

# AI technologies | Maximising benefits, minimising potential harm

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## **In this talk...**

- The relationship between AI and Data Science
- CAIPP as an interdisciplinary centre
- Mapping the domain of the social, ethical and legal effects of AI
- Cases and strategies for maximising benefit and minimising harm

## **AI, Data and Data Science**

- There are not simple agreed-upon definitions of either data science or AI.
- AI is changing data.

## **Data was...**

- given for a purpose
- static
- able to be corrected or deleted

# Data now...

- Data is given but it is also extracted
- Data is inferred
- I know less about what data others hold about me, what it's for, how it was constructed...

## **I have less control as a data subject**

- Tyranny of the minority
- My data is 'exchanged' for essential services by effective monopolies
- It's hard to ask a company to correct or delete data if I don't know it exists or I don't understand what it means
- Data is a form of wealth that is very unevenly distributed

## **So for the individual**

- Data has become much more dynamic, much more empowering, very efficiently harvested
- And I have less knowledge about it and less control over it than people used to

# AI is changing business and government

- It is providing insights, new types of products and services.
- It is allowing us to assess intentions, risks... more accurately and on the fly.
- It is allowing us to target resources in ways we couldn't before.

## **But...**

- The information ecology can be as uncertain for governments and businesses as it is for individuals.
- inaccuracy, bias, lack of transparency are problems for organisations just as for individuals, but organisations have different levels of motivation to solve those problems.

## **IA is democratising data for both individuals and organisations**

- I don't have to be a statistician to use statistics for very complex tasks
- But at the same time I might not know very much about how or how well those tools are making those decisions.





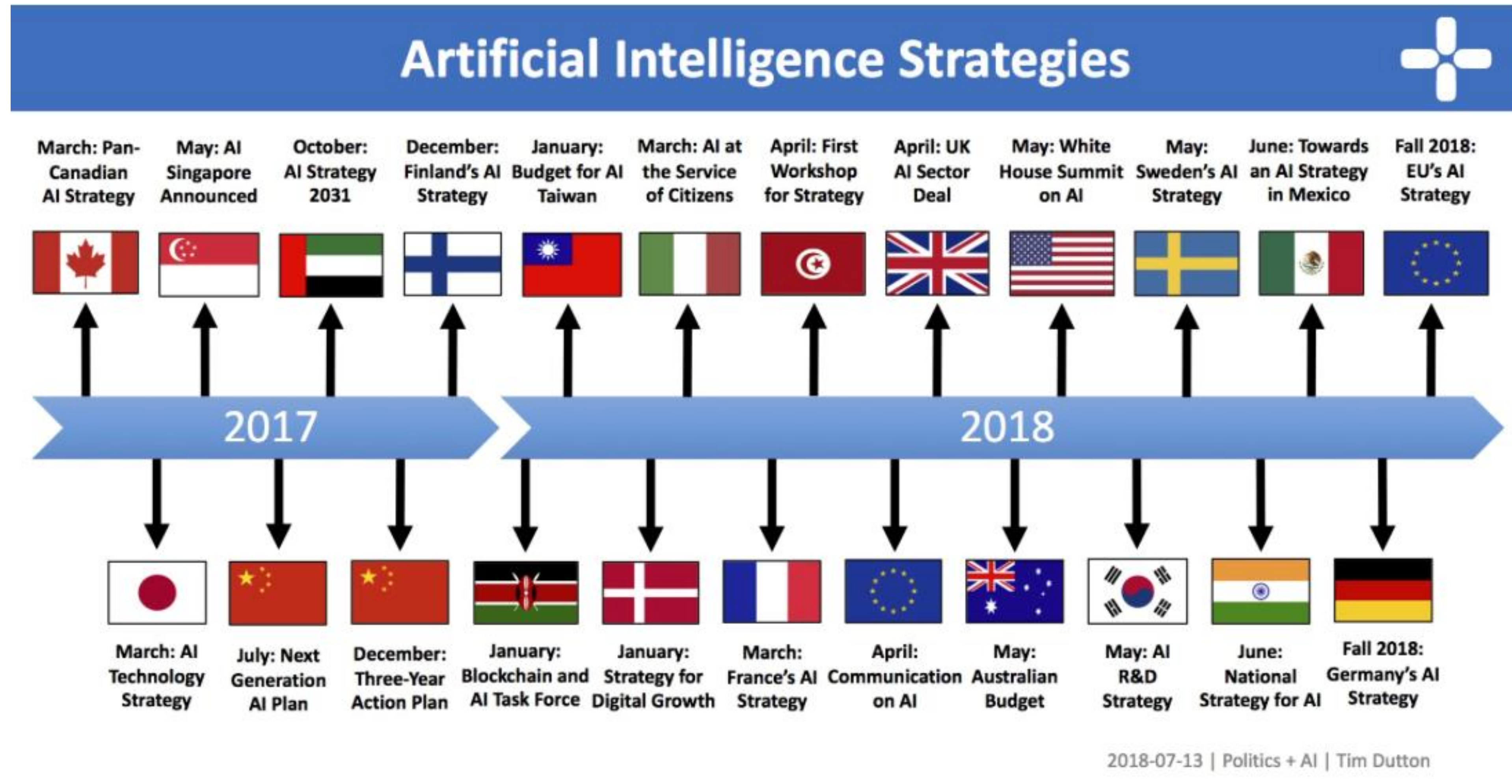
CENTRE FOR  
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and Public Policy

*Te tari Rorohiko Atamai,  
Kaupapa Here Tūmatanui*

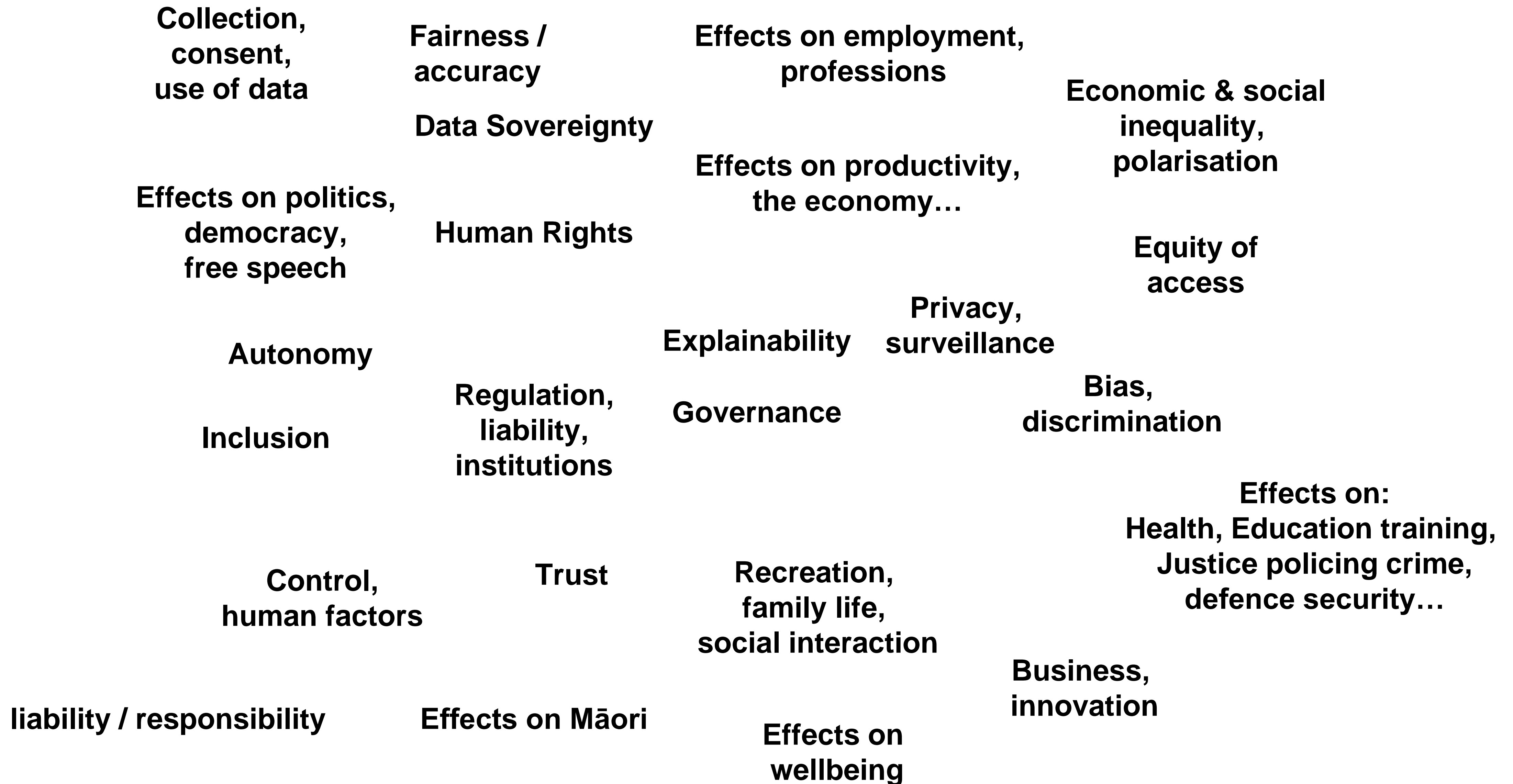
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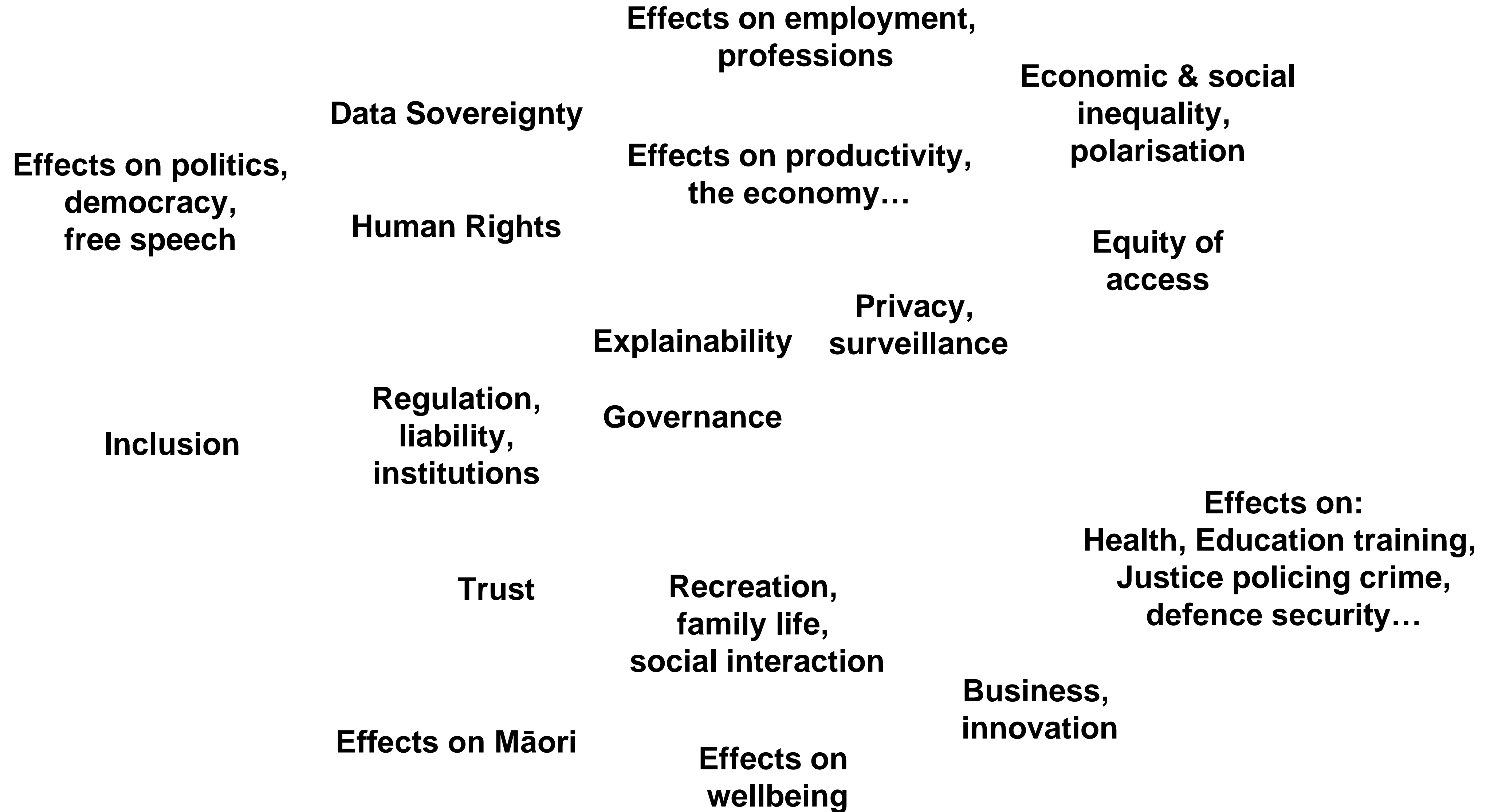
# The domain of social, ethical, legal research into AI



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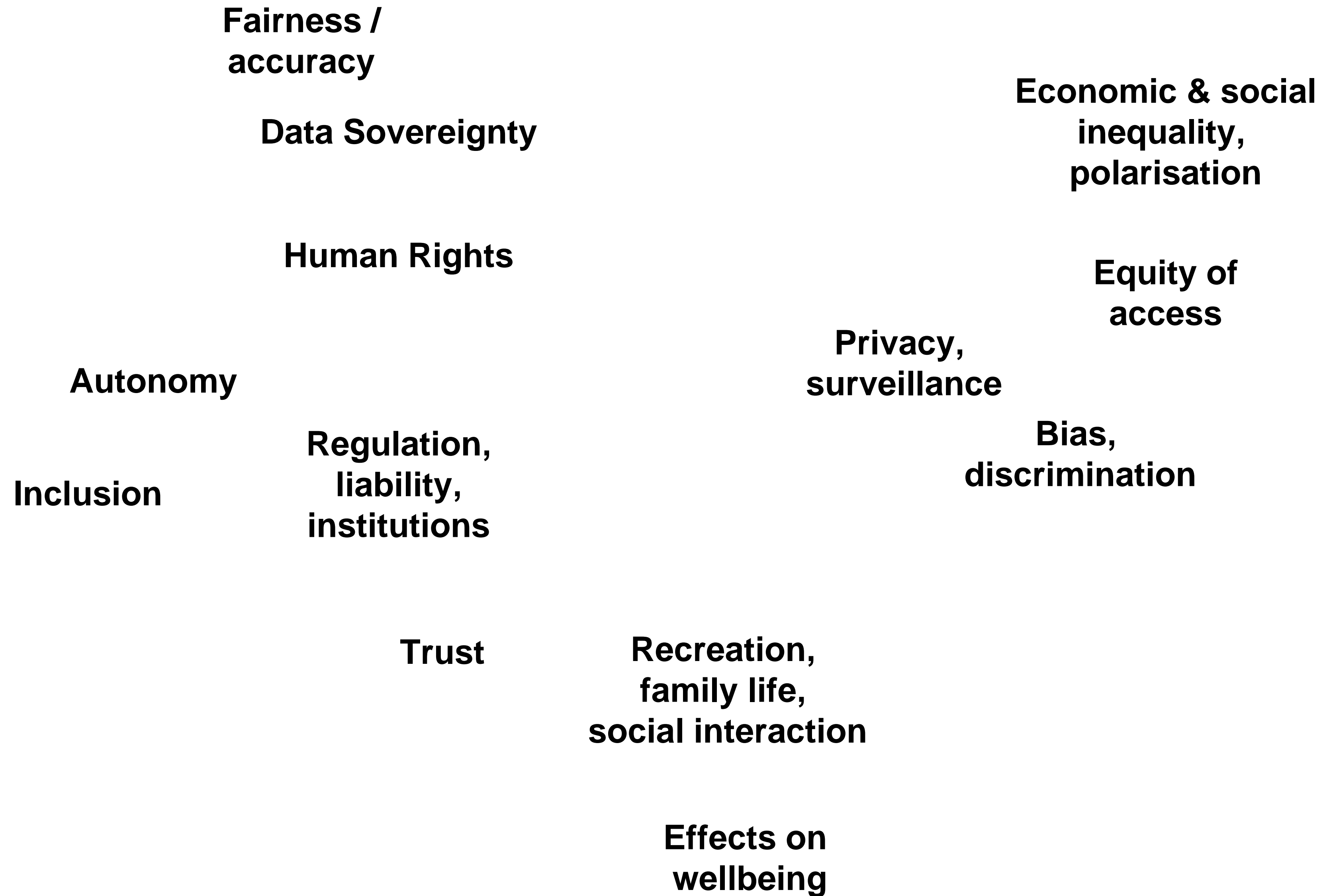


# Effects on





# How AI affects individuals



# Data-centric research

**Collection,  
consent,  
use of data**

**Fairness /  
accuracy**

**Effects on employment,  
professions**

**Data Sovereignty**

**Effects on productivity,  
the economy...**

**Human Rights**

**Equity of  
access**

**Privacy,  
surveillance**

**Inclusion**

**Regulation,  
liability,  
institutions**

**Governance**

**Bias,  
discrimination**

**Trust**

**Recreation,  
family life,  
social interaction**

**liability / responsibility**

**Business,  
innovation**

# Algorithm-centric research

**Fairness /  
accuracy**

**Explainability**

**Privacy,  
surveillance**

**Regulation,  
liability,  
institutions**

**Governance**

**Bias,  
discrimination**

**Control,  
human factors**

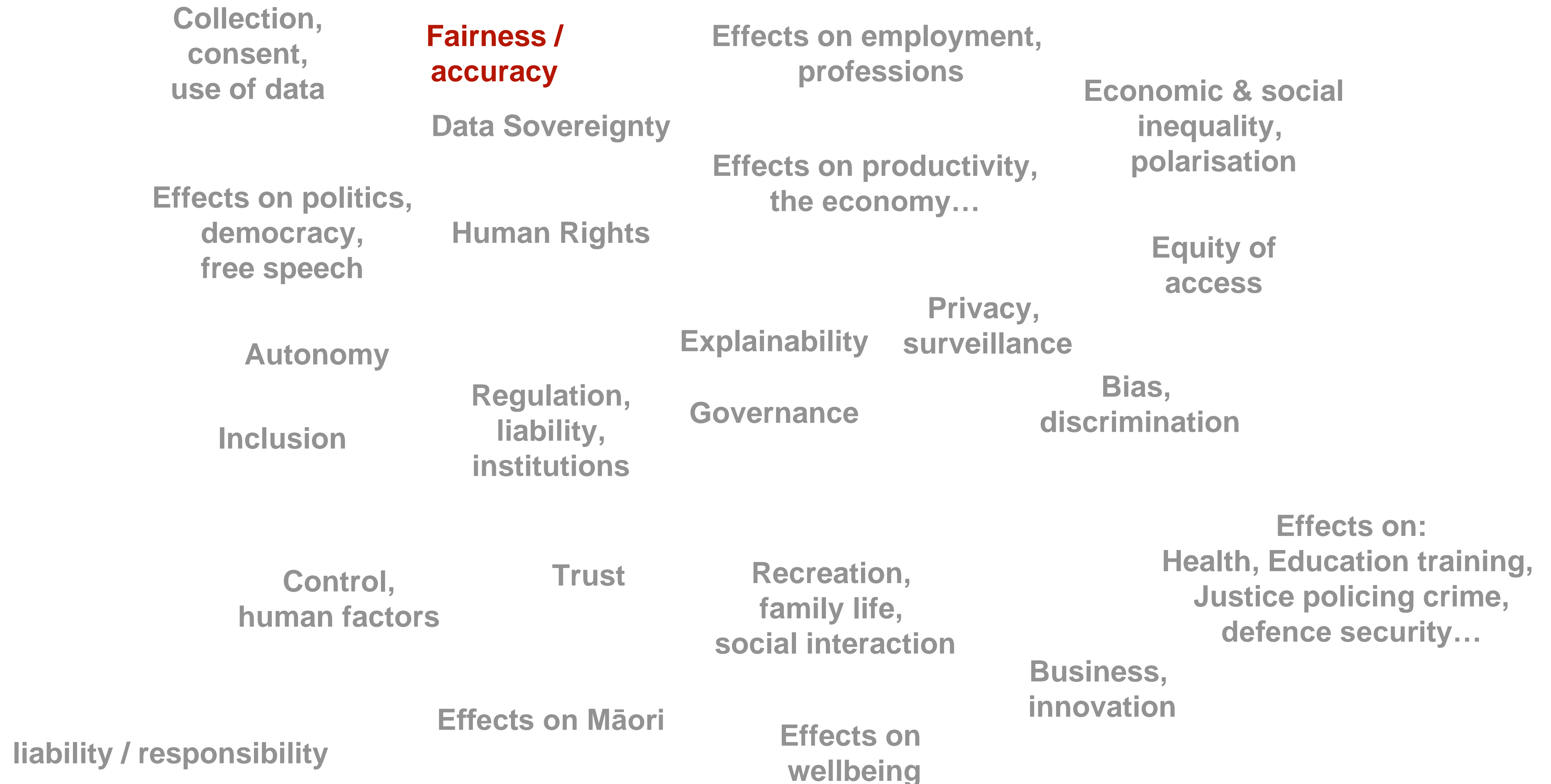
**Trust**

**liability / responsibility**

**Business,  
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# The domain of social, ethical, legal research into AI



# Artificial Intelligence and Law in New Zealand

**Fairness /  
accuracy**

**Effects on employment,  
professions**

**Economic & social  
inequality,  
polarisation**

**Effects on productivity,  
the economy...**

**Explainability**

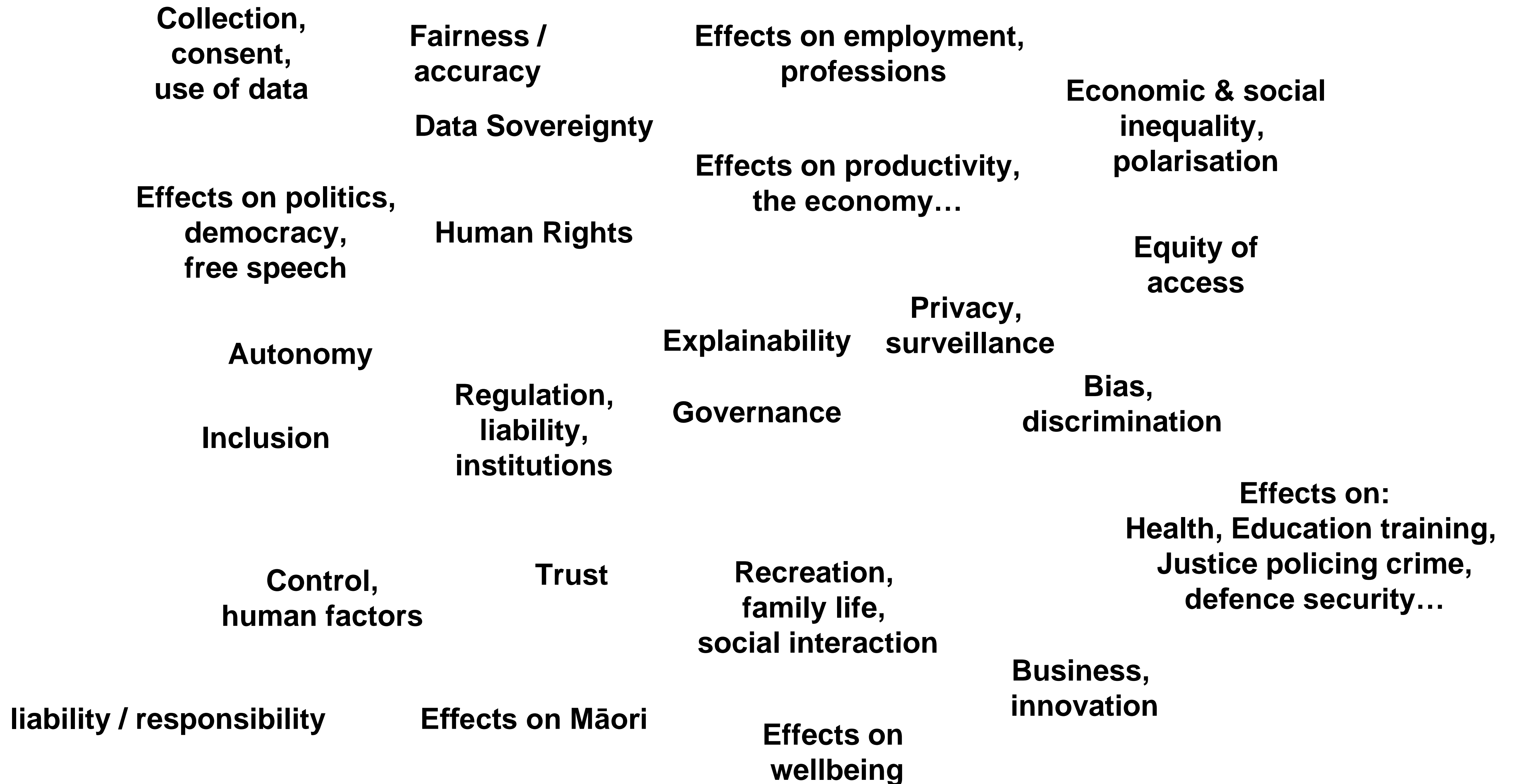
**Regulation,  
liability,  
institutions**

**Bias,  
discrimination**

**Control,  
human factors**

**Justice policing crime,**

# The domain affected by **GDPR**





# The domain affected by GDPR

**Collection,  
consent,  
use of data**

**Human Rights**

**Privacy,**

**Explainability**

**Regulation,**

**Bias,  
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**Trust**

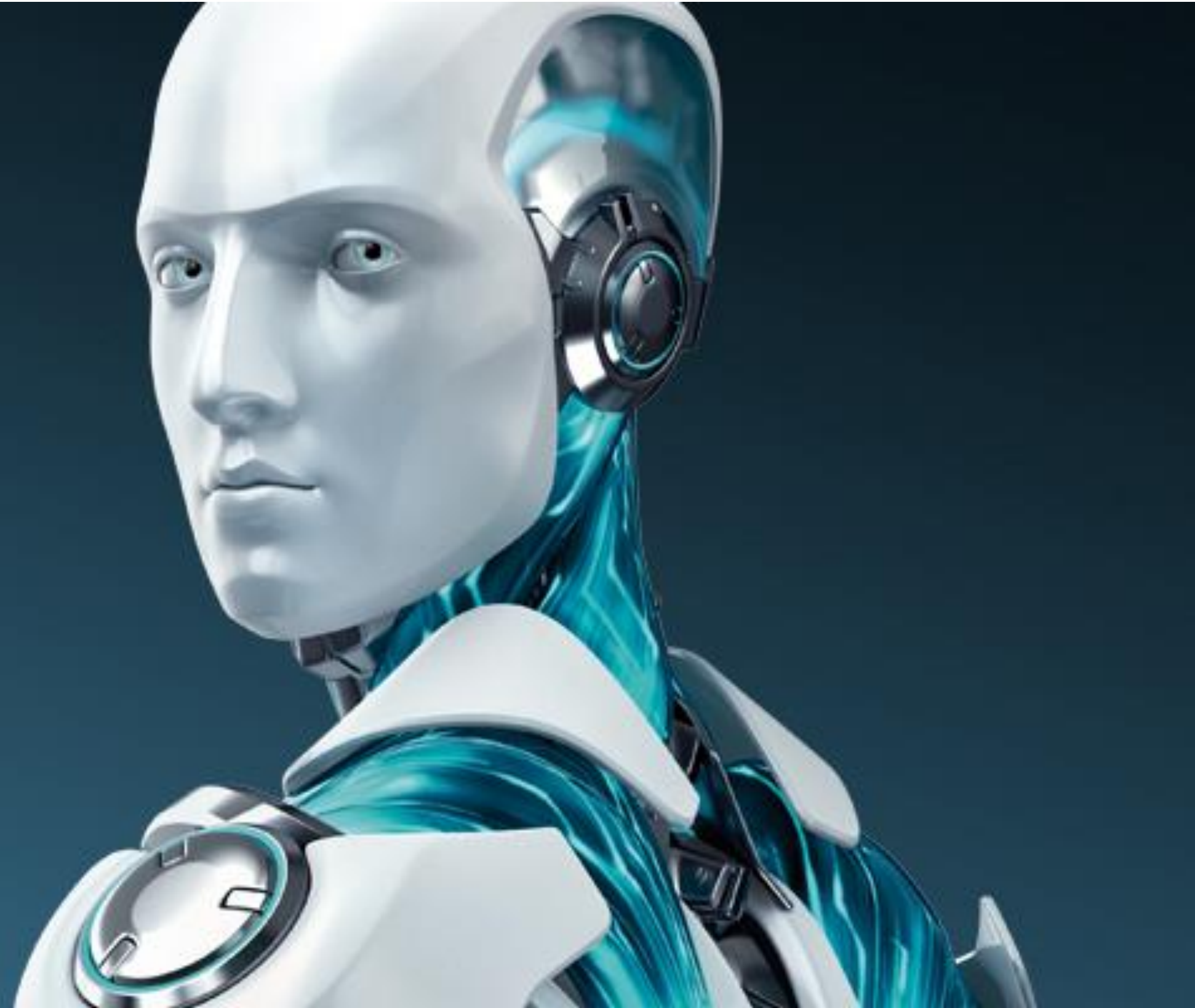
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So we know the question we want to answer—

How do we use data in a way that is fair, for public benefit, and trusted.

# Regulation and AI

Of, by or for  
AI?





# Do we need 'AI law'?

'the policy discussion should start by considering whether the existing regulations already adequately address the risk, or whether they need to be adapted to the addition of AI.' (US National Science and Technology Council)

# Not all problems are (entirely) new problems

DRUNK DRIVING WAS LESS OF A PROBLEM BEFORE CARS. IF YOU GOT ON YOUR HORSE DRUNK AND FELL ASLEEP, IT COULD JUST WALK HOME.



AND IF YOU TRIED TO RIDE INTO A TREE, THE HORSE COULD BE LIKE "NO."

FORGET HUMAN DRIVERS—  
THAT'S THE BENCHMARK WE SHOULD BE JUDGING SELF-DRIVING CARS AGAINST.



# Driverless car makers could face jail if AI causes harm

AI technologies which harm workers could lead to their creators being prosecuted, according to the British government.

16:33, UK,  
Tuesday 25 September 2018

Responding to a written parliamentary question, government spokesperson Baroness Buscombe confirmed that existing health and safety law "applies to artificial intelligence and machine learning software".

"I'm sceptical both that industry's own tests will be deep and comprehensive enough to catch important issues, and that the regulator is expert enough to meaningfully scrutinise them for rigour," said Michael Veale, researcher in responsible public sector machine learning at University College London.

# Right to reasons

## **Official Information Act 1982**

Section 23 (1): where a department or Minister of the Crown makes a decision or recommendation in respect of any person in his or its personal capacity, that person has the right to be given a written statement of... (c) *the reasons for the decision or recommendation.*



# Elements of reasons

- System functionality – ex ante
- Specific decision – ex post
  
- Experts in how the software works
- Experts in the sort of decision being made (criminologists, social scientists, etc)
- Non-experts!

# Explanation not bafflegab

‘The resulting systems can be explained mathematically, however the inputs for such systems are abstracted from the raw data to an extent where the numbers are practically meaningless to any outside observer.’

Dr Janet Bastiman, evidence to UK Parliament Science and Technology  
Ctte (2017)



May 2018



## Principles for the safe and effective use of data and analytics

The use of data and analytics must have clear benefits for New Zealanders. Data and data analytics are tools that support decision-making and it's essential that in collecting and using public data, government agencies consider, and can demonstrate, positive public benefits.

This includes:

- considering the views of all relevant stakeholders
- ensuring all associated policies and decisions have been evaluated for fairness and potential bias and have a solid grounding in law
- embedding a te ao Māori perspective through a Treaty-based partnership approach.

# Accuracy and validation

The *Daubert* test (q.v. *Calder* in NZ)

- Relevant and reliable?
- Scientifically valid and applicable to the facts in issue?
- Known and potential error rate?
- Published and peer-reviewed?

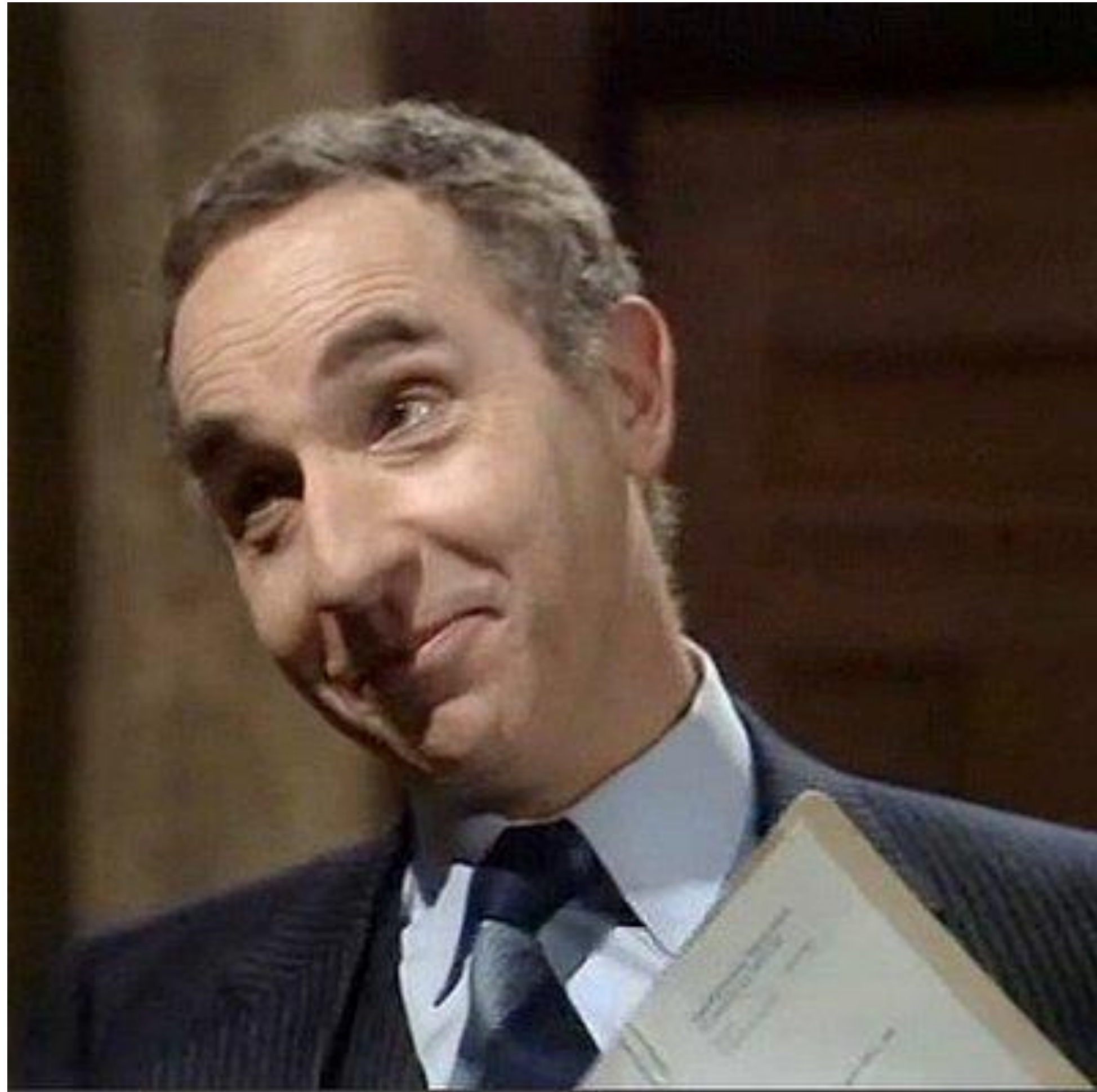


# Not all errors are equal



- 'Black defendants who did not reoffend... were nearly twice as likely to be misclassified as higher risk compared to their white counterparts (45 percent vs. 23 percent)'.
- 'white defendants who reoffended... were mistakenly labeled low risk almost twice as often as black reoffenders (48 percent vs. 28 percent)'.

# Beware of quick and easy fixes



- The Politician's Syllogism
- We must do something
- 'This' is something
- Therefore we must do 'this'

# Keeping a human in the mix

‘When it comes to decisions that impact on people’s lives – judicial decisions etc- then a human should be accountable and in control of those.’

Noel Sharkey, Moral Maze, 18 Nov 2017



# Belt and braces, or false reassurance?



- Supervisor vs driver reaction time
- Inert but alert?
- Decisional atrophy



“Automation bias” or “algorithmic aversion”

‘It remains to be seen, however, how an algorithm might influence custody officer decision-making practices in future. Might some (consciously or otherwise) prefer to abdicate responsibility for what are risky decisions to the algorithm, resulting in deskilling and ‘judgmental atrophy’? Others might resist the intervention of an artificial tool. Only future research will determine this.’

- Oswald, Grace, Urwin and Barnes. ‘Algorithmic risk assessment policing models’ *Information & Communications Technology Law* (2018)

## Real empowerment, or passing the buck?

- Individual data subjects are not empowered to make use of the kind of algorithmic explanations they are likely to be offered
- Individuals mostly too time-poor, resource-poor, and lacking in the necessary expertise to meaningfully make use of these rights
- Individual rights approach not well suited when algorithms create societal harms, such as discrimination against racial or minority groups.
- Lilian Edwards and Michael Veale, 'Slave to the Algorithm? Why a 'right to an explanation' is probably not the remedy you are looking for.'

# Impossible standards, or settling for too little?

