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MASTER THESIS

**Optical Flow, Helmholtz Decomposition and
Autoencoders for the characterization of slow-waves**

A new approach for temporospatial pattern detection in neuroimaging

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Declaration of Authorship

I, Michael GERSTENBERGER, declare that this thesis titled, “Optical Flow, Helmholtz Decomposition and Autoencoders for the characterization of slow-waves” and the work presented in it are my own. I confirm that:

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- Where I have consulted the published work of others, this is always clearly attributed.
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Abstract

Optical Flow, Helmholtz Decomposition and Autoencoders for the characterization of slow-waves

Neural signal transduction at different levels of alertness occurs in distinct shapes (Eger, 1981). One may observe stimulus dependent activity during wakefulness but also spontaneous patterns of activation that dominate in stages of deep sleep (Brown et al., 2012). Similarly, fast neuronal firing is replaced by slow, traveling waves of activation during anaesthesia (Steriade, Nunez, and Amzica, 1993; Celotto et al., 2020). Cortical slow waves can be captured by fluorescence microscopy of GCaMP channel activity in transgenic mice at up to 100Hz.

Different types of slow waves exist which can incorporate rather localized or more widespread regions of cortex Bernardi et al., 2018. Neocortical slow waves are distinguished from delta waves (Steriade, Nunez, and Amzica, 1993). In principle, they are thought to occur spontaneously in various regions, however it was also noted that synchrony with somatosensory states and more importantly signals in subcortical structures exists (Stroh et al., 2013). Neocortico-hippocampal synchrony is related to memory formation (Maingret et al., 2016). Recently important differences between cortical sleep slow waves and their counterpart during anaesthesia were identified (Nghiem et al., 2018). This shows that slow waves represent an important yet highly variable neural phenomenon and indicates that methods are required that allow to systematically distinguish slow waves based on their properties.

Townsend and Gong (Townsend and Gong, 2018) suggest to characterize temporospatial properties of neural recordings by the detection of patterns in Dense Optical Flow. It shows, however, that the use is accompanied by two challenges: First it shows that focal brightness changes result in sources and sinks in the vector fields. Second successful estimation of Optical Flow is challenged by the quality of the contrast it is applied on. Here it is demonstrated how both can be addressed and Dense Optical Flow used successfully to characterize slow waves. The presented procedure allows for a scale independent split of optogenetic recordings into several segments that include neocortical slow waves. This allows for an event related analysis and the computation of a contrast that is independent from anatomical structures and meets the smoothness requirements of Dense Optical Flow. It is further demonstrated that Helmholtz-Decomposition can be used to distinguish between the effect of local sources of neural activation and global patterns of neural flow. Finally, it is shown how the measured high-level features can be analysed using variational autoencoders. The approach reveals a complex topology of the distribution of different features in latent slow-wave-space. These features include the direction of flow and the distribution of sources, phase and amplitude. It shows that the combination of different types of events differs between stages of anaesthesia.

In summary a new technique to characterize slow-waves is presented that can help to improve our understanding of neural processing in the brain under anaesthesia. The use of Optical Flow, Helmholtz-Decomposition and autoencoders to characterize and distinguish different types of slow waves appears promising. It bears potentials to study the relationship between distributions of events and other features including anaesthetics, experimental stimuli or animal behavior.

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List of Abbreviations

LAH List Abbreviations Here
WSF What (it) Stands For

Chapter 1

Introduction

Anaesthesia inhibits neural activity and alters the signals in the brain. Depending on the dosage of the anaesthetic in use different patterns can be observed (Eger, 1981). At intermediate concentrations many agents cause propagating low-frequency waves that occur in close succession and manifest as slow oscillations in EEG (Steriade, Nunez, and Amzica, 1993; Eger, 1981). Similar patterns can be observed during deep sleep. Neocortical slow waves differ in shape and size and with respect to specific temporo-spatial features: Most importantly the source location of recurrent activation, the direction and speed of spread and different sets of functionally connected regions can be observed (Brown et al., 2012). These features can be measured using modern imaging techniques that capture the fluorescence of genetically modified GCamp channels.

The development of new experimental technologies to capture GCamp fluorescence with an increasing temporal resolution of up to 100Hz offers new opportunities for the study of the neural circuitry and dynamic state changes of the brain under anaesthesia. At the same time, they also impose new challenges that result from the complexity of the recorded data. These challenges must be addressed by a well-controlled experiment or suitable approaches for data analysis.

Strict experimental control simplifies data processing. Under certain experimental conditions, high amplitude slow waves with a good signal to noise ratio and contrast to the hemodynamic autofluorescence occur in a regular pattern. For ketamine induced anaesthesia (100mg/kg), regular patterns of neocortical slow waves with highly similar shapes, durations and amplitudes occur. This simplifies the separation of events and the subsequent analysis (Celotto et al., 2020). However, such strict control means that the variance of

the captured slow waves is highly reduced.

More general findings regarding the structure of neocortical slow waves are only possible in less controlled settings. This means, however, that suitable methods for data processing and analysis must be established that are both robust and suitable to capture the dynamics of neocortical slow waves of different shapes and signal strengths.

Hence a new approach was developed that allows to distinguish and characterize neocortical slow waves of various shapes and forms. It reveals a complex topology of events and shows how different levels of isoflurane change the way signals travel in the brain. The development was accompanied by two major questions: (1) What are slow waves and (2) how can their temporospatial properties be measured and characterized. To answer the earlier question and provide a working-definition for neural slow waves that allows to distinguish different events, relevant slow wave literature was systematically reviewed. To tackle the latter question it was investigated how Dense Optical Flow can be used to measure the strengths and direction of neural flow and how Helmholtz-Decomposition can help to distinguish it from neural sources and areas of extended recurrent activity. As not only features but characteristic patterns are of interest it was further assessed how autoencoders can be used for dimensionality reduction of the measured properties. In the following the results of named literature review are presented first (section 2). It shows that anaesthetics change important properties of neurons most prominently their excitability which explains why bistable states of network activity that can be observed in neocortex empirically (see section 2.2). While important differences exist between anaesthesia and sleep, slow waves can be observed in both states. One may distinguish corticothalamic slow waves in the delta range and neocortical slow waves of a duration above one second (see section 2.4). The latter kind of slow waves is of main interest here. They are understood as extended periods of recurrent neural activation that manifests as above baseline activity (see section 2.6).

The methods to analyse slow waves are explained in the second part. First a new approach is presented that allows to track hemoglobin rich blood in vessels of various size. It is shown that high amplitude oscillations result from intracranial bloodflow and band-stop filtering can be used to reduce this effect. Second an overview of the processing steps is provided. The procedure of splitting events is reported, then Dense Optical Flow is introduced and it

is demonstrated how Helmholtz Decomposition can be used to distinguish between the effect of local brightness changes and global flow. Subsequently the use of autoencoders for unsupervised characterization of slow waves is discussed and the hyperparameters of the models used are presented.

The results of the approach are shown in the third part. First simple statistics regarding the dataset, selected features and the experimental conditions are reported. Then the relationship between properties of waves is summarized for representative events. Finally the results of the autoencoders are reported.

In the last part the findings regarding the methods and the achieved results are discussed. Optical-flow, Helmholtz decomposition and autoencoders are a suitable tool to study slow waves. Feature engineering helps to select relevant and robust features. Optical flow can be used in combination with Helmholtz-Decomposition to distinguish between local sources and global pathways of neural flow. Autoencoders represent an easy to use tool for the analysis of high level features. The results confirms the existence of fundamentally different types of slow waves. The approach allows for a fine grained distinctions and reveals a complex topology of static and dynamic properties in latent slow wave space.

Chapter 2

Slow waves

2.1 Overview of types, functions and modes of action

There is a large body of literature that addresses slow oscillations in the brain. Slow waves in the delta range (0.5-4 Hz) can be identified in electroencephalograms (EEG) in deep sleep and under anaesthesia. Neurons that switch between up and down states in the respective frequency are present in thalamus indicating that this kind of slow waves is of thalamic origin (Brown et al., 2012, p. 1110). As they interact with cortex they are referred to as thalamocortical slow waves here. Since their discovery by Steriade et. al (1993) a different type of slow waves that occur with a frequency below 1Hz have been studied extensively as well: Neocortical slow oscillations. As they do not necessarily occur in regular time intervals they are referred to as neocortical slow waves here (see also section 2.6). Both electrophysiological recordings as well as two-photon and wide field fluorescence microscopy have been used to investigate neocortical slow waves (Niethard et al., 2018; Celotto et al., 2020). Neocortical slow waves with a typical duration of more than one second are the main subject of the empirical analysis presented here.

On one side neocortical and thalamocortical slow waves are recognized as distinct phenomena (Brown et al., 2012, p. 1110). On the other side it was found that both signals relate can occur in an orchestrated way. One interpretation is that neocortical slow oscillations bind together spindles and delta waves (Brown et al., 2012, p. 1110). While named signals are of different origin one could hence consider them to be the signature of a single distributed process. To better understand how distinct oscillatory patterns in the brain relate to neocortical slow waves is crucial to summarize our knowledge about them. Hence selected examples for literature on slow waves, sleep and the effect of anaesthetics are briefly discussed in this chapter. Herein the aim is

not to provide a full literature review but to present relevant information that is necessary to discern events, measure their properties and interpretate the results achieved ¹.

Anaesthesia with isoflourane has several effects on the properties and the spiking behavior of neurons. The modes of action of anaesthetics such as isoflourane have been studied extensively (Qazzaz and Winlow, 2017; Moghadam et al., 2019; Eger, 1981; Jenkins, Franks, and Lieb, 1999). Although the precise mechanisms are not fully understood it shows that anaesthetics change several properties of neurons and reduce their excitability which explains widespread quiescence at deep levels of anaesthesia. At intermediate levels bistable states can be observed in which slow waves of distinct shapes occur (see section 2.2).

The question how bistable states arise from the interaction of cells in neural networks has been addressed using spiking network models. They provide an explanation for the occurrence of neocortical slow waves during deep sleep and under anaesthesia. The circumstance that several models exist highlights not only that different explanations are possible but also that different kinds of processes might exist (Nghiem et al., 2018). Bistable states occur both during sleep and during anaesthesia. However important characteristics of the dynamics of neocortical slow waves differ between these states. Hence it was argued that two different types of slow waves occur during sleep and anaesthesia (see section 2.3).

Evidence from several decades of research highlights the role that thalamic circuits play for the generation of rhythmical activity, the awake state and slow wave sleep (Brown et al., 2012). In contrast to anaesthesia sleep is a vegetative state. The decrease in excitation that can be explained by the interaction of anaesthetics and membrane proteins is arguably achieved by processes that include the suppression of excitatory signals from the ascending reticular activation system (ARAS) during deep sleep. Thalamic nuclei play a crucial role in the generation of neural oscillations including spindles and slow waves in the delta range. While slow wave sleep is arguably promoted by several mechanisms including the cardiac cycle and the release of melatonin that alters neural excitability, sleep spindles of thalamic origin are assumed to be the trigger for a transition to deep sleep (Montagna, 2005, p. 347). Besides, specific regions have been identified in thalamus that oscillate

¹For a complete review on neocortical slow waves see (Neske, 2016).

in the delta range (Steriade and Deschenes, 1984). Sleep spindles also precede many of the neocortical slow waves that were observed on the level of single cells using two-photon imaging (Niethard et al., 2018). This contrasts with the assumption that neocortical slow waves form in a spontaneous and fully independent manner in neocortex if the excitability is reduced (see section 2.4).

A putative function of neocortical slow waves is memory consolidation. The assumed modes of action have been described in the hippocampo neocortical dialog model (Buzsáki, 1996). Many results support the hypothesis that memory replay occurs during sleep and that it includes both hippocampus and neocortex. One may hypothesize that neocortical slow waves represent the neural signature of an important part of this process. Memory consolidation is arguably the most prominent cognitive function of slow waves (see section 2.5).

As explained above slow waves differ with respect to several features. This holds both for the duration and frequency at which they occur and the assumed sites of origin. Neocortical slow waves can emerge from various regions. EEG studies in humans indicate that prefrontal-orbitofrontal regions act as the preferred source location (Brown et al., 2012, p. 1110). First steps to distinguish slow waves in a data driven approach were carried out by Bernardi et al. (2018). Two types of sleep slow waves may be distinguished in EEG. This distinction is based on a metric that includes a shape parameter and a term that indicates whether slow waves are rather local or widespread. The distribution of these types of slow waves differs during different stages of deep sleep. It highlights the importance of means to discern slow waves based on their shape and spatial pattern of activation. A working definition that allows to detect and separate slow waves is derived from the developed understanding of slow waves (see section 2.6).

2.2 Effects of isoflourane

Anaesthetics alter the spiking behaviour of cortical neurons. Under deep anaesthesia the activity of cells is highly reduced and quiescence can be observed for most units. However, the modes of action that cause the respective changes of neural properties differ between anaesthetic agents. In addition, the effects can depend on the exact concentration of the drug in use and low

dosages may, in some cases, even have opposing effects (Moghadam et al., 2019). As isoflourane was used to acquire the dataset at hand its effects on individual cells is shortly summarized highlighting important differences to other anaesthetics when necessary. It is known that alterations of the properties of individual cells change the dynamics of neural interaction on a population level. By adjusting the properties of cells in simple models of neural networks accordingly bistable states can be reproduced that resemble slow waves under anaesthesia (Jercog et al., 2017).

The exact pharmacological mechanisms of action of inhalant anaesthetics such as isoflourane remain uncertain (Miller, Theodore, and Widrich, 2020). Effects on several ion channels have been reported including both chemically gated Cl^- channels (GABA receptors and glycine receptors) and K^+ channels (Glutamate receptors). For example isoflourane is known to reduce the hyperpotentiation that results from Cl^- influx as a consequence of in vitro GABA administration (Jenkins, Franks, and Lieb, 1999). Hence it can be assumed to have excitatory effects in the nervous system as it decreases the inhibition due to GABA. However isoflourane also has inhibitory effects as it suppresses K^+ channel currents resulting in smaller electrically triggered peak amplitudes of action potentials (Buljubasic et al., 1992). Besides a potentiation of glycine receptors is assumed alongside other neurochemical mechanisms that alter the excitability of neurons (Biotechnology Information, 2021). In vivo studies indicate that the net-effect of isoflourane appears to be inhibitory for all relevant dosages. In this respect isoflourane contrasts with other anaesthetics (including e.g. halothane and ketamine) that show concentration dependence (Moghadam et al., 2019). It shall be noted however that differences in the density of different types of receptors exist in different areas of the brain. While isoflourane can be assumed to decrease the excitability of neurons and inhibit neural signaling at all concentrations this effect could differ between brain areas.

Single cell recordings reveal decisive effects of isoflourane on the properties and the spontaneous behavior of neurons. Moghadam et al. (2019) performed a comparative study of systemic and volatile general anaesthetics in single cell cultures and the isolated brain of *lymnaea stagnalis*. The substances tested include the volatile agents sodium pentobarbital, sodium thiopentone, ketamine on one side as well as halothane, enflurane and isoflourane on the other. It was found that isoflourane causes a gradual decline

of both amplitude and frequency of spontaneous action potentials. Differences between the six types of neurons studied were found to be marginal. Upon stimulation neurons remained silent at all examined concentrations of isoflourane. In contrast a gradual decline of evoked action potentials showed for inceasing levels of enflourane. Using either halothane or barbiturates the authors were also able to produce bistable states in vitro during which periods of rapid spiking and quiescence alternate spontaneously.

Besides named alternations in spiking, effects on subthreshold properties of neurons have been identified. Increasing levels of isofofourane can cause a decrease in the membrane time constant i.e. the duration between stumulus onset and 63% potential change of the cell membrane. Under normal conditions it takes longer for the neuron to reach maximal voltage as compared to anaesthesia with isoflourane. This is reflected in the estimated membrane capacitance that analogously decreases. The membrane capacitance is especially interesting because of the role it plays in the integration of electrical inputs (Golowasch and Nadim, 2014). The most likely explanation for the apperent reduction is however an increase in the leakage current. This is because it is (1) rather unplausible for the capcitanace of the membrane to change significantly as its surface area does not change and (2) because of the abovementioned interaction between membrane proteins and anaesthetics (Qazzaz and Winlow, 2017). Neurons act as temporal integrators and fire if the combined voltage of input spikes that occur simultaneously or in short succession exceeds threshold. If the leakage current is stronger the timing of input spikes becomes more critical as the membrane potential goes back to baseline more quickly which may prohibit charge accumulation. Hence it can be hypothesized that anaesthetics could affect the temporal integration of signals.²

While named effects explain the inhibition of neural firing under anaesthesia for different amounts of isoflourane decisive effects arise on the population level. Eger (1981) systematically studied the EEG patterns in the awake state and during anaesthesia with isoflourane for five different dosages at up to 2.9% in humans. For very light anaesthesia (iso = .56%) low voltage fast activity can be observed. At a light surgical level (isp = .96% and 1.78%) slow oscillations are present that change from more regular to irregular patterns

²Note that other anaesthetics were found to increase the membrane capacitance

and alternating patterns with high amplitude oscillations at a moderate surgical level (iso = 2.2%). For deep anaesthesia only occasional low voltage activity shows (iso = 2.9%). Isoflourane administration leads to a gradual shift from a stable awake state to bistable states and finally deep anaesthesia where quiescence dominates.

It shows that inhalant anaesthetics interact with ion channels in the neural cell membrane. The overall effect is a reduced excitability that causes quiescence at high dosages. In contrast intermediate concentrations lead to a gradual change from a steady up state to more or less regular slow waves that occur sporadically at deep anaesthesia. Slow oscillations can be observed in the brain at intermediate levels of anaesthesia.

2.3 Mechanistic models

The effects of anesthesia that are presented here provide the background for a mechanistic interpretation of slow wave activity. The dynamics that arise from changes of neural properties can be modeled using spiking artificial neural networks. Several networks have been proposed.

For example, features of slow waves can be produced with a simple model that employs Adaptive Exponential Integrate And Fire cells as shown by Nghiem et al. (Nghiem et al., 2018): Simulations of a network that consists of 80% excitatory cells and 20% fast spiking inhibitory neurons produce sequences of continuous activity (up-states) and widespread absence of action potentials (down-states). A shift from down-states to an up state can be triggered by background noise while "spike-frequency adaptation on excitatory cells produces a self-inhibition that, destabilizing the up state, causes a reset to the down state" (Nghiem et. al 2018, p. 2). Spiking network models can reproduce alternating sequences of up states and down states, an important characteristic of population level activity under anaesthesia.

A model that explains sleep slow waves on the basis of ion channels is the averaged neuron model. Accordingly slow waves occur because of an increase in Ca^{2+} concentration in the cell that can trigger Ca^{2+} gated K^{+} channels. During the slow wave up state intracellular Ca^{2+} is assumed to activate concentration dependant K^{+} channels which result in a membrane hyperpolarization due to K^{+} efflux. This hyperpolarization prohibits action potentials and hence leads to the transition to a down state. "These results suggested

that activity-dependent K⁺ channels, such as K_{Ca}, might be crucial for the termination of Up states" (Neske, 2016). synaptic depression.???? It is hypothesized that "The bursting phase of the SWS firing pattern is initiated by Ca²⁺ entry mainly through the NMDA receptor (NMDAR) and voltage-gated Ca²⁺ channels" (<https://onlinelibrary.wiley.com/doi/full/10.1002/bies.201700105>). According to the model incoming signals from glutaminergic neurons are hence able to trigger up states by activation of NMDA receptors whereas a transition to the down state is explained by accumulation of Ca²⁺ in the cells. (Shi, Millius, and Ueda, 2019).

While simple mechanistic models explain important features of anaesthesia on a population level they rely require the formalization of processes that are in their entirety not fully understood. It was highlighted that the exact mechanisms of action of anaesthetics remain uncertain. Generalizing over different agents and dosages is not necessarily justified and the dynamic of neural signalling might be affected by differences in the distribution of receptors for different neurotransmitters. Besides the complex anatomy of the brain including the various pathways that connect different regions uni- or bidirectionally is typically not reflected in simple mechanistic models. More precise measurements of the pathways and dynamics of neural signal transduction during slow wave anaesthesia are necessary to understand bistable states of the brain. This is especially important also because it was recently shown that slow wave sleep differs significantly from anaesthesia.

2.4 Sleep slow waves and related signals

Recently it was argued that the dynamics of slow wave sleep and slow wave anaesthesia differ substantially (Nghiem et al., 2018). Jercog et al. (2017) found that the length of up states and the subsequent down state is correlated during urethane induced anaesthesia in rats for clearly synchronized periods where high-amplitude, slow fluctuations are present in local field potentials. Coefficients indicate a very weak relationship ($r = .2$). However, the correlation was found to be consistently positive across experiments whereas the correlation with later down-state periods (time-lag > 2) is close to zero (Jercog et al., 2017). This pattern in the dynamic of neural firing can be reproduced using simple mechanistic models (see section ??). During human non-REM sleep, however, no such relationship holds. As a tendency, long down states

are followed by short up states instead as indicated by a very weak but significant negative correlation ($r = -.04$). A negative peak in the temporal cross correlation exists around zero (Nghiem et al., 2018). This gave rise to the argument that sleep-slow waves are fundamentally different from their counterpart during anaesthesia (Nghiem et al., 2018). Hence the mechanisms that enable slow wave sleep are more closely examined in this section.

2.4.1 Three systems for sleep regulation

It is long known that the anatomical substrate of sleep is a distributed system rather than a single region in the brain (Akert, 1965). Mammalian sleep represents a physiological process that is arguably regulated by the interaction between various control mechanisms. In general, one may categorize these processes into three classes each of which affects the activity of cortical neurons directly or via intermediate events.

- (1) First there are processes that decrease neural excitability rather unapacif-ically by a release or an accumulation of disperse chemical messengers such as the neurohormone melatonin.
- (2) Second processes exist that incorporate electrical signalling via action potentials. They include the neurotransmitter systems that correspond to the ascending reticular arousal system (ARAS). Ascending projections from other areas (e.g. thalamic nuclei) that increase the neocortical arousal in a spatially rather unspecifically manner can be subsumed under this category as well.
- (3) Third there are processes that lead to oscillations in the firing rate of distinct neural populations. The latter can include spike rates that follow the circadian rhythm and alter the both the release of neurohormones and the excitation of cortex via spiking activity. However oscillatory activity also occurs in the range of delta slow waves and below.

As it is not possible to discuss all features of sleep here, important examples are presented that fall under named categories. The aim is to characterize the circumstances under which sleep slow waves occur³.

Melatonin is the messenger of a system for sleep regulation that falls under

³For an good review with a focus on sleep that includes a discussion about the origins of different kinds of rhythms in the brain see Brown et al. (2012).

the first category. In healthy subjects, melatonin concentrations alternate according to the cardiac cycle (Montagna, 2005). Melatonin can pass the blood brain barrier and can hence diffuse to the central nervous system where it accumulates (Aulinas, 2019). It acts as a neurohormone and has an inhibitory effect on the excitability of cortical neurons via different modes of action. Melatonin receptors in the neural membrane have been identified. Besides it was found that melatonin interacts with voltage-sensitive Ca^{2+} channel and inhibits the release of neurotransmitters as well as synaptic transmission (Choi et al., 2014). As it decreases the neural excitability of cortical neurons it shares some of the effects of mild anaesthesia. Melatonin reduces the excitability of disperse populations of neurons in the central nervous system promoting drowsiness and sleep.

The suprachiasmatic nuclei (SCN) represent an example for the third of the abovementioned categories. It interacts with the melatonin system. Melatonin is produced by pinealocytes that compose 95% of the cells in the pineal gland (Aulinas, 2019). The pineal gland resides outside the blood brain barrier and extends hypothalamus ventrally and represents an interface between neural signalling and the endocrine system. It is innervated by centres in the brain stem. Most importantly it is known to receive signals which are related to retinal activity from the SCN. The SCN is capable not only of creating a light-dependant circadian rhythm, but also of maintaining an entrained rhythm (Koella, 1984). The latter could be demonstrated in electrophysiological recordings in vitro. Isolated SCN show firing rates that alternate in a 24 hour pattern (De Berardis et al., 2011). Arguably this explains changes of melatonin release that follows the circadian cycle.

While melatonin is evidently part of sleep regulation the transition between the awake state and slow wave sleep is arguably initiated by other mechanisms. Mammalian sleep can be triggered by electrical stimulation of the thalamus. First experiments were carried out using cats while later experiments confirmed that the effect exists for dogs as well (Akert, Koella, and Hess Jr, 1951; Akimoto et al., 1956). A stimulation of the intralaminar thalamus of cats with a duration of 30-60 seconds and pulses at 4-8Hz leads to a transition to sleep. This transition occurs in different stages. Sleep spindles occur while the animal is still awake. After several minutes a full transition to slow wave sleep can be observed (Akert, Koella, and Hess Jr, 1951). These early studies indicate that thalamocortical sleep spindles play an important

role for the transition to sleep.

Early lesion studies indicated that brainstem nuclei are required for both normal awake activity and REM sleep. In animals with a Cerveau isolé preparation no normal sleep patterns can be observed. Instead one may synchro- nized firing that manifests in slow waves with a main frequency of around 1 Hz (Kawamura and Domino, 1968). In named preparation ascending fibers between pons and midbrain are dissected. Effectively large parts of the retic- ular formation and several nuclei that use specific neurotransmitters are dis- connected from midbrain, thalamus and cortex. In contrast an encéphale isolé preparation in which the brain and the spine are separated does not have named effects but normal sleep patterns can be observed. This high- lights the importance of brain stem for the regulation of sleep and alertness. Named nuclei include serotonergic neurons in the raphe nucleus, nore- pinephrine neurons in locus coeruleus and glutamate neurons in the pedun- culopontine nucleus. They represent the most important excitatory neuro- transmitters. These nuclei are known as the source of the ascending reticular activation system that is known to promote alertness (Brown et al., 2012).

Similar to target sites in thalamus electrical stimulation of the raphe nuclei induces sedation and sleep in rats (Kostowski et al., 1969). The raphe nuclei are brain stem areas that represent the origins of the ascending serotonergic pathways. Together with several neighboring cores they provide excitatory input to large populations of cortical neurons and hence facilitate firing via different pathways. If the activity of the raphe nuclei is reduced by electrical stimulation the arousal of neocortex arguably consequently decreases (see 2.4.2). Mechanisms that lead to a transition to sleep modulate the activity of ascending pathways of the ARAS including the serotonergic system which increase cortical activation.

It shows that the mechanisms that regulate sleep decrease the excitability and activation of cortical neurons under the contribution of chemical messengers including melatonin and an inhibition of the ascending arousal system. Sub- cortical nuclei can produce oscillatory patterns that reflect the circadian cycle whereas thalamus plays a key role in the transition from an awake state to deep sleep. Deep sleep represents a bistable and is characterized by a switch between up and down states of cortical activity. Thalamic nuclei are known to produce sleep spindles which are considered a switch for the transition

between light and deep sleep (Montagna, 2005). Besides sleep spindles, thalamic regions generate rhythmical activity that oscillates in the delta range. Neocortical slow waves that arise during sleep arguably interact with these rhythms. The underlying processes are more closely examined in the next section.

2.4.2 Neural oscillators and relays for ascending activation

Both the transition to the bistable state of slow wave sleep and fluctuations in the delta range include ascending signals that originate from subcortical areas. Several centres have been identified that generate oscillatory patterns of spiking activity. These patterns include (1) thalamocortical sleep spindles, (2) hippocampal short wave ripples (3) delta waves of thalamic origin and (4) neocortical slow waves. Distinct pacemakers exist which neural oscillators and cause named rhythms in the brain.

Rostral midline thalamus: A relay for ascending activation

The circumstance that slow waves were present in the *Cerveau isolé* preparation indicates that slow waves are initiated in brain regions above pons. Thalamus is considered a relay for ascending signals in the awake state. It shows that certain nuclei that act as neural oscillators especially under certain conditions, most notably sleep. "During these states, the behavior of thalamic cells is characterized by long-lasting hyperpolarizations and phasic burst discharges recurring rhythmically" (Steriade and Deschenes, 1984, p. 21). Chirurgical removal of thalamus in cats was found to lead to a decrease of the amount of non-REM sleep from 38% to 11.8%. In contrast diencephalic cats still showed sleep spindles (Montagna, 2005). Further support comes for example from fatal familial insomnia a deadly neurodegenerative disease. Findings indicate that thalamic atrophy causes a lack of sleep spindles and delta sleep which implies that "thalamus is the origin of slow wave sleep" (Montagna, 2005, p. 339).

While thalamic nuclei contain neurons that oscillate in the rhythms of slow wave sleep and it is undoubtedly essential for the regulation of sleep it is unclear if delta oscillations are generated by distinct areas of thalamus. Evidence suggests, however, that Rostral midline thalamus (RMT) acts as a relay for ascending arousal signals that originate from the reticular formation and

abolish slow waves. Stimulation of the midbrain reticular formation in cats suppresses slow thalamic rhythms of hyperpolarizing episodes (Steriade and Deschenes, 1984). RMT is part of the nonspecific thalamic system. It contains nuclei that project to cortex in a spatially unspecific manner. While specific roles of some cores can be attributed one can assume RMT to alter the excitability of cortical neurons in a spatially rather unspecific way. (Vertes, Linley, and Hoover, 2015). Phasic desynchronization of the neocortex could only be observed upon stimulation of the midline thalamic area in *cerveau isole* preparations of cats (Kawamura and Domino, 1968). Rostral midline thalamus relays signals of the ARAS that abolish slow waves.

Lateral posterior nucleus: Source of delta oscillations While it was previously found that thalamic delta waves are absent in diencephalic cats where cortex is surgically disconnected from lower structures (Villablanca, 2004), Dossi et. al (1992) identified thalamocortical in the posterior lateral posterior nucleus of thalamus (LPN) that show delta oscillations (here: 0.5 - 4 Hz) even after disconnection from related cortical areas. Intracellular recordings show highly rhythmic oscillations in membrane potential at around 0.5 Hz. These oscillations occur spontaneously. When exceeding a certain threshold each of these oscillations was found to generate exactly one action potential at peak depolarization. Moreover it is shown that self-sustained delta oscillations can be elicited by electrical stimulation in short rhythmic pulses (Dossi, Nunez, and Steriade, 1992). These findings (1) clearly that LPN is a thalamocortical center that generates delta waves and (2) that rhythmic excitation entrains neurons to exhibit a delta rhythm themselves.

More recently the T-type calcium channels have been found to play a key role in thalamocortical circuits that generate slow waves in the delta range. EEG delta waves are practically absent during NREM sleep in knockout mice that lack the α_1G subunit of T-type calcium channel. Moreover the duration of NREM sleep is generally reduced (Lee, Kim, and Shin, 2004). However selective modification of T-type calcium channels in RMT were found to have similar effects (Brown et al., 2012). This potentially indicates an abolishing effect due to changes in the thalamic sites of the ARAS.

As thalamic cores and cortex are strongly connected it was argued that they represent a common functional unit for the generation of delta waves. The

circumstance that rhythmical stimulation elicits delta waves indicates that interactions between oscillations in thalamus and cortex exist. Potentially they bind activity in thalamocortical networks. However the oscillators that produce slow waves in the delta range rely in thalamus.

Thalamic reticular nucleus: Generator for sleep spindles

Another thalamic nucleus is considered the source of sleep spindles. Sleep spindles manifest as patterns of damped oscillations in EEG. Sleep spindles are well visible in the frequency band between 7 and 15 Hz (Niethard et al., 2018). These patterns in EEG arguably arise from synchronous firing of cortical pyramidal cells. It is known however that sleep spindles originate from the thalamic reticular nucleus (Lüthi, 2014). Thalamocortical sleep spindles are considered to be a switch for the transition between light and deep sleep (Montagna, 2005). In many cases they cooccur with activity of cortical cells measured by two photon fluorescence microscopy (Niethard et al., 2018). The reticular thalamic nucleus is considered the pacemaker for sleep spindles that interact with slow waves.

Spindle activity occurs not only during sleep but also under anaesthesia with barbiturates.

Hippocampal formation: Origin of sharp wave ripples

Another area in the brain that is known to act as a neural oscillator resides in the hippocampal formation. Hippocampus represents the oldest part of cortex and exhibits a highly structured nature that has been extensively studied. An important feature of the organization of hippocampus is the overall flow of signals that follows the scheme of the tri-synaptic way. Signals from entorhinal cortex travel in circular patterns through hippocampus. Activation propagates from entorhinal cortex to the dentate gyrus, subsequently to the CA3 subfield, via the Schaffer collaterals further to subfield CA1 and finally back to the entorhinal cortex. This network topology is incorporated in spiking neural networks that can reproduce important features of hippocampal sharp wave ripples (Aussel et al., 2018).

Hippocampal sharp wave ripples can be produced in vitro. Long term potentiation of neurons that form the recurrent networks that include subfield CA3 is known to facilitate the generation of hippocampal sharp wave ripples

. Repeated high- frequency stimulation of a particular area in subfield CA1 induces long term potentiation and enables short wave ripples both in area CA3 and CA1. These findings do not only indicate the source of origin of hippocampal sharp wave ripples but indicate also its involvement in processes of memory formation (Behrens et al., 2005).

Neocortex: Source of neocortical slow waves

Neocortical slow waves occur spontaneously during deep sleep and deep anaesthesia in irregular rhythms with a frequency of one Hz and below. Differences have been found for ketamine induced anaesthesia and urethane where frequencies were found to be slightly lower (Steriade, Nunez, and Amzica, 1993). Neocortical slow waves have been described as reoccurring bursts of action potentials of neurons in several regions of cortex. Simultaneous acquisition of electrocorticograms (ECOG) shows that these bursts cooccur with an oscillation in the electrical potential on the cortical surface (Steriade, Nunez, and Amzica, 1993). Arguably these oscillations relate to the neocortical slow waves [$< 1\text{Hz}$] that can be measured using EEG during deep sleep. Simultaneous ECoG and widefield fluorescence imaging shows that up states typically cooccur with several oscillations of the ECoG potential (Stroh et al. 2013; see figure 2.1).

Simultaneous recording of ECOG and local field potentials (LFP) in the form of a electrothalamigram indicates the presence of a corresponding thalamic signal (Steriade, Nunez, and Amzica, 1993). The hypothesis that neocortex contains networks that act as a neural oscillator has been confirmed both empirically and by means of simulation (see section 2.2). However it was also found that optical stimuli and the resulting signal that originates from the retina are suitable to elicit slow waves. Besides optogenetic stimulation of the lateral geniculate body, the thalamic relay for signals from the retina, causes slow waves. Neocortical slow waves can hence not only arise because of random noise that manifests as spontaneous firing but can also be triggered by perceptual inputs and excitatory projections from subcortical sites. Moreover it was shown that slow waves evoked in cortex strictly precede a signal that can be measured in thalamus. (Stroh et al., 2013). Taken together this indicates that cortex is the site of origin for this type of slow waves. Oscillations arguably arise spontaneously while the interaction with

other signals can determine the precise time of the occurrence of neocortical slow waves.

Two photon fluorescence microscopy reveals a differential behavior of different neurons during up states of slow wave anaesthesia. It shows that neurons do not all behave in the same way but peak activity cooccurs with sleep spindles only for some neurons. In average the fluorescence signal correlates with slow waves measures using ECoG. As mentioned above the electrophysiological response does however not directly reflect the mean signal. Different neurons have specific response properties and cooccur with other oscillations in the brain in a selective manner (see also 2.5).

Interaction effects

Interaction effects between signals that arise from thalamus and those in cortex have been identified. Thalamus is evidently the source of sleep spindles and delta waves, the most important patterns of slow wave sleep. However it shall be mentioned that thalamus is strongly interconnected with cortex. Electrical stimulation of the reticular core of thalamus was found to cause an increased in amplitude of delta waves in hippocampus. In contrast rhythmic stimulation of cortical area 7 evokes periodic oscillations in thalamic neurons that sustain over extended periods of time after stimulation offset. The resulting thalamic rhythm can in return enhance synchronization (Steriade and Deschenes, 1984, p. 21). This finding indicates not only that entrained rhythms exist in thalamus but also that oscillations in thalamus and both archicortex and neocortex strongly relate to each other.

Sleep spindles, hippocampal sharp wave ripples and neocortical slow waves are arguably not fully independent. For example the spindle amplitude is correlated ($r = .3$) with the df/f signal in wake active cells. A weaker yet significant correlation is found for wake inactive cells. The cooccurrence of slow oscillations and sleep spindles is also correlated with effects on the calcium fluorescence. A higher percentage change can be observed when both signals coincide as compared to the case of sleep spindles alone being present during a slow wave in calcium fluorescence (Niethard et al., 2018). Coincidences between hippocampal short wave ripples and slow waves and their role for memory consolidation are discussed in the next section.

Figure ?? illustrates the location of oscillatory nuclei which evidently play a

role in slow wave sleep and shows the ventral and dorsal pathway of the ascending reticular activation system. The SCN that arguably plays an important role in the regulation of sleep according to the circadian cycle is additionally indicated. Note that oscillatory centres in the rostral medulla that orchestrate breathing during sleep are not indicated (Kubin, 2019).

While the named structures are arguably the source of the respective oscillation it must be noted that signals interact. They do not occur fully independently. Up and down states of neocortical slow waves arise in cortex while they can be caused by subcortical afferences. Empirical data indicates that oscillations in archiocortex and thalamus are more or less tightly bound to neocortical slow waves (Niethard et al., 2018). The presumed role of cortico-hippocampal synchrony for memory consolidation is discussed in the next section.

2.5 Sleep slow waves and memory consolidation

Sleep slow waves arguably represent an important mechanism for memory consolidation. According to the hippocampal-neocortical dialogue model of slow waves, the interaction of hippocampal sharp wave ripples and neocortical slow waves fosters memory consolidation (Buzsáki 1989, Walker et al. 2009): Recently it was shown that generating neocortical slow waves in prefrontal networks such that they are coupled to the occurrence of short wave ripples in the hippocampus increases the performance of rats in a memory task (Maingret 2018). Coherently, a correlation of slow wave activity and the brain-derived neurotrophic factor was found for humans (Duncan, 2013). Spontaneously occurring slow waves arguably play a similar role for memory. They predominantly occur phase locked to sleep spindles (Demanuele et al. 2017). However, the exact mechanisms that orchestrate this synchrony are unknown (Sanda et al. 2020).

The hippocampo neocortical dialog model... Entorhinal cortex is the target structure of the Van Essen diagram. Visual information can be assumed to be processed in a hierarchical manner along the ventral pathway. Where response properties in early visual cortices relate to physical features of the visual percept such as simple shapes that occur in a specific receptive field

neurons evidently code for more abstract concepts in higher areas. This indicates that hippocampus receives a highly processed and arguably relatively abstract, sparse code.

Hippocampus can be understood as an autoassociative memory. Unique structure. Schaffer collaterals. Long term potentiation occurs. However long term memory encoded in neocortex. Movements can be elicited in motor cortex whereas sensory perceptions can be triggered by stimulation of somatosensory areas. Higher motor cortices code for more complex actions. The question the hippocampo-neocortico dialogue model aims to answer is how experiences that are encoded in hippocampus are consolidated by the interaction with neocortex. Slow waves arguably play an important role in this respect.

Evidence suggests that the synchrony between slow waves and hippocampal sharp wave ripples during sleep enhances learning. This was demonstrated in a stimulation paradigm where slow waves are triggered such that they coincide with sleep spindles measured in the hippocampal formation (Buzsáki 1989, Walker et al. 2009). ??More information necessary here??. It is however not fully understood why increasing the synchrony between two signals that can oscillate independently fosters memory consolidation.

Empirical data shows that memory replay occurs both in hippocampus and the associated cortical manifestations of the previously encoded learning experience. Ji and Wilson (2007) investigated contingencies between behavior and slow wave sleep to study the interaction between hippocampus and neocortex in the acquisition of long term memories. They studied the response properties of neurons in rats in an experimental paradigm that includes maze running in an eight shaped maze. Place cells were identified that fire for different regions of the maze in ?? where ???. Depending on the direction of running different patterns in the neural response. The sequence of peak activities of the recorded neurons codes for the direction in which the rat passes through the maze. It was found that these sequences are repeated in hippocampus during slow wave sleep. Most interestingly, however, memory replay did not only take place in archicortex but signals appeared to be exported to cortex. The same sequences of neural responses were measurable in cortex ??where exactly?? (Ji and Wilson, 2007).

As mentioned above source modeling of sleep slow waves indicates cingulate fiber trajectories. Moreover, it was found that at least two different types

of slow waves exist that potentially relate to distinct synchronization processes (Bernardi 2018). Because of its high spatial resolution fluorescence microscopy can provide more fine-grained distinctions and may potentially reveal trajectories of neural signal transduction during slow wave anesthesia. This highlights the importance of methods that allow to capture the variance of temporospatial patterns of neural slow waves.

2.6 Neocortical slow waves: Towards a working definition

A distinction between slower and shorter waves has already been made in the pioneering electroencephalographic study almost a century ago (Berger, 1929, p. 550). Shortly later the term delta wave was coined to describe oscillations in the recorded voltage in the respective frequency band. Delta waves are defined for a frequency of 1 to 4 Hz by some authors (Kubin, 2019) while others define the delta range for oscillations between 0.5 to 4Hz (Dossi, Nunez, and Steriade, 1992). The term slow wave is also used to describe oscillatory patterns in EEGs in the delta range. "Slow-wave activity (SWA) is defined as the EEG power in the slow wave frequency band" (Furrer et al., 2019, p. 1). Hence the term delta wave is sometimes used as a synonym for slow waves. However, when speaking of slow waves it is oftenly referred to slow wave sleep or anaesthesia where delta waves occur largely independent from higher frequency components in EEG.

Depending on the context slow waves relate to a pattern in EEG or oscillations that can be measured in a methodologically rather agnostic way. For example changes in the spike rates of neurons are also referred to as slow waves (Jercog et al., 2017). This shows that the term slow wave is used in relation to the neural phenomenon, its functions and its signatures that can be measured using various methods.

Both delta waves and neocortical slow waves represent slow waves in the wide sense. For example Celotto et. al (2020) denote neocortical slow oscillations as slow waves. Furrer et al. (2019) use the term as a synonym for delta waves instead. As explained in the previous sections both phenomena can occur independently from one another. It is hence advisable to use distinct terms. Steriade et. al (1993) suggested the label neocortical slow oscillation. It was noted, however, that the up states and down states do not necessarily appear in a rhythmic pattern (Brown et al., 2012). In addition the transition

to an up state appears as a slowly travelling wave. As the term is more descriptive this phenomenon is hence referred to as neocortical slow wave as opposed to thalamocortical slow waves here. Sleep slow waves show more irregular patterns in EEG and could potentially include signatures of both neocortical slow waves and thalamocortical slow waves in the delta range (Steriade, Nunez, and Amzica, 1993).

Slow waves are understood as short periods of above baseline activity that occur in bistable states such as deep sleep and anaesthesia which are characterized by reduced neural excitability. Slow waves reflect the up states that are interlaced with down states of neural activity (Jercog et al., 2017). They are dynamic patterns of recurrent activity of large potentially distinct populations of neurons. Slow waves can differ with respect to the pathway of neural flow especially in early stages and with respect to the set of recruited cells where recurrent activity prevails. They appear as propagating waves in widefield and two-photon calcium imaging in vivo (Celotto et al., 2020; Niethard et al., 2018). This arguably reflects a runaway process: If activity exceeds the threshold neurons fire and cause postsynaptic potentials which can elicit action potentials in the target neurons (Nghiem et al., 2018). Because of the spatial arrangement of neurons in cortex where close sites are preferably connected more strongly one can arguably observe travelling waves. Travelling waves can even be observed in neocortical slices in vitro (Wu, Huang, and Zhang, 2008). Cortex is known to be of a highly recurrent nature (Gămănuț et al., 2017). Arguably this is why one can observe extended periods of activation as these neurons can be assumed to directly or indirectly project back to the one that initially fired. Hence, slow waves can reveal regions of high functional connectivity. However also cingulate fiber trajectories were identified in EEG during sleep (Murphy et al., 2009). Neocortical slow waves could also travel through cortex via the highly structured fiber bundles of cingulum that connects entorhinal cortex and the bordering hippocampus with with neocortical sites.

If hippocampal short wave ripples correlate with cortical up-states one may speak of a hippocampo neocortical slow wave event. Analogously one could form subcategorizations for slow waves that incorporate deviant sets of connected neurons. One may hypothesize that neocortical slow waves incorporate signals that do not only show functionally connected populations of neurons but pathways of dynamical flow.

Here data from widefield fluorescence imaging is analyzed. This raises the question how slow oscillations in calcium imaging differs from their counterpart in EEG or ECoG. The simultaneous acquisition of ECoG and wide field GCaMP imaging reveals how both signals relate to each other. Slow waves in the mean fluorescence signal cooccur with slow waves in the electrophysiological recording (Stroh et al., 2013). The respective correlation could also be demonstrated on the basis of individual cells using two photon imaging (Niethard et al., 2018). It shall be noted however that Spontaneously occurring neocortical slow waves differ with respect to the frequency and the

As slow waves are understood as short periods of above baseline activity they can incorporate multiple oscillations. These subsegments could potentially relate to distinct neural events. The circumstance that slow waves can include several oscillations is especially important for the detection and separation of events. It means, however, that events could not be separated based on local minima alone (Celotto et al., 2020). However, it is reasonable to assume that multi peak slow waves exist. Conceptualizing slow waves as above baseline activity bears potentials for revealing the temporal dynamics of slow waves that incorporate several stages. It also means that new methods are required for the analysis the dynamic of slow waves.

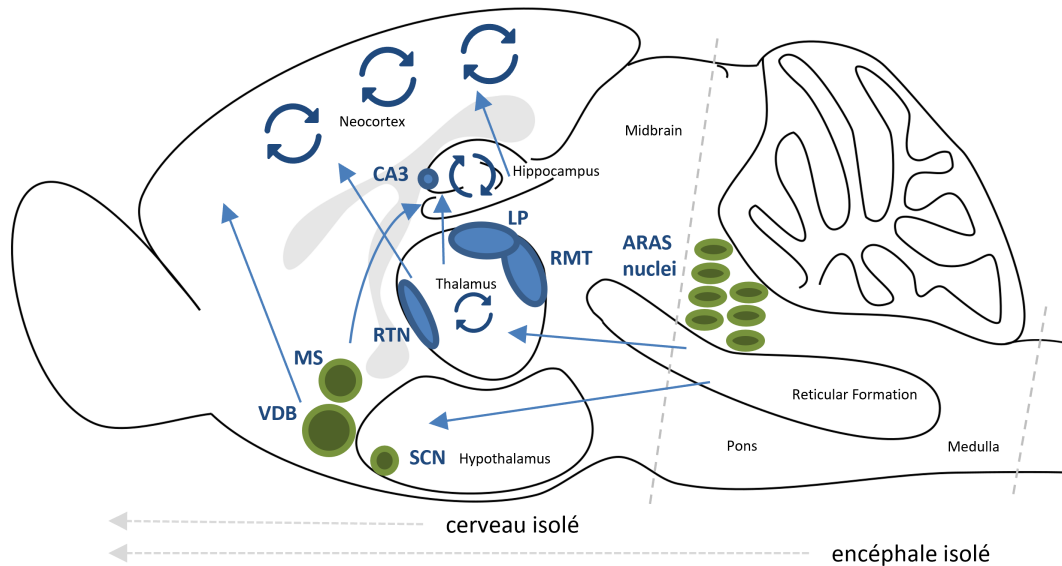


Figure 2.1: Oscillatory centres and the ascending reticular activation system. Circular arrows illustrate sites that act as neural oscillators. Thalamocortical slow waves in the delta range are assumed to originate from the lateral posterior nucleus of thalamus. Thalamocortical sleep spindles are produced in the reticular thalamic nucleus. Rostral midline thalamus (RMT) arguably acts as a relay for the ARAS and abolishes delta waves. It projects to cortex in a non-specific manner. Hippocampal sharp-wave-ripples arise in the subfield CA3 of hippocampus proper. Slow cortical oscillations ($<1\text{Hz}$) and high frequency waves arguably arise in neocortex. Straight arrows indicate the lateral and dorsal pathways of the ascending reticular activation system. Note that centres in the basal forebrain (BF) may suppress the generation of hippocampal short-wave-ripples. Arrows do not relate to the anatomical location of projection fibres and indicate no preferential sites within the target structure. The transections of the preparations “cerveau isolé” and “encephale isolé” are indicated by dotted lines. Slow waves occur in the cerveau isolé while normal sleep patterns have been observed for encephala isolé. Abbreviations: Reticular thalamic nucleus (RTN), Lateral posterior nucleus (LP) Medial nucleus of the preoptic area MnPO and ventrolateral nucleus of the preoptic area (VLPO), Basal forebrain including the Medial septum (MS) and the Vertical Limb of the Diagonal Band (VDB), suprachiasmatic nucleus (SCN), rostral midline thalamus (RMT), Nuclei of the ascending reticular activation system (ARAS nuclei) including serotonergic neurons in the raphe nucleus, norepinephrine neurons in locus coeruleus and glutamate neurons in the pedunculopontine nucleus. While oscillatory centres are isolated inter projections allow for interaction that can cause patterns to occur at the same time. Own depiction inspired by (Brown et al., 2012, p. 1099).

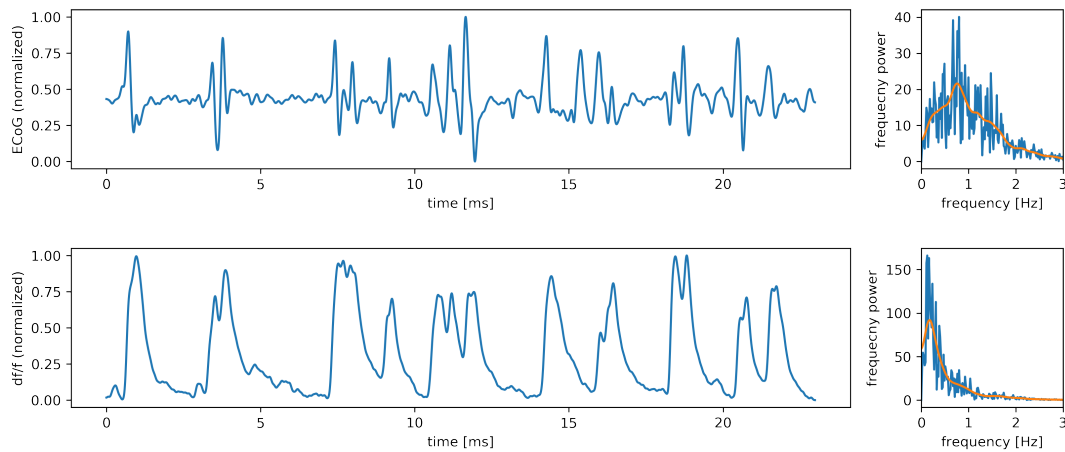


Figure 2.2: Frequencies of slow waves differ between measuring techniques. Data was recovered from supplementary figure 4A, Stroh et al. (2013) for secondary analysis. Signals are smoothed using a gaussian filter and subsequently normalized. Baseline correction was applied for the hemodynamic calcium signal (lower left). The resulting noise reduced signals were investigated with respect to the frequencies present. Upper row: The normalized Electrocardiogram (left) and the frequencies power revealed by fourier transform (right). Lower row: The normalized signal change of the hemodynamic response (left) and the corresponding frequency spectrogram (right). The frequency power is plotted in blue whereas a smoothed version is plotted in orange. It shows that the peak frequency of the hemodynamic signals relies below 0.2 Hz whereas the peak frequency of the calcium signal is above 0.7. Note also that slow waves of both signals are aligned but several ECoG oscillations occur during an hemodynamic event.

Chapter 3

Methods

3.1 A new approach to track intracerebral bloodflow

Typically GCaMP signals are recorded at a sampling rate of 25Hz (Celotto et al., 2020). In contrast the dataset at hand was recorded at 100Hz. High-speed recordings bear potentials for the analysis of neural signals beyond the frequency of neocortical slow waves. However, new methods must be established to identify what high frequency components relate to. This is especially important because widefield fluorescence recordings are confounded with an error that results from the hemodynamic autofluorescence. The results of a new approach indicate that high high frequencies in the df/f signal relate to particles of hemoglobin rich blood that flow through intracerebral bloodvessels of various size.

The approach presented here allows to measure intracerebral bloodflow without the need for the injection of tracers. Typically, microbeads are used to detect the flow rates and speed of intracerebral blood (Kim and Shin, 2019). The study of intracranial bloodflow bears potentials for a better understanding of pathologic conditions and basic neurobiological functions of the brain. For example it was shown that the microvessel density decreases in brain tumors and characteristic changes in the hemoglobin concentration can be observed (Lee et al., 2014). However intracerebral bloodflow is also addressed in the neuroscience in the study of neurometabolic and neurovascular coupling (Devor et al., 2012). New approaches to track the bloodflow in the brain could help to better understand the interaction between the activity of neurons and bloodflow.

To determine what high frequency components of the fluorescence signal relate to several processing steps were employed (see ??). First the mean image was calculated and subtracted from each frame of the recorded videos. The

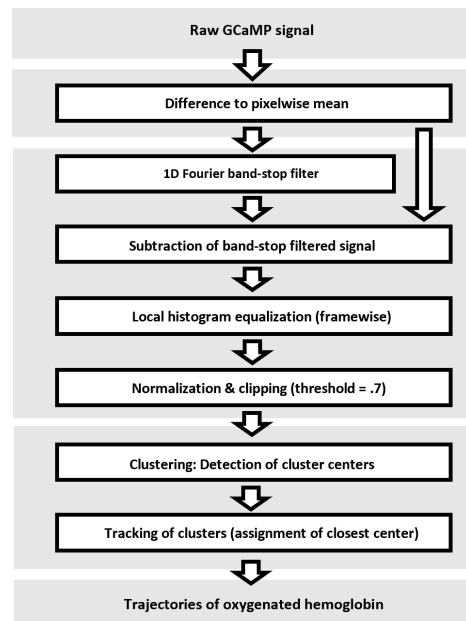


Figure 3.1: Processing pipeline for the detection of bloodflow

signal in time for each pixel was then bandstop filtered and the difference to the original signal was calculated. Bandstop filtering was achieved by transforming the vectors to the fourier domain and setting the frequency components in the desired range to zero before applying inverse fourier transform. Adaptive histogram equalization was used to improve the contrast of each frame after clipping all values below a given threshold. In the resulting videos clusters of bright foreground pixels move on decisive trajectories (see figure 3.3).

Cluster tracking is used to measure the pathways of these temporospatial patterns. Objects in binary images that are potentially connected are oftenly labeled using watershed distance transform (Arganda-Carreras and Legland, 2016)¹. As the data was (1) not binarized but clipping was used and (2) has a low relative resolution which results in a rough outline does not allow for the computation of meaningful watersheds a different approach was used to detect clusters of connected foreground pixels. This approach represents a mean shift technique.

¹Distance transform assigns the value to the closest background pixel to each foreground pixel. Different algorithms exist that allow for fast approximations. The resulting image represents a heightmap if the distance values are interpreted as depth. A simple flooding algorithm can be used to determine the watersheds: A simulated rise of the water level fills basins in the heightmap. The borders between adjacent basins are the watersheds that separate the detected objects.

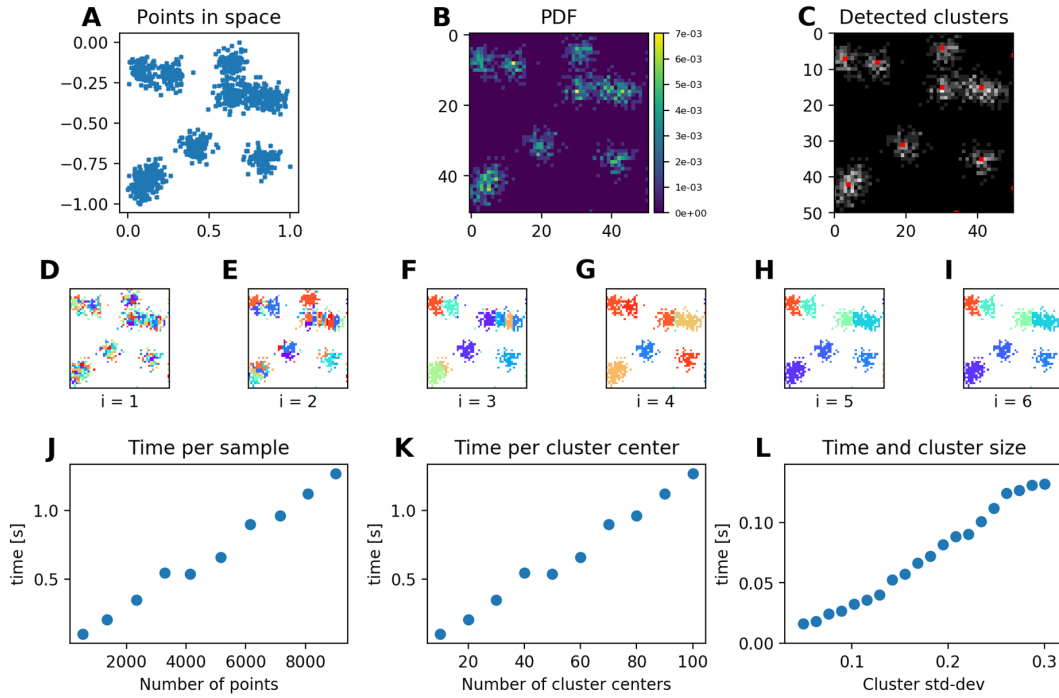


Figure 3.2: A mean shift approach for clustering pixels. After clipping the pre-processed hemodynamic signal, groups of bright, rather isolated foreground pixels are visible. These pixels can be considered as evidence for the presence of oxygenated hemoglobin particles. To detect the location of named clusters of pixels a variation of mean shift clustering was implemented that works with density matrices. Panel A: A distribution of points in \mathbb{R}^2 . Cluster centers were sampled using a 2D gaussian. For each cluster center points were added using a second gaussian probability distribution centered at the respective location. Panel B : A 2D histogram of the distribution in panel A indicating the probability density function. Panel C: The detected cluster centers. Panels D-E: Assignment of pixels to clusters for the first seven iterations beginning with $i=1$. After iteration seven convergence is achieved. Panel J: Time per sample. Panel K: Time per cluster Panel L: Time per cluster size.

The clipped images were interpreted as probability density functions that indicate the presence of oxygenated hemoglobin. Clusters are detected by iteratively moving the density at a given location towards the center of gravity of a patch around this location. The algorithm stops if convergence is achieved or a maximal number of iterations is reached. The environment could be a square or a circular patch. The impact could be weighted using a 2D gaussian kernel to reduce the impact of distant points on the center of gravity. For the results shown in figure 3.3 a simple rectangular patch was used.

Important properties were studied with simulated data. Figure 3.2 summarizes the results. The computation of 2D histograms can be achieved in $O(n)$ as iterating over the data once is sufficient. As the size of this histogram can

be chosen freely the computation time does not directly depend on the number of samples. In the simulation data for each cluster was sampled using the same 2D gaussian with a given standard deviation. Given that convergence can be achieved it can hence be assumed that in average the algorithm takes the same amount of time for each cluster. The simulation confirmed a linear relationship (3.2K). Increasing the number of samples per cluster center also scales linearly with the processing time. A different picture shows for variations of the standard deviation of the cluster centers. A nonlinear relationship can be assumed.

Clustering was performed for each frame separately. Tracking was achieved by assigning the closest cluster center of the subsequent frame (maximal distance 10 px). It shows that particles move on decisive trajectories. Deviant results can be achieved for different frequencies. If bandstop filtering is applied in the range of the heart rate shorter trajectories are detected. The clusters appeared to be larger indicating bloodflow in bigger vessels. In contrast for smaller vessels a more filigree pattern can be observed. It can be hypothesized that these patterns relate to blood that travels at different speeds in vessels of different size.

3.2 Dense Optical Flow

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3.3 Helmholtz-Decomposition

3.4 Helmholtz-Decomposition of Dense Optical Flow

3.5 The processing pipeline

Widefield fluorescence microscopy makes use of selective differences in the reflectance of light of a specific wavelength that results from conformational

changes of ion channels in the neural membrane. This approach is challenged by the hemodynamic autofluorescence due to the blood oxygen level. The fluorescence of oxygenated and deoxygenated hemoglobin depends on the frequency of the light used for illumination. At a wavelength of 405nm the differences are marginal. Nonetheless a hemodynamic error signal remains in the data. Fluctuations are especially visible in recordings with 100 Hz sampling rate as it is high enough to capture potential fluctuations in the fluorescence due to breathing and the heartbeat. This is especially relevant for deep anaesthesia as the GCaMP signal to hemodynamic noise ratio is higher.

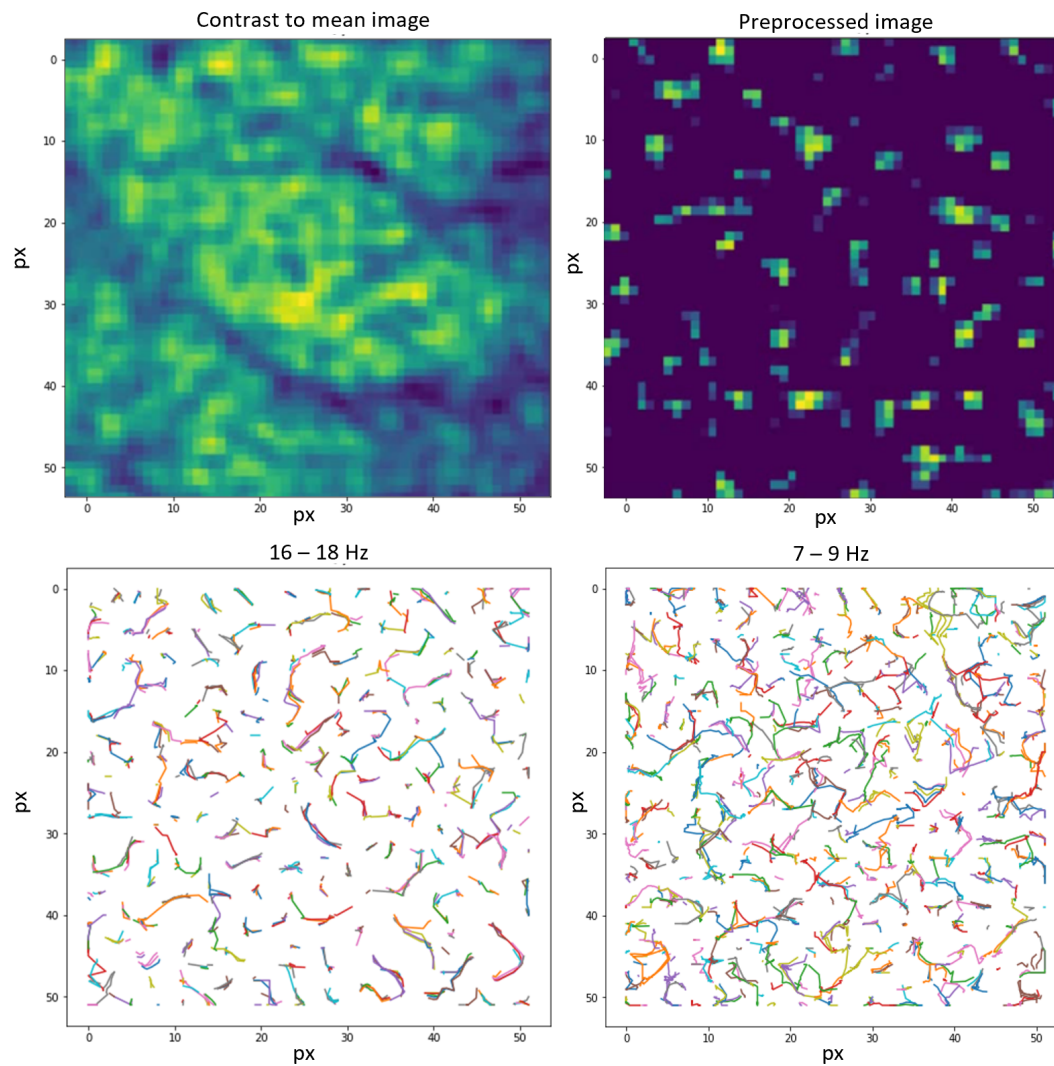


Figure 3.3: Cluster tracking reveals trajectories of oxygenated hemoglobin. Upper left: A sample frame for the original contrast to the pixelwise mean. Upper right: The preprocessed frame after application of clipping. Lower left: Detected trajectories for filtering of frequencies in the range of the heartbeat. Lower right: Detected trajectories of bloodflow for filtering of low frequencies.

See also figure 3.2

Chapter 4

Results

4.1 Main Section 1

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4.1.1 Subsection 1

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4.1.2 Subsection 2

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4.2 Main Section 2

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Chapter 5

Discussion

5.1 Main Section 1

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5.1.1 Subsection 1

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5.1.2 Subsection 2

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Appendix A

Frequently Asked Questions

A.1 How do I change the colors of links?

The color of links can be changed to your liking using:

```
\hypersetup{urlcolor=red}, or
```

```
\hypersetup{citecolor=green}, or
```

```
\hypersetup{allcolor=blue}.
```

If you want to completely hide the links, you can use:

```
\hypersetup{allcolors=.}, or even better:
```

```
\hypersetup{hidelinks}.
```

If you want to have obvious links in the PDF but not the printed text, use:

```
\hypersetup{colorlinks=false}.
```

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