

Lecture 12:

Generating functions, moments & separating classes

(1)

We will see that when working with probability measures over a complicated space (Ω, \mathcal{F}) it will be useful to be able to characterize a probability P on (Ω, \mathcal{F}) by analyzing the value of $\int f dP$ computed over a range of test functions $f: \Omega \rightarrow \mathbb{R}$.

Characteristic functions and moment generating functions are an example of this.

MGF's, CF's and Complex generating functions

Definition:

Let (Ω, \mathcal{F}, P) be a probability space and X be a random variable (taking values in \mathbb{R}).

For $t \in \mathbb{R}$ and $z \in \mathbb{C}$ define

$$\begin{aligned} M_X(t) &:= E(e^{tX}) && \leftarrow \text{Moment generating function of } X \text{ (MGF).} \\ G_X(z) &:= E(e^{zX}) && \leftarrow \text{Complex generating function of } X. \\ \phi_X(t) &:= E(e^{itX}) && \leftarrow \text{Characteristic function for } X \text{ (CF).} \end{aligned}$$

In general, if μ is a measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ define

$$M_\mu(t) := \int_{\mathbb{R}} e^{tx} d\mu(x)$$

$$G_\mu(z) := \int_{\mathbb{R}} e^{zx} d\mu(x).$$

$$\phi_\mu(t) := \int_{\mathbb{R}} e^{itx} d\mu(x)$$

Since $x \mapsto e^{zx}$ takes values in \mathbb{C} we need (2) to say a word about integrating complex valued functions.

If $f: \Omega \rightarrow \mathbb{C}$ then one can decompose it into real and imaginary parts:

$$f(w) = \underbrace{\operatorname{Re} f(w)} + i \underbrace{\operatorname{Im} f(w)}_{\text{functions mapping } \mathbb{R} \rightarrow \mathbb{R}.}$$

If μ is a measure on (Ω, \mathcal{F}) then

$$\int_{\Omega} f(w) d\mu(w) := \underbrace{\int_{\Omega} \operatorname{Re} f(w) d\mu(w)} + i \underbrace{\int_{\Omega} \operatorname{Im} f(w) d\mu(w)}_{\nearrow}$$

all the properties of $\int_{\Omega} f(w) d\mu(w)$ extend to the complex case with minor changes

when these two are defined i.e.
 $\operatorname{Re}, \operatorname{Im} \in \mathcal{Q}(\Omega, \mathcal{F}, \mu)$.

The usefulness of these generating functions come from 3 facts:

1) ϕ_X & G_X (and M_X sometimes) characterizes the distribution of X .

E.g. if you have two r.v.s X & Y then $X = Y$ iff $\phi_X(t) = \phi_Y(t) \quad \forall t \in \mathbb{R}$.

Note: This is analogous to c.d.f.s and densities.

2) The generating functions for sums of independent r.v.s is easy to calculate. i.e. If X_1, \dots, X_n are independent r.v.s all defined on (Ω, \mathcal{F}, P) then

$$\phi_{X_1 + \dots + X_n}(t) = \phi_{X_1}(t) \cdots \phi_{X_n}(t).$$

Note: The corresponding operation for densities is hard, i.e. the density of $X_1 + \dots + X_n$ is a n -fold convolution of the densities of each X_i .

3) If you know $M_X(t)$, $\phi_X(t)$ or $G_X(z)$ you can compute the moments $E(X^k)$ by differentiating.

Note: if μ is a measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ s.t. (3)

$$d\mu = s dx$$

then $\phi_\mu(t)$ is just the Fourier transform of s . It is actually more natural to think of $\phi_\mu(t)$ as inverse Fourier transform of $s(x)$ where x represents frequency.

This explains fact 3) since the FT diagonalizes convolution.

Relating ϕ_x , M_x and G_x

There are two examples of generating functions that are useful to keep in your mind.

Let $Z \sim N(0, 1)$. Then

$$\phi_z(t) = e^{-t^2/2}$$

$$M_z(t) = e^{t^2/2}$$

$$G_z(z) = e^{z^2/2}$$

Also let Y be a r.v. with density $s(x) = \frac{1}{\pi(1+x^2)}$ w.r.t. dx . Then

$$\phi_y(t) = e^{-|t|}$$

$$M_y(t) = \begin{cases} 1 & \text{if } t=0 \\ \infty & \text{o.w.} \end{cases}$$

$$G_y(z) = \begin{cases} e^{-|z|} & \text{if } \operatorname{Re} z = 0 \\ \infty & \text{if } \operatorname{Re} z \neq 0 \text{ & } \operatorname{Im} z = 0 \\ \text{Not defined} & \text{o.w.} \end{cases}$$

Note: Y is a Cauchy r.v..

Looking at the case of $Z \sim N(0, 1)$ we have (4)
 $M_X(it) = \phi_X(t)$. However this can't hold in general since $M_Y(it) \neq \phi_Y(t)$. To understand the difference we need to analyze G_X .

Definition: If \mathcal{S} is a metric space, with metric d , and $A \subset \mathcal{S}$ define
 $A^\circ :=$ the open interior of A ← union of all open sets $C \subset A$
 $\bar{A} :=$ the closure of A ← intersection of all closed sets containing A
 $\partial A := \bar{A} - A^\circ$

$$d(x, A) := \inf \{ d(x, y) : y \in A \}$$



Also for any subset of \mathbb{C} let

$$\operatorname{Re} A := \{ \operatorname{Re} z : z \in A \}$$

$$\operatorname{Im} A := \{ \operatorname{Im} z : z \in A \}$$

Definition:

For any r.v. X let

$$\mathcal{D}_X := \{ u + iz \in \mathbb{C} : E(e^{uz}) < \infty \}$$

= the cylinder in \mathbb{C} with base $\{u \in \mathbb{R} : M_X(u) < \infty\}$

Theorem ($\operatorname{Re} \mathcal{D}_X$ is an interval)

If X is a r.v. then $\operatorname{Re} \mathcal{D}_X$ is an interval containing 0 (closed, open or half open) and M_X is convex on $\operatorname{Re} \mathcal{D}_X$.

Remark: This thm is true for $\operatorname{Re} \mathcal{D}_\mu$ & M_μ when μ is a measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ but now the interval can be empty (in which case it will not contain 0).

Proof:

Since $t \mapsto e^{tx}$ is convex we have

$$e^{[\alpha t_1 + (1-\alpha)t_2]x} \leq \alpha e^{t_1 x} + (1-\alpha)e^{t_2 x}$$

$\forall t_1, t_2 \in \mathbb{R}$ & $\alpha \in [0, 1]$.

$$\begin{aligned} \therefore M_X(\alpha t_1 + (1-\alpha)t_2) &= E(e^{[\alpha t_1 + (1-\alpha)t_2]X}) \\ &\stackrel{\text{By } 3}{\leq} \alpha E(e^{t_1 X}) + (1-\alpha)E(e^{t_2 X}) \\ &= \alpha M(t_1) + (1-\alpha)M(t_2) \end{aligned}$$

$\therefore M_X$ is convex.

Now suppose $t_1, t_2 \in \text{Re } D_X$. Then $M_X(t_1) < \infty$,

$M_X(t_2) < \infty$ and

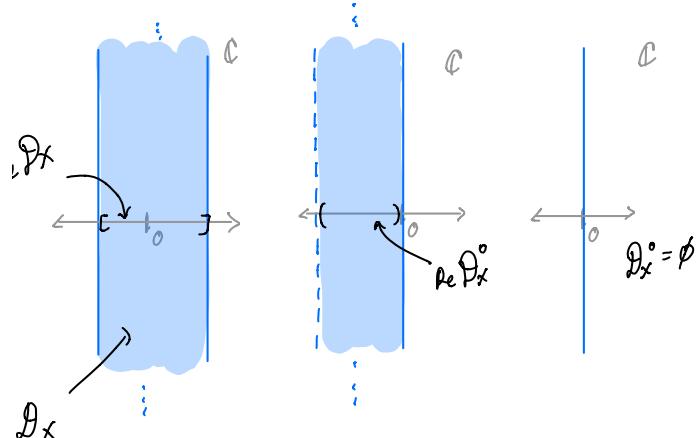
$$\begin{aligned} t_1 \leq t \leq t_2 \Rightarrow M_X(t) &= \alpha M_X(t_1) + (1-\alpha)M_X(t_2) < \infty \\ &\quad \text{Writing } t = \alpha t_1 + (1-\alpha)t_2 \text{ for some } \alpha \in [0, 1] \\ \Rightarrow t &\in \text{Re } D_X. \end{aligned}$$

$\therefore \text{Re } D_X$ is an interval containing 0,

since clearly $M_X(0) = 1$.

QED

so for any r.v. X D_X could look something like this:



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Theorem: (The Analyticity of G_X over D_X^o)

Let X be a r.v. (mapping into \mathbb{R}) such that $D_X^o \neq \emptyset$. Then $\forall z \in D_X^o$

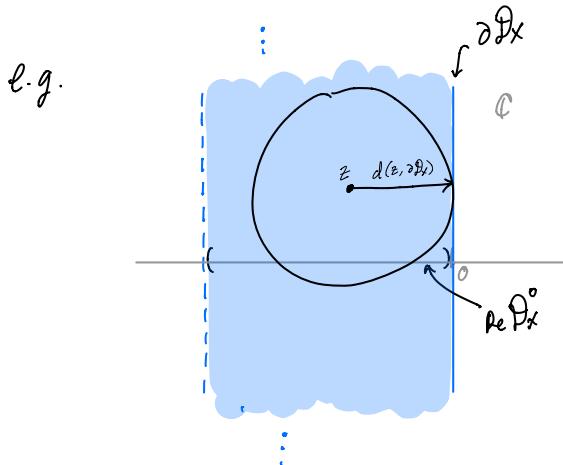
i) $E|X^n e^{zx}| < \infty$ for $n = 1, 2, \dots$

$$\text{ii) } G_X(z) = \sum_{n=0}^{\infty} E(X^n e^{zx}) \frac{(z-z)^n}{n!}$$

for all z in the open ball of \mathbb{C} centered at z with radius $d(z, \partial D_X)$.

iii) G_X is infinitely differentiable on D_X^o with complex derivative

$$\frac{d^n}{dz^n} G_X(z) = E(X^n e^{zx}).$$



Proof:

First note that for any $z, \zeta \in \mathbb{C}$ we have

$$e^{\zeta X} = e^{zX} e^{(\zeta-z)X} = \sum_{n=0}^{\infty} e^{zX} \frac{X^n (\zeta-z)^n}{n!}$$

and

$$\begin{aligned} (*) \quad \sum_{n=0}^{\infty} \left| e^{zX} \frac{X^n (\zeta-z)^n}{n!} \right| &\leq |e^{zX}| \sum_{n=0}^{\infty} \frac{|X(\zeta-z)|^n}{n!} \\ &= e^{uX} e^{r|X|} \\ &\quad \text{where } u := \text{Re } (\zeta) \\ &\quad r := |\zeta - z| \\ &\leq e^{(u-r)X} + e^{(u+r)X} \\ &\quad \text{since } e^{-r|X|} = \begin{cases} e^{-rX} & \text{if } X \in \mathbb{R} \\ e^{rX} & \text{if } X \notin \mathbb{R} \end{cases} \end{aligned}$$

Notice that when $z, \zeta \in \mathcal{D}_X^0$ and
 $r := |z - \zeta| < d(z, \partial \mathcal{D}_X^0)$ then

$$u \pm r := \operatorname{Re} z \pm |z - \zeta| \in \operatorname{Re} \mathcal{D}_X^0$$

so that

$$E(e^{(u-r)x}) = M_x(u-r) < \infty$$

$$E(e^{(u+r)x}) = M_x(u+r) < \infty$$

$$\therefore E\left(\sum_{n=0}^{\infty} \left|e^{zx} \frac{x^n (\zeta-z)^n}{n!}\right|\right) < \infty \quad (**)$$

$\parallel \leftarrow$ By monotone convergence in Big 3

$$\sum_{n=0}^{\infty} E\left|e^{zx} \frac{x^n (\zeta-z)^n}{n!}\right|$$

$$\therefore E\left|e^{zx} \frac{x^n (\zeta-z)^n}{n!}\right| = \frac{|\zeta-z|^n}{n!} E(|e^{zx} x^n|) < \infty$$

for all $n = 1, 2, \dots$ so i) holds

To show ii) notice that

$$\sum_{n=0}^N e^{zx} \frac{x^n (\zeta-z)^n}{n!} \xrightarrow{N \rightarrow \infty} \sum_{n=0}^{\infty} e^{zx} \frac{x^n (\zeta-z)^n}{n!} \quad \text{P-a.e.}$$

and DCT applies with upper bound given

by the LHS of (*) which is integrable by (**).

$$\therefore G_x(\zeta) = E(e^{\zeta x})$$

$$= E\left(\lim_N \sum_{n=0}^N e^{zx} \frac{x^n (\zeta-z)^n}{n!}\right)$$

$$\stackrel{\text{DCT}}{=} \lim_N \sum_{n=0}^N E\left(e^{zx} x^n\right) \frac{(\zeta-z)^n}{n!}$$

$$= \sum_{n=0}^{\infty} E\left(e^{zx} x^n\right) \frac{(\zeta-z)^n}{n!}$$

This gives ii).

Finally iii) follows directly from ii).

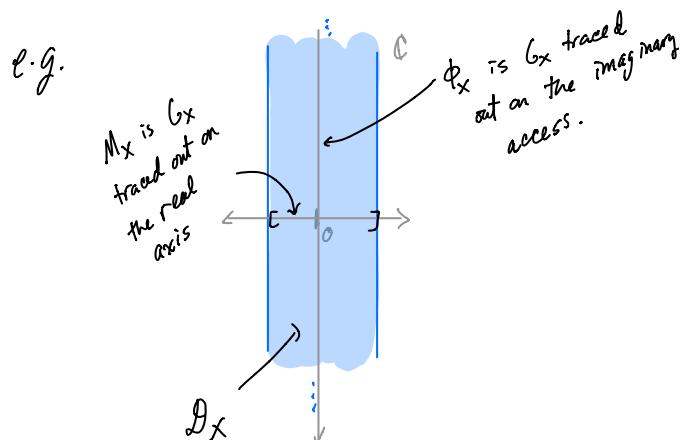
QED

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Now we can understand the relationship
btwn ϕ_x and M_X :

- $G_x(z)$ always exist and is finite on \mathcal{D}_X
since $u+iv \in \mathcal{D}_X \Rightarrow M_X(u) = E(e^{ux}) < \infty$
 $\Rightarrow E(|e^{(u+iv)x}|) = E(e^{ux}) < \infty$
 $\Rightarrow |G_x(u+iv)| < \infty$.

- \mathcal{D}_X always contains $\{it : t \in \mathbb{R}\}$
since $0 \in \operatorname{Re} \mathcal{D}_X$
- $\phi_x(t) = G_x(it) \quad \forall t \in \mathbb{R}$ and
 $M_X(it) = G_x(t) \quad \forall t \in \operatorname{Re} \mathcal{D}_X$



- $G_x(z) =$ the unique analytic extension
of ϕ_x on $\{it : t \in \mathbb{R}\}$ to \mathcal{D}_X .
only when $\operatorname{Re} \mathcal{D}_X \neq \{0\}$ \Downarrow = the unique analytic extension
of M_X on $\operatorname{Re} \mathcal{D}_X$ to \mathcal{D}_X

This follows by a complex analysis result:

Thm: Suppose $D \subset \mathbb{C}$ is open and connected.
if f and g are differentiable complex-valued
functions defined on D which agree on
distinct $z_1, z_2, \dots \in D$ s.t. $\lim_{n \rightarrow \infty} z_n \in D$ then

$$f(z) = g(z) \quad \forall z \in D.$$

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- Now suppose we have a formula for ⑨
 $M_X(t)$ s.t. $\operatorname{Re} \mathcal{D}_X \neq \emptyset$ and a extension $H(z)$ defined on \mathcal{D}_X s.t.

$$H(t) = M_X(t) \quad \forall t \in \operatorname{Re} \mathcal{D}_X.$$

Then if H is complex differentiable it must be that

$$H(z) = G_X(z), \quad \forall z \in \mathcal{D}_X \text{ and}$$

$$H(it) = \phi_X(t), \quad \forall t \in \mathbb{R}.$$

- This also works if you can compute

$$\alpha_n := E|X|^n$$

when $\beta_n := E|X|^n$ decay fast enough so $\sum_{n=0}^{\infty} \beta_n \frac{t^n}{n!}$ has a non-zero radius of convergence. In which case

$$M_X(t) = \sum_{n=0}^{\infty} \alpha_n \frac{t^n}{n!}$$

for all t in an open neighborhood of 0 (use similar arguments for the thm on G_X).

This completely determines G_X , and thus ϕ_X , by analytic extension to \mathcal{D}_X .

Note: once we show that $\phi_X(t)$ completely characterizes the distribution of X we will have:

The moments $\{E|X|^n\}_{n \geq 1}$ characterize the distribution of X only when $\sum_{n=0}^{\infty} E|X|^n \frac{t^n}{n!}$ has a non-zero radius of convergence.

e.g. If $X \sim N(0, 1)$ then one can derive that $M_X(t) = e^{t^2/2}$ & $\operatorname{Re} \mathcal{D}_X = \mathbb{R}$. Here are two extensions defined on $\mathcal{D}_X = \mathbb{C}$:

$$H_1(z) = e^{z^2/2} \leftarrow \text{not analytic}$$

$$H_2(z) = e^{z^2/2} \leftarrow \text{analytic}$$

$\therefore G_X(z) = H_2(z)$ but not $H_1(z)$ and

$$\phi_X(t) = H_2(it) = e^{-t^2/2} \text{ but not } H_1(it) = e^{t^2/2}$$

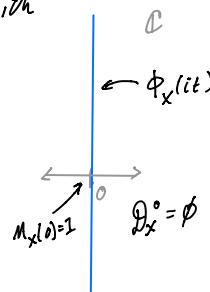
e.g. if Y is a Cauchy R.V.

then $\operatorname{Re} \mathcal{D}_X^\circ = \emptyset$ which

explains why we can't get

ϕ_X or G_X from M_X .

$$\begin{aligned} \phi_X(it) &= G_X(it) \\ &= e^{-|it|} \end{aligned}$$



e.g. suppose $X \geq 0$ which satisfies

$$EX^n = E|X|^n = n!$$

Can we infer what ϕ_X is?

$$\text{since } \sum_{n=0}^{\infty} n! \frac{t^n}{n!} < \infty \quad \forall t \in (-1, 1),$$

$M_X(t)$ is finite on $(-1, 1) = \operatorname{Re} \mathcal{D}_X$

$$\begin{aligned} \therefore M_X(t) &= G_X(t) = \sum_{n=0}^{\infty} E(X^n) \frac{t^n}{n!} \\ &= \sum_{n=0}^{\infty} t^n = \frac{1}{1-t} \end{aligned}$$

when $t \in \operatorname{Re} \mathcal{D}_X = (-1, 1)$.

Since $\frac{1}{1-t}$ is an analytic extension of M_X to \mathcal{D}_X we must have

$$\phi_X(t) = G_X(it) = \frac{1}{1-it}$$

This r.v. is an exponential r.v. where

$$dP_X^{-1} = e^{-x} I_{[0, \infty)}(x) dx$$

Another consequence of our analyticity theorem for G_X is that both ϕ_X and M_X give the moments of X . (11)

Since ϕ_X is always defined and not every r.v.s has finite moments of all orders it suggests that

The larger n is s.t. $E|X|^n < \infty$ \iff $\phi_X(t)$ is at $t=0$

Corollary: (Moments from M_X or ϕ_X)

If X is a r.v. s.t. $0 \in \text{Re } \phi_X^0$ then $E|X|^n < \infty$ for all n and

$$M_X^{(n)}(0) = (-i)^n \phi_X^{(n)}(0) = E(X^n).$$

Proof:

Note that $G_X(t) = M_X(t)$ when $t \in \mathbb{R}$ and the complex derivatives of an analytic function equal the "directional derivatives".

$$\therefore \frac{d^n}{dt^n} M_X(t) \Big|_{t=0} = \frac{d^n}{dz^n} G_X(z) \Big|_{z=0} = E(X^n).$$

A similar argument holds for ϕ_X . \square

Getting Moments from ϕ_X (12)

What about the case when $0 \notin \text{Re } \phi_X^0$? Now either $E|X|^n$ doesn't decay fast enough to be summable, or they are ∞ for all large n .

We need a more fine tuned argument.

Studying Taylor's thm one gets

$$(*) \quad \left| e^{itx} - \sum_{n=0}^N \frac{(itx)^n}{n!} \right| \leq \frac{|tx|^{N+1}}{(N+1)!}$$

$\forall t \in \mathbb{R}$ & $N \geq 0$.

Note: this bound gives a slightly sub-optimal bound on the regularity of $\phi_X(t)$ near $t=0$ when $E|X|^{N+1} < \infty$. See Billingsley p. 343 for the more fine tuned result.

Now (*) already gives

$$\left| \phi_X(t) - \sum_{n=0}^N E(X^n) \frac{(it)^n}{n!} \right| \leq \frac{|t|^{N+1}}{(N+1)!} E(|X|^{N+1})$$

Theorem: (Moments from ϕ_X)

If X is a r.v. that satisfies $E|X|^{n+1} < \infty$ for some $n = 1, 2, \dots$ then ϕ_X is n times differentiable and

$$\phi_X^{(m)}(t) = E((iX)^m e^{itX}).$$

If $m \leq n$.

Proof: start with $m=1$.

$$\begin{aligned} \frac{\phi_X(t+\varepsilon) - \phi_X(t)}{\varepsilon} &= E\left(\frac{e^{itX} e^{i\varepsilon X} - e^{itX}}{\varepsilon}\right) \\ &= E\left(e^{itX} \frac{e^{i\varepsilon X} - 1}{\varepsilon}\right) \end{aligned}$$

Therefore

$$\frac{\phi_X(t+\varepsilon) - \phi_X(t)}{\varepsilon} - E(iX e^{itX}) = E\left(e^{itX} \frac{e^{i\varepsilon X} - 1 - i\varepsilon X}{\varepsilon}\right)$$

bdd in magnitude
by $\frac{|e^{i\varepsilon X}|^2}{\varepsilon^2}$ from (*)

$\therefore \lim_{\varepsilon \rightarrow 0} \text{RHS} = 0$ by DCT.

$$\therefore \phi'_X(t) = E(iX e^{itX}).$$

Repeating the argument gives the result.
QED.

Separating Classes

To show $\phi_X(t)$ characterizes the distribution of X notice that the values of $\phi_X(t)$ can be considered as computing $E f(X)$ over the class of test functions

$$f \in \{\sin(t \cdot) : t \in \mathbb{R}\} \cup \{\cos(t \cdot) : t \in \mathbb{R}\}$$

$$\text{Since } e^{itX} = \cos(tX) + i \sin(tX).$$

It will be useful to study this problem from a more general perspective.

Assumption For the rest of this section suppose \mathcal{S} is a metric space with metric d .

Definition:

- If \mathcal{S} is a complete and separable metric space then \mathcal{S} is called a Polish space
- $C(\mathcal{S}) := \{\text{continuous maps } f: \mathcal{S} \rightarrow \mathbb{R}\}$
- $C_b(\mathcal{S}) := \{\text{bdd and continuous maps } f: \mathcal{S} \rightarrow \mathbb{R}\}$
- $C_c(\mathcal{S}) := \{\text{compactly supported continuous maps } f: \mathcal{S} \rightarrow \mathbb{R}\}$
- $Lip_K(\mathcal{S}) := \{f: \mathcal{S} \rightarrow \mathbb{R} \text{ s.t. } |f(x) - f(y)| \leq K d(x, y) \forall x, y \in \mathcal{S}\}$
- $C^k(\mathbb{R}^d) := \{k\text{-times differentiable maps } f: \mathbb{R}^d \rightarrow \mathbb{R}\}$

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Definition:

Let \mathcal{P} be a collection of probability measures on $(\mathcal{S}, \mathcal{B}(\mathcal{S}))$ where \mathcal{S} is a metric space.

A collection of functions \mathcal{M} separates \mathcal{P} if

i) $\mathcal{M} \subset C_b(\mathcal{S})$

ii) $\int f dP = \int f dQ \quad \forall f \in \mathcal{M} \Rightarrow P = Q \text{ on } (\mathcal{S}, \mathcal{B}(\mathcal{S}))$

If \mathcal{M} separates all probability measures on $(\mathcal{S}, \mathcal{B}(\mathcal{S}))$ then we say \mathcal{M} is a separating class for $(\mathcal{S}, \mathcal{B}(\mathcal{S}))$.

Note: This extends to r.v.s taking values in \mathcal{S} .

Then ii) becomes

$$E f(X) = E f(Y) \quad \forall f \in \mathcal{M} \Rightarrow X = Y$$

Theorem: ($Lip_K(\mathcal{S})$ separates)

Suppose \mathcal{S} is a metric space with metric d . Then $Lip_1(\mathcal{S})$ is a separating class for $(\mathcal{S}, \mathcal{B}(\mathcal{S}))$.

Proof:

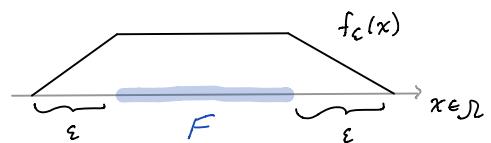
Let P and Q be two probability measures on $(\mathcal{S}, \mathcal{B}(\mathcal{S}))$ s.t.

$$\int f dP = \int f dQ \quad \forall f \in \mathcal{C}$$

Let $F \in \mathcal{F}$ be a closed set. It will be sufficient to show $P(F) = Q(F)$ by π -uniqueness over the closed sets.

For $\varepsilon > 0$ define

$$f_\varepsilon(x) := \left(1 - \frac{d(x, F)}{\varepsilon}\right)^+$$



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Notice that f_ε is bdd (clearly) and Lipschitz (15)
continuous since

$$|f_\varepsilon(x) - f_\varepsilon(y)| \leq \left| \frac{d(x, F)}{\varepsilon} - \frac{d(y, F)}{\varepsilon} \right| \stackrel{\text{since}}{\leq} \frac{|(1-\varepsilon)^+ - (1-\varepsilon)^+|}{\varepsilon} \leq \frac{|x-y|}{\varepsilon}$$

$\leq \frac{d(x, y)}{\varepsilon}$ left as an exercise.

$$\therefore f_\varepsilon \in \mathcal{C} \quad \forall \varepsilon > 0.$$

Moreover

$$(*) \quad I_F(x) \leq f_\varepsilon(x) \leq I_{F^\varepsilon}(x)$$

since $x \in F$ implies $f_\varepsilon(x) = 1$ and
 $x \notin F^\varepsilon := \{y : d(y, F) < \varepsilon\}$ implies $f_\varepsilon(x) = 0$

Now integrate over the terms in (*) to get

$$P(F) \leq \int_{\mathbb{R}} f_\varepsilon dP = \int_{\mathbb{R}} f_\varepsilon dQ \leq Q(F^\varepsilon)$$

↑ since $f_\varepsilon \in \mathcal{C}$

If F is close then $F^\varepsilon \downarrow F$ as $\varepsilon \rightarrow 0$
(since $d(x, F) = 0$ & F closed $\Rightarrow x \in F$).

$$\therefore P(F) \leq \lim_{\varepsilon \downarrow 0} Q(F^\varepsilon) = Q(F).$$

Similarly one obtains $Q(F) \leq P(F)$

$\therefore P(F) = Q(F)$ & closed $F \subset \mathbb{R}$
as was to be shown

QED.

To do:

- Stone-Weierstrass condition for spanning class.

- \mathcal{M} compact metric space
- $\mathcal{M} \subset C(\mathbb{R})$
- \mathcal{M} closed under addition & mult.
- const func & \mathbb{M}
- $\forall w_1, w_2 \in \mathbb{R}$ s.t.
 $w_1 \neq w_2$ then $\exists f \in \mathcal{M}$
s.t. $f(w_1) \neq f(w_2)$
- closed under conj

$\Rightarrow \mathcal{M}$ dense in $C_b(\mathbb{R})$
with sup norm
 $\Rightarrow \mathcal{M}$ separates

- show $C_0^\infty(\mathbb{R}^d)$ separates $(\mathbb{R}^d, B(\mathbb{R}^d))$.

will be useful for general CLT

- Show that we can get a $C_0^\infty(\mathbb{R}^d)$ function for the density of $\sum n_m z_m$.

- show $\{x \mapsto e^{i\langle k, x \rangle} : k \in \mathbb{R}^d\}$
separates $(\mathbb{R}^d, B(\mathbb{R}^d))$.

- conclude characteristic func
(exended to vectors) characterizes X .

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