

Prob. Space: tripple (Ω, \mathcal{F}, P)

Ω : a set "sample space", $\omega \in \Omega$
 'sample point'.

\mathcal{F} : collection of subsets of Ω called
 'event points'.

(i) $\emptyset, \Omega \in \mathcal{F}$

(ii) closed under complementation &
 countable unions/intersection.

$$A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}, A_i \in \mathcal{F} \Rightarrow$$

$$\forall i \in \mathbb{N} \cup A_i, \cap A_i \in \mathcal{F}$$
~~$$\bigcap_{i=1}^{\infty} A_i \in \mathcal{F}$$~~

$$P: \mathcal{F} \rightarrow [0, 1]$$

Probability Measure



(i) $P(\emptyset) = 0, P(\Omega) = 1$.

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

(iii) $\langle A_i \rangle$ disjoint $\Rightarrow P(\cup A_i) = \sum_i P(A_i)$

Countable additivity

Additional properties of P :

$$(i) A \subset B \Rightarrow P(A) \leq P(B)$$

$$(ii) P(\cup A_i) \leq \sum P(A_i)$$

$$(iii) P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Theories of probability
- Kolmogorov - measure
- Girsanov - measure

$$P(A) =$$

Kolmogorov - measure
theoretic formulation

where do you get P ?

(i) usually specified for an algebra \mathcal{F}_0 and extended to $\sigma(\mathcal{F}_0)$

$\sigma(\mathcal{F}) =$ smallest σ -field containing \mathcal{F}
 $\mathcal{E}_\sigma =$ intersection of σ -fields containing \mathcal{F}

(ii) (a) principle of insufficient reason
Darwin - Fowler

(b) worst-case

Mathematical foundation of
information theory

$$-\sum_{i \in \{1, \dots, n\}} p_i \log p_i$$

$$\sum p_i H_i = C$$

$$\sum p_i = 1, p_i \geq 0$$

$$p_i = \frac{e^{-\beta H_i}}{\sum_j e^{-\beta H_j}} \quad (\text{Gibb's dist.})$$

Gaussian is actually the worst case.

$$- \int p(x) \log p(x) dx$$

$$\int x^2 p(x) dx = \sigma^2$$

$$\int x p(x) dx = 0, \quad p(\cdot) \geq 0, \quad \int p(x) dx = 1$$

(c) physical reasoning

(d) subjective probability.

(e) simplicity. ##

~~Agree~~ Agree/disagree

Assume

(f) measurement

Random variables

$$(\Omega, \mathcal{F}, P) \xrightarrow{X} (E, \mathcal{E})$$

$$X^{-1}(A) \triangleq \{\omega : X(\omega) \in A\},$$

$A \in \mathcal{E}$ (collection of subsets of E)

$$\boxed{\triangleq \text{def}}$$

\mathcal{E} is a σ -algebra

Need : $X^{-1}(A) \in \mathcal{F}$ for $A \in \mathcal{E}$. If this holds, X is said to be an E -valued random variable.

¶:

Usually \mathcal{E} 'obvious'.

'Borel' : $\{x : d(y, x) < \tau\}$, 'completed'.

σ -algebra on open & closed circles are same.

Define prob. measure μ on (E, \mathcal{E}) by :

$$\mu(A) = P\left(\underbrace{\{\omega : X(\omega) \in A\}}_{\omega \in \Omega}\right)$$

$$(i) \mu(\emptyset) = 0, \mu(E) = 1.$$

$$(ii) \mu(A^c) = 1 - \mu(A).$$

$$\begin{aligned} \mu(A^c) &= P(\langle \omega : X(\omega) \in A^c \rangle) \\ &= P(\langle \omega : X(\omega) \in A \rangle^c) \\ &= 1 - P(\langle \omega : X(\omega) \in A \rangle) \\ &= 1 - \mu(A) \end{aligned}$$

$$(iii) A_1, A_2 \dots \text{ disjoint in } \mathcal{E}.$$

$$\Rightarrow \mu\left(\bigcup_i A_i\right) = \sum_i \mu(A_i)$$

$$\begin{aligned} \mu\left(\bigcup_i A_i\right) &= P(\langle \omega : X(\omega) \in \bigcup_i A_i \rangle) \\ &= P\left(\bigcup_i \langle \omega : X(\omega) \in A_i \rangle\right) \end{aligned}$$

~~is disjoint~~

$$= \sum_i P(\langle \omega : X(\omega) \in A_i \rangle)$$

$$= \sum_i \mu(A_i)$$

μ is called the law of X .

$E = \mathbb{R}$ define $F(x) = \mu((-\infty, x])$, $x \in \mathbb{R}$.
 $F(\cdot)$ carries complete information about μ .
 $\mu((a, b)) = F(b) - F(a)$

1. Claim: $\lim_{x \rightarrow -\infty} F(x) = 0$,

2. $\lim_{x \rightarrow +\infty} F(x) = 1$

Countable additivity \Rightarrow

$$P\left(\bigcup_i A_i\right) = \lim_{N \uparrow \infty} P\left(\bigcup_{i=1}^N A_i\right)$$

$$P\left(\bigcap_i A_i\right) = \lim_{N \uparrow \infty} P\left(\bigcap_{i=1}^N A_i\right)$$

$$B_1 = A_1, B_2 = A_2 \setminus A_1, \dots, B_n = A_n \setminus \left(\bigcup_{i < n} A_i\right)$$

$$\bigcup_i A_i = \bigcup_i B_i, \quad \langle B_i \rangle \text{ disjoint}$$

$$\bigcup_{i=1}^n A_i = \bigcup_{i=1}^n B_i$$

$$P\left(\bigcup_i A_i\right) = P\left(\bigcup_i B_i\right) = \sum_i P(B_i)$$

$$= \lim_{n \rightarrow \infty} \sum_{i=1}^n P(B_i) = \lim_{n \rightarrow \infty} P\left(\bigcup_{i=1}^n B_i\right)$$

$$= \lim_{n \rightarrow \infty} P\left(\bigcup_{i=1}^n A_i\right)$$

strictly decreasing

$$\lim_{x \rightarrow -\infty} F(x) = \lim_{x \rightarrow -\infty} \mu((-\infty, x])$$

$$= \lim_{x \rightarrow -\infty} \mu\left(\bigcap_{y \geq x} (-\infty, y]\right)$$

$$= \mu\left(\bigcap_y (-\infty, y]\right) = \mu(\emptyset) = 0$$

$$3. \lim_{x \rightarrow \infty} F(x) = 1. \quad \lim_{x \rightarrow \infty} F(x) = \lim_{x \rightarrow \infty} \mu[x, \infty) = \lim_{x \rightarrow \infty} \mu\left(\bigcap_{y \leq x} [y, \infty)\right) = \mu\left(\bigcap_y [y, \infty)\right) = \mu(\emptyset) = 0$$

$$\lim_{y \downarrow x} F(y) = F(x). \quad (\text{because,})$$

$$\lim_{y \downarrow x} F(y) = \mu\left(\bigcap_{y > x} (-\infty, y]\right) = \mu((-\infty, x]) = F(x)$$

(y decreasing to x)

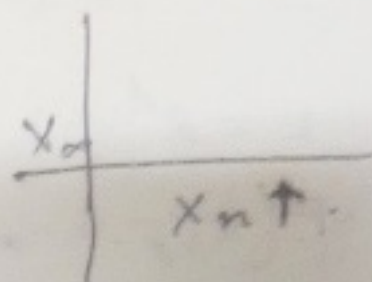


$$4. F(\cdot) \text{ increasing } (A \leq B \Rightarrow \mu(A) \leq \mu(B))$$

$$= \mu((-\infty, x]) = F(x)$$

5. $\lim_{y \uparrow \uparrow x} F(y)$ exists and is $\leq F(x)$ and is $\leq F(x)$

($\because F(\cdot)$ nondecreasing and bounded above by 1)



$$x_\infty = \text{lub} \langle x_n \rangle$$

$$x_n \rightarrow x_\infty$$

$$\lim_{y \uparrow \uparrow x} F(y)$$

$$= \lim_{y \uparrow \uparrow x} \mu([-\infty, y])$$

$$= \mu((-\infty, x)) \leq \mu([-\infty, x]) = F(x)$$

if $\lim_{y \uparrow \uparrow x} F(y) < F(x)$

$$\mu(\langle x \rangle) = \mu([-\infty, x]) - \lim_{y \uparrow \uparrow x} \mu([-\infty, y])$$

$$= F(x) - \lim_{y \uparrow \uparrow x} F(y)$$

6. $F(\cdot)$ has countably many jumps.

$z_n \triangleq \#$ of jump of size $\geq \frac{1}{n}$ then

$z_n \leq n$. let $\tilde{z}_n = \langle \text{jumps of size} \geq \frac{1}{n} \rangle$

$$z_n = |\tilde{z}_n| \leq n$$

$\therefore \bigcup_n Z_n$ is countable.

$$E \subseteq \mathbb{R}^d.$$

$$F(x_1, \dots, x_d) = \mu\left((-\infty, x_1] \times (-\infty, x_2] \times \dots \times (-\infty, x_d] \right)$$

$$\left. \begin{aligned} \lim_{x_i \downarrow -\infty} F(x_1, \dots, x_d) &= 0 \\ \lim_{x_1 \uparrow \infty \dots x_d \uparrow \infty} F(x_1, \dots, x_d) &= 1 \end{aligned} \right\}$$

Law & distribution are ^{almost} equivalent
 more general.

$\lambda(\cdot)$ is a positive measure (Ω, \mathcal{F}) ,
 $\mu(\cdot)$ a signed measure on (Ω, \mathcal{F}) .

$\lambda(\cdot)$ Lebesgue measure on \mathbb{R}^d , μ is above

Suppose $\lambda(A) = 0 \Rightarrow \mu(A) = 0$.

$$\mu \ll \lambda.$$

μ is said to be absolutely continuous

w.r.t. λ .

Theorem:

There exists $p(\cdot) : \mathbb{R}^d \rightarrow [0, \infty)$.

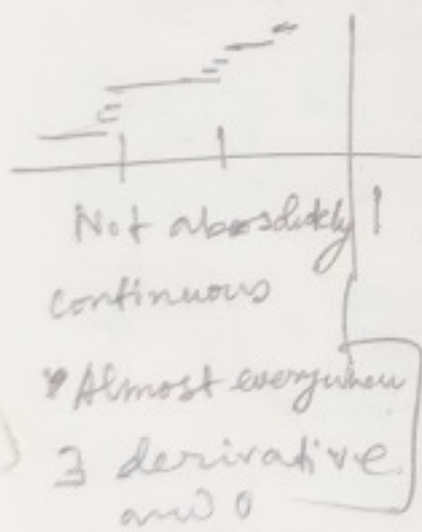
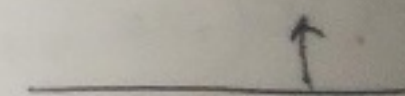
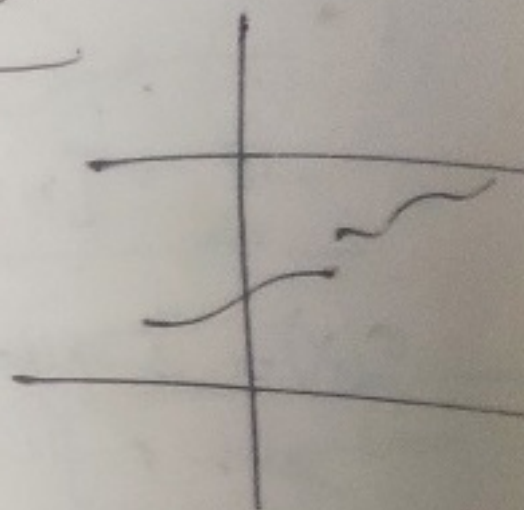
s.t. $\mu(A) = \int_A p(x) dx$.

$p(\cdot)$ is called the density of μ .

(Thus $F(x) = \int_{-\infty}^x p(y) dy$)

μ is called the law of X

define $\delta_{x_0}(A) = 1$ if $x_0 \in A$
 $= 0$ if $x_0 \notin A$.
 (Dirac function)



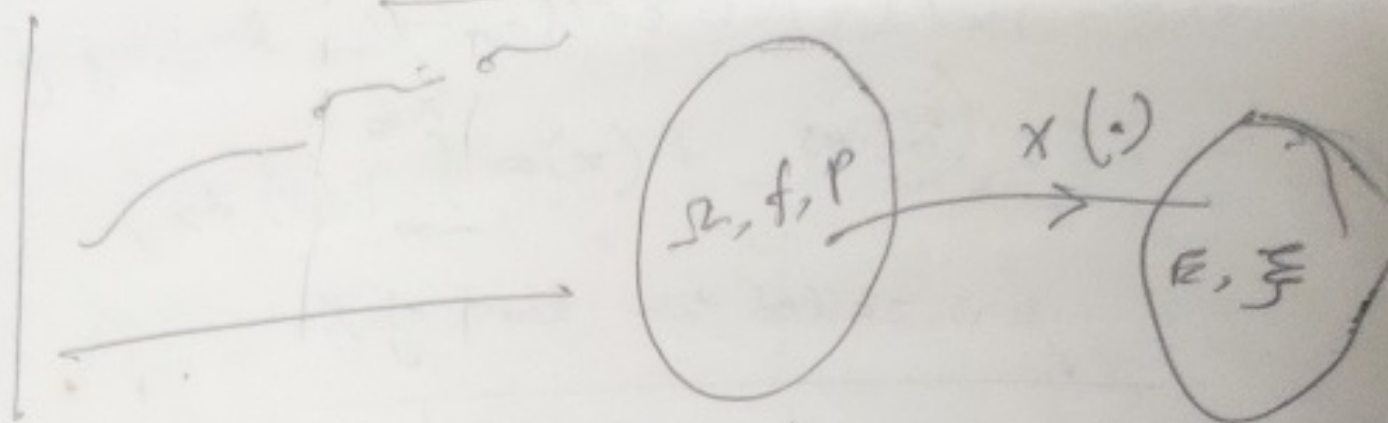
(Continuous)
Devil's staircase

Only absolutely continuous functions satisfy the laws of calculus: f_n is integral of its derivative.

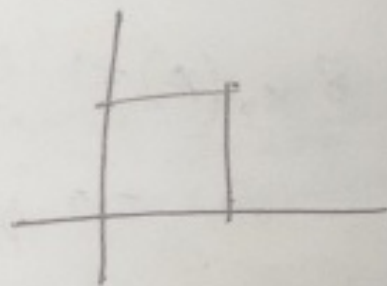
where \neq derivative form Cantor set

$$\mu_2 \ll \mu_1 \Leftrightarrow \mu_1(A) = 0 \Rightarrow \mu_2(A) = 0$$

$$E_{\mu_2}[X] = E_{\mu_1}[\wedge X]$$



$$\Omega = [0, 1]$$



$\sigma(X) =$ smallest
sub- σ field of

\mathcal{F} ~~which~~ which contains $X^{-1}(A)$, $A \in \Sigma$

(i.e., smallest sub- σ field of \mathcal{F}
w.r.t. which X is measurable)

$=$ intersection of all sub- σ field
of \mathcal{F} w.r.t. which X is
measurable

$$\sigma = \{ X^{-1}(A), A \in \mathcal{E} \}$$

Set of σ 's are preserved

$$X_1^{-1}(A) \leftrightarrow X_2^{-1}(A)$$

both uniformly distributed

Algebraic picture:

Canonical space:

$$\Omega = \mathbb{R}, f = \mathbb{R}_S,$$

$$P = \mu$$

$$X(\omega) = \omega$$

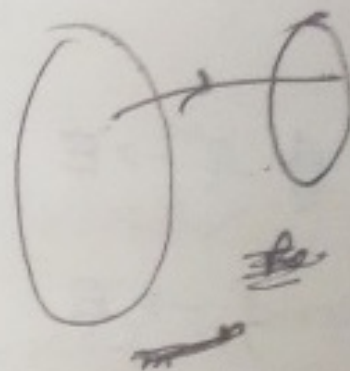
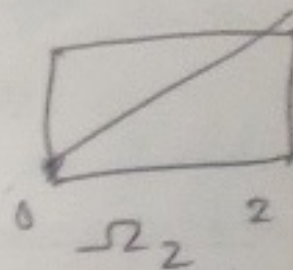
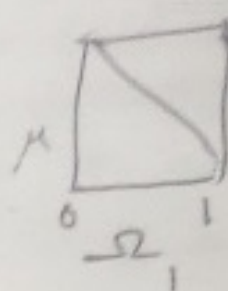
$$\sigma(X_\alpha, \alpha \in I)$$

= smallest σ -algebra w.r.t.

which all X_α are measurable,
 $\alpha \in I$

\Rightarrow Needs to check that it's a σ -field.

$$\text{under } \sigma \quad \bigg| \quad X^{-1}(A)$$



Quantum \rightarrow Non-commutative algebra

$$X \quad (X_1 \dots X_n) \\ (X_1, X_2, \dots)$$

$$\sigma(X_\alpha, \alpha \in I) = \bigcup_{J \subset I} \sigma(X_\alpha, \alpha \in J)$$

$J \subset I$

J countable

~~this is~~

[only those things
are measu
that are
countable]

Prob Measurement in
continuous space.

$$X: \Omega \rightarrow E$$

$$f \rightarrow F$$

$$(\Omega, P) \rightarrow (E, \Sigma)$$

Royden

Real Analysis