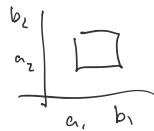


Last time  
inversion formula.



$$\underline{x} \in \mathbb{R}^d$$

$$\varphi_{\underline{x}}(t) = E e^{it(\underline{x}-\underline{u})}$$

$$A = \bigcup_{i=1}^A [a_i, b_i]$$

$$P(x \in A) = \lim_{T \rightarrow \infty} (2\pi)^{-d} \lambda(A) \cdot \int_{[-T, T]^d} \varphi_u(t) \varphi_x(t) dt.$$

Lebesgue measure aka Volume.  
 $\lambda$

$U \sim \text{uniform}(A)$ .

move - from  $U$  to  $t$ . Because  $\varphi_U(t) = E e^{itU}$

$$t \in \mathbb{R}^d$$

$$t \underline{x} = \sum_{i=1}^d t_i x_i$$

Inversion formula

$$\underline{Y} = \underline{X} - \underline{U}, \quad U \sim \text{uniform}(A), \quad (X, U) \text{ i.i.d.}$$

$$\varphi_Y(t) = E(e^{it \cdot (\underline{x}-\underline{u})}) = E(e^{itx} \cdot e^{i(-t)U})$$

$$= \varphi_x(t) \cdot \varphi_u(-t).$$

$$f_y(0) = \frac{\varphi(x \in A)}{\lambda(A)}$$

let  $Y$  be Random Vector. with "nice" density and

$\varphi_Y(t)$  is the ch func.

then  ~~$\lim_{T \rightarrow \infty} (2\pi)^{-d} \lambda(A) \int_{[-T, T]^d} \varphi_Y(t) dt = f_Y(0)$~~

↗  
if we use Fubini  
mass accumulates at 0.

$$f_u(v) = \frac{1}{\lambda(A)}, \quad v \in A$$

$$v = \begin{pmatrix} v_1 \\ \vdots \\ v_d \end{pmatrix} \quad U_k \sim \text{uniform}(a_k, b_k)$$

$$\underline{U} = \left( \underline{U}_k \right)_{1 \leq k \leq d}$$

$$\varphi_U(t) = \prod_{k=1}^d \varphi_{U_k}(t_k)$$

$$t = \begin{pmatrix} t_1 \\ \vdots \\ t_d \end{pmatrix}$$

$$\varphi_U(t) = \prod_{k=1}^d \varphi_{U_k}(t_k) = \prod_{k=1}^d \int_{a_k}^{b_k} e^{it_k y} \cdot \frac{1}{b_k - a_k} dy = \frac{e^{\sum t_k y}}{i \sum t_k (b_k - a_k)}$$

$$t = \begin{pmatrix} t_1 \\ \vdots \\ t_d \end{pmatrix} \quad \int_{y=a_k}^{i t_k y} e^{i t_k y} \cdot \frac{1}{b_k - a_k} dy = \left. \frac{e^{i t_k y}}{i t_k (b_k - a_k)} \right|_{a_k}$$

two r.v. have same c.f. hence same distribution

Application:  $\bar{X}_1, \dots, \bar{X}_d$  are Ind.

$$\text{iff } \Psi_{\bar{X}}(t) = \prod_{k=1}^d \Psi_{X_k}(t_k)$$

$$\bar{X} = \begin{pmatrix} \bar{X}_1 \\ \vdots \\ \bar{X}_d \end{pmatrix}, \quad t = \begin{pmatrix} t_1 \\ \vdots \\ t_d \end{pmatrix}$$

how do we know  $\Psi_{\bar{X}}(t) = \prod_{k=1}^d \Psi_{X_k}(t_k) \rightarrow \prod X_k$ ?

Because  $\Psi$  is c.f.

take  $Y_1, \dots, Y_d$  and  $Y_k \stackrel{\text{def}}{=} X_k, \quad k=1, \dots, d$ .

and  $Y_1, \dots, Y_d$  are Ind.

$$\Psi_Y(t) = \prod_{k=1}^d \Psi_{Y_k}(t_k) = \prod_{k=1}^d \Psi_{X_k}(t_k) = \Psi_{\bar{X}}(t).$$

$$Y = \begin{pmatrix} Y_1 \\ \vdots \\ Y_d \end{pmatrix}$$

Recall  $\bar{X}, Y$  are r.v.  $\in \mathbb{R}^d$

$$\Psi_{\bar{X}+Y}(t) = \Psi_{\bar{X}}(t)\Psi_Y(t). \quad \forall t \in \mathbb{R}^d$$

$\bar{X}, Y$  are not ind.

		1	2	3	
		1	2	3	
Dist of $X$	1	$\frac{1}{3}$	0	$\frac{2}{3}$	$\frac{1}{3}$
	2	$\frac{1}{3}$	$\frac{1}{3}$	0	$\frac{1}{3}$
Dist of $Y$	3	0	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$
		$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	

not Ind.

what does it mean to have vectors to converge in distribution?  
eg weak convergence  $\bar{X}_n \Rightarrow \bar{X}$

Def  $X_n \in \mathbb{R}^d, n=1, 2, \dots$

$$X_n \Rightarrow X_\infty \text{ if } E\delta(X_n) \rightarrow E\delta(X), \quad \forall f \in C_b(\mathbb{R}^d)$$

Theorem The following are equivalent TFAE

$$(1) \quad X_n \Rightarrow X_\infty$$

$$(2) \quad \lim_{n \rightarrow \infty} P(\bar{X}_n \in K) \leq P(X \in K), \quad K \text{ is closed. } \subset \mathbb{R}^d$$

$$(3) \quad \lim_{n \rightarrow \infty} P(\bar{X}_n \in O) \geq P(X \in O), \quad O \text{ is open. } \subset \mathbb{R}^d$$

use for interior A.

boundary  
↓

(3)  $\lim_{n \rightarrow \infty} P(X_n \in O) \geq P(X \in O)$ ,  $O$  is open.  $\subset$  in  
use for interior  $A$ .

(4)  $P(X_n \in A) \rightarrow P(X_\infty \in A) \quad \forall A \in \mathcal{B}(\mathbb{R}^d)$  and  $P(X_\infty \in \bar{A}) = 0$

Boundary  
↓

$A^\circ$  the interior set,  $\bar{A}$  closed set.

$A^\circ \subseteq A \subseteq \bar{A}$

$\bar{A} = \bar{A} \setminus A^\circ$

$$\therefore P(X_\infty \in A^\circ) = P(X_\infty \in \bar{A})$$

how to Prove 1  $\Rightarrow$  2  $\&$  3. Prob. space

in 1 dim we used let  $X_n \Rightarrow X_\infty$  then  $\exists Y_n, Y_\infty$  so that  $Y_n \xrightarrow{n \rightarrow \infty} Y_\infty$   $Y_n \stackrel{D}{=} X_n$ ,  $n \geq 1$   
got  $Y$  from  $X$  used so call quantile function.  
 $Y_n = F_{X_n}^{-1}(U)$  where  $U \sim \text{uniform}(0,1)$

$$Y_\infty \stackrel{D}{=} X_\infty$$

in ONE Dime.

$$F_Z(x) \rightarrow F_{X_\infty}^{(x)}, P(X_\infty = x) = 0.$$

All most All are zero except countable

Can't do it with  $V = 2^+$

solved by skorokhod

if  $X \sim Y$  is dist then And  $X \stackrel{\text{close.}}{\sim} Y$  a.s.

$$X \sim Y \quad | \quad (x, y) \quad x \stackrel{D}{=} x, y \stackrel{D}{=} y$$

$x = y$ :  $\therefore$  point should be on the diagonal.