

MM104/MM106/BM110

Statistics and Data Presentation

Lecture 5:

Sampling distributions

Binomial Distribution

Distribution of the sample proportion

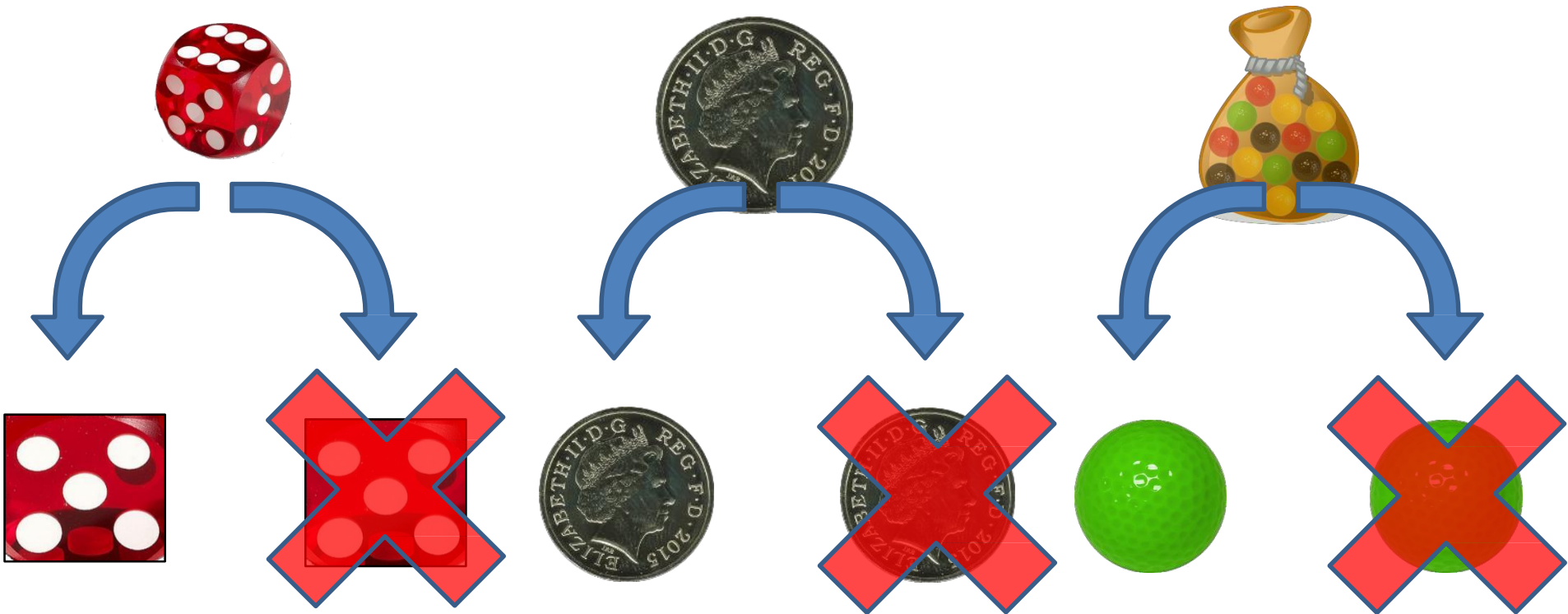
Chris Robertson

Binomial distribution

Binomial Random Variables

It describes cases when each attempt (*trial*) can have only *two* outcomes: success or failure.

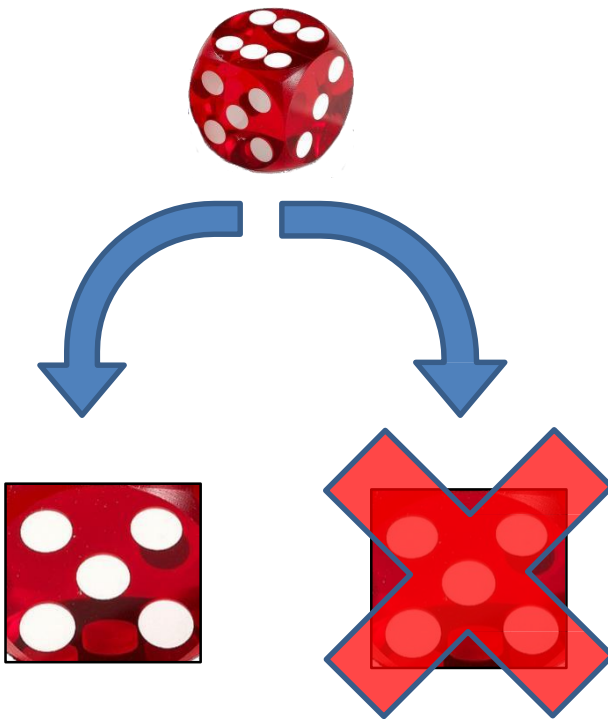
Discrete Random Variable



Number of successes \longrightarrow discrete random variable!!

Binomial Random Variables

Rolling a die and counting if you roll a 5 or not
5 – success; 1,2,3,4,6 - fail

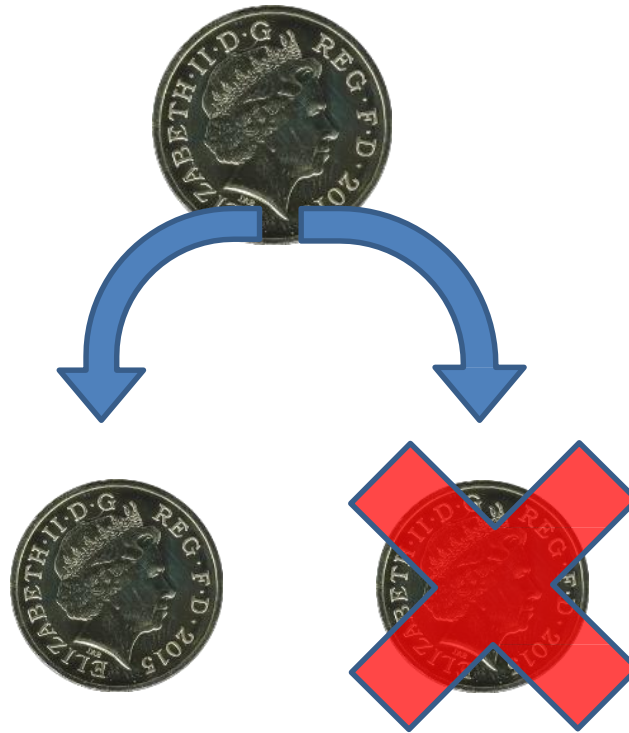


Number of successes \longrightarrow discrete random variable!!

Binomial Random Variables

Tossing a coin

Head – success; Tail - fail

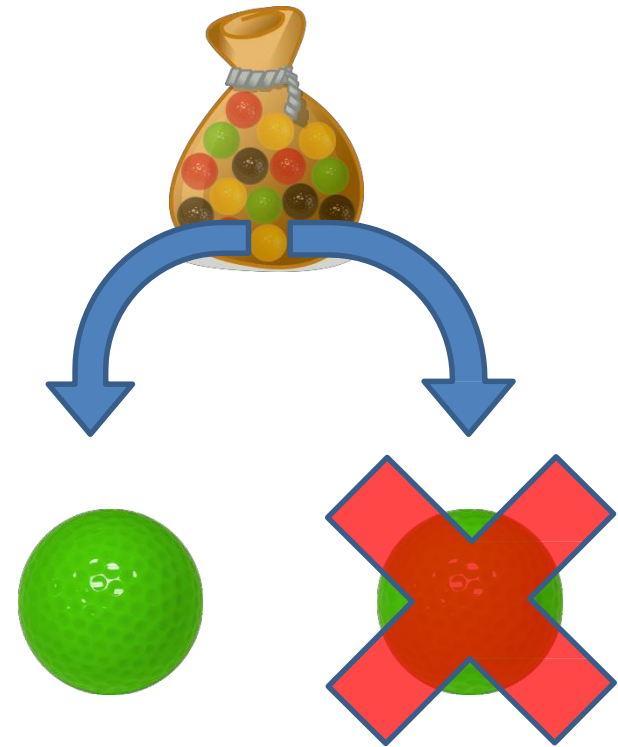


Number of successes \longrightarrow discrete random variable!!

Binomial Random Variables

Drawing a coloured ball from a bag

Green Ball – success; any other colour - fail



Number of successes \longrightarrow discrete random variable!!

Binomial Random Variables

- Distribution associated with X , the number of “successes” in n independent trials, each having the same probability p of success.
- Many important examples,
 - elections (a trial is a vote, a success is “voting for candidate X ”).
 - Testing for a disease –
 - success have disease; failure no disease
 - Gambling – success win ; fail lose stake
- Mean of the distribution is np and standard deviation $\sqrt{np(1 - p)}$.

You'll learn the mathematical details next year!

Binomial Random Variables

Mean of the distribution is np and standard deviation $\sqrt{np(1-p)}$.

If you toss a coin 10 times and the probability of a tail is 0.5;

$$\mu = 10 \times 0.5 = 5$$
$$\sigma = \sqrt{10 \times 0.5 \times 0.5} = \sqrt{2.5} = 1.5811$$

If you play roulette 100 times and bet £1 on a single number each time the probability of winning is 1/37 (36 numbers plus 0 where the bank wins;

$$\mu = 100 \times \left(1/37\right) = 2.7$$
$$\sigma = \sqrt{100 \times \left(1/37\right) \times \left(36/37\right)} = \sqrt{2.6297} = 1.6216$$

Binomial Random Variables

- If the number of trials (n) is large then the distribution of the number of successes is symmetric and bell-shaped
- Binomial distribution converges to normal distribution for large n .
- Large means $np > 5$; $n(1 - p) > 5$
- So we can use the normal distribution to find probabilities for proportions in large samples

The exact distribution is known as the binomial distribution – next year if studying Maths/Statistics

Binomial Random Variables

- Number of times, x , picking red ball out of $n=25$ trials (15 balls, out of which 6 are red).



— Success: picking a red ball.

— Success probability: $p = \frac{6}{15} = 0.4$

— Binomial, mean: $np = 25 * 0.4 = 10$

(Number of times I expect, on average,
to get a red ball out of the 25 trials)

— Standard deviation:

$$\sqrt{np(1 - p)} = 2.449$$

Example

- In a large company, the number of hours that full-time employees work in any week is known to be Normally distributed with $\mu = 30h$ and standard deviation $\sigma = 3.9h$. If there are 276 employees in the company, what is the mean number of employees being paid over-time each week (assuming payment of overtime is based on the rule outlined before i.e more than 31.5 hours per week)?

Success: finding one employee working overtime.

Probability of success: $p = P(X > 31.5) = 0.35026$

If the number of employees is $n=276$, the average number of times the “work overtime” event (i.e. a “success”) happens is:

$$np = 276 * P(x > 31.5) = 276 * 0.35026 = 93.672$$

Distribution of the sample proportion

Sampling distribution of sample proportion

If x is the number of individuals with a characteristic of interest within a sample of size n from the population, then

sample proportion $\hat{p} = \frac{x}{n}$

is a random variable – i.e. the proportion of the sample with the characteristic.

p is the population proportion – proportion in the population with the characteristic, which is estimated by \hat{p} , the sample proportion.

Sampling distribution of sample proportion

Example: Voters in a poll.

- From a population of n voters, number of voters for option 1, x , is like **number of successes** for option 1.

 Random variable!!

- The **proportion** of voters that voted for option 1 (i.e. proportion of successes), $\hat{p} = x/N$, is also a random variable.

Sampling distribution of sample proportion

- If you have binomial data with mean np and standard deviation $\sqrt{np(1-p)}$, where n is the sample size and p is the success probability in the population.
 - Then \hat{p} , the sample proportion, is a random variable, which is the proportion of successes.
 - If $np > 5$ **and** $n(1-p) > 5$ the sampling distribution for \hat{p} can be approximated by a **normal distribution** with mean p and standard error $\sqrt{p(1-p)/n}$.

i.e. the proportion of successes can be treated as a normal random variable!

Example

- Number of times, x , picking red ball out of $n=25$ trials (15 balls, out of which 6 are red, i.e. $\hat{p}=6/15=0.4$).



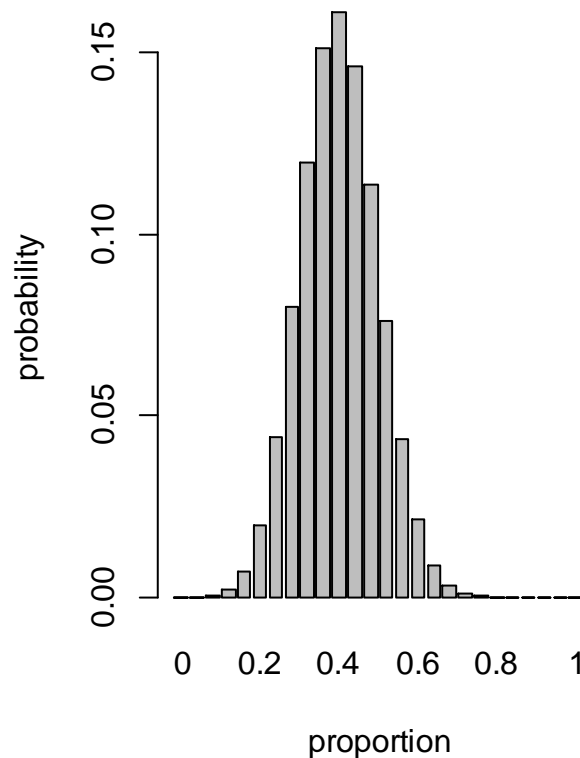
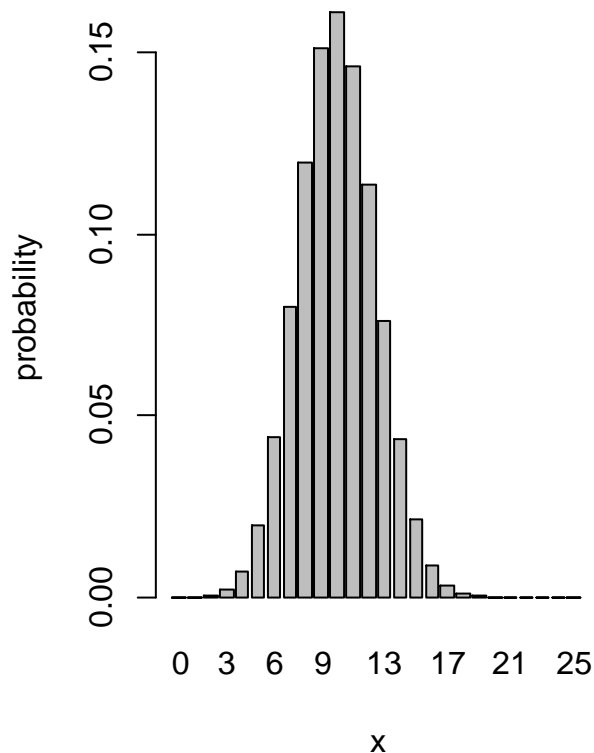
$$\begin{array}{c} \downarrow \\ n\hat{p} = 10, n(1 - \hat{p}) = 15 \\ \downarrow \end{array}$$

\hat{p}
APPROX.
NORMAL!!

— Sample proportion, \hat{p} , approx. normal with mean $\hat{p} = 0.4$ and standard error $\sqrt{\hat{p}(1 - \hat{p})/n} = \sqrt{0.4(1 - 0.4)/25} = 0.098$

Example Distributions

- Essentially the same distribution, but with horizontal axis rescaled:



Example – probability calculations

— In the Brexit vote, 48.1 % of the electorate voted for “Remain”.
On the assumption that allegiances have remained unchanged,
what is the probability that in a repeated vote

(a) a randomly chosen elector will vote for “Remain”?

(b) Fewer than 50% of a sample of 300 voters
will vote for “Remain”?

Example – probability calculations

— In the Brexit vote, 48.1 % of the electorate voted for “Remain”.
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what is the probability that in a repeated vote

(a) a randomly chosen elector will vote for “Remain”?

(b) Fewer than 50% of a sample of 300 voters
will vote for “Remain”?

This bit of the question is asking
you to calculate the probability
about a proportion in the sample
– even though it is expressed in
terms of percentage

Percentage 0-100

Proportion 0-1

Percentage = $100 \times \text{Proportion}$

This bit is just asking you
for the population
proportion from the
question. It is the
population as the vote
was the whole population
of voters.

Example – probability calculations

— In the Brexit vote, 48.1 % of the electorate voted for “Remain”.
On the assumption that allegiances have remained unchanged,
what is the probability that in a repeated vote

(a) a randomly chosen elector will vote for “Remain”? Success prob. $p = 0.481$

(b) Fewer than 50% of a sample of 300 voters
will vote for “Remain”?

$np > 5$ and $n(1 - p) > 5$
(so can use the normal
distribution to get the
probability)

i.e. $P(\hat{p} < 0.50)$

\hat{p} variable: fraction
votes for “Remain”

$$\sigma = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} = 0.0288$$

$\hat{p} = 0.481$, Mean (μ)

Standard deviation

Example – probability calculations

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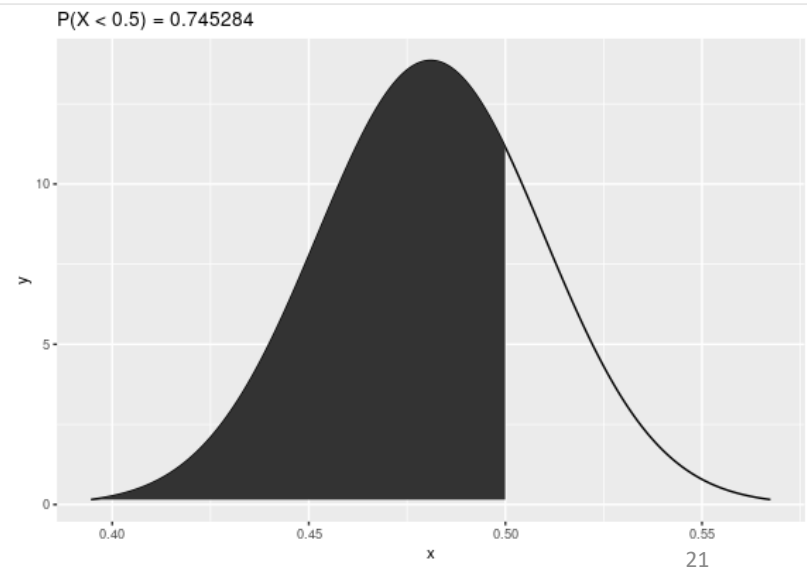
Normal t

Tail
☒ Lower
☐ Upper
☐ Both

x
0.5

Mean
0.481

sd
0.0288



Example – probability calculations

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$\hat{p} = 0.481,$

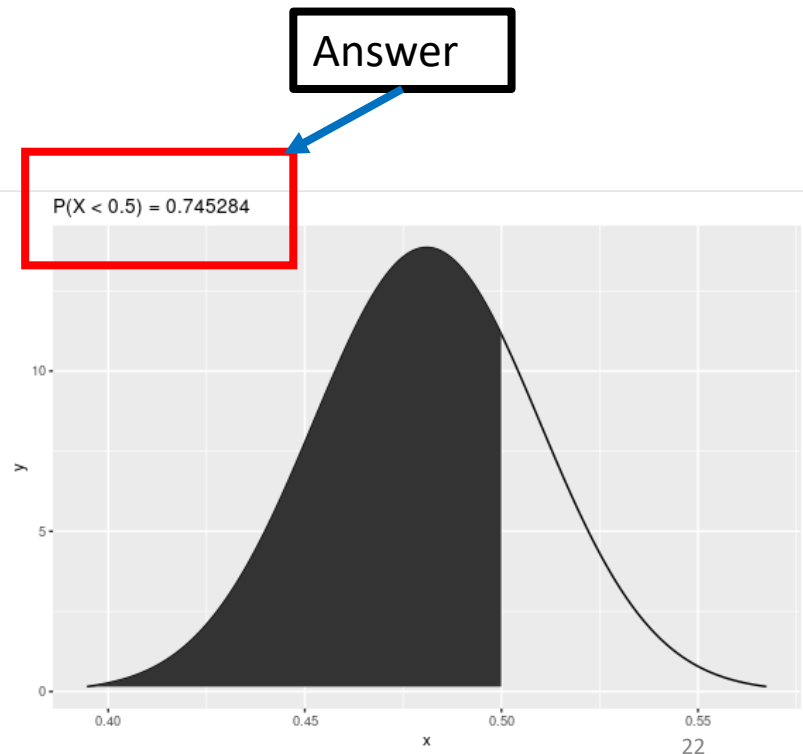
Normal **t**

Tail
☒ Lower
☐ Upper
☐ Both

x
0.5

Mean
0.481

sd
0.0288



Example – Change the sample size

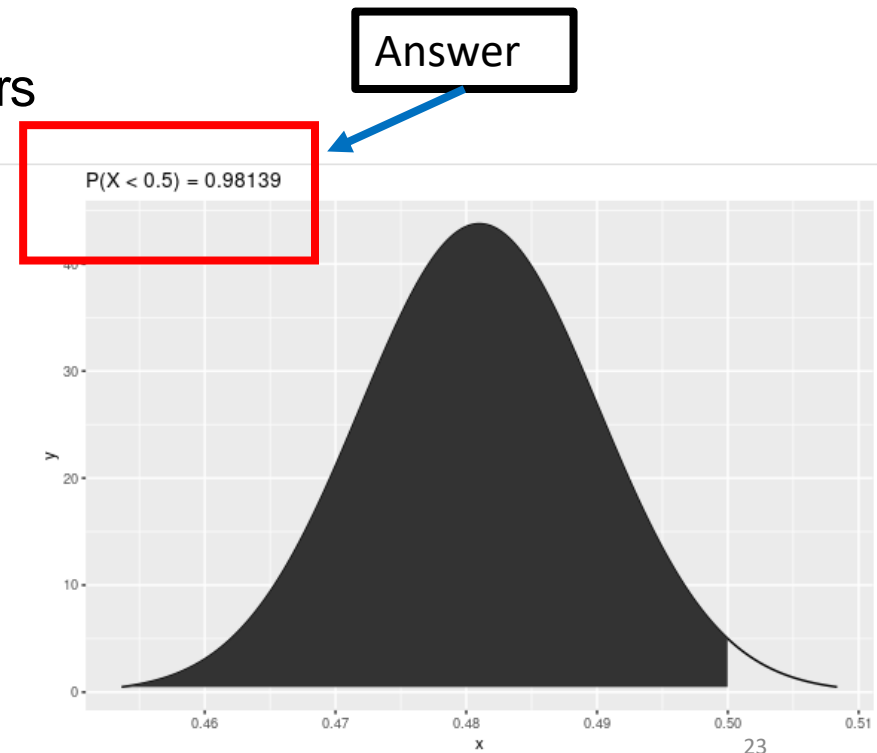
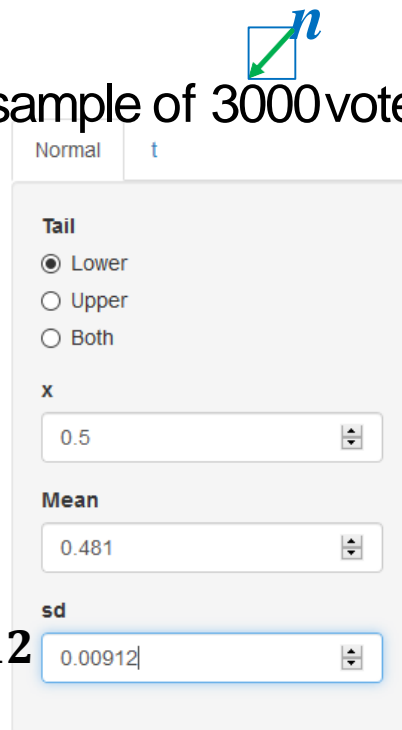
— In the Brexit vote, 48.1 % of the electorate voted for “Remain”.
On the assumption that allegiances have remained unchanged,
what is the probability that in a repeated vote

(a) a randomly chosen elector will vote for “Remain”?

(b) Fewer than 50% of a sample of 3000 voters
will vote for “Remain”?

i.e. $P(\hat{p} < 0.50)$

$$\sigma = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} = 0.00912$$

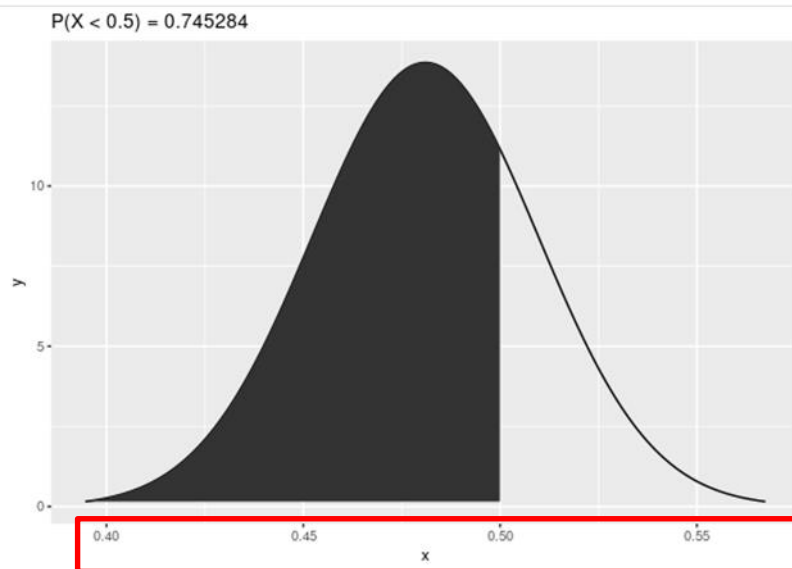


Example – Increasing Sample Size

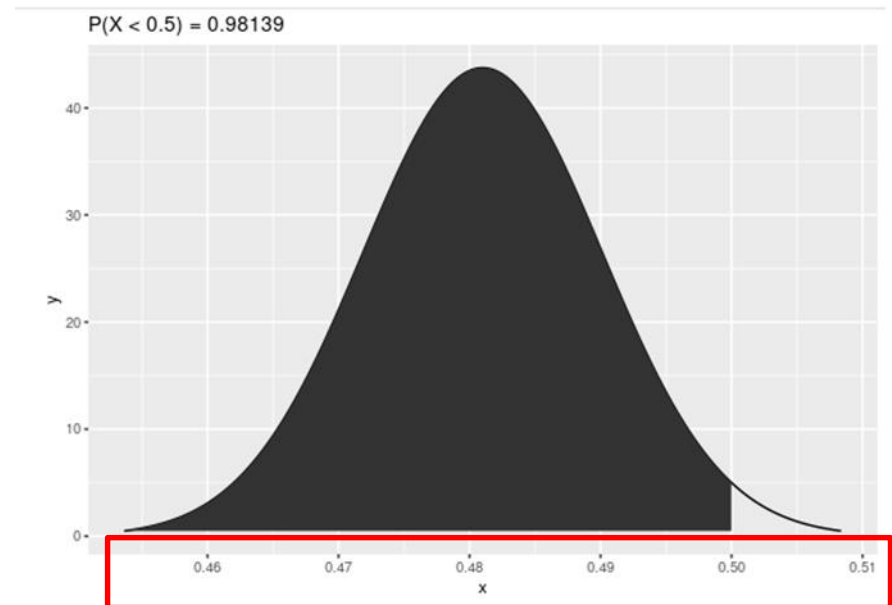
STANDARD ERROR DECREASES WITH INCREASING SAMPLE SIZE

$$\hat{p} = 0.481,$$
$$\sigma = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} = 0.0288$$

$$\hat{p} = 0.481,$$
$$\sigma = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} = 0.00912$$



$n = 300$



$n = 3000$

Note the x axis scales on the graphs – there is much less of a range on the right.

Sampling distributions

Probability calculation

- **Example:**

- A factory production line has been producing components with 14% of them failing the acceptance test, and getting scrapped. A random sample of 700 components from the line is taken for test. If the defectives rate is still 14%, what is the probability that 16.4% or more of components in the sample are defective?

“Success” prob. $\hat{p} = 0.14$

Sample size $n = 700$

$$n\hat{p} > 5 \text{ and}$$

$$n(1 - \hat{p}) > 5$$

(Approx. Normal)

(new variable=fraction defectives in the sample)

$$\sigma = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} = 0.01311$$

Calculate
 $P(X > 0.164)$?

Sampling distributions

Probability calculation

- **Example:**

- A factory production line has been producing components with 14% of them failing the acceptance test, and getting scrapped. A **random sample of 700 components** from the line is taken for test. If the defectives rate is **still 14%**, what is the **probability that 16.4% or more of components** in the sample are defective?

$$P(X > 0.164) ?$$

“Success” prob. $\hat{p} = 0.14$

Sample size $n = 700$

$$\sigma = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} = 0.01311$$

Normal ☒ t

Tail

☐ Lower

☒ Upper

☐ Both

x

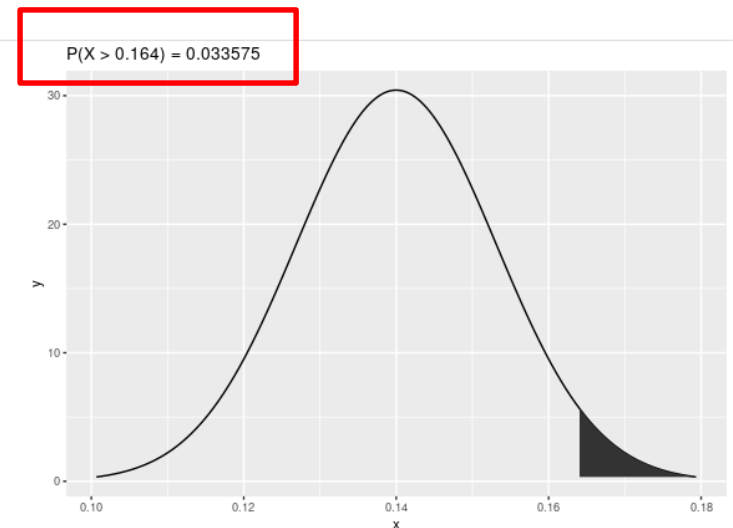
0.164

Mean

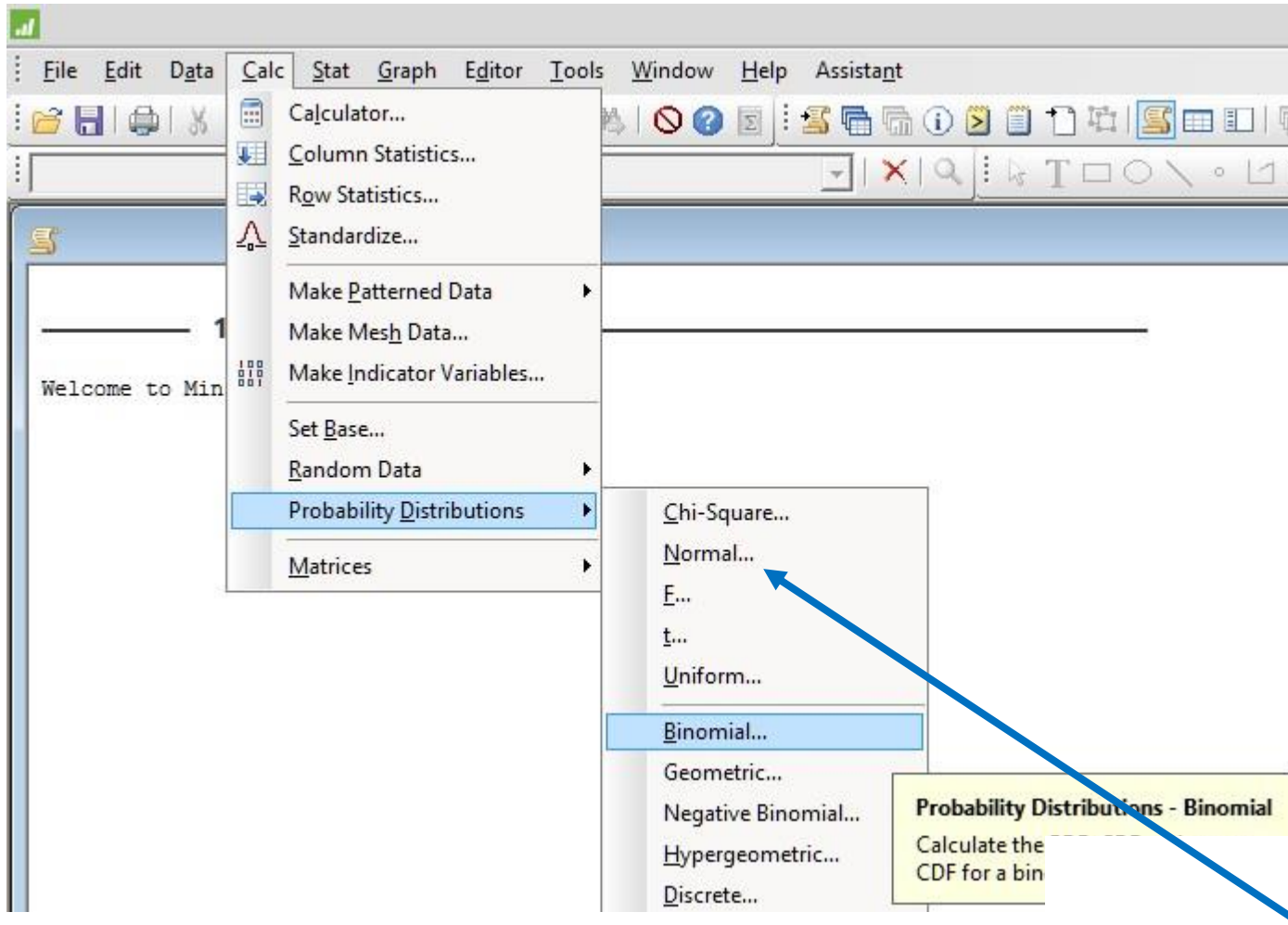
0.14

sd

0.01311



Probability calculations in Minitab



You can use Minitab to make probability distribution calculations if you wish – you do not have to use the tables in myplace

There are also many online calculators.

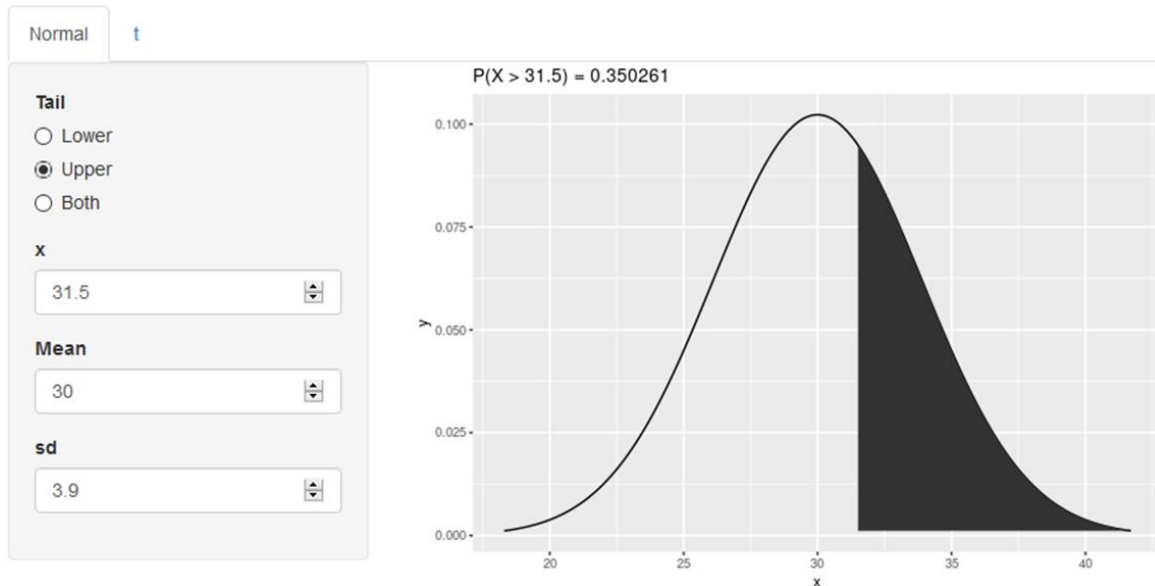
Pick Normal

Probability calculations in Minitab

- Examples:

Normal, $\mu = 30$ and $\sigma = 3.9$

$P(X < 31.5)$?



The dialog box for the Normal Distribution calculator is shown. It has the following settings:

- Probability density:** ☐
- Cumulative probability:** ☒
- Inverse cumulative probability:** ☐
- Mean:** 30
- Standard deviation:** 3.9
- Input column:** (empty)
- Optional storage:** (empty)
- Input constant:** 31.5
- Optional storage:** (empty)

 Buttons for 'Select', 'Help', 'OK', and 'Cancel' are visible at the bottom.

$$P(X < 31.5) = 1 - P(X > 31.5)$$

$$= 1 - 0.350261 = 0.649739$$

Cumulative Distribution Function

Normal with mean = 30 and standard deviation = 3.9

x	P(X ≤ x)
31.5	0.649739

Key Points

- The sampling distribution of the sample proportion follows a normal distribution in large samples
- The mean of the sampling distribution of the sample proportion is the population proportion

$$\mu_{\hat{p}} = p$$

- The standard deviation of the sampling distribution is known as the standard error of the sample mean

$$\sigma_{\hat{p}} = \sqrt{\frac{p(1 - p)}{n}}$$

- Large means $np > 5$; $n(1 - p) > 5$