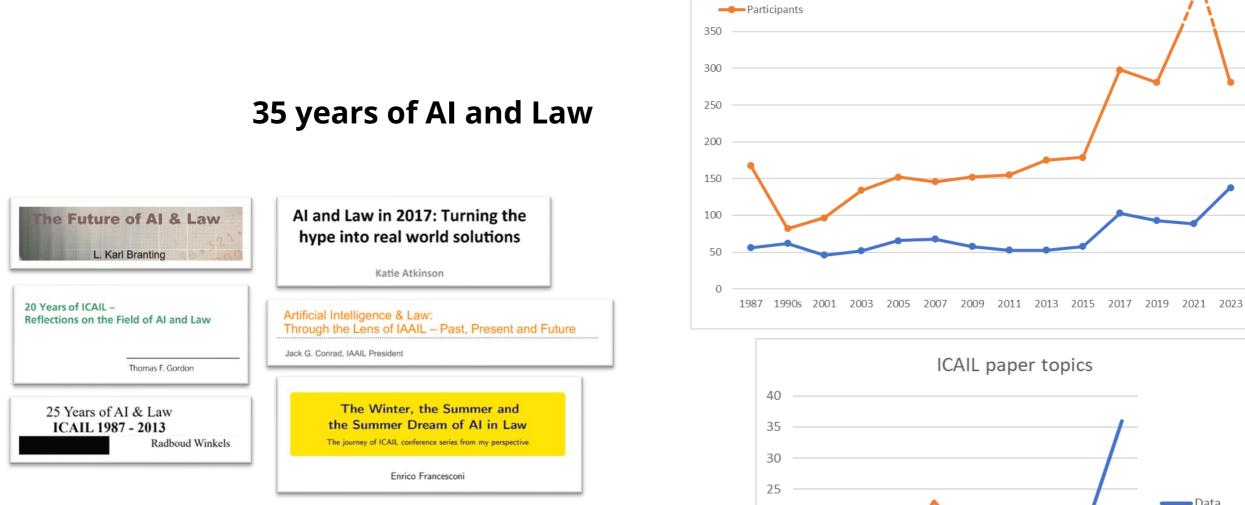
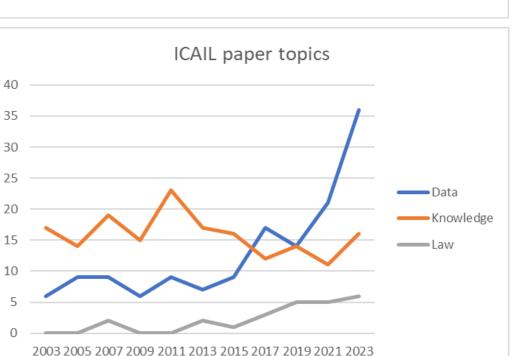


Associate Professor Artificial Intelligence (Utrecht University) Full Professor Data Science and the Judiciary (Tilburg University)



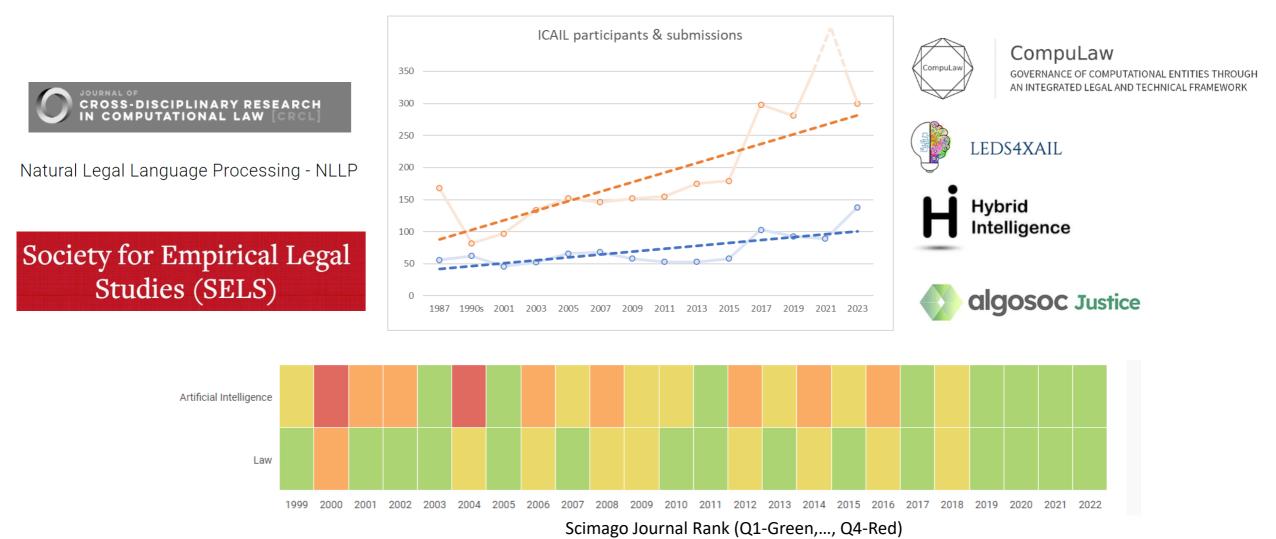


ICAIL participants & submissions

-----Submissions

1240

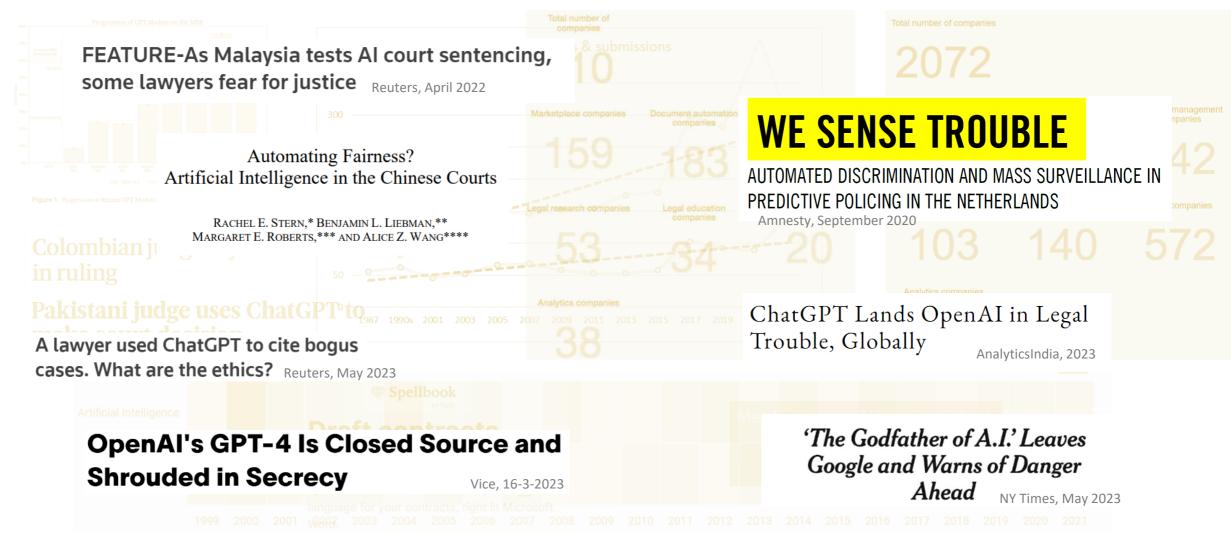
#### AI & Law is doing great in 2023



#### AI & Law is doing great in 2023



#### AI & Law is doing great in 2023 (right??)



#### Futurama - Fox





Robocop - MGM

## An "algorithmic drama"

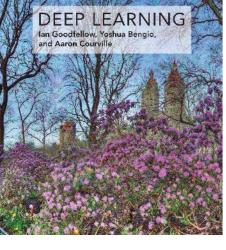


Ziewitz (2016). Governing algorithms: Myth, mess, and methods. *Science, Technology, & Human Values* 

people.dsv.su.se/~jpalme/reports/right-wrong.html



Robocar Poli – Roi Visual





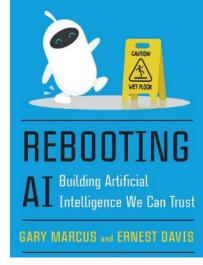


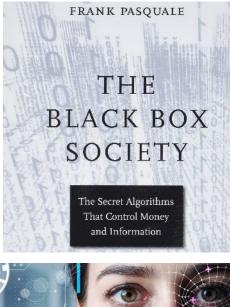


- Data-driven deep learning vs. knowledge-driven reasoning
- Techno-optimism vs. techno scepticism
- Building AI vs. Regulating AI



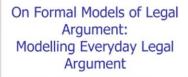
Human-centered Artificial Intelligence







Governing the Digital Society



Henry Prakken



Bart Verheij

Giovanni Sartor

# Stepping away from the drama – A way forward for AI & Law

- 1. Combine knowledge & data
  - Use new techniques without forgetting about the old ones
- 2. Evaluate how AI & Law is being used in practice
  - Develop and broadly evaluate AI & Law applications
- 3. Combine multiple disciplines
  - Law, AI, and beyond





- Collaboration between police and universities
  - Research, develop & evaluate AI for real police problems, in actual police context
- 20 PhDs
  - Majority also works at police
  - 2/3 CS/AI background, 1/3 other (information systems, public management, law, communication studies)

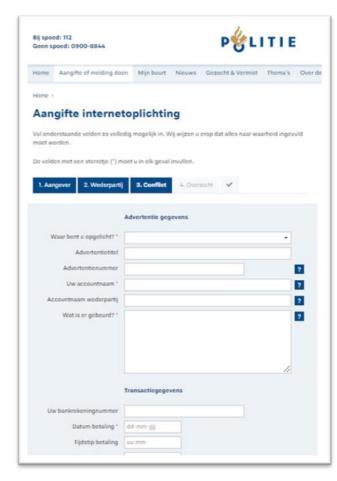
#### **Example 1: AI for citizen complaint/report intake**

- Trade fraud: false webshops, malicious traders on Ebay
  - 40,000+ reports of alleged online fraud per year
  - Not all fraud: wrong product, not paid
- Automatically recommend to file report or not
  - Citizen fills in a form w. details & free text story
  - Possible fraud or not?

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#### AI for intake - data & knowledge

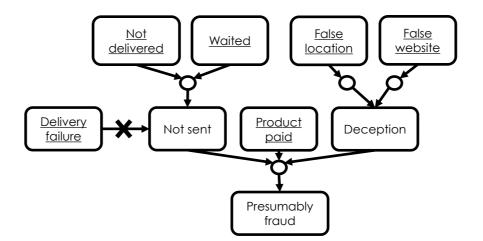
- Combine data- and knowledge-driven Al
  - Relevant legal rules are known, bounded domain
  - Free-text interpretation needs data-driven Al



Ashley & Walker (2013) Toward constructing evidence-based legal arguments using legal decision documents and machine learning. *ICAIL 2013*  Schraagen, Testerink, Odekerken, Bex (2018) Argumentation-driven information extraction for online crime reports. *LeDAM 2018*  Mumford, Atkinson, & Bench-Capon (2022). Reasoning with Legal Cases: A Hybrid ADF-ML Approach. *Jurix 2022*.

#### AI for intake – legal model

#### Legal model



Computational argumentation Rules w. exceptions based on DCC & police policy rules Schraagen, Bex (2019) Extraction of semantic relations in noisy user-generated law enforcement data, *IEEE Semantic Computing (ICSC)*.

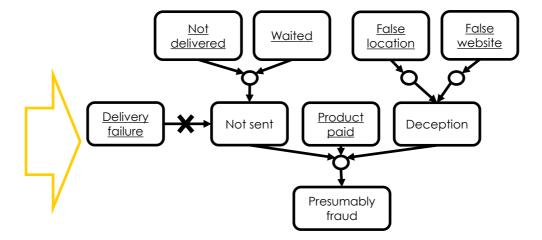
Schraagen, Brinkhuis & Bex (2017) Evaluation of Named Entity Recognition in Dutch Online Criminal Complaints. *DESI VII @ ICAIL 2017* 

#### AI for intake – free text

#### Complaint form

#### Legal model

Fictitious example report 1 I would like to report fraud. I recently saw a bicycle for sale on eBay and contacted the advertiser. He said he lived far away, so he would send me the bike. I paid him in good faith, but have still not received anything. I saw on Facebook he lives nearby.



Computational argumentation Rules w. exceptions based on DCC & police policy rules

D. Odekerken, F. Bex, A. Borg, B. Testerink (2022) Approximating Stability for Applied Argumentbased Inquiry. *Intelligent Systems with Applications*.

#### AI for intake - combining IR and argumentation

# Extracting observations from complaint form

Inferring possible fraud (or not)

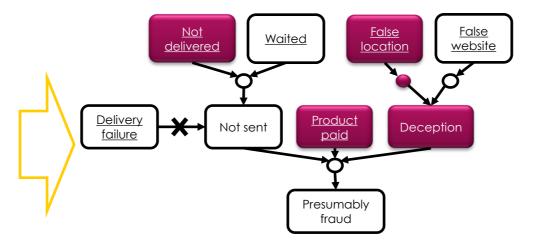
Paid

Fictitious example report 1 I would like to report fraud. I recently saw a bicycle for sale on eBay and contacted the advertiser. He said he lived far away, so he would send me the bike. I paid him in good faith, but have still not received anything. I saw on Facebook he lives nearby.

False location

Not delivered

Basic information extraction



Computational argumentation Rules w. exceptions based on DCC & police policy rules

D. Odekerken, F. Bex, A. Borg, B. Testerink (2022) Approximating Stability for Applied Argumentbased Inquiry. *Intelligent Systems with Applications*.

#### AI for intake – asking the right questions

# Extracting observations from complaint form

#### Inferring possible fraud (or not)

Asking for missing observations

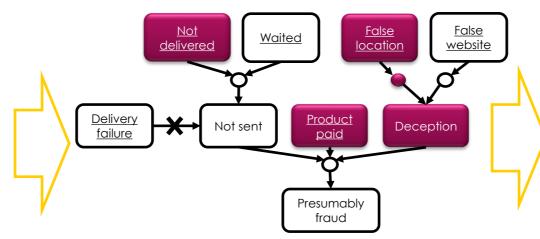
Paid

Fictitious example report 1 I would like to report fraud. I recently saw a bicycle for sale on eBay and contacted the advertiser. He said he lived far away, so he would send me the bike. I paid him in good faith, but have still not received anything. I saw on Facebook he lives nearby.

False location

Not delivered

Basic information extraction



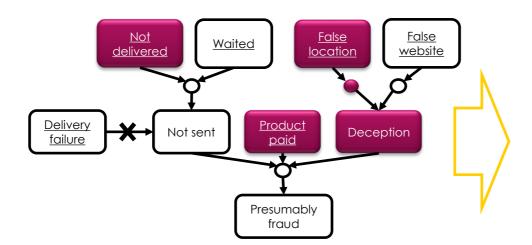


Computational argumentation Rules w. exceptions based on DCC & police policy rules Approximation algorithms Can new info still change the conclusion (and if so which)?

A. Borg & F. Bex (2021) Explaining Arguments at the Dutch National Police. *Explainable AI for Law* (*XAILA*).

#### AI for intake - explanations

#### Inferring possible fraud (or not)



#### Response

Thank you for your complaint. In your case, the system has concluded that it is not a case of fraud, since you did not wait for at least 5 days. We recommend you do not file an official report at this point.

Computational argumentation Rules w. exceptions based on DCC & police policy rules

#### Explanations

*Explaining (non-)acceptance in terms of arguments and counterarguments* 

#### Al for intake - evaluation

- Evaluate accuracy, user satisfaction
- Investigate citizen trust in automatic recommendations
  - How do users perceive recommendations by the system?
  - Do explanations matter?

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E. Nieuwenhuizen, A. Meijer, F. Bex, S. Grimmelikhuijsen Explanations increase citizen trust in police algorithmic recommender systems: Findings from two experimental tests. *Under Review* 

#### Al for intake – citizen trust & explanations

- Do citizens trust the system with and without an explanation?
  - Controlled experiments 1700+ participants
- Not fraud still file an official report? (trusting behaviour)?
  - No explanation (control): 40-60% still filed report
  - With explanation: only 20-35% still filed report

# AI, transparency and citizen trust

- Transparency
  - About the system/decision: XAI
  - About the use of systems in the organisation
  - About AI and regulation
- How do citizens react when AI is more contentious?
  - Predictive policing
- What's the influence of basic trust in police?
  - US vs. Netherlands

#### Example 2: AI for (explainable) text classification

- Police generate, use and analyse lots of text data
  - Citizen reports, Incident reports, Lab reports, Social Media, Seized Data Carriers
- Text classification for search, for use in Al systems

Fictitious example report 1 I recently saw a bicycle for sale online and contacted the advertiser. He said he lived far away, so he would send me the bike. I paid him in good faith, but have still not received anything. I saw on Facebook he lives nearby.

Fictitious example report 2 I wanted to buy champagne from John Doe via Ebay. Up to now, I have not received anything, and he does not respond to my emails, so I haven't transferred the money yet.

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$\left[ \right]$	Not paid	

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	9:57 Att - 24 Jan 2015	
	BBC World Service	
No Threat	75 years ago this week, atomic bombs were dro on Hiroshima and Nagasaki, Japan, killing coun	
	civilians.	

Ribeiro, Singh & Guestrin (2016). "Why should i trust you?" Explaining the predictions of any classifier. In *ACM SIGKDD*. Branting et al. (2019) Semi-supervised methods for explainable legal prediction. *ICAIL 2019*.

Tan, Zhang, Zhang, & Li (2020). The sentencing-element-aware model for explainable term-of-penalty prediction. *NLPCC 2020*.

# AI for explainable text classification

- Being able to explain why something was classified
  - Model testing and improvement
  - Transparency and accountability
  - Use in legal (criminal) cases

Fictitious example report 1
I recently saw a bicycle for sale online and
contacted the advertiser. He said he lived far
away, so he would send me the bike. I paid
him in good faith, but have still not received
anything. I saw on Facebook he lives nearby.

Fictitious example report 2 I wanted to buy champagne from John Doe via Ebay. Up to now, I have not received anything, and he does not respond to my emails, so I haven't transferred the money yet.

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$\bigcap$	Not paid	٦

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<b></b>	BBC World Service wbbcworldservice		
o Threat	75 years ago this week, atomic bombs		
	on Hiroshima and Nagasaki, Japan, kil civilians.	ing countiess	

E. Herrewijnen, D. Nguyen, J. Mense & F. Bex (2021) Machine-annotated Rationales: Faithfully Explaining Text Classification. *AAAI Explainable Agency in AI Workshop*.

#### **Explainable text classification - Rationales**

- Explaining text classification
  - Using *machine generated rationales* (highlighted sentences)

Fictitious example report 1 I recently saw a bicycle for sale online and contacted the advertiser. He said he lived far away, so he would send me the bike. I paid him in good faith, but have still not received anything. I saw on Facebook he lives nearby.

Fictitious example report 2 I wanted to buy champagne from John Doe via Ebay. Up to now, I have not received anything, and he does not respond to my emails, so I haven't transferred the money yet.



Not paid

Fictitious example report 1 I recently saw a bicycle for sale online and contacted the advertiser. He said he lived far away, so he would send me the bike. I paid him in good faith, but have still not received anything. I saw on Facebook he lives nearby.

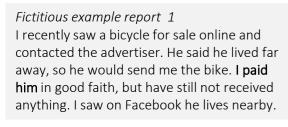
Fictitious example report 2

I wanted to buy champagne from John Doe via Ebay. Up to now, I have not received anything, and he does not respond to my e-mails, **so I** haven't transferred the money yet.

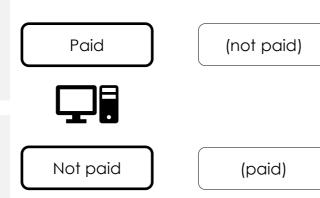
M. Robeer, F. Bex & A. Feelders (2021) Generating Realistic Natural Language Counterfactuals. *Findings of EMNLP 2021*.

#### **Explainable text classification - Counterfactuals**

- Explaining text classification
  - By generating *counterfactual example text*



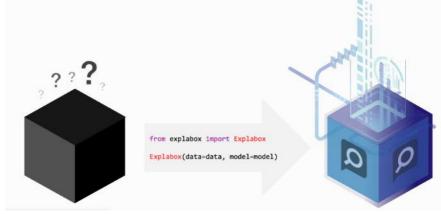
*Fictitious example report 2* I wanted to buy champagne from John Doe via Ebay. Up to now, I have not received anything, and he does not respond to my emails, **so I haven't transferred the money yet**.



Fictitious example report 1 I recently saw a bicycle for sale online and contacted the advertiser. He said he lived far away, so he would send me the bike. I did not pay him in good faith, but have still not received anything. I saw on Facebook he lives nearby.

Fictitious example report 2 I wanted to buy champagne from John Doe via Ebay. Up to now, I have not received anything, and he does not respond to my e-mails, **so I already transferred the money**.

# **Explainable AI for legal decisions**



https://explabox.readthedocs.io/

- Open-source libraries & toolkit for AI model inspection
  - Data statistics
  - XAI: rationales, counterfactuals, LIME/SHAP
  - Robustness: spelling mistakes, typo's
  - Biases: names, gender, etc.
- A holistic view on the AI system
  - What kind of data? How (good) does the system perform? Why does the system do what it does?

Tolan, Miron, Gómez & Castillo (2019) Why Machine Learning May Lead to Unfairness: Evidence from Risk Assessment for Juvenile Justice in Catalonia. *ICAIL 2019* 

Alikhademi et al. (2022) A review of predictive policing from the perspective of fairness. *AI & Law Journal* 

Sevim, Şahinuç & Koç (2023) Gender bias in legal corpora and debiasing it. *Natural Language Engineering* 



#### Rules, tools, and metrics

Brussels, 21.4.2021 COM(2021) 206 final

2021/0106 (COD)

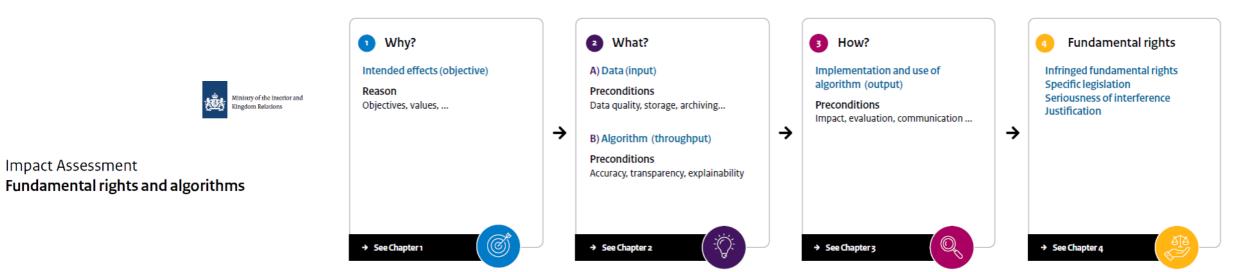
#### ACM Conference on Fairness, Accountability, and Transparency (ACM FAccT)

A computer science conference with a cross-disciplinary focus that brings together researchers and practitioners interested in fairness, accountability, and transparency in socio-technical systems.

Proposal for a

REGULATION OF THE EUROPEAN PARLIAMENT AND OF THE COUNCIL

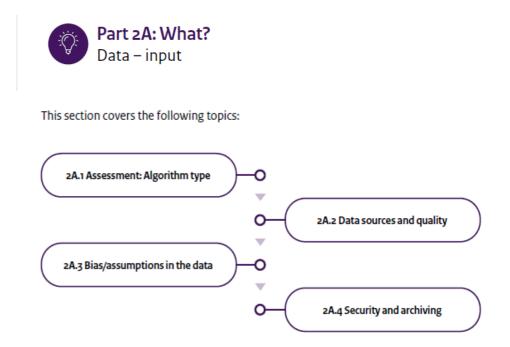
#### LAYING DOWN HARMONISED RULES ON ARTIFICIAL INTELLIGENCE (ARTIFICIAL INTELLIGENCE ACT) AND AMENDING CERTAIN UNION LEGISLATIVE ACTS



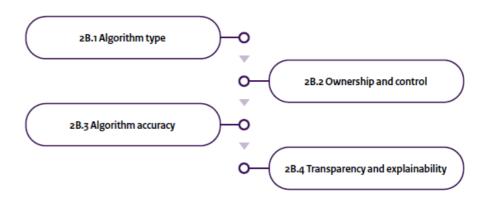


# Explabox as assessment aid

- Use information from Explabox for assessment
  - What kind of data? How (good) does the system perform? Why does the system do what it does?









#### Rules, tools, and metrics

- Tools & metrics
  - What use are they? Intended and actual effects?
  - New roles and responsibilities in organisations

MICHAEL POWER **The Audit Society** Rituals of Verification



Wieringa (2020) What to account for when accounting for algorithms: a systematic literature review on algorithmic accountability. *FAT\* 2020*.



**Explainable AI for legal decisions** 

Proposal for a

Brussels, 21.4.2021 COM(2021) 206 final

2021/0106 (COD)

REGULATION OF THE EUROPEAN PARLIAMENT AND OF THE COUNCIL

LAYING DOWN HARMONISED RULES ON ARTIFICIAL INTELLIGENCE (ARTIFICIAL INTELLIGENCE ACT) AND AMENDING CERTAIN UNION LEGISLATIVE ACTS

- Rules: Operationalising transparency and contestability in the law
  - Equality of arms

Quattrocolo et al. (2020) Technical solutions for legal challenges: equality of arms in criminal proceedings. *Global Jurist*.





**Explainable AI for legal decisions** 

Proposal for a

Brussels, 21.4.2021 COM(2021) 206 final

2021/0106 (COD)

REGULATION OF THE EUROPEAN PARLIAMENT AND OF THE COUNCIL

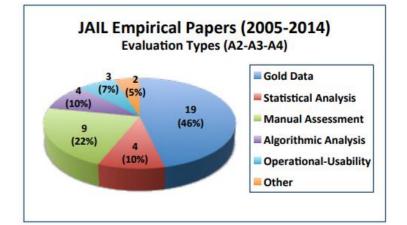
LAYING DOWN HARMONISED RULES ON ARTIFICIAL INTELLIGENCE (ARTIFICIAL INTELLIGENCE ACT) AND AMENDING CERTAIN UNION LEGISLATIVE ACTS

- Rules: Operationalising transparency and contestability in the law
  - Equality of arms
  - Evaluating evidence and motivating decisions

Bibal et al. (2021) Legal requirements on explainability in machine learning. *Al & Law Journal*  Almada (2019) Human intervention in automated decision-making: Toward the construction of contestable systems. *ICAIL 2019*. Atkinson, Bench-Capon & Bollegala (2020) Explanation in AI and law: Past, present and future. *AI Journal*. F.J. Bex (2011) *Arguments, Stories and Criminal Evidence: A Formal Hybrid Theory*.

#### **Evaluating AI in practice**

- AI & Law Tech development & evaluation
  - Argumentation
  - Natural Language Processing
- How to evaluate?
  - Real systems in real user context



Atkinson, Collenette, Bench-Capon, Dzehtsiarou (2021) Practical tools from formal models: the ECHR as a case study. *ICAIL 2021*. Odekerken & Bex (2020) Towards transparent human-in-the-loop classification of fraudulent web shops. *JURIX 2020*  Van Binsbergen, Liu, Van Doesburg & Van Engers (2020) eFLINT: a domain-specific language for executable norm specifications. *ACM SIGPLAN Conference on Generative Programming*.

Conrad & Zeleznikow (2015) The role of evaluation in AI and law: an examination of its different forms in the AI and law journal. *ICAIL 2015*.

## AI for detecting mobile phone usage while driving

• Al does initial filtering for pictures of cars/drivers who seem to be holding a phone, officer then checks the picture.



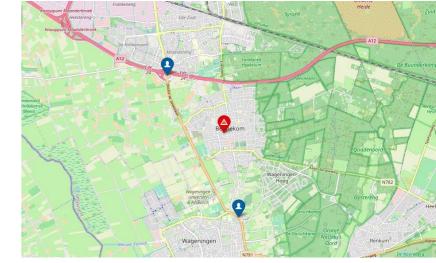
E. Fest et al. (2023) Values? Camera? Action! An ethnography of an AI camera system used by the Netherlands Police. *Under review*.

# AI for detecting mobile phone usage while driving

- Best-practice in value-sensitive design
  - Data protection and anonimization
  - Training models with representative datasets
  - Develop and retain control in-house
- In practice:
  - New windscreen foil on cars
  - Officers share photographs with other officers to get second opinion
- Continuous training of both AI and human!



# AI for police interception



- Notification of crime (e.g. robbery, smash & grab) and fleeing suspects
- Using knowledge about suspect behaviour, roads, etc., predict the suspect's route
- "Just like I thought"
  - Expert dispatchers only followed the recommendations of the system if they coincided with their own intuitions
  - Explanations hardly influence whether they trust/follow the recommendation

Van Droffelaar, I.S., Kwakkel, J.H., Mense, J.P., Verbraeck, A. (2022) Simulation-optimization configurations for fugitive interception. *Proceedings of the 2022 Winter Simulation Conference*. F. Selten, M. Robeer, S. Grimmelikhuijsen (2022) 'Just like I thought'. Street-level bureaucrats trust AI recommendations if they confirm their professional judgement. *Public Administration Review*.

#### AI, Law and beyond

- 1. Combine knowledge & data
  - Trade fraud complaint intake
- 2. Evaluate how AI & Law is being used in practice
  - Citizen interaction & trust
  - XAI techniques in practice
  - Police officers using systems
- 3. Combine multiple disciplines
  - Public management: empirically investigating AI audits
  - Law: XAI, decision motivation and equality of arms

# **Combining knowledge and data**

- Legal information extraction
- Non-statistical models to reason with data
- Combining machine learning and knowledge representation approaches
  - ML to extract information, KR to reason
  - Both ML and KR for one (complex) task
  - Solving KR problems with ML models
  - Constraining ML models using KR models

Craandijk & Bex (2020) Deep Semantics. IJCAI 2020.

Li, Liu, Chen and Rudin (2018) Deep Learning for Abstract Argumentation Learning for Case-Based Reasoning Through Prototypes, AAAI 2018

Li & Srikumar (2019) Augmenting Neural Networks with First-order Logic. ACL 2019. Gan et al. (2021) Judgment Prediction via Injecting Legal Knowledge into Neural Networks AAAI 2021

**Argumentation Structure Prediction** in CJEU Decisions on Fiscal State Aid - Santin et al.

Automatic Identification and Empirical Analysis of Legally Relevant Factors – Gray et al.

Argument Mining with Graph Representation Learning – Zhang et al.

Improving Translation of Case Descriptions into Logical Fact Formulas using LegalCaseNER – Zin et al.

Computable Contracts by Extracting Obligation Logic Graphs – Servantez et al.

Justification, stability and relevance for case-based reasoning with incomplete focus cases – Odekerken et al.

Model- and data-agnostic justifications with a fortiori casebased argumentation - Peters et al.

Analogical Reasoning, Generalization, and Rule Learning for Common Law Reasoning – Blass & Forbus

Combining a Legal Knowledge Model with Machine Learning for Reasoning with Legal Cases – Mumford et al.

Beyond Readability with RateMyPDF: A Combined Rule-based and Machine Learning Approach to Improving Court Forms – Steenhuis et al.

# **Evaluate AI & Law in practice**

- Innovative applications
  - Evaluate operational-usability by "disinterested domain experts"
- Work together with stakeholders from practice
- Evaluate with proxy users
- Work with easily accessible user groups
  - Legal education
  - Academics

sustain.AI: a Recommender System to analyze Sustainability Reports – Hillebrand et al.

Image Analysis Approach to Trademark Congestion and Depletion – Haim & Kesari

"What's wrong with this product?" - Detection of product safety issues based on information consumers share online – Fuchs et al.

A Methodology for Building Augmented Intelligence Tools for Laypeople to Increase Access to Justice – Westermann & Benyekhlef

Beyond Readability with RateMyPDF: A Combined Rule-based and Machine Learning Approach to Improving Court Forms – Steenhuis et al.

Conrad & Zeleznikow (2015) The role of evaluation in AI and law: an examination of its different forms in the AI and law journal. *ICAIL 2015*.

# **Combine multiple disciplines**

- Al for Law
  - Apply AI for law, studying law with AI
- Law for Al
  - Legal-by-design
  - Legal aspects of AI for Law
- Include other disciplines
  - Broaden the AI & Law ecosystem

Araujo et al. (2020) In AI we trust? Perceptions about automated decision-making by artificial intelligence. *AI & Society*  Yalcin et al. (2023) Perceptions of Justice By Algorithms. *Al & Law Journal*  Barysė & Sarel (2023) Algorithms in the court: does it matter which part of the judicial decision-making is automated? *AI & Law Journal*  Using Agent-Based Simulations to Evaluate Bayesian Networks for Criminal Scenarios - van Leeuwen et al.

Do agents dream of abiding by the rules? Learning norms via behavioral exploration and sparse human supervision – Fratrič et al.

Uncovering Trauma in Genocide Tribunals: An NLP Approach Using the Genocide Transcript Corpus – Schirmer et al.

On predicting and explaining asylum adjudication – Katsikouli et al.

Gender Disparities in Child Custody Sentencing in Spain: a Data Driven Analysis – Riera et al.

The Perfect Victim: Computational Analysis of Judicial Attitudes towards Victims of Sexual Violence – Habba et al.

Computational Accountability -Hulstijn

Effects of XAI on Legal Process – Nielsen et al.

Rebuilding 'ethics' to govern 'AI': How to re-set the boundaries for the legal sector? - Unver

#### AI, law and beyond: building a transdisciplinary ecosystem

