How big is your trial?

Tony Brady

The first clinical trial?

"I took 12 patients in the scurvy aboard the Salisbury at sea. Their cases were as similar as I could have them. 2 were ordered a quart of cider a day 2 others 25 gutts of elixir vitriol 2 others 2 sponfulls of vinegar 2 others a course of sea water 2 others the bigness of a nutneg 2 others 2 oranges and 1 lemon a day

The consequence was the most sudden and visible good perceived from the use of oranges and lemons."

James Lind, 1753

Different perspectives

• Scientific

- How many patients are needed:
 - to get firm evidence of a treatment difference? (powerbased)
 - to estimate treatment difference well? (precision-based)
- Sponsor
 - How many patients do we need to carry out a trial at minimum time, cost and effort?
- Ethics committee
 - How soon can the trial be stopped to avoid some patients getting inferior treatment?

· In general want smallest trial possible

BUT

- Should be large enough to have a realistic chance of detecting the treatment effect
- Should be large enough to convince clinicians that it's results are reliable
- Small trials with little hope of detecting a realistic treatment effect are unethical
- Unfortunately undersized trials are very common

Scientific approach to trial size

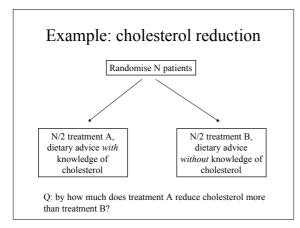
- What is the main purpose of the trial?
 - Treatment A is better than treatment BTreatment A is no better or worse than
 - treatment B (equivalence trial)
- What is the principal measure of patient outcome?
 - Continuous
 - Categorical
 - Survival time

- How will the data be analysed to detect a treatment difference?
 - Two-sample t-test of difference in means
 - 95% CI for difference in proportion
 - survival analysis
- Power-based (A is better than B):
- How small a treatment effect is it important to 'detect' and with 'what degree of certainty'?

Degree of certainty			
Error	Description	Pr(error)	aka
Type I	You think the treatments are different but they're not really	α	Significance level
Type II	You think the treatments are the same but in fact they're different	β	Power = 1- β

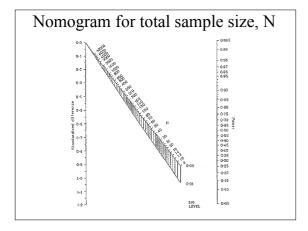
Size of treatment effect

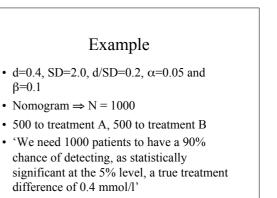
- Smallest effect which is clinically important to detect
- Will depend on treatment highly invasive or toxic treatments may need to show larger effect to be useful
- Most treatment effects are small or moderate in size





- Main outcome: difference between groups in mean cholesterol at end of study (d = mean_B - mean_A)
- Important to detect difference,d of 0.4 mmol/l
- Other studies indicate SD(cholesterol)=2.0 mmol/l
- Standardised difference = d/SD = 0.2
- Set α=0.05 and β=0.1
- Lookup N from nomogram





Paired design

- Reduce sample size by measuring cholesterol once at beginning and once at end of trial
- Look at within-person differences
- SD(individual differences) = 1.28
- Standardised difference d/SD = 0.4/1.28 = 0.31 , $\alpha{=}0.05$ and $\beta{=}0.1$
- N= 420 (210 per group)

Categorical data

- Main outcome is difference in proportions for dichotomous response
- $p_{\rm A}$ = proportion responding on treatment A $p_{\rm B}$ = proportion responding on treatment B
- Standardised difference = $(p_A p_B)/SD$

 $SD = \sqrt{\overline{p}(1-\overline{p})}$ where $\overline{p} = \frac{p_A + p_B}{2}$

Example: beta-blocker trial

- Trial to investigate whether preoperative betablockade prevents myocardial ischaemia
- Expect 20% of high risk patients to experience ischaemia
- Reduction to 15% on beta-blockers would be clinically important to detect
- d = (0.2 0.15) = 0.05
- SD= $\sqrt{(0.175 \times 0.825)} = 0.38$
- Standardised difference, d/SD = 0.13
- α =0.05, β =0.2
- From nomogram, N=1800 (900 per group)

Possible alternatives

- p_A=20%, p_B=18%, α=0.01, β=0.05
 So N=28000
 - Scientifically reasonable, but too costly
- p_A=20%, p_B=5%, α=0.05, β=0.5 - So N=70
 - Over-optimistic, very unlikely to be conclusive

Survival data

- Usually base power calculation on proportion surviving at single time point (e.g. 5 years)
- Can base calculation on median survival time but this isn't always observed in a trial
- Power calculation based on logrank test should be used with care - may underestimate N

Compliance / withdrawals

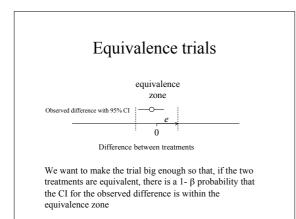
- Lack of compliance or withdrawals from randomised treatment may lead to smaller observed treatment difference
- · Larger sample is need to maintain power
- Let Q_A = proportion withdrawing or noncompliant from group A
- Let Q_B = proportion withdrawing or noncompliant from group B
- Suppose withdrawals and non-compliers from each group have a similar outcome to the other group

Compliance cont.

- Difference observed reduced by a factor $(1-Q_A-Q_B)$
- Need to increase total sample size by a factor $1/(1-Q_{\rm A}-Q_{\rm B})^2$
- If $Q_A = Q_B = 10\%$, need to increase sample size by a factor of $1/0.8^2 = 1.56$

Equivalence trials

- · Approach to sample size is usually precision based rather than power based
- Need to define the largest difference, e, in outcome between the 2 treatments we would accept as equivalence
- Then we need to recruit enough patients so that there is a 1- β chance that the confidence interval for the treatment difference excludes values larger than e



Example

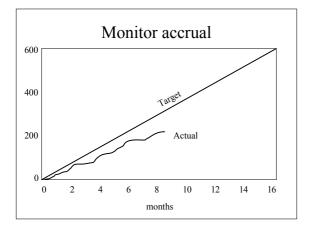
- · New beta-blocker drug being compared to well established beta-blocker for lowering blood pressure
- Prepared to accept disparity of $\pm 2 \text{ mmHg in mean}$ blood pressure at end of trial, e = 2
- · SD diastolic blood pressure is 14 mmHg
- N = 52 x (SD²/ e^2) for α =0.05, β =0.1
- So N = 2550
- 'If the two beta-blocker drugs are truly equivalent a trial of 2550 patients will have a 90% chance of excluding a difference larger than 2 mmHg from the 95% CI'

Is the trial practicable?

- Assess patient accrual rate:
 - Initial estimates may often be optimistic (e.g. don't take into account exclusion criteria)
 - Accrual period should ideally be kept short
- Consider a multi-centre trial r_{0} (0) r_{1} (0) r_{1} (1) r_{1} (1) r_{1} (1) r_{2} (1) r_{1} (1) r_{1} (1) r_{2} (1) r_{1} (- Increase the type II error
 - Increase the type I error
- Not reasonable to make extreme changes, but there may be some room for compromise

Pilot studies

- · Check feasibility / organisation
- · Check anticipated accrual rate
- · Provide estimates of parameters needed for sample size calculations (e.g. SD, placebo response rate)
- · Assess compliance / withdrawal rates



Practical issues

- Choice of $p_{\rm A}$ $p_{\rm B}$ or mean_A mean_B
- Compromise between scientific goals and feasible accrual
- Trialists often over-optimistic about availability of patients
- Abandon a trial if power is inadequate
- Trialists often underestimate the required size of trial
- · Avoid having too many treatments
- Avoid being too restrictive about patient entry