Criteria for defining interictal epileptiform discharges in EEG
A clinical validation study

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Abstract

Objective
To define and validate criteria for accurate identification of EEG interictal epileptiform discharges (IEDs) using (1) the 6 sensor space criteria proposed by the International Federation of Clinical Neurophysiology (IFCN) and (2) a novel source space method. Criteria yielding high specificity are needed because EEG over-reading is a common cause of epilepsy misdiagnosis.

Methods
Seven raters reviewed EEG sharp transients from 100 patients with and without epilepsy (diagnosed definitively by video-EEG recording of habitual events). Raters reviewed the transients, randomized, and classified them as epileptiform or nonepileptiform in 3 separate rounds: in 2, EEG was reviewed in sensor space (scoring the presence/absence of each IFCN criterion for each transient or classifying unrestricted by criteria [expert scoring]); in the other, review and classification were performed in source space.

Results
Cutoff values of 4 and 5 criteria in sensor space and analysis in source space provided high accuracy (91%, 88%, and 90%, respectively), similar to expert scoring (92%). Two methods had specificity exceeding the desired threshold of 95%: using 5 IFCN criteria as cutoff and analysis in source space (both 95.65%); the sensitivity of these methods was 81.48% and 85.19%, respectively.

Conclusions
The presence of 5 IFCN criteria in sensor space and analysis in source space are optimal for clinical implementation. By extracting these objective features, diagnostic accuracy similar to expert scorings is achieved.

Classification of evidence
This study provides Class III evidence that IFCN criteria in sensor space and analysis in source space have high specificity (>95%) and sensitivity (81%–85%) for identification of IEDs.

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Visual analysis is still the gold standard in clinical EEG. However, most recordings are interpreted by neurologists without expert certification. Interictal epileptiform discharges (IEDs) are typically found in patients with epilepsy. However, interpreting EEG is based on pattern recognition, and reading EEG is still more of an art than a science. The body of knowledge is passed down by word of mouth from teacher to student, and EEG training is like an extensive process of probabilistic classification learning, where teachers provide reinforcement for the visually identified patterns.

Overinterpretation of EEG is a major contributor to misdiagnosis: 25%-30% of patients seen at epilepsy centers for drug-resistant seizures do not have epilepsy. Misdiagnosis has multiple detrimental consequences for patients.

Normal sharp transients are often misread as IEDs. Interictal EEG epochs, 10-20 seconds long, were selected by 2 authors (M.A.K. and S.B.) who did not participate in rating the samples according to the index tests (IFCN criteria and IRa). Sharp transients previously marked (during the diagnostic evaluation process) were then reviewed by the 2 authors, and the first marking confirmed by both experts. Because the results of the EEG interpretations were dichotomous, we aimed at including a similar number of patients with epilepsy and nonepileptic paroxysmal events so that the pretest probability was close to 50%. For the expected specificity of 95%, with a significance level of 5% and a power of 0.8, we needed at least 68 patients. EEG was recorded with the standardized 25 electrode array of the IFCN using a sampling frequency of 256 samples/s. Interictal EEG epochs, 10-20 seconds long, were selected by 2 authors (M.A.K. and S.B.) who did not participate in rating the samples according to the index tests (IFCN criteria and sensor space). Sharp transients previously marked (during the diagnostic evaluation process) were then reviewed by the 2 authors, and the first marking confirmed by both experts. Because the results of the EEG interpretations were dichotomous, we aimed at including a similar number of patients with epilepsy and nonepileptic paroxysmal events so that the pretest probability was close to 50%. For the expected specificity of 95%, with a significance level of 5% and a power of 0.8, we needed at least 68 patients. EEG was recorded with the standardized 25 electrode array of the IFCN using a sampling frequency of 256 samples/s. Interictal EEG epochs, 10-20 seconds long, were selected by 2 authors (M.A.K. and S.B.) who did not participate in rating the samples according to the index tests (IFCN criteria and sensor space). Sharp transients previously marked (during the diagnostic evaluation process) were then reviewed by the 2 authors, and the first marking confirmed by both experts. Because the results of the EEG interpretations were dichotomous, we aimed at including a similar number of patients with epilepsy and nonepileptic paroxysmal events so that the pretest probability was close to 50%. For the expected specificity of 95%, with a significance level of 5% and a power of 0.8, we needed at least 68 patients.

Methodology

Study design and EEG recordings
We included EEG recordings from consecutive patients who underwent long-term video-EEG monitoring (LTM) at Aarhus University Hospital (Denmark) and at the Danish Epilepsy Centre (Dianalund, Denmark) between January 2012 and September 2017. Inclusion criteria were patients who had habitual paroxysmal events recorded during LTM (classified as epileptic seizures or as nonepileptic paroxysmal events) and who had sharp transients (epileptiform or nonepileptiform) in their interictal recordings. For patients with epileptic seizures, there was an additional criterion: the IEDs recorded in the interictal period had to be concordant with the recorded ictal event (i.e., bilateral synchronous IEDs for generalized seizures in patients with generalized epilepsies and focal IEDs in the same lobar localization as the ictal recordings acquired in patients with focal onset seizures). Participants were included if they were aged 1 year or older. We excluded patients with diagnostically inconclusive LTM despite recorded habitual events and patients with both epileptic seizures and nonepileptic paroxysmal events. Because the results of the EEG interpretations were dichotomous, we aimed at including a similar number of patients with epilepsy and nonepileptic paroxysmal events so that the pretest probability was close to 50%. For the expected specificity of 95%, with a significance level of 5% and a power of 0.8, we needed at least 68 patients.

standard protocol approvals, registrations, and patient consents
The regional ethics committee reviewed and cleared this project, in which only deidentified patient data were collected. According to Danish regulations, this study did not need written informed consent from the patients because it involved a retrospective analysis of anonymized data. The study was registered in ClinicalTrials.gov (Unique Protocol ID: CIED; identifier: NCT03533374). We report the study according to the Standards for the Reporting of Diagnostic Accuracy Studies guidelines.
Procedures

Seven raters (H.T., L.D., P.G.L., R.S., R.W., S.R., and V.S.H.) with experience in reading EEG (median 14 years; range: 2–37 years) evaluated the samples, blinded to all other data (including the gold standard). The 2 authors who selected the samples for evaluation (as described above) did not participate in rating the samples. Previous studies suggested that the optimal size for a group of experts scoring IEDs in EEG was between 6 and 10.\(^2\)\(^9\) Raters independently evaluated the EEG samples in 3 different rounds (more than 1 month apart), in sensor space and in source space, the samples presented in different randomized order in each round. The raters did not know how many samples were from patients with epilepsy.

Sensor space refers to the traditional way of inspecting EEG, using montages of scalp EEG electrodes (sensors), which is a 2-dimensional representation. In the first round, raters visually extracted the EEG features corresponding to each IFCN criterion in sensor space\(^2\)\(^3\) and scored the presence or absence of each criterion, without concluding whether the waveform was epileptiform. These criteria were as follows: (1) di- or tri-phasic waves with sharp or spiky morphology (i.e., pointed peak); (2) different wave duration than the ongoing background activity; either shorter or longer; (3) asymmetry of the waveform: a sharply rising ascending phase and a more slowly decaying descending phase, or vice versa; (4) the transient is followed by an associated slow after-wave; (5) the background activity surrounding IEDs is disrupted by the presence of the IEDs; and (6) distribution of the negative and positive potentials on the scalp suggests a source of the signal in the brain, corresponding to a radial, oblique, or tangential orientation of the source.\(^2\)\(^1\)

Figure 1 is an infographic summarizing the IFCN criteria for IEDs in sensor space. The raters inspected the selected epochs containing the marked sharp transients using a commercially available EEG reader (BESA Research 6.1); they were allowed to switch between montages in sensor space (longitudinal and transversal bipolar and common average), change gain, time scale (starting at 30 mm/s), and digital filters. Single clicking on the peak of the sharp transients displayed a 3-dimensional voltage map, including isoelectric lines.

In the second round, raters inspected the EEG samples in source space in a different randomized order. Using a spatial filtering method\(^2\)\(^4\),\(^2\)\(^5\) the scalp EEG samples were transformed into a standardized source space corresponding to 25 regions of interest in the brain. In this process, a generalized linear operator combines the scalp EEG signals in a specific way to enhance the activity from one region while reducing the volume-conducted overlap from other areas.\(^2\)\(^4\),\(^2\)\(^5\) An icon at the left side of each displayed source trace symbolizes the corresponding brain region. Both superficial cortical (radially oriented) and deep (tangentially oriented) sources (in sulci) are included into the brain regions to localize the irritative zone at a sublobar level. A detailed technical description of this is available from Dryad (Supplementary material 1, available on Dryad, doi.org/10.5061/dryad.xsj3tx99w).

For the identification of IEDs in source space, the following operational definition was used by the raters: Epileptiform discharges are abnormal sharp transients, with characteristic (1) changes in time and (2) distribution in space, (3) excluding artifacts and normal variants. To qualify as an IED, all
of these features had to be present. Raters were provided with the following instructions: “Changes in time: the sharp transient is clearly distinguishable from the ongoing (background) activity, based on its amplitude, duration and morphology. Check this in the source channel(s) where the discharge is visible. Changes in space: the location of the signal should make sense anatomically. Check this across all channels: Regions with the same polarity should be close together. Opposite polarity may be observed in the other hemisphere or more remote regions. Caveat 1: temporal lobe traces may show opposite polarities. Caveat 2: Tangential IEDs in fissures can show polarity reversal between neighboring regions. If unclear, compare time-lag between the peaks in different channels to observe presence of propagation as an additional indicator of epileptiform transients. Presence of propagation is optional (i.e., not compulsory for all spikes).” Supplementary material 2 (available on Dryad, doi.org/10.5061/dryad.xsj3tx99w) contains an infographic summarizing how IEDs are identified in source space.

Finally, the experts scored the marked waveforms as IEDs or nonepileptiform sharp transients, unrestricted by any criteria, and solely based on their expertise, as if evaluated in their clinical practice (expert scorings).

Before starting scoring of the test samples, the raters went through open-label training samples (5 recordings with IEDs and 5 with nonepileptiform sharp transients). The training samples and the written instructions given to the raters in the training period are available on Dryad (Supplementary material 2, doi.org/10.5061/dryad.xsj3tx99w). The whole data set can be downloaded using this link: datadryad.org/stash/share/Fj1923GZtwudCKU2_ltcFinVpcVOHhYudHWlkwzQ538.

Evaluation of the extracted features
For each extracted feature, the consensus majority scorings were used for determining diagnostic accuracy. For the IFCN criteria in sensor space, we used cutoff values of 2–6 fulfilled criteria for qualifying a waveform as an IED. Analysis in source space and expert scoring resulted in dichotomous scores (IEDs or nonepileptiform sharp transients).

The gold standard (clinical diagnostic reference standard) for classification of patients as having epilepsy or having nonepileptic paroxysmal events was derived from the video-EEG recordings of the patients’ habitual clinical events.

We used the standard formulas: sensitivity = TP/(TP + FN), specificity = TN/(TN + FP), and accuracy = (TP + TN)/all patients (where FN = false negative, FP = false positive, TN = true negative, and TP = true positive).

Statistical analysis
Based on the sensitivities and specificities obtained at cutoff values of 2–6 IFCN criteria in sensor space, we calculated the estimated receiver operating characteristic (ROC) curve.30 We calculated 95% confidence intervals for sensitivity, specificity, and accuracy using the Wilson method,31 and we compared them between the IED identification methods using the McNemar test based on the true positive, the true negative, and both.32

We calculated IRA using Gwet agreement coefficient (AC1) to avoid the paradoxes of kappa. IRA was interpreted according to the conventional criteria: poor (<0.02), fair (0.2–0.4), moderate (0.4–0.6), substantial (0.6–0.8), and almost perfect agreement (>0.8). We compared IRAs of the IED identification methods using bootstrap.34 The analyses were performed using Stata version 15.1.

Classification of evidence
This diagnostic accuracy study was designed to answer the following question: What is the accuracy of the IFCN criteria in sensor space, and of analysis in source space, for identifying epileptiform EEG discharges?

This study provides Class III evidence that IFCN criteria in sensor space and analysis in source space have high specificity (>95%) and sensitivity (81%–85%) for identification of IEDs. The study was retrospective and blinded. All EEG samples, selected according to the criteria described in detail in the Methods section, were analyzed and evaluated. All patients had the diagnostic reference standard inferred from an independent data set and method (recordings of the patients’ habitual clinical episode) that is the most reliable diagnostic test for distinguishing epileptic from nonepileptic paroxysmal events. The included patients had a broad spectrum of diagnoses, with and without epilepsy. The following 3 procedures were performed by different experts, using different methods and blinded for each other: selection of the EEG samples, evaluation of the index test, and evaluation of the diagnostic reference standard (gold standard).

Data availability
Individual deidentified EEG data in EDF format (European Data Format) with the marked sharp transients of all patients included in this study will be shared, along with the following related documents: demographic data of the patients, gold standard of each patient, detailed technical description of signal reconstruction in source space, and STARD flow diagrams. Unrestricted access to these data will be made available from the day of the online publication of the article, until 2030, using a publicly accessible repository on Dryad (doi.org/10.5061/dryad.xsj3tx99w).

Results
We assessed EEG samples from 100 patients (60 female; age: 2–89 years, median: 31 years) who fulfilled the inclusion criteria. According to the reference standard, 54 patients had epilepsy (focal: 45; generalized: 7; both focal and generalized: 2), and 46 patients had nonepileptic paroxysmal events (psychogenic nonepileptic seizures: 31; parasomnia: 7; paroxysmal movement disorders: 4; syncope: 4).
Examples of IEDs and nonepileptiform sharp transients are shown in figure 2. The STARD diagrams and additional examples are available from Dryad (Supplementary materials 3 and 4, available on Dryad, doi.org/10.5061/dryad.xsj3tx99w).

The area under the ROC curve for IFCN criteria in sensor space was 0.977 for the consensus majority scorings (figure 3). The area under ROC curves for the individual raters was between 0.916 and 0.965 (median: 0.941) (Supplementary material 5, available on Dryad, doi.org/10.5061/dryad.xsj3tx99w). Table 1 shows the sensitivity, specificity, and accuracy of identifying IEDs using the IFCN criteria in sensor space with cutoff values between 2 and 6, the analysis in source space, and the expert scorings. The ROC curve showed that 3 methods (IFCN criteria with cutoff values of 4 and 5 and analysis in source space) achieved high diagnostic performance, similar to expert scorings (figure 3), and there was no statistically significant difference in the accuracy of these methods ($p > 0.157$). Using a cutoff value of 5 IFCN criteria resulted in significantly higher specificity ($p = 0.025$) but significantly lower sensitivity ($p = 0.005$) compared with the cutoff value of 4 IFCN criteria. All other comparisons in sensitivity and specificity between the 3 methods with high diagnostic performance and the expert scorings remained below the level of statistical significance.

Two methods achieved specificity higher than the expected threshold of 95%: IFCN criteria with a cutoff value of 5 and analysis in source space (figure 3 and table 1).

Agreement coefficients of the methods of identifying IEDs were between 0.490 and 0.830 (table 1). IRA of the 3 methods with high diagnostic performance was moderate to substantial (AC1: 0.490–0.608), and there was no significant difference in IRA between them ($p > 0.900$).

On their own, none of the IFCN criteria qualifies a transient as epileptiform or not (it is the number of the fulfilled criteria that are diagnostic). Nevertheless, to assess how reliable visual analysis of the individual criteria was, we calculated IRA of each criterion (table 2). IRA of criteria numbers 1, 2, and 4 was substantial; it was moderate for criterion 6 and fair for criteria numbers 3 and 5.

Using the consensus expert scorings as a reference standard (rather than the clinical diagnostic gold standard), the highest accuracies were achieved using a threshold of 5 IFCN criteria in sensor space and evaluating the signals in source space—both methods had an accuracy of 93%. A table showing these results is available from Dryad (Supplementary material 6, available on Dryad, doi.org/10.5061/dryad.xsj3tx99w).

**Discussion**

We found that the set of 6 IFCN criteria for identifying IEDs had high value of the area under the ROC curve (close to 1), demonstrating the high diagnostic ability of these criteria to discriminate between IEDs and nonepileptiform sharp transients. Three methods achieved high diagnostic performance, with accuracy similar to expert scorings: analysis in sensor space using cutoff values of 4 or 5 IFCN criteria and analysis in source space. The IRA of these methods was moderate to substantial.

It has been suggested that 4 IFCN criteria should be fulfilled to qualify a transient as an IED. Although the cutoff value of

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**Figure 2** Examples of IED (A–C) and nonepileptiform sharp transient (D–F)

Montages: longitudinal bipolar (A and D), common average (B and E), and source space (C and F). Voltage maps are shown in G and H. The transient shown in A–C with voltage map in G (oblique distribution corresponding to a source in the right temporal pole) fulfills all IFCN criteria in sensor space, except for criterion 5, and hence, it qualifies as an IED; in source space, it also qualifies as an IED, and the propagation from the temporal pole to the basal temporal region can be observed (C). The transient shown in D–F with voltage map in H only fulfills 2 IFCN criteria in sensor space (1 and 6), and hence, it does not qualify as an IED; in source space, one can observe that the transient belongs to the background activity from the right basal temporal region (fragmented during drowsiness); the orientation of the voltage map corresponds to a source in the right basal temporal region (fulfilling criterion 6). Voltage maps are useful in distinguishing IEDs from artifacts (figure 1); however, nonepileptiform transients originating from the brain can show voltage distributions similar to IEDs. IED = interictal epileptiform discharge; IFCN = International Federation of Clinical Neurophysiology.
4 criteria had high accuracy, the specificity was only 85%, which is below the threshold of 95% expected for clinical implementation. Only 2 methods had specificity over 95%: analysis in sensor space using 5 IFCN criteria as cutoff and analysis in source space. The sensitivity of these 2 methods was 81% and 85%, respectively.

There was substantial IRA on extracting 3 EEG features corresponding to IFCN criteria 1, 2, and 4: the spiky morphology, wave duration distinct from the background activity, and the presence of the slow wave following the transient. Assessment of the voltage maps (criterion 6) had moderate IRA, whereas assessment of the wave asymmetry and the disruption of the background activity (criteria 3 and 5) had only fair IRA. According to the IFCN operational definition, for qualifying as an IED, a transient has to fulfill a number of these criteria, regardless which criteria are fulfilled. When analyzing the IRA of qualifying a transient as an IED using 4 or 5 fulfilled criteria as cutoff, the agreement coefficient was 0.490 (moderate) and 0.608 (substantial), respectively.

In addition to the traditional evaluation in sensor space, we have assessed a novel approach of inspecting EEG signals in source space. In our opinion, it provides a clearer view onto the signals of the underlying brain areas compared with the scalp signals in sensor space (figure 2), which are more blurred by

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**Table 1** Sensitivity, specificity, accuracy, and IRA of the methods for identifying IEDs in sensor space and source space compared with expert scorings

<table>
<thead>
<tr>
<th>Method</th>
<th>Sensitivity % (95% CI)</th>
<th>Specificity % (95% CI)</th>
<th>Accuracy % (95% CI)</th>
<th>AC1 (95% CI)</th>
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<tbody>
<tr>
<td>IFCN criteria in sensor space</td>
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<tr>
<td>Cutoff (no. of fulfilled criteria)</td>
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<tr>
<td>≥2</td>
<td>100 (93.40–100)</td>
<td>8.70 (2.42–20.79)</td>
<td>58 (47.71–67.80)</td>
<td>0.83 (0.76–0.88)</td>
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<tr>
<td>≥3</td>
<td>100 (93.40–100)</td>
<td>63.04 (47.55–76.79)</td>
<td>83 (74.18–89.77)</td>
<td>0.64 (0.53–0.74)</td>
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<tr>
<td>≥4</td>
<td>96.30 (87.25–99.55)</td>
<td>84.78 (71.13–93.66)</td>
<td>91 (83.60–95.80)</td>
<td>0.60 (0.53–0.69)</td>
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<tr>
<td>≥5</td>
<td>81.48 (68.57–90.75)</td>
<td>95.65 (85.16–99.47)</td>
<td>88 (79.98–95.80)</td>
<td>0.49 (0.38–0.60)</td>
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<tr>
<td>≥6</td>
<td>66.67 (52.53–78.91)</td>
<td>100 (92.29–100)</td>
<td>82 (73.05–88.97)</td>
<td>0.60 (0.50–0.70)</td>
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<tr>
<td>Source space</td>
<td>85.19 (72.88–93.38)</td>
<td>95.65 (85.16–99.47)</td>
<td>90 (82.38–95.10)</td>
<td>0.59 (0.51–0.68)</td>
</tr>
<tr>
<td>Expert scorings</td>
<td>90.74 (79.70–96.92)</td>
<td>93.48 (82.10–98.63)</td>
<td>92 (84.84–96.48)</td>
<td>0.58 (0.50–0.67)</td>
</tr>
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</table>

Abbreviations: AC1 = Gwet agreement coefficient; CI = confidence interval; IED = interictal epileptiform discharge; IFCN = International Federation of Clinical Neurophysiology; IRA = interrater agreement.
overlap from remote regions. This property might be the reason for the slightly higher sensitivity and the high specificity, comparable to the best performing method in sensor space. Besides identification of IEDs, source space montages also help in localization, because they show the brain regions generating the discharges (figure 2), and aid in the separation of sublobar activities, for example, involving different parts of the temporal lobe, as documented previously.24,25 Thus, interpreting EEG in source space may be more intuitive and easier to explain to young trainees.25

The previous IFCN definitions of IEDs were conceptual, lacking specific rules for clinical implementation. Maulsby proposed a set of practical guidelines for assessment of IEDs. Many of them translate to the current criteria: “Every spike-looking wave is an artifact unless there are one or more good reasons for suspecting otherwise” (similar to criterion 3 in source space); IEDs “occupy a definable electrical field on the scalp” (similar to criterion 6 in sensor space); IEDs “are followed by a slow wave or series of slow deflections” (criterion 4 in sensor space); ignore sharp transients “which can be logically explained by simple alterations in voltage of the existing background rhythm” (criterion 2 in sensor space). Most of the current criteria can be traced back to Maulsby proposals and to those of Pierre Gloor.15

Because of the need for a reliable reference standard, we analyzed interictal EEG epochs extracted from LTM recordings. The odds for recording sharp transients (either epileptiform or not) are higher in long-term recordings compared with short, routine EEG recordings. Nevertheless, the specific aspect of EEG interpretation addressed in this study was the distinction between epileptiform and nonepileptiform sharp transients, and therefore, our results can be generalized to all instances where sharp transients are observed in EEG recordings of any duration.

A limitation in this study was that we did not assess the pattern of repetition (similarity of the wave morphology and frequency of occurrence) throughout the recordings. Because the IFCN criteria address features related to a single discharge and do not include a criterion on the repetition of the waveforms, we designed the study to reflect that. Moreover, in clinical practice, one needs to classify a transient as epileptiform or not, even when only one or few discharges are observed in a routine EEG recording. Another limitation is that patients with generalized epilepsies are underrepresented in our series, which is a reflection of the general pattern of referring patients for diagnostic LTM. However, distinguishing generalized IEDs from nonepileptiform sharp transients is a lesser problem in clinical practice.2,6

In this study, we opted for an unequivocal diagnostic gold standard: classification of the patients’ habitual paroxysmal clinical episodes, based on the video-EEG recordings. This resulted in a potential risk of spectrum bias because patients without habitual episodes during the monitoring and patients with inconclusive LTM were excluded. We considered it more important to have an unequivocal gold standard than to include all patients.

Raters evaluated the data set 3 different times, and repeated evaluation is a potential source of bias. We attempted to decrease this bias by presenting the samples in different, randomized order, by separating the evaluation rounds by at least 1 month, and by keeping the raters blinded to the gold standard until all 3 rounds of evaluation were completed.

The rate of IEDs in routine EEG recordings obtained in non-epileptic individuals without progressive neurologic disorder has been reported as 0.5%–2.6% in adults and 2.2%–3.5% in children. A much higher rate of 12% was found in a patient population where most patients had progressive neurologic disorders (known to have a higher rate of IEDs) but no history of seizure.37 All our patients had a history of paroxysmal episodes, but none of our patients without epilepsy had progressive neurologic disorder. The sensitivity of single, routine EEG recording is relatively low (32%–59%).4,38,39 However, our data set was derived from LTM. Repeated routine recordings and long-term video-EEG recordings increase the sensitivity of EEG for IEDs to 80%–90%.1,4,39

Although at first glance, the set of IFCN criteria might seem too time consuming to apply in routine clinical practice, with some experience, they are actually rapidly and naturally applied during visual EEG assessment. We provide here infographics (figure 1) to facilitate this. Systematic application of these criteria is especially important in difficult cases, and it provides guidance when disagreement occurs between readers.

In our experience, assessing EEG in source space is even more rapid than the traditional way of inspecting EEG in sensor space. Although the computation is complex, for the EEG reader, switching to source space is not time consuming (it takes the same time as switching from one montage to the other).

The currently available software tools for automated spike detection typically have low specificity (many false detections) but relatively high sensitivity.40 Therefore, there is

<table>
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<th>Table 2 IRA of the 6 IFCN criteria, scored individually</th>
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<td>IFCN criterion</td>
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Abbreviations: AC1 = Gwet agreement coefficient; CI = confidence interval; IFCN = International Federation of Clinical Neurophysiology; IRA = interrater agreement.
a need for validating the automatically detected spikes by experts, who provide the high specificity by rejecting the false detections. However, interrater reliability among experts is variable. The criteria described in this article will be useful for guiding neurologists when validating the automatically detected spikes. By grouping similar types of spikes into clusters, the software tools can significantly reduce the time burden for reviewing the detected spikes.40

There is a need for more precise definition of the 2 EEG features that had only fair IRA (asymmetry of the waveform and disruption of the background activity). Further improvement of the methods could be achieved by adding to the criteria features related to the repetition of the discharge in the recording. The features described in this article are potential targets for development of automated algorithms for spike detection. We grant unrestricted access to the deidentified EEG samples collected for this study to facilitate further development in this field.

In conclusion, the presence of EEG features corresponding to 5 fulfilled IFCN criteria in sensor space and analysis in source space have high diagnostic accuracy. Their implementation in clinical practice will contribute to decreasing overinterpretation of EEG and misdiagnosis of epilepsy.

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Appendix

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<th>Name</th>
<th>Location</th>
<th>Contribution</th>
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Criteria for defining interictal epileptiform discharges in EEG: A clinical validation study

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