

## MATH 3100 FALL 2020. LECTURE SUMMARIES

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### 1. 8/25

Section 1.1 in the textbook.

- (1) Sample space — long abstract definition which encompasses all possible mathematical models of randomness we are going to see in the course
- (2) Examples of sample spaces — coin tossing, dice rolling.
- (3) We are discussing finite sample spaces so far. Out of finite sample spaces, a special case is formed by *finite sample spaces with equally likely outcomes*. In them, we have  $P(\omega) = \frac{1}{\#\Omega}$  for all  $\omega \in \Omega$ , and  $P(A) = \frac{\#A}{\#\Omega}$  for all events  $A$ .
- (4) Repeated experiments, sample space  $\Omega^n = \Omega \times \dots \times \Omega$  (Cartesian power), where

$$\Omega^n = \{(a_1, \dots, a_n) : a_i \in \Omega\}$$

is the space of ordered  $n$ -tuples of elements from  $\Omega$ . The sample space  $\Omega^n$  models the experiment corresponding to  $\Omega$ , repeated (independently)  $n$  times.

- (5) Finer point. In uncountable sample spaces, usually it is not possible to define  $P$  consistently for all subsets. Therefore, we need to restrict the definition of event to “good” subsets of  $\Omega$ .

### 2. 8/27

Section 1.2 in the textbook.

- (1) Random sampling. We stay in the scenario with finite sample spaces, equally likely outcomes.
- (2) We discuss three main sampling schemes of  $k$  objects out of  $n$  objects.
- (3) If we sample with replacement and order matters, then  $\#\Omega = n^k$
- (4) If we sample without replacement and order matters, then

$$\#\Omega = n(n-1)\dots(n-k+1) = \frac{n!}{(n-k)!}.$$

If  $k = n$ , we talk about random permutations of  $n$  objects.

- (5) If we sample without replacement and order does not matter, then

$$\#\Omega = \frac{n(n-1)\dots(n-k+1)}{k!} = \frac{n!}{(n-k)!k!} = \binom{n}{k}.$$

- (6) Hypergeometric distribution. Imagine we have  $n$  objects separated into a number of types  $1, \dots, L$ , and there are  $m_j$  objects of type  $j$ . So that  $m_1 + \dots + m_L = n$ . Sample  $k$  objects at random from  $n$ . The probability that in this sample there are  $p_j$  objects of type  $j$  is equal

to

$$\frac{\binom{m_1}{p_1} \cdots \binom{p_L}{m_L}}{\binom{n}{k}},$$

where  $p_1 + \dots + p_L = k$ .

### 3. 9/1

Sections 1.3 and 1.4 in the textbook.

- (1) Geometric distribution  $P(k) = p^{k-1}(1-p)$ ,  $k = 1, 2, \dots$ . This is an example of an infinite  $\Omega$ . Here  $\Omega$  is countable. Countable and finite sample spaces have a special unifying name, “discrete sample spaces”.
- (2) Geometric series, its sum =  $\frac{\text{first term}}{1 - \text{ratio}}$ .
- (3) Continuous uniform distribution on  $[0, 1]$  — another example of an infinite  $\Omega$ . This  $\Omega$  is uncountable.
- (4)  $P(A)$  behaves like area of the event, both in continuous uniform case and in general (in some sense).
- (5) Operations on events and their probabilities: decomposition, complement, monotonicity, inclusion-exclusion.

### 4. 9/3

Sections 1.4 and 1.5 in the textbook.

- (1) Operations on events and their probabilities, and corresponding examples: decomposition, complements, monotonicity, inclusion-exclusion.
- (2) For monotonicity, a proof that we will see T with probability 1, after repeatedly tossing a fair coin.
- (3) For inclusion-exclusion, discussed a hard problem of computing the probability that no one has their own hat, if the hats are randomly permuted.
- (4) Random variable is a function on the sample space. This is the second fundamental definition of the course.
- (5) Discussed the definition, examples of random variables on discrete and continuous sample spaces.
- (6) Probability mass function (for discrete random variables). Probability distribution.

### 5. 9/8

Conditional probability and Bayes’ rule (Sections 2.1 and 2.2)

- (1) Definition of conditional probability  $P(A | B)$
- (2) Multiplication rule  $P(AB) = P(B)P(A | B)$ .
- (3) Law of total probability  $P(A) = \sum_{k=1}^N P(A | B_k)P(B_k)$ , where  $\Omega = \bigcup_{k=1}^N B_k$  is a partition of the sample space.
- (4) Bayes’s formula

$$P(B | A) = \frac{P(A | B)P(B)}{P(A)} = \frac{P(A | B)P(B)}{P(A | B)P(B) + P(A | B^c)P(B^c)}.$$

## 6. 9/10

Independence (section 2.3).

- (1) Some hints on the most challenging problems from Problem Set 3.
- (2) Independence. Independence algebraically means product rule.
- (3) Independence of two events.
- (4) Mutual independence and pairwise independence of several events.
- (5) Electric circuits example.
- (6) Model of arbitrary many independent events with  $P(A_j) = \frac{1}{2}$  on  $\Omega = [0, 1]$ .
- (7) Independence of random variables.

## 7. 9/15

Section 2.4 and some parts from 2.5.

- (1) Recall independence.
- (2) Independent events from independent events, for example,  $AB^c$  and  $C^c$  are independent if  $A, B, C$  are mutually independent.
- (3) Independent trials. Sample space.
- (4) Proof that all the probabilities sum to one.
- (5) Bernoulli, binomial, geometric distributions.
- (6) Conditional independence (brief discussion).
- (7) Hypergeometric distribution.

## 8. 9/17

Section 3.1.

- (1) Probability mass function, pmf (for discrete distributions). Examples, properties.
- (2) Probability density function, pdf (for continuous distributions). Examples, properties.
- (3) (Continuous) uniform distribution.
- (4) Pdf as a derivative / infinitesimal description.
- (5) Example with a uniform point in a disc and the pdf for  $R$ , the distance from the point to 0.