

标准正态分布的双边 Laplace 变换

$$P_Y(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}, y \in \mathbb{R}$$

$$\mathcal{L} \circ P_Y(y) \stackrel{\text{def}}{=} \int_{-\infty}^{+\infty} e^{-py} dy \cdot \left(\frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} \right)$$

$P_Y(y)$ 在实数轴上收敛，可以定义双边 \mathcal{L} .

即 $\boxed{\mathcal{L} \circ P_Y(y) = e^{\frac{p^2}{2}}$

$T \sim N(0,1)$ $N(0,1)$ 的双边 \mathcal{L} .

以下记 $T \sim N(0,1)$ 为 $P_T(y)$

$X \sim$ 任意分布 为 $P_X(x)$

$$\mathcal{L} \circ P_Y(y) = \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2+2py}{2}} dy$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\left(\frac{y^2+2py+p^2}{2}\right) + \frac{p^2}{2}} dy$$

$$= e^{\frac{p^2}{2}} \cdot \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(y+p)^2}{2}} dy = e^{\frac{p^2}{2}}$$

22 [例] \mathcal{L} 变换与原点矩的关系.

$$\lim_{P \rightarrow 0} \mathcal{L} \circ P_x(x) = \lim_{P \rightarrow 0} \int_{-\infty}^{+\infty} dx \cdot e^{-Px} \circ P_x(x)$$

$$= \int_{-\infty}^{+\infty} dx \circ \left[\lim_{P \rightarrow 0} e^{-Px} \right] \circ P_x(x)$$

$$= \int_{-\infty}^{+\infty} dx \circ 1 \circ P_x(x) = \int_{-\infty}^{+\infty} f_x(x) dx = 1$$

$$\therefore \boxed{\lim_{P \rightarrow 0} \mathcal{L} \circ P_x(x) = 1 = E(x^0)}$$

(概率的规范性)

$$\lim_{P \rightarrow 0} \left(-\frac{d}{dp} \right) \circ \mathcal{L} \circ P_x(x) = \lim_{P \rightarrow 0} \left(-\frac{d}{dp} \right) \circ \int_{-\infty}^{+\infty} dx \cdot e^{-Px} \circ P_x(x)$$

$$= \lim_{P \rightarrow 0} \int_{-\infty}^{+\infty} dx \circ \boxed{-\frac{d}{dp} \circ e^{-Px}} \circ P_x(x)$$

$$= \lim_{P \rightarrow 0} \int_{-\infty}^{+\infty} dx \circ \boxed{e^{-Px} \circ x} \circ P_x(x)$$

$$= \int_{-\infty}^{+\infty} dx \circ \boxed{\lim_{P \rightarrow 0} e^{-Px}} \circ x \circ P(x) = 1$$

$$= \int_{-\infty}^{+\infty} x \circ P(x) dx = E(x)$$

$$\therefore \boxed{\lim_{P \rightarrow 0} \left(-\frac{d}{dp} \right) \circ \mathcal{L} \circ P_x(x) = E(x)}$$

可进一步推广有

$$\lim_{p \rightarrow 0} (-1)^n \circ \left(\frac{d^n}{dp^n} \right) \circ L \circ P_x(x) = E(X^n)$$

可以简化写成

$$\boxed{(-1)^n \left[L \{ P_x(x) \} \right] \Big|_{p=0} = E(X^n)}$$

或 $F_x(p) \stackrel{\text{def}}{=} L \{ P_x(x) \}$

$$\boxed{(-1)^n F_x^{(n)}(0) = E(X^n)}$$

即 对 随机变量 X

① 先求 L ② 再求 n 阶导 ③ 再乘 $(-1)^n$ ④ 再令 $p \rightarrow 0$

求 X 的 n 阶原点矩.

① 先取 n 次幂 ② 再求期望.

卷积与随机变量的和

$$z \stackrel{\text{def}}{=} x_1 + x_2$$

$$P_{x_1+x_2}(z) = P_z(z) = \iint_{\{(x_1, x_2) | x_1 + x_2 = z\}} P_x(x_1) \cdot P_x(x_2) dx_1 dx_2$$

$$P_{x_1+x_2}(z) = P_z(z) = \int_{-\infty}^{+\infty} P_x(x_1) (z - x_1) dx_1$$

$$\therefore P_{x_1+x_2}(z) = P_x(x_1) * P_x(x_2)$$

↑
卷积(据定义)

即 和的密度 = 密度的卷积
可以推广到多个(连续和~连续卷积)

用算子表示

$$\underbrace{P \circ \sum_{i=1}^n}_{f(x)} \circ X_i = \underbrace{(\star) \circ P}_{\text{根元}} \circ X_i$$

先求和，再做根元平测度映射

} 作用于随机变量 X_i

先做根元平测度映射，再求和

$$\sum_{i=1}^n = P^{-1} \circ (\star) \circ P$$

即 卷积与求和 作为

算符 通过 根元平测度映射 P.

是相似的

$$\sum_{i=1}^n \sim (\star)_{i=1}^n$$

乘积与随机变量的和.

$$\sum_{i=1}^n \sim (\star) = [P \circ \sum_{i=1}^n] \circ x_i = [(\star) \circ P] \circ x_i$$

$$\sum_{i=1}^n = P^{-1} \circ (\star) \circ P$$

$$\prod_{i=1}^n \sim (\#) : [\prod_{i=1}^n \circ L] \circ f(x) = [L \circ (\#) \circ f] \circ x$$

$$(\#) = L^{-1} \circ \prod_{i=1}^n \circ L$$

$$\sum_{i=1}^n = P^{-1} \circ (\star) \circ P$$

⊗

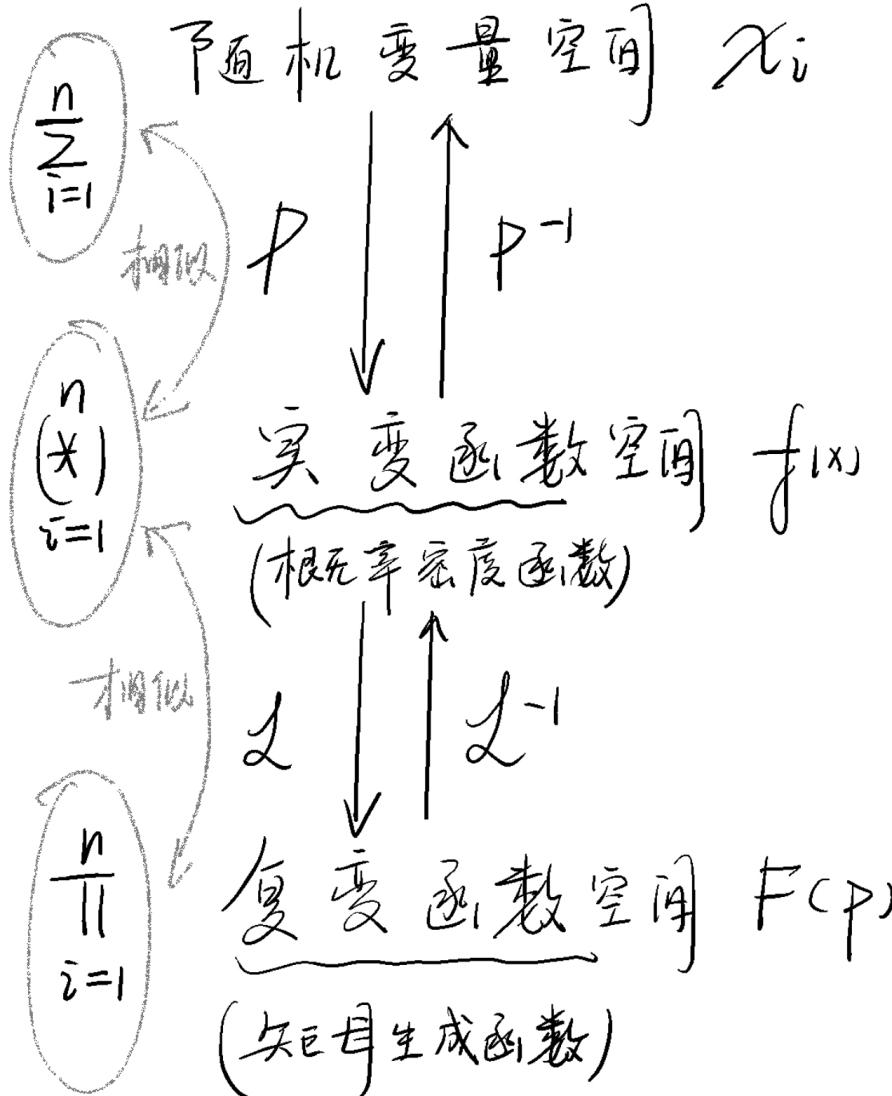
$$(\star) = L^{-1} \circ \prod_{i=1}^n \circ L$$

↓

$$\sum_{i=1}^n = P^{-1} \circ L^{-1} \circ \prod_{i=1}^n \circ L \circ P$$

即

$$P\left(\sum_{i=1}^n x_i\right) = L^{-1} \circ \prod_{i=1}^n \circ L \{P(x_i)\}$$



X_i : 独立同分布, $i=1, 2, \dots, n$
共有 n 个 $E(X_i) = \mu$ $D(X_i) = \sigma^2$

$$Y_i = \frac{X_i - \mu}{\sigma}, \text{ 各自归一化}$$

$$E(Y_i) = 0 \quad D(Y_i) = 1$$

$$\bar{X}_i = \frac{1}{n} \sum X_i, \text{ 平均. } E(\bar{X}_i) = \mu \quad D(\bar{X}_i) = \frac{\sigma^2}{n}$$

$$T_n = \frac{\bar{X}_i - \mu}{\left(\frac{\sigma}{\sqrt{n}}\right)} \text{ 平均后归一化 } E(T_n) = 0 \quad D(T_n) = 1$$

$$T_n = \frac{\left(\frac{1}{n} \sum Y_i + \mu\right) - \mu}{\frac{\sigma}{\sqrt{n}}} = \frac{\sum Y_i}{\sqrt{n}}$$

$$\boxed{T_n = \frac{\sum Y_i}{\sqrt{n}}} \neq$$

$$T_n = \frac{\sum Y_i}{dn}$$

將 $F(s)$ 以 Taylor 展開 ($s=0$ 处)

$$P \circ T_n = P \circ \sum_{i=1}^n o \left(\frac{Y_i}{dn} \right)$$

$$= \underbrace{\left(\sum_{i=1}^n o \left(\frac{Y_i}{dn} \right) \right)}_{\text{即 } P(T_n)}$$

$$= L^{-1} \circ \underbrace{\prod_{i=1}^n o \left(\frac{Y_i}{dn} \right)}_{\text{即 } P(T_n)}$$

$$\text{即 } P(T_n) = L^{-1} \circ \left(\underbrace{L \{ P \left(\frac{Y_i}{dn} \right) \}}_n \right)$$

$$F(s) \stackrel{\text{def}}{=} L \{ P \left(\frac{Y_i}{dn} \right) \}$$

$$F(s) = F(0) + F'(0)s + \frac{1}{2}F''(0)s^2 + o(s^2)$$

$$F(0) = E \left(\left(\frac{Y_i}{dn} \right)^0 \right) = 1 \quad Y_i \text{ 服从 } \text{Exp}$$

$$F'(0) = -E \left(\left(\frac{Y_i}{dn} \right)^1 \right) = -\frac{1}{dn} E(Y_i) = 0$$

$$F''(0) = E \left(\left(\frac{Y_i}{dn} \right)^2 \right) = \frac{1}{n} E(Y_i^2) = \frac{1}{n}$$

$$\therefore F(s) = 1 + \frac{1}{2n}s^2 + o(s^2)$$

先分析 $F(s)$

所有鋪墊都已完成

梳理一下思路

$$P^o T_n = P^o \sum_{i=1}^n o X_i = (*) o P^o X_i$$

$$= L^{-1} o \prod_{i=1}^n o L^o P^o X_i$$

$$= L^{-1} o \exp o \left[\ln o \prod_{i=1}^n o \underbrace{L^o P^o X_i}_{\text{II}} \right]$$
$$F(s) = 1 + \frac{1}{2n} s^2 + o(s^2)$$

接下來求

$$\begin{aligned} & \ln \left(1 + \frac{1}{2n} s^2 + o(s^2) \right)^n \\ &= n \ln \left(1 + \frac{1}{2n} s^2 + o(s^2) \right) \end{aligned}$$

$$\lim_{n \rightarrow \infty} n \ln \left(1 + \frac{1}{2n} s^2 \right)$$

$$= n \times \frac{s^2}{2n} = \frac{s^2}{2}$$

$$\therefore P^o T_n = \underset{n \rightarrow \infty}{L^{-1} \{ e^{\frac{s^2}{2}} \}} = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}$$

-開始證了

$$L \{ P_R(y) \} = e^{\frac{s^2}{2}}$$

$$\lim_{n \rightarrow \infty} P\left(a < \frac{\bar{x}_i - \mu}{\left(\frac{\sigma}{\sqrt{n}}\right)} < b\right)$$

$$\int_a^b \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy.$$

Central Limit Theorem!

铺垫

$$① \mathcal{L}\left\{ \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} \right\} = e^{-\frac{s^2}{2}}$$

$$② \lim_{s \rightarrow 0} (-1)^n \frac{d^n}{ds^n} \circ \mathcal{L} \circ P_0 z_i = E(z_i^n)$$

s 是 \mathcal{L} 后的复变量

$$③ \frac{1}{\sqrt{n}} \bar{Y}_i = \frac{\bar{x}_i - \mu}{\sigma} \quad \boxed{\text{[} \bar{x}_i \text{ 的归一化]}}$$

$$z_i = \frac{Y_i}{\sqrt{n}} \quad \boxed{\text{[} T_n \text{ 展开各项]}}$$

$$\overline{T_n} = \frac{\overline{\bar{x}_i} - \mu}{\left(\frac{\sigma}{\sqrt{n}}\right)} = \sum_{i=1}^n z_i$$

\uparrow
[\bar{x}_i 的归一化]

总体思路

$$\lim_{n \rightarrow \infty} \circ P \circ T_n = \lim_{n \rightarrow \infty} \circ P \circ \sum_{i=1}^n Z_i$$

$$= \lim_{n \rightarrow \infty} \circ \left(\prod_{i=1}^n \right) \circ P \circ Z_i$$

$$= \lim_{n \rightarrow \infty} \circ L^{-1} \circ \prod_{i=1}^n \circ L \circ P \circ Z_i$$

$$= \lim_{n \rightarrow \infty} \circ L^{-1} \circ \exp \circ \ln \circ \prod_{i=1}^n \circ L \circ P \circ Z_i$$

$$= L^{-1} \circ \exp \circ \lim_{n \rightarrow \infty} \circ \ln \circ \prod_{i=1}^n \circ L \circ P \circ Z_i$$

对这一部分 Taylor (在 $s=0$ 邻域)

展开到 2 阶并转化为原点矩计算

① 首先证明

$$\mathcal{L} \{ \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} \} = e^{\frac{s^2}{2}}$$

为后续 $\mathcal{L}^{-1} \{ e^{\frac{s^2}{2}} \}$ 准备

② 再证明 $\lim_{s \rightarrow 0} (-1)^n \frac{d^n}{ds^n} \mathcal{L} \circ P \circ Z_i = E(Z_i^n)$

为后续 对 $\mathcal{L} \{ P(Z_i) \}$ Taylor 展开
做准备

③ 随机变量归一化 $\begin{cases} T_i = \frac{X_i - \mu}{\sigma} \\ Z_i = \frac{T_i}{\sqrt{n}} \end{cases}$

为了 CLT 定理叙述漂亮.

$\begin{cases} \text{Taylor 系数简洁} \end{cases}$