

Other Continuous Distributions (Skewed Distributions)

91

+ Gamma distribution:

$$f(x; \alpha, \beta) = \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta}, \quad x \geq 0$$

- $\beta = 1$, \checkmark the standard pdf

$$f(x; \alpha) = \frac{1}{\Gamma(\alpha)} x^{\alpha-1} e^{-x}, \quad x \geq 0$$

where the gamma function $\Gamma(\alpha)$ is defined for $\alpha > 0$

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx \quad \rightarrow \text{Tables for values}$$

\bullet $\Gamma(n) = (n-1)!$ $\forall n \in \mathbb{N}$.

\checkmark the cdf

$$F(x; \alpha) = \int_0^x \frac{1}{\Gamma(\alpha)} y^{\alpha-1} e^{-y} dy \quad \rightarrow \text{Table for values}$$

is called the incomplete gamma function

- the mgf is

$$M_{\underline{X}}(t) = \frac{1}{(1 - \beta t)^\alpha}$$

- the mean and variance:

$$E(\underline{X}) = \mu = \alpha\beta, \quad V(\underline{X}) = \sigma^2 = \alpha\beta^2$$

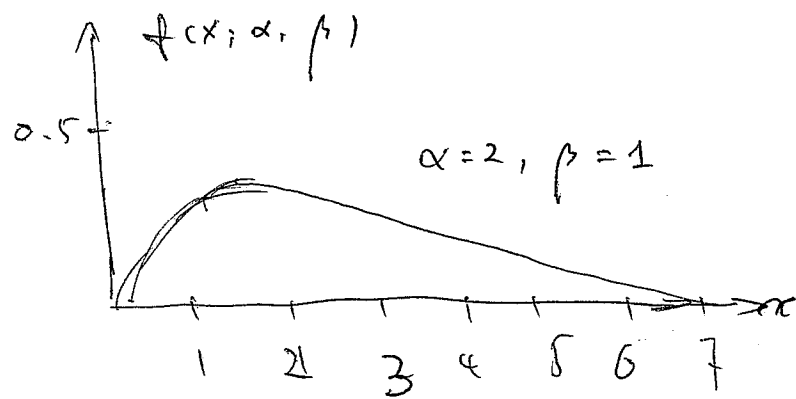
- $\alpha = 1$: \rightarrow the exponential distribution

$$f(x; \beta) = \frac{1}{\beta} x^{-x/\beta}, \quad x \geq 0$$

the cdf $F(x; \beta) = 1 - e^{-x/\beta}, \quad x \geq 0$

* Application

- Gamma distribution: 10% the reaction time \bar{X} (seconds) of a randomly selected individual to a certain stimulus ~~has~~ is supposed to have a standard gamma distribution with $\alpha = 2$. $E(\bar{X}) = \alpha$, $\sqrt{V(\bar{X})} = 2$



20% the survival time \bar{X} in weeks of a randomly selected male mouse exposed to 240 rads of gamma radiation is supposed to have a gamma distribution with $\alpha = 8$ and $\beta = 15$.

$$\left\{ \begin{array}{l} E(\bar{X}) = 8 \times 15 = 120 \text{ weeks} \\ V(\bar{X}) = 8 \times (15)^2 = 1800 = \sigma^2 \\ \Rightarrow \sigma_{\bar{X}} = 42.43 \text{ weeks} \end{array} \right.$$

- Exponential distribution ($\alpha = 1$): 10% the ~~response time~~ number of days \bar{X} between successive calls to a 24-hour "suicide hotline".

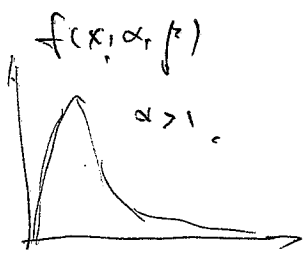
* Chi-squared distribution ($\alpha = \nu/2, \beta = 2$)

$$f(x; \nu) = \frac{1}{2^{\nu/2} \Gamma(\nu/2)} x^{\frac{\nu}{2}-1} e^{-x/2}, \quad x \geq 0$$

⇒ it is the basis for a number of procedures in statistical inference, and is intimately related to normal distributions.

* Weibull distribution:

$$f(x; \alpha, \beta) = \frac{\alpha}{\beta^\alpha} x^{\alpha-1} e^{-\left(\frac{x}{\beta}\right)^\alpha}, \quad x \geq 0$$



- $\alpha = 1$, ⇒ it reduces to exponential distribution.

- the mean and variance

$$E(\bar{X}) = \mu = \beta \Gamma\left(1 + \frac{1}{\alpha}\right), \quad \sigma^2 = V(\bar{X}) = \beta^2 \left[\Gamma\left(1 + \frac{2}{\alpha}\right) - \left[\Gamma\left(1 + \frac{1}{\alpha}\right) \right]^2 \right]$$

- the cdf:

$$F(x; \alpha, \beta) = 1 - e^{-\left(\frac{x}{\beta}\right)^\alpha}, \quad x \geq 0.$$

- Application: better fits to observed data.

1) model engine emission of various pollutants

2) model the corrosion weight loss for a magnesium alloy plate immersed in an inhibited aqueous solution of $MgBr_2$.

* lognormal distribution.

Def: A nonnegative rv \bar{X} is said to have a lognormal distribution if the rv

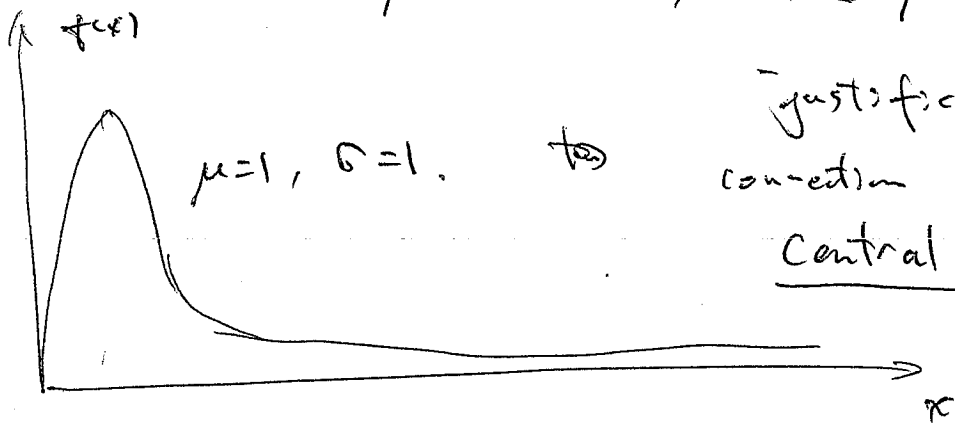
$$Y = \ln(\bar{X})$$

has a normal distribution (with parameter μ and σ^2). - The resulting pdf of \bar{X} is

$$f(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi} \sigma x} e^{-[\ln(x) - \mu]^2 / (2\sigma^2)}, \quad x > 0.$$

- The mean and variance:

$$E(\bar{X}) = e^{\mu + \frac{\sigma^2}{2}}, \quad V(\bar{X}) = e^{2\mu + \sigma^2} (e^{\sigma^2} - 1)$$



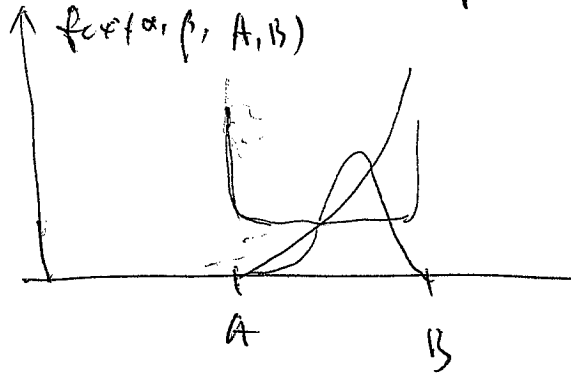
justification in connection with the Central Limit Theorem

- the cdf of \bar{X} can be expressed in terms of the cdf of a standard normal rv Z .

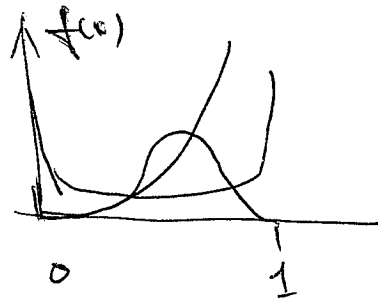
$$\begin{aligned} F(x; \mu, \sigma) &= P(\bar{X} \leq x) = P(\ln(\bar{X}) \leq \ln(x)) \\ &= P\left(Z \leq \frac{\ln(x) - \mu}{\sigma}\right) = \Phi\left(\frac{\ln(x) - \mu}{\sigma}\right). \end{aligned}$$

→ Beta distribution

$$f(x; \alpha, \beta, A, B) = \frac{1}{B-A} \cdot \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \left(\frac{x-A}{B-A}\right)^{\alpha-1} \left(\frac{B-x}{B-A}\right)^{\beta-1}, \quad x \in [A, B]$$



→ $A=0, B=1$: standard beta distribution



→ The mean and variance

$$E(\bar{X}) = \mu = A + (B-A) \frac{\alpha}{\alpha+\beta}$$

$$V(\bar{X}) = \sigma^2 = \frac{(B-A)^2 \alpha \beta}{(\alpha+\beta)^2 (\alpha+\beta+1)}$$

- it provides positive density only for \bar{X} in an interval of finite length.