quantile:
$$x_p = \inf\{x|F_X(x) \geq p\} = \sup\{x|F_X(x) < p\}$$

why this is important? --> Actually the distribution is not normal, higher probability at the extreme data and less data in the middle.

relate to long-term return and risk

kurtosis: fat tail,

skewness: positive and negative --> about positions in risk

management

$$S(x) = E \frac{(X - \mu_x)^3}{\sigma_x^3},$$

kurtosis>3, 高峰肥尾,离群值多

kurtosis<3,离群值少

$$K(x) = E \frac{(X - \mu_x)^4}{\sigma_x^4}.$$

moments of a random variable

t distribution

$$E(X)=0, \quad v>1$$

$$Var(X) = \frac{v}{v-1}, \quad v > 2$$

$$S(X) = 0, v > 3$$

$$K(X) = \frac{6}{v-4}, \quad v > 4.$$

chi-square distribution

$$E(X) = k, \quad Var(X) = 2k$$

symmetry

1. Test for H_0 : symmetry v.s. H_1 : asymmetry:

$$S^* = \frac{\widehat{S}(x)}{\sqrt{6/T}} \sim N(0,1)$$

if normality holds.

thick tail

2. Test for

 $H_0: K(x) = 3(\text{thick tail}) \ v.s. \ H_1: K(x) \neq 3.$

$$K^* = \frac{\widehat{K}(x) - 3}{\sqrt{24/T}} \sim N(0, 1)$$

if normality holds.

normal distribution

$$JB = (K^*)^2 + (S^*)^2 \sim \chi_2^2$$

if normality holds, where χ_2^2 denotes a chi-squared distribution with 2 degrees of freedom.

$$H_0: X \sim N(\mu, \sigma^2) \ v.s.H_1: X \not\sim N(\mu, \sigma^2).$$

Goodness-of-fit Tests

estimation

empirical distribution, better than JB <-- use more information

$$F_n(x) = \frac{1}{n} \sum_{t=1}^n I\{r_t \le x\} = \frac{1}{n} \sum_{i=1}^n I\{r_{(i)} \le x\}$$

$$= \begin{cases} 0 & \text{if } x < r_{(1)} \\ \frac{i}{n} & \text{if } r_{(i)} \le x < r_{(i+1)}, \ i = 1, \dots, n-1 \\ 1 & \text{if } x \ge r_{(n)}. \end{cases}$$

methods

Kolmogorov-Smirnov (D)

Anderson-Darling (A - sq)

Cramr-von Mises (W - sq)

