

Chapter

# 24

## The normal distribution

Syllabus reference: 5.9

- Contents:
- A The normal distribution
  - ~~B Probabilities using a calculator~~
  - C The standard normal distribution ( $Z$ -distribution)
  - D Quantiles or  $k$ -values

## 24A – The Normal Distribution

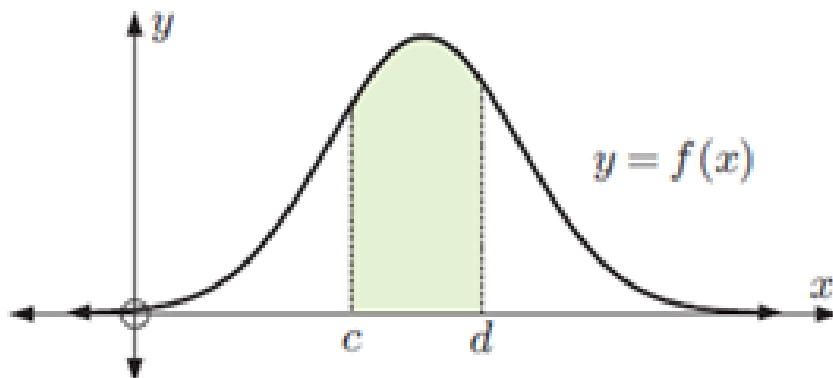
From Ch. 23: **discrete** random variables

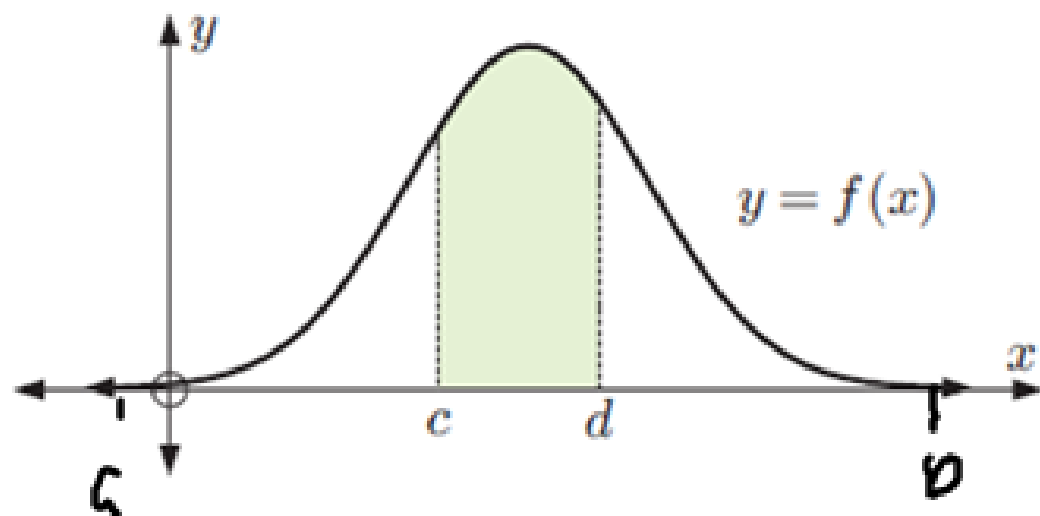
- examined binomial probability distributions where  $X$  could take non-negative integer values,  $x = 0, 1, 2, 3, \dots$

For a **continuous** random variable  $X$ ,  $x$  can be any real value.

Use a **probability density function** to define the probability distribution.

Probabilities are found by calculating areas under the probability density function curve.





For a continuous random variable  $X$ , the **probability density function** is a function  $f(x)$  such that  $f(x) \geq 0$  on its entire domain.

If the domain of the function is  $a \leq x \leq b$ , then  $\int_a^b f(x) dx = 1$ .

The probability that  $X$  lies in the interval  $c \leq X \leq d$  is  $P(c \leq X \leq d) = \int_c^d f(x) dx$ .

For a continuous variable  $X$ , the probability that  $X$  is *exactly* equal to a particular value is zero:

$$P(X = x) = 0 \text{ for all } x.$$

For example, the probability that an egg will weigh *exactly* 72.9 g is zero.

If you were to weigh an egg on scales that weigh to the nearest 0.1 g, a reading of 72.9 g means the weight lies somewhere between 72.85 g and 72.95 g. No matter how accurate your scales are, you can only ever know the weight of an egg within a range.

So, for a continuous variable we can only talk about the probability that an event lies in an **interval**.

A consequence of this is that  $P(c \leq X \leq d) = P(c < X \leq d) = P(c \leq X < d) = P(c < X < d)$ .

This would not be true if  $X$  was discrete.

# THE NORMAL PROBABILITY DENSITY FUNCTION

If  $X$  is normally distributed then its probability density function is

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

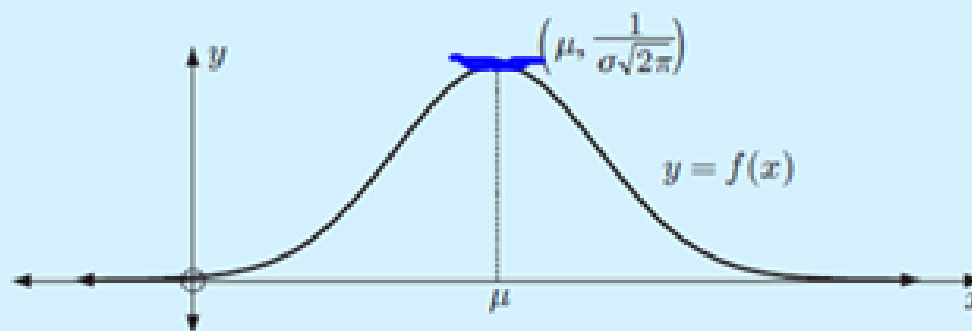
where  $\mu$  is the mean and  $\sigma^2$  is the variance of the distribution.

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

$$f(x) = a e^{-bx^2}$$

Chain Rule

- The curve  $y = f(x)$ , which is called a **normal curve**, is symmetrical about the vertical line  $x = \mu$ .
- As  $x \rightarrow \pm\infty$  the normal curve approaches its asymptote, the  $x$ -axis.
- $f(x) > 0$  for all  $x$ .
- Since the total probability must be 1,  $\int_{-\infty}^{\infty} f(x) dx = 1$ .



- More scores are distributed closer to the mean than further away. This results in the typical **bell shape**.

## THE GEOMETRIC SIGNIFICANCE OF $\mu$ AND $\sigma$

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

we obtain  $f'(x) = \frac{-1}{\sigma^2\sqrt{2\pi}} \left(\frac{x-\mu}{\sigma}\right) e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$

$\therefore f'(x) = 0$  only when  $x = \mu$ , and this corresponds to the point on the graph when  $f(x)$  is a maximum.

Differentiating again, we obtain  $f''(x) = \frac{-1}{\sigma^2\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \left[ \frac{1}{\sigma} - \frac{(x-\mu)^2}{\sigma^3} \right]$

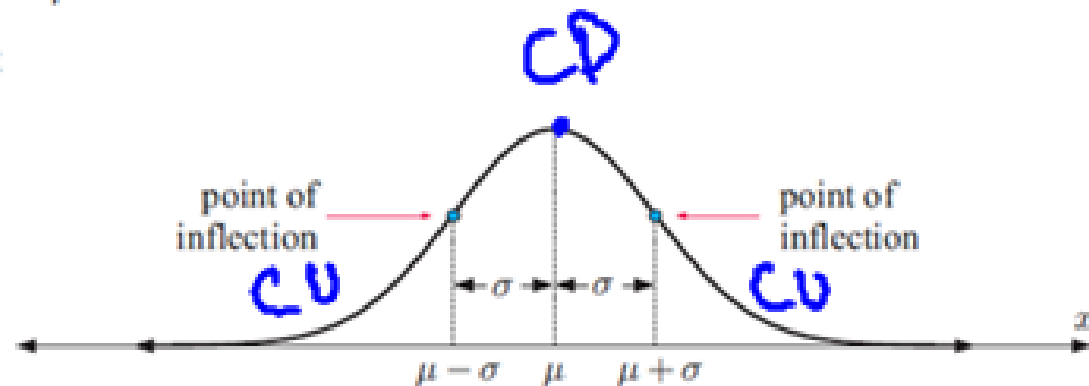
$\therefore f''(x) = 0$  when  $\frac{(x-\mu)^2}{\sigma^3} = \frac{1}{\sigma}$

$\therefore (x-\mu)^2 = \sigma^2$

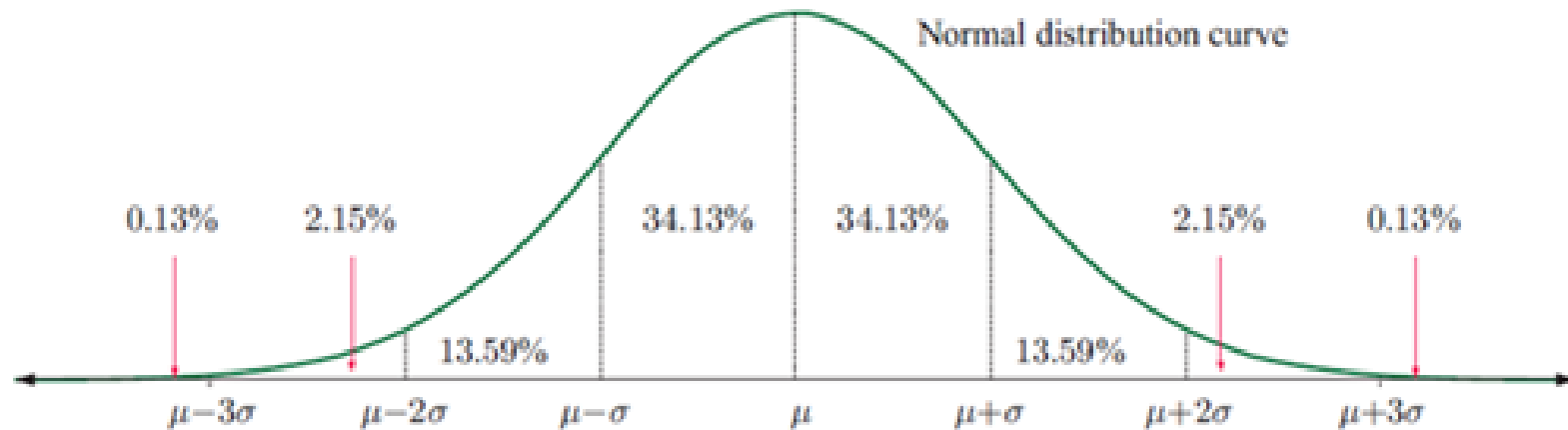
$\therefore x - \mu = \pm\sigma$

$\therefore x = \mu \pm \sigma$

So, the points of inflection are at  $x = \mu + \sigma$  and  $x = \mu - \sigma$ .



For a normal distribution with mean  $\mu$  and standard deviation  $\sigma$ , the percentage breakdown of where the random variable could lie is shown below.

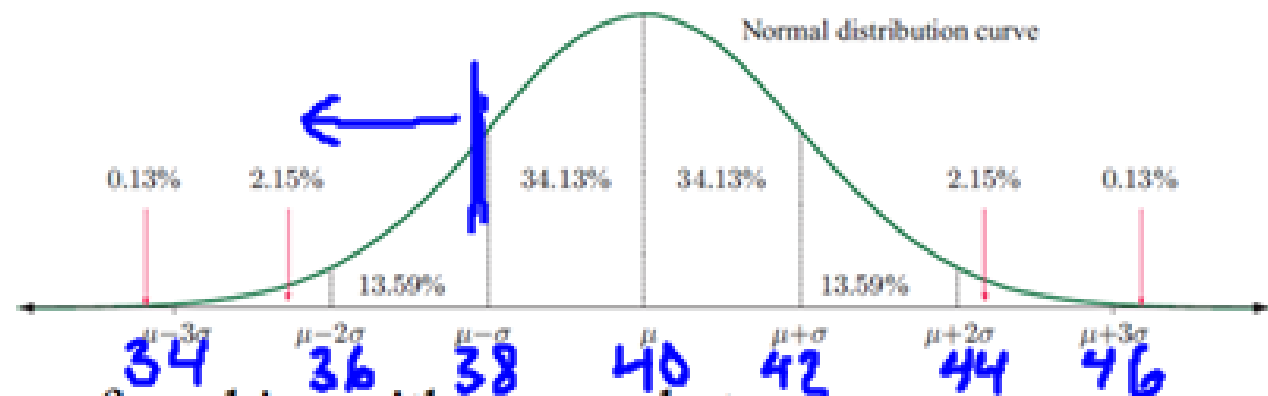


Notice that:

- $\approx 68.26\%$  of values lie between  $\mu - \sigma$  and  $\mu + \sigma$
- $\approx 95.44\%$  of values lie between  $\mu - 2\sigma$  and  $\mu + 2\sigma$
- $\approx 99.74\%$  of values lie between  $\mu - 3\sigma$  and  $\mu + 3\sigma$ .

Example:

In an Oreo factory, the masses of cookies are normally distributed with a mean of 40 g and a standard deviation of 2 g.



(a) Find the percentage of cookies with a mass between 36 g and 42 g.

$$13.59\% + 34.13\% + 34.13\% = 81.85\%$$

(b) Find the probability that the mass of a cookie is between 38 g and 46 g.

$$0.1359 + 0.1359 + 0.0215 = 0.2933$$

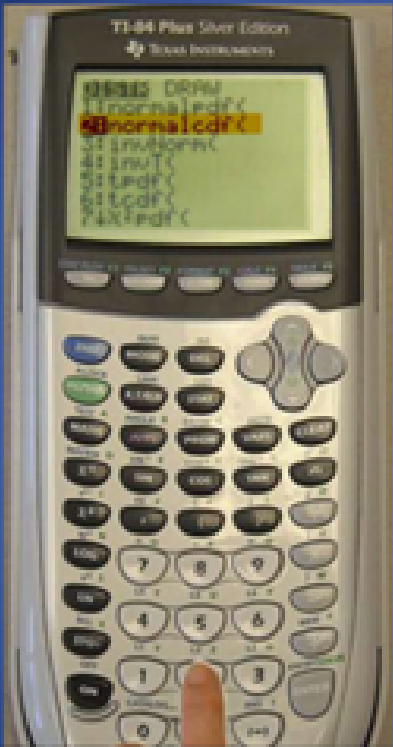
(c) Find the value of  $k$  such that approximately 16% of the cookies are below  $k$  g.

$$k = 38 \text{ g}$$



## 24 B – Probabilities Using a Calculator

Need to use the calculator to find probabilities that are not just regions of width of  $\sigma$  from the mean.



The image shows a TI-84 Plus Silver Edition calculator. The screen displays a list of probability distribution functions: normalcdf, normalpdf, invnorm, invnormC, invnormT, invnormC, invnormC, invnormC, and invnormC. The 'normalcdf' function is highlighted in yellow. A hand is visible at the bottom, pointing at the calculator's keypad.

IMPORTANT!!  
You must choose the normalcdf function, not the normalpdf. Do not ever use normalpdf!!

Example:

is modelled by a Normal dis. function  
mean

Suppose  $X \sim N(50, 3^2)$ .

variance: st. dev = 3

This means  $X$  is normally distributed with mean 50 and standard deviation 3.

Find:

(a)  $P(40 \leq X \leq 65)$

(b)  $P(X \geq 35)$

(c)  $P(X \leq 42)$



$$P(X \geq 35) = 1 - P(X \leq 35)$$

```
lower: -1E99  
upper: 35  
μ: 50  
σ: 3  
Paste
```

```
normalcdf(-1E99, 35, 50, 3)  
2.87105E-7  
1-Ans  
.99999997129
```

## 24 C – The Standard Normal Distribution (Z-Distribution)

Every normal  $X$ -distribution can be transformed into the standard normal distribution, also called the  $Z$ -distribution using the transformation

$$z = \frac{x - \mu}{\sigma}$$

No matter what the values of  $\mu$  and  $\sigma$  are of the  $X$ -distribution, the  $Z$ -distribution will always have mean 0 and standard deviation 1.

$Z \sim N(0, 1)$

modelled by a Norm. dist  
St. dev.  
mean

Read pg. 640 for the  $Z$ -transformation...

The **probability density function** for the  $z$ -distribution is:

$$f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} \quad -\infty < z < \infty$$

If  $x$  is an observation from a normal distribution with mean  $\mu$  and standard deviation  $\sigma$ , the  **$z$ -score** of  $x$  is the number of standard deviations  $x$  is from the mean.

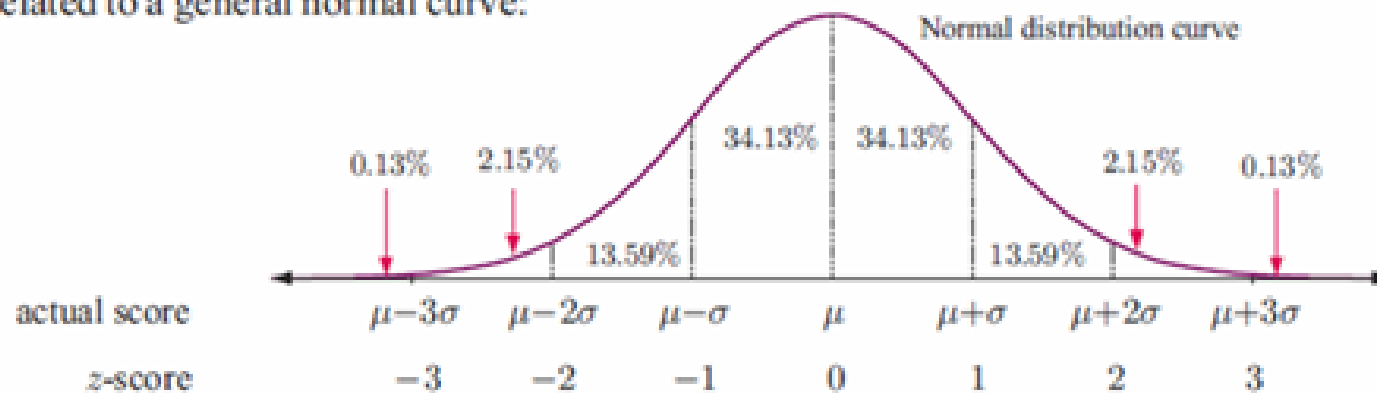
$$z = \frac{x - \mu}{\sigma}$$

For example, if  $z = 1.34$ , then  $x$  is 1.34 standard deviations to the right of the mean.

If  $z = -0.31$ , then  $x$  is 0.31 standard deviations to the left of the mean.

## Pg. 641

This diagram shows how the  $z$ -score is related to a general normal curve:

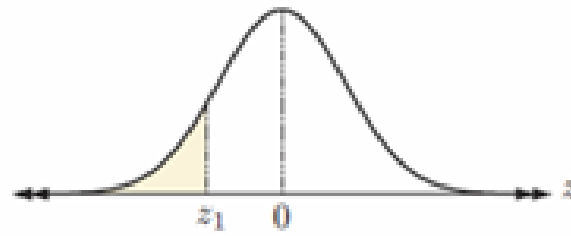
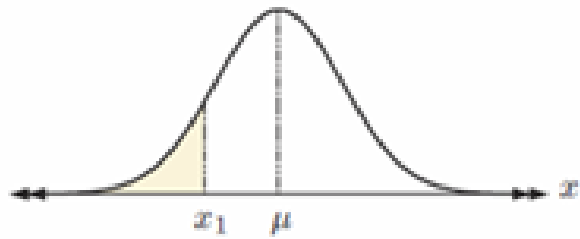


This means that for  $X \sim N(\mu, \sigma^2)$  and  $Z \sim N(0, 1^2)$  and for any  $x_1, x_2 \in \mathbb{R}$ ,  $x_1 < x_2$ , with corresponding  $z$ -values  $z_1 = \frac{x_1 - \mu}{\sigma}$  and  $z_2 = \frac{x_2 - \mu}{\sigma}$ , then:

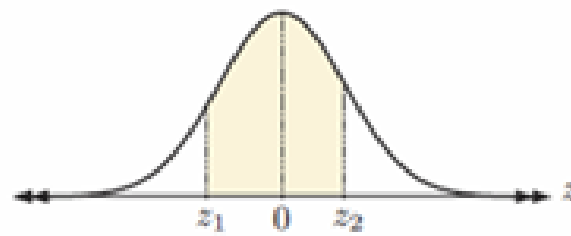
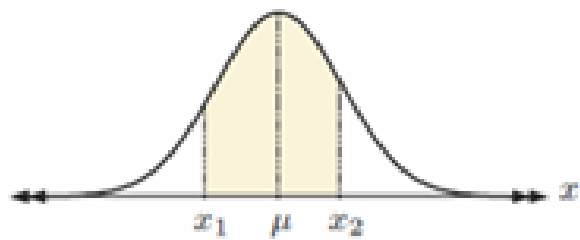
- $P(X \geq x_1) = P(Z \geq z_1)$



- $P(X \leq x_1) = P(Z \leq z_1)$



- $P(x_1 \leq X \leq x_2) = P(z_1 \leq Z \leq z_2)$



z-scores are useful when comparing two populations with different means and standard deviations (as long as both distributions are approximately normal).

$$z = \frac{x - \mu}{\sigma}$$

Example:

Pg. 642

1 The table shows Emma's midyear exam results. The exam results for each subject are normally distributed with the mean  $\mu$  and standard deviation  $\sigma$  shown in the table.

- Find the z-score for each of Emma's subjects.
- Arrange Emma's subjects from 'best' to 'worst' in terms of the z-scores.

Subject	Emma's score	$\mu$	$\sigma$
English	48	40	4.4
Mandarin	81	60	9
Geography	84	55	18
Biology	68	50	20
Maths	84	50	15

English

$$z = \frac{48 - 40}{4.4} = 1.82$$

Mand

$$z = \frac{81 - 60}{9} = 2.33$$

Geo

$$z = \frac{84 - 55}{18} = 1.61$$

Bio

$$z = \frac{68 - 50}{20} = 0.9$$

math

$$z = \frac{84 - 50}{15} = 2.27$$

## Pg. 643

- 5** If  $Z$  is the standard normal distribution, find the following probabilities using technology. In each case sketch the regions.
- a**  $P(-0.86 \leq Z \leq 0.32)$
  - d**  $P(Z \leq -0.53)$
  - g**  $P(Z > 4)$



## 24D – Quantiles or $k$ -values

When finding quantiles, we are given a probability and are asked to calculate the corresponding measurement.

For example, the masses of Oreo cookies are normally distributed with a mean of 40 g and a standard deviation of 2 g.

Find the value of  $k$  such that approximately 16% of the cookies are below  $k$  g.

We need to find  $k$  such that  $P(X \leq k) = 0.16$ .

Use the **inverse normal function** on the calculator:

## Pg. 645

- 3** Suppose  $X \sim N(38.7, 8.2^2)$ . Illustrate with a sketch and find  $k$  such that:
- a**  $P(X \leq k) = 0.9$
  - b**  $P(X \geq k) = 0.8$

**\*\*** If we are trying to find an unknown mean  $\mu$  or standard deviation  $\sigma$ , then we **always** need to convert to z-scores.

Pg. 646

- 1** The IQs of students at school are normally distributed with a standard deviation of 15. If 20% of students have an IQ higher than 125, find the mean IQ of students at school.

## Pg. 647

- 5**
- a** Find the mean and standard deviation of a normally distributed random variable  $X$ , given that  $P(X \geq 80) = 0.1$  and  $P(X \leq 30) = 0.15$ .
  - b** In a Mathematics examination it was found that 10% of the students scored at least 80, and no more than 15% scored under 30. Assuming the scores are normally distributed, what proportion of students scored more than 50?