

FIT1043 Introduction to Data Science

## Module 5

# Data Analysis Process

# Case Studies

Lecture 10 – Part II

# Background: Different Data Types

- Spatial data (also known as geospatial data):
  - is information about a physical object that can be represented by numerical values in a geographic coordinate system.
- Temporal data:
  - is data that varies over time.
- Spatio-Temporal data:
  - is the integration of space and time.

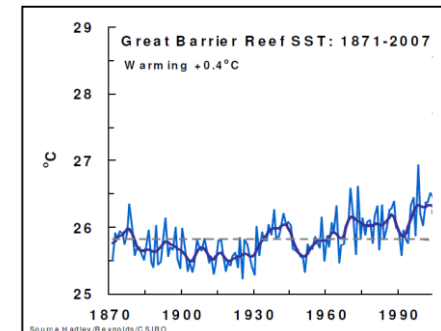
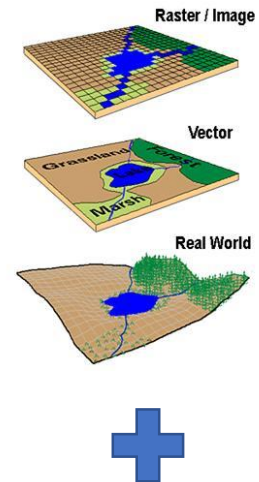


Image planet.botany.uwc.ac.za

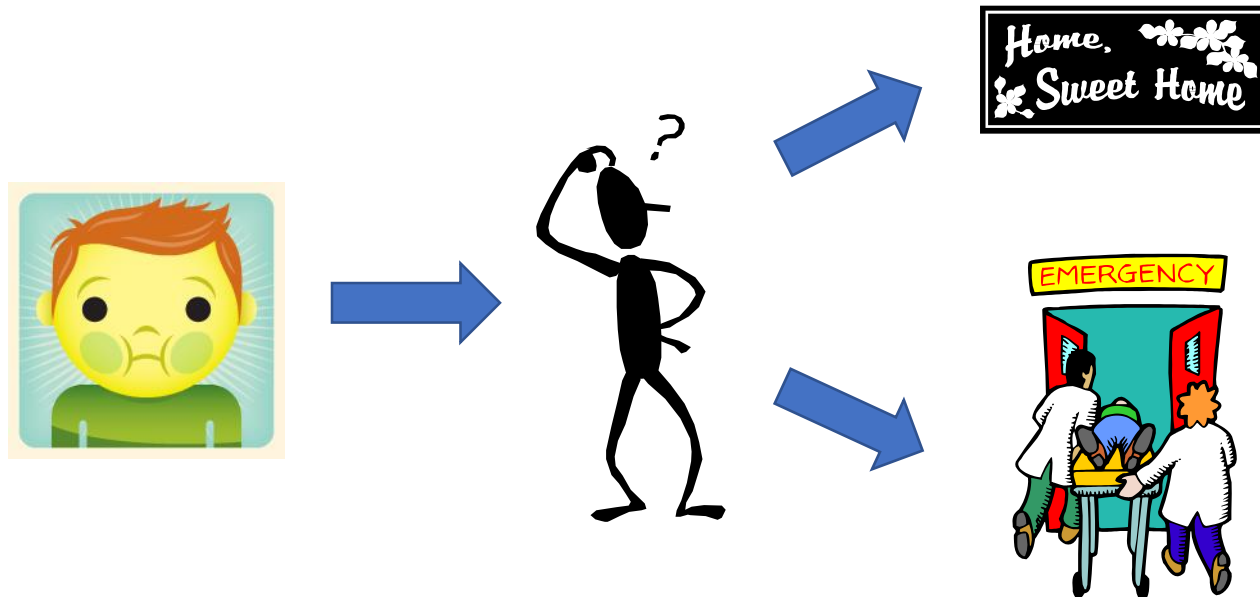
# Data Analysis Case Studies

## Diagnosis of Pneumonia



# Diagnosing pneumonia is difficult

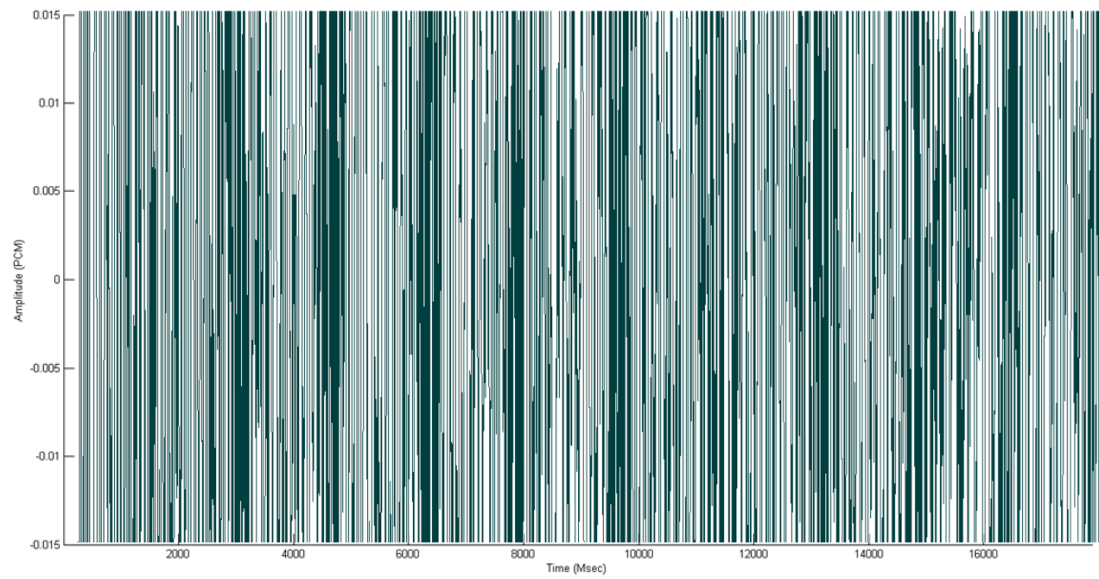
- Pneumonia is the single biggest killer of children.
- It kills 1.5 million children under 5 per year, mostly in developing countries.
- The key to change this lies in early diagnosis.

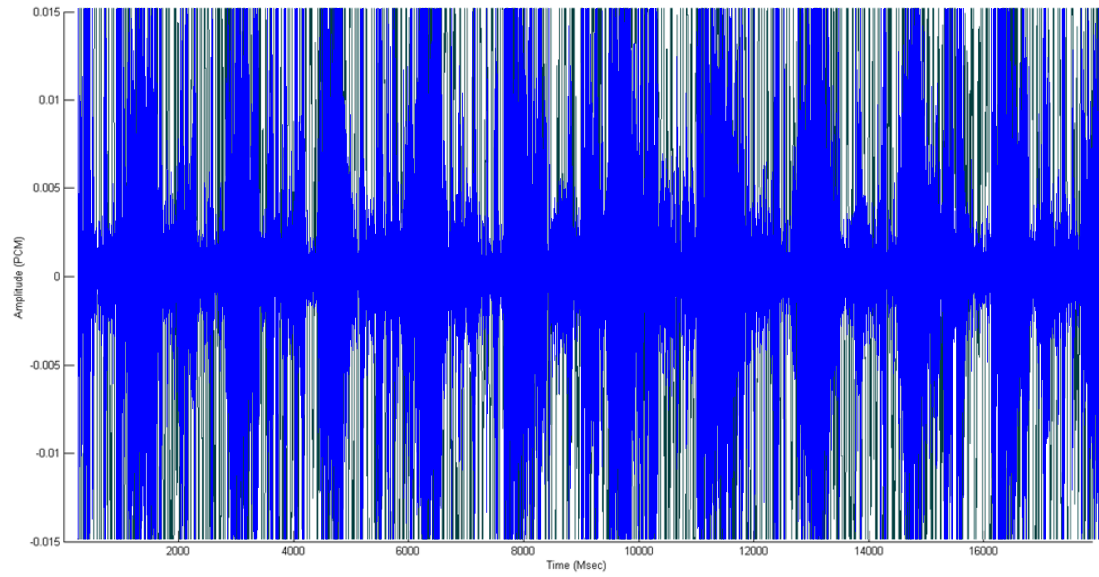


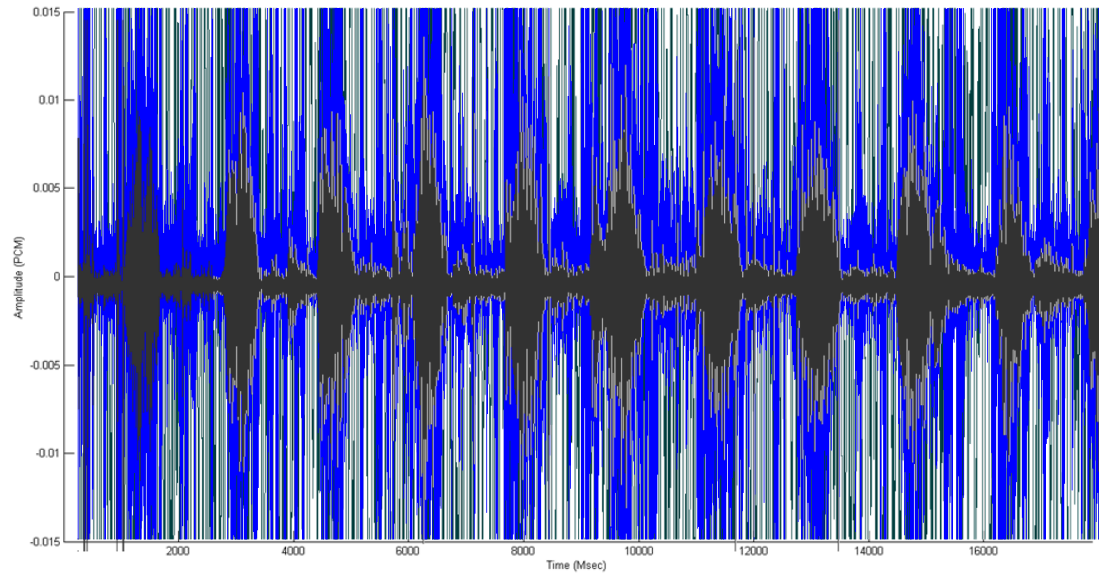
# IoT based solution: StethoCloud



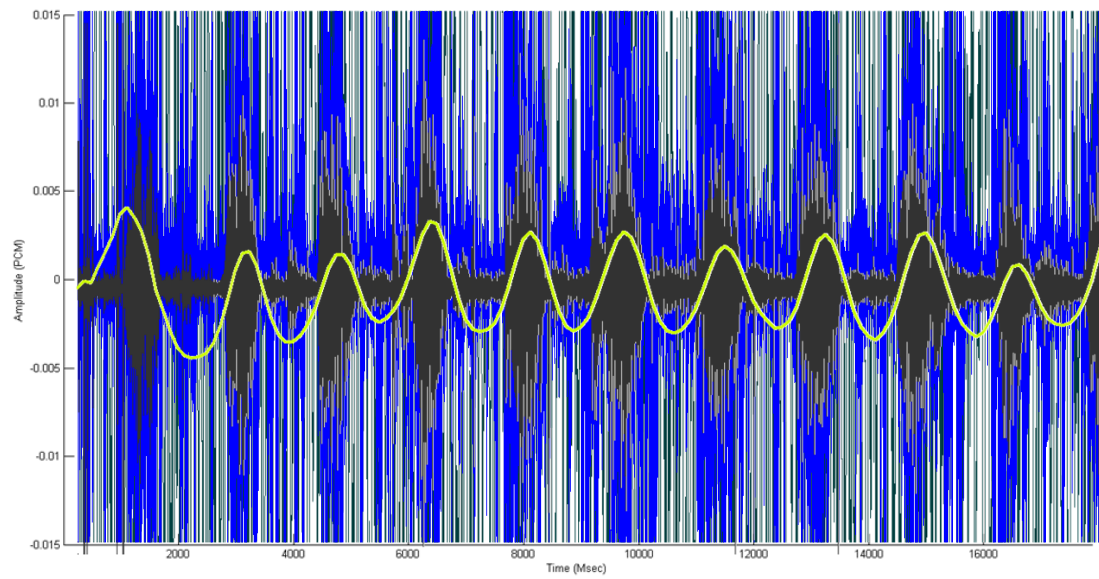
Respiratory Data is collected from six different locations

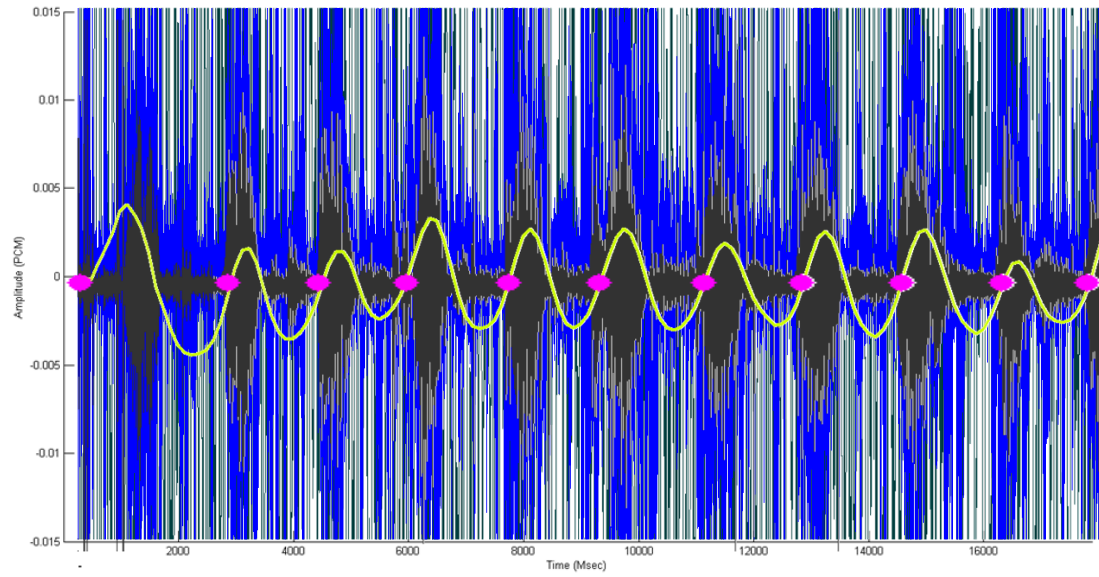








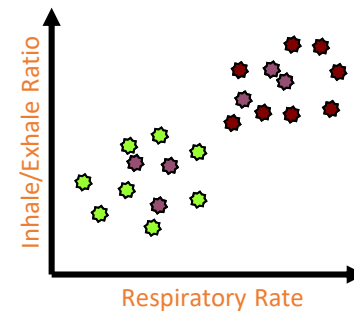




# Classification and Clustering

- Classification

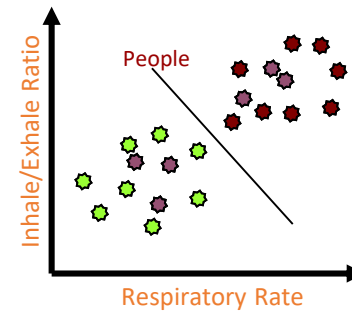
- Have labels for some points



# Classification and Clustering

- Classification

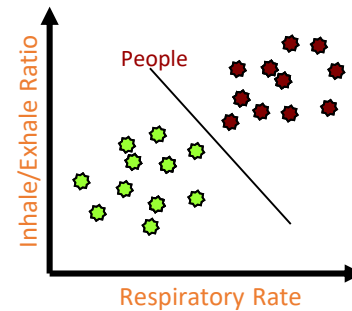
- Have labels for some points
- Want a “rule” that will accurately assign labels to new points



# Classification and Clustering

- Classification

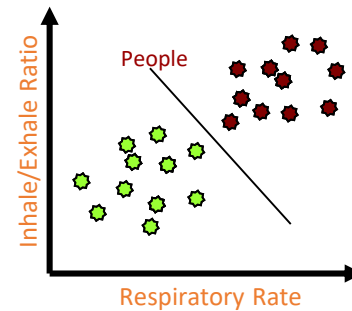
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# Classification and Clustering

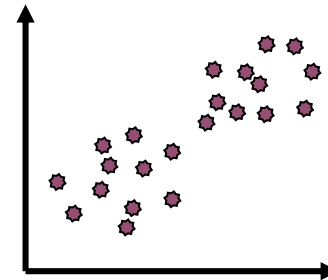
- **Classification**

- Have labels for some points
- Want a “rule” that will accurately assign labels to new points



- **Clustering**

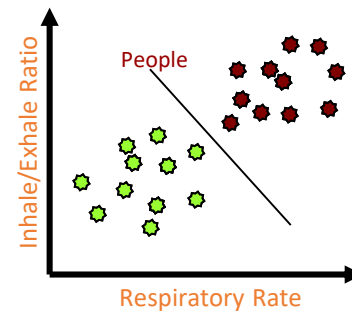
- No labels



# Classification and Clustering

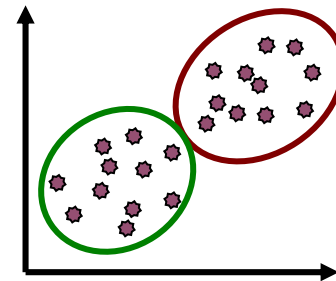
- **Classification**

- Have labels for some points
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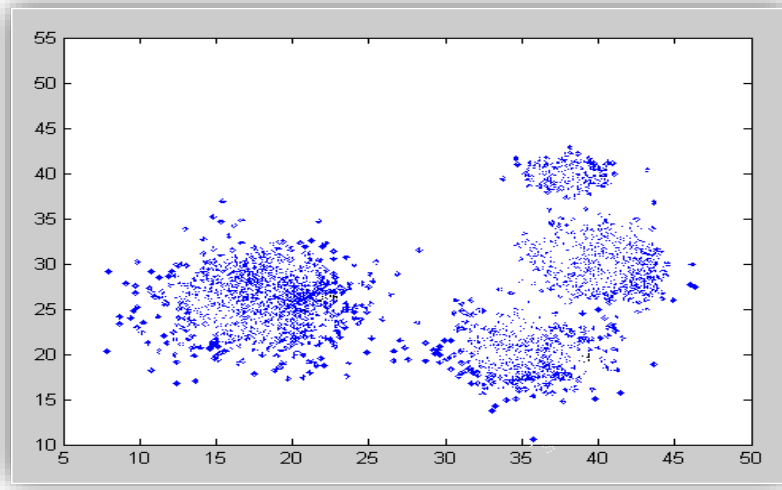


- **Clustering**

- No labels
- Group points into clusters based on how “near” they are to one another
- Identify structure in data

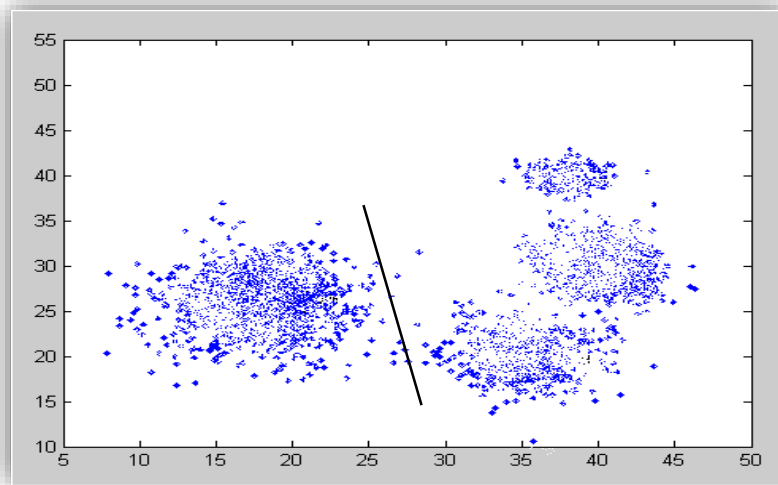


# This is what the data might look like





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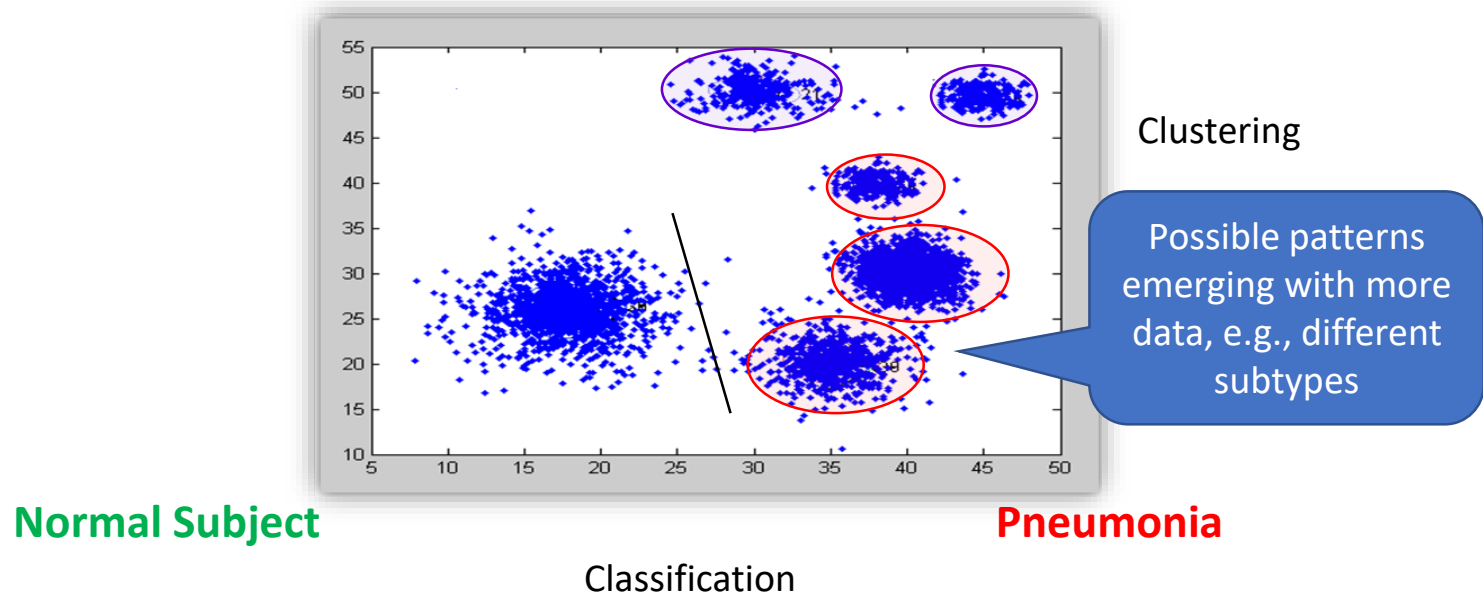


**Normal Subject**

**Pneumonia**

Classification

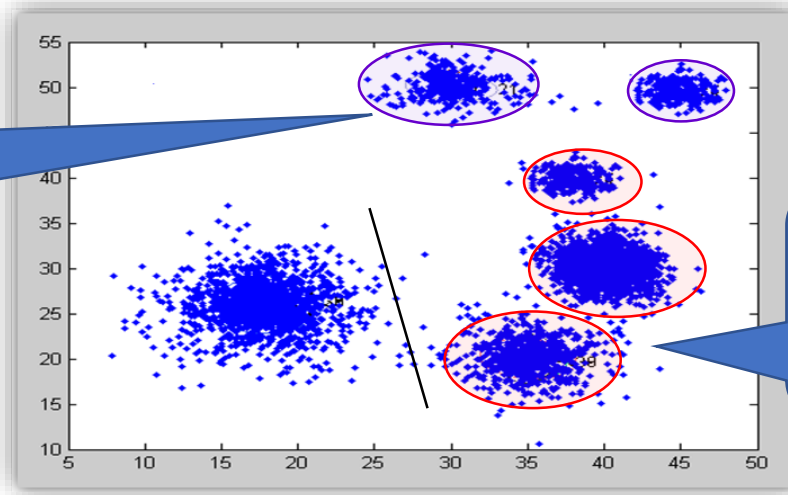
# This is what the data might look like



# This is what the data might look like

## Other Diseases

This technique can directly be used for other diseases, e.g. asthma



Clustering

Possible patterns emerging with more data, e.g., different subtypes

Normal Subject

Pneumonia

Classification

# Data Analysis Case Studies

## Car-Racing Driver

### Distraction Detection



Image emotiv.com

# Why is it important?

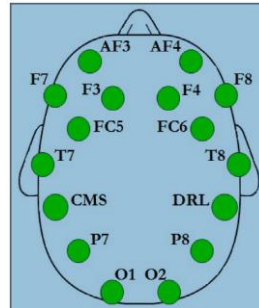
- Monitoring driver attention has a direct effect on decreasing injury/fatality rates, and improving his performance.
- In **car racing** environments reaction times are short, and distraction leads to a reduction in driver's performance during a race.



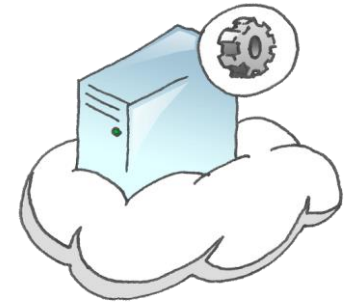
# Brain Computer Interface (BCI) Approach



Emotiv EPOC headset



14 Brain EEG Signals  
+ Gyroscope



# Experimental Setup

- Simulation



- Talking to passenger
- Mobile call, Recording call
- Solving simple mental arithmetic challenges placed on the road, assessed by the driving through the chosen answer from a range of alternatives on screen

# EEG data analysis

- Classification
  - Random forest has best results
  - 73.5% detection accuracy with EEG
  - 81% detection accuracy with EEG + Gyroscope

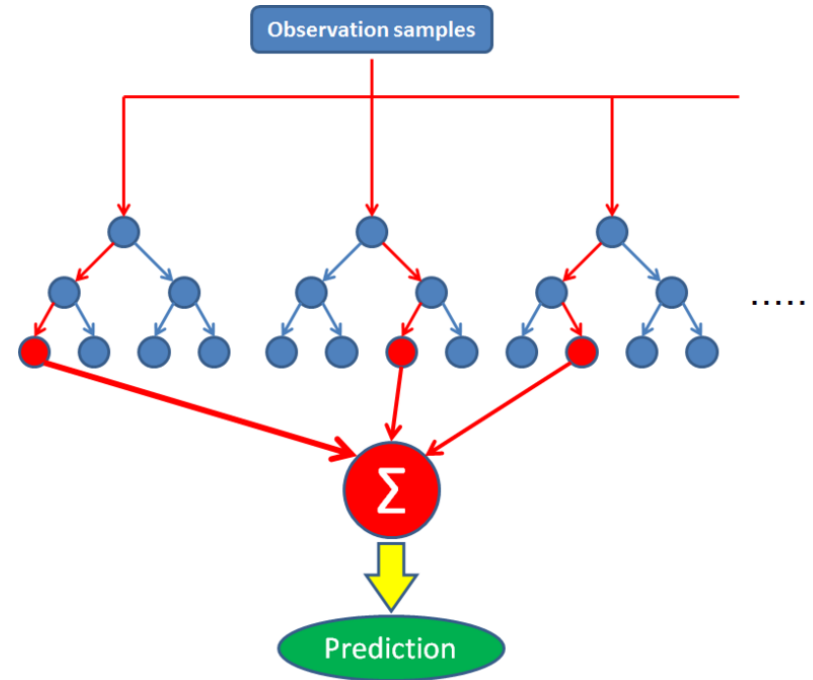


Image cdn-ak.f.st-hatena.com



# Data Analysis Case Studies

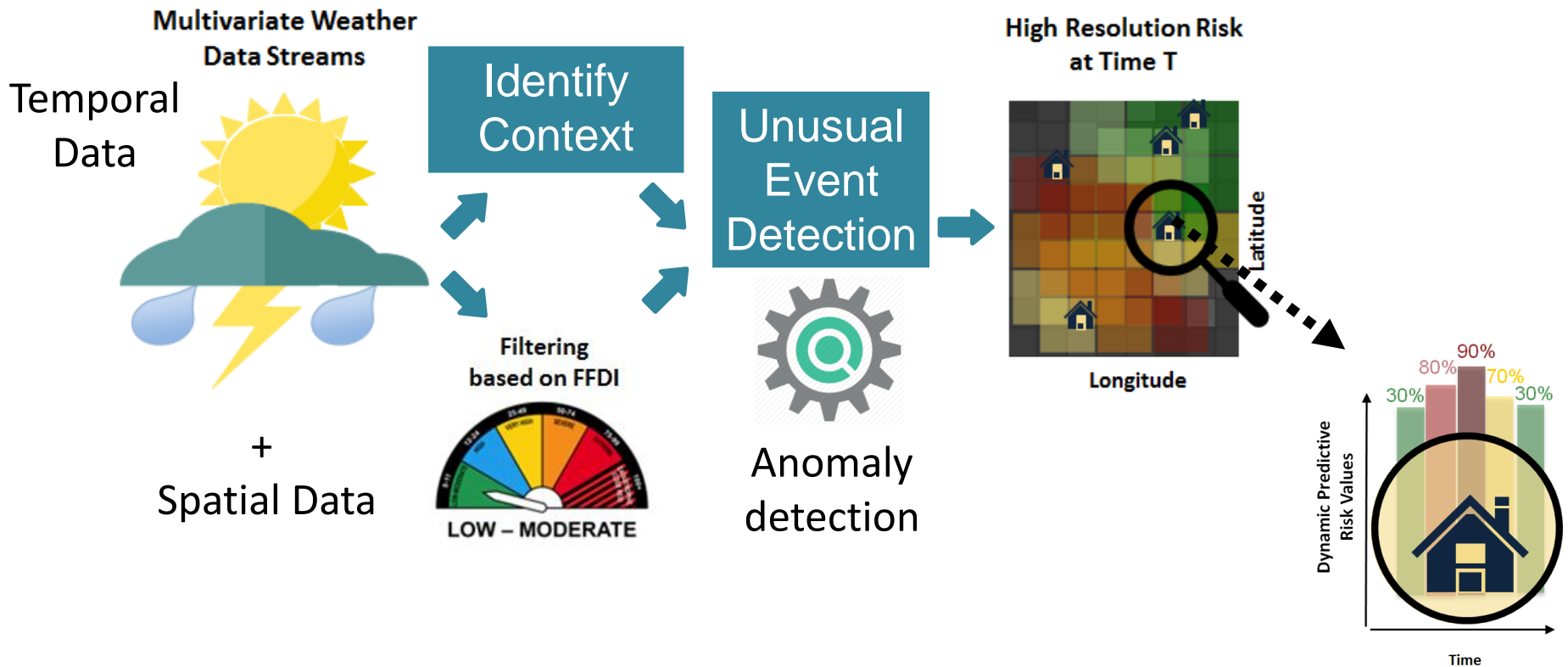
## Disaster Management



# Why bushfires?

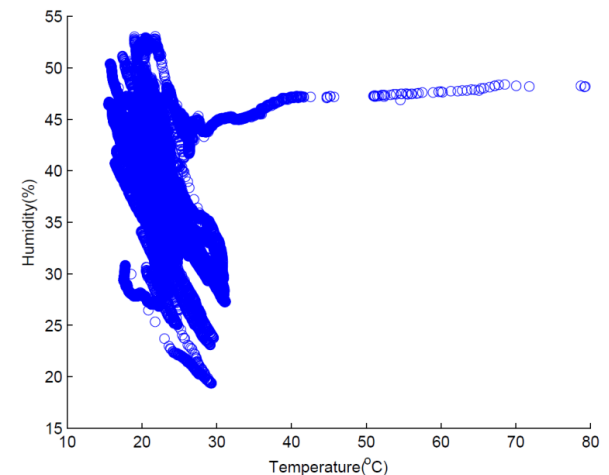
- Bushfires have shaped the Australian landscape for many years. With urban expansion and changing weather patterns the threat to homes and families is increasing.
- Australia experiences bushfires as the most damaging disasters.
- Examples:
  - Ash Wednesday bushfires 1983
  - Black Saturday 2009
- Ability to **predict** the risk of bushfires is crucial in helping emergency services in their decision-making processes, to mitigate and reduce the impact of such events.

# Dynamic bushfire risk prediction



# Anomaly Detection

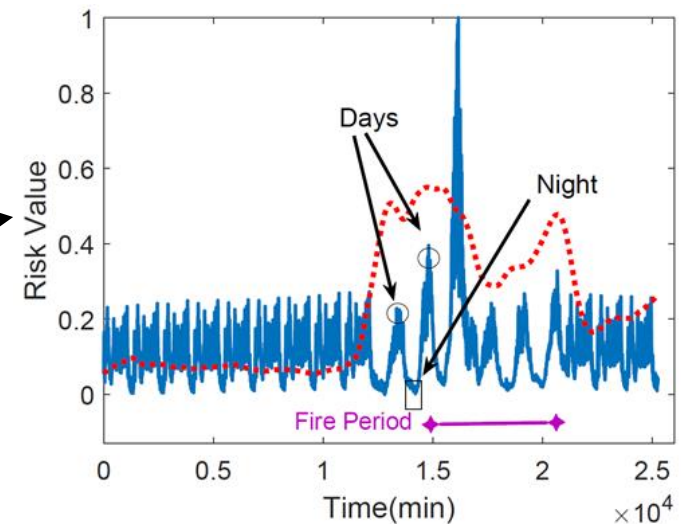
- Anomaly: data points that are inconsistent to the normal data points
- Anomaly detection: the process of finding anomalous patterns in data sets
- In this application, anomalies are relevant to the episodes of time with **high bushfire risk**



# Case study- 2013 Blue Mountains Fire, NSW

- Why Blue Mountains?
  - 11 days of intense bushfires
  - 2 fatalities
  - 248 houses destroyed
  - 183.4M AUD in insurance claims
- Data Available
  - ~ 45 thousands houses
  - historical weather measurement

# Case study- 2013 Blue Mountains Fire, NSW

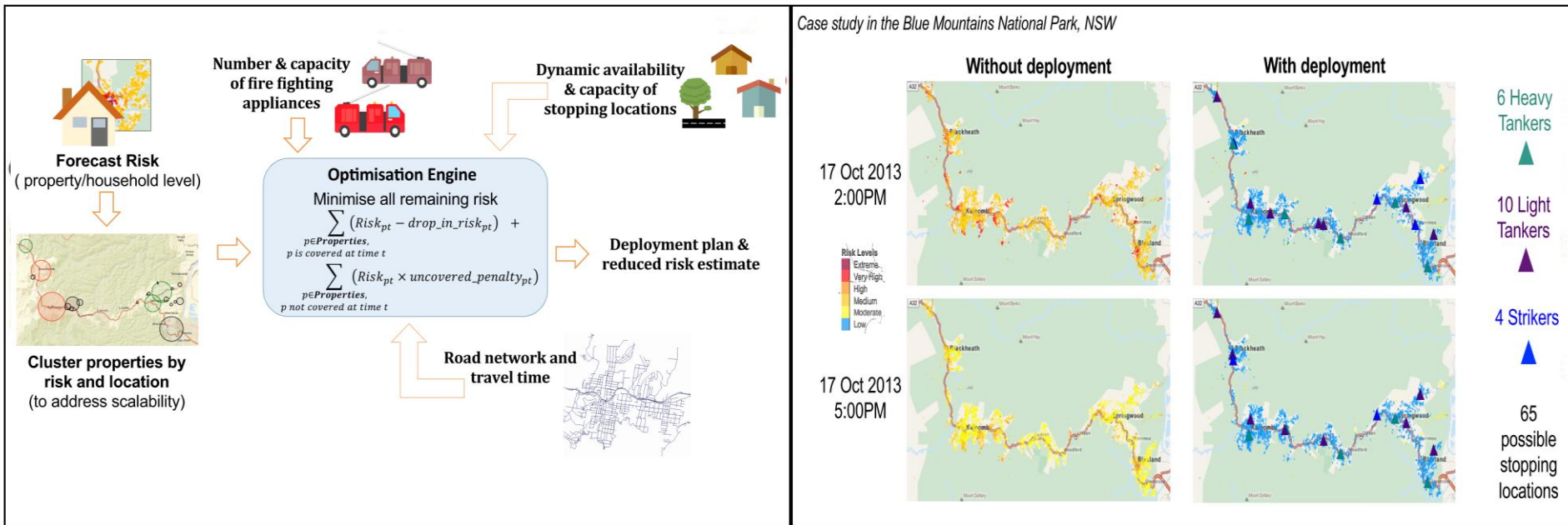


..... Our Method Risk Values

— FFDI Risk Values

# Fire Fighting Appliance Pre-Deployment

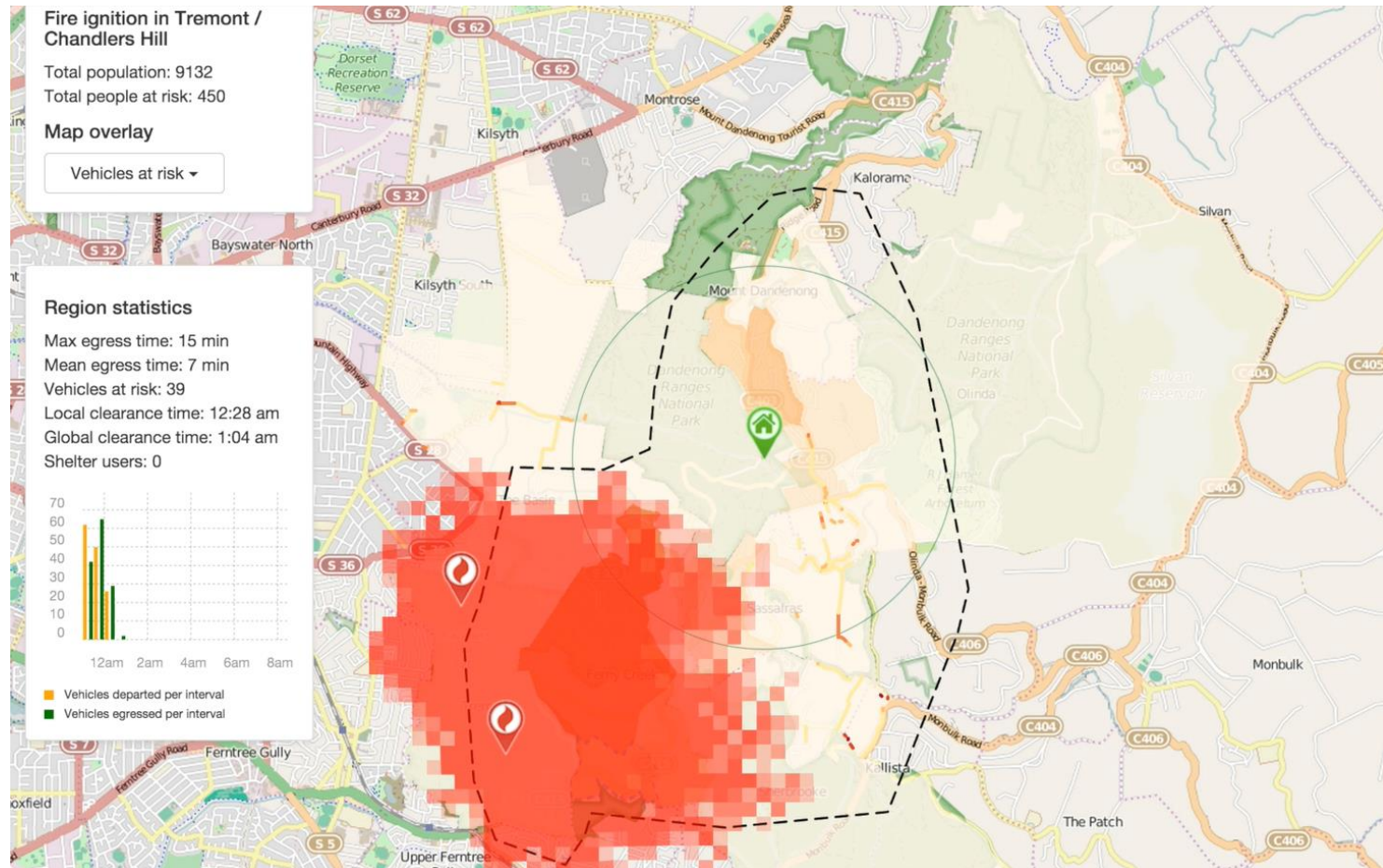
- Optimise locations of available fire fighting resources subject to **forecast risk** and roadside **stopping locations** with **time dependent availability**.





# Evacuation Planner

- Spatio-Temporal fire expansion





# Unit Schedule: Next Week

Module	Week	Content
1	1	Overview and look at projects (Job) roles, and the impact
	2	
2	3	Data business models / application areas
3	4	Characterising data and "big" data Data sources and case studies
	5	
4	6	Resources and standards Resources case studies
	7	
5	8	Data analysis theory Regression and decision trees Data analysis process
	9	
	10	
6	11	<b>Issues in data management</b> GUEST SPEAKER & EXAM INFO.
	12	