

Chapter 4. Part 3

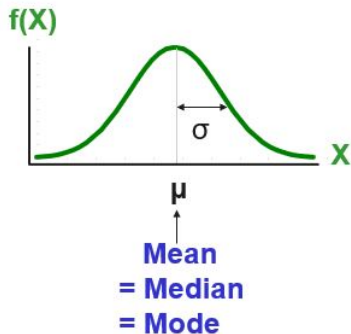


February 11, 2022

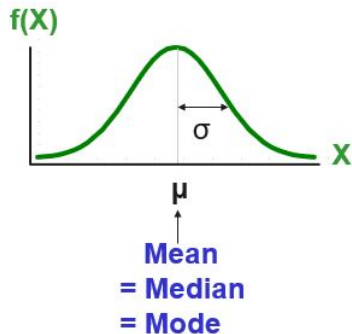
1 4.6 The Normal Distribution

2 4.8 Exponential Distribution

4.6 The Normal Distribution

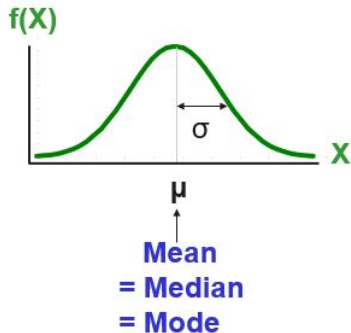


4.6 The Normal Distribution



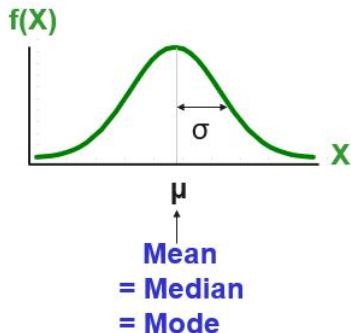
- Bell Shaped

4.6 The Normal Distribution



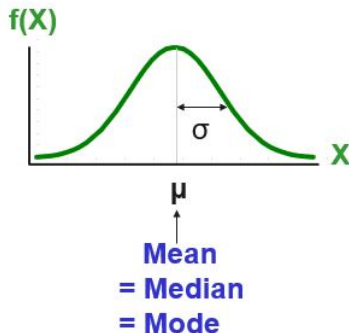
- Bell Shaped
- Symmetrical

4.6 The Normal Distribution



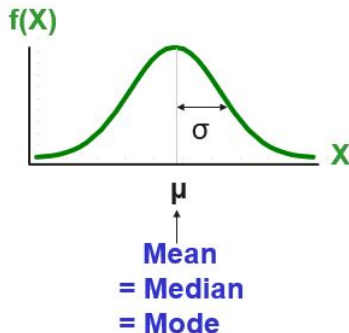
- Bell Shaped
- Symmetrical
- Mean, Median and Mode are equal.

4.6 The Normal Distribution



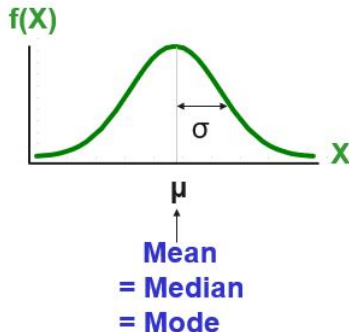
- Bell Shaped
- Symmetrical
- Mean, Median and Mode are equal.
- Location is determined by the mean, μ .

4.6 The Normal Distribution



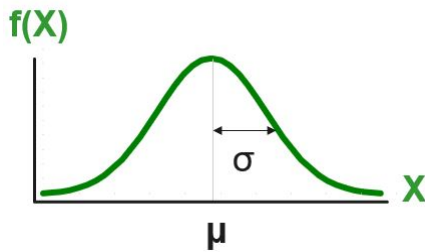
- Bell Shaped
- Symmetrical
- Mean, Median and Mode are equal.
- Location is determined by the mean, μ .
- Spread is determined by the standard deviation, σ .

4.6 The Normal Distribution



- Bell Shaped
- Symmetrical
- Mean, Median and Mode are equal.
- Location is determined by the mean, μ .
- Spread is determined by the standard deviation, σ .

Normal Distribution Shape

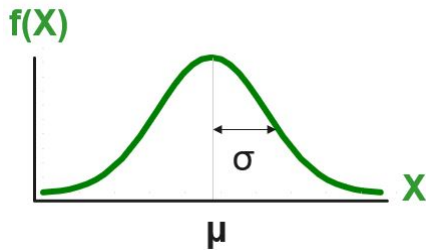


Increase μ shifts the distribution right.

Decrease μ shifts the distribution left.

Changing σ increases or decreases the spread.

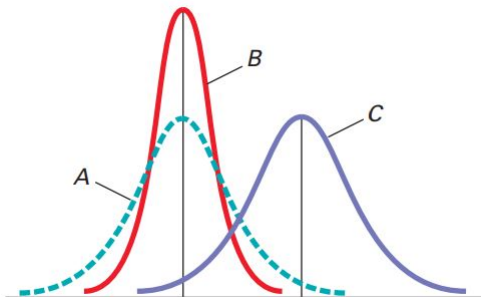
Normal Distribution Shape



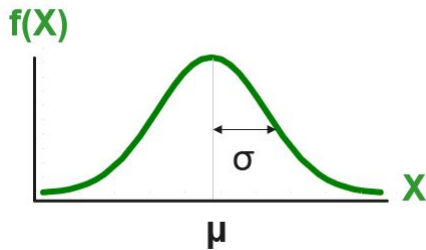
Increase μ shifts the distribution right.

Decrease μ shifts the distribution left.

Changing σ increases or decreases the spread.



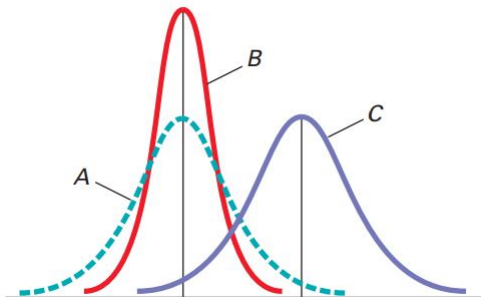
Normal Distribution Shape



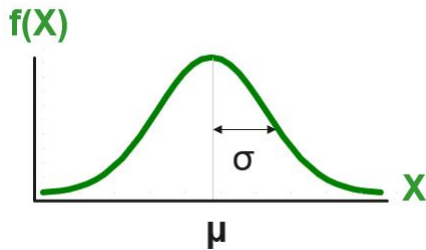
Increase μ shifts the distribution right.

Decrease μ shifts the distribution left.

Changing σ increases or decreases the spread.



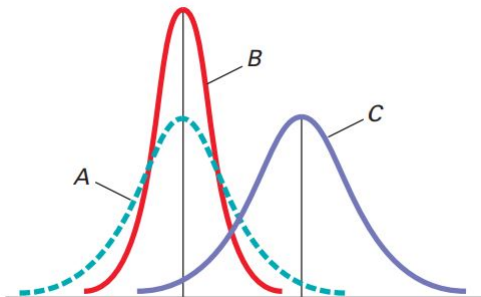
Normal Distribution Shape



Increase μ shifts the distribution right.

Decrease μ shifts the distribution left.

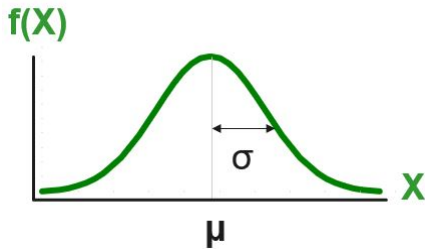
Changing σ increases or decreases the spread.



A and B have the same mean but different standard deviations.

B and C have different means and different standard deviations.

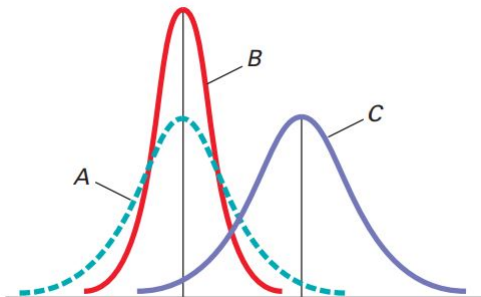
Normal Distribution Shape



Increase μ shifts the distribution right.

Decrease μ shifts the distribution left.

Changing σ increases or decreases the spread.



A and B have the same mean but different standard deviations.

B and C have different means and different standard deviations.

The normal probability density function of a normal distributed variable X with mean μ and variance σ^2 :

$$f(X) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(X-\mu)^2}{2\sigma^2}} = \frac{1}{\sqrt{2\pi}\sigma} e^{-0.5\left(\frac{X-\mu}{\sigma}\right)^2}$$

where

- e = the mathematical constant approximated by 2.71828
- π = the mathematical constant approximated by 3.14159
- μ = the population mean
- σ = the population standard deviation
- X = any value of the continuous variable

The normal probability density function of a normal distributed variable X with mean μ and variance σ^2 :

$$f(X) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(X-\mu)^2}{2\sigma^2}} = \frac{1}{\sqrt{2\pi}\sigma} e^{-0.5\left(\frac{X-\mu}{\sigma}\right)^2}$$

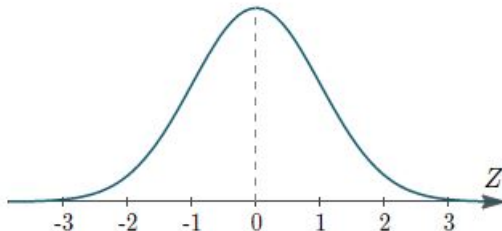
where

- e = the mathematical constant approximated by 2.71828
- π = the mathematical constant approximated by 3.14159
- μ = the population mean
- σ = the population standard deviation
- X = any value of the continuous variable

Standardized Normal Distribution

A normal distribution with $\mu = 0$ and $\sigma = 1$ is called **standardized normal distribution**, denoted by Z :

$$f(Z) = \frac{1}{\sqrt{2\pi}} e^{-0.5Z^2}$$



Suppose that X has normal distribution with μ, σ^2 . Set

$$Z = \frac{X - \mu}{\sigma}.$$

Then Z is the standardized normal random variable.

Suppose that X has normal distribution with μ, σ^2 . Set

$$Z = \frac{X - \mu}{\sigma}.$$

Then Z is the standardized normal random variable.

Example

If X is distributed normally with mean of 10 and standard deviation of 2, the Z value for $X = 11$ is:

Suppose that X has normal distribution with μ, σ^2 . Set

$$Z = \frac{X - \mu}{\sigma}.$$

Then Z is the standardized normal random variable.

Example

If X is distributed normally with mean of 10 and standard deviation of 2, the Z value for $X = 11$ is:

Suppose that X has normal distribution with μ, σ^2 . Set

$$Z = \frac{X - \mu}{\sigma}.$$

Then Z is the standardized normal random variable.

Example

If X is distributed normally with mean of 10 and standard deviation of 2, the Z value for $X = 11$ is:

$$Z = \frac{X - \mu}{\sigma} = \frac{11 - 10}{2} = 0.5$$

Suppose that X has normal distribution with μ, σ^2 . Set

$$Z = \frac{X - \mu}{\sigma}.$$

Then Z is the standardized normal random variable.

Example

If X is distributed normally with mean of 10 and standard deviation of 2, the Z value for $X = 11$ is:

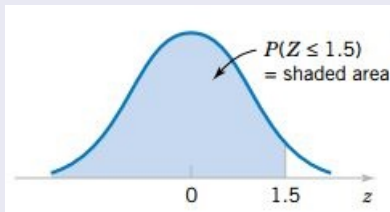
$$Z = \frac{X - \mu}{\sigma} = \frac{11 - 10}{2} = 0.5$$

Definition

The **cumulative standardized distribution** of Z is

$$\Phi(a) = P(Z \leq a) = P(Z < a) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-0.5z^2} dz$$

given in Appendix Table III (page 708).



Example

$$\Phi(1.5) = 0.9332, \quad \Phi(Z \leq 0) = 0.5, \quad \Phi(Z < -0.5) = 0.3085$$

Example

Compute $P(0.5 < Z \leq 1.2)$.

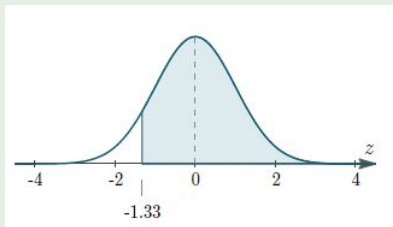
Example

Compute $P(0.5 < Z \leq 1.2)$.

$$P(0.5 < Z \leq 1.2) = \Phi(1.2) - \Phi(0.5) = 0.885 - 0.691 = 0.194$$

Example

Compute $P(Z > -1.33)$.



$$P(Z > -1.33) = 1 - P(Z \leq -1.33) = 1 - \Phi(-1.33) = 1 - 0.092 = 0.908$$

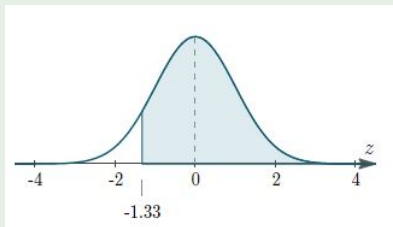
Example

Compute $P(0.5 < Z \leq 1.2)$.

$$P(0.5 < Z \leq 1.2) = \Phi(1.2) - \Phi(0.5) = 0.885 - 0.691 = 0.194$$

Example

Compute $P(Z > -1.33)$.



$$P(Z > -1.33) = 1 - P(Z \leq -1.33) = 1 - \Phi(-1.33) = 1 - 0.092 = 0.908$$

Suppose X has a normal distribution with mean μ and standard deviation σ .

Normalize $Z = \frac{X - \mu}{\sigma}$.

$$P(X < b) = P\left(Z < \frac{b - \mu}{\sigma}\right)$$

$$P(X > a) = P\left(Z > \frac{a - \mu}{\sigma}\right)$$

$$P(a < X < b) = P\left(\frac{a - \mu}{\sigma} < Z < \frac{b - \mu}{\sigma}\right)$$

Example

Suppose X is normal with mean 12 and standard deviation 2. Find $P(X > 13.2)$.

Example

Suppose X is normal with mean 12 and standard deviation 2. Find $P(X > 13.2)$.

We have

$$\begin{aligned} P(X > 13.2) &= P\left(Z > \frac{13.2 - 12}{2}\right) = P(Z > 0.6) \\ &= 1 - P(Z < 0.6) = 1 - \Phi(0.6) = 1 - 0.726 = 0.274 \end{aligned}$$

Example

Suppose X is normal with mean 12 and standard deviation 2. Find $P(X > 13.2)$.

We have

$$\begin{aligned} P(X > 13.2) &= P\left(Z > \frac{13.2 - 12}{2}\right) = P(Z > 0.6) \\ &= 1 - P(Z < 0.6) = 1 - \Phi(0.6) = 1 - 0.726 = 0.274 \end{aligned}$$

Example

A machine pours beer into 16 oz. bottles. Experience has shown that the number of ounces poured is normally distributed with a variance of 2.69 ounces. Find the probabilities that the amount of beer the machine will pour into the next bottle will be between 15.5 and 16 ounces.

Example

A machine pours beer into 16 oz. bottles. Experience has shown that the number of ounces poured is normally distributed with a variance of 2.69 ounces. Find the probabilities that the amount of beer the machine will pour into the next bottle will be between 15.5 and 16 ounces.

Answer: Let X be the amount of beer the machine will pour into the next bottle.

Example

A machine pours beer into 16 oz. bottles. Experience has shown that the number of ounces poured is normally distributed with a variance of 2.69 ounces. Find the probabilities that the amount of beer the machine will pour into the next bottle will be between 15.5 and 16 ounces.

Answer: Let X be the amount of beer the machine will pour into the next bottle. Then X is normal with mean $\mu = 16$ and standard deviation $\sigma = \sqrt{2.69} = 1.64$. We want to find the probability $P(15.5 < X < 16)$.

Example

A machine pours beer into 16 oz. bottles. Experience has shown that the number of ounces poured is normally distributed with a variance of 2.69 ounces. Find the probabilities that the amount of beer the machine will pour into the next bottle will be between 15.5 and 16 ounces.

Answer: Let X be the amount of beer the machine will pour into the next bottle. Then X is normal with mean $\mu = 16$ and standard deviation $\sigma = \sqrt{2.69} = 1.64$. We want to find the probability $P(15.5 < X < 16)$.

$$\begin{aligned} P(15.5 < X < 16) &= P\left(\frac{15.5 - 16}{1.64} < Z < \frac{16 - 16}{1.64}\right) \\ &= P(-0.305 < Z < 0) = \Phi(0) - \Phi(-0.305) = 0.5 - 0.38 = 0.12 \end{aligned}$$

Example

A machine pours beer into 16 oz. bottles. Experience has shown that the number of ounces poured is normally distributed with a variance of 2.69 ounces. Find the probabilities that the amount of beer the machine will pour into the next bottle will be between 15.5 and 16 ounces.

Answer: Let X be the amount of beer the machine will pour into the next bottle. Then X is normal with mean $\mu = 16$ and standard deviation $\sigma = \sqrt{2.69} = 1.64$. We want to find the probability $P(15.5 < X < 16)$.

$$\begin{aligned} P(15.5 < X < 16) &= P\left(\frac{15.5 - 16}{1.64} < Z < \frac{16 - 16}{1.64}\right) \\ &= P(-0.305 < Z < 0) = \Phi(0) - \Phi(-0.305) = 0.5 - 0.38 = 0.12 \end{aligned}$$

Definition

The random variable X that equals the distance between successive events of a Poisson process with mean number of events $\lambda > 0$ per unit interval is an **exponential** random variable with parameter λ .

Definition

The random variable X that equals the distance between successive events of a Poisson process with mean number of events $\lambda > 0$ per unit interval is an **exponential** random variable with parameter λ . The probability density function of X is

$$f(x) = \lambda e^{-\lambda x}, \quad 0 \leq x < \infty.$$

Definition

The random variable X that equals the distance between successive events of a Poisson process with mean number of events $\lambda > 0$ per unit interval is an **exponential** random variable with parameter λ . The probability density function of X is

$$f(x) = \lambda e^{-\lambda x}, \quad 0 \leq x < \infty.$$

Definition

The random variable X that equals the distance between successive events of a Poisson process with mean number of events $\lambda > 0$ per unit interval is an **exponential** random variable with parameter λ . The probability density function of X is

$$f(x) = \lambda e^{-\lambda x}, \quad 0 \leq x < \infty.$$

Theorem

If the random variable X has an exponential distribution with parameter λ . Then

$$\mu = E(X) = \frac{1}{\lambda}, \quad \sigma = \frac{1}{\lambda}.$$

Definition

The random variable X that equals the distance between successive events of a Poisson process with mean number of events $\lambda > 0$ per unit interval is an **exponential** random variable with parameter λ . The probability density function of X is

$$f(x) = \lambda e^{-\lambda x}, \quad 0 \leq x < \infty.$$

Theorem

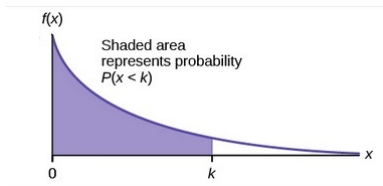
If the random variable X has an exponential distribution with parameter λ . Then

$$\mu = E(X) = \frac{1}{\lambda}, \quad \sigma = \frac{1}{\lambda}.$$

Cumulative Exponential Distribution

Suppose X is exponential with parameter λ . For $k > 0$

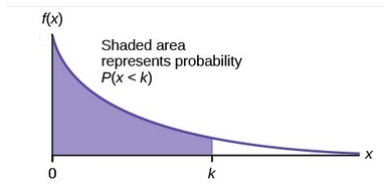
$$P(X \leq k) = \int_0^k \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_0^k = 1 - e^{-\lambda k}$$



Cumulative Exponential Distribution

Suppose X is exponential with parameter λ . For $k > 0$

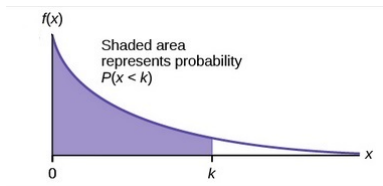
$$P(X \leq k) = \int_0^k \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_0^k = 1 - e^{-\lambda k}$$



Cumulative Exponential Distribution

Suppose X is exponential with parameter λ . For $k > 0$

$$P(X \leq k) = \int_0^k \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_0^k = 1 - e^{-\lambda k}$$



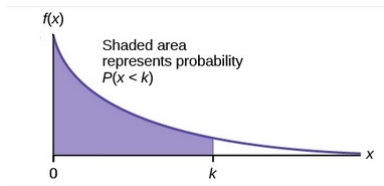
$$P(X > k) = \int_k^{\infty} \lambda e^{-\lambda x} dx = e^{-\lambda k}$$

$$P(k_1 < X < k_2) = \int_{k_1}^{k_2} \lambda e^{-\lambda x} dx = e^{-\lambda k_1} - e^{-\lambda k_2}$$

Cumulative Exponential Distribution

Suppose X is exponential with parameter λ . For $k > 0$

$$P(X \leq k) = \int_0^k \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_0^k = 1 - e^{-\lambda k}$$



$$P(X > k) = \int_k^{\infty} \lambda e^{-\lambda x} dx = e^{-\lambda k}$$

$$P(k_1 < X < k_2) = \int_{k_1}^{k_2} \lambda e^{-\lambda x} dx = e^{-\lambda k_1} - e^{-\lambda k_2}$$

Example

The time between customer arrivals at a furniture store has an approximate exponential distribution with mean of 9 minutes. If a customer just arrived, find the probability that the next customer will not arrive for at least 15 minutes.

Example

The time between customer arrivals at a furniture store has an approximate exponential distribution with mean of 9 minutes. If a customer just arrived, find the probability that the next customer will not arrive for at least 15 minutes.

Answer: Let X the time between customer arrivals. Then X has an exponential distribution with parameter $\lambda = \frac{1}{9}$. We want to compute $P(X > 15)$. We have

$$P(X > 15) = \int_{15}^{\infty} \frac{1}{9} e^{-x/9} dx = -e^{-\frac{1}{9}x} \Big|_{15}^{\infty} = e^{-15/9} = 0.189$$

Example

The time between customer arrivals at a furniture store has an approximate exponential distribution with mean of 9 minutes. If a customer just arrived, find the probability that the next customer will not arrive for at least 15 minutes.

Answer: Let X the time between customer arrivals. Then X has an exponential distribution with parameter $\lambda = \frac{1}{9}$. We want to compute $P(X > 15)$. We have

$$P(X > 15) = \int_{15}^{\infty} \frac{1}{9} e^{-x/9} dx = -e^{-\frac{1}{9}x} \Big|_{15}^{\infty} = e^{-15/9} = 0.189$$

The amount of time between successive TV watching by first graders follows an exponential distribution with a mean of 5 hours. Find the probability that a given first grader spends less than 4 hours between successive TV watching.

The amount of time between successive TV watching by first graders follows an exponential distribution with a mean of 5 hours. Find the probability that a given first grader spends less than 4 hours between successive TV watching.

Answer: Let X the time between successive TV watching. Then X is exponential with $\lambda = 0.2$. We want to compute $P(X < 4)$.

The amount of time between successive TV watching by first graders follows an exponential distribution with a mean of 5 hours. Find the probability that a given first grader spends less than 4 hours between successive TV watching.

Answer: Let X the time between successive TV watching. Then X is exponential with $\lambda = 0.2$. We want to compute $P(X < 4)$.

The amount of time between successive TV watching by first graders follows an exponential distribution with a mean of 5 hours. Find the probability that a given first grader spends less than 4 hours between successive TV watching.

Answer: Let X the time between successive TV watching. Then X is exponential with $\lambda = 0.2$. We want to compute $P(X < 4)$.

$$P(X < 4) = \int_0^4 0.2e^{-0.2x} dx = 1 - e^{-4*0.2} = 0.55$$

The amount of time between successive TV watching by first graders follows an exponential distribution with a mean of 5 hours. Find the probability that a given first grader spends less than 4 hours between successive TV watching.

Answer: Let X the time between successive TV watching. Then X is exponential with $\lambda = 0.2$. We want to compute $P(X < 4)$.

$$P(X < 4) = \int_0^4 0.2e^{-0.2x} dx = 1 - e^{-4*0.2} = 0.55$$