

Ch. 5 - The Normal Distribution



1 Definition and Properties

Normal Distribution:

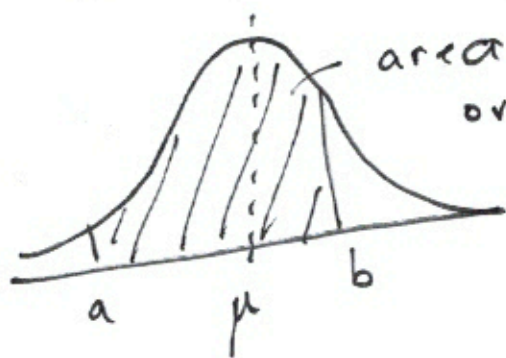
- If a continuous random variable X is said to follow a normal distribution with mean μ and standard deviation σ , we write

$$X \sim N(\mu, \sigma^2)$$

The normal distribution has a continuous density, which is summarized by two **parameters** (numerical characteristics of a model). The location of the distribution is determined by the value of the mean (μ) and spread is determined by the standard deviation (σ). The density function for a normal random variable is

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad -\infty < x < \infty$$

(Note: we rarely use this in calculations for STAT251)



area under the normal curve
over $[a, b]$ gives percentage
of observations that
have values between
 a and b .

2 Finding Probabilities

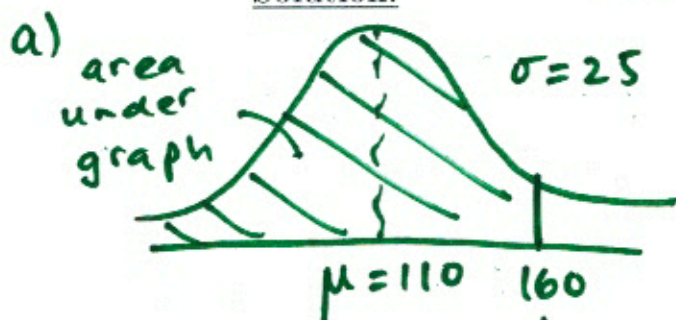
Areas under the normal density can be found using: computer software and statistical tables. Statistical tables only give areas under the $N(0,1)$ curve. (Statistical table posted on course website)

Example 2. Scores on a standard IQ test follow approximately the Normal model with mean $\mu = 110$ and standard deviation $\sigma = 25$.

- (a) What percentage of people have IQ scores below 160?
- (b) What percentage of people have scores between 90 and 120?
- (c) How high is the IQ such that only 0.15% of people fall above?

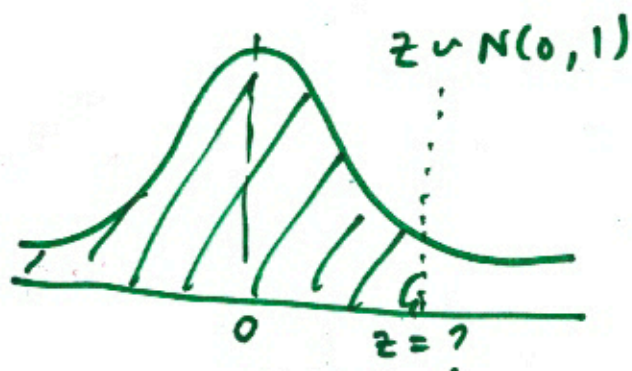
Solution:

Let X rep. IQ score.



$$P(X < 160) = ?$$

$$z = \frac{x - \mu}{\sigma} = \frac{160 - 110}{25} = 2$$



$$P(X < 160) =$$

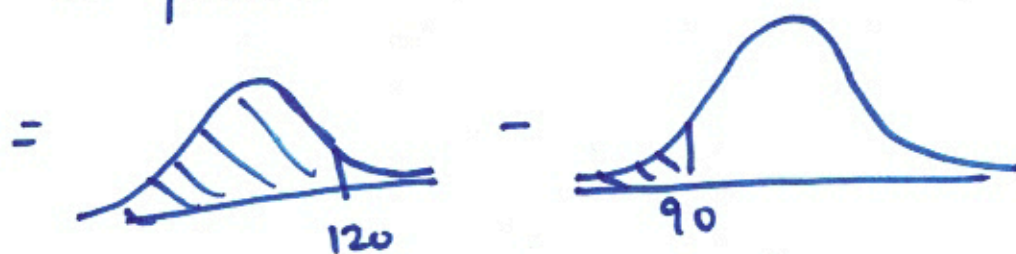
$$P\left(\frac{X - \mu}{\sigma} < \frac{160 - 110}{25}\right)$$

$$= P(z < 2)$$

$$= 0.9772 \quad \text{look at (table)}$$

b)

$$P(90 < X < 120) = ?$$



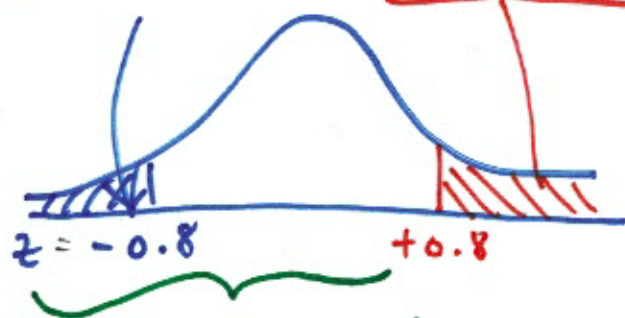
$$\begin{aligned} P(90 < X < 120) &= P(X < 120) - P(X < 90) \\ &= P\left(z < \frac{120 - 110}{25}\right) - P\left(z < \frac{90 - 110}{25}\right) \\ &= P(z < 0.4) - P(z < -0.8) \end{aligned}$$

$$P(z < 0.4) = \underline{0.6554}$$

(look in z-table)

$$P(z < -0.8) = \underbrace{1 - P(z < 0.8)}_{= 0.2119} = 1 - 0.7881$$

$$= \underline{\underline{0.2119}}$$

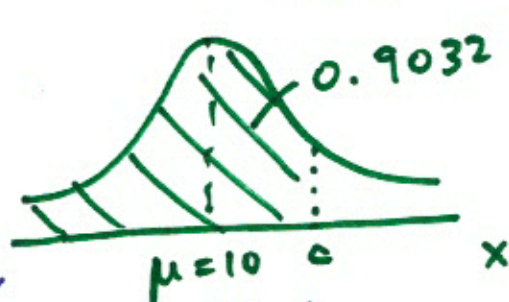


$$P(z < 0.8)$$

$$P(90 < X < 120) = 0.6554 - 0.2119 = 0.4435$$

Example 3. Given $X \sim N(10, 25)$, find c such that $P(X < c) = 0.9032$

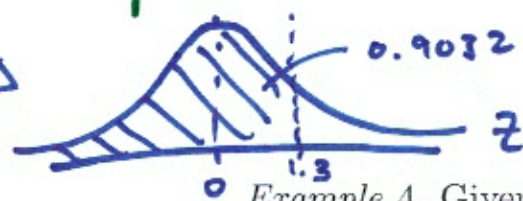
Solution:



$$P(Z < \frac{c - 10}{5}) = P(Z < 1.3)$$

$$\Rightarrow \frac{c - 10}{5} = 1.3$$

$$c = (1.3 \times 5) + 10 = 16.5$$



$$\sigma^2 = 4 \rightarrow \sigma = 2$$

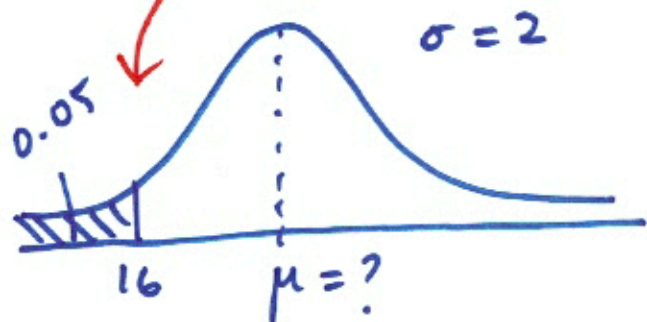
Example 4. Given $X \sim N(\mu, 4)$ and $P(X < 16) = 0.05$. Find μ .

Solution:

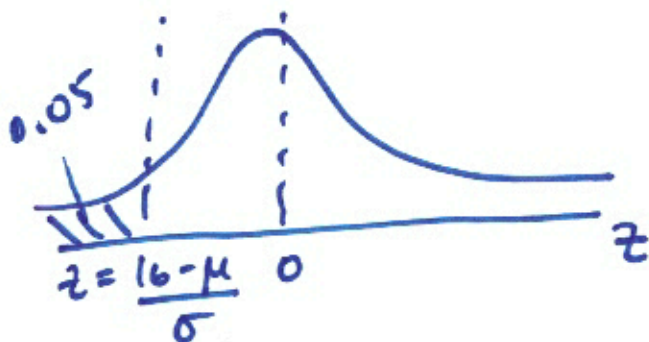
$$P(X < 16) = P\left(\frac{X - \mu}{\sigma} < \frac{16 - \mu}{2}\right) = P\left(Z < \frac{16 - \mu}{2}\right) = 0.05$$

$$\Rightarrow \frac{16 - \mu}{2} = -1.645$$

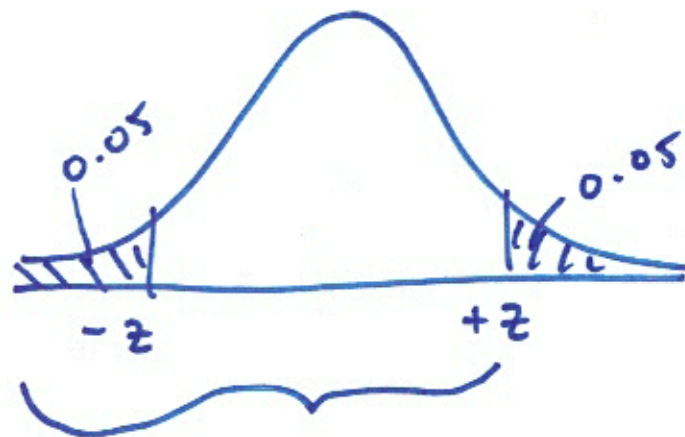
$$\mu = (1.645 \times 2) + 16 = 19.29$$



look up $1 - 0.05 = 0.95$



5



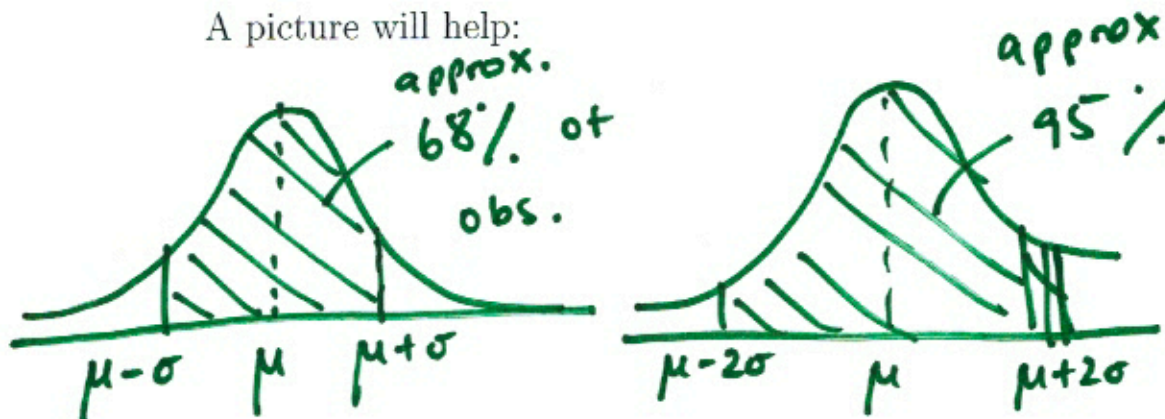
68-95-99.7 Rule:

A useful rule for the normal distribution is the **68-95-99.7** or **empirical rule**:

If $X \sim N(\mu, \sigma)$ then approximately,

- 68% of the observations will fall within 1 standard deviation of the mean
- 95% of the observations will fall within 2 standard deviations of the mean
- 99.7% of the observations will fall within 3 standard deviations of the mean

A picture will help:

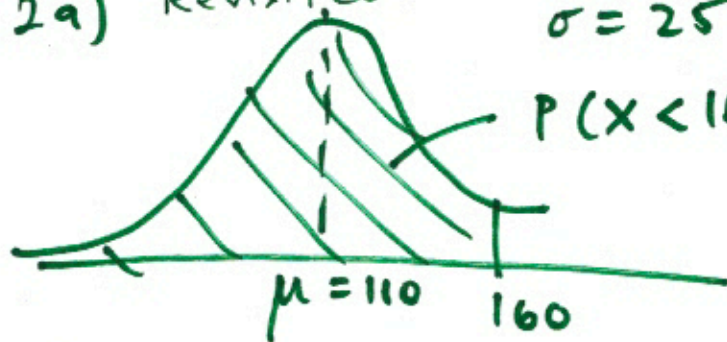


Let's revisit example 2(a) and verify we get approximately the same answer using this method...

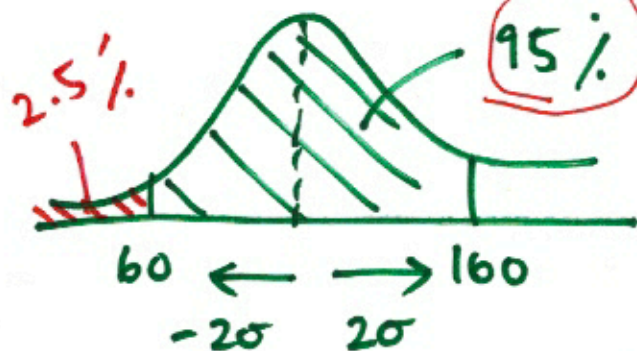
2a) Revisited

$$\sigma = 25$$

$$P(X < 160) = ?$$



$$z = \frac{160 - 110}{25} = 2$$



approx by the 68-95-99.7% rule

$$\text{So } 95\% + 2.5\% = 97.5\%$$

approx. 97.5% of values less than 160

3 Independent Normal Samples

The sums (and differences) of independent normal variables are also normal

- If $X \sim N(10, 25)$ and $Y \sim N(5, 16)$ are independent, then $X + Y$ is also normal.

$$\begin{aligned} E(X + Y) &= E(X) + E(Y) \\ &= 10 + 5 = 15 \end{aligned}$$

$$\begin{aligned} \text{Var}(X + Y) &= \text{Var}(X) + \text{Var}(Y) \quad X, Y \text{ independent} \\ &= 25 + 16 = 41 \end{aligned}$$

$$\Rightarrow X + Y \sim N(15, 41)$$

What about $X - Y$?

$$\Rightarrow X - Y \sim N(5, 41)$$

$$E(X - Y) = E(X) - E(Y) = 5$$

$$\text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y)$$

- If X_1, X_2, \dots, X_n are n independent observations of normal variable $X \sim N(\mu, \sigma^2)$, then

$$X_1 + X_2 + \dots + X_n \sim N(n\mu, n\sigma^2)$$

$$\begin{aligned} E(X_1 + \dots + X_n) &= E(X_1) + E(X_2) + \dots + E(X_n) \\ &= \mu + \mu + \dots + \mu \end{aligned}$$

$$\begin{aligned} &= n\mu \\ \text{Var}(X_1 + \dots + X_n) &= n \text{Var}(X) \end{aligned}$$

Example 5. A certain company manufactures stereo systems. The times required to pack the stereos can be described by a Normal model with mean of 9 minutes and standard deviation of 1.5 minutes. The times for the boxing stage can be modelled as Normal with a mean of 6 minutes and standard deviation of 1 minute. Assume that the two packing times are independent.

- What is the probability that packing two consecutive systems takes over 20 minutes?
- What percentage of the stereo systems take longer to pack than to box? Assume that the packing and boxing times are independent.

Solution:

Let P_1 = time pack 1st system.

P_2 = time pack 2nd system.

$T = P_1 + P_2$ total time to pack two systems

→ we are told both r.v \sim Normal

→ times independent.

$$P_i \sim N(\underline{9}, 1.5^2) \quad i = 1, 2$$

$$T \sim N(\underset{\uparrow}{?}, \underset{\uparrow}{?})$$

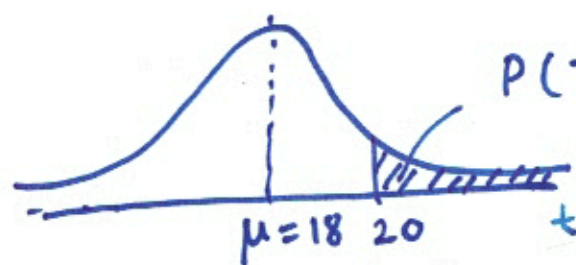
$$\begin{aligned} E(T) &= E(P_1 + P_2) = E(P_1) + E(P_2) \\ &= 9 + 9 = \underline{18} \text{ minutes.} \end{aligned}$$

$$\begin{aligned} \text{Var}(T) &= \text{Var}(P_1 + P_2) = \text{Var}(P_1) + \text{Var}(P_2) \\ &\quad \text{times independent} \end{aligned}$$

$$= 1.5^2 + 1.5^2 = 4.50$$

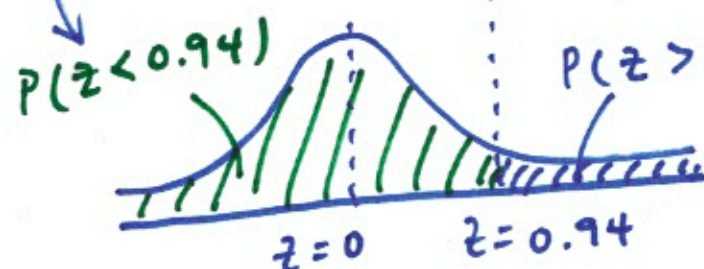
$$SD(T) = \sqrt{4.50} = 2.12 \text{ min.}$$

$$T \sim N(18, 2.12^2)$$



$$P(T > 20) = ?$$

$$z = \frac{20 - 18}{2.12} = 0.94$$



$$P(T > 20) = P\left(\frac{T - \mu}{\sigma} > \frac{20 - 18}{2.12}\right)$$

$$= P(Z > 0.94)$$

$$= 1 - P(Z < 0.94)$$

$$= 1 - 0.8264$$

$$= 0.1736$$

There is a 17.36% that it will take over 20 min to pack two systems.

b) we want to estimate the percentage of stereo systems that take longer to pack than to box.

P = time packing a system

B = time boxing a system.

D = P - B difference in times to pack and box.

$$P(P > B) = P(P - B > 0) = P(D > 0)$$

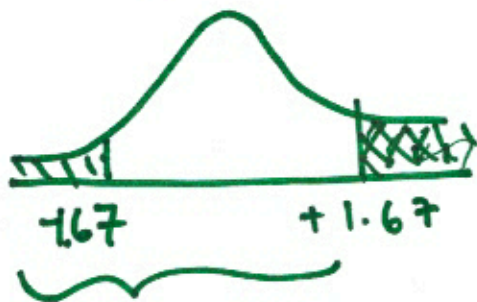
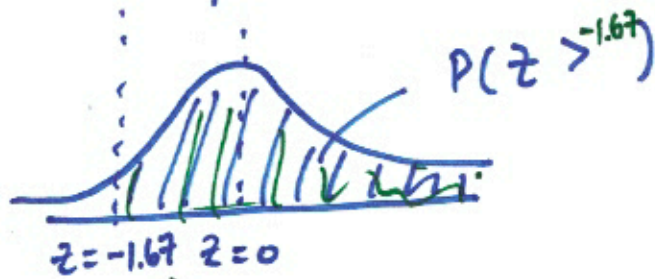
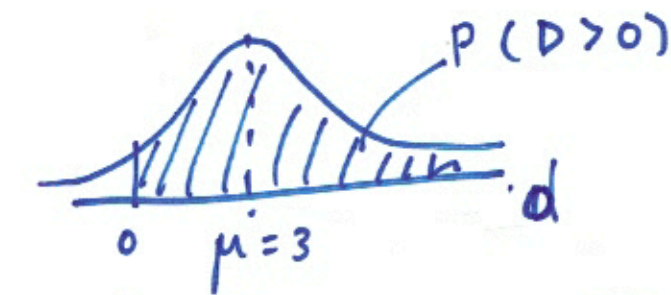
$$D \sim N(?, ?)$$

$$E(D) = E(P - B) = E(P) - E(B) = 9 - 6 = 3 \text{ min.}$$

$$\begin{aligned} \text{Var}(D) &= \text{Var}(P - B) = \text{Var}(P) + \text{Var}(B) \\ &= 1.5^2 + 1^2 = 3.25 \end{aligned}$$

$$SD(D) = \sqrt{3.25} = 1.8 \text{ minutes.}$$

$$D \sim N(3, 1.80^2)$$



$$\begin{aligned} P(D > 0) &= P\left(z > \frac{0 - 3}{1.8}\right) \\ &= P(z > -1.67) \\ &= 1 - P(z < -1.67) \\ &= 1 - (1 - P(z < 1.67)) \\ &= P(z < 1.67) \\ &= 0.9525 \end{aligned}$$

About 95.25% of all stereo systems require more time for packing than boxing.

The Sample Mean

Let X_1, X_2, \dots, X_n be a random sample of n independent observations. If $X_i \sim N(\mu, \sigma^2)$, then the sample mean \bar{X} also follows a normal distribution and

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$\bar{X} = \frac{X_1 + \dots + X_n}{n}$$

Example 6. A random sample of size 15 is taken from a normal distribution with mean 60 and standard deviation 4. Find the probability that the mean of the sample is less than 58.

Solution:

$$X \sim N(60, 4^2)$$

$$\bar{X} = \frac{X_1 + \dots + X_n}{n}$$

$$\bar{X} \sim N\left(60, \frac{4^2}{15}\right)$$

$$\Rightarrow P(\bar{X} < 58) = P\left(z < \frac{58 - 60}{\sqrt{4^2/15}}\right)$$

$$= P(z < -1.936)$$

$$= 1 - P(z < 1.936)$$

$$= 1 - 0.9732$$

$$= 0.0268$$



look in table

$$\begin{aligned}
 E(\bar{X}) &= E\left(\frac{X_1 + \dots + X_n}{n}\right) = \frac{1}{n} [E(X_1) + \dots + E(X_n)] \\
 &= \frac{1}{n} [\mu + \mu + \dots + \mu] \\
 &= \frac{1}{n} [n\mu] = \mu
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(\bar{X}) &= \text{Var}\left(\frac{X_1 + \dots + X_n}{n}\right) \\
 &= \frac{1}{n^2} [\text{Var}(X_1) + \dots + \text{Var}(X_n)] \quad \text{indep.} \\
 &= \frac{1}{n^2} (\sigma^2 + \dots + \sigma^2) = \frac{1}{n^2} n\sigma^2 \\
 &= \frac{\sigma^2}{n}
 \end{aligned}$$

$$\therefore \bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$