

# Mathematical statistics

February 22<sup>nd</sup>, 2018

Lecture 5: Statistics and sampling distribution (cont.)

**Week 1** .....● Probability reviews

**Week 2** .....● **Chapter 6: Statistics and Sampling Distributions**

**Week 4** .....● Chapter 7: Point Estimation

**Week 7** .....● Chapter 8: Confidence Intervals

**Week 10** .....● Chapter 9: Test of Hypothesis

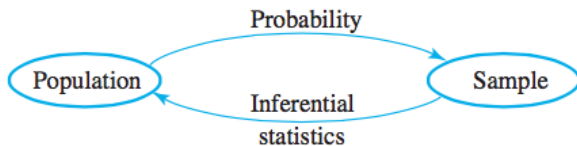
**Week 14** .....● Regression

6.1 Statistics and their distributions

6.2 The distribution of the sample mean

6.3 The distribution of a linear combination

Order 6.1  $\rightarrow$  6.3  $\rightarrow$  6.2



## Definition

The random variables  $X_1, X_2, \dots, X_n$  are said to form a (simple) random sample of size  $n$  if

- 1 the  $X_i$ 's are independent random variables
- 2 every  $X_i$  has the same probability distribution

## Definition

A statistic is any quantity whose value can be calculated from sample data

- prior to obtaining data, there is uncertainty as to what value of any particular statistic will result → a statistic is a random variable
- the probability distribution of a statistic is referred to as its *sampling distribution*

# Example of a statistic

- Let  $X_1, X_2, \dots, X_n$  be a random sample of size  $n$
- The sample mean of  $X_1, X_2, \dots, X_n$ , defined by

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n},$$

is a statistic

- When the values of  $x_1, x_2, \dots, x_n$  are collected,

$$\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n},$$

is a realization of the statistic  $\bar{X}$

# Example of a statistic

- Let  $X_1, X_2, \dots, X_n$  be a random sample of size  $n$
- The random variable

$$T = X_1 + 2X_2^2 + 3X_5^3$$

is a statistic

- When the values of  $x_1, x_2, \dots, x_n$  are collected,

$$t = x_1 + 2x_2^2 + 3x_5^3,$$

is a realization of the statistic  $T$

# Questions for this chapter

Given a random sample  $X_1, X_2, \dots, X_n$ , and

$$T = a_1X_1 + a_2X_2 + \dots + a_nX_n$$

- If we **know** the distribution of  $X_i$ 's, can we obtain the distribution of  $T$ ?
- If we **don't know** the distribution of  $X_i$ 's, can we still obtain/approximate the distribution of  $T$ ?



# Example 1

## Problem

Consider the distribution  $P$

$x$	$10$	$15$	$20$
$p(x)$	$0.2$	$0.3$	$0.5$

Let  $\{X_1, X_2\}$  be a random sample of size 2 from  $P$ , and  $T = X_1 + X_2$ .

- 1 Compute  $P[T = 100]$
- 2 Derive the probability mass function of  $T$
- 3 Compute the expected value and the standard deviation of  $T$

## Example 2

### Problem

Let  $\{X_1, X_2\}$  be a random sample of size 2 from the exponential distribution with parameter  $\lambda$

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

and  $T = X_1 + X_2$ .

What is the distribution of  $T$ ?

For continuous random variable:

$$F_X(t) = P(X \leq t) = \int_{-\infty}^t f(x) dx$$

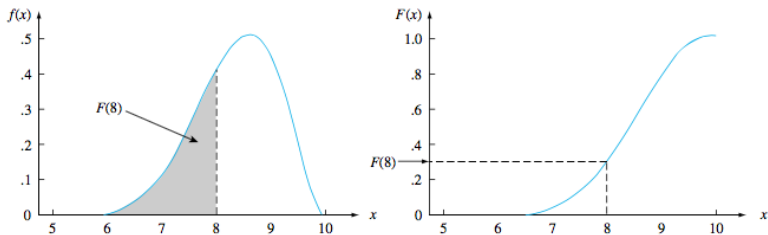


Figure 4.5 A pdf and associated cdf

Moreover:

$$f(x) = F'(x)$$

## Example 2

### Problem

Let  $\{X_1, X_2\}$  be a random sample of size 2 from the exponential distribution with parameter  $\lambda$

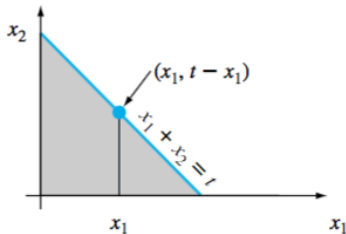
$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

and  $T = X_1 + X_2$ .

- 1 Compute the cumulative density function (cdf) of  $T$

## Example 2

$$\begin{aligned} F_{T_o}(t) &= P(X_1 + X_2 \leq t) = \iint_{\{(x_1, x_2): x_1 + x_2 \leq t\}} f(x_1, x_2) dx_1 dx_2 \\ &= \int_0^t \int_0^{t-x_1} \lambda e^{-\lambda x_1} \cdot \lambda e^{-\lambda x_2} dx_2 dx_1 = \int_0^t (\lambda e^{-\lambda x_1} - \lambda e^{-\lambda t}) dx_1 \\ &= 1 - e^{-\lambda t} - \lambda t e^{-\lambda t} \end{aligned}$$



## Example 2b

### Problem

Let  $\{X_1, X_2\}$  be a random sample of size 2 from the exponential distribution with parameter  $\lambda = 2$

$$f(x) = \begin{cases} 2e^{-2x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

and  $T = X_1 + X_2$ .

- 1 Compute the cumulative density function (cdf) of  $T$
- 2 Compute the probability density function (pdf) of  $T$

- 1 If the distribution and the statistic  $T$  is simple, try to construct the pmf of the statistic (as in Example 1)
- 2 If the probability density function  $f_X(x)$  of  $X$ 's is known, the
  - try to represent/compute the cumulative distribution (cdf) of  $T$

$$\mathbb{P}[T \leq t]$$

- take the derivative of the function (with respect to  $t$ )

# Example 1\*

## Problem

Consider the distribution  $P$

$x$	40	45	50
$p(x)$	0.2	0.3	0.5

Let  $\{X_1, X_2\}$  be a random sample of size 2 from  $P$ , and  $T = X_1 - X_2$ .

- 1 Derive the probability mass function of  $T$
- 2 Compute the expected value and the standard deviation of  $T$



## Example 2\*

### Problem

Let  $\{X_1, X_2\}$  be a random sample of size 2 from the exponential distribution with parameter  $\lambda$

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

and  $T = X_1 + 2X_2$ .

- 1 Compute the cumulative density function (cdf) of  $T$
- 2 Compute the probability density function (pdf) of  $T$

## Linear combination of normal random variables

## Theorem

Let  $X_1, X_2, \dots, X_n$  be independent normal random variables (with possibly different means and/or variances). Then

$$T = a_1X_1 + a_2X_2 + \dots + a_nX_n$$

also follows the normal distribution.

What are the mean and the standard deviation of  $T$ ?

- $E(T) = a_1E(X_1) + a_2E(X_2) + \dots + a_nE(X_n)$
- $\sigma_T^2 = a_1^2\sigma_{X_1}^2 + a_2^2\sigma_{X_2}^2 + \dots + a_n^2\sigma_{X_n}^2$

## Moment generating function

# Moment generating function

## Definition

The moment generating function (mgf) of a continuous random variable  $X$  is

$$M_X(t) = E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} f_X(x) dx$$

Reading: 3.4 and 4.2

# Moment generating function

## Property

*Two distributions have the same pdf if and only if they have the same moment generating function*

# Moment generating function

Distribution	Moment-generating function $M_X(t)$
Bernoulli $P(X = 1) = p$	$1 - p + pe^t$
Geometric $(1 - p)^{k-1} p$	$\frac{pe^t}{1 - (1 - p)e^t}$ $\forall t < -\ln(1 - p)$
Binomial $B(n, p)$	$(1 - p + pe^t)^n$
Poisson $\text{Pois}(\lambda)$	$e^{\lambda(e^t - 1)}$
Uniform (continuous) $U(a, b)$	$\frac{e^{tb} - e^{ta}}{t(b - a)}$
Uniform (discrete) $U(a, b)$	$\frac{e^{at} - e^{(b+1)t}}{(b - a + 1)(1 - e^t)}$
Normal $N(\mu, \sigma^2)$	$e^{t\mu + \frac{1}{2}\sigma^2 t^2}$
Chi-squared $\chi_k^2$	$(1 - 2t)^{-\frac{k}{2}}$
Gamma $\Gamma(k, \theta)$	$(1 - t\theta)^{-k}; \forall t < \frac{1}{\theta}$
Exponential $\text{Exp}(\lambda)$	$(1 - t\lambda^{-1})^{-1}, (t < \lambda)$
	$\Gamma(\nu + 1) \Gamma(\nu)$

# Moment generating function

## Definition

Let  $X_1, X_2$  be a 2 independent random variables and  $T = X_1 + X_2$ , then

$$M_T(t) = M_{X_1}(t)M_{X_2}(t)$$

Hint:

$$M_T(t) = E(e^{tT}) = E(e^{t(X_1+X_2)}) = E(e^{tX_1} \cdot e^{tX_2})$$



## Example 3

### Problem

*Given that the mgf of a Poisson variables with mean  $\lambda$  is*

$$e^{\lambda(e^t-1)}$$

*Suppose  $X$  and  $Y$  are independent Poisson random variables, where  $X$  has mean  $a$  and  $Y$  has mean  $b$ . Show that  $T = X + Y$  also follows the Poisson distribution.*