

Automating the Modelling of Transformative Artificial Intelligence Risks

"An Epistemic Framework for Leveraging Frontier AI Systems to Upscale Conditional Policy Assessments in Bayesian Networks on a Narrow Path towards Existencial Safety"

A thesis submitted at the Department of Philosophy for the degree of $Master\ of\ Arts\ in\ Philosophy\ \ \ \ Economics$

Author: Supervisor:

Valentin Jakob Meyer Valentin.meyer@uni-bayreuth.de Matriculation Number: 1828610 Tel.: +49 (1573) 4512494 Pielmühler Straße 15

Pielmühler Straße 15 52066 Lappersdorf Word Count:
30.000
Source / Identifier:
Document URL

Dr. Timo Speith

Contents

Preface					
\mathbf{A}	Abstract				
O	Outline(s): Table of Contents				
1	\mathbf{Intr}	oduction	7		
	1.1	Abstract	7		
	1.2	Motivation: Problem Statement	7		
	1.3	Motivation: Research Question	7		
	1.4	Scope: Aim & Context of the Research	7		
	1.5	Significance of the Research: Theory of Change	7		
	1.6	Thesis Statement & Position: (Aim of the Paper)	8		
	1.7	Overview: Structure & Approach of the Paper (Roadmap — Theory of			
		Change)	8		
	1.8	Table of Contents	8		
	1.9	Problem Statement — Motivation	8		
	1.10	Aim of the Paper	9		
		1.10.1 Research Question & Scope	9		
		1.10.2 Significance of the Research	9		
	1.11	Theory of Change — Approach & Structure of the Paper	9		
		1.11.1 The Coordination Crisis in AI Governance	10		
		1.11.2 1.2 Research Question and Scope	10		
		1.11.3 1.3 The Multiplicative Benefits Framework	10		
		1.11.4 1.4 From Theory to Practice: A Roadmap	10		
	1.12	Overview / Table of Contents	10		
2		text	11		
	2.1	Background Considerations			
	2.2	Literature, Concepts & Terminology			
		2.2.1 DAG / BayesNets			
		2.2.2 State of the art (MTAIR) — Explanation	12		
	2.3	Theoretical Foundations			
		2.3.1 AI Existential Risk: The Carlsmith Model			
		2.3.2 The Epistemic Challenge of Policy Evaluation	12		
		2.3.3 Argument Mapping and Formal Representations	12		
		2.3.4 Bayesian Networks as Knowledge Representation	12		
		2.3.5 The MTAIR Framework: Achievements and Limitations	12		
		2.3.6 "A Narrow Path": Conditional Policy Proposals in Practice	12		

		2.3.7 (Intro) Example — Rain/Sprinkler/Lawn	12
	2.4	Methodology	13
		2.4.1 From World Models to Computational Representation	13
	2.5	1	13
		2.5.1 Rain-Sprinkler-Grass DAG	13
3	\mathbf{AM}	TAIR 1	L 5
	3.1		15
	3.2		15
			15
		· · · · · · · · · · · · · · · · · · ·	16
			16
			18
	3.3	Inference & Extensions	19
		3.3.1 Probabilistic Inference Engine	19
		3.3.2 Policy Evaluation Interface	19
		3.3.3 Prediction Market Integration Module	19
	3.4	Results & Analysis: From Theory to Application	19
			19
		3.4.2 Computational Performance Analysis	19
			19
		3.4.4 Comparative Analysis of AI Governance Worldviews	19
		3.4.5 Policy Impact Evaluation: Proof of Concept	19
	3.5	Insights & Findings	19
		3.5.1 Automated Modeling Pipeline — From Academic Papers to Polit-	
		ical Strategy	19
		3.5.2 Project Scaling — Workflow Pipeline & Automation	19
		3.5.3 Computational Complexity — Computational Tractability 2	20
		3.5.4 External Validation — Manual Extraction & Processing 2	21
4	Disc	cussion 2	22
5		0	23
	5.1		23
	5.2	Implications & Impact — Uptake, Feedback Loops, Uptake & Success –	30
	- 0		23
	5.3	Known Unknowns & Unknown Unknowns — Input Data Example: Mod-	
	_ ,	0 1	24
	5.4	Discussion: Implications and Limitations	24
6	Con	nclusion	25
	6.1	The Current State of Things & How to Continue	25
	6.2		25
		6.2.1 Assessing Policy Effects:	25
		6.2.2 Conditional Probability:	25
		6.2.3 Methodology:	25
		6.2.4 Purpose:	25
	6.3	Outlook — Outlook & Next Steps / Further Research	26
		6.3.1 Scaling Up:	26

		6.3.2	Collaboration:	26
		6.3.3	Technological Enhancements:	26
		6.3.4	Potential Impact:	26
		6.3.5	Limitations of the Analysis	26
		6.3.6	Policy Implications & Recommendations	26
		6.3.7	Areas for Future Research	26
		6.3.8	Open Questions — Central/Remaining Questions & Feedback $ \ldots $	26
		6.3.9	Outlook — Outlook & Next Steps / Further Research	27
	6.4	Conclu	usion: Toward an Adaptive AI Governance Framework	27
		6.4.1	Key Contributions and Findings	27
		6.4.2	Limitations of the Current Implementation	27
		6.4.3	Policy Implications and Recommendations	27
		6.4.4	Future Research Directions	27
		6.4.5	Concluding Reflections	27
_				•
ŀr	ontm	atter		28
		Ackno	owledgments	28
Рr	efato	ory An	paratus: Illustrations and Terminology — Quick References	29
			es	29
			phics & Figures	30
		_	reviations	30
				31
			er requirements"	31
				31
			Anything that makes it easier to understand	31
	(,	, 0	
7	Qua	rto Sy	ntax	33
	7.1	Headir	ngs & Potential Headings	33
ъ.		,		0.0
Вı	bliog	raphy	(References)	36
A	ppei	ndices	3	37
	PP o.			•
A	App	endice	es	37
	A.1	Appen	dix A: Technical Implementation Details	37
	A.2	Appen	dix B: Model Validation Procedures	37
	A.3	Appen	dix C: Case Studies	37
	A.4	Appen	dix D: Ethical Considerations	37
_		7. 4		00
ĸ	ann	$\operatorname{endix} A$	\	38

Preface

This is a Quarto book.

To learn more about Quarto books visit https://quarto.org/docs/books.

Abstract

Outline(s): Table of Contents

Introduction

10% of Grade: ~ 14% of text ~ 4200 words ~ 10 pages

- introduces and motivates the core question or problem
- provides context for discussion (places issue within a larger debate or sphere of
- states precise thesis or position the author will argue for
- provides roadmap indicating structure and key content points of the essay

1.1 Abstract

[x] introduces and motivates the core question or problem

1.2 Motivation: Problem Statement

AIs are (on the path to) getting smarter than humans Humanities (de-centralized) collective information processing, sense-making AND c BUT new technology/AI also enables new modes/methods of collective information pr therefore this thesis/AMTAIR aspires to be an epistemic tool that enables/empower

1.3 Motivation: Research Question

- [] provides context for discussion (places issue within a larger debate or sphere of relevance)

1.4 Scope: Aim & Context of the Research

1.5 Significance of the Research: Theory of Change

-[] states precise thesis or position the author will argue for

1.6 Thesis Statement & Position: (Aim of the Paper)

AMTAIR is a new epistemic tool/framework that enables/empowers humanities collective information processing, comprehension and thereby enables the necessary coordination ... (to compete with AIs?)

[] provides roadmap indicating structure and key content points of the essay

1.7 Overview: Structure & Approach of the Paper (Roadmap — Theory of Change)

1.8 Table of Contents

1.9 Problem Statement — Motivation

Continued AI Progress:

• Rapid advancements in AI technology increase both potential benefits and risks.

Existential Risks (AI X-Risk):

• Advanced AI systems could pose significant threats if misaligned with human values.

Complexity Challenges:

• The intricate nature of AI systems complicates policy formulation and understanding.

Limitations of Current Approaches:

- MTAIR's Reliance on Human Labor:
 - Modeling Transformative AI Risks (MTAIR) is constrained by manual cognitive efforts.
- Need for Automation:
 - Scaling and automating risk modeling is essential to keep pace with AI developments.

Opportunity:

• Leveraging new technologies to enhance our ability to model and mitigate AI risks.

1.10 Aim of the Paper

1.10.1 Research Question & Scope

Can frontier AI technologies be utilized to automate the modeling of transformative AI risks, so as to allow for the prediction of policy impacts?

Frontier AI Technology: Today's most capable AI systems (e.g. GPT4 level LLMs)

Scaling Up: Automating the previously "manual" cognitive labor

Modeling: Formalizing the world views underlying arguments

Transformative AI: Level of AI capabilities defined by severe impact on the world

Safety & Governance Literature: Publications, reports etc. concerned with risks from AI Automated Estimation: Non-manual (AI systems + scaffolding), quantified evaluations

Probability Distributions: Formal expressions of the expectations over future worlds Conditional Trees of Possible Worlds: "If ... then..." reasoning over ways things may play out

Forecasting Policy Impacts: Qualitative & quantitative evaluation of expected outcomes

1.10.2 Significance of the Research

1.11 Theory of Change — Approach & Structure of the Paper

Multiplicative Benefits:

• Automation × Live Prediction Market Integrations × Policy Impact Evaluations

Explanation:

Automation:

- Increases efficiency and scalability of risk modeling.
 - Live Prediction Markets:
- Provides up-to-date, collective intelligence to inform models.
 - Policy Impact Evaluations:
- Improves the accuracy and relevance of policy assessments.

Outcome:

• Enhanced ability to develop effective policies that mitigate AI risks.

Visual Aid:

• A diagram illustrating how each component amplifies the others, leading to greater overall impact.

- 1.11.1 The Coordination Crisis in AI Governance
- 1.11.2 1.2 Research Question and Scope
- 1.11.3 1.3 The Multiplicative Benefits Framework
- 1.11.4 1.4 From Theory to Practice: A Roadmap
- 1.12 Overview / Table of Contents

Context

```
### 20% of Grade: \sim 29% of text \sim 8700 words \sim 20 pages
```

- demonstrates understanding of all relevant core concepts
- explains why the question/thesis/problem is relevant in student's own words (suppor
- situates it within the debate/course material
- reconstructs selected arguments and identifies relevant assumptions
- describes additional relevant material that has been consulted and integrates it wi
 - 1. successively (chunk my chunk) introduce concepts/ideas and 2. ground each with existing literature

- 2.1 Background Considerations
- 2.2 Literature, Concepts & Terminology
- 2.2.1 DAG / BayesNets
- 2.2.2 State of the art (MTAIR) Explanation Carlsmith Model (Analytica)
- 2.3 Theoretical Foundations
- 2.3.1 AI Existential Risk: The Carlsmith Model
- 2.3.2 The Epistemic Challenge of Policy Evaluation
- 2.3.3 Argument Mapping and Formal Representations
- 2.3.4 Bayesian Networks as Knowledge Representation
- 2.3.5 The MTAIR Framework: Achievements and Limitations
- 2.3.6 "A Narrow Path": Conditional Policy Proposals in Practice
- 2.3.7 (Intro) Example Rain/Sprinkler/Lawn

/ Rain/Sprinkler/Lawn DAG / BayesNet — Extended Example

•••

Own Position/Argument: AMTAIR ... Own Rain/Sprinkler/Lawn DAG / BayesNet Implementation

2.4 Methodology

2.4.1 From World Models to Computational Representation

Research Design Overview

Formalizing World Models from AI Safety Literature

Directed Acyclic Graphs: Structure and Semantics

Quantification Approaches for Probabilistic Judgments

Inference Techniques for Complex Networks

Integration with Prediction Markets and Forecasting Platforms

2.5 Practical Real World Examples

MTAIR / Carlsmith Model (Analytica) — Explanation (— is motivation: should come first)

Kialo

BayeServer

BayesNet — Extended Example

2.5.1 Rain-Sprinkler-Grass DAG

Code + documentation

Testing crossreferencing grapics Figure 7.1.

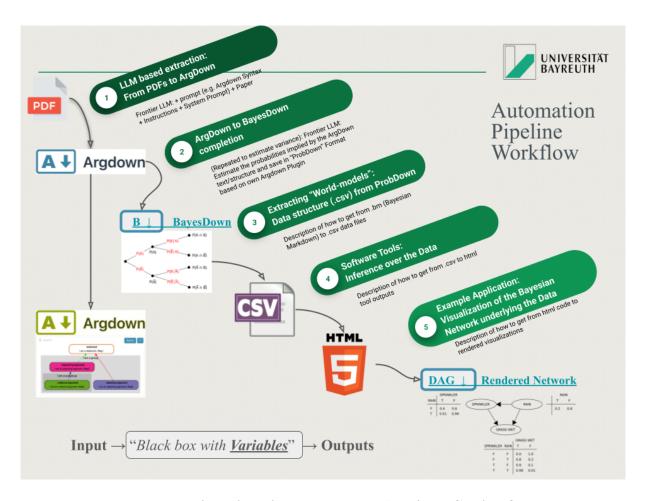


Figure 2.1: AMTAIR Automation Pipeline from CITATION

AMTAIR

20% of Grade: \sim 29% of text \sim 8700 words \sim 20 pages

- provides critical or constructive evaluation of positions introduced
- develops strong (plausible) argument in support of author's own position/thesis
- argument draws on relevant course material claim/argument
- demonstrate understanding of the course materials incl. key arguments and core conc
- claim/argument is original or insightful, possibly even presents an original contri

3.1 Rain-Sprinkler-Grass Example

3.2 Software Implementation

3.2.1 System Architecture (and Data Flow) Overview

AMTAIR Prototype Demonstration (Public Colab Notebook)

AMTAIR Prototype: Automating Transformative AI Risk Modeling

Executive Summary

Purpose Within the Master's Thesis

Relevance to AI Governance

Notebook Structure and Workflow

Project Context and Purpose

Notebook Overview and Pipeline

Connection to Master's Thesis

Instructions — How to use this notebook:

Key Concepts:

Example Workflow:

Troubleshooting:

Environment Setup and Data Access

- 0.1 Prepare Colab/Python Environment Import Libraries & Packages
 - 0.2 Connect to GitHub Repository
 - 0.3 File Import

3.2.2 Automated Extraction Pipeline:

3.2.3 The Two-Stage Extraction from Carlsmith

- **Process overview:** Explain the separation of structure and probability extraction
- **Stage 1: Structure extraction**
- **Process details:** Outline the steps for extracting argument structure
- **CODE EXAMPLE: ** Show key function for ArgDown parsing
- **Visualization:** Demonstrate structural extraction for Carlsmith model
- **Stage 2: Probability integration**
- **Process details:** Explain how probability information is incorporated
- **Question generation:** Show how appropriate questions are derived from structure
- **CODE EXAMPLE: ** Show key function for BayesDown enhancement
- **Visualization:** Demonstrate probability extraction for Carlsmith model
 - 1.0 Sources (PDF's of Papers) to ArgDown (.md file)

Sources to ArgDown: Structured Argument Extraction

Process Overview

What is ArgDown?

- 1.1 Specify Source Document (e.g. PDF)
- 1.2 Generate ArgDown Extraction Prompt

- 1.3 Prepare LLM API Call
- 1.4 Make ArgDown Extraction LLM API Call
- 1.5 Save ArgDown Extraction Response
- 1.6 Review and Check ArgDown.md File
- 1.6.2 Check the Graph Structure with the ArgDown Sandbox Online
- 1.7 Extract ArgDown Graph Information as DataFrame
- 1.8 Store ArgDown Information as 'ArgDown.csv' file
- 2.0 Probability Extractions: Arg Down (.csv) to Bayes Down (.md + plugin JSON syntax)

ArgDown to BayesDown: Adding Probability Information

Process Overview

What is BayesDown?

- 2.1 Probability Extraction Questions 'ArgDown.csv' to 'ArgDown_WithQuestions.csv'
- 2.2 'ArgDown_WithQuestions.csv' to 'BayesDownQuestions.md'
- 2.3 Generate BayesDown Probability Extraction Prompt
- 2.3.1 BayesDown Format Specification

Core Structure

Rain-Sprinkler-Lawn Example

- 2.4 Prepare 2nd API call
- 2.5 Make BayesDown Probability Extraction API Call
- 2.6 Save BayesDown with Probability Estimates (.csv)
- 2.7 Review & Verify BayesDown Probability Estimates
- 2.7.2 Check the Graph Structure with the ArgDown Sandbox Online
- 2.8 Extract BayesDown with Probability Estimates as Dataframe
- 3.0 Data Extraction: BayesDown (.md) to Database (.csv)

BayesDown to Structured Data: Network Construction

Extraction Pipeline Overview

Theoretical Foundation

Role in Thesis Research

- 3.1 ExtractBayesDown-Data_v1
- 3.1.2 Test BayesDown Extraction
- 3.1.2.2 Check the Graph Structure with the ArgDown Sandbox Online
- 3.3 Extraction
 - 3.3 Data-Post-Processing
 - 3.4 Download and save finished data frame as .csv file

3.2.4 Network Construction and Visualization

4.0 Analysis & Inference: Bayesian Network Visualization

Bayesian Network Visualization Approach

Visualization Philosophy

Connection to AMTAIR Goals

Implementation Structure

Phase 1: Dependencies/Functions

Phase 2: Node Classification and Styling Module

Phase 3: HTML Content Generation Module

Phase 4: Main Visualization Function

Quickly check HTML Outputs

Conclusion: From Prototype to Production

Summary of Achievements

Limitations and Future Work

Connection to AMTAIR Project

6.0 Save Outputs

Saving and Exporting Results

Convert .ipynb Notebook to MarkDown

3.3 Inference & Extensions

- 3.3.1 Probabilistic Inference Engine
- 3.3.2 Policy Evaluation Interface
- 3.3.3 Prediction Market Integration Module

3.4 Results & Analysis: From Theory to Application

- 3.4.1 Extraction Quality Assessment
- 3.4.2 Computational Performance Analysis
- 3.4.3 Case Study: The Carlsmith Model Formalized
- 3.4.4 Comparative Analysis of AI Governance Worldviews
- 3.4.5 Policy Impact Evaluation: Proof of Concept

3.5 Insights & Findings

3.5.1 Automated Modeling Pipeline — From Academic Papers to Political Strategy

Success of Automation:

- Demonstrated feasibility of automated model extraction. Improved Forecasting:
- Enhanced accuracy with real-time data integration. Policy Analysis:
- Identified impactful policies through conditional forecasting. Scalability Achieved:
- Efficient processing of extensive data sets. Addressed Challenges:
- Overcame limitations of manual modeling.

3.5.2 Project Scaling — Workflow Pipeline & Automation

Scaling Opportunities:

- Horizontal: Incorporate more data sources.
- Vertical: Add detailed variables.

New Capabilities:

- Advanced analytics.
- Real-time data integration.

Requirements:

- Software Setup: Robust infrastructure.
- Financial: Funding for APIs and compute resources.

Impact:

- Broader, more comprehensive models.
- Enhanced policy analysis.

${\bf 3.5.3 \quad Computational \ Complexity -- Computational \ Tractability}$

Challenges:

• High computational demands of complex models.

Solutions:

- Clustering Worldviews:
- Group similar perspectives to simplify models.
- Correlation Management:
- Adjust for variable interdependencies.
- Efficient Algorithms:
 Use optimized sampling methods like Monte Carlo.

Outcome:

• Achieved efficiency without sacrificing accuracy.

Link to Theory of Change:

• Scalability amplifies policy impact.

3.5.4 External Validation — Manual Extraction & Processing

Purpose:

- Assess accuracy of automated methods. Comparison:
- Automation Strengths:
- Speed, consistency.
- Human Strengths:
- Nuanced understanding. Findings:
- Automation excels in data handling.
- Human oversight enhances quality. Conclusion:
- Optimal results from combining AI with expert input.

Discussion

```
### 10% of Grade: ~ 14% of text ~ 4200 words ~ 10 pages
```

- discusses a specific objection to student's own argument
- provides a convincing reply that bolsters or refines the main argument
- relates to or extends beyond materials/arguments covered in class

Discussion — Exchange, Controversy & Influence

5.1 Challenges & Problems — Red Teaming Problems, Failures & Downsides

Potential Failures:

- Data Issues: Inaccurate or biased inputs.
- Model Limitations: Oversimplifications.
- Tech Risks: AI misinterpretations. Red Teaming:
- Stress-testing models to find weaknesses. Impact on Theory of Change:
- Identifying points of failure strengthens the approach.

5.2 Implications & Impact — Uptake, Feedback Loops, Uptake & Success – Green Teaming –

Potential Outcomes:

- First-Order: Reduced AI risks through better policies.
- Second-Order: Enhanced collaboration.
- Third-Order: Framework applied to other global risks. Feedback Loops:

- Continuous model improvement.
- Adaptive policy-making. Green Teaming:
- Strategies to maximize positive impacts.

5.3 Known Unknowns & Unknown Unknowns — Input Data Example: Modeling Author Worldviews from Bibliographies Instead of Individual Papers

Potential Outcomes:

- First-Order: Reduced AI risks through better policies.
- Second-Order: Enhanced collaboration.
- Third-Order: Framework applied to other global risks. Feedback Loops:
- Continuous model improvement.
- Adaptive policy-making. Green Teaming:
- Strategies to maximize positive impacts.

5.4 Discussion: Implications and Limitations

Red-Teaming Results: Identifying Failure Modes
Enhancing Epistemic Security in AI Governance
Scaling Challenges and Opportunities
Integration with Existing Governance Frameworks
Known Unknowns and Deep Uncertainties

Conclusion

6.1 The Current State of Things & How to Continue

```
### 10% of Grade: ~ 14% of text ~ 4200 words ~ 10 pages
```

- summarizes thesis and line of argument
- outlines possible implications
- notes outstanding issues / limitations of discussion
- points to avenues for further research
- overall conclusion is in line with introduction

6.2 Summary — Key Takeaways & Findings

6.2.1 Assessing Policy Effects:

Evaluating how different policies alter P(Doom).

6.2.2 Conditional Probability:

Calculating P(Doom | Policy Alpha).

6.2.3 Methodology:

Update model parameters based on policy implementation. Recompute probabilities accordingly.

6.2.4 Purpose:

Inform policymakers of potential policy effectiveness.

Prioritize interventions that significantly reduce risks.

6.3 Outlook — Outlook & Next Steps / Further Research

6.3.1 Scaling Up:

• Include more variables and data sources.

6.3.2 Collaboration:

• Partner with policymakers and researchers.

6.3.3 Technological Enhancements:

• Employ advanced AI techniques.

6.3.4 Potential Impact:

• Influence global AI governance.

6.3.5 Limitations of the Analysis

6.3.6 Policy Implications & Recommendations

6.3.7 Areas for Future Research

6.3.8 Open Questions — Central/Remaining Questions & Feedback

Questions:

- How can we improve automation accuracy?
- What challenges exist in policy implementation?
- How do we mitigate AI model biases?
- How can interdisciplinary efforts enhance outcomes?

Feedback:

• Invite thoughts, critiques, and suggestions.

- 6.3.9 Outlook Outlook & Next Steps / Further Research
- 6.4 Conclusion: Toward an Adaptive AI Governance Framework
- 6.4.1 Key Contributions and Findings
- 6.4.2 Limitations of the Current Implementation
- 6.4.3 Policy Implications and Recommendations
- 6.4.4 Future Research Directions
- 6.4.5 Concluding Reflections

Frontmatter

Acknowledgments

- Academic supervisor (Prof. Timo Speith) and institution (University of Bayreuth)
- Research collaborators, especially those connected to the original MTAIR project
- $\bullet\,$ Technical advisors who provided feedback on implementation aspects
- Funding sources and those who provided computational resources or API access
- Personal supporters who enabled the research through encouragement and feedback

Prefatory Apparatus: Illustrations and Terminology — Quick References

List of Tables

Table 1: Table name
Table 2: Table name
Table 3: Table name

- Figure 1.1: The coordination crisis in AI governance visualization of fragmentation
- Figure 2.1: The Carlsmith model DAG representation
- Figure 3.1: Research design overview workflow diagram
- Figure 3.2: From natural language to BayesDown transformation process
- Figure 4.1: ARPA system architecture component diagram
- Figure 4.2: Visualization of Rain-Sprinkler-Grass_Wet Bayesian network screenshot
- Figure 5.1: Extraction quality metrics comparative chart
- Figure 5.2: Comparative analysis of AI governance worldviews network visualization
- Table 2.1: Comparison of approaches to AI risk modeling
- Table 3.1: Probabilistic translation guide for qualitative expressions
- Table 4.1: System component responsibilities and interactions
- Table 5.1: Policy impact evaluation results summary metrics

List of Graphics & Figures

List of Abbreviations

esp. especially
f., ff. following
incl. including
p., pp. page(s)
MAD Mutually Assured Destruction

- AI Artificial Intelligence
- AGI Artificial General Intelligence
- ARPA AI Risk Pathway Analyzer
- DAG Directed Acyclic Graph
- LLM Large Language Model
- MTAIR Modeling Transformative AI Risks
- P(Doom) Probability of existential catastrophe from misaligned AI
- CPT Conditional Probability Table

Glossary

- **Argument mapping**: A method for visually representing the structure of arguments
- BayesDown: An extension of ArgDown that incorporates probabilistic information
- Bayesian network: A probabilistic graphical model representing variables and their dependencies
- Conditional probability: The probability of an event given that another event has occurred
- Directed Acyclic Graph (DAG): A graph with directed edges and no cycles
- Existential risk: Risk of permanent curtailment of humanity's potential
- Power-seeking AI: AI systems with instrumental incentives to acquire resources and power

- **Prediction market**: A market where participants trade contracts that resolve based on future events
- **d-separation**: A criterion for identifying conditional independence relationships in Bayesian networks
- Monte Carlo sampling: A computational technique using random sampling to obtain numerical results

Checklists

"Usual paper requirements"

- introduce all terminology
 - go through text, make sure all terms are defined, explained (and added to the list of Abbr.) when first mentioned
- readership is intelligent and interested but has no prior knowledge

(Format:) \sim Anything that makes it easier to understand

- short sentences
- paragraphs (one idea per paragraph)
- simplicity
- !limit use of passive voice!
- use active voice, even prefer I over we!
- minimise use of "zombi nouns" (don't turn verbs/adjectives to nouns!)
- "find words that can be cut"
- the paper can focus on one aspect of the presentation
- "open door policy" for (content) questions
- ~ demonstrate ability for novel research
- "solve research question with the tools accessible to you"
- "show something that has not been shown before / should be publishable in principle"
- new idea (or criticism) "in this field"
- Outline idea THEN reading with a purpose (answering concrete questions)

- "Only" confirm that no body has published the exact same idea on the same topic
- pretty much determined by presentation & proposal but narrow down further (& choose supervisor?)

Quarto Features Incompatible with LaTeX (Below)

Quarto Syntax

Figures

Testing crossreferencing grapics Figure 7.1.
Testing crossreferencing grapics Figure 7.2.

Citations

```
Soares and Fallenstein [4]
[4] and [3]
Blah Blah [see 3, pp. 33–35, also 2, chap. 1]
Blah Blah [3, 33–35, 38-39 and passim]
Blah Blah [2, 3].
Growiec says blah [2]
```

7.1 Headings & Potential Headings

verbatim code formatting for notes and ideas to be included (here)

Also code blocks for more extensive notes and ideas to be included and checklists

```
- test 1.
```

- test 2.

- test 3.

2. second

3. third

Blockquote formatting for "Suggested Citations (e.g. carlsmith 2024 on ...)" and/or claims which require a citation (e.g. claim x should be backed-up by a ciation from the literature)

Here is an inline note.¹ Here is a footnote reference,² Here's some raw inline HTML:

page 1

¹Inlines notes are easier to write, since you don't have to pick an identifier and move down to type the note.

²Here is the footnote.

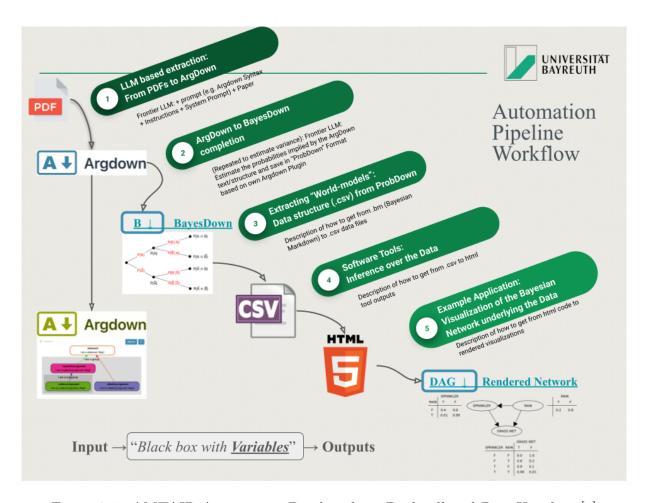
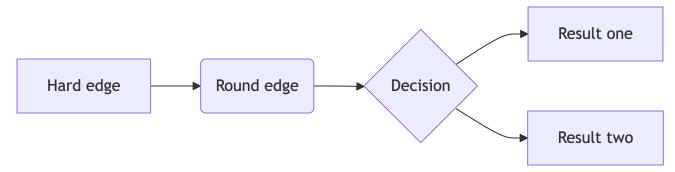


Figure 7.1: AMTAIR Automation Pipeline from Bucknall and Dori-Hacohen [1]



Figure 7.2: Caption/Title 2

page 2



Testing crossreferencing grapics Figure 7.1.

Bibliography (References)

- [1] Benjamin S. Bucknall and Shiri Dori-Hacohen. "Current and Near-Term AI as a Potential Existential Risk Factor". In: *Proceedings of the 2022 AAAI/ACM Conference on AI, Ethics, and Society.* AIES '22: AAAI/ACM Conference on AI, Ethics, and Society. Oxford United Kingdom: ACM, July 26, 2022, pp. 119–129. ISBN: 978-1-4503-9247-1. DOI: 10.1145/3514094.3534146. URL: https://dl.acm.org/doi/10.1145/3514094.3534146 (visited on 11/13/2024).
- [2] Jakub Growiec. "Existential Risk from Transformative AI: An Economic Perspective". In: Technological and Economic Development of Economy (2024), pp. 1–27.
- [3] Donald E. Knuth. "Literate Programming". In: Computer Journal 27.2 (May 1984), pp. 97–111. ISSN: 0010-4620. DOI: 10.1093/comjnl/27.2.97. URL: https://doi.org/10.1093/comjnl/27.2.97.
- [4] Nate Soares and Benja Fallenstein. "Aligning Superintelligence with Human Interests: A Technical Research Agenda". In: (2014).

Appendix A

Appendices

- A.1 Appendix A: Technical Implementation Details
- A.2 Appendix B: Model Validation Procedures
- A.3 Appendix C: Case Studies
- A.4 Appendix D: Ethical Considerations

TestText

$\begin{array}{c} \mathbf{Appendix} \ \mathbf{B} \\ \mathbf{appendix} \mathbf{A} \end{array}$

testtext

List of Figures

2.1	Five-step AMTAIR automation pipeline from PDFs to Bayesian networks .	14
7.1	Five-step AMTAIR automation pipeline from PDFs to Bayesian networks .	34
7.2	Short 2 caption	34



Affidavit

Declaration of Academic Honesty

Hereby, I attest that I have composed and written the presented thesis

Automating the Modelling of Transformative Artificial Intelligence Risks

independently on my own, without the use of other than the stated aids and without any other resources than the ones indicated. All thoughts taken directly or indirectly from external sources are properly denoted as such.

This paper has neither been previously submitted in the same or a similar form to another authority nor has it been published yet.

BAYREUTH on the May 21, 2025

VALENTIN MEYER.