A commentary on randomized clinical trials: How to produce them with a good level of evidence

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TO RANDOMIZE OR TO SAMPLE?

When properly designed, conducted, and reported, the randomized clinical trial (RCT) represents the gold standard study in the evaluation of health interventions. However, it can produce biased results if there is no methodological rigor. The RCT design is considered the gold standard because the randomization of different groups can provide results without bias between groups exposed to different treatment conditions. The random assignment to treatment groups aims to ensure that the characteristics of the participants which may affect the results are balanced.

Fifty years after, the publication of the first RCT, the technical meaning of the term randomization continues...
to confuse some researchers. Several journals continue to publish “randomized” trials without a full understanding of the procedure. The term random does not mean the same as haphazard but has a precise technical meaning. Random allocation presupposes that each patient has a known chance, usually an equal chance as other participants, of being given a treatment option to be tested, but the treatment to be given cannot be predicted. If there are two treatments, the simplest method of random allocation is the one that gives each patient an equal chance of receiving either treatment; it is equivalent to flipping a coin. A common approach is simply to randomize the treatments according to the dates of birth of patients, the hospital registration numbers or dates of enrollment in the study, for example, to provide one treatment to those with even dates and the other one to those with odd dates. Although all these approaches are in principle unbiased, since they are not related to patient characteristics, problems arise from the openness and knowledge of the allocation system. Because the treatment is known when a patient is considered for participation in the clinical trial, this knowledge may influence the decision to recruit that patient and thereby produce groups that are not comparable.

Randomization depends on two processes: Generation of an unpredictable designation sequence and the concealment of this sequence until the intervention occurs. The generation or allocation of the sequence is appropriate if the sequences can prevent selection bias, for example, randomized computer-generated numbers, random number table, envelope drawing, coin flipping, card shuffling, dice throwing, etc.

Randomization based on a single sequence of random assignments is known as simple randomization. Simple randomization can be trusted to generate similar numbers in the two trial groups and to generate groups that are roughly comparable in terms of known (and unknown) prognostic variables. Restricted randomization describes any procedure to control the randomization to achieve balance between groups in size or characteristics. Blocking is used to ensure that comparison groups will be of approximately the same size. Stratified randomization is achieved by performing a separate randomization procedure within each of two or more subsets of participants (e.g., those defining age, smoking, or disease severity). Stratification by the center is common in multicenter trials.

HOW TO AVOID BIAS

According to Pannucci and Wilkins (2010), bias can occur in the planning, data collection, analysis, and publication phases of research. Understanding research bias allows readers to critically and independently review the scientific literature and avoid treatments which are suboptimal or potentially harmful. A thorough understanding of bias and how it affects study results is essential for the practice of evidence-based science.

The concealment of the allocation sequence is appropriate if the patients and researchers involved cannot foresee the designation, for example, numbered or coded drug containers with identical appearance and prepared by an independent pharmacy, randomization performed in a location away from test location, sequentially numbered, sealed and opaque envelopes, etc. Many researchers mistakenly consider the sequence generation process as randomization and disregard blinding. However, without adequate concealment, even randomized, unpredictable assignment sequences can be corrupted. Moreover, the generation of unpredictable sequences is probably irrelevant if they are not hidden from those involved in the recruitment of patients.

Blinding (or masking) should not be confused with allocation concealment. The allocation concealment is intended to prevent selection bias, protecting the designation sequence before and until allocation occurs. It can always be successfully implemented. However, blinding seeks to avoid determination bias, protects the sequence after allocation, and cannot always be implemented. Performance bias may occur if additional treatment interventions are provided preferably for a group. Blinding of patients and of those involved in the application of interventions prevents this bias and also protects against placebo differences in responses between the groups. Detection bias arises if the knowledge of the patient's name influences the evaluation of the results. This is avoided by blinding those who assess the results.

According to the Acceptance Program Guidelines, the clinical trial should be double-blind, including a randomized selection of individuals using a parallel or a crossover design. The subjects may be admitted and must be fully informed of the study type. An informed consent should be obtained. According to Martinez-Ricarte et al., the use of a parallel control group allows to evaluate the effectiveness, equivalence or superiority of the study treatment.

Assessing the quality of randomization of 250 controlled trials and 33 meta-analyses and analyzing the association between these evaluations and the estimated effects of treatment, Schulz concluded that trials in which the allocation sequence was inadequately concealed produced higher estimates of treatment effects than trials in which authors reported adequate concealment (odds ratio exaggerated, on average, by 30–40%). However, trials without proper generation sequence led to an estimation of treatment effects similar to those of trials with adequate generation. Thus, the procedure for generating sequence
has a lower overall role in preventing bias than the procedure for concealment. This observation makes sense, since having a random sequence (unpredictable) should make little difference without an adequate concealment. In their review, the authors also concluded that studies that were not double-blind yielded larger estimates of effects than double-blind trials (odds ratio exaggerated, on average, by 17%). Double-blinding and avoidance of exclusions after trial entry are the most important methods for reducing bias. Although the strength of this effect falls short of that for allocation concealment, double blinding appears to prevent bias. Randomization controls the selection bias, and the double-blind design controls the observer bias.

**USING THE CONSOLIDATED STANDARDS OF REPORTING TRIAL**

To understand the results of a randomized controlled trial, readers should understand its design, conduction, analysis, and interpretation. This goal can be achieved only through the complete transparency of the authors. Despite several decades of educational efforts, RCT and reports need to be improved. Researchers and editors developed the Consolidated Standards of Reporting Trials (CONSORT) report to help authors improve reporting using a checklist and a flow diagram. The items on the checklist relate to the content of the Title, Abstract, Introduction, Methods, Results, and Discussion. The revised checklist [Table 1] includes 22 items that were selected because empirical evidence indicates that not reporting the information is associated with biased estimation, (biased) effects of treatment or because the information is essential to judge the reliability or relevance of the results. The flow diagram is intended to represent the progress of all participants through the RCT [Figure 1]. It describes information for four stages of a trial: Enrollment, intervention allocation, monitoring, and analysis. It explicitly includes the number of participants according to each intervention group, each group behavior during the study and the participants who were included in the primary analysis of data.

**HOW TO MANAGE DEVIATIONS AND FOLLOW-UP LOSS**

The random assignment to treatment groups aims to ensure that the characteristics of the participants which may affect the results are balanced. However, the advantages of this experimental design can be affected by absences, withdrawals and losses of participants, which may void the initial equivalence of control and experimental groups. To manage these deviations, two strategies are commonly used: (1) The principle of analysis by intention to treat (ITT) states that any person should be analyzed as if he/she had completely followed the planned project and (2) Per protocol analysis, which proposes to include only those volunteers who joined the designated intervention and concluded the default action, without any deviation from the original protocol.

The analytical strategy known as ITT has been considered the best approach to preserve the integrity of randomization and strengthen the internal validity of the test and is defended by CONSORT guidelines. ITT is technical analysis of RCTs in which the final results of patients are compared within the groups to which they were randomized at baseline, regardless of whether they have been treated, left the study (drop-outs) or for any reason have not obeyed the initial protocol. Thus, analysis of ITT preserves the similarity between the treatment groups with respect to prognosis, except for the causal variation. For example: In an assay in which medical and surgical treatments for chronic stable angina pectoris are compared, some patients assigned for surgical intervention died before surgery. If these deaths are not attributed to surgical intervention, using an analysis by ITT, surgery may seem to have a false low mortality.

It is also necessary to consider that in clinical trials of substance abuse, many of the missing data are due to loss of follow-up, that is, individuals who discontinued the study after randomization to treatment and whose data were lost later. Using the last observation or the worst possible outcome in substance abuse testing may be valid because much of the observed abandonment may be due to recurrence, adverse effects or no change in the response. A high rate of discontinuation of the drug poses a problem because it can dilute the true treatment effect, reducing the power of comparison.

When the treatment is effective, but the losses are substantial, the analysis following the ITT principle underestimates the magnitude of the effect of the treatment that occurs in adherent patients. The subsequent potential loss of study power can be cut by increasing the sample size. Losses smaller than 10% of the total number of randomized patients reduce the chances of bias in the analysis.

**SAMPLE SIZE**

The estimation of sample size is a key issue in RCTs. Its purpose is to enroll an adequate number of subjects with a given confidence on the number that may be affected by sampling error. Thus, the researcher will get the data in a shorter period, cost-effectively and following ethical principles. The estimation of sample size is essential to avoid the occurrence of errors Types I and II. The size can be...
Table 1: Consolidated standards of reporting trials checklist

| Paper section and topic                  | Item  | Description |
|-----------------------------------------|-------|-------------|
| Title and abstract                      | 1     | How participants were allocated to interventions (e.g., “random allocation,” “randomized,” or “randomly assigned”) |
| Introduction background                 | 2     | Scientific background and explanation of rationale |
| Methods participants                    | 3     | Eligibility criteria for participants and the settings and locations where the data were collected |
| Interventions                           | 4     | Precise details of the interventions intended for each group and how and when they were actually administered |
| Objectives                              | 5     | Specific objectives and hypotheses |
| Outcomes                                | 6     | Clearly defined primary and secondary outcome measures and, when applicable, any methods used to enhance the quality of measurements (e.g., multiple observations, training of assessors) |
| Sample size                             | 7     | How sample size was determined and, when applicable, explanation of any interim analyses and stopping rules |
| Randomization - sequence generation     | 8     | Method used to generate the random allocation sequence, including details of any restrictions (e.g., blocking, stratification) |
| Randomization - allocation concealment  | 9     | Method used to implement the random allocation sequence (e.g., numbered containers or central telephone), clarifying whether the sequence was concealed until interventions were assigned |
| Randomization - implementation         | 10    | Who generated the allocation sequence, who enrolled participants, and who assigned participants to their groups |
| Blinding (masking)                      | 11    | Whether or not participants, those administering the interventions, and those assessing the outcomes were blinded to group assignment. When relevant, how the success of blinding was evaluated |
| Statistical methods                     | 12    | Statistical methods used to compare groups for primary outcome (s); Methods for additional analyses, such as subgroup analyses and adjusted analyses |
| Results participant flow                | 13    | Flow of participants through each stage (a diagram is strongly recommended). Specifically, for each group report the numbers of participants randomly assigned, receiving intended treatment, completing the study protocol, and analyzed for the primary outcome. Describe protocol deviations from study as planned, together with reasons |
| Recruitment                             | 14    | Dates defining the periods of recruitment and follow-up |
| Baseline data                           | 15    | Baseline demographic and clinical characteristics of each group |
| Numbers analyzed                        | 16    | Number of participants (denominator) in each group included in each analysis and whether the analysis was by “intention-to-treat.” State the results in absolute numbers when feasible (e.g., 10/20, not 50%). |
| Outcomes and estimation                 | 17    | For each primary and secondary outcome, a summary of results for each group, and the estimated effect size and its precision (e.g., 95% confidence interval) |
| Ancillary analyses                      | 18    | Address multiplicity by reporting any other analyses performed, including subgroup analyses and adjusted analyses, indicating those prespecified and those exploratory |
| Adverse events                          | 19    | All important adverse events or side effects in each intervention group |
| Discussion interpretation              | 20    | Interpretation of the results, taking into account study hypotheses, sources of potential bias or imprecision and the dangers associated with multiplicity of analyses and outcomes |
| Generalizability                        | 21    | Generalizability (external validity) of the trial findings |
| Overall evidence                        | 22    | General interpretation of the results in the context of current evidence |

estimated using a mathematical formula which will depend on the purpose, nature and parameters investigated in the RCT. However, the decision to choose the appropriate values of the parameters required in the formula is not always simple. It is, therefore, crucial that the authors present the estimated sample size through statistical principles.

**PLANNING STATISTICAL ANALYSIS**

The importance of the correct use of statistical analysis lies in the researcher’s effort to better interpret, organize and analyze data from his/her search. In addition, through the statistics, it is possible to draw conclusions, make predictions for the population and assist in decision making. In a clinical trial, after identifying the groups to be compared, it is necessary to define the dependent variable response that will be submitted to hypothesis testing. It is usual to set as the hypothesis of interest the lack of difference between groups, known as the null hypothesis. The alternative hypothesis is a second statement which contradicts the null hypothesis, that is, the alternative hypothesis is the lack of equality between the groups. These two cases cover all possible values (0–1) for the statistical hypothesis test, and then one of the two statements is true. The null hypothesis is rejected if the P value is “large.” The P value indicates the probability that Type I error has occurred. In the medical field, it is considered that a P value lower than 0.05 indicates that there are significant differences between the groups.
However, this statistical significance does not necessarily imply clinical significance.\cite{21}

An alternative to statistical analyses that are based on the $P$ value is the size effect analysis that aims to determine the clinical significance of the effect found. It is not limited to dichotomous outcomes (significant or not significant).\cite{22} In other words, this statistical model is an appropriate measure to determine the clinical significance of the clinical procedure proposed by the RCT. In addition, it will enable to determine whether the sample size was adequate to get enough statistical power.\cite{23} Thus, through the use of size effect analysis, it is possible to identify whether the observed differences are small, medium, or large.\cite{24}

**FINAL CONSIDERATIONS**

Certain characteristics that produce consequences for modern reviews must be taken into account when developing secondary studies, such as systematic reviews, and assessing the quality of primary studies, such as RCTs. One of these is that former clinical trial results may be less reliable. The majority of these studies were conducted at a time when the methodology of the assays was less stringent than now. Thus, they cannot be compared with more modern studies regarding the results and how to measure them.\cite{25}

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