

# The ECG Challenge

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# Topics covered in this presentation

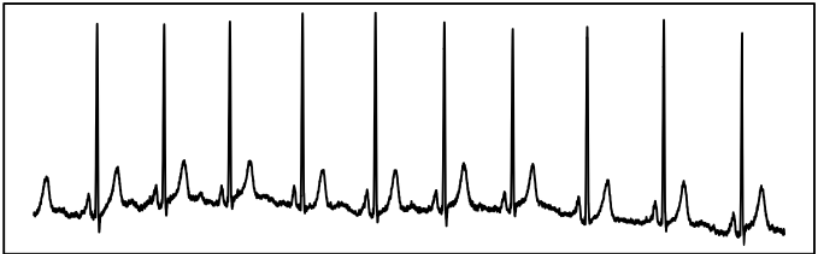
1. Introduction
2. Overview of the Problem
3. Challenges Faced So Far
4. Manipulating the Data
5. Predictions So Far
6. Conclusions and Further Work

# Introduction

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# Background Information

- ECG stands for an **electrocardiogram**.
- An ECG looks at a person's heart's **rate**, **rhythm** and **electrical activity**.
- When would you get an ECG?
  - If you have symptoms of a **heart attack** or **coronary heart disease**.
  - If you have a **pre-existing** heart condition.
  - While taking certain **medication**.
- **Willem Einthoven** is credited with inventing the ECG



**Figure 1:** An example of a graph from an ECG reading.

# Overview of the Problem

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# Aim For This Challenge

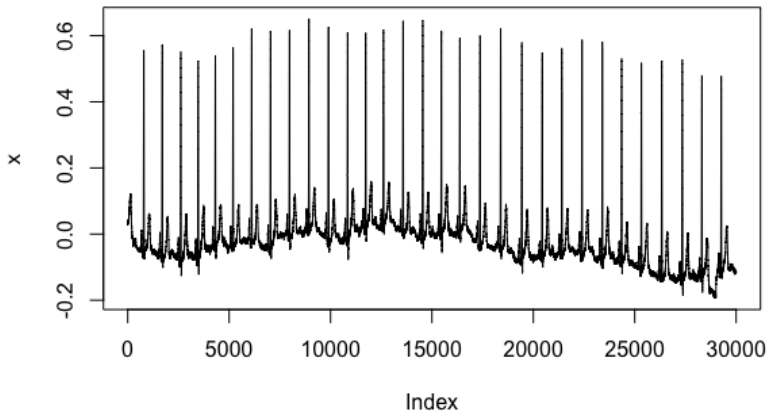
## Our Aim

- Create a model to diagnose two heart conditions:
  1. **Myocardial infarction** (MI), commonly known as a heart attack.
  2. **Cardiomyopathy**, the general term for diseases of the heart muscle.

## Data Given To Us

- Training data:
  - ECGs for **115 patients**.
  - **Heart condition** (Healthy, MI, Cardiomyopathy).
- Test data:
  - Just the ECGs for **100 patients**.
- Our model should classify the training data correctly.

## Example for a patient



**Figure 2:** Plot of patient 1 from training data.

## Challenges Faced So Far

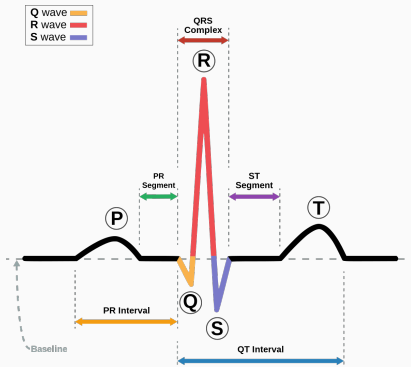
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## Extracting features

What indicates issues with the heart?

- **T-wave inversion**, the t-wave goes down instead of up.
- **QRS interval** is lengthened.
- **ST elevation**, the heartbeat doesn't return to baseline quickly after QRS complex.



**Figure 3:** Diagram of a single heartbeat

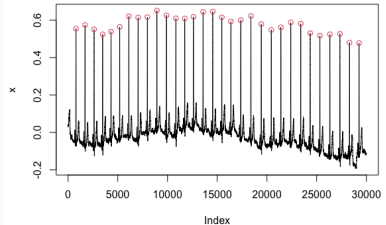
# Manipulating the Data

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# Splitting Into Individual Heartbeats

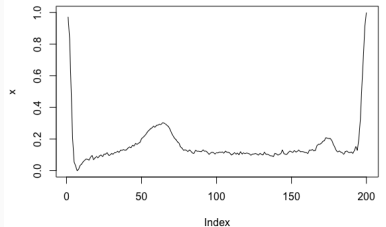
## Determining maximums

- Set a **threshold** that the point must be above.
- Look for a point **higher** than the points either side of it.



## Splitting at the maximums

- Split vector at the **maximums**
- **Interpolate** each heartbeat to be of length 200
- Scale y-axis to be in **[0, 1]**



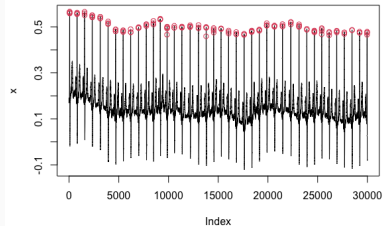
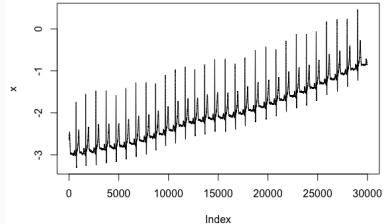
# Issues When Splitting ECGs

## Drift in ECG machine

- Model drift with `lm` function in `r`.
- Use coefficients  $c = y\text{-intercept}$  and  $m = \text{gradient}$  to remove drift.
- $x \leftarrow x - m * i - c$

## Multiple indices for each peak

- Create function to check the if index `points are close` together.
- Tolerance used was `100`.



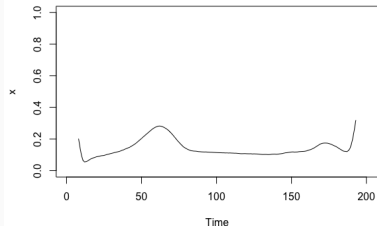
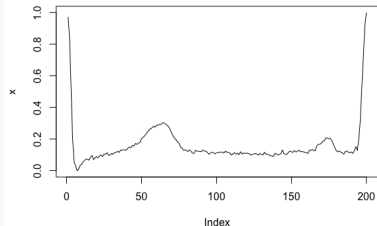
# Smoothing the Functions

## Method for smoothing function

- Use a **moving averages** method.
- Each new point is the average of the **15 data closest points** to it.

## Issues with this method

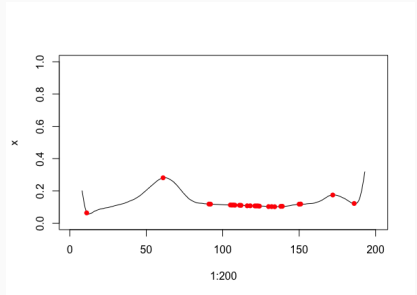
- First and last 7 points don't get values.
- Trade off when selecting window size.
  - Higher window size gives a **smoother curve**.
  - Higher window size can lead to **loss of features** in the graph.



# Why Has All This Been Useful?

## Using the derivative

- Now we have smoothed the function we can **take the derivative**.
- Plotted are the points where the **derivative is equal to zero**
- These minima and maxima can be used to **extract features**.



## T-wave inversion

Look at the **second derivative** at the second turning point.

## QRS interval

Look at the **length of the QRS interval**.

$$\text{QRS Interval} = \text{position of first TP} + (200 - \text{position of last TP})$$

## Predictions So Far

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## Baseline predictions

- Baseline prediction from looking at the original data without manipulation.
- Look at the variance of the data for each patient.
- Gave us a prediction accuracy of 65% on the test set.

## Classification Trees

- As with many applications in medicine this problem lends itself well to decision trees.
- This allow factors that are definite indicators of a certain condition to be accounted for.
- Random Forests can be used to improve accuracy of predictions.
- This gave us a prediction accuracy of 67%.



## Conclusions and Further Work

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## Conclusions

- The key part of this challenge has been **extracting features** from the data.
- Decision trees and thus **Random Forests** seem to be a good model to use for predicting heart conditions.

## Work still to do

- Continue looking for features to extract using **exploratory plots** and ways to extract them **automatically**.
- Explore other machine learning methods, potentially multinomial logistic regression?
- Use cross validation to compare these methods.

# Thank You For Listening

**Any Questions?**