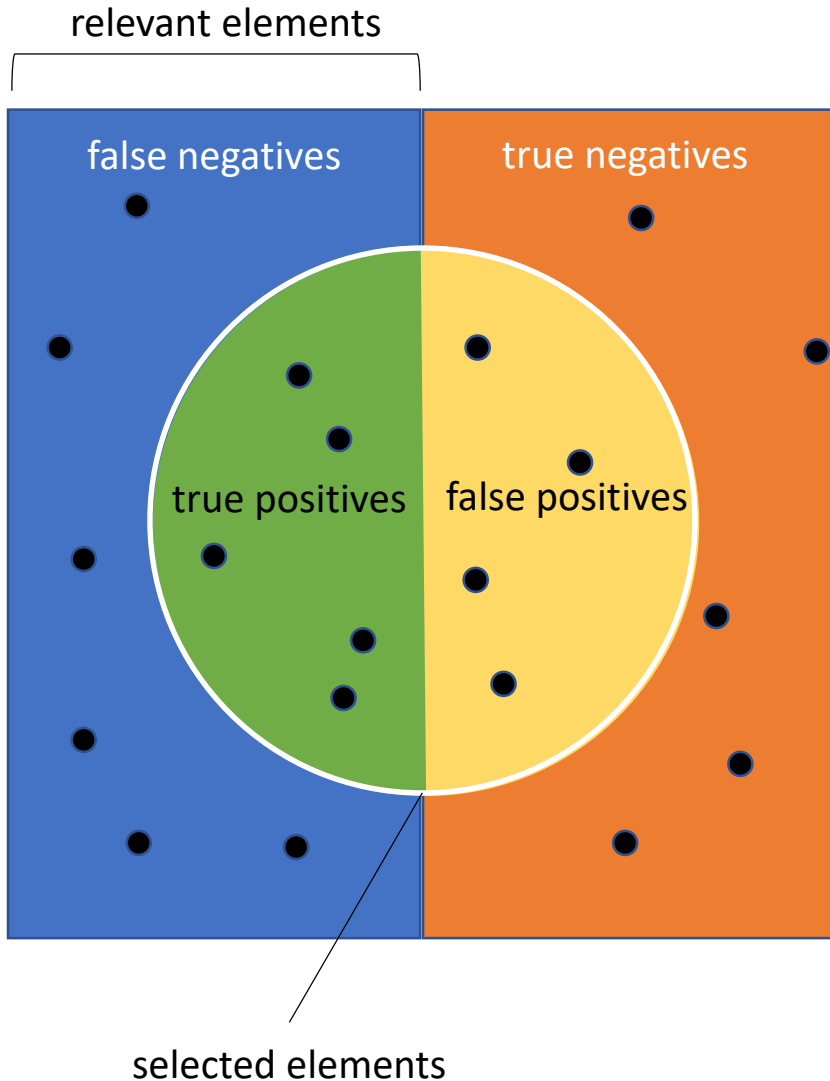


# Analytics

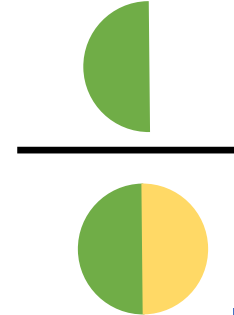
# How it should work



# Accuracy = Precision AND Recall



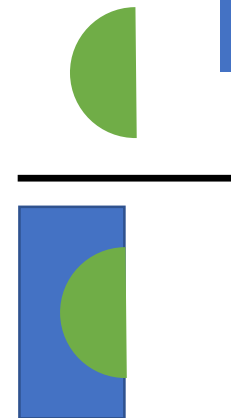
Precision =



How many selected elements are relevant?

$$\text{Accuracy} = \frac{T_p + T_n}{T_p + T_n + F_p + F_n}$$

Recall =



How many relevant elements are selected?

# Siloed analytics

- Should be business improving not just call centre focused (helps drive ROI)
- Old fashioned technological approach – open, not ring fenced
  - Contrast SaaS platforms with full range of APIs and existing integration
- Useful for non content benchmarking and analysis (eg. AHT/Outcome agent)
- Fundamental Attribution Error
  - Csats generally high with agent and lower with process/outcome

# The Seven Key Steps of Data Analysis

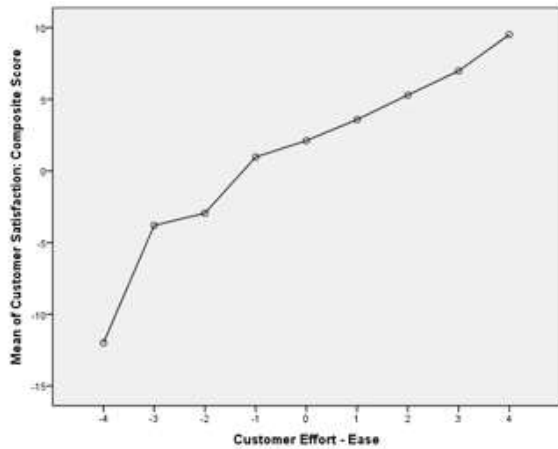
- Decide on the objectives
- Identify business levers/frameworks
- Data collection
- Data cleaning
- Data modelling
- Grow a data team
- Optimise and repeat

# Frameworks – understanding customer experience

## Rational Brain:

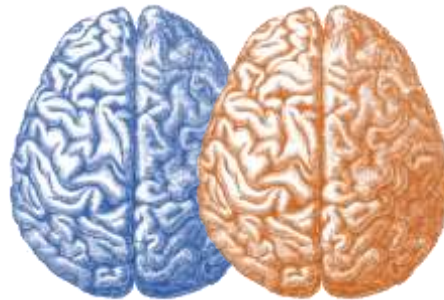
Task Resolution – did I get my job done?

Effort – and how hard was it to get it done?



## System 1

Automatic  
Fast  
Intuitive  
Emotional

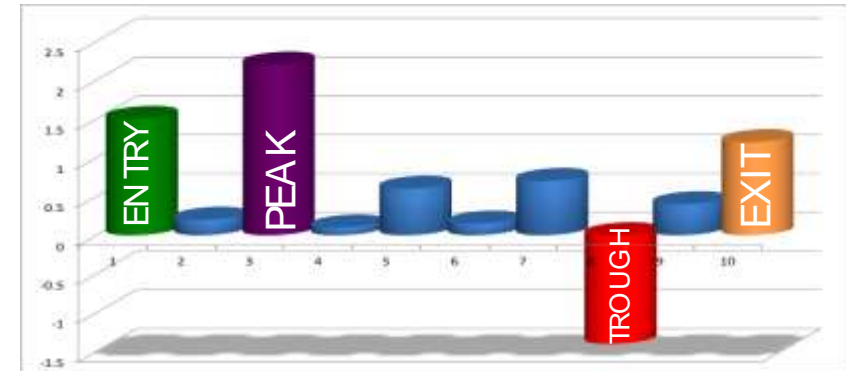
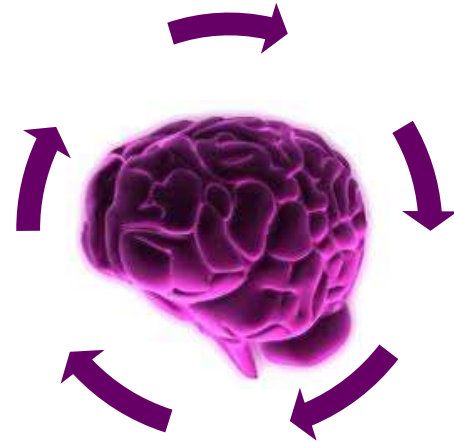


## Irrational Brain:

Judgemental heuristics

## System 2

Effortful  
Deliberate  
Logical  
Rational



# Rol

- Sell more, sell to more, cost less (increased Csat etc are 'nice to have')
- Sell more/to more/churn less are harder to justify in my experience
- Demonstrable cost reduction trumps the rest
- Demand deflection/reduction
- AHT reduction
- Support cost reduction (quality/compliance/attrition/TTC)
- 50 seat call centre could easily find €0.9m p.a. savings for an outlay of 40-45% of that or a 6 month payback
- SaaS models and RESTful APIs driving implementation times in days/weeks rather than months with reduced professional service fees

20% increase Csat  
5% demand reduction  
25% efficiency increase in AHT  
41% of contact shown to be avoidable

# Contact categorisation – northern European public transport

- Used SaaS analytic platform to apply machine learning to the customer's existing data
- Nearly 50% of the data was being categorised imprecisely.
- This analysis was used to generate three bespoke machine learning models:
  - Fresh topics – to refresh the existing issue tree
  - Sentiment – to capture customer mood at certain points
  - Intent – actionable emotional clues
- Running the same data through the new categorisation model provided startling results. All the 'bucket' categories (50% by volume initially) had disappeared and the tagging was much more relevant and accurate. Of the original 136 categories, 30% were made redundant as they were too broad and 18% retired completely as they were out of date or used infrequently. The 80 replacement categories were all relevant and consistently applied.



# Impact of adding sentiment to accurate categorisation

- Reliable early warning system
- Opportunity for revising customer journey
- Highlighted agent disparities
- Redesign staff coaching plans
- Enables real time next best action for known contact/situation problem areas

# Interesting examples of analytics

- Impact of on brand conversations on sales outcomes
  - Impact of Agent skills – which is the most important?
- 
- Most\* CXOs have never listened to customer calls....

\*unqualified assessment!