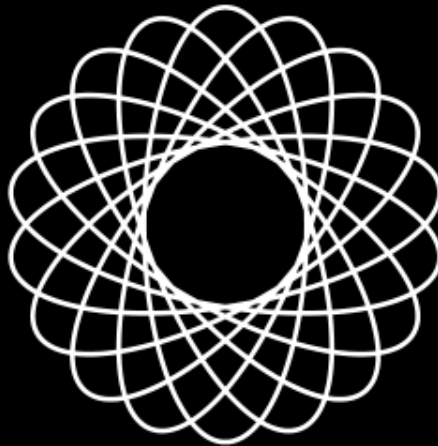


DATA SCIENCE





STATISTICAL CONCEPTS

What does Statistics cover?

Sample v/s Population

Probability Theory

Probability Distribution Concepts



Types of Distributions

1. **Discrete**
2. **Continuous**



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ❑ A coin flip



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year
- ☐ The number of people claiming fire insurance in a month



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year
- ☐ The number of people claiming fire insurance in a month

Kinds of discrete distributions:



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year
- ☐ The number of people claiming fire insurance in a month

Kinds of discrete distributions:

- ☐ Binomial or Bernoulli



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year
- ☐ The number of people claiming fire insurance in a month

Kinds of discrete distributions:

- ☐ Binomial or Bernoulli
- ☐ Negative Binomial



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year
- ☐ The number of people claiming fire insurance in a month

Kinds of discrete distributions:

- ☐ Binomial or Bernoulli
- ☐ Negative Binomial
- ☐ Geometric



Discrete Distribution

If a random variable can have only discrete outcomes, we have a discrete probability distribution

Examples:

- ☐ A coin flip
- ☐ The number of customers walking into a store on any given day
- ☐ The number of times a machine breaks down in a year
- ☐ The number of people claiming fire insurance in a month

Kinds of discrete distributions:

- ☐ Binomial or Bernoulli
- ☐ Negative Binomial
- ☐ Geometric
- ☐ Poisson



Binomial Distribution



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:

- ☐ Gender of babies delivered in a hospital



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:

- ☐ Gender of babies delivered in a hospital
- ☐ Fatal side effect deaths for a Schedule H drug



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:

- ☐ Gender of babies delivered in a hospital
- ☐ Fatal side effect deaths for a Schedule H drug

The common feature across all these is:



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:

- ☐ Gender of babies delivered in a hospital
- ☐ Fatal side effect deaths for a Schedule H drug

The common feature across all these is:

1. There are only two possible outcomes: Win or Lose, 1 or 0, Male or Female



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:

- ☐ Gender of babies delivered in a hospital
- ☐ Fatal side effect deaths for a Schedule H drug

The common feature across all these is:

1. There are only two possible outcomes: Win or Lose, 1 or 0, Male or Female
2. There are no external factors influencing the probability of each outcome over time



Binomial Distribution

The Binomial Distribution is an example of a probability distribution of a discrete random variable

Examples:

- ☐ Gender of babies delivered in a hospital
- ☐ Fatal side effect deaths for a Schedule H drug

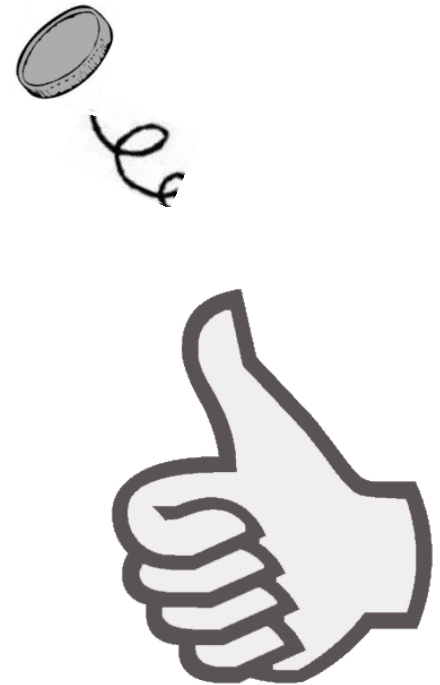
The common feature across all these is:

1. There are only two possible outcomes: Win or Lose, 1 or 0, Male or Female
2. There are no external factors influencing the probability of each outcome over time
3. The chances of each outcome are independent of previous results



Binomial Distribution

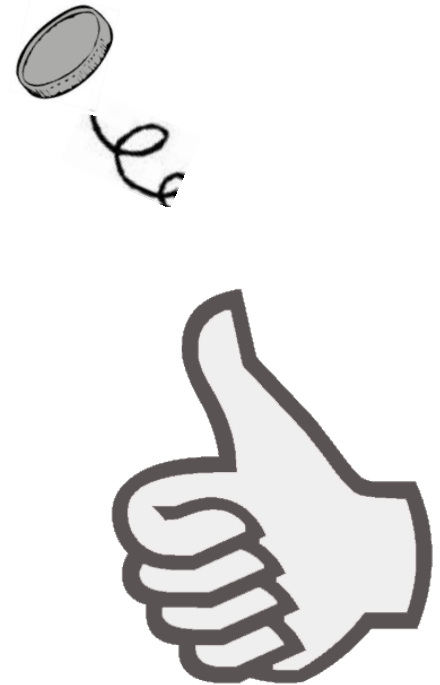
Simple non-business example: a coin toss



Binomial Distribution

Simple non-business example: a coin toss

- Outcome: **Random**



Binomial Distribution

Simple non-business example: a coin toss

- Outcome: **Random**
- Possible outcomes: **Two**



Binomial Distribution

Simple non-business example: a coin toss

- Outcome: **Random**
- Possible outcomes: **Two**
- External factors influencing outcome probability over time: **No**



Binomial Distribution

Simple non-business example: a coin toss

- Outcome: **Random**
- Possible outcomes: **Two**
- External factors influencing outcome probability over time: **No**
- Chances of each outcome independent of previous trials: **Yes**



Binomial Distribution

Toss a coin once: Bernoulli trial



Binomial Distribution

Toss a coin once: **Bernoulli trial**

Toss a coin 10 times:

- Measure # of *heads* observed: outcome follows a **binomial distribution**



Binomial Distribution

Toss a coin once: **Bernoulli trial**

Toss a coin 10 times:

- Measure # of *heads* observed: outcome follows a **binomial distribution**
- How many *heads* will we see? 0? 1? 2? ...10?



Binomial Distribution

Toss a coin once: **Bernoulli trial**

Toss a coin 10 times:

- Measure # of *heads* observed: outcome follows a **binomial distribution**
- How many *heads* will we see? 0? 1? 2? ...10?
- Are all of these outcomes equally likely?



Binomial Distribution

We calculate these probabilities using a mathematical formula:

probability distribution function



Binomial Distribution

We calculate these probabilities using a mathematical formula:

probability distribution function

For a binomial distribution:

PDF:
$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$



Binomial Distribution

We calculate these probabilities using a mathematical formula:

probability distribution function

For a binomial distribution:

PDF:
$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

Where:

x = Outcome



Binomial Distribution

We calculate these probabilities using a mathematical formula:

probability distribution function

For a binomial distribution:

PDF:
$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

Where:

x = Outcome

n = Trials



Binomial Distribution

We calculate these probabilities using a mathematical formula:

probability distribution function

For a binomial distribution:

PDF:
$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

Where:

x = Outcome

n = Trials

p = Probability of success on each trial



Binomial Distribution

Probability of 4 *heads*?

$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$



Binomial Distribution

Probability of 4 *heads*?

$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$p(4) = (10! / 4! * 6!) * (0.5)^4 * (0.6)^6 = 0.205$$



Binomial Distribution

Probability of 4 *heads*?

$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$p(4) = (10! / 4! * 6!) * (0.5)^4 * (0.6)^6 = 0.205$$

$$p(8) = ?$$



Binomial Distribution

Probability of 4 *heads*?

$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$p(4) = (10! / 4! * 6!) * (0.5)^4 * (0.6)^6 = 0.205$$

$$p(8) = ?$$

There are built in functions in statistical tools that generate these calculations



Binomial Distribution

Probability of 4 *heads*?

$$P(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$p(4) = (10! / 4! * 6!) * (0.5)^4 * (0.6)^6 = 0.205$$

$$p(8) = ?$$

There are built in functions in statistical tools that generate these calculations

In Excel: **BINOM.DIST**



Binomial Distribution

=BINOM.DIST(|

BINOM.DIST(**number_s**, trials, probability_s, cumulative)



Binomial Distribution

=BINOM.DIST(|

BINOM.DIST(**number_s**, trials, probability_s, cumulative)

In our example:

number_s = 4

trials = 10

probability_s = 0.5

cumulative = False



Binomial Distribution

=BINOM.DIST(|

BINOM.DIST(**number_s**, trials, probability_s, cumulative)

In our example:

number_s = 4

trials = 10

probability_s = 0.5

cumulative = False

=BINOM.DIST(4,10,0.5,false)|



Binomial Distribution

Example: Quality Control



Binomial Distribution

Example: Quality Control

In a manufacturing unit, the process has a defect rate of 10%.



Binomial Distribution

Example: Quality Control

In a manufacturing unit, the process has a defect rate of 10%.

As part of the QC process, you randomly evaluate products in lots of 10.



Binomial Distribution

Example: Quality Control

In a manufacturing unit, the process has a defect rate of 10%.

As part of the QC process, you randomly evaluate products in lots of 10.

Supposing from a random lot you get 2 defects,

how likely is this outcome due to random chance?



Binomial Distribution

Example: Quality Control

In a manufacturing unit, the process has a defect rate of 10%.

As part of the QC process, you randomly evaluate products in lots of 10.

Supposing from a random lot you get 2 defects,

how likely is this outcome due to random chance?

`=BINOM.DIST(2,10,0.1,FALSE) = 0.19`



Coming Up

Binomial Distribution Example:

How a Finance Manager may use Binomial Distribution to create a contingency fund based on likelihood



Binomial Distribution

Accounts Receivable Balances for All Divisions as of October 25, 2011

Past due balances and expected cash flow based on balance due date

Generated: October 25, 2011 at 12:08 PM



Month	Paid	AR Balances	Running AR Total
2012 January	\$44,220.30	\$188,809.50	\$213,696.46
2012 February	\$114,300.00	\$768,978.05	\$982,674.51
2012 March	\$271,091.61	\$1,568,852.37	\$2,551,526.88
2012 April	\$147,800.00	\$905,515.20	\$3,457,042.08
2012 May	\$65,100.00	\$373,712.30	\$3,830,754.38
2012 June	\$47,570.70	\$320,351.10	\$4,151,105.48
2012 July	\$0.00	\$24,098.92	\$4,175,204.40
2012 August	\$0.00	\$5,856.91	\$4,181,061.31
2012 October	\$0.00	\$1,642.92	\$4,182,704.23



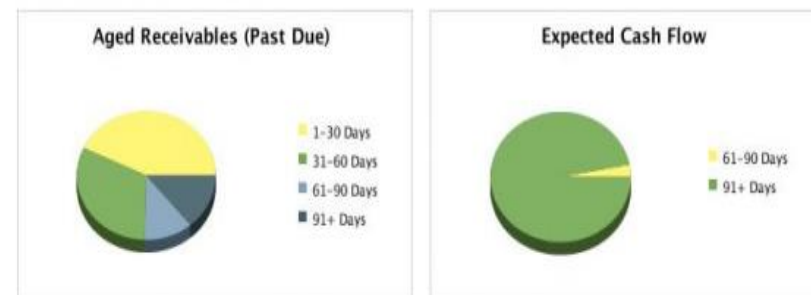
Binomial Distribution

As Finance Manager you are reviewing AR balances:

Accounts Receivable Balances for All Divisions as of October 25, 2011

Past due balances and expected cash flow based on balance due date

Generated: October 25, 2011 at 12:08 PM



Month	Paid	AR Balances	Running AR Total
2012 January	\$44,220.30	\$188,809.50	\$213,696.46
2012 February	\$114,300.00	\$768,978.05	\$982,674.51
2012 March	\$271,091.61	\$1,568,852.37	\$2,551,526.88
2012 April	\$147,800.00	\$905,515.20	\$3,457,042.08
2012 May	\$65,100.00	\$373,712.30	\$3,830,754.38
2012 June	\$47,570.70	\$320,351.10	\$4,151,105.48
2012 July	\$0.00	\$24,098.92	\$4,175,204.40
2012 August	\$0.00	\$5,856.91	\$4,181,061.31
2012 October	\$0.00	\$1,642.92	\$4,182,704.23



Binomial Distribution

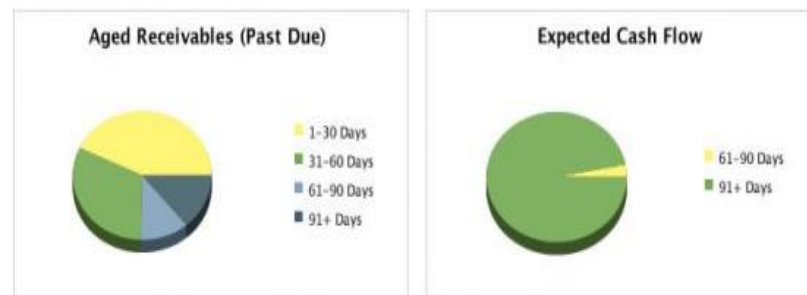
As Finance Manager you are reviewing AR balances:

- Based on past data, on average 40% of customers are more than 60 days late with payments

Accounts Receivable Balances for All Divisions as of October 25, 2011

Past due balances and expected cash flow based on balance due date

Generated: October 25, 2011 at 12:08 PM



Month	Paid	AR Balances	Running AR Total
2012 January	\$44,220.30	\$188,809.50	\$213,696.46
2012 February	\$114,300.00	\$768,978.05	\$982,674.51
2012 March	\$271,091.61	\$1,568,852.37	\$2,551,526.88
2012 April	\$147,800.00	\$905,515.20	\$3,457,042.08
2012 May	\$65,100.00	\$373,712.30	\$3,830,754.38
2012 June	\$47,570.70	\$320,351.10	\$4,151,105.48
2012 July	\$0.00	\$24,098.92	\$4,175,204.40
2012 August	\$0.00	\$5,856.91	\$4,181,061.31
2012 October	\$0.00	\$1,642.92	\$4,182,704.23



Binomial Distribution

As Finance Manager you are reviewing AR balances:

- Based on past data, on average 40% of customers are more than 60 days late with payments
- Assuming a total of 150 customers, you want to create a contingency where > 50% of customers are late

Accounts Receivable Balances for All Divisions as of October 25, 2011

Past due balances and expected cash flow based on balance due date

Generated: October 25, 2011 at 12:08 PM



Month	Paid	AR Balances	Running AR Total
2012 January	\$44,220.30	\$188,809.50	\$213,696.46
2012 February	\$114,300.00	\$768,978.05	\$982,674.51
2012 March	\$271,091.61	\$1,568,852.37	\$2,551,526.88
2012 April	\$147,800.00	\$905,515.20	\$3,457,042.08
2012 May	\$65,100.00	\$373,712.30	\$3,830,754.38
2012 June	\$47,570.70	\$320,351.10	\$4,151,105.48
2012 July	\$0.00	\$24,098.92	\$4,175,204.40
2012 August	\$0.00	\$5,856.91	\$4,181,061.31
2012 October	\$0.00	\$1,642.92	\$4,182,704.23



Binomial Distribution

As Finance Manager you are reviewing AR balances:

- Based on past data, on average 40% of customers are more than 60 days late with payments
- Assuming a total of 150 customers, you want to create a contingency where > 50% of customers are late
- The size of the contingency fund is directly proportional to the probability of > 50% of customers being late

Accounts Receivable Balances for All Divisions as of October 25, 2011

Past due balances and expected cash flow based on balance due date

Generated: October 25, 2011 at 12:08 PM



Month	Paid	AR Balances	Running AR Total
2012 January	\$44,220.30	\$188,809.50	\$213,696.46
2012 February	\$114,300.00	\$768,978.05	\$982,674.51
2012 March	\$271,091.61	\$1,568,852.37	\$2,551,526.88
2012 April	\$147,800.00	\$905,515.20	\$3,457,042.08
2012 May	\$65,100.00	\$373,712.30	\$3,830,754.38
2012 June	\$47,570.70	\$320,351.10	\$4,151,105.48
2012 July	\$0.00	\$24,098.92	\$4,175,204.40
2012 August	\$0.00	\$5,856.91	\$4,181,061.31
2012 October	\$0.00	\$1,642.92	\$4,182,704.23



Binomial Distribution

As Finance Manager you are reviewing AR balances:

- Based on past data, on average 40% of customers are more than 60 days late with payments
- Assuming a total of 150 customers, you want to create a contingency where > 50% of customers are late
- The size of the contingency fund is directly proportional to the probability of > 50% of customers being late
- You need a sense of the likelihood of > 50% being late in any given month

Accounts Receivable Balances for All Divisions as of October 25, 2011

Past due balances and expected cash flow based on balance due date

Generated: October 25, 2011 at 12:08 PM



Month	Paid	AR Balances	Running AR Total
2012 January	\$44,220.30	\$188,809.50	\$213,696.46
2012 February	\$114,300.00	\$768,978.05	\$982,674.51
2012 March	\$271,091.61	\$1,568,852.37	\$2,551,526.88
2012 April	\$147,800.00	\$905,515.20	\$3,457,042.08
2012 May	\$65,100.00	\$373,712.30	\$3,830,754.38
2012 June	\$47,570.70	\$320,351.10	\$4,151,105.48
2012 July	\$0.00	\$24,098.92	\$4,175,204.40
2012 August	\$0.00	\$5,856.91	\$4,181,061.31
2012 October	\$0.00	\$1,642.92	\$4,182,704.23



Binomial Distribution

- Probability of $> 50\%$ of customers being late in any month?



Binomial Distribution

- Probability of > 50% of customers being late in any month?

$P(76 \text{ being late}) + P(77 \text{ being late}) + P(78 \text{ being late}) + \dots P(150 \text{ being late})$



Binomial Distribution

- Probability of > 50% of customers being late in any month?

$P(76 \text{ being late}) + P(77 \text{ being late}) + P(78 \text{ being late}) + \dots P(150 \text{ being late})$

- This is a tedious manual calculation



Binomial Distribution

- Probability of > 50% of customers being late in any month?

$P(76 \text{ being late}) + P(77 \text{ being late}) + P(78 \text{ being late}) + \dots P(150 \text{ being late})$

- This is a tedious manual calculation
- A faster method is to use the concept of **cumulative probability**



Binomial Distribution

In Excel:

=BINOM.DIST(S,T,P, **FALSE**) :



Binomial Distribution

In Excel:

=BINOM.DIST(S,T,P, **FALSE**) :

the random chance probability of seeing EXACTLY S successes in T Bernoulli trials, when the probability of success on any trial is P



Binomial Distribution

In Excel:

=BINOM.DIST(S,T,P, **FALSE**) :

the random chance probability of seeing EXACTLY S successes in T Bernoulli trials, when the probability of success on any trial is P

=BINOM.DIST(S,T,0, **TRUE**):



Binomial Distribution

In Excel:

=BINOM.DIST(S,T,P, **FALSE**) :

the random chance probability of seeing EXACTLY S successes in T Bernoulli trials, when the probability of success on any trial is P

=BINOM.DIST(S,T,0, **TRUE**):

the random chance probability of seeing LESS THAN EQUAL TO S successes in T Bernoulli trials, when the probability of success on any trial is P



Binomial Distribution

Probability of seeing $> 50\%$ of customers being late



Binomial Distribution

Probability of seeing $> 50\%$ of customers being late

= Probability of seeing > 75 customers being late



Binomial Distribution

Probability of seeing > 50% of customers being late

= Probability of seeing > 75 customers being late

= $1 - \text{BINOM.DIST}(75, 150, 0.4, \text{TRUE})$



Binomial Distribution

Probability of seeing > 50% of customers being late

= Probability of seeing > 75 customers being late

= $1 - \text{BINOM.DIST}(75, 150, 0.4, \text{TRUE})$

= 0.005



Coming Up

Discrete Distributions:

Hypergeometric Distribution

Hypergeometric Distribution

A hypergeometric distribution is generated when you have Bernoulli trials, **but selections are not replaced**

$$P\{X = k\} = \frac{\binom{K}{k} \binom{N-K}{n-k}}{\binom{N}{n}}$$

Where

$$\binom{A}{B} = \frac{A!}{(A-B)!B!}$$



Hypergeometric Distribution

Example: HR Policies and Diversity

A company wants to encourage diversity in its management ranks. Of the 18 employees eligible for promotion into middle management, 9 are women.



Hypergeometric Distribution

Example: HR Policies and Diversity

A company wants to encourage diversity in its management ranks. Of the 18 employees eligible for promotion into middle management, 9 are women.

The promotions are announced: 8 are promoted of which 3 are women.
Is there a problem?



PROMOTED
?



Hypergeometric Distribution

Example: HR Policies and Diversity

A company wants to encourage diversity in its management ranks. Of the 18 employees eligible for promotion into middle management, 9 are women.

The promotions are announced: 8 are promoted of which 3 are women.
Is there a problem?



PROMOTED
?



In this case, we use Hypergeometric Distribution rather than Binomial



Hypergeometric Distribution

Example: HR Policies and Diversity

A company wants to encourage diversity in its management ranks. Of the 18 employees eligible for promotion into middle management, 9 are women.

The promotions are announced: 8 are promoted of which 3 are women.
Is there a problem?



PROMOTED
?



In this case, we use Hypergeometric Distribution rather than Binomial

=hypgeom.dist(3,8,9,18,false) = 0.24



Other Binomial Type Distributions

Negative Binomial



Other Binomial Type Distributions

Negative Binomial

- Used to find out the number of trials needed to get X successes



Other Binomial Type Distributions

Negative Binomial

- Used to find out the number of trials needed to get X successes
- Example: What is the probability that the 30th purchase in my store will happen with the 100th customer, when the probability of purchase for any customer is 20%?



Other Binomial Type Distributions

Negative Binomial

- Used to find out the number of trials needed to get X successes
- Example: What is the probability that the 30th purchase in my store will happen with the 100th customer, when the probability of purchase for any customer is 20%?

=NEGBINOM.DIST (number_f, number_s, probability_s, cumulative)



Other Binomial Type Distributions

Negative Binomial

- Used to find out the number of trials needed to get X successes
- Example: What is the probability that the 30th purchase in my store will happen with the 100th customer, when the probability of purchase for any customer is 20%?

=NEGBINOM.DIST (number_f, number_s, probability_s, cumulative)

=NEGBINOM.DIST(70,30,0.2,FALSE) = 0.001



Other Binomial Type Distributions

Geometric Distribution



Other Binomial Type Distributions

Geometric Distribution

- Probability of the first success in the r th trial



Other Binomial Type Distributions

Geometric Distribution

- Probability of the first success in the r th trial
- Example: Supposing there is a defect rate of 2% with some mechanical component being produced. What is the probability that a QC Inspector will need to review at most 20 items before finding a defect?



Other Binomial Type Distributions

Geometric Distribution

- Probability of the first success in the r th trial
- Example: Supposing there is a defect rate of 2% with some mechanical component being produced. What is the probability that a QC Inspector will need to review at most 20 items before finding a defect?

=NEGBINOM.DIST (number_f, number_s, probability_s, cumulative)



Other Binomial Type Distributions

Geometric Distribution

- Probability of the first success in the r th trial
- Example: Supposing there is a defect rate of 2% with some mechanical component being produced. What is the probability that a QC Inspector will need to review at most 20 items before finding a defect?

=NEGBINOM.DIST (number_f, number_s, probability_s, cumulative)

=NEGBINOM.DIST(19,1,0.02,true) = 0.33



Coming Up

Discrete Distributions:

Poisson Distribution

Poisson Distribution



Poisson Distribution

Another discrete probability distribution used to model number of events occurring in a time frame



Poisson Distribution

Another discrete probability distribution used to model number of events occurring in a time frame

Examples:

- ☐ Number of insurance claims in a month



Poisson Distribution

Another discrete probability distribution used to model number of events occurring in a time frame

Examples:

- ☐ Number of insurance claims in a month
- ☐ Disease spread in a day



Poisson Distribution

Another discrete probability distribution used to model number of events occurring in a time frame

Examples:

- ☐ Number of insurance claims in a month
- ☐ Disease spread in a day
- ☐ Number of telephone calls in an hour



Poisson Distribution

Another discrete probability distribution used to model number of events occurring in a time frame

Examples:

- ☐ Number of insurance claims in a month
- ☐ Disease spread in a day
- ☐ Number of telephone calls in an hour
- ☐ Number of patients needing emergency services in a day



Poisson Distribution

These conditions apply to correctly use a Poisson Distribution:



Poisson Distribution

These conditions apply to correctly use a Poisson Distribution:

1. Events have to be counted as whole numbers



Poisson Distribution

These conditions apply to correctly use a Poisson Distribution:

1. Events have to be counted as whole numbers
2. Events are independent: so if one event occurs, it does not impact the chances of the second event occurring



Poisson Distribution

These conditions apply to correctly use a Poisson Distribution:

1. Events have to be counted as whole numbers
2. Events are independent: so if one event occurs, it does not impact the chances of the second event occurring
3. Average frequency of occurrence for the given time period is known



Poisson Distribution

These conditions apply to correctly use a Poisson Distribution:

1. Events have to be counted as whole numbers
2. Events are independent: so if one event occurs, it does not impact the chances of the second event occurring
3. Average frequency of occurrence for the given time period is known
4. Number of events that have already occurred can be counted



Poisson Distribution

Poisson Probabilities are calculated as:

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$



Poisson Distribution

Poisson Probabilities are calculated as:

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

where Lambda is the mean number of occurrences in a given interval of time



Poisson Distribution

Poisson Probabilities are calculated as:

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

where Lambda is the mean number of occurrences in a given interval of time

Notice here that there is no n (sample size) impact



Poisson Distribution

Example: Call Centre Management

You are a Manager in a call center with a staff of 55 people, who on average handle 330 calls in an hour. A holiday is coming up and 5 resources want leave. You estimate the 50 remaining resources can manage 20% greater calls, but want to plan for the chance of greater than 20% increased call volume.



Poisson Distribution

Example: Call Centre Management

You are a Manager in a call center with a staff of 55 people, who on average handle 330 calls in an hour. A holiday is coming up and 5 resources want leave. You estimate the 50 remaining resources can manage 20% greater calls, but want to plan for the chance of greater than 20% increased call volume.



- What are the chances that number of calls on that day will go up by more than 20%?



Poisson Distribution

Example: Call Centre Management

You are a Manager in a call center with a staff of 55 people, who on average handle 330 calls in an hour. A holiday is coming up and 5 resources want leave. You estimate the 50 remaining resources can manage 20% greater calls, but want to plan for the chance of greater than 20% increased call volume.



- What are the chances that number of calls on that day will go up by more than 20%?

$$\lambda = (330)/55 = 6 \text{ calls an hour;}$$



Poisson Distribution

Example: Call Centre Management

You are a Manager in a call center with a staff of 55 people, who on average handle 330 calls in an hour. A holiday is coming up and 5 resources want leave. You estimate the 50 remaining resources can manage 20% greater calls, but want to plan for the chance of greater than 20% increased call volume.



- What are the chances that number of calls on that day will go up by more than 20%?

$$\lambda = (330)/55 = 6 \text{ calls an hour;}$$

$$20\% \text{ greater calls with 5 less resources} = (330 \times 1.2)/50 = 7.2 (=7) \text{ calls an hour}$$



Poisson Distribution

Example: Call Centre Management

You are a Manager in a call center with a staff of 55 people, who on average handle 330 calls in an hour. A holiday is coming up and 5 resources want leave. You estimate the 50 remaining resources can manage 20% greater calls, but want to plan for the chance of greater than 20% increased call volume.



- What are the chances that number of calls on that day will go up by more than 20%?

$$\lambda = (330)/55 = 6 \text{ calls an hour;}$$

$$20\% \text{ greater calls with 5 less resources} = (330 \times 1.2)/50 = 7.2 (=7) \text{ calls an hour}$$

- We need **Probability of seeing 8** or more calls an hour when **average is 6**. Calculated manually or by referencing the Poisson Distribution table



Poisson Distribution

Example: ATM Machine



Poisson Distribution

Example: ATM Machine

At an ATM machine in a particular location, a bank notes that there an average of 80 withdrawals a day, with an average transaction amount of \$40. The bank needs to stock the ATM machine with appropriate levels of cash.



Poisson Distribution

Example: ATM Machine

At an ATM machine in a particular location, a bank notes that there an average of 80 withdrawals a day, with an average transaction amount of \$40. The bank needs to stock the ATM machine with appropriate levels of cash.

- Assuming that they start with a zero balance, what is the most appropriate amount of cash that needs to be stocked, for a 4 day period?



Poisson Distribution

Example: ATM Machine

At an ATM machine in a particular location, a bank notes that there an average of 80 withdrawals a day, with an average transaction amount of \$40. The bank needs to stock the ATM machine with appropriate levels of cash.

- Assuming that they start with a zero balance, what is the most appropriate amount of cash that needs to be stocked, for a 4 day period?
- A customer service KRA is to keep customer complaints to less than 10%



Binomial or Poisson?



Binomial or Poisson?

1. If a *mean / average probability* of an event happening per unit time / per page / per mile cycled etc., is given, and you are asked to calculate a probability of n events happening in a given time / number of pages / number of miles cycled, then the **Poisson Distribution** is used.



Binomial or Poisson?

1. If a *mean / average probability* of an event happening per unit time / per page / per mile cycled etc., is given, and you are asked to calculate a probability of n events happening in a given time / number of pages / number of miles cycled, then the **Poisson Distribution** is used.
2. If an *exact probability* of an event happening is given, or implied, in the question, and you are asked to calculate the probability of this event happening k times out of n , then the **Binomial Distribution** must be used.



Binomial or Poisson?

Binomial Distribution

describes the distribution of binary data from a **finite sample**. Thus it gives the probability of getting r events out of n trials.



Binomial or Poisson?

Binomial Distribution

describes the distribution of binary data from a **finite sample**. Thus it gives the probability of getting r events out of n trials.

Poisson Distribution

describes the distribution of binary data from an **infinite sample**. Thus it gives the probability of getting r events in a population.



Recap

Types of Distributions

Discrete:



Recap

Types of Distributions

Discrete:

- Binomial



Recap

Types of Distributions

Discrete:

- Binomial
- Hypergeometric



Recap

Types of Distributions

Discrete:

- Binomial
- Hypergeometric
- Negative Binomial



Recap

Types of Distributions

Discrete:

- Binomial
- Hypergeometric
- Negative Binomial
- Geometric



Recap

Types of Distributions

Discrete:

- Binomial
- Hypergeometric
- Negative Binomial
- Geometric
- Poisson



Continuous Distributions



Continuous Distributions

Continuous distributions are applicable when an event can take on any value within a given range



Continuous Distributions

Continuous distributions are applicable when an event can take on any value within a given range

Examples:



Continuous Distributions

Continuous distributions are applicable when an event can take on any value within a given range

Examples:

☐ Height of males in Bangalore



Continuous Distributions

Continuous distributions are applicable when an event can take on any value within a given range

Examples:

- ☐ Height of males in Bangalore
- ☐ Average waiting time per patient at a hospital



Continuous Distributions

Continuous distributions are applicable when an event can take on any value within a given range

Examples:

- ☐ Height of males in Bangalore
- ☐ Average waiting time per patient at a hospital
- ☐ Per Capita Income



Continuous Distributions

Continuous distributions are applicable when an event can take on any value within a given range

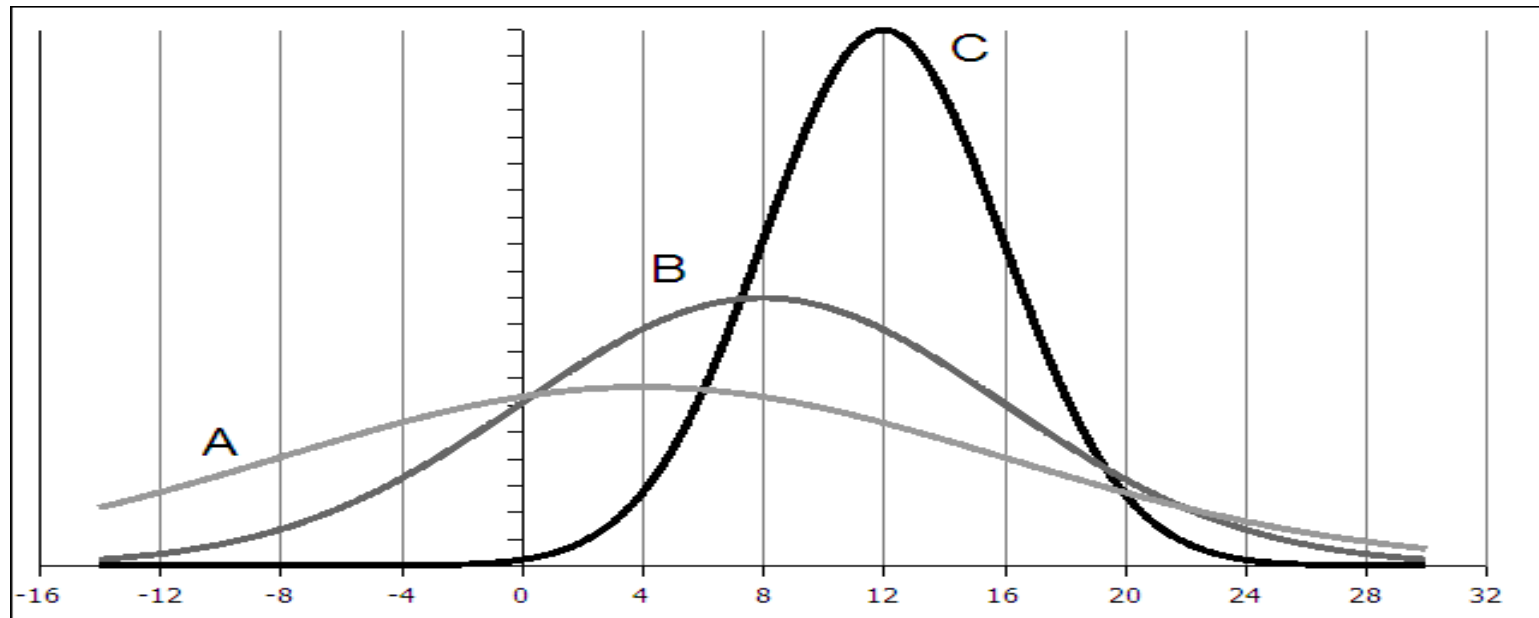
Examples:

- ☐ Height of males in Bangalore
- ☐ Average waiting time per patient at a hospital
- ☐ Per Capita Income

Normal Distribution is the most common kind of a continuous probability distribution due to its useful applications in Statistics

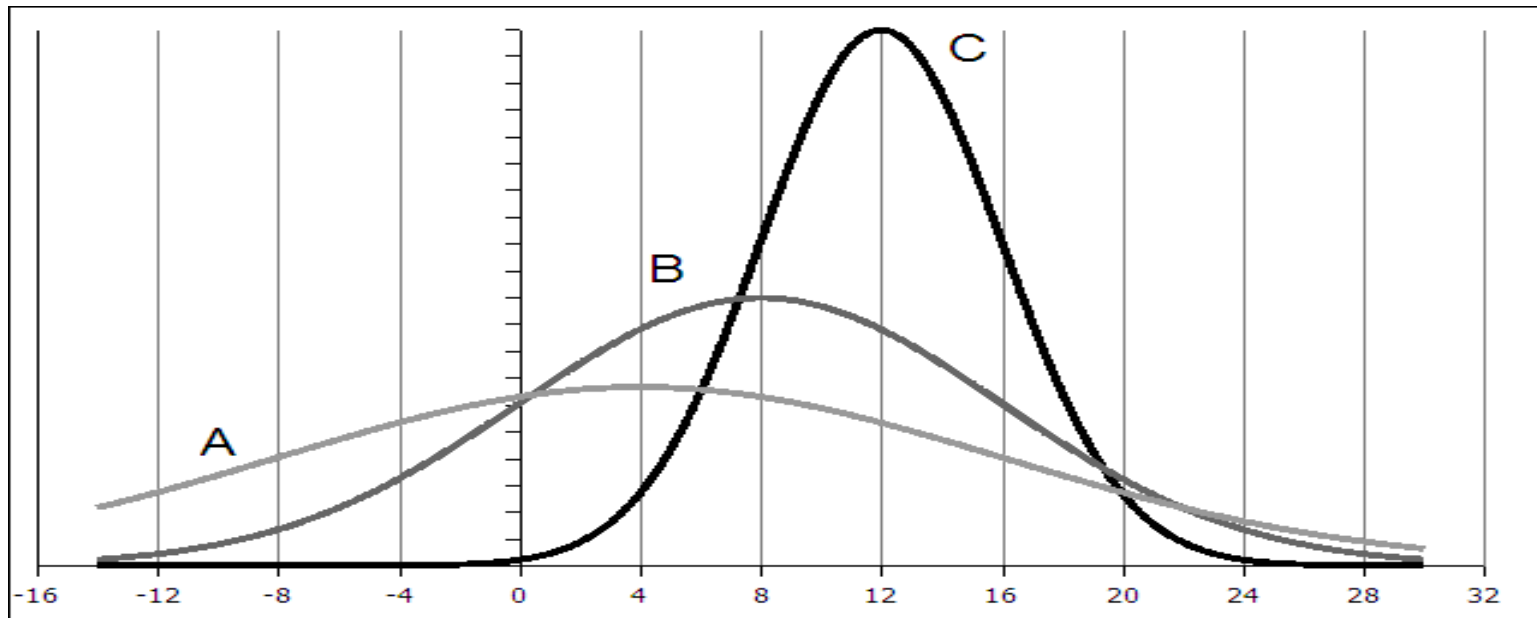


Normal Probability Distribution



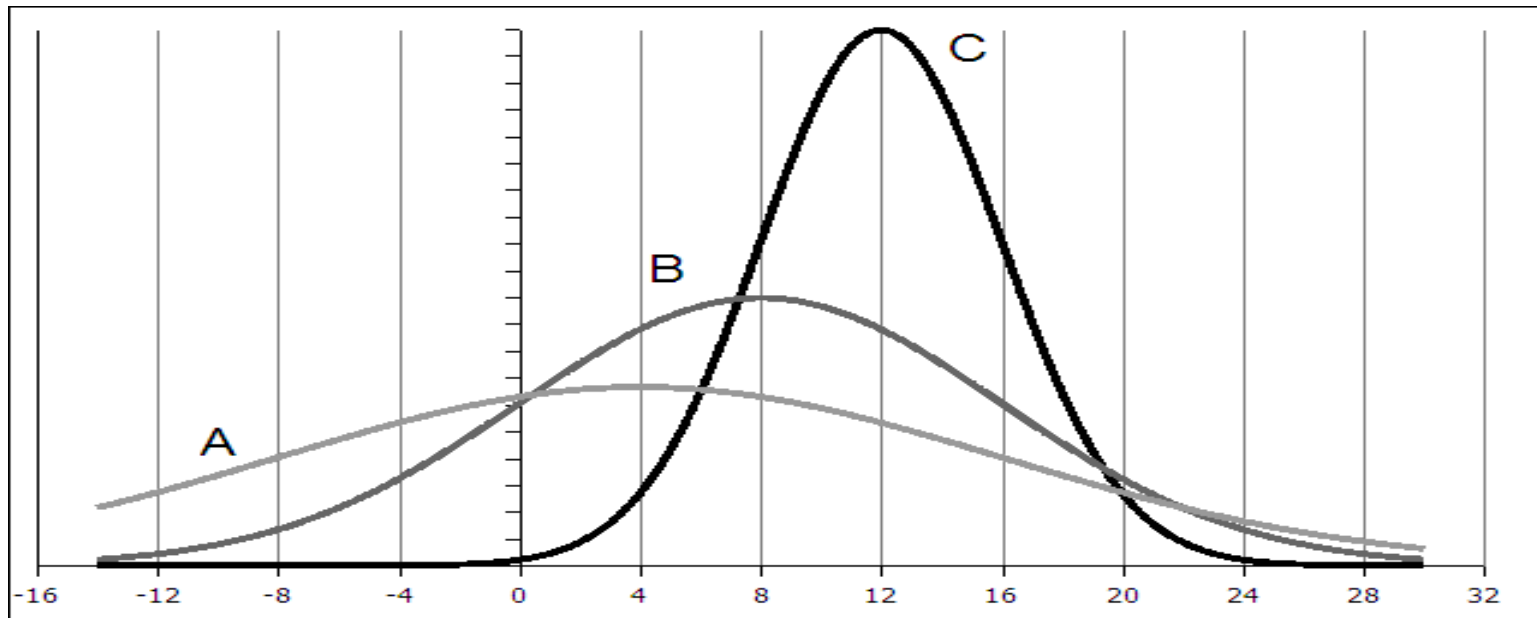
Normal Probability Distribution

1. Symmetric about the (single) mean



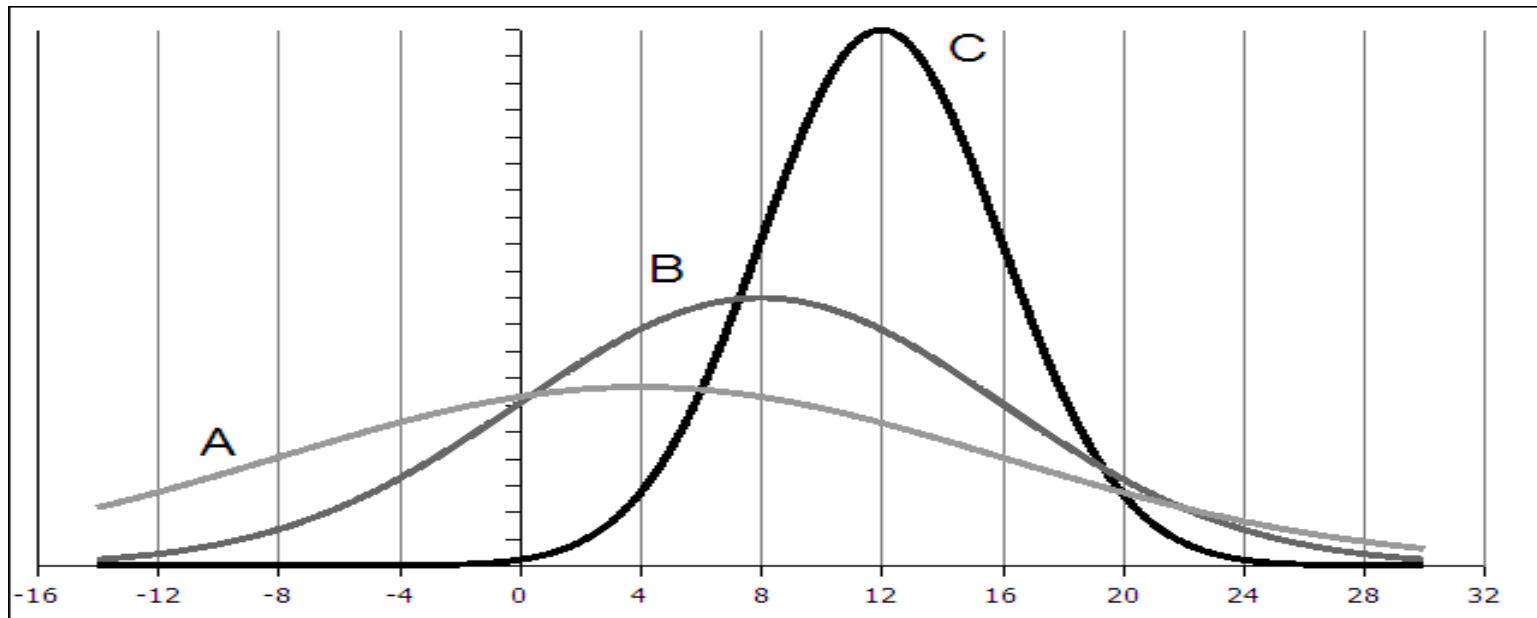
Normal Probability Distribution

1. Symmetric about the (single) mean
2. Mean = Median = Mode



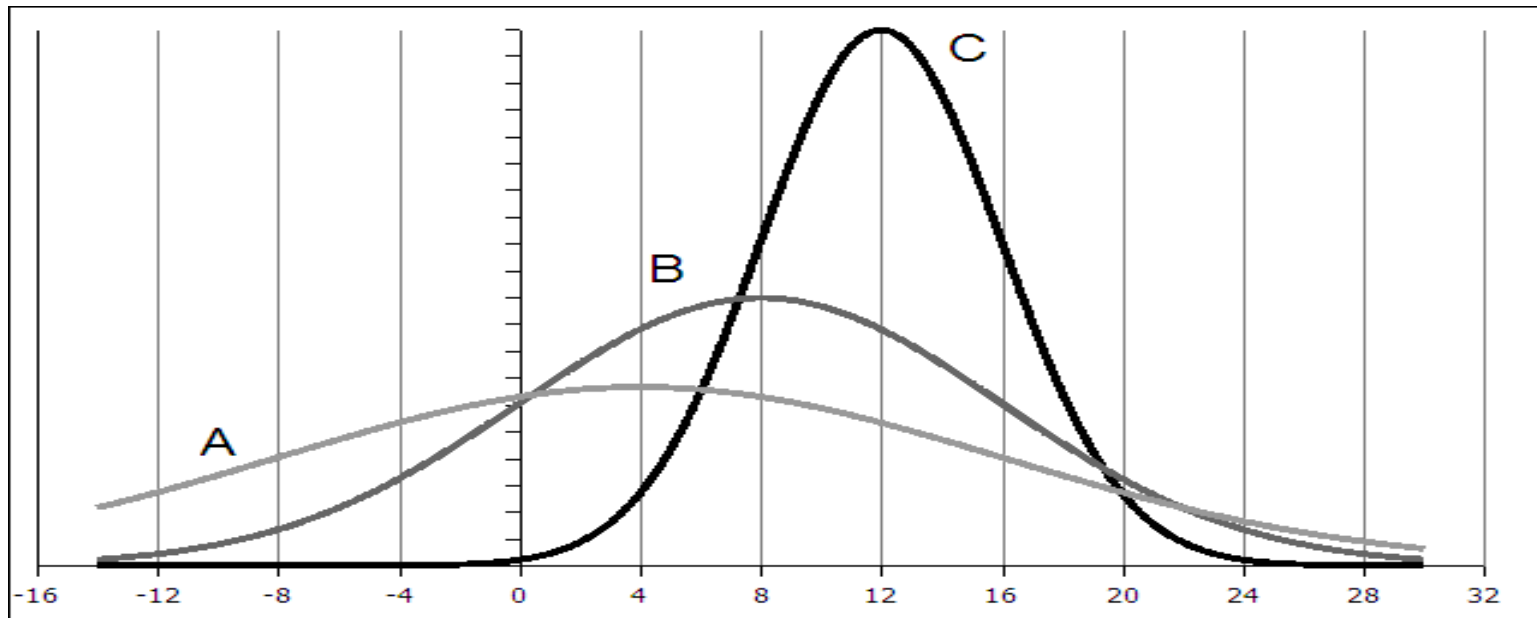
Normal Probability Distribution

1. Symmetric about the (single) mean
2. Mean = Median = Mode
3. The two tails extend indefinitely and never touch the axis



Normal Probability Distribution

1. Symmetric about the (single) mean
2. Mean = Median = Mode
3. The two tails extend indefinitely and never touch the axis



What are the differences between the curves?



Normal Probability Distribution

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

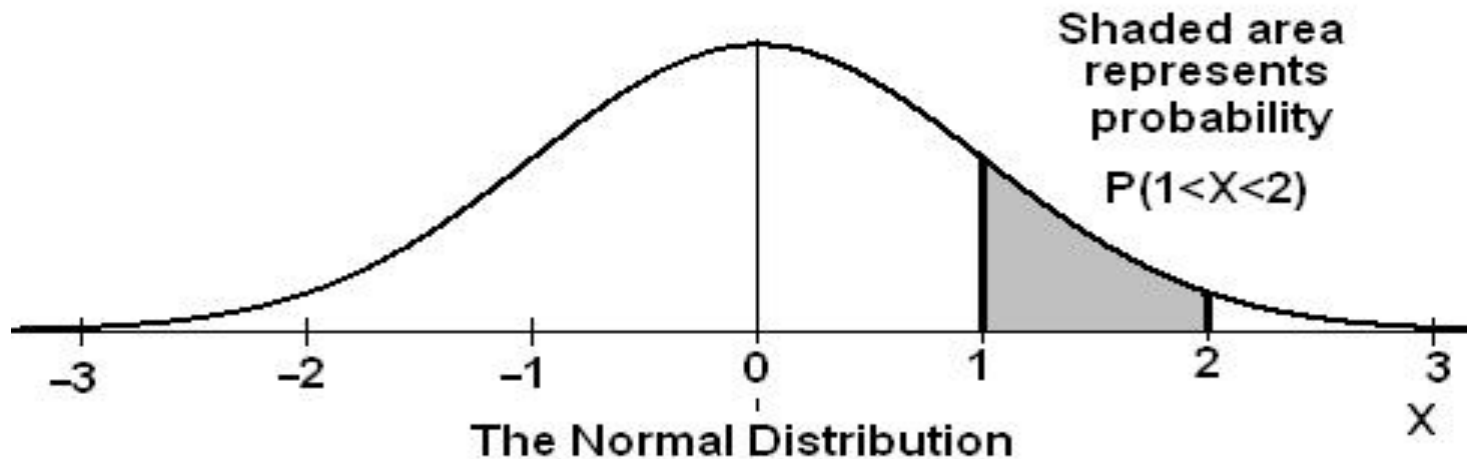
- where:
- y = vertical height of a point on the normal distribution
 - x = distance along the horizontal axis
 - σ = standard deviation of the data distribution
 - μ = mean of the data distribution
 - e = exponential constant = 2.71828...
 - π = pi = 3.14159....



Normal Distribution

Area under the curve:

The total area under a normal probability curve is always 1. This property allows us to think of the area as probability, and therefore we can compute probability two values on the curve



Normal Distribution

Example: Jigsaw Students IQ Testing



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random student from the 100, what are the chances he / she has an IQ > 115 ?



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random student from the 100, what are the chances he / she has an $IQ > 115$?

We know:

IQ Score: Random variable



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random student from the 100, what are the chances he / she has an $IQ > 115$?

We know:

IQ Score: Random variable

Distribution: Continuous normal



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random student from the 100, what are the chances he / she has an IQ > 115 ?

We know:

IQ Score: Random variable

Distribution: Continuous normal

Mean = 108



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random student from the 100, what are the chances he / she has an IQ > 115 ?

We know:

IQ Score: Random variable

Distribution: Continuous normal

Mean = 108

Std Dev: 7



Normal Distribution

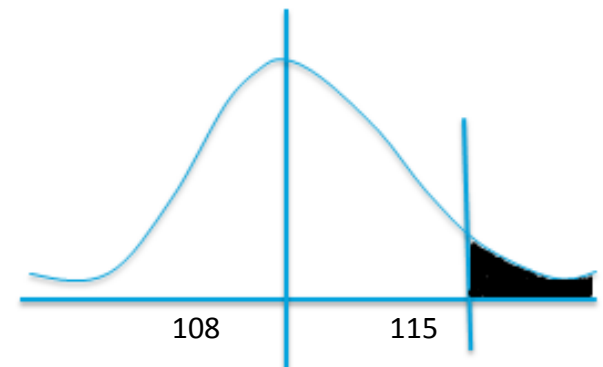
Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random person from the 100 students. What are the chances that that student has an IQ > 115 ?

We need:

$P(\text{Score} > 115)$



Normal Distribution

Example: Jigsaw Students IQ Testing

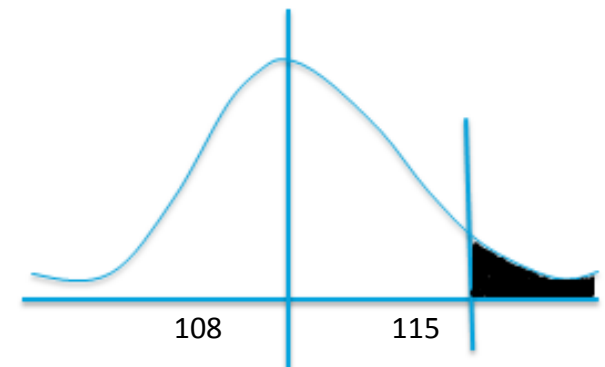
We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random person from the 100 students. What are the chances that that student has an IQ > 115 ?

We need:

$P(\text{Score} > 115)$

Which is nothing but $1 - P(\text{SCORE} \leq 115)$



Normal Distribution

Example: Jigsaw Students IQ Testing

We test 100 students and find that IQ is normally distributed with an average of 108, with a std deviation of 7.

Supposing you pick a random person from the 100 students. What are the chances that that student has an IQ > 115 ?

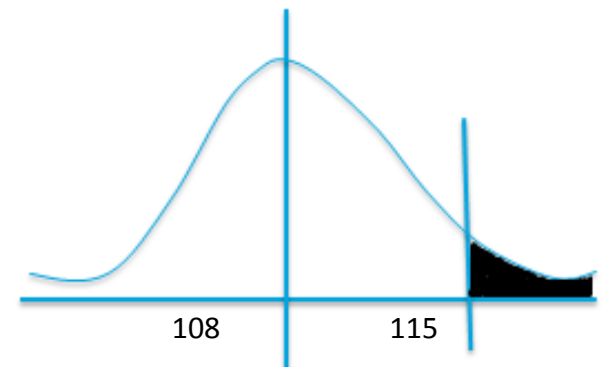
We need:

$P(\text{Score} > 115)$

Which is nothing but $1 - P(\text{SCORE} \leq 115)$

We can of course rely on Excel:

NORMDIST(Outcome, Mean, Std Dev, Cuml)



Normal Distribution

Example: Product Guarantee

A manufacturer wants to state a guarantee for performance of their product, in hours, so that product failure rates on the basis of hours of performance are restricted to less than 5%.

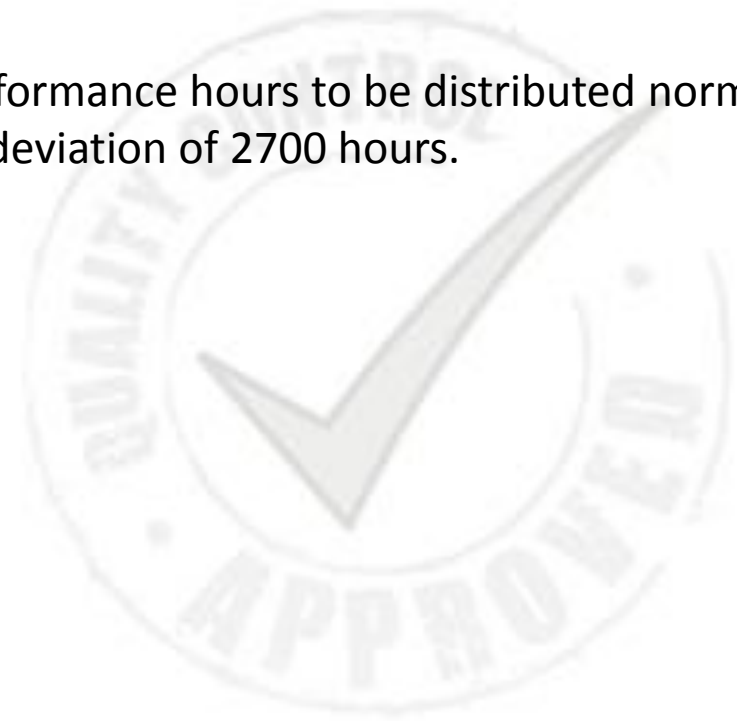


Normal Distribution

Example: Product Guarantee

A manufacturer wants to state a guarantee for performance of their product, in hours, so that product failure rates on the basis of hours of performance are restricted to less than 5%.

They test a 1000 samples, and find average performance hours to be distributed normally, with an average life of 71,450 hours, and a std deviation of 2700 hours.



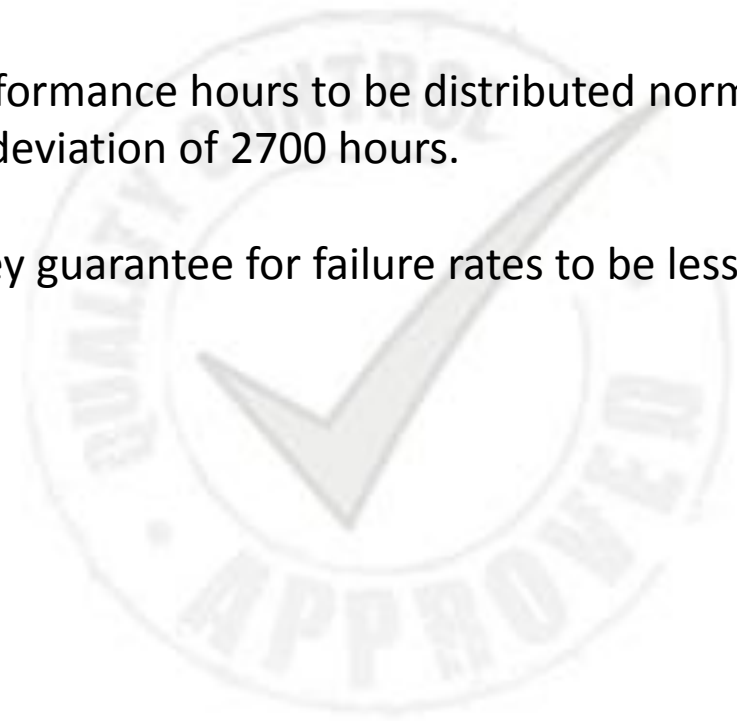
Normal Distribution

Example: Product Guarantee

A manufacturer wants to state a guarantee for performance of their product, in hours, so that product failure rates on the basis of hours of performance are restricted to less than 5%.

They test a 1000 samples, and find average performance hours to be distributed normally, with an average life of 71,450 hours, and a std deviation of 2700 hours.

What number of performance hours should they guarantee for failure rates to be less than 5%?



Normal Distribution

Example: Product Guarantee

A manufacturer wants to state a guarantee for performance of their product, in hours, so that product failure rates on the basis of hours of performance are restricted to less than 5%.

They test a 1000 samples, and find average performance hours to be distributed normally, with an average life of 71,450 hours, and a std deviation of 2700 hours.

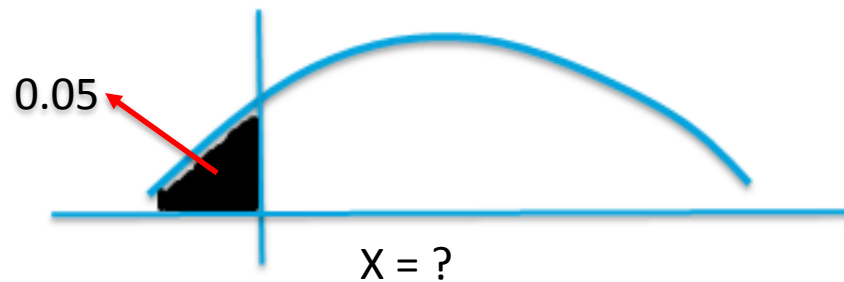
What number of performance hours should they guarantee for failure rates to be less than 5%?

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



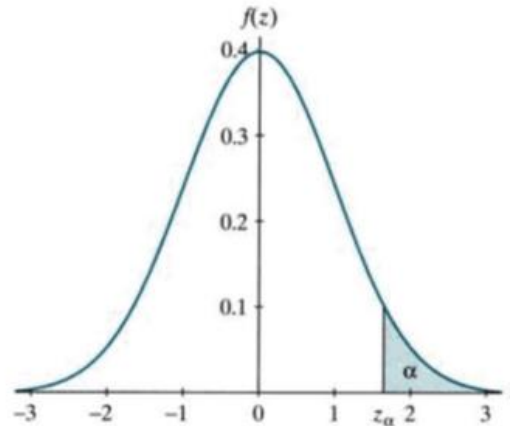
Normal Distribution

We could also calculate probabilities of multiple values of hours and identify the hours at which probability is less than 5%



Normal Distribution

But, there is an easier way to calculate the X value that will give us a specific probability – using a **standard normal probability table**



$$P(Z > z_{\alpha}) = \alpha$$

$$P(Z > z) = 1 - \Phi(z) = \Phi(-z)$$

z_{α}	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.4960	0.4920	0.4880	0.4840	0.4801	0.4761	0.4721	0.4681	0.4641
0.1	0.4602	0.4562	0.4522	0.4483	0.4443	0.4404	0.4364	0.4325	0.4286	0.4247
0.2	0.4207	0.4168	0.4129	0.4090	0.4052	0.4013	0.3974	0.3936	0.3897	0.3859
0.3	0.3821	0.3783	0.3745	0.3707	0.3669	0.3632	0.3594	0.3557	0.3520	0.3483
0.4	0.3446	0.3409	0.3372	0.3336	0.3300	0.3264	0.3228	0.3192	0.3156	0.3121
0.5	0.3085	0.3050	0.3015	0.2981	0.2946	0.2912	0.2877	0.2843	0.2810	0.2776
0.6	0.2743	0.2709	0.2676	0.2643	0.2611	0.2578	0.2546	0.2514	0.2483	0.2451
0.7	0.2420	0.2389	0.2358	0.2327	0.2296	0.2266	0.2236	0.2206	0.2177	0.2148
0.8	0.2119	0.2090	0.2061	0.2033	0.2005	0.1977	0.1949	0.1922	0.1894	0.1867



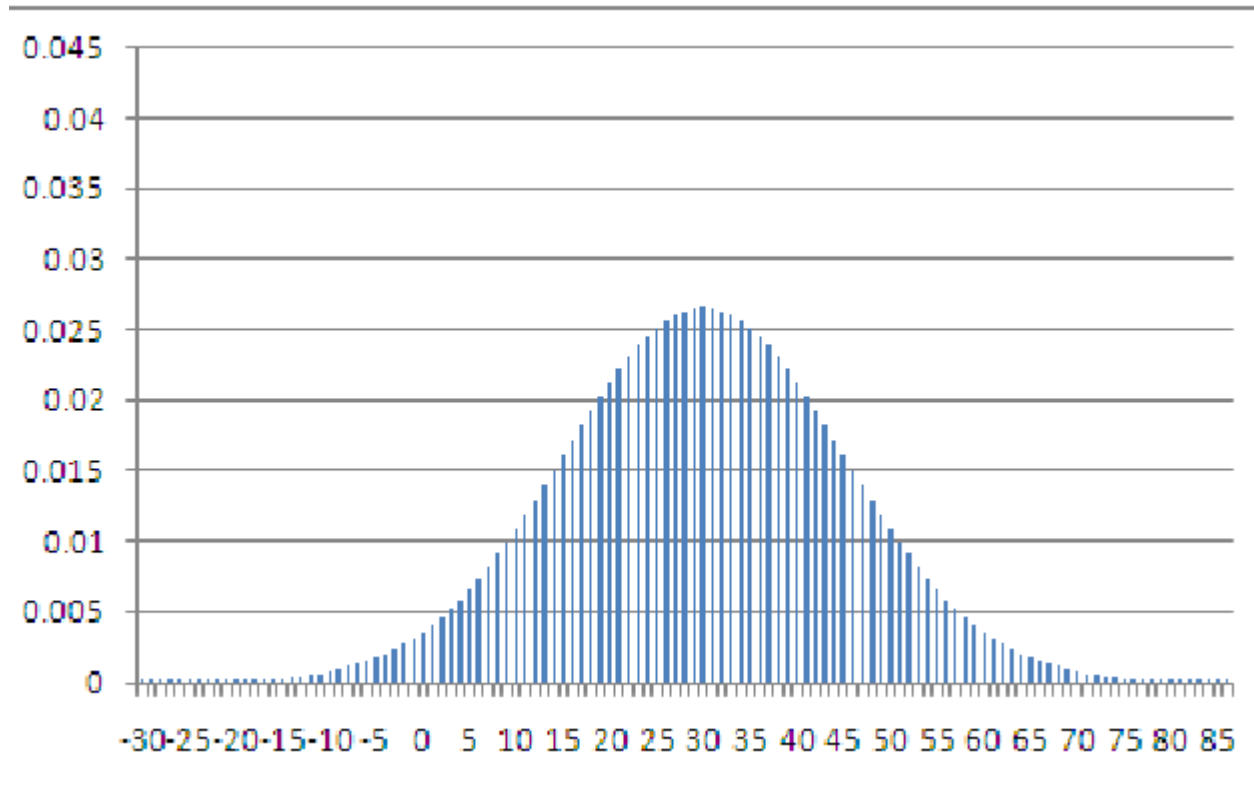
Coming Up

Continuous Distributions:

Standard Normal Distribution

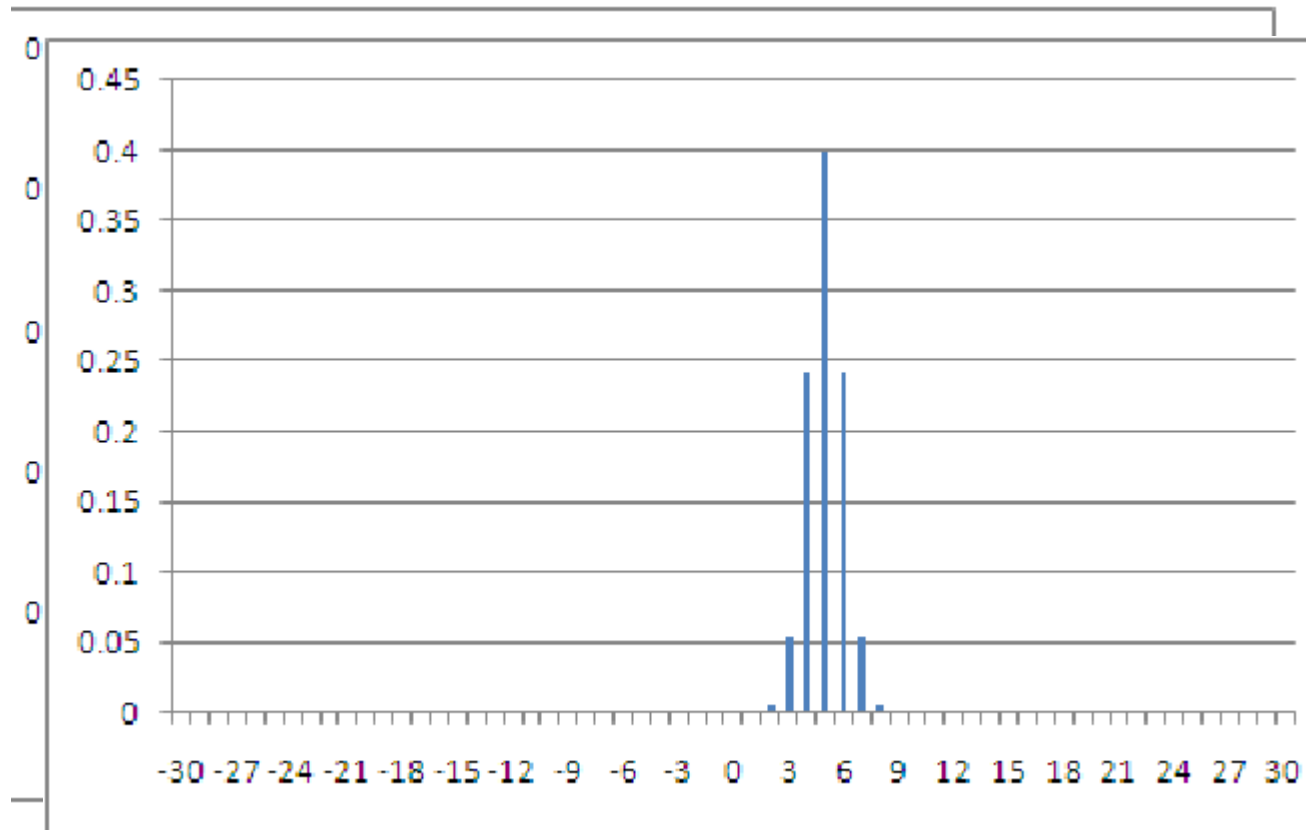
Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



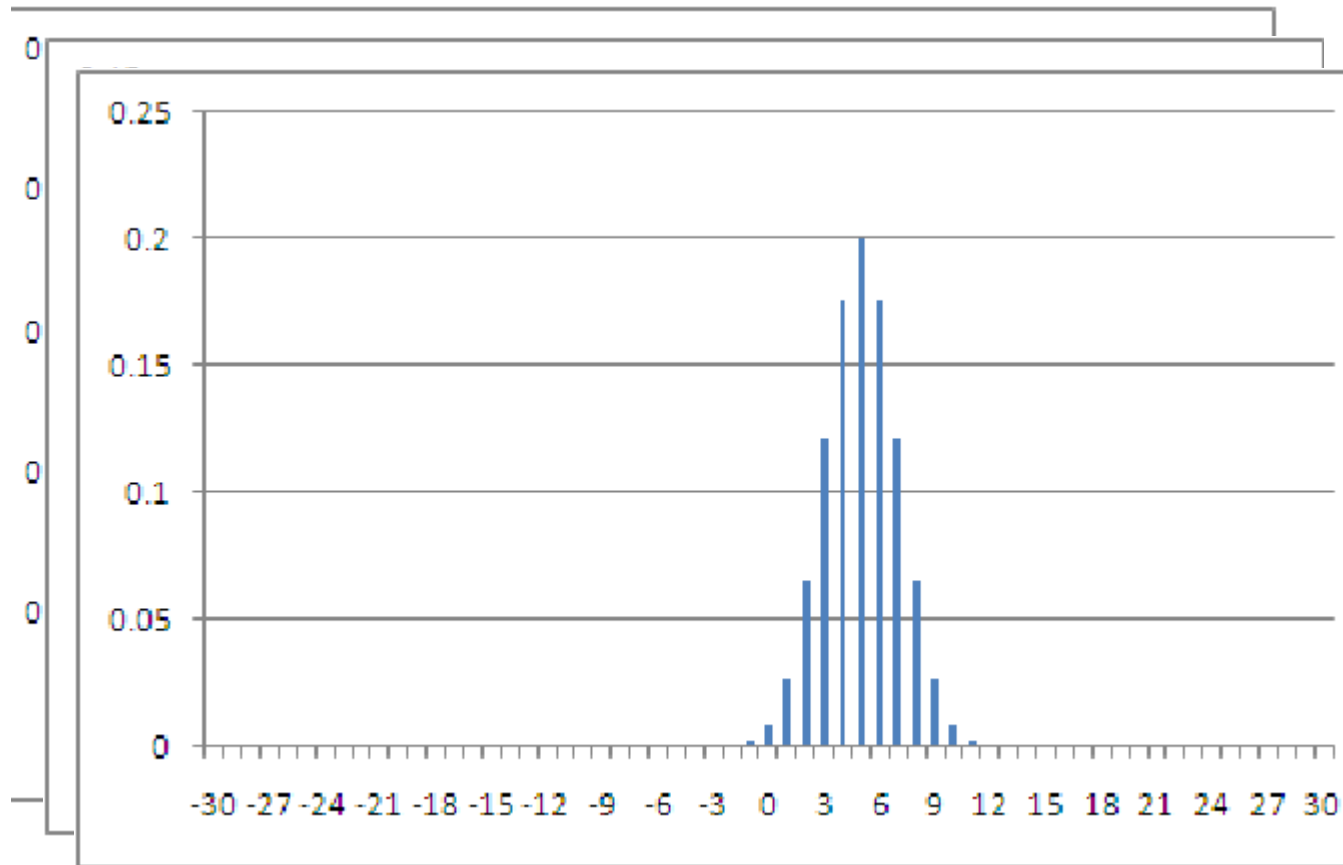
Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



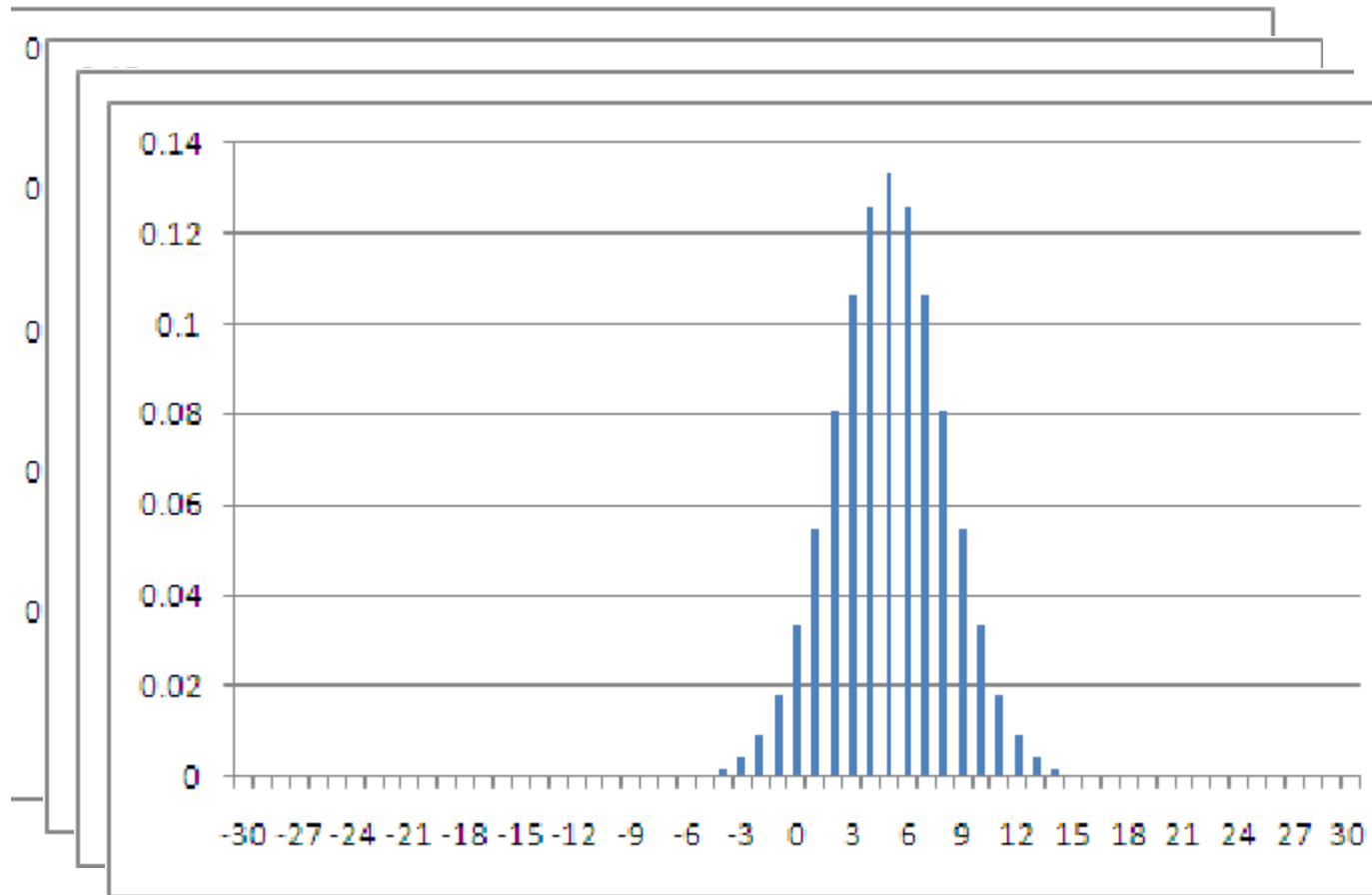
Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



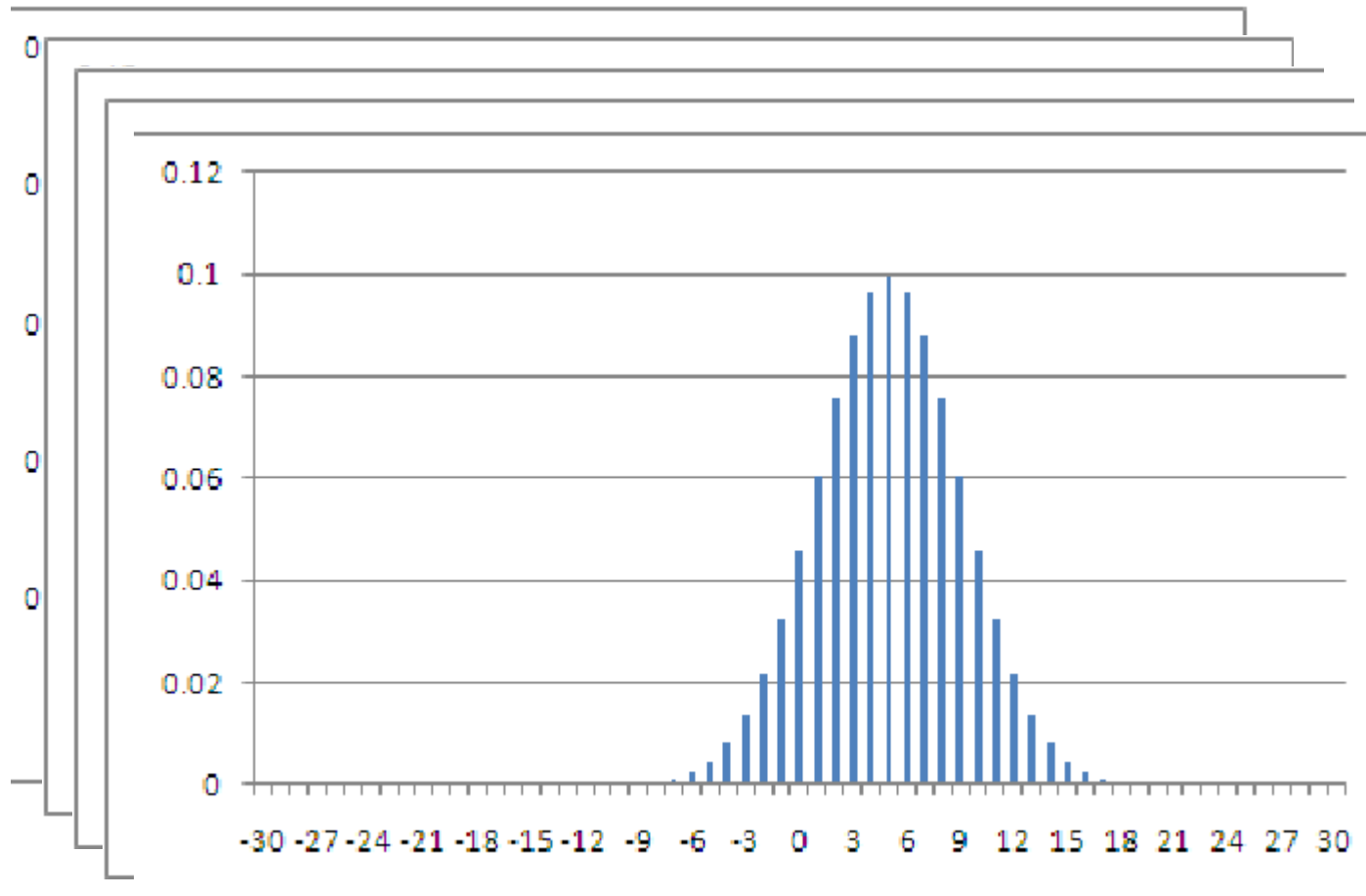
Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



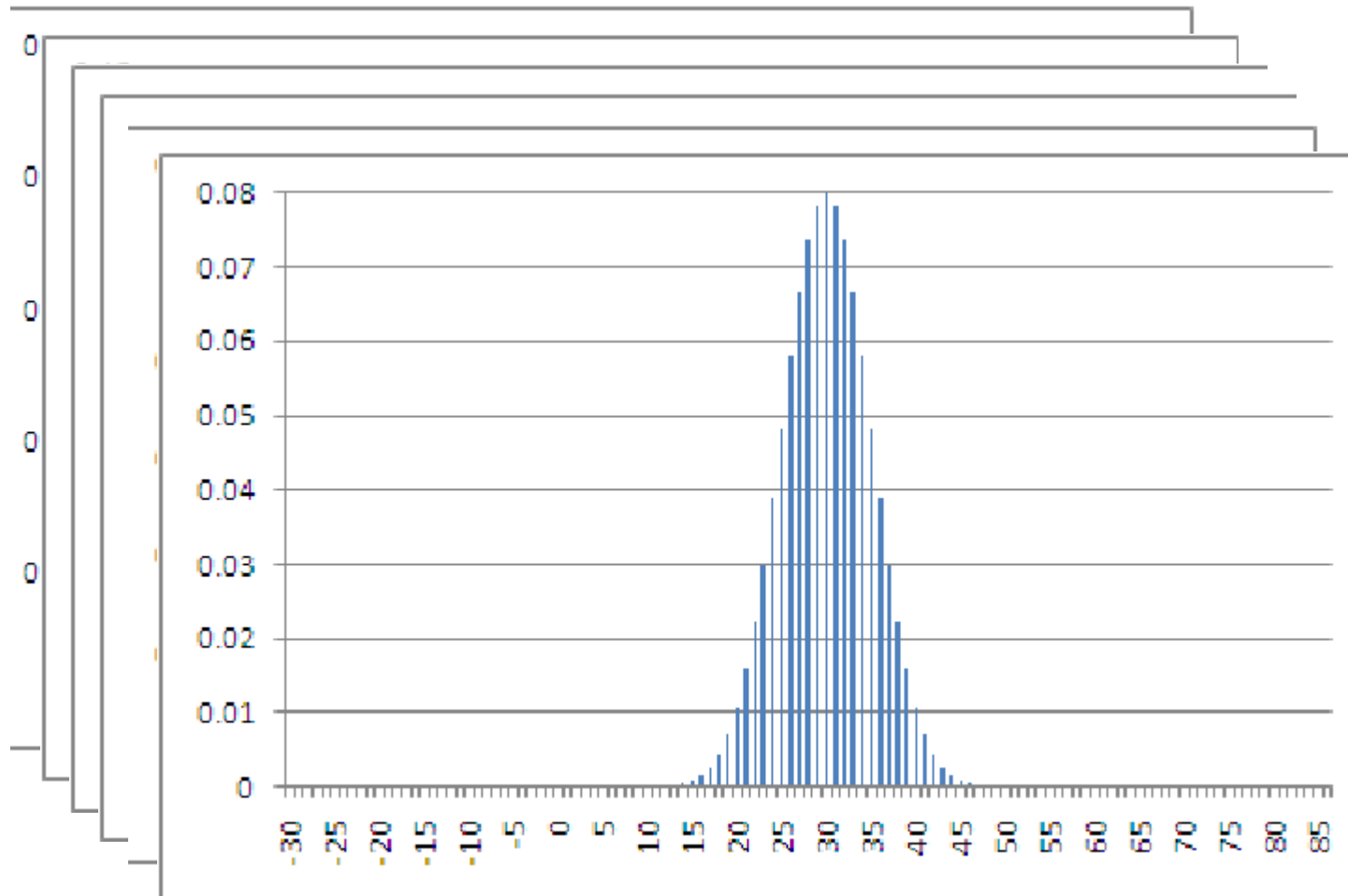
Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



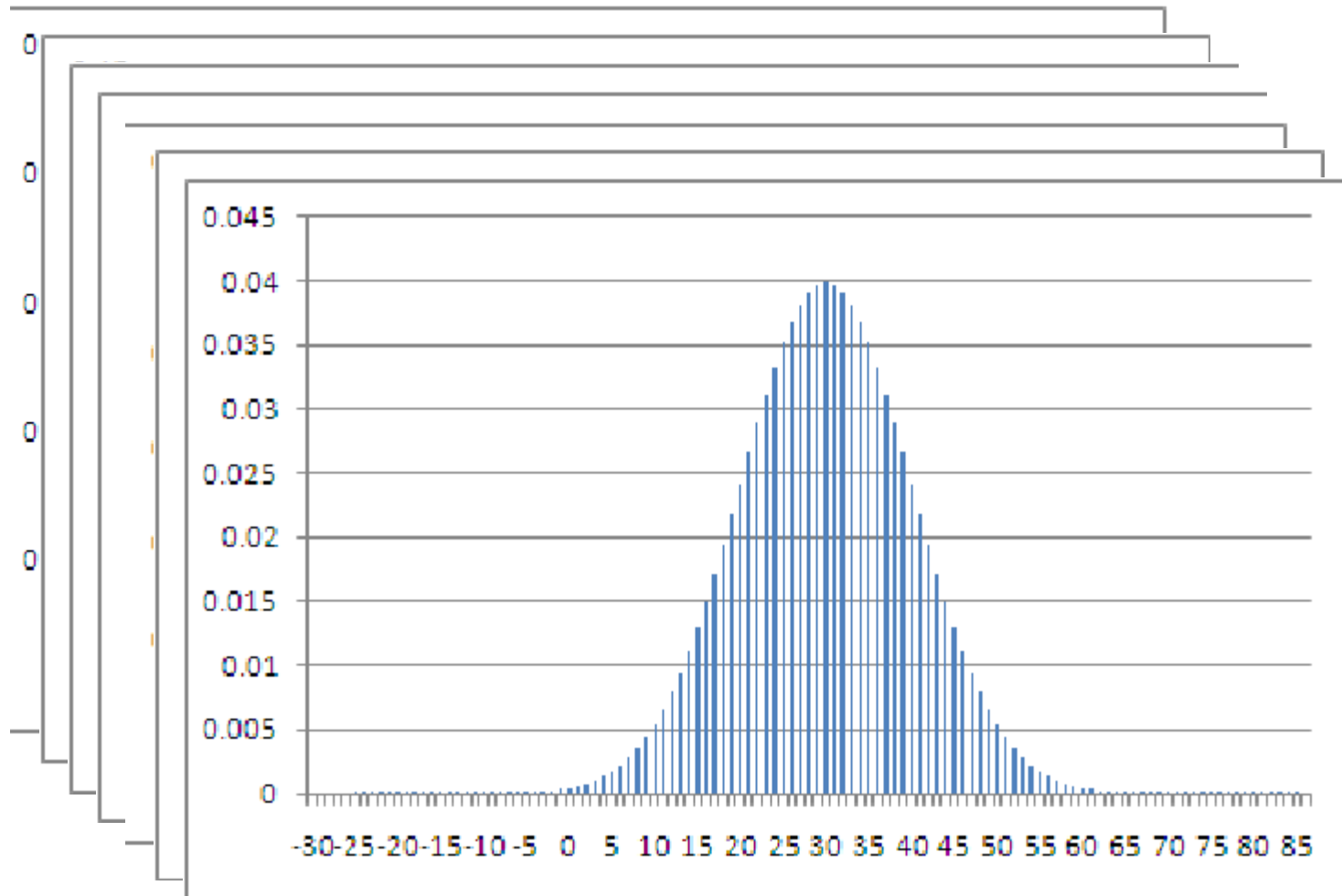
Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



Normal Distribution

We can compute the probabilities for any Mean, Std Dev and an X



Standard Normal Distribution

Supposing we were looking at three different normal distributions:



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)

Dist B (1890, 135)



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)

Dist B (1890, 135)

Dist C (0.26, 0.03)



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)

Dist B (1890, 135)

Dist C (0.26, 0.03)

We want to look at the probabilities of outcomes from each of these distributions as follows:



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)

Dist B (1890, 135)

Dist C (0.26, 0.03)

We want to look at the probabilities of outcomes from each of these distributions as follows:

A: Outcome $X \leq 210$



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)

Dist B (1890, 135)

Dist C (0.26, 0.03)

We want to look at the probabilities of outcomes from each of these distributions as follows:

A: Outcome $X \leq 210$

B: Outcome $X \leq 1620$



Standard Normal Distribution

Supposing we were looking at three different normal distributions:

Dist A (Mean = 250, Std Dev = 20)

Dist B (1890, 135)

Dist C (0.26, 0.03)

We want to look at the probabilities of outcomes from each of these distributions as follows:

A: Outcome $X \leq 210$

B: Outcome $X \leq 1620$

C: Outcome $X \leq 0.2$



Standard Normal Distribution

The probability of each of these outcomes = 0.023 – why?

		✕ ✓ f_x		=NORM.DIST(0.2,0.26,0.03,TRUE)		
	B	C	D	E	F	G
A		0.023				
B		0.023				
C		0.023				

Is it a coincidence?



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:

Outcome X = Mean – 2 * Std Deviations



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:

Outcome $X = \text{Mean} - 2 * \text{Std Deviations}$

a) $210 = 250 - (2 * 20)$



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:

Outcome $X = \text{Mean} - 2 * \text{Std Deviations}$

a) $210 = 250 - (2 * 20)$

b) $1620 = 1890 - (2 * 135)$



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:

Outcome $X = \text{Mean} - 2 * \text{Std Deviations}$

- a) $210 = 250 - (2 * 20)$
- b) $1620 = 1890 - (2 * 135)$
- c) $0.2 = 0.26 - (2 * 0.03)$



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:

Outcome X = Mean – 2 * Std Deviations

- a) $210 = 250 - (2 * 20)$
- b) $1620 = 1890 - (2 * 135)$
- c) $0.2 = 0.26 - (2 * 0.03)$

- In other words: each outcome is 2 std deviations less than its mean



Standard Normal Distribution

- There is a pattern in the outcome value relative to the mean and std deviation in each of the three distributions:

Outcome $X = \text{Mean} - 2 * \text{Std Deviations}$

- a) $210 = 250 - (2 * 20)$
- b) $1620 = 1890 - (2 * 135)$
- c) $0.2 = 0.26 - (2 * 0.03)$

- In other words: each outcome is 2 std deviations less than its mean
- For any normal distribution, an outcome X can be expressed in standardized units – number of its std deviations away from its mean



Standard Normal Distribution

Why is it standardized units?

We are converting any normal distribution into a normal distribution with a mean of 0, and a std deviation of 1

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



Standard Normal Distribution

What does this imply?



Standard Normal Distribution

What does this imply?

As long as something is the same deviation from any mean, p values will be the same



Standard Normal Distribution

What does this imply?

As long as something is the same deviation from any mean, p values will be the same

In other words,

If an outcome in a normal distribution is say + 1 std deviation away from its mean (irrespective of the actual values of mean and std dev), its probability will be the same



Standard Normal Distribution

- The standard normal distribution is a special case of the normal distribution with a mean of 0 and a std deviation of 1 (why is mean 0?)



Standard Normal Distribution

- The standard normal distribution is a special case of the normal distribution with a mean of 0 and a std deviation of 1 (why is mean 0?)
- All normal distributions can be converted to std normal by the following formula

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



Standard Normal Distribution

- The standard normal distribution is a special case of the normal distribution with a mean of 0 and a std deviation of 1 (why is mean 0?)
- All normal distributions can be converted to std normal by the following formula

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

Where X is the value of the random variable in the original normal distribution, μ is the mean, and σ is the std. deviation of the original normal distribution



Standard Normal Distribution

- The standard normal distribution is a special case of the normal distribution with a mean of 0 and a std deviation of 1 (why is mean 0?)
- All normal distributions can be converted to std normal by the following formula

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

Where X is the value of the random variable in the original normal distribution, μ is the mean, and σ is the std. deviation of the original normal distribution

$$y_z = \frac{1}{\sigma\sqrt{2\pi}} e^{\frac{-z^2}{2}}$$

where: y_z = vertical height on the standard normal distribution

z = as defined above



Standard Normal Distribution

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986



Standard Normal Distribution

IQ Example:

Average IQ scores for Jigsaw students is 108, std deviation is 7.



Standard Normal Distribution

IQ Example:

Average IQ scores for Jigsaw students is 108, std deviation is 7.

1. What is the probability that a random student will have a score < 120 ?



Standard Normal Distribution

IQ Example:

Average IQ scores for Jigsaw students is 108, std deviation is 7.

1. What is the probability that a random student will have a score < 120 ?
2. What is the probability that a random student will have a score between 110 and 115?



Standard Normal Distribution

IQ Example:

Average IQ scores for Jigsaw students is 108, std deviation is 7.

1. What is the probability that a random student will have a score < 120 ?
2. What is the probability that a random student will have a score between 110 and 115?
3. What is the probability that a random student will have a score of < 105 ?



Standard Normal Distribution

What is the probability that a random student will have a score < 120 ?

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986




Standard Normal Distribution

What is the probability that a random student will have a score < 120 ?

$$Z = (120 - 108)/7 = 1.71$$

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986



Standard Normal Distribution

Probability that a random student will have a score between 110 and 115?

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986




Standard Normal Distribution

Probability that a random student will have a score between 110 and 115?

$Z = \text{between } (110-108)/7 \text{ and } (115-108)/7 = \text{between } 0.28 \text{ and } 1$

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986



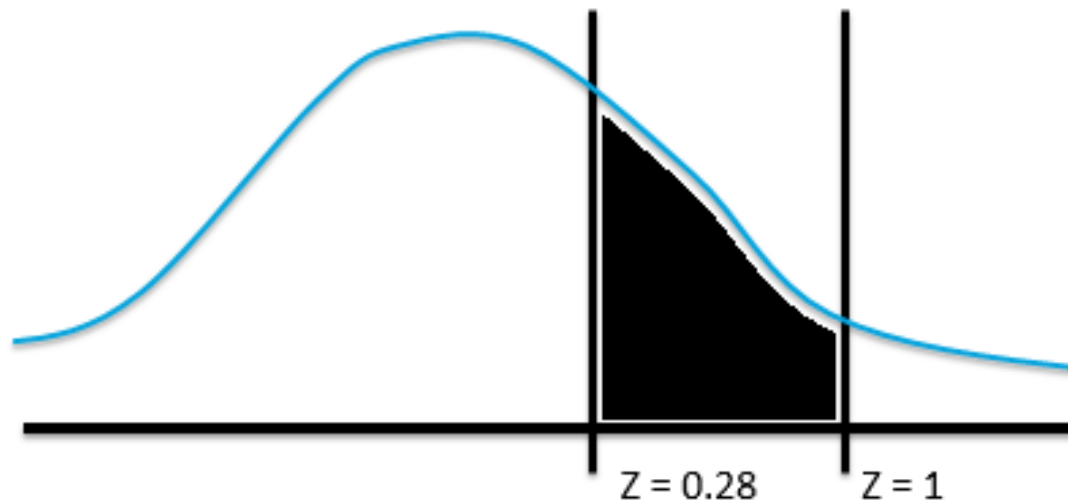
Standard Normal Distribution

What is the probability a random student will have a score between 110 and 115?

Based on the table:

$$P(z < 1) = 0.8413$$

$$P(z < 0.28) = 0.6103$$



Standard Normal Distribution

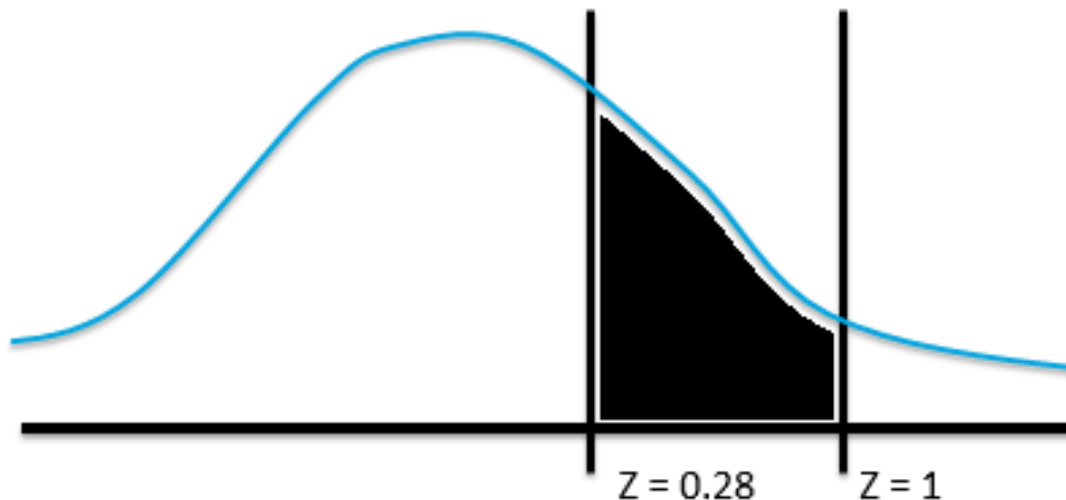
What is the probability a random student will have a score between 110 and 115?

Based on the table:

$$P(z < 1) = 0.8413$$

$$P(z < 0.28) = 0.6103$$

$$\text{So } P(z \text{ between } 0.28 \text{ and } 1) = 0.8413 - 0.6103 = \mathbf{0.23}$$



Standard Normal Distribution

What is the probability that a random student will have a score of < 105 ?

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986




Standard Normal Distribution

What is the probability that a random student will have a score of < 105 ?

$$Z = (105 - 108) / 7 = -0.42$$

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986



Standard Normal Distribution

$$P(z < -0.42) = p(z > 0.42)$$

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986



Standard Normal Distribution

$$P(z < -0.42) = p(z > 0.42)$$

$$p(z > 0.42) = 1 - p(z < 0.42) = 1 - 0.6628$$

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986




Standard Normal Distribution

$$P(z < -0.42) = p(z > 0.42)$$

$$p(z > 0.42) = 1 - p(z < 0.42) = 1 - 0.6628$$

$$\text{So, } p(\text{score} < 105) = 0.3372$$

Tables of the Normal Distribution



Probability Content
from $-\infty$ to Z

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986



Probability Distributions

Case Study Examples





Recap

- What does Statistics cover?

Recap

➤ What does Statistics cover?

- Summary Statistics
- Inferential Statistics

Recap

- What does Statistics cover?
 - Summary Statistics
 - Inferential Statistics
- Sample v/s Population

Recap

- What does Statistics cover?
 - Summary Statistics
 - Inferential Statistics
- Sample v/s Population
- Probability Theory

Recap

- What does Statistics cover?
 - Summary Statistics
 - Inferential Statistics
- Sample v/s Population
- Probability Theory
- Probability Distribution Concepts

Recap

- What does Statistics cover?
 - Summary Statistics
 - Inferential Statistics
- Sample v/s Population
- Probability Theory
- Probability Distribution Concepts
- Types of Distributions

Recap

- What does Statistics cover?
 - Summary Statistics
 - Inferential Statistics
- Sample v/s Population
- Probability Theory
- Probability Distribution Concepts
- Types of Distributions
 - Discrete

Recap

- What does Statistics cover?
 - Summary Statistics
 - Inferential Statistics
- Sample v/s Population
- Probability Theory
- Probability Distribution Concepts
- Types of Distributions
 - Discrete
 - Continuous