#### Transformations of Distributions

#### The binomial distribution

- Sequence of n independent trials of an experiment.
- Two possible outcomes for each trial:
   "Success" or "Failure"
- p = P(Success on any one trial)

Let X = # successes in n trials.

 $X \sim bin(n, p)$  pmf?

## Suppose n = 4.

P(X = 3) = P(SSSF or SSFS or SFSS or FSSS)



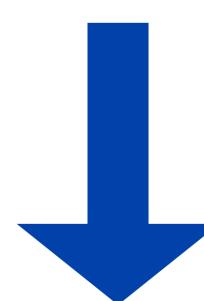
disjoint

$$= P(SSSF) + P(SSFS) + P(SFSS) + P(FSSS)$$



independent

$$= p \cdot p \cdot p \cdot (1 - p) + p \cdot p \cdot (1 - p) \cdot p + \cdots$$



counting

$$= 4p^3 (1 - p)$$

## n trials, x successes

There are 
$$\binom{n}{x} = \frac{n!}{x!(n-x)!}$$
 different

ways to arrange the x S's in the n positions.

$$P(X = x) = \binom{n}{x} p^{x} (1 - p)^{n-x} I_{\{0,1,...,n\}}(x)$$

 $X \sim bin(n, p)$ 

#### Transformations

Suppose that  $X \sim bin(n, p)$ .

What is the distribution of Y = n-X?

$$P(X = x) = {n \choose x} p^{x} (1 - p)^{n-x} I_{\{0,1,...,n\}}(x)$$

#### Just do it:

$$P(Y = y) = P(n - X = y)$$

$$= P(X = n - y)$$

$$= {n \choose n - y} p^{x} (1 - p)^{n - (n - y)} I_{\{0,1,...,n\}} (n - y)$$

#### Transformations

$$P(Y = y) = \binom{n}{n-y} p^{x} (1-p)^{n-(n-y)} I_{\{0,1,...,n\}} (n-y)$$

$$n - y = 0, 1, ..., n \Rightarrow y = 0, 1, ..., n$$

$$\binom{5}{3} = \frac{5!}{3!2!} = \frac{5!}{2!3!} = \binom{5}{2}$$

$$P(Y = y) = {n \choose y} (1 - p)^y p^{n-y} I_{\{0,1,...,n\}}(y)$$

 $Y \sim bin(n, 1 - p)$ 

#### Continuous Transformations

## For X discrete or continuous, the cumulative distribution function (cdf) Is denoted by F(x) and is defined by

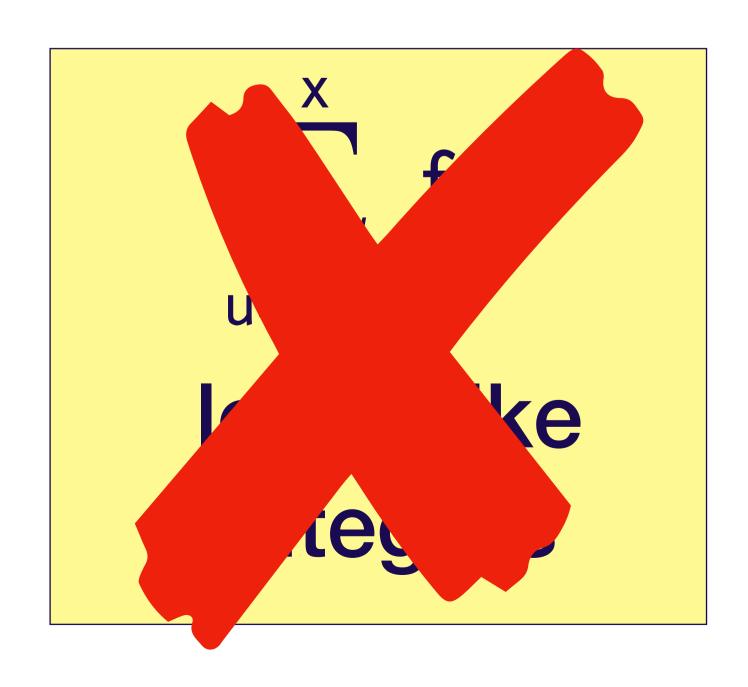
$$F(x) = P(X \le x)$$

X discrete:

$$F(x) = \sum_{u \le x} P(X = u) = \sum_{u \le x} f(u)$$

X continuous:

$$F(x) = \int_{-\infty}^{x} f(u) du$$



#### X continuous:

# Fundamental Theorem of Calculus

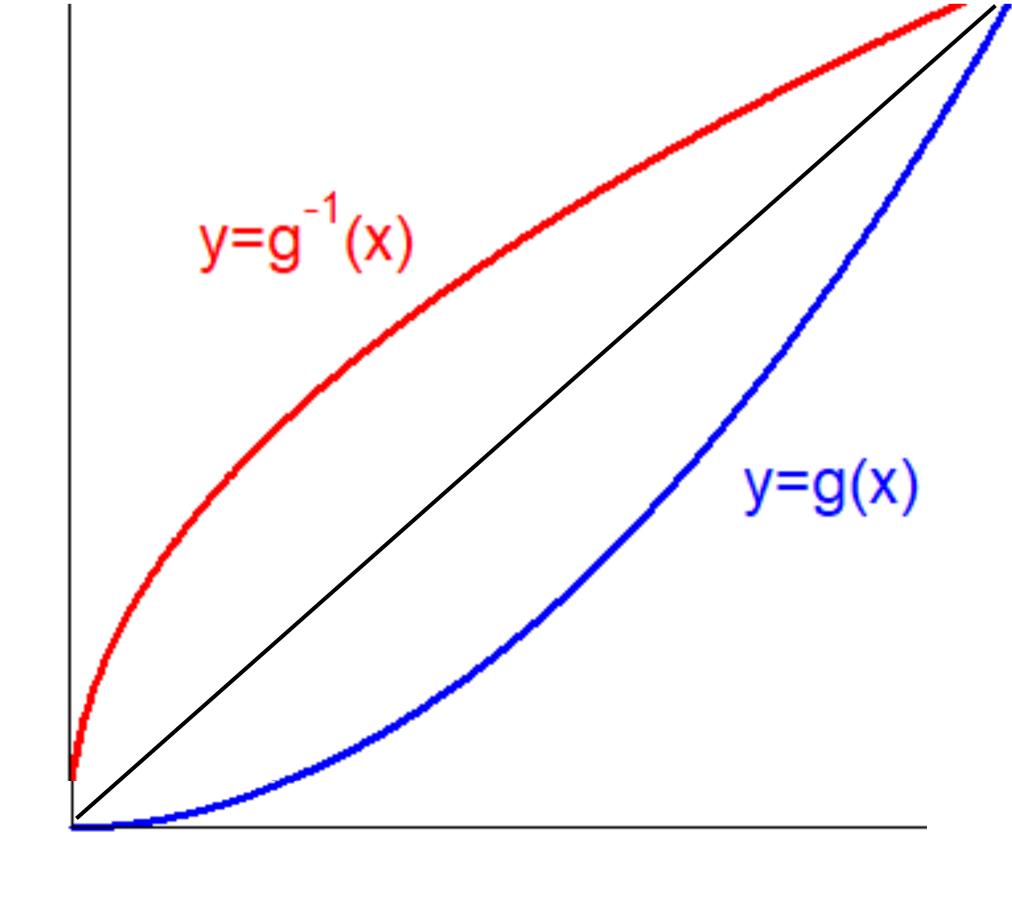
$$f(x) = \frac{d}{dx}F(x)$$

- Suppose that X is continuous with pdf  $f_X(x)$ .
- Let g(x) be a continuous, differential, and invertible function.
- The g is either strictly increasing or strictly decreasing.

The inverse of an increasing function

is increasing.

same for decreasing



# X with pdf $f_X(x)$ , Y=g(X) invertible

## Case: g increasing

$$F_{Y}(y) = P(Y \le y) = P(g(X) \le y)$$

$$= P(g^{-1}(g(X)) \le g^{-1}(y))$$

$$= P(X \le g^{-1}(y))$$

$$= F_{X}(g^{-1}(y))$$

$$\begin{split} f_Y(y) &= \frac{d}{dy} F_Y(y) \ = \frac{d}{dy} F_X(g^{-1}(y)) \\ &= f_X(g^{-1}(y)) \cdot \frac{d}{dy} g^{-1}(y) \end{split}$$

# X with pdf $f_X(x)$ , Y=g(X) invertible

## Case: g decreasing

$$\begin{aligned} F_Y(y) &= P(Y \le y) = P(g(X) \le y) \\ &= P(g^{-1}(g(X)) \ge g^{-1}(y))) \\ &= P(X \ge g^{-1}(y)) = 1 - P(X < g^{-1}(y)) \\ &= 1 - P(X \le g^{-1}(y)) = 1 - F_X(g^{-1}(y)) \end{aligned}$$

$$\begin{split} f_Y(y) &= \frac{d}{dy} F_Y(y) \ = \frac{d}{dy} \left( 1 - F_X(g^{-1}(y)) \right) \\ &= - f_X(g^{-1}(y)) \left( \frac{d}{dy} g^{-1}(y) \right) \end{split} \text{negative}$$

- X with pdf  $f_X(x)$ , Y=g(X) invertible
- g increasing or decreasing

$$f_{Y}(y) = f_{X}(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right|$$

Example: Suppose that X has a gamma distribution with shape parameter  $\alpha$  and inverse scale parameter  $\beta$ .

$$f(x) = \frac{1}{\Gamma(\alpha)} \beta^{\alpha} x^{\alpha - 1} e^{-\beta x} I_{(0, \infty)}(x)$$

 $X \sim \Gamma(\alpha, \beta)$ 

$$X \sim \Gamma(\alpha, \beta)$$

$$f(x) = \frac{1}{\Gamma(\alpha)} \beta^{\alpha} x^{\alpha - 1} e^{-\beta x} I_{(0,\infty)}(x)$$

What is the distribution of Y=cX for constant c>0 a constant?

$$y = g(x) = cx \Rightarrow x = g^{-1}(y)$$

$$f_{Y}(y) = f_{X}(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right|$$

$$= \frac{1}{\Gamma(\alpha)} \beta^{\alpha}(x/c)^{\alpha - 1} e^{-\beta x/c} I_{(0,\infty)}(x) \cdot |1/c|$$

$$f_{Y}(y) = \frac{1}{\Gamma(\alpha)} \beta^{\alpha}(y/c)^{\alpha-1} e^{-\beta y/c} I_{(0,\infty)(y/c)} \cdot |1/c|$$

$$0 < y/c < \infty \stackrel{c > 0}{\Rightarrow} 0 < y < \infty$$

$$f_{Y}(y) = \frac{1}{\Gamma(\alpha)} (\beta/c)^{\alpha} x^{\alpha-1} e^{-\beta x/c} I_{(0,\infty)(y)}$$

$$\Rightarrow$$
  $Y \sim \Gamma(\alpha, \beta/c)$