

Helly's selection theorem:

If  $F_1, F_2, \dots$  are CDFs then there is a subsequence  $n_\varepsilon$  and a nondecreasing right-continuous function  $F$  so that  $F_{n_\varepsilon}(x) \rightarrow F(x)$  as  $\varepsilon \rightarrow 0$ , at all continuity points of  $F$ .

$$F_{n_\varepsilon} \Rightarrow_\sigma F \quad \text{vague convergence}$$

---

Def: We say that  $F_1, F_2, \dots$  is tight if for any  $\varepsilon > 0$  we can find  $M > 0$  so that  $P(|X_n| > M) < \varepsilon$

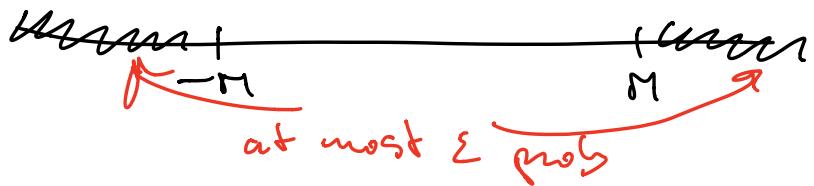
$$\text{or } F_n(-M) - 1 - F_n(M) < \varepsilon$$

---

Then: the sequence  $F_1, F_2, \dots$  is tight if and only if any subsequential vague limit is limitindistinguish.

Proof:  $\implies$  Suppose that  $F_1, F_2, \dots$  is tight. Suppose that  $F$  is a vague limit of a subsequence. Because  $F_1, F_2, \dots$  is tight, for  $\varepsilon > 0$  we can find  $M$  so that

$$F_n(-M) + 1 - F_n(M) < \varepsilon$$



Choose  $s < -M$ ,  $r > M$  so that  $s$  and  $r$  are continuity points of  $F$ .

At these values  $F_{n_\varepsilon}$  will converge to  $F$  and hence  $F(s) < 1 - F(r) \leq \varepsilon$ .



$$\rightarrow \lim_{n \rightarrow \infty} F(s) \leq \varepsilon \quad 1 - \varepsilon \leq F(r) \leq 1$$

We can do that for any  $\varepsilon > 0 \rightsquigarrow$

$$\lim_{x \rightarrow -\infty} F(x) = 0 \quad \lim_{x \rightarrow \infty} F(x) = 1.$$

 Try to do it on your own!

How can we check tightness?

Moment bounds + Markov

E.g. If  $E|X_n| < C$  for all  $n \geq 1$

$$P(|X_1| > m) \leq \frac{c}{m} \rightsquigarrow X_1, X_2, \dots \text{ tight}$$

Convergence in distribution can be characterised using the following metrics

Levy metric:

$$\delta(F, G) = \inf \left\{ \varepsilon > 0 : F(x-\varepsilon) - \varepsilon \leq G(x) \leq F(x+\varepsilon) \right\} \text{ for all } x \in \mathbb{R}$$

Ky-Fan metric

$$d(F, G) = \inf \left\{ \varepsilon > 0 : P(|X - Y| > \varepsilon) \leq \varepsilon \text{ with } X \sim F, Y \sim G \right\}$$

### Characteristic Function

Def.: The characteristic function of a r.v.  $X$  is the function  $\varphi(t) = E \left[ e^{itX} \right] = E \left[ \cos(t \cdot X) \right] + i \cdot E \left[ \sin(t \cdot X) \right]$ .

Claim: This is always well-defined.

$$\mathcal{L}_X(f) = \int_{\mathbb{R}} e^{itx} dQ_X^{(x)}$$

Fourier transform  
of  $Q_X$

$$M_X(t) = E[e^{tX}] \quad \text{could be } \infty$$

exponential moment generating function of  $X$

Thm.:  $\mathcal{L}_X$  identifies  $Q_X$

$$\begin{aligned} \lim_{T \rightarrow \infty} \frac{1}{2\pi} \int_{-T}^T \frac{e^{-ita} - e^{-itb}}{it} \mathcal{L}_X(t) dt &= \\ &= P(X \in (a, b)) + \frac{1}{2} P(X=a) + \frac{1}{2} P(X=b) \\ &= \frac{1}{2} (P(X \in (a, b)) + P(X \in [a, b])) \end{aligned}$$

Proof:

$$\left| \frac{e^{-ita} - e^{-itb}}{it} \right| = \left| \int_a^b e^{-ity} dy \right| \leq |b-a|$$

$$\frac{1}{2\pi} \int_{-T}^T \left| \frac{e^{-ita} - e^{-itb}}{it} \right| \int_{\mathbb{R}} e^{itx} Q_X(dx) dt \xrightarrow{T \rightarrow \infty} \int_{\mathbb{R}} e^{-ity} dy = 1$$

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} \left( \int_{-T}^{T} \frac{e^{it(x-a)} - e^{it(x-b)}}{it} dt \right) Q_x(dx)$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \left( \left( \int_{-T}^{T} \frac{\sin(t(x-a))}{t} dt - \int_{-T}^{T} \frac{\sin(t(x-b))}{t} dt \right) \right) Q_x(dx)$$

$$R(\vartheta, A) = \int_{-T}^{T} \frac{\sin(\vartheta t)}{t} dt = 2 \int_0^{T|\vartheta|} \frac{\sin(x)}{x} dx$$

$$= 2 \cdot \text{sgn}(\vartheta) \int_0^{\pi/2} \frac{\sin x}{x} dx$$

As  $T \rightarrow \infty$   $\int_0^{\pi/2} \frac{\sin x}{x} dx \rightarrow \frac{\pi}{2}$

(with complex analysis:  $\ln \frac{e^{iz}-1}{z}$ )

$$R(\vartheta, A) \rightarrow \frac{\pi}{2} \text{sgn}(\vartheta)$$

$$\int_{-T}^{T} \frac{\sin(t(x-a))}{t} dt - \int_{-T}^{T} \frac{\sin(t(x-b))}{t} dt$$

as  $T \rightarrow \infty$  this will converge to

$$\begin{cases} \frac{2\pi}{\pi} & a < x < b \\ 0 & x=a \text{ or } x=b \\ 0 & x < a \text{ or } x > b \end{cases}$$

$$\sup_{\gamma} \left| \int_0^{\gamma} \frac{\sin(x)}{x} dx \right| < \infty$$

By the bounded convergence theorem  
we can take the limit inside the  
integral

$$\lim_{T \rightarrow \infty} \frac{1}{2\pi} \int_{\mathbb{R}} \left( \int_{-T}^T \frac{\sin(t(x-a))}{t} dt - \int_{-T}^T \frac{\sin(t(x-b))}{t} dt \right) \psi_x(t) dt$$

$$= P(X \in (a, b)) + \frac{1}{2}(P(X=a) + P(X=b))$$

Then : If  $\int_{\mathbb{R}} |\psi_x(t)| dt < \infty$  then

$X$  has a PDF

$$f_X(y) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{-ity} \psi_x(t) dt$$

Proof :

$$P(X \in (a, b)) + \frac{1}{2}(P(X=a) + P(X=b)) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{ict_a} - e^{ict_b}}{it} \psi_x(t) dt$$

$$\leq \frac{b-a}{2\pi} \int_{\mathbb{R}} |\psi| dt$$

This shows that  $P(X=a)=0$  for all  $a \in \mathbb{R}$ .

$$\begin{aligned} P(X \in (a, b)) &= \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \int_a^b e^{-ity} \psi(f) dt dy \end{aligned}$$

$$= \int_a^b \underbrace{\frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-ity} \psi(f) dt}_{\text{this is the PDF of } X} dy$$

*This is the PDF of  $X$*  □

If  $\psi_X(t) = \psi_Y(t)$  for all  $t \in \mathbb{R}$

then  $X \stackrel{d}{=} Y$ .

Ex: If  $\psi_X(t) \in \mathbb{R}$  for all  $t$

then  $X \stackrel{d}{=} -X$ .

$$\psi_{-X}(t) = E[e^{-it(-X)}] = E[\cos(tx)] - i E[\sin(tx)]$$

$$\varphi_X(t) =$$

$$E[e^{itX}] \cancel{= E[\cos(X)]}$$

If  $X, Y$  are independent then

$$\begin{aligned} \varphi_{X+Y}(t) &= E[e^{it(X+Y)}] = E[e^{itX} \cdot e^{itY}] \\ &\stackrel{\text{independence}}{=} E[e^{itX}] E[e^{itY}] \\ &= \boxed{\varphi_X(t) \varphi_Y(t)} \end{aligned}$$

In the CLT we want to understand the limit distribution of  $\frac{S_n - n\mu}{\sqrt{n}\sigma}$ .

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

$$\begin{aligned} \varphi_{aY+b}(t) &= E[e^{(aY+b)it}] = E[e^{ibt} \cdot e^{iatY}] \\ &= e^{ibt} E[e^{iatY}] = e^{ibt} \varphi_Y(at) \end{aligned}$$

$$\varphi_{S_n}(t) = \varphi_{X_1 + \dots + X_n}(t) = \varphi_{X_1}(t) \cdot \dots \cdot \varphi_{X_n}(t) = (\varphi_{X_1}(t))^n$$

$$\varphi_{\frac{S_n - \mu}{\sqrt{n\sigma^2}}}(t) = e^{-\frac{\mu t}{\sqrt{n\sigma^2}}} \left[ \varphi_{X_1}\left(\frac{t}{\sqrt{n\sigma^2}}\right) \right]^n$$

If  $X \sim N(0, 1)$  then  $\varphi_X(t) = e^{-\frac{t^2}{2}}$

$$E\left(e^{itX}\right) = \int_{\mathbb{R}} e^{itx} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = e^{-\frac{t^2}{2}}$$

$$\frac{1}{\sqrt{2\pi}} e^{itx - \frac{x^2}{2}} = \frac{1}{\sqrt{2\pi}} e^{+\frac{t^2}{2} + itx - \frac{x^2}{2} - \frac{t^2}{2}} \\ \cong \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-it)^2} \cdot e^{-\frac{t^2}{2}}$$

$$\int \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-it)^2} dx = 1$$

Numerous proof: complex analysis (residue theorem)