

Monday 9 November 2020

Lecture 16

Define $L^1 := \{f: \int_{\mathbb{R}} |f(x)| dx < \infty\}$ all the functions in this set are called "L1 integrable" or "absolutely integrable"

Are PDF's in the set L^1 ? $\swarrow \searrow$ $f(x) = x^2 \notin L^1$

If $f \in L^1 \Rightarrow \exists \hat{f}$, the "Fourier transform" of f :

$$\hat{f}(\omega) = \int_{\mathbb{R}} e^{-i2\pi\omega x} f(x) dx = \mathcal{F}[f]$$

This is called "forward Fourier Transform" or "Fourier analysis". x is called the "time domain" and ω is called the "frequency domain". One of Fourier's ideas is that functions in L^1 can be decomposed into a sum of sines and cosines with different frequencies, amplitudes, and phases $|f(\omega)|$, and phase shifts, $\text{Arg}[f(\omega)]$. ω

Further, if $\hat{f} \in L^1$, then we can do a "reverse/inverse Fourier transform" to restore our original function f :

$$f(x) = \int_{\mathbb{R}} e^{i2\pi\omega x} \hat{f}(\omega) d\omega = \mathcal{F}^{-1}[\hat{f}]$$

This is called the "inverse Fourier transform" or "Fourier synthesis".

Fourier Inversion thm: If f and \hat{f} are in L^1 , then f and \hat{f} are 1:1

We define the Characteristic function (chf) for r.v. X as:

$$\begin{aligned}\phi_X(t) &:= E[e^{itX}] = \int_{\mathbb{R}} e^{itx} f_X(x) dx \\ &= \sum_{x \in \mathbb{R}} e^{itx} P_X(x)\end{aligned}$$

This is the Fourier transform with a different frequency unit $t = -2\pi\omega$.

The reason why we bother to take this crazy-looking transformation is that there are really powerful properties of the chf that will enable us to solve problems. Here are the main properties:

$\textcircled{P_0}$ $\phi_X(0) = E[e^{i(0)X}] = E[1] = 1 \quad \forall X, \forall t$

$\textcircled{P_1}$ $\phi_X(t) = \phi_Y(t) \Leftrightarrow X \stackrel{d}{=} Y$ "Uniqueness"

$\textcircled{P_2}$ If $Y = aX + b$ where $a, b \in \mathbb{R}$ $\phi_Y(t) = E[e^{it(ax+b)}] = E[e^{iatx} e^{itb}]$
 $= e^{itb} E[e^{it'x}] = e^{itb} \phi_X(t') = e^{itb} \phi_X(at)$

$\textcircled{P_3}$ X_1, X_2 are independent. $T = X_1 + X_2$

$$\phi_T(t) = E[e^{it(X_1+X_2)}] = E[e^{itX_1} e^{itX_2}] = E[e^{itX_1}] E[e^{itX_2}] = \phi_{X_1}(t) \phi_{X_2}(t)$$

$\textcircled{P_4}$ "Moment Generation"

We are able to interchange differentiation and integration here

$$\phi'_X(t) = \frac{d}{dt} [E[e^{itX}]] = E\left[\frac{d}{dt} [e^{itX}]\right] = E[ix e^{itX}]$$

$$\phi'_X(0) = E[ix] \Rightarrow E[X] = \frac{\phi'_X(0)}{i}$$

$$\phi''_X(t) = \frac{d}{dt} [E[ix e^{itX}]] = E\left[ix \frac{d}{dt} [e^{itX}]\right] = E[i^2 x^2 e^{itX}]$$

$$\phi''_X(0) = E[i^2 x^2] \Rightarrow E[X^2] = \frac{\phi''_X(0)}{i^2} \dots \dots \dots \phi^{(n)}_X(0) = E[i^n x^n]$$

$$E[X^n] = \frac{\phi^{(n)}_X(0)}{i^n}$$

P5 $\phi_x(t) \in [-1, 1]$ i.e. the chf exist $\forall x, \forall t$

⇐

$$|\phi_x(t)| \leq 1 \quad \text{proof: } |E[e^{itx}]| = \left| \int_{-\infty}^{\infty} e^{itx} f_x(x) dx \right| \leq \int_{-\infty}^{\infty} |e^{itx} f_x(x)| dx \\ \leq \int_{-\infty}^{\infty} |e^{itx}| |f_x(x)| dx = \int_{-\infty}^{\infty} |\cos(tx) + i \sin(tx)| f_x(x) dx \\ = \int_{-\infty}^{\infty} \sqrt{\cos^2(tx) + \sin^2(tx)} f_x(x) dx = \int_{-\infty}^{\infty} f_x(x) dx = 1 \quad \checkmark$$

P6 Inversion: If $\phi_x(t) \in L^1$ then $f_x(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-itx} \phi_x(t) dt$

P7 Levy's CDF formula

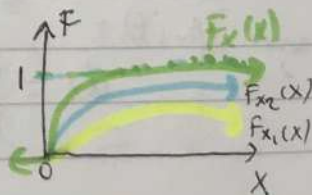
Probability

$$P(x \in [a, b]) = \frac{1}{2\pi i} \int_{-\infty}^{\infty} \frac{e^{-ita} - e^{-itb}}{it} \phi_x(t) dt$$

P8 Levy's Continuity Thm

Consider a sequence of RV's X_1, X_2, \dots, X_n . We define " X_n converges in distribution to x " (denoted $X_n \xrightarrow{d} x$) as:

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_x(x) \quad \forall x \text{ "pointwise convergence"}$$



$$\lim_{n \rightarrow \infty} \phi_{X_n}(t) = \phi_x(t) \Rightarrow X_n \xrightarrow{d} x.$$

If n large $\phi_{X_n}(t) \approx \phi_x(t) \Rightarrow X_n \xrightarrow{d} x$

Define $M_X(t) = E[e^{tx}]$, the moment generating function (mgf)

Properties:

P_0 $M_X(0) = 1 \quad \forall X$

P_1 $M_X(t) = M_Y(t) \Rightarrow X \stackrel{d}{=} Y$

P_2 $Y = aX + b, \quad M_Y(t) = e^{tb} M_X(at)$

P_3 X_1, X_2 indep $T = X_1 + X_2, \quad M_T(t) = M_{X_1}(t) M_{X_2}(t)$

P_4 $E[X^n] = M_X^{(n)}(0)$

I don't have P_5 . Thus, sometime mgf's don't exist at all and sometimes it doesn't exist for certain values of t .

Chf's can do everything mgf's can do and more!! Thus, you don't need mgf's!

Ex 1 $X \sim \text{Gamma}(\alpha, \beta)$

$$\phi_X(t) = E[e^{itx}] = \int_0^\infty e^{itx} \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \mathbb{1}_{x>0} dx$$

$$\frac{\beta^\alpha}{\Gamma(\alpha)} \int_0^\infty x^{\alpha-1} e^{-(\beta-it)x} dx = \frac{\beta^\alpha}{\Gamma(\alpha)} \frac{\Gamma(\alpha)}{(\beta-it)^\alpha} = \left(\frac{\beta}{\beta-it} \right)^\alpha$$

Ex 2 $X_1 \sim \text{Gamma}(\alpha_1, \beta)$ indep of $X_2 \sim \text{Gamma}(\alpha_2, \beta), T = X_1 + X_2$

P_2

$$\phi_T(t) \stackrel{P_3}{=} \phi_{X_1}(t) \phi_{X_2}(t) = \left(\frac{\beta}{\beta-it} \right)^{\alpha_1} \left(\frac{\beta}{\beta-it} \right)^{\alpha_2} = \left(\frac{\beta}{\beta-it} \right)^{\alpha_1 + \alpha_2}$$

$P_1 \Rightarrow T \sim \text{Gamma}(\alpha_1 + \alpha_2, \beta)$