

Quantiles

popper12: choose d for 1-5

Let F be a given cumulative distribution and let p be any real number between 0 and 1. The **(100p)th percentile** of the distribution of a continuous random variable X is defined as

$$F^{-1}(p) = \min\{x | F(x) \geq p\}.$$

For continuous distributions, $F^{-1}(p)$ is the smallest number x such that $F(x) = p$.

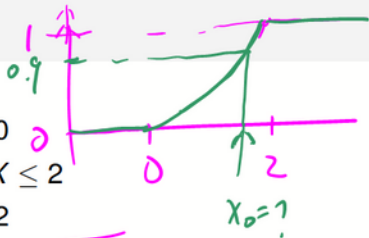


Determine the Percentiles

Given a cdf,

$$P(X \leq c) = P(X < c)$$

$$F(x) = \begin{cases} 0 & X < 0 \\ \frac{1}{8}x^3 & 0 \leq X \leq 2 \\ 1 & X > 2 \end{cases}$$



1. Determine the 90th percentile.

set $F(x_0) = 0.9$, solve for x_0 .

2. Determine the 50th percentile.

set $\frac{1}{8}x_0^3 = 0.90$
 $\Rightarrow x_0 = (8 \times 0.90)^{\frac{1}{3}}$

set $\frac{1}{8}x_0^3 = 0.5$, solve for x_0 .

$$x_0 = (8 \times 0.5)^{\frac{1}{3}}$$

- 3. Find the value of c such that $P(X \leq c) = 0.75$.

$P(X \leq c) = F(c)$

set $\frac{1}{8}c^3 = 0.75$, solve for c .

MATH 3339

Statistics for the Sciences

Sec 5.4; 5.5

Wendy Wang
wwang60@central.uh.edu

Lecture 10 - 3339

Outline

for discrete.

$$E(X) = \sum x P(X=x)$$

- 1 Expected Values
- 2 Exponential Distribution
- 3 Gamma Distribution
- 4 Normal Distribution
- 5 The Empirical Rule

Expected Values for Continuous Random Variables



The **expected** or **mean value** of a continuous random variable X with pdf $f(x)$ is

$$E(X) = \int_{-\infty}^{\infty} xf(x)dx.$$

More generally, if h is a function defined on the range of X , $h(x)$

$$E(\underline{h(X)}) = \int_{-\infty}^{\infty} \underline{h(x)}f(x)dx.$$

for example, $h(x) = x^2$

$$E(x^2) = \int_{-\infty}^{\infty} (x^2) f(x) dx$$

$$\text{Var}(X) = E((X - \mu)^2) = E(X^2) - (E(X))^2$$

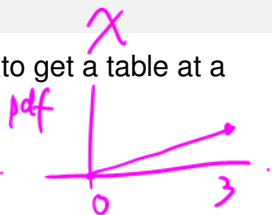
$$\text{sd}(X) = \sqrt{\text{Var}(X)}$$

Example From Quiz 8

$$C' = 0$$

Let X be the amount of time (in hours) the wait is to get a table at a restaurant. Suppose the cdf is represented by

$$F(X) = \begin{cases} 0 & x < 0 \\ \frac{x^2}{9} & 0 \leq x \leq 3 \\ 1 & x > 3 \end{cases}$$



Use the cdf to determine $E[X]$.

$$\rightarrow f(x) = F'(x) = \begin{cases} 0, & x < 0, \text{ or } x > 3 \\ \frac{2x}{9}, & 0 \leq x \leq 3 \end{cases}$$

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx$$

$$= \int_{-\infty}^0 (x \cdot 0) dx + \int_0^3 (x \cdot \frac{2x}{9}) dx + \int_3^{\infty} (x \cdot 0) dx$$

$$= \int_0^3 \frac{2x^2}{9} dx = \left. \frac{2}{27} x^3 \right|_0^3 = \frac{2}{27} (3^3 - 0) = 2.$$

The Exponential Distribution

binomial(1, p)

X is said to have an **exponential distribution** with parameter λ ($\lambda > 0$) if the pdf of X is:

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

Where λ is a rate parameter, we write $X \sim \text{Exp}(\lambda)$. The cdf of a exponential random variable is:

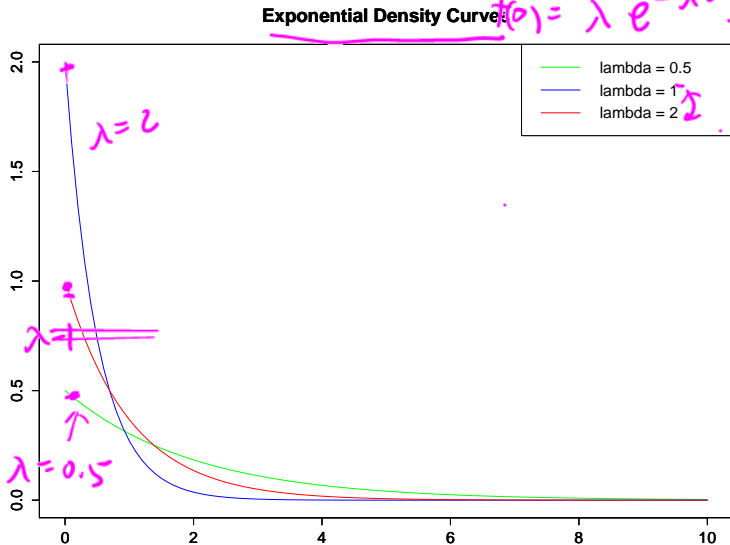
$$F(x) = \begin{cases} 0 & x < 0 \\ 1 - e^{-\lambda x} & x \geq 0 \end{cases}$$

The mean of the exponential distribution is $\mu_x = E(X) = \frac{1}{\lambda}$ the standard deviation is also $\frac{1}{\lambda}$.

Exponential Density Curves

$$f(x) = \lambda e^{-\lambda x}$$

$$f(0) = \lambda e^{-\lambda \cdot 0} = \lambda$$



Exponential Distribution Related to the Poisson Distribution

λ
continuous

\uparrow
discrete

- The exponential distribution is frequently used as a model for the distribution of times between the occurrence of successive events until the first arrival.
- Suppose that the number of events occurring in any time of length t has a Poisson distribution with parameter λt .
- Where λ , the rate of the event process, is the expected number of events occurring in 1 unit of time.
- The number of occurrences are in non overlapping intervals and are independent of one another.
- Then the distribution of elapsed time between the occurrence of two successive events is exponential with parameter λ .

Example

- Suppose you usually get 3 phone calls per hour.
- 3 phone calls per hour means that we would expect one phone call every $\frac{1}{3}$ hour so $\mu = \frac{1}{3}$.
- Compute the probability that a phone call will arrive within the next hour.

Example # of phone calls/hr \sim poisson ($\lambda = 3/\text{hr.}$)

- Suppose you usually get 3 phone calls per hour.
- 3 phone calls per hour means that we would expect one phone call every $\frac{1}{3}$ hour so $\mu = \frac{1}{3}$.
- Compute the probability that a phone call will arrive within the next hour.

X : time it takes for the next phone call

$$X \sim \text{exp}(\lambda = 3) \quad E(X) = \frac{1}{\lambda} = \frac{1}{3}$$

$$P(X \leq 1) = P_{\text{exp}}\left(\frac{1}{3}, 3\right)$$

$$P(X=1) = 0;$$

R code

```
> pexp(1, 3)
[1] 0.9502129
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- To find the probability of an exponential distribution in R: pexp(x, λ).
- To find the percentile (quantile) in R: qexp(x, λ).

Examples

Applications of the exponential distribution occurs naturally when describing the waiting time in a homogeneous Poisson process. It can be used in a range of disciplines including queuing theory, physics, reliability theory, and hydrology. Examples of events that may be modeled by exponential distribution include:

- The time until a radioactive particle decays
- The time between clicks of a Geiger counter
- The time until default on payment to company debt holders
- The distance between roadkills on a given road
- The distance between mutations on a DNA strand
- The time it takes for a bank teller to serve a customer
- The height of various molecules in a gas at a fixed temperature and pressure in a uniform gravitational field
- The monthly and annual maximum values of daily rainfall and river discharge volumes

Example from Quiz 8

- $\frac{1}{\lambda} = \mu = 6 \Rightarrow \lambda = \frac{1}{6} \quad X \sim \exp(\lambda = \frac{1}{6})$
1. Suppose the time a child spends waiting at for the bus as a school bus stop is exponentially distributed with mean 6 minutes. Determine the probability that the child must wait at least 9 minutes on the bus on a given morning.

$$\begin{aligned} P(X \geq 9) &= 1 - P(X < 9) = 1 - P(X \leq 9) \\ &= 1 - \text{pexp}(9, \frac{1}{6}) \end{aligned}$$

2. Suppose the time a child spends waiting at for the bus as a school bus stop is exponentially distributed with mean 4 minutes. Determine the probability that the child must wait between 3 and 6 minutes on the bus on a given morning.

$$\begin{aligned} X &\sim \exp(\lambda = \frac{1}{4}) \\ P(3 \leq X \leq 6) &= P(X < 6) - P(X < 3) \\ &= \text{pexp}(6, \frac{1}{4}) - \text{pexp}(3, \frac{1}{4}) \end{aligned}$$

The "Memoryless" Property

Another application of the exponential distribution is to model the distribution of component lifetime.

- Suppose component lifetime is exponentially distributed with parameter λ .
- After putting the component into service, we leave for a period of t_0 hours and then return to find the components still working; what now is the probability that it last at least an addition t hours?
- We want to find $P(T \geq t + t_0 | T \geq t_0) = P(T \geq t)$

The Gamma Function

The gamma function $\Gamma(\alpha)$ is defined by:

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx$$

$$\underline{\underline{\alpha=1}} \quad \Gamma(1) = \int_0^{\infty} x^{1-1} e^{-x} dx = \int_0^{\infty} e^{-x} dx$$
$$\rightarrow \exp(-x) \Big|_0^{\infty} = 1$$

Properties of the Gamma Function

The most important properties of the gamma function are the following:

1. For any $\alpha > 1$, $\Gamma(\alpha) = (\alpha - 1)\Gamma(\alpha - 1)$
2. For any positive integer, n , $\Gamma(n) = (n - 1)!$
3. $\Gamma(\frac{1}{2}) = \sqrt{\pi}$

$$\Gamma(\frac{3}{2}) = (\frac{3}{2} - 1) \Gamma(\frac{3}{2} - 1) = 0.5 \Gamma(0.5) = \frac{\sqrt{\pi}}{2}$$

$\alpha = \frac{3}{2}$

The PDF of a Gamma Distribution

A continuous random variable X is said to have a **gamma distribution** if the pdf of X is

$$f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

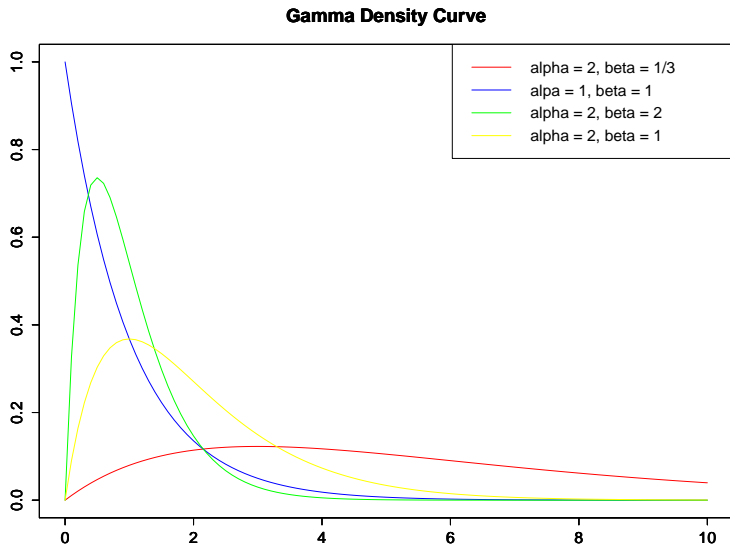
where parameters α and β satisfy $\alpha > 0$, $\beta > 0$.

Gamma Distribution Related to the Poisson

- Gamma distribution is a distribution that arises naturally in processes for which the waiting times between events are relevant.
- It can be thought of as a waiting time between Poisson distributed events, until k arrivals.
- Thus the scale parameter can also be thought of as the inverse of the rate parameter (λ), $\frac{1}{\lambda}$.
- Then $\alpha = k$ and $\beta = \frac{1}{\lambda}$
- In R, $P(X \leq x) = \text{pgamma}(x, \alpha, \frac{1}{\beta})$

or, $\text{pgamma}(x, \alpha, \text{scale} = \beta)$

Gamma Density Curve



Applications of the Gamma Distribution

The gamma distribution can be used a range of disciplines including queuing models, climatology, and financial services. Examples of events that may be modeled by gamma distribution include:

- The amount of rainfall accumulated in a reservoir
- The size of loan defaults or aggregate insurance claims
- The flow of items through manufacturing and distribution processes
- The load on web servers
- The many and varied forms of telecom exchange

Example

$$\lambda = 5 / \text{min}$$

Suppose that the telephone calls arriving at a particular switchboard follow a Poisson process with an average of 5 calls coming per minute. What is the probability that up to a minute will elapse until 2 calls have come in to the switchboard?

$$X \sim \text{gamma}(\alpha=2, \beta=\frac{1}{5})$$

- Average of 5 calls coming per minute means that $\beta = \frac{1}{5}$.
- Until 2 calls have come into the switchboard means that $\alpha = 2$.

$$P(X \leq 1 \text{ min}) = \text{pgamma}(1, 2, 5)$$

(Note: In the original image, a curved arrow points from the '1' in the formula to the '1 min' in the text, and an upward arrow points from the '5' to the 'λ' in the top equation.)

Mean and Variance of the Gamma Distribution

The mean and variance of a random variable X having the gamma distribution are:

$$\begin{aligned} E(X) &= \mu = \alpha\beta \\ \text{Var}(X) &= \sigma^2 = \alpha\beta^2 \end{aligned}$$

$$= \alpha \cdot \beta^2$$

Example of Gamma Distribution

$$\mu = 24$$

$$sd = 12$$

Suppose that a transistor of a certain type is subjected to an accelerated life test, the lifetime Y (in weeks) has a gamma distribution with a mean of 24 and a standard deviation of 12.

1. Find the values of α and β .

$$\alpha \beta = 24$$

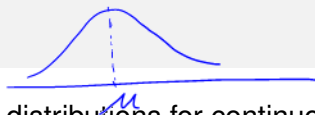
$$\alpha \beta^2 = 12^2 = 144.$$

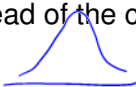
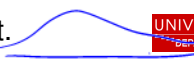
$$\beta = 6$$

$$\alpha = 4$$

2. Find $P(Y \leq 24) = \text{pgamma}(24, 4, \frac{1}{6})$

The Normal distributions

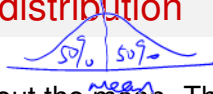


- Common type of probability distributions for continuous random variables.
- The highest probability is where the values are centered around the mean. Then the probability declines the further from the mean a value gets.
- These curves are symmetric, single-peaked, and bell-shaped.
- The mean μ is located at the center of the curve and is the same as the median.
- The standard deviation σ controls the spread of the curve.
- If σ is small then the curve is tall and slim. 
- If σ is large then the curve is short and fat. 

Normal distributions important to statistics?

- Normal distributions are good descriptions for some distributions of real data.
- Normal distributions are good approximations to the results of many kinds of chance outcomes.
- Many statistical inference procedures based on Normal distributions work well for other roughly symmetric distributions.

Facts about the Normal distribution



- The curve is symmetric about the mean. That is, 50% of the area under the curve is below the mean. 50% of the area under the curve is above the mean.
- The spread of the curve is determined by the standard deviation.
- The area under the curve is with respect to the number of standard deviations a value is from the mean.
- Total area under the curve is 1.
- Area under the curve is the same as probability within a range of values.



- If X follows a Normal distribution with mean μ and standard deviation σ we would write it as $X \sim N(\mu, \sigma)$.

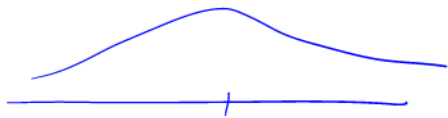
$$P(-1 < X < 3) = \text{Area under the curve}$$



$$\mu = 1$$

$$\sigma = 0.5$$

$$X \sim N(\underline{1}, \underline{0.5})$$



$$\mu = 0$$

$$\sigma = 2.5$$

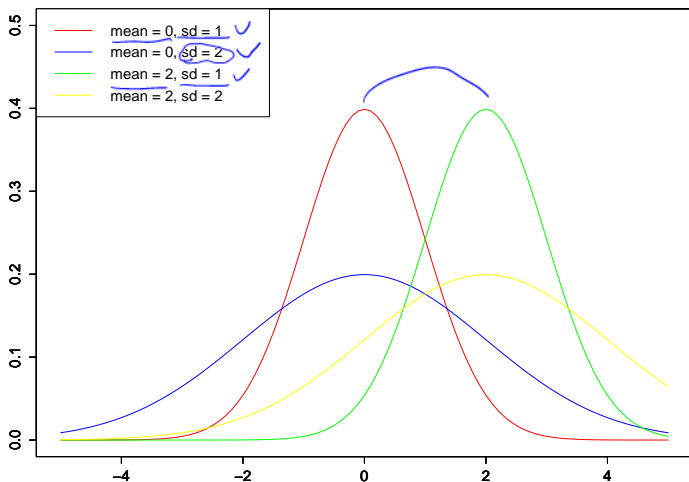
$$X \sim N(\overset{\mu}{0}, \overset{\sigma}{2.5})$$

Density Function

This is the graph of the density function.

μ : location parameter
 σ : scale parameter

Density Curves for Normal Distributions



PDF of a Normal Distribution

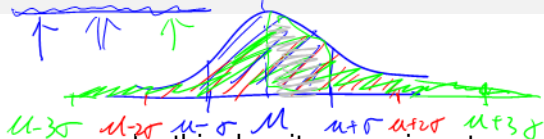
A continuous random variable X is said to have a **Normal distribution** with parameters μ and σ (or μ and σ^2), where $-\infty < \mu < \infty$ and $0 < \sigma$, if the pdf of X is:

$$\underline{f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2}}$$

For all $-\infty < x < \infty$.

$$\int_{-\infty}^{\infty} f(x) dx = \int \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = 1$$

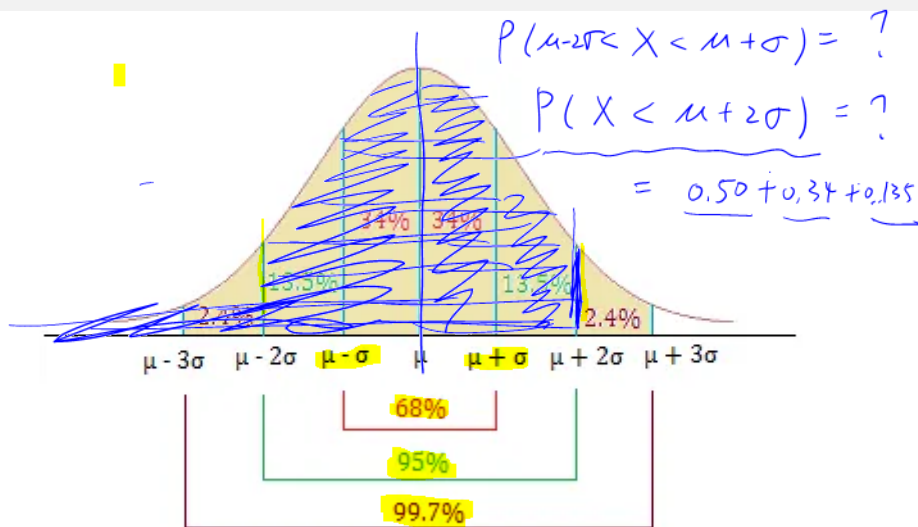
The Empirical Rule or 68-95-99.7 Rule



- Unfortunately to find the area under this density curve is not as easy to compute. Thus we can use the following approximate rule for the area under the **Normal** density curve.
- In the Normal Distribution with mean μ and standard deviation σ :
 - 68% of the observations fall within 1 standard deviation σ of the mean μ . $P(\mu - \sigma < X < \mu + \sigma) = 0.68$
 - 95% of the observations fall within 2 standard deviations 2σ of the mean μ . $P(\mu - 2\sigma < X < \mu + 2\sigma) = 0.95$
 - 99.7% of the observations fall within 3 standard deviations 3σ of the mean μ . $P(\mu - 3\sigma < X < \mu + 3\sigma) = 0.997$

$$P(\mu < X < \mu + \sigma) = 0.34$$

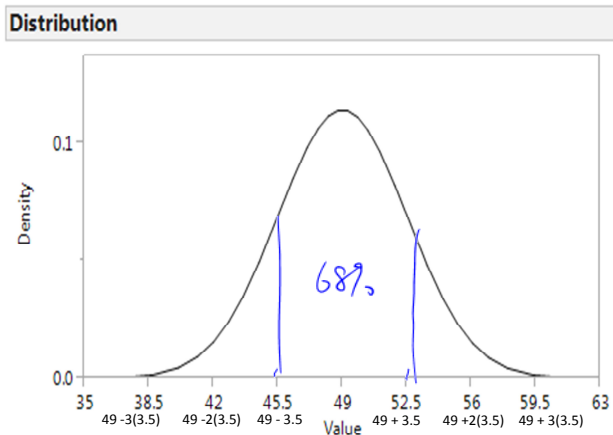
The 68-95-99.7 rule for Normal distributions



Empirical rule

MPG of Prius

The MPG of Prius has a Normal distribution with mean $\mu = 49$ mpg and standard deviation $\sigma = 3.5$ mpg.

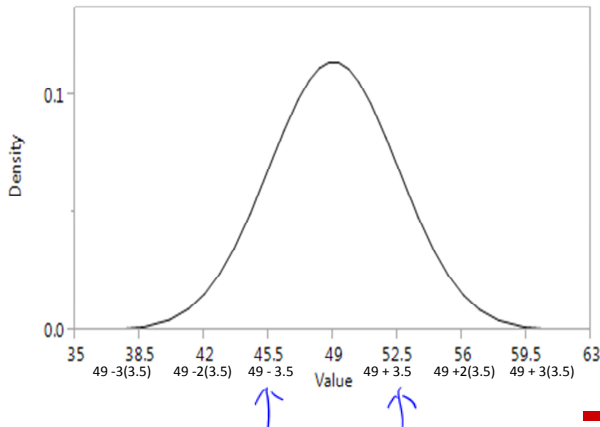


MPG of Prius

About what percent have between 45.5 and 52.5 mpg?

$$P(45.5 < X < 52.5) = 0.68$$

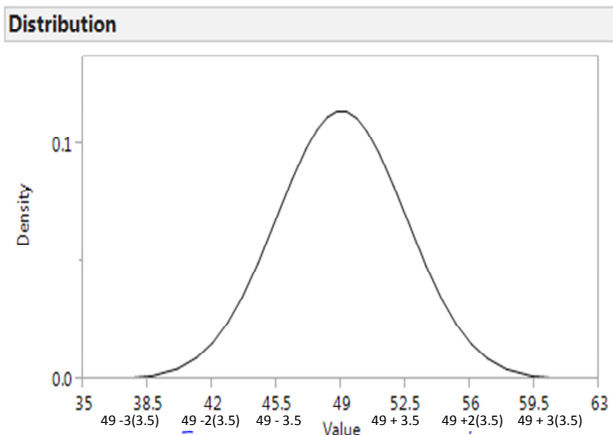
Distribution



MPG of Prius

About what percent have between 42 and 56 mpg?

$$0.95 = P(\mu - 2\sigma < X < \mu + 2\sigma)$$

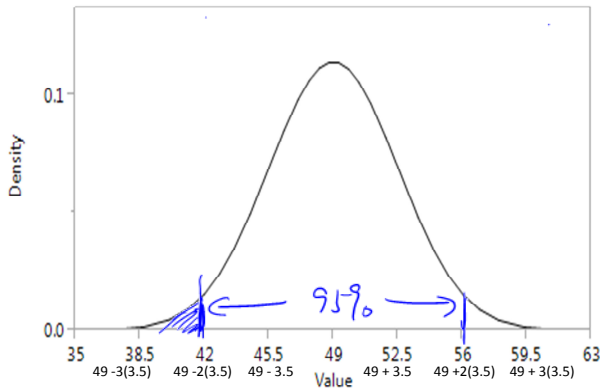


MPG of Prius

About what percent have less than 42 mpg?

$$P(X < 42) = \frac{1 - 95\%}{2} = 2.5\%$$

Distribution



Orange Juice

$$X \sim N(\mu = 4.7, \sigma = 0.40)$$

An orange juice producer buys all his oranges from a large orange grove. The amount of juice squeezed from each of these oranges is approximately normally distributed, with a mean of 4.70 ounces and a standard deviation of 0.40 ounce.

1. What is the probability that an orange from this orange grove has between 3.9 and 5.5 ounces of juice?

$$P(3.9 < X < 5.5) = P(\mu - 2\sigma < X < \mu + 2\sigma) = 0.95$$

$5.5 = 4.7 + 0.4 * 2$

2. What is the probability that an orange from this orange grove has less than 4.7 ounces of juice?

$$P(X < 4.7) = P(X < \mu) = 0.5$$

3. Approximately ~~95%~~ of the oranges have juice between what two middle values? 99.7% ($X - 3\sigma$, $X + 3\sigma$)

$$(4.7 - 3 * 0.4, 4.7 + 3 * 0.4)$$

4. What percent of values less than 4?

$$P(X < 4) = ?$$

$$4 = 4.7 - 0.7$$