Key concepts and/or techniques:

- bivariate normal distribution and its properties
- 1. marginal distributions are normal
- 2. conditional distributions are normal
- 3. independence ← Uncorrelation
- ► find the distribution of the function of RVs, i.e., determine the pmf or pdf of the functions of RVs

Definition

Let X and Y be 2 continuous RVs and have the joint pdf

$$f(x,y) = \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-
ho^2}} \exp[-\frac{1}{2}q(x,y)], x \in \mathbb{R}, y \in \mathbb{R},$$

$$q(x,y) = \frac{1}{1-\rho^2} \left[\left(\frac{x-\mu_X}{\sigma_X} \right)^2 - 2\rho \left(\frac{x-\mu_X}{\sigma_X} \right) \left(\frac{y-\mu_Y}{\sigma_Y} \right) + \left(\frac{y-\mu_Y}{\sigma_Y} \right)^2 \right] \ge 0$$

where $\mu_X, \mu_Y \in \mathbb{R}$, $\sigma_X, \sigma_Y > 0$ and $|\rho| < 1$. Then X and Y are said to be bivariate normally distributed.

Key components: Scaled exponential function with a quadratic and negative function as its exponent.

1. Marginal distributions are normal:

$$X \sim N(\mu_X, \sigma_X^2), Y \sim N(\mu_Y, \sigma_Y^2)$$

Conditional distributions are normal:

$$X|Y = y \sim N\left(\mu_X + \frac{\sigma_X}{\sigma_Y}\rho(y - \mu_Y), (1 - \rho^2)\sigma_X^2\right)$$
$$Y|X = x \sim N\left(\mu_Y + \frac{\sigma_Y}{\sigma_X}\rho(x - \mu_X), (1 - \rho^2)\sigma_Y^2\right)$$

3. Independence ← Uncorrelation

[Function of One Random Variable]

Let X be a RV of either discrete or continuous type with its pmf or pdf denoted by f(x). Consider a function of X, say Y = u(X). Then Y is also a RV and has its pmf or pdf.

How to compute the pmf or pdf of Y?

In what follows, we consider the case where Y = u(X) is a one-to-one mapping.

1. For discrete RV, when Y = u(X) be a one-to-one mapping with inverse X = v(Y). Then the pmf of Y is

$$g(y) = f[v(y)]$$
 for $y \in \overline{S_Y}$

2. For continuous RV, when Y = u(X) is continuous, strictly decreasing or increasing and has inverse function X = v(Y), whose derivative $\frac{dv(y)}{dy}$ exists, the pdf of Y, denoted by g(y),

$$g(y) = f(v(y)) \left| \frac{dv(y)}{dy} \right|$$

STA2001 Probability and Statistics (I)

Lecture 19

Tianshi Chen

The Chinese University of Hong Kong, Shenzhen

Example 2, page 174

Let X have the pdf

$$f(x) = 3(1-x)^2$$
, $0 < x < 1$.

Consider $Y = (1 - X)^3$ and calculate the pdf of Y, g(y).

Example 2, page 174

$$Y=u(X)=(1-X)^3 \longrightarrow ext{continuous, strictly decreasing}$$
 Inverse function $\longrightarrow X=v(Y)=1-Y^{rac{1}{3}}$

- 1. The sample space of Y is $\overline{S_Y} = (0,1)$, since 0 < x < 1.
- 2.

$$g(y) = f(v(y)) \left| \frac{dv(y)}{dy} \right| \quad \text{where} \quad \frac{dv(y)}{dy} = -\frac{1}{3}y^{-\frac{2}{3}}$$
$$= 3(1 - (1 - y^{\frac{1}{3}}))^2 \left| -\frac{1}{3}y^{-\frac{2}{3}} \right| = 1,$$
$$0 < y < 1 \quad Y \sim U(0, 1)$$

Theorem 5.1-1, page 175

Given a random distribution, it is possible to construct a RV such that this RV has the given random distribution.

Theorem[Random Number Generator]

Let $Y \sim U(0,1)$ and F(x) have the properties of a cdf of a continuous RV with F(a) = 0, F(b) = 1. Moreover, F(x) is strictly increasing such that $F(x) : (a,b) \to [0,1]$, where a could be $-\infty$, b could be ∞ . Then $X = F^{-1}(Y)$ is continuous RV with cdf F(x).

Theorem 5.1-1, page 175

Proof: Idea — we need to show
$$P(X \le x) = F(x)$$

$$P(X \le x) = P(F^{-1}(Y) \le x) = P(Y \le F(x))$$
 since $\{y | F^{-1}(y) \le x\} = \{y | y \le F(x)\}.$

Note that

$$Y \sim U(0,1) \Longrightarrow P(Y \leq y) \xrightarrow{0 < y < 1} \int_0^y 1 dz = y$$

Therefore,

$$P(X \le x) = P(Y \le F(x)) = F(x)$$

Remarks

Theorem 5.1-1 can be used to construct a random number generator for distributions with strictly increasing cdf based on the random generator for a uniform distribution.

Random number generator

- 1. generator a random number y from U(0,1)
- 2. Take $x = F^{-1}(y)$

Then x is a random number generated from the distribution or RV with cdf F(x).

Question

Assume that we know how to generate a random number from $Y \sim U(0,1)$.

Can we generate a random number from the exponential distribution with parameter θ , whose pdf is given by

$$f(x) = \frac{1}{\theta}e^{-\frac{x}{\theta}}, \quad x \ge 0$$

Question

Assume that we know how to generate a random number from $Y \sim U(0,1)$.

Can we generate a random number from the exponential distribution with parameter θ , whose pdf is given by

$$f(x) = \frac{1}{\theta}e^{-\frac{x}{\theta}}, \quad x \ge 0$$

Can we run computer simulation to collect the waiting time until the first customer that arrives at a cinema or a hospital?

The cdf of X is

$$F(x) = P(X \le x) = \int_0^x \frac{1}{\theta} e^{-\frac{t}{\theta}} dt = 1 - e^{-\frac{x}{\theta}}, \quad x \ge 0,$$

which is strictly increasing.

- 1. If we know how to generate a random number from $Y \sim U(0,1)$, say the random number generated is y.
- 2. $x = F^{-1}(y)$ is the random number generated from the exponential distribution with θ , where

$$x = F^{-1}(y) = -\theta \ln(1 - y), \quad y \in (0, 1)$$

Histogram for continuous distribution

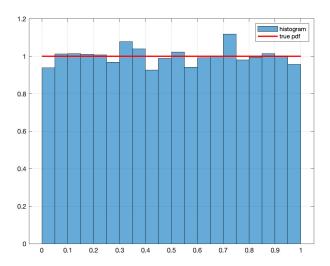
The simplest form of a histogram is constructed as follows

- divide (or "bin") the sample space of the distribution into a sequence of adjacent, non-overlapping and equally spaced subintervals.
- treat each subinterval as an event, then count how many observed numerical outcomes fall into each subinterval and calculate the relative frequency
- 3. draw a rectangle erected over the bin with height equal to the relative frequency divided by the width of each subinterval.

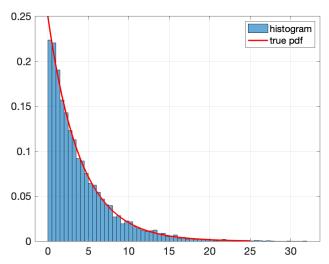
Remark:

▶ Note that the area of the histogram is equal to 1, thus histogram gives an approximation of the probability density function of the underlying random variable.

The histogram of 10000 random numbers y generated from U(0,1)



The histogram of 10000 random numbers x for exponential distribution with $\theta=4$



Matlab script for the random number generator:

```
figure(1);
y=rand(10000,1); % generate 10000 random number y from U(0,1)
histogram(y,'normalization','pdf') % draw the histogram of
10000 y
hold on;
plot(0:0.01:1,ones(1,length(0:0.01:1)),'r',
'linewidth',2) % plot the true pdf of U(0,1)
grid on;
legend('histogram', 'true pdf')
figure(2);
theta=4;
x=-theta*log(1-y); % generate 10000 random number x
histogram(x, 'normalization', 'pdf') % draw the histogram of
10000 ×
hold on;
plot(0:0.01:25, exp(-(0:0.01:25)/theta)/theta, 'r',
'linewidth', 2) % plot the true pdf of exponential distribution with
theta=4
grid on;
```

Theorem 5.1-2, page 176

[Theorem]

Suppose that X is a continuous RV with $\overline{S_X}=(a,b)$, and moreover, its cdf F(x) is strictly increasing. Then the RV Y, defined by Y=F(X), has a uniform distribution, that is, $Y\sim U(0,1)$.

Theorem 5.1-2, page 176

Proof:

Since F(a) = 0, F(b) = 1 and F(x) is strictly increasing,

Y = F(X) with the range $\overline{S_Y} = (0,1)$.

Consider the cdf of Y:

$$P(Y \le y) = P(F(X) \le y), \quad y \in (0,1)$$

Theorem 5.1-2, page 176

Since F(x) is strictly increasing, $\{F(X) \le y\} \iff \{X \le F^{-1}(y)\}.$

$$P(Y \le y) = P(F(X) \le y) = P(X \le F^{-1}(y)), \quad 0 < y < 1.$$

Since $P(X \le x) = F(x)$, we have

$$P(Y \le y) = F(F^{-1}(y)) = y$$
, $0 < y < 1 \longrightarrow \text{cdf of } U(0,1)$

Example 2, page 174, continued

Let X have the pdf

$$f(x) = 3(1-x)^2$$
, $0 < x < 1$

Consider $Y = (1 - X)^3$ and then $Y \sim U(0, 1)$.

The result can be obtained from Theorem 5.1-2. Since

$$F(x) = 1 - (1 - x)^3$$

is strictly increasing, then

$$F(X) = 1 - (1 - X)^3 \sim U(0, 1)$$

which implies $(1 - X)^3 = [1 - F(X)] \sim U(0, 1)$.

The case: Y = u(X) not one-to-one

[Function of One Random Variable]

Let X be a RV of either discrete or continuous type with its pmf or pdf denoted by f(x). Consider a function of X, say Y = u(X). Then Y is also a RV and has its pmf or pdf.

How to compute the pmf or pdf of Y?

When Y = u(X) is not one-to-one, there is no general result.

Question

Assume that X is a continuous RV with pdf

$$f(x) = \frac{1}{\pi(1+x^2)} \quad x \in (-\infty, \infty)$$

Let $Y = X^2$. Find the pdf of Y.

Clearly,
$$\overline{S_Y} = [0, \infty)$$

Let the cdf of Y be G(y). Then

$$G(y) = P(Y \le y), \quad y \in [0, \infty)$$
$$= P(X^2 \le y) = P(-\sqrt{y} \le X \le \sqrt{y}) = \int_{-\sqrt{y}}^{\sqrt{y}} f(x) dx$$

$$G(y) = P(Y \le y), \quad y \in [0, \infty)$$

$$= P(X^{2} \le y) = P(-\sqrt{y} \le X \le \sqrt{y}) = \int_{-\sqrt{y}}^{\sqrt{y}} f(x) dx$$

$$= \int_{-\sqrt{y}}^{\sqrt{y}} \frac{1}{\pi(1 + x^{2})} dx = 2 \int_{0}^{\sqrt{y}} \frac{1}{\pi(1 + x^{2})} dx$$

$$\Rightarrow g(y) = G'(y) = \frac{2}{\pi(1 + y)} \times \frac{1}{2} \times \frac{1}{\sqrt{y}} = \frac{1}{\pi(1 + y)\sqrt{y}}$$