MCID — The Minimal Clinically Important Difference Assigns Significance to Outcome Effects

At medical school we were taught that 70–80% of the information that leads to diagnosis comes from the patient’s anamnesis. The severity of the subjective symptoms and disabilities drives the patient to seek medical help and has a major influence on treatment and intervention, despite “objective” findings such as alterations on a radiograph, for example, in a knee affected by osteoarthritis (OA)1. Based on this insight, a huge number of patient-rated outcome instruments have been developed in the past 2 decades2. Nevertheless, the significance of therapeutic effects is still quantified by statistical methods alone in many study reports, especially in pharmacological ones, even if the effects are labeled as “clinically significant”3.

Beyond statistically and distribution-based quantification of effect significance, the dimension of an effect’s importance and significance, which includes the patient’s subjective perception of pain and function, reaches a higher sphere because it is closer to the central subject of interest in medicine, the patient. “It is recommended that the patient’s perspective be given the most weight, because these are patient-related outcome measures, although the clinician’s perspective is considered important as well”4. The founders of the concepts developed consequently to give this alternative meaning for outcome effects were Jaeschke and Redelmeier5,6.

Jaeschke was the first investigator to ask patients to rate their subjectively perceived change of health or symptoms between baseline and followup5. The responses on this “transition” item were related to the score differences of an outcome instrument within the same time period — the so-called anchor-based technique. The mean score difference (between baseline and followup) of the transition group responses “almost the same,” “a little better,” and “somewhat better” was defined as the minimal clinically important difference (MCID) for improvement — I call it the “crude” mean change method5.

In rheumatology, Redelmeier developed this concept further to become the so-called “corrected” mean change method6. The 280 included patients with knee OA scored at least 30/100 mm on pain visual analog scale and underwent individually tailored conservative (nonsurgical) outpatient management.

Specific positive characteristics of the study are that the MID were stratified by followup (12, 26, and 52 weeks) and adjusted by baseline variables sex, age, pain severity, 6-min walking test, waitlisted for joint replacement (yes/no), and unilateral/bilateral symptoms. The effects of 2 different anchors, walking level and knee symptoms, were examined.

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The KOOS score differences were correlated to the transition/anchor responses, and M(C)ID analysis was performed only if the correlation was ≥ 0.40\(^4\). The M(C)ID were determined by the 2 mean change methods as well as 2 receiver-operation characteristics curve (ROC) analyses. The first ROC, the “Youden” method, maximizes the sum of sensitivity plus specificity, and the second, the 80% specificity method\(^5,6,9\), works with a fixed minimum of specificity.

Analysis partly revealed wide variation of M(C)ID depending on the 4 chosen methods. By definition, the crude and the corrected mean change methods differed by the mean score difference of the “not changed” response group. In some analyses, it was close to zero [e.g., M(C)ID for improvement of function]; in others, not [e.g., M(C)ID for improvement of pain at 26 weeks]. The 2 ROC methods resulted in similar levels in some analyses and in quite different ones in others. The differing results highlight the effects and importance of the \textit{a priori} chosen method to determine M(C)ID and the underlying assumptions. This issue is discussed at great length as part of the study of Mills, \textit{et al}\(^11\).

The 2 mean change methods provide a measurement that is obtained from a sample of several individuals and is a mean value. In contrast, the M(C)ID obtained by the ROC methods is a single empirical result. From all the individual score differences of the sample lying on the ROC curve, the ROC method chooses one for which the sum of the sensitivity plus the specificity to differentiate “slightly better” from “not changed” is maximal\(^9\). This characteristic of a single point estimate that maximizes differentiation questions by the ROC method is that it provides a valid, minimal effect estimate of perceived changes in health.

Beside the chosen concepts, additional methods to quantify, for example, the use of relative instead of absolute effect parameters may yield M(C)ID that are easier to apply to future studies. The use of percentages of the baseline score is 1 possibility, as also discussed in the study by Mills, \textit{et al}\(^11\). Another might be the use of effect sizes, i.e., the mean change (between baseline and followup) in either absolute score points divided by the SD of the whole group at baseline or of the group’s score differences or the differences pooled between 2 groups\(^3,4,12\).

The study by Mills, \textit{et al} demonstrates that many different attitudes, assumptions, and theoretical considerations exist, which results in different concepts and empirical quantifications of M(C)ID\(^11\). Confusion is complete if anchor-based methods are mixed up with distribution-based ones that use statistical computation methods alone\(^4,9\). Discussion of which method is best for which condition is ongoing in the scientific world\(^4\).

One solution could be based on a consensus to return to the patients, who are the focus of our work, and consequently, to the original concepts of Jaeschke and Redelmeier\(^5,6\). Subtraction of the mean score difference of the “not changed” group from that of the “slightly better/worse” group according to Redelmeier might be elaborated further and be less biased than the crude estimate of Jaeschke, who, incidentally, “did not claim to actually derive a truly ‘minimal’ difference estimate”\(^4\). Multivariate adjustment for baseline variables and characterizing/determination/quantification M(C)ID by sample-based effect sizes offer possibilities that might reduce different sorts of bias to obtain valid M(C)ID rather than single point estimates.

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