

Let (Ω, \mathcal{F}, P) be a probability space. A function, g , is called an indicator function if it takes the values 0 and 1 only. The set of points, $E = \{\omega : g(\omega) = 1\}$, is called "the set indicated by g " and " g " is the indicator of E ". For $E \in \mathcal{F}$ we write I_E for the indicator of E , so $I_E(\omega) = 1$ if $\omega \in E$ and is zero otherwise. We define the class of elementary random indicator functions, I_E , for $E \in \mathcal{F}$, to be the elementary random variables and we define,

$$\int_{\Omega} I_E dP \equiv P(E).$$

A random variable is simple iff it is a finite linear combination of elementary random variables. It is clear that a simple random variable takes only finitely many distinct values. Let f be a simple random variable taking the distinct values, $\alpha_1, \dots, \alpha_n$. Then we may write

$$f(\omega) = \sum_{i=1}^n \alpha_i I_{E_i} \quad (1)$$

where $E_i = \{\omega : f(\omega) = \alpha_i\}$. Since the α_i 's are different of one another, $E_i \cap E_j = \emptyset$ for $i \neq j$ and each E_i is non-empty. Any function which

may be written as $\sum_{i=1}^n \alpha_i I_{E_i}$ where $E_i \in \mathcal{F}$, $1 \leq i \leq n$, $\alpha_i \in \mathbb{R}$.
 According to it is a simple random variable. $\sum_{k=1}^n p_k I_{F_k}$ where the p_k 's are the distinct values that $\sum_{k=1}^n \alpha_k I_{E_k}$ takes, each $F_i \in \mathcal{F}$, $F_i \cap F_j = \emptyset$ for $i \neq j$. We define, for $F = \sum_{k=1}^n p_k I_{F_k}$.

A function of the form $\sum_{i=1}^n \alpha_i I_{E_i}$, with $E_i \in \mathcal{F}$, $1 \leq i \leq n$ may be written in many different ways as a linear combination of indicator functions. There is a lemma which proves that such a function can be written as a linear combination of indicator functions of disjoint sets in \mathcal{F} — non-uniquely — and that the integral defined in the manner of (2) above is well defined and the integrals are all equal to that defined in (2).

A simple random variable is positive if its range lies in $[0, \infty)$.

Theorem
 Let $f: \Omega \rightarrow [0, \infty]$ be (measurable). There exist simple random variables, (Δ_n) , each of which are positive and

$$(1) 0 \leq \Delta_1 \leq \Delta_2 \leq \Delta_3 \leq \dots \leq f$$

$$(2) \Delta_n(\omega) \rightarrow f(\omega) \quad \forall \omega \in \Omega$$

Pr for $n=1, 2, \dots$ and $1 \leq i \leq n2^n$ let

$$E_{n,i} = F^{-1} \left(\left[\frac{i-1}{2^n}, \frac{i}{2^n} \right) \right)$$

$$F_n = F^{-1}([n, \infty])$$

set $\Delta_n = \sum_{i=1}^{n2^n} \left(\frac{i-1}{2^n} \right) I_{E_{n,i}} + n F_n$. By definition

$E_{n,i}$ and F_n are in \mathcal{F} for each n and each i , and for fixed n these sets are disjoint. Property (1) follows because of the "halving" occurring in the construction and the way the values of Δ_n are defined on $E_{n,i}$ and F_n .

Property (2) follows because if $F(\omega) = \infty$ then $\Delta_n(\omega) = n$ and " $\Delta_n(\omega) \rightarrow \infty$ " as $n \rightarrow \infty$. If $F(\omega) < \infty$ then for large enough n , $F(\omega) \in E_{n,i}$ for some suitable choice of i , so $|F(\omega) - \Delta_n(\omega)| < 1/2^n$, enough.

Integrals of Simple Random Variables

Let $f = \sum_{i=1}^n \alpha_i I_{E_i}$, $E_i \in \mathcal{F}$, $E_i \cap E_j = \emptyset$, $i \neq j$, and $\alpha_1, \dots, \alpha_n$ are the distinct values of f . We define

$$\int f dP = \int f I_F dP = \sum_{i=1}^n \alpha_i P(E_i \cap F)$$

(convention $\infty \cdot 0 = 0$)
If g is also simple, $g = \sum_{j=1}^m \beta_j I_{F_j}$... the usual stuff

being true for the F_j 's and β_j 's ... Then $f+g$ is clearly a simple function and, by adapting "that lemma" alluded to earlier one can show that

$$\int (f+g) dP = \int f dP + \int g dP$$

Evidently, for $\lambda \in \mathbb{R}$, $\int \lambda f dP = \lambda \int f dP$ and for disjoint sets, $E, F \in \mathcal{F}$,

$$\int f dP = \int_E f dP + \int_{E^c} f dP.$$

If $f \geq 0$, f simple then, obviously, $\int f dP \geq 0$ and $\int f dP = 0 \iff f = 0$ on a set of probability 1.

Integration of Non-Negative Random Variables

Let $f: \Omega \rightarrow [0, \infty]$ be a random variable. We

define,

$$\int f dP = \sup \left\{ \int \Delta dP : \Delta \text{ is simple, } 0 \leq \Delta \leq f \right\}$$

and for $E \in \mathcal{F}$,

$$\int_E f dP = \sup \left\{ \int \Delta dP : \Delta \text{ is simple, } 0 \leq \Delta \leq f \right\}$$

Note that if f is simple then this is consistent with our previous definition because, f is itself simple and $\int f dP \geq \int \Delta dP$ for every simple $\Delta \leq f$ — (because the integral of +ve things is +ve)

Theorem Let f, g etc be random variables, E, F, G, \dots etc events in \mathcal{F} . Then,

Remember, we deal with positive functions.

(a) If $f \leq g$ then $\int_{\Omega} f dP \leq \int_{\Omega} g dP$

(b) If $A, B \in \mathcal{F}$, $A \subseteq B$ then $\int_A f dP \leq \int_B f dP$

(c) If $0 \leq f$, $c \in [0, \infty)$ then $\int c f dP = c \int f dP$

(d) If $f(\omega) = 0$ on a set of probability 1, then $\int f dP = 0$

for every $E \in \mathcal{F}$.

(e) If $P(E) = 0$ then $\int_E f dP = 0$ (even if $f = \infty$ on E)

(f) $\int f dP = \int f I_E dP$

PF (a) If λ is simple, $\lambda \leq f \leq g$ then $\lambda \leq g$

no $\int \lambda dP \leq \int g dP$. Taking the supremum over all

such λ leads to $\int f dP \leq \int g dP$.

(b) If $A \subseteq B$ and $f \geq 0$ then $f I_A \leq f I_B$ and by

(a), $\int f I_A dP \leq \int f I_B dP$. Now, if $\lambda \leq f$ and λ is simple

then λI_E is simple and $\lambda I_E \leq f I_E$. So,

$$\int \lambda dP \leq \int f I_E dP.$$

Taking the supremum over all $\lambda \leq f$ gives,

$$\int f dP \leq \int f I_E dP.$$

On the other hand, if λ is simple and $\lambda \leq f I_E$ then $\lambda \leq f$ and $\lambda = \lambda I_E$. If we choose λ so that $\int \lambda dP > \int f I_E dP - \epsilon$

(this assumes $\int_{\Omega} f I_{E^c} dP < \infty$) then, as $\Delta = \Delta I_E$ and $\Delta \ll f$,

$$\int_{\Omega} \Delta dP = \int_{\Omega} \Delta dP \ll \int_{\Omega} f dP, \text{ so } \int_{\Omega} f I_E dP < \int_{\Omega} \Delta dP + \epsilon.$$

This shows $\int_{\Omega} f dP = \int_{\Omega} f I_E dP$, proves (b) and concludes

(b). If $\int_{\Omega} f I_E dP = \infty$ then for each $n \in \mathbb{N}$ there is a

simple random variable, Δ_n , $\Delta_n \ll f I_E$ with $\int_{\Omega} \Delta_n dP \geq n$.

But $\Delta_n = \Delta_n I_E$ and $\int_{\Omega} \Delta_n dP = \int_{\Omega} \Delta_n dP$ showing $\int_{\Omega} f dP = \infty$.

(c) If $c = 0$ it's obvious. If $c > 0$ and Δ is simple

$\Delta \ll f$ then $c \Delta \ll c f$, and $c \Delta$ is simple too, as is

$c \Delta I_E$. Now,

$$\int_{\Omega} c \Delta dP \geq \int_{\Omega} c \Delta dP = c \int_{\Omega} \Delta dP, \text{ so } \int_{\Omega} c \Delta dP \geq c \int_{\Omega} f dP.$$

On the other hand, if t is simple, $t \ll c f$ then $\frac{t}{c} \ll f$.

Now $\frac{t}{c}$ is simple and

$$\int_{\Omega} f dP \geq \int_{\Omega} \frac{t}{c} dP = \frac{1}{c} \int_{\Omega} t dP, \text{ taking the sup over } t$$

gives $\int_{\Omega} f dP \geq \frac{1}{c} \int_{\Omega} c f dP$. This proves c.

(d) Let $E = \{\omega : f(\omega) = 0\}$ and $\alpha = \inf\{f(\omega) : \omega \in E^c\}$

then $\Delta = \alpha I_{E^c}$ is a simple random variable and is the

largest simple random variable less than f . So

$$\int_{\Omega} f dP = \int_{\Omega} \Delta dP = 0 \cdot P(E) + \alpha \cdot P(E^c)$$

$$= 0 \cdot 1 + \alpha \cdot 0$$

$$= 0 \text{ (even if } \alpha = \infty)$$

By (b), $f I_E \ll f$ so $\int_{\Omega} f I_E dP \leq \int_{\Omega} f dP = 0$, for $f \in \mathcal{F}$.

② If $P(E) = 0$ then $\int_E f dP$ is zero on a set of probability 0. So $\int_E f dP = 0$ by ①, and the result follows from ①.

The Monotone Convergence Theorem

Let (f_n) be a sequence of random variables, suppose

$$(i) \quad 0 \leq f_1 \leq f_2 \leq \dots \leq \infty \quad \forall \omega \in \Omega,$$

$$(ii) \quad f_n(\omega) \rightarrow f(\omega) \text{ as } n \rightarrow \infty \text{ for each } \omega \in \Omega.$$

Then f is a random variable and $\int f_n dP \rightarrow \int f dP$.

Pr We know $(\int f_n dP)$ is an increasing sequence. Let

$$\lim_n \int f_n dP = \alpha \leq \infty.$$

It is known that f is measurable (ask about this CB).

Since $f_n \leq f$ we have $\int f_n dP \leq \int f dP \quad \forall n, \forall \Omega$,

$$\alpha \leq \int f dP.$$

Let Δ be simple, $0 \leq \Delta \leq f$ and let $C \in (0, 1)$.

Define $E_n = \{\omega : f_n(\omega) \geq C\Delta(\omega)\}$, $n \in \mathbb{N}$. Each

$E_n \in \mathcal{F}$. Also, $E_1 \subseteq E_2 \subseteq E_3 \dots$ and $\Omega = \bigcup_n E_n$

because the f_n 's are increasing & converge to $f(\omega) \geq \Delta(\omega)$.

If $f(\omega) = 0$ then $\Delta(\omega) = 0$ and $f_n(\omega) = 0$ no $\omega \in E_1$. If $f(\omega) > 0$ then $f(\omega) > C\Delta(\omega)$ ($f \geq \Delta$), there will be an n such that $f_n(\omega) \geq C\Delta(\omega)$, here $\omega \in E_n$. Also,

$$\int f_n dP \geq \int_{E_n} f_n dP \geq C \int_{E_n} \Delta dP.$$

letting $n \rightarrow \infty$ we get

$$\alpha = \lim_n \int_{E_n} f_n dP \geq \lim_n c \int_{E_n} dP.$$

Fact: Do we need to prove this? The set function

$$\exists E \in \mathcal{F} \rightarrow \int_{E_n} dP$$

is a (finite) measure on \mathcal{F} .

Assuming this,

$$\alpha \geq c \int_{\Omega} dP$$

for every $c \in (0, 1)$. This shows $\alpha \geq \int_{\Omega} dP$ and hence

$\alpha \geq \int_{\Omega} f dP$. So the result is proved.

Corollary

Let f and g be random variables. Then

$$\int_{\Omega} (f+g) dP = \int_{\Omega} f dP + \int_{\Omega} g dP.$$

Pf From an earlier theorem we can find simple r.v's

$\Delta_n \uparrow f$ and $t_n \uparrow g$ with, according to the MCOm

$$\int_{\Omega} \Delta_n dP \rightarrow \int_{\Omega} f dP \text{ and } \int_{\Omega} t_n dP \rightarrow \int_{\Omega} g dP. \text{ Since } \Delta_n + t_n \uparrow f+g$$

the result follows because the integral is additive on

simple r.v's.

Fatou's Lemma

If (f_n) is a sequence of r.v.'s

$$\int_{\Omega} \liminf_n f_n dP \leq \liminf_n \int_{\Omega} f_n dP$$

PF Set $g_k = \bigvee_{l \geq k} f_l$ then $g_k \leq f_k$ obviously and g_k 's

$0 \leq g_1 \leq g_2 \leq \dots$ That g_k is a r.v. is "well known".

Now $g_k \rightarrow \liminf_n f_n$ (by definition) so, by monotone

convergence,

$$\int_{\Omega} g_k dP \rightarrow \int_{\Omega} \liminf_n f_n dP.$$

But $\int_{\Omega} g_k dP \leq \int_{\Omega} f_k dP$ and so

$$\lim_k \int_{\Omega} g_k dP = \lim_k \int_{\Omega} \inf_{l \geq k} f_l dP \leq \liminf_k \int_{\Omega} f_k dP, \text{ showing}$$

the result is true.

The Dominated Convergence Theorem.

Suppose (f_n) is a sequence of r.v.'s such that:
(i) $\lim_n f_n(\omega) = f(\omega)$ exists $\forall \omega \in \Omega$

(ii) There is a r.v. g , with $g \geq 0$ and $f_n \leq g \forall n$ and $\int_{\Omega} g dP < \infty$.

Then, f is a r.v., $\int_{\Omega} f dP < \infty$, $\lim_n \int_{\Omega} |f - f_n| dP = 0$ and as a consequence, $\lim_n \int_{\Omega} f_n dP = \int_{\Omega} f dP$.

Let since $f \leq g$ then $\int_{\Omega} f d\mu \leq \int_{\Omega} g d\mu < \infty$, and that f is a r.v. is "well known". Now $|f_n - f| \leq 2g$ and the sequence $2g - |f_n - f|$ converges to $2g$. By Fatou's lemma

$$\int_{\Omega} \liminf (2g - |f_n - f|) d\mu = \int_{\Omega} 2g d\mu < \liminf \int_{\Omega} (2g - |f_n - f|) d\mu$$

$$= \int_{\Omega} 2g d\mu + \liminf \left(- \int_{\Omega} |f_n - f| d\mu \right)$$

$$= \int_{\Omega} 2g d\mu - \limsup \int_{\Omega} |f_n - f| d\mu$$

Subtracting $\int_{\Omega} 2g d\mu$ (which is finite!)

$$0 \geq \limsup \int_{\Omega} |f_n - f| d\mu$$

Since $\forall n \int_{\Omega} |f_n - f| d\mu \geq 0$ this forces the limsup and lim inf to be equal. So

$$\lim \int_{\Omega} |f_n - f| d\mu = 0.$$

Since (again we not proved this...)

$$\left| \int_{\Omega} f d\mu - \int_{\Omega} f_n d\mu \right| = \left| \int_{\Omega} (f - f_n) d\mu \right| < \int_{\Omega} |f - f_n| d\mu$$

This shows that $\lim \int_{\Omega} f_n d\mu = \int_{\Omega} f d\mu$

□

(+) ok in checking, we don't yet know that $|f - f_n|$ is a r.v.