

Critical evaluation

After completing this tutorial, you will be able to:

- Assess the validity and usefulness of clinical trials.
- Describe the meaning of some common terms used to quantify the benefits and harms of a medicine.
- Outline how evidence from clinical trials is evaluated and applied by NHS prescribing committees.

Why this subject matters...

As a new hospital pharmacist you probably won't be undertaking detailed critical evaluation of trials or systemic reviews every day. However, you will be weighing up evidence from all sorts of different sources such as the BNF, journals, national guidelines or online discussion forums to help you to make decisions. You need the best quality evidence to help you care for individual patients, update your practice, or write a guideline. So, you must be able to assess the validity and usefulness of different types of data with confidence.

Definition

What do you think we mean by the term critical evaluation (sometimes known as critical appraisal)? One definition is that it is a process of carefully and systematically examining research to judge whether the data are **valid** and **useful**.

We assess **validity** by considering how robust the data are: is the clinical trial scientifically sound?

We can evaluate **usefulness** by thinking about what the results mean for our patients in the real world.



Most of this tutorial is devoted to examining validity and usefulness in more detail.

Validity and bias

To evaluate validity we need to consider whether the research was done properly. All possible measures should be used to reduce the risk of **bias** in a clinical trial. The methods section of a paper should explain exactly what steps have been taken and the results should be fully and unambiguously reported.

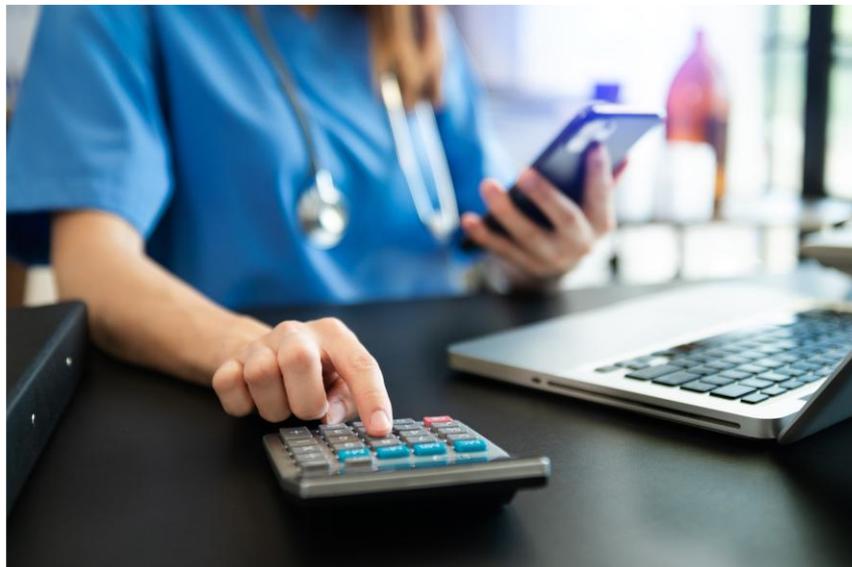
There are several useful tools available to help you assess the validity of a trial including;

- [SPIRIT-CONSORT 2025 Checklists](#)
- [Critical Appraisal Skills Programme \(CASP\) Randomised Controlled Trial Checklist](#)

For simplicity, the following points describe a trial of placebo (the control) versus drug treatment, but they also apply to trials that compare drug treatments (e.g. drug A versus drug B).

- The number of participants (**sample size**) needs to be planned carefully. It's a balancing act between how many participants can be enrolled considering ethical, financial and time constraints, but making sure there are enough so the study is adequately '**powered**'.

The **power** of a trial is the likelihood that it will detect a difference between 2 groups when one genuinely exists. The ideal power for a trial is at least 80%. This means that if the trial was repeated 100 times a statistically significant treatment effect would be seen in 80 of them. The power of a study increases with sample size.



A **power calculation** is essential for determining the right number of participants and ideally it should be presented in the final published paper. It should specify the primary endpoint (the main result being measured) and account for the expected outcome. If the difference between treatment and placebo is expected to be small, a larger sample size will be needed.

The calculation will also need include an allowance for participants who might **drop out**. For example: 'To detect a reduction in hospital stay of 3 days with a power of 80%, we calculated a sample size of 75 patients per group was needed, given an anticipated dropout rate of 10%'.

Trials that are **underpowered** or have a small sample size can still be useful, but their results might not be definitive. Larger trials or a meta-analysis may be needed to be more confident in the results. We also need to watch out for trials that are **overpowered**. If the sample size is too large, even very small effects of little clinical importance can be reported as statistically significant (see p-values below). This can give a misleading impression of a treatment's effectiveness.

- **Selection** of participants for a trial should be **random**. This stops the researcher from choosing their preferred patient population, which could make the results look better than they are. For example, a researcher may ask every third patient that comes to a clinic to participate, rather than choosing the ones who they think will respond well.



- **Allocation** of volunteers to placebo and treatment groups should be **concealed** from the researchers. This is a different concept to 'blinding'. It is recommended for all trials, including unblinded (open-label) trials. It prevents selection bias by ensuring the researchers do not influence which patients get the study treatment. For example, early studies of diphtheria vaccine showed a higher death rate in patients in the vaccine group than in the placebo group. This was because the sickest patients were chosen to receive the vaccine and the healthier patients were given a placebo. The best way of ensuring allocation concealment is to use a **centralised** service, where randomisation is carried out independently at a site away from the trial location (e.g. hospital pharmacy).
- Ideally as many people as possible involved in the trial should be **'blinded'** (or masked) to whether volunteers are receiving placebo or treatment. The opposite is **'open-label'**, when everyone knows what the volunteer is receiving. **'Double-blinded'** usually means

the investigators and the volunteers do not know which arm of the study each volunteer is in, and **'triple-blinded'** means the committee monitoring the data also do not know.

However blinding is not always possible – for example with drugs that cause distinct side effects which may make it easier for the participant or researcher to figure out what is being taken (e.g. peppermint oil capsules cause rectal burning) or if the treatment has a complicated dosage regimen (e.g. warfarin dosed according to INR results). One way around this is to use something called a **'PROBE'** design: prospective, randomised, open-label, blinded endpoint evaluation where the people doing the evaluation of the endpoints do not know which group the volunteers have been assigned to.

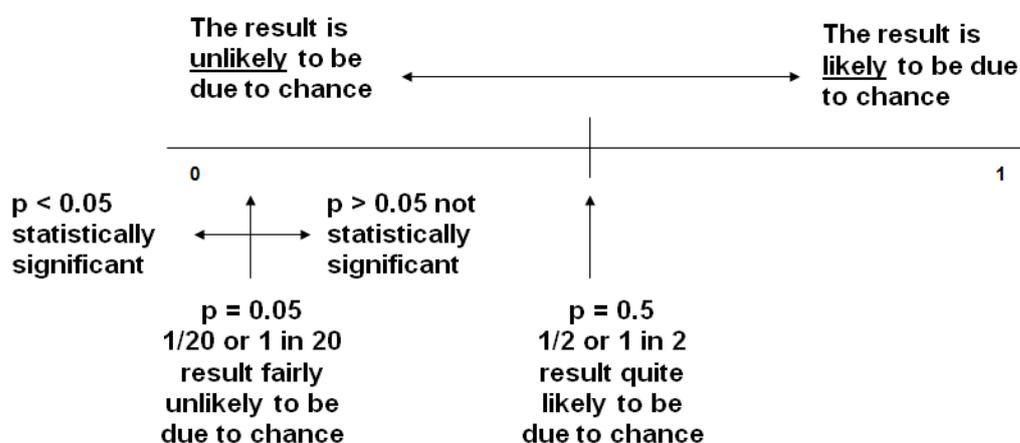
- The **baseline characteristics** of the groups under study should be as similar as possible. This helps to ensure that any effect seen in the treatment group is due to the treatment and not to pre-existing differences between the groups. When the baseline characteristics of the groups are very similar, it is a good sign that allocation to groups was truly random. The demographics of the groups should be clearly described in the study paper.
- Apart from the treatment or placebo, patients should be **treated identically** during the trial; they should receive the same number of blood tests, X-rays, and clinic appointments etc.
- **Participant flow** should be clearly reported, showing exactly what happened to each one. A good report should explain if and why volunteers did not receive the treatment allocated, or if they were lost to follow-up, dropped out or were excluded after the trial began. If this leads to imbalances between the groups, it is known as attrition bias. It is also important to know which and how many trial participants were included in the final analysis. There are two main ways to analyse the data: **'on-treatment'** or **'per protocol'** analysis where only those available for follow up are included, or **'intention-to-treat' (ITT)** analysis where all participants who underwent randomisation are included in the groups they were originally assigned to, no matter what happened during the trial. ITT analysis is generally favoured because it reduces bias and is more like real life, where people change their minds, or change or stop treatments. This gives a more realistic picture of the treatment's effectiveness.

Validity and chance

In critically evaluating a paper you also need to ask whether the results of a trial are valid or whether they occurred by chance. Statistical tests are used to assess this. The most commonly encountered terms are p-values and confidence intervals.

A **p-value** is the probability that a difference will be seen between two interventions in a trial when, in fact, there is no actual difference between the two interventions. In other words, it's an indication of whether the result occurred by chance. Probability is measured on a scale of 0 to 1 where an impossible event is given 0 and an event that is certain to happen is given 1. In drug trials, by convention, $p < 0.05$ is regarded as being statistically significant. It means that there is a less than 1 in 20 chance that you have observed a difference between your study drug and placebo when there is no actual difference between them.

The p-value – could the result have occurred by chance?



Adapted from original courtesy of The Critical Appraisal Skills Programme (CASP) www.casp-uk.net

When evaluating trial data, it's important not to rely solely on p-values, but to consider whether the results are important. For instance, a trial might show that an antihypertensive drug improved blood pressure readings by 2mmHg per year, but would this be clinically important, even if it was statistically significant?

P-values are easily misinterpreted, and can be overtrusted and misused. The threshold of 0.05 to claim statistical significance is questionable, and many experts would advocate use of a lower threshold, e.g. 0.005.

It's also important to realise that p-values depend on the sample size and don't consider the size of an effect or its clinical relevance. So the effect may be small and clinically unimportant, but the p-value can still be "significant" if the sample size is large. On the other hand, an effect can be large, but fail to meet the $p < 0.05$ criterion if the sample size is small.

We also need to consider that p-values are based only on data from a sample of people, and the results you get for that sample may not be the results you would get with a different sample.

Because of these limitations we should look at other statistical values.

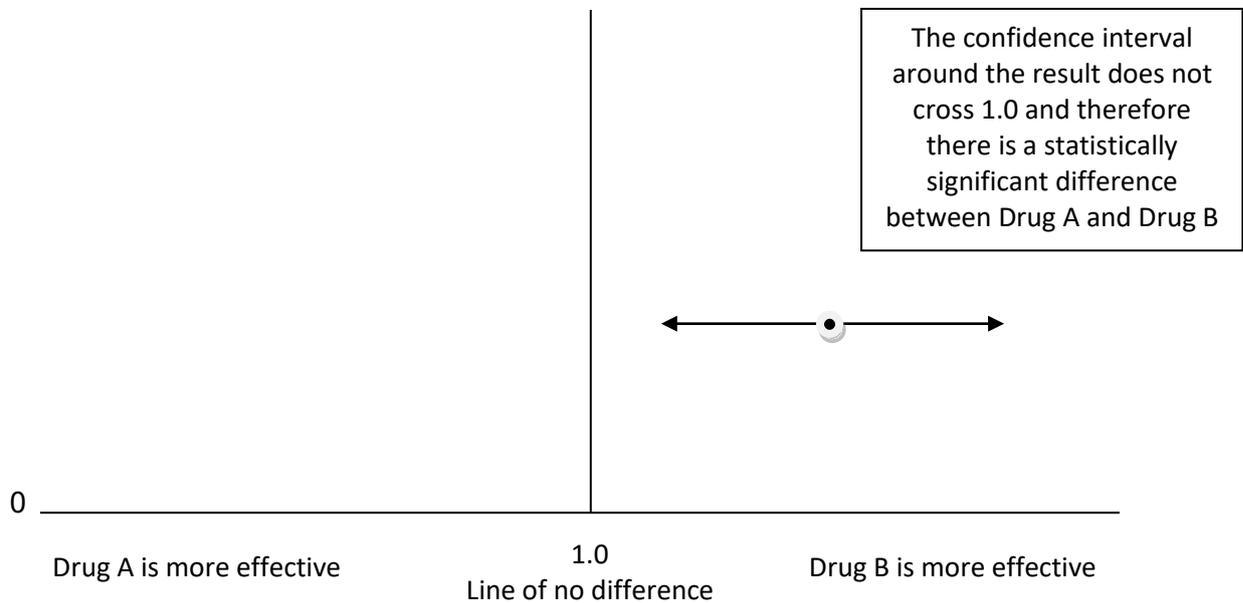
Confidence intervals can give us a measure of the certainty of a result. A clinical trial only tests a drug on a small group of people. The confidence interval tells us how likely it is that the results from that small group would still hold true for the entire population. They are expressed as a range of possible results, within which we expect the actual result to lie. A narrower confidence interval means the result is more precise and reliable, while a wider interval suggests less certainty.

By convention, 95% confidence intervals (95% CI) are normally used in drug trials, but you may also encounter 90 or 99%. A 95% CI means that you can be 95% sure that the true result lies within the range quoted, or, expressed another way, that there is a 1 in 20 (i.e. 5%) chance that the true value lies outside the range quoted. For example, if a study finds a drug lowers blood pressure by 10mmHg points with a 95% CI of 8 to 12, it means you can be 95% confident that the true effect on the whole population is somewhere between an 8- and 12-point reduction.

Confidence intervals also show if the difference between interventions is statistically significant or not. When dealing with results which are expressed as ratios (e.g. relative risk, hazard ratio, odds ratio), if the confidence intervals do not contain 1.0 then the result is statistically significant.

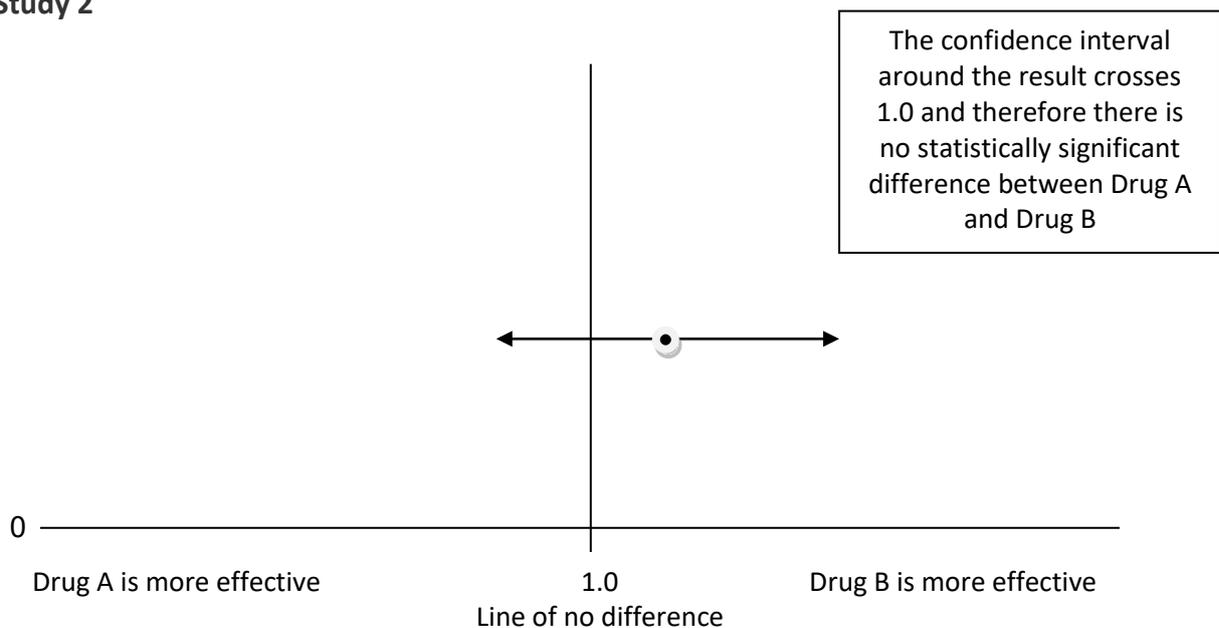
For example, consider the following results of 2 studies comparing Drug A and Drug B in reducing the risk of stroke. In the first study, the odds ratio is reported as 1.25 (95% CI 1.05 to 1.45) in favour of drug B.

Study 1



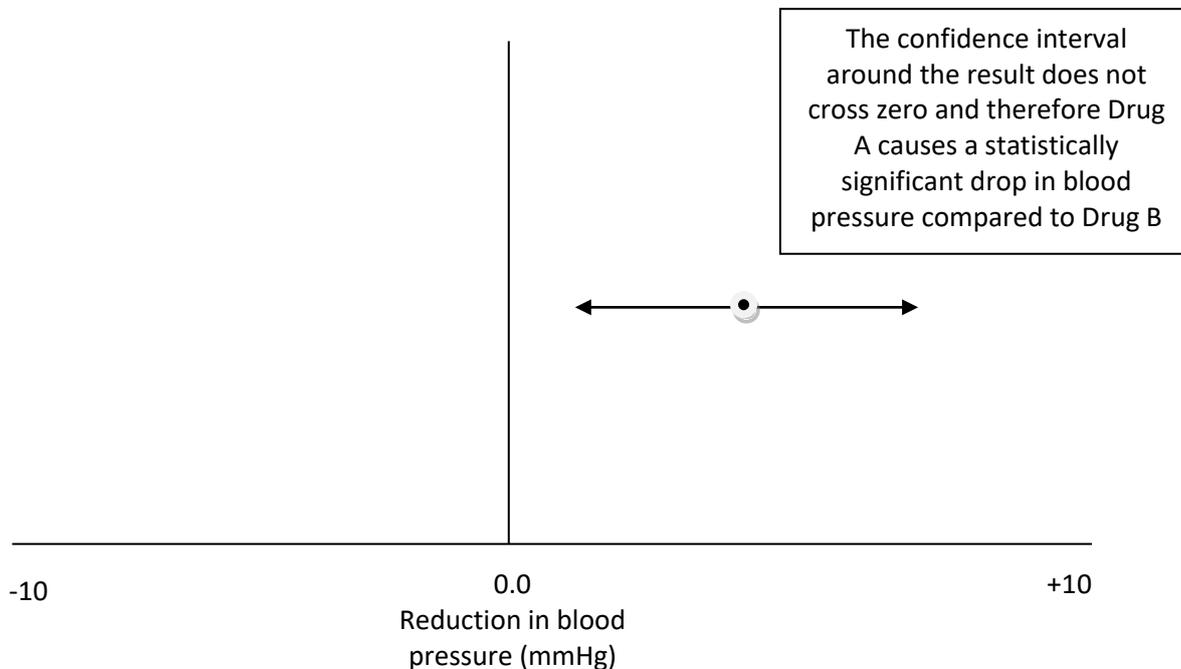
In the second study the odds ratio is reported as 1.10 (95% CI 0.90 to 1.30).

Study 2



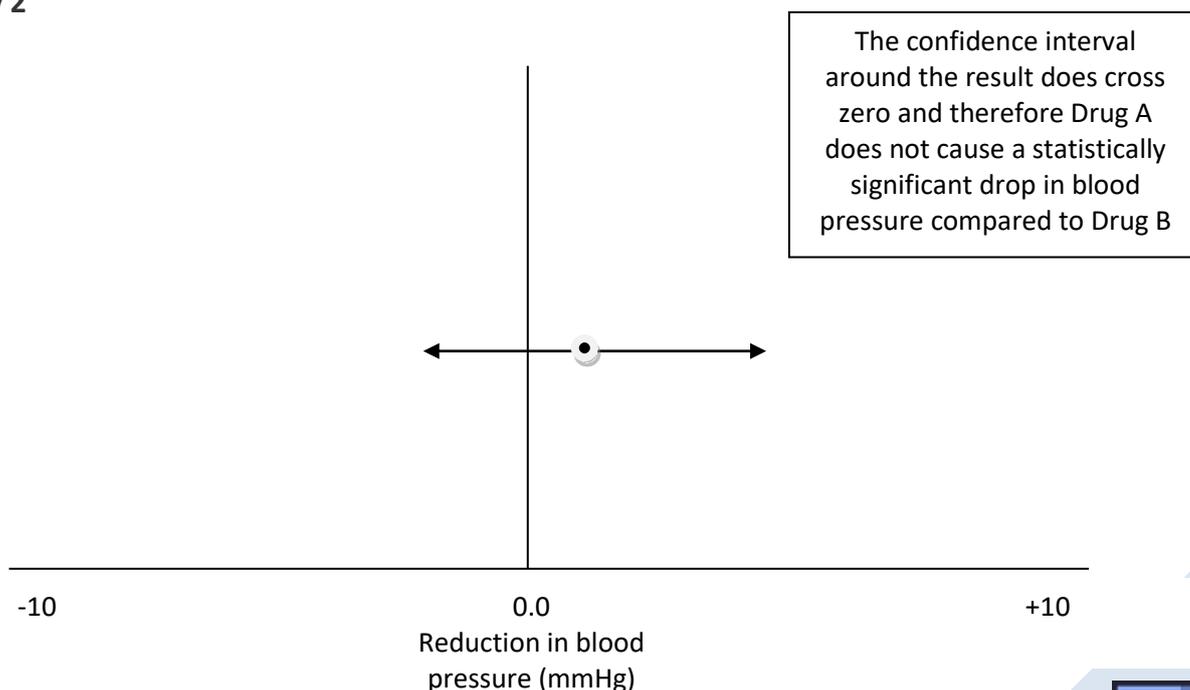
If you have a result not expressed as a ratio such as an absolute difference in blood pressure, then if the confidence intervals do not contain zero the result is statistically significant. For example, consider the following results of 2 studies investigating Drug A for hypertension versus Drug B. In study 1 Drug A produced a mean drop in blood pressure of 5mmHg (95% CI +1 to +7mmHg) more than Drug B.

Study 1



In study 2 of Drug A and Drug B, Drug A caused a mean drop in blood pressure of 1mmHg (95% CI -2 to +4 mmHg) compared to Drug B.

Study 2



As with most statistics, it's highly unlikely you'll need to do calculate confidence intervals yourself as this will be done by statisticians. But it does help to have an understanding of how they're calculated.

Information on how to calculate confidence intervals can be found in the bulletin 'Statistics in Divided Doses' [number 3](#) and [number 8](#). Or

you can watch this video <https://www.youtube.com/watch?v=R8fSAeuMvA>.



Statistics in Divided Doses

July 2005 No 8

Confidence intervals

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Some revision of confidence intervals

What do we already know about confidence intervals?

Observations from samples of subjects in clinical trials are used to draw inferences about the population from which those samples are drawn (SID 2). Due to the effects of random variation between the subjects and measurement errors, such observations have an inherent level of uncertainty, which can usually be quantified by calculating the relevant confidence interval (SID 3, 4, 7).

95% confident that the population value lies within this interval).

Example - Comparing antihypertensives

Two groups of men, diagnosed as having a rare type of hypertension, were randomised to receive either drug A or drug B in a clinical trial designed to compare the antihypertensive effects of the two drugs. The results are shown in Table 1.

Table 1

Observations	Drug A	Drug B
Number of subjects	50	50
Mean reduction (mmHg) in systolic BP	45	35
Standard deviation (SD) (mmHg)	20	18

The difference in systolic blood pressure reduction is 10mmHg in favour of drug A. In order to calculate the 95% CI we first need to calculate the standard error of the difference (SED) between the two drugs (SID 4).

Usefulness and risk

In assessing the usefulness of a trial there are two important considerations: are the results of the trial clinically important and what is the size of the benefit (and any harm)?

It is essential to judge the **clinical importance** of a result. We can do this by checking if the outcome measured was disease-oriented or patient-oriented. For example, if a new osteoporosis drug increases bone mineral density we would call this a **disease-oriented outcome** (DOO). If the drug reduces the risk of fractures we call this a **patient-oriented outcome** (POO). So an increase in bone mineral density of 1%, for example, may be statistically significant but if the rate of fractures is not improved is this clinically important?

To quantify the size of the benefit and harms of a new treatment, some of the key terms you'll see are:

- Absolute risk reduction (ARR)
- Relative risk (RR) and hazard ratio (HR)
- Relative risk reduction (RRR)
- Number needed to treat (NNT) and Number needed to harm (NNH)



We're going to use an example to help you understand what these terms mean and how to calculate them. Consider a fictitious randomised, double-blind trial of 'anotheraban' versus placebo and the prevention of stroke over 2 years. Each group contains 2,000 volunteers. At the end of the study, the number of patients suffering a stroke in the anotheraban group is 120 and the number in the placebo group is 160.

The **absolute risk** (AR) of an event is simply the chance it will happen. To work out the absolute risk of a stroke in each group, divide the number of strokes by the number of volunteers:

$$\text{AR placebo group} = 160 \div 2000 = 0.08 \text{ or } 8\%$$

$$\text{AR anotheraban group} = 120 \div 2000 = 0.06 \text{ or } 6\%$$

The **absolute risk reduction** (ARR) is the difference in risk of stroke between the two groups:

$$\text{ARR} = 0.08 - 0.06 = 0.02 \text{ or } 2\%$$

This means that anotheraban reduces the absolute risk of a stroke by 2%.

Relative risk (RR) tells us how many times more or less likely an event will occur in the treatment group relative to the placebo group. It is the risk of the outcome in the treatment group divided by the risk in the placebo group. From the above, AR anotheraban is 0.06 and AR placebo is 0.08, so:

$$RR = 0.06 \div 0.08 = 0.75$$

As this result is less than 1.0, anotheraban has made the risk of stroke less likely compared to placebo.

Some studies may use the term **hazard ratio** (HR) instead of relative risk – the two terms are broadly equivalent but hazard ratios are useful when the risk is not constant over time. It is weighted for the number of patients in the trial at different time points.

The **relative risk reduction** (RRR) is an alternative way of expressing the difference in risk; it is the reduction in risk of an event in the treatment group relative to the risk in the placebo group. Looking at the figures above you can see that the risk in the placebo group is 8% and the risk in the anotheraban group is 6%. Anotheraban has reduced the risk of a stroke by a quarter, from 8% to 6%. This is the relative risk reduction. Mathematically it is calculated like this:

$$RRR = (AR \text{ placebo group} - AR \text{ anotheraban group}) \div (AR \text{ placebo group})$$

$$RRR = (0.08 - 0.06) \div (0.08) = 0.25 \text{ or } 25\%$$

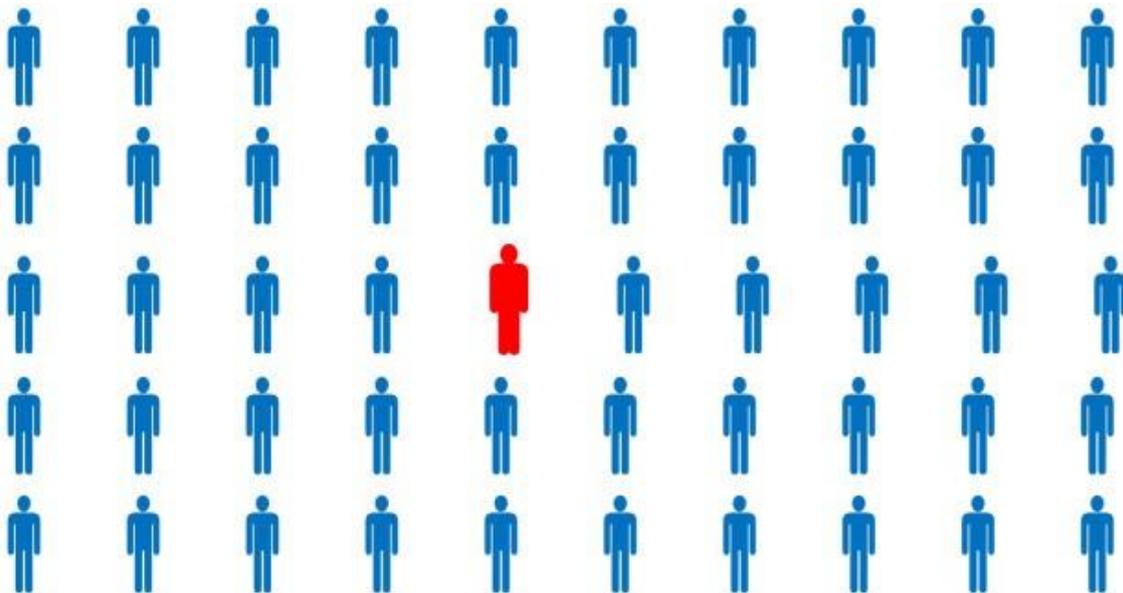
Pharmaceutical companies often use relative risk reduction because it makes a treatment's benefits sound more impressive. When you're looking at the results, think about this and try and work out the absolute risk reduction instead, as this gives a more realistic picture of a drug's effect.

Usefulness and NNT/NNH

The **number needed to treat** (NNT) is a useful statistic because it provides a practical and easy-to-understand measure of a treatment's effectiveness. NNT describes the number of patients that we would need to treat with anotheraban for 2 years to prevent one stroke. It is calculated as shown below, remembering that if you've been using percentages throughout the calculation then use 100 as the numerator to make the maths work. NNTs are normally rounded up to whole numbers:

$$\text{NNT} = 1 \div \text{ARR} = 1 \div 0.02 \text{ or } 100\% \div 2\% = 50$$

This means that 50 patients need to be treated with anotheraban for 2 years to prevent one stroke.



So what does this mean in terms of usefulness? NNT is directly meaningful to prescribers and patients which can help with decision-making. In general the smaller the NNT the better the treatment, with 1 being the ideal NNT. But this doesn't mean we should necessarily reject a drug with a high NNT. There are no rules about what an acceptable maximum NNT would be, and it will depend on several factors, such as the severity of the condition being treated, costs, side effects and individual values and preferences. We also need to realise that NNTs are open to interpretation. So some prescribers may see an NNT of 50 over 2 years as a considerable benefit, whereas others may see the benefit as small.

Comparing NNTs can help us choose between drug treatments. So, if anotheraban has an NNT of 50 while yetanotheraban has an NNT of 35, we might choose yetanotheraban as it seems more effective.

However, it is important to balance effectiveness or benefits of treatments against safety or potential harms. For this we can look at the **number needed to harm** (NNH).

In the same trial, 2 patients in the placebo group and 82 in the anotheraban suffer from life-threatening bleeding. We can describe how many patients we would need to treat for one to suffer from major bleeding (harm) using the NNH. We first need to work out the **absolute risk increase** (ARI) of major bleeding:

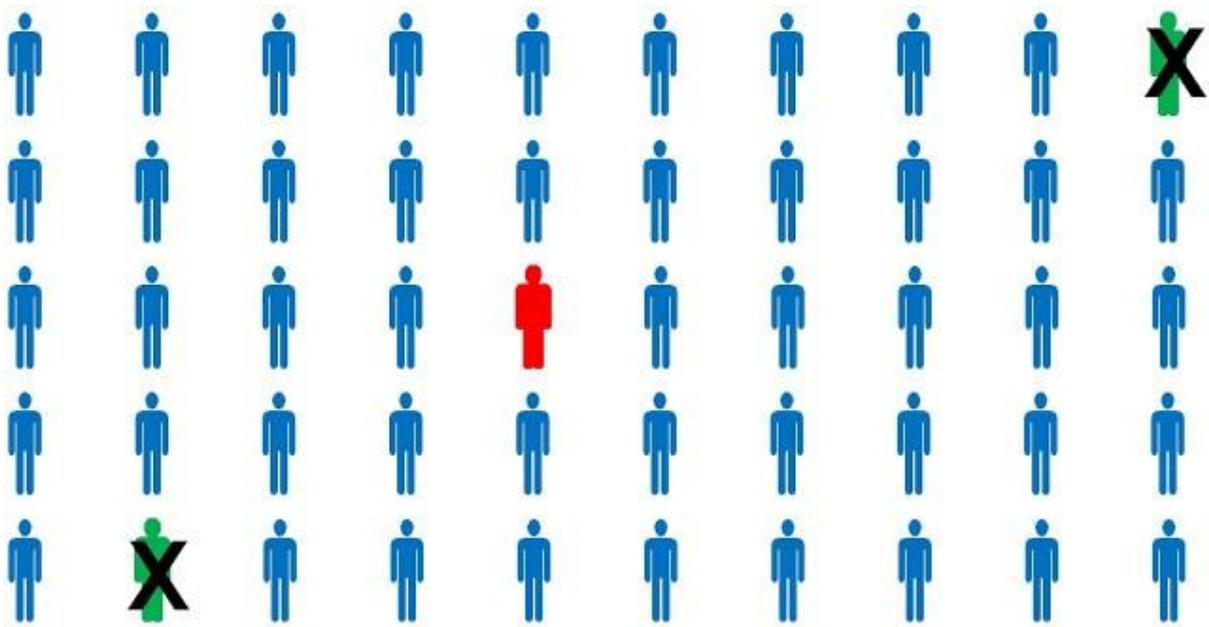
$$\text{ARI} = \text{AR anotheraban group} - \text{AR placebo group}$$

$$\text{ARI} = (82 \div 2000) - (2 \div 2000) = 0.04 \text{ or } 4\%$$

Then the calculation is similar to that used for NNTs:

$$\text{NNH} = 1 \div \text{ARI} = 1 \div 0.04 \text{ or } 100\% \div 4\% = 25$$

NNHs are normally rounded down. A higher NNH is generally more favourable, but as with NNT, there are no set rules about what an acceptable value might be.



Commissioners, or the people buying a service in which anotheraban was to be used, would need to consider these calculations carefully. Is it worth treating 50 patients for 2 years to prevent one stroke when for every 25 patients treated, one of them will have a major bleed?

Making the decision

So once we've assessed the validity and usefulness of a clinical trial paper, how do we then put that knowledge into practice and make a decision about the use of the medicine? An effective, agreed process is needed to review the drug, taking into consideration more than just cost. Most areas of the UK are covered by Area Prescribing Committees (APCs) or Medicines Optimisation Committees (MOCs), and their remit includes balancing the potential benefits and harms of different medicines.

One way of deciding between different therapies is known by the acronym 'STEPS':

Safety: What are the risks for patients? Look at adverse events from trials. Are there any groups of patients who were excluded from the trials who might be likely to receive the drug in practice?

Tolerability: Do patients remain on therapy? Examine the withdrawal rates from trials.

Effectiveness: How effective was the drug in trials? Is it clinically significant?

Price: Not just acquisition cost, but cost of administration equipment or blood tests, for example.

Simplicity: Is the device complicated? Is one drug given orally and a comparator by infusion?

It is important that this decision-making process is based upon clear criteria and is documented appropriately. This is especially important now that, in England at least, the NHS Constitution promises patients 'the right to expect local decisions on funding of other drugs and treatments to be made rationally following a proper consideration of the evidence'. If the local NHS decides not to fund a drug or treatment that the patient and their doctor think would be beneficial for them, then the patient also has the right to have that decision explained to them.

Being practical with critical evaluation

Critical evaluation skills are important for all healthcare professionals to support the application of evidence-based medicine. It's very likely that you'll need to use these skills to some extent whatever path your career takes, to influence healthcare professionals and patients to make the best decisions, and you may not always realise that this is what you're doing!

When you're undertaking critical evaluation more formally it's sometimes in response to a request for a medicine to be added to a local formulary. Often these requests are linked to a newly identified need, the potentially inappropriate use of a medicine, a safety issue, or funding. With this in mind, here are some tips on how to save time and make your work a cut above the rest:

1. See what's already available

You could spend several days retrieving papers and evaluating them, only to find that someone else has already done the work. For example, the [Specialist Pharmacy Service](#) publishes a list of new product evaluations that are freely available to NHS staff. The list is updated monthly and can be found on their website (type "new product evaluations" in the search box and find the latest version). Databases such as Medline and Embase will help you find reviews in journals. The [TRIP](#) database and the [Cochrane Library](#) can also be useful. Using someone else's work does not mean that you have to use it as it stands, but it gives you another point of view and at least helps you with your literature search.



2. Find out what the issues are

The best evaluations pick out the real issues right from the word 'go'. You could do a wonderful assessment of the evidence to support the efficacy of a new antidepressant

only to find that the key issue was its improved safety, and so your work wouldn't be very relevant. Identifying the issues can save you time too.

3. **Involve clinical experts**

By an extension to the above point, it is very clear that some evaluations are not done with the 'inside clinical knowledge' that can make them really sharp and relevant. Work with clinical pharmacists and specialist doctors by asking them to comment on your work. Remember that with critical evaluation you are aiming to assess 'usefulness' and hoping to change or affirm practice: you're more likely to do this well if you invite experts to work with you.

4. **Evaluate, don't summarise**

Regrettably it can happen quite often that evaluations are little more than summaries. What's the difference? Evaluations look at validity and usefulness: they place the evidence in a clinical context and point out any limitations. They help the reader use the evidence.

5. **Think 'big picture'**

Once you understand the issues in clinical practice, you may find that you need to expand your remit. Maybe you have been invited to appraise one drug, but is it the whole drug class that really needs looking at? You may have been asked to evaluate one particular paper, but perhaps you need an audit of current local practice first? Is it an evaluation that's needed or a clinical guideline, or both?

Information sources

The UK [Critical Appraisal Skills Programme](#) (CASP) offers a series of practical checklists to help you systematically evaluate clinical evidence including RCTs. There are some short e-learning courses on the CASP website too.

The [NICE Medicines and Prescribing community](#) provides support for delivering quality, safety and efficiency in the optimisation of medicines. Visit their website for evidence-based summaries about new medicines or unlicensed medicines.

You might like to compare the evidence-based reviews of medicines (technology appraisals) and clinical guidelines on the [NICE website](#), with those available in Scotland and Wales. Try the [Scottish Intercollegiate Guidelines Network](#) (SIGN), the [Scottish Medicines Consortium](#) (SMC) and the [All Wales Medicines Strategy Group](#) (AWMSG) for information on the effectiveness of new and existing medicines.

The [Cochrane Library](#) also has a collection of evidence-based reviews that are regularly updated, including The Cochrane Database of Systematic Reviews.

The [TRIP](#) database is a gateway to many evidence-based resources including those from professional bodies e.g. Royal Colleges.

Be careful about conducting a general internet search when looking for critical appraisals. If you do, you may like to look at our brief guide to [evaluating websites about medicines](#).