

Probabilistic Techniques

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Lecture 8

Continuation of L1-L7
of Martin Tancer.

- Probability: formalised \times in the world out there
a method of enumeration

- incomplete information
- future is unknown
- quantum mechanics

Literature on probability

- P. Diaconis and B. Skyrms: Ten Great Ideas about Chance

(Measurement, Judgment, Psychology, Frequency, Mathematics, Inverse Inference, Unification, Algorithmic, Randomness, Physical Chance, Induction) - no enumer.!

- G.J. Székely: Paradoxes in Probability Theory and Mathematical Statistics

- J. Haigh: Probability: A Very Short Introduction

[p. 109: "The subject of Probability is wholly free from real paradoxes."]

- E.T. Jaynes: Probability Theory: The Logic of Science (available online) • Edwin T. Jaynes (1922-1998)

- G. Shafer and V. Vovk: Game-Theoretic Foundations for Probability and Finance (prelim. version available online)

[Ch. 13.5 Getting Rich Quick with The Art of Choice]

- P. Billingsley: Probability and Measure (av. Outline) (2)
- Patrick Billingsley (1925-2011) - am. mathem. & stage and screen actor

A result on independence

Prob. space: (Ω, Σ, P_r) where $\Sigma \subset \mathcal{P}(\Omega)$ is a σ -algebra on Ω and $P_r: \Sigma \rightarrow [0, 1]$ is σ -additive and $P_r(\Omega) = 1$. ($\Rightarrow P_r(\emptyset) = 0$) (finitely add. ?!)

For example, $\Omega = \mathbb{N} = \{1, 2, 3, \dots\}$, $\Sigma = \mathcal{P}(\mathbb{N})$, $c_1, c_2, \dots \in [0, 1]$ s.t. $\sum_{i=1}^{\infty} c_i = 1$, and $P_r(A) := \sum_{i \in A} c_i$.

- Countable discrete prob. space - CDPS

Events $A_1, A_2, \dots, A_n \in \Sigma$, $n \in \mathbb{N}$, are independent:

$$P_r\left(\bigcap_{i=1}^n A_i\right) = \prod_{i=1}^n P_r(A_i).$$
 (Infinitely many)

Events $A_i \in \Sigma$, $i \in \mathbb{N}$,

such that \forall finite $I \subset \mathbb{N}$: $P_r\left(\bigcap_{i \in I} A_i\right) = \left(\frac{1}{2}\right)^{|I|}$.

That is, A_1, A_2, \dots are mutually independent tosses, flippings of a fair coin (head and tail have prob. $\frac{1}{2}$). Is there a CDPS with infinitely many independent coin tosses?

We will show that it is not.

Lemma Suppose that $A_1, A_2, \dots, A_n \in \Sigma$ are s.t. $\forall I \subset [n] (I := \{1, 2, \dots, n\}) : Pr(\bigcap_{i \in I} A_i) = (\frac{1}{2})^{|I|}$.

Then $\forall I \subset [n] : Pr(\bigcap_{i \in I} A'_i) = (\frac{1}{2})^{|I|}$, where for $i \in [n], A'_i = A_i$ or $A'_i = \bar{A}_i := \Omega \setminus A_i$.

Proof. By induction on the number $c \in \mathbb{N}_0$ of complemented events \bar{A}_i . For $c=0$ - assumed. For

$c > 0$; for instance $c=2 : Pr(A_2 \bar{A}_4 A_6 \bar{A}_7 A_9) =$
and $|I|=5$ (n as product)

$= Pr(A_2 A_6 \bar{A}_7 A_9) - Pr(A_2 A_4 A_6 \bar{A}_7 A_9)$
 $= (\frac{1}{2})^4 - (\frac{1}{2})^5 = (\frac{1}{2})^5$



Theorem In no countable discrete prob. space there are infinitely many independent coin tosses.

Proof. Suppose the space is given by $C_i \in [0,1], i \in \mathbb{N}$, and $A_n \subset \mathbb{N}, n \in \mathbb{N}$, are indep coin tosses. Let $j \in \mathbb{N}$ be arbitrary, so then $Pr(\{j\}) = C_j$. We define events $B_n \subset \mathbb{N}, n \in \mathbb{N}$, s.t. $B_n := \begin{cases} A_n \dots j \in A_n \\ \bar{A}_n = (\mathbb{N} \setminus A_n \dots j \notin A_n) \end{cases}$. Then

$\forall n \in \mathbb{N}: \{j\} \subset B_1 \cap B_2 \cap \dots \cap B_n$ and therefore (4)
 $c_j = P_r(\{j\}) \leq P_r\left(\bigcap_{i=1}^n B_i\right) \stackrel{\text{Lemma}}{=} \left(\frac{1}{2}\right)^n \rightarrow 0, n \rightarrow \infty.$

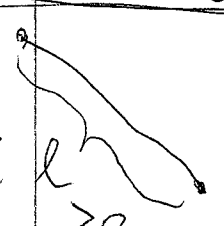
Hence $\forall j \in \mathbb{N}: c_j = 0$. This contradicts that
 $\sum_{j=1}^{\infty} c_j = 1.$ ⊠

Independence is a tricky notion (in applications not in the formalisation), see pp. 24-25 (2.1-2.2) in R. O'Donnell's LN (link on H. Tamer's page).

Thus, to model indep. coin tosses we have to turn to the uncountable prob. space $[0,1]$ where $\Omega = [0,1]$ (the unit real interval), $\Sigma \subset \mathcal{P}([0,1])$ are Lebesgue measurable sets, and $P_r: \Sigma \rightarrow [0,1]$ is the normalised Lebesgue measure (so that $P_r([0,1]) = 1$). Or a similar space. UCPS I_{meas}

tion three results/problems/paradoxes on UCPS. (This is outside the syllabus.)

① Buffon's needle problem, or Auflon's noodle problem
 Dropping randomly (i.e.) a needle of length $l > 0$



In the plane \mathbb{R}^2 ~~is~~ split in stripes by parallel lines of width d . We assume that $l \leq d$. // $\mathbb{E}X$, where $X :=$ the # of intersections (of the needle with a line) $\in \{0, 1\}$.

$P_i = P_V(\text{needle intersects a line}) = ?$

needle in 3 parts: $x_1 + x_2 + x_3 \Rightarrow X = x_1 + x_2 + x_3$, by linearity of expectation: $\mathbb{E}X = \mathbb{E}x_1 + \mathbb{E}x_2 + \mathbb{E}x_3$

$x_i \Rightarrow \mathbb{E}x_i = \mathbb{E}x_1, \mathbb{E}x_2 = \mathbb{E}x_2, \mathbb{E}x_3 = \mathbb{E}x_3$
 \Rightarrow For $X' := x_1' + x_2' + x_3'$ one has that $\mathbb{E}X' = \mathbb{E}X$

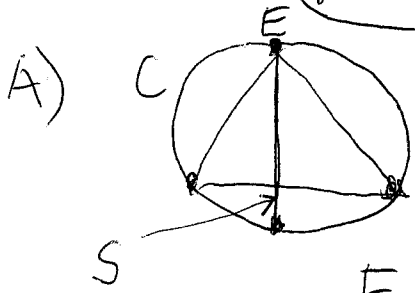
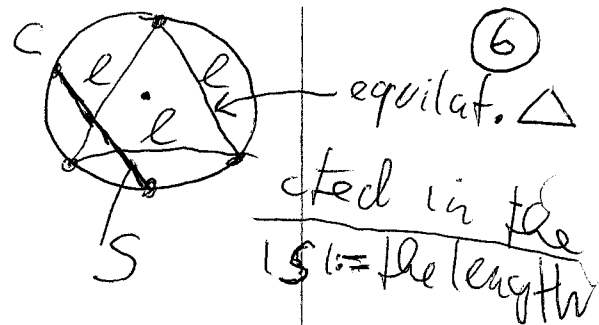
a needle \Rightarrow a noodle (of the same length!) Best noodle is the circle of length l , thus with radius $r = \frac{l}{2\pi}$.

radius $r = \frac{l}{2\pi}$. $\mathbb{E}X = \mathbb{E}X'$.

$\Rightarrow \mathbb{E}X' = 2 \cdot \frac{2r}{d} = \frac{l}{d} \cdot \frac{2}{\pi} = \mathbb{E}X = P$ \square

(2) Bertrand paradox (1889)

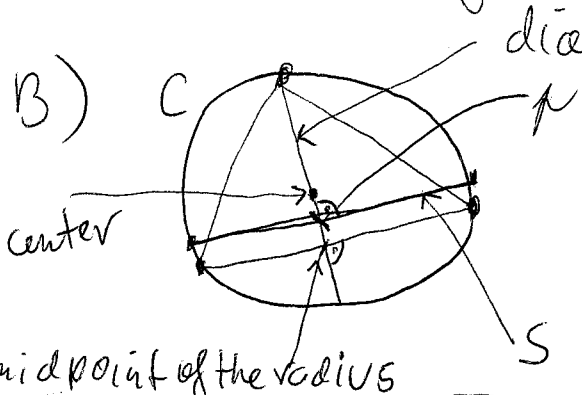
A random (?) chord S is selected in circle C . $P := P_r(|S| > l) = ?$



The equil. Δ is rotated so that one of its vertices coincides with ~~the~~ endpoint E of S

E of $S \Rightarrow P = \frac{1}{3}$

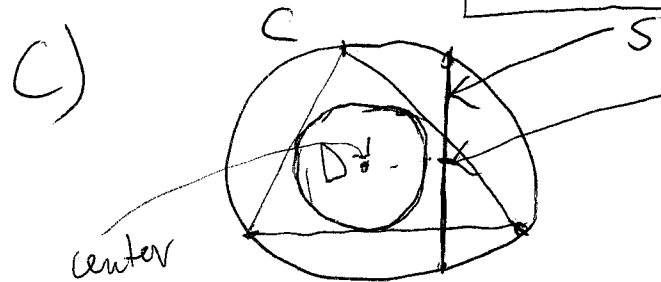
random endpoints method



The equil. Δ has side \perp to the diameter, the side bisects the diameter

random diameter method

the radius $\Rightarrow P = \frac{1}{2}$



the midpoint M of S

$|S| > l \Leftrightarrow M \in D$

random midpoint method $\text{area}(D) = \frac{1}{4} \text{area}(C)$

$\Rightarrow P = \frac{1}{4}$

$\frac{1}{3}, \frac{1}{2}, \frac{1}{4}, \dots, ?$

(3) The theorem of Pavel Valtr, and my problem related to it.

Let $C_n = \frac{1}{n} \binom{2n-2}{n-1}$ be the n -th Catalan number,

$(C_n)_{n \geq 1} = (1, 1, 2, 5, 14, 42, \dots)$

P. Valtr, 1995: $P_n = \Pr \left(\text{convex chain} \mid \text{convex n-gon} \right) = \frac{1}{C_n} \quad (7)$

(= $1/C_5 = \frac{1}{14}$)

(prob. that n random points in a square form a convex chain, if they already form a convex n -gon)

Clearly, $P_3 = \Pr \left(\text{convex chain} \mid \text{convex 3-gon} \right) = \frac{1}{2} = \frac{1}{C_3}$, due to

the symmetry \Downarrow .

PROBLEM
(OPEN)

Prove in the same way the general case

The lower bound on the Ramsey number $R(q, q) \in \mathbb{N}$ is revisited. Recall that $R(q) := \min N$ s.t. $\forall G = (V, E)$ with $|V| \geq N$: $G \supset \text{q-clique}$ or $G \supset \text{q-independent set}$.

Theorem (Erdős)

$(q \geq 3) \Rightarrow R(q) \geq \frac{q}{2e} 2^{q/2}$

Proof. $G = (V, E)$ with $|V| = N, q \in \mathbb{N}$. Then

the # (of G s.t. $G \supset \text{q-clique}$) $\leq \binom{N}{q} 2^{\binom{q}{2}}$; same bound holds for q -indep. set. Thus if N is s.t. $2 \binom{N}{q} 2^{\binom{q}{2}} < 2^{\binom{N}{2}} =$ the # (all $G = (V, E)$ with $V = \{1, 2, \dots, N\}$), then $R(q) > N$.

We need to solve the inequality. We use the estimate: $\binom{N}{q} \leq \left(\frac{eN}{q}\right)^q$. Then N is as we want if

$$\left(\frac{2N}{2}\right)^2 < \frac{1}{2} 2^{\binom{2}{2}} \iff N < \frac{2}{e} 2^{\frac{2-1}{2} - \frac{1}{2}} \geq \frac{2}{2e} 2^{1/2} \quad (8)$$

- purely enumerative proof; probability is only a way of counting things. Similarly, for $p \in [0,1]$ the random graph $G(p) = G(n,p)$ on n vertices is just a set of pairs

$$G(n,p) = \left\{ (E, w(E)) \mid E \subset \mathcal{P}\left(\binom{[n]}{2}\right) \right\}, \text{ where}$$

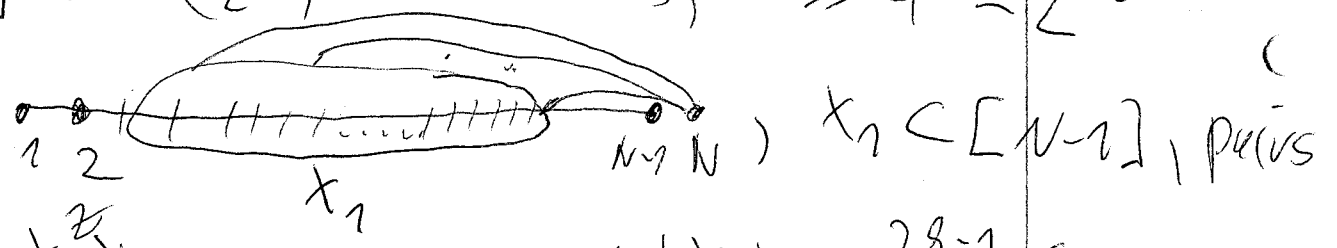
$$\binom{[n]}{2} = \{A \mid A \subset \{1,2,\dots,n\}, |A|=2\} \text{ and } w(E) = p^{|E|} (1-p)^{\binom{[n]}{2} - |E|}$$

Then we can do some weighted counting on $G(n,p)$...

$R(\beta)$ is defined (why not consider just $G \supset \text{graph?}$)

Theorem $R(\beta) \leq 4^\beta, \beta \in \mathbb{N}$.

Proof. $\chi: \binom{[N]}{2} \rightarrow \{\text{red, blue}\}, N \geq 4^\beta = 2^{2\beta}$



$\{x, N\}, x \in X_1$, monochromatic, and $|X_1| \geq 2^{2\beta-1}$. Same argument with $X_1 \setminus \{\max(X_1)\}$ and $\max(X_1)$ in place of $[N-1]$ and N , we get $X_2 \subset X_1$ s.t. all $\{x, \max(X_1)\}$ are χ -monochr. and $|X_2| \geq 2^{2\beta-2}, \dots \implies$ Sets $[N] \supset X_1 \supset X_2 \supset \dots \supset X_{2\beta}$.

Let $\Upsilon := \{N > \max(X_1) > \max(X_2) > \dots > \max(X_{2\beta})\}$ - is χ -max-monochr.; for $A \in \binom{[N]}{2}$, $\chi(A)$ depends only on $\max(A)$

$\Rightarrow \exists Z \subset Y: |Z| \geq n+1$ and $\chi|_Z = \text{const.}$ \square (9)

(We actually proved that $R(n) \leq 4^{n-1}$)

Big and hard open problem: Improve the bases of the exponential bounds ($\sqrt{2}$ and 4)

$$a(n) (\sqrt{2})^n < R(n) < b(n) 4^n, \quad n \in \mathbb{N},$$

(where $a, b: \mathbb{N} \rightarrow \mathbb{R}_{>0}$ are subexponential functions ($\forall c > 1: a(x), b(x) < c^{\sqrt{x}}$ for $x > x_0 = x_0(c)$).

Current records in bounds on $R(n)$:

$$R(n) > (1+o(1)) \frac{\sqrt{2}}{e} n 2^{n/2} \quad (\text{J. Spencer, by 1975 LLL})$$

$$R(n) < e^{-c \log^2 n} \binom{2n}{n} \quad (\text{A. Sah, 2020})$$

Probab. method tool $\sim \frac{c}{\sqrt{n}} 4^n$

using quasirandomness.

$c, c' > 0$
Lovász
Local
Lemma

(LLL also means Lenz-Lenztra

(Lovász algorithm) $A_1, A_2, \dots, A_n \in \Sigma$ is mutually independent of $A_1, \dots, A_n: \forall I \subset [n]: \Pr(A_n \cap A_I) =$

$\Pr(A) \Pr(\bigcap_{i \in I} A_i)$. Depends on digraph

A_1, \dots, A_n is $D = ([n], E) = \dots$

General LLL $A_1, \dots, A_n \in \mathcal{A}$, $D = ([n], E)$

digraph of A_1, \dots, A_n , $x_i \in [0, 1]$ for $i \in [n]$ a

$\Pr(A_i \leq x_i \mid \bigcap_{(i,j) \in E} (1-x_j))$. Then

$$\Pr\left(\bigcap_{i=1}^n A_i\right) \geq$$

$$\prod_{i=1}^n (1-x_i) > 0.$$

Proof. Next time

Thank you!