

Coin tossing : Fair coin tossed n times \rightarrow iid / id. dist
 $\text{prob}(H) = \text{prob}(T) = \frac{1}{2}$

2^n possible outcomes \rightarrow each has prob 2^{-n}

What is prob of exactly k heads? $\binom{n}{k} / 2^n$

What is the prob of n heads in $2n$ tosses?

$$\text{Ans} = \frac{(2n!)}{(n!)^2 2^n} \approx \frac{1}{\sqrt{\pi n}} (1 + \delta_n) \quad \text{where } \delta_n \rightarrow 0 \text{ as } n \rightarrow \infty$$

As $2n \rightarrow \infty$ Prob of $n \frac{1}{2}$ heads $\rightarrow 0$.

So clearly convergence to $\frac{1}{2}$ is not happening in this sense
 Weak law of Large numbers:

Let Ω_n be the set of $\{0, 1\}$ sequences of length n
 an individual sequence is written ω .

Prop let $X : \Omega_n \rightarrow \mathbb{R}$ be any function & let $\varepsilon > 0$

$$P(\{\omega | |X(\omega)| \geq \varepsilon\}) \leq \frac{1}{\varepsilon^2} E(X^2)$$

$$P(\{\omega | |X(\omega)| \geq \varepsilon\}) = P(\{\omega | X^2(\omega) \geq \varepsilon^2\})$$

$$= \sum_{\omega: X^2(\omega) \geq \varepsilon^2} \frac{1}{2^n} \leq \sum_{\omega: X^2(\omega) \geq \varepsilon^2} \frac{X^2(\omega)}{\varepsilon^2} \frac{1}{2^n} \leq \frac{1}{\varepsilon^2} \sum_{\omega \in \Omega_n} \frac{X^2(\omega)}{2^n}$$

$$= \frac{1}{\varepsilon^2} E(X^2). \blacksquare$$

Now define $X_j(\omega) = \begin{cases} 1 & \text{if } \omega[j] = H \\ 0 & \text{if } \omega[j] = T \end{cases}$

$$S_n(\omega) = \sum_{i=1}^n X_i(\omega)$$

$$\text{e.g. } E(X_i) = \frac{1}{2} \quad E(X_i X_j) = \frac{1}{4} \text{ for } i \neq j$$

$$E((X_i - \frac{1}{2})(X_j - \frac{1}{2})) = 0 \text{ for } i \neq j \quad E(X_i - \frac{1}{2})^2 = \frac{1}{4}$$

$$S_n - \frac{n}{2} = \sum_{i=1}^n (X_i - \frac{1}{2})$$

$$E(S_n - \frac{n}{2})^2 = E\left(\frac{S_n - n}{n}\right)^2 = \frac{1}{n^2} E\left(\sum_{i,j} (X_i - \frac{1}{2})(X_j - \frac{1}{2})\right)$$

$= \frac{1}{4n}$ (cross-terms vanish)

$$\text{Now. } P(\{\omega | |\frac{S_n(\omega)}{n} - \frac{1}{2}| \geq \varepsilon\}) \leq E\left(\frac{S_n - n}{n}\right)^2 / \varepsilon^2 \quad (\text{Chebyshev})$$

$$P(\dots) \leq \frac{1}{4n\varepsilon^2}$$

$$\lim_{n \rightarrow \infty} P(\{\omega | |\frac{S_n(\omega)}{n} - \frac{1}{2}| \geq \varepsilon\}) = 0$$

Strong law of large numbers

Let Ω be infinite sequences of 0's & 1's $w \in \Omega$.

Define $X_j : \Omega \rightarrow \mathbb{R}$ as before.

$$S_n(w) = \sum_{j=1}^n X_j(w)$$

$$\lim_{n \rightarrow \infty} \frac{S_n(w)}{n} = \frac{1}{2} ?? \text{ Absolutely false!!}$$

$$\Pr(\{\omega \mid \lim_{n \rightarrow \infty} \frac{S_n(\omega)}{n} = \frac{1}{2}\}) = 1 \quad \text{limit is inside.}$$

Do we know how to make sense of this?

How do we assign probabilities to subsets of Ω ?

What is the probability of the sequence

HTHTHT...? Ans $\rightarrow 0$.

So any element of Ω has prob. 0. & we cannot get probabilities of all sets by looking at probabilities of singletons or atoms.

What is the prob of $\Pr(HT^\infty) = 0$

$\Pr(THT^\infty) = 0$

$\Pr(T^2HT^\infty) = 0$

:

Prob exactly one H = 0

Prob exactly 2 H = 0

Prob exactly n , H = 0

Prob at least 1 H is $= 0 + 0 + \dots + 0 + \dots = 0$ Huh??

Prob all T = 0 so Prob at least 1 H is 1 not 0!?

We forgot $(HT)^\infty$ and similar sequences. What we proved is Prob (finite # of H) = 0.

For what sequences can we assign prob?

Notation $s \in \Omega_n$ for some n $s \leq w$; s prefix of $w \in \Omega$.

$$S^{\uparrow} := \{ \omega / s \leq \omega \} \quad \text{a wingless } \\ P(s^{\uparrow}) = 2^{-\text{len}(s)} \quad \text{len}(s): \text{length of } s.$$

But $\{ \omega / \lim_{n \rightarrow \infty} \frac{s_n}{n} = \frac{1}{2} \}$ is not a wingless!
How do we assign measures to such sets?

What are the basic axioms we want?

(i) $P: (\Omega, \mathcal{F}, P) \rightarrow \text{probability triple}$

$\Omega \rightarrow \text{sample space } \mathcal{F} \rightarrow \text{null sets } P \rightarrow \text{prob measure}$

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

(i) Want $P(\emptyset) = 0$ & $P(\Omega) = 1$ so $\emptyset, \Omega \in \mathcal{F}$

(ii) $P(A) + P(A^c) = 1$ so $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$

* (iii) $A, B \in \mathcal{F}$ & $A \cap B = \emptyset \Rightarrow P(A) + P(B) = P(A \cup B)$

in fact $\{A_i / i \in \mathbb{N}\}$ $A_i \in \mathcal{F} \Rightarrow \bigcup_i A_i \in \mathcal{F}$.

$\forall i \neq j \quad A_i \cap A_j = \emptyset \Rightarrow P(\bigcup_i A_i) = \sum_i P(A_i).$

Uncountable sums don't make sense
as we saw with coin tossing.

Def A σ -algebra $\mathcal{F} \subseteq P(\Omega)$ is a family of sets s.t.

(i) $\emptyset \in \mathcal{F}$ (ii) $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$

(iii) $A_i \in \mathcal{F}$ for $i \in \mathbb{N} \Rightarrow \bigcup_i A_i \in \mathcal{F}$.

Examples (i) $P(\Omega)$ (ii) $\{ \emptyset, A, A^c, \Omega \}$

Need more interesting examples

Prop let $\{ \mathcal{F}_\alpha / \alpha \in A \}$ be any family of σ -algebras on Ω .
Then $\bigcap_{\alpha \in A} \mathcal{F}_\alpha$ is a σ -algebra.

Proof. If $A \in \bigcap_{\alpha \in A} \mathcal{F}_\alpha$ then $\forall \alpha \quad A \in \mathcal{F}_\alpha \Rightarrow A^c \in \mathcal{F}_\alpha \Rightarrow$
 $A^c \in \bigcap_{\alpha \in A} \mathcal{F}_\alpha$. Similarly for $\bigcup_i A_i$.

Thus given any family \mathcal{H} of subsets of Ω there is a
least σ -algebra containing \mathcal{H} = $\bigcap_{\alpha \in A} \mathcal{F}_\alpha$ all σ -alg containing \mathcal{H} .

(iii) Let $(a, b) = \{x \mid a < x < b\}$ be the family of open intervals of \mathbb{R} . The σ -algebra generated is called the Borel algebra.

(iv) Let (M, d) be a metric space, we define $r \in \mathbb{R}^{>0}, x \in M, B_r(x) := \{u \in M \mid d(x, u) < r\}$. These are called open balls. The σ -algebra generated by the open balls is the Borel algebra of M .

(v) Coin tossing space (Ω, \mathcal{F}) : the σ -algebra used is the one generated by the wins/glasses. The same σ -algebra is generated by cylinders

Measures

$$C(n_1, \dots, n_k; b_1, \dots, b_k) := \\ \{\omega \mid \forall 1 \leq i \leq k \quad \omega_i = b_i\}.$$

Measures Given (Ω, \mathcal{F}) a measure is a map $\mu : \mathcal{F} \rightarrow \mathbb{R}^{>0}$ s.t.

$$(i) \quad \mu(\emptyset) = 0 \quad (ii) \quad A \subseteq B \Rightarrow \mu(A) \leq \mu(B)$$

$$(iii) \quad A_i \in \mathcal{F}, \text{ pairwise disjoint} \Rightarrow \mu(\bigcup_i A_i) = \sum_i \mu(A_i).$$

If $\mu(\Omega) < \infty$ we say μ is a finite measure.

If $\mu(\Omega) = 1$ we say μ is a probability measure

If $\mu(\Omega) \leq 1$ we say μ is a subprobability measure

If $\mu(\Omega) = \infty$ but \exists a countable family A_i s.t.

$\bigcup_i A_i = \Omega$ & $\forall i \quad \mu(A_i) < \infty$ we say

μ is σ -finite.

Examples (1) For coin tossing space

$$\Pr(\text{S}) = 2^{-1/2} \quad \text{somewhat extended.}$$

(2) $(\Omega, \mathcal{P}(\Omega))$ & $\delta_{\omega_0}(A) = \begin{cases} 1 & \text{if } \omega_0 \in A \\ 0 & \text{if } \omega_0 \notin A \end{cases}$ DIRAC measure or point mass.

(3) $(\mathbb{R}, \text{Borel})$ $\lambda : \text{Borel sets} \rightarrow \mathbb{R}^{>0}$ by "length"
Lebesgue measure But what is it?

How to define measure on \mathbb{R} ?

$$\lambda((a, b)) = b - a$$

$$\lambda([a, b]) = \lambda((a, b]) = \lambda([a, b]) = b - a.$$

$$\lambda(\mathbb{Q}) = ?$$

One approach: Given $S \subseteq \mathbb{R}$ we define a cover of S to be a set of intervals $\mathcal{I} = \{(a_i, b_i)\}_{i \in I}$ s.t. $\bigcup_{i \in I} (a_i, b_i) \supseteq S$.

$$\lambda(S) \leq \sum_{i \in I} \lambda((a_i, b_i)) \text{ so } \lambda(S) \leq \inf_{\mathcal{I} \text{ covers } S} \lambda(\mathcal{I})$$

Now $\mathcal{I} \nsubseteq \mathbb{Q}$. Consider \mathbb{Q} : enumerate $\mathbb{Q} = q_1, q_2, q_3, \dots, q_n, \dots$

Give me a "budget" of $\epsilon \in \mathbb{R}$ to cover \mathbb{Q} :

$$(q_1 - \frac{\epsilon}{4}, q_1 + \frac{\epsilon}{4}), (q_2 - \frac{\epsilon}{8}, q_2 + \frac{\epsilon}{8}), (q_3 - \frac{\epsilon}{16}, q_3 + \frac{\epsilon}{16}), \dots$$

length of the cover is $\sum_{i=1}^{\infty} \epsilon (\frac{1}{2})^i = \epsilon$.

Note $\epsilon \rightarrow 0$ so $\inf = 0$ & $\lambda(\mathbb{Q}) = 0$.

Why not define $\lambda(S) = \inf_{\mathcal{I} \text{ covers } S} \lambda(\mathcal{I})$?
Call this λ^*

B This is not countably additive

$$\text{Prop } \lambda^*(\bigcup_i A_i) \leq \sum \lambda^*(A_i)$$

Proof Choose any $\epsilon > 0$ since $\lambda^*(A_i)$ is defined as the inf over covering intervals we can find \mathcal{I}_i s.t.

$$\lambda(I_1) \leq \lambda^*(A_1) + \frac{\epsilon}{2}$$

$$\dots \lambda(I_n) \leq \lambda^*(A_n) + \frac{\epsilon}{2^n}$$

$$\therefore \lambda(\mathcal{I}) \leq \sum \lambda^*(A_i) + \epsilon \text{ where } \mathcal{I} = \text{union of } I_i.$$

Now \mathcal{I} covers $\bigcup_i A_i$ so $\lambda^*(\bigcup_i A_i) \leq \lambda(\mathcal{I}) \leq \sum \lambda^*(A_i) + \epsilon$

$$\text{Thus } \lambda^*(\bigcup_i A_i) \leq \sum \lambda^*(A_i).$$

The ϵ -elbow room.

Def An outer measure is a function $\mu^*: \mathcal{P}(\Omega) \rightarrow \mathbb{R}^{>0}$ s.t.

- (i) $\mu^*(\emptyset) = 0$
- (ii) $A \subseteq B \Rightarrow \mu^*(A) \leq \mu^*(B)$
- (iii) $\mu^*(\cup_i A_i) \leq \sum_i \mu^*(A_i)$.

So we have constructed an outer measure which is defined on all sets but it is not a measure.

Theorem Given Ω & $\mu^*: \mathcal{P}(\Omega) \rightarrow \mathbb{R}^{>0}$ an outer measure we call a set E good if $\forall A \subseteq \Omega \quad \mu^*(A) = \mu^*(A \cap E) + \mu^*(A \cap E^c)$. The collection of good sets forms a σ -algebra & μ^* restricted to this σ -algebra is a measure.

Proof Omitted.

Vitali's example ~~Let us see~~ Then There is no non-trivial translation-invariant measure that can be defined on all subsets of \mathbb{R} .

Proof Step Let $x \sim y := x - y \in \mathbb{Q}$. Then \sim is an eq. rel.
Define V by choosing one member from each equivalence class in the interval $[0, 1]$.

$V \subseteq [0, 1]$ so $\mu(V) \leq \mu([0, 1]) = 1$.

$V+g := \{v+g | v \in V\} \quad g \in \mathbb{Q} \cap [0, 1]$

Claim: $V+g_1 \cap V+g_2 = \emptyset \quad \text{if } g \neq g'$.

Suppose $x \in V+g_1 \cap V+g_2$ then $\exists v_1, v_2 \in V$ s.t.

$$x = v_1 + g_1 \quad \text{and} \quad x = v_2 + g_2 \quad \text{so}$$

$$v_1 = v_2 + g_2 - g_1 \quad \text{so} \quad v_1 \sim v_2 \quad \text{so}$$

Claim For any $x \in (0, 1)$ ~~There exists~~ $\exists r \in (-1, 1)$ s.t. $x \in V+r$: $\forall x \in (0, 1) \quad \exists v \in V$ s.t. $x \sim v$ so.

~~the~~ $x-v \in \mathbb{Q}$, let $r = x-v$ so $x \in V+r$. Since both $x, v \in (0, 1)$ $x-v \in (-1, 1)$.

$$S := \bigcup_{r \in (-1, 1) \cap \mathbb{Q}} V+r \quad ; \quad S \supseteq (0, 1) \quad \text{and} \quad S \subseteq (-1, 2)$$

Now $\mu(V+r) = \mu(V)$ since μ is A. i.e.v.

$V+r$ are pwo & ~~so~~. Suppose $\mu(V) = \alpha > 0$

$$\mu(\emptyset S) = \sum_{x \in S} \alpha = \infty \text{ but } \mu(S) \leq 3 \otimes.$$

Suppose $\mu(V) = 0$ then

$$\mu(S) = \sum 0 = 0 \text{ but } \mu(S) \geq 1 \otimes.$$

So $\mu(V)$ cannot be defined. \blacksquare

These ~~are~~ are

This gives us the example we are looking for outer ~~as~~ measure failing to be countably additive.

So we have to choose well defined sets.

The Borel sets can be measured. A much larger class of sets (the Lebesgue sets) can be measured as well but it is impossible to measure all sets.

—x—

Some properties of measures:

Prop (i) If $A_1 \subseteq A_2 \subseteq A_3 \subseteq \dots \subseteq A_n \subseteq \dots$ & $A = \bigcup_i A_i$ then
 $\lim_{i \rightarrow \infty} \mu(A_i) = \mu(A)$

(ii) If $A_1 \supseteq A_2 \supseteq \dots \supseteq A_n \supseteq \dots$ & $A = \bigcap_i A_i$ then
 $\lim_{i \rightarrow \infty} \mu(A_i) = \mu(A)$ if $\mu(A_1) < \infty$.

Proof Recall $\# A \subseteq B \Rightarrow \mu(A) \leq \mu(B)$.

$$B = A \cup (B \setminus A) \text{ so } \mu(B) = \mu(A) + \mu(B \setminus A) \geq \mu(A).$$

(i) We define a new family B_i by induction

$$B_1 = A_1, \quad B_{n+1} = A_{n+1} \setminus A_n \quad B_i \text{ are pwo}$$

$$\& \bigcup_i B_i = \bigcup_i A_i = A \quad \& \bigcup_{i=1}^n B_i = A_n \text{ so } \mu(A) = \sum_{i=1}^{\infty} \mu(B_i)$$

$$\lim_{n \rightarrow \infty} \mu(A_n) = \lim_{n \rightarrow \infty} \sum_{i=1}^n \mu(B_i) = \sum_{i=1}^{\infty} \mu(B_i) = \mu(A)$$

$\lim_{n \rightarrow \infty} \mu(A_n) = \lim_{n \rightarrow \infty} \sum_{i=1}^n \mu(B_i) = \sum_{i=1}^{\infty} \mu(B_i) = \mu(A)$

(ii) Similarly for down continuity. But why $\mu(A_1) < \infty$?

Consider $A_i = (i, \infty)$ $\lambda(A_i) = \infty$ for all i so

$$\lim_{i \rightarrow \infty} \lambda(A_i) = \infty \text{ but } \bigcap_i A_i = \emptyset = A \text{ so } \mu(A) = 0.$$

* Gas about Choquet capacity.

For any countable family of sets
 $\mu(\bigcup_i B_i) \leq \sum_i \mu(B_i).$

Integration: what are we integrating?

def A function $f: (X, \Sigma) \rightarrow (\mathbb{R}, \mathcal{B})$ is measurable if $\forall B \in \mathcal{B} f^{-1}(B) \in \Sigma$.

Functions that are not measurable are too wild.

Unfortunately even null is not enough e.g.

$$\int_{-\infty}^{\infty} 1 dx = \infty$$

$$\int_{-\infty}^{\infty} x dx = ?$$

lets not be too ambitious.

(Step 1). $\mathbf{1}_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$ X_A same notation.

$$\int_X \mathbf{1}_A d\mu := \mu(A)$$

Next $s: (X, \Sigma) \rightarrow \mathbb{R}$ is a simple function if it has a finite range so $s(X) = \{a_1, \dots, a_n\}$.

$$A_i = s^{-1}(a_i)$$

$$s = \sum_{i=1}^n a_i \cdot \mathbf{1}_{A_i} \quad \text{note } A_i \cap A_j = \emptyset \text{ if } i \neq j.$$

$$\int_X s d\mu = \sum_{i=1}^n a_i \cdot \mu(A_i)$$

what if $a_1 < 0$, $\mu(A_1) = \infty$ & $a_2 > 0$ & $\mu(A_2) = \infty$?

def We say s is integrable if whenever $a \in \text{range}(s)$ $a \neq 0 \Rightarrow \mu(s^{-1}(a)) < \infty$.

Then $\int_X s d\mu$ formula is sensible & taken as the def.

If $\{f_i\}_{i \in \mathbb{N}}$ is a family of nbl functions
then f defined as pt wise limit is also measurable

$$f(x) := \lim_{i \rightarrow \infty} f_i(x).$$

Given a non-negative nbl function $f: X \rightarrow \mathbb{R}$
there is a family of simple functions s_i such that
 $s_i \leq s_{i+1} \leq f$ & $\{s_i\}$ converges ptwise to f .

Suppose f is a non-negative real valued nbl f.
We say f is integrable if the everywhere non-neg
simple functions below f are integrable & their
integrals are bounded. In that case we define

$$\int_X f d\mu = \sup_{\substack{s \leq f \\ \text{simple}}} \int_X s d\mu \quad \text{where } \delta_{x_0}$$

Ex $(X, \mathcal{E}, \delta_{x_0})$ $\int s(x) d\delta_{x_0} = s(x_0) \delta_{x_0}(s^{-1}(x_0)) = s(x_0)$

$$\int_X f d\delta_{x_0}$$

Now let s be any simple function below f
so $\forall x \quad s(x) \leq f(x)$ so $\int_X s d\delta_{x_0} = s(x_0) \leq f(x_0)$.
Consider the function $t(x) = \begin{cases} f(x) & \text{if } x = x_0 \\ 0 & \text{otherwise} \end{cases}$ This is simple sf.

$$\text{so } \int t d\delta_{x_0} = f(x_0) \quad \text{so}$$

$$\sup_{s \leq f} \int s d\delta_{x_0} = f(x_0).$$

$$\int_X f d\delta_{x_0} = f(x_0).$$

Properties $\int (f+g) d\mu = \int f d\mu + \int g d\mu \quad f \leq g \Rightarrow \int f d\mu \leq \int g d\mu$

$\forall x \in X. \quad 0 \leq f_1(x) \leq f_2(x) \leq \dots \leq \infty \quad \& \quad \int f_n d\mu = f(x)$ Then

$$f \text{ is nbl} \quad \int f d\mu = \sup_n \int f_n d\mu = \lim_{n \rightarrow \infty} \int f_n d\mu$$