

Ch-3

Random Variable and Expectation

Random Variable

Expectation

Conditional prob.

Conditional Expectation

prob. of Conditional Expectation.

Ch-3

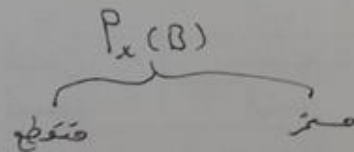
Random Variable and Expectation

Def r.v.

a random variable X (or extended r.v.) on a prob. space (Ω, \mathcal{F}, P) is a Borel measurable fun. from Ω to \mathbb{R} (or $\bar{\mathbb{R}}$)

i.e $X: (\Omega, \mathcal{F}) \xrightarrow{\text{image}} (\mathbb{R}, \mathcal{B}(\mathbb{R})) \ni$

$$P_X(B) = P\{\omega: X(\omega) \in B\}, B \in \mathcal{B}(\mathbb{R})$$



example

let $\Omega = \{HH, HT, TH, TT\}$

and F the set of all subsets of Ω

μ on F is a prob. measure

$$\mu\{HH\} = \mu\{HT\} = \mu\{TH\} = \mu\{TT\} = \frac{1}{4}$$

let x be a function defined on Ω such that

$$x\{HH\} = 0, x\{HT\} = 1, x\{TH\} = 1, x\{TT\} = 2$$

$$x: \Omega \rightarrow \mathbb{R}$$

$$x: \Omega \rightarrow \{0, 1, 2\} = \Omega_1$$

let F_2 be the set of all subsets of Ω_2 , then
 (Ω_2, F_2) is a measurable space

let P_k be the measure on F_2

$$P_k\{0\} = \mu\{HH\} = \frac{1}{4}$$

$$P_k\{1\} = \mu\{HT, TH\} = \frac{1}{2}$$

$$P_k\{2\} = \mu\{TT\} = \frac{1}{4}$$

Then x is a random variable

ملاحظة: إذا كان لدينا space نضع ان نترك اي صغير نتو اي عليه

random x

measure

Point fun. يكون على

set fun.

Borel measure
 ايضا نضع

سؤال /

تعريف random variable X في (Ω, \mathcal{F}, P)

Q/ (i) what do we mean by a random variable and by the corresponding probability dist. fun.
what is the expected value of X .

(ii) Given $\Omega = \{TT, FF, FT, TF\}$
 $X =$ the number of T in each event
Find $P_X(X=i)$, $i=0,1,2$
and then calculate $E(X)$.

Def. the distribution fun. (on prob. space)

the dist. fun. of the r.v. X is the fun.
 $F = F_X$ from \mathbb{R} to $[0, 1]$ given by

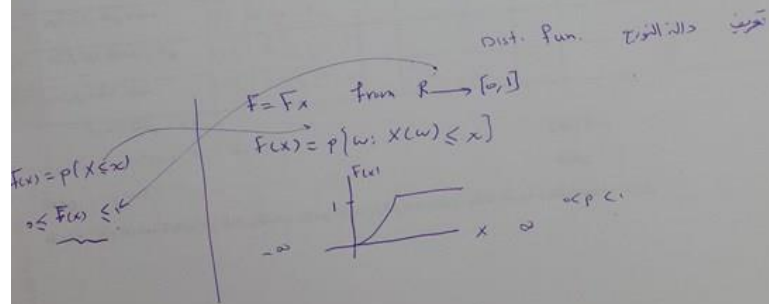
$$F(x) = P\{\omega: X(\omega) \leq x\}, x \text{ real}$$

$$\Rightarrow F(b) - F(a) = P\{\omega: a < X(\omega) \leq b\}, a, b \in \mathbb{R}, a < b$$

$$= P_X(a, b]$$

F is a distribution function corresponding to l.s.m. p.f.
to L.S. m. p.f.

(F is increasing, right continuous, $F(x) \rightarrow 1$ as
 $x \rightarrow \infty$ and $F(x) \rightarrow 0$ as $x \rightarrow -\infty$).



Def: if x is a.r.v. on (Ω, \mathcal{F}, P) then

$$E(x) = \int_{\Omega} x \, dP \quad \text{provided by the integral exists.}$$

if g is Borel measurable fun. from \mathbb{R} to \mathbb{R}
and $Y = g \circ X$

$$\text{then } E(Y) = \int_{\mathbb{R}} g(x) \, dF(x)$$

where. F is the distribution fun. of x

$$\left(= \int_{\mathbb{R}} g \, dP_x \text{ if exists} \right).$$

Remark

A normally distributed r.v. has the useful property that the dist is completely determined by the mean and the variance.

i.e $E(x) = \mu$, $V(x) = \sigma^2$

and $f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

Definition and Comments

Discrete ^{ধাপ}

Def.

let X be a r.v. on (Ω, \mathcal{F}, P) , X is said to be Simple.

iff: X can take only finitely many possible value. then

X discrete iff the set of values of X is finite or countably

infinite if X is discrete and the values of X , $\{x_n\}$, can

be arranged so that $x_n < x_{n+1} \quad \forall n$

then the dist. fun. F is a step fun. with a discontinuity
at each x_n , of magnitude. ^{০০}

$p_n = P\{X = x_n\}$, F is constant between the x_n and x_{n+1}
and take the upper value at each discontinuity.

① if $X_{n-1} < a < x_n \leq b < X_{n+1}$ then

$$F(b) - F(a) = P\{a < X \leq b\} = P_n$$

② if x is arbitrary determined by the prob. fun. of x are completely determined by the prob. fun. P_x , defined by:

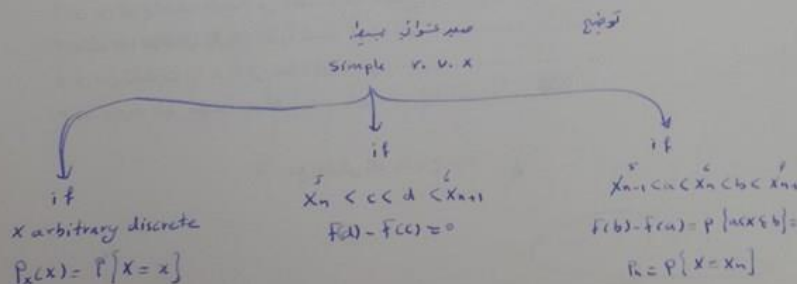
$$P_x(x) = P\{X=x\}, \quad x \in R$$

B Explicitly ^{صريح} ^{مستقرا}

$$P_x(B) = \sum_{x \in B} P_x(x) \quad \left\{ \begin{array}{l} \text{(a countable sum } P_x \text{ is except} \\ \text{except at the } x_n) \end{array} \right.$$

③ if $X_n < c < d < X_{n+1}$ then

$$F(d) - F(c) = 0$$



the r.v. X is said to be absolutely cont.'s iff there is
 anon-negative real valued Borel measurable fun. f on \mathbb{R}
 such that

$$\int_{-\infty}^x f(t) dt, \quad x \in \mathbb{R}.$$

we call f the density or the density fun. of X .

since $F(x) \rightarrow 1$ as $x \rightarrow \infty$ we have

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

if X is absolutely cont.'s with density f it follows
 that

$$P_X(B) = \int_B f(x) dx, \quad \forall B \in \mathcal{B}(\mathbb{R})$$

for the measure μ defined by

$$\mu(B) = \int_B f(x) dx, \quad B \in \mathcal{B}(\mathbb{R})$$

satisfies $\mu(a, b] = f(b) - f(a)$, $a < b$

as μ is the Lebesgue-Stieltjes measure
corresponding to f , ~~there~~ hence
~~also write~~

hence $\mu = P_x$

thus ^{absolutely continuous} absolute continuity of x means

$P_x \ll L.m.$

^{absolutely continuous} \ll \Leftrightarrow \ll

Any non-negative Borel measurable fun. f on \mathbb{R}

with

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

is the density of some ^{abs} absolutely cont. r.v. x

Def.

The r.v. X is said to be cont.s iff
its dist. fun. F is cont.s on all
of $\mathbb{R} \iff X$ is cont.s iff

$$P\{X=x\}=0, \forall x$$

Def: Random vector

An n -dimensional random vector

on a prob. space (Ω, \mathcal{F}, P) is a Borel
measurable map from Ω to \mathbb{R}^n

two number x and y are ^{indep} picked at random between 0 and 1. assume that x and y are indep. and that each is uniformly distributed (i.e. x and y have densities f_1 and f_2 given by $f_1(x) = f_2(x) = 1$, $0 \leq x \leq 1$, and 0 o.w.)
let z be the product xy and let us find the
dist. of z

take $\Omega = \mathbb{R}^2$

$$F = \mathcal{B}(\mathbb{R}^2)$$

$$X(x, y) = x$$

$$Y(x, y) = y$$

$$\text{and } f(x, y) = f_1(x) \cdot f_2(y)$$

$$\therefore P\{(x, y) \in B\} = \int_B f(x, y) dx dy$$

$$\Rightarrow P(B) = \int_B f_1(x) \cdot f_2(y) dx dy$$

$$\begin{aligned} \therefore F(z) &= P(z \leq z) = P\{(x, y) : \overbrace{xy}^B \leq z\} \\ &= \int_{xy \leq z} f_1(x) f_2(y) dx dy \end{aligned}$$

$\therefore 0 \leq x, y \leq 1$ i.e. x, y between 0, 1 (with prob. 1)

$$f(x) = 1 \quad 0 < x < 1$$

$$f(y) = 1 \quad 0 < y < 1$$

$$f(x, y) = f(x) f(y) = 1 \quad \begin{matrix} 0 < x < 1 \\ 0 < y < 1 \end{matrix}$$

$$\begin{aligned} z &= xy & \text{let} & & y &= \frac{z}{x} = \frac{z}{g} \\ g &= x & \Rightarrow & & x &= g \end{aligned}$$

\Rightarrow

$$|J| = \begin{vmatrix} \frac{dy}{dz} & \frac{dy}{dg} \\ \frac{dx}{dz} & \frac{dx}{dg} \end{vmatrix} = \begin{vmatrix} \frac{1}{g} & -\frac{z}{g^2} \\ 0 & 1 \end{vmatrix} = \frac{1}{g} (1 - (-\frac{z}{g^2})(0)) = \frac{1}{g} - 0 = \frac{1}{g}$$

$$J = \frac{1}{g}$$

$$|J| = \frac{1}{g}$$

$$\therefore f(z, g) = \begin{cases} \frac{1}{g} & , 0 < z < 1 \\ & , z < g < 1 \\ 0 & \text{w.} \end{cases}$$

2 dimensions
area $\sim f(z, g)$ w.

$$f(z) = \int_z^1 \frac{1}{g} dg = \ln g \Big|_z^1 = -\ln z$$

$$f(z) = \begin{cases} -\ln z & , 0 < z < 1 \\ 0 & \text{w.} \end{cases}$$

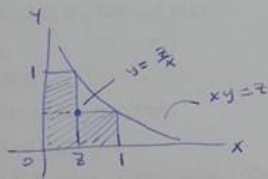
$$\begin{aligned} &\begin{matrix} 0 < x < 1 \\ 0 < y < 1 \\ 0 < xy < 1 \\ 0 < z < 1 \end{matrix} & \begin{matrix} x = \frac{z}{g} = y \\ 0 < x < 1 \\ 0 < y < 1 \\ 0 < g < 1 \end{matrix} \\ & & \begin{matrix} 1, 1 = 1 \\ y = 0 \\ z = y = 0 \\ y = 1 \\ z = y \end{matrix} \end{aligned}$$

$$\therefore F(z) = 1 \text{ for } z \geq 1$$

$$\text{and } F(z) = 0 \text{ for } z \leq 0$$

$$\text{since } f_1(x) \cdot f_2(x) = 1 \text{ for } 0 < x \leq 1, 0 < y \leq 1 \\ \text{and } 0 \text{ o.w.}$$

$f(z)$, ($0 < z < 1$) is the area under the curve xy as fig.1 where $y = \frac{z}{x}$ if $z < x < 1$



$$F(z) = z + \int_z^1 \frac{z}{x} dx = z - z \ln z, \quad 0 < z < 1$$

$$f(z) = F'(z) = 1 - (z \cdot \frac{1}{z} + \ln(z) \cdot 1)$$

$$= 1 - 1 - \ln z$$

$$= -\ln z, \quad 0 < z < 1 \\ 0 \quad \text{o.w.}$$

let X, Y and Z be indep. r.v.'s each is normally distributed with $m=0$, $\sigma=1$; i.e X, Y and Z each have the density

$$g(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-m)^2}{2\sigma^2}} \quad \text{with } m=0, \text{ and } \sigma=1$$

let $w = (x^2 + y^2 + z^2)$, (take the positive square root so that $w \geq 0$)

Find the dist. fun. of w .

Sol

we take $\Omega = \mathbb{R}^3$,

$$F = \mathcal{B}(\mathbb{R}^3)$$

$$X(x, y, z) = x$$

$$Y(x, y, z) = y$$

$$Z(x, y, z) = z$$

and

$$P(B) = \iiint_B f(x, y, z) dx dy dz$$

$$\text{where } f(x, y, z) = f(x)g(y)g(z)$$

we

$$f(x, y, z) = g(x) \cdot g(y) \cdot g(z)$$

$$= (2\pi)^{-\frac{3}{2}} \exp\left\{-\frac{1}{2}(x^2 + y^2 + z^2)\right\}$$

then
things

$$F(w) = P\{W \leq w\} = P\{x^2 + y^2 + z^2 \leq w\} \text{ if } w \geq 0$$

$$\therefore f(w) = \iint_{x^2 + y^2 + z^2 \leq w} (2\pi)^{-\frac{3}{2}} \exp\left\{-\frac{1}{2}(x^2 + y^2 + z^2)\right\} dx dy dz$$

نحوه محاسبه این انتگرال را در ادامه خواهیم دید.

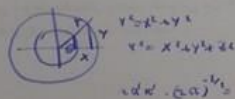
in spherical coordinates.

$$F(w) = \int_0^{2\pi} d\phi \int_0^\pi d\theta \int_0^w (2\pi)^{-\frac{3}{2}} \exp\left\{-\frac{1}{2}r^2\right\} \cdot r^2 \sin\theta dr$$

$$= (2\pi)^{-\frac{3}{2}} (2\pi) (2) \int_0^w r^2 e^{-\frac{1}{2}r^2} dr$$

$\therefore w$ is absolutely continuous with density

$$f(w) = \begin{cases} \frac{2}{\sqrt{2\pi}} w^{\frac{1}{2}} e^{-\frac{1}{2}w} & , w \geq 0 \\ 0 & , w < 0 \end{cases}$$



Theorem

let x_1, x_2, \dots, x_n be indep. r.v.s on (S, \mathcal{F}, P)
if all x_i are non-negative or if $E x_i$ is finite, $< \infty$
for all i , then $E(x_1 x_2 \dots x_n)$ exists and equals
 $E(x_1) \cdot E(x_2) \cdot \dots \cdot E(x_n)$

proof ① if all $x_i \geq 0$ then non-negative

$$E(x_1, \dots, x_n) = \int_{\mathcal{R}} x_1 \dots x_n dP_x(x_1, \dots, x_n) \quad \xrightarrow{x=x_1, \dots, x_n}$$

since P_x is the product of P_{x_i} (using Fubini's theorem)
we get

$$E(x_1, \dots, x_n) = \int_{\mathcal{R}} x_1 dP_{x_1}(x_1) \cdot \int_{\mathcal{R}} x_2 dP_{x_2}(x_2) \dots \int_{\mathcal{R}} x_n dP_{x_n}(x_n)$$
$$E(x_1, \dots, x_n) = E(x_1) \cdot E(x_2) \cdot \dots \cdot E(x_n)$$

② if all x_i finite then.

$$E(|x_1, \dots, x_n|) = \prod_{i=1}^n E(x_i) < \infty$$

Thus Fubini's theorem may be applied just as
in the first part of the proof.

Lemma Cauchy-Schwarz inequality

Let x and y have finite second moments then:

$$[E(xy)]^2 \leq E(x^2) E(y^2)$$

Proof:

Corollary: $| \rho_{xy} | \leq 1$, this equality holds iff:
 $P(Y = cx) > 0$ for some constant c .

theorem

Cauchy, Chebyshev's - schwartz inequality

let x and y have finite second moments then.

$$[E(xy)]^2 = |E(xy)|^2 \leq E x^2 \cdot E y^2$$

proof :

corollary. $|r_{xy}| \leq 1$, this equality holds iff:
 $P\{Y=cX\}=1$ for some constant c .

conditional expectation.

Theorem. let x and y be two dimensional r.v. each
then $u_1(x), u_2(y)$ be a fun. of r.v. then

$$(i) \quad E \{ u_1(y) + u_2(y) | x=x \} = E \{ u_1(y) | x=x \} + E \{ u_2(y) | x=x \}$$

$$(ii) \quad E \{ u_1(x) \cdot u_2(y) | x=x \} = u_1(x) \cdot E \{ u_2(y) | x=x \}$$

proof.

Note if x and y independent then.

(i) $f(x, y) = f(x)$

(ii) $p(a < x < b \mid Y=y) = p(a < x < b)$

(iii) $E(X \mid Y=y) = E(X)$

Theorem let x and y be a two dimensional r.v. then
if $E\{u(y)\}$ exists we have

$$E\{u(y)\} = E\{E\{u(y) \mid X=x\}\}$$

ومن يثبت الخمول (24-3) التفسير
 التبعيات الثاني وقطاع الشعب في
 نعو النمط العشوائي. أما قيمة معامل
 ما بين (1 < R < 2) وهذا يعني
 المدائن فهي تنحصر ما بين
 إما اختبار المصفوفات
 نمط التوزيع

theorem.

$$\text{Var}(Y) = E \left\{ \text{Var}[Y|X=x] + \text{Var}[E(Y|X=x)] \right\}$$

proof

proof

$$E(y) = E \{ E[y|x=x] \}$$

$$E(x) = E \{ E[x|y=y] \}$$

example,

$$\text{let } f(x, y) = x + y \quad , 0 < x < 1 \\ , 0 < y < 1$$

$$\text{find (1) } E\{X|Y=y\}$$

$$(4) E\{X^2 Y^2 | Y = \frac{1}{2}\}$$

$$(2) E\{X | Y = \frac{1}{2}\}$$

$$(5) \text{Var}[X | Y = \frac{1}{2}]$$

$$(3) E\{X + 2X^2 | Y = \frac{1}{2}\}$$

example

let $X \sim \text{Ber}(p)$. suppose that.
 $E\{Y|X=0\}=1$ and $E\{Y|X=1\}=2$
what is EY .

example, let $f(x) = \begin{cases} \frac{x}{3} & , x=1/n \\ 0 & \text{o.w.} \end{cases}$

and let $Y|X=x \sim b(x, \frac{1}{2})$

Find (1) EY

(2) $f(x, y)$.

show that. $E[\text{Var}[Y/X=x]] = \text{Var}(Y)$ iff $E[Y/X=x]$ is independent of x .

* show that $x+y$ and $x-y$ are uncorrelated iff $\sigma_x^2 = \sigma_y^2$

* if $\sigma_x^2 = \sigma_y^2$ for any r.v.s x and y then show that

$$\rho_{x, x+y} = \frac{\text{Cov}(x, x+y)}{\sqrt{\text{Var}(x) \text{Var}(x+y)}} = \sqrt{\frac{1}{2}(1+\rho_{xy})}$$

* show that for any indep. r.v.s x and y then

$$\text{Var}(Y) = \mu_y^2 \sigma_x^2 + \mu_x^2 \sigma_y^2 + \sigma_x^2 \sigma_y^2$$

* if x and y r.v. and independent and if $U_1(x)$ and $U_2(y)$ are fun. for the two r.v. then.

$$E[U_1(x), U_2(y)] = E[U_1(x)] \cdot E[U_2(y)]$$

* let the joint p.d.f. of x and y be given by

$$f(x,y) = 1 - a(1-2x)(1-2y) \quad \begin{matrix} 0 < x < 1 \\ 0 < y < 1 \end{matrix}$$

prove or disprove that x and y are indep. iff they are uncorrelated

Bini's theorem

assume the hypothesis of the product measure theorem

which says that: let $(\Omega_1, \mathcal{F}_1, \mu_1)$ be a measure space

with μ_1 σ -finite on \mathcal{F}_1

and let Ω_2 be a set with σ -field ~~assume that~~

~~for each~~ $\omega_1 \in \Omega_1$

assume that for each $\omega_1 \in \Omega_1$

we are given a measure $\mu(\omega_1, \cdot)$ on \mathcal{F}_2

assume that $\mu(\omega_1, B)$, besides being a measure
in B for each fixed $\omega_1 \in \Omega_1$, is B -measurable
in ω_1 for each fixed $B \in \mathcal{F}_2$.

Assume that $\mu(\omega_1, \cdot)$ are uniformly σ -finite
(i.e. Ω_2 can be written as $\bigcup_{n=1}^{\infty} B_n$).

where for some positive (finite) constant K_n
we have $\mu(\omega_1, B_n) \leq K_n$, $\forall \omega_1 \in \Omega_1$

[The case in which $\mu(\omega_1, \cdot)$ are uniformly bounded
i.e. $\mu(\omega_1, \Omega_2) \leq K < \infty$]

→ Then there is a unique measure μ on $\mathcal{F} = \mathcal{F}_1 \times \mathcal{F}_2$

such that

$$\mu(A \times B) = \int_A \mu(\omega_1, B) d\omega_1, \quad \forall A \in \mathcal{F}_1, B \in \mathcal{F}_2$$

استنتاج نظرية مقياس المنتج
بقانون فون نيومان

let $f: (\Omega, \mathcal{F}) \rightarrow (\bar{\mathbb{R}}, \mathcal{B}(\bar{\mathbb{R}}))$

if f non-negative.

then

$= \int_{\Omega_2} f(w_1, w_2) \mu(w_1, dw_2)$ exists and defines
a Borel measurable $f(w_1)$

also

$$\int_{\Omega} f d\mu = \int_{\Omega_2} \left(\int_{\Omega_1} f(w_1, w_2) \mu(w_1, dw_2) \right) \mu(dw_1)$$

$E\{g_1(x)\} \leq E\{g_2(x)\}$, if $g_1(x) \leq g_2(x)$.

proof $|E(g(x))| \leq E(|g(x)|)$