

**Theorem 1.**  $\varphi_{\mathcal{N}(0,1)}(t) = e^{-\frac{t^2}{2}}$        $\varphi_{\mathcal{N}(a,\sigma^2)}(t) = e^{ita - \frac{t^2\sigma^2}{2}}$

**Theorem 2.**  $X$  random variable in  $\mathbb{R}^d$ ,  $\mathbb{E}|X_1|^{k_1} \dots |X_d|^{k_d} < \infty$ , then  $\frac{\partial^{k_1}}{\partial t_1^{k_1}} \dots \frac{\partial^{k_d}}{\partial t_d^{k_d}} \varphi_X(t)$  exists and equals  $i^{|k|} \mathbb{E}X_1^{k_1} \dots X_d^{k_d}$ .

*Remark 3.*  $X_1, \dots, X_d$  independent, then  $\varphi_{X_1+\dots+X_d}(t) = \varphi_{X_1}(t) \dots \varphi_{X_d}(t)$ , but opposite is not true in general.

**Theorem 4.**  $X_1, \dots, X_d$  independent iff  $\forall_{t \in \mathbb{R}^d} \varphi_{(X_1, \dots, X_d)}(t) = \varphi_{X_1}(t_1) \dots \varphi_{X_d}(t_d)$ .

**Theorem 5** (Lévy-Cramer). 1. If  $\mu_n, \mu$  probability measures on  $\mathbb{R}^d$  and  $\mu_n \Rightarrow \mu$ , then  $\forall_{t \in \mathbb{R}^d} \varphi_{\mu_n}(t) \rightarrow \varphi_\mu(t)$ .

2. If  $\mu_n$  probability measure on  $\mathbb{R}^d$  and there exists function  $\varphi : \mathbb{R}^d \rightarrow \mathbb{C}$  such that  $\forall_{t \in \mathbb{R}^d} \varphi_{\mu_n}(t) \rightarrow \varphi(t)$  and  $\varphi$  is continuous at 0, then there exists a probability measure  $\mu$  on  $\mathbb{R}^d$  such that  $\varphi = \varphi_\mu$  and  $\mu_n \Rightarrow \mu$ .

**Corollary 6.**  $\mu_n \Rightarrow \mu$  iff  $\varphi_{\mu_n} \rightarrow \varphi_\mu$  pointwise.

**Theorem 7** (Inverse Fourier Theorem). Suppose that  $\mu$  is a probability measure on  $\mathbb{R}^d$  and  $\varphi_\mu \in L^1(\mathbb{R}^d)$  such that  $\int_{\mathbb{R}^d} |\varphi_\mu(x)| dx < \infty$ , then  $\mu$  has the density  $g$  given by the formula

$$g(x) = \frac{1}{(2\pi)^d} \int_{\mathbb{R}^d} \varphi_\mu(t) e^{-i\langle t, x \rangle} dt.$$