

$$X^{-1}(B) = \left\{ \omega : \underbrace{I_A(\omega)}_{0 \text{ or } 1} \in B \right\}$$

$$X^{-1}(B) = \begin{cases} \emptyset & \text{if } 0 \notin B \text{ and } 1 \notin B \\ A^c & \text{if } 0 \in B \text{ and } 1 \notin B \\ A & \text{if } 0 \notin B \text{ and } 1 \in B \\ \Omega & \text{if } 0 \in B \text{ and } 1 \in B \end{cases}$$

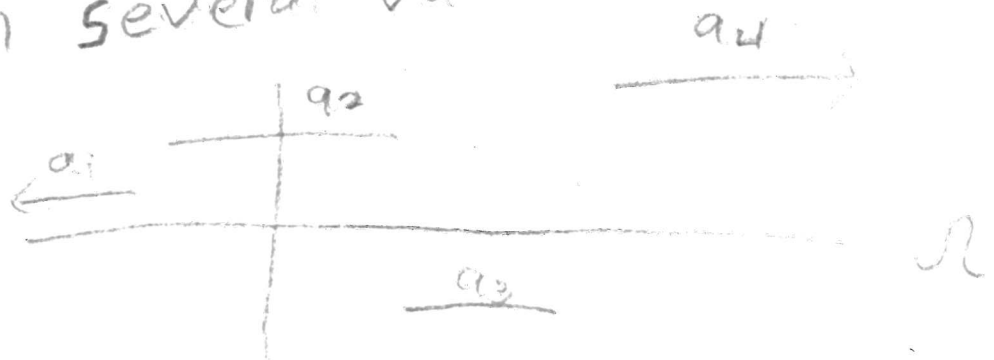
(If $0 \in B$ and $1 \notin B$, then

$$X^{-1}(B) = \left\{ \omega : I_A(\omega) \in B \right\} = \left\{ \omega : I_A(\omega) = 0 \right\} = A^c.)$$

Note: $\sigma(I_A) = \{ \emptyset, A, A^c, \Omega \}$

and $A \in \mathcal{F}$ iff $A^c \in \mathcal{F}$.

ex) Suppose X is a function that takes on several values.



e.g. $X(\omega) = a_1 I_{A_1}(\omega) + a_2 I_{A_2}(\omega) + a_3 I_{A_3}(\omega) + a_4 I_{A_4}(\omega)$

where $A_i = \{\omega : X(\omega) = a_i\}$.

Such a step function is a linear combination of indicators.

Pr 85 [2] A function is simple if it is a linear combination of indicators

$$f = \sum_{i=1}^K \alpha_i I_{A_i} \quad \text{for some positive}$$

integer K (so finite range), ^{measurable} ~~simple~~

13] A simple function is a RV iff each $A_i \in \mathcal{F}$.

End Exam 1 material

Begin Exam 2 material

Integration and Expected Value

§5, 15-18, 21
 ~~~~~  
 ch 3 ch 4

P63  $\Rightarrow$  Fix  $(\Omega, \mathcal{F}, P)$ , A simple random variable is a function  $X: \Omega \rightarrow \mathbb{R}$

such that the range of  $X$  is finite

and  $\{X=x\} = \{\omega: X(\omega)=x\} \in \mathcal{F} \quad \forall x \in \mathbb{R}$ ,

( $X$  is a discrete RV with finite support.)

ex)  $X = \sum_{i=1}^n x_i I_{A_i}$ ,  $A_i \in \mathcal{F}$  is a

simple RV.

ex)  $A_n$   $n \geq 1$  disjoint,  $X = \sum_{i=1}^{\infty} x_i I_{A_i}$ ,

where  $x_i \neq x_j$  for  $i \neq j$ , is not

a simple RV since  $X$  has infinite range.

2) Suppose <sup>events</sup>  $A_1, \dots, A_n$  are disjoint and

$\bigcup_{i=1}^n A_i = \Omega$ . Let  $X = \sum_{i=1}^n x_i I_{A_i}$ .

The expected value of  $X =$

Use for proofs

$$E[X] = \sum_{i=1}^n x_i P(A_i) = \sum_x x P(X=x)$$

which is a finite sum since  $X$  is a simple RV.

### 3) Existence and Uniqueness of $E[X]$ :

Existence:

Suppose  $X$  takes on <sup>distinct</sup> values

$x_1, \dots, x_m$  ( $m$  need not equal  $n$ ).

$$\text{Let } B_i = \{X = x_i\} = \{\omega : X(\omega) = x_i\} \text{ for}$$

$i=1, \dots, m$ . Then the  $B_i$  are disjoint

$$\text{and } \bigcup_{i=1}^m B_i = \Omega.$$

$$\text{Thus } X = \sum_{i=1}^m x_i I_{B_i} \text{ and } E(X) = \sum_{i=1}^m x_i P(B_i)$$

$$= \sum_{i=1}^m x_i P(X=x_i).$$

$$\text{Uniqueness: } \sum_{i=1}^n x_i P(A_i) =$$

$$\sum_x \sum_{i: x_i=x} x_i P(A_i) = \sum_x x P(\cup_{i: x_i=x} A_i)$$

$x_i = x$  in this sum       $\{x=x\}$

finite sum

$$= \sum_x x P(x=x),$$

Note: Although many partitions  $A_i$ 's exist, they all yield the same value of  $E(X)$ .

(Existence in  $\mathbb{R}$  shows  $X$  can be written as in 2)

4) Let  $X_n, X$  and  $Y$  be simple RVS.

a)  $-\infty < E[X] < \infty$ .

linearity

b)  $E[aX + bY] = aE(X) + bE(Y)$

c) If  $X \leq Y$ , then  $E(X) \leq E(Y)$

d) If  $\{X_n\}$  is uniformly bounded and  $X = \lim_n X_n$  on a set of prob 1,

then  $E(X) = \lim_n E(X_n)$

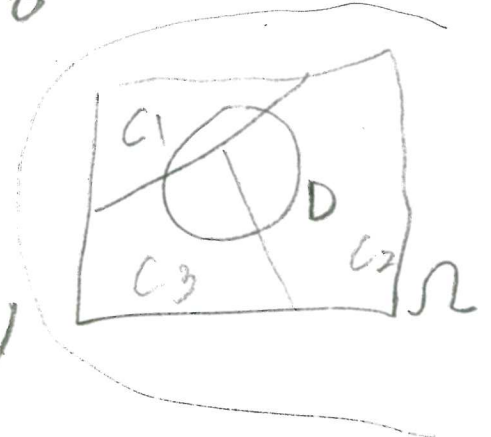
(Special case of LDCT)

e)  $E[t(x)] = \sum_x t(x) P(X=x)$ .  $t$  a real valued function 26.5

f) If  $X$  is nonnegative,  $X \geq 0$ ,

then  $E[X] = \sum_i P(X > x_i) = \int_0^\infty P(X > x) dx$

$= \int_0^\infty [1 - F(x)] dx$ .



Survival function  $S(x)$

$E[XY] = E[X]E[Y]$ .

some proofs

a)  $E[X] = \sum_x x P(X=x)$ . The result holds since  $X$  has finite range  $x_1, \dots, x_m$ .  
 bounded  $\in [0, \infty]$

$\min(x_i) \leq E[X] \leq \max(x_i)$ .

(c)  $E(X) = \sum_i x_i P(A_i)$  and  $E(Y) = \sum_j y_j P(B_j)$ .

skip

Since  $X \leq Y \forall \omega$ ,  $x_i \leq y_j$  if  $A_i \cap B_j \neq \emptyset$   
 disjoint

Venn diagram

$\therefore E[X] = \sum_i x_i P(A_i) = \sum_i \sum_j x_i P(A_i \cap B_j) \leq \sum_i \sum_j y_j P(A_i \cap B_j) = \sum_j y_j P(B_j) = E[Y]$

c) Let  $W = Y - X \geq 0$ .

$$E(W) = \sum_{\substack{w \\ \geq 0}} w P(W=w) \geq 0$$

$$\therefore 0 \leq E[Y - X] = E[Y] - E[X] \quad \leftarrow \text{by b)}$$

or  $E(X) \leq E(Y)$ .

e) If  $X = \sum_{i=1}^n x_i I_{A_i}$ , then

$$t(X) = \sum_{i=1}^n t(x_i) I_{A_i}$$

$t$  as a simple RV

$$\begin{aligned} \therefore E[t(X)] &= E[W] = \sum w P(W=w) \\ &= \sum_{i=1}^n t(x_i) P(A_i) \end{aligned}$$

g)  $XY = \sum_i x_i I_{A_i} \sum_j y_j I_{B_j} =$

$$\sum_i \sum_j x_i y_j I_{A_i \cap B_j} \quad \text{is a SRV}$$

( $I_{A_i} I_{B_j} \in \{0,1\}$  and  $I_{A_i} I_{B_j} = 1$  iff  $I_{A_i} = I_{B_j} = 1$  iff  $I_{A_i \cap B_j} = 1$ )

So  $E[XY] = \sum_i \sum_j x_i y_j P(A_i \cap B_j) =$  ind

$$\sum_i \sum_j x_i y_j P(A_i) P(B_j) =$$

$$\sum_i x_i P(A_i) \sum_j y_j P(B_j) = E(X)E(Y)$$

Note:  $\underbrace{I_{A_1} I_{A_2} \dots I_{A_m}}_{\in \{0,1\} \text{ and } 1 \text{ iff } \bigcap_{i=1}^m A_i = \Omega} = I_{A_1 \cap A_2 \cap \dots \cap A_m} = I_{\bigcap_{i=1}^m A_i}$

b) Let  $X = \sum_i x_i I_{A_i}$  and  $Y = \sum_j y_j I_{B_j}$ .

Then  $aX + bY = a x_i + b y_j$  for  $\omega \in \underbrace{A_i \cap B_j}_{\text{partition } \Omega}$

So  $aX + bY = \sum_i \sum_j (a x_i + b y_j) I_{A_i \cap B_j}$  is a SRV

with  $E[aX + bY] = \sum_i \sum_j (a x_i + b y_j) P(A_i \cap B_j) =$

$$\sum_i a x_i \sum_j P(A_i \cap B_j) + \sum_j b y_j \sum_i P(A_i \cap B_j)$$

$$= a \sum_i x_i P(A_i) + b \sum_j y_j P(B_j) = a E(X) + b E(Y)$$

□