## Fractal Geometry

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## I. Stochastic Calculus

**Definition.** Given a measure space  $(\Omega, \mathcal{F}, \mathbb{P})$ , a measurable function  $f : \Omega \to \mathbb{R}$  is called a **random variable**.

**Definition.** A **stochastic process**  $X = \{X_t\}_{t \in T}$  is a collection of random variables defined on some probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ .

Typically  $t \in \mathbb{Z}^+$  or  $t \in \mathbb{R}^+$  (including 0); t is a discrete or continuous time parameter. Given some  $\omega \in \Omega$  the map  $t \mapsto X_t(\omega)$  is called a **realization** or **path** of this process. We will regard  $\{X_t\}_{t\geq 0}$  as a random element in some path space, equipped with a proper  $\sigma$ -algebra and probability.

Consider  $X_t(\omega)$  as a function  $X:[0,\infty)\times\Omega\to\mathbb{R}$  equipped with the product  $\sigma$ -algebra.

**Definition.** The **distribution** of a stochastic process is the collection of all its finite-dimensional distributions.

Two processes *X* and *Y* can be "the same" in different senses:

**Definition.** Two process  $X = \{X_t\}_{t \geq 0}$  and  $Y = \{Y_t\}_{t \geq 0}$  are called **distinguishable** if almost all their sample paths agree; in other words,  $\mathbb{P}(X_t = Y_t, 0 \leq t < \infty) = 1$ . We say that Y is a **modification** of X if for each  $t \geq 0$  we have  $\mathbb{P}(X_t = Y_t) = 1$ . Finally, X and Y are said to have the **same distribution** if all the finite dimensional distributions agree. In other words, if for all  $n \in \mathbb{N}$  and  $0 \leq t_1 < t_2 < \cdots < t_n < \infty$ , we have  $(X_{t_1}, \dots, X_{t_n}) \stackrel{d}{=} (Y_{t_1}, \dots, Y_{t_n})$ .

*Example.* Let X be a continuous stochastic process and N a Poisson point process on  $[0, \infty)$ . Then define

$$Y_t := \begin{cases} X_t & : t \notin N \\ X_t + 1 & : t \in N \end{cases}$$

Thus  $\mathbb{P}(X_t = Y_t) = 1$  for all t, so X is a modification of Y. However,  $\mathbb{P}(X_t = Y_t, t \ge 0) = 0$ , so that X and Y are not indistinguishable.

A filtration formalizes the idea of "information acquired over time".

**Definition.** Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space. A **filtration** is a non-decreasing family  $\{\mathcal{F}_t\}_{t\geq 0}$  of sub- $\sigma$ -algebras of  $\mathcal{F}$  so that  $\mathcal{F}_s \subseteq \mathcal{F}_t \subseteq \mathcal{F}$  for  $0 \leq s < t < \infty$ . We write  $F_{\infty} = \sigma(\bigcup_{t>0} \mathcal{F}_t)$ .

Let  $\{X_t\}_{t\geq 0}$  be a stochastic process. The filtration generated by  $\{X_t\}_{t\geq 0}$  is  $\{\sigma(X_s:0\leq s\leq t)\}_{t\geq 0}$ , in other words  $\mathcal{F}_t$  is the smallest  $\sigma$ -algbra which makes  $X_s$  measurable for all  $s\in [0,t]$ .

**Definition.** A stochastic process  $\{X_t\}_{t\geq 0}$  is called **adapted** to a filtration  $\{\mathcal{F}_t\}_{t\geq 0}$  if  $X_t$  is  $\mathcal{F}_t$ -measurable for every  $t\geq 0$ .

The filtration generated by  $\{X_t\}_{t\geq 0}$  is the smallest filtration which makes  $(X_t)_{t\geq 0}$  adapted.

A filtration  $\{\mathcal{F}_t\}_{t\geq 0}$  is said to satisfy the "usual condition" if

- 1. It is right-continuous:  $\lim_{s\to t^+} := \bigcap_{s>t} \mathcal{F}_s = \mathcal{F}_t$
- 2.  $\mathcal{F}_0$  contains all the  $\mathbb{P}$ -null events in  $\mathcal{F}$ .

## 1 Martingale Theory

Consider a filtered space  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \in S})$  where  $S = \mathbb{N}$  or  $S = \mathbb{R}^+$ .

**Definition.** A random time T is called a stopping time if  $\{T \leq t\} \in \mathcal{F}_t$  ("we know it happens when it happens").

*Example.* (i) Constants are trivial stopping times.

- (ii) Last hit a constant before *N* is not a stopping time
  - **1.1 Proposition.** If S, T are stopping times,  $T \vee S$ ,  $T \wedge S$ , T + S are stopping times.

PROOF That  $T \wedge S$  and  $T \vee S$  are stopping times are trivial. For T + S,  $\{T + S > t\} = \{T = 0, S > t\} \cup \{0 < T \le t, T + S > t\} \cup \{T > t\}$ . It suffices to prove that

$$\{0 < T \le t < T + S > t\} = \bigcup_{\substack{r \in \mathbb{Q}^+ \\ 0 < r < t}} \{r < T \le t, S > t - r\}.$$

If there exists r with  $r < T \le t$ , then S > t = r and S + T > r + (t - r) = t, so  $\supseteq$  holds. Conversely, if  $0 < T \le t$  and  $T + S \ge t$ ; then there exists  $r \in \mathbb{Q}$  such that r < T and r + S > t. Hence  $r < T \le t$  and S > t - r.

**Definition.** The  $\sigma$ -algebra generated by a stopping time T is the collection of all the events A for which  $A \cap \{T \leq t\} \in \mathcal{F}_t$  for every  $t \geq 0$ . This is the "information you collect until the stopping time".

Exercise: show that the collection given in the definition above is actually a  $\sigma$ -algebra.

We write  $X_{T \wedge t}$  is a random variable evaluated at time  $T \wedge t$  (or T); in other words,  $(X_{T \wedge t})(\omega) = X_{T \wedge t}(\omega)$ . Then  $\{T_{T \wedge t}\}_{t \geq 0}$ , or  $X^T$ , is a stochastic process stopped at time t. **Definition.**