

Probability

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1. Basic theory: prob. spaces, random variables, expectation, convergence concepts, independence, conditioning, limit theorems.
2. Markov chains: classification of states, stationary distributions.
3. Martingales: inequalities, convergence theorems.

3x35 tests. Feller

Basic: Elementary prob. theory...
Chung +
Hoel, Port, Stone (1st & 3rd)



Ritman

Next level: First Course - Chung
Probability - Bauer
An intro. to prob. & measure
- Parthasarathy

Breiman - Probability ✓

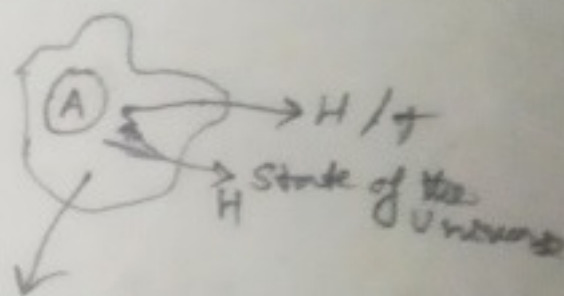
Williams - ~~tests~~ Weighing the odds
Probability in martingale

P is a probability measure (a measure on the measurable space (Ω, \mathcal{F}) non-negative measure with total mass 1)
 Probability Space: (Ω, \mathcal{F}, P)

\mathcal{F} : a σ -field
 subsets of Ω containing itself, elements of \mathcal{F} are called events

Toss a coin \rightsquigarrow Head
 \rightsquigarrow Tail
 \rightsquigarrow Stand on edge

Consider all possible ~~conditions~~ initial position velocity surface details air friction \vdots
 set of parameters that determine the coin's motion \rightarrow this is called the sample space Ω .



Ω : a set
 elements of $\Omega, \omega \in \Omega$ sample point
 Ω is called 'sample space'
 $A \subset \Omega$ to which we want to assign

probability ~~(number between 0 and 1)~~ (number between 0 and 1)

\mathcal{F} is a set of all $A \subset \Omega$ to which we want to assign a probability $P(A)$

~~desirable~~ desirable (\mathcal{F} is collection of the subsets of Ω , called ~~events~~ events)

$A \in \mathcal{F}$, 'event' (subset of the sample space)

If an event A is of the type $A = \{\omega \in \Omega \mid R(\omega)\}$ for some property $R(\cdot)$, we may write $P(R)$ for $P(A)$.
 $= \{ \text{points of } \Omega \text{ that lead to head} \}$ or $A^c = \{ \text{points of } \Omega \text{ that lead to tail} \}$

the act of tossing the coin at a particular time in a particular surrounding in a particular manner picks (sample) a point from this space Ω the only thing that

desired properties of \mathcal{F} :

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

$\emptyset \in \mathcal{F}$

(ii) $A, B \in \mathcal{F} \Rightarrow 'A \cup B'$

$'A \cap B'$

$A \cup B \in \mathcal{F}, A \cap B \in \mathcal{F}.$

More generally, $A_1, A_2, \dots \in \mathcal{F} \Rightarrow \bigcap_i A_i,$

$\bigcup_i A_i \in \mathcal{F}$

(iii) $A \in \mathcal{F} \Rightarrow 'not A' \in \mathcal{F}, i.e., A^c \in \mathcal{F}$

Remarks: (i) $\bigcap_i A_i \in \mathcal{F},$

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

(ii) can consider weaker requirement

$A_1, \dots, A_n \in \mathcal{F} \Rightarrow \bigcup_{i=1}^n A_i \in \mathcal{F}.$

closed

under union &
intersection

\mathcal{F} is ~~set~~ said to be a σ -algebra.

σ -field

usually denotes some
countable operation

\mathcal{f} containing \emptyset , Ω & is closed under complementation, countable unions and intersections

one often need not consider all subsets.

P : 'probability of ...' $P: \mathcal{f} \rightarrow [0, 1]$

$P(A)$: probability of A .

Desired properties of P :

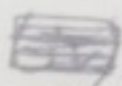
(i) $P(\Omega) = 1, P(\emptyset) = 0$.

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

(iii) A_1, A_2, \dots, A_n disjoint,

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

Countable additivity



Remark: If we require only that

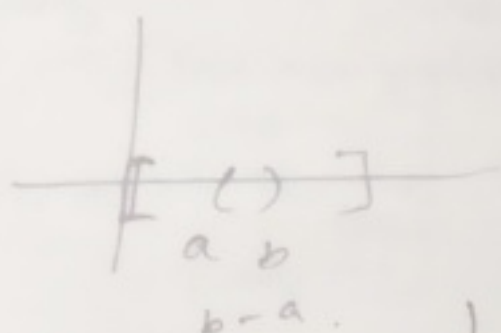
$$P\left(\bigcup_{i=1}^n A_i\right) = \sum_{i=1}^n P(A_i), \quad \{A_i\} \text{ disjoint, finite additivity}$$

to each event $A \in \mathcal{f}$, is assigned a number $\in [0, 1]$, called its probability denoted by $P(A)$. We expect that \mathcal{f} and $P(\cdot)$ as a map from \mathcal{f} to $[0, 1]$ should satisfy certain requirements - must be defined

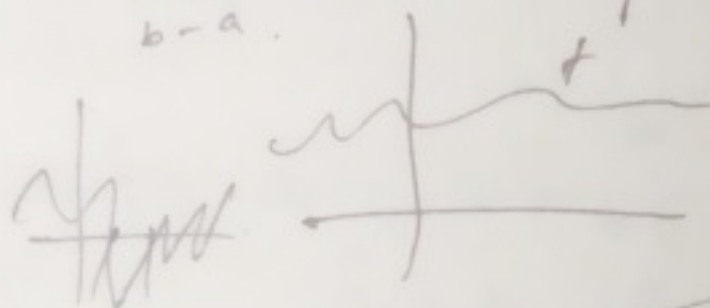
(Ω, \mathcal{F}, P) probability space.

(i) - (iii) defines P as a positive measure on (Ω, \mathcal{F}) with total mass 1.

'measure' length, area, volume.



examples of measures.



$$\mu(A) = \int_A f(x) dx.$$

negative measure

signed measure.

Ω, \mathcal{F}, ν

$\Omega = [0, 1]$ $\mathcal{F} = ?$

Require: \mathcal{F} contains all intervals.

Given \mathcal{f}_0 , a collection of subsets of Ω ,

we can define $\sigma(\mathcal{f}_0)$ = the smallest σ -algebra containing \mathcal{f}_0 .

= the intersection of σ -algebras containing f_0

Need (i) \exists a σ -algebra containing f_0 —
take all subsets of Ω

(ii) \blacktriangleright observe: Arbitrary intersections of σ -algebras are σ -algebras.
(Unions are not)

Let $\mathcal{F}_\alpha, \alpha \in I$, be σ -algebras on Ω

$$\hat{\mathcal{F}} = \bigcap_{\alpha} \mathcal{F}_\alpha.$$

$$(i) \quad \Omega \in \mathcal{F}_\alpha \quad \forall \alpha \Rightarrow \Omega \in \hat{\mathcal{F}},$$

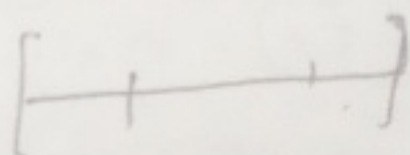
similarly, $\emptyset \in \hat{\mathcal{F}}$.

$$(ii) \quad A \in \hat{\mathcal{F}} \Rightarrow A \in \mathcal{F}_\alpha \quad \forall \alpha \Rightarrow A^c \in \mathcal{F}_\alpha \quad \forall \alpha$$

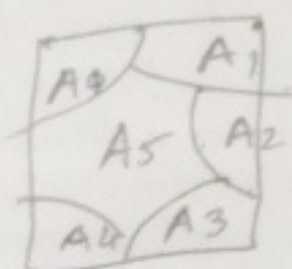
$$(iii) \quad A_1, A_2, \dots \in \hat{\mathcal{F}} \Rightarrow A_i \in \mathcal{F}_\alpha \quad \forall \alpha, i$$
$$\Rightarrow \bigcup_i A_i \in \mathcal{F}_\alpha \quad \forall \alpha \Rightarrow \bigcup_i A_i \in \hat{\mathcal{F}}.$$

\Rightarrow the two definitions are equivalent.

$\sigma(f_0)$ = smallest σ -algebra containing f_0 = intersection of σ -algebras containing f_0 .



$$f_0 = \{A_1, A_2, \dots, A_6\}$$



$$f_0 = \{A_1, A_2, \dots, A_6\}$$

$$\sigma(f_0) = \{\emptyset, \Omega, A_1, \dots, A_6, A_1 \cup A_2, \dots, A_1 \cup A_2 \cup A_3, \dots\}$$

S , $d(\dots, \dots)$ metric. is.

$$d(x, y) \geq 0$$

$$d(x, z) \leq d(x, y) + d(z, y)$$

$$d(x, y) = 0 \text{ iff } x = y$$

$$\{x \mid d(x, x_0) < r\}$$

Ball of radius r , ~~center~~
centre x_0

Borel σ -algebras. (for open sets)

$$R \mid M((a, b)) = b - a$$

Ω

Example of a non-measurable set



\mathcal{F}_0 algebra $\rightarrow \sigma(\mathcal{F}_0)$ σ algebra

P on $(\Omega, \mathcal{F}_0) \rightarrow P$ on (Ω, \mathcal{F})

need extension
theorems.

'Completion' $(\Omega, \mathcal{F}, P) \xrightarrow{\text{let } \hat{\mathcal{F}}}$

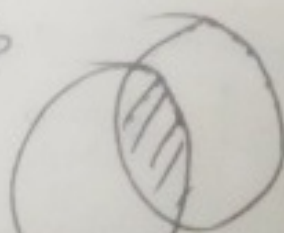
Let $\hat{\mathcal{F}} = \mathcal{F} \cup \{\text{arbitrary subsets of sets } A \in \mathcal{F} \text{ s.t.}$

additional properties
of P

$$P(A) = 0\}$$

$$(ii) A \subseteq B \Rightarrow P(A) \leq P(B)$$

~~additional prop~~



$$(iv) P\left(\bigcup_i A_i\right) \leq \sum_i P(A_i)$$

$$(2) P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Example of compulsion R.

$$m((a, b)) = b - a$$

Lebesgue σ -algebra

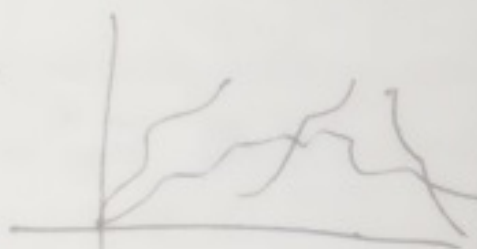
$$P(\emptyset) = 0$$

So, A is null event if $P(A) = 0$.

sure event if $P(A) = 1$.

Statement S is said to be true almost
surely if $P(S \text{ true}) = 1$ a.s.

$$\underline{X = Y}$$



$$r_1, r_2, \dots$$

$$r_n = \frac{\epsilon}{2^n}$$

Sum up to $< \epsilon$

if X is a set, a nonempty class \mathcal{R} of subsets of X is called a ring if it is closed under the formation of set theoretic differences & finite unions. If moreover $X \in \mathcal{R}$, then \mathcal{R} is called an algebra. A nonempty subclass of $\mathcal{P}(X)$ is a ring in the earlier sense iff it is a subring of $\mathcal{P}(X)$ in algebraic sense.

The class $\mathcal{P}(X)$ of X (power set) is a ring in usual algebraic notation

if, $\begin{cases} + \equiv \Delta \\ \cdot \equiv \cap \end{cases}$ defined

→ requirement is that it is ~~an~~ a σ -algebra
 (Ω, \mathcal{F}, P) It's assumed that σ -algebra
 is Borel algebra.

Borel algebra requires the
 notion of neighbourhood.

~~(A, F)~~ (Ω, \mathcal{F}) — measurable
 space

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

desired properties of X :

want to assign prob. to sets

$$\{\omega \mid X(\omega) \in A\}, A \subseteq E$$

= a set

$$\{\omega \in \Omega : X(\omega) \in A\}$$

Desired properties of Σ = \langle subsets A ^{Range, not domain}

$A \subseteq E$ under
 consideration)

(i) ~~$\emptyset \in \Sigma$~~ $\emptyset \in \Sigma, \phi \in \Sigma$

(ii) $A \in \Sigma \Rightarrow A^c \in \Sigma$

(iii) $A_i \in \Sigma \forall i \Rightarrow \bigcup_i A_i \in \Sigma$

Let (E, \mathcal{E}) be another measurable space and $X: (\Omega, \mathcal{F}, P) \rightarrow (E, \mathcal{E})$
 \mathcal{E} should be a σ -algebra $\left[\begin{array}{l} \text{a r.v. (i.e. a measur. map)} \end{array} \right]$

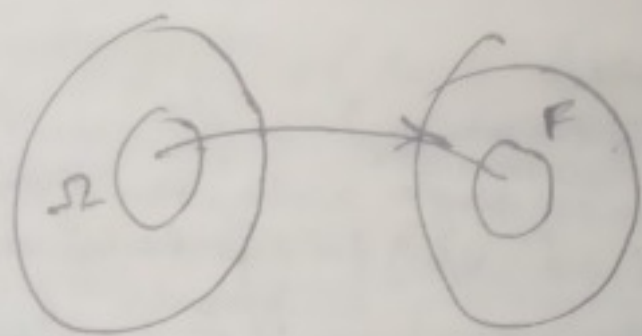
Want $\{\omega: X(\omega) \in A\}$ has a prob.
 i.e., $\{\omega: X(\omega) \in A\} \in \mathcal{F}$.

$$\Omega = [0, 1] \quad A \notin \mathcal{F} \leftarrow$$

$$X(\omega) = \begin{cases} 1 & \text{on } A \\ 0 & \text{on } A^c \end{cases}$$

$$\{\omega: X(\omega) \in (-.5, .5)\} = A, \text{ we can't assign}$$

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



Require: $\forall A \in \mathcal{E}, X^{-1}(A) \in \mathcal{F}$.

$$X^{-1}(A) \in \mathcal{F} \quad \forall A \in \mathcal{E}.$$

\rightarrow X is 'measurable'.

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

X is said to be an E -valued random variable if $X^{-1}(A) \triangleq \{\omega : X(\omega) \in A\} \in \mathcal{F}$
 $\forall A \in \mathcal{E}$

Can define a prob. measure μ on

$$(E, \mathcal{E}) \text{ by: } \mu(A) \\ = P(\underbrace{\langle \omega : X(\omega) \in A \rangle}_{\equiv}) \\ = P(X \in A)$$

$\mu = \text{the law of } X$

- (i) ^(Equally likely outcomes) Principle of ~~insufficient~~ ^{reason} ~~reason~~ ^{insufficient}
- (ii) Physical reasoning [since single realization of a r.v. X , i.e. $X(\omega)$ for a particular $\omega \in \Omega$ the probability space (Ω, \mathcal{F}, P) is a hypothetical entity and its choice is by no means unique.]
- (iii) worst case - max. entropy

$$\Omega = \langle a_1, \dots, a_n \rangle$$

$$\int P(x) \log(P(x)) dx$$

E. T. Jaynes

(Non-Bayesian)

entropy method
max. ~~info~~ ~~interpretation~~

The image μ of P under X is a probability measure on (E, \mathcal{E}) , called the law of X and denoted by $\mathcal{L}(X)$.

The events $\{\omega \mid X(\omega) \in A\}$ for $A \in \mathcal{E}$ form a sub- σ -field of \mathcal{F} called the σ -field generated by X and denoted by $\sigma(X)$. More generally, given a family of $X_\alpha, \alpha \in I$, of random variables on (Ω, \mathcal{F}, P) taking values in measurable spaces $(E_\alpha, \mathcal{E}_\alpha), \alpha \in I$, respectively, the σ -field generated by $X_\alpha, \alpha \in I$, denoted by $\sigma(X_\alpha, \alpha \in I)$, is the smallest sub- σ -field w.r.t. which they are all measurable.

Now an experiment such as tossing a coin or rolling a die picks a point from the sample space and maps it to another space E . ($E = \{\text{head, tail}\}$ and $\{1, 2, \dots, 6\}$ resp.) Thus it's a map $X: \Omega \rightarrow E$. Since our idea is to have sets of the type $\{\omega \mid X(\omega) \in B\}$ to be events (i.e. elements of \mathcal{F}) for a suitable collection of subsets $B \subseteq E$, considerations analogous to ^{the} above for f suggest that we equip E with a σ -field \mathcal{E} and require a map $X: (\Omega, \mathcal{F}) \rightarrow (E, \mathcal{E})$ to be measurable. Such a map is called E -valued r.v.