Over the past decades in modern medicine, there has been a shift from statistical significance to clinical relevance when it comes to interpreting results from clinical trials. A concept that is increasingly being used as a surrogate for clinical relevance and effect size calculation is the minimum clinically important difference (MCID). In this paper, an overview is presented of the most important aspects of the MCID concept used in research trials and a discussion of what this means for the neurological patient in clinical trials and daily practice is given. Is the MCID the best outcome measure cut-off to be implemented?

Introduction

The need for evidence-based medicine is greater than it has ever been before. In accordance, there is a great need for outcome measures which fulfil modern clinimetric requirements such as validity, reliability and responsiveness and which solely represent a level of assessing outcome as postulated by the International Classification of Functioning, Disability, and Health concept [1].

Rating scales are frequently used as an outcome measurement in both daily neurological practice as well as many clinical trials. An example is the Modified Rankin Scale, an ordinal scale consisting of seven items running from no symptoms (0) to death (6), a commonly used outcome measurement in stroke trials like the ‘Mr Clean’ study regarding endovascular treatment of ischaemic stroke [2].

Another example is the Unified Parkinson’s Disease Rating Scale (UPDRS) which was used as a primary outcome in the Adagio study for the effect of rasagiline in Parkinson’s disease [3]. The UPDRS is also an ordinal scale with a range of 0 to 176 points in which a higher score indicates more severity.

The list with similar examples in daily practice and clinical trials is endless. However, when researchers do not correlate a statistically significant study result with its clinical relevance point of view, this could lead to misinterpreting results showing falsely positive or negative findings and exposing patients unnecessarily to (lack of) therapies [4]. Fortunately, over the past decades there has been a shift from statistical significance to clinical relevance when it comes to interpreting results from clinical trials [5–7].

Although the term ‘clinical relevance’ seems to be straightforward, it is not easy to define and to quantify. Who decides what is clinically relevant, the patient and/or the clinician? And how does one deal with different views on clinical relevance between patients? As an example, imagine two patients (A and B), both bedridden due to Guillain–Barré syndrome (GBS). Patient A is an elderly patient who considers ‘being able to walk with aid’ as a clinically relevant improvement. Patient B is a young adult who considers ‘being able to compete in professional sports again’ as a clinically relevant change. Both are affected by the same disease and are functioning on a similar level. Yet, they have a different interpretation of the term ‘clinical relevance’ and will have different...
goals for their treatment. The same could be applied to
differences between clinicians in terms of interpret-
ing the significance of the minimum clinically impor-
tant difference (MCID).

Even if a consensus could be reached on what
would be a clinically relevant change, a proper out-
come measure would still be needed to be able to
detect that change. When these changes are vast and
our study population is large, then an instrument
will have little to no problem detecting such a
change. However, changes are more likely to be sub-
tle, and sometimes can be shrouded by the natural
fluctuating disease course or other confounding fac-
tors. In these cases, proper outcome measures are
needed to detect these smaller but clinically relevant
changes and differentiate these changes from fluctuat-
ing ‘noise’ variations seen in illnesses. Subsequently,
consensus needs to be reached on how to determine
when a change is relevant enough to call the trial a
success and not simply let this be driven by the
P value hypothesis approach. In essence, is a trial
successful by only looking at a statistically significant
P value, like the Adagio study [3]? Or should a stan-
dardized cut-off point be used, like being able to
walk independently as was done in the GBS trials
[8]? Another option would be to determine the
MCID, as was done in a follow-up analysis of the
results of the ICE trial in patients with inflammatory
neuropathies [6].

In this paper, an overview is provided of the MCID
concept striving to help neurologists become more
familiar with this entity. The origin of the MCID, its
variable faces, and methods on how to determine its
cut-off as well as pitfalls when applying this concept
are discussed.

Finally, recommendations regarding its use in future
clinical studies and trials are provided.

Minimum clinically important difference:
origin

The term MCID was first defined by Jaeschke et al.
as ‘the smallest difference in score in the domain of
interest which patients perceive as beneficial and
which would mandate, in the absence of troublesome
side effects and excessive cost, a change in the
patient’s management’ [9]. Over time, a mix of various
definitions has been adopted for the MCID concept,
such as the minimal important difference, the mini-
mally important change, the minimally detectable dif-
fERENCE, the minimum detectable change etc. [10,11].
All these adaptations have a common denominator
aiming to quantify changes that are considered clini-
cally relevant.

Another related term is the patient acceptable
symptom state (PASS), which is defined as ‘the
value beyond which patients consider themselves
well’ [12]. Tubach et al. state that the MCID deals
with the concept of improvement (feeling better)
whereas the PASS deals with the concept of wellbe-
ing (feeling good), thus also being complementary to
the MCID.

Minimum clinically important difference:
methods

The MCID concept is generally categorized into two
main streams: the anchor-based method and the distri-
bution-based method [13,14]. Extensive information
regarding these (and newer) methods have been pub-
ished in several excellent reviews [15–17].

Anchor-based method

Anchor-based methods involve comparing the change
in the situation of a patient as captured by an out-
come measure to an external criterion. This external
criterion is often a patient’s own categorization of
their personal change, e.g. after an intervention.
Examples of this method are using a pain score like
the visual analogue scale, or a patient global impres-
sion of change scale (much worse, somewhat worse,
about the same, somewhat better, much better).

Most often researchers will look at the change in a
single patient over time, the so-called ‘within-patient’
change [18]. In a study population, the group who
scored ‘somewhat better’ or ‘much better’ is of interest
since these people have informed the researcher they
have clinical improvement (from the patient’s point of
view). The next step is to look at the (median) change
of the score of the instrument used that is representing
the level of assessment of interest, which is often con-
sidered as the minimum change that correlates with
clinical improvement.

Another anchor-based method is looking between
patients at a single point in time, the so-called ‘be-
tween-patient’ difference [18]. Patients are grouped
based on their rating on the external criterion: e.g.
pain (I have no pain, I have moderate pain, I have
extreme pain). Next, one would look at the (median)
scores of the instrument of interest in these groups
and then determine the MCID as the difference
between the median score of the groups ‘I have mod-
erate pain’ and ‘I have no pain’.

Less commonly adaptations are a combination of
within-person and between-person and a method in
which patients rate their health state in comparison to
other patients [19].
Distribution-based method

Distribution-based MCID methods are built upon the statistical properties of a study’s result [20]. These include both the effect size, where the mean change of the individual is divided by the variability of either the whole group or the subset of stable subjects, and the reliability change index, a statistic rooted in the standard error of measurement (SEM) [21]. The SEM is a measure of the variation of observed scores due to measurement error compared to the ‘true’ score. Any changes which are below the SEM could be due to a measurement error ‘noise’ rather than to a truly occurring change. The number of SEMs needed to qualify a change in a patient’s score to be meaningful is not yet fully established [16,17]. The most compelling argument has been provided by Wyrwich et al., who state that the ‘one-SEM criterion holds promise for identifying clinically meaningful intra-individual change’ [21,22].

The standard deviation (SD) is a measure that is used to quantify the amount of variation or dispersion of a set of data values. In multiple studies with different outcome measures there seemed to be a universally applicable rule of thumb that the MCID is equal to 0.5 SD [23].

Minimum clinically important difference: pitfalls

When using and interpreting MCID values, one should always take several considerations into account, of which some are addressed in the following.

Use of ordinal scales

The use of ordinal scales is widespread across modern medicine, even though their shortcomings have been known for a long time [24]. Modern techniques like the Rasch model can help transform ordinal scales to interval-based scales [25,26]. Ordinal scales are often treated as an interval-based scale. However, one of the major problems with ordinal scales is that the amount of change required to go from 2 to 4 on a 10-point scale is not always the same as to go from 6 to 8. Therefore, the lack of a fixed unit using a metric such as the Rasch model can help transform ordered data to interval data. The term modern is relative, the first publication on its theory dating from 1960 [25]. It is being applied increasingly in research in modern medicine over the past decades (Fig. 1). The Rasch model states that ‘a person having a greater ability than another person should have the greater probability of solving any item of the type in question, and similarly, one item being more difficult than another means that for any person the probability of solving the second item is the greater one’. So, in short, the person with a higher ability (thus being less ill) has a higher chance of getting a higher score.

Small but clinically relevant changes

Distribution-based methods define the MCID as a change bigger than the expected variation or error in measurement. It is not beyond imagination, however, to suggest a clinically relevant change that lies within the variation of a measurement, especially when the variation of a measurement has a wide range. All these relevant changes will not be noted because they are considered changes due to variation of the measurement rather than to clinical real change.

Static value for a dynamic concept

Ideally, one would be able to use a fixed MCID cut-off for each instrument for all patients in clinical studies. However, since the MCID is a dynamic concept, this is not a realistic proposition. Patients’ (individual) MCID will vary by the severity of their illness, their social status, their own concepts of health and improvement etc. Furthermore, the MCID can be different for different treatments in the same patient group. For a surgical intervention with high risk and long recovery, a patient will expect more improvement for it to be clinically relevant and to warrant this treatment than for example when the patient only needs to make a minor adjustment in his/her lifestyle. So, when examining a new treatment, one cannot just blindly copy an MCID determined in a different study with different treatments.

The MCID does not account for cost–benefit

Although the classic definition by Jaeschke et al. did incorporate cost (‘in the absence of ... excessive cost’), the MCID values determined nowadays do not take into account the cost of a specific treatment. In the light of the widespread pressure on funds for healthcare, it seems justified to take the cost of a certain amount of change into account as well, before establishing a change to be an MCID.

Minimum clinically important difference: Rasch improvements

The Rasch model is a modern technique that helps transform ordinal data to interval data. The term modern is relative, the first publication on its theory dating from 1960 [25]. It is being applied increasingly in research in modern medicine over the past decades (Fig. 1). The Rasch model states that ‘a person having a greater ability than another person should have the greater probability of solving any item of the type in question, and similarly, one item being more difficult than another means that for any person the probability of solving the second item is the greater one’. So, in short, the person with a higher ability (thus being less ill) has a higher chance of getting a higher score.
See also Fig. 2 for a graphic illustration of this concept. This means that a confirmation on an item depends on both the difficulty of the item and the ability of the person. For more in-depth information, readers are referred to other publications [27,28].

The Rasch model can help improve the MCID. One way is to transform the ordinal data to interval data, as stated above. This removes the obstacle of score changes being different across the scale (an example is a change from 2 to 4 is not necessarily the same as from 8 to 10 on an ordinal scale), since on an interval scale the increment along the scale is equal for each step. Applying the Rasch model is mostly beneficial if the original (ordinal) scale does not behave in a linear fashion.

Additionally, after applying the Rasch model, one can determine the individual standard error (SE) of the ability estimate of each individual patient. This allows for individually determined MCIDs based on the patient’s own SE. This is important because the SE of the ability of a patient varies across the range of said ability. In general, a patient with either a low or high ability score has a higher SE, whilst a patient with an average score will have a lower SE. This shows that one cannot just take a single MCID cutoff value to be applied to all patients equally. For more detail on this, see an extensive report by Hobart and Cano, in particular Chapter 8 [29].

**Recommendations and future perspectives**

For clinical research trials to advance, one should always consider if the pre-defined primary outcome truly is clinically relevant to our patients and that the method of determining that outcome is up to current modern clinimetric standards.

Before determining the MCID of a rating scale, one should be cautious of the nature of the scale. If it is an ordinal scale (behaving in a non-linear fashion), one should apply the Rasch model to create an interval-based scale, as mentioned above. This will also allow for individual MCIDs to be determined, furthering the ability of establishing responsiveness in individual patients and capturing their voice more accurately.

Researchers using previously established MCID cutoff values when setting up new trials are also strongly discouraged if the intervention, outcome measure and
patient population are not similar to the settings in which the MCID cut-off was established.

The current paper provides the neurological community a brief overview on the meaning of the MCID. Despite all the limits, pitfalls and controversy on how to determine the MCID, its concept is of great importance in modern clinical trials. As long as no consensus is reached on which method to use, an anchor-based method alongside a distribution-based method is recommended, and they should be seen as complementary to each other rather than separately. Furthermore, by examining both methods, one can also compare the two methods in terms of their dynamics and results. Finally, by applying an anchor-based method with for example a self-evaluation of one’s health, one can also capture the opinion of the patient. After all, is not the patient’s perspective on their own health, be it improvement, deterioration or maintenance, that which should be the most important outcome measure in clinical trials?

A relatively new trend regarding this aspect is the inclusion of patients (and their caregivers) in clinical trial design as part of the study team. This can lead to a better understanding of the scope over which a disease can impact a patient’s life. In doing so, previously neglected outcome domains, such as fatigue or sleep disturbances, can be identified and corrected [30–32].

Perhaps a next step would be to let the patients themselves tell us what they would deem as a sufficient amount of change in their health status for a treatment to be qualified as successful. This would surely be easier than constructing intricate surrogate markers with dubious clinimetric qualities which most clinicians struggle to comprehend. Simply talking and, more importantly, listening to our patients, however, is something every clinician excels at and perhaps researchers should start doing more and more.

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