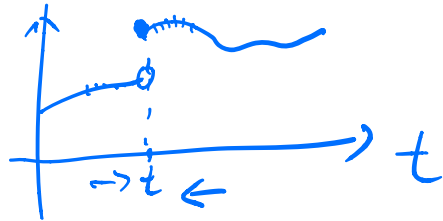


Continuous time stoch. processes (A deeper look)

$\left\{ \begin{array}{l} \text{continuous stoch. process} \\ \text{continuous-time stoch. processes. } X_t \quad t \in \mathbb{R}_+ \end{array} \right.$
 $\hookrightarrow \left(\begin{array}{l} X(\cdot, \omega) \text{ is a function on } \mathbb{R}_+ \\ \text{fix } \omega, \end{array} \right.$

"Cadlag": right-continuous & has left limit.



$$\forall t \quad \lim_{s \rightarrow t^+} X_s = X_t$$

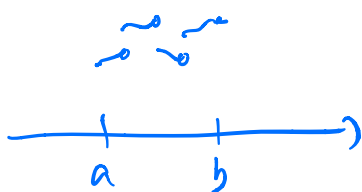
$$\lim_{s \rightarrow t^-} X_s \text{ exists.}$$

$C[0, \infty)$: space of all conti-functions

$D[0, \infty)$: - - - - - cadlag functions.

Rmks 1) in any compact time interval
for each $\varepsilon > 0$

a cadlag function has at most finitely many
dis-cont. of size $\geq \varepsilon$.



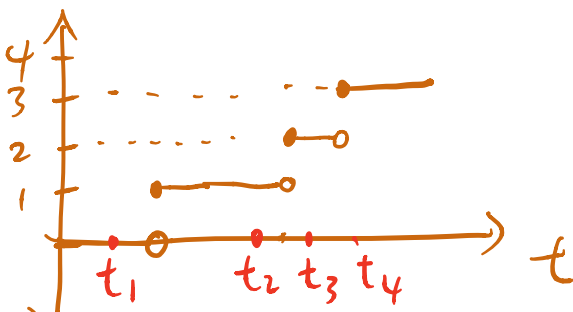
(If not true, you would have
right limit point or left limit point)

- 2) a cadlag function can have at most countably many discontinuities.
(take $\varepsilon = \frac{1}{n}$)

• Examples of cadlag stock processes

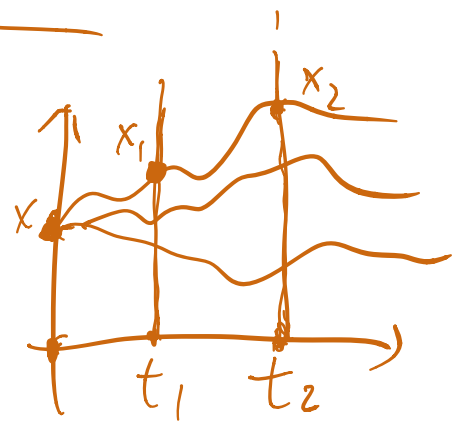
1) Brownian motion

2) Poisson process.



Poisson(λ)

$$h(\lambda, k) = e^{-\lambda} \frac{\lambda^k}{k!}$$



$$\frac{p(t, x, x_1)}{p(t_2 - t_1, x_1, x_2)}$$

$$P(X_{t_1} = k_1, \dots, X_{t_n} = k_n)$$

$$= \{ h(\lambda t_1, k_1) h(\lambda(t_2 - t_1), k_2 - k_1) \dots h(\lambda(t_n - t_{n-1}), k_n - k_{n-1})$$

if $k_1 < k_2 < \dots < k_n$

called Poisson process w. parameter λ

• Filtration: a collection of σ -algebras $\{\mathcal{F}_t\}$, satisfying

$$\mathcal{F}_s \subseteq \mathcal{F}_t \subseteq \mathcal{F}, \quad \forall s \leq t,$$

$\swarrow (\Omega, \mathcal{F}, \mathbb{P})$

• A stochastic process X is adapted to a filtration $\{\mathcal{F}_t\}$ if X_t is \mathcal{F}_t -measurable for all $t \geq 0$.

• Given $\{X_t\}$ generates a filtration.

$$\mathcal{F}_t = \sigma(X_s : s \leq t)$$

the smallest σ -alg

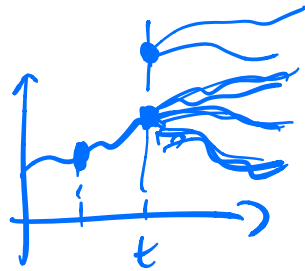
s.t. X_s is \mathcal{F}_t -measurable for all $s \leq t$

• Markov process;

A stoch process X adapted to $\{\mathcal{F}_t\}$
 is a Markov process, wrt $\{\mathcal{F}_t\}$ if

$$\mathbb{E}[f(X_{t+s}) | \mathcal{F}_t] = \mathbb{E}[f(X_{t+s}) | X_t]$$

\uparrow history before t
 \uparrow



• Martingale

X adapted to $\{\mathcal{F}_t\}$ is "martingale"
 w.r.t. $\{\mathcal{F}_t\}$. if

$$\mathbb{E}[\underline{X(t+s)} | \mathcal{F}_t] = X(t)$$

\uparrow
 $\forall t, s \geq 0.$

"fair game"

Lem: BM is Markov & Martingale.

Pf: $\mathcal{F}_t = \sigma(B_s, s \leq t)$

$$\mathbb{E}(B_{t+s} | \mathcal{F}_t)$$

$$= \mathbb{E} (B_{t+s} - B_t + B_t \mid \mathcal{F}_t)$$

$$= \mathbb{E} (\underline{B_{t+s}} - \underline{B_t} \mid \underline{\mathcal{F}_t}) + \mathbb{E} (B_t \mid \mathcal{F}_t)$$

$$= \mathbb{E} (\underline{B_{t+s}} - B_t) + \mathbb{E} (B_t \mid \mathcal{F}_t)$$

$$= 0 + B_t = B_t \Rightarrow \text{martingale}$$

ref: Timo Seppäläinen's notes,
prop 2.31 b)

Def (stopping time)

A random variable τ with values in $[0, \infty]$
 is an $\{\mathcal{F}_t\}$ -stopping time if

$$\underline{\{ \tau \leq t \}} \in \mathcal{F}_t \quad \forall t \geq 0$$

properties: $\tau_1 \vee \tau_2, \tau_1 \wedge \tau_2$
 are stopping times.

