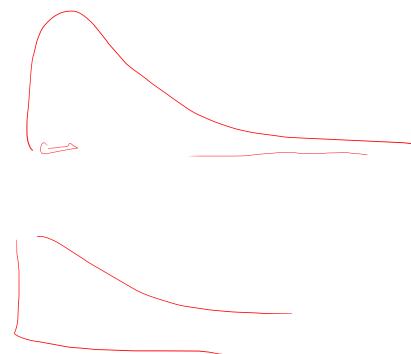


# CONTINUOUS RANDOM VARIABLES

A few other distributions

# Skewed (to the right) data

- Many real-life datasets involve random variables that are skewed to the right.
- Examples:
  - Survival times of patients after diagnosis
  - Time to relapse after recovery
  - Time to failure of a medical device
  - Time between phone calls to a hotline



# The gamma distribution

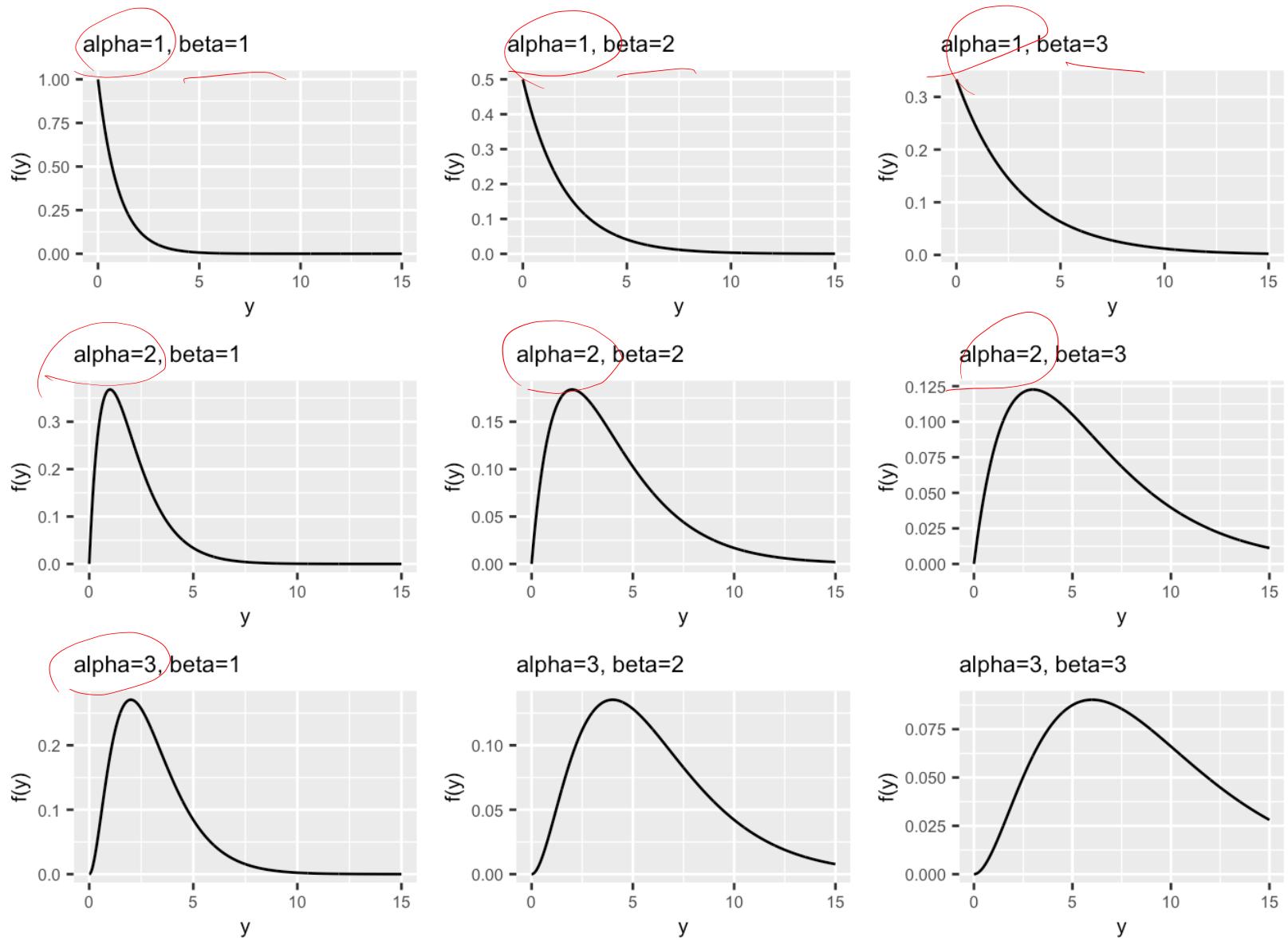
- Definition 6: A random variable  $Y$  has a GAMMA DISTRIBUTION with parameters  $\alpha > 0$  and  $\beta > 0$  if its pdf is

$$f(y) = \frac{y^{\alpha-1} e^{-y/\beta}}{\beta^\alpha \Gamma(\alpha)} \text{ for } y \geq 0,$$

where

$$\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy$$

- $\Gamma(\cdot)$  is the “gamma function”:
  - $\Gamma(1) = 1$  (check this!)
  - $\Gamma(\alpha) = (\alpha - 1)\Gamma(\alpha - 1)$  for any  $\alpha > 1$  (integration by parts)
  - $\Gamma(n) = (n - 1)!$  If  $n$  is a positive integer.
- Shorthand:  $Y \sim \text{gamma}(\alpha, \beta)$
- $\alpha$ : “shape parameter”;  $\beta$ : “scale parameter”



# The gamma distribution

- Check: Is the gamma pdf a proper pdf?

$\rightarrow$  Does it integrate to 1?

$$\int_0^\infty \frac{y^{\alpha-1} e^{-y/\beta}}{\beta^\alpha \Gamma(\alpha)} dy = \frac{1}{\beta^\alpha \Gamma(\alpha)} \int_0^\infty y^{\alpha-1} e^{-y/\beta} dy$$

$$z = \frac{y}{\beta} \Leftrightarrow y = \beta z \quad dz = \frac{1}{\beta} dy \\ dy = \beta dz$$

$$= \frac{1}{\beta^\alpha \Gamma(\alpha)} \int_0^\infty (\beta z)^{\alpha-1} e^{-z} \beta dz$$

$$= \frac{1}{\beta^\alpha \Gamma(\alpha)} \cdot \beta^{\alpha-1} \cdot \beta \left[ \int_0^\infty z^{\alpha-1} e^{-z} dz \right] = \frac{1}{\beta^\alpha \Gamma(\alpha)} \cancel{\beta^\alpha} \cancel{\Gamma(\alpha)} = 1$$

YES!

$\Gamma(\alpha)$  (defined)!

$$f(y) \geq 0 \quad y^{\alpha-1} e^{-y/\beta}$$

# The gamma distribution

- Theorem 7: If  $Y \sim \text{gamma}(\alpha, \beta)$  then  $E[Y] = \alpha\beta$  and  $\text{Var}(Y) = \alpha\beta^2$ .

$$\begin{aligned}
 E[Y] &= \int_0^\infty y \frac{y^{\alpha-1} e^{-y/\beta}}{\beta^\alpha \Gamma(\alpha)} dy = \int_0^\infty \frac{y^\alpha e^{-y/\beta}}{\beta^\alpha \Gamma(\alpha)} dy \\
 &= \beta \int_0^\infty \frac{y^\alpha e^{-y/\beta}}{\beta^{\alpha+1} \Gamma(\alpha+1)/\alpha} dy = \alpha \beta \int_0^\infty \frac{y^\alpha e^{-y/\beta}}{\beta^{\alpha+1} \Gamma(\alpha+1)} dy \\
 &\quad \text{PDF of } \text{Gamma}(\alpha+1, \beta) \\
 &= \alpha \beta
 \end{aligned}$$

G.F.F.  
 $\Gamma(\alpha+1) = \alpha \Gamma(\alpha)$   
 $\Gamma(\alpha) = \frac{\Gamma(\alpha+1)}{\alpha}$

# The gamma distribution

- Theorem 8: If  $Y \sim \text{gamma}(\alpha, \beta)$  then the mgf of  $Y$  is

$$m(t) = \left( \frac{1}{1-\beta t} \right)^\alpha, \text{ for } t < \frac{1}{\beta}$$

~~check book~~

check book

# The gamma distribution

- Example 10: Suppose a random variable  $Y$  has pdf

$$f(y) = cy^4 e^{-y/3}, \quad y > 0$$

What must the constant  $c$  be? Find  $E[Y]$  and  $Var(Y)$ .

$$\text{Gamma}(5, 3)$$

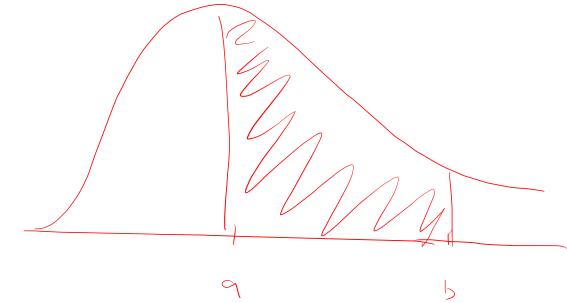
$$c = \frac{1}{\int_0^\infty y^4 e^{-y/3} dy} = \frac{1}{3^5 \Gamma(5)} = \frac{1}{3 \cdot 3 \cdot 3 \cdot 3 \cdot 3 \cdot 4 \cdot 3 \cdot 2 \cdot 1}$$

$$E[Y] = \alpha\beta = 5 \cdot 3 = 15$$

$$Var(Y) = \alpha\beta^2 = 5 \cdot 3^2 = 45$$

$$\int_0^\infty c y^4 e^{-y/3} dy$$

# Calculating probabilities for the gamma distribution



- If  $Y \sim \text{gamma}(\alpha, \beta)$ , there is no closed-form solution for quantities  $P(a < Y < b)$  (unless  $\alpha$  is an integer).
- Numerical approximation:
  - pgamma() function in R
  - Online calculators
- **WARNING!** Not all representations of the gamma function are parameterized the same way! (Sometimes “their  $\beta$ ” is the reciprocal of “our  $\beta$ .”) It doesn’t have to be a problem – just make the substitution – but you have to look out for it!

$$e^{-y/\beta}$$

~~β~~

# Calculating probabilities for the gamma distribution

- Example 11: If  $Y \sim \text{gamma}(3,5)$ , calculate  $P(10 < Y < 20)$ .
- FIRST: Check expression for pdf!
- It matches our pdf in R!
- From R :

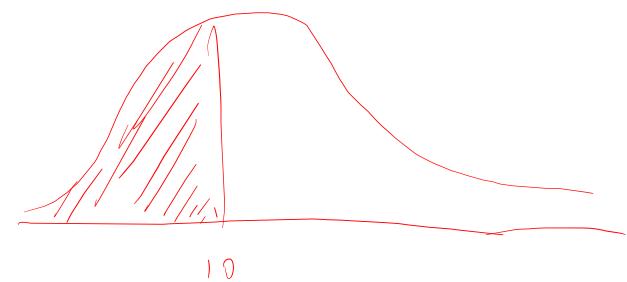
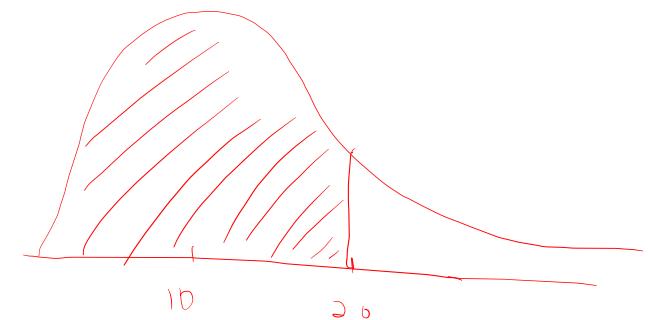
```
> pgamma(20, shape=3, scale=5)
```

```
[1] 0.7618967
```

```
> pgamma(10, shape=3, scale=5)
```

```
[1] 0.3233236
```

$$P(10 < Y < 20) = 0.762 - 0.323$$



$$E[Y] = \alpha \beta \quad V[Y] = \alpha \beta^2$$

## The chi-square distribution

- Definition 7: A random variable  $Y$  has a CHI-SQUARE DISTRIBUTION with  $\nu$  degrees of freedom if  $Y \sim \text{gamma}(\nu/2, 2)$ . ( $\nu$  must be a positive integer)
- Shorthand:  $Y \sim \chi^2(\nu)$
- If  $Y \sim \chi^2(\nu)$ , then  $E[Y] = \nu$  and  $Var(Y) = 2\nu$ .

$$\alpha = \frac{\nu}{2} \quad \beta = 2$$

$$\sim \chi^2(\nu)$$

$$\frac{\nu}{2}, 2 = \nu \quad \frac{\nu}{2} \text{ d.f.} = 2\nu$$

# The exponential distribution

- Definition 8: A random variable  $Y$  has an EXPONENTIAL DISTRIBUTION with parameter  $\beta$  if  $Y \sim \text{gamma}(1, \beta)$ . ( $\beta > 0$ )
- Shorthand:  $\underline{Y \sim \exp(\beta)}$
- If  $Y \sim \exp(\beta)$  then  $E[Y] = \beta$  and  $Var(Y) = \beta^2$

$$\mathbb{E}[\cdot] = \alpha \beta$$

$\nabla \approx \alpha \beta$

$$| \beta = \beta$$
$$1 \cdot \beta^2 = \beta^2$$

# The beta distribution

- Some random variables are constrained to have support only on  $\underline{[0,1]}$ .
- This is typical when the random variable is a measured proportion.
- Definition 9: A random variable  $Y$  has a BETA DISTRIBUTION with parameters  $\alpha$  and  $\beta$  if its pdf is

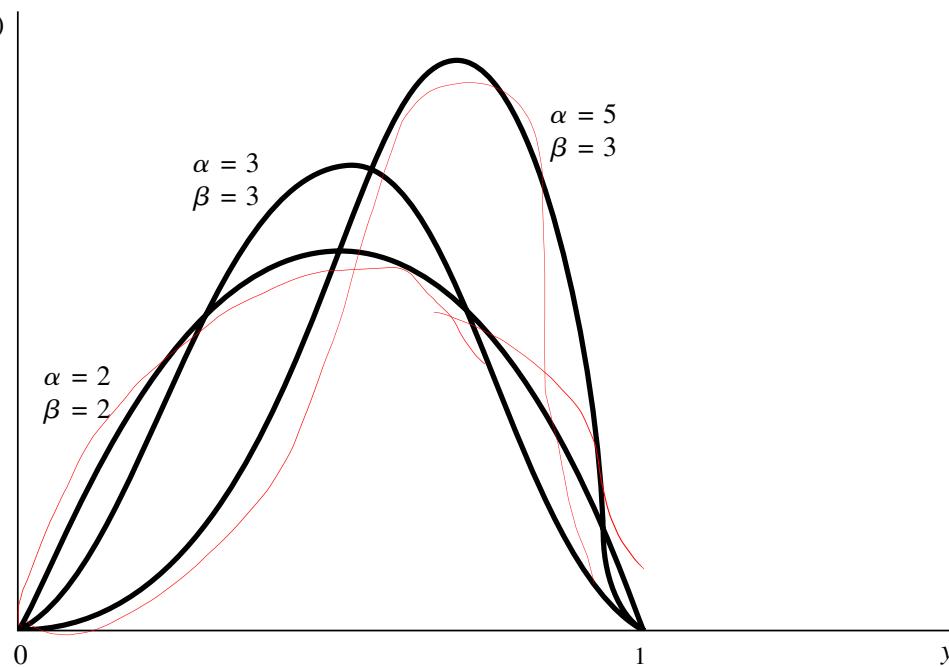
$$f(y) = \frac{y^{\alpha-1}(1-y)^{\beta-1}}{B(\alpha, \beta)} \text{ for } 0 \leq y \leq 1$$

- Here,  $B(\alpha, \beta) = \int_0^1 y^{\alpha-1}(1-y)^{\beta-1} dy = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$
- Shorthand:  $Y \sim \text{beta}(\alpha, \beta)$

# The beta distribution

- The beta distribution, again, constrained to be in  $[0,1]$ , is quite versatile and flexible.

FIGURE 4.17  
Beta density  
functions



# The beta distribution

- If  $Y \sim \text{beta}(\alpha, \beta)$  then  $E[Y] = \frac{\alpha}{\alpha+\beta}$  and  $\text{Var}(Y) = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$
- For computing probabilities  $P(a < Y < b)$  (where  $0 \leq a \leq b \leq 1$ ), it is possible to do this analytically if  $\alpha$  and  $\beta$  are both integers.  
(Otherwise, numerical approximations!)