



THE UNIVERSITY OF
ALABAMA IN HUNTSVILLE

Challenges in Developing Requirements for AI-based Systems: A Systems Engineering Perspective

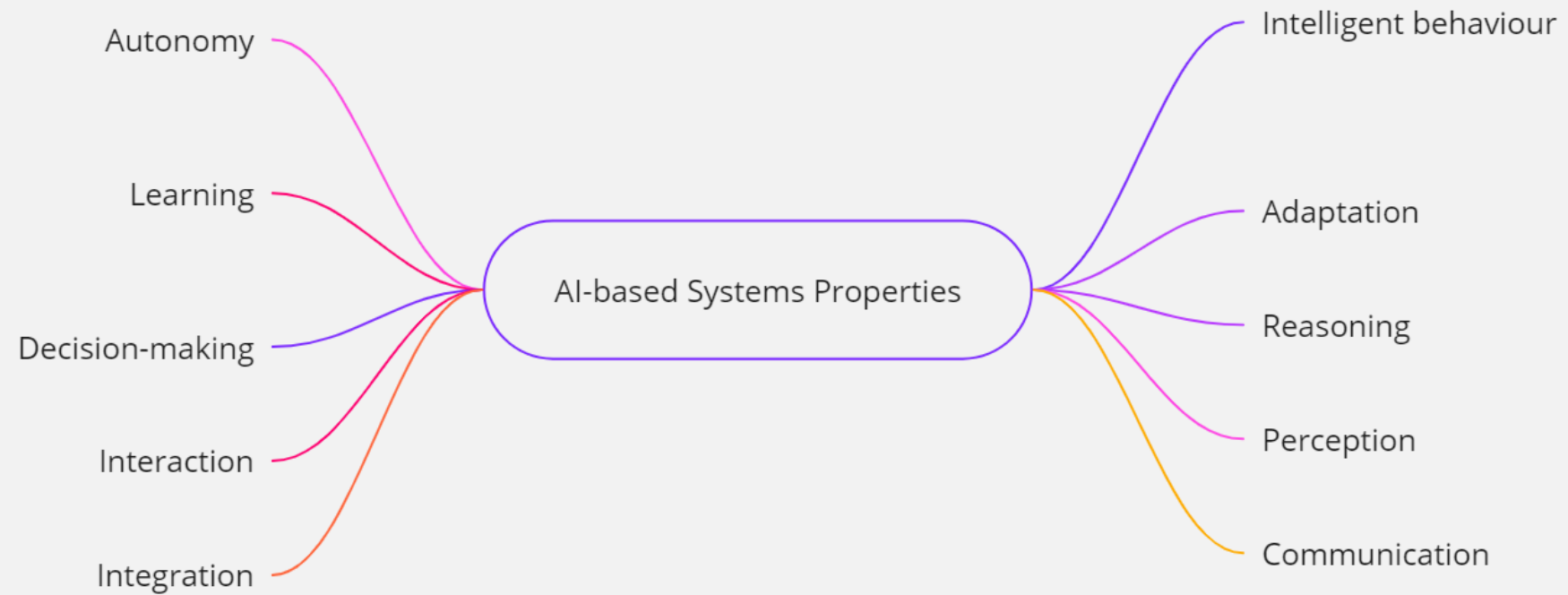
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Introduction to AI-based Systems

❖ AI-based systems

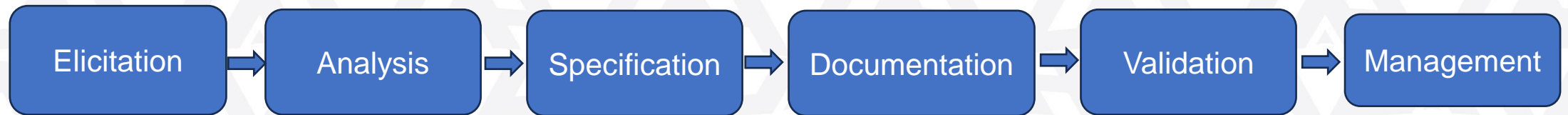
- Software and/or Hardware systems.
- Data-centric
 - ✓ During development and/or deployment
- Rule-based, algorithmic, heuristics, etc., to exhibit intelligence.
- Several vague, quality attributes
 - ✓ Explainability, privacy, ethics, resilience, etc.

Characteristics of AI-based Systems



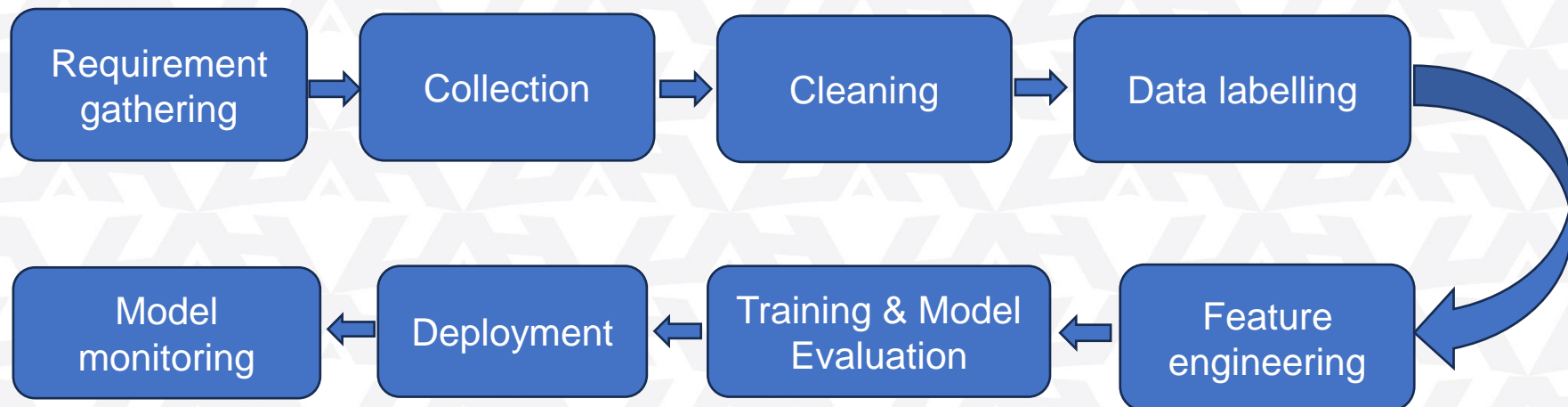
Requirements Engineering

❖ Requirement engineering involves:



(Extracted from Inayat et al. 2015)

❖ RE4AI



(Extracted from Amershi et al. 2019)

Challenges in building requirements for AI-based systems

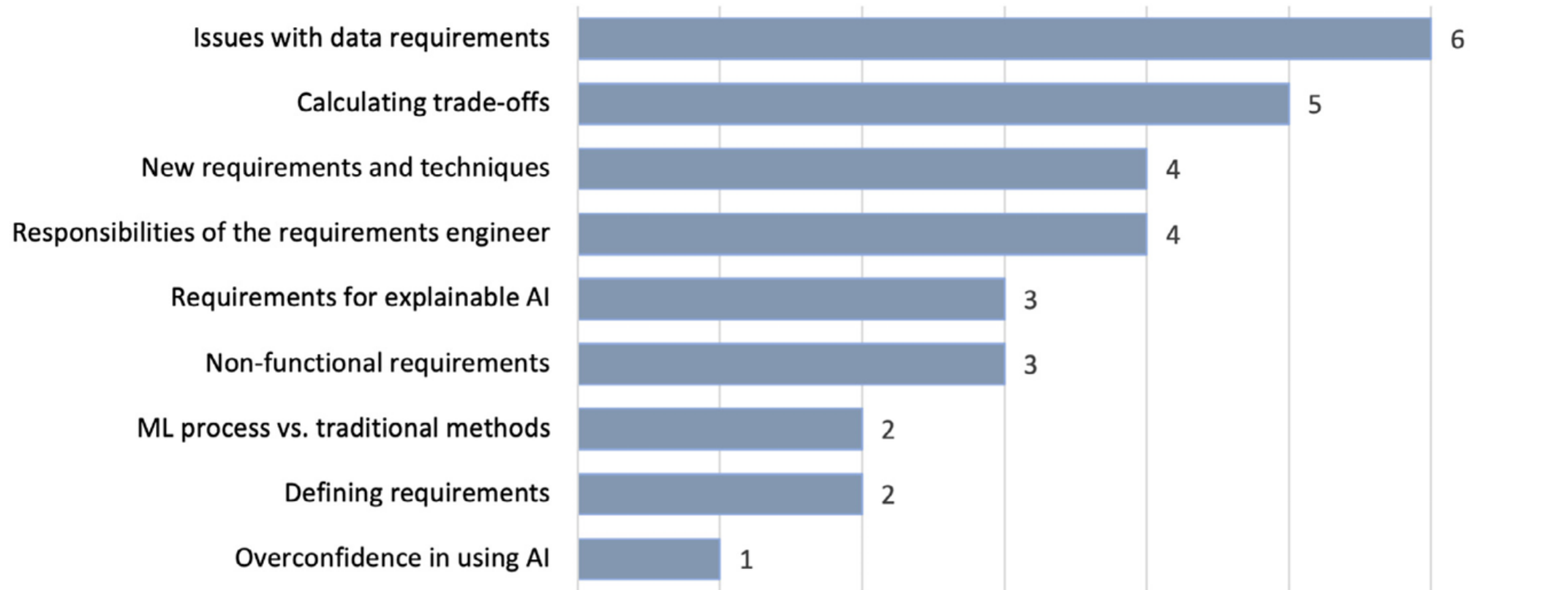


Figure 1: Number of RE4AI issues as appeared in the literature. Source: Ahmad et al. 2023

Data requirements

- ❖ Data-generation expense – Shin et al. (2019)
- ❖ Data availability and quality – Weihrauch et al. (2018)
- ❖ Training and testing of data – Nakamichi et al. (2020)
- ❖ Lack of structure and rules appropriate to train an AI system using the available data – Sandkuhl (2019).

Data Requirements

Data Quality

Data Structure

Data Format

Figure 2: Part of data-related AI requirements to be considered as proposed by Ahmad et al.(2023)

Calculating Trade-offs

- ❖ Confusion in making important choices.
 - Conflicting choices (positive or negative).
 - Cost implications, and impact of letting go of one of the choices.
- ❖ For example,
 - Trading privacy for transparency?

Emergence of new requirements

❖ New requirements -

- Data, ethics, trust, and transparency.
- Requires clear specifications for adequate integration into the current practices.

Requirements engineer responsibilities

- ❖ Lack of capacity to handle large amounts of data.

Non-functional requirements (NFR)

❖ Traditional SE NFR

- Challenges adapting to fit AI-based systems due to its unpredictable nature.
- Some NFRs might become less important in AI-based systems.
- For example,
 - ✓ Fairness and transparency are very important in AI-based systems while compatibility and modularity now have little or no importance.

Machine learning vs Traditional Methods

❖ Traditional SE

- Has a well-established RE process
- Outcomes of Traditional SE are mostly predictable

❖ AI-based systems

- Outcome can be unpredictable
- Models require dataset training, and testing (Arpteg et al. 2018).

Requirement definitions

- ❖ Vagueness of some attributes
 - Explainability, privacy, ethics, resilience, etc.
- ❖ Lack of clarity in defining vague attributes
 - For example, how can we define fairness? How do we explain a term that has multiple meanings to an AI-based system?

Current Approaches

❖ RE4AI

- GORE
- UML

❖ SE4AI

- Need approaches that can be scaled to AI-based systems, specifically in problem formulation.

❖ Gore-Oriented Requirement Engineering (GORE)

- Goal modeling and reasoning
 - ✓ Take humans to mars
 - ✓ Maintain speed
- Lacks semantics for action verbs
- Direct logical conflicts can be identified
 - ✓ Implicit conflicts are challenging.
- No explicit representation of design, and its traceability to goals

Current Approaches

❖ UML

- Easy to use
- Not flexible to model NFR and business rules

Summary

- ❖ We present the challenges of developing requirements based on existing literature.
- ❖ We found 9 most challenges as identified by various researchers.
- ❖ Current approaches to solving the challenges are also discussed with their limitations.
- ❖ Further studies will further use goal-oriented requirement engineering (GORE) and introduce requirement specifications for AI-based systems in addressing most of the identified challenges



Thank You!!!