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# EXERCISE II

Since the sequence  $A_1 \subset A_2 \subset \cdots$  is expanding, the union can be written as:  $A = \bigcup_{n=1}^{\infty} A_n \quad \text{where each } A_n \text{ event grows larger}$ 

Notice that  $\ell(A_n) = \ell(A_1 \cup A_2 \cup \cdots)$  is increasing because  $A_1 \subset A_2 \subset \cdots$ Then  $\ell(A_1) \leq \ell(A_2) \leq \cdots$  and  $\lim_{n \to \infty} \ell(A_n)$  exists.

By the definition of the probability measure, the idea of continuity of probability can be seen:  $P(U + n) = \lim_{n \to \infty} P(A_n)$ 

Similarly, since the sequence  $A_1 \supset A_2 \supset \cdots$  is contracting the union can be written as  $B = \bigcap_{n=1}^{\infty} A_n$  where each  $A_n$  event becomes smaller

If one event contains the other, its probability can not increase, then  $R(A_n) = P(A_s \cap A_2 \cap \cdots)$  and  $P(A_s) \ge P(A_2) \ge \cdots$  and  $P(A_n) = P(A_n) \ge P(A_n)$ 

Then by the definition of probability measure  $P\left(\bigcap_{n=1}^{\infty}A_{n}\right)=\lim_{n\to\infty}P(A_{n})$ 

# EXERCISE 12

Given that  $B_n$  is a contracting sequence  $(B_1 \supset B_2 \supset \cdots)$ , the results from Exercise 1.1 show that  $P(\bigcap_{n=1}^{\infty} B_n) = \lim_{n \to \infty} P(B_n)$   $= \lim_{n \to \infty} P(A_n \cup A_{n+1} \cup \cdots)$ By the boole inequality  $P(B_n) \subseteq P(A_n) + P(A_{n+1}) + \cdots$ Assume that  $\bigcap_{n=1}^{\infty} P(A_n) \subset \infty$ Notice that when  $n \to \infty$ , the terms in  $\bigcap_{n=1}^{\infty} P(A_n)$  get progressively  $S_n = \lim_{n \to \infty} P(B_n) = 0$   $P(\bigcap_{n=1}^{\infty} B_n) = \lim_{n \to \infty} P(B_n) = 0$ 

#### EXERCISE 1.3

X is a r.v.  $\sigma(X)$  - measurable, by that for any Borel set A  $\{X \in A\} = X^{-1}(A) \in \mathcal{F}$ Considering  $\{f(X) \in B\}$  where  $\mathcal{B}$  is a Borel set  $\{f(X) \in B\} = X^{-1}(f^{-1}(B))$ Notice that  $X^{-1}(f^{-1}(B)) \in \sigma(X)$  because f is Borel. And  $\sigma(X)$  contains all events that depend on X including  $\{X \in A\}$  where A is a Bovel set. And that implies that f(X) is  $\sigma(X)$ -measurable.

#### EXERCISE 1.4

(1) Assume  $x_1 \angle x_2$ , then the event  $9x \angle x_1$   $3 \Rightarrow 9x \angle x_2$ 3

meaning  $P(x \angle x_1) \angle P(x \angle x_2)$ which implies  $f_{\xi}(x_1) \angle f_{\xi}(x_2)$ Thus  $f_{\xi}(x)$  is decreasing.

1) By the definition,  $f_{\xi}(x_n) = P(\xi = x_n)$ And by Exercise 1.1  $f_{\xi}(x_n) = P(\xi = x_n) = \lim_{n \to \infty} P(\xi = x_n) = \lim_{n \to \infty} f_{\xi}(x_n)$ 

Proving that for is right continuous.

(1) As  $x \to -\infty$  the probability of  $X \le x$  goes to 0 Therefore  $\lim_{n \to \infty} f_{\xi}(x) = f(Q) = 0$ 

Also, by Exercise 1.1  $\lim_{n\to\infty} f_{\xi}(n) = \lim_{n\to\infty} P(g \le n) = P(\Omega) = 1$ 

## EXERCISE 1.5

By definition,

$$f_{\xi}(x) = P(\xi \leq x) = \int_{-\infty}^{x} f_{\xi}(x) dx$$

Assuming that  $f_{\xi}(n)$  is defined. Now, assuming  $f_{\xi}(n)$ 's continuity and that  $f_{\xi}$  is differentiable:  $\int_{\xi} f_{\xi}(n) = f_{\xi}(n)$ on

## EXERCISE 1.6

By definition: 
$$f_{g}(x) = f(\chi \leq x) = \sum_{x_i = x} f(\chi = x_i)$$

As (s,t] is an interval that does not contain  $x_1,x_2,\cdots$ , the probability of X taking any value is 0.

Therefore, for any sexet:

$$f_{g}(t) - f_{g}(s) = P(s < g \leq t) = 0$$

Now, the size of the jump is given by:

$$f_{\xi}(n_i) - f_{\xi}(\bar{x_i}) = f(\xi \leq x_i) - f(\xi \leq x_i) = f(\xi = x_i)$$

the value before  $x_i$ 

#### EXERCISE 1.8

The Schwarz inequality is given by 
$$[E(\S\eta)]^2 = E(\S^2) \cdot E(\eta^2)$$
 The objective is to prove that 
$$E(\S^2) \text{ exists}$$
 Then if  $N=1$  is choosen: 
$$[E(\S\cdot 1)]^2 = E(\S^2) \cdot E(1^2) = E(\S^2) \subset \infty$$

## EXERCISE 19

$$E(\eta^2)$$
 can be written as:  
 $E(\eta^2) = \int_0^\infty t^2 df_n(t)$  where  $f_n(t) = P(\eta \le t)$   
Integrating by parts:

Since  $f_{\eta}(t) \to 1$  as  $t \to \infty$ , the integrable is finite and it is proven that  $E(\eta^2)$  is square integrable.

# EXERCISE 1.10

Given that 
$$B_1, B_2, \cdots$$
 is a sequence or disjoint sets,  
 $A = (A \cap B_1) \cup (A \cap B_2) \cup \cdots$   
By the property of additivity:  
 $P(A) = P((A \cap B_1) \cup (A \cap B_2) \cup \cdots)$   
 $= \sum_{n=1}^{\infty} P(A \cap B_n)$   
 $= \sum_{n=1}^{\infty} P(A \cap B_n) \cdot P(B_n)$ 

## EXERCISE 1.11

If A and B are independent, then P(A|B) = P(A) and vice versa. Therefore, if  $P(B) \neq 0$ , A and B are independent if and only if P(A|B) = P(A)Since  $P(B) \neq 0$  we can multiply by P(B) on both sides:  $P(A) = P(A \cap B) \implies P(A) \cdot P(B) = P(A \cap B) \cdot P(B)$  P(B)and find  $P(A \cap B) = P(A) \cdot P(B)$