



Hermeneutic Ontology Engineering: LLM-Assisted Schema Induction for Oral History Knowledge Organisation

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What testimony sounds like

NYHAN

...it is the physicality of the operation that always strikes me. That is something we have lost.

INTERSUBJECTIVE —
the interviewer's remark
shapes what is told next

MALLOY

Indeed, we have lost that. ... I used to think that the slot accessed some long tunnel to the Goddard Space Flight Center, but when I was looking at the map I saw that the Space Center was actually not that close ... They must have gone to a truck or something.

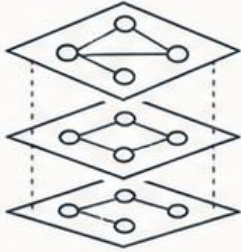
SITUATED —
remembered from now,
corrected by a map seen
years later

PERFORMATIVE —
meaning made live:
'must have gone to a truck'

Oral history records not simply what happened, but how individuals construct meanings from the past.

Research questions

01



RQ1 – MODELLING & REPRESENTATION

How can network structures represent the narrative and temporal dimensions of oral histories?

02



RQ2 – ANALYSIS & PATTERNS

What narrative patterns of knowledge creation, mobility and influence emerge when we analyse these recollected relationships?

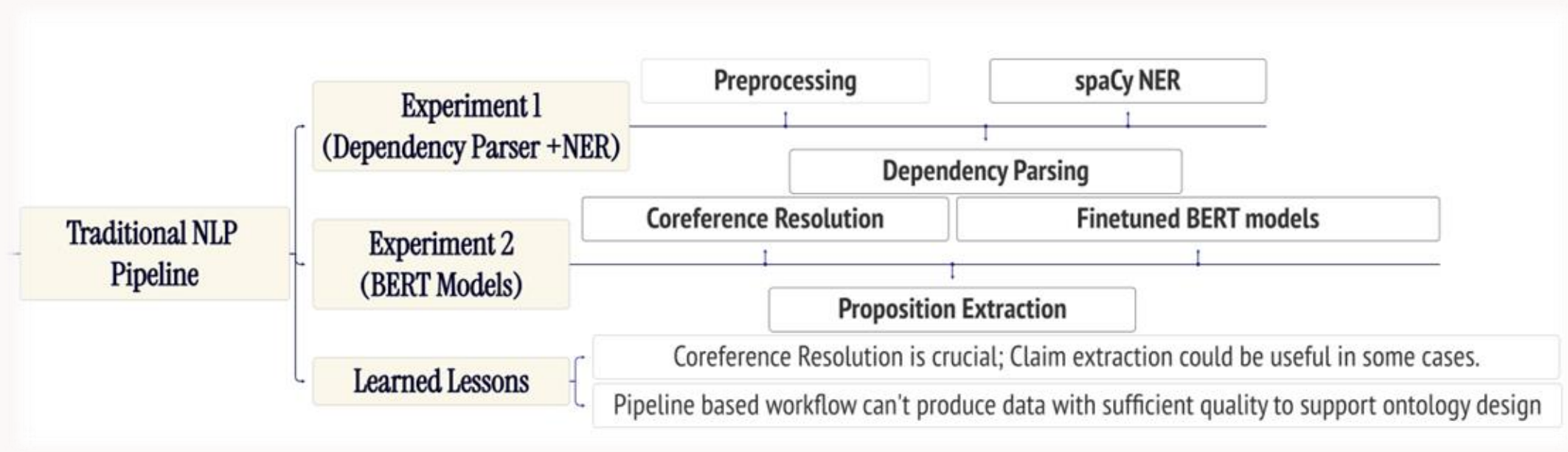
03



RQ3 – BRIDGING METHODS

How knowledge graphs create a defensible and transparent bridge between qualitative narrative and quantitative network analysis?

Information Extraction from Transcript

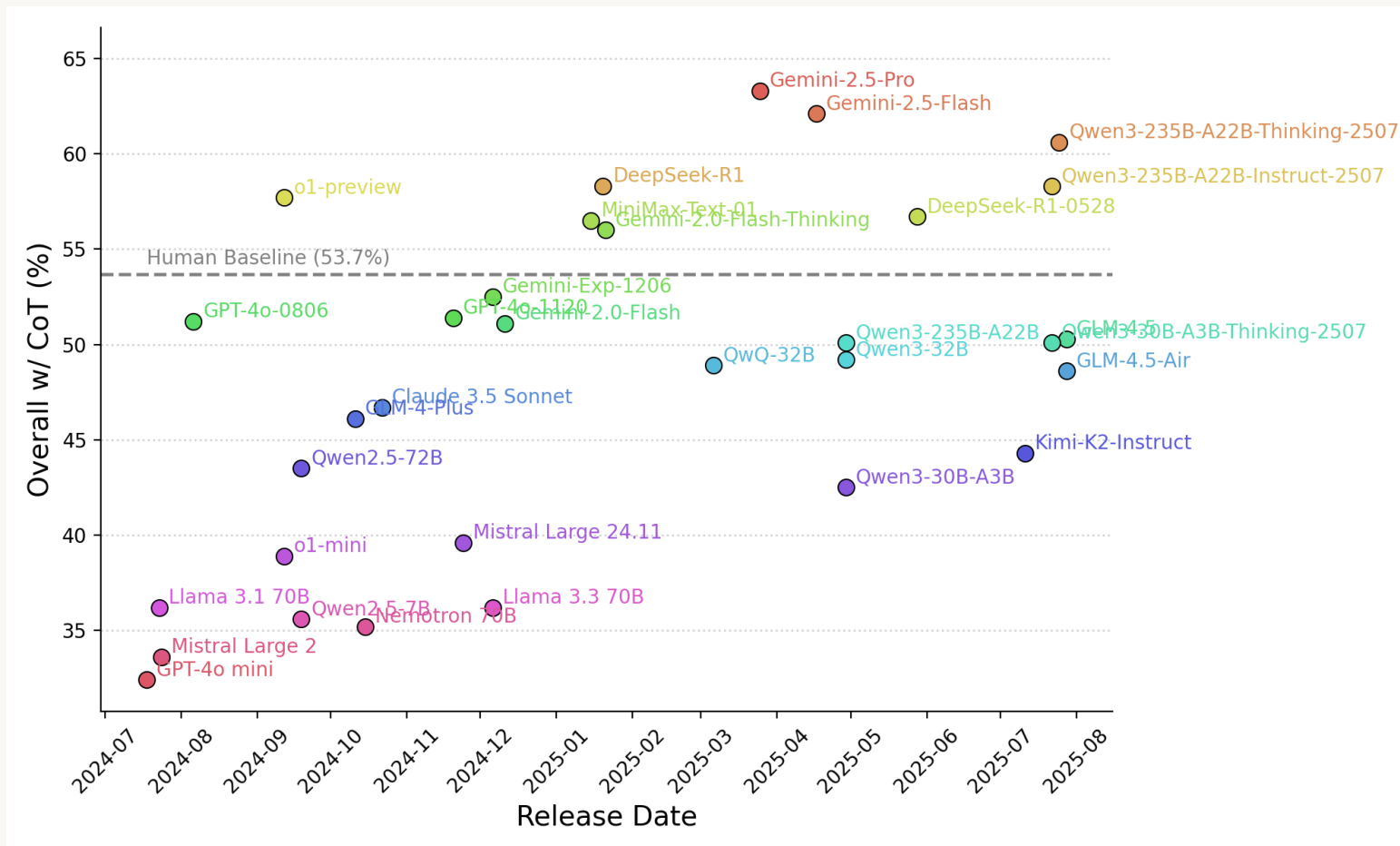


Traditional NLP Pipeline (Traditional ≈ before 2021)

What generic AI

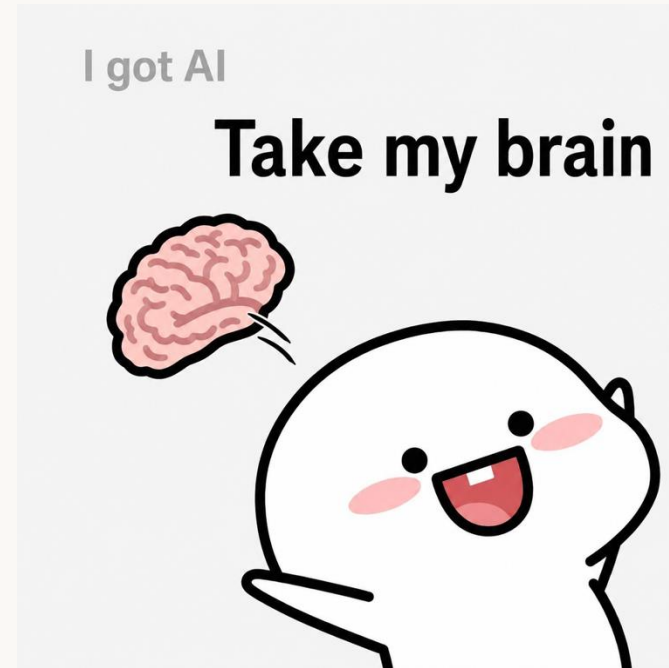
LongBench v2

Benchmarking
Deeper
Understanding and
Reasoning on
Realistic Long-
context Multitasks



I / IV – THE PROBLEM OF REMEMBERING

What generic AI ?



What generic AI smooths over

Three ways a generic pipeline fails testimony —

01 PARAMETRIC KNOWLEDGE OVERRIDE

“I had access to the University of California's libraries.”

— MALLOY



Malloy — educatedAt → University of California

NEVER SPOKEN

a ‘Wikipedia-shaped’ pattern from pretraining
overrides the sentence in front of it

02 FLATTENING THE HEDGE

“My first significant grant was,
if I remember correctly, for \$242,000,
from the NEH.”

— NITTI



Nitti — receivedGrant → \$242,000 (NEH)

a bare triple has no field for epistemic stance —
the doubt is erased

03 CONTEXT ROT

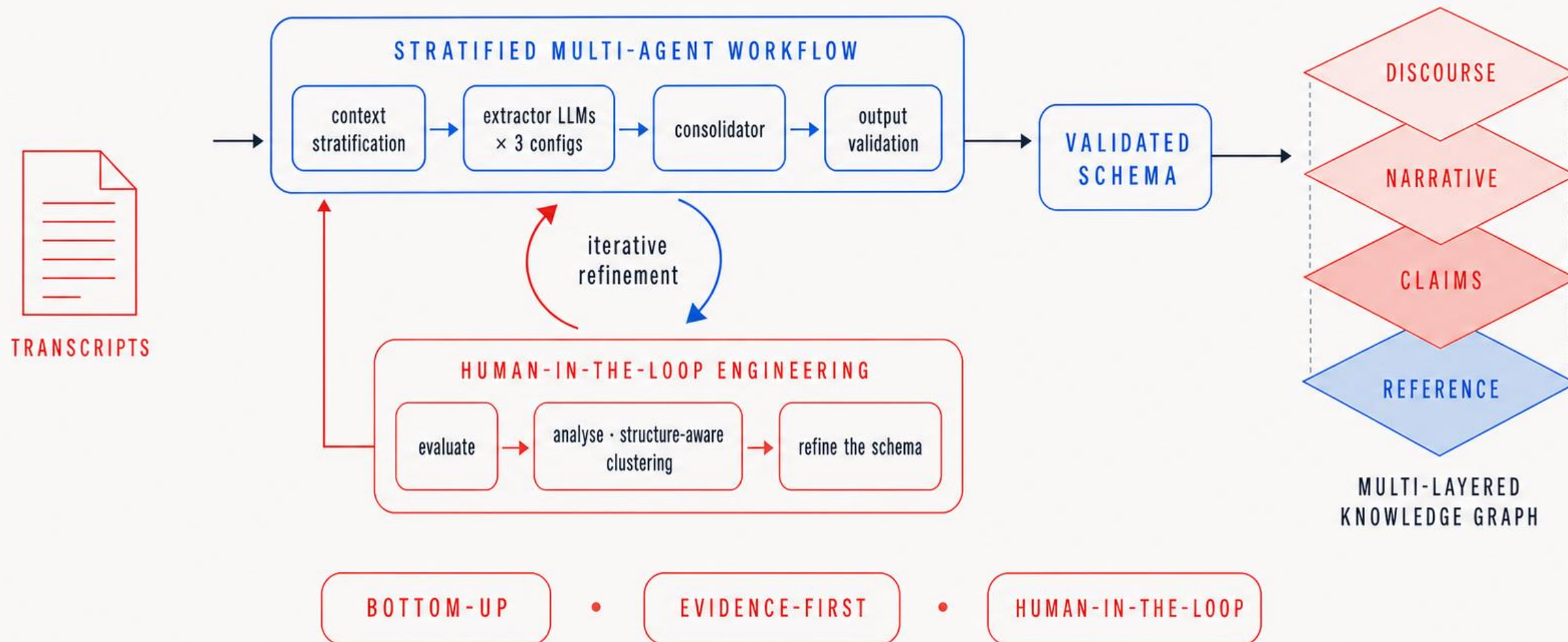


a full-length transcript dilutes attention:
details blend, speaker attributions fail

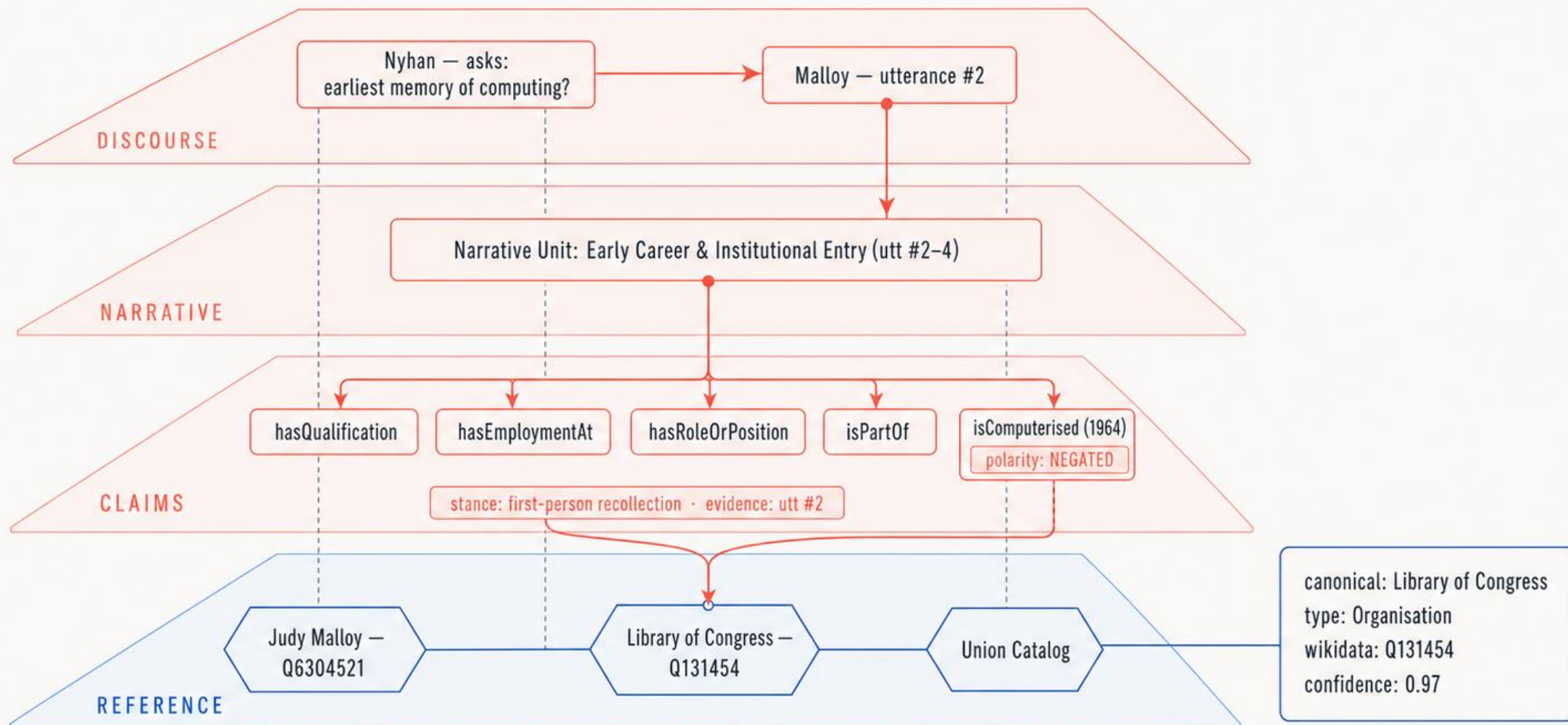
fatal where meaning lives in nuances
scattered across the whole interview

LLM-driven Hermeneutics Framework

The ontology as an instrument of reading, not a verdict — the LLM as a tireless reader, not an oracle.



Layers of information



One utterance → five verifiable claims — hedges, roles and negations kept. Nothing stored as bare fact.

Claims, not facts

A FLAT FACT

Malloy — worked at → Library of Congress

speaker, stance, evidence: gone

“

〈 Malloy — hasEmploymentAt → Library of Congress 〉

”

STANCE

polarity: affirmed

type: direct_experience

certainty: certain

temporal: distant_memory

PROVENANCE

speaker: Judy Malloy · interviewee

utterance #2

My first job ... was at the Library of Congress ...

VERIFICATION

confirmed — 0 of 3 refute votes

extractor: claude-opus-4-8

CONTESTED MEMORY, PRESERVED — WHERE DID THE TEI BEGIN?

〈 TEI — takesPlaceAt → Modern Languages Association 〉

said by RUTIMANN

certainty: probable

≠

〈 TEI — takesPlaceAt → Vassar College 〉

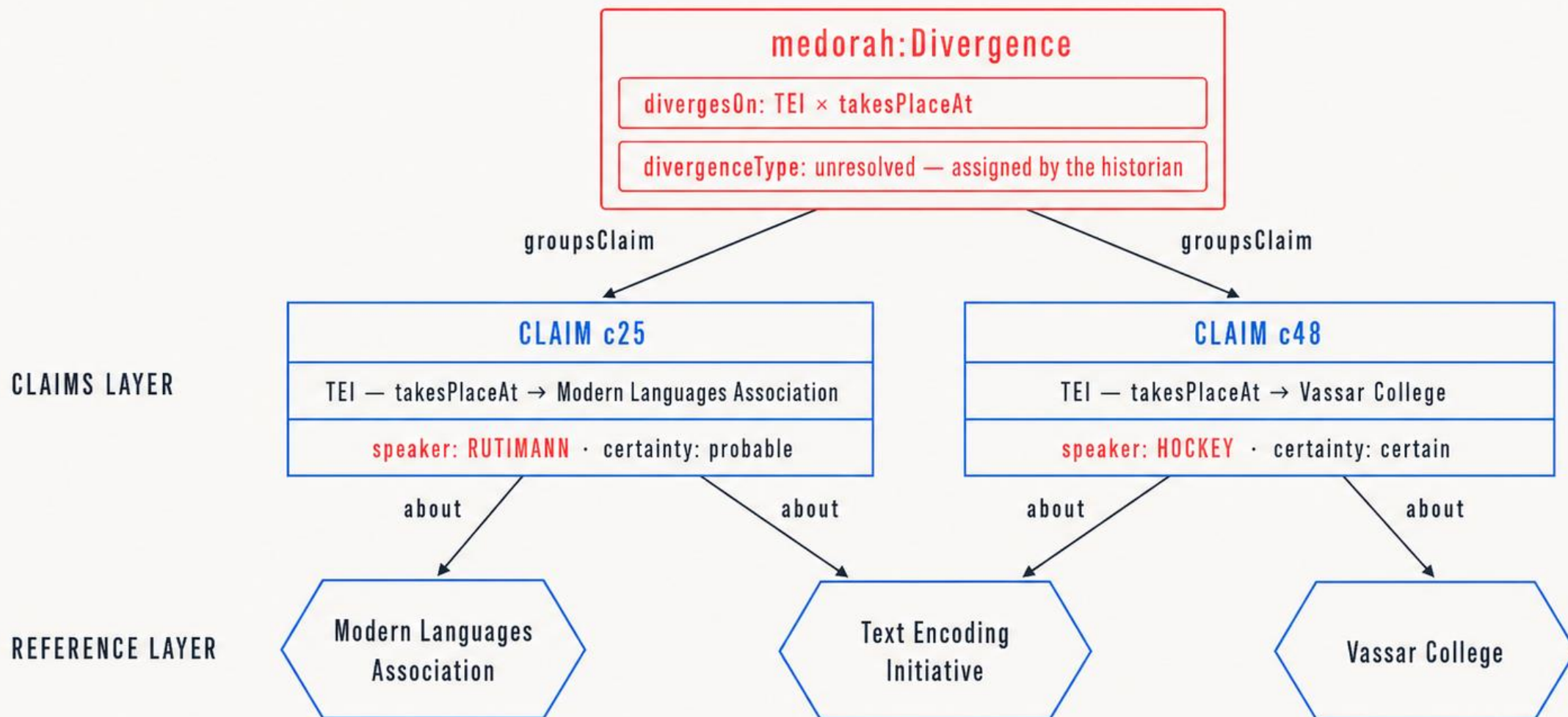
said by HOCKEY

certainty: certain

Both stay — attributed, stanced, queryable. The graph preserves the divergence; the historian interpretes it.

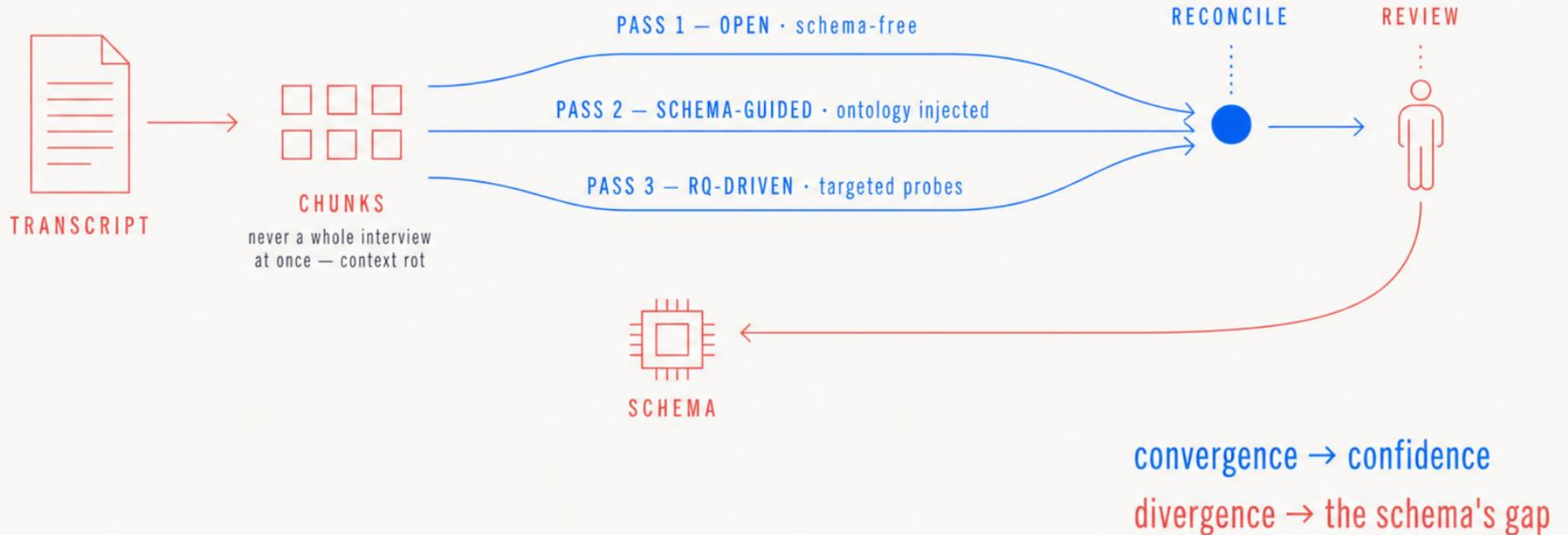
The graph knows how it knows.

The schema of a disagreement



Disagreement becomes a first-class, queryable object — noticed by the machine, typed by the historian, resolved by no one.

Three passes, designed to disagree

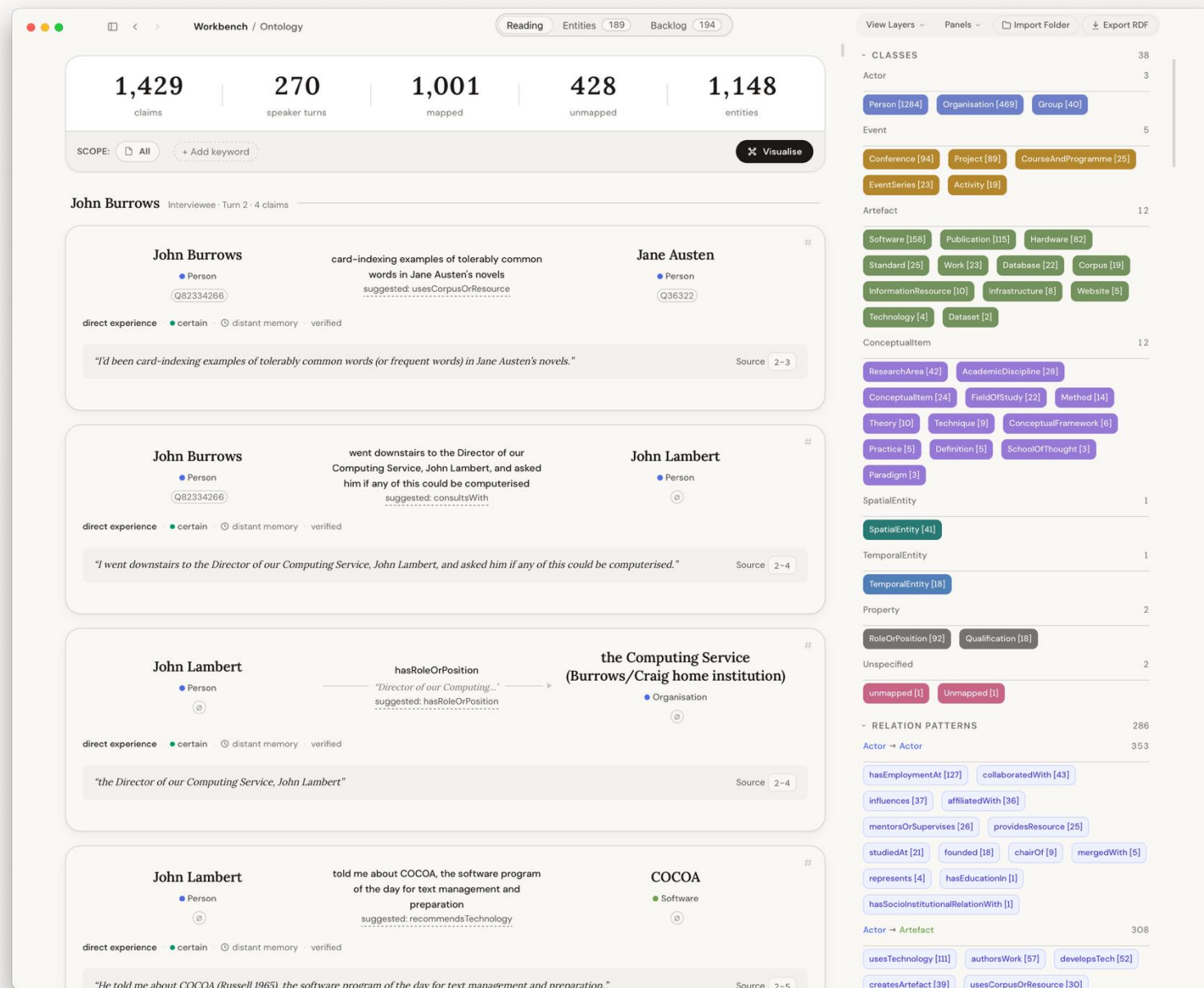


An interactive workbench

Judgement needs an instrument.

Macro View: The sidebar provides aggregated analytics, revealing frequent entities, relationships, and structural patterns across the entire dataset.

Micro View: The main panel grounds every extraction in its original narrative context, showing the speaker and the



An interactive workbench

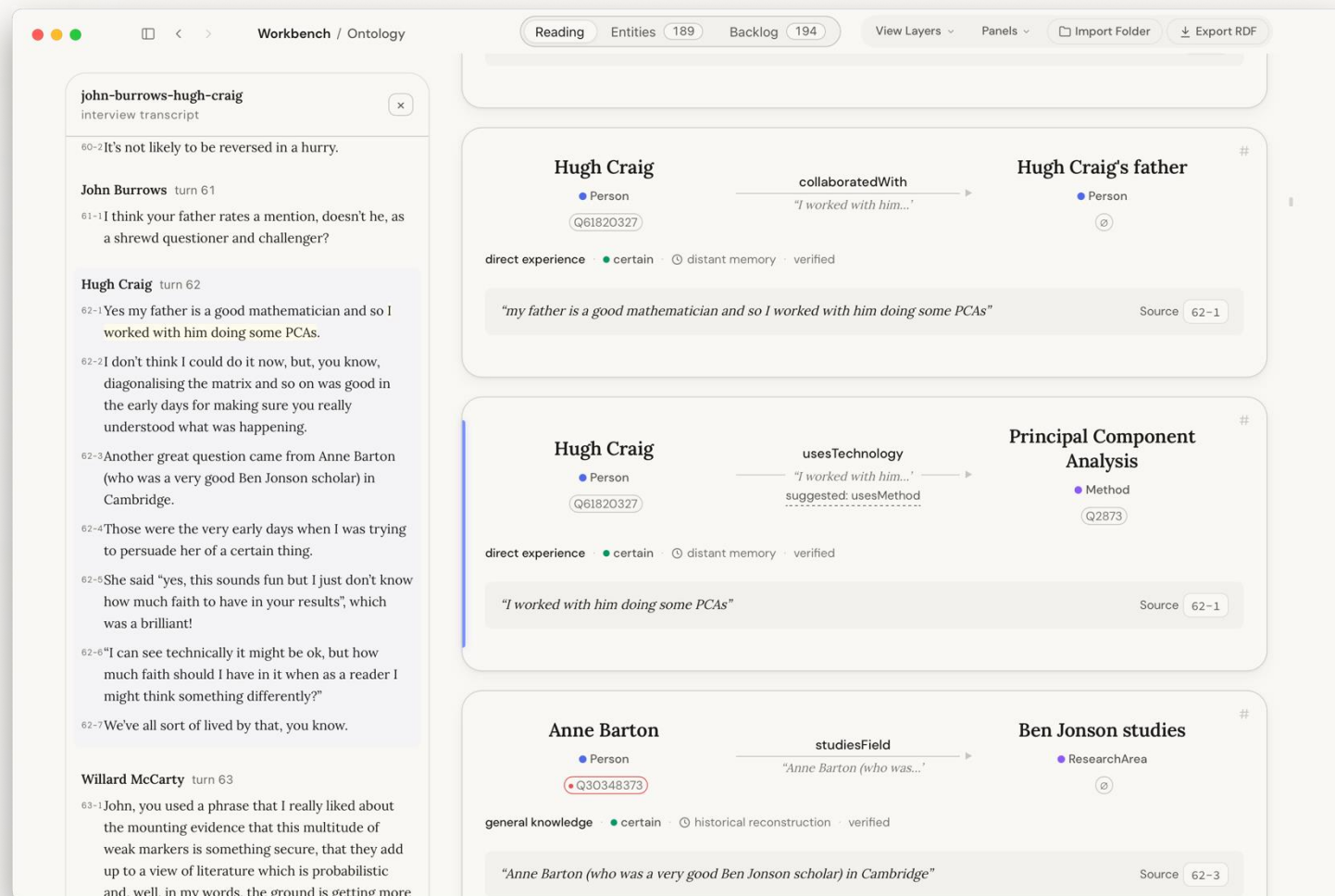
Judgement needs an instrument.

provenance-linked inspection — every claim beside its source span

structural analytics — how relations distribute

predicate clustering — candidate templates to accept or reject

OPEN SOURCE



From 160 raw types to a governed ontology

160

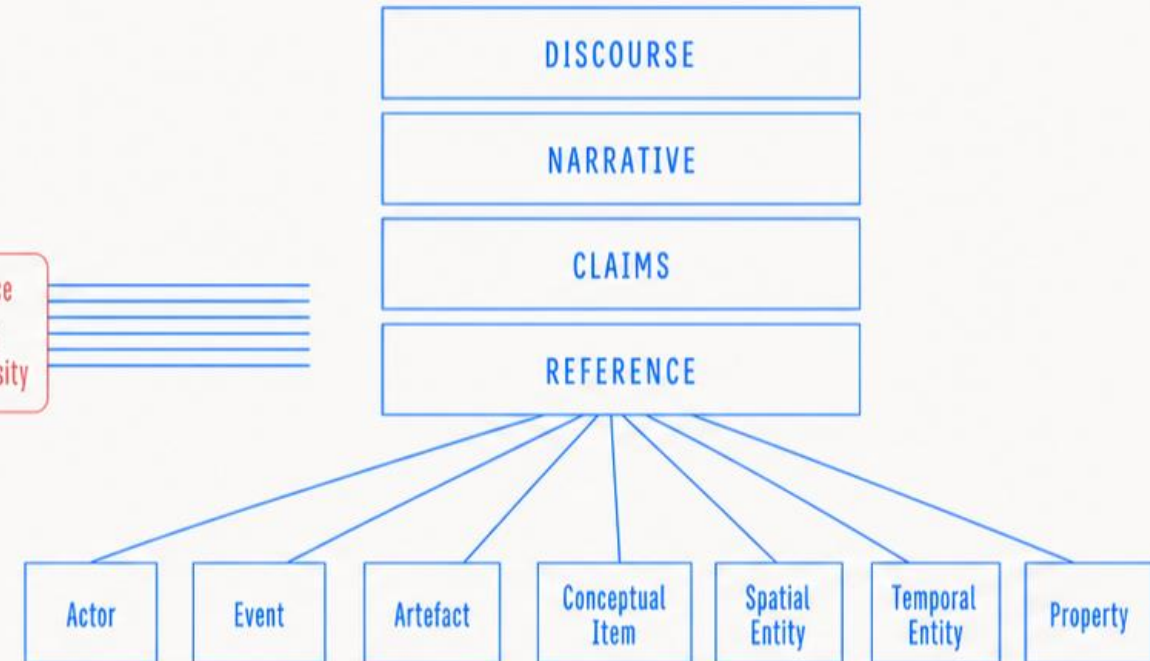
raw entity types, schema-free pass — real labels:



eight labels, one idea — the model mints a new type for every new word

22

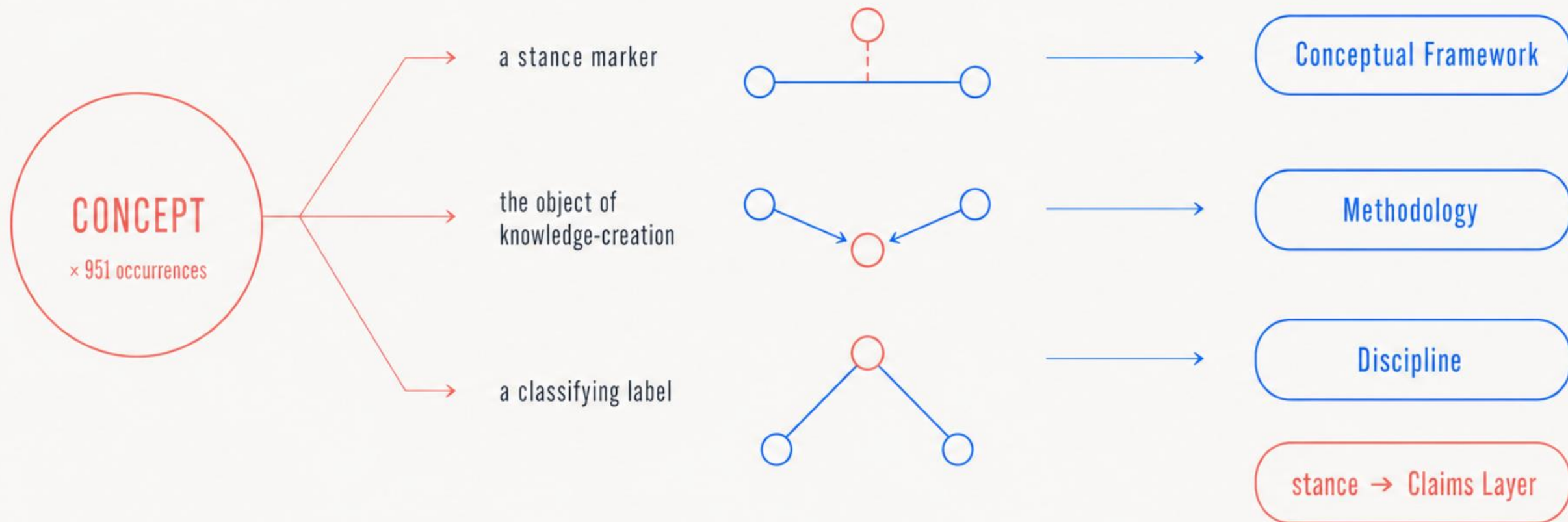
classes — 7 top-level + 15 sub



11 relation patterns · informal alignment: CIDOC-CRM · FOAF · Dublin Core

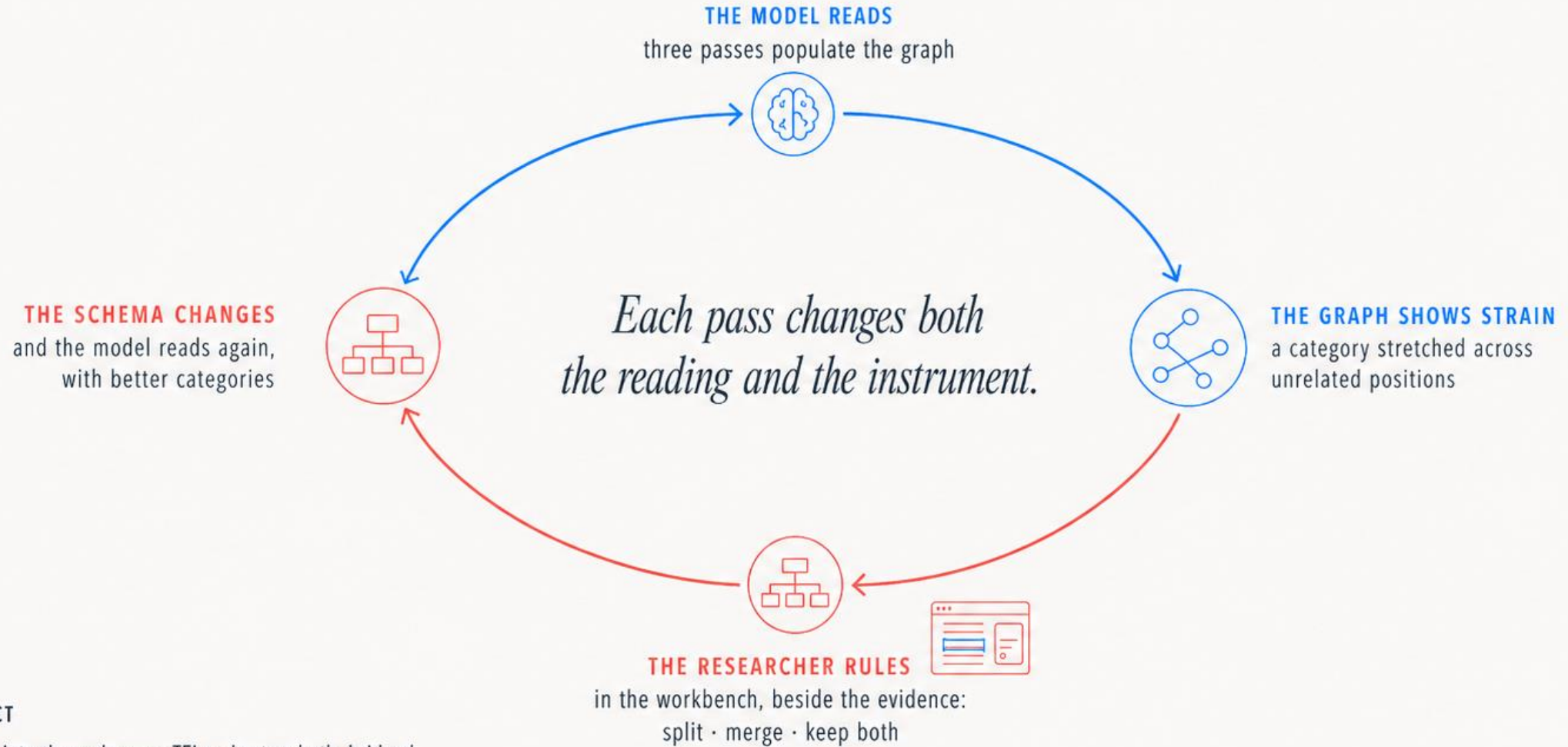
formal mappings deferred — deliberately: the schema is still moving.

The case of "Concept"



Invisible in frequency. Obvious in graph position.

The hermeneutic loop, in practice

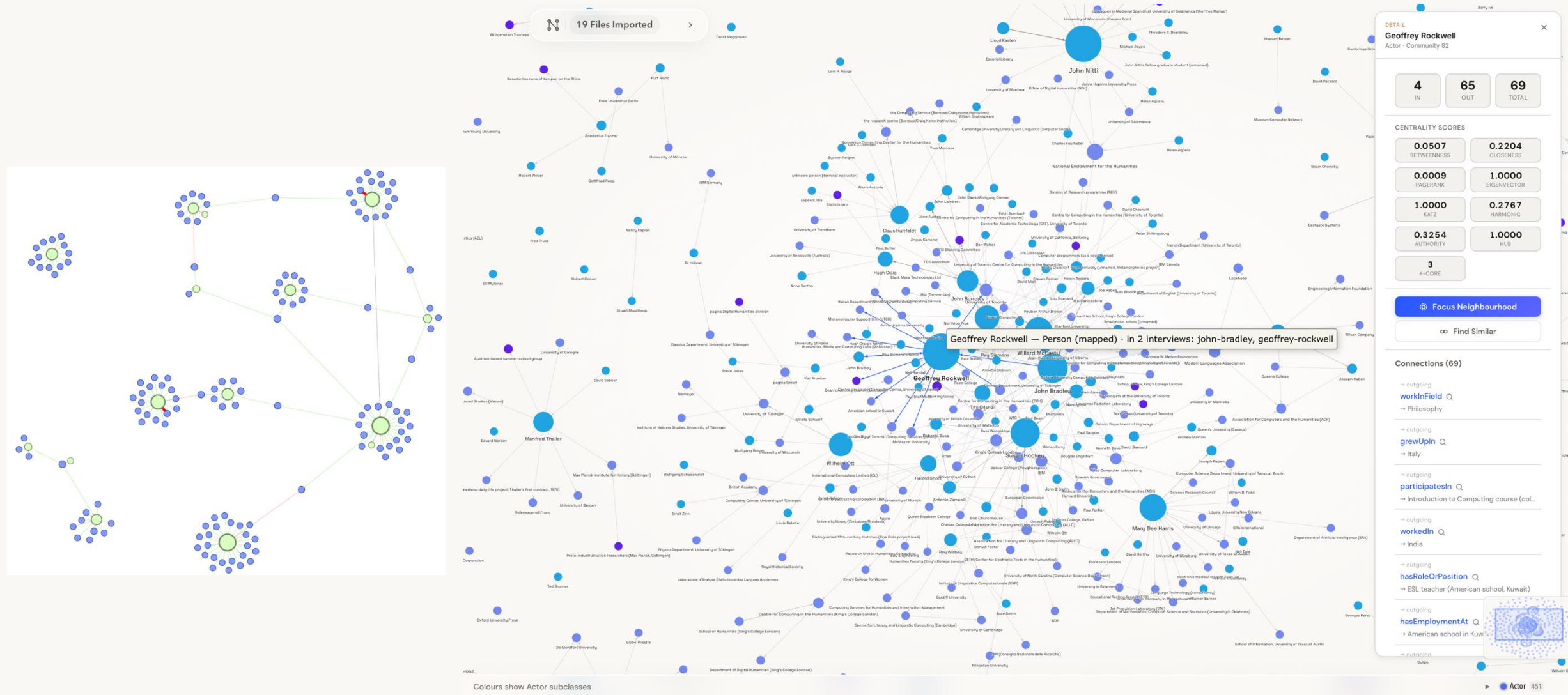


IN THIS PROJECT

'Concept' → split into three classes · TEI → kept as both, bridged

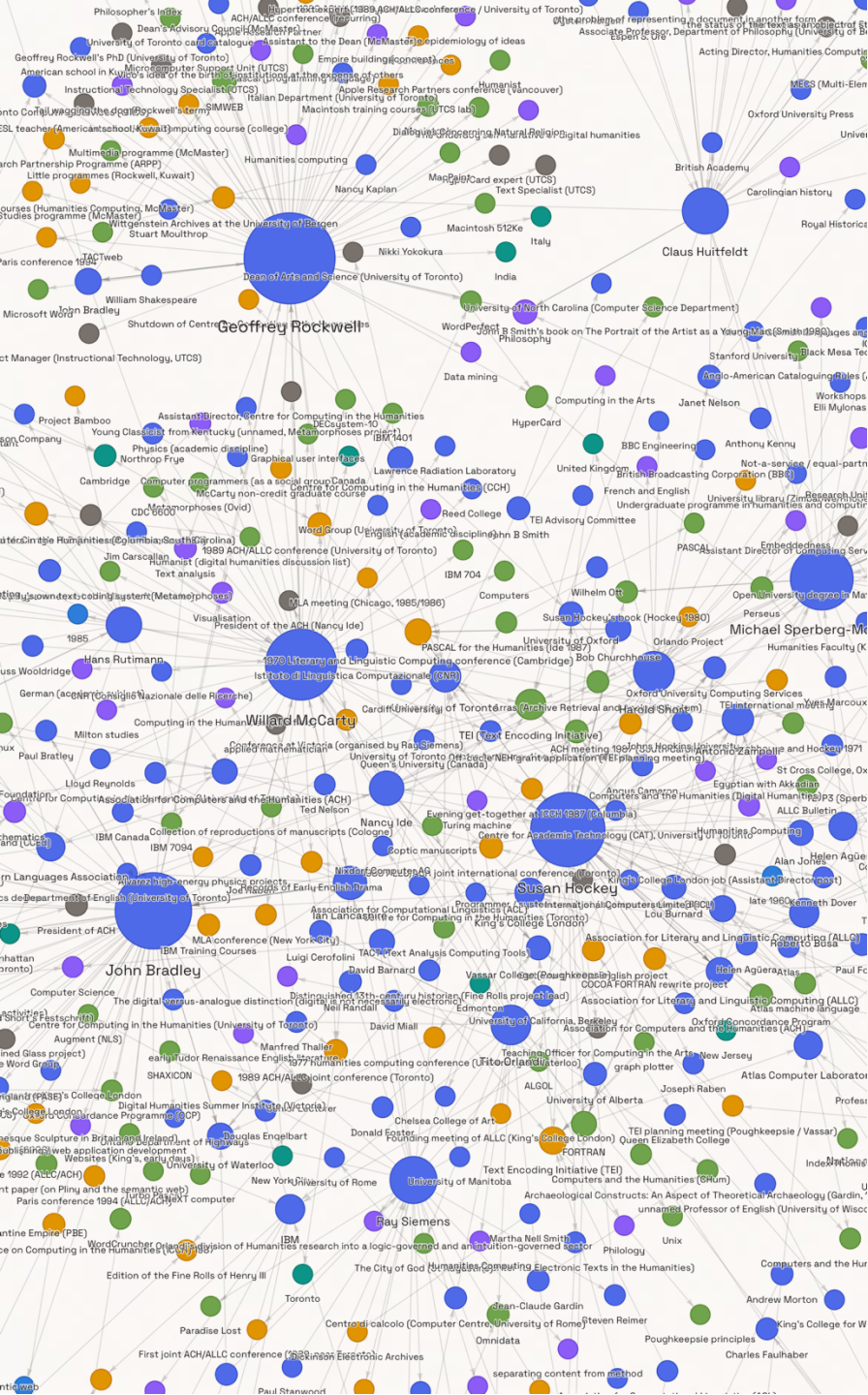
Extraction stops consuming the schema – and starts testing it.

Expert Interpretation & Automated Extraction



Manual Annotation

LLM driven extraction with human in the loop



IV / IV — WHAT THE STRUCTURE REVEALED

The populated knowledge graph

Same questions, two readings — hand-coded 2025 vs the claims graph 2026.

HAND-CODED '25

CLAIMS GRAPH '26

PEOPLE RECALLED

124

209

CLUSTERING

0.0975

0.0388

still star-shaped

TRIANGLES

57

51

MEMORIES MEET THROUGH

people

P · P · P 33

institutions

I · P · P 23

CENTRAL TECHNOLOGY

IBM 360/20 • TUSTEP

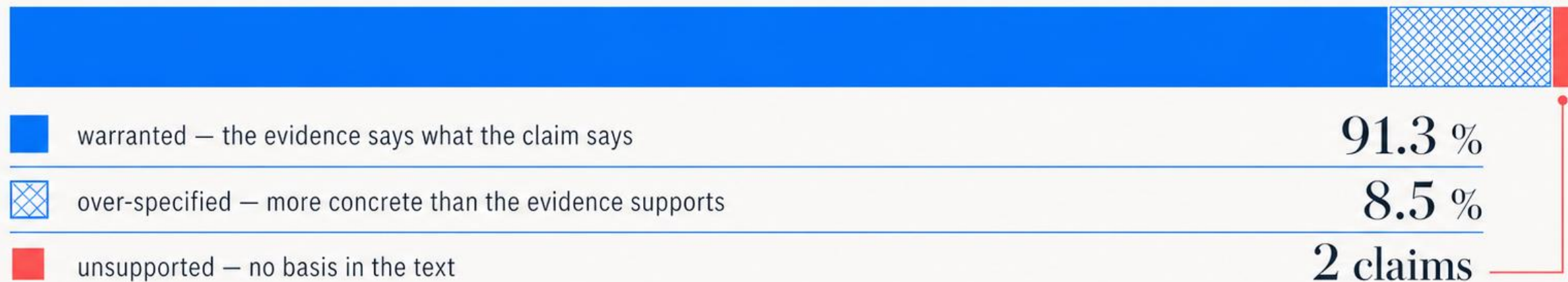
IBM 7094 • FORTRAN

Busa & Zampolli — never interviewed, still anchoring the remembered field.

35.7 % Wikidata-anchored • 25 hearsay • 169 probable — every cell of this table is queryable, with provenance.

Reliable, not exhaustive

1,429 claims. Every one read back against the transcript by hand.



WHAT THE 8.5 % ACTUALLY IS — INFLATION, NOT INVENTION

*...the usual 1980s route of having
a Sinclair Spectrum and Commodore 64...*

— GABRIEL EGAN



Egan — usesTechnology → Sinclair Spectrum

owning is not using — the direction is right, the verb says one notch too much.

Selective by design — we extract what the evidence anchors, not everything a reader would notice.

What this reframes

01

Categories as hypotheses

interpretive commitments, held lightly and revised – not truth claims

02

Quality with contestation

bottom-up induction, bounded extraction, provenance – traceable, and still plural

03

Extraction as evaluation

the LLM as an instrument that tests the schema – not a consumer of it

And a lower wall between KOS designers and the communities whose knowledge is organised.

RESOURCES · OPEN SOURCE

Open code, open **ontology**

THE ONTOLOGY

medorah_ontology

github.com/articoder/medorah_ontology

THE WORKBENCH & TOOLKITS

medorah_nlp

github.com/articoder/medorah_nlp



 **MeDoraH**

idohl.org/medorah/about