Statistical and Mathematical Methods for Data Analysis

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Textbooks

- **Probability & Statistics for Engineers & Scientists**, Ninth Edition, Ronald E. Walpole, Raymond H. Myer

- **Elementary Statistics: Picturing the World,** 6th Edition, Ron Larson and Betsy Farber

- **Elementary Statistics,** 13th Edition, Mario F. Triola

Reference books

- **Probability and Statistical Inference, Ninth Edition,** Robert V. Hogg, Elliot A. Tanis, Dale L. Zimmerman
- **Probability Demystified**, Allan G. Bluman
- **Practical Statistics for Data Scientists: 50 Essential Concepts,** Peter Bruce and Andrew Bruce
- **Schaum's Outline of Probability,** Second Edition, Seymour Lipschutz, Marc Lipson
- **Python for Probability, Statistics, and Machine Learning,** José Unpingco

References

Readings for these lecture notes:

- **Probability & Statistics for Engineers & Scientists**, Ninth edition, Ronald E. Walpole, Raymond H. Myer

These notes contain material from the above book.

Mathematical Expectation

If **two coins** are tossed **16 times** and ***X* is the number of heads** that occur per toss, then the values of ***X*** are **0, 1**, and **2**.

Suppose that the experiment yields **no heads**, **one head**, and **two heads** a total of **4, 7, and 5** times, respectively.

$$\frac{(0)\,(4) + (1)\,(7) + (2)\,(5)}{16} = 1.06$$

- This is an **average value** of the data and yet it is not a possible outcome **of {0, 1, 2}.**
- Hence, an average is **not necessarily a possible** outcome for the experiment
- For instance, a **salesman's average monthly** income is **not likely** to be equal to any of his **monthly paychecks**.

x	f	fx
0	4	0
1	7	7
2	5	10
	$\sum f = 16$	$\sum fx = 17$
$\bar{x} = \frac{17}{16}$ =1.06		

x	P(X)	xP(X)
0	$\frac{4}{16}$	0
1	$\frac{7}{16}$	$\frac{7}{16}$
2	$\frac{5}{16}$	$\frac{10}{16}$
	$\sum P(X) = \frac{16}{16} = 1$	$\sum xP(X) = \frac{17}{16} = 1.0625$

- The numbers **4/16, 7/16, and 5/16** are the fractions of the total tosses resulting in **0, 1, and 2 heads**, respectively. These **fractions** are also the **relative frequencies** for the different values of *X* in our experiment.
- In fact, then, we can **calculate the mean**, or **average**, of a set of data by knowing the distinct values that occur and their **relative frequencies**, **without any knowledge of the total number of observations in our set of data**.

- Therefore, if **4/16, or 1/4**, of the tosses result in **no heads**, **7/16** of the tosses result in **one head**, and **5/16** of the tosses **result in two heads**, the **mean number of heads per toss** would be **1.06** no matter whether the total number of **tosses were 16, 1000, or even 10,000.**

- This method of **relative frequencies** is used to calculate the **average number of heads** per toss of two coins that we might **expect in the long run**.

- We shall refer to this average value as the **mean of the random variable *X*** or the **mean of the probability distribution of *X*** and write it as μ_x or simply as μ when it is clear to which random variable we refer.

- It is also common among statisticians to refer to this **mean** as the **mathematical expectation**, or **the expected value** of the **random variable *X***, and denote it as ***E*(*X*).**

Mean or Expected value of X

Theorem: Let X be a random variable with probability distribution f(x). The **mean**, or **expected value**, of X is

$$\mu = E(X) = \sum_x xf(x)$$

if **X is discrete**, and

$$\mu = E(X) = \int_{-\infty}^{\infty} xf(x)\,dx$$

if **X is continuous**

Example : A lot containing **7 components** is sampled by a quality inspector; the lot contains **4 good components** and **3 defective components**. A **sample of 3** is taken by the inspector. Find the **expected value** of the number of **good components** in this sample.

N = 7

n = 3

k = 4

$P(X = x) = h(x; N, n, k) = ({}_kC_x)({}_{N-k}C_{n-x})/({}_NC_n)$, $\max\{0, n-(N-k)\} \le x \le \min\{n, k\}$

Let *X* represent the number of good components in the sample.

$\max\{0, n-(N-k)\} = \max\{0, 3-(7-4)\} = \max(0, 0) = 0$

$\min\{n, k\} = \min(3, 4) = 3$

x	P(X = x)	x P(X)
0	$\frac{1}{35}$	0
1	$\frac{12}{35}$	$\frac{12}{35}$
2	$\frac{18}{35}$	$\frac{36}{35}$
3	$\frac{4}{35}$	$\frac{12}{35}$
	$\sum P(X) = \frac{16}{16} = 1$	$\sum xP(X) = \frac{60}{35} = 1.7143$

Cont.

- Thus, if a sample of **size 3 is selected** at random over and over again from a lot of **4 good components** and **3 defective** components, it will contain, on average, **1.7 good components**.

Example: A salesperson for a medical device company has **two appointments** on a given day.

At the first appointment, he believes that he has a **70% chance** to make the deal, from which he can **earn $1000** commission if successful. On the other hand, he thinks he only has a **40% chance** to make the deal at the second appointment, from which, if successful, he can **make $1500**. What is his **expected commission** based on his own probability belief? Assume that the appointment results are **independent** of each other.

First appointment:

Probability of **commission** = **0.70, Commission = 1000**

Probability of **no commission** = 1- 0.70 = **0.30, Commission = 0**

Second appointment:

Probability of **commission** = **0.40, Commission = 1500**

Probability of **no commission** = 1- 0.40 = **0.60, Commission = 0**

Since appointment results are **independent.**

Let X denotes total commission from appointment 1 and appointment 2

x	P(X = x)	xP(X)
0 + 0 = 0	(0.30) (0.60) = 0.18	0
1000 + 0 = 1000	(0.70) (0.60) = 0.42	420
0 + 1500 = 1500	(0.30)(0.40) = 0.12	180
1000 + 1500 = 2500	(0.70)(0.40) = 0.28	700
	$\sum P(X) = 1$	$\sum xP(X)$ = \$1300

Example : Let ***X*** be the random variable that denotes the life in hours of a certain electronic device. The **probability density function** is

$$f(x) = \begin{cases} \dfrac{20{,}000}{x^3}, & x > 100, \\ 0, & \text{elsewhere} \end{cases}$$

Find the expected life of this type of device.

$\mu = E(X) = \int_{-\infty}^{\infty} xf(x)\,dx$

if **X is continuous**

$$\mu = E(X) = \int_{100}^{\infty} x \times \frac{20{,}000}{x^3}\,dx$$

$$= \int_{100}^{\infty} \frac{20{,}000}{x^2}\,dx$$

$$= 20{,}000 \int_{100}^{\infty} x^{-2}\,dx$$

$$= -20{,}000\,[x^{-1}]_{100}^{\infty} = -20{,}000\left(\frac{1}{\infty} - \frac{1}{100}\right)$$

$$= 200$$