

Fourier inversion

Characteristic function:

$$\varphi(t) = E[e^{itX}] \text{ if density } \int_{-\infty}^{\infty} e^{itx} f(x) dx = \hat{f}(t)$$

Fourier inversion formula:

$$f(x) = \lim_{T \rightarrow \infty} \frac{1}{2\pi} \int_{-T}^T \varphi(t) e^{-itx} dt$$

If $\varphi(t)$ is integrable, then

$$f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi(t) e^{-itx} dt$$

Example: $X = a$ (constant)

$$\varphi(t) = E[e^{itX}] = e^{ita} = \cos ta + i \sin ta.$$

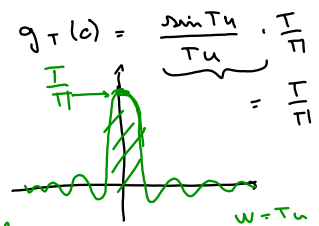
Motivation for inversion formula: Consider

$$\begin{aligned} I_T &= \frac{1}{2\pi} \int_{-T}^T \varphi(t) e^{-itx} dt & \varphi(t) &= \int_{-\infty}^{\infty} e^{ity} f(y) dy \\ &= \frac{1}{2\pi} \int_{-T}^T \left(\int_{-\infty}^{\infty} e^{ity} f(y) dy \right) e^{-itx} dt \\ &= \frac{1}{2\pi} \int_{-T}^T \left(\int_{-\infty}^{\infty} e^{it(y-x)} f(y) dy \right) dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \left[\int_{-T}^T e^{it(y-x)} dt \right] f(y) dy \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} f(y) \left(\int_{-T}^T e^{it(y-x)} dt \right) dy \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} f(y) \left[\frac{e^{it(y-x)}}{i(y-x)} \right]_{t=-T}^T dy \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} f(y) \left[\frac{e^{iT(y-x)} - e^{-iT(y-x)}}{i(y-x)} \right] dy \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} f(y) \frac{2 \sin[T(y-x)]}{y-x} dy \end{aligned}$$

$$\sin z = \frac{e^{iz} - e^{-iz}}{2i}$$

$$g_T(u) = \frac{\sin Tu}{\pi u}$$

Take a look at $g_T(u)$



$$= \int_{-\infty}^{\infty} f(y) g_T(y-x) dy \rightarrow f(x)$$

contribution when $y=x$

Also

$$\begin{aligned} \int_{-\infty}^{\infty} g_T(u) du &= \int_{-\infty}^{\infty} \frac{\sin Tu}{\pi u} du \\ &= \int_{-\infty}^{\infty} \frac{\sin w}{\pi \frac{w}{T}} \frac{dw}{T} = \int_{-\infty}^{\infty} \frac{\sin w}{\pi w} dw \\ &= \underline{\underline{1}} \end{aligned}$$

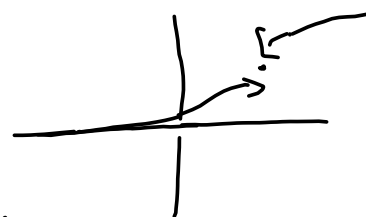
W = Tu, dw = T du

Theorem: Assume that X is a random variable with a density f and a characteristic function φ . Then

$$f(x) = \lim_{T \rightarrow \infty} \frac{1}{2\pi} \int_{-T}^T \varphi(t) e^{-itx} dt$$

If φ is integrable, then

$$f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi(t) e^{-itx} dt.$$



If F is a distribution function, define

$$\bar{F}(x) = \frac{F(x) + F(x-)}{2}$$

Levy's Inversion Formula: Let X be a random variable with characteristic function φ and distribution function F .

For $a < b$,

$$\begin{aligned} \bar{F}(b) - \bar{F}(a) &= \lim_{\varepsilon \rightarrow 0} \int_{-\infty}^{\infty} \underbrace{\frac{e^{-ibt} - e^{-iat}}{-2\pi it}}_{h(t)} \varphi(t) e^{-\frac{\varepsilon^2 t^2}{2}} dt \\ &= \lim_{T \rightarrow \infty} \int_{-T}^T \underbrace{\frac{e^{-ibt} - e^{-iat}}{-2\pi it}}_{h(t)} \varphi(t) dt. \end{aligned}$$

$$\int_{-T}^T \frac{e^{-ibt} - e^{-iat}}{-2\pi it} \varphi(t) dt = \int_{-\infty}^{\infty} \mathbb{1}_{[-T, T]} h(t) dt$$

$$\int_{-\infty}^{\infty} h(t) e^{-\frac{\varepsilon^2 t^2}{2}} dt$$

Let ξ be a $N(0,1)$ random variable independent of Σ and define

$$\Sigma_\epsilon = \Sigma + \epsilon \xi \quad \text{for } \epsilon > 0.$$

Then Σ_ϵ is "smooth" random variable with a density f_ϵ .

Then

$$P[a < \Sigma_\epsilon < b] = \int_a^b f_\epsilon(x) dx = \int_a^b \left[\frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi_\epsilon(t) e^{-itx} dt \right] dx$$

Note $\varphi_\epsilon(t) = \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}}$ (integrable)

$$\begin{aligned} & \frac{1}{2\pi} \int_a^b \left[\int_{-\infty}^{\infty} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} e^{-itx} dt \right] dx \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \left[\int_a^b \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} e^{-itx} dx \right] dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} \left[\int_a^b e^{-itx} dx \right] dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} \left[\frac{e^{-itx}}{-it} \right]_{x=a}^{x=b} dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} \left[\frac{e^{-itb} - e^{-ita}}{-it} \right] dt \\ &= \int_{-\infty}^{\infty} \frac{e^{-itb} - e^{-ita}}{-2\pi it} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} dt \end{aligned}$$

hence

$$\lim_{\epsilon \rightarrow 0} P[a < \Sigma_\epsilon < b] = \lim_{\epsilon \rightarrow 0} \int_{-\infty}^{\infty} \frac{e^{-itb} - e^{-ita}}{-2\pi it} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} dt$$

What happens with the limit on the left?

$$\lim_{\epsilon \rightarrow 0} P[a < \Sigma_\epsilon < b] = \lim_{\epsilon \rightarrow 0} E \left[\mathbb{1}_{(a,b)}(\Sigma_\epsilon) \right]$$

$$\stackrel{DCT}{=} E \left[\lim_{\epsilon \rightarrow 0} \mathbb{1}_{(a,b)}(\Sigma_\epsilon) \right] = \overline{F}(b) - \overline{F}(a)$$

1 if $\Sigma \in (a,b)$
 if $\Sigma = a$, then we are on the left with prob $\frac{1}{2}$
 if $\Sigma = b$, " " " " " "

hence

$$\overline{F}(b) - \overline{F}(a) = \lim_{\epsilon \rightarrow 0} \int_{-\infty}^{\infty} \frac{e^{-itb} - e^{-ita}}{-2\pi it} \varphi(t) e^{-\frac{\epsilon^2 t^2}{2}} dt$$

Corollary: Two different distributions cannot have the same characteristic function.

Proof: Assume not, and F, G be two distributions with the same char. funcl. Φ

$$\text{Then } \overline{F}(b) - \overline{F}(a) = \overline{G}(b) - \overline{G}(a)$$

Letting $a \rightarrow \infty$, we get $\overline{F}(b) = \overline{G}(b)$. But at all point x , $F(x) = \lim_{y \downarrow x} \overline{F}(y) = \lim_{y \rightarrow x} \overline{G}(y)$. The same for G .

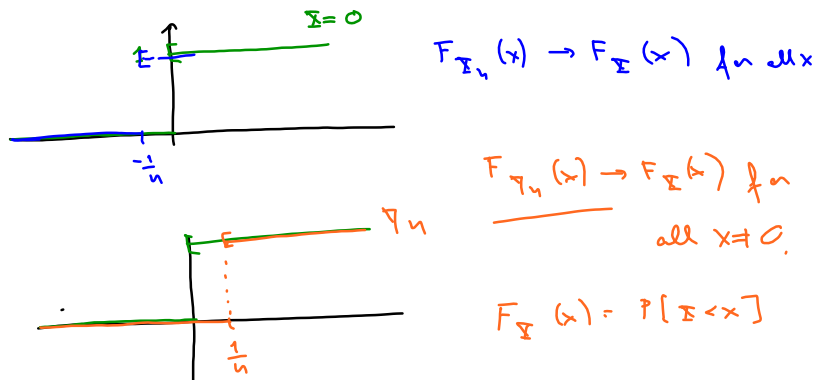
Convergence in distribution

Recall: X_n converges to X in distribution if

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x) \text{ in all points } x \text{ where } F_X \text{ is continuous.}$$

Example: $X_n = -\frac{1}{n}$ should converge to $X = 0$

$$Y_n = \frac{1}{n} \text{ --- " --- } X = 0$$



$$F_{X_n}(x) \rightarrow F_X(x) \text{ for all } x$$

$$F_{Y_n}(x) \rightarrow F_X(x) \text{ for all } x \neq 0.$$

$$F_X(x) = P[X < x]$$

Lemma: Assume that $\{X_n\}$ is a sequence of random variables with distribution function F_n . If F is the distribution function of X and

$$\lim_{n \rightarrow \infty} F_n(x) = F(x)$$

for all x in a dense set of points D , then X_n converges to X in distribution.

Proof: Let x be a continuity point of F . We need to show that $F_n(x) \rightarrow F(x)$. Let $\epsilon > 0$ be given. Since D is dense and F is cont at x , we can find points $a, b \in D$ such that

$$a < x < b \quad \begin{array}{c} a \quad x \quad b \\ | \quad | \quad | \\ \hline \end{array}$$

$$\text{such that } \underline{F(a) > F(x) - \epsilon} \text{ and } F(b) < F(x) + \epsilon$$

But then

$$\limsup_{n \rightarrow \infty} F_n(x) \leq \limsup_{n \rightarrow \infty} F_n(b) = F(b) < F(x) + \epsilon$$

$$\liminf_{n \rightarrow \infty} F_n(x) \geq \liminf_{n \rightarrow \infty} F_n(a) = F(a) > F(x) - \epsilon$$

$$\text{Hence } \limsup F_n(x) = \liminf F_n(x) = F(x) \text{ and}$$

$$\text{thus } \lim_{n \rightarrow \infty} F_n(x) = F(x).$$

$$\underline{E[f(X_n)] \rightarrow E[f(X)] \text{ for all bounded cont } f.}$$