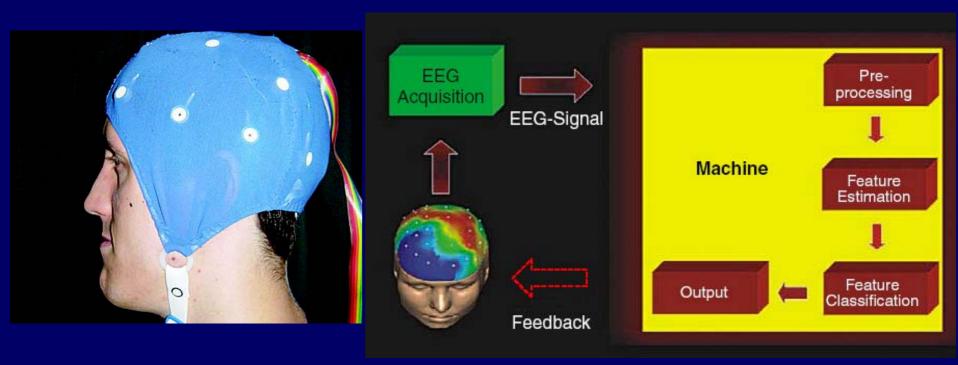
DSP概論: Biomedical Signal Processing

台大生醫電資所/電機系 李百祺

Brain-Machine Interface

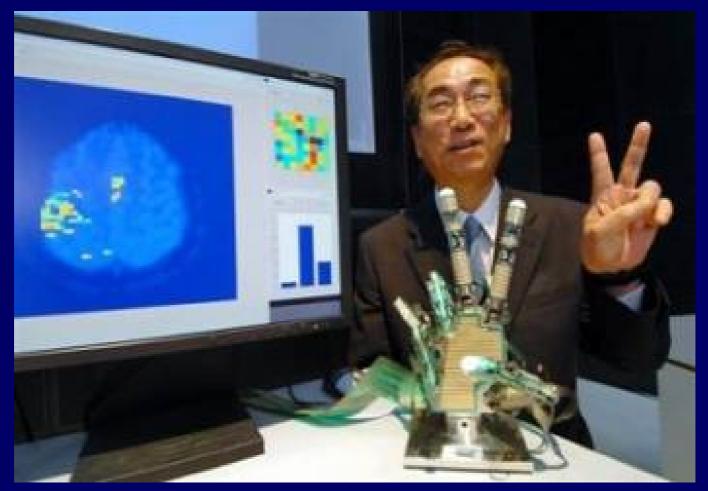


Brain-Machine Interface

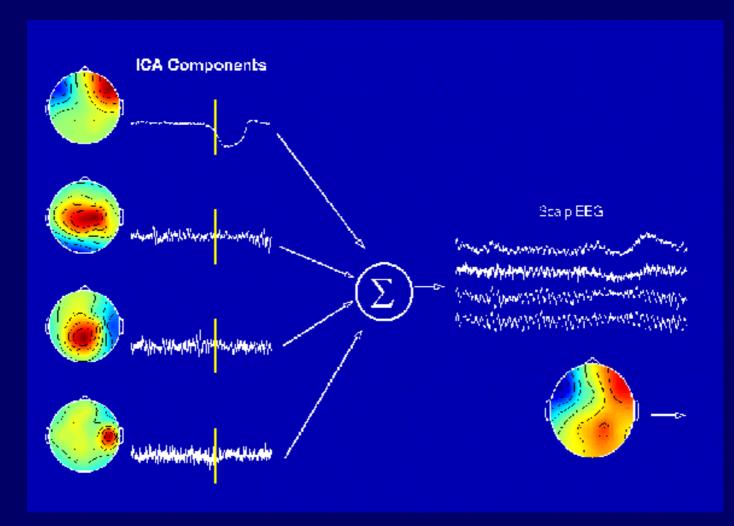


Brain-Machine Interface

(By ATR-Honda)



The Need for Biomedical Signal Processing



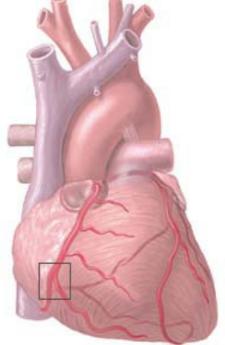
What is it?

- <u>Biomedical Signal Processing</u>: Application of *signal processing methods*, such as filtering, Fourier transform, spectral estimation and wavelet transform, to *biomedical problems*, such as the analysis of cardiac signals, the breathing cycle,...etc.
- <u>A broader aspect</u>: Biomedical imaging, genomic signal processing,...etc.

 Heart attack: Coronary artery disease,
 blockage of blood supply to the myocardium.

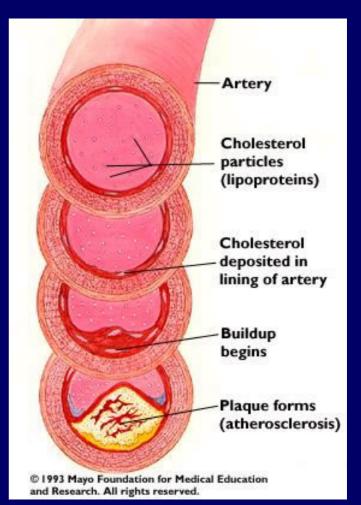


Blockage in right coronary artery



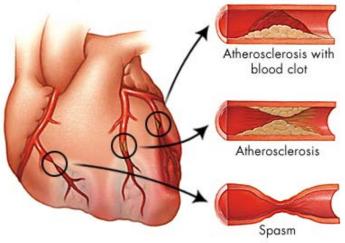
adam.com

• Plaque: A gradual buildup of fat (cholesterol) within the artery wall.



- Symptoms:
 - Chest pressure with stress, heart burn, nausea, vomiting, shortness of breath, heavy sweating.
 - Chest pain, heart attack, arrhythmias.

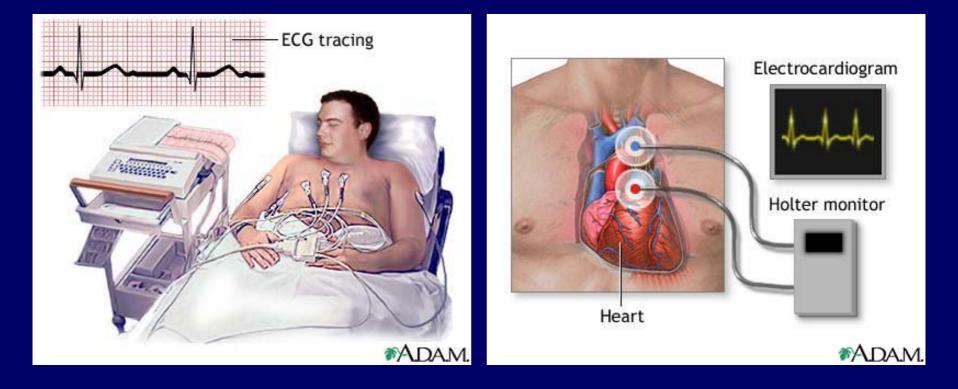




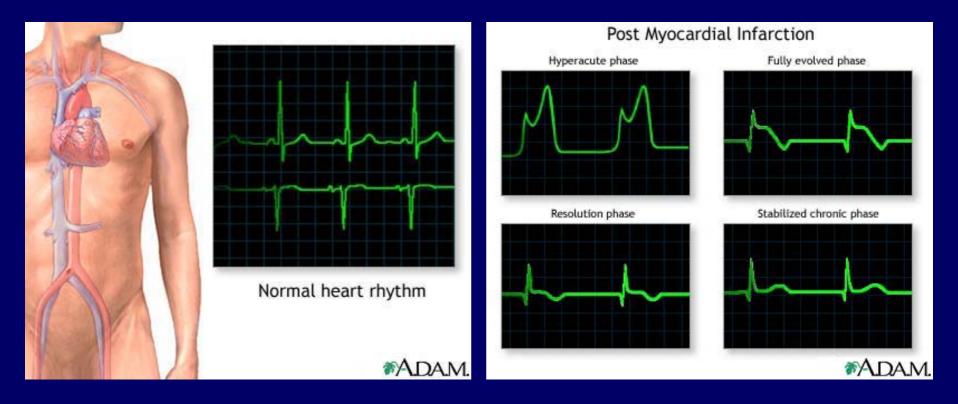
• Diagnosis:

- Prehospital electrocardiography (ECG).
- Continuous/serial ECG.
- Exercise stress ECG.
- Biochemical tests and biomarkers.
- Sestamibi myocardial perfusion imaging.
- Echocardiography.
- Computer-based decision aids.

Medical Diagnosis: ECG



Medical Diagnosis: ECG



Medical Diagnosis: ECG

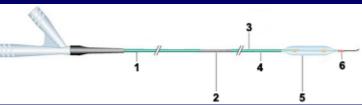
Atrioventricular block ECG tracing

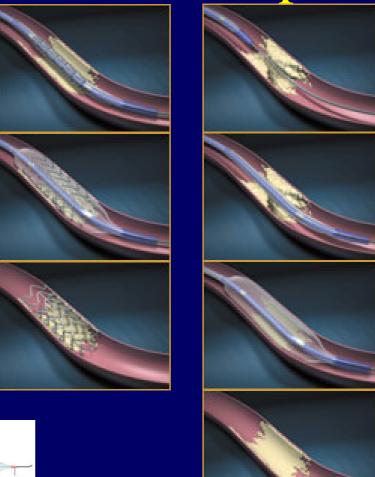


Ventricular tachycardia

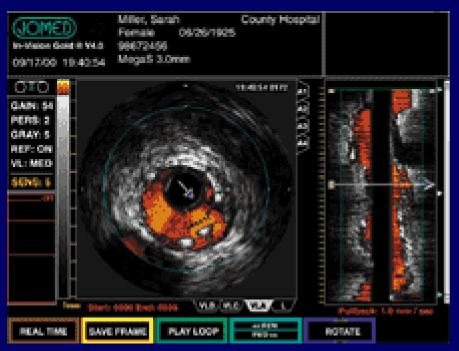


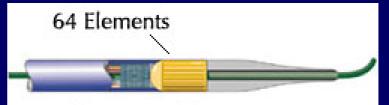
- Treatment:
 - Angioplasty.
 - Stent implantation.
 - Atherectomy.
 - Coronary bypass surgery.
 - Intravascular radiotherapy.
 - Excimer laser.





Imaging:
Ultrasound.

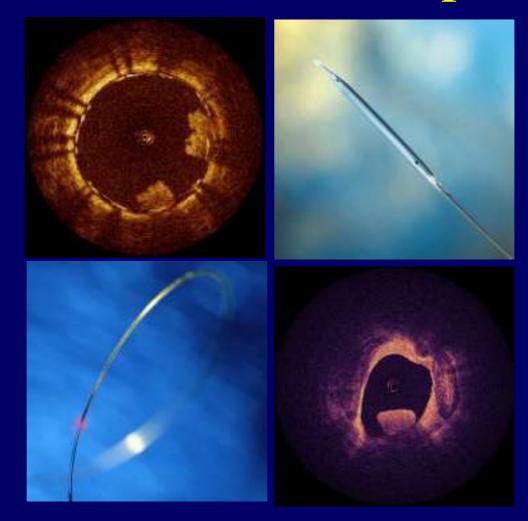




Solid-State Catheter Design



• Imaging: – Optics.



Biomedical Signals: Broader Definition

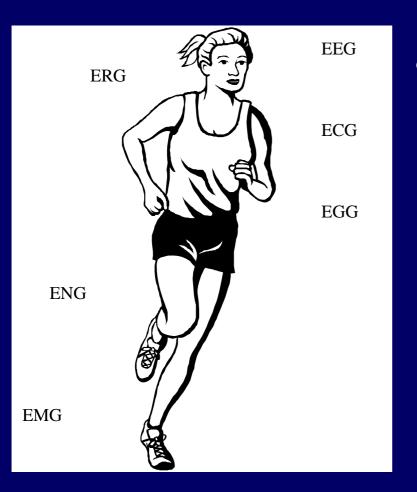
- Signals as a result of physiological activities in the body:
 - Electrical and Non-electrical
- Invasive/Non-invasive interrogation of an external field with the body
- Diagnosis and therapy
- → <u>Will focus mostly on bioelectric signal.</u>

Outline

- Bioelectrical signals:
 - Excitable cells
 - Resting/action potential
- ECG, EEG,...etc
- Applications of signal processing techniques

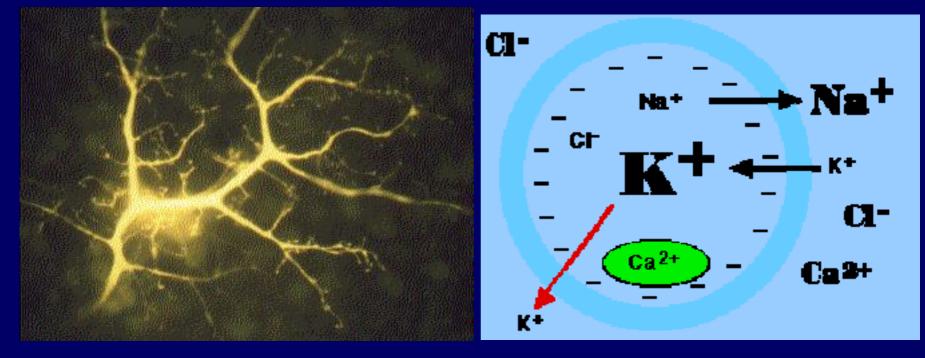
 Sampling, filtering, data compression,...etc
- Non-stationary nature of biomedical signals

Bioelectrical Signals



• The bioelectric signals represent many physiological activities.

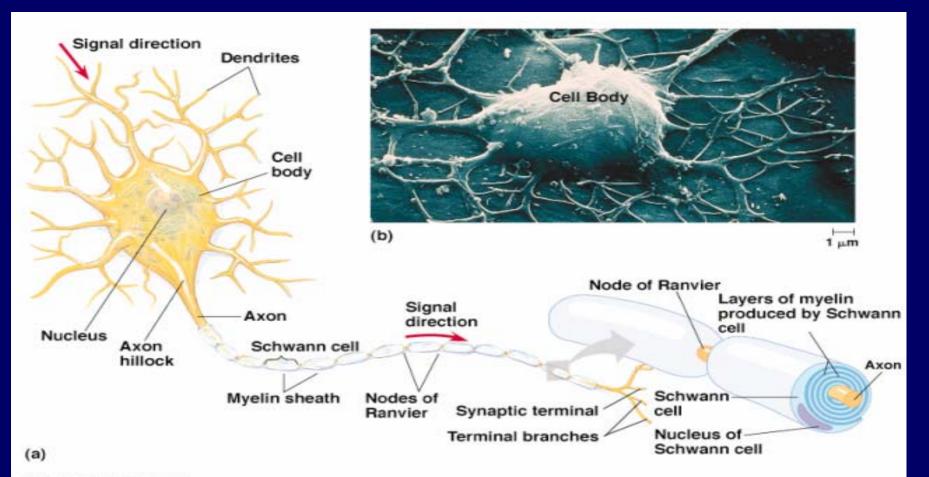
Excitable Cells



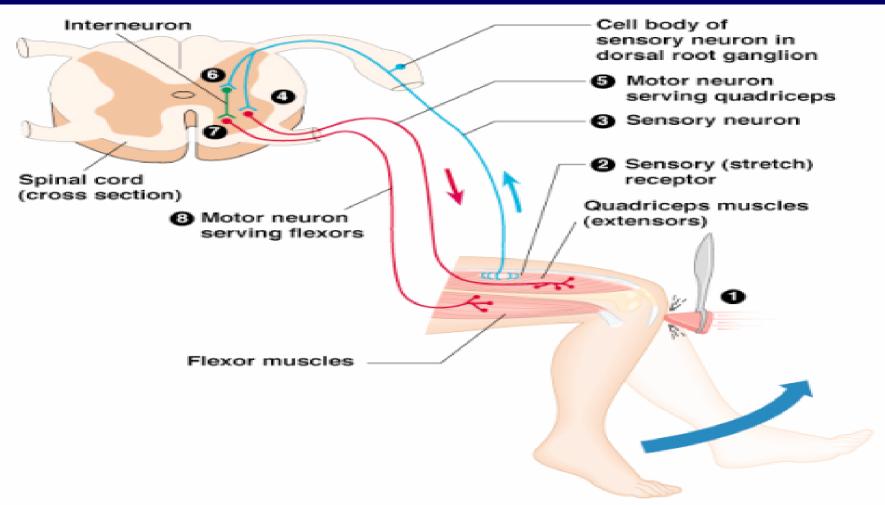
Neuron (Rabbit Retina)

Ionic Relations in the Cell

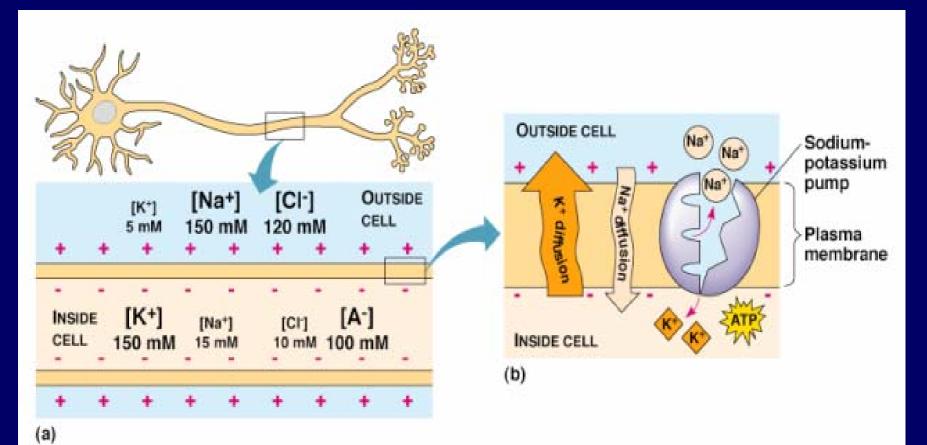
Structural unit



Functional unit

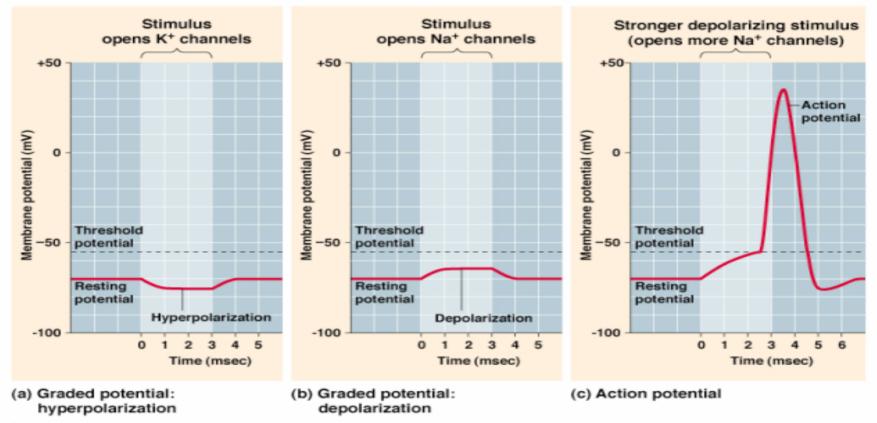


Neural signaling (I)



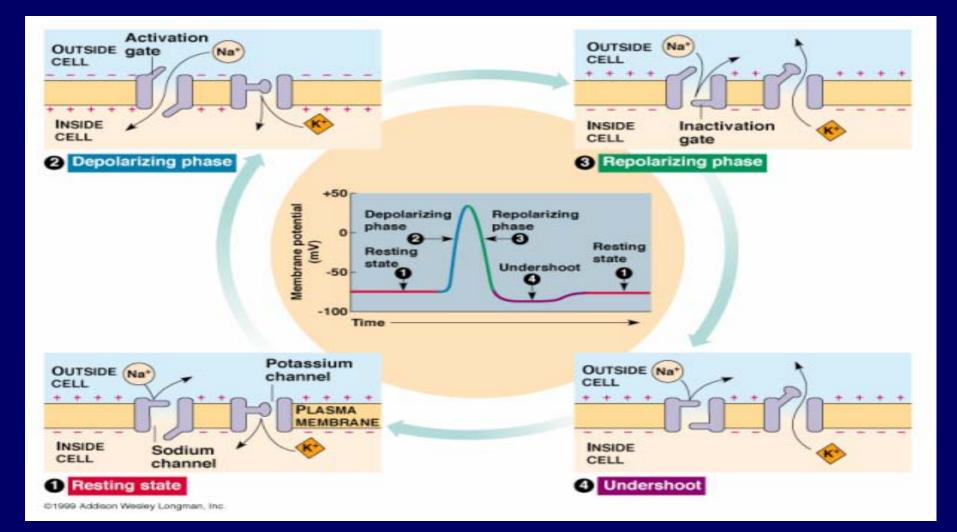
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Neural signaling (II)

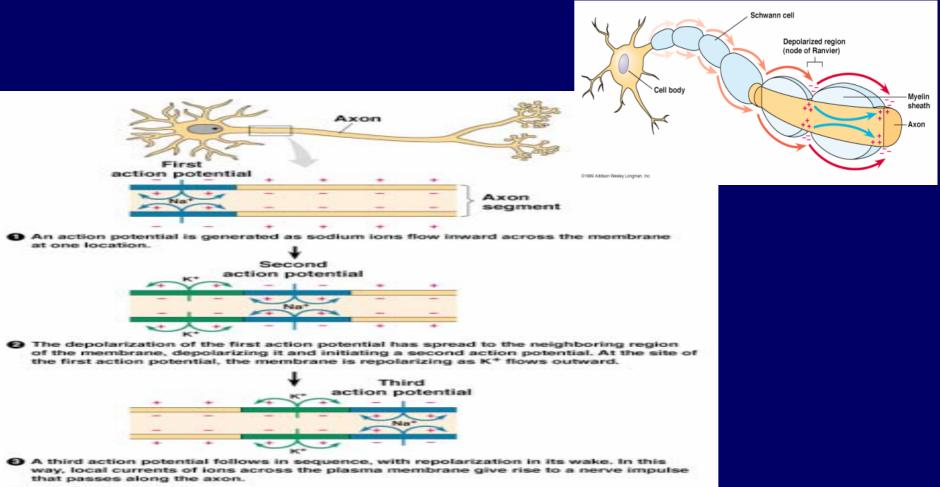


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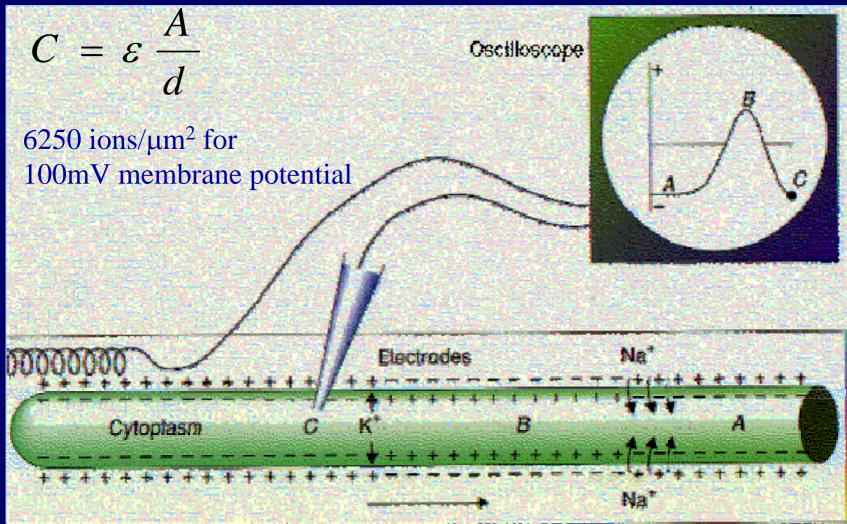
Neural signaling (III)



Neural signaling (IV)



Measurements of Action Potential



Goldman Equation

$$E = \frac{RT}{F} \ln \left\{ \frac{P_{K}[K]_{o} + P_{Na}[Na]_{o} + P_{Cl}[Cl]_{i}}{P_{K}[K]_{i} + P_{Na}[Na]_{i} + P_{Cl}[Cl]_{o}} \right\}$$

- E: Equilibrium resting potential
- *R*: Universal Gas Constant (8.31 J/(mol*K))
- *T*: Absolute temperature in K
- *F*: Faraday constant (96500 C/equivalent)
- $P_{\rm M}$: Permeability coefficient of ionic species M.

Example: Ion Concentration

Species	Intracellular (millimoles/L)	Extracellular (millimoles/L)
Na ⁺	12	145
K^+	155	4
Cl-	4	120

(For frog skeletal muscle)

Example: Equilibrium Resting Potential for frog skeletal muscle

- $P_{\rm N_a}$: 2 X 10⁻⁸ cm/s
- $P_{\rm K}$: 2 X 10⁻⁶ cm/s
- $P_{\rm Cl}: 4 \ge 10^{-6} \text{ cm/s}$
- *E*=-85.3 mV

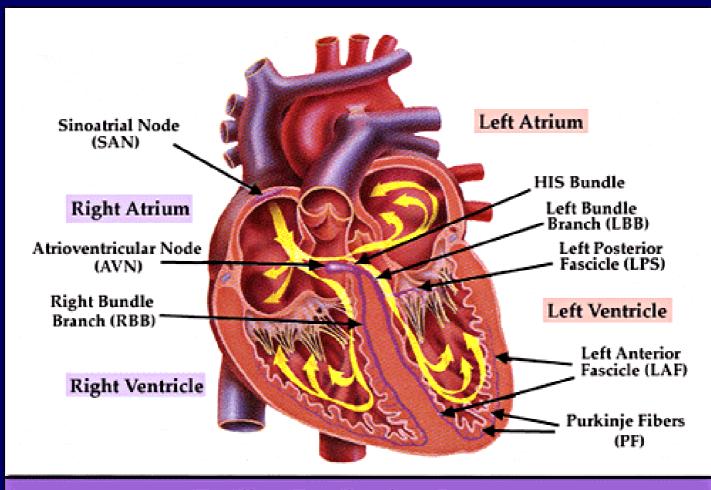
Electrocardiogram (ECG)

ECG

- One of the main methods for assessing heart functions.
- Many cardiac parameters, such as heart rate, myocardial infarction, and enlargement can be determined.
- Five special groups of cell:

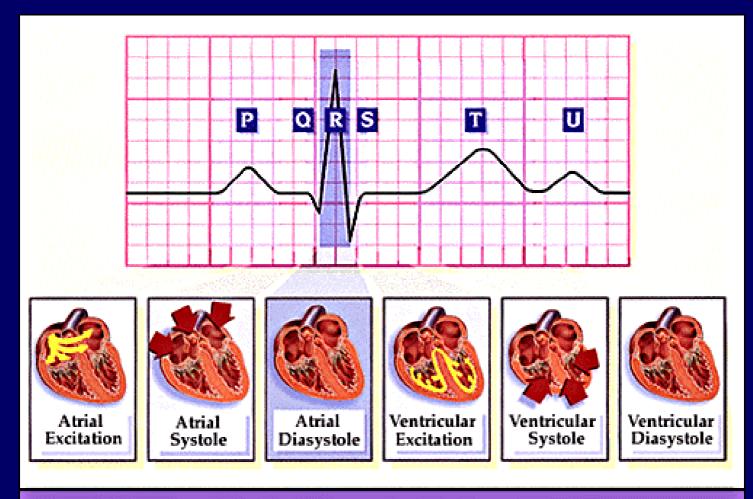
– SA, AV, common bundle, RBB and LBB.

ECG



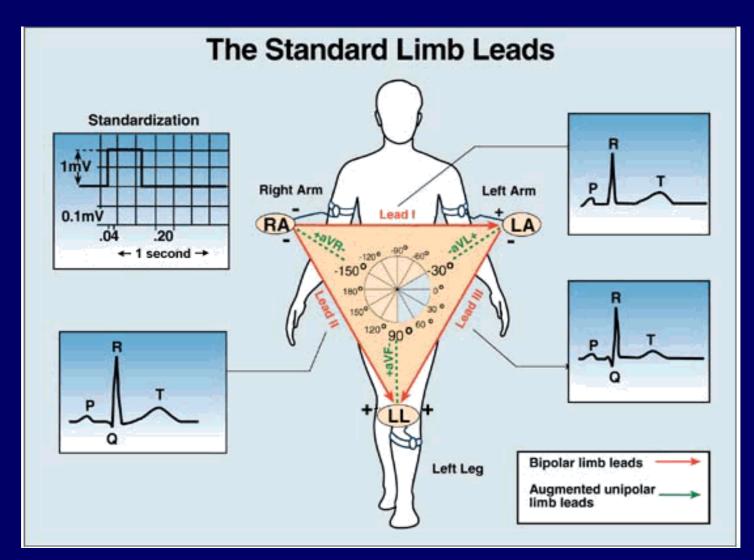
Cardiac Conduction System

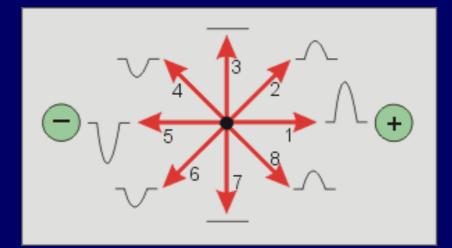
ECG

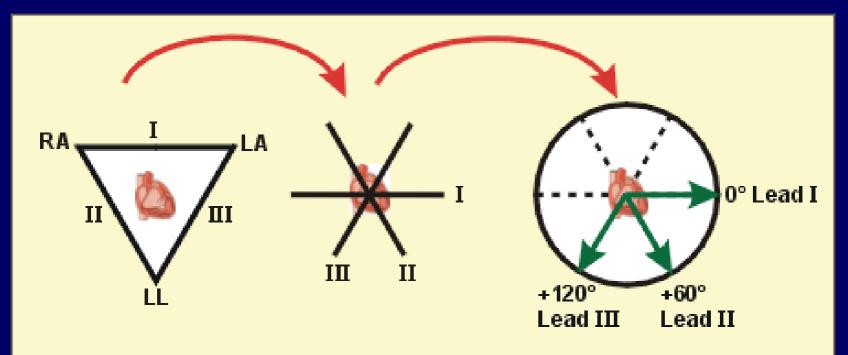


Electrical and Mechanical Events

ECG Leads

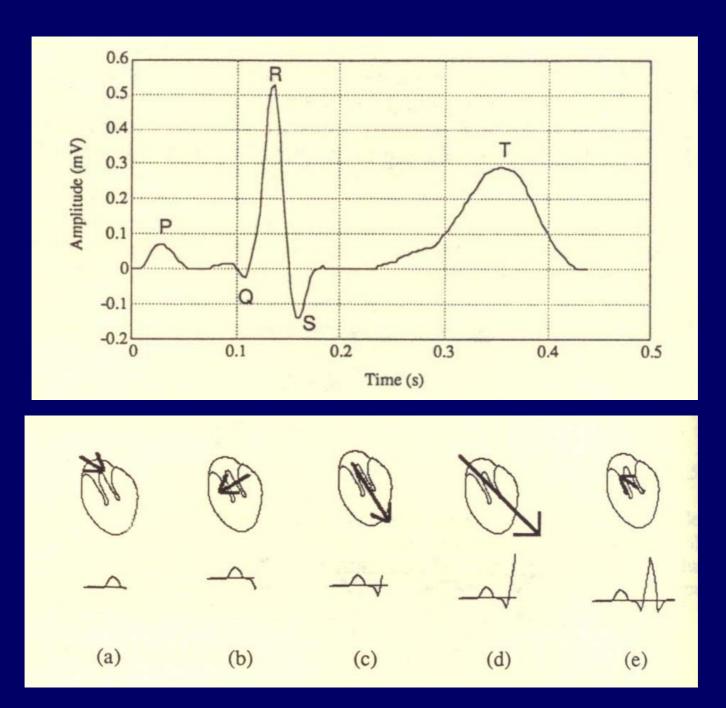




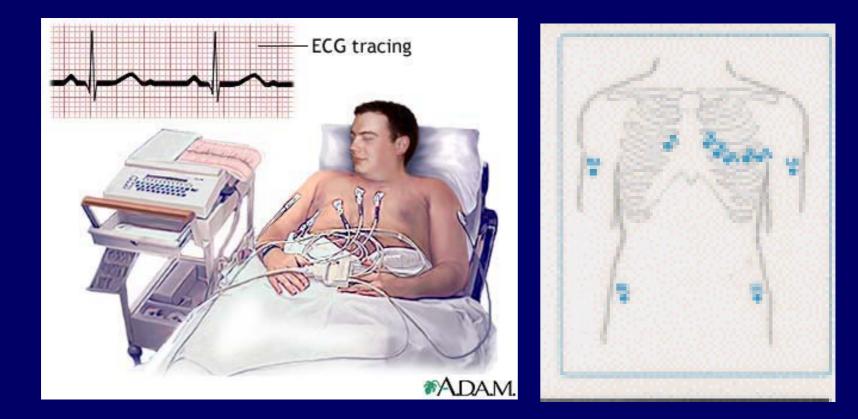


Einthoven's Triangle

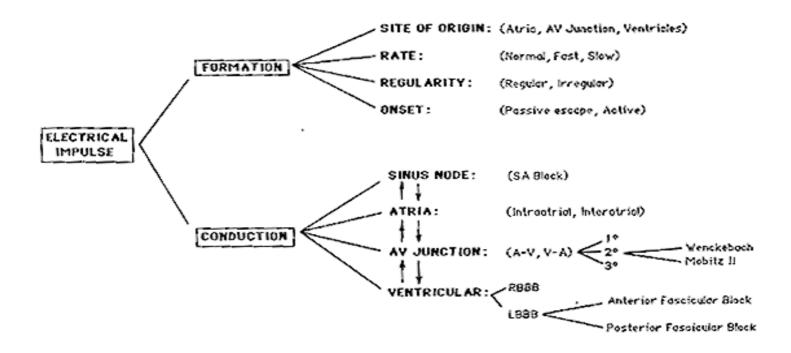
Axial Reference System

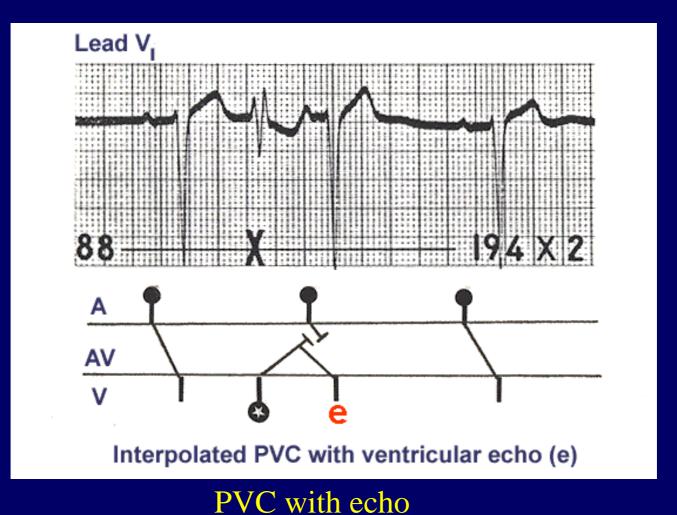


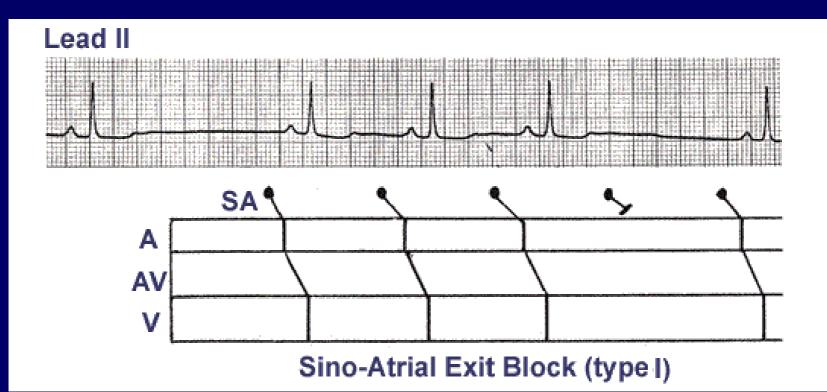
ECG Leads



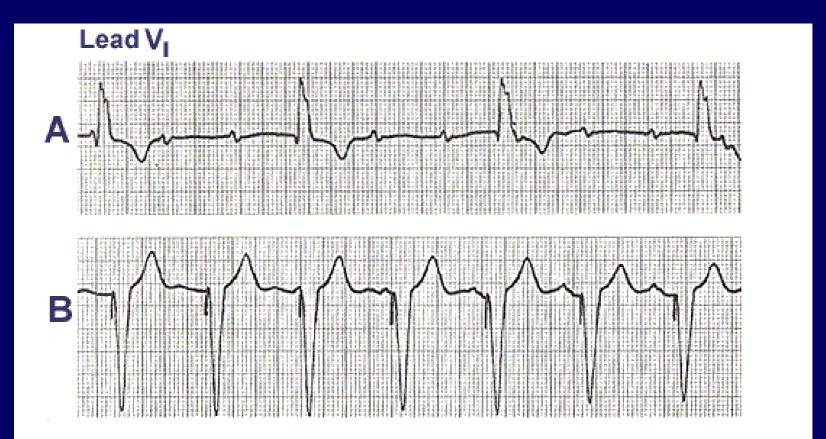
HOW TO THINK ABOUT ARRHYTHMIAS AND CONDUCTION DISTURBANCES



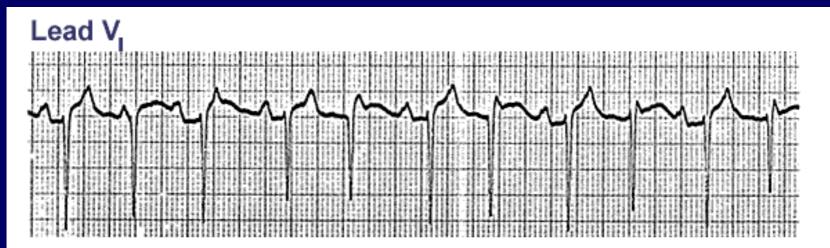




Conduction: SA Block (Type I)

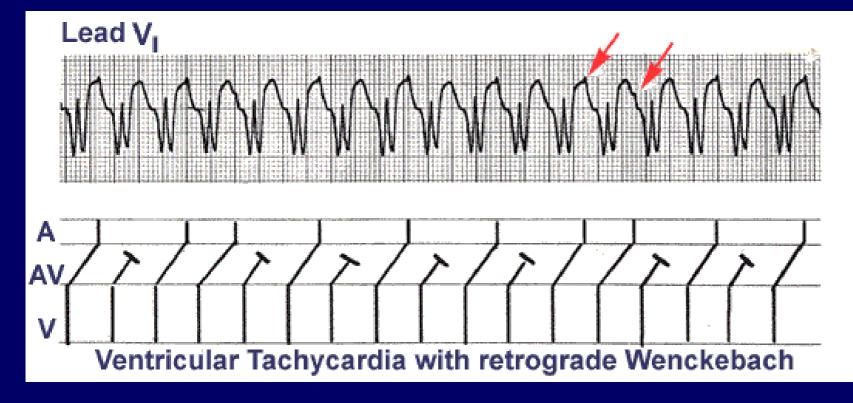


Conduction: Complete AV Block

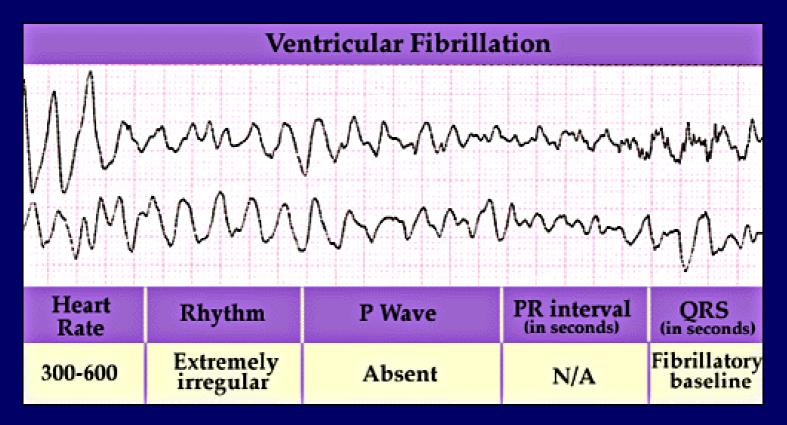


Atrial tachycardia with variable AV block

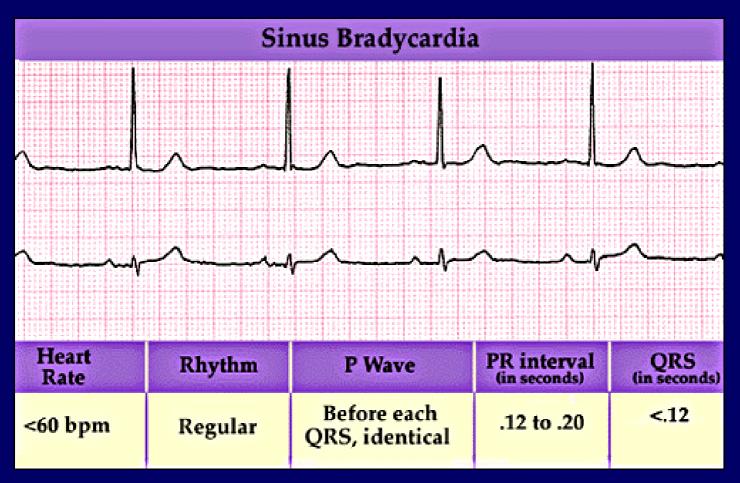
Rate: Atrial Tachycardia (160 bpm)



Rate: Ventricular Tachycardia

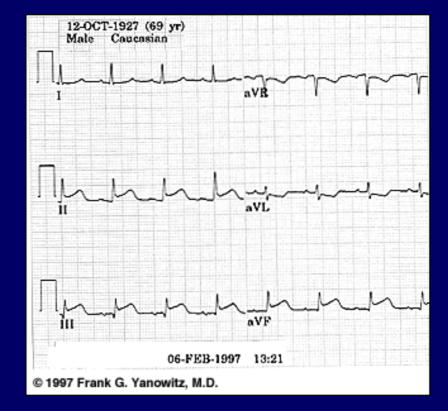


Rate: Ventricular Fibrillation

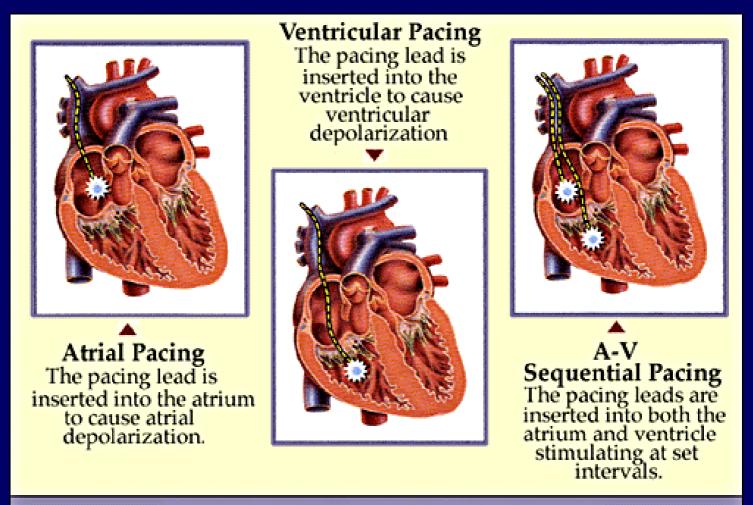


Rate: Sinus Bradycardia

- Other abnormalities:
 Myocardial infarction
 - Atrial/Ventricular enlargement
 - ST segment elevation



Pace Makers



Pacemaker Lead Wire Replacement

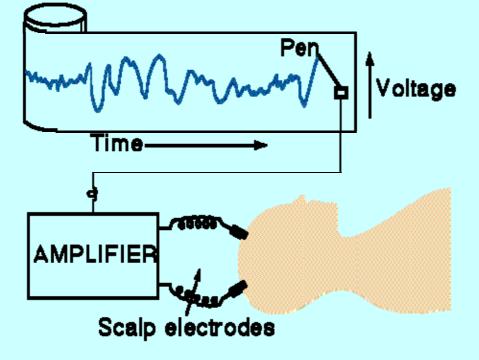
Electroencephalogram (EEG)

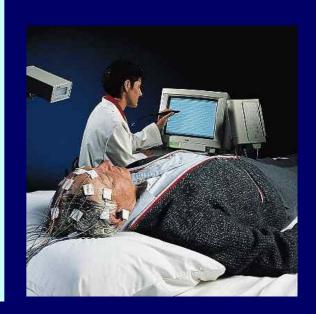
EEG

- Electrical potential fluctuations of the brain.
- Under normal circumstances, action potentials in axons are asynchronous.
- If simultaneous stimulation, projection of action potentials are detectable.
- The analysis is based more on frequency than morphology.

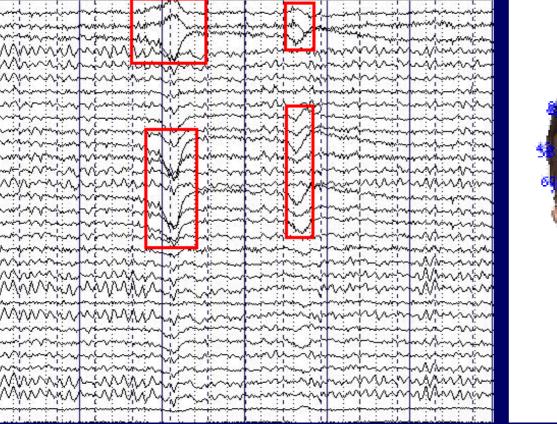
EEG: Instrument

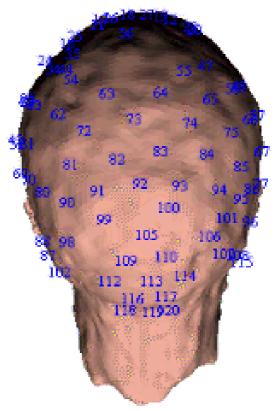
Electroencephalography Recording System



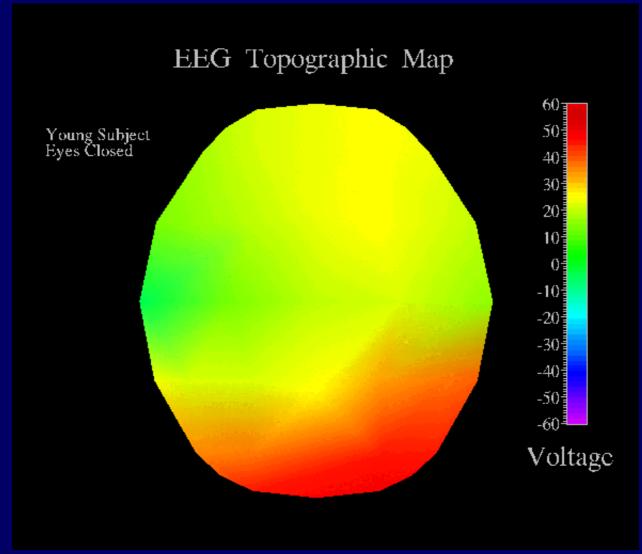


EEG: Spatial and Temporal Characteristics

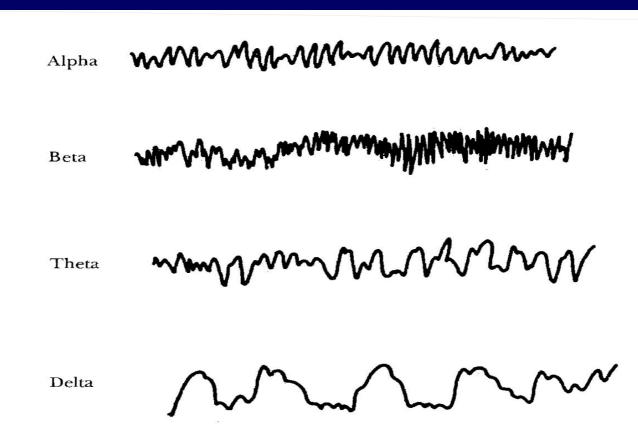


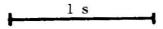


EEG: Presentation



EEG Classification





EEG Classification

- Alpha:
 - 8 to 13Hz.
 - Normal persons are awake in a resting state.
 - Alpha waves disappear in sleep.
- Beta:
 - 14 to 30Hz.
 - May go up to 50Hz in intense mental activity.
 - Beta I waves: frequency about twice that of the alpha waves and are influenced in a similar way as the alpha waves.
 - Beta II waves appear during intense activation of the central nervous system and during tension.

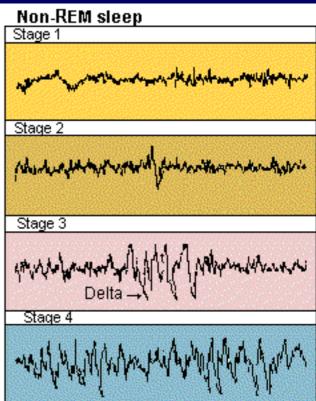
EEG Classification

- Theta waves:
 - 4 to 7Hz.
 - During emotional stress.
- Delta waves
 - Below 3.5Hz.
 - Deep sleep or in serious organic brain disease.

EEG Applications

- Epilepsy.
- Dream:

Waking	
A. Alert (beta rhytm)	
18 have a start for the product of the start	
B. Relaxed with eyes closed (alpha rhytm)	
ؾ؋ ڂؾٳؙڂڟۏڣڂڟ ؾڹۅٳۑۮڂ ^{ۑڛ} ۅۮڂۼۑڟڶڰ ^ۅ ۮڹؽۑڟڣۯڔۑؽڰڹۯڂ؆ڸڹڂڕڞٳ؇	
REM Sleep	



Other Biomedical Signals

- Electrical:
 - Electroneurogram (ENG)
 - Electromyogram (EMG)
 - Electroretinogram (ERG)
 - Electrogastrogram (EGG).

Other Non-Electrical Biomedical Signals

- Circulatory system
 - Blood pressure
 - Heart sound
 - Blood flow velocity
 - Blood flow volume

Other Non-Electrical Biomedical Signals

- Respiratory system
 - Respiratory pressure
 - Gas-flow rate
 - Lung volume
 - Gas concentration

Applications of Signal Processing Techniques

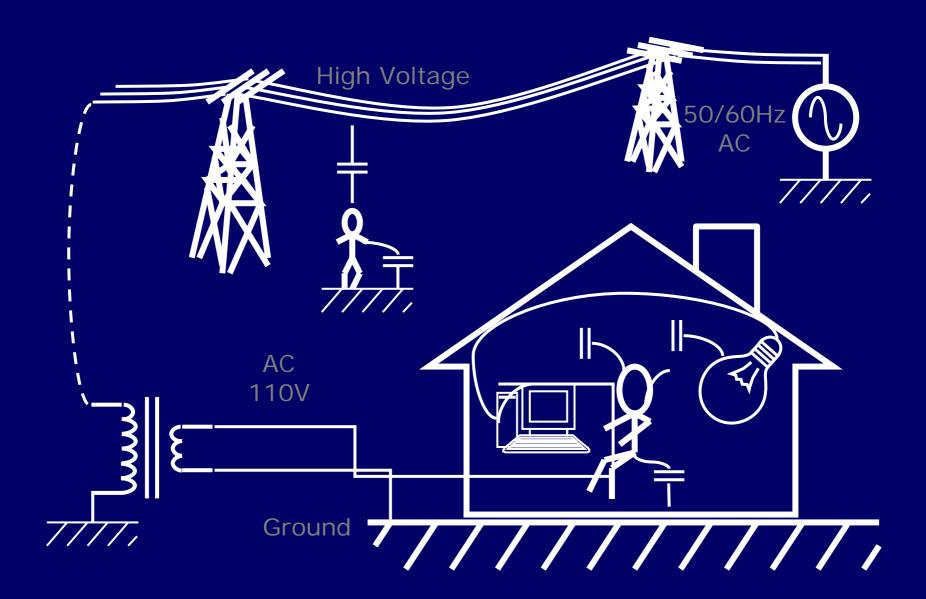
Sampling

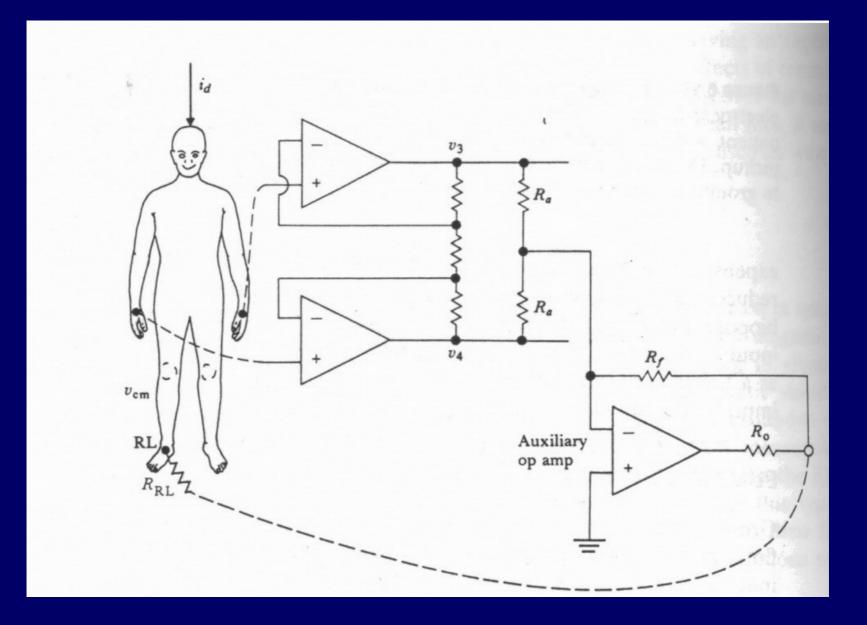
- Digital analysis and presentation of biomedical signals.
- Sampling requirements.
 - Low frequencies.
 - Frequency ranges of different physiological signals may be overlapping.
 - Electronic noise and interference from other physiological signals.
 - Very weak (maybe μV level), the pre-amp circuit is often very challenging.

Filtering

- Digital filters are used to keep the in-band signals and to reject out-of-band noise.
- Low-pass, band-pass, high-pass and band-reject.
- Similar to those of other applications.

Noise Sources of ECG

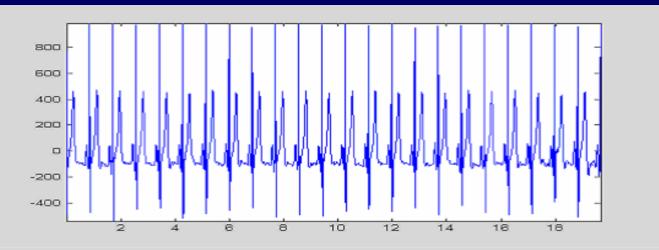


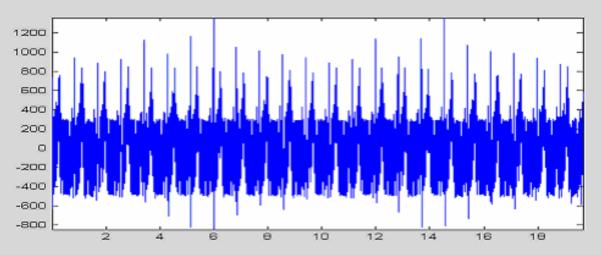


Noise Sources

- Inherent
- Instrumentation
- Environment
- From the body
- •

Ideal Signal Vs. Signal with Powerline Noise

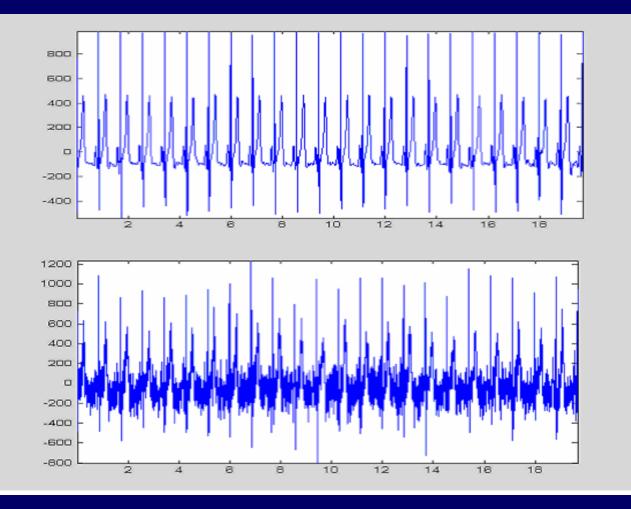




Ideal Signal Vs. Signal with Powerline Noise

- Powerline interference consists of 60Hz tone with random initial phase.
- It can be modeled as sinusoids and its combinations.
- The characteristics of this noise are generally consistent for a given measurement situation and, once set, will not change during a detector evaluation. Its typical SNR is in the order of 3dB.

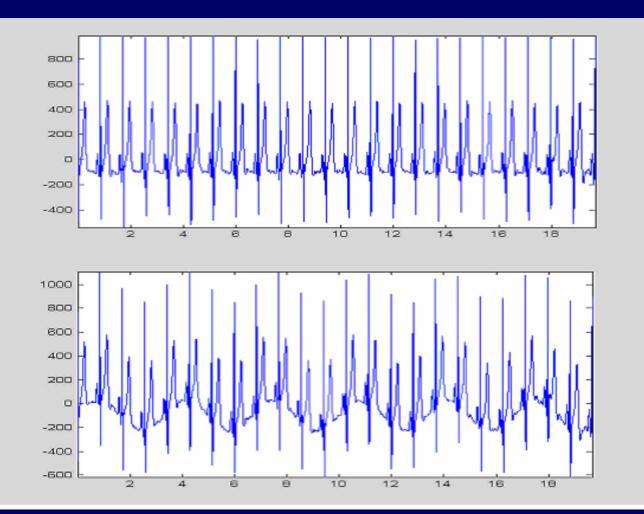
Ideal Signal Vs. Signal with Electromyographic Noise



Ideal Signal Vs. Signal with Electromyographic Noise

- EMG noise is caused by muscular contractions, which generate millivolt-level potentials.
- It is assumed to be zero mean Gaussian noise. The standard deviation determines the SNR, whose typical value is in the order of 18dB.

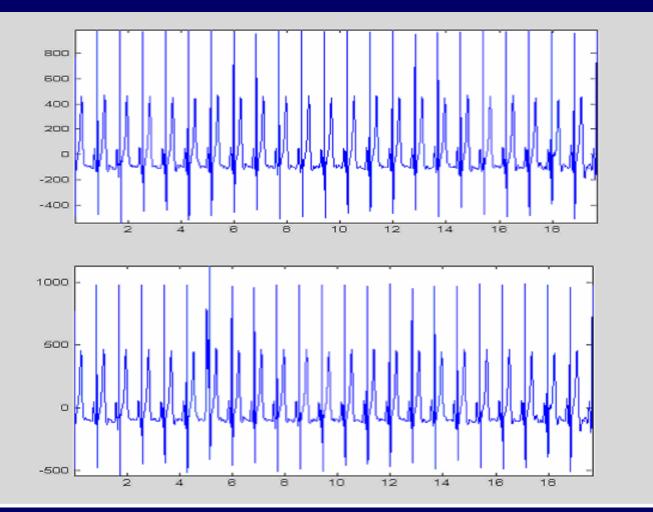
Ideal Signal Vs. Signal with Respirational Noise



Ideal Signal Vs. Signal with Respirational Noise

- Respiration noise considers both the sinusoidal drift of the baseline and the ECG sinusoidal amplitude modulation.
- The drift can be represented as a sinusoidal component at the frequency of respiration added to the ECG signal.
- The amplitude variation is about 15 percent of peak-to-peak ECG amplitude. It is simulated with a sinusoid of 0.3Hz frequency with typical SNR 32dB.
- The modulation is another choice of representing respiration noise. It can be simulated with 0.3Hz sinusoid of 12dB SNR.

Ideal Signal Vs. Signal with Motion Artifacts



Ideal Signal Vs. Signal with Motion Artifacts

- Motion artifact is caused by displacements between electrodes and skin due to patients' slow movement.
- It is simulated with an exponential function that decays with time.
- Typically the duration is 0.16 second and the amplitude is almost as large as the peak-to-peak amplitude.
- The phase is random with a uniform distribution.

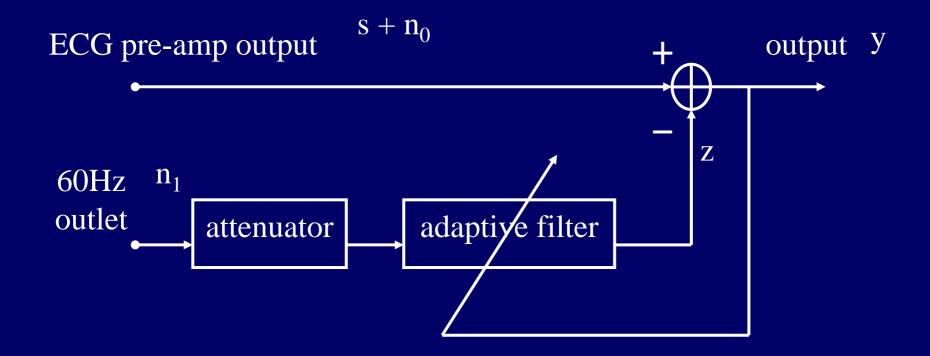
Noise Removal

• The four types of noises are mostly sinusoidal or Gaussian. The sinusoidal noises are usually removed with a notch filter. Other distortions are zeroed out using the moving average.

Adaptive Noise Cancellation

- Noise from power line (60Hz noise).
- The noise is also in the desired frequency range of several biomedical signals (e.g., ECG), notch filter is required.
- Adaptive filtering: The amplitude and exact frequency of the noise may change.

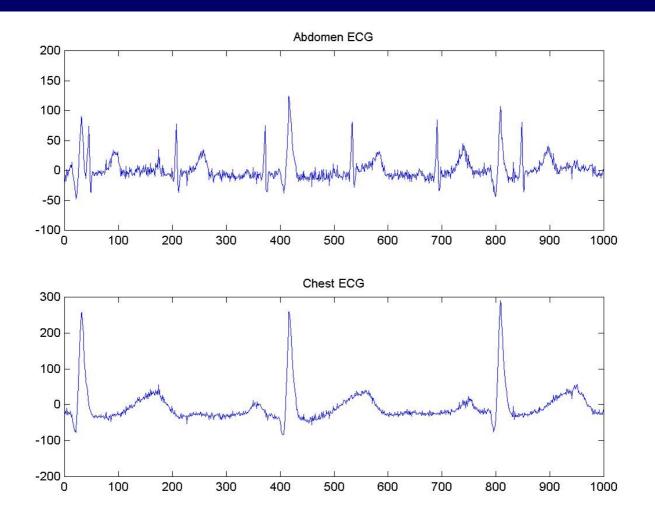
Adaptive Filter



Adaptive Filter

 $y(nT) = s(nT) + n_0(nT) - z(nT)$ $y^2 = s^2 + (n_0 - z)^2 + 2s(n_0 - z)$ $E[y^2] = E[s^2] + E[(n_0 - z)^2] + 2E[s(n_0 - z)] = E[s^2] + E[(n_0 - z)^2]$ $\min E[y^2] = E[s^2] + \min E[(n_0 - z)^2]$

Adaptive Filtering for Fetal ECG



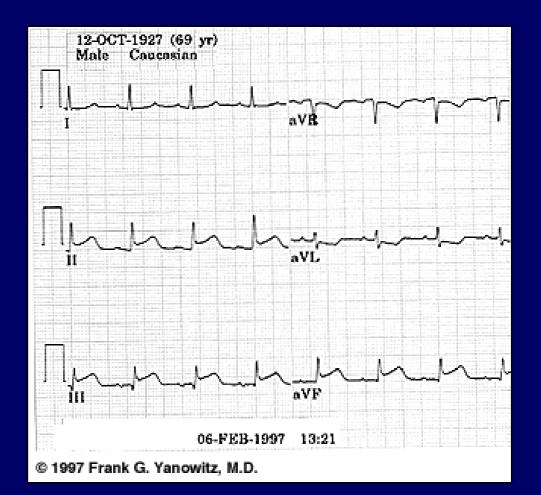
Pattern Recognition

- Abnormal physiological signals vs. the normal counterparts.
- An average of several known normal waveforms can be used as a template.
- The new waveforms are detected, segmented and compared to the template.
- Correlation coefficient can be used to quantify the similarity.

$$\rho = \frac{\sum_{i=1}^{N} (T_i - \mu_T) (X_i - \mu_X)}{\sqrt{\sum_{i=1}^{N} (T_i - \mu_T)^2} \sqrt{\sum_{i=1}^{N} (X_i - \mu_X)^2}}$$

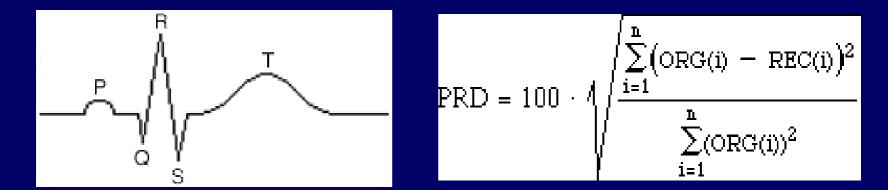
Pattern Recognition

Ex. ECG

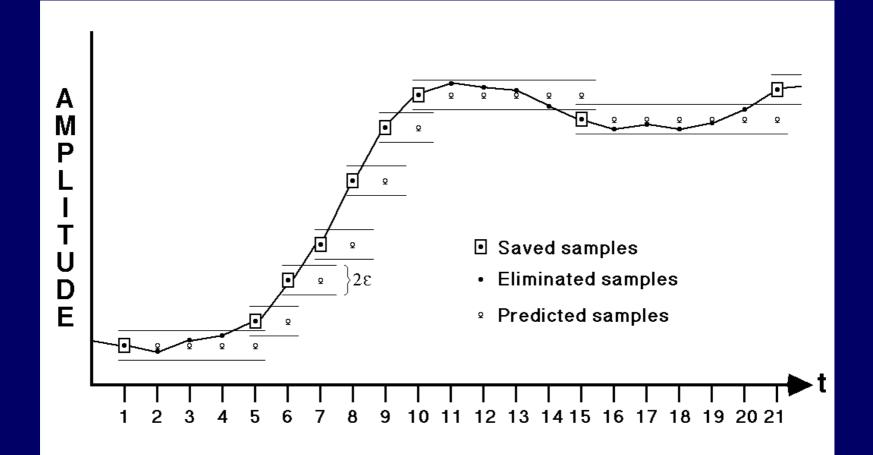


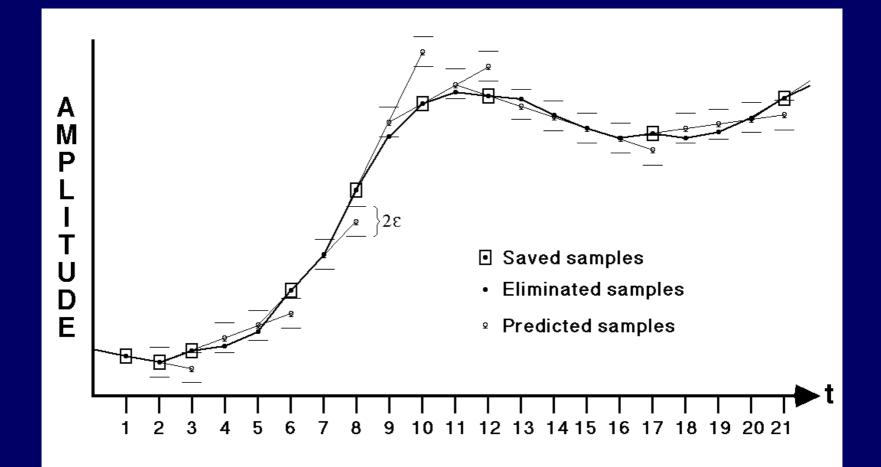
Data Compression

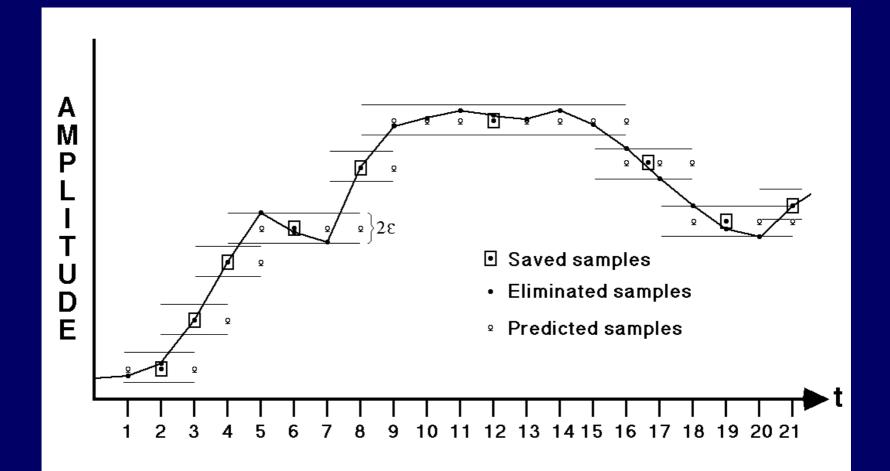
- For large amount of data (e.g., 24 hour ECG).
- Must not introduce distortion, which may lead to wrong diagnosis.
- Formal evaluation is necessary.



WGAQQQQQRBCCCCCHZY



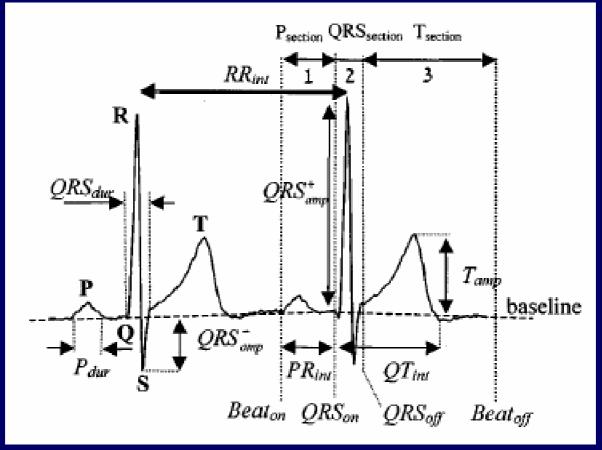




Is straightforward implementation sufficient for biomedical signals?

Characteristics of Biomedical Signals (I): Weak \neq Unimportant

• The information is in the details:







JPEG Compression



4302 Bytes



2245 Bytes



1714 Bytes



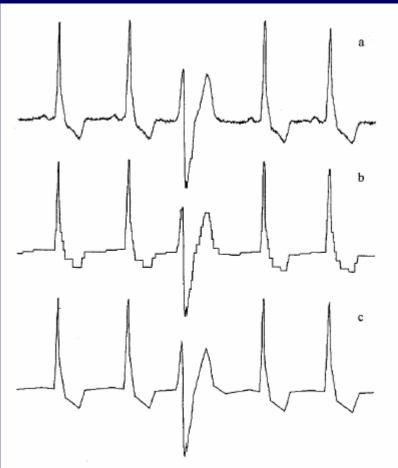
4272Bytes





1708 Bytes

2256 Bytes Wavelet Compression

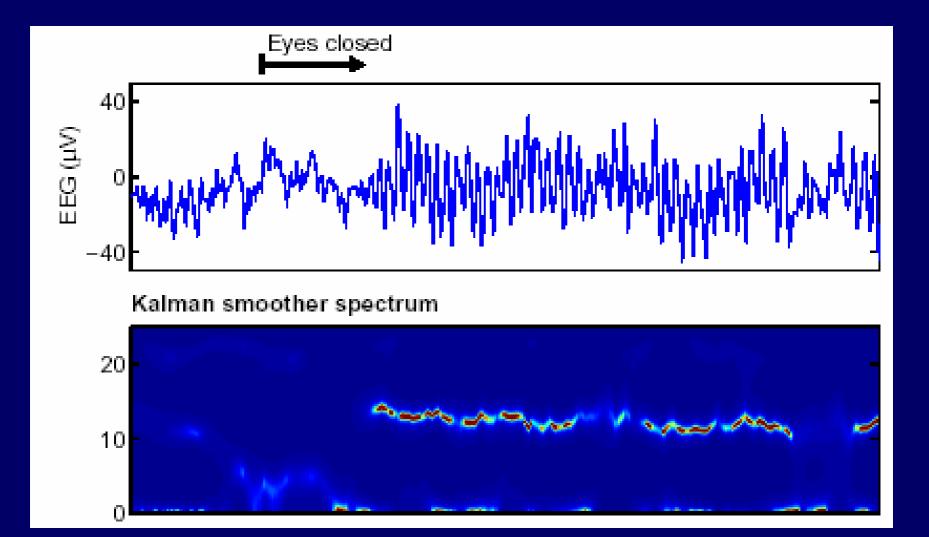


Characteristics of Biomedical Signals (II): Nonstationarity

Fourier Transform : $X(\omega) = \int_{-\infty}^{+\infty} x(t)e^{-j\omega t} dt$

- Fourier transform requires signal stationarity.
- Biomedical signals are often time-varying.
 - Short-time Fourier analysis
 - Time-frequency representation
 - Cyclo-stationarity

Nonstationarity: An EEG Example



Spectral Estimation for Nonstationary Signals

 Fourier Transform → Short Time Fourier Transform

$$X(\omega) = \int_{-\infty}^{+\infty} x(t) e^{-j\omega t} dt$$
$$\bigcup$$

$$X(\omega,a) = \int_{-\infty}^{+\infty} x(t)g(t-a)e^{-j\omega t}dt$$

Another Example: Signal Processing for Blood Velocity Estimation (Please refer to the class notes.)

Other Important Biomedical Applications

- Biomedical imaging:
 X-ray, CT, MRI, PET, OCT, Ultrasound,...
- Genomic signal processing
- ...,etc

Term Project Applications of PCA (Principle Component Analysis) in Biomedical Signal Processing Due 6/25/2007, Hardcopy in my mailbox (BL-425)