



German  
Research Center  
for Artificial  
Intelligence

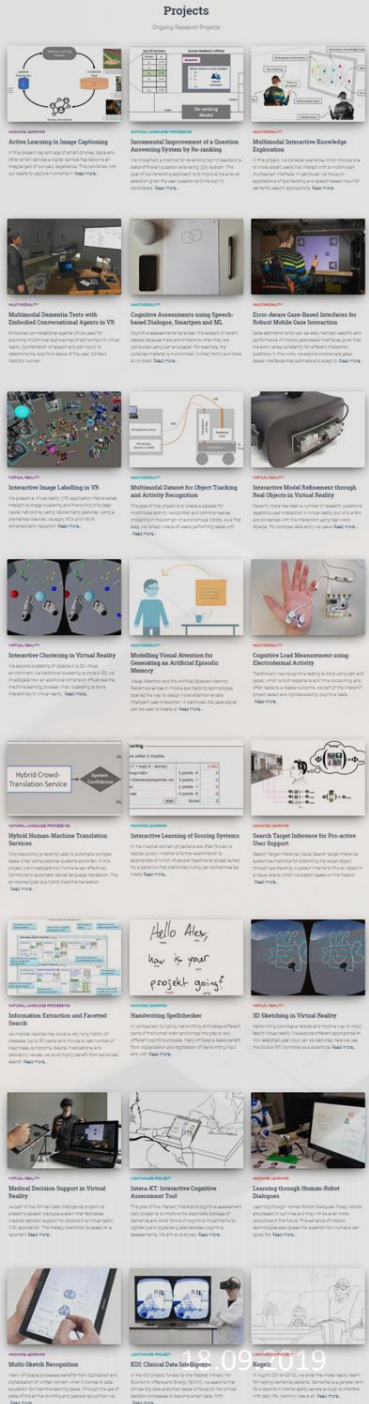


# Interactive and Explainable Machine Learning

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# Interactive Machine Learning Lab

 [iml.dfki.de](http://iml.dfki.de)



# People

## Academic Research

**Researcher**




German Research Center for Artificial Intelligence



**PhD Student**




SAARBRÜCKEN GRADUATE SCHOOL OF COMPUTER SCIENCE



Daniel Sonntag  
PRINCIPAL RESEARCHER & RESEARCH FELLOW






Hans-Jürgen Profitlich  
SENIOR SOFTWARE ENGINEER








Fabrizio Nunnari  
SENIOR RESEARCHER






Michael Barz  
RESEARCHER, PHD STUDENT


Alexander Prange  
RESEARCHER




Rajarshi Biswas  
RESEARCHER



Marimuthu Kalimuthu  
RESEARCHER



Mira Niemann  
RESEARCH ASSISTANT

Cognitive assessments, analysis of speech and electrodermal activity.



Muaz Muhammad  
RESEARCH ASSISTANT

Interactive machine learning in VR



Divyam Saran  
RESEARCH ASSISTANT

Multimodal human-machine interaction



Maximilian Diversi  
RESEARCH ASSISTANT

Web-UI development and data



Carolin Grieser  
RESEARCH ASSISTANT

Cognitive assessments



Abraham Ezema  
RESEARCH ASSISTANT



Duy Nguyen  
RESEARCH ASSISTANT



Haris Isbal  
RESEARCH ASSISTANT

**NATURAL LANGUAGE PROCESSING**  
**Incremental Improvement of a Question Answering System by Re-ranking**  
We implement a method for re-ranking top-10 results of a state-of-the-art question answering (QA) system. The goal of our re-ranking approach was to improve the answer selection given the user question and the top-10 candidates. We focus on improving deployed QA systems that do not allow re-ranking or which re-ranking [Read more...](#)

**MULTIMODALITY**  
**Multimodal Interactive Knowledge Exploration**  
In this project, we consider scenarios which include one or more expert users that interact with a multimodal/multitouch interface. In particular, we focus on explorations of eye tracking and gesture-based input for semantic search applications, e.g., interactive knowledge exploration. We aim at developing interaction techniques for individual modalities, combinations thereof [Read more...](#)

**MULTIMODALITY**  
**Error-Aware Gaze-Based Interfaces for Robust Mobile Gaze Interaction**  
Gaze estimation error can severely hamper usability and performance of mobile gaze-based interfaces given that the error varies constantly for different interaction positions. In this work, we explore error-aware gaze-based interfaces that permit and adapt to gaze estimation error implicitly. We implement a simple error-aware user interface for gaze-based selection and [Read more...](#)

**MULTIMODALITY**  
**Modelling Visual Attention for Generating an Artificial Episodic Memory**  
Visual Attention and the Artificial Episodic Memory Research advances in mobile eye-tracking technologies opened the way to design novel attention-aware intelligent user interaction. In particular, the gaze signal can be used to create or improve artificial episodic memories for offline processing, but also real-time processing. We investigate different methods [Read more...](#)

**NATURAL LANGUAGE PROCESSING**  
**Hybrid Human-Machine Translation Services**  
Crowdsourcing is recently used to automate complex tasks when computational systems alone fail. In this project, we investigate how humans can effectively contribute to automate natural language translation. The envisioned goal is a hybrid machine translation service that harmonically adapts machine translation models to new domains by employing human computation to [Read more...](#)

**MACHINE LEARNING**  
**Search Target Inference for Pro-active User Support**  
Search Target Inference Visual Search target inference subsumes methods for predicting the target object through eye tracking. A person intends to find an object in a visual scene which we predict based on the fixation behavior. Knowing about the search target can improve intelligent user interaction. Introducing the tag of [Read more...](#)

**VIRTUAL REALITY**  
**Medical Decision Support in Virtual Reality**  
As part of the Clinical Data Intelligence project we present a speech dialogue system that facilitates medical decision support for doctors in a virtual reality (VR) application. The therapy graduation is based on a recurrent neural network model that incorporates the examination history of patients. A central supervised patient database [Read more...](#)

**MACHINE LEARNING**  
**Learning through Human-Robot Dialogues**  
Learning through Human-Robot Dialogues Today, robots are present in our lives and they will be even more ubiquitous in the future. The advance of robotic technologies also poses the question how humans can guide the behavior of a robot and, in particular, how non-experts can advise robots to learn new [Read more...](#)



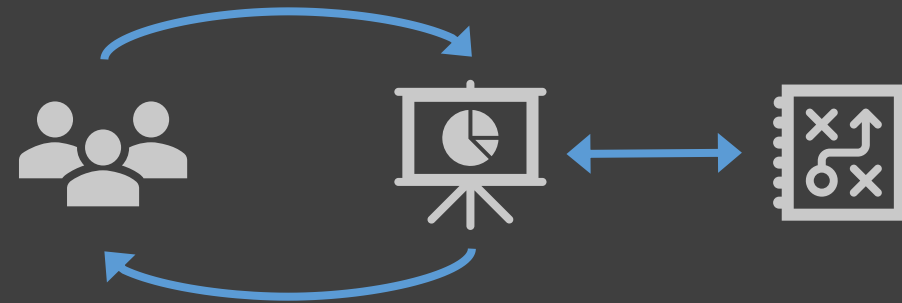
What is it and why is it important?

# Interactive and Explainable Machine Learning

# Interactive Machine Learning

Intelligent User Interface

Algorithms

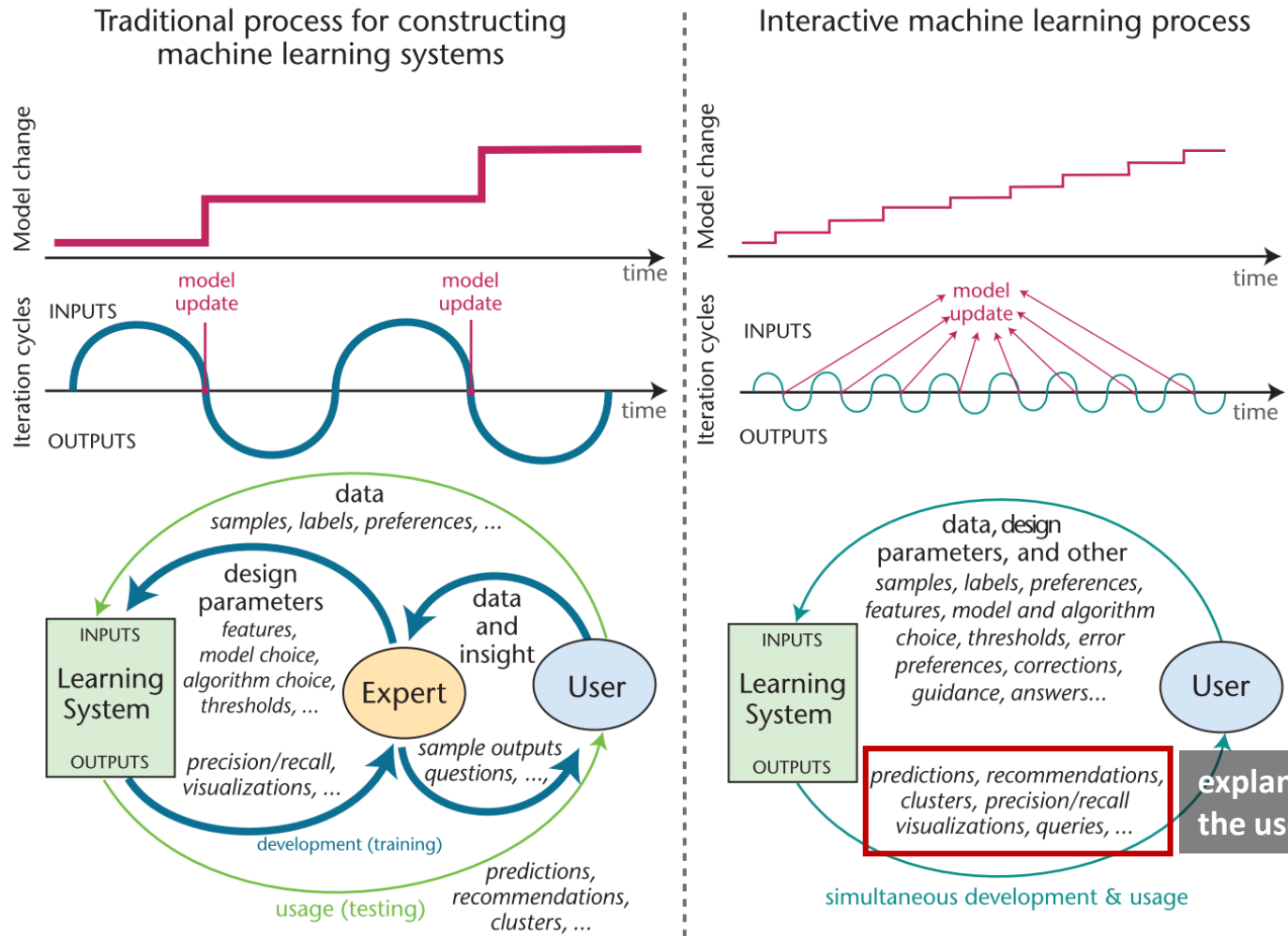


Interactive Machine Learning (IML) is the design and implementation of algorithms and intelligent user interface (IUI) frameworks that facilitate machine learning (ML) with the help of human interaction

**Mixed Initiative Interaction**

The ML system and the domain expert engage in a two-way dialogue.

# Interactive Machine Learning



Amershi, S., Cakmak, M., Knox, W. B., & Kulesza, T. (2014). Power to the People: The Role of Humans in Interactive Machine Learning. *AI Magazine*, 35(4), 105. <https://doi.org/10.1609/aimag.v35i4.2513>

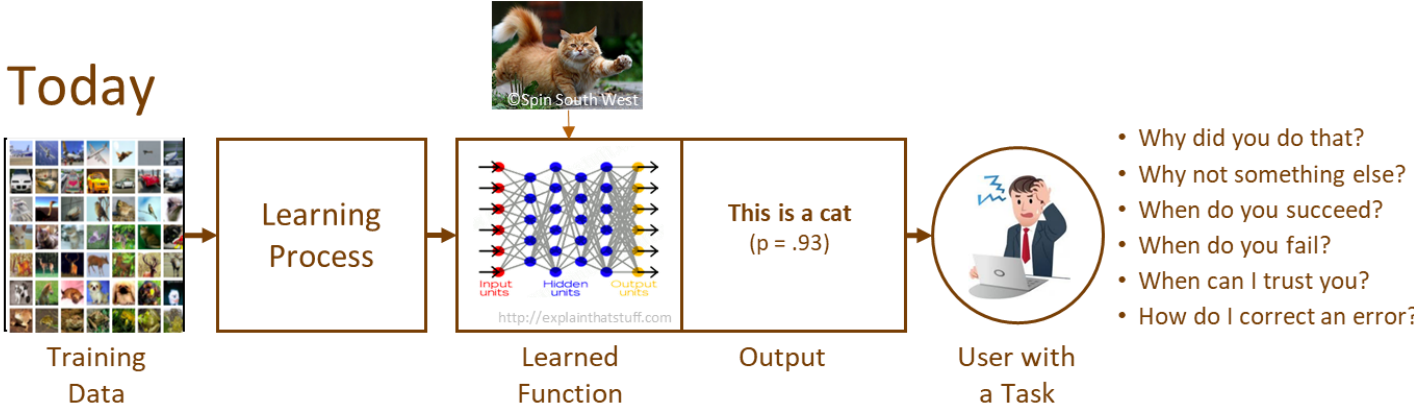
# Explainable Machine Learning

- Mitigate bias in ML
- Evaluate machine learning models
- Render ML models more interpretable
- Interpretation of black-box models
- Enhance trust in ML
- Causality of predictive models
- Visual analytics

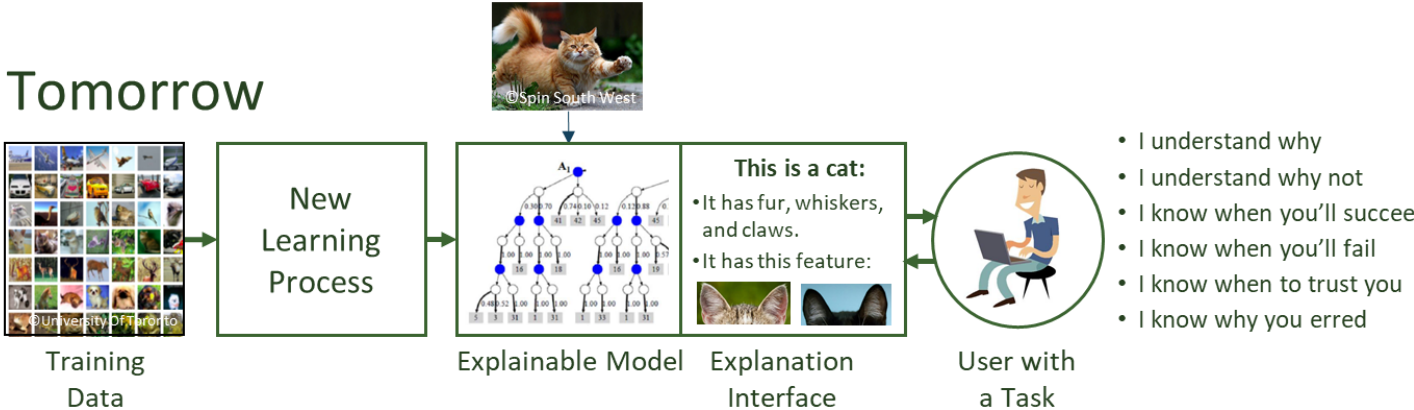
Explainable Machine Learning is a subfield of explainable artificial intelligence (XAI) which concerns transparency and human interpretability in machine learning (ML).

# Explainable Machine Learning

## Today



## Tomorrow

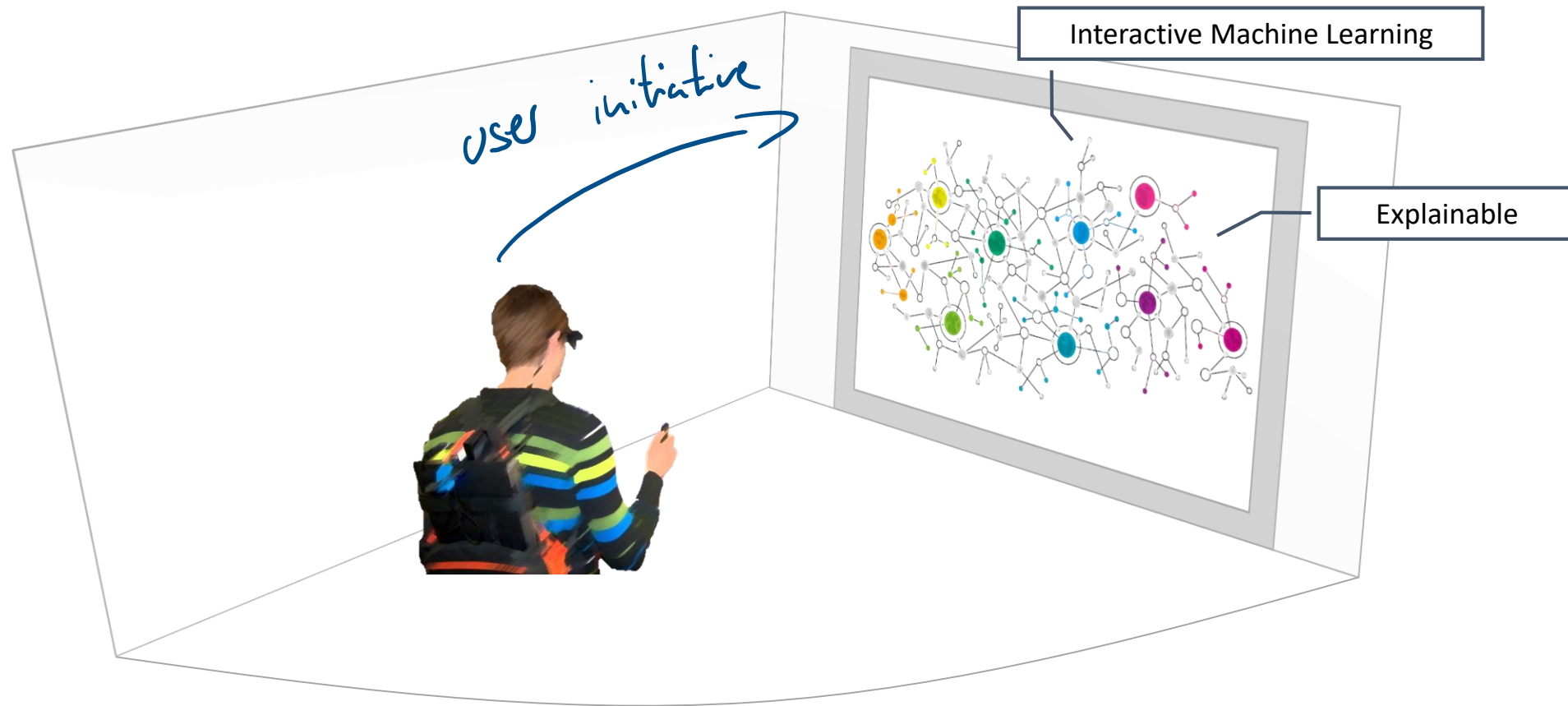


Gunning, D., & Aha, D. (2019). DARPA's Explainable Artificial Intelligence (XAI) Program. AI Magazine, 40(2), 44–58. <https://doi.org/10.1609/aimag.v40i2.2850>





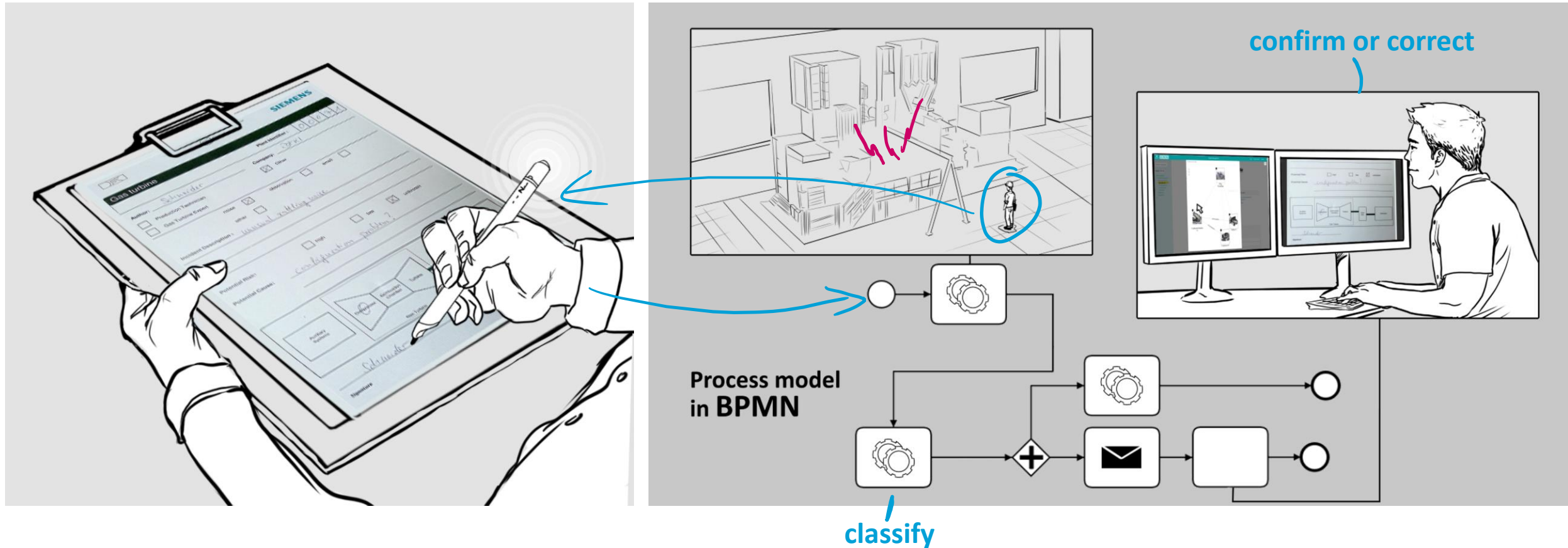
# IML and XAI Systems: Application Sample 1



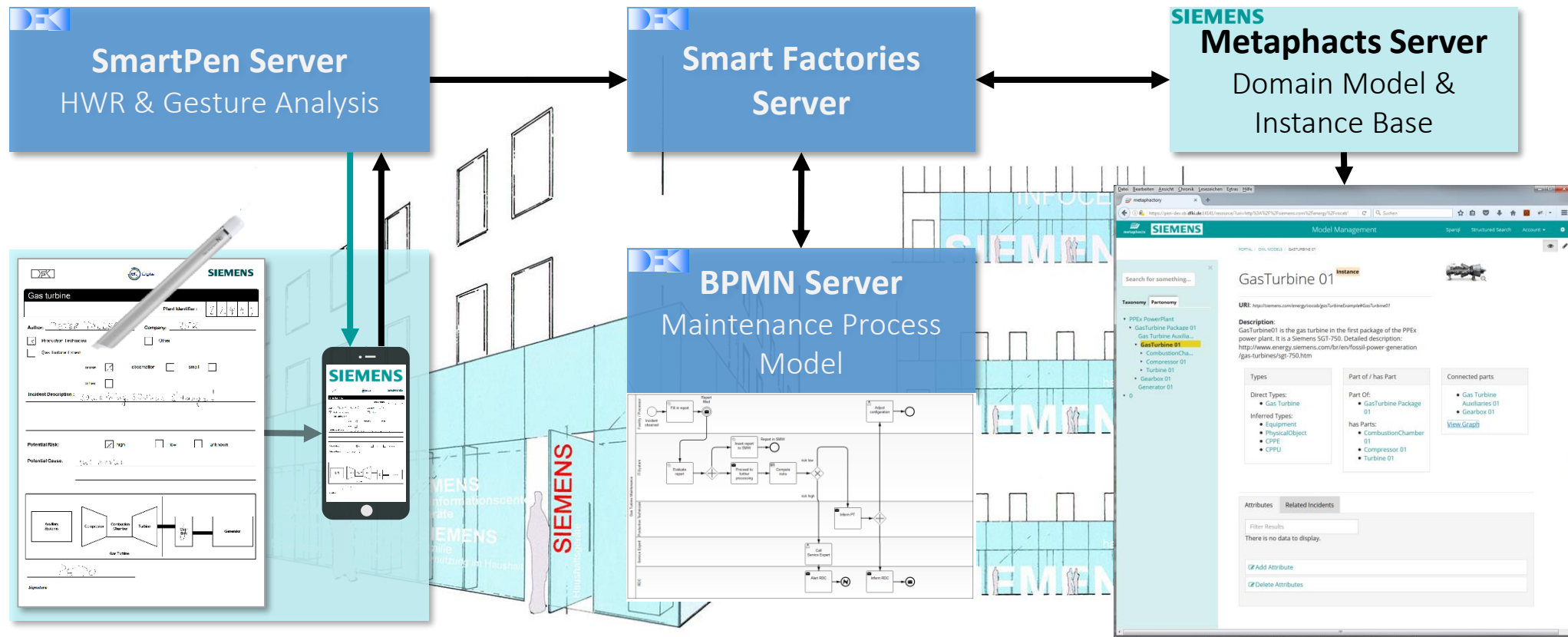
user initiative

# Interactive Report Analysis in Smart Factories

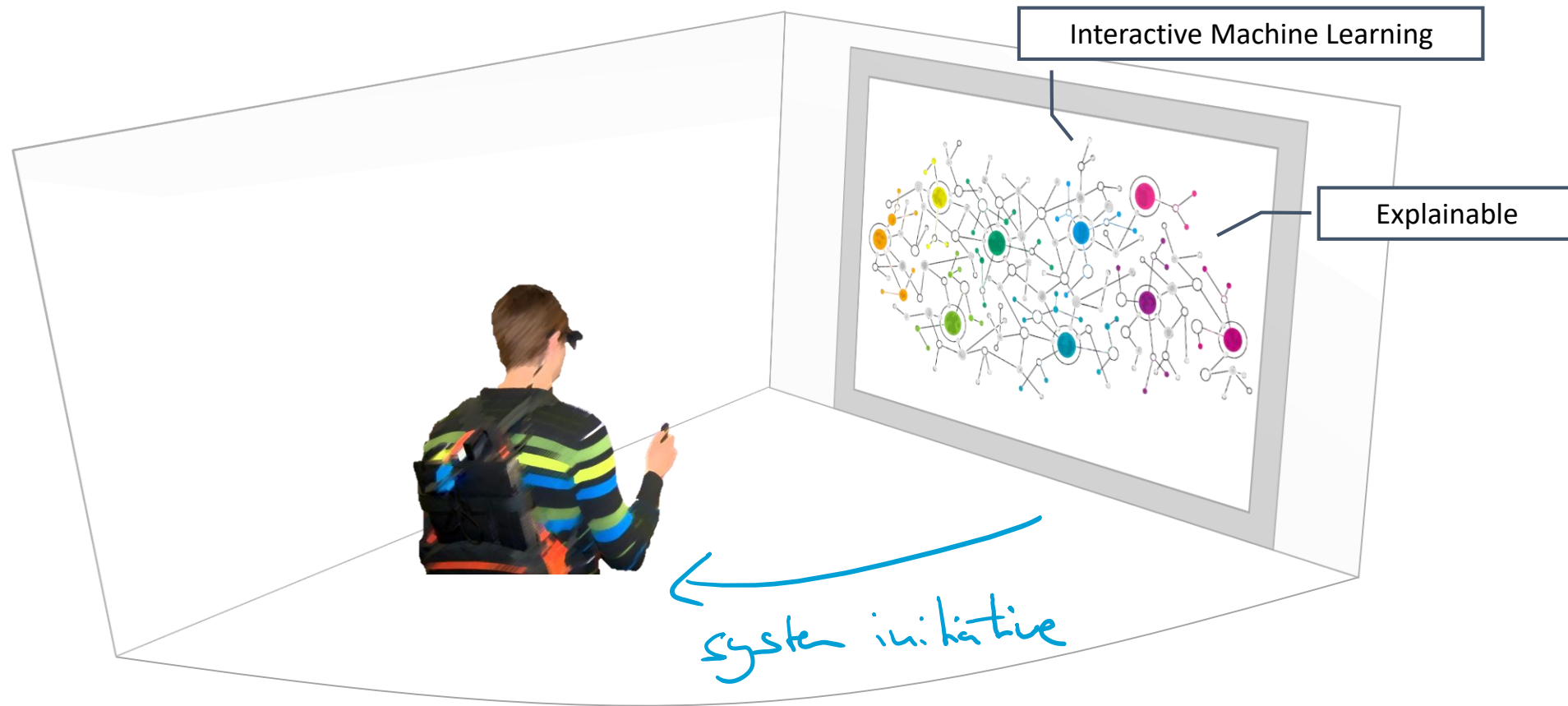
Barz, M., Poller, P., Schneider, M., Zillner, S., & Sonntag, D. (2017). Human-in-the-Loop Control Processes in Gas Turbine Maintenance. In V. Marik, W. Wahlster, T. I. Strasser, & P. Kadera (Eds.), *Industrial Applications of Holonic and Multi-Agent Systems - 8th International Conference, HoloMAS 2017, Lyon, France, August 28-30, 2017, Proceedings* (pp. 255–268). [https://doi.org/10.1007/978-3-319-64635-0\\_19](https://doi.org/10.1007/978-3-319-64635-0_19)



# Interactive Report Analysis in Smart Factories

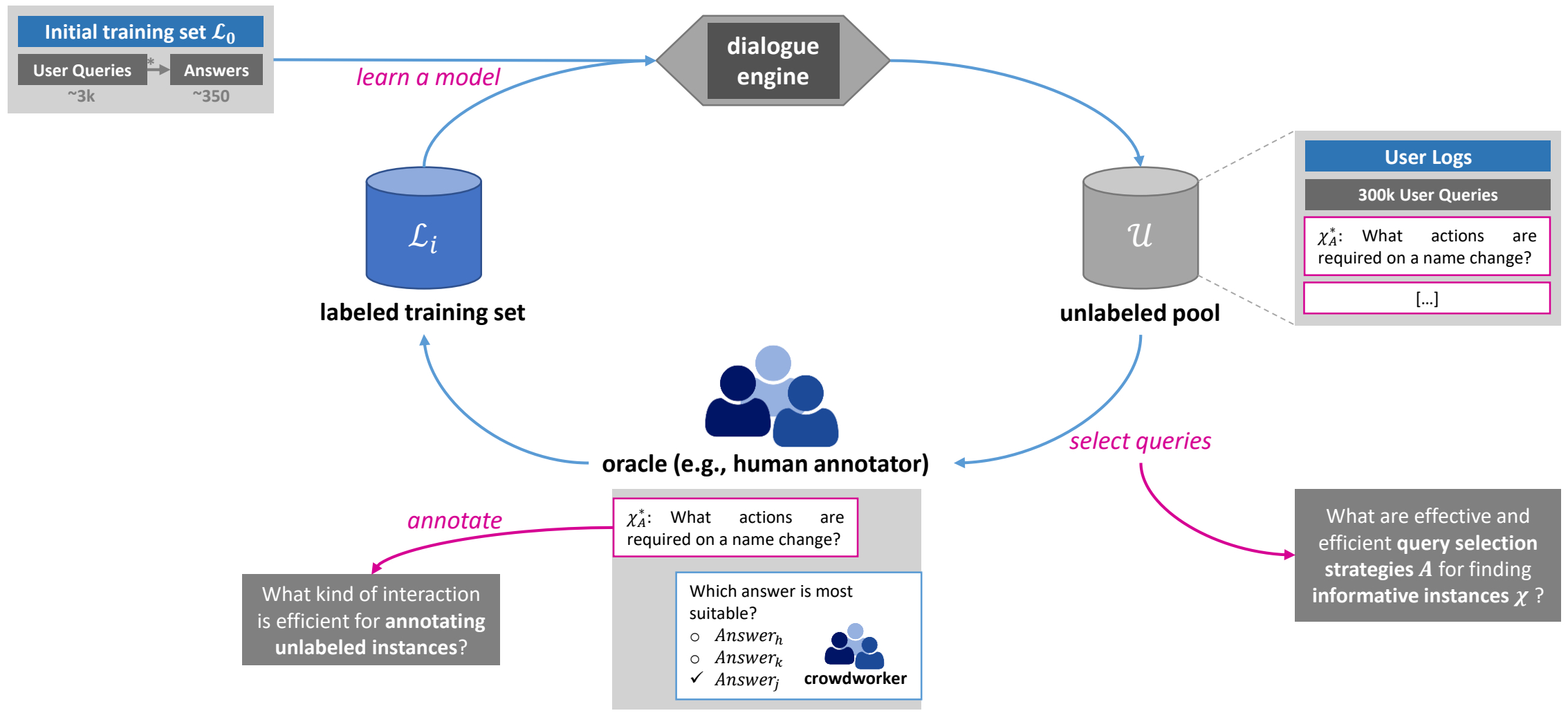


# IML and XAI Systems: Application Sample 2

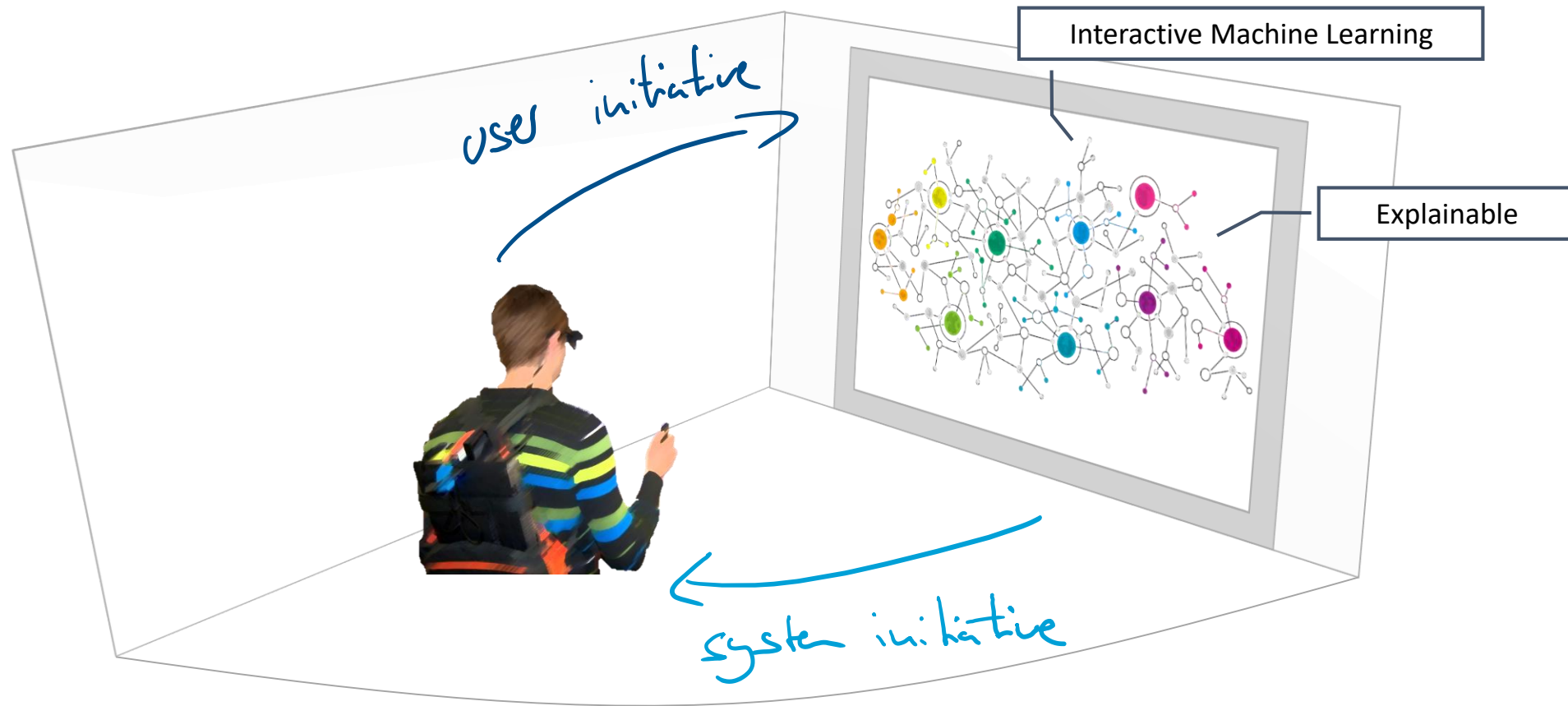


← system initiative

# Improve CRM Chatbot using Active Learning



# IML and XAI Systems: Application Sample 3



*user initiative*

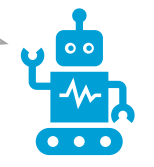
“Why did you predict dog?”



### visual explanations for golden\_retriever



“Is it a cat or a dog?”



*system initiative*

Class	Probability
golden_retriever	0.334
cocker_spaniel	0.203
sussex_spaniel	0.132
tabby_cat	0.047
otterhound	0.041

*dog*  
*cat*



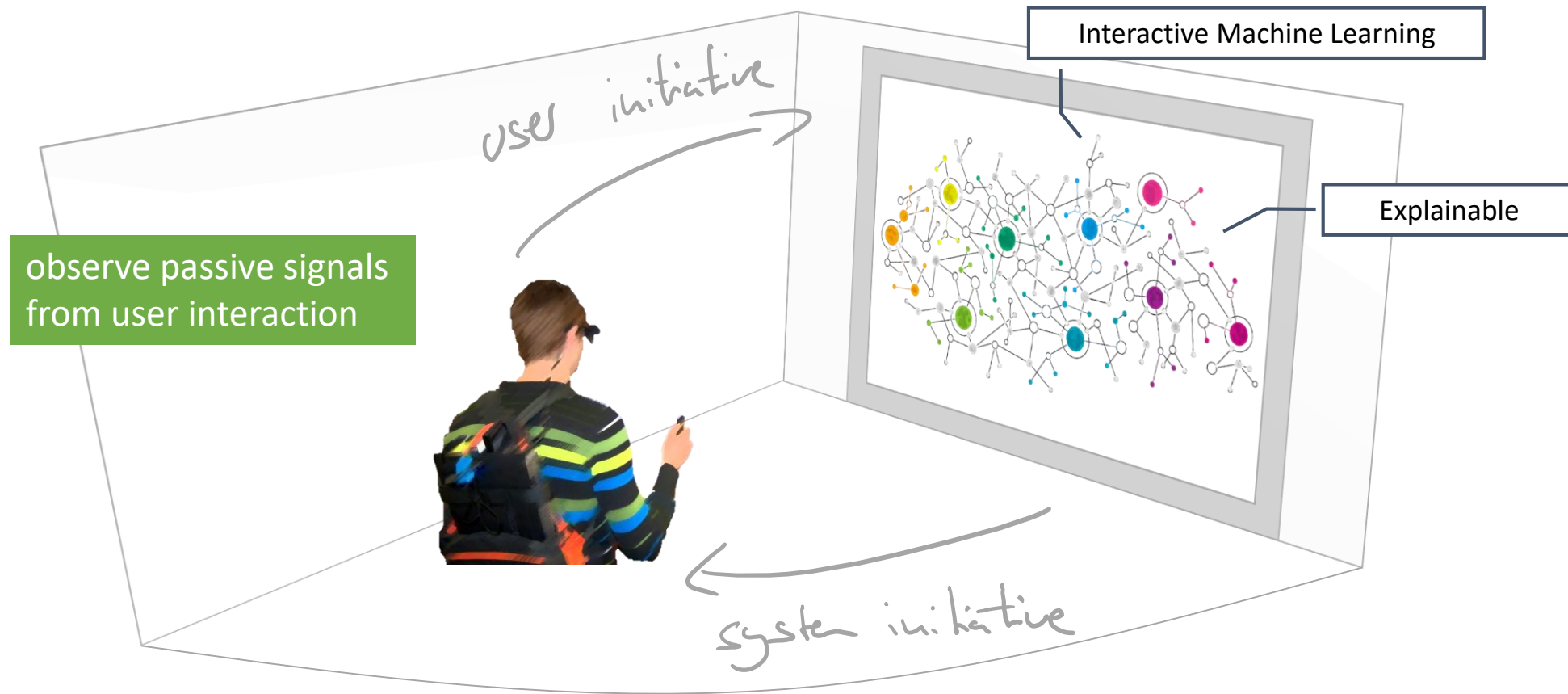
### visual explanations for tabby\_cat

**mixed initiative dialogue about the model**

can be used to resolve uncertainties of the model about its outcomes and of the user about how the model works



# IML and XAI Systems: Application Sample 4



# Passive collection of training data for document retrieval

learn to classify document relevance by observing expert users

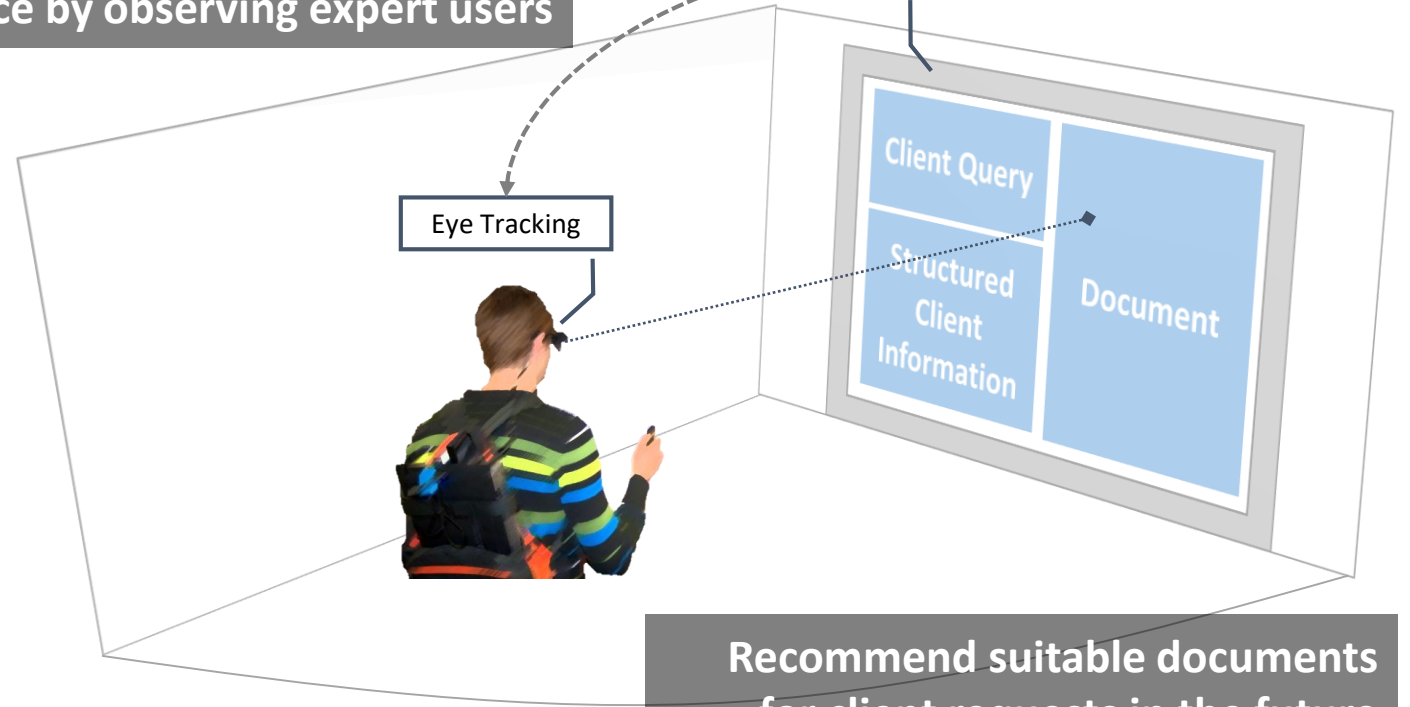
Interactive Knowledge Exploration

Eye Tracking

Task | Answer Client Query

Which document is relevant?

Which additional information is relevant?



Recommend suitable documents for client requests in the future

# Summary

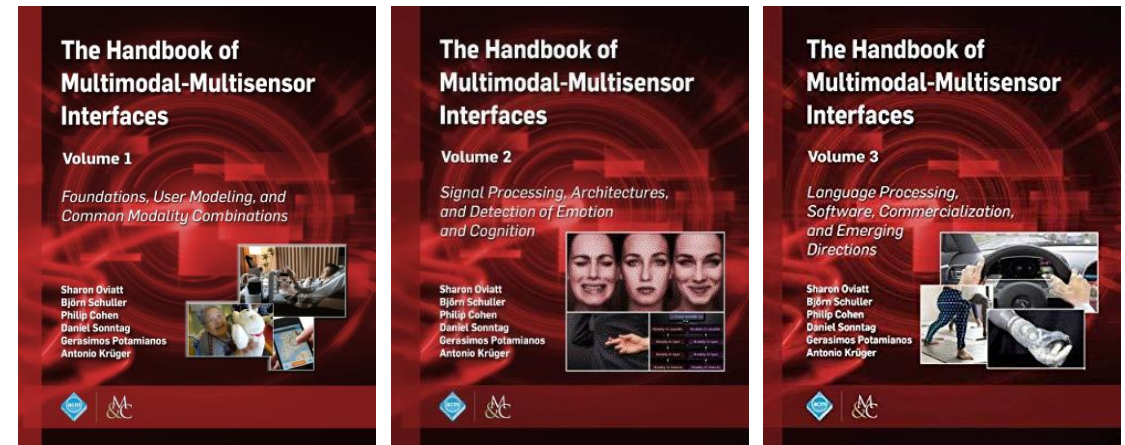
## Take Away

In an IML / XAI system, the user and the system benefit from each other

- Mixed Initiative Interaction enables incremental learning and leads to more robust predictions.
- XAI improves transparency, trust and, thus, the interpretability of a model by the user.

## Further Reading

ACM book series on multimodal interfaces and machine learning.





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Thanks for your Attention!

Questions?

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