **hyperfinite Lévy process** if there is a set $\Omega' \subset \Omega$ of Loeb measure 1 such that $L(\omega, t)$ is finite for all $\omega \in \Omega'$ and all finite $t \in T$.

(There are better criteria available!)

3. Splitting infinitesimals

In many situations it would be convenient to separate the infinitesimal increments of L from the noninfinitesimal, but it is impossible to do this in an internal way.

It is, however, possible to find infinitesimals η which are so large that infinitesimal increments larger than η only contribute to the process in a negligible way. Such infinitesimals η are called splitting infinitesimals.

Formal definition: $\eta \approx 0$ is a splitting infinitesimal if

$$\text{S-lim}_{\delta \rightarrow 0} \left(\frac{1}{\delta t} \sum_{\eta \leq |a| < \delta} |a|^2 p_a \right) = 0$$

4. Nonlinear stochastic integrals.

We want to define stochastic integrals

$$I(\omega, t) = \sum_{\Delta \in \mathcal{F}} \varphi(\omega, \Delta L_\Delta, \Delta)$$

Where φ acts nonlinearly on ΔL_Δ .

We need some conditions on φ :

(i) $\varphi(\omega, 0, \Delta) = 0$

(ii) φ is nonanticipating in the sense that

$$\varphi(\omega, x, \Delta) = \varphi(\omega^1, x, \Delta) \text{ if } \omega(r) = \omega^1(r) \text{ for all } r \leq \Delta.$$

(iii) Integrability and differentiability conditions
on φ

The strategy is to decompose $I(\omega, t)$ into simpler parts which we know how to handle.

Let us calculate! We choose a splitting infinitesimal η :

$$I(\omega, t) = \sum_{\Delta < t} \varphi(\omega, \Delta L_\Delta, \Delta) =$$

$$\sum_{\Delta < t} \varphi(\omega, \Delta L_\Delta^{>n}, \Delta) + \sum_{\Delta < t} \varphi(\omega, \Delta L_\Delta^{\leq n}, \Delta)$$

$$= \sum_{\Delta < t} \{ \varphi(\omega, \Delta L_\Delta^{>n}, \Delta) - \nabla \varphi(\omega, 0, \Delta) \cdot \Delta L_\Delta^{>n} \}$$

$$+ \sum_{\Delta < t} \varphi(\omega, \Delta L_\Delta^{\leq n}, \Delta) + \sum_{\Delta < t} \nabla \varphi(\omega, 0, \Delta) \cdot \Delta L_\Delta^{>n}$$

$$\approx \sum_{\Delta < t} \{ \varphi(\omega, \Delta L_\Delta^{>n}, \Delta) - \nabla \varphi(\omega, 0, \Delta) \Delta L_\Delta^{>n} \}$$

$$+ \sum_{\Delta < t} \{ \varphi(\omega, 0, \Delta) + \nabla \varphi(\omega, 0, \Delta) \cdot \Delta L_\Delta^{\leq n} + \frac{1}{2} \sum_{i,j} \frac{\partial^2 \varphi}{\partial x_i \partial x_j}(\omega, 0, \Delta) \Delta L_i^{\leq n}(\Delta) \Delta L_j^{\leq n}(\Delta) \}$$

$$+ \sum_{\Delta < t} \nabla \varphi(\omega, 0, \Delta) \Delta L_\Delta^{>n}$$

$$= \sum_{\Delta < t} \{ \varphi(\omega, \Delta L_\Delta^{>n}, \Delta) - \nabla \varphi(\omega, 0, \Delta) \Delta L_\Delta^{>n} \}$$

$$+ \sum_{\Delta < t} \nabla \varphi(\omega, 0, \Delta) \Delta L_\Delta + \frac{1}{2} \sum_{\Delta < t} \sum_{i,j} \frac{\partial^2 \varphi}{\partial x_i \partial x_j}(\omega, 0, \Delta) \Delta L_i^{\leq n}(\Delta) \Delta L_j^{\leq n}(\Delta)$$

Ordinary stochastic integral

We thus have:

$$I(\omega, t) = \sum_{\Delta < t} \{ \varphi(\omega, \Delta L_{\Delta}^{>n}, \Delta) - \nabla \varphi(\omega, 0, \Delta) \Delta L_{\Delta}^{>n} \} \\ + \sum_{\Delta < t} \nabla \varphi(\omega, 0, \Delta) \Delta L_{\Delta} + \frac{1}{2} \sum_{\Delta < t} \sum_{i,j} \frac{\partial^2 \varphi}{\partial x_i \partial x_j} (\omega, 0, \Delta) \Delta L_i^{<n}(\Delta) \Delta L_j^{<n}(\Delta)$$

and need to control the first and the last term on the right.

First term: Using Taylor's formula and the definition of splitting infinitesimal, we can prove that this term is finite and equal to

$$\text{S-lim } \underset{\delta \downarrow 0}{\circ} \left(\sum_{\Delta < t} \{ \varphi(\omega, \Delta L_{\Delta}^{>\delta}, \Delta) - \nabla \varphi(\omega, 0, \Delta) \Delta L_{\Delta}^{>\delta} \} \right)$$

Hence it makes "standard sense".

Last term: Using martingale theory, we can prove that it is finite and infinitely close to

$$\frac{1}{2} \sum_{i,j} C_{ij} \sum_{\Delta=0}^t \frac{\partial^2 \varphi}{\partial x_i \partial x_j} (\omega, 0, \Delta) \Delta t$$

where is the covariance matrix

$$C_{ij} = \frac{1}{\Delta t} E [\Delta L_i^{<n}(t) \Delta L_j^{<n}(t)]$$

Thus we end up with

$$\begin{aligned}
 I(\omega, t) &= \sum_{\Delta=0}^t \varphi(\omega, \Delta L_\Delta, \Delta) \approx \\
 &\approx \sum_{\Delta < t} \left\{ \varphi(\omega, \Delta L_\Delta^{>n}, \Delta) - \nabla \varphi(\omega, 0, \Delta) \Delta L_\Delta^{>n} \right\} \\
 &+ \int_0^t \nabla \varphi(\omega, 0, \Delta) dL_\Delta + \frac{1}{2} \sum_{i,j} C_{ij} \int_0^t \frac{\partial^2 \varphi}{\partial x_i \partial x_j}(\omega, 0, \Delta) ds
 \end{aligned}$$

We may think of this a Sum Formula for infinitely many, mostly infinitesimal terms $\varphi(\omega, \Delta L_\Delta, \Delta)$.

This suggests the idea of a corresponding Product Formula for $\prod_{\Delta < t} \gamma(\omega, \Delta L_\Delta, \Delta)$ where $\gamma(\omega, \Delta L_\Delta, \Delta)$ is (mostly) infinitely close to 1.

5. The Product Formula

Assume that $\gamma(\omega, 0, \delta) = 1$. To find an expression for $\prod_{\Delta < t} \gamma(\omega, \Delta L_\Delta, \delta)$, we write

$$\prod_{\Delta < t} \gamma(\omega, \Delta L_\Delta, \delta) = e^{\sum_{\Delta < t} \ln(\gamma(\omega, \Delta L_\Delta, \delta))}$$

and apply our previous work to the function $c_\theta(\omega, \Delta L_\Delta, \delta) = \ln(\gamma(\omega, \Delta L_\Delta, \delta))$. We get

$$\begin{aligned} & \prod_{\Delta < t} \gamma(\omega, \Delta L_\Delta, \delta) \approx \\ & \approx \prod_{\Delta < t} \left\{ \gamma(\omega, \Delta L_\Delta^{>n}, \delta) e^{-\nabla \gamma(\omega, 0, \delta) \cdot \Delta L_\Delta^{>n}} \right\} \times \\ & \times e^{\int_0^t \nabla \gamma(\omega, 0, \delta) dL_\Delta + \frac{1}{2} \sum_{i,j} C_{i,j} \int_0^t \left[\frac{\partial^2 \gamma}{\partial x_i \partial x_j} - \frac{\partial \gamma}{\partial x_i} \frac{\partial \gamma}{\partial x_j} \right](\omega, s) ds} \end{aligned}$$

6. Geometric Lévy processes

A geometric Lévy process is a solution of a stochastic differential equation

$$d\bar{X}_t = \bar{X}_t \cdot \Gamma(\omega, dL_t, t)$$

In our terms

$$\Delta \bar{X}_t = \bar{X}_t \cdot \Gamma(\omega, \Delta L_t, t)$$

or, equivalently,

$$\bar{X}_{t+\Delta t} = \bar{X}_t (1 + \Gamma(\omega, \Delta L_t, t))$$

By induction

$$\bar{X}_t = \bar{X}_0 \prod_{0 < s < t} (1 + \Gamma(\omega, \Delta L_s, s))$$

Using the product formula with $\gamma(\omega, \Delta L_s, s) = 1 + \Gamma(\omega, \Delta L_s, s)$, we get

$$\begin{aligned} \bar{X}_t &\approx \bar{X}_0 \prod_{0 < s < t} (1 + \Gamma(\omega, \Delta L_s^{\gamma}, s)) e^{-T_x(\omega, 0, s) \Delta L_s^{\gamma}} \\ &\times \sqrt{e^{\int_0^t T_x(\omega, 0, s) dL_s + \frac{\sigma^2}{2} \int_0^t (T_{xx} - T_x^2)(\omega, 0, s) ds}} \end{aligned}$$

which extends previous standard results.

E. Transforming increments.

Given a hyperfinite Lévy process with increments A and transition probabilities $\{p_a\}_{a \in A}$, we can create a new one by applying a function φ to the elements in A . The new process will have increments $\{\varphi(a)\}_{a \in A}$ and transition probabilities p_a . It is given by

$$\varphi L(\omega, t) = \sum_{s < t} \varphi(\Delta L_s) \approx$$

$$\approx \nabla \varphi(0) \cdot L(t) + \frac{t}{2} \sum_{i,j} C_{i,j}^n \frac{\partial^2 \varphi}{\partial x_i \partial x_j}(0)$$

$$+ \sum_{s=0}^t \{ \varphi(\Delta L_s^{>n}(\omega, s)) - \nabla \varphi(0) \Delta L_s^{>n}(\omega, s) \}$$

8. Transforming probabilities.

We may also change a hyperfinite Lévy process by keeping the increments and changing the transition probabilities.

If the new probabilities q_a are given by

$$q_a = \gamma(a) p_a$$

we need

$$\sum_{a \in A} \gamma(a) p_a = 1$$

The density of the new measure Q with respect to the old measure P on the timeline $T_t = \{ \delta \in T : \delta < t \}$ is

$$D(\omega, t) = \prod_{\delta=0}^t \gamma(\Delta L(\omega, \delta))$$

Applying the product formula one gets:

$$D(\omega, t) = \prod_{\delta < t} \gamma^{\rho} (\Delta L^n(\omega, \delta)) e^{-\nabla \varphi(0) \Delta L^n(\omega, \delta)}$$

$$\times e^{\lambda t + \nabla \psi(0) \cdot L(\omega, t) + \frac{t}{2} \sum_{i,j} C_{ij} \left[\frac{\partial \psi}{\partial x_i}(\omega) - \frac{\partial \psi}{\partial x_i}(0) \right] \frac{\partial \psi}{\partial x_j}(\omega)}$$

Where $\lambda := \frac{\psi(0) - 1}{\Delta t}$ is finite.

There are similar (but more complex) applications to equivalent martingale measures.