Lecture 2 (p43-49)

Constructing stochastic processes in continuous time

Motivation: Consider a stock with S=100. How much do we expect it to fluctuate?

	1 day	1 month	1 year
Reasonable range for S	100 ± 1	100 ± 5	100 ± 15
AS	1	5	15
DE	1	20	250
DS	1	J ₂₀ ≈ 1	$\frac{15}{\sqrt{250}} \approx 1$

Below we construct a stochastic process with variance t (st. dev. VE). Fix a time interval [0,T].

Stage 1: Let X'=0. At t=0, toss a coin.

If heads, let X1 = VT.

if tail, let x' = - JT.

Stage 2: Let X = 0. Toss a coin at t=0.

Head $\Rightarrow X_{\perp}^2 = \sqrt{\frac{1}{2}}$

Tail $\Rightarrow \chi_{\pm}^2 = -\sqrt{\frac{T}{2}}$

Repeat at t= = adding/subtracting 1.



Stagen: Let $X_0^n = 0$. At each time $t_k = \frac{k}{n}T$ toss a coin. $X^{n} = X^{n} + Y_{k}, \text{ where } Y_{k} = \begin{cases} \sqrt{\frac{1}{n}} & \text{prob. } \frac{1}{2} \\ \sqrt{\frac{1}{n}} & \text{prob. } \frac{1}{2} \end{cases}$ Clearly, $E[X_{t_i}^n] = E[Y_0 + Y_1 + \dots + Y_{k-1}] = 0$. Also, $Var(X_{t_k}^n) = Var(Y_0) + ... + Var(Y_{k-1}) =$ $= \frac{1}{5} \cdot k = \frac{1}{5}$ When n->00, we obtain Brownian motion (or Wiener More formally, we make the following definition. Def 4.1 A stochastic process W is a Wiener process (Brownian motion) if i) W =0 ii) it has independent increments, i.e Wt-Wt, and Wt-Wt are independent if t,<t2<t3<t4 (ii) W_-W_ is N(0, Jt-s') iv) We is continuous.

Thm 4.2 timble is of infinite variation and nowhere differentiable.

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Our next goal: To define integrals

\$ 9, dlys, where g_t is a stochastic process

"determined" by { W_s : $0 \le s \le t$ }.

Def. 4.3 Let X be a stochastic process.

- i) An event A is f_{\pm}^{X} -measurable $(A \in f_{\pm}^{X})$ if it is possible to determine whether A has happened or not based on observations of $\{X_{s}; 0 \le s \le t\}$.
- ii) If a random variable \mathbb{Z} can be determined given observations of $\{X_s: 0 \leqslant s \leqslant t\}$, then we also write $\mathbb{Z} \in \mathbb{F}_t^{\times}$.
- iii) A stochastic process Y_t with $Y_t \in \mathcal{F}_t^{\times}$ for all $t \geqslant 0$ is adapted to the filtration \mathcal{F}_t^{\times} .

$$\underline{\dot{E}}_{X}$$
: 1. $A = \{X_s \leq 7 \text{ for all } s \leq 9\} \in \mathcal{F}_q^X$
2. $Z = \int_0^5 X_s ds \in \mathcal{F}_5^X$

3. Y = sup Ws is adapted to &

4. Y= SUP Ws is not adapted to xw.

Def 4.4 A process g_t belongs to L^2 if if g_t is adapted to g_t it g_t g_t

$Ex: \int E[w_s^2] ds = \int s ds = \frac{t^2}{2} < \infty, so W \in L^2.$

Stochastic integration

Assume gel2.

1. If g is simple, i.e.
$$g_s = g_{tk}$$
 for $s \in \{t_k, t_{k+1}\}$
where $0 = t_0 < t_1 < ... < t_n = t$, define
$$\int_{s=0}^{t} g_s dW_s := \int_{k=0}^{n-1} g_{tk} \left(W_{tk+1} - W_{tk}\right)$$

2. For a general
$$g \in d^2$$
, approximate g with simple g^n such that $\int_0^1 [(g_s^n - g_s)^2] ds \to 0$ as $n \to \infty$. Define $\int_0^1 g_s dW_s := \lim_{n \to \infty} \int_0^n g_s^n dW_s$ (limit in L^2)

Remarks 1. One can show that the limit

- · does not depend on the approximating sequence.
- 2. Forward increments are used!
- 3. Riemann-Stieltje integration not possible since W has paths of infinite variation.

Prop. 4.5 Assume g EL. Then

(i)
$$E\left[\int_{0}^{t} g_{s} dW_{s}\right] = 0$$

(ii)
$$E[(\frac{t}{s}g_s dW_s)^2] = \int E[g_s^2] ds$$
 (Ito isometry)

Pf: Assume g is simple (general case follows by 5) approximation) (i) $E[Sg_sdW_s] = E[Sg_t(W_{t_{k+1}} - W_{t_k})]$ $= \underbrace{\sum_{k=1}^{N-1} \left\{ \int_{\mathbb{R}^{N}} \left(W_{t_{k+1}} - W_{t_{k}} \right) \right\}}_{:}$ = Egt Elgt indep. = Egt Elyther - Wth = $(ii) E[(fg_sdW_s)^2] = E[(fg_sdW_s)^2]$ $= \sum_{k=0}^{N-1} E\left[g_{t_{k}}^{2} \left(W_{t_{k+1}} - W_{t_{k}}\right)^{2}\right] + 2 \underbrace{E\left[g_{t_{k}} - W_{t_{k+1}} - W_{t_{k}}\right]}_{j < k} \left(W_{t_{j+1}} - W_{t_{j}}\right) \left(W_{t_{k+1}} - W_{t_{k}}\right)$ indep. $= \sum_{k=0}^{N-1} E[g_{t_k}] E[W_{t_{k+1}} - W_{t_k}]^2 + 2 \sum_{j < k} E[g_{t_j} g_{t_k} (W_{t_j} - W_{t_j})] E[W_{t_k}]^2$ $= \int E[g_s^2] ds$ Ex: Calculate JWs dWs. Solution: Let $g_t^n = W_{t_n}$ for $t \in [t_k, t_{k+1})$ Then $\int_{0}^{\infty} E[(g^{n} - W_{s})^{2}] ds = \sum_{k=0}^{\infty} \int_{0}^{\infty} E[(W_{s} - W_{t})^{2}] ds$ $= \underbrace{\sum_{k=1}^{N-1} \left(\underbrace{t_{k+1} - t_{ik}}_{2} \right)}_{} \rightarrow 0 \text{ as } \Delta t \rightarrow 0$

Thus grapproximates the integrand.

$$\sum_{k=0}^{N-1} W_{t_{k}} (W_{t_{k+1}} - W_{t_{k}}) = \frac{1}{2} \sum_{k=0}^{N-1} (W_{t_{k+1}}^2 - W_{t_{k}}^2 - W_{t_{k+1}}^2)$$

$$= \frac{1}{2} W_{t}^2 - \frac{1}{2} \sum_{k=0}^{N-1} (W_{t_{k+1}} - W_{t_{k}})^2$$
We have
$$E \left[\left(\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 - t \right)^2 \right] = E \left[\left(\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 - (t_{k+1} - t_{k}) \right)^2 \right]$$

$$= \sum_{k} E \left[\left(\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 - (t_{k+1} - t_{k}) \right) \right] + \sum_{k=0}^{N-1} E \left[\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 - (t_{k+1} - t_{k}) \right]$$

$$= \sum_{k} Var \left(\left(\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 - (t_{k+1} - t_{k}) \right) - \left(E \left[\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 \right] \right)$$

$$= \sum_{k=0}^{N-1} \left(\left(t_{k+1} - W_{t_{k}} \right)^4 - \left(E \left[\left(W_{t_{k+1}} - W_{t_{k}} \right)^2 \right] \right)$$

$$= \sum_{k=0}^{N-1} \left(\left(t_{k+1} - t_{k} \right)^2 - \left(t_{k+1} - t_{k} \right)^2 \right)$$

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 $\int W_s dW_s = \frac{1}{2}W_t^2 - \frac{t}{2}$