

# **Diameter distribution models**

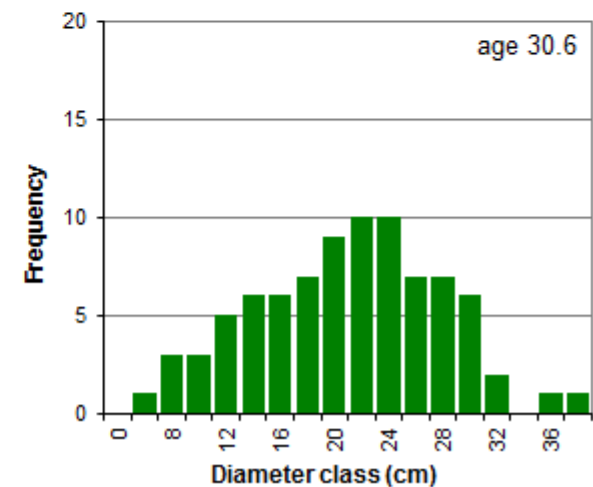
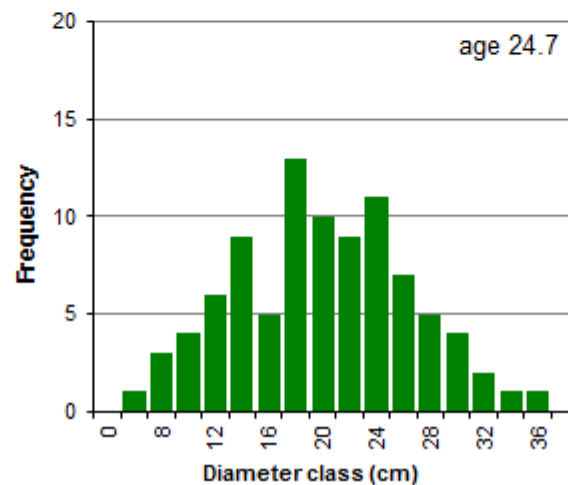
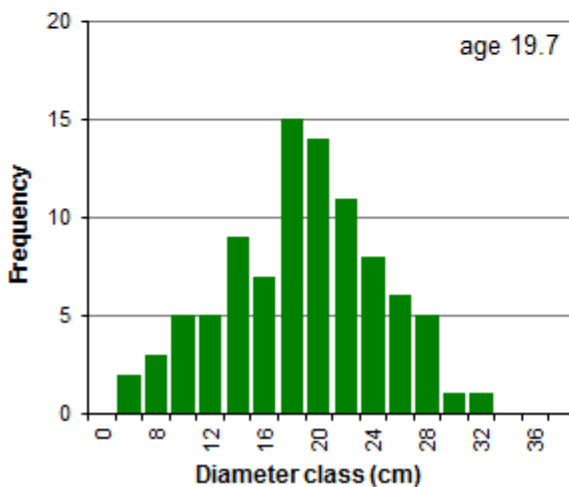
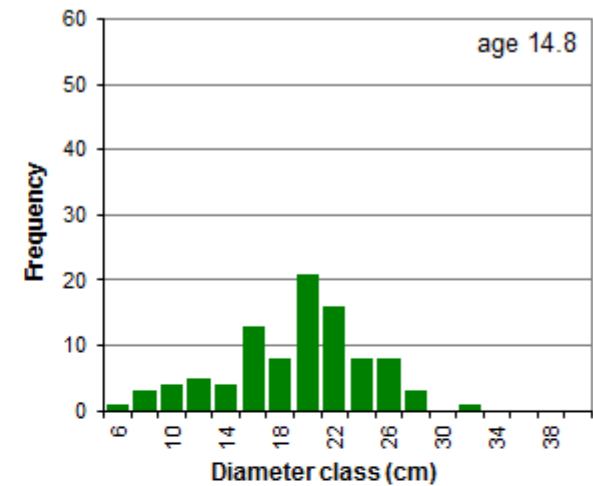
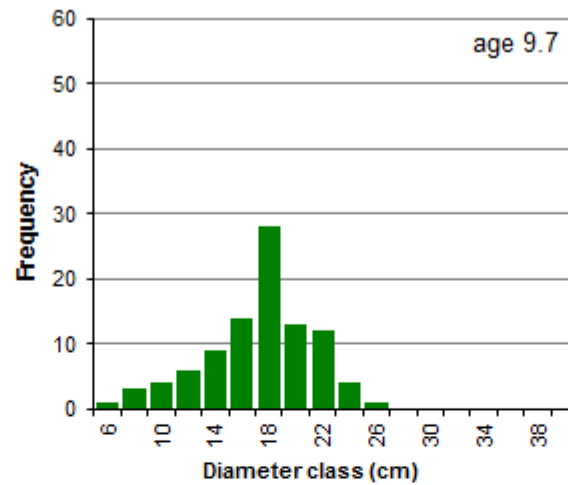
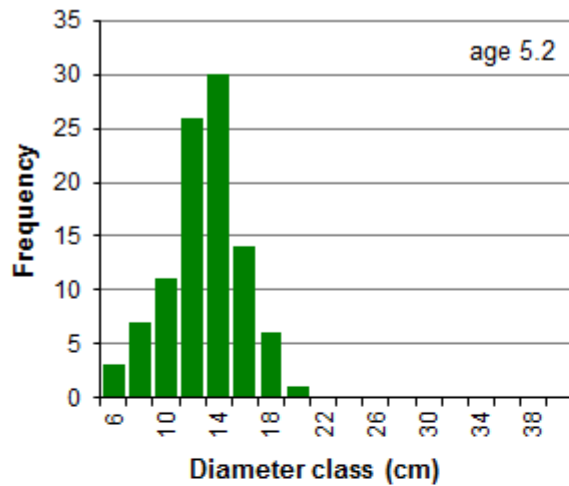
# Diameter distribution models

- ✗ The idea behind diameter distribution models is:
  - ✓ To start by simulating the growth of some variables (principal variables):
    - dominant height
    - number of trees per ha
    - stand basal area estimation
    - some variables characterizing the diameter distribution such as the minimum diameter, some percentile of the diameter distribution or the variance of diameters (depending on the pdf used for diameter distribution)

# Diameter distribution models

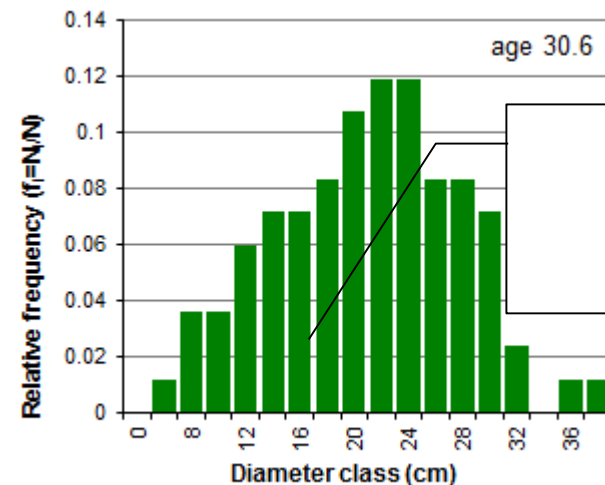
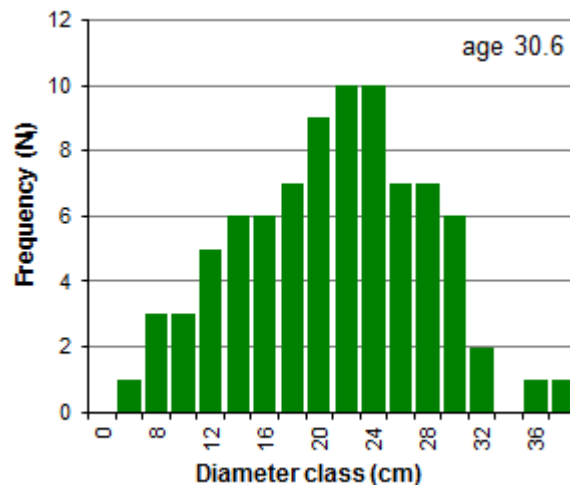
- ✗ The idea behind diameter distribution models is:
  - ✓ to estimate the distribution of trees by diameter classes (diameter distribution)
  - ✓ Usually the simulation of diameter distribution implies the need to predict other variables, namely minimum diameter and some percentile in the upper part of the distribution
  - ✓ to estimate stand volume (total and merchantable) from the diameter distribution, by using tree volume equations

# Diameter distributions of a permanent plot over time



# Diameter distribution in relative frequencies

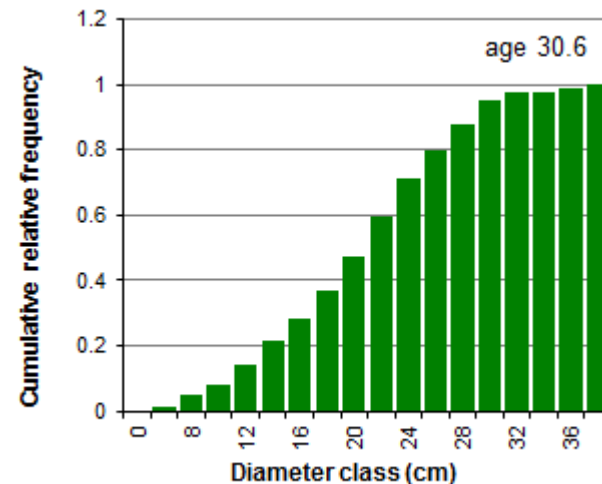
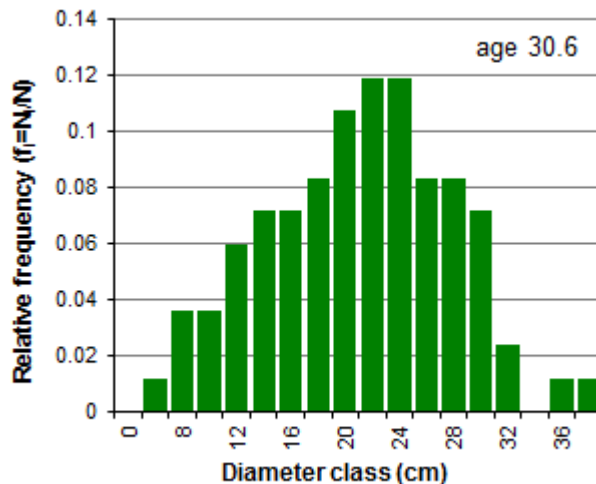
- ✗ The diameter distributions may be expressed in terms of relative frequencies by expressing the frequency each diameter class ( $N_i$ ) relative to the total number of trees per ha ( $N$ )



$$\sum f_i = 1$$

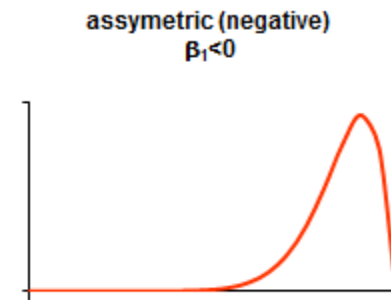
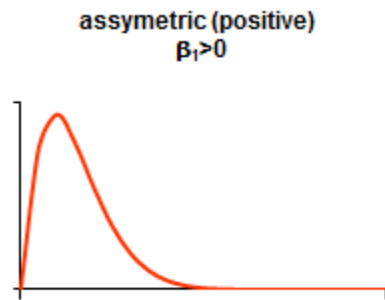
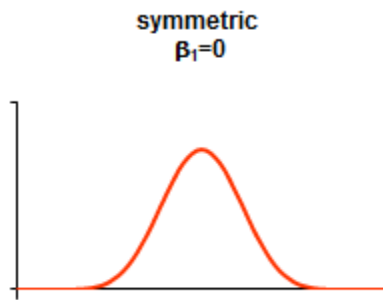
# Diameter distribution in cumulative relative frequencies

✗ The cumulative relative frequency of a diameter distribution leads to the empirical distribution function



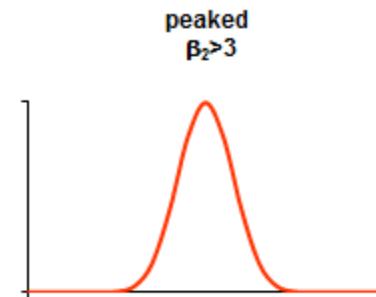
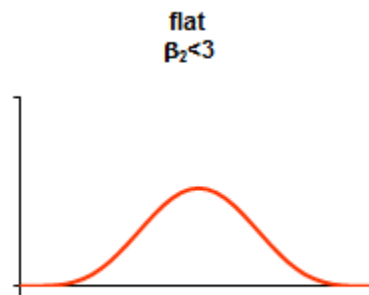
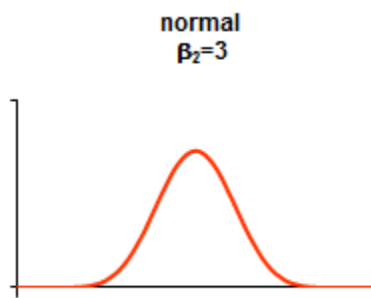
# Modelling diameter distributions

- ✗ A typical diameter distribution for pure, even-aged stands is unimodal and slightly skewed
- ✗ Skewness coefficient ( $\beta_1$ ) is used to measure the symmetry of a distribution



# Modelling diameter distributions

- ✗ Diameter distributions for pure, even-aged stands can also be more or less flat
- ✗ Kurtosis coefficient ( $\beta_2$ ) is used to measure flatness or peakdness of a distribution





# Modelling diameter distributions

- ✗ Diameter distributions can be modelled by a variety of mathematical functions from the probability density functions (pdfs) type
- ✗ Probability density functions express the relative likelihood for a random variable to take on a given value
- ✗ The probability density function is nonnegative everywhere, and its integral over the entire space is equal to one
- ✗ The probability that the random variable takes a value  $< x$  is equal to the integral of the pdf from the start to  $x$

# Probability density functions (pdfs)

- ✗ A pdf is described by a mathematical expression that contains parameters
- ✗ The values of the parameters give a different shape to the pdf
- ✗ For instance, a normal distribution, the most well know pdf, has the following expression

$$f(x; \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

that includes two parameters, the mean ( $\mu$ ) and the standard deviation ( $\sigma$ ), and is designated by  $N(\mu, \sigma)$

# Probability density functions (pdfs)

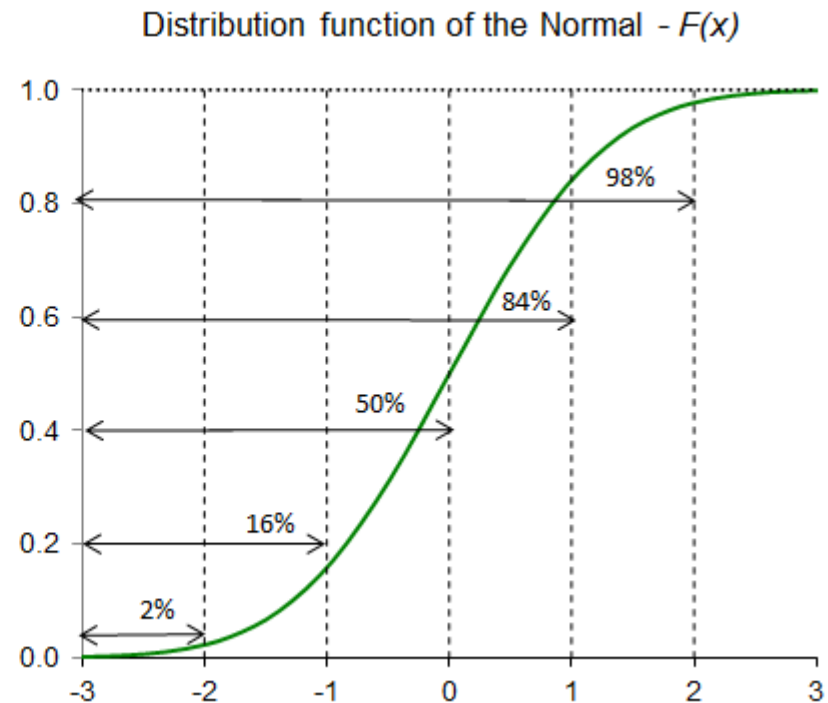
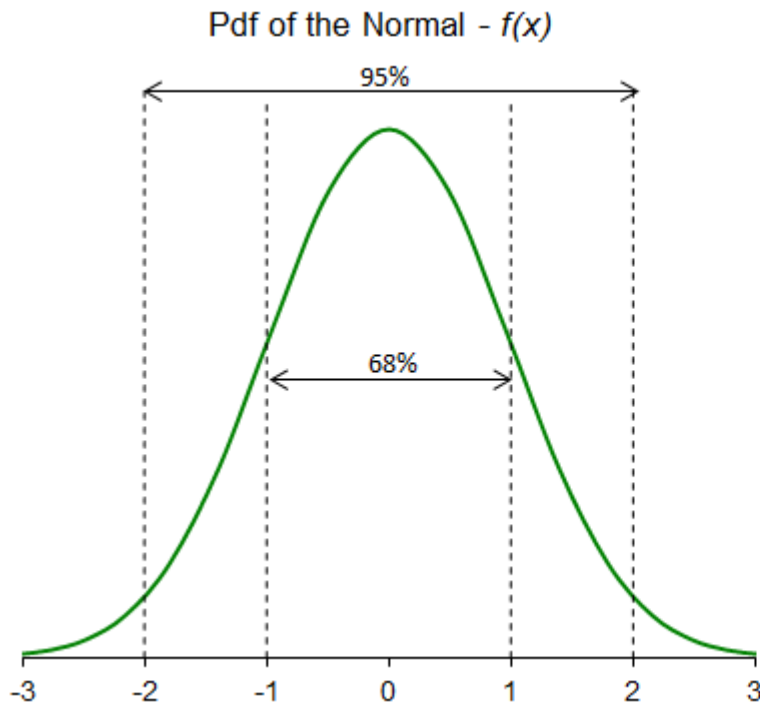
- ✗ Integrating the pdf produces the cumulative distribution function that, for the normal distribution is

$$F(x; \mu, \sigma) = \text{Prob}(X \leq x) = \int_{-\infty}^x \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

- ✗ The Normal distribution is not appropriate to model diameter distributions because of its symmetry

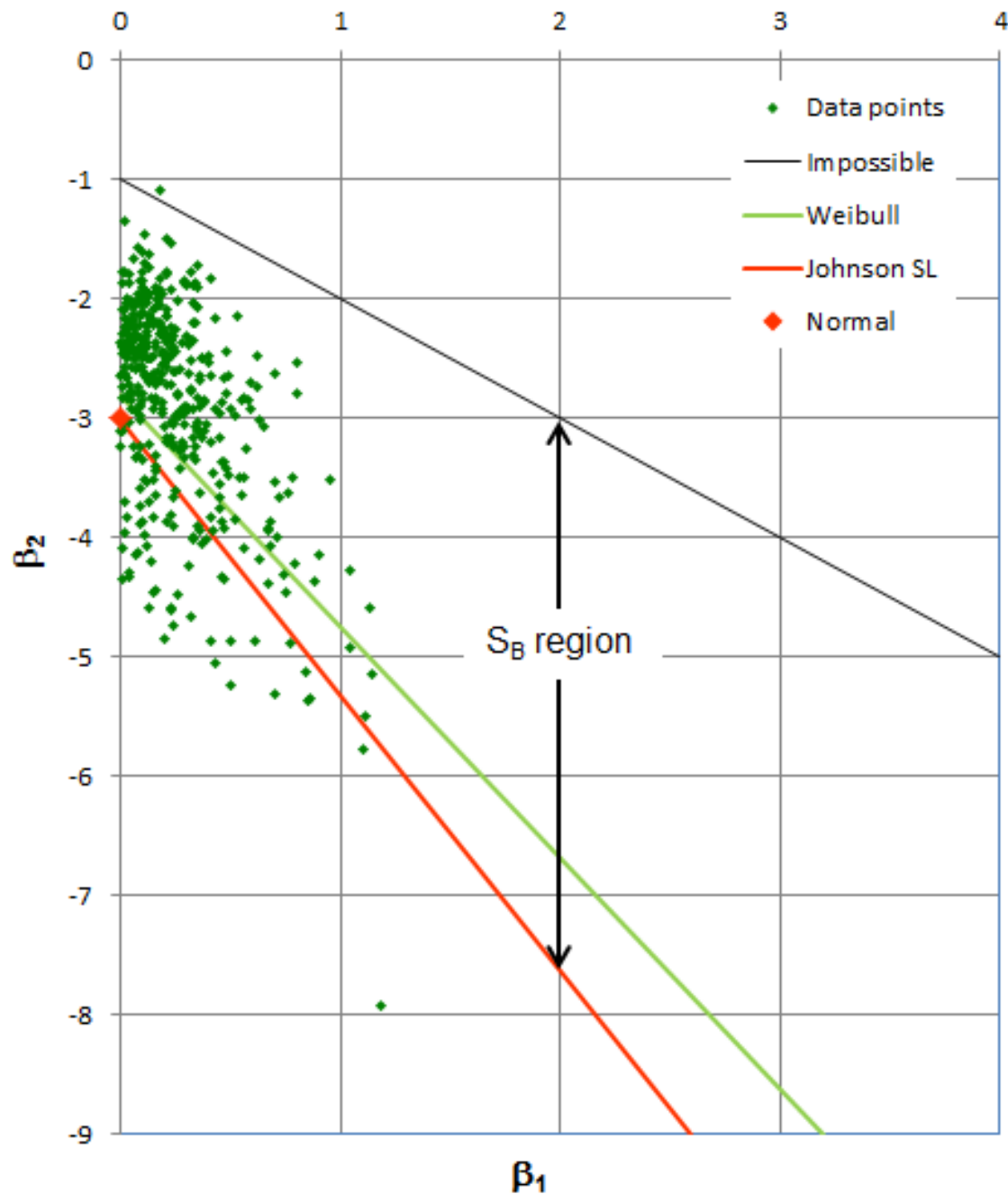
# Probability density functions (pdfs)

✗ The normal distribution is symmetric around the mean ( $\mu$ )



# Probability density functions (pdfs)

- ✗ Other pdfs, that can take different values for the pair  $(\beta_1, \beta_2)$ , have been used for diameter distribution modelling



$(\beta_1, \beta_2)$  values for the pdfs most used for diameter distribution modeling and for a set of eucalyptus permanent plots

Estimatos for  $\beta_1$  and  $\beta_2$ :

$$\sqrt{b_1} = \sqrt{n} \frac{m_3}{m_2^{3/2}}$$

$$b_2 = n \frac{m_4}{m_2^2}$$

$$m_j = \frac{\sum_{i=1}^n (x_i - \bar{x})^j}{n}, \quad j = 2, 3, 4$$

$n$  - number of observations

# Weibull pdf

✗ The Weibull, one of the most used pdfs in diameter distribution modeling, is a three-parameter pdf

$$f(x) = \frac{c}{b} \left( \frac{x-a}{b} \right)^{c-1} \exp \left[ - \left( \frac{x-a}{b} \right)^c \right] \quad (a \leq x < \infty)$$

= 0 otherwise

a - location parameter (related to the  $d_{\min}$ )

b - scale parameter (>0)

c - shape parameter (>0; if  $c > 1$  implies a inverse J shape; if  $c = 3.6$  is close to Normal;  $c < 3.6$  is right skewed; if  $c > 3.6$  is left skewed)

$a+b$  is close to percentile 63% ( $P_{63}$ ) of the distribution

# Weibull pdf

- ✗ Integrating the pdf produces the cumulative distribution function for the Weibull distribution

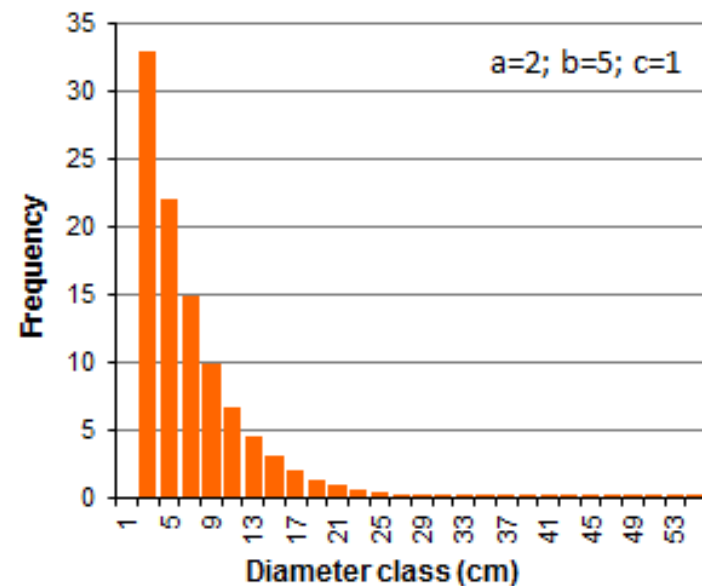
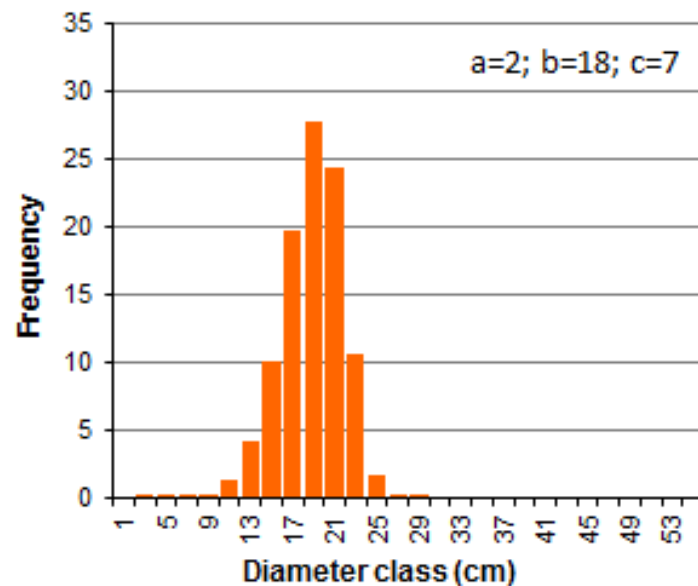
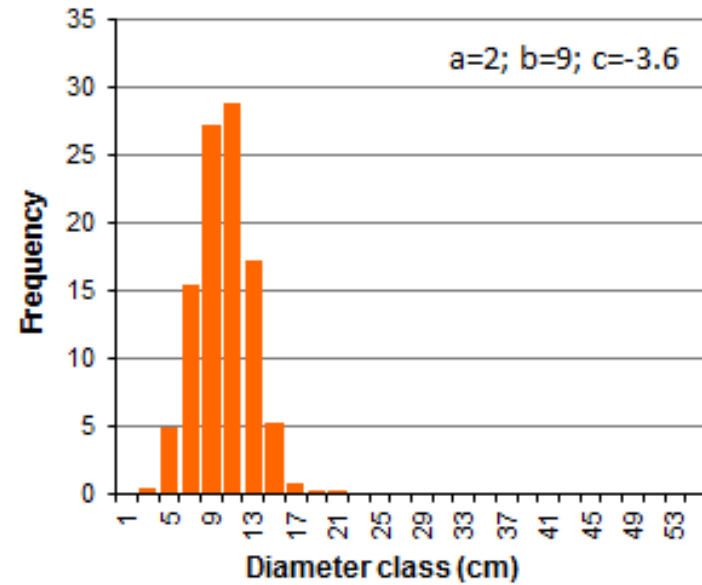
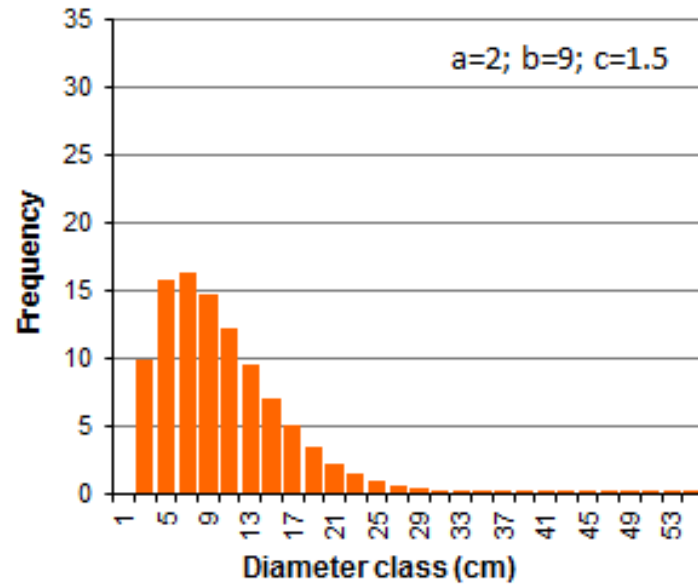
$$F(x) = 1 - \exp\left[-\left(\frac{x-a}{b}\right)^c\right] \quad (a \leq x < \infty)$$

= 0 otherwise

- ✗ The Weibull distribution has the advantage of having a closed integral form which makes it very tractable



# Weibull distribution - examples



# The Johnson's SB system of pdfs

✗ The system of random variables generated by

$$Z = \gamma + \delta \ln\left(\frac{X - \varepsilon}{\varepsilon + \lambda - X}\right)$$

$$\varepsilon < X < \varepsilon + \lambda$$

$$-\infty < \gamma < \infty$$

$$-\infty < \varepsilon < \infty$$

$$\lambda > 0$$

is called the Johnson's SB system of distributions

✗ It is very flexible and can take several shapes

# Johnson's SB system - examples

