

$(\Omega, \mathcal{F}, \mathbb{P})$

- \mathcal{F} sigma-algebra \Rightarrow
- 1) $\mathcal{F} \neq \emptyset$
 - 2) $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$
 - 3) $\{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F} \Rightarrow \bigcup_{i \in \mathbb{N}} A_i \in \mathcal{F}$

Corollary

Corollario \rightarrow 1) \mathcal{F} closed under any finite or countable sequence of operations of sets ($\cup, \cap, \setminus, \Delta, c$)

\uparrow symmetric difference:
 $A \Delta B = (A \cup B) \cap (A \cap B)^c$
 $= (A \cup B) \setminus (A \cap B)$

2) $\emptyset, \Omega \in \mathcal{F}$

$\hookrightarrow \{\emptyset, \Omega\}$ is the smallest σ -algebra possible

$\hookrightarrow 2^\Omega = \{A \mid A \subseteq \Omega\}$ is the largest σ -algebra possible

$\hookrightarrow \{\emptyset, \Omega\} \subseteq \mathcal{F} \subseteq 2^\Omega \forall \mathcal{F} \sigma$ -algebra.

3) $\mathcal{F}_\theta \sigma$ -algebra $\forall \theta \in \Theta$ (not necessarily countable)

$\Rightarrow \bigcap_{\theta \in \Theta} \mathcal{F}_\theta \sigma$ -algebra

\uparrow It is the largest σ -field contained in $\mathcal{F}_\theta \forall \theta \in \Theta$.

4) If G is a family of subsets ($G \subseteq 2^\Omega$),

denote $\sigma(G) = \bigcap_{\mathcal{F} \supseteq G} \mathcal{F} \Rightarrow \sigma(G)$ is the smallest σ -field that contains G .

$\uparrow \sigma(G) \equiv \sigma$ -algebra generated by G .

5) Unions of σ -algebras are not necessarily σ -algebras.

Intersections are.

$\hookrightarrow \mathcal{F}_1 = \{\emptyset, \{1,2\}, \{2,3\}, \Omega\}$

$\mathcal{F}_2 = \{\emptyset, \{1,2\}, \{1,3\}, \Omega\}$

Notación $\rightarrow \bigvee_{\theta \in \Theta} \mathcal{F}_\theta = \sigma(\bigcup_{\theta \in \Theta} \mathcal{F}_\theta)$

\mathbb{P} -probabilidad. $\rightarrow \mathbb{P}: \mathcal{F} \rightarrow [0, 1]$

$\{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F}$ mutually disjoint $\Rightarrow \mathbb{P}(\bigcup_{i \in \mathbb{N}} A_i) = \sum_{i \in \mathbb{N}} \mathbb{P}(A_i)$

$\mathbb{P}(\Omega) = 1$

Corollary \rightarrow 1) $\mathbb{P}(\emptyset) = 0$.

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$$\{A_i\}_{i=1}^n \subseteq \mathcal{F} \Rightarrow P(\bigcap_{i=1}^n A_i) \leq \sum_{i=1}^n P(A_i)$$

$$\{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F} \Rightarrow P(\bigcup_{i \in \mathbb{N}} A_i) \leq \sum_{i \in \mathbb{N}} P(A_i)$$

$$4) \{A_i\}_{i=1}^n \subseteq \mathcal{F} \text{ or } \{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F} \quad \text{if } P(A_i) = 0 \quad \forall i \Rightarrow P(\bigcup_i A_i) = 0$$

$$5) P(A^c) = 1 - P(A)$$

6) $\{A_i\} \subseteq \mathcal{F}$ finite or countable subset

$$P(A_i) = 1 \quad \forall i \Rightarrow P(\bigcap_i A_i) = 1$$

$$7) A \subseteq B \Rightarrow P(B \setminus A) = P(B) - P(A) \Rightarrow P(A) \leq P(B)$$

$$8) A_1 \subseteq A_2 \subseteq \dots \in \mathcal{F} \Rightarrow \lim_{n \rightarrow \infty} P(A_n) = P(\bigcup_{n=1}^{\infty} A_n)$$

$$\{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F} \Rightarrow \lim_{n \rightarrow \infty} P(\bigcap_{i=1}^n A_i) = P(\bigcap_{i \in \mathbb{N}} A_i)$$

} Continuity of probability.

$$A_1 \supseteq A_2 \supseteq \dots \in \mathcal{F} \Rightarrow \lim_{n \rightarrow \infty} P(A_n) = P(\bigcap_{n=1}^{\infty} A_n)$$

$$A_1, A_2, \dots \in \mathcal{F} \Rightarrow \lim_{n \rightarrow \infty} P(\bigcap_{i=1}^n A_i) = P(\bigcap_{i=1}^{\infty} A_i)$$

* Under the assumption that for $A, B \in \mathcal{F}$ disjoint, $P(A \cup B) = P(A) + P(B)$ we have that 8) $\Leftrightarrow \sigma$ -additivity. $(P(\bigcup_{i \in \mathbb{N}} A_i) = \sum_{i=1}^{\infty} P(A_i))$

* $\{A_i\}_{i \in \mathbb{N}}$ independent if $\forall n \geq 2, i_1 < i_2 < \dots < i_n$

$$P(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_n}) = P(A_{i_1}) \cdot \dots \cdot P(A_{i_n})$$

Corollary \Rightarrow 1) $\{B_i\}_{i \in \mathbb{N}}$ are ind + $\forall B_1, B_2, \dots \Rightarrow B_i \in \{\emptyset, A_i, A_i^c, \Omega\}$

$A \perp B$	$A^c \perp B^c$
$A^c \perp B$	$A \perp \emptyset$
$A \perp B^c$	$A \perp \Omega$

2) It is possible that $A_{i_1}, \dots, A_{i_{n-1}}$ are ind + $\{i_1, \dots, i_{n-1}\} \subseteq \{1, \dots, n\}$ but $\{A_{i_1}, \dots, A_{i_n}\}$ are dependent

3) $P(\bigcap_{i=1}^n A_i) = \prod_{i=1}^n P(A_i) \quad \forall n \geq 2 \not\Rightarrow$ independence for the whole independence

$$A_1 = \{1, 2, 3, 4\}$$

$$A_2 = \{3, 4, 5, 6\}$$

$$A_3 = \{4, 6, 7, 8\}$$

$$\Omega = \{1, \dots, 8\}; P(\{3, 4\}) = 1/8$$

Clearly not independent

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Exercise $\rightarrow P(B_i) = \prod_{i=1}^n P(B_i) \quad \forall B_i \in \{\emptyset, A_i, A_i^c, \Omega\}$
 $\Rightarrow A_1, A_2, \dots$ indep

- \rightarrow 5) A, B disjoint and independent $\Rightarrow P(A) = 0$ or $P(B) = 0$.
- 6) $P(A) \in (0, 1) \Rightarrow A \perp B \quad \forall B \in \mathcal{F}$
- 7) $A_i \perp_{i \in \mathbb{N}}$ indep. $P(\bigcap_{i=1}^{\infty} A_i) = \prod_{i=1}^{\infty} P(A_i)$.

Borel-Cantelli Lemma $\rightarrow \{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F}$

- $\cdot \sum_{i \in \mathbb{N}} P(A_i) < \infty \Rightarrow P(\bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} A_m) = 0$.
 $\uparrow A_n$ i.o. \leftarrow infinitely often
- $\cdot \sum_{i \in \mathbb{N}} P(A_i) = 1 \Rightarrow P(A_n \text{ i.o.}) = 1$.

Borel sets of \mathbb{R} \rightarrow smallest σ -algebra that contains $(a, b) \quad \forall a, b \in \mathbb{R}$.

Product σ -algebra $\rightarrow \mathcal{F}_1 \otimes \mathcal{F}_2 = \sigma(\{A \times B \mid A \in \mathcal{F}_1, B \in \mathcal{F}_2\})$

★ Measure

(Ω, \mathcal{F}) measurable space

* $\mu : \mathcal{F} \rightarrow \mathbb{R}^+ \cup \{0, \infty\}$.

$\{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F}$ disjoint $\Rightarrow \mu(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \mu(A_i)$
 $\mu(\emptyset) = 0$.

* A measure is called σ -finite if

- 1) $\exists \{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F}$ mutually disjoint $\rightarrow \bigcup_{i \in \mathbb{N}} A_i = \Omega$
 $\rightarrow \mu(A_i) < \infty \quad \forall i \in \mathbb{N}$.
- 2) $\exists A_1 \subseteq A_2 \subseteq \dots \in \mathcal{F}, \bigcup_{i \in \mathbb{N}} A_i = \Omega \rightarrow \mu(A_i) < \infty \quad \forall i \in \mathbb{N}$

Remark $\rightarrow \mu$ σ -finite, $\{A_i\}_{i \in \mathbb{N}} \subseteq \mathcal{F}$ mutually disjoint with $\bigcup_{i=1}^{\infty} A_i = \Omega, \mu(A_i) < \infty$

* $\lambda_i = \mu(A_i), \mu(A) = \sum_{i=1}^{\infty} \lambda_i P_i(A_i)$

\rightarrow Any σ -finite measure is a non-negative linear combination of probability measures.

$\hookrightarrow 0 < \mu(\Omega) < \infty \Rightarrow P(A) = \frac{\mu(A)}{\mu(\Omega)} \quad \lambda = \mu(\Omega) \Rightarrow \mu(A) = \lambda P(A)$

④

Lebesgue measure \rightarrow it's a measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$
 $\mathcal{B}(\mathbb{R})$ = Borel of \mathbb{R}

that satisfies $\ell(a, b) = b - a$.

$$* \ell(A) = 0 \iff \exists \{a_i, b_i\}_{i \in \mathbb{N}} \text{ s.t. } \bigcup_{i \in \mathbb{N}} (a_i, b_i) \supseteq A$$

$$a_i < b_i \quad \sum_{i=1}^{\infty} (b_i - a_i) < \epsilon$$

$$* \ell \text{ is } \sigma\text{-finite. } \leftarrow \ell((n-1, n]) = 1$$

$$\bigcup_{n \in \mathbb{Z}} (n-1, n] = \mathbb{R}$$

Product measures $\rightarrow (\Omega_1, \mathcal{F}_1, \mu_1), (\Omega_2, \mathcal{F}_2, \mu_2)$ σ -finite measure spaces.

$\mu_1 \otimes \mu_2$ is called the product measure and it is the unique

$$\text{measure on } \mathcal{F}_1 \otimes \mathcal{F}_2 \rightarrow \mu_1 \otimes \mu_2 (A_1 \times A_2) = \mu_1(A_1) \mu_2(A_2)$$

$$\forall A_1 \in \mathcal{F}_1, A_2 \in \mathcal{F}_2$$

$$* \mathbb{E}(X) = \mathbb{E}\left(\int_0^{\infty} dt\right) = \mathbb{E}\left(\int_{\mathbb{R}^+} \mathbb{I}_{[0, X]}(t) dt\right) \stackrel{?}{=} \int_{\mathbb{R}^+} \mathbb{E}(\mathbb{I}_{[0, X]}(t)) dt$$

$$X > 0$$

$$\uparrow \mathbb{I}_{\{X > t\}}$$

$$= \int_0^{\infty} \mathbb{P}(X > t) dt$$

Martes 26 enero

Notation $\rightarrow \bar{\mathbb{R}} = \mathbb{R} \cup \{-\infty, \infty\}$

$X: \Omega \rightarrow \bar{\mathbb{R}}$ \leftarrow (extended) random variable.

$\hookrightarrow X$ is a random variable if $\forall t \in \bar{\mathbb{R}}$

$$\{X \leq t\} = \{\omega : X(\omega) \leq t\} = X^{-1}([-\infty, t]) \in \mathcal{F}$$

\uparrow we want measurable events.

Theorem $\rightarrow X$ is a r.v.

$\Rightarrow \forall B$ Borel subset of \mathbb{R} , we have that

$$\{X \in B\} = \{\omega : X(\omega) \in B\} = X^{-1}(B) \in \mathcal{F}$$

$$X^{-1}(B^c) = (X^{-1}(B))^c$$

$$X^{-1}\left(\bigcup_{\alpha} B_{\alpha}\right) = \bigcup_{\alpha} X^{-1}(B_{\alpha})$$

$$X^{-1}\left(\bigcap_{\alpha} B_{\alpha}\right) = \bigcap_{\alpha} X^{-1}(B_{\alpha})$$

$$* (\Omega_1, \mathcal{F}_1) \xrightarrow{X} (\Omega_2, \mathcal{F}_2) \xrightarrow{Y} (\Omega_3, \mathcal{F}_3)$$

$X^{-1}(A_2) \in \mathcal{F}_1 \quad \forall A_2 \in \mathcal{F}_2 \Rightarrow X$ is a measurable transformation.

So is Y if $Y^{-1}(A_3) \in \mathcal{F}_2 \quad \forall A_3 \in \mathcal{F}_3$.

⑤. * $f: \mathbb{R} \rightarrow \mathbb{R}$ is a Borel function if

$$(\mathbb{R}, \mathcal{B}(\mathbb{R})) \xrightarrow{f} (\mathbb{R}, \mathcal{B}(\mathbb{R})) \text{ measurable.}$$

* A random variable is a measurable transformation $(\Omega, \mathcal{F}) \xrightarrow{X} (\mathbb{R}, \mathcal{B}(\mathbb{R}))$.

Corollary \rightarrow If $\{X_n\}_{n \in \mathbb{N}}$ is a sequence of r.v.

\Rightarrow so are $\sup \{X_n\}_{n \in \mathbb{N}}$, $\inf \{X_n\}_{n \in \mathbb{N}}$, $\limsup_{n \rightarrow \infty} X_n$, $\liminf_{n \rightarrow \infty} X_n$
and, if the limit \exists , then so is $\lim_{n \rightarrow \infty} X_n$.

$$\lim_{n \rightarrow \infty} X_n = \inf_{n \geq 1} \sup_{m \geq n} X_m$$

\rightarrow f Borel $\Rightarrow f(X)$ r.v.
 X r.v.

More generally, if $f: (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n)) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$ Borel

X_1, \dots, X_n r.v.

$\Rightarrow f(X_1, \dots, X_n)$ r.v.

$\mathcal{B}(\mathbb{R}^n) = \mathcal{B}(\mathbb{R}) \otimes \dots \otimes \mathcal{B}(\mathbb{R})$ not by definition, it's a theorem.

* $E(\mathbb{I}_A) = P(A)$

$$E\left(\sum_{i=1}^n a_i \mathbb{I}_{A_i}\right) = \sum_{i=1}^n a_i P(A_i)$$

X	P
a_1	$P(A_1)$
a_2	$P(A_2)$
\vdots	\vdots
a_n	$P(A_n)$
0	$1 - \sum_{i=1}^n P(A_i)$

Simple (measurable) function

* $0 \leq X$, $X_n = \sum_{i=1}^{n2^n} \frac{i-1}{2^n} \mathbb{I}_{\{\frac{i-1}{2^n} < X < \frac{i}{2^n}\}} \leq X$

On $\{X \leq n2^n\}$, $0 \leq X - X_n \leq \frac{1}{2^n} \xrightarrow{n \rightarrow \infty} 0$

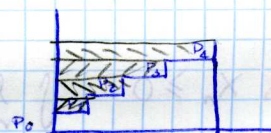
$X_n \leq X_{n+1}$

$$X_n \rightarrow X \quad E(X_n) = \sum_{i=1}^{n2^n} \frac{i-1}{2^n} P\left(\frac{i-1}{2^n} < X \leq \frac{i}{2^n}\right)$$

* $X \geq 0 \Rightarrow E(X) = \lim_{n \rightarrow \infty} E(X_n)$

$$E(X) = \int_0^\infty P(X > t) dt = \int_0^\infty (1 - F_X(t)) dt.$$

$$= \int_{\mathbb{R}^+} X(\omega) dP(\omega) = \int_{\mathbb{R}^+} x dF_X(x) = \int x dP_X(x)$$



⑥. * $X \leq 0$.

$$\begin{aligned} E(X) &= -E(-X) = -\int_{\mathbb{R}^+} P(-X > t) dt = -\int_{\mathbb{R}^-} P(X < t) dt = -\int_{\mathbb{R}^-} P(X < t) dt, \\ &= -\int_{(-\infty, 0]} F_X(t) dt. \end{aligned}$$

* If $E(X^+) = \int X \mathbb{I}_{\{X > 0\}} dP = E(X \mathbb{I}_{\{X > 0\}}) = \int_{\mathbb{R}^+} (1 - F_X(x)) dx = \int_{[0, \infty)} x dF_X(x) < \infty$
 or $E(X^-) = \int X \mathbb{I}_{\{X < 0\}} dP = E(X \mathbb{I}_{\{X < 0\}}) = -\int_{(-\infty, 0]} F_X(x) dx = \int_{(-\infty, 0]} x dF_X(x) > -\infty$.

\Rightarrow we say that $E(X)$ \exists (it could be ∞ or $-\infty$).

* $E(X) = \int_{\mathbb{R}} x dF_X(x) = \int x dP = -\int_{-\infty}^0 F_X(x) dx + \int_0^{\infty} (1 - F_X(x)) dx$.

* If μ σ -finite, $\mu(A) = \sum_{i=1}^{\infty} \lambda_i P_i(A)$

$$E(X) = \int X d\mu$$

if $X \geq 0 \Rightarrow \int X d\mu = \sum_{i=1}^{\infty} \lambda_i E_i(X)$

* if $\int X^+ d\mu < \infty$ or $\int (-X)^- d\mu < \infty$

$$\Rightarrow \int X d\mu = \int X^+ d\mu - \int (-X)^- d\mu$$

* $\mu_X(A) = \mu(X^{-1}(A))$

Properties of integration:

1. $\int_A d\mu = \int \mathbb{I}_A d\mu = \mu(A)$

2. $\int aX d\mu = a \int X d\mu$

3. $\int (X+Y) d\mu = \int X d\mu + \int Y d\mu$, provided that

not both are infinite with different signs.

4. $X \leq Y \Rightarrow \int X d\mu \leq \int Y d\mu$

5. $0 \leq X_n \uparrow X \Rightarrow \int X_n d\mu \xrightarrow{n \rightarrow \infty} \int X d\mu$

\uparrow Monotone convergence theorem

5'. $X_n \geq 0 \Rightarrow \int \sum_{n=1}^{\infty} X_n d\mu = \sum_{n=1}^{\infty} \int X_n d\mu$

6. $|X_n| \leq Y, \int Y d\mu < \infty$

$\mu\{X_n \not\rightarrow X\} = 0 \Rightarrow \int X_n d\mu \xrightarrow{n \rightarrow \infty} \int X d\mu$

\uparrow Dominated convergence theorem

7. $X_n \geq 0 \Rightarrow \int \liminf_{n \rightarrow \infty} X_n d\mu \leq \liminf_{n \rightarrow \infty} \int X_n d\mu$

\uparrow Fatou's Lemma

⑦. Theorem \rightarrow Fubini / Tonelli $\rightarrow f: (\Omega_1 \times \Omega_2, \mathcal{F}_1 \otimes \mathcal{F}_2) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$

$$(\Omega_i, \mathcal{F}_i, \mu_i), i \in \{1, 2\}.$$

$$f \geq 0 \Rightarrow \int f d\mu_1 \otimes \mu_2 = \int_{\Omega_2} \int_{\Omega_1} f(x, y) d\mu_1(x) d\mu_2(y) \\ = \int_{\Omega_1} \int_{\Omega_2} f(x, y) d\mu_2(y) d\mu_1(x)$$

$$\hookrightarrow E(X) = \int_{\mathbb{R}} \int_{\Omega} \mathbb{I}_{\{X(\omega) > t\}} dt dP(\omega) \\ = \int_{\mathbb{R}} \int_{\Omega} \mathbb{I}_{\{X(\omega) > t\}} dP(\omega) dt = \int_0^{\infty} P(X > x) dx.$$

\uparrow We just have to see if $\mathbb{I}_{\{X(\omega) > t\}}$ is jointly measurable.

$$\{(\omega, t) \mid X(\omega) > t\} = \bigcup_{q \in \mathbb{Q}} \underbrace{\{X(\omega) > q\}}_{\in \mathcal{F}} \times \underbrace{\{q > t\}}_{\in \mathcal{B}(\mathbb{R})} \in \mathcal{F} \otimes \mathcal{B}(\mathbb{R})$$

* For general f , if either $\int f^+ d\mu_1 \otimes \mu_2 < \infty$ or $\int (-f)^- d\mu_1 \otimes \mu_2 < \infty$.
 \Rightarrow the same holds.

$$* \int_{\Omega_2} f(x, y) d\mu_2(y) \quad \mathcal{F}_1\text{-measurable function of } x$$

$$\int_{\Omega_1} f(x, y) d\mu_1(x) \quad \mathcal{F}_2\text{-measurable function of } y.$$

Conditional expectation.

$$(\Omega, \mathcal{F}, P) \quad \mathcal{F}_0 \subseteq \mathcal{F}$$

$$Y = E(X | \mathcal{F}_0) \quad Y \text{ is } \mathcal{F}_0 \text{ measurable.}$$

$$Y \in \mathcal{F}_0$$

$$\{Y \leq t\} \in \mathcal{F}_0 \quad \forall t.$$

$$\forall A \in \mathcal{F}_0, E(Y \mathbb{I}_A) = E(X \mathbb{I}_A)$$

Theorem $\rightarrow E(X | \mathcal{F}_0)$ is well defined and is determined with probability 1.
 provided that $E(X) \exists$, ($E(X^+) < \infty$ or $E(X^-) > -\infty$)

* The conditional expected value has the same properties as ordinary expected value.

Def $\rightarrow \sigma(X)$: smallest σ -field for which X is measurable.

* $\{X^{-1}(B) \mid B \in \mathcal{B}(\mathbb{R})\}$ is already a σ -algebra.

* for every r.v. Z measurable respect to $\sigma(X)$, $\exists f$ Borel $\exists Z = f(X)$.

$$\hookrightarrow E(Y | X) = E(Y | \sigma(X)) = f(X).$$

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Another form of Borel-Cantelli:

$\hookrightarrow \mathbb{E} \left\{ \sum_{n=1}^{\infty} \mathbb{I}_{A_n} \right\} < \infty \Rightarrow \mathbb{P} \left\{ \sum_{n=1}^{\infty} \mathbb{I}_{A_n} < \infty \right\} = 1$

$= \infty \Rightarrow \mathbb{P} \{ \text{idem} \} = 0, A_1, A_2, \dots \text{ incl.}$

* $\{X_n\}_{n \in \mathbb{N}}$ independent if $\forall k \geq 2, \forall t_1, \dots, t_k,$

$\mathbb{P} \{ X_1 \leq t_1, \dots, X_k \leq t_k \} = \prod_{i=1}^k \mathbb{P} \{ X_i \leq t_i \}.$

2) $\forall k \in \mathbb{N}, B_1, \dots, B_k \in \mathcal{B}(\mathbb{R})$

$\mathbb{P} \{ X_1 \in B_1, \dots, X_k \in B_k \} = \prod_{i=1}^k \mathbb{P} \{ X_i \in B_i \}.$

3) $\forall k, \text{ Borel-functions } f_1, \dots, f_k.$

$\mathbb{E} \{ f_1(X_1) \dots f_k(X_k) \} = \prod_{i=1}^k \mathbb{E} \{ f_i(X_i) \}$

4) $\mathbb{E} \left\{ e^{i \left(\sum_{j=1}^k \theta_j X_j \right)} \right\} = \prod_{j=1}^k \mathbb{E} \left\{ e^{i \theta_j X_j} \right\}, \forall k, \theta_1, \dots, \theta_k \in \mathbb{R}.$

5) $X_k \geq 0, k=1, 2, \dots, \mathbb{E} \left\{ e^{-\sum_{k=1}^n \theta_k X_k} \right\} = \prod_{k=1}^n \mathbb{E} \left\{ e^{-\theta_k X_k} \right\}.$

o (End of review).

$N \sim \text{Poisson}(\lambda) \Rightarrow \mathbb{P}(N=n) = \frac{e^{-\lambda} \lambda^n}{n!}$

$\begin{pmatrix} X_1 \\ \vdots \\ X_n \end{pmatrix} \sim \text{Multinomial}(n, p_1, \dots, p_k) \Rightarrow \mathbb{P}(X_1=n_1, \dots, X_k=n_k) = \frac{(n_1 + \dots + n_k)!}{n_1! \dots n_k!} p_1^{n_1} \dots p_k^{n_k}$

* Assume $N \sim \text{Poisson}(\lambda)$

$\begin{pmatrix} Y_1 \\ \vdots \\ Y_k \end{pmatrix} | N=n \sim \text{Multinomial}(n, p_1, \dots, p_k)$

Question \rightarrow find the joint distribution of $\begin{pmatrix} Y_1 \\ \vdots \\ Y_k \end{pmatrix}$ (unconditional).

$\mathbb{P}(Y_1=n_1, \dots, Y_k=n_k) = \sum_{N \in \mathbb{N}^{(k)}} \underbrace{\mathbb{P} \{ Y_1=n_1, \dots, Y_k=n_k, N=n \}}_{= 0 \text{ if } n \neq n_1 + \dots + n_k}$

$= \sum_{N \in \mathbb{N}^{(k)}} \mathbb{P} \{ Y_1=n_1, \dots, Y_k=n_k | N=n \} \mathbb{P} \{ N=n \}$

$= \mathbb{P} \{ Y_1=n_1, \dots, Y_k=n_k | N = \sum_{i=1}^k n_i \} \mathbb{P} \{ N = \sum_{i=1}^k n_i \}$

$= \left\{ \frac{(n_1 + \dots + n_k)!}{n_1! \dots n_k!} p_1^{n_1} \dots p_k^{n_k} \right\} \frac{e^{-\lambda} \lambda^{\sum_{i=1}^k n_i}}{(\sum_{i=1}^k n_i)!}$

$\lambda = \sum_{i=1}^k p_i \lambda$
 $\lambda^{\sum_{i=1}^k n_i} = \prod_{i=1}^k \lambda^{n_i}$

$= \prod_{i=1}^k \frac{e^{-\lambda p_i} (\lambda p_i)^{n_i}}{n_i!} \leftarrow \text{producto de } k \text{ Poisson } (\lambda p_i) \text{ indep.}$

$\Rightarrow Y_i \sim \text{Poisson}(\lambda p_i), Y_1, \dots, Y_k \text{ indep.}$

④

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* $X \sim \text{Poisson}(\lambda)$, $Y \sim \text{Poisson}(\mu)$, $X \perp Y$

$\Rightarrow X + Y \sim \text{Poisson}(\lambda + \mu)$

* By induction, if $\{N_i\}_{i=1}^n \sim \text{Poisson}(\lambda_i)$ ind

$\Rightarrow \sum_{i=1}^n N_i \sim \text{Poisson}(\sum_{i=1}^n \lambda_i)$

* Under the same condition, $N_1, \dots, N_k \mid \sum_{i=1}^k N_i = n \sim ?$

$$P\{N_1 = n_1, \dots, N_k = n_k \mid \sum_{i=1}^k N_i = n\} = \begin{cases} 0 & n \neq n_1 + \dots + n_k \\ \frac{P(N_1 = n_1, \dots, N_k = n_k)}{P(\sum_{i=1}^k N_i = n)} & n = n_1 + \dots + n_k \end{cases}$$

$$\frac{P(N_1 = n_1, \dots, N_k = n_k)}{P(\sum_{i=1}^k N_i = \sum_{i=1}^k n_i)} = \frac{\prod_{i=1}^k \frac{e^{-\lambda_i} \lambda_i^{n_i}}{n_i!}}{e^{-\sum_{i=1}^k \lambda_i} \frac{(\sum_{i=1}^k \lambda_i)^{\sum_{i=1}^k n_i}}{(\sum_{i=1}^k n_i)!}}$$

$$= \frac{(n_1 + \dots + n_k)!}{n_1! \dots n_k!} \prod_{i=1}^k \left(\frac{\lambda_i}{\lambda_1 + \dots + \lambda_k}\right)^{n_i}$$

$\therefore N_1, \dots, N_k \mid \sum_{i=1}^k N_i = n \sim \text{Multinomial}(n, p_1, \dots, p_k)$; $p_i = \frac{\lambda_i}{\lambda_1 + \dots + \lambda_k}$

* $Y \sim \text{Poisson}(0)$ ← what will happen?



$N \sim \text{Poisson}(\lambda)$; $\lambda \in \mathbb{R}^+$

Choose N balls and throw them in the k urns.

$\therefore Y \sim \text{Poisson}(0) \Leftrightarrow P(Y=0) = 1$

* $N \sim \text{Poisson}(\lambda) \Rightarrow E(N) = \text{Var}(N) = \lambda$

$$E(z^N) = \sum_{n=0}^{\infty} e^{-\lambda} \frac{\lambda^n}{n!} z^n = e^{-\lambda} \sum_{n=0}^{\infty} \frac{(\lambda z)^n}{n!} = e^{-\lambda(1-z)}$$

* $N_\lambda \sim \text{Poisson}(\lambda)$

$$\Rightarrow \lim_{\lambda \rightarrow \infty} P\{N_\lambda \leq n\} = \lim_{\lambda \rightarrow \infty} \sum_{i=0}^n e^{-\lambda} \frac{\lambda^i}{i!} = 0$$

$$\Rightarrow \lim_{\lambda \rightarrow \infty} P\{N_\lambda > n\} = 1$$

* $\sum_{i=1}^{\infty} \lambda_i = \infty$, $N_i \sim \text{Poisson}(\lambda_i)$

$\Rightarrow \sum_{i=1}^{\infty} N_i \sim \text{Poisson}(\sum_{i=1}^{\infty} \lambda_i)$, $\sum_{i=1}^k N_i \uparrow \sum_{i=1}^{\infty} N_i$

$$P\{\sum_{i=1}^{\infty} N_i > n\} = \lim_{k \rightarrow \infty} P\{\sum_{i=1}^k N_i > n\} = 1$$

$$A_k = \{\sum_{i=1}^k N_i > n\}; A_k \subseteq A_{k+1}$$

$$\bigcup_{k=1}^{\infty} A_k = \{\sum_{i=1}^{\infty} N_i > n\} \Rightarrow \lim_{k \rightarrow \infty} P(A_k) = P(\bigcup_{k=1}^{\infty} A_k)$$

(10)

$$* P \left\{ \sum_{i=1}^{\infty} N_i > n \right\} = 1 \quad \forall n$$

$$\Rightarrow P \left(\bigcap_{n=0}^{\infty} \left\{ \sum_{i=1}^{\infty} N_i > n \right\} \right) = 1.$$

$$= P \left(\sum_{i=1}^{\infty} N_i = \infty \right)$$

$$\because N_1, N_2, \dots \text{ ind, } N_i \sim \text{Poisson}(\lambda_i), \quad \sum_{i=1}^{\infty} \lambda_i = \infty$$

$$\Rightarrow P \left(\sum_{i=1}^{\infty} N_i = \infty \right) = 1$$

* For this reason, we define $N \sim \text{Poisson}(\infty)$ if $P(N = \infty) = 1$.

* If $\sum_{i=1}^{\infty} \lambda_i < \infty$

$$\Rightarrow P \left\{ \sum_{i=1}^k N_i = n \right\} = e^{-\sum_{i=1}^k \lambda_i} \frac{\left(\sum_{i=1}^k \lambda_i \right)^n}{n!} \xrightarrow{k \rightarrow \infty} e^{-\sum_{i=1}^{\infty} \lambda_i} \frac{\left(\sum_{i=1}^{\infty} \lambda_i \right)^n}{n!}$$

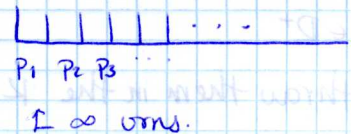
Conclusion

$$n \sim \text{Poisson}(\lambda_n); \quad \lambda_n \xrightarrow{n \rightarrow \infty} \lambda$$

$$\Rightarrow Y_n \xrightarrow{d} Y \sim \text{Poisson}(\lambda)$$

Moreover, if N_1, N_2, \dots iid con $N_i \sim \text{Poisson}(\lambda_i)$

$$\Rightarrow \sum_{i=1}^{\infty} N_i \sim \text{Poisson} \left(\sum_{i=1}^{\infty} \lambda_i \right)$$

*  $\sum_{i=1}^{\infty} p_i = 1$

Choose N balls, where $N \sim \text{Poisson}(\lambda)$, $\lambda \in \mathbb{R}^+$

Throw them independently in the urns, where the probability of falling in urn i is p_i .

$$\Rightarrow P(Y_1 = n_1, Y_2 = n_2, \dots | N = n) = \frac{n!}{\prod_{i=1}^{\infty} n_i} \prod_{i=1}^{\infty} p_i^{n_i}$$

For this purpose, we define $0^0 = 1$.

$$\Rightarrow Y_1, Y_2, \dots \text{ ind r.v. with } Y_i \sim \text{Poisson}(\lambda p_i)$$

$\uparrow Y_1, \dots, Y_k$ ind $\forall k \geq 2$?

$$Y_1, \dots, Y_k, \sum_{i=k+1}^{\infty} Y_i | N = n \sim \text{Multinomial} \left(n, p_1, \dots, p_k, \sum_{i=k+1}^{\infty} p_i \right)$$

$$\sim \text{Multinomial} \left(n, p_1, \dots, p_k, 1 - \sum_{i=1}^k p_i \right)$$

$\uparrow k+1$ urns.

$$\Rightarrow Y_1, \dots, Y_k, \sum_{i=k+1}^{\infty} Y_i \text{ ind, } Y_i \sim \text{Poisson}(\lambda p_i)$$