Simulation of spreading depolarization trajectories in cerebral cortex: Correlation of velocity and susceptibility in patients with aneurysmal subarachnoid hemorrhage

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Abstract

Cortical spreading depolarization (SD) is fundamental mechanism observed in neuronal tissue which is provoked by critically low substrate influx. SD is characterized by the near-complete breakdown of transcellular ion gradients measurable as a pronounced drop in the membrane potential in neuronal tissue during many pathological states. So called slow potential change (SPC) propagates at the velocities between 2 and 9 mm/min. Typically the SD velocities correlate with the SD susceptibility in many animal models. For patients in neurocritical care, the Co-Operative Studies on Brain Injury Depolarization (COSBID) recommends several variables to quantify the SD occurrence and susceptibility, without providing the method for measurement of SD velocity. Therefore we developed an algorithm to estimate the SD velocities. The model is based on the semi-lunar arc representing the SD-wavefront which propagates across geometrically discretized brain surface. The surface was reconstructed from MRI scans to provide the spatial input. Temporal component is given by the differences between the SPC hit-times measured on the single electrodes of subdural six-channel electrode strip. After the onset of aneurysmatic subarachnoid hemorrhage (aSAH) the patients were admitted to intensive care unit and were subjected to electrocorticographic (ECoG) monitoring over the period of fifteen days. From the total of 70 available patients the analysis was technically feasible on 12 of them. The results contained great number of fitted trajectories. The output parameter, the SD-velocity (median 3.6(2.8 4.8) mm/min), was calculated from the velocities of every single trajectory of the corresponding event. Estimated SD-velocity was correlated against the independent clinical variables related to the SD susceptibility. Statistically significant correlations supported the validity of here proposed method. Not a single trajectory was found when electrode strip was placed on a lesion. Significantly frequent inability to render trajectories was present in the sub-group with the duration of isoelectric SDs >4min compared against the sub-group with the duration <4min, without significant difference in velocity between the two. It seems the SD velocity depends on local conditions of the tissue, in contrast to the propagation complexity which depends on the SD frequency and duration. Low rate of success in reconstructing the brains with pronounced anatomical changes as well as the inability of algorithm to exploit the SD-events with less than three active electrodes, are the limitations of this method.
Abstract (German)

Spreading Depolarization (SD) is the fundamental mechanism of an almost complete breakdown of transcellular ion gradients with resulting drastic drop in membrane potential, which occurs in neuronal tissue at a variety of neurological diseases. The SD spreads in the cortex with a speed between 2 and 9 mm/min and is measured as so-called slow potential change (SPC). Typically, the speed and susceptibility of tissue to SD correlate with the speed and susceptibility of SD in many animal models. For monitoring patients, Co-Operative Studies on Brain Injury Depolarizations (COSBID) recommends more variables for quantification of SD occurrence and the susceptibility of SD, but no reliable method for speed measurement existed until now. Therefore, we developed an algorithm for estimation of the SD speed. The used model is based on the propagation of the semilunar wavefront of the SD on the geometrically discrete cortical surface. This was reconstructed from the MRT images and serves as a spatial input parameter. As temporal input parameter, the differences between the arrival times of the SD at individual contacts of the six-channel electrode strip serve as data. The data was collected from 70 patients suffering from an aneurysmatic subarachnoidal bleeding and were examined in the framework of COSBID. The analysis was only possible in 12 cases. The results contained a large number of model trajectories, except in cases where the electrode strip was placed on a lesion. The initial parameter, the SD speed (median 3.6 (2.8 4.8) mm/min), was calculated from the individual trajectory speeds. This was validated against independent susceptibility-related clinical variables with high significance. Above the duration of 4 min, the model fitting of isoelectric SDs was significantly unsuccessful, so that the speeds between the two subgroups could not be differentiated. It seems that the SD speed is affected by local tissue parameters and the propagation complexity by the SD frequency and duration. Low failure rate in the discretization of the cortical surface due to the altered anatomy and algorithm-based minimum number of three active electrodes are the weaknesses of this method.
Affidavit

I, Denny Milakara certify under penalty of perjury by my own signature that I have submitted the thesis on the topic "Simulation of spreading depolarization trajectories in cerebral cortex: Correlation of velocity and susceptibility in patients with aneurysmal subarachnoid hemorrhage". I wrote this thesis independently and without assistance from third parties, I used no other aids than the listed sources and resources.

All points based literally or in spirit on publications or presentations of other authors are, as such, in proper citations (see "uniform requirements for manuscripts (URM)" the ICMJE www.icmje.org) indicated. The section on methodology (in particular practical work, laboratory requirements, statistical processing) and results (in particular images, graphics and tables) corresponds to the URM (s.o.) and are answered by me. My contribution in the selected publication for this dissertation corresponds to those that are specified in the following joint declaration with the responsible person and supervisor

The importance of this affidavit and the criminal consequences of false affidavit (section 156, 161 of the Criminal Code) are known to me and I understand the rights and responsibilities stated therein.

Date

Signature

Detailed Declaration of Contribution

Denny Milakara had the following share in the following publication:

Publication: Denny Milakara, Cristian Grozea, Markus Dahlem, Sebastian Major, Maren K.L. Winkler, Janos Lückl, Michael Scheel, Vasilis Kola, Karl Schonknecht, Svetlana Lublinsky, Alon Friedman, Jed A. Hartings, Johannes Woitzik, Jens P. Dreier, Simulation of spreading depolarization trajectories in cerebral cortex: Correlation of velocity and susceptibility in patients with aneurysmal subarachnoid hemorrhage, NeuroImage Clin 2017, Vol 16 (524-538), DOI: 10.1016/j.nicl.2017.09.005, PMCID: PMC5602748

Contribution in detail: Denny Milakara autonomously: designed this method, analyzed the imaging data and analyzed prepared neurophysiological data, implemented the software and generated both the numeric results as well as graphic results, partially developed the algorithm, developed and implemented the algorithm validation method and wrote the initial version of the manuscript.

Signature, date and stamp of supervising University teacher

Signature of the doctoral candidate
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Simulation of spreading depolarization trajectories in cerebral cortex: Correlation of velocity and susceptibility in patients with aneurysmal subarachnoid hemorrhage

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In many cerebral grey matter structures including the neocortex, spreading depolarization (SD) is the principal mechanism of the near-complete breakdown of the transcellular ion gradients with abrupt water influx into neurons. Accordingly, SDs are abundantly recorded in patients with traumatic brain injury, spontaneous intracerebral hemorrhage, aneurysmal subarachnoid hemorrhage (aSAH) and malignant hemispheric stroke using subdural electrode strips. SD is observed as a large slow potential change, spreading in the cortex at velocities between 2 and 9 mm/min. Velocity and SD susceptibility typically correlate positively in various animal models. In patients monitored in neurocritical care, the Co-Operative Studies on Brain Injury Depolarizations (COSBID) recommends several variables to quantify SD occurrence and susceptibility, although accurate measures of SD velocity have not been possible. Therefore, we developed an algorithm to estimate SD velocities based on reconstructing SD trajectories of the wave-front’s curvature center from magnetic resonance imaging scans and time-of-SD-arrival differences between subdural electrode pairs. We then correlated variables indicating SD susceptibility with algorithm-estimated SD velocities in twelve aSAH patients. Highly significant correlations supported the algorithm’s validity. The trajectory search failed significantly more often for SDs recorded directly over emerging focal brain lesions suggesting in humans similar to animals that the complexity of SD propagation paths increase in tissue undergoing injury.

A R T I C L E  I N F O

Keywords:
Cytotoxic edema
Ischemia
Spreading depression
Stroke
Subarachnoid hemorrhage
Traumatic brain injury

A B S T R A C T

In many cerebral grey matter structures including the neocortex, spreading depolarization (SD) is the principal mechanism of the near-complete breakdown of the transcellular ion gradients with abrupt water influx into neurons. Accordingly, SDs are abundantly recorded in patients with traumatic brain injury, spontaneous intracerebral hemorrhage, aneurysmal subarachnoid hemorrhage (aSAH) and malignant hemispheric stroke using subdural electrode strips. SD is observed as a large slow potential change, spreading in the cortex at velocities between 2 and 9 mm/min. Velocity and SD susceptibility typically correlate positively in various animal models. In patients monitored in neurocritical care, the Co-Operative Studies on Brain Injury Depolarizations (COSBID) recommends several variables to quantify SD occurrence and susceptibility, although accurate measures of SD velocity have not been possible. Therefore, we developed an algorithm to estimate SD velocities based on reconstructing SD trajectories of the wave-front’s curvature center from magnetic resonance imaging scans and time-of-SD-arrival differences between subdural electrode pairs. We then correlated variables indicating SD susceptibility with algorithm-estimated SD velocities in twelve aSAH patients. Highly significant correlations supported the algorithm’s validity. The trajectory search failed significantly more often for SDs recorded directly over emerging focal brain lesions suggesting in humans similar to animals that the complexity of SD propagation paths increase in tissue undergoing injury.

Abbreviations: AC, alternating current; ADC, apparent diffusion coefficient; aSAH, aneurysmal subarachnoid hemorrhage; COSBID, Co-Operative Studies on Brain Injury Depolarizations; CT, computed tomography; DC, direct current; DWI, diffusion-weighted imaging; E, electrode; EGCG, electrocorticography; FLAIR, fluid-attenuated inversion recovery; HU, Hounsfield units; ICH, intracerebral hemorrhage; IOS, intrinsic optical signal; MCA, middle cerebral artery; MHS, malignant hemispheric stroke; MPRAGE, magnetization prepared rapid gradient echo; MRI, magnetic resonance imaging; NO, nitric oxide; PTDD, peak total SD-induced depression duration of a recording day; R_diff, radius difference; SAH, subarachnoid hemorrhage; SD, spreading depolarization; SPG, slow potential change; TBI, traumatic brain injury; TOAD, time-of-SD-arrival-difference; V_diff, velocity difference; WFN, World Federation of Neurosurgical Societies; 3D, three dimensional

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1. Introduction

It is now increasingly recognized that spreading depolarization (SD) is the principal mechanism of the mass edema of neurons in many gray matter structures of the brain including the neocortex (Dreier and Reifffurth, 2017). Its hallmark is the abrupt, near-complete, toxic breakdown of the transcellular ion gradients, which creates an osmotic driving force for influx of water (Dreier et al., 2013). Whether or not the SD-induced mass edema, also termed cytotoxic edema, progresses toward cellular injury, is determined by the local tissue conditions and their repercussions on mechanisms involved in the SD process. The term SD continuum describes the changing characteristics of the wave dependent on the local tissue conditions (Dreier and Reifffurth, 2015; Hartings et al., 2017b). The full SD continuum is, for example, observed in a single SD wave when it originates in the center of focal ischemia and subsequently invades first the penumbra and then the surrounding well-nourished tissue against the gradients of oxygen, glucose and perfusion.

SD-induced cytotoxic edema is observed in animals using electron microscopy (Van Harreveld and Khattab, 1967) and the quaternary ammonium salt method (Mazel et al., 2002; Perez-Pinzon et al., 1995; Windmuller et al., 2005) as an abrupt decrease of the extracellular volume fraction and increase in tortuosity, which describes the average path length for diffusion between two points in the extracellular compartment. It is moreover visualized as swelling of the neuronal somata and dendritic beading using two-photon microscopy (Murphy et al., 2008; Obeidat et al., 2000; Runhta et al., 2015; Steffenes et al., 2015; Takano et al., 2007). These structural changes hinder the mobility of water in both the intra- and extracellular compartment during SD no matter whether SD passes through well-perfused or ischemic tissue (Budde and Frank, 2010; Mazel et al., 2002; Perez-Pinzon et al., 1995).

On this basis, SD causes abrupt decline in the apparent diffusion coefficient (ADC) of water in diffusion-weighted magnetic resonance imaging (MRI) scans (Cain et al., 2017; de Crespigny et al., 1998; de Crespigny et al., 1999). According to Leão's original notion that SD is the principal response of neurons to a prolonged episode of cerebral ischemia (Leão, 1947; Marshall, 1959), this ADC decline represents the gold standard for diagnosis of acute ischemia in the cortex in clinical neurology (Dreier and Reifffurth, 2015). It corresponds well with this notion that electrophysiological evidence of SDs has been found in practically 100% of patients with malignant hemisphere stroke (MHS) (Dohmen et al., 2008; Wozitik et al., 2013), 70–80% of patients with poor-grade aneurysmal subarachnoid hemorrhage (aSAH) (Dreier et al., 2009; Dreier et al., 2006), 60–70% of patients with intracerebral hemorrhage (ICH) (Fabricius et al., 2006; Hellbek et al., 2017) and 50–60% of patients with severe traumatic brain injury (TBI) (Fabricius et al., 2006; Hartings et al., 2011a). Animal experiments suggest that, similar to electrographic seizures, SDs never occur spontaneously in healthy brain (Dreier and Reifffurth, 2017; Hartings et al., 2017b).

Current gold standard for monitoring SDs in patients with TBI, ICH, aSAH and MHS is electrocorticography (ECoG) with a collinear subdural electrode strip (Dreier et al., 2017) (Fig. 1A). SD is observed as a large, abrupt negative electrical potential shift in the direct current (DC) frequency range of the ECoG below ~0.05 Hz, also called slow potential change (SPC) (Fig. 1B) (Dreier et al., 2009; Hartings et al., 2011b; Oliveira-Ferreira et al., 2010). This emanates from differences in depolarization between soma and dendrites (Makarova et al., 2010). The energy-dependent local duration of the DC shift indicates the local duration of mass depolarization and cytotoxic edema and, hence, the risk of injury at the recording site (Back et al., 1994; Dijkhuizen et al., 1999; Hinzman et al., 2015; Nallet et al., 1999; Oliveira-Ferreira et al., 2010). Changes in spontaneous activity are simultaneously recorded in the alternate current (AC) frequency range of the ECoG above ~0.5 Hz. In electrically active tissue, SD typically leads to cessation of the spontaneous activity due to sodium channel inactivation, collapse of ion gradients and suppression of synaptic transmission (Kager et al., 2002; Lindquist and Shuttleworth, 2017; Sawant-Pokam et al., 2017). This has been called spreading depression of cortical activity (Fig. 1B) (Leão, 1944).

Clusters of SDs are typically linked to complex global and/or focal neurological deficits (Dreier and Reifffurth, 2015; Dreier et al., 2006). Serial neuroimaging studies suggested that such clusters are associated with ischemic lesion progression, especially when accompanied by prolonged depression of activity (Dreier et al., 2012; Dreier and Reifffurth, 2015; Dreier et al., 2006). Accordingly, SDs in electrically inactive (= isoelectric) tissue, so-called isoelectric SDs (Fig. 1B), were associated with poor outcome in patients with TBI (Hartings et al., 2011a) and, in parallel, SDs with prolonged depression periods were linked to worse outcome in patients with aSAH (Dreier et al., 2012; Winkler et al., 2017). Temporal clusters of SDs and persistent depression of spontaneous cortical activity can afford even remote detection of ischemic zones because SDs propagate widely (Dreier et al., 2017; Oliveira-Ferreira et al., 2010; Winkler et al., 2017).

Leão, however, also suggested that a single SD-induced spreading depression of activity in normal and eloquent brain tissue is the pathophysiological correlate of the mostly harmless migraine aura (Leão and Morison, 1945). Case series using imaging of changes in regional cerebral blood flow (rCBF) or its surrogates and magnetoencephalography supported this notion (Bowyer et al., 2001; Hadijkhanjik et al., 2001; Olesen et al., 1981; Woods et al., 1994). In previous imaging studies of patients undergoing migraine aura, SD velocities of 2.2 (2.1, 2.5 (1st quartile, 3rd quartile)) and 3.5 ± 1.1 mm/min were measured using either the 123xenon intracarotid injection method or functional MRI of changes in the blood oxygen level-dependent (BOLD) signal (Hadijkhanjik et al., 2001; Lauritzen et al., 1983). In a study of patients with MHS undergoing decompressive hemicraniectomy, SD velocities ranged between 1.7 and 9.2 mm/min using laser speckle imaging of rCBF and imaging of the intrinsic optical signal (IOS) in the operating room (Wozitik et al., 2013). However, in patients undergoing neuro-monitoring in critical care, assessment of SD velocities has not been possible until now. While time-of-SD-arrival differences (TOAD) between adjacent electrodes are known (Strong et al., 2002), the orientation of the SD wavefront relative to the collinear electrode strip and the length of the propagation path along the brain surface between electrodes are unknown. Therefore, in order to estimate SD velocity, we here developed an algorithm for reconstructing SD trajectories of the wave-front’s curvature center based on MRI scans and TOADs between different electrode pairs. We then compared the median algorithm-estimated SD velocities of twelve aSAH patients with different variables indicating the tissue's susceptibility to SD such as the median interval between SDs or the peak numbers of SDs, spreading depressions and isoelectric SDs following the recent recommendations of the Co-Operative Studies on Brain Injury Depolarizations (COSBID) (Dreier et al., 2017). Statistically significant correlations supported the validity of the here proposed method.

2. Materials and methods

2.1. General

Seventy aSAH patients were prospectively enrolled in COSBID between 04/2008 and 09/2012 at two centers (Campus Benjamin Franklin and Campus Virchow Klinikum, Charité University Medicine Berlin, Berlin, Germany) according to the following inclusion criteria: (i) aSAH of World Federation of Neurosurgical Societies (WFNS) grade I–V; (ii) age (≥ 18 years); (iii) ruptured saccular aneurysm proven by computed tomography (CT)-angiography or digital subtraction angiography; (iv) symptom onset within the preceding 72 h; (v) either surgical treatment of the aneurysm via craniotomy or, in coiled patients, burr hole trepanation for placement of a ventilar drain or oxygen sensor, which allows simultaneous placement of a subdural electrode strip (Bruce and Bizi, 2000; Dreier et al., 2009; Eross et al., 2009).
Exclusion criteria for patient monitoring were SAH due to other causes (e.g., trauma, fusiform or mycotic aneurysm), admission in a clinical state with unfavorable prognosis (e.g., wide, nonreactive pupils for >1 h), bleeding diathesis or pregnancy, unavailability of the monitoring equipment and refusal of the patient or legal representative to participate in the study. The research protocol was approved by the local ethics committee of the Charité University Medicine Berlin. Either informed consent or surrogate informed consent was obtained for all patients. Research was conducted in accordance with the Declaration of Helsinki. The subdural electrode strip was targeted to the vascular territory of the aneurysm-carrying vessel because this is often covered with blood and, thus, a predilection site for delayed cerebral ischemia (Dreier et al., 2017).

Twelve of the 70 patients met further screening criteria for inclusion in the present study: (vi) at least one MRI without extensive pathology in the present study: (vi) at least one MRI without extensive pathology that would preclude successful cortical surface reconstruction using FreeSurfer (Martinos Center for Biomedical Imaging, Charlestown, MA, USA, http://surfer.nmr.mgh.harvard.edu/); (vii) at least one CT scan that displayed the collinear electrode strip; and (viii) the presence of SDs hitting at least 3 electrodes of the subdural strip, required for the algorithm. Demographic data for the 12 patients are shown in Table 1.

Aneurysmal SAH was diagnosed through interdisciplinary assessment of CT scans by a neuroradiologist and a neurosurgeon. Hemorrhage was graded according to the original Fisher scale (Fisher et al., 1980), and clinical presentation on admission according to the WFNS scale. A study neurologist or neurosurgeon performed a neurological and general medical evaluation on admission. Baseline demographic data and clinical signs and symptoms of the initial hemorrhage were recorded. The aneurysm was assessed using four-vessel digital subtraction angiography, or a more restricted study when indicated.

After aneurysm treatment by either clip ligation or endovascular coil embolization, all patients were transferred to the intensive care unit where the continuous neuromonitoring data were acquired for up to 15 days. Glasgow Coma Score, blood gases, glucose and electrolytes were documented at least every 6 h. A thorough neurological examination was performed at least daily. Oral nimodipine was given prophylactically. Transcranial Doppler-sonography was performed daily as described previously (Dreier et al., 2012). MRI exam protocol included a T2-weighted fluid-attenuated inversion recovery (FLAIR) sequence, a T1-weighted sequence pre- and post-gadolinium, a T1-weighted 3D high resolution sequence (i.e. magnetization prepared rapid gradient echo (MPRAGE)), and a diffusion weighted imaging (DWI) sequence. The first MRI was performed 24–48 h after aneurysm treatment in order to assess the initial structural brain injury, the second MRI around day 7 and the third one on the day of electrode withdrawal (~day 15) to assess the occurrence of delayed ischemic strokes during the ECoG monitoring period. In addition, serial CT scans were performed at times of clinical deterioration. At the conclusion of the monitoring period, the electrode strip was removed at the bedside by gentle traction.

2.2. ECoG acquisition

ECoG was recorded with a subdural collinear electrode strip carrying six disk-shaped platinum electrodes, each 4 mm in total diameter with 2 mm diameter of the contact surface and 10 mm distance between electrode centers (Ad-Tech Medical, Racine, Wisconsin, USA). The strip was placed on cortex after craniotomy (Fig. 1A) or through an extended burr-hole and connected to a GT-205 amplifier (0.01–100 Hz). The near-DC/AC-ECoG signal was sampled at 200 Hz with a PowerLab 16/SP analog/digital converter and LabChart software (all by ADInstruments, New South Wales, Australia). In parallel the electrode strip was connected to a DC-coupled BrainAmp amplifier (0–100 Hz) (Brain Products GmbH, Munich, Germany) and the
recorded DC/AC-ECoG data imported into the LabChart software. All recorded signals were analyzed using monopolar montage against a subdermal platinum needle electrode placed over the hemisphere ipsilateral to the recording strip. DC/AC-ECoG data were used whenever possible. However, they were replaced by the near-DC/AC-ECoG data in case of recording problems, such as, for example, saturation of the DC amplifier due to drifting. Following current recommendations, SDs in electrically active tissue received the epithet “spreading depression”. By contrast, SDs measured in a zone of electrically inactive tissue were denoted with the adjective “isoelectric” (Fig. 1B) (Dreier et al., 2017).

2.3. Overview of the modeling approach

In order to reconstruct the potential SD trajectories and velocities based on the TOADs, we chose to restrict the trajectories to the ones for which the following assumptions held: (1) following the typically observed pattern in animal experiments (Dahlem and Müller, 2004; Kaufmann et al., 2017; Kneer et al., 2014; Santos et al., 2014; Scholl et al., 2017), the shape of the SD wave-front can be approximated by a circular arc that propagates tangentially along the discretized brain surface; (2) the radius of this arc is fairly constant; (3) the velocity of the arc’s center is fairly constant; (4) the SD center travels on geodesics (shortest paths) from the point where it reaches one electrode to the point at which it reaches the next electrode in temporal succession.

The trajectory reconstruction was a model fitting procedure that searched in the class of models consistent with the assumptions as given above. These assumptions are simplifying. Therefore, there were real cases in which no model existed for the observed data and the trajectory reconstruction hence failed. In practically all successful searches the assumptions were not constraining enough to reduce the ambiguity to the point where a single trajectory fit the data; therefore, the reconstruction produced groups of many potential trajectories. In this context, we use “simulation” to describe the process of iterative extension (from electrode to electrode, in the order a given SD reached the electrodes) of the set of potential trajectories, because it corresponds to simulating/extending all those trajectories in parallel - not completing one trajectory to the last electrode hit before starting describing the key points of the next possible trajectory. The logical parallelism implemented by grouping the partial trajectories by their end point accelerates the search of the valid, observations-conform trajectories in the huge space of the geometrical trajectories. Assumption number (4) above is very strong, and is equivalent to requiring that the SD wave-front travelled at the lowest velocity to explain the observed TOADs. Without this assumption, any trajectory would be possible, although many of those would require high velocities that are not biologically plausible.

Between each two electrodes A and B (identified by their position on the cortex), hit in succession by the SD, all trajectories on the discretized cortical surface were (virtually) retained that were between the vertex (Fig. 1C) a in the neighborhood of A and the vertex b in the neighborhood of B (a and b were candidate positions for the center of the SD), provided that the distance between a and A approximately matched the distance between b and B (their difference was upper-bounded by a search parameter, R_diff). This condition enforced the second assumption above. The velocity estimated for the trajectory segment a-b was given by the times the SD hit A and B and the geodesic distance on the cortical surface between a and b, according to our fourth assumption.

Trajectory segments a-b (first electrode pair) and b-c (second electrode pair) were joined into an a-c trajectory if and only if their velocities approximately matched (maximum difference was upper-bounded by a search parameter, V_diff). This filtering enforced our third assumption. The process was repeated iteratively, extending the partial trajectories with every new electrode-to-electrode hop and retaining only those trajectories for which a suitable extension existed.

2.4. Imaging data preprocessing

Cortical reconstruction, including both gyri and sulci, was performed with FreeSurfer. In brief, this processing comprises removal of non-brain tissue (Segonne et al., 2004), automated Talairach transformation, segmentation of the subcortical white matter and deep grey matter volumetric structures (Fischl et al., 2002), intensity normalization, tessellation of the grey/white matter boundary, topology correction and surface deformation (Dale et al., 1999; Fischl and Dale, 2000) to produce representations of cortical thickness, which are calculated as the closest distance from the grey/white matter boundary to the grey matter/cerebrospinal fluid boundary at each vertex on the surface. The reconstructed discrete brain surface generated by FreeSurfer remains identical to the geometry of the resampled MRI images. The reconstructed cortical surface can then be converted into a geometric mesh by exporting it as an ASCII file with FreeSurfer’s single-line command. The structure of such ASCII files is very similar to the output format, which we used for the simulation after simple conversion with Matlab. As a result, the cortical surface is mathematically modeled by a grid of unequal triangles (= tessellation) (Fig. 1C). Such a surface is termed a discrete non-Euclidean or discrete two-manifold surface. The distance between two points (vertices) on a two-manifold surface is given as the shortest distance along the surface, called geodesic distance. We used Matlab (MathWorks Inc., Natick, MA, USA) for data processing and simulation, and the Dijkstra shortest path algorithm (contained in the BOOST Graphic Library (BGL) toolbox) for computing geodesic distances.

The electrodes from the CT images were obtained by simple thresholding since the brain tissue ranges between 20 and 120 Hounsfield Units (HU) and metallic parts have values over 3000 HU. The surface of isolated electrodes was geometrically reconstructed by Matlab’s ‘isosurface’ function. To localize the electrodes on the geometric mesh representing the cortical surface, we first rigidly co-registered the CT scan with electrodes to the MRI scan reoriented and resampled with FreeSurfer. Secondly, we estimated the centroids of every reconstructed electrode using Matlab’s ‘regionprops’ function. Finally, we identified the vertices on the cortical mesh with the closest Euclidean distance to the centroids. The final result was a discrete cortical surface in form of a triangulated mesh with the known electrode positions. This provided the spatial information for the simulation.

2.5. Geometric data preprocessing

We up-sampled the geometric mesh because of the rounding error that resulted from the inequality of triangles. To reduce the long computation time we ran the simulation using a patch (Fig. 1C). This patch was cut from the cortical mesh with 30 mm radius around each electrode. To ensure that the space between electrodes and any point on the patch’s edge was never smaller than 30 mm, we computed the shortest paths between each neighboring electrode and checked the distances for every vertex on the path. Every patch was up-sampled by doubling the number of the vertices twice (Table 1) according to the spline interpolated 4-split method (Shirman, 1990). Up-sampling slightly increased the number of could-fit solutions and improved the results from the path search by Dijkstra’s algorithm. The increase in number of vertices by up-sampling for factor K also increased the computation time required by a polynomial algorithm of order P for K^P times. The original mesh generated by FreeSurfer often contains small unnaturally edgy areas. We applied the Laplace operator to smooth the curvature and set the parameters in order to retain the original lengths of the edges and, thus, the original topology (Desbrun et al., 1999). Only the angles between edges were changed so that the smoothing predominantly affected the global curvature. From a vertex selected as a disc’s center, we calculated shortest paths to every other vertex of the patch.
Table 1
Demographic data are given in the left part. ACA, anterior cerebral artery; ACoA, anterior communicating artery; MCA, middle cerebral artery; PGoA: posterior communicating artery. The patch section shows the properties of the geometric mesh for each patient. The variations in total patch area in the first column reflect the individual variations of surface folding at a constant distance (30 mm) between electrodes and any vertex on the edge of a given patch. The second column contains the cortical thickness in the patch. The quality of the geometric discretization before and after the up-sampling is reflected by the vertex count in the third and fourth column, and in the mean triangle edge length in the fifth and sixth column, respectively. The hemisphere section shows individual differences in total surface area and cortical volume of the individual brain hemispheres from which the single patches were extracted. The ECoG section gives the number of all recorded and analyzed SDs of the 12 patients and those SDs with at least three active electrodes, which was required for the algorithm.

| No. | Age (years), sex | WFNS grade | Fisher grade | Location of aneurysm | Intervention | Early focal brain lesion at electrode strip (either infarct or ICH) | Delayed focal brain lesion (days: 0-3) | Recording time [h] | Median SD velocity based on the reduced hit-sequences | Median interval to previous SD | PTDDD Peak total number of SDs of a recording day | Peak total number of spreading depressions of a recording day | Peak total number of isoelectric SDs of a recording day |
|-----|-----------------|-------------|--------------|----------------------|--------------|-------------------------------------------------|------------------------------------|---------------------|---------------------------------|-----------------------------|-----------------------------------------------|-----------------------------------------------|-----------------------------------------------|
| 1   | 56, f 4         | 3           | MCA          | Clipping             | y            | y y y                                           | 287.1                              | 4.11                 | 34.8                            | 100.5                       | 36.1                           | 36.1                           | 5.0                          |
| 2   | 50, f 2         | 3           | PCoA         | Clipping             | n            | n n                                            | 253.1                              | 2.42                 | 158.1                           | 106.4                       | 11.9                           | 11.9                           | 0.0                          |
| 3   | 68, f 4         | 3           | PCoA         | Coiling             | y            | y y                                            | 258.7                              | 5.1                  | 27.0                            | 1103.8                      | 47.1                           | 13.2                           | 36.1                         |
| 4   | 50, f 3         | 3           | ACoA         | Clipping             | y            | y y                                            | 228.7                              | 6.72                 | 58.2                            | 647.3                       | 34.7                           | 25.1                           | 9.6                          |
| 5   | 64, f 4         | 3           | ACoA         | Clipping             | n            | n n                                            | 278.2                              | 3.2                  | 136.2                           | 158.3                       | 12.5                           | 12.5                           | 1.0                          |
| 6   | 58, f 1         | 3           | ACoA         | Clipping             | n            | y n                                            | 287.3                              | 3.84                 | 19.8                            | 389.6                       | 37.1                           | 24.7                           | 12.4                         |
| 7   | 48, m 5         | 3           | ACoA         | Coiling             | n            | y y                                            | 303.8                              | 2.43                 | 408.5                           | 206.5                       | 20.1                           | 20.1                           | 0.0                          |
| 8   | 31, m 2         | 3           | ACoA         | Clipping             | n            | n y                                            | 202.1                              | 2.6                  | 616.6                           | 410.0                       | 20.9                           | 20.9                           | 5.1                          |
| 9   | 47, f 4         | 3           | MCA          | Clipping             | y            | y y                                            | 269.3                              | 3.22                 | 103.8                           | 502.4                       | 21.4                           | 13.2                           | 21.4                         |
| 10  | 44, f 4         | 3           | PCoA         | Clipping             | y            | y y                                            | 239.9                              | 3.28                 | 31.8                            | 1407.0                      | 53.5                           | 30.2                           | 53.5                         |
| 11  | 70, f 4         | 3           | ACoA         | Clipping             | y            | y y                                            | 245.1                              | 2.47                 | 138.3                           | 184.3                       | 8.1                            | 8.1                            | 4.0                          |
| 12  | 61, f 1         | 3           | PCoA         | Clipping             | n            | y n                                            | 236.1                              | 3.18                 | 267.2                           | 179.0                       | 12.1                           | 12.1                           | 7.1                          |

| No. | Patch | Hemisphere | Area [cm²] | Thickness [mm] | Vertex count | Vertex count up-sampled | Mean edge length before up-sampling [mm] | Mean edge length after up-sampling [mm] | Recorded SDs | Simulated SDs |
|-----|-------|------------|------------|---------------|--------------|-------------------------|------------------------------------------|----------------------------------------|--------------|----------------|
2.6. SD trajectory search procedure – parameters and details

The search for candidate trajectories of a given SD was performed in a stepwise fashion by considering sub-trajectories for each pair of neighboring electrodes that showed a time difference between SPC onset indicating a time of arrival delay of the respective SD (= TOAD). The number of steps varied depending on the number of TOADs which ranged between two and five, or, respectively, a minimum of three and a maximum of six active electrodes. A variable wave-front between sub-trajectory sets was possible within the limits given by the maximally allowed wave-front curvature radius difference (R_diff as introduced above) and wave-front velocity difference (V_diff as introduced above), respectively. Since the path search presumed a constant curvature radius and velocity of the wave-front within a single sub-trajectory, R_diff and V_diff allowed the simulation to compensate for small changes in velocity, wave-front curvature and spatial shear between two neighboring sub-trajectory sets and a given electrode. A stepwise search for the sets of sub-trajectories between individual electrode pairs also allowed the simulation to compensate for more complex shapes of trajectories than the simple ones based exclusively on the shortest path search.

Pre-set maxima of R_diff and V_diff were used to filter the connecting candidates belonging to two consecutive sub-trajectory sets. Candidates exceeding either of these two parameters were excluded. Given a sufficiently large maximum radius, the region adjacent to each electrode was considered with all vertices within the preset disk radius. The bipartite distances between each pair of vertices were pre-computed and the trajectories were then iteratively “grown” starting from zero-length partial trajectories and proceeding stepwise according to the order of incidence. The zero-length trajectory seeds were all the vertices in the region of the first electrode hit by the SD (step 1). At step N all vertices around the corresponding electrode were considered as possible disk centers separately for each velocity subrange. All possible pairs between currently possible partial trajectory endings around the (N − 1)th hit electrode and the vertices in the region of the Nth hit electrode were considered with exclusion of pairs at which (i) the segment velocity did not fit the fixed velocity subrange, or (ii) the distance to the (N − 1)th electrode differed too greatly from the distance to the Nth electrode.

The accepted velocity range was set between 0 and 15 mm/min per sub-trajectory set with a tolerance range V_diff between the sets. V_diff was set in turn at three different values (0.25, 0.5, and 1 mm/min). The accepted wave-front curvature radius was set between 0 and 5 mm per single set with a tolerance range R_diff between the sets. R_diff was set at two different values (0.5 and 1 mm). Finer granulation of V_diff values was not necessary since the number of solutions dropped dramatically at lower tolerance values. The wave-front radii of the simulation tended to be very conservative in contrast to the velocity between the sub-trajectory sets. The reason to set a lower limit for R_diff at 0.5 mm was the average edge length at around 0.25 mm as shown in Table 1. The upper limits for both V_diff (in mm/min) and R_diff (in mm) were set at value 1. We repeatedly simulated for every pair of control parameters V_diff and R_diff, respectively. These numerical choices were a trade-off between computation time and the method’s sensitivity. Furthermore, the selected upper limit was a trade-off between realistic and unrealistic trajectory candidates based on our knowledge from animal experiments. Every trajectory found was possible under the modeling assumptions and consistent with the observed TOADs between electrode pairs.

The velocity of every SD was then calculated as a weighted arithmetic mean of the estimated velocities over the electrode-to-electrode segments, with the TOADs as weights. The SD velocity over an electrode-to-electrode segment was calculated as the arithmetic mean of the sub-trajectories’ velocities on that segment, after quantization toward the central values of the velocity sub-ranges. The velocity sub-ranges were defined for V_diff = 1 mm/min as ± 1 mm/min intervals.

| No. | Patch | Area [mm²] | Thickness [mm] | Hemisphere | Average edge length before up-sampling [mm] | Average edge length after up-sampling [mm] | Simulated SDs | Recorded SDs |
|-----|-------|-----------|---------------|------------|------------------------------------------|------------------------------------------|-------------|-------------|
| 10  |       | 71.92     | 0.77          |            | 146.79                                   | 0.46                                     | 1037        | 611 (58.9%) |
| 11  |       | 75.48     | 0.74          |            | 146.16                                   | 0.43                                     |             |             |
| 12  |       | 72.46     | 0.62          |            | 146.16                                   | 0.43                                     |             |             |

Table 1 (continued)

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around fifteen central values from 1 to 15 mm/min in 1 mm/min steps, resulting in 2 mm/min intervals with 50% overlap between every two neighbor ones.

### 2.7. Validation of results

In order to validate the exploratory accuracy of the simulation model, we withheld part of the available information by removing one of the electrode hits by a given SD and estimated the electrode hit times on a random selection of 100 trajectories from the ones proposed, taking the predicted SD radius and the predicted SD propagation velocity into account. Some of the trajectories generated for this reduced set of electrodes could no longer make a hit on the withheld electrode. The first measure of quality of the modeling was the percentage of cases where the predicted trajectory still predicted the withheld electrode being reached by the SD area during its propagation. The second measure of quality concerned the precision with which the electrode hit time matched the observed time for the withheld electrode corrected for velocity. This can be given as a spatial error between the trajectory from the actual simulation and the validation trajectory.

### 2.8. Statistics

Data are given as median (1st, 3rd quartile). In the figures, the whiskers (error bars) above and below the box indicate the 90th and 10th percentiles. Statistical tests are given in the text. 

**Table 2**

This table shows the number of successfully reconstructed SD events per subject and per combination of simulation parameters. 

| Subject | Simulated SDs | $R_{diff} = 1$ | $V_{diff} = 1$ | $R_{diff} = 0.5$ | $V_{diff} = 0.25$ | $R_{diff} = 0.5$ | $V_{diff} = 0.25$ |
|---------|---------------|----------------|---------------|----------------|----------------|---------------|----------------|
| **Full hit-sequences** |
| 1 | 129 | 112 (86.8%) | 105 (81.4%) | 94 (72.9%) | 112 (86.8%) | 104 (80.6%) | 88 (68.2%) |
| 2 | 7 | 4 (57.1%) | 4 (57.1%) | 3 (42.9%) | 4 (57.1%) | 4 (57.1%) | 3 (42.9%) |
| 3 | 86 | 65 (75.6%) | 32 (37.2%) | 26 (30.3%) | 65 (75.6%) | 31 (36.0%) | 26 (30.3%) |
| 4 | 25 | 23 (92.0%) | 21 (84.0%) | 21 (84.0%) | 22 (88.0%) | 21 (84.0%) | 21 (84.0%) |
| 5 | 45 | 25 (55.6%) | 24 (53.3%) | 24 (53.3%) | 26 (57.8%) | 24 (53.3%) | 25 (51.1%) |
| 6 | 41 | 32 (78.0%) | 30 (73.2%) | 30 (73.2%) | 32 (78.0%) | 30 (73.2%) | 29 (70.7%) |
| 7 | 5 | 4 (80.0%) | 3 (60.0%) | 2 (40.0%) | 4 (80.0%) | 2 (40.0%) | 2 (40.0%) |
| 8 | 31 | 30 (96.8%) | 29 (93.5%) | 26 (83.9%) | 30 (96.8%) | 28 (90.3%) | 26 (83.9%) |
| 9 | 72 | 27 (37.5%) | 25 (34.7%) | 20 (27.8%) | 27 (37.5%) | 25 (34.7%) | 20 (27.8%) |
| 10 | 124 | 67 (54.0%) | 65 (52.4%) | 61 (49.2%) | 67 (54.0%) | 64 (51.6%) | 60 (48.4%) |
| 11 | 25 | 15 (60.0%) | 13 (52.0%) | 12 (48.0%) | 15 (60.0%) | 13 (52.0%) | 12 (48.0%) |
| 12 | 21 | 19 (90.5%) | 18 (85.7%) | 18 (85.7%) | 19 (90.5%) | 18 (85.7%) | 18 (85.7%) |
| **All** | 631 | 374 (61.2%) | 314 (51.4%) | 283 (46.3%) | 374 (61.2%) | 309 (50.6%) | 274 (44.8%) |

In 12 patients, our simulations found potential trajectories (‘could fit’ solutions) for 374 of 611 SDs (61.2%) using $V_{diff} = 1$ mm/min and $R_{diff} = 1$ mm. Using lower tolerance values the number of successful simulations decreased (Table 2). The median count of ‘could fit’ solutions per SD was $4.18 \times 10^{16}$ (1st quartile, 3rd quartile: $1.43 \times 10^{15}$, $3.58 \times 10^{16}$ trajectories). 

For the 374 SDs, the median of the trajectory velocity was 3.6 (2.9, 5.2) mm/min. Since the simulation found ‘could fit’ trajectories for only measurements within the same patient. (2) In some cases, pooled analyses with “n” equal to the number of SDs were performed in addition when this provided additional information about the structure of the data as explained in the discussion. In these analyses linear mixed models were then applied with patient as random factor. To approach normal distribution, values were logarithmically transformed for these analyses related to SDs.

Standard ECoG analyses (Dreier et al., 2017) were performed by MW, blinded to the clinical courses, neuroimaging findings and simulation results. VK assessed the serial neuroimaging for type of lesion, location with respect to electrode strip and time point of infarct or hemorrhage occurrence, blinded to the clinical courses, ECoG analyses and simulation results. DM determined the TOADs and analyzed the DC data under supervision of JD, and developed the algorithm in collaboration with CG and MD. DM’s analysis of the DC shifts was based on the recently published criteria by Hartings and colleagues (Hartings et al., 2017a). DM, MW and VK were blinded to each other. JD and PM performed the statistical analysis.

### 3. Results

In 12 patients, our simulations found potential trajectories (‘could fit’ solutions) for 374 of 611 SDs (61.2%) using $V_{diff} = 1$ mm/min and $R_{diff} = 1$ mm. Using lower tolerance values the number of successful simulations decreased (Table 2). The median count of ‘could fit’ solutions per SD was $4.18 \times 10^{16}$ (1st quartile, 3rd quartile: $1.43 \times 10^{15}$, $3.58 \times 10^{16}$ trajectories). Fig. 2 shows the results of the trajectory search for three SDs in patient 1. In these examples, overlapping ‘could fit’ trajectories are color-coded as a heatmap.

For the 374 SDs, the median of the trajectory velocity was 3.6 (2.9, 5.2) mm/min. Since the simulation found ‘could fit’ trajectories for only
from 374 to 502 of 611 (82.2%) for V_di and R_di SD in which the depolarization hit-sequences occurred (Fig. 3A). Branching describes a type of simplification in which the simulation found two branches were of the same length. Under this condition, the model moved either the shorter branch or the one that occurred later when the depolarization hit-sequences were not in the middle of the strip (electrode 2 (E2), E3, E4 or E5) and continued to the middle of the strip (electrode 2 (E2), E3, E4 or E5) and continued. The third and most important measure of quality was the biological validation using variables that were not involved in the simulation procedure. For this purpose, we investigated whether the median estimated velocities based on the full and reduced hit-sequences correlated with typical variables related to the tissue's susceptibility to SD. Notably, a high negative correlation was found between the median SD velocity based on the reduced hit-sequences and the median interval between SD and the previous SD (Fig. 4A). Significant correlations were also found with the peak total SD-induced depression duration of a recording day (PTDDD) (Dreier et al., 2017) and the peak numbers of SDs (Fig. 4D), spreading depressions (Fig. 4E) and isoelectric SDs (Fig. 4F) of a recording day. No significant correlation was found with age. The median SD velocities based on the full hit-sequences correlated significantly with the median interval between SD and the previous SD (Spearman rank order correlation coefficient: \( \rho = 0.58, P < 0.045, n = 12 \) patients) and the peak number of SDs of a recording day (Spearman rank order correlation coefficient: \( 0.62, P = 0.031, n = 12 \) patients) but not with the other variables (number of correlations = 3). The median wave-front velocities assuming an ideal linear spread along the recording strip, using either the inter-electrode space of 10 mm or the shortest path along the brain surface, did not correlate with any of these variables (Fig. 4B).

We also investigated whether SD velocities based on the reduced hit-sequences and the SD interval between SD and the previous SD correlated when all successfully simulated SDs were pooled. Though this correlation was statistically significant, it was relatively weak (Spearman rank order correlation coefficient: \( -0.10, P = 0.035, n = 497 \) SDs). Moreover, in the double logarithmic linear mixed model, taking into account single SDs and correcting for cluster effects, influences of speed and SD interval were not significant (\( P = 0.113 \)).

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The median longest DC shift among all electrodes of a given SD was 5.9 (4.1, 7.2) mV (n = 12 patients) but not with the other variables (number of correlations = 3). The median wave-front velocities assuming an ideal linear spread along the recording strip, using either the inter-electrode space of 10 mm or the shortest path along the brain surface, did not correlate with any of these variables (Fig. 4B).
patients). DC duration and DC rise time showed a strongly positive correlation (Fig. 4G). This was also highly significant for the pooled data (Fig. 4H); the linear mixed model revealed a highly significant correlation (Fig. 4G). This was also highly significant for the pooled data (Fig. 4H); the linear mixed model revealed a highly significant correlation with previous observations in patients with aSAH (Winkler et al., 2017). Moreover, they showed a significantly higher number of DC shifts lasting longer than 4 min (124 out of 174 isoelectric SDs (71.3%) versus 80 out of 336 spreading depressions (23.8%), \( P < 0.001, \) Chi-Square Test) similar to previous observations in patients with aSAH and TBI (Hartings et al., 2011b; Oliveira-Ferreira et al., 2010).

Isoelectric SDs with DC shifts > 4 min were only observed in patients in whom the electrode strip was overlying either a primary or secondary focal brain lesion (n = 7) but not in patients in whom the electrode strip was located remote from focal brain lesions (n = 5) (\( P = 0.010, \) Mann-Whitney Rank Sum Test, Fig. 5B, Table 1). The simulation based on the reduced hit-sequences failed to find possible SD trajectories in a significantly higher proportion of these isoelectric SDs with DC shift durations > 4 min compared with the remaining SDs (\( P < 0.001, \) Chi-Square Test, Fig. 5C). This suggests a higher proportion of SDs with more complex propagation patterns in this subgroup. Figs. 5D and 6 illustrate these results using example SDs measured in the recording area during the evolution of delayed ischemic infarcts. The median velocity of isoelectric SDs with DC shifts > 4 min was 3.4 (2.5, 3.9, n = 6 patients) mm/min and not different to that of the remaining SDs.

In addition to the occurrence of isoelectric SDs with DC shift durations > 4 min, recording areas with primary or secondary focal brain lesions (n = 7 patients) also differed significantly from recording areas remote from focal brain lesions (n = 5 patients) in the PTDDD (502.4 (355.2, 875.6) versus 179.0 (158.5, 206.5), \( P < 0.018, \) Mann-Whitney Rank Sum Test), whereas differences in the other variables, as shown in Fig. 5A, did not reach statistical significance.
4. Discussion

Collinear subdural electrode strips are used in neurocritical care rather than electrode grids though grids would ease the calculation of SD velocities. This is because strips can be removed at the bedside by gentle traction whereas withdrawal of grids requires another neurosurgical intervention in the operating room with the known potential risks for the patient. Yet in order to estimate SD velocities in neurocritical care, an algorithm was developed here. For the biological validation of this, we exploited the well-established relationship between SD velocity and the tissue’s susceptibility to SD based on animal experiments. For example, the hippocampal area CA3 of immature rabbits is less susceptible to SD than area CA1, which was attributed to a higher Na,K-ATPase activity of area CA3 (Haglund et al., 1985). The Na,K-ATPase blocker ouabain was then applied at a concentration in which it completely inhibits the α2/α3 isoforms and partially inhibits the α1 isoform (Major et al., 2017). This not only evened out the difference in SD susceptibility between the two areas but also facilitated the invasion of SD from area CA1 to area CA3 (Haglund and Schwartzkroin, 1990). A significantly increased SD velocity in parallel with an increased susceptibility to SD was found in genetically modified mice, in which astrocyte-directed inactivation of Cx43 decreased astrocytic gap junctional communication (Theis et al., 2003). A similar correlation was also found in mouse models of familial hemiplegic migraine type 1, a rare Mendelian model disease of migraine with aura (Cain et al., 2017; van den Maagdenberg et al., 2004). Further, this link was noted in acquired metabolic disruptions. For example, pathologically low concentrations of nitric oxide (NO) enhanced the susceptibility to SD in vivo and in brain slices in rodents (Petzold et al., 2008; Petzold et al., 2005), and conversely, elevated NO reduced the propagation velocity of
SD in the chicken retina (Ulmer et al., 1995). Because of the experimentally and clinically well-established decline in NO availability following aSAH, the latter may have particular relevance to our patient population investigated in this study (Dreier, 2011; Pluta et al., 2009). Also various drugs including anesthetics with concurrent effects on SD velocity and susceptibility might have relevance to our patient population (Dhir et al., 2012; Kudo et al., 2008; Marrannes et al., 1988). The same applies to female hormones and their modulatory effects because the majority of aSAH patients are women (Eikermann-Haerter et al., 2009). Finally, SD velocity and susceptibility showed concurrent changes with age. Thus, young rats were more prone to develop SDs and SD velocities were higher (Hablitz and Heinemann, 1989; Hablitz and Heinemann, 1989; Maslarova et al., 2011; Menyhart et al., 2015). In the present study, the influence of age was however not statistically significant. This may have several reasons: only adults were included, the study was relatively small and other modulatory factors may be stronger in this severely diseased patient population.

The propagation process of SD is discussed controversially as reviewed recently (Dreier and Reifthuth, 2015). Accordingly, also the question of how SD velocity and susceptibility are mechanistically linked has not been fully elucidated. However, the propagation process is generally regarded as a reaction/diffusion rather than a simple diffusion mechanism (Zandt et al., 2013). It is thus assumed that neurons release neuroactive substances such as potassium or glutamate, which diffuse to adjacent neurons where they trigger a self-propagating regenerative process. This concept entails the re-induction of SD at every spot in the tissue reached by the SD wave and might therefore well explain why SD velocity and susceptibility correlate.

4.1. SD propagation shows anisotropy

The propagation of SD usually shows deviations from concentricity in both liss- and gyrencephalic brains (Kaufmann et al., 2017; Santos et al., 2017). An apparent tropism of SD for superficial layers, rather than deeper layers of cortex, has been repeatedly reported (Basarsky et al., 1998; Bogdanov et al., 2011; Grafstein, 1956; Herreras and Somjen, 1993; Richter and Lehmenkühler, 1993). Cytoarchitectonic peculiarities also affect the propagation in the horizontal plane (Chen et al., 2006; Eiselt et al., 2004; Leiô and Morison, 1945). The exact mechanisms underlying this anisotropy have remained enigmatic but myelin content, neuron-astrocyte ratio and vascular anatomy have all been advanced as possible explanations (Fujita et al., 2016; Merkler et al., 2009; Santos et al., 2014). The present observations in humans are in line with this anisotropy in animals. In particular, we found a relatively high proportion of SDs that branch or propagate heterogeneously, as found previously with imaging of the gyrencephalic swine brain (Santos et al., 2014). The frequency of branching necessitated use of the algorithm with the reduced rather than the full SD hit-sequences.
Intervals between repeated SD waves also vary with propagation patterns and susceptibility. For instance, the first SD in the otherwise healthy gyrencephalic brain of cats usually spreads with an elliptical wavefront over the ipsilateral cerebral hemisphere whereas succeeding SDs often remained within the originating gyrus, were slower, more fragmented and varied in number (James et al., 1999). A similar decline in susceptibility to SD after the first event was also observed in rodents (Chen et al., 2006; Kaufmann et al., 2017; Sukhotinsky et al., 2011). Accordingly, imaging studies in migraineurs suggested that an isolated SD often affects large parts of the hemisphere (Hadjikhani et al., 2001; Lauritzen, 1994; Olesen et al., 1981; Woods et al., 1994) although the representation fields responsible for the visual percept seem to be relatively small (Dahlem and Hadjikhani, 2009). However, repeated SDs in humans were often restricted to a single gyrus as assessed with laser speckle imaging in the operating room in patients with MHS (Woitzik et al., 2013).

Nonetheless, velocity and interval to the previous SD correlated only weakly in our study when every SD was considered in form of a pooled SD-related (versus patient-related) analysis. A strong correlation was merely found between these variables in the patient-related analysis based on medians, thus averaging the data. This result supports strong constitutional (e.g. genetic, disease- or drug-related) components with general influence on both variables rather than a direct dependence between these two variables. In this context it is worth noting that the recorded interval to the preceding SD depends above all on the local conditions where a given SD arises (Dreier et al., 2017; Winkler et al., 2017), but depends only secondarily on the transmission of the SD from its site of origin to the recording site. By contrast, the recorded velocity is merely determined by the conditions at the recording site.

In contrast to the correlation of velocity and interval to the previous SD, not only medians of DC durations and DC rise times but also pooled data showed highly significant correlation. This suggests a direct dependence of these two variables, which are merely determined by the local tissue conditions at the recording site. This statistical result is consistent with observations that the ultraslow negative potential of prolonged SDs does not usually ascend abruptly in subdural recordings (Dreier et al., 2017).

4.2. More complex propagation patterns of isoelectric SDs with prolonged DC shifts

It has been established in numerous animal studies that the local
duration of the DC shift is energy-dependent, indicates the local duration of the cytotoxic edema and, hence, the risk of injury at the recording site (Back et al., 1994; Dijkstra et al., 1999; Hartings et al., 2003; Nallet et al., 1999; Oliveira-Ferreira et al., 2010). Moreover, it has been noted for decades that zones with prolonged negative DC shifts become electrically inactive (Astrup et al., 1981; Hossmann, 1994; Koroleva and Bures, 1996; Oliveira-Ferreira et al., 2010). For the first time, we here provide corresponding statistical evidence in patients, based on MRI findings and ECoG recordings, that isoelectric SDs with prolonged DC shifts exclusively occur in recording areas undergoing irreversible damage. The velocity of such SDs was not significantly different from the velocity of other SDs. Largely overlapping SD velocities between well-nourished and energetically compromised tissue correspond well with previous in vitro and in vivo studies in rodents (Aitken et al., 1998; Bere et al., 2014a; Farkas et al., 2010; Jarvis et al., 2001). These findings render it unlikely that changes in SD velocity are useful as an early warning sign of impending ischemia in the individual patient. However, trajectory reconstruction of those SDs failed significantly more often, according to the notion based on animal experiments that their propagation paths can become exceedingly complex (Bere et al., 2014b; Nakamura et al., 2010; Scholl et al., 2017). In concert with other parameters such as the SD frequency and the durations of depressions and DC shifts (Dreier et al., 2017), this could possibly be used as a tool to monitor alarming development in the individual patient. The strongest deviation from concentricity is their reverberation due to continuous cycling. This was first demonstrated in rats in vivo by Shibata and Bures who performed a surgical lesion of the cortex to impose a pathway of cycllical propagation (Shibata and Bures, 1972). A similar approach also resulted in cycling in the isolated chicken retina (Martins-Ferreira et al., 1974). Cycling of SD was moreover observed around electrically stimulated areas, epileptic foci or ischemic zones (Koroleva and Bures, 1979; Nakamura et al., 2010). It has been advanced as an explanation for the periodicity in clusters of SDs (Nakamura et al., 2010; Santos et al., 2017; Scholl et al., 2017).

4.3. Limitations

A limitation of our study was the small size with 12 patients. This was due to the fact that the MRI segmentation sequence of FreeSurfer is optimized for analyzing human brains free of pathology. However, an exceedingly large number of the 70 screened patients had significant pathologies including grey matter edema or missing parts of the skull. This often rendered FreeSurfer unable to geometrically reconstruct the cortical surface. Even manual correction of the segmentation frequently failed. Moreover, the simulation required at least three active electrodes but patients often showed either no SDs or SDs hitting less than three recording electrodes.

We should also like to indicate that we only used a simple arc as wave front on the discretized brain surface although the front of SD typically shows a depolarized zone of time-varying width both in liss- and gyrencephalic animal cortex whose spatial extent locally depends (i) on the necessity for the energy-dependent repolarization after the local SD onset, (ii) on the local velocity, and (ii) probably also on the curvature of the local gyriification pattern (Dahlem and Müller, 2004; Kaufmann et al., 2017; Kneer et al., 2014; Santos et al., 2014; Scholl et al., 2017). It is, however, unlikely that the widths of the depolarized zones significantly impact on the velocities because, in animals, SDs under normoxic and hypoxic conditions strongly differ in widths of the depolarized zones but SD velocities are similar (Aitken et al., 1998). Yet the shape of the wave fronts might still deviate more from a simple arc in zones undergoing injury. This could be among the reasons why the simulation more often failed to find possible SD trajectories for isoelectric SDs with DC shifts > 4 min than for other SDs. The class of trajectories that we considered was limited according to the four assumptions we made. Whereas the first three assumptions have a biological justification, assumption (4) that the trajectory can abandon a geodesic ("change direction") only when the SD hits an electrode is rather artificial and is a compromise between the limited data available for model fitting and the richness of the models.

Another limitation was that the onset of SD at a given electrode can be subjective. However, all simulations were completed before the statistical analysis started. Further, simulation and statistical analysis were performed by different investigators so that the statistical analysis could not influence the TOADs on which the simulations were based.

5. Conclusions

We established a novel algorithm enabling the estimation of SD velocities in patients monitored in neurocritical care. The validity of the algorithm was supported by the fact that the fundamental relationship between SD velocity and susceptibility was similar to that reported in experimental animals. Our findings establish the opportunity to exploit this variable as part of the multimodal assessment in neurocritical care. For example, it was found in previous studies that the peak number of SDs of a recording day was significantly higher in patients who later developed post-hemorrhagic epilepsy (Dreier et al., 2012) and that isoelectric SDs and a high PTDDD were associated with poor patient outcomes (Dreier et al., 2011a; Winkler et al., 2017). Based on the correlations identified here, it would be interesting to investigate whether a high median SD velocity is also associated with these risks. However, in order to render this variable available to a larger patient population, better methods for geometrical reconstruction of brains with severe pathology would be required.

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Curriculum Vitae

Mein Lebenslauf wird aus datenschutzrechtlichen Gründen in der elektronischen Version nicht veröffentlicht.
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List of publications

Thesis for Master of Science (Magister rerum naturalium; Mag. rer. nat)

Prächirurgische Gedächtnislateralisation mittels fMRT
Advisor: Prof. Dr. R. Beisteiner at Faculty for Life Sciences of University in Vienna, Austria, January 2006; DOI: 10.13140/RG.2.2.25571.76328

First author:

Simulation of spreading depolarization trajectories in cerebral cortex: correlation of velocity and susceptibility in patients with aneurysmal subarachnoid hemorrhage.
Denny Milakara, Cristian Grozea, Markus Dahlem, Sebastian Major, Michael Scheel, Maren K.L Winkler, Clemens Reiffurth, Peter Martus, Jed A. Hartings, Johannes Woltzlik, Jens Dreier
Neuroimage Clinical, Volume 16, Pages 524-538; DOI:10.1016/j.nicl.2017.09.005

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