

Study Notes of Matrix and Tensor

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Preface

The notes mainly refer to:

- Introduction to Mathematical Statistics 8th Edition
- [lecture note](#)
- [Study Guide](#)

Chapter 1

Probability and Distributions

1.1 Introduction

Definition 1.1

If an experiment can be repeated under the same conditions it is a random experiment. The set of every possible outcome of an experiment is the sample space, denoted \mathcal{C} .

Remark. For an experiment, the sample space is not unique. For example, When talking about the temperature in an area, we can define the sample space as $\mathcal{C} = (-\infty, \infty)$ or $\mathcal{C} = [a, b]$. For a specific random experiment, we can use different sample spaces to describe it. However, it is worth studying how to describe it with an appropriate sample space.

Note/Definition. Notationally, we denote the elements of the sample space with lower case letters such as a, b, c . Subsets of the sample space are *events* and we denote them with upper case letters such as A, B, C .

Definition 1.2

If an experiment is performed N times and a specific event occurs f times, then f is the frequency of the event and f/N is the relative frequency of the event.

1.2 Sets

1.3 The Probability Set Function

We need to define a set function that assigns a probability to the events (subsets of sample space \mathcal{C}). We denote the collection of events as \mathcal{B} . If \mathcal{C} is finite set, then we hope to assign a probability to all events (that is, to define a probability set function on the power set of \mathcal{C}). More generally, we require that \mathcal{B} (the collection of events) to satisfy: (1) the sample space \mathcal{C} itself is an event, (2) the complement of every event is again an event, and (3) every countable union of events is again an event. Symbolically, this means (1) $\mathcal{C} \in \mathcal{B}$, (2) if $A \in \mathcal{B}$ then $A^c \in \mathcal{B}$, and (3) if $A_1, A_2, \dots \in \mathcal{B}$ then $\bigcup_{n=1}^{\infty} A_n \in \mathcal{B}$. Combining (2) and (3), we see by DeMorgan's Law (for countable unions) that

if $A_1, A_2, \dots \in \mathcal{B}$ then $\cap_{n=1}^{\infty} A_n \in \mathcal{B}$. So the collection of events \mathcal{B} is closed under complements, countable unions, and countable intersections. Such a collection of sets form a σ -algebra.

Definition 1.3

A collection of events $\{A_n | n \in I\}$ (where I is some indexing set) such that $A_i \cap A_j = \emptyset$ is a mutually exclusive collection of events.

Definition 1.4

Let \mathcal{C} be a sample space and let \mathcal{B} be the set of all events (thus, \mathcal{B} is a σ -field). Let P be a real-valued function defined on \mathcal{B} . Then P is a probability set function if P satisfies the following three conditions:

(1) $P(A) \geq 0$ for $A \in \mathcal{B}$.

(2) $P(\mathcal{C}) = 1$.

(3) If $\{A_n\}$ is a mutually exclusive collection of events, then $P(\cup_{n=1}^{+\infty} A_n) = \sum_{n=1}^{+\infty} P(A_n)$.

Theorem 1.1

For each event $A \in \mathcal{B}$, $P(A) = 1 - P(A^c)$.

Theorem 1.2

The probability of the null set is zero; that is, $P(\emptyset) = 0$.

Theorem 1.3

If A and B are events such that $A \subset B$, then $P(A) \leq P(B)$.

Theorem 1.4

For each event $A \in \mathcal{B}$ we have $0 \leq P(A) \leq 1$.

Theorem 1.5

If A and B are events in \mathcal{C} , then $P(A \cup B) = P(A) + P(B) - P(A \cap B)$.

Theorem 1.6

Let $\{A_n\}$ be a nondecreasing sequence of events (ie. $A_n \subseteq A_{n+1}$). Then

$$\lim_{n \rightarrow \infty} P(A_n) = P(\lim_{n \rightarrow \infty} A_n) = P(\cup_{n=1}^{\infty} A_n).$$

Let $\{A_n\}$ be a nonincreasing sequence of events (ie. $A_n \supseteq A_{n+1}$). Then

$$\lim_{n \rightarrow \infty} P(A_n) = P(\lim_{n \rightarrow \infty} A_n) = P(\cap_{n=1}^{\infty} A_n).$$

Theorem 1.7

Let $\{A_n\}$ be an arbitrary sequence of events. Then

$$P(\cup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} P(A_n).$$

1.4 Homework

Exercise 1.1

Show that the moment generating function of the random variable X having the pdf $f(x) = \frac{1}{3}$, $-1 < x < 2$, zero elsewhere, is

$$M(t) = \begin{cases} \frac{e^{2t}-e^{-t}}{3t} & t \neq 0 \\ 1 & t = 0. \end{cases}$$

1.5 Reference

- [lecture note](#)
- [Probability and Distributions](#)
- [Sample space is unique?](#)
- [proof of 1.3](#)

Chapter 2

Multivariate Distributions

2.1 Homework

Exercise 2.1

Let the joint pdf of X and Y be given by

$$f(x, y) = \begin{cases} \frac{2}{(1+x+y)^3} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{elsewhere.} \end{cases}$$

(a) Compute the marginal pdf of X and the conditional pdf of Y , given $X = x$. (b) For a fixed $X = x$, compute $E(1 + x + Y|x)$ and use the result to compute $E(Y|x)$.

Exercise 2.2

Let X_1, X_2, X_3 be iid with common pdf $f(x) = \exp(-x)$, $0 < x < \infty$, zero elsewhere. Evaluate:

(a) $P(X_1 < X_2 | X_1 < 2X_2)$.

(b) $P(X_1 < X_2 < X_3 | X_3 < 1)$.