

lecture 10 : Jump-Diffusion Processes

Note Title

4/17/2012

- "diffusion" \Rightarrow Brownian motion component
(integral)
- we shall consider processes with finitely many jumps in each finite time interval.
- fundamental jump process is the **Poisson process**

Recall : {

- all jumps of a Poisson process are of size 1
- the **Compound Poisson process** is like a Poisson process except that the jumps are of random size.

Recall facts about the Poisson Process (pp)

- to construct a PP we begin with a sequence τ_1, τ_2, \dots of independent r.v $\sim \exp(\lambda)$ (the "interarrival times")

$$S_n = \sum_{k=1}^n \tau_k : \text{the time of the } n^{\text{th}} \text{ jump.}$$

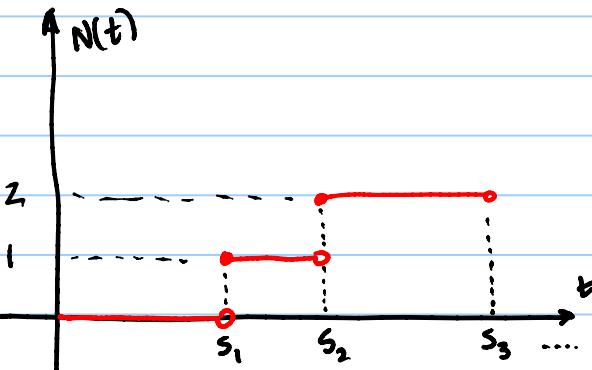
- the PP : $N(t)$ counts the number of jumps that occur at or before time t .

$$N(t) = \begin{cases} 0 & \text{if } 0 \leq t < s_1 \\ 1 & \text{if } s_1 \leq t < s_2 \\ \dots \end{cases}$$

- right-continuous process $(N(t) = \lim_{s \downarrow t} N(s))$

- $\mathcal{F}_t = \sigma(N(s); s \leq t)$

- we say $N(t)$ has intensity λ



The Distribution of PP's increments

$$\cdot S_n = \sum_{k=1}^n \zeta_k \quad \left. \begin{array}{l} \\ \zeta_1, \zeta_2, \dots \sim \exp(\lambda), \text{ indep} \end{array} \right\} \Rightarrow S_n \sim \Gamma(n, \lambda)$$

more precisely: the density of S_n is:

$$g_n(s) = \frac{(\lambda s)^{n-1}}{(n-1)!} \lambda e^{-\lambda s} \quad \text{for } s \geq 0$$

(notice that for $n=1 \Rightarrow g_1(s) = \lambda e^{-\lambda s} \Rightarrow \exp(\lambda)$)

• $N(t) \sim \text{Poisson}(\lambda t)$

$$P(N(t) = k) = \frac{(\lambda t)^k}{k!} \cdot e^{-\lambda t} \quad k=0, 1, 2, \dots$$

Theorem Let $0=t_0 < t_1 < t_2 < \dots < t_n$ be given.

The increments $N(t_j) - N(t_{j-1}), N(t_2) - N(t_1), \dots, N(t_n) - N(t_{n-1})$

are stationary and independent and

$$P(N(t_{j+1}) - N(t_j) = k) = \frac{\lambda^k (t_{j+1} - t_j)^k}{k!} e^{-\lambda(t_{j+1} - t_j)}$$

$$W_t - W_s \sim N(0, t-s) \sim W_{t-s}$$

⇒ in general for any $s < t$:

$$N(t) - N(s) \sim N(t-s) \sim \text{Poisson}(\lambda(t-s))$$

Martingale Property

- $N(t)$ is not a martingale (easy)
- $M(t) = N(t) - \lambda t$: the compensated PP is martingale

Compound Poisson Process (CPP)

- $N(t)$: PP(λ) (Poisson process with intensity λ)
- Y_1, Y_2, \dots iid r.v. with $p = EY_i$
(Y_1, Y_2, \dots independent of each other and $N(t)$)

⇒ the CPP is : $Q(t) = \sum_{i=1}^{N(t)} Y_i \quad t \geq 0$

⇒ the jumps of CPP occur at the same time with the jumps of PP

⇒ the jump size of PP are always 1, however the jumps of CPP are of size $Y_1, Y_2, \dots, Y_n, \dots$

- like PP, the increments of CPP are independent and stationary

$$Q(t) - Q(s) = \sum_{i=N(s)+1}^{N(t)} Y_i \sim \sum_{i=1}^{N(t-s)} Y_i = Q(t-s)$$

(because $N(t) - N(s) \sim N(t-s)$)

$$\begin{aligned} E[Q(t)] &= E\left[\sum_{i=1}^{N(t)} Y_i\right] \\ &= E\left[E\left[\sum_{i=1}^{N(t)} Y_i \mid N(t)\right]\right] \\ &= E[p \cdot N(t)] = p\lambda t \end{aligned}$$

- The compensated CPP is a martingale

$$\rightsquigarrow Q(t) - p\lambda t$$

(easy to check)

Jump Processes and Their Integrals

(Ω, \mathcal{F}, P) $\{\mathcal{F}_t\}_{t \geq 0}$: filtration

- we say that W is a Brownian Motion relative to $\{\mathcal{F}_t\}$ if
 - W_t is \mathcal{F}_t measurable
 - $W_t - W_s$ is independent of \mathcal{F}_s
- we say that N is a PP relative to $\{\mathcal{F}_t\}$ if
 - $N(t)$ is \mathcal{F}_t measurable
 - $N(t) - N(s)$ is independent of \mathcal{F}_s

(same goes for CPP)

Jump Process

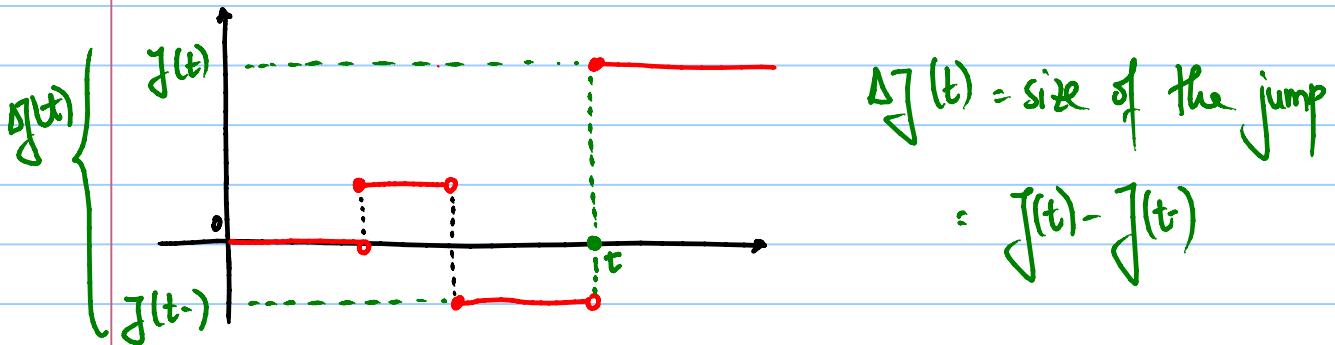
• we wish to define : $\int_0^t \phi_s dX_s$

where X can have jumps.

• assume all processes are adapted to $\{\mathcal{F}_t\}$

$$X(t) = X(0) + I(t) + R(t) + J(t)$$

- $I(t) = \int_0^t \Gamma(s) dW(s)$: stochastic integral
- $R(t) = \int_0^t \Theta(s) ds$: Riemann integral
- $X^c(t) = X(0) + I(t) + R(t)$: the continuous part of X
 $\rightsquigarrow \langle X^c \rangle_t = \int_0^t \Gamma^2(s) ds$: quadratic var. of X^c
 $(dX_t^c \cdot dX_t^c = \Gamma^2(t) dW_t \cdot dW_t = \Gamma^2(t) dt)$
- $J(t)$: adapted, right-continuous **pure jump process**
 $\rightsquigarrow J(0) = 0$; $J(t) = \lim_{s \downarrow t} J(s)$
 \rightsquigarrow the left continuous version of such a process
 will be denoted $J(t-)$



- assume J has finitely many jumps on each $(0, \bar{t}]$.

Def: A process $\overbrace{X(t)}^{X^c(t)} = X(0) + I(t) + R(t) + J(t)$ is called a **jump process**.

- the jump process X is adapted, right continuous
- the left continuous version of X is

$$X(t^-) = X(0) + I(t) + R(t) + J(t^-)$$

- the jump size of X at time t is :

$$\Delta X(t) = X(t) - X(t^-) = J(t) - J(t^-) = \Delta J(t)$$

Def: let X be a jump process, ϕ adapted process

$$\rightsquigarrow \int_0^t \phi(s) dX(s) = \int_0^t \phi(s) \Gamma(s) dW_s + \int_0^t \phi(s) \Theta(s) ds + \sum_{s \leq t} \phi(s) \Delta J(s)$$

or

$$\rightsquigarrow \phi(t) dX(t) = \underbrace{\phi(t) \Gamma(t) dW(t) + \phi(t) \Theta(t) dt}_{\phi(t) dX^c(t)} + \phi(t) dJ(t)$$

$$\int_0^t \phi_s dX_s$$

- example : $X(t) = N(t) - \lambda t$ (compensated PP)

for this process $J(t) = 0$; $R(t) = -\lambda t$, $J(t) = N(t)$

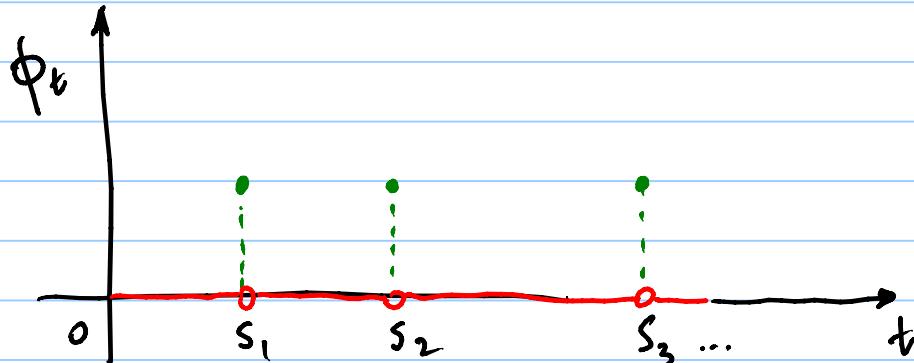
def $\phi(t) = \Delta N(t) = \begin{cases} 1 & \text{if } N \text{ jumps at } t \\ 0 & \text{otherwise} \end{cases} = N(t) - N(t^-)$

$$\left\{ \begin{array}{l} \int_0^t \phi(s) dX(s) = \int_0^t \phi(s) dR(s) = -\lambda \int_0^t \phi(s) ds = 0 \quad R(t) = -\lambda t \\ \int_0^t \phi(s) dJ(s) = \int_0^t \phi(s) dN(s) = \sum_{0 \leq s \leq t} (\Delta N(s))^2 = N(t) = \sum_{s \leq t} \phi(s) \cdot \Delta N(s) \\ \int_0^t \phi(s) dX(s) = -\lambda \int_0^t \phi(s) ds + \int_0^t \phi(s) dN(s) = N(t) \end{array} \right.$$

Theorem : Assume that the jump process $X(t)$ is a martingale, $\phi(t)$ is left-continuous and adapted with

$$E \int_0^t \phi^2(s) ds < \infty \quad (\forall) t \geq 0$$

Then the stochastic integral $\int_0^t \phi(s) dX(s)$ is also a martingale.



Quadratic Variation

recall the definition $\langle X \rangle_t = \lim_{\|\Pi\| \rightarrow 0} Q_\Pi(X)$

$$\text{where } Q_\Pi(X) = \sum_{j=0}^{m-1} (X(t_{j+1}) - X(t_j))^2$$

$$\Pi = \{t_0, t_1, \dots, t_n\} \quad 0 = t_0 < t_1 < \dots < t_n = T$$

cross-variation $\langle X_1, X_2 \rangle(t) = \lim_{\|\Pi\| \rightarrow 0} C_\Pi(X_1, X_2)$

$$\text{where } C_\Pi(X_1, X_2) = \sum_{j=0}^{m-1} (X_1(t_{j+1}) - X_1(t_j))(X_2(t_{j+1}) - X_2(t_j))$$

Theorem : Let $X(t) = X^c(t) + I(t) + R(t) + J(t)$ jump process

$$\Rightarrow \langle X \rangle(t) = \langle X^c \rangle(t) + \langle J \rangle(t)$$

$$= \int_0^t \Gamma^2(s) ds + \sum_{s \leq t} (\Delta J(s))^2$$

similarly for the cross-variation process we have :

$$\begin{aligned} \langle X_1, X_2 \rangle(t) &= \langle X_1^c, X_2^c \rangle(t) + \langle J_1, J_2 \rangle(t) \\ &= \int_0^t \Gamma_1(s) \Gamma_2(s) ds + \sum_{s \leq t} \Delta J_1(s) \Delta J_2(s) \end{aligned}$$

• in differential notation:

$$dX_1(t) = dX_1^c(t) + dJ_1(t); \quad dX_2(t) = dX_2^c(t) + dJ_2(t)$$

$$\begin{aligned} \Rightarrow dX_1(t) \cdot dX_2(t) &= dX_1^c(t) \cdot dX_2^c(t) + \underbrace{dX_1^c(t) dJ_2(t)}_{=0} + \\ &+ \underbrace{dX_2^c(t) dJ_1(t)}_{=0} + dJ_1(t) dJ_2(t) \end{aligned}$$

Remark: if $M(t) = N(t) - \lambda t$ compensated PP
 W : Brownian Motion

$$\Rightarrow \langle M, W \rangle_t = -\lambda \underbrace{dt \cdot dW_t}_{=0} + \underbrace{dN(t) \cdot 0}_{=0} = 0$$

\Rightarrow we shall see later that $\langle M, N \rangle = 0$ implies that
N and W are in fact independent.

Stochastic Calculus for Jump Processes

Ito formula : $f \in C^2$

$$f(X(t)) = f(X(s)) + \int_0^t f'(X(u)) dX_u^c + \frac{1}{2} \int_0^t f''(X(u)) d\langle X^c \rangle_u$$
$$+ \sum_{s \leq t} [f(X(s)) - f(X(s-))]$$

example : The Geometric Poisson Process

$X(t) \rightsquigarrow$ jump process

$$S(t) = S(0) \exp \left\{ N(t) \log(r+1) - \lambda r t \right\} =$$
$$= S(0) e^{-\lambda r t} (r+1)^{N(t)}$$

$$\rightsquigarrow S(t) = S(0) f(X(t)) \text{ where } f(x) = e^x$$

Ito formula gives :

$$S(t) = S(0) - \lambda r \int_0^t S(u) du + \sum_{s \leq t} [S(u) - S(u-)]$$

$$\left(\begin{aligned} \text{recall } X^c(t) &= -\lambda r t \implies dX^c(t) = -\lambda r dt \\ &\implies d\langle X^c \rangle(t) = 0 \end{aligned} \right)$$

• if there is a jump at time $u \rightsquigarrow$

$$S(u) = (\sigma+1) S(u^-)$$

$$\Rightarrow S(u) - S(u^-) = \sigma S(u^-)$$

• if there is no jump at time u : $S(u) - S(u^-) = 0$

$$\Rightarrow \sum'_{u \leq t} [S(u) - S(u^-)] = \sum'_{u < t} \sigma S(u^-) \Delta N(u) = \int_0^t \sigma S(u^-) dN(u)$$

$$(\text{recall } \Delta N(u) = \begin{cases} 1 & \text{if there is a jump at } u \\ 0 & \text{otherwise} \end{cases})$$

$$\Rightarrow S(t) = S(0) - \lambda \sigma \int_0^t S(u) du + \sigma \int_0^t S(u^-) dN(u)$$

or equivalently

$$\begin{aligned} S(t) &= S(0) - \lambda \sigma \int_0^t S(u^-) du + \sigma \int_0^t S(u^-) dN(u) \\ &= S(0) + \sigma \int_0^t S(u^-) dM(u) \end{aligned}$$

where $M(t) = N(t) - \lambda t$ (Compensated PP)

$$dM(t) = dN(t) - \lambda dt$$

Corollary : let W be a Brownian motion and N a PP with intensity λ , both defined on the same prob. space $(\Omega, \mathcal{F}, \mathbb{P})$ and relative to the same filtration. Then W and N are independent.

Proof : define $Y(t) = \exp \left\{ u_1 W(t) + u_2 N(t) - \frac{1}{2} u_1^2 t - \lambda (e^{u_2} - 1)t \right\}$

$X(t)$ jump proc.

- write the Itô formula for $f(X(t)) = e^{X(t)}$

- show that $Y(t)$ is martingale

$$\Rightarrow \mathbb{E} e^{u_1 W(t) + u_2 N(t)} = e^{\frac{1}{2} u_1^2 t} \cdot e^{\lambda t (e^{u_2} - 1)}$$

moment generating funct of
 $N(0, t)$ and Poisson (λt)

Ito formula for multiple Jump Processes

$$\begin{aligned}
 f(t, X_1(t), X_2(t)) &= f(0, X_1(0), X_2(0)) + \int_0^t f_t(s, X_1(s), X_2(s)) ds \\
 &\quad + \int_0^t f_{X_1}(s, X_1(s), X_2(s)) dX_1^c(s) + \int_0^t f_{X_2}(s, X_1(s), X_2(s)) dX_2^c(s) \\
 &\quad + \frac{1}{2} \int_0^t f_{X_1 X_1}(s, X_1(s), X_2(s)) d\langle X_1^c \rangle(s) + \\
 &\quad + \int_0^t f_{X_1 X_2}(s, X_1(s), X_2(s)) d\langle X_1^c, X_2^c \rangle(s) + \\
 &\quad + \frac{1}{2} \int_0^t f_{X_2 X_2}(s, X_1(s), X_2(s)) d\langle X_2^c \rangle(s) \\
 &\quad + \sum'_{0 < s \leq t} [f(s, X_1(s), X_2(s)) - f(s, X_1(s-), X_2(s-))]
 \end{aligned}$$

Corollary : (Product Rule)

$$\begin{aligned}
 X_1(t)X_2(t) &= X_1(0)X_2(0) + \int_0^t X_2(s) dX_1^c(s) + \int_0^t X_1(s) dX_2^c(s) \\
 &\quad + \langle X_1^c, X_2^c \rangle(t) + \sum'_{0 < s \leq t} [X_1(s)X_2(s) - X_1(s-)X_2(s-)] \\
 &= \underline{X_1(0)X_2(0)} + \underline{\int_0^t X_2(s-) dX_1(t)} + \underline{\int_0^t X_1(s-) dX_2(s)} + \boxed{\langle X_1, X_2 \rangle(t)}
 \end{aligned}$$

Change of measure (Girsanov's Th) general case.

Recall the Girsanov's Th (case with no jumps)

(Ω, \mathcal{F}, P) $W = B.M$, X adapted to \mathcal{F}^W

$$Z_t^X = \mathbb{E}(X)_t = \exp \left\{ X_t - \frac{1}{2} \langle X \rangle_t \right\}$$

- is a solution of the following SDE

$$dZ_t = Z_t dX_t \quad \text{with } Z_0 = 1$$

(Doleans-Dade exponential of X w.r.t. W)

- Z_t^X is local martingale

- a prob. measure $Q \sim P$ exists $dQ = Z^X dP$
such that for any M , P local martingale

$$\tilde{M}_t = M_t - \langle M, X \rangle_t \quad \text{is a}$$

continuous Q -local martingale

- in particular, if

$$Z(t) = \exp \left\{ - \int_0^t r(s) dW(s) - \frac{1}{2} \int_0^t r^2(s) ds \right\}$$

or $dZ(t) = - \bar{r}(t) Z(t) dW(t) = Z(t) dX^c(t)$

where $X^c(t) = - \int_0^t r(s) dW(s)$ $\langle X^c \rangle(t) = \int_0^t r^2(s) ds$

$$\Rightarrow Z(t) = \exp \left\{ X^c(t) - \frac{1}{2} \langle X^c \rangle(t) \right\}$$

- in the stochastic calculus with jumps the above equation becomes

$$\textcircled{a} \quad dZ^x(t) = Z^x(t-) dX(t)$$

where $X(t) = X^c(t) + J(t)$

- the solution to \textcircled{a} is the same as R-N in the continuous case except that whenever there is a jump in X there will be a jump in Z^x

$$\Delta Z^x(t) = Z^x(t-) \Delta X(t)$$

$$\rightsquigarrow Z^x(t) = Z^x(t-) + \Delta Z^x(t)$$

$$= Z^x(t-) (1 + \Delta X(t))$$

- So if X is a jump process,

$$Z^x(t) = \Sigma(X)(t) \text{ becomes}$$

$$= \exp \left\{ X^c(t) - \frac{1}{2} \langle X^c \rangle(t) \right\} \cdot \prod_{s \leq t} (1 + \Delta X(s))$$

- this process is the solution to the

SDE :

$$Z^x(t) = 1 + \int_0^t Z^x(s-) dX(s)$$

Change of measure for a PP process

• $N(t) = \text{PP on } (\Omega, \mathcal{F}, \mathbb{P})$

$$\mathbb{E} N(t) = \lambda t \quad (\lambda = \text{intensity})$$

• assume the stock price is modelled by a GPP

$$S(t) = S(0) \exp \{ N(t) \log(0+1) - \lambda t \}$$

we have seen already that

$$\begin{aligned} S(t) &= S(0) - \lambda \sigma \int_0^t S(u-) du + \sigma \int_0^t S(u-) dN(u) \\ &= S(0) + \sigma \int_0^t S(u-) dM(u) \end{aligned}$$

where $M(t) = N(t) - \lambda t$: compensated PP
(martingale)

∴ $S(t)$ is martingale as well.

- assume now the equation of the stock price is

$$S(t) = S(0) \exp \{ \alpha t + N(t) \log(\sigma+1) - \lambda \sigma t \}$$

$$\Rightarrow dS(t) = \alpha S(t) dt + \sigma S(t) dM(t)$$

$$\begin{aligned} d(e^{-rt} S(t)) &= ((\alpha - r) S(t) dt + \sigma S(t) dM(t)) e^{-rt} \\ &= \sigma \left[\frac{\alpha - r}{\sigma} S(t) dt + S(t) dM(t) \right] e^{-rt} \\ &= \sigma S(t) \left[\underbrace{\frac{\alpha - r}{\sigma} dt + dM(t)}_{d\tilde{M}(t)} \right] e^{-rt} \end{aligned}$$

- We would like \tilde{M} to be a compensated PP under \tilde{P} with intensity $\tilde{\lambda}$

$$\begin{aligned} \tilde{M}(t) &= M(t) + \frac{\alpha - r}{\sigma} t \\ &= N(t) - \lambda t + \frac{\alpha - r}{\sigma} t \\ &= N(t) - \tilde{\lambda} t \end{aligned}$$

where $\tilde{\lambda} = \lambda - \frac{\alpha - r}{\sigma}$: new intensity

Remark : we need $\tilde{\lambda} = \lambda - \frac{\alpha - r}{\sigma} > 0$ otherwise

there is arbitrage !

- we want the change of measure that transforms $N(t)$ from a PP(λ) to a PP($\tilde{\lambda}$)

- the density that makes this change possible is

$$Z(t) = e^{(\lambda - \tilde{\lambda})t} \cdot \left(\frac{\tilde{\lambda}}{\lambda} \right)^{N(t)} \quad \left(\frac{\tilde{\lambda}}{\lambda} = \lambda - \frac{\alpha - r}{\sigma} \right)$$

- easy to check that :

$$dZ(t) = \frac{\tilde{\lambda} - \lambda}{\lambda} Z(t) dM(t)$$

$\Rightarrow Z(t)$ is martingale under \tilde{P} and $EZ(t) = 1$

Theorem : Under prob. measure \tilde{P} $0 \leq t \leq T$
 $(d\tilde{P} = Z_t dP)$, the process $N(t)$ is a PP
with intensity $\tilde{\lambda}$

Pricing a European Call in a Jump Model

- asset driven by a Poisson process

$$\begin{aligned} S(t) &= S(0) \exp \left\{ \alpha t + N(t) \log(\sigma+1) - \lambda \sigma t \right\} \\ &= S(0) e^{(\alpha - \lambda \sigma)t} (\sigma+1)^{N(t)} \end{aligned}$$

$$\Rightarrow dS(t) = \alpha S(t) dt + \sigma S(t) dM(t)$$

where $M(t) = N(t) - \lambda t$: compensated PP

$V(T) = (S(T) - K)^+$: payoff at time T

- assume $\lambda > \frac{\alpha - r}{\sigma}$ (to rule out arbitrage)

$$\Rightarrow \tilde{\lambda} = \lambda - \frac{\alpha - r}{\sigma}$$

$$\tilde{P}(A) = \int_A Z(T) d\tilde{P} \quad \text{for all } A \in \mathcal{F}$$

$$\text{where } Z_T = e^{(\lambda - \tilde{\lambda})T} \left(\frac{\tilde{\lambda}}{\lambda} \right)^{N(T)}$$

(Z_t is a solution of : $dZ(t) = \frac{\tilde{\lambda} - \lambda}{\lambda} Z(t) dM(t)$)

\tilde{P} : risk neutral measure

- under \tilde{P} : $\tilde{M}(t) = N(t) - \tilde{\lambda}t$ is compensated PP (martingale)

$$dS(t) = rS(t) dt + \sigma S(t) d\tilde{M}(t)$$

$$d(e^{-rt} S(t)) = r e^{-rt} S(t) dt + \sigma e^{-rt} S(t) d\tilde{M}(t)$$

\Rightarrow discounted stock price is a martingale under \tilde{P}

$$\Rightarrow S(t) = S(0) e^{(r - \tilde{\lambda}\sigma)t} \left(\frac{\tilde{\lambda}}{\lambda} \right)^{N(t)}$$

- let $V(t)$: the price of the European call with payoff v_T

$\rightsquigarrow e^{-rt} V(t)$ is a martingale under \tilde{P}

$$\begin{aligned} e^{-rt} V(t) &= \tilde{E} \left[e^{-rT} V(T) \mid \mathcal{F}_t \right] \\ &= \tilde{E} \left[e^{-rT} (S(T) - K)^+ \mid \mathcal{F}_t \right] \end{aligned}$$

$$\text{where } S(T) = S(t) e^{- (r - \tilde{\lambda}) (T-t)} \frac{N(T) - N(t)}{(r+1)}$$

$$\begin{aligned} \Rightarrow V(t) &= \tilde{E} \left[e^{-r(T-t)} \left(S(t) e^{- (r - \tilde{\lambda}) (T-t)} \frac{N(T) - N(t)}{(r+1)} - K \right)^+ \mid \mathcal{F}_t \right] \\ &= c(t, S(t)) \quad \uparrow \text{Markov property.} \end{aligned}$$

where

$$c(t, x) = \sum_{j=0}^{\infty} e^{-r(T-t)} \left(x e^{- (r - \tilde{\lambda}) (T-t)} \frac{j}{(r+1)} - K \right)^+ \cdot \frac{\tilde{\lambda}^j (T-t)^j}{j!} e^{-\tilde{\lambda}(T-t)}$$

(since $N(T) - N(t) \sim N(T-t) \sim \text{Poisson}(\tilde{\lambda}(T-t))$)
under \tilde{P}

so $e^{-rt} c(t, S(t)) : \tilde{P}$ martingale

Ito
 \rightsquigarrow we need to compute $d(e^{-rt} c(t, S(t)))$ and set the dt term equal to 0.

$$dS(t) = \underbrace{(r - \tilde{\lambda}\sigma)S(t)}_{dS^c(t)} dt + \sigma S(t-) dN(t)$$

$dS^c(t) \rightsquigarrow$ the cont. part.

- on the other hand if the stock jumps at t

$$\Delta S(t) = S(t) - S(t-) = \sigma S(t-) \rightsquigarrow S(t) = (r+i)S(t-)$$

$$\Rightarrow e^{-rt} c(t, S(t)) = c(0, S(0)) + \int_0^t e^{-ru} \left[-rc(u, S(u)) du + c_t(u, S(u)) du + \right. \\ \left. + c_x(u, S(u)) dS^c(u) \right] + \sum_{u \leq t} e^{-ru} [c(u, S(u)) - c(u, S(u-))]$$

(note that c_{xx} term is missing because $d\langle S^c \rangle(t) = 0$)

$$\Rightarrow e^{-rt} c(t, S(t)) = c(0, S(0)) + \int_0^t e^{-ru} \left(-rc(u, S(u)) + c_t(u, S(u)) + \right. \\ \left. + (r - \tilde{\lambda}\sigma) S(u) c_{xx}(u, S(u)) \right) du +$$

$$\int_0^t e^{-ru} [c(u, (r+i)S(u-)) - c(u, S(u-))] d[N(t) - \tilde{\lambda}u + \tilde{\lambda}u]$$

not a martingale yet !!!

- to make it a martingale you need to compensate it with $-\tilde{\lambda}t$

$$\int [\dots] d[N(u) - \tilde{\lambda}u] + \int [] d(\tilde{\lambda}u)$$

$\underbrace{M_u}_{\tilde{\lambda}u}$

\Rightarrow we need t_0 set the dt term = 0.

$$-rc(t, S(t)) + c_f(t, S(t)) + (r - \tilde{\lambda}\tilde{r})S(t)C_x(t, S(t)) + \\ + \tilde{\lambda}(c(t, (r+1)S(t)) - c(t, S(t))) = 0$$

\rightsquigarrow replace $S(t)$ with the dummy var x :

$$-rc(t, x) + c_f(t, x) + (r - \tilde{\lambda}\tilde{r})x C_x(t, x) + \\ + \tilde{\lambda}(c(t, (r+1)x) - c(t, x)) = 0$$

"differential-difference" eq.

to complete the pricing of this option we must construct a hedging portfolio:

at time 0 invest $X(0) = c(0, S(0))$

at time t : invest in stock & money market so that:
 $X(t) = c(t, S(t))$

or equivalently: $e^{-rt} X(t) = e^{-rt} c(t, S(t))$

\rightsquigarrow match the differentials:

$$d(\bar{e}^{-rt} c(t, S(t))) = \bar{e}^{-rt} [c(t, (r+1)S(t-)) - c(t, S(t-))] d\tilde{M}(t)$$

$$d(X(t)) = \Gamma(t-) dS(t) + r [X(t) - \Gamma(t) S(t)] dt$$

$$\begin{aligned} \Rightarrow d(\bar{e}^{-rt} X(t)) &= \bar{e}^{-rt} [-rX(t) dt + dX(t)] \\ &= \bar{e}^{-rt} [\Gamma(t-) dS(t) - r \Gamma(t) S(t) dt] \\ &= \bar{e}^{-rt} r \Gamma(t-) S(t-) d\tilde{M}(t) \end{aligned}$$

$$\Rightarrow \Gamma(t-) = \frac{c(t, (r+1)S(t-)) - c(t, S(t-))}{r S(t-)}$$

• this hedging position we should hold at all times whether they are jump times or not.

$$\text{i.e. define } \Gamma(t) = \frac{c(t, (r+1)S(t)) - c(t, S(t))}{r S(t)}$$

$$\Rightarrow \bar{e}^{-rt} X(t) = X(0) + \int_0^t \bar{e}^{-ru} [c(u, (r+1)S(u-)) - c(u, S(u-))] d\tilde{M}(u)$$

Change of measure for the CFP

$$Y_i = \{y_1, y_2, \dots, y_m\} \quad p_m = P(Y_i = y_m) > 0$$

$$Q(t) = \sum_{i=1}^{N(t)} Y_i = \sum_{m=1}^M y_m \cdot N_m(t)$$

where N_1, N_2, \dots, N_M are indep. PP.

with intensities $\lambda p_1, \lambda p_2, \dots, \lambda p_m$

$$Z_1(t) = e^{(\lambda_1 - \tilde{\lambda}_1)t} \left(\frac{\tilde{\lambda}_1}{\lambda_1} \right)^{N_1(t)} \quad \lambda_1 = \lambda \cdot p_1$$

$$Z(t) = \prod_{m=1}^M Z_m(t)$$