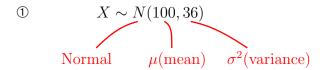
## Ch5 Normal Distribution



A continuous random variable X is said to follow a Normal distribution with mean  $\mu$  and standard deviation  $\sigma$  if its probability density function is

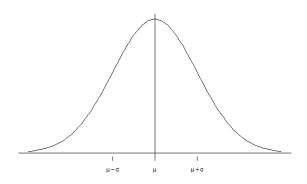
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad -\infty < x < \infty$$

We rarely use this in calculation for STAT 251.

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

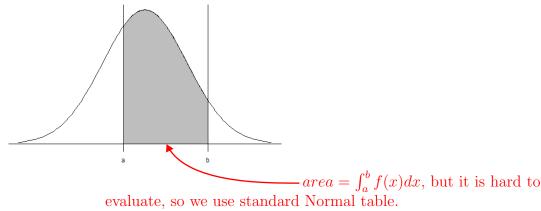
The Normal distribution with  $\mu = 0$  and  $\sigma = 1$  is called a <u>standard Normal distribution</u>. We denote the standard Normal variable by  $Z \sim N(0, 1)$ .

3



Normal distribution is bell-shaped and symmetrical about the line  $x=\mu$ . The mean, median and mode (where pdf attains max value) coincide.

4 If  $Z \sim N(0,1) \leftarrow$  standard Normal table then the probability that Z lies between a and b is given by the area



Note: A copy of the standard Normal table is posted on Connect.

Eg

$$P(Z < -2.1) = ? 0.0179$$
 from table

 $P(Z>-2.12)=1-P(Z\leq -2.12)=1-0.0170=0.983 \rightarrow \text{Note: Instead of F(-2.12)},$  it is customary to use  $\Phi(-2.12)$  for Normal distributions

$$P(1.0 \le Z \le 2.09) = \Phi(2.09) - \Phi(1.0) = 0.9817 - 0.8413 = 0.1404$$

Now that you know how to calculate probability given the z-scores, make sure you know how to find z-scores given the probability (as in the next question)?

Find a such that P(Z < a) = 0.7

Ans: 
$$[P(Z < 0.525) = 0.7 \text{ from table}] \Rightarrow P(Z < a) = P(Z < 0.525) \Rightarrow a = 0.525$$

Most of the time, the question tells us  $X \sim N(100, 36)$  and the aim is say to find  $P(X \le 90)$ .

How?

We first express in terms of Z and then use the standard Normal table to get the probability.

$$Z = \frac{X - \mu}{\sigma} \leftarrow Important!$$

Eg.  $X \sim N(10, 25)$ . Find P(X < 13).

$$P(X < 13) = P(Z < \frac{13-10}{5}) = P(Z < 0.6) = 0.7257$$
 (from table)  
 $\sigma = 5$  since  $\sigma^2 = 25$ 

What about  $P(X > 13) = 1 - P(X \le 13) = 1 - 0.7257 = 0.2743$ 

What about

$$P(11 \le X \le 13) = P(\frac{11 - 10}{5} \le Z \le \frac{13 - 10}{5})$$
$$= \Phi(0.6) - \Phi(0.2)$$
$$= 0.7257 - 0.5793$$
$$= 0.1463$$

What about 
$$P(X < -5) = P(Z < \frac{-5-10}{5}) = P(Z < -3) = 0.0013$$

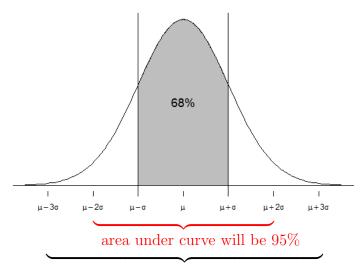
⑤ All Normal distribution curves satisfy the 68-95-99.7% rule:

68% of the observations will fall within 1 standard deviation of the mean (that is,  $P(\mu - \sigma < X < \mu + \sigma) \approx 0.68)$ 

approximate
95% of the observations will fall within 2 standard deviations

99.7% of the observations will fall within 3 standard deviations

A picture will help:



area under curve from  $(\mu - 3\sigma, \mu + 3\sigma)$  is approx. 99.7%

## Some examples:

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

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

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

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

$$= 16.5$$

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

$$P(X < 16) = 0.05$$

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

$$= P(Z < -1.645)$$

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

$$16 - \mu = -1.645 \times 2$$

$$\mu = 16 + 3.29$$

$$= 19.29$$

© The sums (and differences) of independent Normal variables are also Normal.

Important!

What does it mean?

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

$$E(X + Y) = E(X) + E(Y)$$

$$= 10 + 5 = 15$$

$$Var(X + Y) = Var(X) + Var(Y) \qquad [X, Yindependent]$$

$$= 25 + 16 = 41$$

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

Try: what is the distribution of X - Y?

X - Y is also Normal.

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

$$= 10 - 5 = 5$$

$$Var(X - Y) = Var(X) + Var(Y) \qquad [X, Yindependent]$$

$$= 25 + 16 = 41 \qquad \text{Note}$$

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

## Example

Given  $X \sim N(100, 16)$  and  $Y \sim N(90, 20)$ . X, Y independent. Find P(X < Y).

Solution: It is important to understand how to solve this simple problem. Hint: bring all the variables to left side.

$$P(X < Y) = P(X - Y < 0)$$
$$X - Y \sim N(10.16 + 20)$$

[ because sum/difference of independent Normal is also Normally distributed.] Must state for exams and assignments

$$\Rightarrow X - Y \sim N(10, 36)$$

$$P(X < Y) = P(X - Y < 0)$$

$$= P(Z < \frac{0 - 10}{6})$$

$$= P(Z < -1.67)$$

$$= 0.0475$$

Try it yourself:

What is P(X + 3Y > 400) using X and Y above?

Ans: 0.0162

Another example:  $X \sim N(2,9)$ , Y = 5X + 1. Find the distribution of Y.

Y is also Normal.

$$E(Y) = E(5X + 1) = 5E(X) + 1 = (5 \times 2) + 1 = 11$$
$$Var(Y) = Var(5X + 1) = 5^{2}Var(X) = 5^{2} \times 9 = 225$$
$$\Rightarrow Y \sim N(11, 225)$$

If  $X_1, X_2, X_3, \ldots, X_n$  are n independent observations of Normal variable  $X \sim N(\mu, \sigma^2),$  then  $X_1 + X_2 + X_3 + \cdots + X_n \sim N(n\mu, n\sigma^2)$ 

Example: Ex. 5.17 in coursenotes (Pg 100).

Let W be the weight of a concrete beam.  $W \sim N(31, 0.5^2)$ 

(a)

$$P(30 \le W \le 32) = P(\frac{30 - 31}{0.5} \le Z \le \frac{32 - 31}{0.5})$$

$$= P(-2 \le Z \le 2)$$

$$= \Phi(2) - \Phi(-2)$$

$$= 0.9772 - 0.0228$$

$$= 0.9544$$

(b) It is important for students to think of the sum of 25 randomly selected beams as

$$T = X_1 + X_2 + X_3 + \dots + X_{25}$$

A student who thinks of T as 25X will get the wrong answer!

$$T \sim N(25 \times 31, 25 \times 0.5^{2})$$

$$T \sim N(775, 6.25)$$

$$P(T > 795) = P(Z > \frac{795 - 775}{\sqrt{6.25}})$$

$$= P(Z > 8)$$

$$\approx 0$$

 $\otimes$  Let  $X_1, X_2, X_3, \ldots, X_n$  be random sample of n independent observations of X. If  $X \sim N(\mu, \sigma^2)$ , then the sample mean  $\bar{X}$  is also a Normal distribution and

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

Example:

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:

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

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

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

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

$$= 0.0268$$

	em: Central Limit	Theorem
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CLT is taught in Chapter 7 but introduced here as it is an appropriate spot:

Let  $X_1, X_2, X_3, \ldots, X_n$  be a random sample of size n taken from  $\underline{\underline{\text{any}}}$  distribution with mean  $\mu$  and variance  $\sigma^2$ . Then, for sufficiently  $\underline{\underline{\text{large}}}$  sample n, the sample mean  $\bar{X}$  is approximately Normal and

$$\bar{X} \stackrel{approx.}{\sim} N(\mu, \sigma^2/n)$$

This is known as central limit theorem.

Large sample will usually be of size at least 50 (your notes on ch7 says at least 20 will do).

