

# Making Big Discoveries Big Data, Al, Machine Learning & Behavioural Science

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### **Presentation Outline**

Al and Behavioural Science

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- Electric vehicles and 'affect heuristic'
- Energy efficiency and 'motivations & barriers'

Online purchase and 'behaviour costs'

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- Patient experience and 'peak-end'
- Hotel experience and 'peak-end'

Five tips to maximise

AI-BeSci

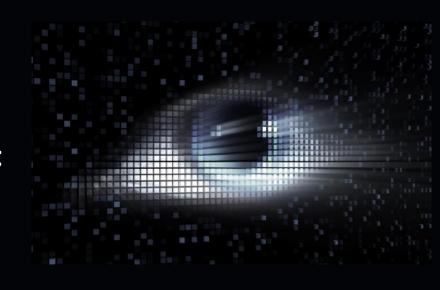
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### Al and machine learning

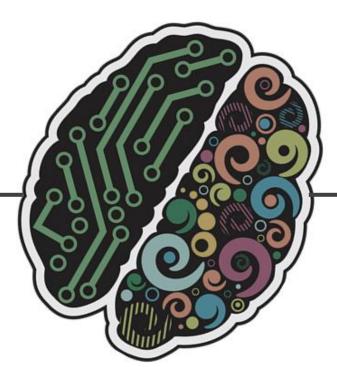
- Supervised machine learning still main form of AI
- MLAs ability to 'see' and detect patterns in data
- But machines remain socialcognitive dummies





#### MLA eyes + BeSci brains

MACHINE LEARNING EYES



BEHAVIOURAL SCIENCE BRAINS

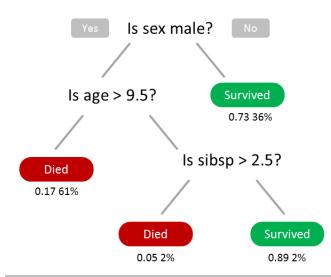
#### Classification and regression trees

 CRT with supervised learning used by tax evasion, national security risk, financial risk, voting, online searching behaviour, buying behaviour, healthcare outcomes, etc.

MLAs often used to 'score', identify and predict

Patterns also provide behavioural insights

#### **Survival of Titanic passengers**



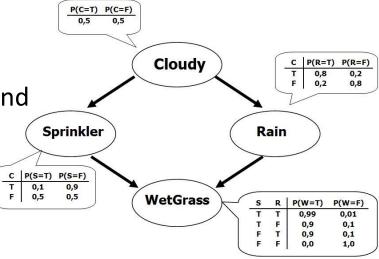
A tree showing survival of passengers on the Titanic ("sibsp" is the number of spouses or siblings aboard). The figures under the leaves show the probability of survival and the percentage of observations in the leaf

### **Bayesian (belief) networks**

• BNs used to identify complex structures, often used when 'deep learning is required', but becoming more widely used

 BNs are able to identify causal structures and recognises inter-dependent nature of relationships

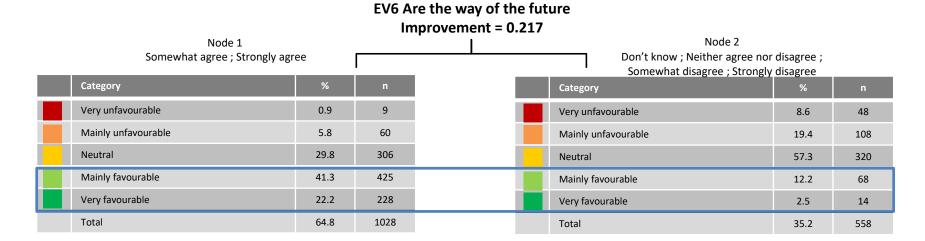
 Good at illustrating cognitive structures of markets, behaviours, etc...



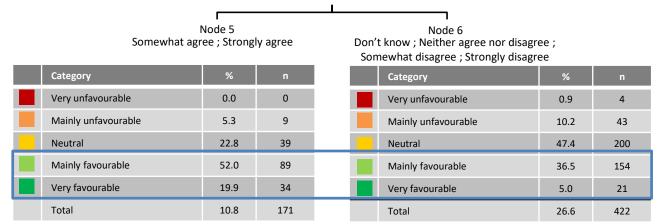


#### EV4 How favourable or unfavourable is your overall opinion or impression of EVs

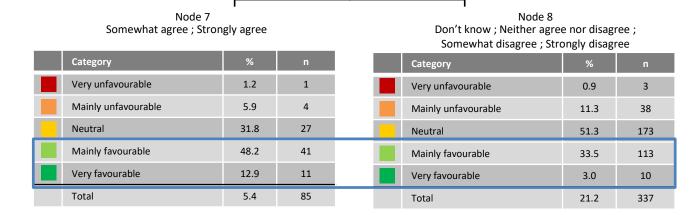
Node 0					
	Category	%	n		
	Very unfavourable	3.6	57		
	Mainly unfavourable	10.6	168		
	Neutral	39.5	626		
	Mainly favourable	31.1	493		
	Very favourable	15.3	242		
	Total	100.0	1586		



#### EV6 Are reliable cars Improvement = 0.028



#### EV6 Have a driving distance range that is suitable for my needs Improvement = 0.011



#### Electric vehicles and the 'affect heuristic'



- 1. Feeling of 'driving the future' will encourage EV adoption
- Pocusing on the 'short-cut' means people are more willing to engage
- Identify and understand heuristics to trigger desired behaviour change



#### Number of energy efficient actions done Providing for myself or my family Having a job that I enjoy Being happy Building a life with my partner or spouse Using energy wisely Making a comfortable home **Building financial security** Keeping myself fit & healthy Improving myself through education/learning Spending time with my family Making & keeping good friends Minimising environmental impacts Contributing to my community

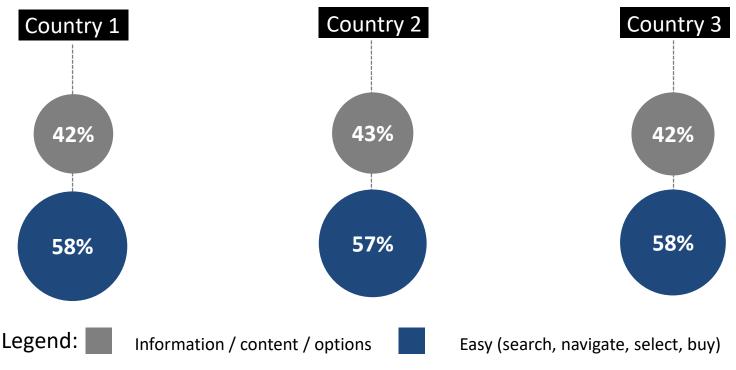
## Home energy efficiency: motivation and barriers

1. Encouraging behaviour is more than motivation

- 2. Understand people in context to encourage change
- Interventions need to focus on more than just motivations, also understand how to reduce barriers



# Impact of customer experience on overall impression (SEA)

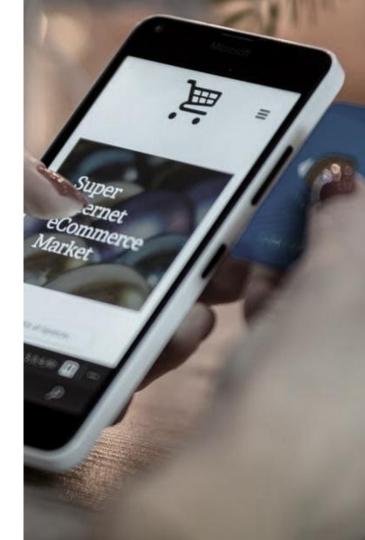


# Online purchase and 'behavioural costs'

Misallocation of marketing resources?

Underestimating 'behavioural costs'

Less leverage in motivation than costs





## Impact of patient experience on patient memories

Experience	Overall impression index	
Politeness and courtesy of admission staff: very good	143	and
Care and treatment from nurses: very good	122	and
Care and treatment from Drs: very good	133	and
Discharge process: very good	143	

### Impact of patient experience on patient memories

Experience	Overall impression index	
Care and treatment from nurses: very good	122	and
Care and treatment from Drs: very good	133	and
Discharge process: good to very poor	114	and
Politeness and courtesy of admission staff: good to very poor	89	

### Patient experience and 'peakend' effect

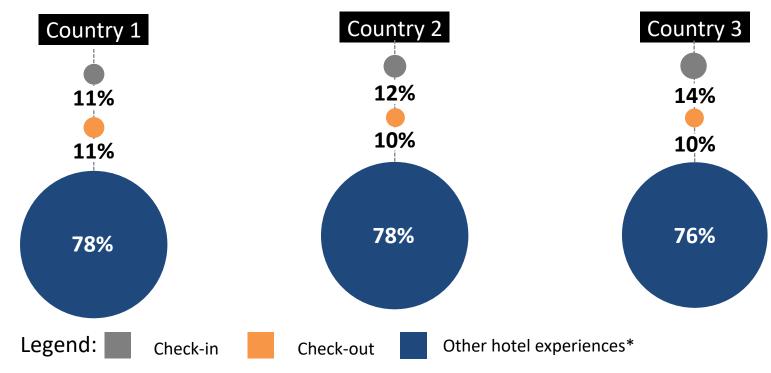
- 1. No observed 'peak' effect
- But strong impact of 'end' effect at patient discharge
- Unexpected strong effect at patient admission
- BeSci principles + context
- **5.** Re-design discharge process



### Hotel customers and peak-end effect



# Impact of customer experience on overall impression (SEA)



<sup>\*</sup> room cleanliness, hotel amenities, dining, etc.

### Hotel experience and 'peak-end' effect

1. No 'end' effect in check-out experience

- 2. Hotel policy addresses check-out negatives
- 3. Opportunity to build positive experience

### The END



# Five tips to maximise AI/BeSci benefits

- 1. Go beyond driver rank
- 2. Let machines find patterns
- 3. Know cognitive mechanisms
- 4. Think person + context
- 5. Accelerate cooperation

