

Convergence in distributionDef

The sequence  $X_n$  of random variables with values in  $\mathbb{R}^d$  converges in distribution to  $X$  if for all  $f: \mathbb{R}^d \rightarrow \mathbb{R}$  bounded and continuous

$$\mathbb{E}(f(X_n)) \longrightarrow \mathbb{E}(f(X)) \text{ as } n \rightarrow \infty.$$

Notation:  $X_n \xrightarrow{d} X$ .

Proposition

Let  $X_n$  and  $X$  be real random variables and let  $F_n(x) = \mathbb{P}(X_n \leq x)$  and  $F(x) = \mathbb{P}(X \leq x)$  be their distribution function.

Then  $X_n \xrightarrow{d} X$  if and only if  $F_n(x) \rightarrow F(x)$  for all  $x$  which are points of continuity of  $F$ .

Central limit theorem (CLT)

Let  $(X_n)_{n=1}^{\infty}$  be independent identically distributed (iid) random variables. Suppose that  $\mathbb{E}(X_1) = \mu$  and  $\text{Var}(X_1) = \sigma^2 < \infty$ .

For  $S_n = X_1 + \dots + X_n$ ,

$$\frac{S_n - n\mu}{\sigma\sqrt{n}} \xrightarrow{d} \mathcal{X}$$

where  $\mathcal{X} \sim N(0,1)$ .

Characteristic function

For a random variable  $X$ , let its characteristic function be

$$\varphi(t) = \mathbb{E}(e^{itX}) \text{ for any } t \in \mathbb{R}.$$

Properties:

- the expectation exists for all  $t \in \mathbb{R}$  and  $|\varphi(t)| \leq 1$
- the same as the Fourier transform of the distribution (measure)/density
- as a consequence, the characteristic function determines the distribution
- Continuity theorem

If  $X_n \xrightarrow{d} X$  and  $\varphi_n(t) = \mathbb{E}(e^{itX_n})$  and  $\varphi(t) = \mathbb{E}(e^{itX})$  then  $\varphi_n(t) \rightarrow \varphi(t)$  for all  $t \in \mathbb{R}$ . If  $\varphi_n(t) \rightarrow \varphi(t)$  for all  $t \in \mathbb{R}$  where  $\varphi_n(t) = \mathbb{E}(e^{itX_n})$  and  $\varphi(t)$  is continuous at 0, then there is an  $X$  s.t.  $X_n \xrightarrow{d} X$  and  $\varphi(t) = \mathbb{E}(e^{itX})$ .

- if  $X \sim N(0,1)$ , then  $\varphi(t) = \mathbb{E}(e^{itX}) = e^{-\frac{t^2}{2}}$ .
- $\varphi(t) = \mathbb{E}(e^{itX}) = 1 + it\mathbb{E}X - \frac{t^2\mathbb{E}(X^2)}{2} + o(t^2)$  if  $\mathbb{E}X, \mathbb{E}(X^2) < \infty$ .

Proof of the CLT

Assume  $\mathbb{E}X_n = 0$ .

We compute

$$\begin{aligned} \mathbb{E}\left(e^{it \frac{S_n}{\sigma\sqrt{n}}}\right) &= \mathbb{E}\left(e^{i \frac{t}{\sigma\sqrt{n}}(X_1 + \dots + X_n)}\right) = \prod_{k=1}^n \mathbb{E}\left(e^{i \frac{t}{\sigma\sqrt{n}} X_k}\right) \\ &\quad \uparrow \\ &\quad \text{indep.} \\ &= \left(\mathbb{E}\left(e^{i \frac{t}{\sigma\sqrt{n}} X_1}\right)\right)^n = \left(1 - \frac{\left(\frac{t}{\sigma\sqrt{n}}\right)^2 \mathbb{E}(X_1^2)}{2} + o\left(\frac{t^2}{n}\right)\right)^n \rightarrow e^{-\frac{t^2}{2}} \quad \square \end{aligned}$$

Domain of attraction

Def

The distribution function  $F$  is in the domain of attraction of the distribution  $\phi$  if  $\exists a_n > 0$  and  $b_n \in \mathbb{R}$  s.t. if  $X_1, X_2, \dots$  are iid with distribution  $F$  and  $S_n = X_1 + \dots + X_n$ , then  $\frac{S_n - b_n}{a_n} \xrightarrow{d} \phi$  where  $\phi$  has distribution  $\phi$ . Notation:  $F \in DA(\phi)$ .

Thm

The distribution  $F$  is in the domain of attraction of the normal dist., iff

- 1)  $F$  has finite variance (in which case the CLT applies) or
- 2)  $F$  satisfies the condition  $\lim_{x \rightarrow \infty} \frac{x^2 P(|X| > x)}{\int_{|y| \leq x} y^2 dF(y)} = 0 \quad (*)$

where  $X$  has distribution  $F$ .

Exercise

Show that if  $F$  has finite variance, then the condition (\*) holds, i.e. in the thm above, condition 1) implies 2).

Stable laws

Goal: find other non-degenerate random variable  $Y$  s.t. there is an iid. sequence  $X_1, X_2, \dots$  with  $S_n = X_1 + \dots + X_n$  and  $a_n > 0$  and  $b_n \in \mathbb{R}$  for which  $\frac{S_n - b_n}{a_n} \xrightarrow{d} Y$  holds.

Example (symmetric)

Let  $X_1, X_2, \dots$  be an iid. sequence with distribution

$$\mathbb{P}(|X_1| > x) = x^{-\alpha} \text{ for } x \geq 1 \text{ and } \mathbb{P}(X_1 > x) = \mathbb{P}(X_1 < -x).$$

where  $\alpha \in (0, 2)$ . Density:  $f(x) = \frac{\alpha}{2} |x|^{-\alpha-1}$  for  $|x| > 1$



Characteristic function:  $\varphi(t) = \mathbb{E}(e^{itX_1})$

$$\begin{aligned} \text{For } t > 0, \quad 1 - \varphi(t) &= \int_1^{\infty} (1 - e^{itx}) \frac{\alpha}{2x^{\alpha+1}} dx + \int_{-\infty}^{-1} (1 - e^{itx}) \frac{\alpha}{2|x|^{\alpha+1}} dx \\ &= \int_1^{\infty} \left(1 - \frac{e^{itx} + e^{-itx}}{2}\right) \frac{\alpha}{x^{\alpha+1}} dx = \alpha \int_1^{\infty} \frac{1 - \cos(tx)}{x^{\alpha+1}} dx \end{aligned}$$

$$= \alpha t^\alpha \int_t^{\infty} \frac{1 - \cos u}{u^{\alpha+1}} du$$

• as  $u \rightarrow \infty$ , the integrand is  $\approx \frac{1}{u^{\alpha+1}}$  integrable for  $\alpha \in (0, 2)$

• as  $u \rightarrow 0$ ,  $\frac{1 - \cos u}{u^{\alpha+1}} \approx \frac{u^2/2}{u^{\alpha+1}} = \frac{1}{2} u^{1-\alpha}$  integrable for  $\alpha \in (0, 2)$

Let  $C = \alpha \int_0^{\infty} \frac{1 - \cos u}{u^{\alpha+1}} du$ . Then  $1 - \varphi(t) = C|t|^\alpha + o(|t|^\alpha)$  as  $t \rightarrow 0$  since  $\varphi(-t) = \varphi(t)$  by the symmetry of the distribution.

Let  $S_n = X_1 + \dots + X_n$ . We show that  $\frac{S_n}{n^{1/\alpha}}$  converges in distr:

$$\mathbb{E}\left(e^{it \frac{S_n}{n^{1/\alpha}}}\right) \stackrel{\text{i.i.d.}}{=} \left(\mathbb{E}\left(e^{i \frac{t}{n^{1/\alpha}} X_1}\right)\right)^n = \varphi\left(\frac{t}{n^{1/\alpha}}\right)^n = \left(1 - \frac{C|t|^\alpha}{n} + o\left(\frac{|t|^\alpha}{n}\right)\right)^n$$

$\rightarrow e^{-C|t|^\alpha}$  as  $n \rightarrow \infty$  which is continuous at  $t=0$ .

By the continuity thm,  $\frac{S_n}{n^{1/\alpha}} \xrightarrow{d} Y$  and  $\mathbb{E}(e^{itY}) = e^{-C|t|^\alpha}$ .  $\square$

## Remarks

- In the example above, we assumed  $\alpha \in (0, 2)$ . For  $\alpha > 2$ , the variance is finite  $\Rightarrow$  CLT holds.
- For  $\alpha = 2$ , the variance is infinite, but condition (\*) holds. Actually,  $\frac{S_n}{\sqrt{n \log n}} \xrightarrow{d} N(0, \sigma^2)$  can be seen by characteristic function.
- For  $\alpha \in (0, 1)$ ,  $n^{1/\alpha} \gg n$ , hence the main contribution of  $S_n$  comes from a few terms of size  $n^{1/\alpha}$ .

## Def

The distribution of the random variables  $X$  and  $Y$  is of the same type if  $\exists a > 0$  and  $b \in \mathbb{R}$  s.t.  $aX + b \stackrel{d}{=} Y$ .

## Def

$Y$  has a stable law if for every positive integer  $k$ ,  $\exists a_k > 0$  and  $b_k \in \mathbb{R}$  s.t. if  $Y_1, \dots, Y_k$  are iid. with the distr. of  $Y$ , then  $\frac{Y_1 + \dots + Y_k - b_k}{a_k} \stackrel{d}{=} Y$ , i.e.  $Y_1 + \dots + Y_k$  and  $Y$  have distributions of the same type.

## Thm

The symmetric stable laws are those which appeared in the example, i.e. the ones with characteristic function  $e - c|t|^\alpha$  for index  $\alpha \in (0, 2)$ .

## Def

The function  $L: \mathbb{R} \rightarrow \mathbb{R}$  is slowly varying if  $\lim_{x \rightarrow \infty} \frac{L(tx)}{L(x)} = 1$  for all  $t > 0$ .

## Thm (non-symmetric)

Let  $X_1, X_2, \dots$  iid with distribution  $\mathbb{P}(X_1 > x) = x^{-\alpha} L(x)$  for some index  $\alpha \in (0, 2)$  and slowly varying  $L$  and skewness  $\kappa = \lim_{x \rightarrow \infty} \frac{\mathbb{P}(X_1 > x) - \mathbb{P}(X_1 < -x)}{\mathbb{P}(X_1 > x)} \in [-1, 1]$ .

Let  $S_n = X_1 + \dots + X_n$ ,  $a_n = \inf \{x : \mathbb{P}(X_1 > x) \leq \frac{1}{n}\}$  and  $b_n = n \mathbb{E}(X_1 \mathbb{1}(X_1 \leq a_n))$ .

Then  $\frac{S_n - b_n}{a_n} \xrightarrow{d} Y$  as  $n \rightarrow \infty$  where  $Y$  is non-degenerate.

The distribution of  $Y$  is stable with index  $\alpha$  and skewness  $\kappa$ .

There are no other stable laws.

Example

Let  $\alpha \in (0, 2)$  and  $K \in [-1, 1]$ .

$$\begin{aligned}
 P(X_1 > x) &= \frac{1+K}{2} x^{-\alpha} \\
 P(X_1 < -x) &= \frac{1-K}{2} x^{-\alpha}
 \end{aligned}
 \left. \begin{array}{l} \text{for } x \geq 1 \end{array} \right\} \text{density } f(x) = \frac{1+K}{2} \alpha |x|^{-\alpha-1} \text{ for } |x| \geq 1$$

By the thm above,  $a_n = n^{1/\alpha}$  and  $b_n = n \int_1^{n^{1/\alpha}} x (f(x) - f(-x)) dx$

$$= n \int_1^{n^{1/\alpha}} K \alpha x^{-\alpha} dx \asymp \begin{cases} n & \alpha > 1 \\ n \log n & \alpha = 1 \\ n^{1/\alpha} & \alpha < 1 \end{cases}$$

Special stable laws (no explicit formulas otherwise)

$\alpha = 2$  normal distr. (no  $K$  dependence)  $\varphi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$

$\alpha = 1, K = 0$  Cauchy distr.  $f(x) = \frac{1}{\pi} \frac{1}{1+x^2}$

$\alpha = 1/2, K = 1$  Levy distribution  $f(x) = \frac{1}{\sqrt{2\pi x^3}} e^{-\frac{1}{2x}}$  for  $x > 0$   
(appears as hitting time distribution for BM)

Thm

$Y$  has stable law iff  $Y$  appears as the limit of  $\frac{X_1 + \dots + X_n - b_n}{a_n}$  for some iid  $X_i, K$

Thm (convergence of types)

Suppose that  $\frac{X_n - b_n}{a_n} \xrightarrow{d} U$  and  $\frac{X_n - \beta_n}{\alpha_n} \xrightarrow{d} V$

for non-degenerate rvs  $X_n, U, V$ ;  $a_n, \alpha_n > 0$ ;  $b_n, \beta_n \in \mathbb{R}$ .

Then  $\frac{a_n}{\alpha_n} \rightarrow A > 0$  and  $\frac{\beta_n - b_n}{\alpha_n} \rightarrow B \in \mathbb{R}$  and  $V \stackrel{d}{=} \frac{U - B}{A}$ .

Proof of thm on stable laws

$\Rightarrow$  Let  $X_1, X_2, \dots$  be iid. with the stable law of  $Y$ .

$\Leftarrow$  Fix an integer  $k$  and let  $S_{nk} = S_n^{(1)} + \dots + S_n^{(k)}$

where  $S_n^{(j)} = X_{(j-1)n+1} + \dots + X_{jn}$  for which  $\frac{S_n^{(j)} - b_n}{a_n} \xrightarrow{d} Y$

Hence  $\frac{S_{nk} - kb_n}{a_n} = \frac{S_n^{(1)} - b_n}{a_n} + \dots + \frac{S_n^{(k)} - b_n}{a_n} \xrightarrow{d} Y_1 + \dots + Y_k$  where  $Y_1, \dots, Y_k$  are iid with distr. of  $Y$

On the other hand,  $\frac{S_{nk} - b_{nk}}{a_{nk}} \xrightarrow{d} Y$ .

By convergence of types:  $\frac{Y_1 + \dots + Y_k - B_k}{A_k} \stackrel{d}{=} Y$  where  $A_k = \lim_{n \rightarrow \infty} \frac{a_{nk}}{a_n}$ ,  $B_k = \lim_{n \rightarrow \infty} \frac{b_{nk} - kb_n}{a_n}$ . □

If  $H: \mathbb{R} \rightarrow \mathbb{R}$  is a non-decreasing function, then let  $H^{-1}(y) = \inf\{s: H(s) > y\}$  be its right continuous inverse.

Prop (technical)

If  $H_n$  are non-decreasing and  $H_n \rightarrow H$  for all points of continuity of  $H$ , then  $H_n^{-1} \rightarrow H^{-1}$  for all points of continuity of  $H^{-1}$ .

Proof of convergence of types thm

With the notation  $F_n(x) = \mathbb{P}(X_n < x)$ ,  $U(x) = \mathbb{P}(U < x)$ ,  $V(x) = \mathbb{P}(V < x)$ , we have  $F_n(a_n x + b_n) \rightarrow U(x)$  for all points of continuity of  $U$ .

By the prop. above and by taking inverses:

$$\frac{F_n^{-1}(y) - b_n}{a_n} \rightarrow U^{-1}(y) \quad \text{and similarly} \quad \frac{F_n^{-1}(y) - \beta_n}{\alpha_n} \rightarrow V^{-1}(y)$$

for all points of continuity of the RHS.

Let  $y_1 < y_2$  with  $U^{-1}(y_1) < U^{-1}(y_2)$  and  $V^{-1}(y_1) < V^{-1}(y_2)$ .

$$\text{Then} \quad \frac{F_n^{-1}(y_i) - b_n}{a_n} \rightarrow U^{-1}(y_i) \quad \text{and} \quad \frac{F_n^{-1}(y_i) - \beta_n}{\alpha_n} \rightarrow V^{-1}(y_i) \quad i=1,2$$

By subtraction,

$$\frac{F_n^{-1}(y_2) - F_n^{-1}(y_1)}{a_n} \rightarrow U^{-1}(y_2) - U^{-1}(y_1) > 0 \quad \text{and} \quad \frac{F_n^{-1}(y_2) - F_n^{-1}(y_1)}{\alpha_n} \rightarrow V^{-1}(y_2) - V^{-1}(y_1) > 0$$

$$\text{By division,} \quad \frac{\alpha_n}{a_n} \rightarrow \frac{U^{-1}(y_2) - U^{-1}(y_1)}{V^{-1}(y_2) - V^{-1}(y_1)} =: A > 0$$

$$\text{With this,} \quad \frac{F_n^{-1}(y_i) - b_n}{a_n} \rightarrow U^{-1}(y_i) \quad \text{and} \quad \frac{F_n^{-1}(y_i) - \beta_n}{a_n} \rightarrow A \cdot V^{-1}(y_i)$$

$$\text{By subtraction,} \quad \frac{\beta_n - b_n}{a_n} \rightarrow U^{-1}(y_1) - A V^{-1}(y_1) =: B$$

$$\text{Since} \quad \frac{X_n - b_n}{a_n} = \frac{\alpha_n}{a_n} \frac{X_n - \beta_n}{\alpha_n} + \frac{\beta_n - b_n}{a_n} \stackrel{d}{\Rightarrow} AV + B \quad \text{and} \quad \frac{X_n - b_n}{a_n} \stackrel{d}{\Rightarrow} U,$$

$AV + B \stackrel{d}{=} U$  as required.  $\square$