





Toulouse, June 2025

# Rethinking Sentencing Research

Scoping Literature Review, Upgraded



# WISHING FOR A DIFFERENT REVIEW

A literature review, but also setting the foundation for modelling sentencing as a structured decision-making environment.

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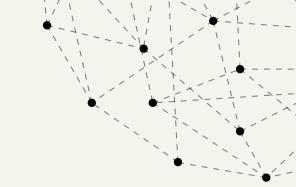
### WHAT DOES SENTENCING SCHOLARSHIP SAY?

→ Identify key themes, methods, jurisdictions, gaps

### WHAT IS SENTENCING AS AN ENVIRONMENT?

→ Longer-term aim: define actors, tools, processes

# Why is this hard?



### SENTENCING SCHOLARSHIP

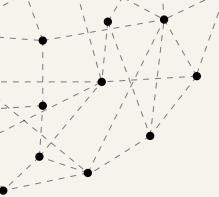
- o Interdisciplinary
- o Methodologically diverse
- Nationally grounded and shaped by legal traditions
- O Views sentencing as shaped by law, but also by institutional, individual, and cultural factors, but varied viewpoints are not always in dialogue with one another

### CHOICE ARCHITECTURE

- o Rooted in behavioural economics and cognitive psychology
- o Focused on how contextual design influences decisions
- o Experimental, often abstracted from institutional specifics
- Views decision-making as bounded, biased,
   and modifiable through design

They differ in vocabulary, methodologies and assumptions about how decisions are made.

The challenge: how to bring them together without overlooking their differences.



# Why not a traditional review?

## LIMITS OF MANUAL REVIEWS

- o Small-scale
- o Language-limited
- o Highly time-consuming and labour-intensive
- o Prone to selection bias
- Lacking in transparency (implicit criteria)
- o III-suited for cross-jurisdictional or interdisciplinary synthesis

### WHAT WE WANT?

- o To capture the widest possible range of sources (disciplines, languages, jurisdictions)
- o To build the foundation for a sentencing architecture
- o Most importantly, an approach that is:
  - o Large-scale
  - o Uniform
  - o Efficient
  - o Structured and reviewable

A review process that is scalable, transparent, and open: capable of mapping the full landscape of sentencing research.

# Our starting point

### WHAT WE STARTED TO DO

o Build a taxonomy of >280 themes (e.g., human rights in sentencing)

o Use ChatGPT to read papers and ...

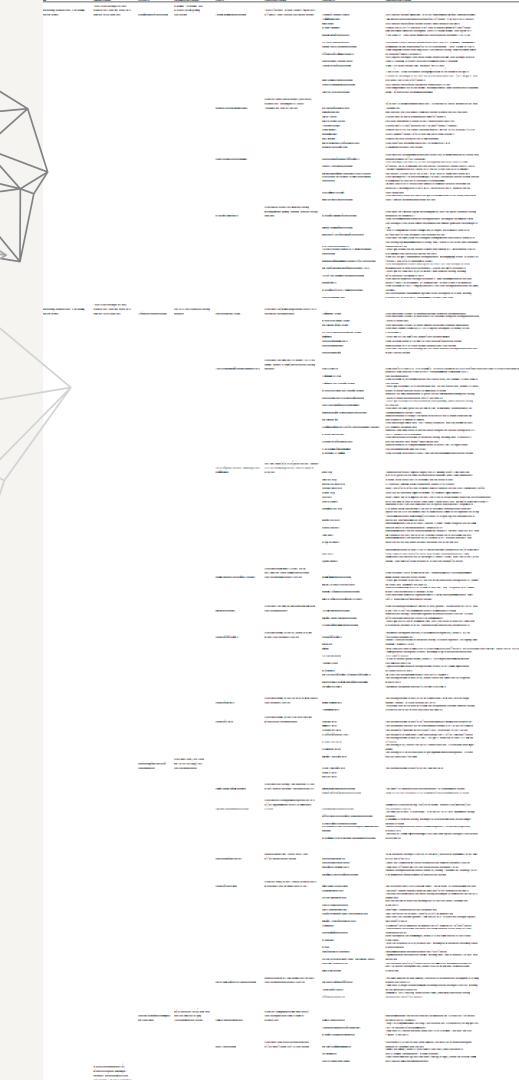
o assign themes

o prepare structured abstracts

o extract additional information (authors, year of publication ...)

o Create a searchable, mappable structure of the literature

We thought this would be straightforward — and that scale could be managed by rules.



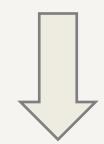
# What happened instead?

- o Weekly team meetings turned into continuous iteration
- O Our spreadsheet grew exponentially in complexity: the more detailed we got, the harder it became to agree

### THE PROBLEMS

- o Loads of variation
  - o Human-human (Inter-rater variation)
  - o Human-Al ("Extra-rater" variation)
  - o Al-Al (Cross-model variation)
- o Themes began to pre-shape findings instead of letting findings emerge

Started of as a way of classifying papers



A shift from classifying papers to understanding the space between them.

# What we're actually doing?

What we kept

- o Structured abstracts (ChatGPT-generated)
- o A lot of other information about the papers
- o Themes & subthemes
- o Methodology
- o Relevance of papers for the Sentrix project

What we added

- o Embeddings to compare papers semantically, not just by assigned themes
- o A shift from "categorising papers" to understanding their relationships

### Structured abstracts

Building comparable narratives

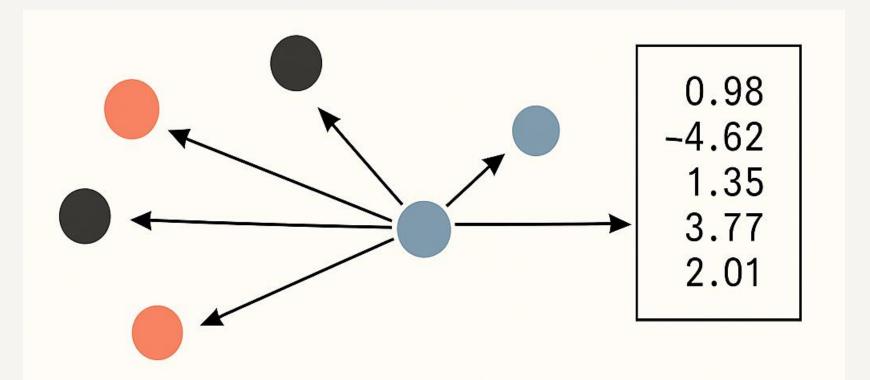
What they are?

- Each paper summarised using a ~400-word structured abstract, generated by ChatGPT
- Based on a detailed prompt with four consistent sections:
  - o Background / Context
  - o Objectives / Focus
  - o Methods / Approach
  - o Findings / Insights

Why they matter?

 Make papers comparable across disciplines, jurisdictions, and methods

# Embeddings



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### WHAT ARE THEY?

= a way of converting text (a paper, theme, or sentence) into a numeric vector - a list of numbers that captures its meaning.

### HOW WE USE THEM?

Texts similar in meaning end up close together in a mathematical space – even if they don't use the same words.

# Embeddings as method

From labels to relationships

### USE OPEN.AI EMBEDDINGS TO:

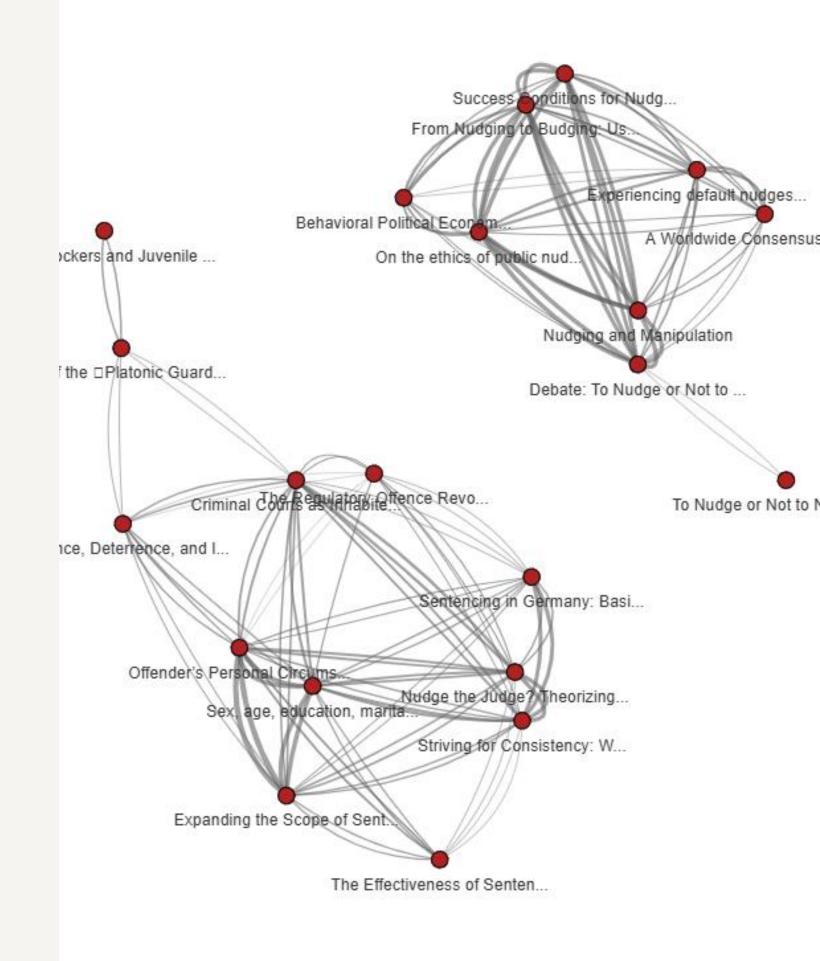
Represent papers, themes, methods and project documents as vectors

Compare them semantically

### hopefully THIS ENABLES

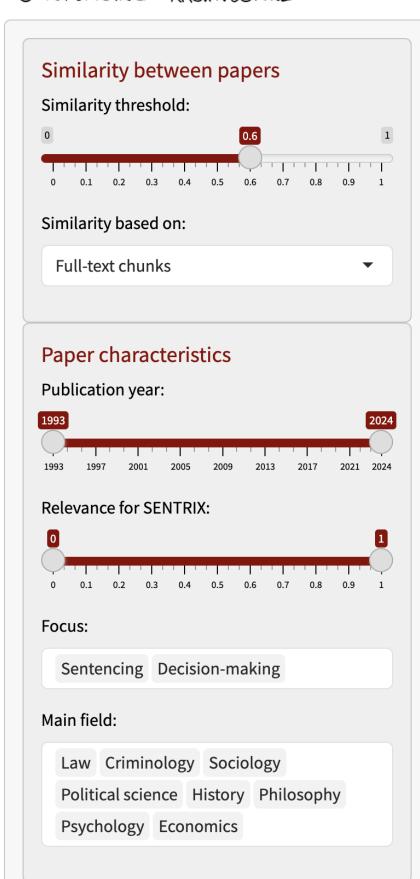
Discovery of hidden structure
Theme-agnostic similarity
Mapping of new conceptual spaces

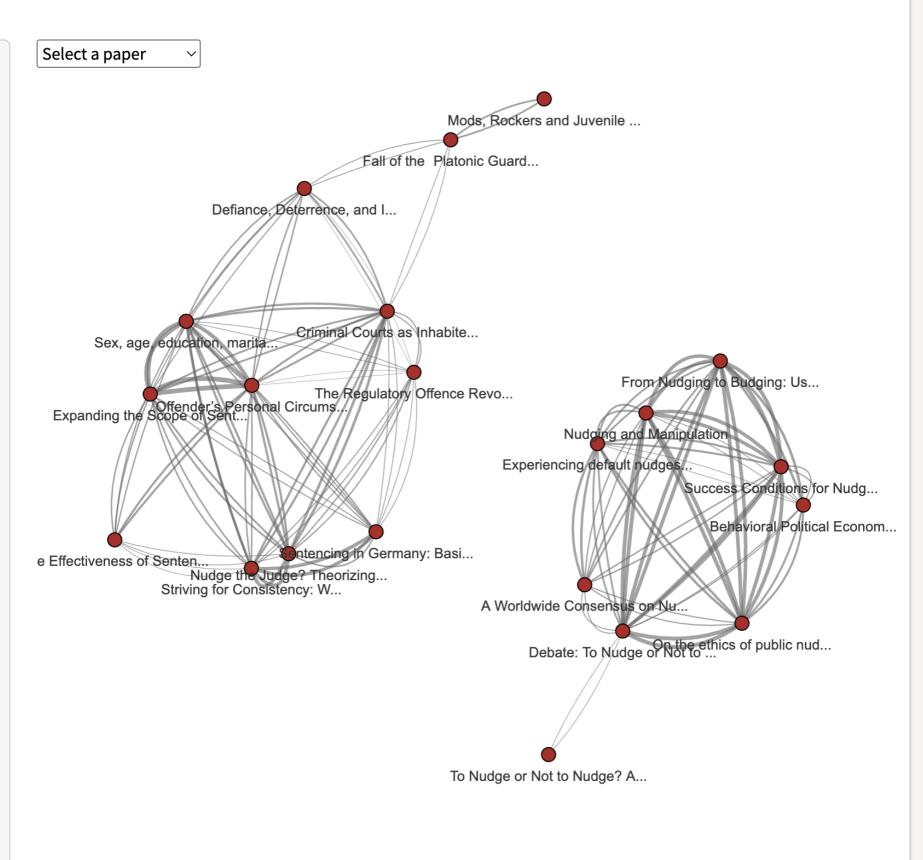
Moves beyond surface labels → captures deep conceptual proximity





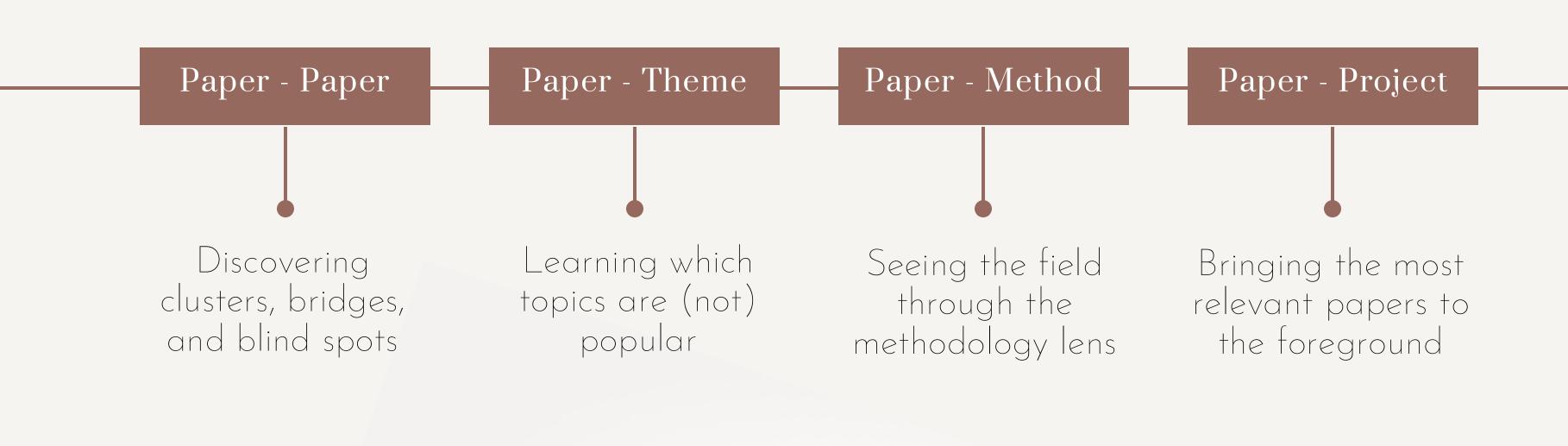
#### SENTENCING ARCHITECTURE





# The analytical engine behind the map

COMPARING PAPERS, THEMES, METHODS, AND ASSESSING RELEVANCE



# Looking ahead

Modelling the sentencing environment

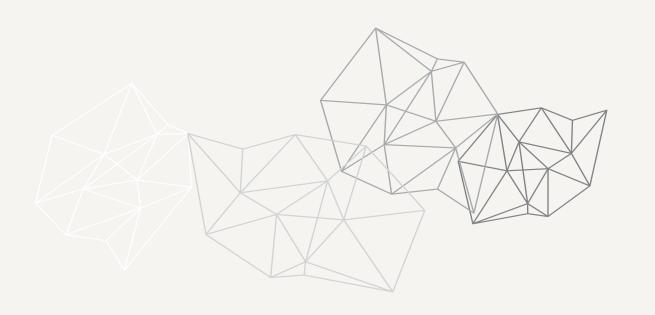
### WHAT ARE WE ASKING?

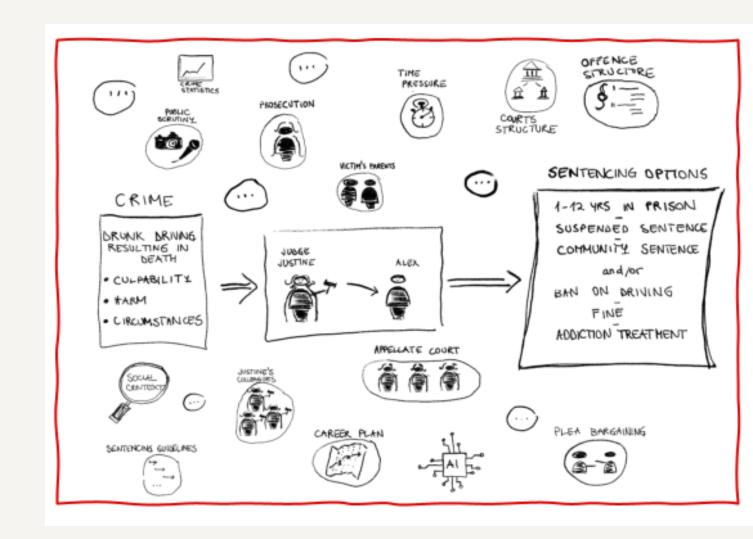
What are the latent elements that structure sentencing as a decision-making environment?

What parts of this environment are well-represented in the literature—and what's missing?

### HOW ARE WE DOING IT?

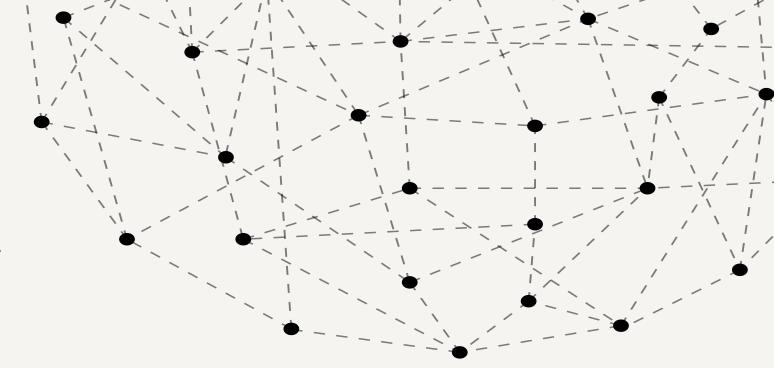
O Create a rich network of papers that can be searched, filtered, and visualized based on a large set of their properties





### Broader value?

A replicable way of doing literature review



Our approach is ...

- o Replicable
- o Transparent
- o Flexible
- o Scalable

Useful beyond sentencing?

- o Other legal domains
- o Other interdisciplinary spaces

Thank you.

# We welcome all and any feedback!

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