Chapter 2

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The Electroencephalogram - A Brief Background

- The human brain is the most complex organic structure, leading to extensive research from molecular to cognitive levels.
- Early electrical activity in the brain was first recorded by Richard Caton in 1875 on rabbits and dogs, showing microvolt-range signals on the cerebral cortex.
- Hans Berger later recorded time-varying, oscillating electrical "brain waves" from the human scalp, noting variations across different scalp locations.
- Berger observed that these waves varied depending on:
- · Neurological health
- Mental states (e.g., attention, relaxation, sleep)
- Berger's work laid the foundation for electroencephalography (EEG), a noninvasive clinical tool used to:
- Understand brain function
- Diagnose brain disorders
- EEG interpretation involves analyzing:
- Frequency
- Amplitude
- Morphology
- Spatial distribution of brain waves
- Despite progress, no comprehensive biological or mathematical model explains all EEG patterns; interpretation remains largely **phenomenological**.
- Historically, EEG was interpreted solely through visual inspection, but this is now supported by digital signal processing (DSP) methods.
- DSP enhances EEG analysis by:
- Improving signal-to-noise ratio (SNR)
- Quantifying signal features
- Extracting hidden information
- Signal processing for EEG falls into two main categories:
- 1. Spontaneous activity analysis (background EEG)
- 2. **Evoked potentials (EPs)** analysis relies heavily on DSP due to the complexity and low amplitude of EP signals.

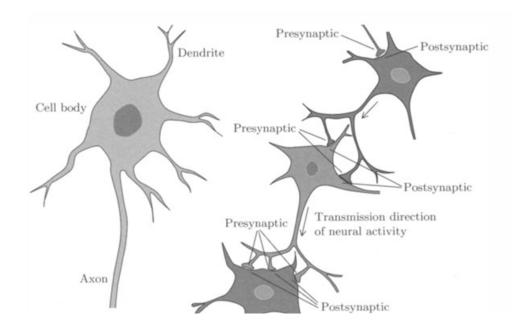
Advancements in Brain Imaging and the Role of EEG

- Brain function research has advanced with imaging modalities such as:
- Positron Emission Tomography (PET)
- Single Photon Emission Computed Tomography (SPECT)
- Magnetic Resonance Imaging (MRI)
- These techniques provide **2D or 3D images** with good spatial resolution and offer additional insights on **anatomy** and **blood flow**, supplementing electrophysiological data.
- Despite reduced dominance in clinical use, **EEG remains vital** for diagnosing:
- Epilepsy
- Sleep disorders
- Dementia
- EEG is crucial for real-time monitoring in:
- · Operating rooms
- Intensive care units
- Coma or encephalopathy patient tracking
- EEG's temporal resolution (fractions of a second) surpasses that of PET, SPECT, and MRI.

- EEG is also cost-effective:
- Requires basic equipment: electrodes, amplifier, PC
- · Minimal technical demands compared to neuroimaging
- Magnetoencephalogram (MEG) is another noninvasive technique:
- Measures weak magnetic fields of neural activity using SQUID sensors
- Advantage: Magnetic fields are less distorted by the skull than electrical signals
- Initially thought to provide **independent information**, MEG is now known to be **strongly interdependent** with EEG signals.
- EEG, MEG, and imaging techniques each have unique strengths and limitations and should be used complementarily.
- Chapter Overview:
- Section 2.1: Overview of the nervous system and brain's electrical activity
- Section 2.2: Common EEG patterns and waveforms
- Section 2.3: Standard EEG recording techniques
- Section 2.4: Key EEG applications

Structure and Function of the Nervous System

- The **nervous system** is responsible for collecting, communicating, and processing information to respond to internal and external changes efficiently.
- It is divided into two main parts:
- Central Nervous System (CNS): Brain and spinal cord
- Peripheral Nervous System (PNS): Connects CNS to body organs and sensory systems
- The CNS and PNS work in coordination:
- Afferent (sensory) nerves carry signals to the CNS
- Efferent (motor) nerves carry responses from the CNS to the body
- A functional division of the nervous system includes:
- Somatic Nervous System: Controls voluntary muscle movement and relays physical sensations
- Autonomic Nervous System: Regulates involuntary activities (e.g., heart, bladder, uterus)
- The autonomic nervous system has two opposing subsystems:
- Sympathetic Nervous System: Active during physical activity or fear (increases heart rate)
- Parasympathetic Nervous System: Dominant during relaxation (decreases heart rate)
- Both subsystems **innervate the same organs** to maintain internal balance.
- **Heart rate variability** arises from this antagonistic interaction and is a key research area in understanding links between neurological diseases and autonomic dysfunction.
- Methods for quantifying heart rate variability are discussed in Chapter 8.





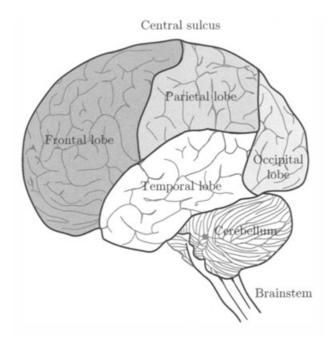
The Neuron – Functional Unit of the Nervous System

- The **neuron** is the basic functional unit of the nervous system, responsible for transmitting information to and from the brain.
- While neurons vary in size, shape, and function, they are collectively referred to as neurons.
- Based on **functionality**, neurons are categorized into three types:
- **Sensory neurons**: Connected to sensory receptors
- Motor neurons: Connected to muscles
- Interneurons: Connected to other neurons
- Structure of a typical neuron (Figure 2.1(a)):
- Soma (cell body): Central part of the neuron
- **Dendrites**: Multiple branches (up to thousands) that receive signals from other neurons (≤ 2 mm in length)
- Axon: Single, long branch that sends output signals (can be >1 m in length)
- **Synapse**: Junction where an axon's terminal contacts another neuron to transfer information.
- Signal transmission:
- Begins in the soma and travels as a pulse-shaped electrical signal (action potential) along the axon
- At the synapse, the **electrical signal is converted to a chemical signal** (neurotransmitter)
- The chemical diffuses across the **synaptic gap**, then reconverted to an electrical signal in the postsynaptic neuron (Figure 2.1(b))
- In the postsynaptic neuron:
- Summation of multiple synaptic inputs occurs
- Signal amplitude depends on number and timing of inputs
- If the summed signal **exceeds a threshold**, the neuron fires an action potential
- Not all inputs are excitatory—some are **inhibitory**, due to specific chemical structures that prevent neuron excitation.

Neural Computation and Signal Transmission

- A **postsynaptic neuron** receives both **excitatory and inhibitory** inputs; its output is determined by how these inputs are **summed together**.
- This summation and response mechanism is termed a **neural computation**, performed billions of times across neurons.
- Electrical activity within neurons differs from scalp measurements:
- Along the axon, signals appear as action potentials—uniform waveforms of identical amplitude
- This is due to the "on/off" property of neurons:
- An action potential either occurs at full amplitude or not at all
- Signal intensity is encoded by the firing rate, not waveform size
- Examples of rate-based signaling:
- **High firing rate in sensory neurons** = Strong pain sensation
- **High firing rate in motor neurons** = Strong muscle contraction
- This firing-rate modulation:
- Is ideal for long-distance transmission
- Is **robust** to local signal failures

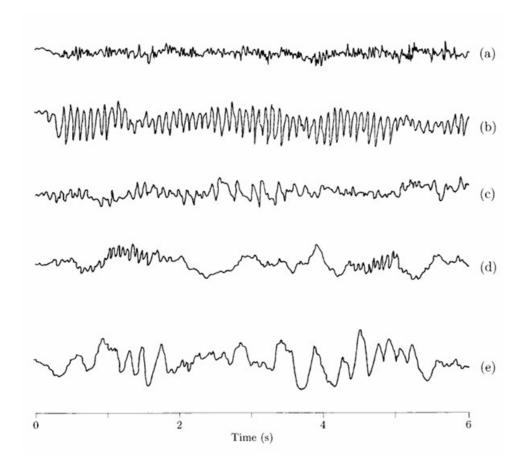
- Is limited by the **refractory period**, during which the neuron cannot fire again
- Neurons form interconnected circuits (neural networks) to process specific types of information.
- Example: Knee-jerk reflex circuit:
- Triggered by a hammer tap stimulating muscle receptors
- Signal travels via afferent pathway to the CNS
- Sensory input activates motor neurons via synapses
- Response signal travels back to muscle, causing contraction and visible knee-jerk reaction



The Cerebral Cortex and Its Functional Regions

- The **cerebral cortex** is the most critical part of the CNS, responsible for:
- Sensation
- Learning
- Voluntary movement
- Speech
- Perception
- It is the **outer layer of the cerebrum**, measuring **2–3 mm thick**, but due to its **highly folded surface** (ridges and valleys), it covers an area of about **2.5 m²** and contains **over 10 billion neurons**.
- The cortex is split into two **symmetrical hemispheres** (left and right), separated by the **sagittal fissure** (central sulcus).
- Each hemisphere is divided into **four lobes** (Figure 2.2):
- Frontal lobe
- Temporal lobe
- Parietal lobe
- Occipital lobe
- Key functional areas:
- Motor cortex: Located just anterior to the central sulcus in the frontal lobe; controls voluntary movement
- Large subareas are dedicated to tasks needing fine control (e.g., speech, facial expressions, finger movements)
- Auditory cortex: In the superior temporal lobe, processes hearing
- Visual cortex: At the posterior occipital lobe, processes vision
- Somatic sensory cortex: Just posterior to the central sulcus in the parietal lobe, processes touch and body sensation
- These are known as primary cortical areas, each with neurons specialized for specific functions.

- Surrounding these are larger secondary areas, essential for advanced mental processing:
- Analyze features like shape, color, size
- Provide associative links between senses
- Integrate new information with past experiences and knowledge



EEG Rhythms and Cortical Electrical Activity

- The combined electrical activity of the **cerebral cortex** is referred to as a **rhythm**, due to the **oscillatory and repetitive** nature of the signals.
- **Single-neuron activity** is **not measurable on the scalp** due to attenuation from tissues like **fluids, bone,** and skin.
- However, **collective activity of millions of neurons**, even several millimeters deep, generates a **measurable electrical field** at the scalp.
- These fields are primarily produced by **synaptic excitation of dendrites**, generating **excitatory postsynaptic potentials**.
- While **intracerebral recordings with microelectrodes** can measure few neurons directly, such methods are **invasive** and not discussed further in this text.
- **EEG rhythms vary widely** and are influenced by the subject's **mental state** (e.g., attention, wakefulness, sleep).
- **Figure 2.3** shows examples of EEG rhythms associated with different states.
- EEG rhythms are defined by:
- Frequency range
- · Relative amplitude
- Amplitude of EEG signals depends on the synchrony of cortical neuron firing:
- Synchronous excitation: High amplitude, rhythmic, coherent waveforms
- Asynchronous excitation: Low amplitude, irregular waveforms
- Note: The number of involved neurons may be the same in both cases.
- **EEG frequency** is influenced by:

- Thalamic input: Thalamic neurons have pacemaker properties, firing rhythmically on their own
- **Cortical feedback**: Local cortical neuron interaction without pacemaker activity can also generate rhythms through **circuit feedback mechanisms**

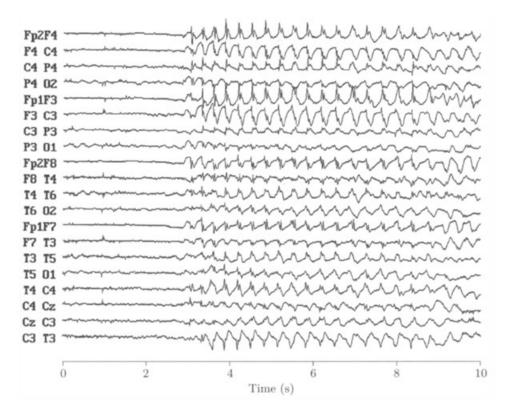
Interpretation of EEG Rhythms and Brain States

- **High-frequency / low-amplitude rhythms** indicate an **active brain** state, typically seen during:
- Alertness
- Dream sleep (REM)
- Low-frequency / high-amplitude rhythms are associated with:
- Drowsiness
- Non-dreaming (deep) sleep
- Reasoning behind this relationship:
- During active states, cortical neurons are highly active but unsynchronized
- Each neuron (or small group) processes different parts of complex tasks
- They fire rapidly, but out of sync with neighbors → results in low synchrony and thus low EEG
 amplitude
- During deep sleep, neurons are not processing information, but are collectively excited by rhythmic input
- This leads to high synchrony and high EEG amplitude
- Despite many unexplained aspects of EEG rhythms, and only hypothetical explanations so far, analyzing and quantifying these rhythm characteristics remains a valuable clinical method for assessing brain functional states.

Common EEG Rhythms and Their Characteristics

- EEG signals recorded from the scalp typically have:
 - \circ **Amplitudes**: From a few microvolts up to ~100 μ V
 - Frequency range: 0.5 to 30–40 Hz
- These rhythms, also called **background rhythms**, are grouped into **five frequency bands**, though the classification is coarse and relative to **age** and **mental state**.
- Newborns show higher frequency content compared to adults.
- **EEG Frequency Bands** (as per [13]):
 - Delta rhythm (< 4 Hz)
 - Large amplitude
 - Common during deep sleep
 - Abnormal in awake adults, may indicate cerebral damage or encephalopathy
 - Theta rhythm (4–7 Hz)
 - Appears during drowsiness and certain sleep stages
 - Alpha rhythm (8–13 Hz)
 - Prominent in relaxed, awake subjects with eyes closed
 - **Suppressed** when eyes are open
 - Highest amplitude in occipital regions
 - Beta rhythm (14–30 Hz)
 - **Fast**, **low-amplitude** rhythm
 - Seen in activated cortex and some sleep stages
 - Observed mainly in frontal and central scalp regions
 - Gamma rhythm (> 30 Hz)
 - Linked to active cortical information processing
 - Can be detected during **finger movements** using sensitive recordings over the **sensorimotor area** [14]
- Rhythm duration:
 - Most rhythms can persist for several minutes
 - Some (e.g., gamma rhythm) may last only a few seconds

- It's important to note that:
 - A single rhythm is **not always continuously present**
 - Long periods of irregular or arrhythmic EEG signals are common

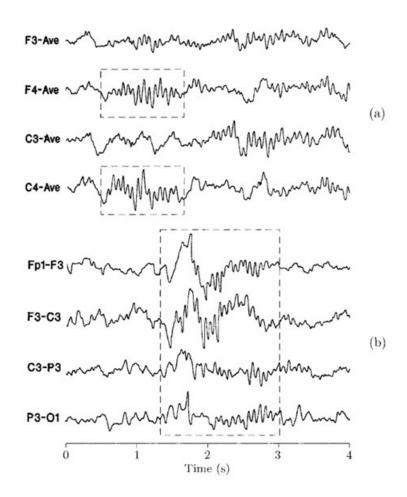


Spikes, Sharp Waves, and Sleep Rhythms in EEG

- Spikes and Sharp Waves (SSWs):
 - o Transient waveforms distinct from the background EEG
 - Occur with irregular, unpredictable timing (paroxysmal activity)
 - Often observed in epileptic patients
 - Typically occur between seizures → termed interictal activity
- Waveform characteristics:
 - o Both feature a steep initial upstroke
 - Differentiated by duration:
 - **Spikes**: 20–70 ms
 - **Sharp waves**: 70–200 ms
 - Morphology is usually monophasic, but biphasic or triphasic shapes can also occur
 - Waveform shape and appearance depend on electrode location
- Spike-wave complexes:
 - Composed of a spike followed by a slow wave
 - Occur as **isolated events** or **in runs**, with repetition rates of <3 to 6 Hz
 - Different repetition rates may relate to specific clinical conditions
 - See Figure 2.4 for an example
- Artifacts:
 - Non-neural activity, like cardiac signals, can mimic SSWs
 - The **QRS complex** of a heartbeat may resemble a spike in EEG recordings
- Sleep Rhythms:
 - Brain operates in three main functional states:
 - 1. Awake
 - 2. Non-REM sleep

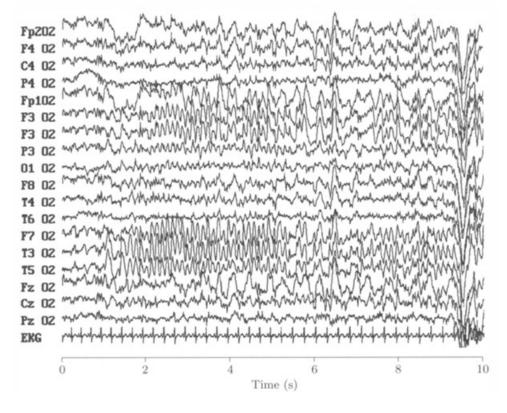
3. REM sleep

- o A person cycles through non-REM and REM sleep several times each night
- Non-REM sleep:
 - o Considered an "idle" brain state
 - Accompanied by slow, large-amplitude EEG rhythms
 - Indicates **high synchrony** among cortical neurons
 - Subdivided into four distinct stages based on depth of sleep
 - See Table 2.1 for classification of sleep stages



Transient EEG Patterns and Ictal EEG

- Sleep-Related Transient Waveforms (see Table 2.1 and Figure 2.5):
 - Vertex waves:
 - Occur in early sleep stages,
 - Triggered by **external stimuli** (e.g., sounds).
 - Sleep spindles:
 - Appear as bursts of alpha-like activity,
 - Duration: **0.5–1 second**.
 - K complexes:
 - Viewed as a fusion of sleep spindles and vertex waves.
- REM Sleep Characteristics:
 - EEG resembles wakeful brain activity, often showing beta rhythms,
 - Associated with **dreaming**,
 - Eye movements (with closed lids) occur rapidly and irregularly,
 - Produce a sawtooth pattern in the EEG when electrodes are near the eyes.



- **Ictal EEG** (during epileptic seizures):
 - Characterized by abnormal rhythmic activity with a sudden amplitude increase (see Figure 2.6),
 - Seizure onset often causes a **sudden frequency shift**, which may evolve into a **spiky** wave pattern,
 - The ictal EEG varies significantly between seizures, making both manual and automatic detection challenging.
- Figure 2.4:
 - Shows a multichannel EEG with spike-wave complexes at a 3-Hz repetition rate,
 - Each channel is derived from **two electrodes**, with placements defined by codes (see **Figure 2.7** for electrode layout).

Categories of EEG Activity Based on Nonstationarity and Sleep Stage Characteristics

• **Table 2.1**: Characteristics of Non-REM and REM Sleep Stages [17]

Sleep Stage	Sleep Depth	Waveforms
1	Drowsiness	Alpha dropouts to vertex waves
2	Light sleep	Vertex waves, sleep spindles, K complexes
3	Deep sleep	Much slowing, K complexes, some spindles
4	Very deep sleep	Much slowing, some K complexes
REM	REM sleep	Desynchronization with faster frequencies

- EEG Activity Categorized by Degree of Nonstationarity (based on [18]):
 - 1. Activity without major temporal changes
 - Normal, spontaneous waking activity at rest (eyes open/closed),
 - Includes stable alpha, beta, and theta rhythms.
 - 2. Slowly time-varying activity
 - Observed during sleep background and postictal states,

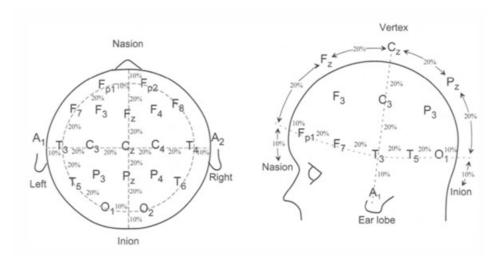
Includes lengthy seizure discharges.

3. Intermittent activity

- Includes intermittent slow rhythms and sleep spindles,
- Activity remains stable over a few seconds.

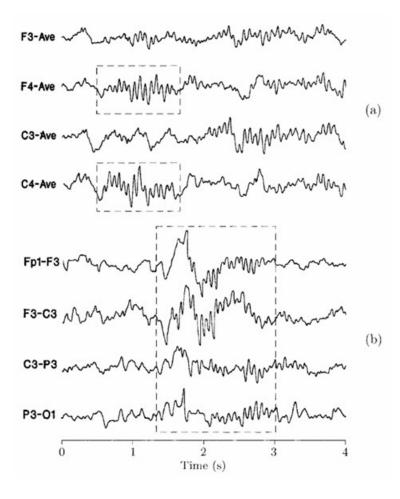
4. Paroxysmal activity

- Includes spikes, sharp waves, spike-wave complexes, 3-Hz spike-wave formations,
- Also includes **K complexes** and **vertex waves** (typically during sleep),
- Represents transient, abrupt phenomena.



Electrode Placement and Spatial Resolution in EEG

- Clinical EEG recordings typically use the International 10/20 system [19], a standardized electrode placement method:
 - Utilizes 21 scalp electrodes,
 - o Positions are based on anatomical reference points,
 - The numbers 10 and 20 refer to percentage distances between electrode sites along the skull perimeter (see Figure 2.7).
- Electrode types:
 - Bipolar electrodes: measure potential differences between two scalp sites.
 - Unipolar electrodes: reference is taken from a distant point or the average of all electrodes.



• Figure 2.5:

- Shows EEG recordings during sleep:
 - (a) Sleep spindles,
 - (b) K complexes, which are fusions of a vertex wave and a sleep spindle.
- Reprinted from Wong [16].

• Limitations of the 10/20 system:

- Sparse spatial coverage:
 - Average interelectrode distance \approx 4.5 cm on an adult head.
- Inadequate for high-resolution brain mapping.

EEG mapping:

- A **spatial analysis** technique that projects EEG activity as a **topographic map** on the scalp [20].
- Using too few electrodes can lead to spatial aliasing, causing inaccurate representations of electrical activity.
- Recommended electrode count for mapping:
 - Studies suggest using 64 or more electrodes to ensure sufficient spatial detail [21].

EEG Sampling Rate Considerations

- The sampling rate for EEG signal acquisition is typically set to at least 200 Hz,
 - This is based on the frequency ranges of standard rhythmic brain activities.
- However, for detailed analysis of transient and evoked waveforms (like evoked potentials),
 - A much higher sampling rate may be necessary.

• Further details are provided in Chapter 4, which discusses EP signal analysis.

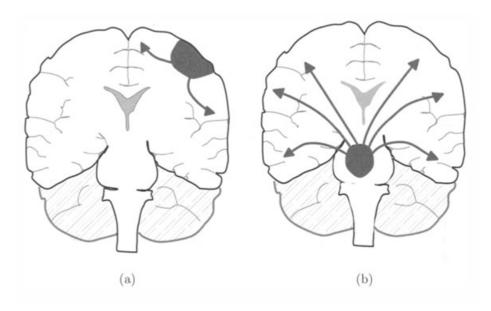
Clinical Applications of EEG

- This section focuses on **two key clinical uses** of EEG:
 - 1. Epilepsy diagnosis and monitoring,
 - 2. Sleep disorder analysis.
- An additional, **research-oriented** application:
 - o Brain-computer interface (BCI) development.
- **Figure 2.7**: *International 10/20 Electrode Placement System*:
 - Based on anatomical reference points:
 - Nasion (top of the nose),
 - Inion (back of the skull).
 - Electrode labels:
 - F = Frontal,
 - \blacksquare P = Parietal,
 - \blacksquare C = Central,
 - T = Temporal,
 - \bullet O = Occipital,
 - A = Auricle.
 - Odd numbers = left side,

Even numbers = right side,

 $\mathbf{z'} = \text{midline}.$

- While the descriptions are **kept brief**, they highlight the **importance of signal processing** in **EEG-based clinical and research applications**.
- For deeper insights, readers are encouraged to consult **specialist literature**.

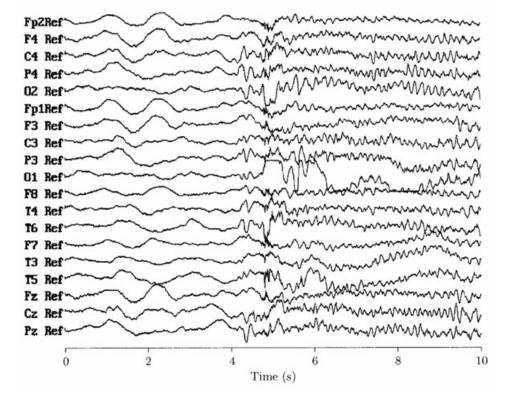


• Figure 2.8:

- (a) Partial seizure:
 - Focus in motor cortex,
 - Symptoms: arm muscle twitches, patient remains conscious.
- (b) Primary generalized seizure:
 - Affects entire brain,
 - Symptoms: spasms and loss of consciousness.
- Diagrams show frontal cross-sections of the brain.

EEG in Epilepsy Diagnosis and Seizure Classification

- Epilepsy is characterized by seizures—sudden bursts of uncontrolled electrical activity in groups of neurons in the cerebral cortex.
- The manifestation of a seizure depends on:
 - The location of the epileptic focus (the impaired group of neurons),
 - How other brain regions become recruited during the seizure.
 - Example: A seizure starting in **sensory areas** may result in **visual or auditory sensations**.
- Normal brain activity relies on a balance between:
 - o Excitatory signals (promote neuronal firing),
 - Inhibitory signals (suppress firing).
- Imbalance between excitatory and inhibitory activity is a key factor in epilepsy:
 - Often caused by neurotransmitter dysfunction:
 - Too much excitatory activity, or
 - Too little inhibitory activity increases seizure likelihood.
 - Antiepileptic drugs aim to restore balance by:
 - Reducing excitatory activity, or
 - Enhancing inhibitory activity.
- Seizure symptoms range widely:
 - Some may cause only minor mental confusion,
 - Others may lead to loss of consciousness,
 - Seizures vary in **frequency** (from rare to multiple per day),
 - o Duration: typically a few seconds to a few minutes.
- Seizure Classification (based on EEG characteristics) [22]:
 - Partial (focal) seizures:
 - Start in a specific area of the brain,
 - Related to a single epileptic focus,
 - Can sometimes be surgically treated by removing the affected cortical region,
 - May evolve into generalized seizures → called partial seizures with secondary generalization.
 - Primary generalized seizures:
 - Involve the entire brain from the onset,
 - Not linked to a single focal point (see **Figure 2.8**).



• Figure 2.9:

- Displays an EEG recorded during the onset of a primary generalized seizure,
- Shows characteristic sudden changes in amplitude and frequency.

EEG in Clinical Epilepsy Diagnosis and Activation Techniques

- Causes of Epilepsy:
 - Linked to various pathological conditions, including:
 - Brain injury,
 - Stroke,
 - Brain tumors,
 - Infections.
 - Genetic factors.
 - In most patients, the cause is unknown.
- EEG: Primary Tool for Epilepsy Diagnosis:
 - Used to determine seizure type and origin.
 - Standard procedure:
 - 30-minute EEG recording in a quiet, dimly lit room.
 - Patient is asked to open and close their eyes to observe EEG changes (e.g., alpha rhythm variations, see p. 34).
- Activation Methods to provoke epileptic waveforms:
 - 1. Hyperventilation patient breathes rapidly and deeply.
 - 2. Photic stimulation patient exposed to flashing strobe light (1–25 Hz).
 - 3. Sleep deprivation sometimes used to enhance detection of seizure-prone activity.
- Interictal EEG (recorded between seizures):
 - May still show epileptiform waveforms indicating seizure tendency.
 - Examples:
 - Spike and sharp waves (SSWs) in a localized brain area (e.g., left temporal lobe) suggest partial seizures.

- Bilateral spike-wave complexes across hemispheres suggest primary generalized seizures.
- However, absence of interictal abnormalities does not exclude epilepsy.

Prolonged EEG Monitoring: Video and Ambulatory EEG

• Seizure EEG recording often requires long-term monitoring, as seizures may not occur during short clinical sessions.

Video EEG:

- o Conducted in-hospital over several days,
- o Combines EEG recordings with video footage of the patient,
- Enables neurologists to correlate EEG changes with visible seizure behavior,
- Improves accuracy in seizure assessment.

• Ambulatory EEG:

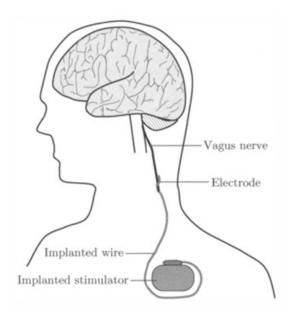
- A more convenient and cost-effective alternative,
- Recorded during the patient's **normal daily life** using a **small digital recorder** (worn around the waist),
- o Typically done for 24 hours or more,
- Captures both waking and sleeping states.

• Practical considerations:

- Requires multiple electrodes attached to the scalp for extended periods,
- May cause itching, leading to scratching, which introduces noise artifacts,
- Other noise sources include:
 - Blinking,
 - Frowning,
- These artifacts can mimic physiological waveforms, making EEG interpretation more challenging.

Long-Term EEG Monitoring and Therapeutic Interventions for Epilepsy

- Long-term EEG monitoring (hospital or ambulatory) generates large volumes of data, making manual review time-consuming.
- Automatic spike and seizure detection is critical for:
 - o Reducing data load,
 - Enhancing interpretation efficiency.
- Detection algorithm design requires:
 - Signal processing techniques for mathematically characterizing:
 - Interictal waves, and
 - Seizure patterns [24–26],
 - Noise and artifact rejection algorithms are also essential.
- Seizure prediction/warning algorithms:
 - o Can provide early alerts to patients wearing ambulatory devices,
 - Allow patients to take safety precautions before a seizure occurs.



• Therapeutic Devices:

- Aim to prevent seizures before they start.
- Vagus Nerve Stimulator (VNS):
 - Most well-known device in this category [27],
 - o Delivers intermittent electrical pulses to the vagus nerve,
 - Has shown antiepileptic effects in some patients, though mechanisms remain unclear.
 - o Implanted surgically, like a cardiac pacemaker,
 - Stimulating electrode is wrapped around the vagus nerve in the neck (see Figure 2.10).
- Current VNS limitations:
 - Operates blindly using preset stimulation patterns,
 - Does not adapt to actual seizure risk in real-time.
- Future development:
 - Focuses on intelligent VNS systems,
 - Will incorporate seizure prediction algorithms using advanced signal processing [28, 29].

Classification of Sleep Disorders

Sleep disorders are common and may arise from **medical or psychological causes**. A widely used classification scheme organizes them into **four main groups** [30]:

1. Insomnia

- Difficulty initiating or maintaining sleep.
- Commonly experienced during **stressful events** (e.g., exams), often **temporary** and **not treated**.
- Associated conditions:
 - **Depression** \rightarrow reduces deep sleep (Stages 3 & 4), leading to daytime fatigue.
 - Alcohol or drug abuse → disrupts sleep patterns.

1. Hypersomnia

- Characterized by excessive sleepiness and daytime somnolence.
- Examples:
 - Narcolepsy: sudden, uncontrollable daytime sleep attacks, despite normal night-time sleep.

Sleep apnea:

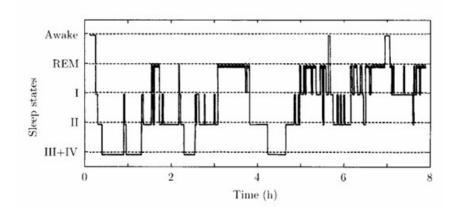
- □ Involves frequent breathing interruptions during sleep (>10 seconds),
- □ Causes snoring-induced awakenings,
- □ Results in disrupted deep sleep and severe daytime fatigue,
- ☐ May also be considered a form of **insomnia**.

1. Circadian Rhythm Disorders

- Affect the timing of sleep-wake cycles.
- o Examples:
 - Jet lag:
 - □ Caused by **travel across time zones**,
 - □ Leads to temporary misalignment of the body clock,
 - □ Symptoms usually **resolve within a week**.
 - Non-24-hour sleep-wake disorder:
 - □ Diurnal rhythm is slightly longer than 24 hours,
 - □ Patient's sleep onset shifts later by ~30 minutes daily,
 - □ Leads to cycling into and out of daytime sleep, making stable routines difficult.

Parasomnia and Automated Sleep Staging

- Parasomnia: Sleep disorders involving abnormal behaviors or experiences during sleep, often without full awakening.
 - Nightmares:
 - Involve threatening dream content,
 - Associated with increased autonomic activity, such as a sudden spike in heart rate.
 - Sleep terrors:
 - Not linked to dreams.
 - Marked by piercing screams or cries,
 - Primarily seen in **children**, typically **resolves with age**.
 - Sleepwalking:
 - Occurs during deep sleep stages,
 - Often related to sleep terror, involving unconscious motor activity.



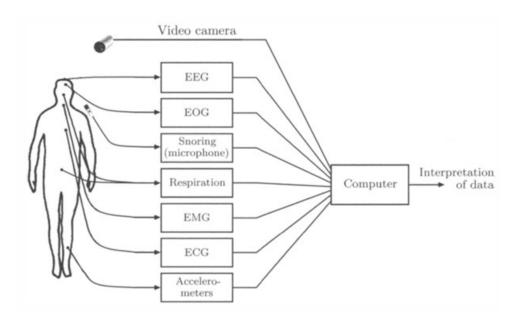
• EEG in Sleep Disorder Diagnosis:

- Each type of sleep disorder has distinct EEG manifestations,
- Accurate diagnosis requires quantitative tracking of sleep stage patterns over time (see Figure 2.11).
- Sleep study setup:

- o Patients stay overnight in a sleep lab,
- Scalp electrodes are used to record the EEG.
- Need for Automation:
 - o Manual sleep staging is labor-intensive,
 - There is a strong need for **automated systems** that can:
 - Detect individual sleep waveforms (e.g., vertex waves, sleep spindles, K complexes),
 - Identify different rhythms (e.g., delta, theta, alpha, beta).
- Design of an automated system should:
 - Mimic neurologist interpretation,
 - Account for both:
 - Spatial distribution of waves across EEG channels,
 - **Temporal distribution** within each channel [31, 32].

Polysomnography and Automated Sleep Analysis

- Sleep analysis is typically performed using polygraphic recordings, which include multiple physiological signals—not just EEG.
- These recordings are known as **polysomnographic** recordings.
- Figure 1.8 (Introduction) illustrates a polysomnographic setup with signals including:
 - o **EEG** (brain activity),
 - o ECG (heart activity),
 - o EMG (muscle activity),
 - o EOG (eye movements),
 - o Blood pressure,
 - Nasal and abdominal respiration.



- Polysomnography may also include:
 - o Video recording of the patient during sleep,
 - Used to observe **behavioral expressions**, such as **sounds** and **body movements** (see **Figure 2.12**).
- Challenges and Solutions:

- Analysis is **complex** due to the **variety of signals** from different physiological
- o Computer-based analysis helps to:
 - Handle complexity,
 - Quantify correlations between signals (e.g., between EEG patterns and respiratory changes).
- Noise and artifact rejection is as essential here as it is in systems for automatic seizure detection [33, 34].

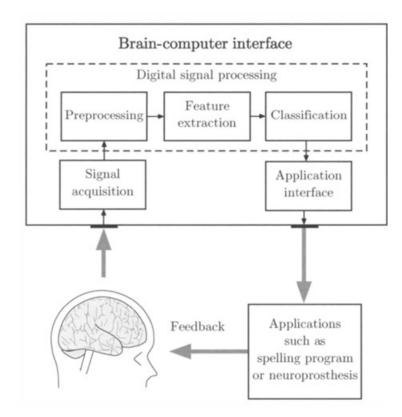
Brain-Computer Interfaces (BCIs) and EEG-Based Communication

- A Brain-Computer Interface (BCI) enables communication and control of the external world without involving peripheral nerves or muscles [35–37].
- Instead of muscle-based outputs like **speech** or **writing**, communication is achieved through:
 - o Spontaneous or
 - Evoked EEG activity.
- Primary beneficiaries of BCI systems:
 - o Individuals with severe neuromuscular disorders,
 - Especially those with complete paralysis (e.g., locked-in syndrome),
 - o BCIs can enable:
 - Use of spelling programs,
 - Operation of **neuroprosthetic devices**.
- Development timeline:
 - o Concept introduced in the early 1970s [38],
 - Significant advances occurred in the 1990s, driven by:
 - Better understanding of **EEG signals**,
 - Progress in computer processing technology [39, 40].
- Two key components in BCI design:
 - 1. The user's mental process:
 - Encodes commands within the EEG signal.
 - 2. The BCI system:
 - Uses advanced signal processing to:
 - □ Analyze EEG signal characteristics,
 - ☐ Translate them into commands for controlling an external device.

Motor Imagery and Adaptation in Brain-Computer Interfaces (BCIs)

- Motor imagery:
 - The mental simulation of movement (e.g., hand or foot),
 - Produces distinct EEG patterns that can encode commands [35, 40, 41],
 - No actual muscle movement occurs—only the neural process is activated.
- BCI Learning Process:
 - The subject is asked to **imagine specific movements**,
 - For each imagined action:
 - **Descriptive features** are extracted from the EEG,
 - These features are used to train a classifier,
 - Repeated trials enable the classifier to learn to differentiate mental commands.
- Post-training use:
 - The classifier translates ongoing motor imagery into device commands (e.g., selecting letters in a spelling program),

- The BCI must function in real time, so signal processing must be fast, without unacceptable delays.
- Continuous Adaptation:
 - o BCI training is not permanent—must be repeated regularly,
 - EEG signals vary with:
 - Time of day,
 - Hormonal levels,
 - Fatigue,
 - The classifier must be updated to maintain accurate performance.
- Two adaptive controllers must coordinate:
 - 1. The user's brain (intention and control),
 - 2. The **BCI system** (signal feature extraction and classification),
 - o Success depends on:
 - The user's ability to produce controllable EEG features,
 - The BCI's ability to **correctly interpret** those features [36].



• Figure 2.13

• Presents a block diagram outlining the main components of a BCI system.

Feature Extraction for BCIs Using Spectral Analysis

- The most common method for feature extraction in BCI systems is spectral power analysis across different EEG frequency bands [42–46].
- While single-channel spectral analysis can be useful, multichannel analysis is preferred because it:
 - o Captures spatial variations in EEG signals,
 - Exploits differences between **brain hemispheres**, which are relevant in **motor** imagery tasks [47].
- Key EEG rhythms used in BCIs:

- o Mu rhythm and
- o Beta rhythm,
- Both originate from the **sensorimotor cortex**—the region responsible for **hand and foot movement control**.
- The selection of frequency bands focuses on capturing these task-relevant rhythms.
- Further details:
 - Spectral feature extraction techniques used in BCI systems are discussed in Chapter 3.

BCI Performance and Alternative Approaches

- BCI performance is commonly evaluated using the information transfer rate, measured in bits per minute.
 - This depends on how accurately the system can classify different mental (imagined) states.
- Current BCI capabilities:
 - Achieve about 10–25 bits/minute,
 - Enables paralyzed users to write around two words per minute,
 - o Still **too slow** for:
 - Controlling complex movements,
 - Operating neuroprosthetic devices effectively.
- Potential performance enhancement:
 - Using surgically implanted microelectrodes,
 - Allows recording from more localized neuron populations,
 - Can significantly boost information transfer rates.
- Alternative BCI design:
 - Instead of analyzing spontaneous EEG, BCIs can use evoked potentials (EPs),
 - EPs are EEG responses triggered by **sensory stimulation** [36, 38, 48–50].