## **Functions of random variables**

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## Transformation of a random variable

#### Distribution of functions of a random variable

Let X be a random variable with cdf  $F_X(x) = P(X \le x)$ , then for any function g, we have Y = g(X) is also a random variable. The distribution of Y can be determined from

$$F_Y(y) = P(Y \le y) = P(g(X) \le y) = P(X \in \mathcal{A})$$

where  $A = \{x : g(x) \le y\}$ .

# Example

#### Distribution of functions of a random variable

If g is **monotone increasing**, the calculation can be simplified. In this case, the function g has also a monotone increasing inverse function  $g^{-1}$ . Then we have

If g is **monotone decreasing**, then we have

# Density of g(X)

# Example

## **Bivariate transformation**

#### Joint distribution

Let X and Y be two random variables with known joint distribution. Consider  $U=g_1(X,Y)$  and  $V=g_2(X,Y)$  for some functions  $g_1$  and  $g_2$ . Using the cdf argument, we can also find the joint distribution of U and V.

Example:

## Jacobian technique

Consider a case when both X and Y are continuous with joint pdf  $f_{X,Y}(x,y)$ , and there is a one-to-one transformation between (X,Y) and (U,V). In this case, we can write  $X=h_1(U,V)$  and  $Y=h_2(U,V)$ , and the joint density of U and V is given by

# **Examples**

# **Examples**

### **Extension to multivariate distributions**

# Sample from normal distributions

## Sample mean and variances

Let  $X_1,\ldots,X_n$  denote iid samples from  $N(\mu,\sigma^2)$  distribution. Let  $\bar{X}_n=n^{-1}\sum_{i=1}^n X_i$  and  $s_n^2=\sum_{i=1}^n (X_i-\bar{X})^2/(n-1)$ .

Then:

### Proofs

#### The *t*-distribution

Let  $Z \sim N(0,1)$  and  $V \sim \chi^2_v$  independent of Z. Then the random variable

$$T = \frac{Z}{\sqrt{V/v}} \sim t_v,$$

the t distribution with v degrees of freedom.

Question: Derive the density of *T*?

#### The *t*-distribution

Let  $X_1, \ldots, X_n$  denote iid samples from  $N(\mu, \sigma^2)$  distribution, with  $\bar{X}_n$  and  $s_n^2$  the sample mean and sample variance. Then

$$T = \frac{\bar{X}_n - \mu}{s_n / \sqrt{n}} \sim t_{n-1}$$