

# QUANTILES

- \* Quantiles are percentiles but not expressed as a percent; e.g. the 90<sup>th</sup> percentile is the 0.9 quantile.
- \* A baby whose head circumference is at the .9 quantile has a head circumference as great as or greater than 90% of all babies.
- \* Random variables have quantiles.

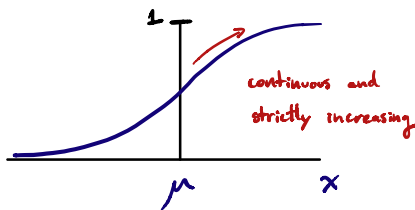
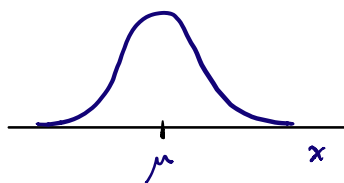
Let  $X$  be a continuous r.v. with a strictly increasing cdf  $F_X$ . Then the  $\theta$ <sup>th</sup> quantile of  $X$  is the value  $z$  which satisfies

$$F_X(z) = \theta.$$

\* So  $X \leq z$  with probability  $\theta$ .

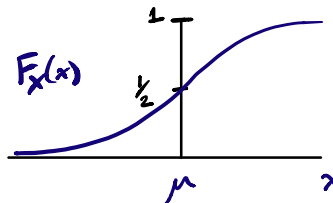
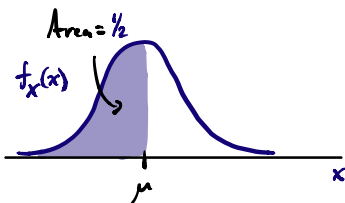
E.g. Suppose  $X \sim \text{Normal}(\mu, \sigma^2)$ .

$$f_X(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi}} \frac{1}{\sigma} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right] \quad F_X(x; \mu, \sigma^2) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} \frac{1}{\sigma} \exp\left[-\frac{(t-\mu)^2}{2\sigma^2}\right] dt$$



Find the 0.5-quantile (called the median).

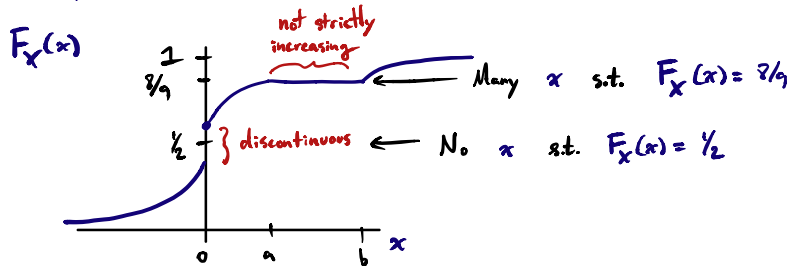
We have  $z = \mu$ , since  $P_X(X \leq \mu) = \frac{1}{2}$ .



If  $F_X$  is not both continuous and strictly increasing, then  $g$  which satisfies  $F_X(g) = \theta$

- may not exist
- may not be unique

E.g. Consider  $F_X$  like:



Better definition comes from the quantile function:

Defn: The quantile function  $Q_X$  of a r.v.  $X$  with cdf  $F_X$  is

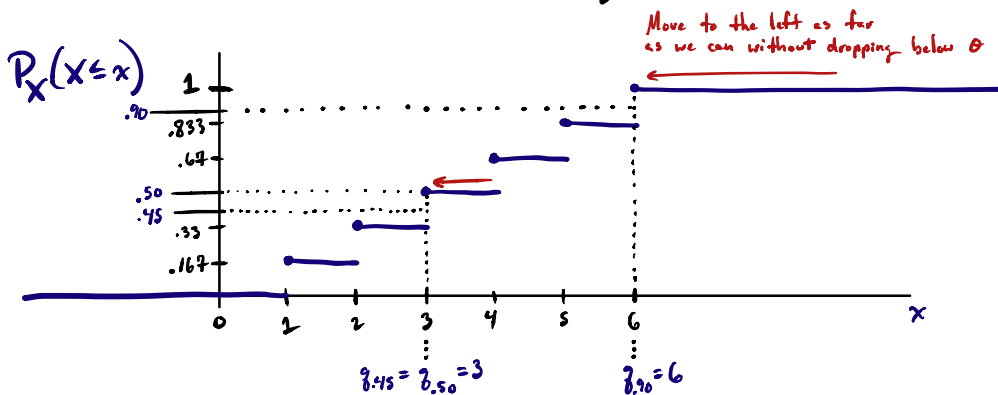
$$Q_X(\theta) = \inf\{x : F_X(x) \geq \theta\}$$

for  $\theta \in (0, 1)$ .

The  $\theta$ -quantile of  $X$  is defined as  $Q_X(\theta)$ .

E.g. Let  $X =$  up-face after rolling a 6-sided die.

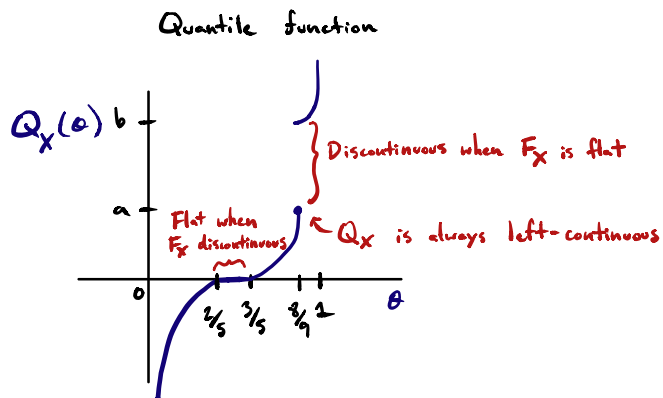
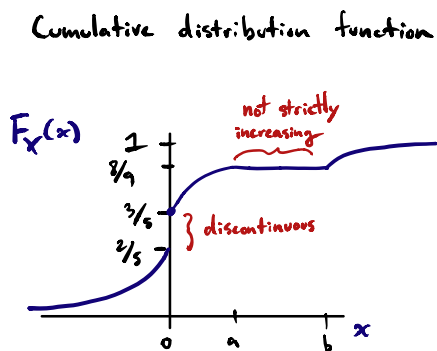
Find the 0.45, 0.50, and 0.70 quantiles.



Interpretation:  $F_{.90} = 6$  : At least 90% of rolls are  $\leq 6$   
 $F_{.50} = 3$  : At least 50% of rolls are  $\leq 3$   
 $F_{.45} = 3$  : At least 45% of rolls are  $\leq 3$

Remark: If  $F_X$  is continuous and strictly increasing, then  $Q_X(\theta) = F_X^{-1}(\theta)$ .

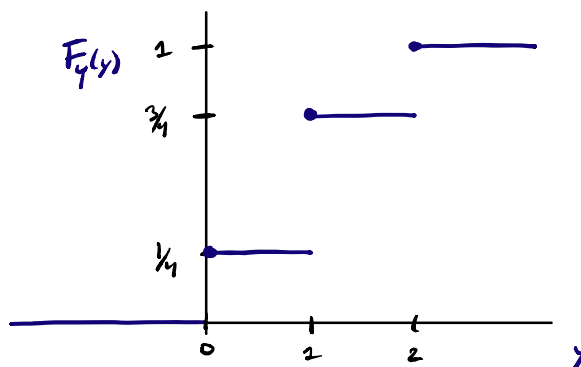
E.g. A cdf with a flat part and a discontinuity.



E.g. Find quantile function of  $Y \sim \text{Binomial}(2, 1/2)$ .

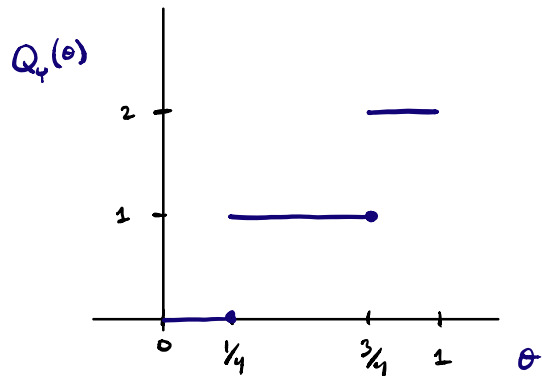
The cdf of  $Y$  is given by

$$F_Y(y) = \begin{cases} 0, & y < 0 \\ 1/4, & 0 \leq y < 1 \\ 3/4, & 1 \leq y < 2 \\ 1, & 2 \leq y \end{cases}$$



The quantile function is given by

$$Q_Y(\theta) = \begin{cases} 2, & \frac{3}{4} < \theta < 1 \\ 1, & \frac{1}{4} < \theta \leq \frac{3}{4} \\ 0, & 0 < \theta \leq \frac{1}{4} \end{cases}$$



Note:  $Q_X : (0,1) \rightarrow \mathcal{X}$ , and  $Q_X$  is left-continuous.

E.g.:  $Q_X$  for Exponential( $\lambda$ ):

Let  $X$  = time in years between eruptions of a volcano

Assume  $X \sim \frac{1}{\lambda} \exp\left[-\frac{x}{\lambda}\right] \mathbb{1}(0 < x < \infty)$ .

(a) Find  $Q_X$ .

We have

$$F_X(x; \lambda) = \begin{cases} 1 - \exp\left[-\frac{x}{\lambda}\right] & 0 < x < \infty \\ 0 & -\infty < x \leq 0 \end{cases} \leftarrow \begin{array}{l} \text{Just focus on support.} \\ \text{Here the cdf is continuous} \\ \text{and strictly increasing.} \end{array}$$

Solve  $F_X\left(\frac{z}{\lambda}; \lambda\right) = \theta$  for  $z$ :

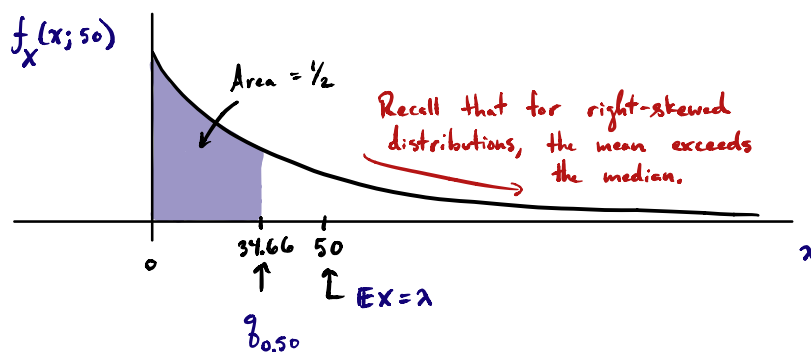
$$1 - \exp\left[-\frac{z}{\lambda}\right] = \theta \quad \Leftrightarrow \quad 1 - \theta = \exp\left[-\frac{z}{\lambda}\right]$$

$$\Leftrightarrow -\lambda \log(1-\theta) = z$$

So  $Q_X(\theta; \lambda) = -\lambda \log(1-\theta)$  for  $\theta \in (0,1)$ .

(b) Suppose  $\lambda = 50$  yrs and find the median time between eruptions.

$$Q_X(.50; 50) = -50 \log(1-.50) = 34.66$$



Ex.  $Q_X$  for Empirical Distribution  $(x_1, \dots, x_n)$ :

Recall the pmf

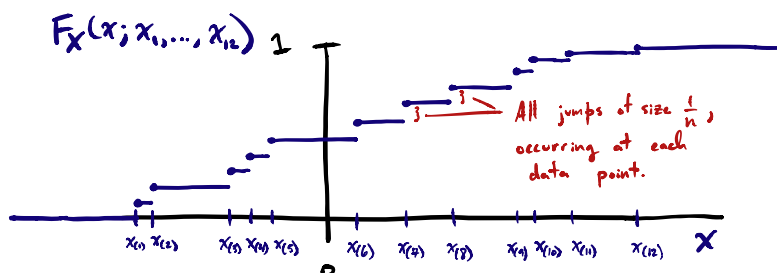
$$p_X(x; x_1, \dots, x_n) = \begin{cases} \frac{1}{n} & \text{for } x \in \{x_1, \dots, x_n\} \\ 0 & \text{otherwise,} \end{cases}$$

which gives the cdf

$$F_X(x; x_1, \dots, x_n) = \frac{\#\{x_1, \dots, x_n \leq x\}}{n} = \frac{1}{n} \sum_{i=1}^n \mathbb{1}(x_i \leq x).$$

Denote by  $x_{(1)}, \dots, x_{(n)}$  the data points when reordered such that  $x_{(1)} < x_{(2)} < \dots < x_{(n)}$ .

The cdf of the empirical distribution based on 12 "data points" looks like:



$$\begin{aligned}
 \text{Then } Q_X(\theta) &= \inf \left\{ x : \frac{\#\{x_1, \dots, x_n \leq x\}}{n} \geq \theta \right\} \\
 &= \min \left\{ x_i : \frac{\#\{x_1, \dots, x_n \leq x_i\}}{n} \geq \theta \right\} \\
 &= x_{(\lceil n\theta \rceil)} \quad \text{for } \theta \in (0, 1),
 \end{aligned}$$

where  $\lceil \cdot \rceil$  is the ceiling function:  $\lceil x \rceil = j$  if  $j-1 < x \leq j$ ,  $j$  an integer.

E.g. Consider the empirical distribution based on the data points

$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$	$x_9$	$x_{10}$
1.98	1.01	0.86	-0.15	1.55	0.20	-0.90	1.10	-0.37	-0.77
$x_{(10)}$	$x_{(9)}$	$x_{(8)}$	$x_{(6)}$	$x_{(4)}$	$x_{(5)}$	$x_{(1)}$	$x_{(7)}$	$x_{(3)}$	$x_{(2)}$

$$\text{Let } X \sim p_X(x; x_1, \dots, x_n).$$

$$\text{Then } \int_{0.25} = x_{(\lceil 10 \cdot 0.25 \rceil)} = x_{(3)} = -0.37$$

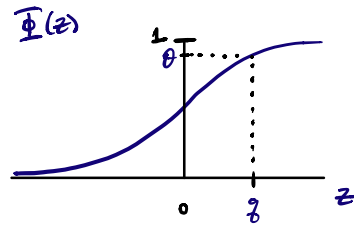
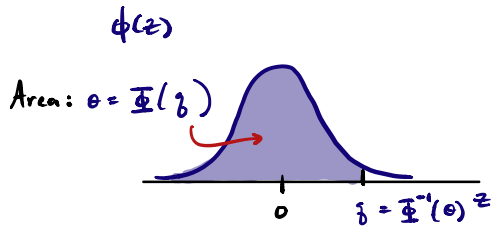
$$\int_{0.50} = x_{(\lceil 10 \cdot 0.5 \rceil)} = x_{(5)} = 0.20$$

$$\int_{0.75} = x_{(\lceil 10 \cdot 0.75 \rceil)} = x_{(8)} = 1.10$$

E.g.  $Q_X$  for Normal( $\mu, \sigma^2$ ):

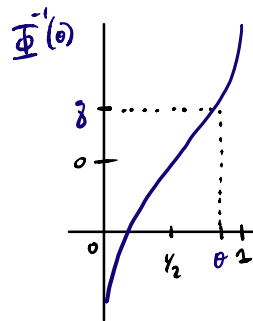
Begin with  $Z \sim \text{Normal}(0, 1)$ .

The cdf of  $Z$  is  $\Phi(z) = \int_{-\infty}^z \phi(t) dt$ ,  $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$ .



\*  $\Phi(z)$  is continuous and strictly increasing, so

$$Q_Z(\theta) = \Phi^{-1}(\theta)$$



\*  $\Phi^{-1}(\theta)$  does not have a simple expression.  
We use tables/computer.

In  $\mathbb{R}$ :  $z_{\text{norm}}(\theta)$

\* If  $X \sim \text{Normal}(\mu, \sigma^2)$  then  $Q_X(\theta) = \mu + \sigma Q_Z(\theta)$

Derivation: Let  $z$  be the  $\theta^{\text{th}}$  quantile of  $X$ . Then

$$\theta = P_X(X \leq z) = P_X\left(\frac{X - \mu}{\sigma} \leq \frac{z - \mu}{\sigma}\right)$$

$$= P_Z(z \leq \frac{g-\mu}{\sigma})$$

$$= \Phi\left(\frac{g-\mu}{\sigma}\right)$$

$$\Leftrightarrow \Phi^{-1}(\theta) = \frac{g-\mu}{\sigma}$$

$$\Leftrightarrow g = \mu + \sigma \Phi^{-1}(\theta)$$

$$\text{So } Q_X(\theta) = \mu + \sigma Q_Z(\theta).$$

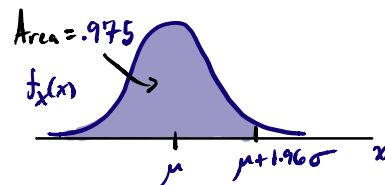
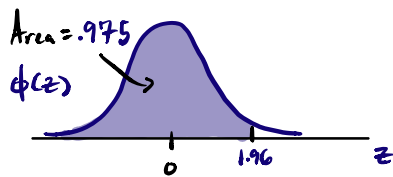
\* Therefore, to get quantiles of  $X$ , we transform quantiles of  $Z$ .

$$Q_X(0.975) = \mu + Q_Z(0.975)\sigma = \mu + \Phi^{-1}(0.975)\sigma = \mu + 1.96\sigma$$

Obtain from a table or  $qnorm(0.975)$  in R

(i) Get  $\text{Normal}(0,1)$  quantile

(ii) scale and shift to get  $\text{Normal}(\mu, \sigma^2)$  quantile



## COMPUTING QUANTILES

R has for every distribution the four functions

- r — : generate realizations from — distribution
- d — : compute value of pmf or pdf of — distribution
- p — : compute value of cdf of — distribution
- q — : compute quantile of — distribution

For quantiles