

Lévy's Continuity Theorem

Fact here: For all  $A > 0$ ,  
 $\mu([-2A, 2A]) \geq A \left| \int_{-1/A}^{1/A} \varphi(t) dt \right| - 1$

Lévy's Continuity Theorem: Let  $\{\mu_n\}$  be a sequence of distributions with characteristic functions  $\varphi_n$ . Assume that there is a function  $\varphi_\infty$  such that  
 (i)  $\varphi_n(t) \rightarrow \varphi_\infty(t)$  for all  
 (ii)  $\varphi_\infty$  is continuous at 0.

Then

- a) There is a distribution  $\mu_\infty$  such that  $\mu_n \Rightarrow \mu_\infty$ .
- b)  $\varphi_\infty$  is the characteristic function of  $\mu_\infty$ .

Proof: Plan: Use Helly's Theorem to prove that there is a convergent subsequence  $\{\mu_{n_k}\}$  and that the limit  $\mu_\infty$  has char. func.  $\varphi_\infty$ . Extend to the entire sequence.

Assume that  $\varepsilon > 0$  is given. Since  $\varphi_\infty$  is continuous at the origin, there is a  $\delta > 0$  such that  $|1 - \varphi_\infty(t)| < \frac{\varepsilon}{4}$  when  $|t| < \delta$ . Now apply

$$\begin{aligned} & \text{the lemma with } A = \frac{1}{\delta} : \\ & \mu_n\left[-\frac{2}{\delta}, \frac{2}{\delta}\right] \geq \frac{1}{\delta} \left| \int_{-\delta}^{\delta} \varphi_n(t) dt \right| - 1 \\ & \geq \frac{1}{\delta} \left| \int_{-\delta}^{\delta} [\varphi_\infty(t) - (\varphi_\infty(t) - \varphi_n(t))] dt \right| - 1 \\ & \geq \frac{1}{2\delta} \left| \int_{-\delta}^{\delta} \varphi_\infty(t) dt \right| - \frac{1}{\delta} \int_{-\delta}^{\delta} |\varphi_\infty(t) - \varphi_n(t)| dt - 1 \\ & \geq 2 \left(1 - \frac{\varepsilon}{4}\right) - \frac{\varepsilon}{2} - 1 = 2 - \frac{\varepsilon}{2} - \frac{\varepsilon}{2} - 1 = \underline{1 - \varepsilon} \end{aligned}$$

Hence  $\mu_n\left[-\frac{2}{\delta}, \frac{2}{\delta}\right] \geq 1 - \varepsilon$  for  $n \geq N$ .

This means that for  $n \geq N$ ,  $F_n\left(\frac{2}{\delta}\right) \geq 1 - \varepsilon$

$$F_n\left(-\frac{2}{\delta}\right) \leq \varepsilon$$

Let  $n \rightarrow \infty$ , then  $F_\infty\left(\frac{2}{\delta}\right) \geq 1 - \varepsilon$ ,  $F_\infty\left(-\frac{2}{\delta}\right) \leq \varepsilon$ .

Thus  $F_\infty(x) \rightarrow 1$  as  $x \rightarrow \infty$ ,  $F_\infty(-x) \rightarrow 0$  as  $x \rightarrow -\infty$ .

By Helly's Theorem, there is a subsequence  $\{\mu_{n_k}\}$  such that  $\mu_{n_k} \Rightarrow \mu_\infty$  for some distribution  $\mu_\infty$ .

Note first that since  $\mu_{n_k} \Rightarrow \mu_\infty$ , then  $\varphi_{n_k} \rightarrow \varphi$  where  $\varphi$  is the char. func.  $\mu_\infty$ .

But  $\varphi_{n_k}(t) \rightarrow \varphi_\infty(t)$ , hence  $\varphi = \varphi_\infty$ . Hence  $\varphi_\infty$  is the characteristic function of  $\mu_\infty$ .

There is a metric  $d$  on the space of all distributions such that  $\mu_n \Rightarrow \mu_\infty$  is equivalent to  $d(\mu_n, \mu_\infty) \rightarrow 0$  (see Prob. 6.34). Assume for contradiction that  $\mu_n \not\Rightarrow \mu_\infty$ , then there must be a  $\varepsilon > 0$ , such that  $d(\mu_n, \mu_\infty) > \varepsilon$  for infinitely many  $n$ . Make a subsequence  $\nu_n$  of these  $n$ 's, i.e.  $d(\nu_n, \mu_\infty) > \varepsilon$ .

Apply the argument above to  $\{\nu_n\}$ , then we get a subsequence  $\{\nu_{n_k}\}$  converging to a limit distribution  $\nu_\infty$ . As before, the char. function of  $\nu_\infty$  is  $\varphi_\infty$ , hence  $\nu_\infty = \mu_\infty$ .

We now have a strategy for proving convergence in distribution: <sup>?</sup>

- (i) Prove that  $\varphi_n(t) \rightarrow \varphi_\infty(t)$  for some function  $\varphi_\infty$   
 (ii) Check that  $\varphi_\infty$  is char. ch. of  $O$ .

Lemma: Assume that  $\{z_n\}$  is a sequence of complex numbers converging to  $z$ . Then

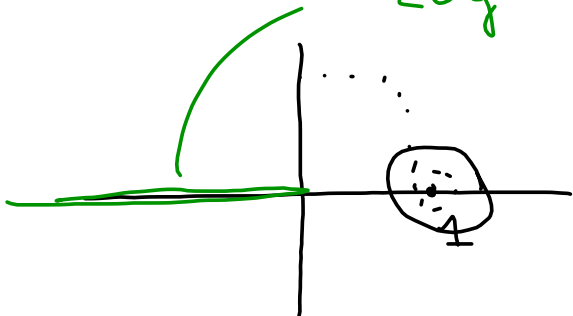
$$\lim_{n \rightarrow \infty} \left(1 + \frac{z_n}{n}\right)^n = e^z$$

Proof:  $\left(1 + \frac{z_n}{n}\right)^n \rightarrow 1$

Note that

$$\begin{aligned} \text{Log} \left(1 + \frac{z_n}{n}\right)^n &= \\ = n \text{Log} \left(1 + \frac{z_n}{n}\right) &= \frac{\text{Log} \left(1 + \frac{z_n}{n}\right)}{\frac{1}{n}} \end{aligned}$$

Log (1+x)  
 $= x - \frac{x^2}{2} + \frac{x^3}{3} - \dots$

$$= \frac{\cancel{\frac{z_n}{n}} - \frac{1}{2} \left(\frac{z_n}{n}\right)^2 + \frac{1}{3} \left(\frac{z_n}{n}\right)^3 - \dots}{\cancel{\frac{1}{n}}} \rightarrow z$$


Take exponential

$$\left(1 + \frac{z_n}{n}\right)^n = e^{n \text{Log} \left(1 + \frac{z_n}{n}\right)} \rightarrow e^z$$

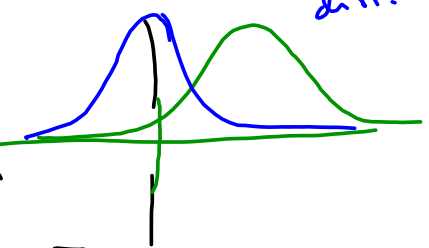
p 161-162: 6.2, 6.3, 6.4, 6.5, (6.6), (6.8) (last year)

Ex. 6.2:  $\varphi$  is the c.f. for  $X$   
 $\bar{\varphi}$  — " — — — — —  $-X$

$$\varphi_{-X}(t) = E[e^{it(-X)}] = E[\overline{e^{itX}}] = \overline{E[e^{itX}]} = \overline{\varphi(t)}$$

Ex 6.3: Show that  $X$  has a symmetric dist. if and only if  $\varphi$  is real and symmetric.   
*↳  $X$  and  $-X$  has the same dist.*

$\varphi_{-X}(t) = \varphi_{-X}(t) = \varphi_X(t)$ , hence  
*same dist.*  $\varphi_X(t) \in \mathbb{R}$



$$\varphi(-t) = E[e^{i(-t)X}] = E[e^{it(-X)}] = \varphi_{-X}(t) = \varphi_X(t)$$

*same dist.*

6.5: Show that if  $\varphi$  is a c.f., so is  $|\varphi|^2$  is a c.f.

$$|\varphi|^2 = \varphi \bar{\varphi}$$

Thinking loud:  $\varphi = \varphi_X$   
 $\bar{\varphi} = \varphi_{-X}$

Pick two independent v.v., both with the dist. of  $X$ :  $X, Y$

$$\begin{aligned} \text{Then } \varphi_{X-Y}(t) &= E[e^{it(X-Y)}] = E[e^{itX} \cdot e^{-itY}] \\ &= E[e^{itX}] E[e^{-itY}] = \varphi_X(t) \varphi_{-X}(t) \\ &= \varphi(t) \bar{\varphi}(t) = |\varphi(t)|^2 \end{aligned}$$

6.6 a)  $X_1, X_2, \dots, X_n$  symmetric <sup>dist.</sup> and independent v.v. Show that  $S = X_1 + X_2 + \dots + X_n$  has sym. dist.

Recall that a dist. is symmetric if and only if  $\varphi$  is real and symmetric.

$$\varphi_S(t) = \varphi_{X_1}(t) \varphi_{X_2}(t) \dots \varphi_{X_n}(t)$$

If all the  $X_i$ 's are symmetric, then all the  $\varphi$ 's are real and symmetric, then  $\varphi_S$  is real and symmetric.

Counterexample when the  $X$ 's aren't independent.

$$\Omega = \{\omega_1, \omega_2, \omega_3\}, P\{\omega_i\} = 1/3$$

$$X_1(\omega_1) = 1, X_1(\omega_2) = -1, X_1(\omega_3) = 0$$

$$X_2(\omega_1) = 1, X_2(\omega_2) = 0, X_2(\omega_3) = -1$$

$$\left. \begin{aligned} P[X_1 + X_2 = 2] &= 1/3 \\ P[X_1 + X_2 = -2] &= 0 \end{aligned} \right\} \text{ not symmetric.}$$

b)  $X_1, X_2, \dots$  i.i.d and  $X_i \neq 0$ .

Show that with prob 1

$$\left. \begin{aligned} \limsup S_n &= \infty \\ \liminf S_n &= -\infty \end{aligned} \right\} \text{ tail events.}$$

Assume that this is the case and let

$$\hat{S}_n = S_n + \varepsilon \xi \quad (\text{where } \xi \text{ is } N(0,1) \text{ independent of the } X_i\text{'s.})$$

Then need to be a fixed interval

(a,b) such that for all n

$$\varepsilon < P[\hat{S}_n \in (a,b)] = \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi_{S_n}(t) e^{-\frac{\varepsilon^2 t^2}{2}} \frac{e^{-itb} - e^{-ita}}{-2it} dt$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi_{X_1}(t)^n e^{-\frac{\varepsilon^2 t^2}{2}} \frac{e^{-itb} - e^{-ita}}{-2it} dt \xrightarrow{LDC} 0$$

a.s ↓  
0  
except when  
 $|\varphi_{X_1}(t)| = 1$ .

$$\varphi(t) = E[e^{itX}]$$

