

## 7.1 Subordinators

If  $X$  is a subordinator then

$$\mathbb{E}(e^{i\theta X}) = -i\delta\theta + \int_{(0, \infty)} (1 - e^{i\theta x}) \Pi(dx), \quad \theta \in \mathbb{R}$$

where necessarily  $\delta \geq 0$ ,  $\int_{(0, \infty)} (1 \wedge x) \Pi(dx) < \infty$

Moreover, we can write

$$X_t = \delta t + \sum_{s \leq t} \Delta X_s$$

$$\text{where } \Delta X_s = X_s - X_{s-}$$

According to theorem 7.1.

$$\text{hence } \mathbb{E}(e^{-\lambda X_t}) = \mathbb{E}(e^{\lambda(-X_t)})$$

where  $\lambda \geq 0$  and jump measure associated with  $-X$  assigns zero mass to  $(0, \infty)$

$$\text{Hence } \int_{|x| \geq 1} e^{\lambda x} \Pi_{(-x)}(dx) < \infty$$

$$\text{and thus } \mathbb{E}(e^{-\lambda X_t}) < \infty \quad \forall t \geq 0.$$

By analytic extension we can also write

$$\mathbb{E}(e^{-\lambda X_t}) = \exp \left\{ -(\lambda \delta t + \int_{(0, \infty)} (1 - e^{-\lambda x}) \Pi(dx)) t \right\}$$

In the future it will be more convenient to talk about (killed) subordinators.

~~Def~~ A process  $\{X_t : t \geq 0\}$  is a (killed) subordinator if

$$X_t = \begin{cases} Y_t & t < \tau_q \\ \partial & \text{otherwise} \end{cases}$$

where  $\{Y_t : t \geq 0\}$  a subordinator,  $\tau_q \perp Y$  and exponentially distributed with parameter  $q \geq 0$ .  
 We understand  $\tau_0 := \infty$  v.p. 1) and  $\partial$  is a cemetery state.

(killed) Subordinator  $X$  has Laplace exponent

" $\partial = \infty$ "

$$\mathbb{E}(e^{-\lambda X_t}) = \exp \left\{ - \left( q + \delta \lambda + \int_{(0, \infty)} (1 - e^{-\lambda x}) \Pi(dx) \right) t \right\}$$

$$\delta \geq 0, \quad \int_{(0, \infty)} (1 \wedge x) \Pi(dx) < \infty$$

$$\mathbb{E}(e^{-\lambda Y_t} \mid t < \tau_q)$$

$$e^{-qt} \mathbb{E}(e^{-\lambda Y_t})$$

## 7.2 Spectrally negative Lévy processes

Recall a SNLP is a L.P. which has  $\Pi(0, \infty) = 0$  and does not have monotone paths. i.e. we exclude from the def<sup>n</sup>  $X_t = \delta t$  with  $\delta \geq 0$  and

$$X_t = -Y_t \text{ where } Y \text{ is a subordinator.}$$

Again from Th 7.4 "  $\int_{|x| \geq 1} e^{\lambda x} \Pi(dx) < \infty$  "

$$\text{Hence } \mathbb{E}(e^{\lambda X_t}) < \infty \quad \forall t \geq 0. \quad \boxed{\forall \lambda \geq 0}$$

A similar argument using analytic extension allows us to deduce that

$$\mathbb{E}(e^{\lambda X_t}) = \exp \left\{ t \left( -a\lambda + \frac{1}{2}\sigma^2\lambda^2 + \int_{(-\infty, 0)} (e^{\lambda x} - 1 - \lambda x \mathbb{1}_{|x| < 1}) \Pi(dx) \right) \right\}$$

$$\lambda = i(-i\lambda)$$

+ (A)

Recall that if  $X$  is a SNLP of BV  
 then necessarily  $X_t = \delta t - \sum_{s \leq t} (-\Delta X_s)$

pure jump  
 subordinator  
 with jump measure  $\nu$   
 $\nu(x, \infty) = \mathbb{1}(-\infty, -x)$   
 $x > 0$

$$\psi(\lambda) = \delta \lambda + \int_{(-\infty, 0)} (e^{\lambda x} - 1) \Pi(dx)$$

$$= \delta \lambda - \int_{(0, \infty)} (1 - e^{-\lambda x}) \nu(dx)$$

$$\mathbb{E}(e^{\lambda X_t}) = e^{\psi(\lambda)t} = e^{\delta \lambda t - t(\text{Laplace exp. of sub})}$$

Something v. important about  $\psi(\lambda) := \frac{1}{t} \log \mathbb{E}(e^{\lambda X_t})$   
 for SNEP is

$$\boxed{\lambda \geq 0}$$

①  $\psi$  is  $C^\infty$

②  $\psi$  is strictly convex

③  $\psi(0) = 0$

④  $\psi(\infty) = \infty$

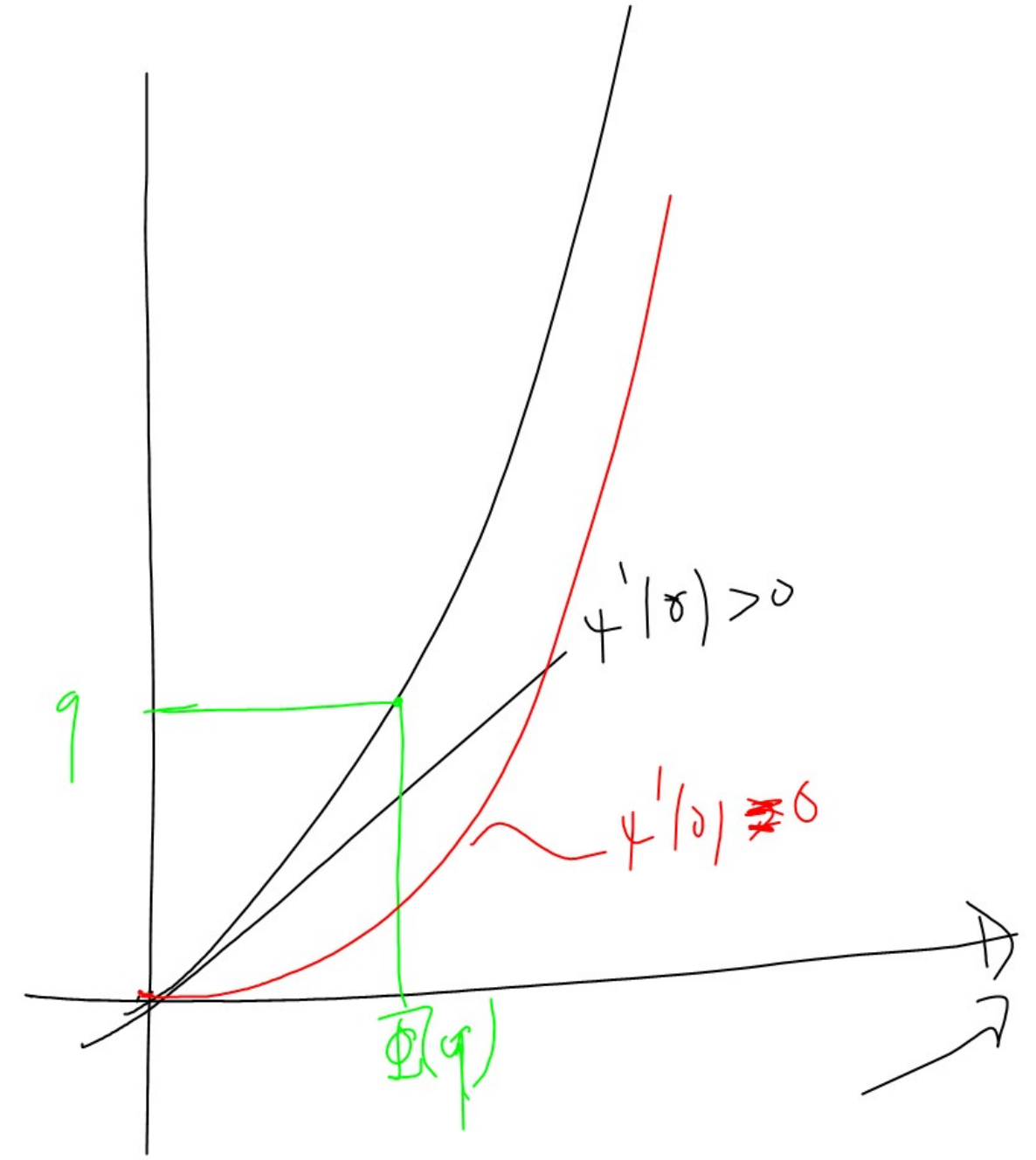
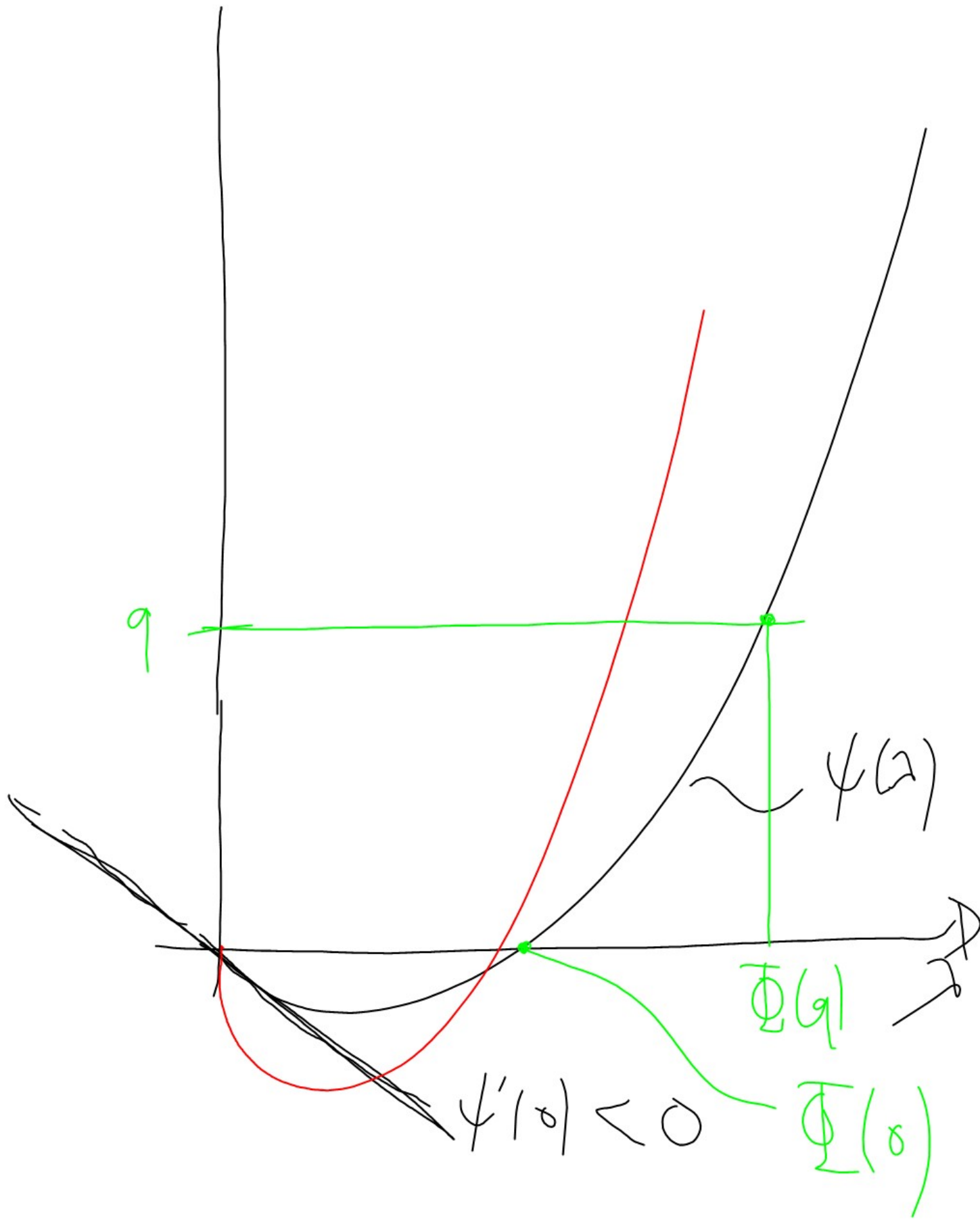
⑤ Note  $\psi'(0) = \mathbb{E}(X_1) \in [-\infty, \infty)$

⑥  $\Phi(q) = \sup \{ \lambda : \psi(\lambda) = q \}$

then [there are at most 2 roots to  $\psi(\lambda) = q$ ]

$\Phi(0) = 0$  if  $\mathbb{E}(X_1) \geq 0$  o/w  $\Phi(0) > 0$

and  $\Phi(q) > 0$  when  $q > 0$





# Exponential martingale for SNLP with applications

8.1  $E_t(\lambda) = e^{\lambda X_t - \psi(\lambda)t}$   $\rightarrow \geq 0$

$$\begin{aligned} E(\tau_{t+s}(\lambda) | \mathcal{F}_t) &= e^{\lambda X_t - \psi(\lambda)t} E\left(e^{\lambda(X_{t+s} - X_t) - \psi(\lambda)s} | \mathcal{F}_t\right) \\ &= e^{\lambda X_t - \psi(\lambda)t} E\left(e^{\lambda X_s} | \mathcal{F}_t\right) e^{-\psi(\lambda)s} \\ &= E_t(\lambda) \end{aligned}$$

mean 1

$E(\lambda)$  is a mgf!

$$E(\lambda) = 1$$

Going to use  $E(\lambda)$  to study

$$\tau_x^+ := \inf \{ t > 0 : X_t > x \}$$

Theorem 8.1 For any SNLP and  $q \geq 0$

$$\mathbb{E} \left( e^{-q \tau_x^+} \mathbb{1}_{(\tau_x^+ < \infty)} \right) = e^{-\Phi(q)x}$$

where  $\Phi(q)$  is the largest root of  $\psi(\lambda) = q$  and  $\psi$  is the Laplace exponent of the SNLP.

~~Proof~~ Spectral negativity means that on  $\{\tau_x^+ < \infty\}$   
 $X_{\tau_x^+} = x$ . Bearing this in mind

$$\mathbb{E} \left( e^{\Phi(q)X_t - qt} \mid \mathcal{F}_{\tau_x^+} \right) = \mathbb{1}_{(\tau_x^+ < t)} e^{\Phi(q)X_{\tau_x^+} - q\tau_x^+} \times \mathbb{E} \left( e^{\frac{\Phi(q)(X_t - X_{\tau_x^+}) - q(t - \tau_x^+)}{\sigma_{\tau_x^+}} \mid \mathcal{F}_{\tau_x^+}} \right)$$

$$[\Phi(q) \geq 0, q \geq 0] \mathbb{E}(\Phi(q)), \text{ using } \psi(\Phi(q)) = q + \mathbb{1}_{(t \leq \tau_x^+)} e$$

Hence 
$$\mathbb{E} \left( e^{\Phi(q) X_t - q t} \mathbb{1}_{\tau_x^+} \right)$$

$$= e^{\Phi(q) X_{t \wedge \tau_x^+} - q(t \wedge \tau_x^+)}$$

Take expectations again:

$$1 = \lim_{t \uparrow \infty} \mathbb{E} \left( e^{\Phi(q) X_{t \wedge \tau_x^+} - q(t \wedge \tau_x^+)} \right)$$

note  $X_{t \wedge \tau_x^+} \leq x$

(again uses special regularity!)

Hence  $\mathbb{P}(\bar{\tau})$  to deduce

$$1 = \mathbb{E} \left[ \lim_{t \uparrow \infty} e^{\Phi(q) X_{t \wedge \tau_x^+} - q(t \wedge \tau_x^+)} \right]$$

$$(\dots) \mathbb{1}_{\{\tau_x^+ < \infty\}} + (\dots) \mathbb{1}_{\{\tau_x^+ = \infty\}}$$

$\downarrow$  use SN!

$$e^{\Phi(q)x - q\tau_x^+}$$

$$\leq e^{\Phi(q)x - qt} \rightarrow 0$$

$$1 = \mathbb{E} \left( e^{\Phi(q)x - q\tau_x^+} \mathbb{1}_{\{\tau_x^+ < \infty\}} \right)$$