

Artificial Intelligence for Medical Data with Python

9 SAMPLE SLIDES

**6th session –
Machine learning algorithms for Image Processing
and other complex medical signals**

UNIVERSITY OF THE
AEGEAN



SCHOOL OF ENGINEERING
DEPARTMENT OF INFORMATION
AND COMMUNICATION
SYSTEMS ENGINEERING

Presenter: Panagiotis Symeonidis

Associate Professor

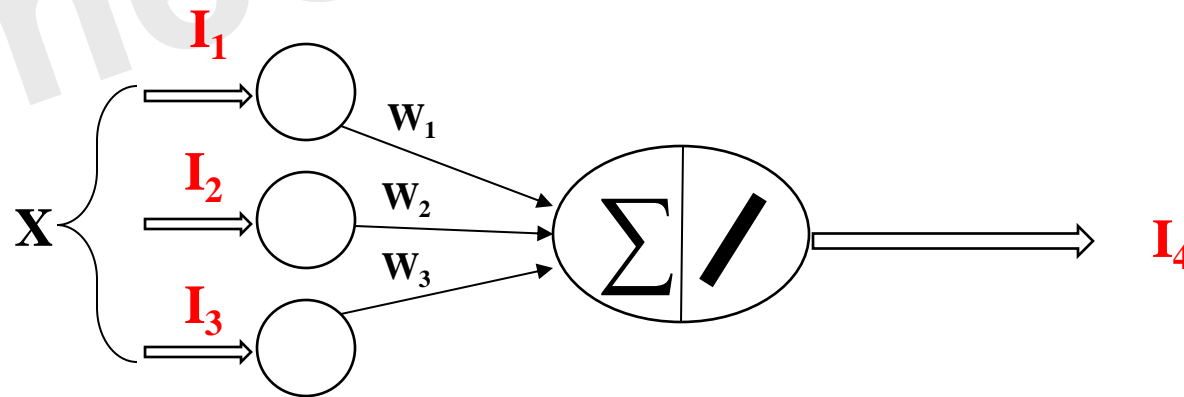
<http://panagiotissymeonidis.com>

psymeon@aegean.gr

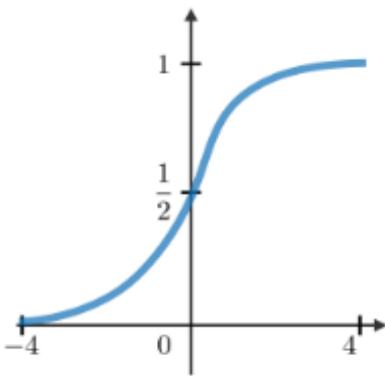
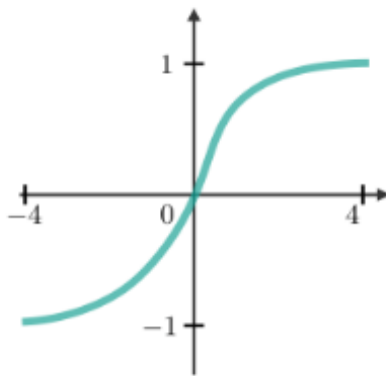
Single-Layer Perceptron for Item-based Collaborative Filtering

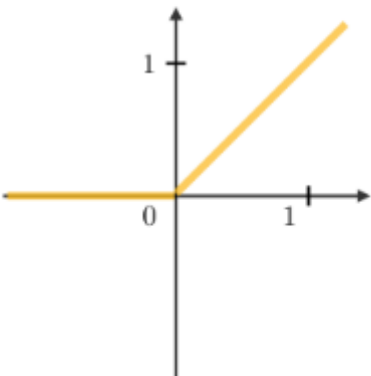
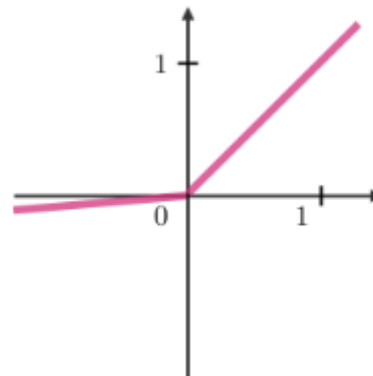
- Predict whether patient U_4 needs Drug I_4

	I_1	I_2	I_3	I_4
U_1	4	1	1	4
U_2	1	4	2	0
U_3	2	1	4	5
U_4	1	4	1	?



Activation Functions

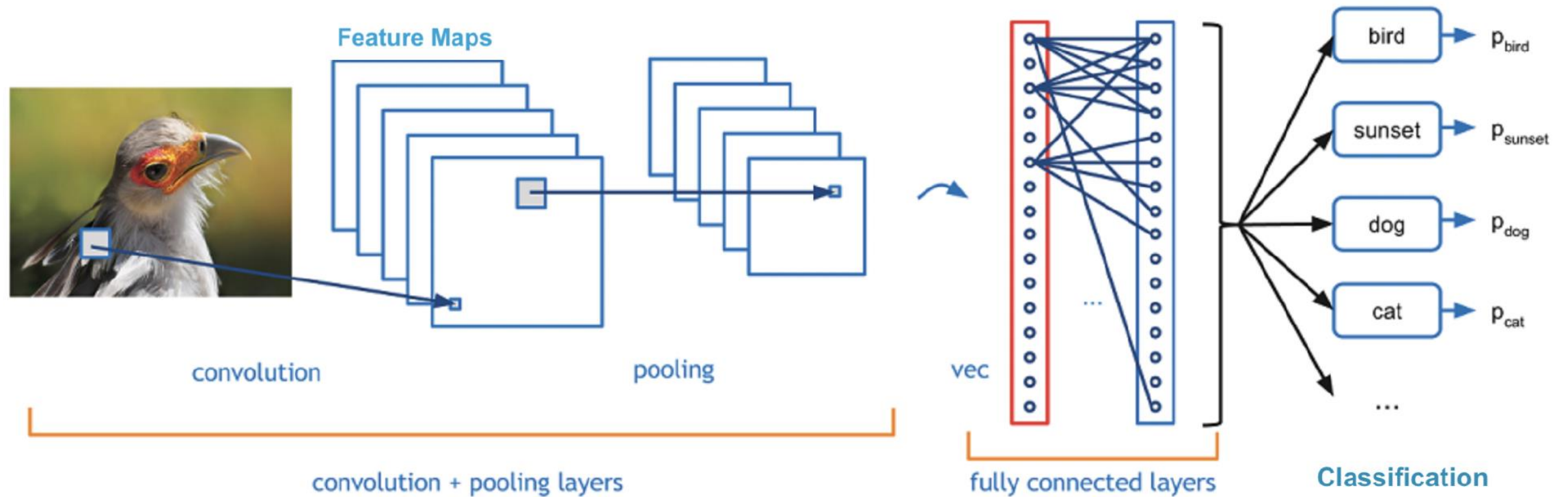
Sigmoid	Tanh
$g(z) = \frac{1}{1 + e^{-z}}$	$g(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$
	

ReLU	Leaky ReLU
$g(z) = \max(0, z)$	$g(z) = \max(\epsilon z, z)$ with $\epsilon \ll 1$
	

Pros and Cons of Activation Functions

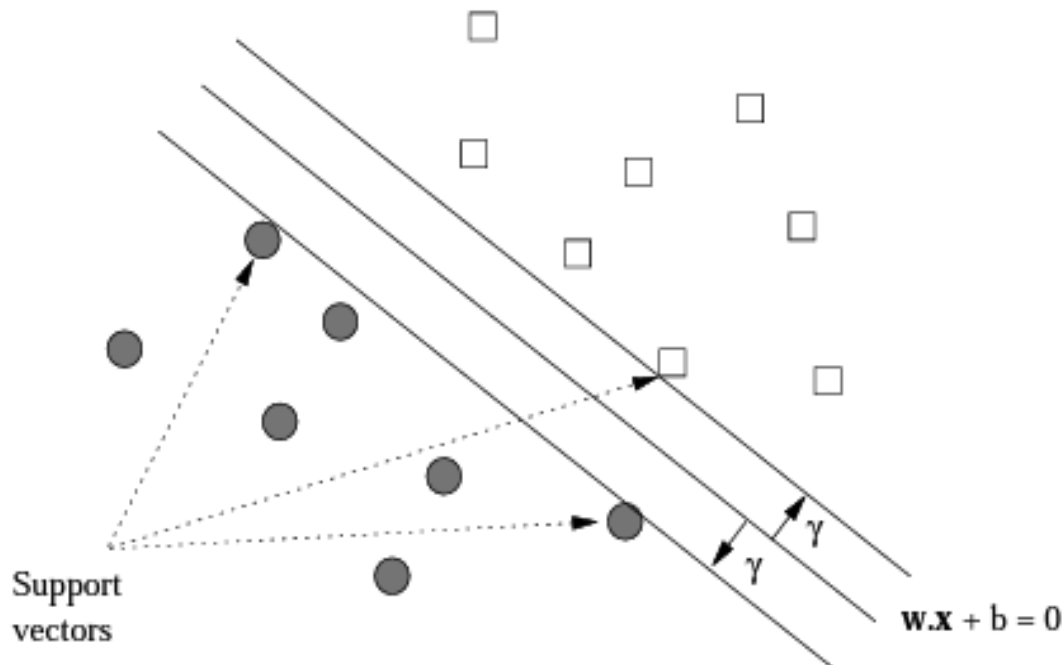
Properties	Sigmoid	Tanh	ReLU	Leaky ReLU
Continuous	yes	yes	Except at $x=0$	yes
Differentiable	yes	yes	Except at $x=0$	yes
Pros	<ul style="list-style-type: none">• Outputs in bounded range• Smooth	<ul style="list-style-type: none">• Outputs in bounded range• Smooth• Symmetric around origin• Zero-centered output	<ul style="list-style-type: none">• Fast• Sparsity• Avoids vanishing gradient• Ideal for deep architectures	<ul style="list-style-type: none">• Fast• Sparsity• Allows recovery of dead neurons
Cons	<ul style="list-style-type: none">• Saturates as we move away from 0 (vanishing gradient)• Computationally expensive• Lack of zero-centered output	<ul style="list-style-type: none">• Saturates as we move away from 0 (vanishing gradient)• Computationally expensive	<ul style="list-style-type: none">• Unbounded output's range (exploding gradients)• Less smooth• Lack of zero-centered output• Zero gradient for very negative inputs (dying ReLU)	<ul style="list-style-type: none">• Unbounded output's range• (exploding gradients)• Needs one more parameters

Example of a Convolutional Neural Network



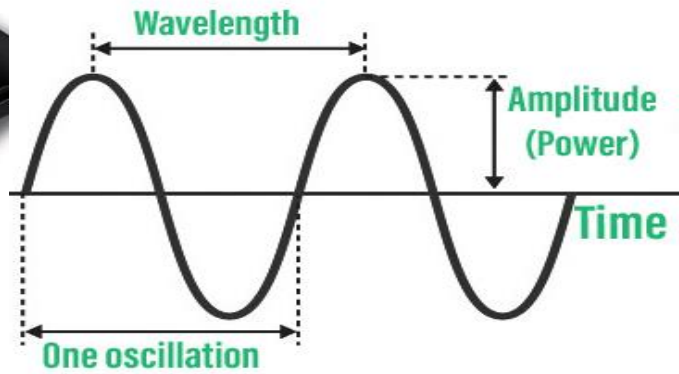
Support Vector Machines

- The goal of a Support Vector Machine (SVM) is to select a hyperplane $w \cdot x + b = 0$ that maximizes the distance γ between the hyperplane and any point of the training set.



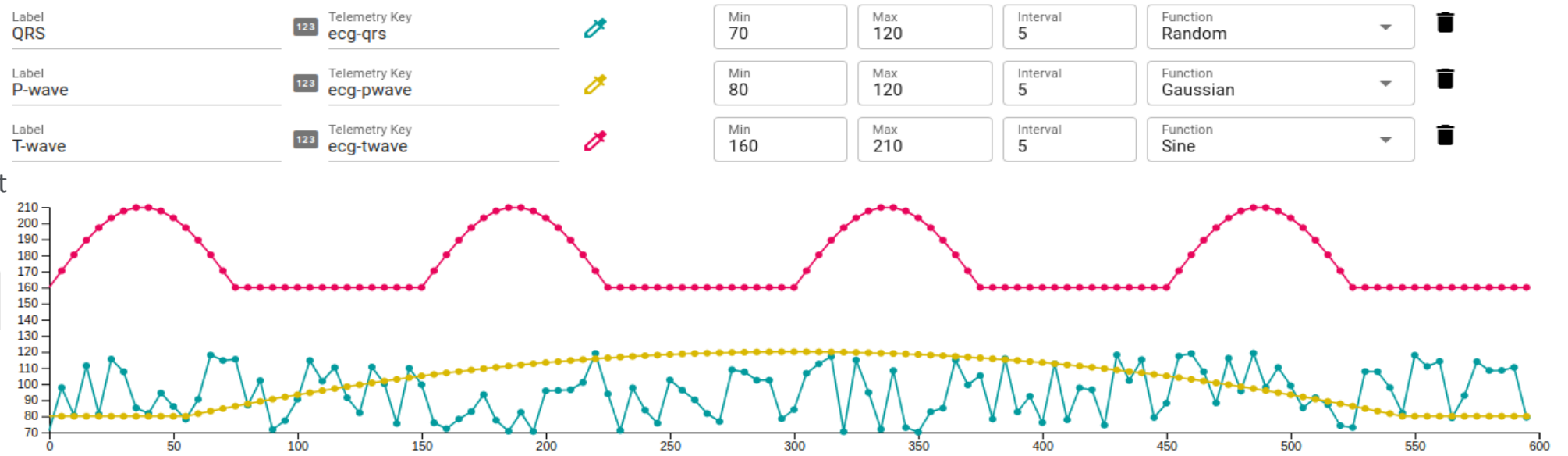
With a large margin, we are more certain to correctly classify points that are in the full data set and not only the training set

Heart Failure Prediction

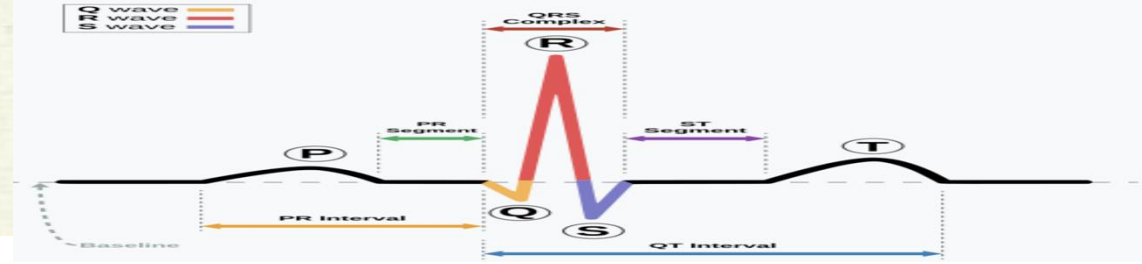


Parameter	Normal Range
ECG QRS width/amplitude	60-110msec/ $\leq 1\text{mV}$
ECG P-wave width/amplitude	80-110ms/ $\leq 0.1\text{mV}$
ECG T-wave width/amplitude	160-200ms/ $\leq 0.25\text{mV}$

A doctor observes a graph that shows the electrical activity (in volts) of the heart of a patient over a period of time (in seconds). Each local maximum of 0.004 volts corresponds to one heartbeat.



What are the main signals of ECG (Electrocardiogram)?

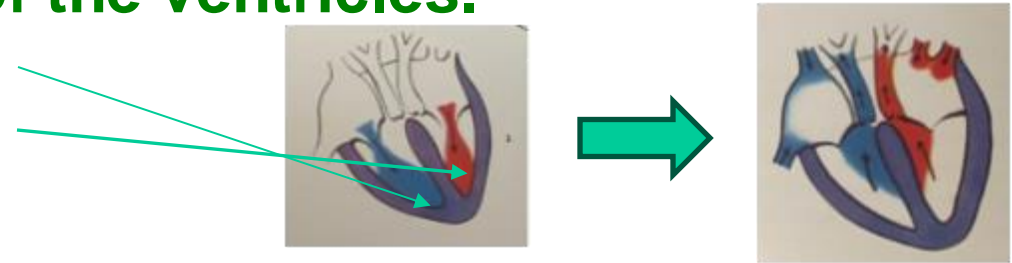


- There are three main components to an ECG:

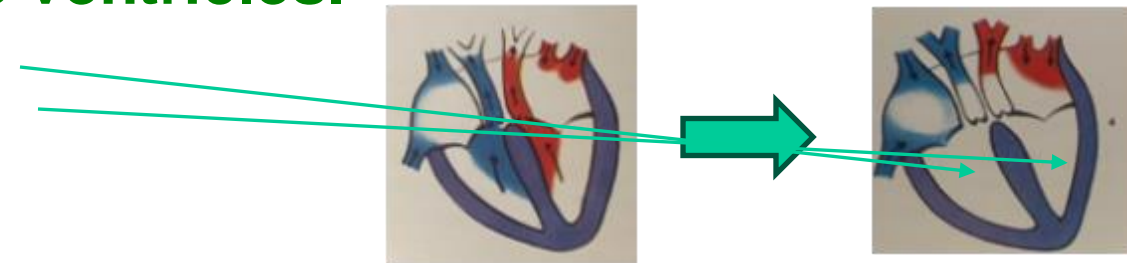
1. The **P wave** represents depolarization of the atria.



2. The **Q R S complex** represents depolarization of the ventricles.



3. The **T wave** represents repolarization of the ventricles.



Heart movements

Detection Example of Ischemic Heart Failure

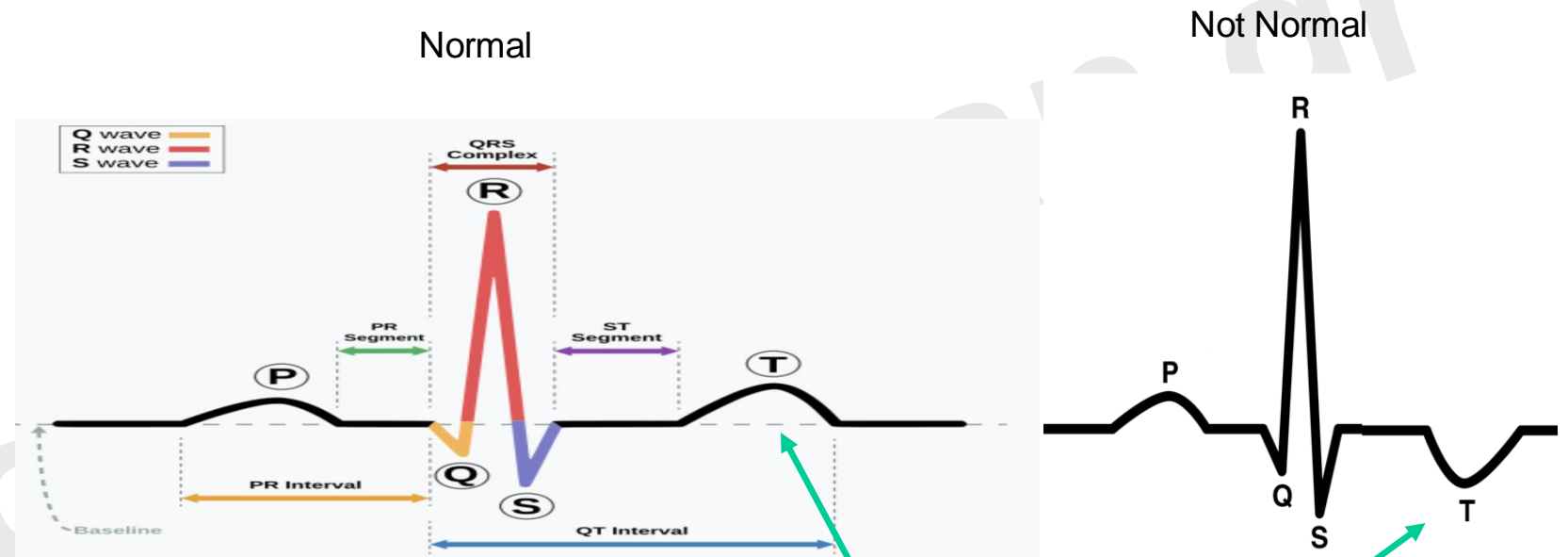
Heart failure:

Ischemia, heart attack and angina pectoris.

Cardiac arrhythmias: –

Tachycardia, bradycardia, ventricular fibrillation, atrial fibrillation, atrial flutter.

Cardiac arrest
arrest



Ischemia happens when we have a narrowing of a large coronary artery. It causes a decrease in blood flow.

Can be found by changes in T wave. T is negative and symmetrical in the leads that are normally positive.