

Classes of Sets

A collection  $\mathcal{F}$  of subsets of  $\Omega \neq \emptyset$  is an **algebra** if

- 1.  $\Omega \in \mathcal{F}$ ,
- 2.  $A \in \mathcal{F} \implies A^c \in \mathcal{F}$ ,
- 3.  $A, B \in \mathcal{F} \implies A \cup B \in \mathcal{F}$  (equivalently,  $A \cap B \in \mathcal{F}$ ).

A collection  $\mathcal{F}$  of subsets of  $\Omega$  is a  **$\sigma$ -algebra** if

- 1.  $\mathcal{F}$  is an algebra,
- 2.  $A_1, A_2, \cdots \in \mathcal{F} \implies \bigcup_{n=1}^\infty A_n \in \mathcal{F}$ .

If  $\mathcal{F}_\theta$ ,  $\theta \in \Theta$  is a collection of  $\sigma$ -algebras on  $\Omega$ , then  $\mathcal{G} = \bigcap_{\theta \in \Theta} \mathcal{F}_\theta$  is a  $\sigma$ -algebra but  $\bigcup_{\theta \in \Theta} \mathcal{F}_\theta$  may not be.

If  $\mathcal{A}$  is a collection of subsets of  $\Omega$ , then the  $\sigma$ -algebra generated by  $\mathcal{A}$  is

$$\sigma(\mathcal{A}) = \bigcap_{\substack{\mathcal{F} \text{ is } \sigma\text{-algebra} \\ \mathcal{A} \subset \mathcal{F}}} \mathcal{F}.$$

A class  $\mathcal{C}$  of subsets of  $\Omega$  is a  **$\pi$ -class** if

- 1.  $A, B \in \mathcal{C} \implies A \cap B \in \mathcal{C}$ .

A class  $\mathcal{L}$  of subsets of  $\Omega$  is a  **$\lambda$ -class** if

- 1.  $\Omega \in \mathcal{L}$ ,
- 2.  $A \in \mathcal{L} \implies A^c \in \mathcal{L}$ ,
- 3.  $A_1, A_2, \cdots \in \mathcal{L}$  **disjoint**  $\implies \bigcup_{n=1}^\infty A_n \in \mathcal{L}$ .

$\lambda$ -class  $\not\implies \sigma$ -algebra.  
 $\pi$ -class and  $\lambda$ -class  $\implies \sigma$ -algebra (but not converse).

**Dynkin’s  $\pi - \lambda$  Theorem:** On a set  $\Omega$ , if  $\mathcal{C}$  is a  $\pi$ -class and  $\mathcal{L}$  is a  $\lambda$ -class such that  $\mathcal{C} \subset \mathcal{L}$ , then  $\sigma(\mathcal{C}) \subset \mathcal{L}$ .

Product Spaces

If  $\mathcal{F}$  is a  $\sigma$ -algebra on a nonempty set  $\Omega$ , then  $(\Omega, \mathcal{F})$  is a **measurable space**. If  $(\Omega_i, \mathcal{F}_i)$ ,  $i = 1, 2, \dots, n$  are measurable spaces, the  $n$ -dimensional **product space** is

$$\prod_{i=1}^n \Omega_i = \{(\omega_1, \omega_2, \dots, \omega_n) : \omega_i \in \Omega_i, \ 1 \leq i \leq n\}.$$

If  $A_i \subset \Omega_i$ , the  $n$ -dimensional **rectangle** is

$$\prod_{i=1}^n A_i = \{(\omega_1, \omega_2, \dots, \omega_n) : \omega_i \in A_i, \ 1 \leq i \leq n\}.$$

The  $n$ -dimensional **product  $\sigma$ -algebra** is

$$\prod_{i=1}^n \mathcal{F}_i = \sigma \left\langle \left\{ \prod_{i=1}^n A_i : A_i \in \mathcal{F}_i, \ 1 \leq i \leq n \right\} \right\rangle.$$

Measures

An extended real-valued function on a class of subsets of a nonempty set  $\Omega$  is a **set function**. A set function  $\mu$  on an algebra  $\mathcal{F}$  on  $\Omega$  is a **measure** if

- 1.  $\mu(A) \in [0, \infty]$ ,  $\forall A \in \mathcal{F}$ ,
- 2.  $\mu(\emptyset) = 0$ ,
- 3.  $A_1, A_2, \cdots \in \mathcal{L}$  **disjoint** with  $\bigcup_{n=1}^\infty A_n \in \mathcal{F}$ , then

$$\mu \left( \bigcup_{n=1}^\infty A_n \right) = \sum_{n=1}^\infty \mu(A_n).$$

If  $\mu(\Omega) = 1$ , then  $\mu$  is a **probability measure**. If  $\exists B_1, B_2 \in \mathcal{F}$  such that  $\Omega = \bigcup_{n=1}^\infty B_n$  and  $\mu(B_i) < \infty, \ \forall i \geq 1$ , then  $\mu$  is  **$\sigma$ -finite**.

**Extensions of Measures:** If  $\mu$  is a  $\sigma$ -finite measure on an algebra  $\mathcal{F}$ , then there is a *unique* extension of  $\mu$  to  $\sigma(\mathcal{F})$ .

**Monotonicity:** If  $A, B \in \mathcal{F}$  such that  $A \subset B$ , then  $P(A) \leq P(B)$ .

**Inclusion-exclusion formula:** If  $A_1, A_2, \dots, A_n \in \mathcal{F}$ , then

$$P \left( \bigcup_{i=1}^n A_i \right) = \sum_{i=1}^n P(A_i) - \sum_{1 \leq i < j \leq n} P(A_i \cap A_j) + \cdots + (-1)^{n-1} P(A_1 \cap \cdots \cap A_n).$$

**Countable subadditivity:** If  $A_1, A_2, \cdots \in \mathcal{F}$  such that  $\bigcup_{n=1}^\infty A_n \in \mathcal{F}$ , then

$$P \left( \bigcup_{n=1}^\infty A_n \right) \leq \sum_{n=1}^\infty P(A_n).$$

**Monotone continuity from below** (mcfb): If  $A, A_1, A_2, \dots \in \mathcal{F}$  such that  $A_n \uparrow A$  (i.e.,  $A_n \subset A_{n+1}$  and  $A = \bigcup A_n$ ), then  $P(A_n) \uparrow P(A)$ .

**Monotone continuity from above** (mcfa): If  $A, A_1, A_2, \dots \in \mathcal{F}$  such that  $A_n \downarrow A$  (i.e.,  $A_n \supset A_{n+1}$  and  $A = \bigcap A_n$ ), then  $P(A_n) \downarrow P(A)$ .

Measurable Transformations

If  $(\Omega_1, \mathcal{F}_1), (\Omega_2, \mathcal{F}_2)$  are measurable spaces, then  $T : \Omega_1 \rightarrow \Omega_2$  is  **$\langle \mathcal{F}_1, \mathcal{F}_2 \rangle$ -measurable** if

$$T^{-1}(A) \equiv \{\omega \in \Omega_1 : T(\omega) \in A\} \in \mathcal{F}_1 \text{ (equivalently, } T^{-1}(\mathcal{F}_2) \subset \mathcal{F}_1).$$

Let  $(\Omega, \mathcal{F}, P)$  be a psp. Then,  $X : \Omega \rightarrow \mathbb{R}$  is a **r.v.** if it is  **$\langle \mathcal{F}, \mathcal{B}(\mathbb{R}) \rangle$ -measurable**.

Checking for measurability:  $T : \mathbb{R} \rightarrow \mathbb{R}$  is  **$\langle \mathcal{B}(\mathbb{R}), \mathcal{B}(\mathbb{R}) \rangle$ -measurable** iff  $T^{-1}(-\infty, r) = \{\omega \in \mathbb{R} : T(\omega) < r\} \in \mathcal{B}(\mathbb{R}) \ \forall r \in \mathbb{R}$ .

If  $T_i$  is  **$\langle \mathcal{F}_i, \mathcal{F}_{i+1} \rangle$ -measurable** for  $i = 1, 2$ , then the composition  $T = T_1 \circ T_2 = T_1(T_2(\cdot))$  is  **$\langle \mathcal{F}_1, \mathcal{F}_3 \rangle$ -measurable**.

If  $f : \mathbb{R}^k \rightarrow \mathbb{R}^p$  is continuous, then  $f$  is  **$\langle \mathcal{B}(\mathbb{R}), \mathcal{B}(\mathbb{R}) \rangle$ -measurable**.

If  $f_1, f_2, \dots, f_n$  are each  **$\langle \mathcal{F}, \mathcal{B}(\mathbb{R}) \rangle$ -measurable** transformations from  $\Omega$  to  $\mathbb{R}$ , then  $f = (f_1, f_2, \dots, f_n)'$  is  **$\langle \mathcal{F}, \mathcal{B}(\mathbb{R}) \rangle$ -measurable**. Also,  $\sum_{i=1}^n f_i, \prod_{i=1}^n f_i, \sup_n f_n, \inf_n f_n, \limsup f_n, \liminf f_n$ , and  $\mathbb{I}_{\{\omega \in \Omega : \lim_{n \rightarrow \infty} f_n(\omega) \text{ finitely exists} \}} \lim_{n \rightarrow \infty} f_n$  are all  **$\langle \mathcal{F}, \mathcal{B}(\mathbb{R}) \rangle$ -measurable**.

Induced Measures & Distribution Functions

If  $(\Omega_1, \mathcal{F}_1), (\Omega_2, \mathcal{F}_2)$  are measurable spaces and  $T : \Omega_1 \rightarrow \Omega_2$  is  **$\langle \mathcal{F}_1, \mathcal{F}_2 \rangle$ -measurable**, then for any measure  $\mu$  on  $(\Omega_1, \mathcal{F}_1)$ , the set function  $\mu T^{-1}$  defined by

$$\mu T^{-1}(A) = \mu \circ T^{-1}(A) = \mu(T^{-1}(A)), \quad \forall A \in \mathcal{F}_2$$

is a measure on  $\mathcal{F}_2$  and is called the **measure induced** by  $T$  under  $\mu$  on  $\mathcal{F}_2$ .

For a r.v.  $X$  on a psp  $(\Omega, \mathcal{F}, P)$ , the probability **distribution** of  $X$  (or the law of  $X$ ), denoted  $P_X$ , is the induced measure of  $X$  under  $P$  on  $\mathcal{B}(\mathbb{R})$ . That is,

$$P_X(A) = P(X^{-1}(A)) = P(\{\omega \in \Omega : X(\omega) \in A\}) = P(X \in A), \quad A \in \mathcal{B}(\mathbb{R}).$$

The **cumulative distribution function** (cdf) of a r.v.  $X$  is

$$F(x) = P_X((-\infty, x]) = P(X \leq x), \quad x \in \mathbb{R}.$$

If  $F$  is the cdf of a r.v.  $X$ , then

- 1.  $F$  is right continuous: if  $x_n \downarrow x_0$  and  $x_n \geq x$  then  $F(x_n) = P_X((-\infty, x_n]) \downarrow P_X((-\infty, x_0]) = F(x_0)$  by mcfa,
- 2.  $F$  is monotone nondecreasing: if  $x \leq y \implies (-\infty, x] \subset (-\infty, y]$  then  $F(x) = P_X((-\infty, x]) \leq P_X((-\infty, y]) = F(y)$  by monotonicity,
- 3.  $\lim_{x \rightarrow \infty} F(x) = 1$  and  $\lim_{x \rightarrow -\infty} F(x) = 0$ : show using argument similar to (1).

If  $F : \mathbb{R} \rightarrow [0, 1]$  satisfies (1)-(3) above, then there exists a r.v.  $X$  on a psp  $(\Omega, \mathcal{F}, P)$  such that  $F$  is the cdf of  $X$ .

Integrals

Any extended real-valued function  $f$  defined on  $\Omega$  can be decomposed into positive  $f^+ = \mathbb{I}_{f \geq 0} f$  and negative  $f^- = -\mathbb{I}_{f < 0} f$  parts. Then,  $f = f^+ - f^-$  and  $|f| = f^+ + f^-$ .

A measurable function  $f : \Omega \rightarrow \overline{\mathbb{R}}$  on  $(\Omega, \mathcal{F}, \mu)$  is  **$\mu$ -integrable** if  $\int_\Omega |f| d\mu < \infty$ . If  $\mu(|f| = \infty) > 0$ , then  $\int_\Omega |f| d\mu \equiv \infty$ . Note  $|f|$  is  $\mu$ -integrable iff both  $f^+$  and  $f^-$  are  $\mu$ -integrable.

If at least one of  $\int_\Omega f^+ d\mu, \int_\Omega f^- d\mu$  is  $< \infty$ , then  $\int_\Omega f d\mu \equiv \int_\Omega f^+ d\mu - \int_\Omega f^- d\mu$  and we say  $\int_\Omega f d\mu$  exists.

If  $f : \Omega \rightarrow \overline{\mathbb{R}}$  is nonnegative and  $\mu$ -integrable, then  $f < \infty$  a.e. $(\mu)$ . If  $f : \Omega \rightarrow \overline{\mathbb{R}}$  is nonnegative, then  $\int_\Omega f d\mu = 0$  iff  $f = 0$  a.e. $(\mu)$ .

If  $(\Omega, \mathcal{F}, \mu)$  is a msp and  $f : \Omega \rightarrow \mathbb{R}$  is measurable and either  $\mu$ -integrable or nonnegative. Then, for any  $A \in \mathcal{F}$ ,

$$\int_A f d\mu = \int_\Omega \mathbb{I}_A f d\mu.$$

Using the DCT (see next section), it can be shown that for **disjoint**  $A_1, A_2, \dots \in \mathcal{F}$  and measurable  $f : \Omega \rightarrow \mathbb{R}$  either  $\mu$ -integrable or nonnegative, then

$$\int_\Omega f \mathbb{I}_{\bigcup_{n=1}^\infty A_n} d\mu = \sum_{n=1}^\infty \int_{A_i} f d\mu.$$

Convergence Theorems

**Monotone Convergence Theorem** (MCT): If  $f_n : \Omega \rightarrow \overline{\mathbb{R}}$  is an increasing sequence of nonnegative measurable functions, i.e.,  $f_n(\omega) \leq f_{n+1}(\omega) \ \forall \omega \in \Omega$  and  $f_n(\omega) \uparrow f(\omega)$  a.e. $(\mu)$ , then  $\int_\Omega f_n d\mu \uparrow \int_\Omega f d\mu$ . That is,

$$\int_\Omega f d\mu = \int_\Omega \lim_{n \rightarrow \infty} f_n d\mu = \lim_{n \rightarrow \infty} \int_\Omega f_n d\mu.$$

**Fatou’s Lemma:** If  $f_n : \Omega \rightarrow \overline{\mathbb{R}}$  is a sequence of nonnegative functions, then

$$\int_\Omega \liminf_{n \rightarrow \infty} f_n d\mu \leq \liminf_{n \rightarrow \infty} \int_\Omega f_n d\mu.$$

**Dominated Convergence Theorem** (DCT): Suppose (1)  $g : \Omega \rightarrow \overline{\mathbb{R}}$  is a nonnegative,  $\mu$ -integrable function; (2)  $|f_n| \leq g$  a.e. $(\mu) \ \forall n \geq 1$ ; and (3)  $f_n \rightarrow f$  a.e. $(\mu)$ . Then,  $f$  is  $\mu$ -integrable and

$$\lim_{n \rightarrow \infty} \int_\Omega |f_n - f| d\mu = 0 \text{ and } \lim_{n \rightarrow \infty} \int_\Omega f_n d\mu = \int_\Omega f d\mu.$$

*Result under weaker conditions, UI and  $\mu(\Omega) < \infty$ :* If  $(\Omega, \mathcal{F}, \mu)$  is a msp with  $\mu(\Omega) < \infty$  and  $f, f_n : \Omega \rightarrow \overline{\mathbb{R}}$  are measurable such that  $f_n \rightarrow f$  a.e. $(\mu)$  and  $\{f_n : n \geq 1\}$  is UI (see next section), then  $f$  is  $\mu$ -integrable and

$$\lim_{n \rightarrow \infty} \int_\Omega f_n d\mu = \int_\Omega f d\mu.$$

Note: for a nonnegative measurable real-valued function  $f$  on a msp  $(\Omega, \mathcal{F}, \mu)$ ,  $\nu(A) = \int_A f d\mu \ \forall A \in \mathcal{F}$  is a measure on  $(\Omega, \mathcal{F})$  and  $f$  is a density of the measure  $\nu$ .

**Scheffe’s Theorem:** Let  $(\Omega, \mathcal{F}, \mu)$  be a msp and for  $n \geq 0$ , let  $\nu_n(A) = \int_A f_n d\mu \ \forall A \in \mathcal{F}$  be finite measures on  $\mathcal{F}$  with densities  $f_n \geq 0$ . If  $\nu_n(\Omega) = \nu_0(\Omega) < \infty$  for all  $n \geq 1$  and  $f_n \rightarrow f$  a.e. $(\mu)$ , then

$$\lim_{n \rightarrow \infty} \int_\Omega |f_n - f_0| d\mu = 0.$$

Also, the “total variation norm” tends to zero:

$$\sup_{A \in \mathcal{F}} |\nu_n(A) - \nu_0(A)| = \frac{1}{2} \int_\Omega |f_n - f_0| d\mu \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Uniform Integrability

Recall if  $f : \Omega \rightarrow \mathbb{R}$  is  $\mu$ -integrable, then by the DCT

$$\lim_{n \rightarrow \infty} \int_{|f| > n} |f| d\mu = \lim_{n \rightarrow \infty} \int_\Omega \mathbb{I}_{|f| > n} |f| d\mu = 0.$$

A family of  $\mu$ -integrable functions  $\{f_\lambda : \lambda \in \Lambda\}$  on a msp  $(\Omega, \mathcal{F}, \mu)$  is **uniformly integrable** (UI) w.r.t.  $\mu$  if

$$\sup_{\lambda \in \Lambda} \int_{|f_\lambda| > t} |f_\lambda| d\mu \rightarrow 0 \text{ as } t \rightarrow \infty.$$

Suppose  $\mathcal{A} \equiv \{f_\lambda : \lambda \in \Lambda\}$  is a collection of  $\mu$ -integrable functions on a msp  $(\Omega, \mathcal{F}, \mu)$ . Then,

- if  $\Lambda$  is a finite set, then  $\mathcal{A}$  is UI,
- if  $\exists \varepsilon > 0$  such that  $\sup \left\{ \int |f_\lambda|^{1+\varepsilon} d\mu : \lambda \in \Lambda \right\} < \infty$ , then  $\mathcal{A}$  is UI,
- if  $|f_\lambda| \leq f$  a.e.  $(\mu)$  and  $\int f d\mu < \infty$ , then  $\mathcal{A}$  is UI,
- if  $\mathcal{A}$  is UI and  $\mu(\Omega) < \infty$ , then  $\exists M > 0$  such that  $\sup \left\{ \int |f_\lambda| d\mu : \lambda \in \Lambda \right\} \leq M$ ,
- if  $\{f_\lambda : \lambda \in \Lambda\}$  and  $\{g_\lambda : \lambda \in \Lambda\}$  are both UI, then  $\{f_\lambda + g_\lambda : \lambda \in \Lambda\}$  is also UI.

## Independence

Let  $(\Omega, \mathcal{F}, P)$  be a psp and  $I$  be a set of indices.

- A collection  $A_i$ ,  $i \in I$  of sets in  $\mathcal{F}$  are **ind** if  $\forall i_1, i_2, \dots, i_n \in I$  distinct indices and fixed  $1 \leq n < \infty$ ,

$$P\left(\bigcap_{j=1}^n A_{i_j}\right) = \prod_{j=1}^n P(A_{i_j}).$$

Note the above has  $2^n - n - 1$  independence conditions!

- Suppose  $\mathcal{G}_i \subset \mathcal{F}$  is a collection of measurable sets for each  $i \in I$ . Then the family of sets  $\{\mathcal{G}_i : i \in I\}$  is called **ind** if any possible collection  $\{A_i : i \in I\}$  of sets are ind, where  $\{A_i : i \in I\}$  is formed by choosing an arbitrary set  $A_i$  from  $\mathcal{G}_i$  for each  $i \in I$ . That is,  $\forall i_1, i_2, \dots, i_n \in I$  distinct indices, fixed  $1 \leq n < \infty$  and  $\forall A_{i_1}, A_{i_2}, \dots, A_{i_n} \in \mathcal{G}_{i_n}$ ,

$$P\left(\bigcap_{j=1}^n A_{i_j}\right) = \prod_{j=1}^n P(A_{i_j}).$$

Note the family of sets  $\{\mathcal{G}_i : i \in I\}$  are ind iff for each finite  $T \subset I$ ,  $\{\mathcal{G}_i : i \in T\}$  are ind.

- A collection of r.v.'s  $X_i$ ,  $i \in I$  on  $(\Omega, \mathcal{F}, P)$  are **ind** if the family  $\{\sigma(X_i) : i \in I\}$  is ind, where

$$\sigma(X_i) = \left\{ X_i^{-1}(B) : B \in \mathcal{B}(\mathbb{R}) \right\} = X_i^{-1}(\mathcal{B}(\mathbb{R}))$$

is the  $\sigma$ -algebra generated by  $X_i$ . That is,  $\forall i_1, i_2, \dots, i_n \in I$  distinct indices, fixed  $1 \leq n < \infty$  and  $\forall B_{i_1}, B_{i_2}, \dots, B_{i_n} \in \mathcal{B}(\mathbb{R})$ ,

$$P(X_{i_1} \in B_{i_1}, X_{i_2} \in B_{i_2}, \dots, X_{i_n} \in B_{i_n}) = \prod_{j=1}^n P(X_{i_j} \in B_{i_j}).$$

In terms of distribution functions,  $X_i, i \in I$  are ind iff  $\forall x_1, x_2, \dots, x_n \in \mathbb{R}$  and  $\forall i_1, i_2, \dots, i_n \in I$  distinct indices,

$$P(X_{i_1} \leq x_1, X_{i_2} \leq x_2, \dots, X_{i_n} \leq x_n) = \prod_{j=1}^n P(X_{i_j} \leq x_j).$$

**Independence of generated  $\sigma$ -algebras:** If  $(\Omega, \mathcal{F}, P)$  is a psp,  $\mathcal{G}_i \subset \mathcal{F}$  is a  $\pi$ -class for each index  $i \in I$ , and the family  $\{\mathcal{G}_i : i \in I\}$  is ind. Then, the family  $\{\sigma(\mathcal{G}_i) : i \in I\}$  is ind.

## Borel-Cantelli Lemmas

If  $(\Omega, \mathcal{F})$  be a measurable space and  $A_1, A_2, \dots \in \mathcal{F}$ , then

$$\limsup_{n \rightarrow \infty} A_n = \overline{\lim} A_n = \bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} A_n \in \mathcal{F},$$

$$\liminf_{n \rightarrow \infty} A_n = \underline{\lim} A_n = \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} A_n \in \mathcal{F}.$$

Also,

$$\liminf_{n \rightarrow \infty} A_n \subset \limsup_{n \rightarrow \infty} A_n,$$

$$("A_n \text{ i.o.}")^c = (\overline{\lim} A_n)^c = \underline{\lim} A_n^c = "A_n^c \text{ eventually,}"$$

$$("A_n \text{ eventually}")^c = (\underline{\lim} A_n)^c = \overline{\lim} A_n^c = "A_n^c \text{ i.o.}"$$

**Borel-Cantelli Lemma:** Let  $(\Omega, \mathcal{F}, P)$  be a psp and  $A_1, A_2, \dots \in \mathcal{F}$ .

1. If  $\sum_n P(A_n) < \infty$ , then  $P(\overline{\lim} A_n) = 0$ .
2. If  $\{A_n\}$  are ind and  $\sum_n P(A_n) = \infty$ , then  $P(\overline{\lim} A_n) = 1$ .

**Borel 0-1 Law:** If  $A_1, A_2, \dots$  are ind events, then

$$P(\overline{\lim} A_n) = P(A_n \text{ i.o.}) = \begin{cases} 0 & \text{iff } \sum_n P(A_n) < \infty, \\ 1 & \text{iff } \sum_n P(A_n) = \infty. \end{cases}$$

## Tail Events & K's 0-1 Law

The **tail  $\sigma$ -algebra** of a sequence of r.v.'s  $X_1, X_2, \dots$  on a psp  $(\Omega, \mathcal{F}, P)$  is

$$\mathcal{T} = \bigcap_{n=1}^{\infty} \sigma(\{X_j : j \geq n\}),$$

where  $\sigma(\{X_j : j \geq n\}) = \sigma(\{X_j^{-1} : B \in \mathcal{B}(\mathbb{R}), j \geq n\})$  is the  $\sigma$ -algebra generated by  $X_j, j \geq n$ .

Any set (event)  $A \in \mathcal{T}$  is a **tail event**.

An extended real-valued r.v.  $T : \Omega \rightarrow \overline{\mathbb{R}} = \mathbb{R} \cup \{-\infty, \infty\}$  is a **tail r.v.** if  $T$  is  $\langle \mathcal{T}, \mathcal{B}(\overline{\mathbb{R}}) \rangle$ -measurable. That is,

$$\forall r \in \mathbb{R}, T^{-1}([-\infty, r)) = \{\omega \in \Omega : T(\omega) < r\} \in \mathcal{T}.$$

For fixed, arbitrary  $m \geq 1$ , a tail event or r.v. is only determined by  $X_n, n \geq m$  (so changing finitely man r.v.'s does not effect  $\mathcal{T}$ ).

Examples:

- $\overline{\lim} X_n, \underline{\lim} X_n$  are extended real-valued tail r.v.'s,
- $\{\lim X_n \text{ finitely exists}\}, \{\overline{\lim} X_n \geq r\}, \{\underline{\lim} X_n < r\}$  are tail events.

**Kolmogorov's 0-1 Law:** Tail events of a sequence  $X_1, X_2, \dots$  of ind r.v.'s have probabilities 0 or 1. That is, if  $\{X_n\}_{n \geq 1}$  are ind and  $A \in \mathcal{T} = \bigcap_n \sigma(\{X_j : j \geq n\})$ , then  $P(A) \in \{0, 1\}$ .

**Corollary:** For a psp  $(\Omega, \mathcal{F}, P)$  and tail  $\sigma$ -algebra  $\mathcal{T}$  defined by a sequence  $X_1, X_2, \dots$  of ind r.v.'s, if  $T : \Omega \rightarrow \overline{\mathbb{R}}$  is a tail r.v. (that is,  $T$  is  $\langle \mathcal{T}, \mathcal{B}(\overline{\mathbb{R}}) \rangle$ -measurable), then  $T$  is degenerate. That is,  $\exists c \in \overline{\mathbb{R}}$  such that  $P(T = c) = 1$ .

## Convergence of r.v.'s

A sequence of r.v.'s  $X_1, X_2, \dots$  on a psp  $(\Omega, \mathcal{F}, P)$  **converge almost surely** to a r.v.  $X_0$  on  $(\Omega, \mathcal{F}, P)$  if

$$P\left(\left\{\lim_{n \rightarrow \infty} X_n(\omega) = X_0(\omega)\right\}\right) = 1.$$

TFAE:

- $X_n \xrightarrow{\text{a.s.}} 0$ ,
- $P(|X_n| > \varepsilon \text{ i.o.}) = 0$  for all  $\varepsilon > 0$ ,
- $\sup_{j \geq n} |X_j - X_0| \xrightarrow{P} 0$  as  $n \rightarrow \infty$ ,
- $\lim_{n \rightarrow \infty} P\left(\bigcap_{j=n}^{\infty} [|X_j - X_0| \leq \varepsilon]\right) = 1$  for all  $\varepsilon > 0$ .

A sequence of r.v.'s  $X_1, X_2, \dots$  on a psp  $(\Omega, \mathcal{F}, P)$  **converge in probability** to a r.v.  $X_0$  on  $(\Omega, \mathcal{F}, P)$  if

$$\lim_{n \rightarrow \infty} P(|X_n - X_0| > \varepsilon) = 0, \quad \forall \varepsilon > 0.$$

TFAE:

- $X_n \xrightarrow{P} 0$ ,
- $\sup_{m \geq n} (|X_m - X_n| > \varepsilon) \rightarrow 0$  as  $n \rightarrow \infty$  for all  $\varepsilon > 0$ ,
- $\forall \{n_j\}$  of  $\{X_n\}$ ,  $\exists \{n_{j_k}\}$  such that  $X_{n_{j_k}} \xrightarrow{\text{a.s.}} X_0$ .

For a sequence of r.v.'s  $X_1, X_2, \dots$  on a psp  $(\Omega, \mathcal{F}, P)$ ,

1. if  $X_n \xrightarrow{\text{a.s.}} X_0$  and  $g : \mathbb{R} \rightarrow \mathbb{R}$  is continuous, then  $g(X_n) \xrightarrow{\text{a.s.}} g(X_0)$ ,
2. if  $X_n \xrightarrow{P} X_0$  and  $g : \mathbb{R} \rightarrow \mathbb{R}$  is continuous, then  $g(X_n) \xrightarrow{P} g(X_0)$ .

A sequence  $X_1, X_2, \dots$  of

$$\mathcal{L}_r(\Omega, \mathcal{F}, P) \equiv \left\{ \text{measurable } X \in \mathbb{R} : \int_{\Omega} |X|^r dP < \infty \right\}$$

functions **converges in  $\mathcal{L}_r$**  to a measurable function  $X$  if

$$\lim_{n \rightarrow \infty} \int_{\Omega} |X_n - X|^r dP = 0.$$

If  $X \in \mathcal{L}_r$ , then  $t^r P(|X| > t) \rightarrow 0$  as  $t \rightarrow \infty$ , that is,  $\uparrow r \implies$  faster convergence. If  $\exists p \in (0, \infty)$  such that  $t^p P(|X| > t) \rightarrow 0$ , then  $X \in \mathcal{L}_r \forall r \in (0, p)$ .

If the r.v.'s  $X_1, X_2, \dots \in \mathcal{L}_r$ , then  $\exists X \in \mathcal{L}_r$  such that  $X_n \xrightarrow{\mathcal{L}_r} X$  iff

$$\sup_{m \geq n} E|X_m - X_n|^r \rightarrow 0 \text{ as } n \rightarrow \infty.$$

For a r.v.  $X$ , the fixed, real-valued  $m(X)$  is a **median** if  $P(X \geq m(X)) \geq 1/2$  and  $P(X \leq m(X)) \geq 1/2$ . Can be defined as  $\inf \{x \in \mathbb{R} : P(\overline{X} \leq x) \geq 1/2\}$ . If  $P(|X| \geq c) < \varepsilon \leq 1/2$  then  $|m(X)| \leq c$ .

**Levy's Inequality:** If  $X_1, X_2, \dots, X_n$  are ind r.v.'s on a psp  $(\Omega, \mathcal{F}, P)$  and  $S_j = \sum_{i=1}^j X_i, 1 \leq j \leq n$ , then  $\forall \varepsilon > 0$ ,

1.  $P\left(\max_{1 \leq j \leq n} [S_j - m(S_j - S_n)] \geq \varepsilon\right) \leq 2P(S_n \geq \varepsilon)$ ,
2.  $P\left(\max_{1 \leq j \leq n} [S_j - m(S_j - S_n)] \geq \varepsilon\right) \leq 2P(|S_n| \geq \varepsilon)$ .

**Levy's Theorem:** If  $X_1, X_2, \dots$  are ind r.v.'s on a psp  $(\Omega, \mathcal{F}, P)$  and  $S_n = \sum_{i=1}^n X_j, n \geq 1$ , then  $S_n$  converges a.s.  $(P)$  iff  $S_n$  converges in probability.

**Khinchine-Kolmogorov Convergence Theorem:** If  $X_1, X_2, \dots$  are ind r.v.'s on a psp  $(\Omega, \mathcal{F}, P)$  with  $E(X_n) = 0$  and  $E(X_n^2) < \infty$  for all  $n \geq 1$  and  $\sum_n E(X_n^2) < \infty$ , then  $S_n = \sum_{i=j}^n X_j, n \geq 1$  converges a.s.  $(P)$  and in  $\mathcal{L}_2$  to some random variable  $S = \sum_n X_n$ . Also,  $E(S) = 0, E(S^2) = \sum_n E(X_n^2)$ .

**Corollary:** If  $X_1, X_2, \dots$  are ind r.v.'s on a psp  $(\Omega, \mathcal{F}, P)$  with  $\sum_n E(X_n) < \infty$  and  $\sum_n \sigma_{X_n}^2 < \infty$ , then  $S_n = \sum_{j=1}^n X_j, n \geq 1$  converges a.s.  $(P)$  to  $S = \sum_n X_n$ .

Two sequences of r.v.'s  $\{X_n\}$  and  $\{Y_n\}$  are **tail equivalent** if  $\sum_n P(X_n \neq Y_n) < \infty$ . If  $\{X_n\}$  and  $\{Y_n\}$  are tail equivalent, then

- By Borel-Cantelli,  $P(\lim(X_n \neq Y_n)) = 0 \implies P(X_n = Y_n \text{ for large } n) = 1$ ,

$$\bullet S_n = \sum_{j=1}^n X_j \xrightarrow{\text{a.s.}} S \iff S'_n = \sum_{j=1}^n Y_j \xrightarrow{\text{a.s.}} S',$$

- If  $b_n \rightarrow \infty$ , then

$$\frac{\sum_{j=1}^n X_j}{b_n} \xrightarrow{\text{a.s.}} 0 \iff \frac{\sum_{j=1}^n Y_j}{b_n} \xrightarrow{\text{a.s.}} 0.$$

**Berry-Esseen Lemma:** If  $X_1, X_2, \dots, X_n$  are ind r.v.'s with  $E(X_i) = 0$  and  $E|X_i|^3 < \infty, 1 \leq i \leq n$ , then  $\forall n \geq 4$ ,

$$\sup_{x \in \mathbb{R}} \left| P\left(\frac{S_n}{\sigma_n} \leq x\right) - \Phi(x) \right| \leq \frac{2.75}{\sigma_n^3} \sum_{i=1}^n E|X_i|^3,$$

where  $S_n = \sum_{j=1}^n X_j, \sigma_n^2 = \text{Var}(S_n) = \sum_{i=1}^n E(X_i^2)$ , and  $\Phi(\cdot)$  is the cdf of a  $\mathcal{N}(0, 1)$  r.v.

**Kolmogorov's 3-Series Theorem:** If  $X_1, X_2, \dots$  are ind r.v.'s on a psp  $(\Omega, \mathcal{F}, P)$ , for fixed  $c > 0$ , define

$$\sum_n P(|X_n| > c), \quad \sum_n E(X_n^{(c)}), \quad \sum_n \text{Var}(X_n^{(c)}),$$

where  $X_n^{(c)} = X_n \mathbb{I}_{|X_n| \leq c}$ . Then,

1. if the 3 series converge for *some*  $c > 0$ , then  $S_n = \sum_{j=1}^n S_j, n \geq 1$  converges a.s.  $(P)$ ,
2. if  $S_n = \sum_{j=1}^n S_j, n \geq 1$  converges a.s.  $(P)$ , then the 3 series converge for *all*  $c > 0$ .

**Corollary:** If  $X_1, X_2, \dots$  are ind r.v.'s on a psp  $(\Omega, \mathcal{F}, P)$  with  $E(X_n) = 0, n \geq 1$ , then

1. if  $\sum_n [E(X_n^{(c)})^2 + E|X_n| \mathbb{I}_{|X_n| > c}] < \infty$  for *some*  $c > 0$ , then  $S_n = \sum_{j=1}^n X_j$  a.s.  $(P)$ ,
2. if  $\sum_n E|X_n|^{\alpha_n} < \infty$  for *some*  $\{\alpha_n\} \subset [1, 2]$ , then  $S_n = \sum_{j=1}^n X_j$  converges a.s.  $(P)$ .

## Useful Inequalities

For positive  $a, b, p, (a + b)^p \leq 2^p(a^p + b^p)$ .

**Markov's:** if  $X$  is a nonnegative r.v. and  $a > 0$ , then  $P(X \geq a) \leq E(X)/a$ .

**Holder's:** If  $1/p + 1/q = 1$ , then for measurable  $f, g, \|fg\|_1 \leq \|f\|_p \|g\|_q$ .

**Jensen's:**  $\forall r \in (0, q), \phi(x) = x^{q/r}$  is convex

$$\implies [E|X|^{q/r}]^{1/q} \geq [E|X|^r]^{1/r}.$$