# Moments and Characteristic Functions Instructor's Notes

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## 1 Moments

#### 1.1 Definition of Moments

Generally, in math, the n-th moment of a real-valued continuous function about center c is: [1]

$$\mu_n = \int_{-\infty}^{\infty} (x - c)^n f(x) dx$$

In particular, for probability density functions f (or cumulative density function F), the moments are given by

$$\mu'_n = E[X^n] = \int_{-\infty}^{\infty} x^n dF(x) = \int_{-\infty}^{\infty} x^n f(x) dx$$

Also we have the definition of the central moment [2]:

$$\mu_n = E[(X - E[X])^n] = \int_{-\infty}^{\infty} (x - \mu)^n f(x) dx$$

Generally central moments are more useful. Not to be confused with mean  $\mu$ .

## 1.2 Description of Moments

The first moment is the mean of a random variable, i.e.

$$\mu = E[X]$$

The second moment is related to the variance of a random variable:

$$Var[X] = E[X^2] - E[X]^2$$

In fact the variance is just the second central moment:

$$Var[X] = \mu_2 = E[(X - E[X])^2]$$

As for the third central moment, a related concept is skewness. Below shows two random variables with the same mean variance however different in skewness[3]:

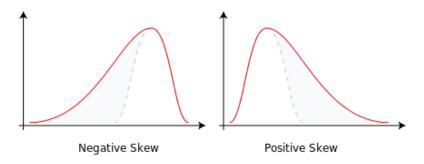


Figure 1: Negative and Positive Skew Diagrams

With all moments up to the order of infinity we can describe the **characteristics** of a probability distribution.

## 2 Characteristic Functions

## 2.1 Moment Generating Functions

**Definition 2.1** Let X be a random variable with probability density function f(x). If there is a positive number h such that

$$\int_{-\infty}^{\infty} e^{tx} f(x) dx$$

exists and is finite for h < t < h, then the function defined by

$$M(t) = E[e^{tX}]$$

is called the moment-generating function of X (or of the distribution of X). [4]

The r-th moment about the origin can be achieved from the moment generating function by evaluating the r-th derivative[5]:

$$M^{(r)}(0) = E[X^r]$$

Also notice the relation between the Taylor Expansion and the moments.

#### 2.2 Characteristic Function

Notice that  $e^{tx}$  is not a "good" function in the sense that it is not bounded and may not converge under some circumstances. Before going to characteristic functions, we first get acquainted with knowledge of complex numbers:

#### 2.2.1 Basic information about complex numbers

Let z=a+bi, where  $a,b\in\mathbb{R}$ , and  $i=\sqrt{-1}$  is the imaginary unit. z is then called a complex number and a,b are called the real and imaginary parts of z, denoted by  $a=\mathrm{Re}(z),b=\mathrm{Im}(z)$  respectively. (Consider i as rotation by  $\frac{\pi}{2}$  counterclockwise in the complex plane)

The conjugate of a complex number  $z=a+bi, a,b\in\mathbb{R}$  is  $\hat{z}=a-bi$ , we also define the modulus (or length) of z to be  $|z|=z\hat{z}$ . Notice that |z| is a non-negative real number. Euler's formula:

$$e^{i\theta} = \cos\theta + i\sin\theta$$

The formula comes from Taylor's Series. It also gives rise to the polar representation of a complex number, i.e.  $z=re^{i\theta}$ , where r is the modulus and  $\theta$  is the phase. From this we also have that  $|e^{i\theta}|=1$  for any  $\theta$ .

#### 2.2.2 Definition of Characteristic Functions

**Definition 2.2** Let X be a random variable and denote by F the cumulative distribution function of X. The characteristic function  $\varphi = \varphi_X$  of X (or of F, in which case we also write  $\varphi_F$ ) is defined by [6]

$$\varphi_X(t) := E[e^{itX}] = \int_{-\infty}^{\infty} e^{itx} dF(x), t \in \mathbb{R}$$

## 2.2.3 Basic Properties [6]

**Theorem 2.1** If X and Y are independent random variables then the characteristic function of their sum is

$$\varphi_{X+Y}(t) = \varphi_X \cdot \varphi_Y.$$

Corollary 2.1.1 The product of two characteristic functions is a characteristic function

**Remark** If X and Y are random variables such that  $\varphi_{X+Y} = \varphi_X \cdot \varphi_Y$ , then in general we do not conclude X and Y are independent. (See page 13 in [6])

**Theorem 2.2** For any  $a, b \in \mathbb{R}$ ,

$$\varphi_{aX+b}(t) = e^{ibt}\varphi_X(at).$$

**Theorem 2.3** Every characteristic function  $\varphi$  has the following properties:

- (i) f(0) = 1,
- (ii) |f(t)| < 1,
- (iii)  $f(-t) = \overline{f(t)}$
- (iv) f is continuous on  $\mathbb{R}$

#### 2.3 Common Distributions and Their Characteristic Functions

## 3 Examples and Applications of Characteristic Functions

To be continued...

### References

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