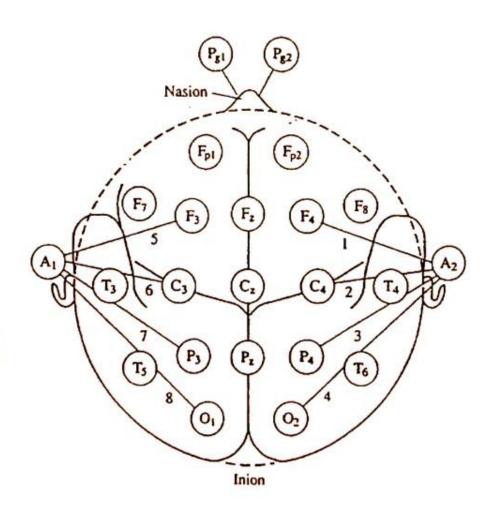
Biosignals and Systems

Lecture 5

EEG

Brief introduction to the EEG

The EEG signal is measured with Ag-AgCl electrodes which are placed in standard positions on the scalp (signal is <100µV due to skull attenuation)





Characteristics of the EEG

- The important information is in the frequency domain.
- The frequency range from 0.5 to 30 Hz has been arbitrarily divided into 5 bands:

Delta 0.5-4Hz

• Theta 4-8 Hz

Alpha 8-13 Hz

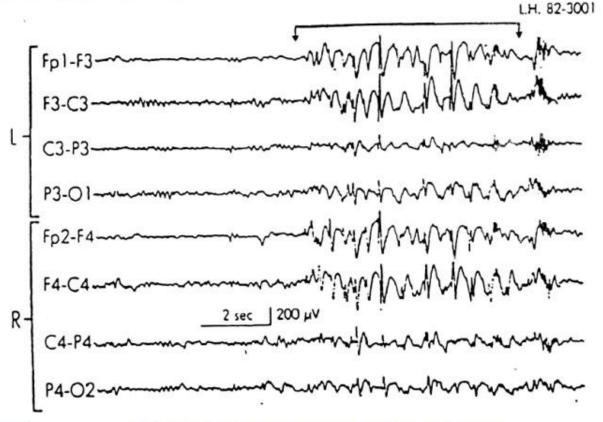
Beta 13-22 Hz

Gamma 22-30 Hz



Diagnostic uses of the EEG

EEG analysis helps in the diagnosis of brain death, epilepsy and sleep disorders





Sleep analysis

- Quality of life is heavily dependent on quality of sleep.
- Between 5 and 10% of the adult population suffers from some form of sleep disorder (insomnia, heavy snoring, Obstructive Sleep Apnoea (OSA), etc...)
- Such people may be referred to a "sleep clinic" by their GP where various signals, including four channels of EEG, the EOG and oxygen saturation, will be recorded throughout the night.
- The EEG and the other signals are printed out and reviewed by a trained sleep technician (requiring 2 to 5 hours for each record).

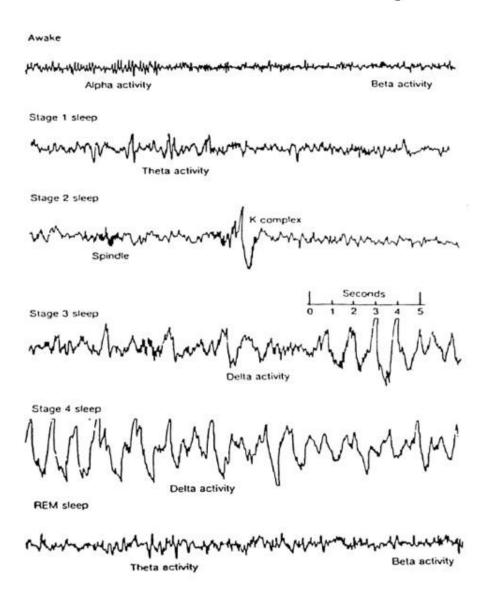


Sleep EEG

 The four channels of sleep EEG are analysed using a rule-based system which assigns consecutive 30-second segments to one of six stages (Wake, Stages 1 to 4, REM sleep).



EEG in different sleep stages



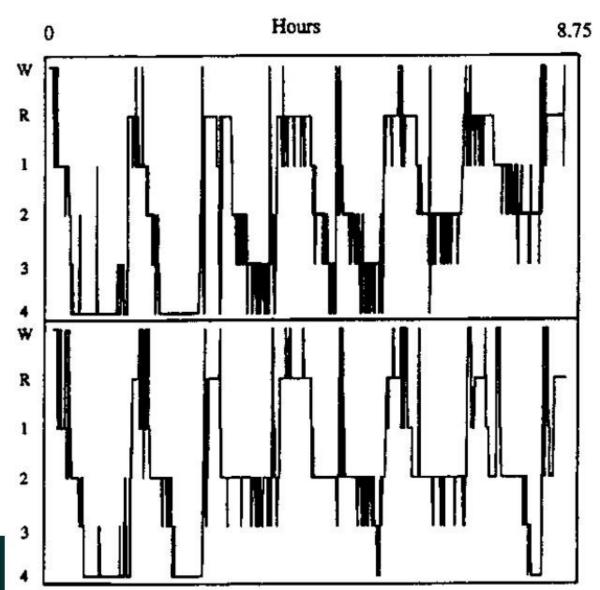


Sleep EEG

- The four channels of sleep EEG are analysed using a rule-based system which assigns consecutive 30-second segments to one of six stages (Wake, Stages 1 to 4, REM sleep).
- For example, two rules for stage 3:
 - an EEG record in which at least 20% but not more than 50% of the epoch consists of waves of frequency 2 Hz or lower which have amplitudes greater than 75 μV peak to peak.
 - sleep spindles may or may not be present in stage 3.



Conventional sleep scoring





The sleep EEG Automating the analysis

 The important information is in the frequency domain.

 Use the Short-term Fourier Transform or an Auto-Regressive (AR) model to extract the frequency-domain information.



AR models for spectral estimation

 The notation AR(p) refers to the autoregressive model of order p. The AR(p) model is written as follows:

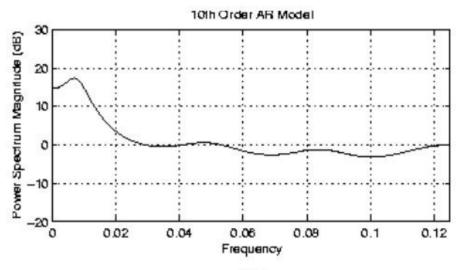
$$X_t = \sum a_i X_{t-i} + \varepsilon_t \quad (1 \le i \le p)$$

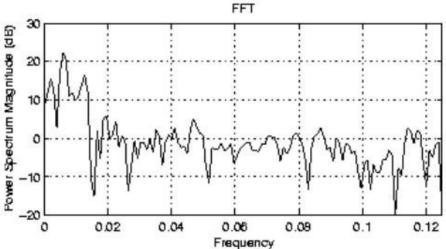
where the a_i 's are the *parameters* of the model and ϵ_t is a white-noise process with zero mean.

 An autoregressive model is essentially an <u>infinite</u> <u>impulse response</u> filter which shapes the white-noise input. The poles are the resonances of the filter and correspond to the spectral peaks in the signal.



AR-model *vs* FFT spectra (EEG signal)











Conventional vs neural computing

Conventional computing

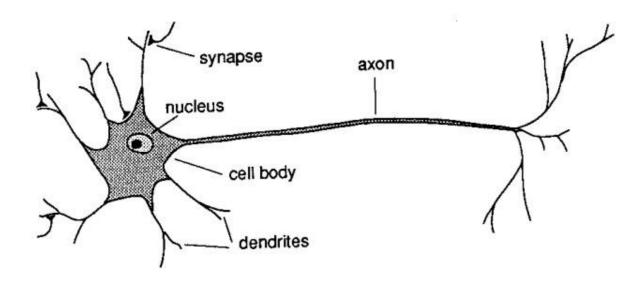
- A set of programmed instructions
 - IF A = B THEN ADD 3 TO RESULT ELSE SUBTRACT 5

Neural networks

 The solution to a problem is *learnt* from a set of examples using error feedback



Neural networks



By analogy with the brain, *artificial* neural networks consist of large numbers of small units ("neurons") with modifiable connections ("synapses"), ordered in feedforward layers



Some definitions

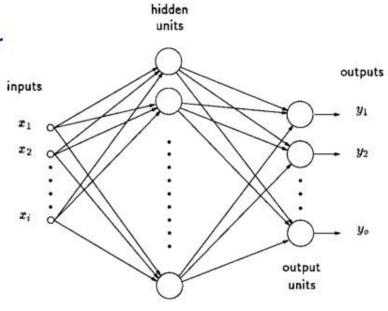
- Neural networks: A computational system which learns from examples how to recognise complex patterns in data, signals or images.
- <u>Learning</u>: The process of modifying the neural network's internal connections (the synaptic weights) until desired output responses are associated with the input patterns in the training set.
- Generalisation: The ability to generate the correct output response for a test pattern, an input pattern not previously seen (i.e. not in the training set).



 The signal recorded from one EEG channel is sampled at 128 Hz and segmented into one-second epochs.

 The input data are the ten AR coefficients for each consecutive one-second epoch.

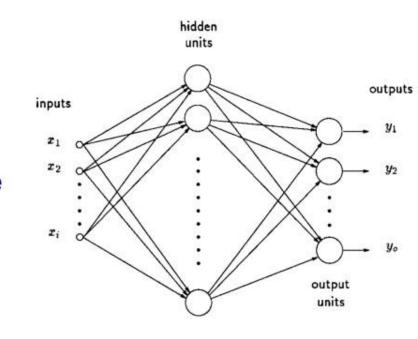
The multi-layer perceptron (MLP) neural network is *trained* to classify the input EEG data into three output classes: awake, light sleep or deep sleep. This involves the gradual modification of the synaptic weights in the hidden-output and input-hidden layers (error minimisation using gradient descent).



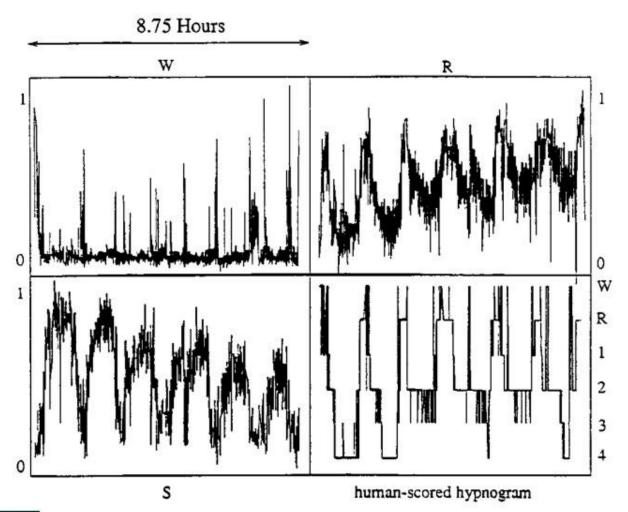
- The training data are collected from human-scored sleep, selecting only "consensus-scored" Wake, REM or stage 1 and stage 4 data (independently scored by three experts).
- Once the network is trained, the intermediate stages for test EEG data are obtained by interpolation.



- The neural network is trained to classify the input EEG data into three output classes: wakefulness, light sleep or deep sleep.
- The trained neural network is used for the automated analysis of the EEG recorded from new patients (test data).







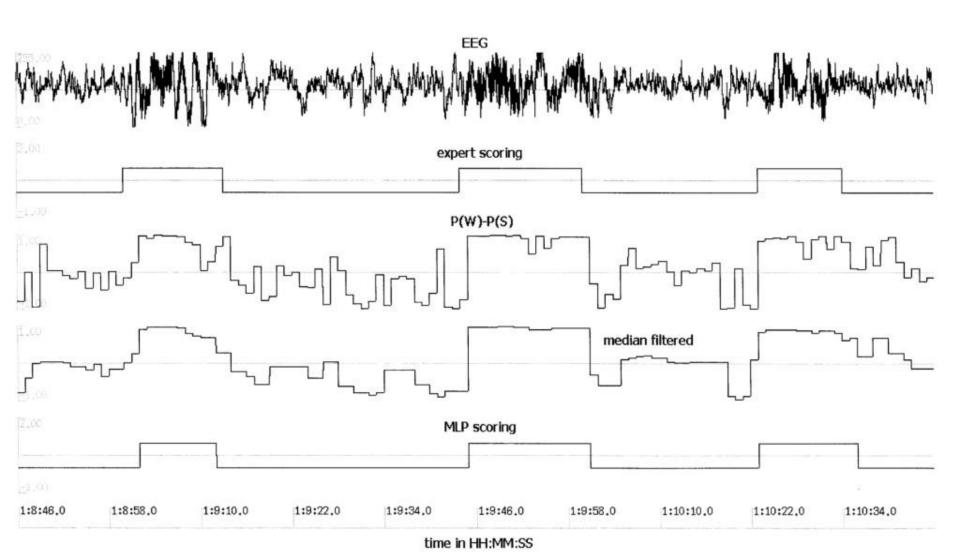


Study of arousals in OSA

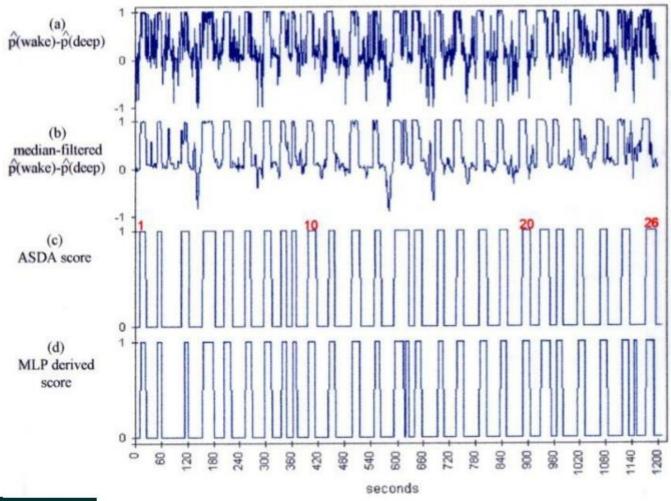
- Seven 20-minute recordings from patients with different levels of OSA used as test data
- EEG scored by one expert according to the American Sleep Disorders Association (ASDA) rules
- P(Wake) P(Deep Sleep) used to represent the sleepwake continuum
- Transients shorter than 3 s discarded using median filtering



Sleep analysis using neural networks (2 mins of data)



Sleep analysis using neural networks (20 mins of data)





- Continuous tracking of sleep-wake continuum using single channel of EEG
- Events on a 2 to 5-second timescale are clearly detected
- This type of analysis is not possible with the traditional sleep scoring methods, which require several channels and only have 30-sec resolution
- Oxford Medical incorporated neural network analysis as an option within their Sleep Analysis software



Sleep analysis using neural networks Oxford BioSignals' first product

- Continuous tracking of sleep-wake continuum from a single-channel of EEG
- Automated analysis makes it possible to quantify severity of OSA (and other sleep disorders) before and after treatment.





