

Conexion de la serie N° f

Exo 1

1) On sait que $\phi_x |H = E(e^{itx})$ et $\phi_x(0) = 1$.

Donc $\phi_x(0) = ce = 1 \Leftrightarrow c = \bar{e}^{-1}$ et $\phi_x |H = e^{-4|H} \quad t \in \mathbb{R}$.

2) Par definition $\phi_x |H = \int_{\mathbb{R}} e^{itx} f_x(x) dx$ donc

$$f_x(x) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{-itx} \phi_x |H dt \quad (\text{la transformation inverse de Fourier})$$

$$= \frac{1}{2\pi} \int_{\mathbb{R}} e^{-4|H - itx} dt = \frac{1}{2\pi} \left[\int_{-\infty}^0 e^{-(4-ix)t} dt + \int_0^{+\infty} e^{-(4+ix)t} dt \right]$$

$$= \frac{1}{2\pi} \left[\frac{1}{4-ix} e^{(4-ix)t} \Big|_{-\infty}^0 + \frac{1}{4+ix} e^{-(4+ix)t} \Big|_0^{+\infty} \right]$$

$$= \frac{1}{2\pi} \left(\frac{1}{4-ix} + \frac{1}{4+ix} \right) = \frac{1}{2\pi} \left(\frac{4+ix + 4-ix}{16+x^2} \right)$$

$$= \frac{1}{2\pi} \frac{8}{16+x^2} = \frac{4}{\pi(16+x^2)} \quad x \in \mathbb{R}$$

Exo 2

On a $\varphi_{x+y} |H = E(e^{it(x+y)}) = E(e^{itx} e^{ity})$

x, y indep $\Rightarrow E(e^{itx}) E(e^{ity}) = \varphi_x |H \varphi_y |H$

On a $f_x(x) = \frac{1}{\sqrt{2\pi}\sigma_x} e^{-\frac{(x-\mu_x)^2}{2\sigma_x^2}} \quad \text{si } x \in \mathbb{R}$.

$$\varphi_x(x) = E(e^{itx}) = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}\sigma_x} e^{itx - \frac{(x-\mu_x)^2}{2\sigma_x^2}} dx$$

$$= \frac{1}{\sqrt{2\pi}\sigma_x} \int_{\mathbb{R}} \exp \left[\frac{-1}{2\sigma_x^2} \left\{ x^2 - 2\mu_x x + \mu_x^2 - 2i\sigma_x^2 t x \right\} \right] dx.$$

$$= \frac{1}{\sqrt{2\pi}\sigma_x} \int_{\mathbb{R}} \exp \frac{-1}{2\sigma_x^2} \left[x^2 - 2(\mu_x + i\sigma_x^2 t)x + \mu_x^2 + 2i\mu_x\sigma_x^2 t + (i\sigma_x^2 t)^2 - 2i\mu_x\sigma_x^2 t - (i\sigma_x^2 t)^2 \right] dx.$$

$$= \frac{1}{\sqrt{2\pi}\sigma_x} \int_{\mathbb{R}} \exp \frac{-1}{2\sigma_x^2} \left(x - \mu_x - \frac{i\sigma_x^2 t}{\sigma_x} \right)^2 e^{i\mu_x t - \frac{t}{2}\sigma_x^2} dx$$

$$\varphi_x(H) = e^{i\mu_x t - \frac{\sigma_x^2}{2} t}$$

$$\text{denn } \varphi_y(H) = e^{i\mu_y t - \frac{\sigma_y^2}{2} t}$$

$$\begin{aligned} 2) \varphi_{x+y}(H) &= \varphi_x(H) \varphi_y(H) = e^{i\mu_x t - \frac{\sigma_x^2}{2} t} e^{i\mu_y t - \frac{\sigma_y^2}{2} t} \\ &= e^{i(\mu_x + \mu_y)t - \frac{(\sigma_x^2 + \sigma_y^2)}{2} t} \end{aligned}$$

$$\text{Also } x+y \sim \mathcal{N}(\mu_x + \mu_y, \sigma_x^2 + \sigma_y^2).$$

Ex 03:

$$1) \text{ an a } P(X=k) = e^{-\lambda} \frac{\lambda^k}{k!} \quad k \in \mathbb{N}$$

$$\begin{aligned} \varphi_X(t) &= E(e^{itX}) = \sum_{k \in \mathbb{N}} e^{itk} P(X=k) = \sum_{k \in \mathbb{N}} e^{itk} e^{-\lambda} \frac{\lambda^k}{k!} \\ &= e^{-\lambda} \sum_{k \in \mathbb{N}} \frac{(\lambda e^{it})^k}{k!} = e^{-\lambda} (e^{\lambda e^{it}} - 1) \end{aligned}$$

$$\begin{aligned}
 2) \quad \ell_Z(H) &= \ell_{\sum_{i=1}^n X_i}(H) = \prod_{i=1}^n \ell_{X_i}(H) \\
 &= \prod_{i=1}^n \frac{\lambda}{e} e^{-\lambda x_i} \\
 &= \frac{\lambda^n}{e^n} e^{-\lambda \sum_{i=1}^n x_i} = \frac{\lambda^n}{e^n} e^{-\lambda Z}
 \end{aligned}$$

Also $Z \sim \mathcal{P}(n\lambda)$.