

Spring School on Artificial Intelligence and Law

*Invited graduate course
at Central South University, Changsha*

Floris Bex, Enrico Francesconi, Bart Verheij, April 2019

Organisers: Juan Li, Minghui Xiong



中南大學
CENTRAL SOUTH UNIVERSITY

Day 3 Tuesday April 16

- 8:30 AI, Law & Data (Floris Bex)
- 10:00 *Break*
- 10:30 AI & Law - Semantic Annotation of Legal Texts (Enrico Francesconi)
- 12:00 *Break*
- 14:30 Arguments, Scenarios and Probabilities in AI & Law (Bart Verheij)
- 16:00 *Break*
- 16:30 Discussion
- 17:30

A Coffeehouse Conversation on the Van den Herik Test

Bart Verheij

Participants in the dialogue: Chris, a lawyer; Floris, a computer scientist; Alex.

Chris: Alex, I want to thank you for suggesting that I read Van den Herik's "Kunnen computers rechtspreken?". It's a wonderful piece and certainly made me think — and think about my thinking.

Alex: Glad to hear it. Are you still as much of a skeptic about applying artificial intelligence to the law as you used to be?

Chris: You've got me wrong. I'm not against artificial intelligence; I think it's wonderful stuff — perhaps a little crazy, but why not? I simply am convinced that you AI advocates have far underestimated the minds of lawyers, and that there are things a computer will never, ever be able to do. For instance, can you imagine a computer writing a volume in the Asser series? The richness in content, the complexity of the considerations —

Alex: Rome wasn't built in a day!

Floris: Hey, are you two going to clue me in as to what this text by Van den Herik is all about? It's only available in Dutch, you know!

Alex: It is about the question whether computers can decide legal cases. Van den Herik discusses the nature of legal decision making and the prospects of automating it. The text is Van den Herik's 1991 inaugural address, delivered upon acceptance of his position in Leiden. It also contains what might be called the Van den Herik test.

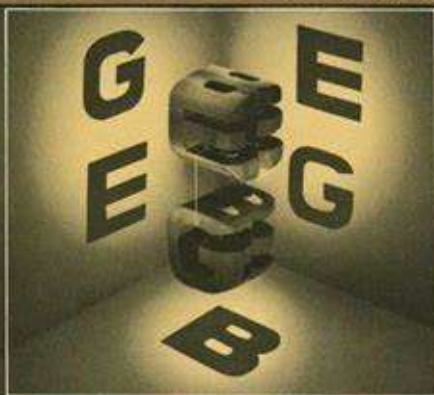
Reference:

Verheij, B. (2007). A Coffeehouse Conversation on the Van den Herik Test. *Liber Amicorum ter Gelegenheid van de 60e Verjaardag van Prof.Dr. H. Jaap van den Herik*, pp. 155-163. Maastricht: Maastricht ICT Competence Center.

WINNER OF THE PULITZER PRIZE

GÖDEL, ESCHER, BACH:

AN ETERNAL GOLDEN BRAID



DOUGLAS R.
HOFSTADTER

A METAPHORICAL FUGUE ON MINDS AND MACHINES
IN THE SPIRIT OF LEWIS CARROLL

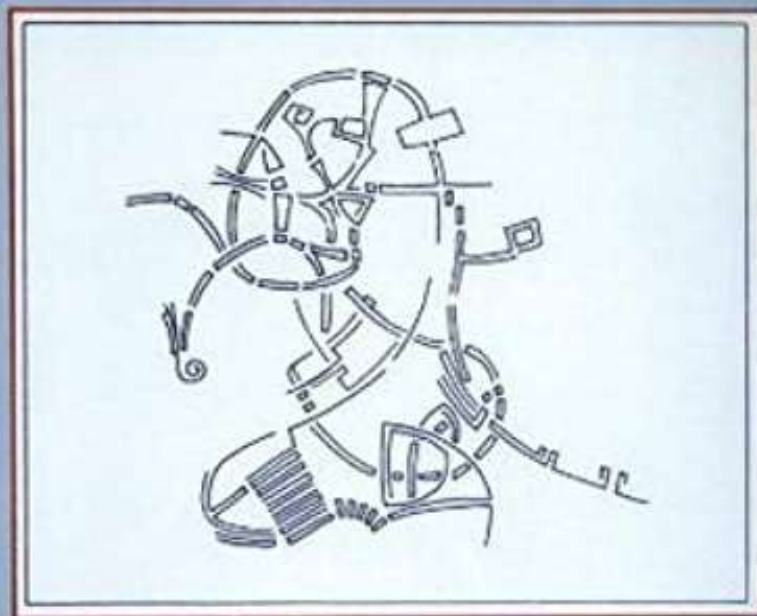
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METAMAGICAL THEMAS:

Questing for the Essence
of Mind and Pattern

DOUGLAS R. HOFSTADTER



*An Interlocked Collection of
Literary, Scientific, and Artistic Studies*

How can forensic evidence be handled effectively and safely?



Expert: "The probability is 1 in 342,000,000 that a nurse's shifts coincide with so many unexplained deaths and resuscitations."

Expert: "*Dat kan geen toeval zijn.*"
(That cannot be by chance.)

Analyses of what went wrong

1. The statistical calculations were erroneous.
Wrongly combining p-values

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Wrongly combining p-values
2. The statistics were erroneous.
Biased data collection

Analyses of what went wrong

1. The statistical calculations were erroneous.
Wrongly combining p-values
2. The statistics were erroneous.
Biased data collection
3. The statistics only show that what happened is rare.
Lack of context

What makes a suspect's guilt convincing?

When the context speaks for itself.

E.g.,

- The murder weapon is found.
- Fingerprints found on the gun match the suspect's.
- The suspect has `shooting hands'.
- The suspect is a known hitman.
- The victim was a drug dealer involved in a gang war.
- ...

What makes a suspect's guilt convincing?

When the context speaks for itself.

E.g.,

- The murder weapon was a .38 Smith & Wesson revolver.
- Fingerprints found on the gun match the suspect's.
- The suspect has 'shooting hands'.
- The suspect is a known hitman.
- The suspect was a drug dealer involved in a gang

...

Concrete facts in context

Goal:

promote rational handling of evidence in courts

Tool needed:

a normative framework

shared between experts and factfinders

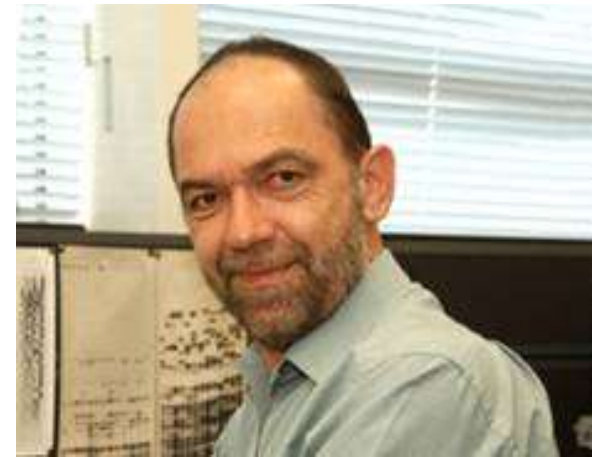
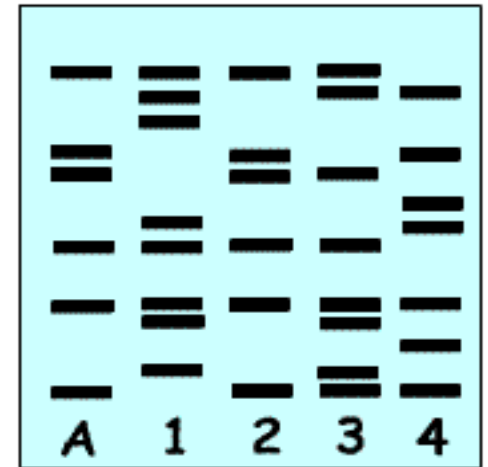
DNA profiling

Successful

High information value

Scientific foundation

Precise statistical information
(Random Match Probability)



DNA profiling

DNA Profile		Allele frequency from database			Genotype frequency for locus		
Locus	Alleles	Times allele observed	Size of database	Frequency		Formula	Number
CSF1PO	10	109	432	$p=$	0.25	$2pq$	0.16
	11	134		$q=$	0.31		
TPOX	8	229	432	$p=$	0.53	p^2	0.28
	8						
THO1	6	102	428	$p=$	0.24	$2pq$	0.07
	7	64		$q=$	0.15		
vWA	16	91	428	$p=$	0.21	p^2	0.05
	16						
			profile frequency= 0.00014				

Roughly
1 in 7000

Random Match Probability

Charles H. Brenner

“The DNA effect”

By the success and nature of DNA the following idea has gained momentum:

Evidence is only valuable when it comes with scientifically supported statistics.

(Cf. the CSI effect;

http://en.wikipedia.org/wiki/CSI_effect)

Proof With and Without Probabilities

Bart Verheij

Institute of Artificial Intelligence and Cognitive Engineering

`www.ai.rug.nl/~verheij`



university of
 groningen

faculty of mathematics
 and natural sciences



Goal:

promote rational handling of evidence in courts

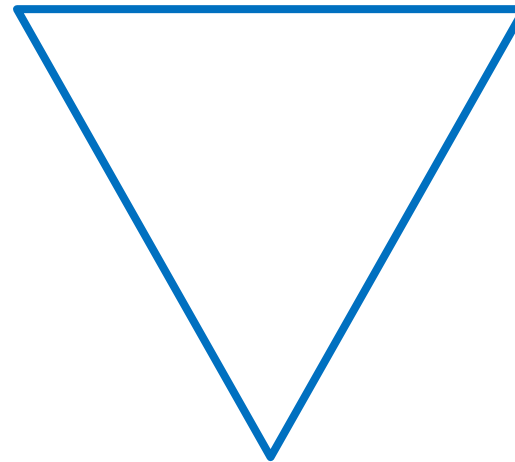
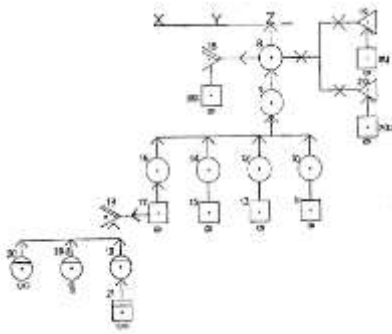
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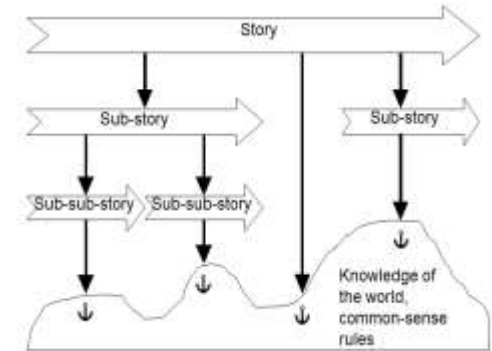
shared between experts and factfinders

Three normative frameworks

Arguments



Scenarios



Probabilities

$$\frac{p(H|E)}{p(\text{not-}H|E)} = \frac{p(E|H)}{p(E|\text{not-}H)} \cdot \frac{p(H)}{p(\text{not-}H)}$$

Posterior odds = Likelihood ratio · Prior odds

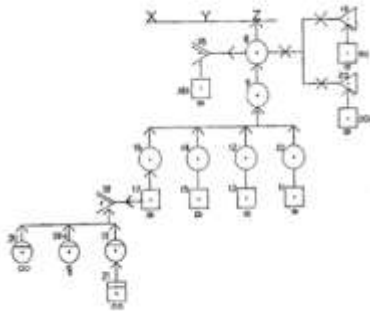
Three normative frameworks

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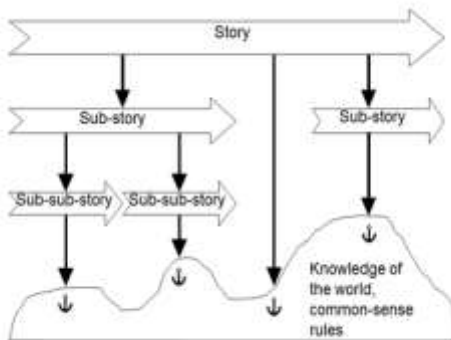
Probabilities

E.g., follow the calculus, don't transpose conditional probabilities, don't forget prior probabilities



Argumentation

E.g., take all arguments into account, both pro and con, assess strength and relative strength, avoid fallacies



Scenarios

E.g., consider alternative scenarios, assess plausibility and coherence, consider which evidence is explained or contradicted

Three normative frameworks

Probabilities

$$\frac{p(H|E)}{p(\text{not-}H|E)} = \frac{p(E|H)}{p(E|\text{not-}H)} \cdot \frac{p(H)}{p(\text{not-}H)}$$

Posterior odds = Likelihood ratio · Prior odds

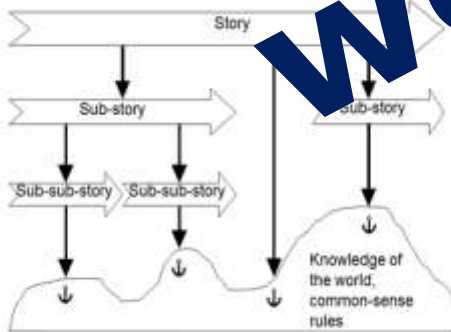
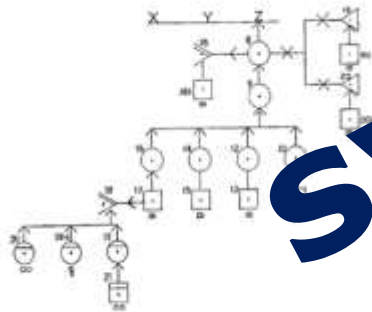
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Argumentation

E.g., take all arguments into account, both pro and con, assess strength and relative strength, avoid fallacies

Scenarios

E.g., consider alternative scenarios, assess plausibility and coherence, consider which evidence is explained or contradicted



systematic
well-regulated

EVIDENTIAL REASONING
Chapter for the Handbook of Legal Reasoning

Marcello Di Bello & Bart Verheij – April 19, 2017

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Conflicting evidence

Arguments Three kinds of attack can be distinguished: rebutting, undercutting and undermining. Three kinds of support can be distinguished: multiple, subordinated and coordinated. Arguments can involve complex structures of supporting and attacking reasons.

Scenarios There may be conflicting scenarios about what has happened. Evidence can be explained by one scenario, but not by another. Scenarios can be contradicted by evidence.

Probabilities Support can be characterized as “probability increase” or “positive likelihood ratio”. Attack can be characterized as “probability decrease” or “negative likelihood ratio”. The conflict between two pieces of evidence can be described probabilistically.

Evidential value

Probabilities The incremental evidential value is measured by probabilistic change. The overall evidential value is measured by the overall conditional probability. The use of evidence with high incremental evidential value has complications.

Arguments The reasons used can be conclusive or defeasible. Arguments can be evaluated by asking critical questions. It can be subject to debate whether a reason supports or attacks a conclusion.

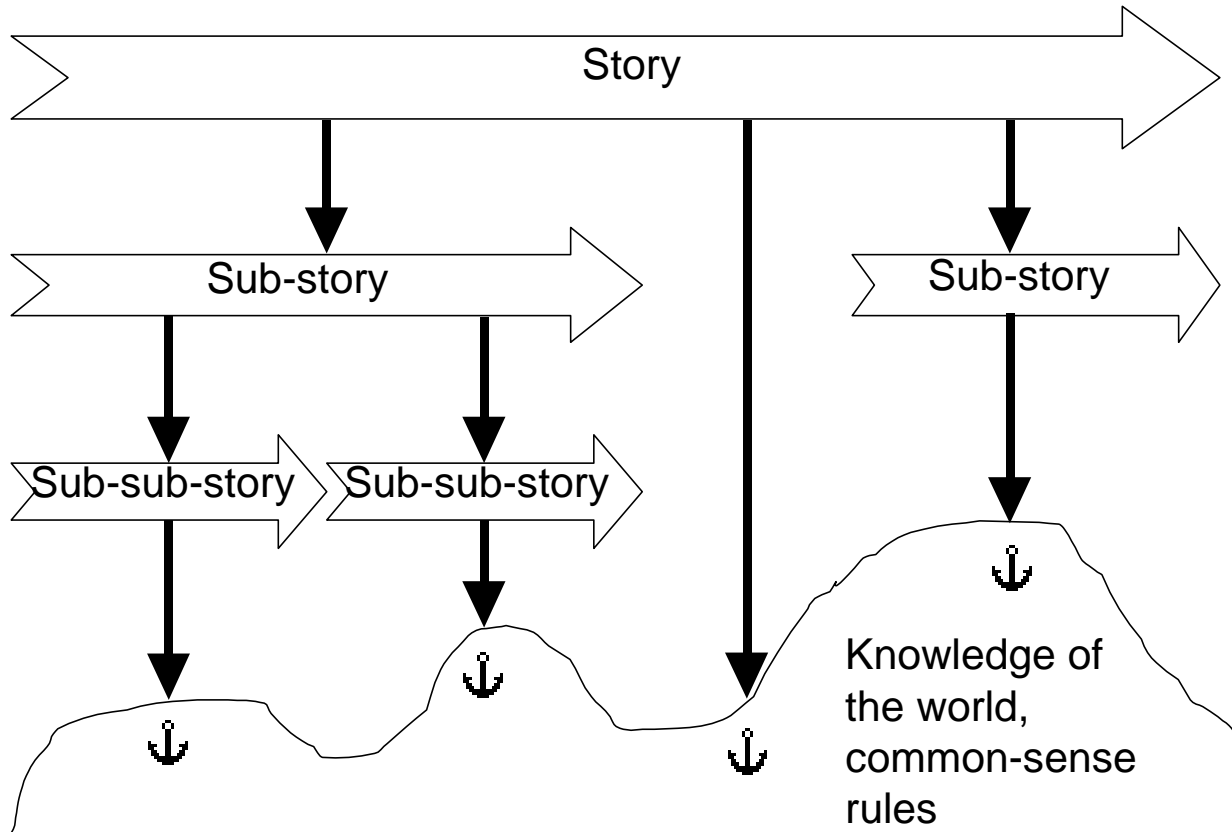
Scenarios Scenarios can be plausible and logically consistent. The more evidence a scenario can explain, the better. The more pieces of evidence a scenario is consistent with, the

Introduction

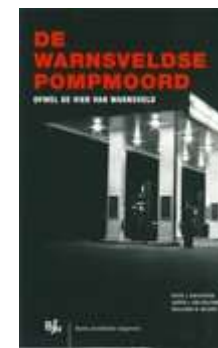
Hybrid models

AI & Law

Anchored narratives



Crombag, H.F.M., van Koppen, P.J., and Wagenaar, W.A. (1992, 1994), *Dubieuze Zaken: De Psychologie van Strafrechtelijk Bewijs. (Dubious Cases. The Psychology of Criminal Evidence.)* (Amsterdam: Contact).



Ten universal rules of evidence

1. The prosecution must present at least **one well-shaped narrative**.
2. The prosecution must present **a limited set of well-shaped narratives**.
3. **Essential components** of the narrative must be **anchored**.
4. **Anchors** for different components of the charge **should be independent** of each other.
5. The trier of fact should give **reasons for the decision** by specifying the narrative and the accompanying anchoring.
6. A fact-finder's decision as to **the level of analysis** of the evidence **should be explained through an articulation of the general beliefs used as anchors**.
7. There should be **no competing story** with equally good or better anchoring.
8. There should be **no falsifications of the** indictment's **narrative** and nested sub-narratives.
9. There should be **no anchoring onto obviously false beliefs**.
10. **The indictment and the verdict should contain the same narrative**.

Wagenaar, W.A., van Koppen, P.J., and Crombag, H.F.M. (1993), *Anchored Narratives. The Psychology of Criminal Evidence* (London: Harvester Wheatsheaf).

Anchored narratives

ANT can be regarded as a mixed approach, with **story-based** and **argument-based** elements.

Verheij, B. (2000). Dialectical Argumentation as a Heuristic for Courtroom Decision Making. *Rationality, Information and Progress in Law and Psychology. Liber Amicorum Hans F. Crombag* (eds. van Koppen, P.J., & Roos, N.), 203-226. Maastricht: Metajuridica Publications.

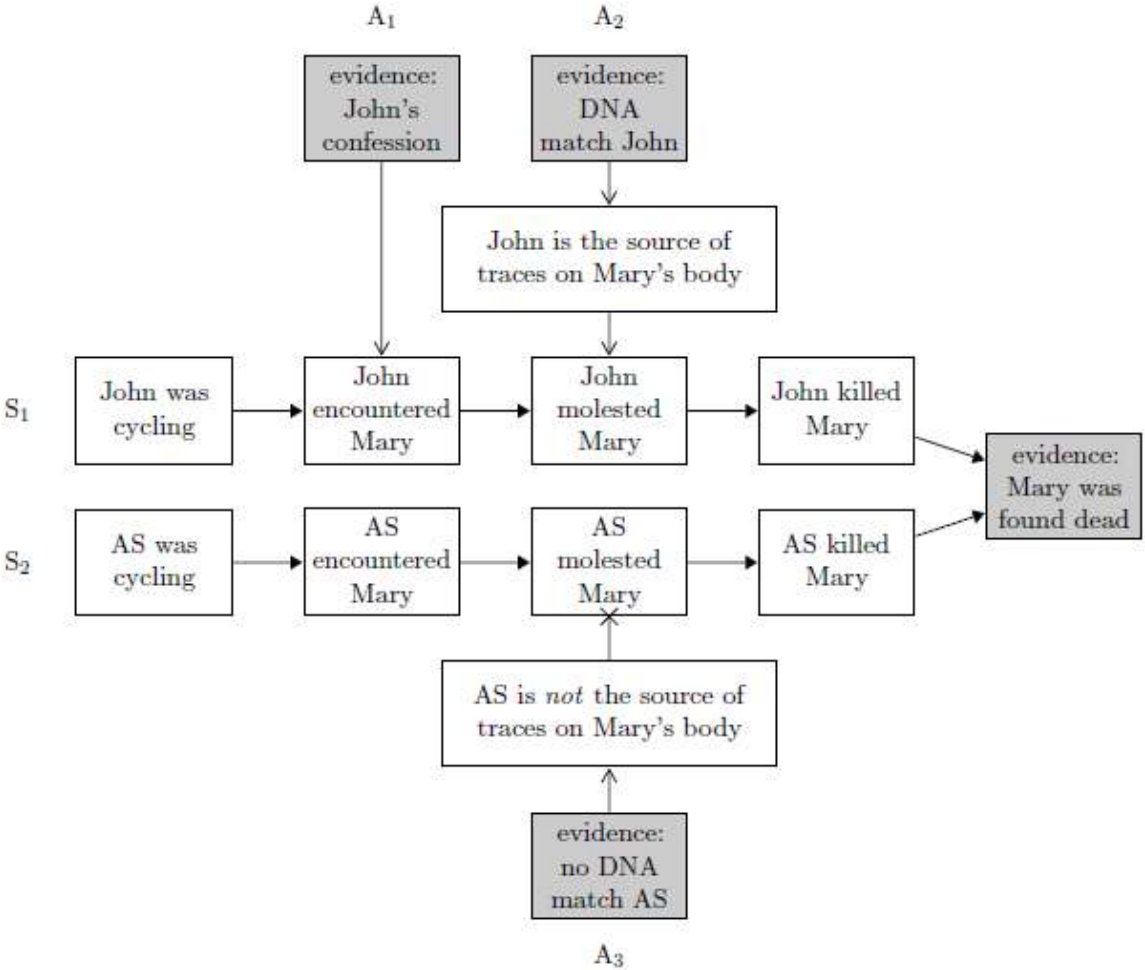
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Arguments and scenarios





Connecting arguments and scenarios: a hybrid theory

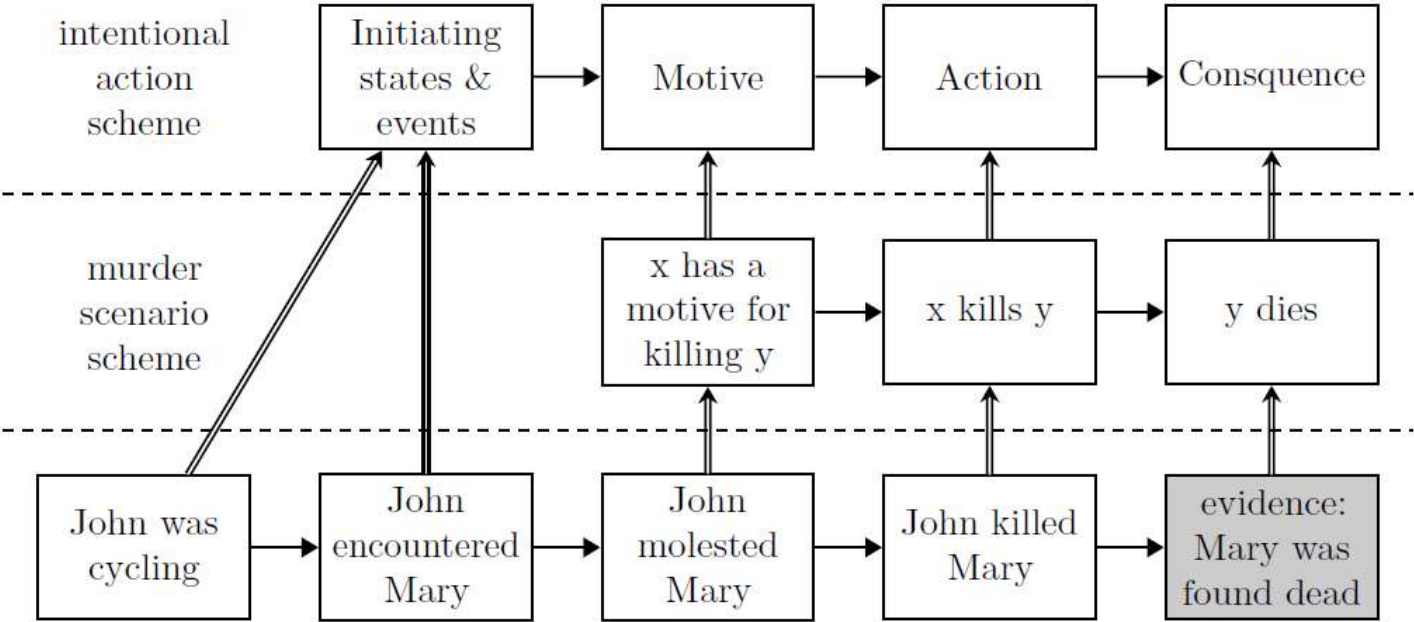


Figure 4: The scenario S_1 as an instance of different scenario schemes

Bayesian networks

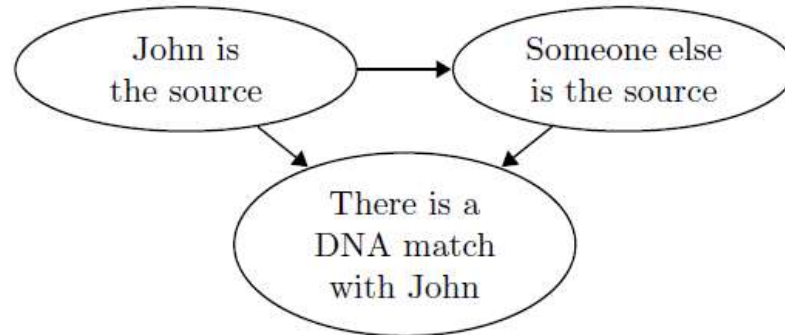


Figure 5: A Bayesian network structure with dependency relations

John is the source

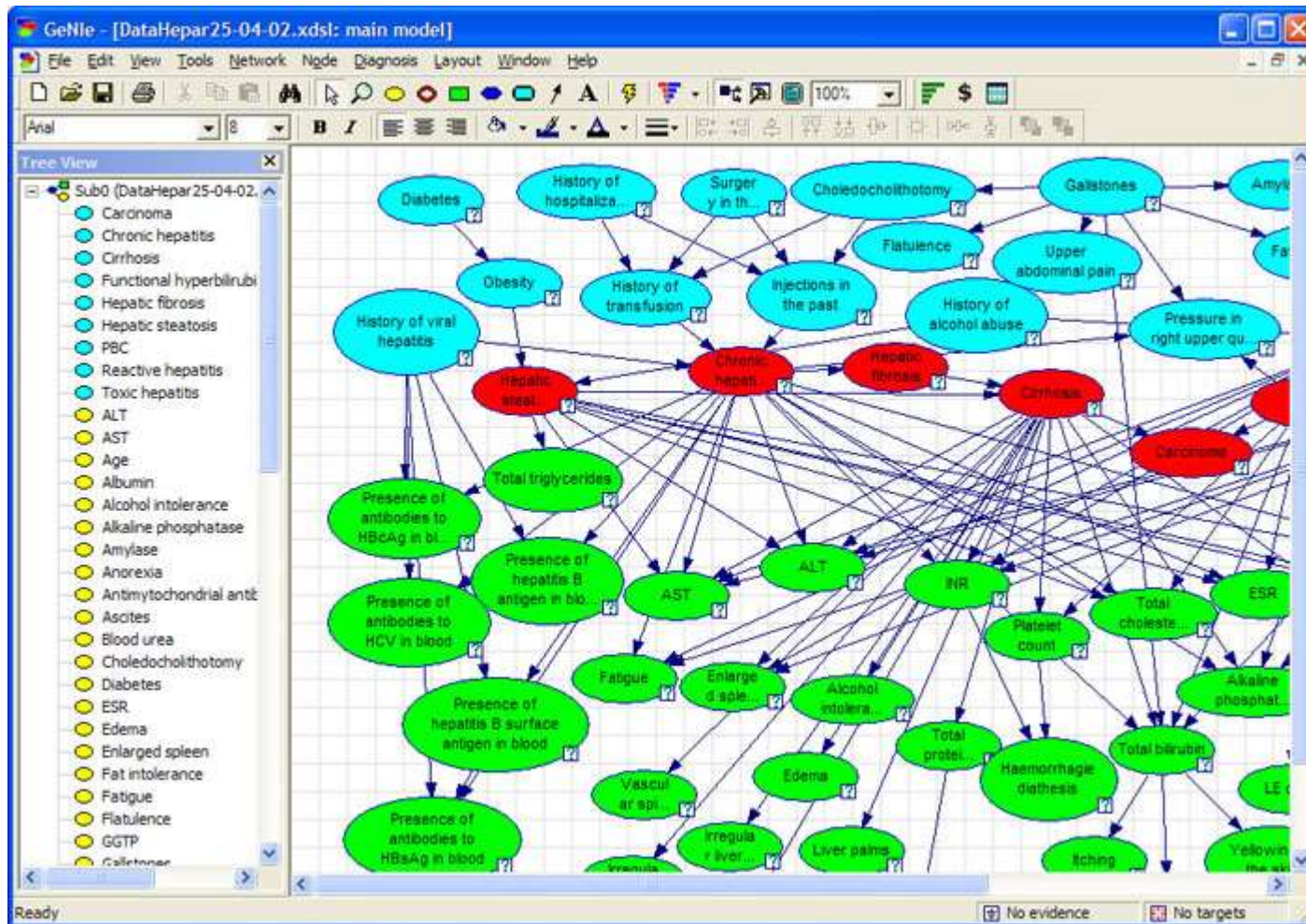
John is the source = false	8000/8001
John is the source = true	1/8001

Someone else is the source

John is the source	false	true
Someone else is the source = false	0	1
Someone else is the source = true	1	0

DNA match

John is the source	false		true	
Someone else	false	true	false	true
DNA match = false	0.5*	$1 - 0.66 \cdot 10^{-21}$	0	0.5*
DNA match = true	0.5*	$0.66 \cdot 10^{-21}$	1	0.5*



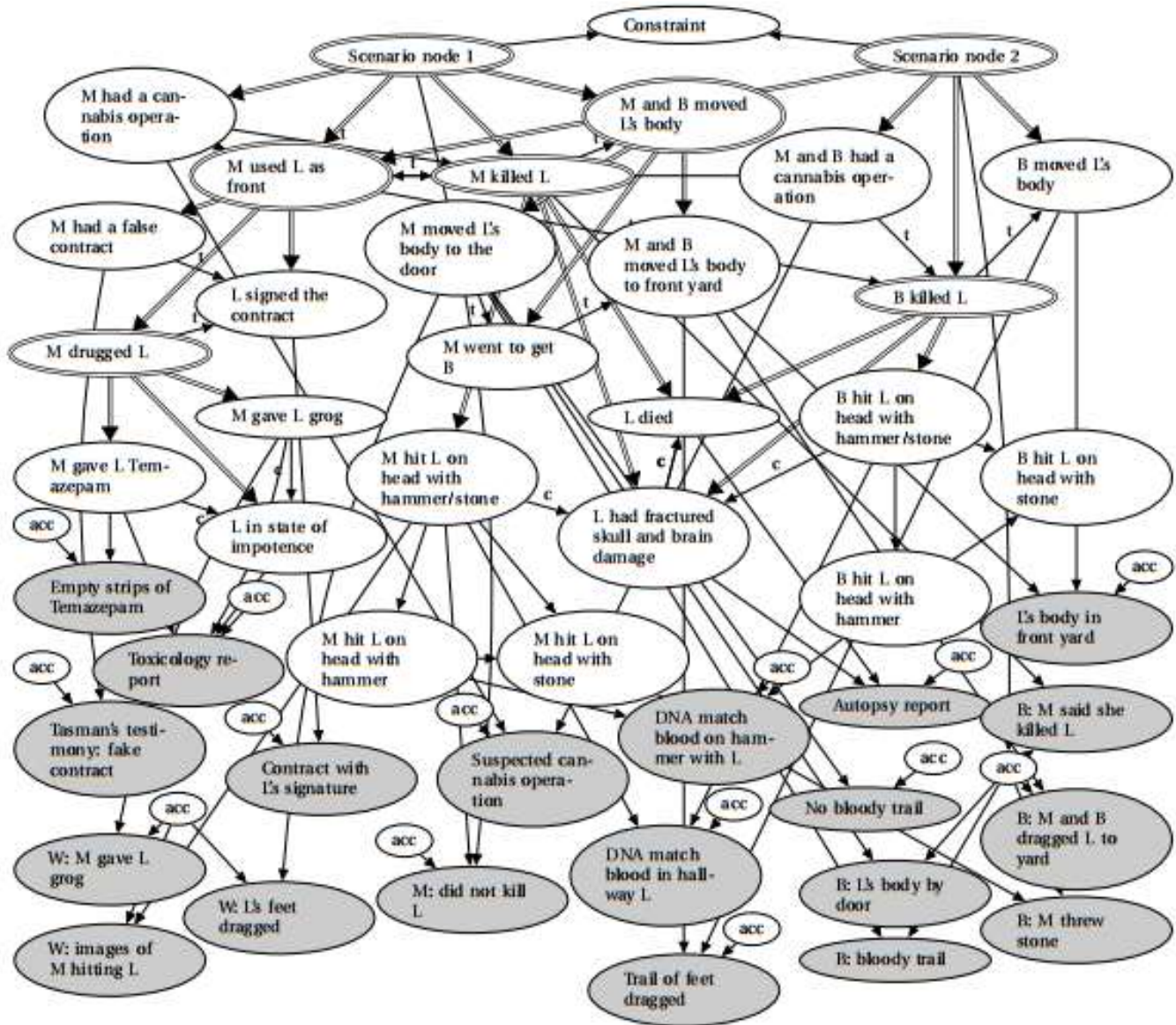


Figure 4.11: Scenario 1 and scenario 2 with evidence. Evidential nodes are indicated as grey nodes.

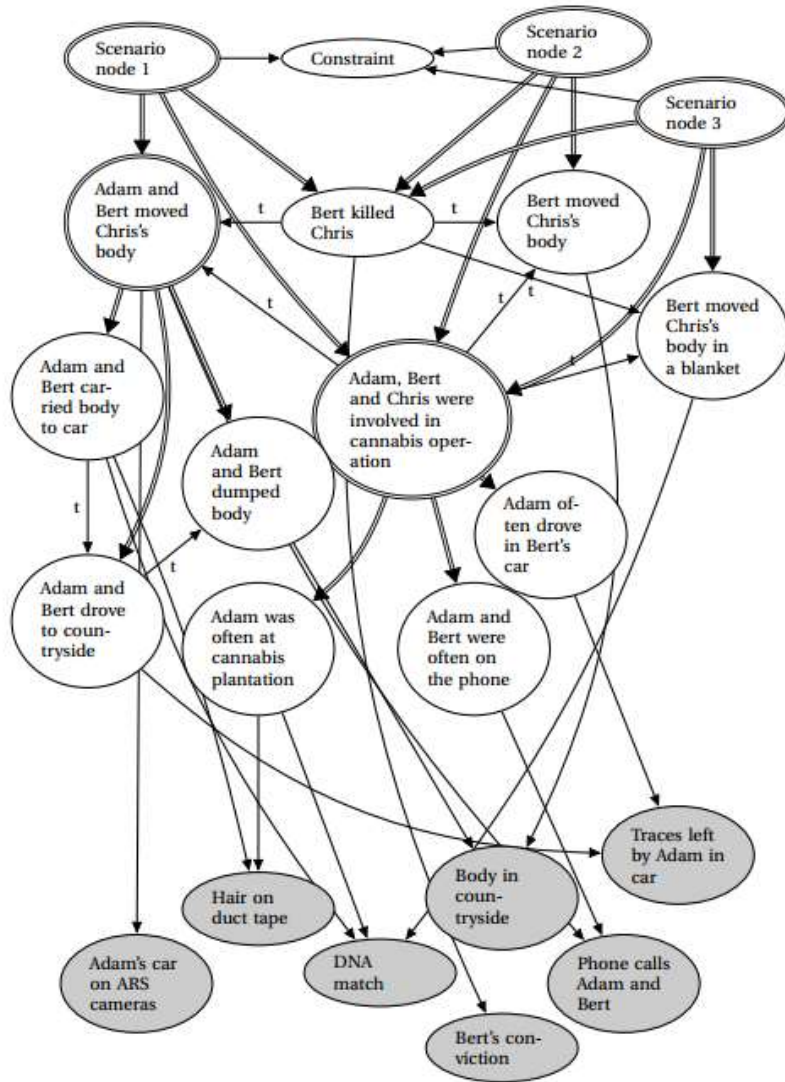


Figure 6.5: A network for the case study: The three scenarios with evidence. Evidential nodes are indicated as grey nodes.

• Scenarios in the network:

- Scenario 1 (prior probability: 0.001, posterior probability: 0.5296):

Scenario: Bert killed Chris, and Adam, Bert and Chris were involved in cannabis operation. Then Adam and Bert moved Chris's body.

Adam, Bert and Chris were involved in cannabis operation: Adam was often at cannabis location and Adam and Bert were often on the phone and Adam often drove in Bert's car.

Adam and Bert moved Chris's body: Adam and Bert carried body to car. Then Adam and Bert drove to countryside. Then Adam and Bert dumped body.

- Scenario 2 (prior probability: 0.001, posterior probability: 0.1180):

Scenario: Bert killed Chris, and Adam, Bert and Chris were involved in cannabis operation. Then Bert moved Chris's body.

Adam, Bert and Chris were involved in cannabis operation: Adam was often at cannabis location and Adam and Bert were often on the phone and Adam often drove in Bert's car.

- Scenario 3 (prior probability: 0.001, posterior probability: 0.2913):

Scenario: Bert killed Chris, and Adam, Bert and Chris were involved in cannabis operation. Then Bert moved Chris's body in a blanket.

Adam, Bert and Chris were involved in cannabis operation: Adam was often at cannabis location and Adam and Bert were often on the phone and Adam often drove in Bert's car.

• Scenario quality

- Scenario 1 is complete and consistent. It contains the supported implausible element Bert killed Chris.
- Scenario 2 is complete and consistent. It contains the supported implausible element Bert killed Chris.
- Scenario 3 is complete and consistent. It contains the supported implausible element Bert killed Chris.

• Evidence related to each scenario

- Evidence for and against scenario 1:

- Adam's car not on ARS cameras: weak evidence to attack scenario 1.
- DNA match: moderate evidence to support scenario 1.
- Hair on duct tape: moderate evidence to support scenario 1.
- Bert's conviction: moderate evidence to support scenario 1.
- Body in countryside: strong evidence to support scenario 1.
- Phone calls Adam and Bert: weak evidence to support scenario 1.
- Traces of Adam in car: weak evidence to support scenario 1.
- All evidence combined: strong evidence to support scenario 1.

- Evidence for and against scenario 2:

- Adam's car not on ARS cameras: weak evidence to attack scenario 2.
- DNA match: moderate evidence to support scenario 2.



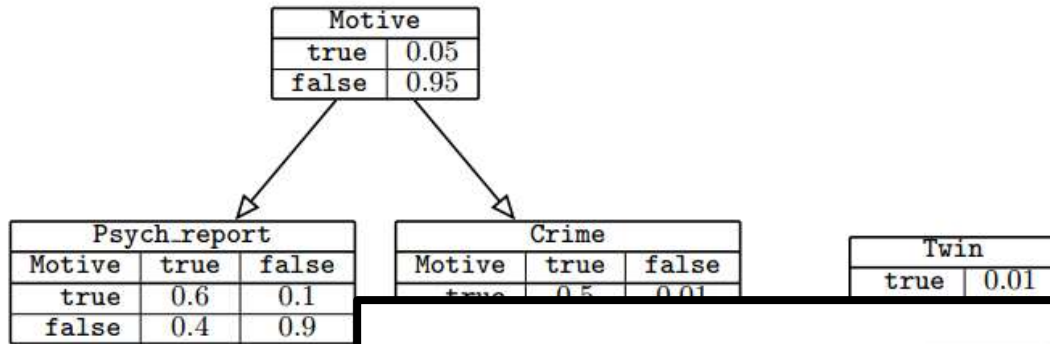


Figure 2: A small BN concerning Motive, Psych_report, Crime, and Twin. Each node is represented by a table showing conditional probabilities.

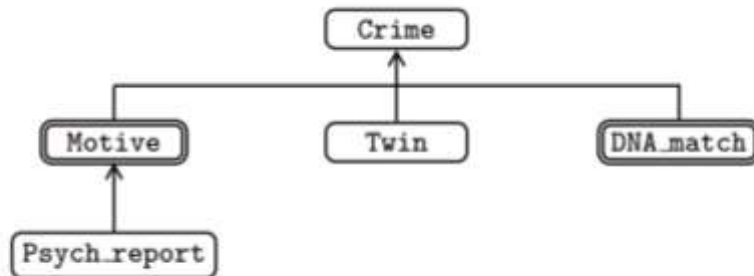


Figure 7: Support variables are depicted as rounded rectangles.

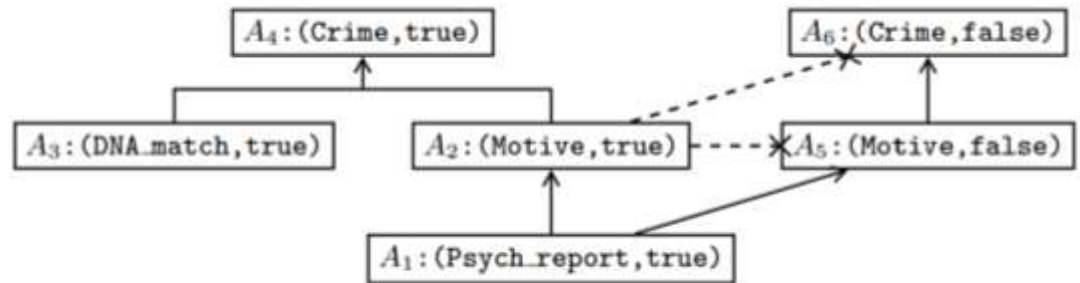


Figure 8: An argument graph resulting from our running example. Arrows show the immediate sub-argument relation. Besides the intuitively correct arguments A_1, \dots, A_4 there are two additional arguments depicted that can also be made but that are successfully rebutted by A_2 . The dashed arrows with crosshair tips show the defeat relation between arguments. Argument A_5 is defeated by A_2 because (Motive, true) is probabilistically stronger (using the likelihood ratio measure of strength in this case) than (Motive, false) based on this evidence. Any conclusion that builds on this second argument (such as A_6) is also defeated.

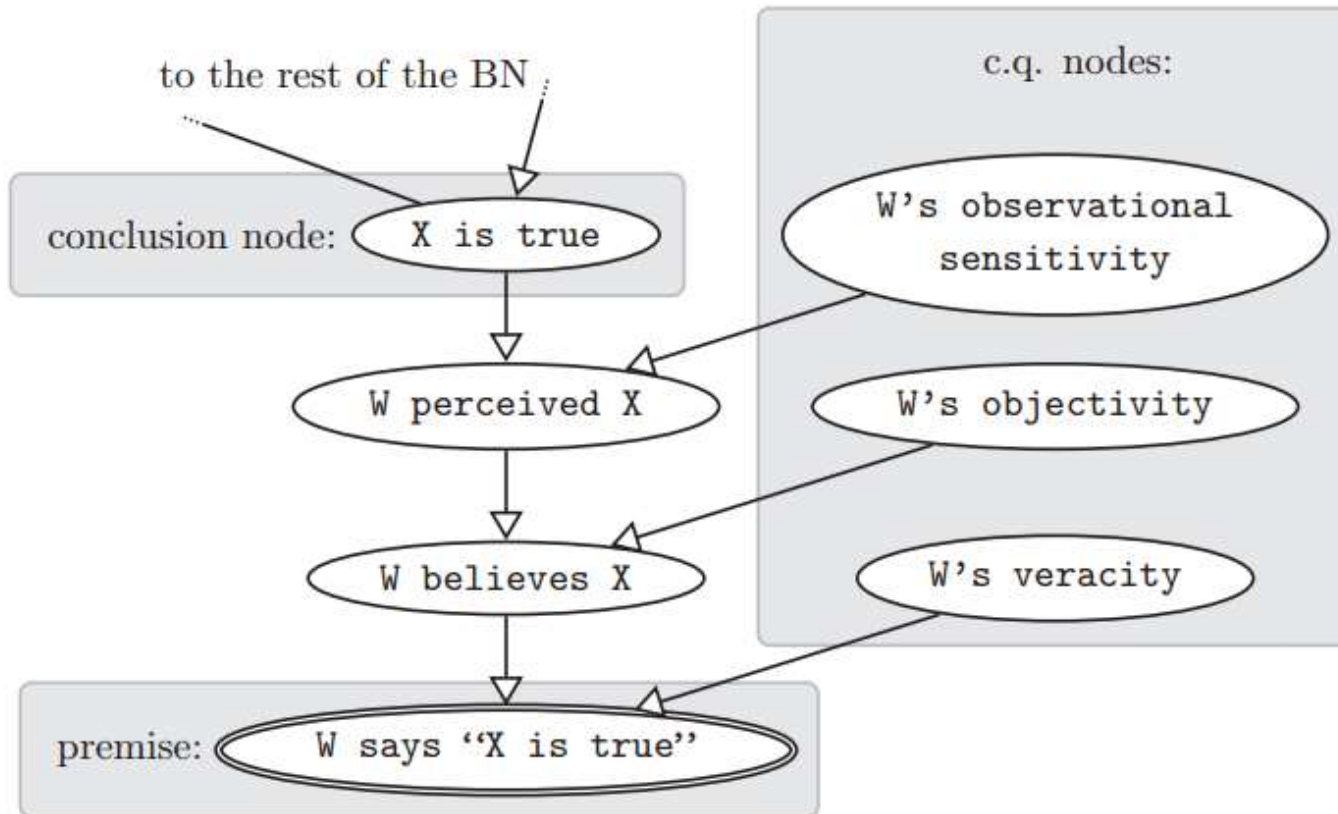


Figure 5.7: Modelling critical questions as a chain of exceptions.

NWO Forensic Science project

- A method to incorporate argument schemes in a Bayesian Network (Timmer, 2017; Timmer et al., 2015a);
- An algorithm to extract argumentative information from a Bayesian Network modeling hypotheses and evidence (Timmer, 2017; Timmer et al., 2016);
- A method to manually design a Bayesian Network incorporating hypothetical scenarios and the available evidence (Vlek, 2016; Vlek et al., 2014);
- A method to generate a structured explanatory text of a Bayesian Network modeled according to this method (Vlek, 2016; Vlek et al., 2016);
- A case study testing the design method (Vlek, 2016; Vlek et al., 2014);
- A case study testing the explanation method (Vlek, 2016).

Bayesian Network modeling with idioms

Strengths

- Explicit complex model (allows for discussion)
- Correct calculations (supported by software)
- Systematic, reusable (idioms)

Issues

- Design (numbers, dependencies, compositionality of idioms)
- Interpretation (formal versus material meaning)

Goal:

promote rational handling of evidence in courts

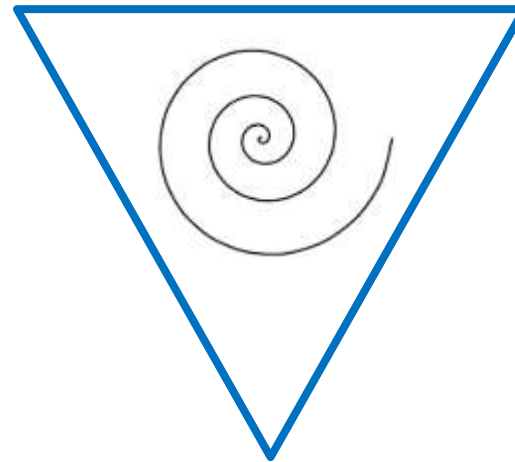
Tool needed:

a normative framework

shared between experts and factfinders



Arguments



Scenarios

Probabilities

Verheij, B. (2017). Proof With and Without Probabilities. Correct Evidential Reasoning with Presumptive Arguments, Coherent Hypotheses and Degrees of Uncertainty. *Artificial Intelligence and Law* 25 (1), 127-154.
<http://dx.doi.org/10.1007/s10506-017-9199-4>

Integrating the three perspectives

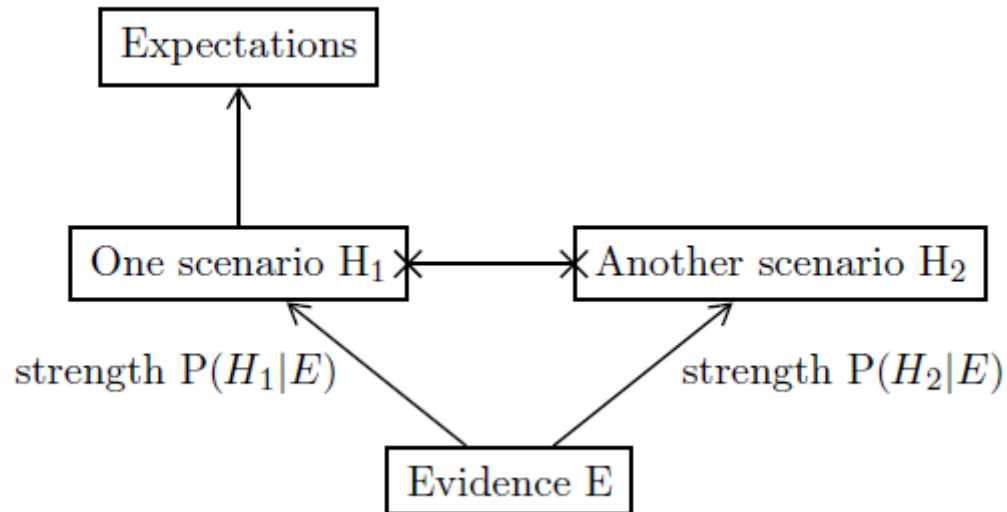
- They are just three different ways of speaking about the same things, each emphasising some specific aspects
- There is no need to idolize any
- There is no need to demonize any



Hypothesis

There exists an
integrated perspective
on
arguments, scenarios and probabilities
as
normative tools for evidential reasoning
in which each has
its **natural and transparent** place.

Arguments, scenarios and probabilities



Verheij, B. (2014). To Catch a Thief With and Without Numbers: Arguments, Scenarios and Probabilities in Evidential Reasoning. *Law, Probability and Risk*, 13, 307-325.

Definition 1. (*Case models*) A case model is a pair (C, \geq) with finite $C \subseteq L$, such that the following hold, for all φ, ψ and $\chi \in C$:

1. $\not\models \neg\varphi$;
2. If $\not\models \varphi \leftrightarrow \psi$, then $\models \neg(\varphi \wedge \psi)$;
3. If $\models \varphi \leftrightarrow \psi$, then $\varphi = \psi$;
4. $\varphi \geq \psi$ or $\psi \geq \varphi$;
5. If $\varphi \geq \psi$ and $\psi \geq \chi$, then $\varphi \geq \chi$.

Case models are 'with and without numbers' in a precise sense:

- the ordering can be derived from a numeric representation;
- it is without numbers since an ordering is a qualitative relation.

Proof With and Without Probabilities

Correct Evidential Reasoning with Presumptive Arguments, Coherent Hypotheses and Degrees of Uncertainty

Bart Verheij

Received: date / Accepted: date

Abstract Evidential reasoning is hard, and errors can lead to miscarriages of justice with serious consequences. Analytic methods for the correct handling of evidence come in different styles, typically focusing on one of three tools: arguments, scenarios or probabilities. Recent research used Bayesian Networks for connecting arguments, scenarios, and probabilities. Well-known issues with Bayesian Networks were encountered: More numbers are needed than are available, and there is a risk of misinterpretation of the graph underlying the Bayesian Network, for instance as a causal model. The formalism presented here models presumptive arguments about coherent hypotheses that are compared in terms of their strength. No choice is needed between qualitative or quantitative analytic styles, since the formalism can be interpreted with and without numbers. The formalism is applied to key concepts in argumentative, scenario and probabilistic analyses of evidential reasoning, and is illustrated with a fictional crime investigation example based on Alfred Hitchcock's film 'To Catch A Thief'.

Verheij, B. (2017). Proof With and Without Probabilities.
Correct Evidential Reasoning with Presumptive Arguments,
Coherent Hypotheses and Degrees of Uncertainty.
Artificial Intelligence and Law 25 (1), 127-154.
<http://dx.doi.org/10.1007/s10506-017-9199-4>.

**One
interpretation
of the evidence**

**Another
interpretation
of the evidence**

$$p(H_1|E)$$

$$p(H_2|E)$$

Evidence

The diagram consists of a central word 'Evidence' at the bottom. Two blue arrows originate from the top of 'Evidence'. The left arrow points upwards and to the left towards the text 'One interpretation of the evidence' and the equation $p(H_1|E)$. The right arrow points upwards and to the right towards the text 'Another interpretation of the evidence' and the equation $p(H_2|E)$.

Proof With and Without Probabilities

An argumentation theory that connects

- presumptive arguments,
- coherent hypotheses, and
- degrees of uncertainty

using classical logic and standard probability theory.

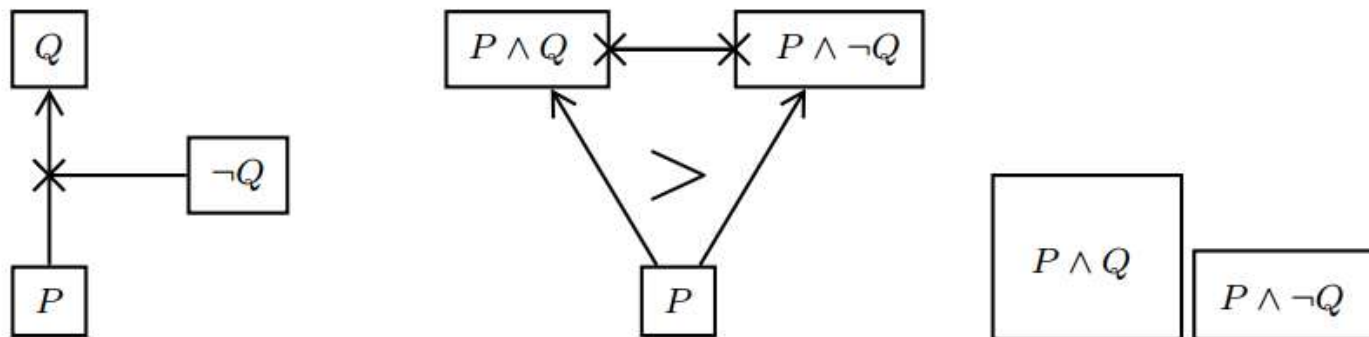


Fig. 1 General idea: an argument with a counterargument (left); arguments for conflicting cases and their comparison (middle); cases and their comparative value (right)

- Patients have reported a sexual assault by their doctor (`patients`).
- The DNA of a trace of semen found on one patient is compared with the DNA in a blood sample taken from the doctor. There is no match (`¬dna-match`).
- The doctor had implanted a drain into his arm, filled with someone else's blood (`implant`).

See https://en.wikipedia.org/wiki/DNA_profiling#Fake_DNA_evidence

By patients, we presume dna-match and guilt

patients $\sim >$ dna-match \wedge guilt

We find \neg dna-match, so now we presume \neg guilt

patients $\wedge \neg$ dna-match $\sim >$ \neg guilt

We find implant, so we presume, in fact conclude, guilt

patients $\wedge \neg$ dna-match \wedge implant $\sim >$ guilt

patients $\wedge \neg$ dna-match \wedge implant \Rightarrow guilt

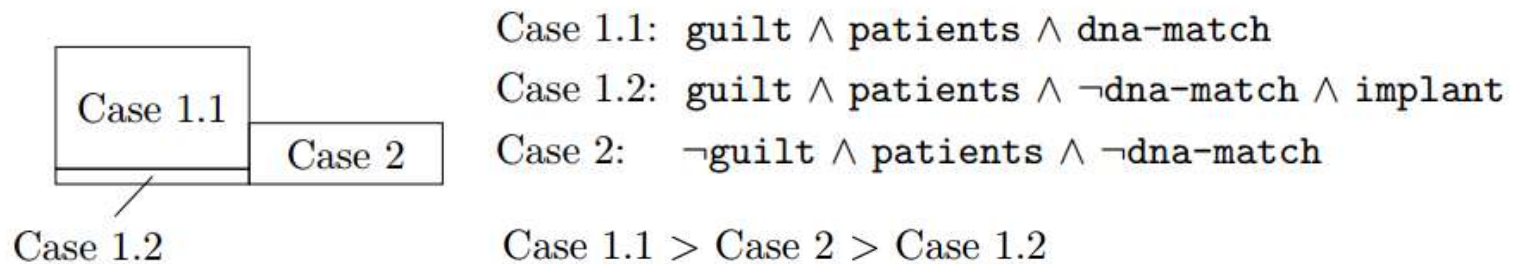


Figure 2: Case model for the example

Case 1.1 > Case 2 > Case 1.2

3 > 2 > 1

$$\Pr(\text{Case1.1}) = 3/(3+2+1) = 50\%$$

$$\Pr(\text{Case2}) = 2/(3+2+1) \sim 33\%$$

$$\Pr(\text{Case1.2}) = 1/(3+2+1) \sim 17\%$$

Case 1.1 > Case 2 > Case 1.2

$$\pi > e > 1$$

$$\Pr(\text{Case1.1}) = \frac{\pi}{\pi+e+1} \sim 46\%$$

$$\Pr(\text{Case2}) = \frac{e}{\pi+e+1} \sim 40\%$$

$$\Pr(\text{Case1.2}) = \frac{1}{\pi+e+1} \sim 14\%$$

Case 1.1 > Case 2 > Case 1.2

very high > low > extremely small

Pr(Case1.1) ~ 99%

Pr(Case2) ~ 1%

Pr(Case1.2) ~ 0.0..01%

Case 1.1 > Case 2 > Case 1.2

very high > low > extremely small

~~Pr(Case1.1) ~ 99%~~
~~Pr(Case2) ~ 1%~~
~~Pr(Case1.2) ~ 0.0..01%~~

(It seems that we don't need the numbers)

Kinds of argument validity

Coherent arguments

$(C, \geq) \models (\varphi, \psi)$ if and only if $\exists \omega \in C: \omega \models \varphi \wedge \psi$.

Conclusive arguments

$(C, \geq) \models \varphi \Rightarrow \psi$ if and only if $\exists \omega \in C: \omega \models \varphi \wedge \psi$ and $\forall \omega \in C: \text{if } \omega \models \varphi, \text{ then } \omega \models \psi$.

Presumptively valid arguments

$(C, \geq) \models \varphi \rightsquigarrow \psi$ if and only if $\exists \omega \in C:$

1. $\omega \models \varphi \wedge \psi$; and
2. $\forall \omega' \in C: \text{if } \omega' \models \varphi, \text{ then } \omega \geq \omega'$.

Arguments

Scenarios

Probabilities

Coherence

Coherence

$p > 0$

Presumptive validity

Plausible

p maximal, $p > t$

Conclusive

Beyond a
reasonable
doubt

$p = 1$

Three kinds of validity

Coherent arguments

$$p(\psi \mid \varphi) > 0$$

$(C, \geq) \models (\varphi, \psi)$ if and only if $\exists \omega \in C: \omega \models \varphi \wedge \psi$.

Presumptive arguments

$$p(\psi \mid \varphi) > t$$

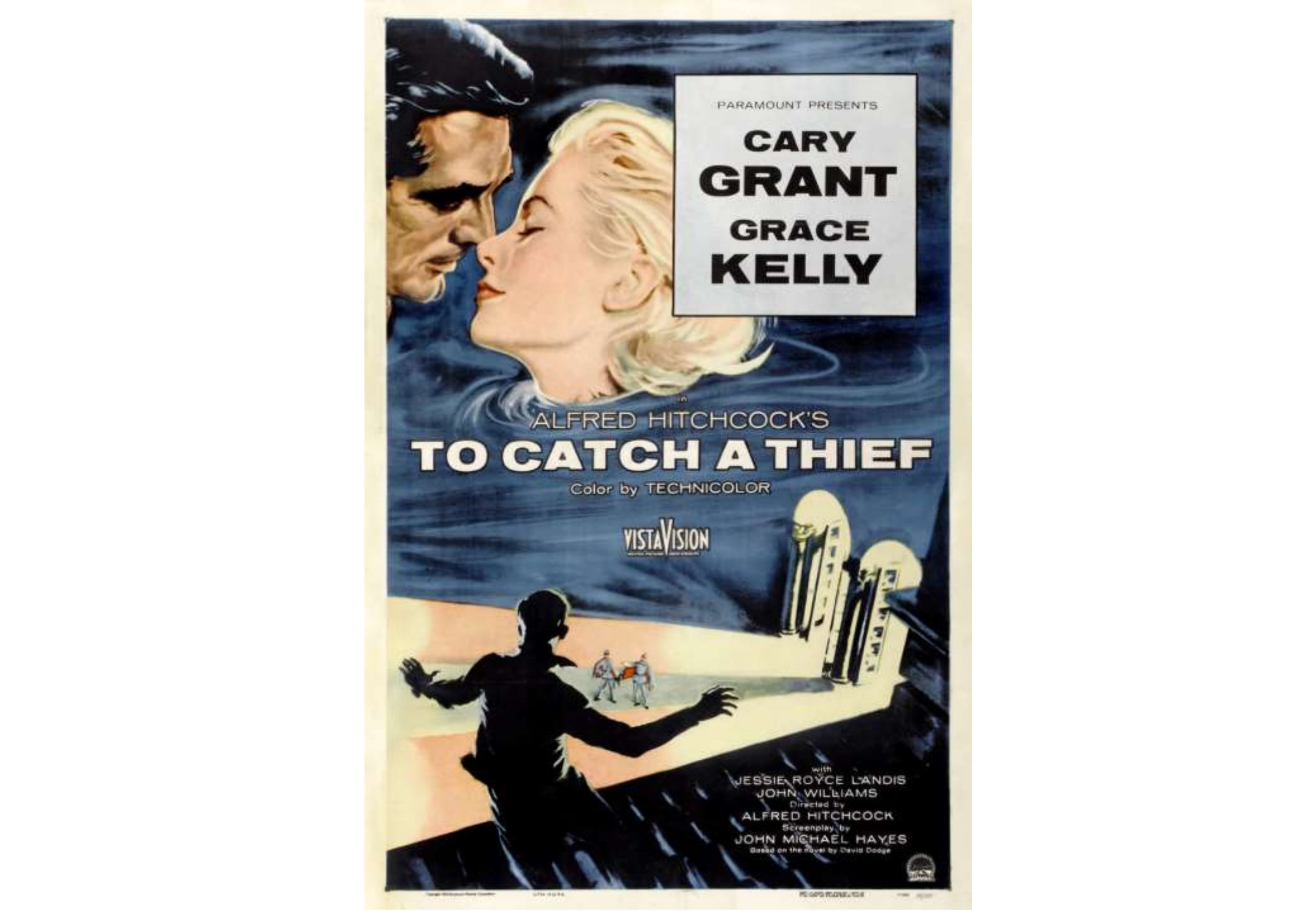
$(C, \geq) \models \varphi \rightsquigarrow \psi$ if and only if $\exists \omega \in C$:

1. $\omega \models \varphi \wedge \psi$; and
2. $\forall \omega' \in C$: if $\omega' \models \varphi$, then $\omega \geq \omega'$.

Conclusive arguments

$$p(\psi \mid \varphi) = 1$$

$(C, \geq) \models \varphi \Rightarrow \psi$ if and only if $\exists \omega \in C: \omega \models \varphi \wedge \psi$ and $\forall \omega \in C$: if $\omega \models \varphi$, then $\omega \models \varphi \wedge \psi$.



PARAMOUNT PRESENTS

**CARY
GRANT**
**GRACE
KELLY**

in
ALFRED HITCHCOCK'S
TO CATCH A THIEF

Color by TECHNICOLOR

VISTAVISION

with
JESSIE ROYCE L'ANDIS
JOHN WILLIAMS
Directed by
ALFRED HITCHCOCK
Screenplay by
JOHN MICHAEL HAYES

Based on the novel by David Dodge







↓ Evidence

Hypotheses

resemblance	robie	\neg robie		
escape				
fight		foussard	\neg foussard	
prosthesis				
arrest			d.	\neg d.
confession			j.	\neg j.
finding				

Hypothesis 1

Hypothesis 2

Hyp. 3

Hyp. 4

d. = daughter

j. = jewelry



Block 1: Robie indeed was the thief

Block 3: Resistance friend Foussard's daughter was the thief

↓ Evidence

Hypotheses

resemblance	robie	\neg robie		
escape				
fight		foussard	\neg foussard	
prosthesis				
arrest			d.	\neg d.
confession			j.	\neg j.
finding				

Hypothesis 1

Hypothesis 2

Hyp. 3

Hyp. 4

d. = daughter

j. = jewelry



robie



resemblance \wedge escape

coherent, presumptive, not conclusive

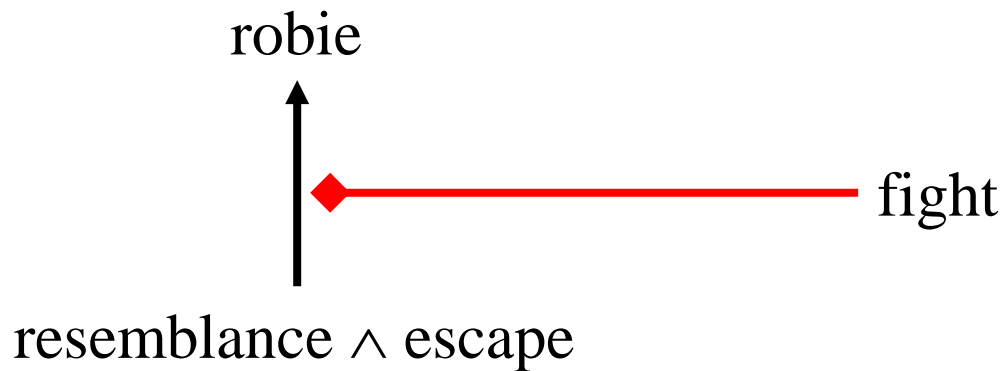
\neg robie



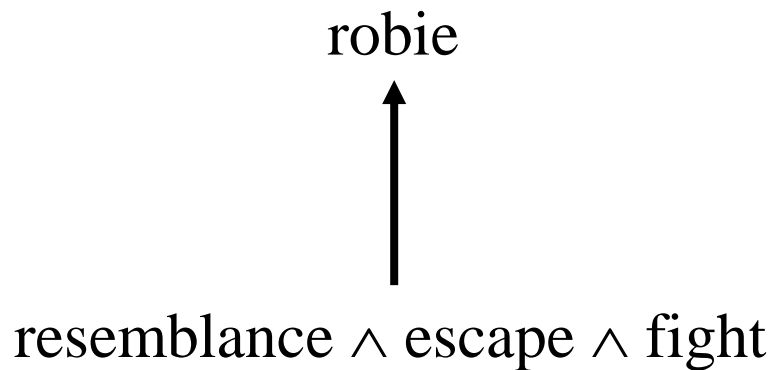
resemblance \wedge escape

coherent, not presumptive, not conclusive

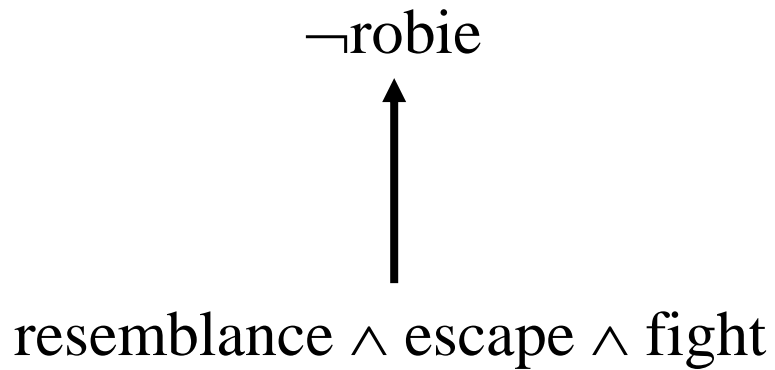




defeating,
rebutting



not coherent,
not presumptive,
not conclusive



coherent,
presumptive,
conclusive

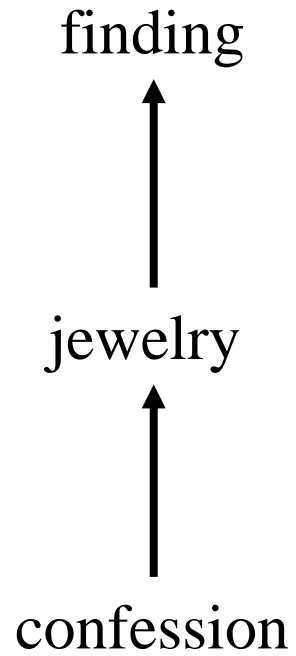


\neg robie \wedge \neg fousard \wedge daughter \wedge jewelry



resemblance \wedge escape \wedge fight \wedge prosthesis \wedge arrest \wedge confession \wedge finding

coherent, presumptive, conclusive



coherent, presumptive, not conclusive

coherent, presumptive, not conclusive

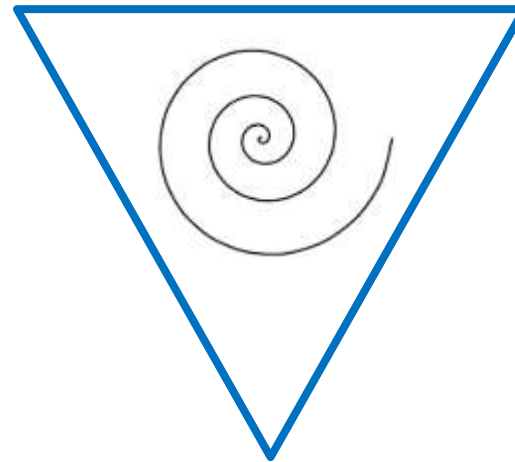


Hypothesis

There exists an
integrated perspective
on
arguments, scenarios and probabilities
as
normative tools for evidential reasoning
in which each has
its **natural and transparent** place.



Arguments



Scenarios

Probabilities

Verheij, B. (2017). Proof With and Without Probabilities. Correct Evidential Reasoning with Presumptive Arguments, Coherent Hypotheses and Degrees of Uncertainty. *Artificial Intelligence and Law* 25 (1), 127-154.
<http://dx.doi.org/10.1007/s10506-017-9199-4>

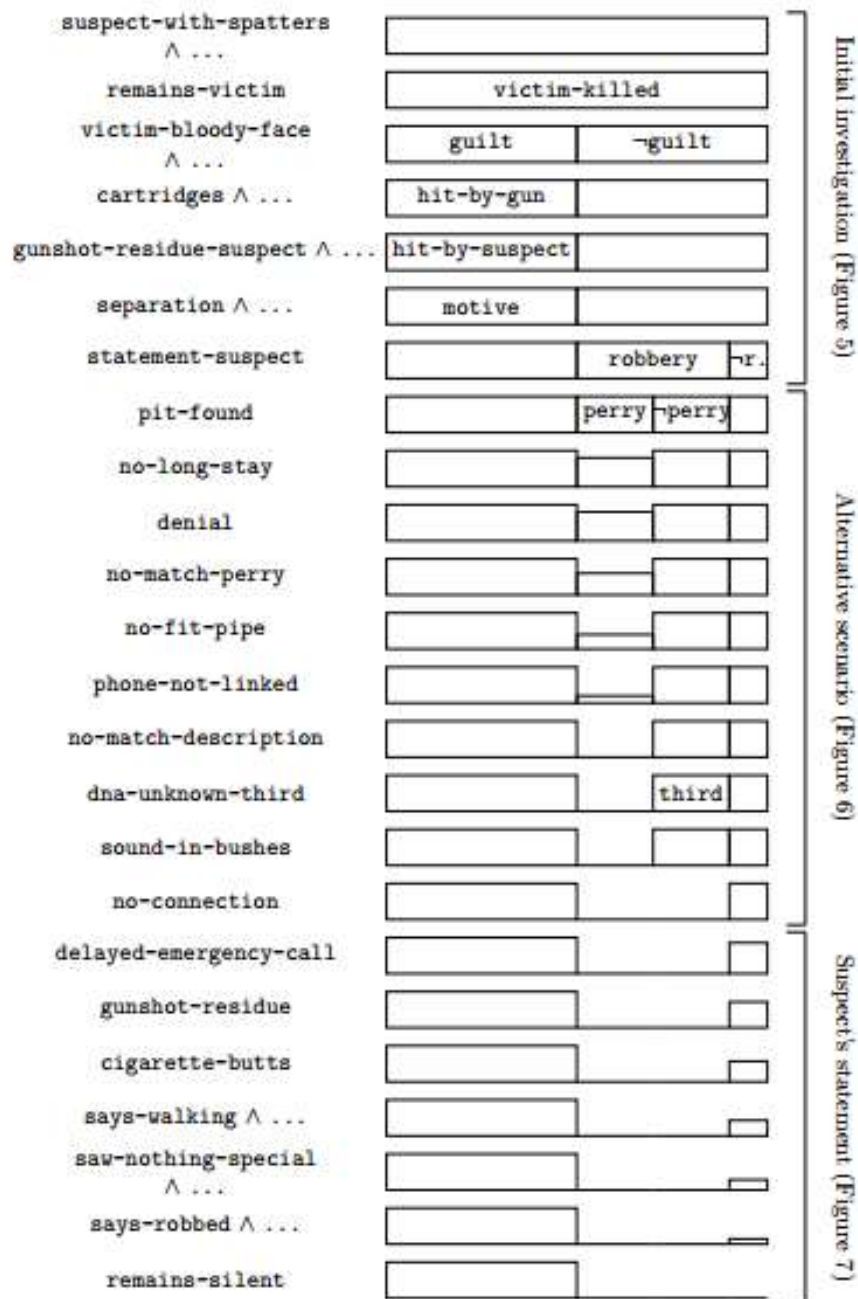


Figure 10: The Appellate Court's reasoning

Introduction

Hybrid models

AI & Law

Artificial intelligence and Law

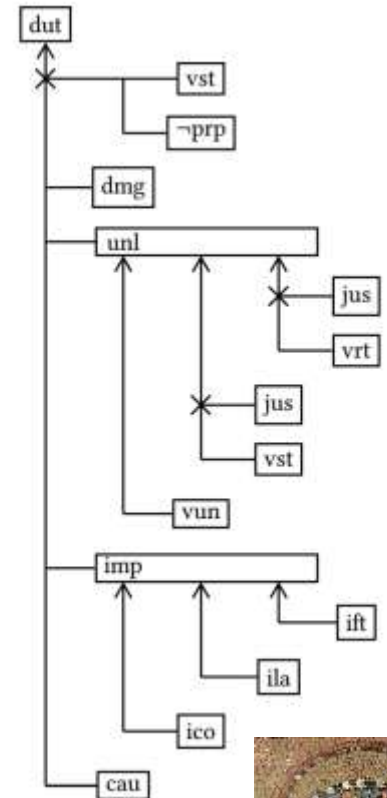
法律人工智能

Legal artificial intelligence

Artificial Intelligence and Law

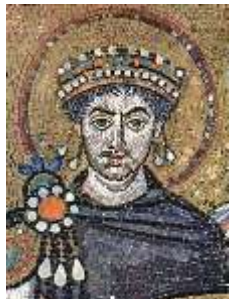
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
-dmg	-dut	-dut	-dut	dut	dut	dut	dut	dut	dut	dut	dut	dut	-dut	-dut	-dut
	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg	dmg
	-unl	unl	unl	unl	unl	unl	unl	unl	unl	unl	unl	unl	-unl	-unl	unl
		-imp	imp	imp	imp	imp	imp	imp	imp	imp	imp	imp			imp
			-cau	cau	cau	cau	cau	cau	cau	cau	cau	cau			cau
	-vrt			vrt	vrt	vrt	-vrt	-vrt	-vrt	-vrt	-vrt	-vrt	vrt	-vrt	
	-vst			-vst	-vst	-vst	vst	vst	vst	-vst	-vst	-vst	-vst	vst	vst
	-vun			-vun	-vun	-vun	-vun	-vun	-vun	vun	vun	vun			
		-ift		ift	-ift	-ift	ift	-ift	-ift	ift	-ift	-ift			
		-ila		-ila	ila	-ila	-ila	ila	-ila	-ila	ila	-ila			
		-ico		-ico	-ico	ico	-ico	-ico	ico	-ico	-ico	ico			
				-jus	-jus	-jus	-jus	-jus	-jus	-jus	-jus	-jus	jus	jus	
				prp	prp	prp									-prp

1 > 2 > 3 > 4 > 5 ~ 6 ~ 7 ~ 8 ~ 9 ~ 10 ~ 11 ~ 12 ~ 13 > 14 ~ 15 ~ 16



Data

Knowledge



The two faces of Artificial Intelligence

Expert systems
Business rules
Open data
IBM's Deep Blue
Complex structure

Knowledge

Formal

Explainability

Adaptive systems
Machine learning
Big data
Natural language processing
Adaptive structure

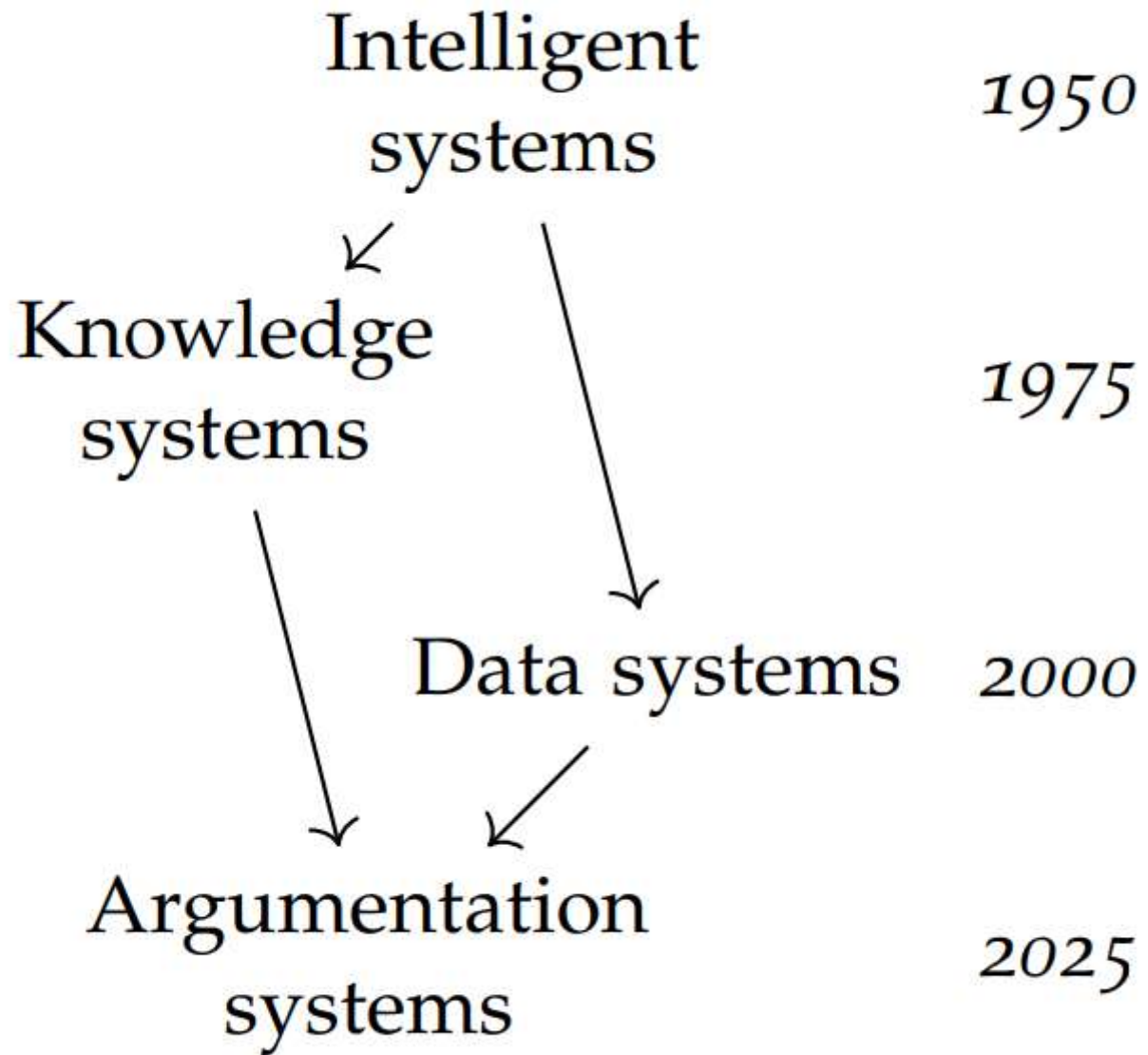
Data tech

Foundation:

probability theory

Scalability

Argumentation technology





Readings

- Verheij, B. (2018). *Arguments for Good Artificial Intelligence*. Groningen: University of Groningen. www.ai.rug.nl/~verheij/oratie/
- Verheij, B. (2017a). Proof With and Without Probabilities. Correct Evidential Reasoning with Presumptive Arguments, Coherent Hypotheses and Degrees of Uncertainty. *Artificial Intelligence and Law* 25 (1), 127-154. www.ai.rug.nl/~verheij/publications/ail2017.htm
- Verheij, B., Bex, F. J., Timmer, S., Vlek, C., Meyer, J. J., Renooij, S., & Prakken, H. (2016). Arguments, Scenarios and Probabilities: Connections Between Three Normative Frameworks for Evidential Reasoning. *Law, Probability & Risk* 15, 35-70. www.ai.rug.nl/~verheij/publications/lpr2016.htm