

MA 573 HW Correction for 3/3/2022

1. Let $t > s$, then,

$$g(t) = E[f(X_t)] = E[E[f(X_t)|X_s]], \text{ by independence.}$$

Since f is convex, we can apply Jensen's inequality, such that,

$$g(t) = E[E[f(X_t)|X_s]] \geq E[f(E[X_t|X_s])].$$

Since X is a martingale, we have, by definition,

$$g(t) \geq E[f(E[X_t|X_s])] = E[f(X_s)] = g(s).$$

Or simply saying, $g(t) \geq g(s)$. Hence, $g(t)$ is increasing.

2. Let $t > s$, then,

$$C(t) = E[e^{-rt}(S_t - K)^+] = E[(S_t e^{-rt} - K e^{-rt})^+].$$

Let $X_t = S_t e^{-rt}$, then,

$$\begin{aligned} C(t) &= E[(X_t - K e^{-rs})^+] \geq E[(X_t - K e^{-rs})^+]; \\ &= E[f(X_t)]; \text{ where } f(x) = (x - K e^{-rs})^+; \\ &\geq E[f(X_s)]; \\ &= E[(X_s - K e^{-rs})^+] = C(s). \end{aligned}$$

Or simply saying, $C(t) \geq C(s)$. Hence, $C(t)$ is increasing.

3. Let $t > s$, then,

$$P(t) = E[(S_t - K)^-] = E[(K - S_t)^+];$$

$$P(t) = E[(K - S_t)^+] \geq E[(K - S_s)^+] = P(s).$$

Or simply saying, $P(t) \geq P(s)$. Hence, $P(t)$ is increasing.

4. Let $t > s$, then,

$$P(t) = E[(S_t - K)^-] = E[(K - S_t)^+].$$

Since we let $r > 0$, then,

$$P(t) = E[(K - S_t)^+] \leq E[(K e^{-rt} - S_t)^+];$$

$$\leq E[(K e^{-rs} - S_t)^+];$$

$$= E[f(S_t)];$$

$$\geq E[f(S_s)];$$

$$= E[(K e^{-rs} - S_s)^+] \leq E[(K - S_s)^+] = P(s).$$

The fact that $P(t) \leq P(s)$ challenges the claim that $P(t)$ increases.