

APPLIED DATA SCIENCE

# ONTOLOGY ALIGNMENT CLASSIFIER BASED ON NLP

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## PROJECT BACKGROUND

01

## ► GIVING STRUCTURE TO THE VOICE OF PATIENTS

Repertorio transforms the *voice of the patient* into **STRUCTURED EVIDENCE**

*Supporting:*

- Clinical research
- Pharmaceutical R&D
- Market access strategies

## ► THE CHALLENGES



### DATA FRAGMENTATION

Critical healthcare evidence is widely scattered across isolated studies, creating disconnected data silos



### SEMANTIC INCONSISTENCY

Different terms and formats limit the meaningful integration of heterogeneous data sources.

## ► FROM PATIENT EVIDENCE TO BETTER DECISIONS

Repertorio generates **Quantified Evidence** to support regulators, pharma, and researchers in key decisions.



### RISK-BENEFIT TRADE-OFFS

Optimizing clinical outcomes  
against safety profiles



### WILLINGNESS TO PAY

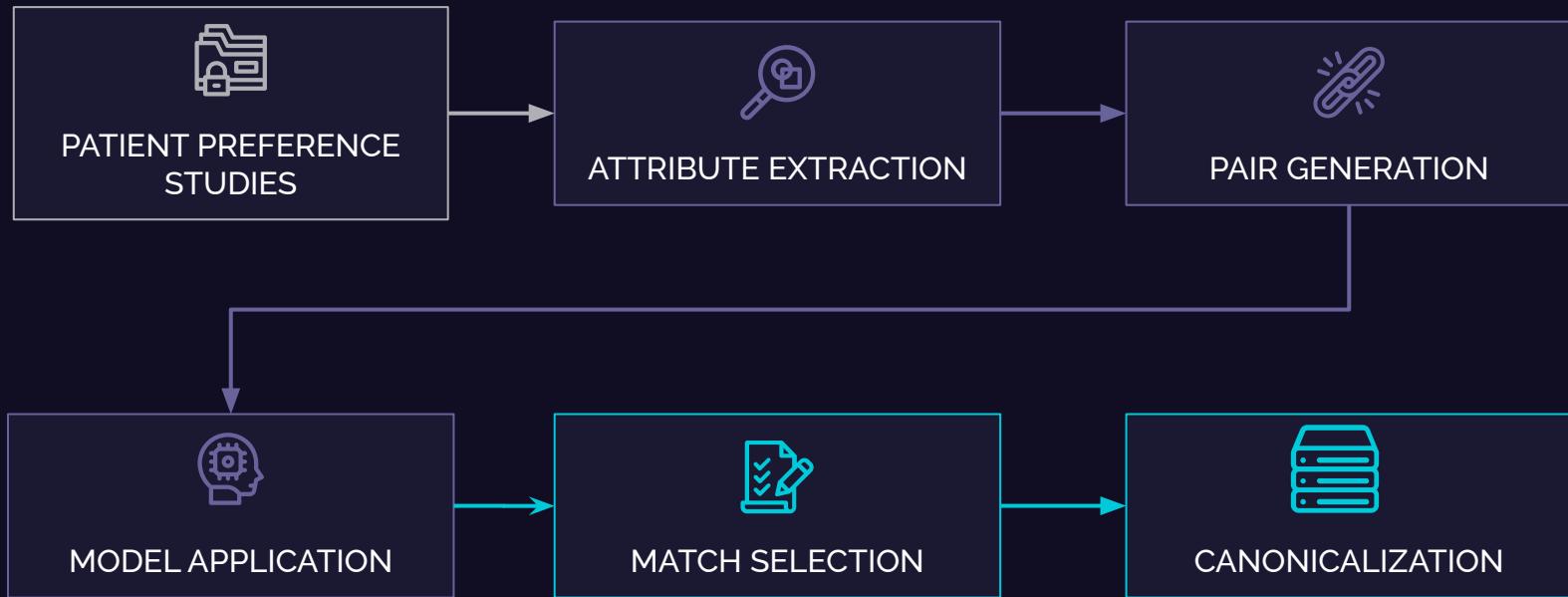
Quantifying value for  
market access



### MAXIMUM ACCEPTABLE RISK

Defining safety boundaries  
for regulators

## ► REPERTORIO FUNCTIONAL DIAGRAM





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## ► PROJECT VALUE PROPOSITION

02

## ► THE VALUE PROPOSITION

Developing an **AI-POWERED SEMANTIC ENGINE**  
that automatically harmonizes  
fragmented data.

Enabling Repertorio to scale the  
analysis of patient preferences for  
faster, **EVIDENCE-BASED DECISIONS**.





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## ► PROJECT GENERAL OBJECTIVE

03

# ► GENERAL OBJECTIVE



## BUILD THE AI ENGINE

Design and deploy the core AI model that aligns heterogeneous labels and concepts across ontologies



## SEMANTIC MAPPING

Provide reliable similarity scores that Repertorio will use to map raw attributes to canonical concepts in their internal ontology

## EXAMPLE

Fatigue  $\leftrightarrow$  Asthenia  $\leftrightarrow$  Lack of energy

# ► THE RESEARCH OBJECTIVE

Developing a Context-Aware Semantic Alignment Engine



## EVALUATE TRANSFORMER EFFICIENCY

Assess how models like BERT and S-BERT perform on complex biomedical data compared to traditional methods.



## OVERCOME LEXICAL LIMITATIONS

Move beyond simple string matching to capture the actual *meaning* and *context* of clinical terms.



## VALIDATE PERFORMANCE

Benchmark the new approach against established standards (OAEI) to ensure reliability.

# ► THE EXPECTED OUTCOME

## **Reliable Semantic Foundation**

Advanced NLP bridges terminology gaps, ensuring consistent, trustworthy data across the platform.

## **Advanced Analytics**

Unlocked by consistent data, the platform will offer enhanced search and meta-analysis for deeper insights

## **Stakeholder Value**

These capabilities will deliver essential indirect benefits for researchers, clinicians, and ultimately, patients.



A dark blue background featuring a light blue line that curves from the top left, passes through a small circle, and then turns right. Another line extends from the bottom right, passing through a circle, and then turns left. A central dark grey rectangle contains the number '04' in a large, light blue font. The overall aesthetic is minimalist and modern.

## ► PROJECT DESIGN

04

# ► THE STAKEHOLDERS

**Direct Stakeholders**  
Repertorio Technical Team

**External Stakeholders**  
Policy Makers  
HTA Bodies  
Patients

**Indirect Stakeholders**  
Academic Researchers  
Pharma & Biotech  
Clinical and R&D Units  
Healthcare Professionals



# ► THE PERSONA

## DEMOGRAPHICS

45 years old • Turin, Italy  
Senior Data Engineer

## TASKS & HABITS

- Writes Python scripts daily.
- Aligns continuously with R&D.

## PAIN POINTS

- Frustrated by brittle scripts.
- Manual data alignment is a bottleneck.



**JOHN SMITH**  
**LEAD DATA ENGINEER**

## NEEDS AND GOALS

- Automated, scalable alignment pipeline.
- Stop being the manual bottleneck.

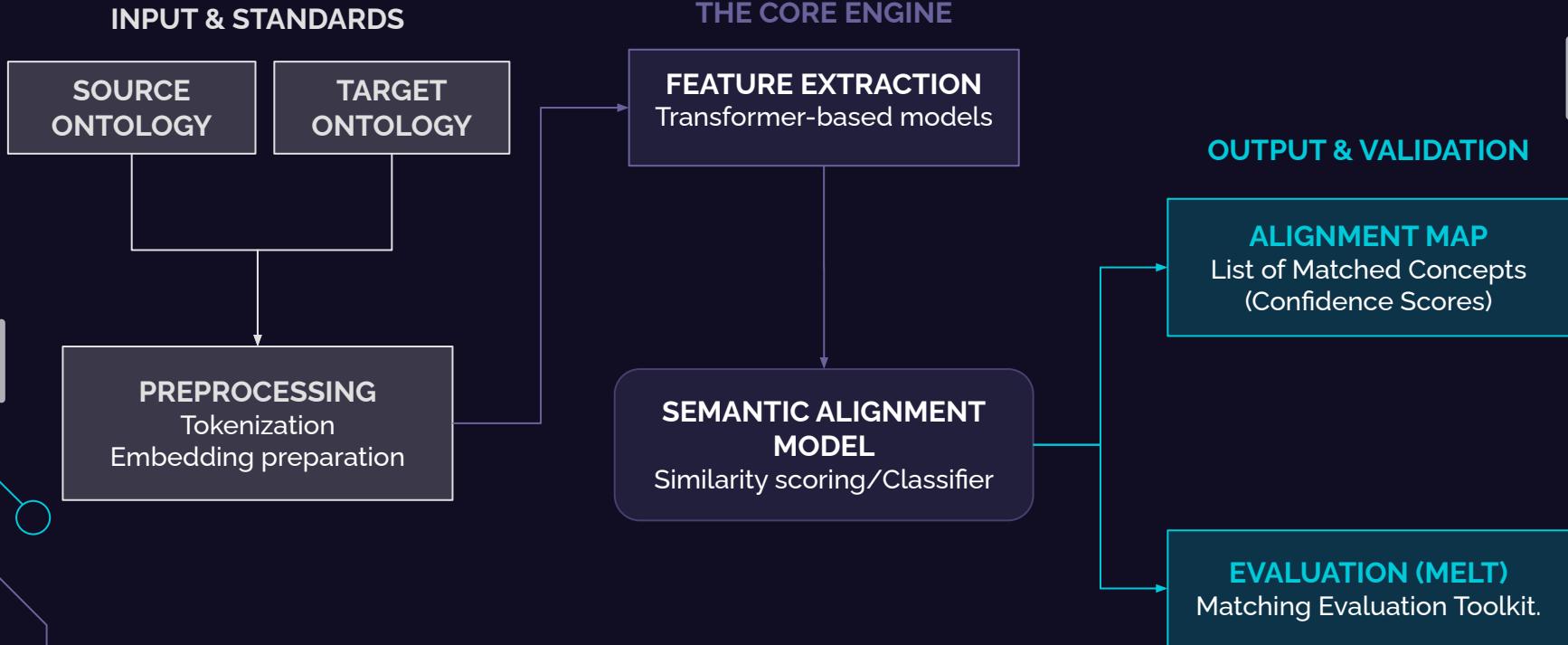
## EXPECTED OUTPUT

- Reliable AI Model (API).
- Clear semantic match scores.

## DATA USAGE

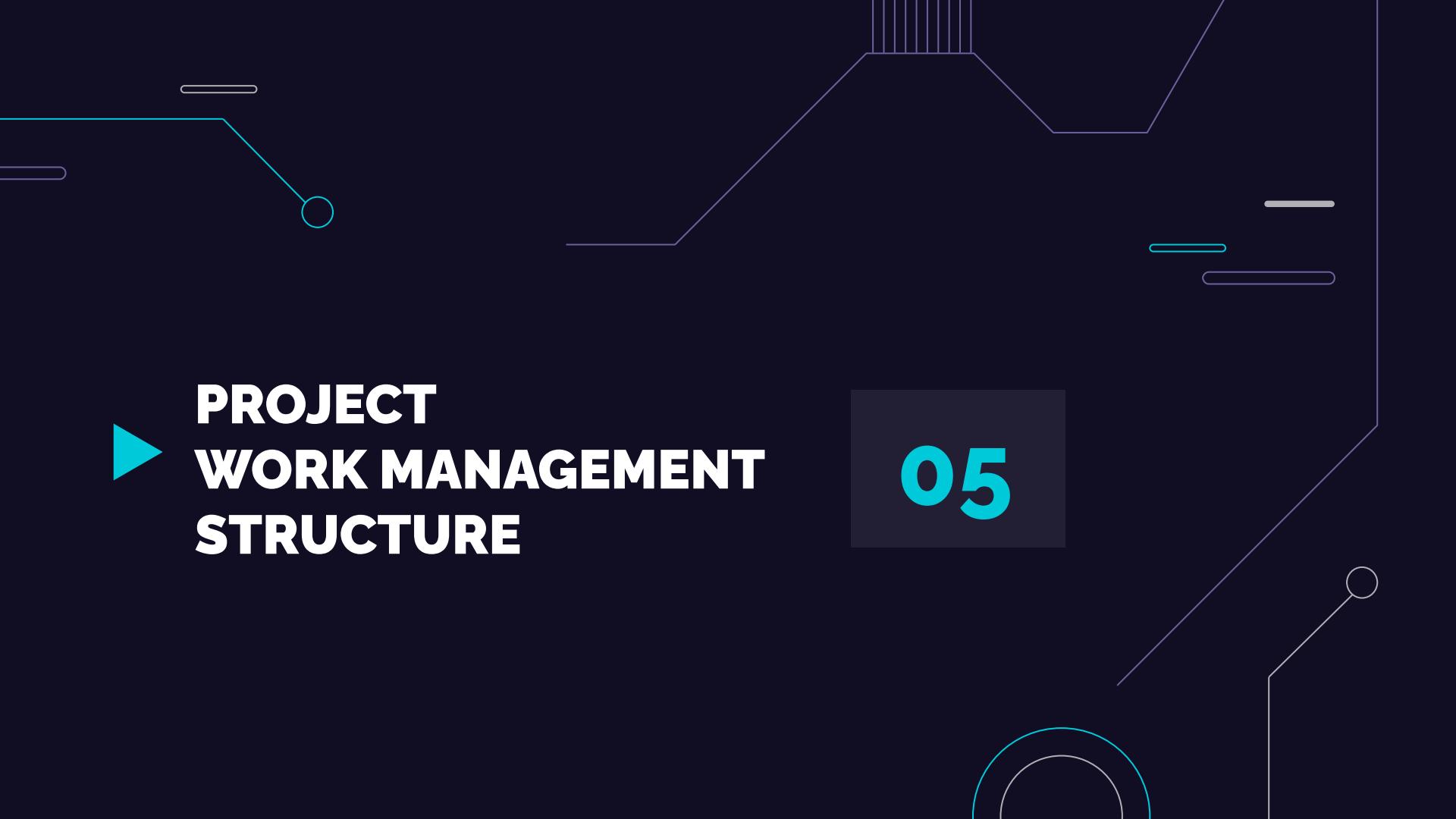
- Input: Messy OAEI / OWL files.
- Output: Clean Knowledge Graph.

# ► SYSTEM FUNCTIONAL DIAGRAM



# ► SYSTEM REQUIREMENTS

	FUNCTIONAL	NON-FUNCTIONAL
MUST HAVE	<p><i>Load</i>: the system must load an OAEI dataset (a track folder).</p> <p><i>Build</i>: the system must automatically build a dataset of pairs (concept A, concept B) and their associated text/graph features, ready for training.</p> <p><i>Train</i>: the system must allow training a binary classifier for ontology alignment.</p> <p><i>Evaluate</i>: the system must be able to evaluate the model on a test set and produce standard metrics.</p> <p><i>Generate</i>: the system must be able to generate an alignment file in one of the standard formats (e.g., Alignment API RDF) starting from the model's predictions</p>	<p><i>Usability/Hardware Accessibility</i>: the system must be executable on machines with standard GPUs.</p> <p><i>Reproducibility</i>: The system must be reproducible: same inputs, same parameters =&gt; same results.</p>
SHOULD HAVE	<p><i>Compare</i>: the system should implement and compare semantic encoding using domain-specific Language Models (e.g., BioBERT, SciBERT) against generic models (e.g., RoBERTa).</p>	<p><i>Transparency</i>: the system should be transparent regarding model version, OAEI tracks used for training, main hyperparameters.</p>
COULD HAVE	<p><i>Inspect</i>: the Domain Expert could be able to view and inspect the proposed alignments, with clear visibility of ambiguous cases.</p>	<p><i>Modularity</i>: the codebase could be structured into distinct modules, allowing the Repertorio team to retrain the model on future internal data.</p>
WON'T HAVE	<p><i>Collect</i>: the system won't collect questionnaires or collect raw patient preference data from hospitals or clinics.</p>	<p><i>Real-time inference</i>: the system won't require real-time processing.</p>



# ► PROJECT WORK MANAGEMENT STRUCTURE

05

# ► THE WORK PACKAGES OVERVIEW

## WP1 — Domain Understanding & High-Level Design

- **T1** → Stakeholder, persona & pipeline analysis
  - **T2** → Requirements definition + High-Level Functional Diagram
  - **T3** → Dataset inspection & MELT standards
  - **T4** → Management Plan Construction
- 

## WP2 — Data Preparation & Dataset Builder

- **T1** → OAEI Loader (ontologies + reference alignment)
  - **T2** → Dataset Builder (positive + negative samples)
  - **T3** → Train/Val/Test split with leakage prevention
- 

## WP3 — Literature Review & Architecture Selection

- **T1** → Literature review
- **T2** → Baseline implementation (TF-IDF + cosine)
- **T3** → Comparative evaluation (precision/recall/F1)
- **T4** → Final architecture selection

## WP4 — Model Training, Tuning & Evaluation

- **T1** → Model implementation & training pipeline
  - **T2** → Hyperparameter tuning and error analysis
  - **T3** → MELT-compatible alignment file generation
- 

## WP5 — Packaging & Final Presentation

- **T1** → Code packaging (modular repo: loader, builder, baseline, model)
- **T2** → Technical documentation + inference examples
- **T3** → Final report + slide deck (Checkpoint 3)
- **T4** → Public release of the full reproducible pipeline

# ► WORK PACKAGES TIMELINE

WP	OBJECTIVE	Weeks									
		1	2	3	4	5	6	7	8	9	10
WP1	Domain understanding & conceptual design	●	●								
WP2	Data loading tools & training dataset construction			●	●						
WP3	SOTA analysis, baseline definition & model selection			●	●	●	●				
WP4	Model implementation, optimization & OAEI/MELT-ready output				●	●	●	●			
WP5	Final delivery, replicability & reporting					●	●	●	●		

# THANK YOU

ANY QUESTIONS?