## Using R, Chapter 6: Normal Distributions The pnorm and qnorm functions.

• Getting probabilities from a normal distribution with mean  $\mu$  and standard deviation  $\sigma$ 

pnorm(x, mean = , sd = , lower.tail= )

If x is a normally distributed random variable, with mean =  $\mu$  and standard deviation =  $\sigma$ , then

$$P(x < x_{\max}) = \text{pnorm}(x_{\max}, \text{mean} = \mu, \text{sd} = \sigma, \text{lower.tail=TRUE})$$

$$P(x > x_{\min}) = \text{pnorm}(x_{\min}, \text{mean} = \mu, \text{sd} = \sigma, \text{lower.tail=FALSE})$$

$$P(x_{\min} < x < x_{\max}) = pnorm(x_{\max}, mean = \mu, sd = \sigma, lower.tail=TRUE)$$
  
- pnorm(x<sub>min</sub>, mean =  $\mu$ , sd =  $\sigma$ , lower.tail=TRUE)

## Examples:

Suppose IQ's are normally distributed with a mean of 100 and a standard deviation of 15.

1. What percentage of people have an IQ less than 125?

pnorm(125, mean = 100, sd = 15, lower.tail=TRUE) = .9522 or about 95%

2. What percentage of people have an IQ greater than 110?

pnorm(110, mean = 100, sd = 15, lower.tail=FALSE) = .2525 or about 25%

3. What percentage of people have an IQ between 110 and 125?

pnorm(125, mean = 100, sd = 15, lower.tail=TRUE)
- pnorm(110, mean = 100, sd = 15, lower.tail=TRUE)
= 0.2047 or about 20%

• Usage for the standard normal (z) distribution ( $\mu = 0$  and  $\sigma = 1$ ).

In the text we first convert x scores to z scores using the formula  $z = (x-\mu)/\sigma$  and then find probabilities from the z-table. These probabilities can be found with the **pnorm** function as well. It is actually the default values for  $\mu$  and  $\sigma$  with the **pnorm** function.

$$\begin{split} P(z < z_{\max}) &= \texttt{pnorm}(\texttt{z}_{\max}) \\ P(z > z_{\min}) &= \texttt{pnorm}(\texttt{z}_{\min}, \texttt{lower.tail=FALSE}) \\ P(z_{\min} < z < z_{\max}) &= \texttt{pnorm}(\texttt{z}_{\max}) - \texttt{pnorm}(\texttt{x}_{\min}) \end{split}$$

• Getting percentiles from a normal distribution with mean  $\mu$  and standard deviation  $\sigma$ 

qnorm(lower tail area, mean= , sd = , lower.tail=TRUE)

Suppose you want to find the x-value that separates the bottom k% of the values from a distribution with mean  $\mu$  and standard deviation  $\sigma$ . We denote this value in the text as  $P_k$ .

 $P_k =$ qnorm(k (in decimal form), mean =  $\mu$ , sd =  $\sigma$ , lower.tail=TRUE)

 $P_{25} =$  qnorm(.25, mean =  $\mu$ , sd =  $\sigma$ , lower.tail=TRUE)

 $P_{90} = \text{qnorm}(.90, \text{mean} = \mu, \text{sd} = \sigma, \text{lower.tail=TRUE})$ 

## Examples:

Suppose IQ's are normally distributed with a mean of 100 and a standard deviation of 15.

1. What IQ separates the lower 25% from the others? (Find  $P_{25}$ .)

$$P_{25}= ext{qnorm(.25, mean = 100, sd = 15, lower.tail=TRUE)}=89.88$$

2. What IQ separates the top 10% from the others? (Find  $P_{90}$ .)

 $P_{90} = ext{qnorm(.90, mean = 100, sd = 15, lower.tail=TRUE)} = 119.22$ 

• Usage for the standard normal (z) distribution ( $\mu = 0$  and  $\sigma = 1$ ).

These are actually the default values for  $\mu$  and  $\sigma$  in the **qnorm** function. So getting z-scores is quite easy.

- $P_k =$  qnorm(k (in decimal form))
- $P_{25}=$  qnorm(.25) = -0.67449 ightarrow -0.67
- $P_{90} =$  qnorm(.90) = 1.28155  $\rightarrow$  1.28