## 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  $\Omega^n = \Omega \times ... \times \Omega$  (Cartesian power), where

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

- 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  $\Omega$ .

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  $1, \ldots, L$ , and there are  $m_j$  objects of type j. So that  $m_1 + \ldots + m_L = n$ . 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  $p_1 + \ldots + p_L = k$ .

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
- (5) Operations on events and their probabilities: decomposition, complement, monotonicity, inclusion-exclusion.

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
- (6) Probability mass function (for discrete random variables). Probability distribution.

Conditional probability and Bayes' rule (Sections 2.1 and 2.2 d)

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

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

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- (1) Some hints on the most challenging problems from Problem Set 3.
- (2)