

Hybrid Human-Al Decision Support for Enhanced Human Empowerment in Dynamic Situations

Human-in-the-Loop AI for Trustworthy Decision Support: Insights from the HumAIne Pilots

ENFIELD Workshop on Human-Centric Al Bucharest, Romania (Hybrid) | 09/09/2025 Vassilis Voulgarakis (OKYS)





The HumAlne project in a nutshell



A 3-year Horizon Europe initiative dedicated to pioneering the collaboration between humans and Alsystems for decision-making in **dynamic, unstructured environments**.

o The project aims to empower AI solution integrators through **the HumAIne Operating System (OS)** a platform that:

- ✓ integrates cutting-edge technologies into a single system to surpass both standalone AI and Human capabilities and
- ✓ transforms the landscape of decision-making and problem-solving across five 'Smart Industry' sectors.









































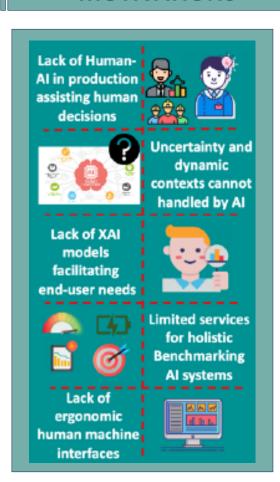
The project's scope



MOTIVATIONS

SOLUTION

EXPECTED IMPACTS



HumAlne Innovative Objectives & Ambition Unified platform for Human-

centred AI decision-making From single-aspect Human-Al to holistic & dynamic human-machine collaboration



Neuro-Symbolic AI for truly human and energy efficient Al Efficient completion of complex tasks combining AI with semantic reasoning



semantics

Active Learning for effective

Swarm Learning for collective

human intelligence

Increased Al accuracy and human

empowerment with collective

decision-making



Human-Machine Interfaces Enhanced visualisations of Al outcomes interaction between human and machines, and shared knowledge.





Benchmarking Suite for Human-Al systems

Multi-aspect benchmarking of **Human-Al applications**



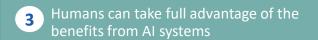
Novel Human-Al pilots and **Use Cases**



Validation of HumAlne KERs in real-life use cases from various high impact domains

Supporting a human-centered and ethical development of digital industrial technologies







Accurate and trusted predictions of HumAlne systems leveraging domain knowledge based on Neuro-Symbolic AI



Keynote's fundamental thematic





Al may risk replacing human judgment if not designed carefully.

- The key issues revolve around:
 - Trust by human factors
 - Transparency of the AI-systems/models
 - Accountability of the Al-systems/models
- So...why human-in-the-loop?
- Affords human factors to take control when AI confidence is low
- Ensure contestability and interpretability of AI-systems/models' outcomes
- ✓ The experts' feedback adapts systems/models dynamically

Al should augment rather than replacing human decisions!

> Human intervention is a design feature, not an afterthought!



The HumAlne OS and why it matters



✓ Active Learning

Humans intervene in AI processes when faced with increased uncertainty

✓ Neuro-Symbolic Learning

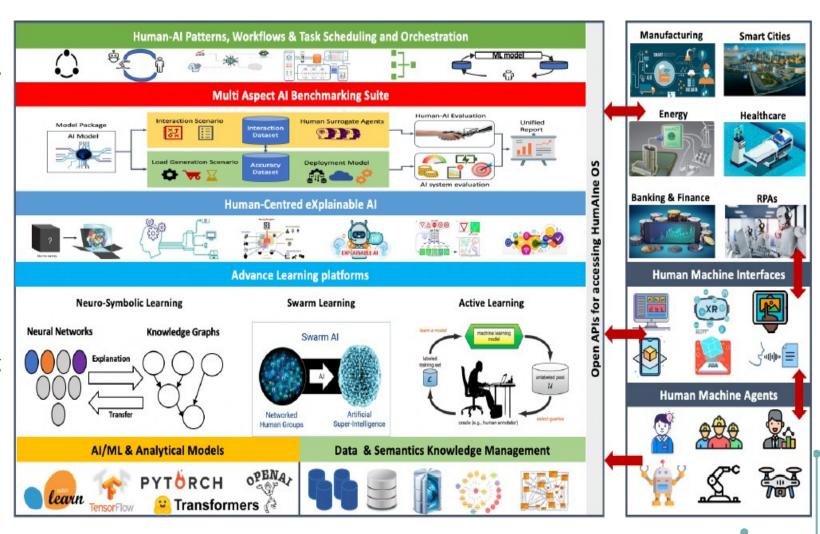
Deep Learning is combined with semantics and rules to complete complex tasks with high accuracy and minimal training

✓ Swarm Learning

Al models are trained in a decentralized and collaborative way across nodes without sharing raw data (privacy preserved)

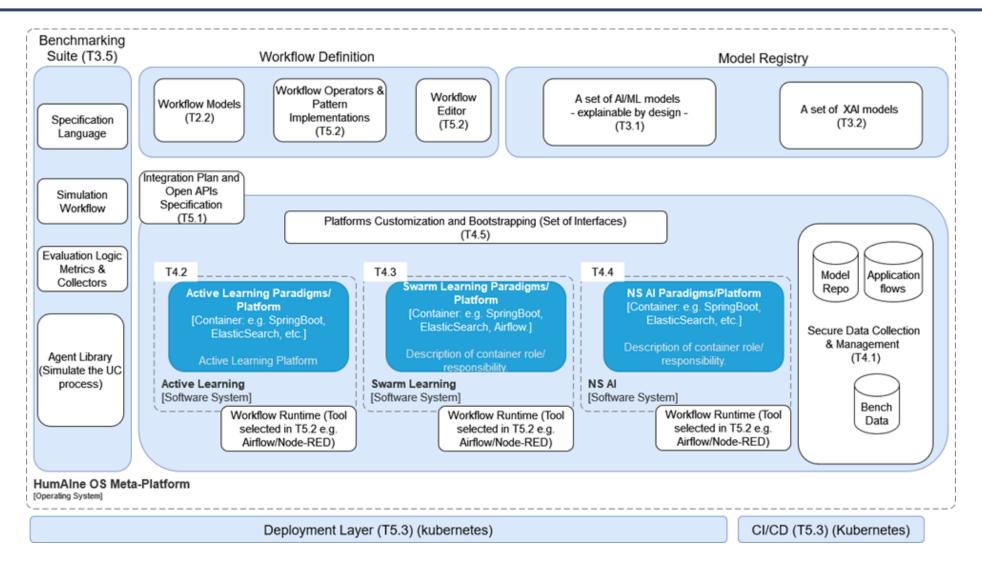
✓ Explainable AI (XAI)

XAI models deliver holistic predictions and allow for users' feedback



The HumAlne OS Architecture







Transforming decision-making across Smart Industries hum Aine





Smart Manufacturing

Scheduling in a Machine-as-a-Service scenario



Smart Energy

Al-assisted decision support tools for system operators



Smart Cities

Robotic automation of processing parking permits



Smart Ticketing

Automating technical ticket resolution



Smart Healthcare (Diabetes)

Living at home without exacerbations & rehospitalisations



Smart Healthcare (Oncology)

Managing patient workflow with Al-assisted tools



Smart Manufacturing Pilot



The context

Challenge:

- Complex production lines with modular tasks need optimal scheduling. Traditional scheduling methods are time-consuming and lack flexibility.
- Data ownership and privacy are key challenges in industrial settings.

AI Paradigm: Swarm Learning

 Distributed machine learning that keeps data local, ensuring privacy. Al model training is decentralized, with each machine as an agent.

Human Role:

• Mixed Reality (MR) UI ensures transparency and allows human operators to adjust goals and oversee AI predictions in real-time.

The Human-in-the-Loop in action!

Swarm agents learn from local data without transferring sensitive information.

MR Interface allows real-time human intervention, adjusting scheduling objectives as needed.

Efficiency improvement: Human oversight ensures accurate alignment with operational goals while maintaining privacy.

Privacy-preserving AI provides valuable insights without compromising data security.



Smart Energy Pilot



The context

Challenge:

 Power grid operations are increasingly complex due to renewable energy integration. Operators need real-time insights to make decisions based on accurate, interpretable predictions.

Al Paradigm: Active Learning + XAI

- Al helps forecast grid states and identify risks.
- Digital Twin simulates scenarios for decisionmaking, while XAI explains AI predictions.

Human Role:

- Operators interact with the AI outputs, leveraging AR/VR HMI for immersive decision support.
- Human feedback ensures that predictions and actions are interpreted correctly in dynamic grid conditions.

The Human-in-the-Loop feature

Digital Twin simulates the grid and helps operators visualize potential scenarios.

XAI explains forecasting errors and provides confidence levels for predictions.

AR/VR HMI allows operators to interact with simulations and AI insights, guiding manual interventions as needed.



Smart Cities Pilot



The context

Challenge:

 Manual processing of resident permit applications is labor-intensive and prone to errors. There's a need for automation, but with human supervision to ensure accuracy.

AI Paradigm: Active Learning

 The system continuously learns from operator feedback to improve accuracy. All assists in identifying fields in PDFs, automating the data validation process.

Human Role:

- Council operators validate AI output and annotate training data, making corrections where necessary.
- Expert feedback directly influences AI learning cycles and improves model precision over time.

The Human-in-the-Loop feature

Active Learning enables operators to intervene when Al uncertainty is detected.

Operators annotate fields in PDFs to correct and finetune AI predictions.

Automation: All automates routine validation checks, saving time and reducing errors.

Outcome: Processing time for a permit reduced from **20 minutes** to **10 seconds**, with humans ensuring ultimate control and accuracy.



Smart Ticketing Pilot



The context

Challenge:

- IT ticket systems face challenges in efficiently categorizing and resolving complex requests
- Human expertise is essential for effective dispatching, but AI can automate routine tasks to improve efficiency.

Al Paradigm: Active Learning + XAI

 Al continuously learns from human feedback to categorize tickets and suggest resolutions. XAI ensures that IT experts understand AI decisions and can intervene when necessary.

Human Role:

• IT experts validate AI predictions, help categorize new issues, and correct any errors improving AI performance over time.

The Human-in-the-Loop feature

Active Learning: Al improves by learning from IT expert feedback on ticket categorization.

XAI: Explains AI decisions, enabling experts to validate and correct outputs.

Outcome: AI models adapt continuously through expert input, reducing manual categorization time. Trust is built via transparent, explainable AI.



Smart Healthcare (Diabetes) Pilot



The context

Challenge:

 Personalized recommendations for diabetes management require real-time adaptation to patient needs. Traditional models rely on expert rules or data.

AI Paradigm: Neuro-Symbolic Learning

 Combines rule-based expert knowledge with data-driven AI learning for robust decision support. NSL helps predict outcomes based on past patient data and expert guidelines.

Human Role:

 Clinicians input expert rules and feedback on model predictions, ensuring personalized and safe care. The system adapts dynamically to clinician feedback, enhancing its accuracy over time.

The Human-in-the-Loop feature

NSL combines data-driven learning with expert feedback to continuously improve diabetes management strategies.

Clinicians can **intervene to adjust the system's decisions** based on individual patient circumstances.

Outcome: Early results show +10% accuracy improvement using NSL over baseline models, thanks to clinician input.



Smart Healthcare (Oncology) Pilot



The context

Challenge:

 Structured reporting in oncology is essential for diagnosis, but annotation is slow and inconsistent. Al models need to be accurate and interpretable, especially when dealing with health data.

Al Paradigm: Active Learning + XAI

 AL reduces manual labeling workload, and XAI improves transparency by explaining model decisions. XAI can flag uncertainty in AI-generated reports, so that radiologists maintain control.

Human Role:

 Radiologists provide input by annotating images and reports, so that AI models learn from correct examples. XAI highlights uncertainty, prompting

The Human-in-the-Loop feature

AL significantly reduces annotation load for clinicians by automating parts of the labeling process.

XAI ensures clinicians can verify AI outputs and identify potential issues with model predictions.

Outcome: Faster report generation, with clinicians still overseeing critical aspects to ensure accuracy and relevance.



Key Principles for Effective Human-in-the-Loop 1/2



Feedback Aggregation & Consolidation:

Collects human feedback across multiple interactions, grouping it to guide system learning.

✓ Example: In **Smart Cities**, operators annotate data in batches, improving AI accuracy over time.

Expert Consensus Consolidation:

When multiple experts provide feedback, the system consolidates diverse opinions to ensure a unified decision.

✓ Example: In Healthcare – Oncology, radiologists validate AI predictions, with any disagreement triggering further review and retraining.

Low-Confidence Alerts:

The system notifies humans when AI confidence drops below a threshold, prompting intervention.

✓ Example: In **Smart Manufacturing**, Al alerts operators when schedules are uncertain, allowing them to adjust goals in real-time.



Key Principles for Effective Human-in-the-Loop 2/2



Continuous Learning & Adaptation:

Al models adapt and improve through ongoing expert interaction, ensuring they evolve with real-world data and feedback.

✓ Example: In **Smart Energy**, operators adjust forecasts via AR/VR interfaces, guiding AI toward better predictions.

Transparency & Explainability:

All decisions are explained clearly to users, fostering trust and allowing for meaningful human intervention.

✓ Example: In **Smart Ticketing**, XAI explains categorization choices, ensuring IT experts can trust and correct the model's decisions.



Key takeaways & lessons learnt





Trust requires explainability and usable interfaces

Al models are trusted more when users understand how they make decisions and can interact with them in an intuitive way.

Adoption depends on empowering, not replacing experts

Al-systems/models perform best when human input helps guide learning, especially in situations of uncertainty or complex decision-making.

Human feedback must be integrated quickly and efficiently into AI models to continuously improve performance without overwhelming users.

Scalability and Flexibility are essential for successful integration

Human-in-the-loop systems must be adaptable and scalable across various industries to meet the diverse needs of different sectors





Hybrid Human-Al Decision Support for Enhanced Human Empowerment in Dynamic Situations

Thank you for your attention!





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