Lecture Notes

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Math 170E Lecture Notes

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1 Lecture 1

The goal of this class is to quantify randomness. The main topics for the term are:

- 1. The fundamentals of probability theory, including conditional probability and enumeration arguments.
- 2. Discrete and continuous random variables.
- 3. Sequences of i.i.d. random variables, including the Weak Law of Large Numbers and the Central Limit Theorem.

1.1 Properties of Probability

Probability theory takes place inside a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Definition. Probability Space

A probability space is a triplet $(\Omega, \mathcal{F}, \mathbb{P})$ satisfying:

- 1. A non-empty set Ω , called the *sample space*.
- 2. A set \mathcal{F} of subsets of Ω satisfying certain properties:
 - Elements of \mathcal{F} are called *events*.
 - Events A_1, A_2, \ldots, A_k are called mutually exclusive if they are pairwise disjoint, i.e. if $i \neq j$ then $A_i \cap A_j = \emptyset$.
 - Events A_1, A_2, \ldots, A_k are called *exhaustive* if their union is the sample space, i.e.

$$\bigcup_{j=1}^{k} A_j = \Omega.$$

- For this class you may ignore \mathcal{F} and assume that all subsets of Ω are events.
- 3. A function $\mathbb{P} \colon \mathcal{F} \to [0,1]$, called a *probability measure*, which satisfies:
 - $\mathbb{P}[\Omega] = 1$, or "the probability that something happens is 1".
 - If A_1, A_2, \ldots, A_n are mutually exclusive events, then

$$\mathbb{P}\left[\bigcup_{j=1}^{n} A_j\right] = \sum_{j=1}^{n} \mathbb{P}[A_j].$$

• If A_1, A_2, \ldots are mutually exclusive events, then

$$\mathbb{P}\left[\bigcup_{j=1}^{\infty} A_j\right] = \sum_{j=1}^{\infty} \mathbb{P}[A_j].$$

Example. Suppose I flip two fair coins. Then the sample space can be written as $\Omega = \{HH, HT, TH, TT\}$. The probability measure should be defined as

$$P[HH] = \frac{1}{4}$$

$$P[HT] = \frac{1}{4}$$

$$P[TH] = \frac{1}{4}$$

$$P[TT] = \frac{1}{4}$$

The probability of getting exactly one head is hence $\mathbb{P}[\{HT, TH\}] = \mathbb{P}[HT] + \mathbb{P}[TH] = \frac{1}{2}$.

Theorem. $\mathbb{P}[\varnothing] = 0$.

Proof. We know that Ω and \varnothing are mutually exclusive, since, $\Omega \cap \varnothing = \varnothing$. Thus

$$\begin{split} \mathbb{P}[\Omega] &= \mathbb{P}[\Omega \cup \varnothing] \\ &= \mathbb{P}[\Omega] + \mathbb{P}[\varnothing], \end{split}$$

and so $\mathbb{P}[\varnothing] = 0$.

Theorem. If $A \subseteq \Omega$ is an event and $A' = \Omega \setminus A$ then

$$\mathbb{P}[A] = 1 - \mathbb{P}[A'].$$

Proof. Since we have that $A' = \Omega \setminus A$, we know that $A' \cap A = \emptyset$, so they are mutually exclusive. Thus we have

$$\begin{split} \mathbb{P}[\Omega] &= \mathbb{P}[A \cup A'] \\ 1 &= \mathbb{P}[A] + \mathbb{P}[A'] \\ \mathbb{P}[A] &= 1 - \mathbb{P}[A']. \end{split}$$

Theorem. If $A \subseteq B$ then

$$\mathbb{P}[B\backslash A] = \mathbb{P}[B] - \mathbb{P}[A].$$

Proof. We know that $B = A \cup (B \backslash A)$ and $A \cap (B \backslash A) = \emptyset$. Hence

$$\mathbb{P}[B] = \mathbb{P}[A \cup (B \backslash A)] = \mathbb{P}[A] + \mathbb{P}[B \backslash A],$$

and the result follows.

Theorem. If $A \subseteq B$ then $\mathbb{P}[A] \leq \mathbb{P}[B]$.

Proof. From the previous theorem we have

$$\mathbb{P}[A] \leqslant \mathbb{P}[A] + \mathbb{P}[B \backslash A] = \mathbb{P}[B].$$

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2 Lecture 2

2.1 Inclusion-Exclusion Principle

Theorem — Inclusion-Exclusion Principle

If $A, B \subseteq \Omega$ are events, then

$$\mathbb{P}[A \cup B] = \mathbb{P}[A] + \mathbb{P}[B] - \mathbb{P}[A \cap B].$$

Proof. Observe that we may write

$$A \cup B = [A \setminus (A \cap B)] \cup [A \cap B] \cup [B \setminus (A \cap B)],$$

where $A \cap B$, $B \setminus (A \cap B)$, and $A \setminus (A \cap B)$ are mutually exclusive. Hence

$$\begin{split} \mathbb{P}[A \cup B] &= \mathbb{P}[A \backslash (A \cap B)] + \mathbb{P}[B \backslash (A \cap B)] + \mathbb{P}[A \cap B] \\ &= (\mathbb{P}[A] - \mathbb{P}[A \cap B]) + (\mathbb{P}[B] - \mathbb{P}[A \cap B]) + \mathbb{P}[A \cap B] \\ &= \mathbb{P}[A] + \mathbb{P}[B] - \mathbb{P}[A \cap B]. \end{split}$$

Theorem — Union Bound

If $A_1, A_2, \ldots, A_n \subseteq \Omega$ are events, then

$$\mathbb{P}\left[\bigcup_{j=1}^{n} A_j\right] \leqslant \sum_{j=1}^{n} \mathbb{P}[A_j].$$

Proof. We proceed via proof by induction. Observe that for n = 1, we have $\mathbb{P}[A_1] \leq \mathbb{P}[A_1]$, which is obviously true. Suppose that this statements holds for some $k \geq 1$. Then

$$\mathbb{P}\left[\bigcup_{j=1}^{k+1} A_j\right] = \mathbb{P}\left[\left(\bigcup_{j=1}^{k} A_j\right) \cup A_{k+1}\right]$$

$$= \mathbb{P}\left[\bigcup_{j=1}^{k} A_j\right] + \mathbb{P}[A_{k+1}] - \mathbb{P}\left[\left(\bigcup_{j=1}^{k} A_j\right) \cap A_{k+1}\right]$$

$$\leqslant \mathbb{P}\left[\bigcup_{j=1}^{k} A_j\right] + \mathbb{P}[A_{k+1}]$$

$$\leqslant \sum_{j=1}^{k} \mathbb{P}[A_j] + \mathbb{P}[A_{k+1}]$$

$$= \sum_{j=1}^{k+1} \mathbb{P}[A_j].$$

Hence the statement holds for k + 1, and so holds for all natural numbers n.

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2.2 Mutual Independence

 $\textbf{Definition.}\ \ Independence$

We say that two events $A, B \subseteq \Omega$ are independent if

$$\mathbb{P}[A \cap B] = \mathbb{P}[A]\mathbb{P}[B].$$

If two events are not independent, then we say that they are dependent.

Definition. Mutual Independence

We say that events $A_1, \ldots, A_n \subseteq \Omega$ are mutually independent if, given any $1 \le k \le n$ and $1 \le j_1 < j_2 \cdots < j_k \le n$ we have

$$\mathbb{P}\left[\bigcap_{\ell=1}^k A_{j_\ell}\right] = \prod_{\ell=1}^k \mathbb{P}[A_{j_\ell}].$$

3 Lecture 3

Theorem — Multiplication Principle

Suppose I run r mutually independent experiments so that

- The 1st experiment has n_1 possible outcomes.
- The 2^{nd} experiment has n_2 possible outcomes.
- ...
- The r^{th} experiment has n_r possible outcomes.

The composite experiment then has $n_1 \cdot n_2 \cdot \cdots \cdot n_r$ outcomes.

In some experiments we care about taking samples of size r from a set of n objects.

- We can seek ordered or unordered samples.
- We can do this with or without replacement.

Theorem. There are n^r possible choices of an *ordered* sample of size r from a set of n objects with replacement.

Proof. We run r experiments corresponding to each choice. For each choice, we have n possible outcomes because we are performing the choices with replacement. The Multiplication Principle tells us that there are $n \cdot n \cdot \dots \cdot n = n^r$ outcomes.

Theorem. There are

$$_{n}P_{r} = \frac{n!}{(n-r)!}$$

ordered samples of size r from a set of n objects without replacement. The number ${}_{n}Pr$ is known as the number of permutations of n objects, taken r at a time.

Proof. Each choice is an independent experiment:

- 1st choice: n outcomes
- 2^{nd} choice: n-1 outcomes
- 3^{rd} choice: n-2 outcomes
- r^{th} : n-(r-1) outcomes

Hence the composite experiment has

$$n \cdot (n-1) \cdot \cdots \cdot (n-r+1) = {}_{n}P_{r}$$

outcomes.

Theorem. There are

$${}_{n}C_{r} = \frac{n!}{(n-r)!r!}$$

unordered samples of size r from a set of n objects without replacement.

Proof. From the previous theorem, there are ${}_{n}P_{r}$ ordered samples of size r from n objects without

replacement. However, we have over counted because our sample will show up r! times (in every possible permutation). Hence we divide by r! to get

$$_{n}C_{r} = \frac{_{n}P_{r}}{r!} = \frac{n!}{(n-r)!r!}.$$

Note. ${}_nC_r = {}_nC_{n-r}$.