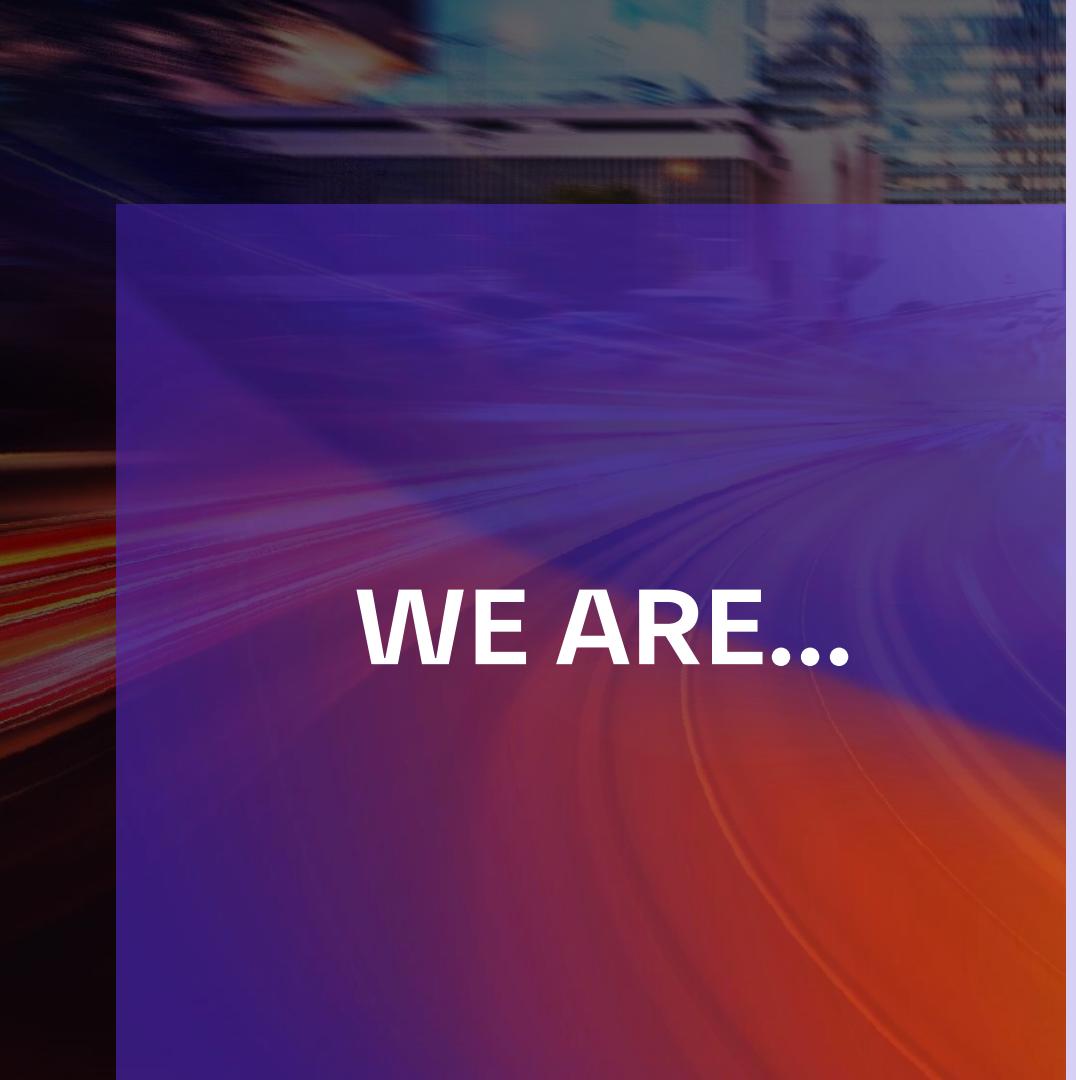


# AI & ASPICE: innovation driver or compliance risk?

Balancing innovation with traceability, reproducibility, and  
process integrity

# AUMOVIO Engineering Solutions



WE ARE...

An engineering and technology provider.

A one stop shop – from the idea to the product.

A gateway to high volume automotive products.

More than 1,600 experienced engineers & specialists.

Developers of technology for future mobility.

# AUMOVIO Engineering Solutions

## Solutions portfolio

### Consulting & specialist support

- Agile development and transformation
- Quality management & process improvement
- Data services
- Simulation engineering
- Security & privacy
- Functional safety management
- Cloud
- Data literacy

### Concept creation

- System conception
- Requirement engineering
- E/E architecture
- Innovation engineering

### Development

- System engineering
- Software engineering
- Electric machine design
- Hardware & mechanical engineering
- Noise-vibration-harshness

### Integration

- System integration
- Virtual integration
- Vehicle integration & workshops
- Software integration

### Testing & simulation

- 3D thermal simulation and structure analysis
- Test consulting services
- Driveline performance simulation
- Brake systems test and validation
- Virtual vehicle testing

### Manufacturing

- Samples, electronics & mechanics
- Series production
- Build-to-print
- Special projects (automotive & beyond)



# Presenters

## AES business center Process Management



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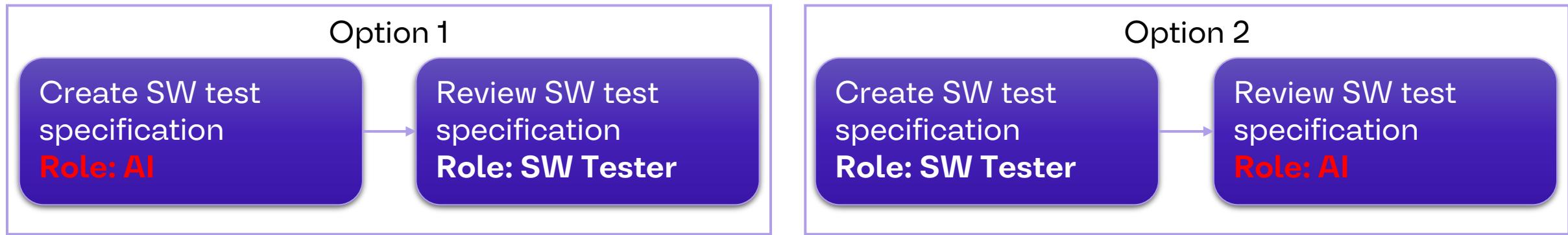
# Would you trust AI in an ASPICE assessment? Exploring the fine line between automation and accountability

*“If AI can do requirements, design, coding, and testing - what’s left for engineers?”*

# Would you trust AI in an ASPICE assessment?

## Exploring the fine line between automation and accountability

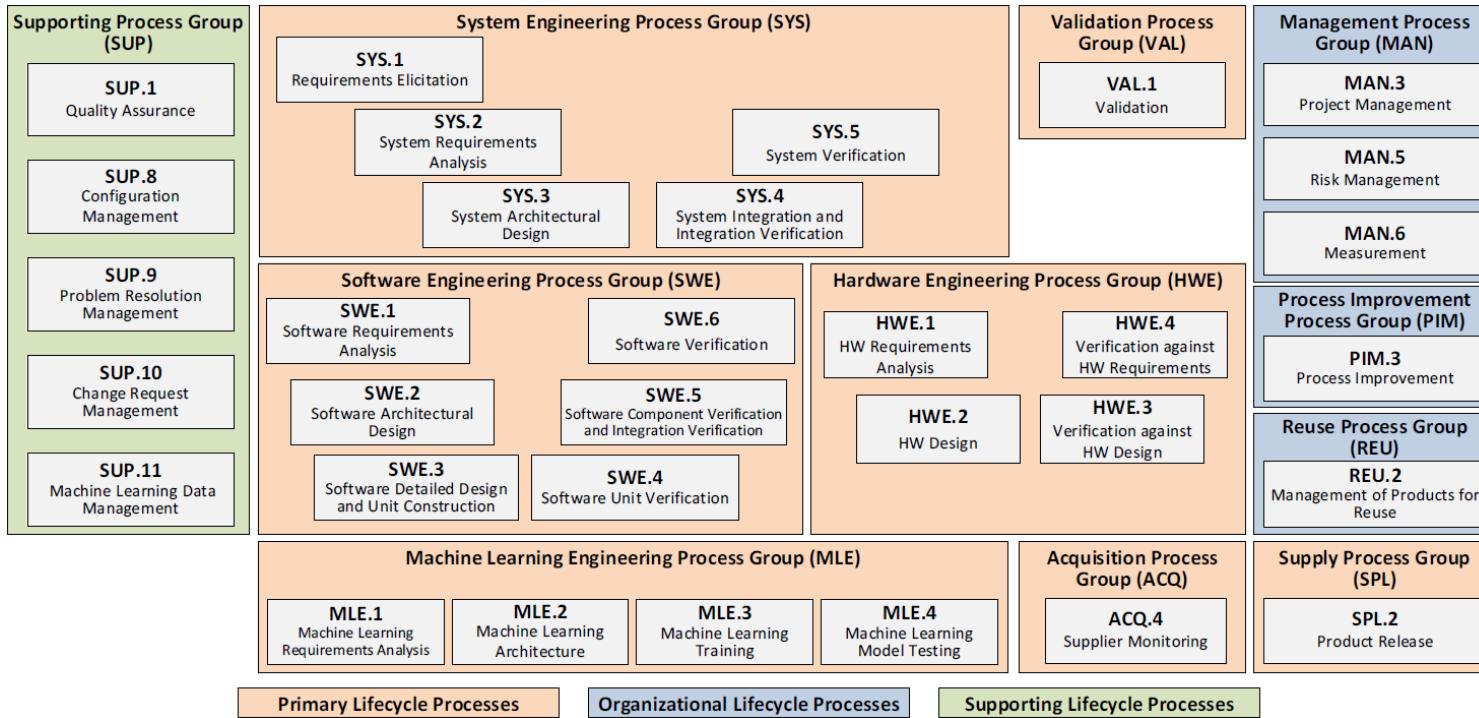
The line between automation and accountability is blurring...



- Would you accept/define “AI” as a role in a process step?
  - + Yes – AI is already reliable and objective
  - No – AI still needs human validation
  - ? Maybe – Depends on the context

# ASPICE: the gatekeeper of trust

## The model behind cap. levels, processes, and evidence



- ASPICE – **Automotive Systems Process Improvement and Capability DETERmination**
- Based on V-model
- Structured into process groups e.g., MAN, SUP, SYS, SW, etc.

### Goals

- Mitigate project/product risk
- Identify process improvements

### Capability Levels

- 0 Incomplete
- 1 Performed
- 2 Managed
- 3 Established
- 4/5 ...

### Traceability

Supports consistency, enables impact analysis, provides coverage, and shows that the product fully covers what was specified

### Documentation

Provides transparency, reproducibility, alignment, clarity

# Why AI is entering the ASPICE world

## AI is becoming a necessity in ASPICE-driven development



- **Software complexity in modern cars**
  - Vehicles are now software-defined platforms
  - Explosion of features → more data, more tests, more traceability
  - Manual processes struggle to keep up with the scale



- **Time-to-market & efficiency pressure**
  - Automotive industry faces **shrinking development cycles**
  - ASPICE compliance adds “overhead” - AI offers **efficiency gains**
  - AI offers acceleration without skipping compliance steps → Cost reduction demands smarter, faster engineering



- **Maturity of AI/ML tooling (AI is already embedded in engineering toolchains)**
  - **Planning:** effort estimation, resource allocation
  - **Requirements:** drafting, refinement, classification
  - **Coding:** generation, review, optimization
  - **Testing:** testcase generation, coverage analysis
  - **Defect analysis:** pattern detection, root cause suggestions
  - **Etc.**

# AI acting in Management Processes

## Example: MAN.3 project management



### Potential AI applications

- Predictive scheduling and effort estimation using historical data
- Resource allocation optimization across parallel projects
- Early warning signals for milestone slippage

### Benefits



- Improved accuracy in forecasts
- Faster re-planning in dynamic contexts
- Reduced manual effort in project tracking

### Risks

- Black-box estimations – team doesn't understand how results are derived
- Risk of over-trusting AI predictions without challenge
- Potential mismatch between AI forecasts and actual organizational capacity



### Assessment concerns

- Plans must remain **evidence-based and explainable**
- Human accountability:** forecasts reviewed & approved by managers
- Traceability:** records must show what was decided based on AI input
- Repeatability:** same input + same tool version → same output

*AI can be your co-pilot, but never your project manager of record.*

# AI acting in Support Processes

## Example: SUP.1 quality assurance



### Potential AI applications

- Automated compliance checks on documents, code, and models
- Continuous monitoring of process adherence (e.g., monitoring reaction time and process fulfilment)
- AI-based anomaly detection in QA metrics (defect rates, coverage gaps)

### Benefits



- Faster detection of process non-conformities
- Reduced manual QA effort, especially in repetitive checks
- Improved coverage by analyzing all artifacts, not just samples
- Earlier feedback to teams – shift-left QA

### Risks



- False positives/negatives leading to wasted effort or missed issues
- Risk of treating AI findings as “truth” without validation
- Over-automation may erode the independence of QA (AI is part of the project and could be biased towards the project)
- Lack of explainability if AI flags issues without rationale



### Assessment concerns

- QA findings must be **objective, evidence-based, and reproducible**
- Independence of QA must remain intact (AI cannot sign off on its own output)
- Records must show how AI findings were reviewed, confirmed, or rejected
- Assessors may ask: “Who ultimately took responsibility for the QA judgement?”

*AI can scan the evidence but never sign off the assurance.*

# AI acting in Development Processes

## Example: requirements & design activities

### Potential AI Applications



- Extract, classify, and refine requirements from stakeholder inputs or natural language.
- Detect ambiguities and inconsistencies early using NLP models.
- Suggest architecture patterns and component decompositions based on functional/non-functional needs.
- Perform consistency checks and interface validation.
- Generate design models and code stubs aligned with requirements.
- Support design-to-code traceability and documentation.

### Benefits



- Faster creation of requirements, architecture, and design work-products.
- Improved traceability and coverage across the V-model.
- Reduced manual effort and human error in repetitive tasks.
- Early detection of gaps and inconsistencies.

### Risks



- Ambiguity: AI-generated requirements or designs may lack clarity or intent.
- Explainability: Decisions made by AI can be hard to justify during assessments.
- Hallucinations: AI may introduce incorrect or non-existent elements.
- Verif. Need: Outputs should be validated for correctness and completeness.

### Assessment Concerns



- ASPICE demands objective, traceable, and reproducible evidence.
- AI outputs must be version-controlled and linked to preceding process items.
- Human oversight is mandatory for approval and accountability.
- Assessors will expect clear documentation of the AI generation pipeline (tool, configuration, prompts, reviewer).

*AI can accelerate development processes, but ASPICE reminds us: speed is valuable only when accountability and evidence remain uncompromised.*

# When is AI evidence acceptable?

## Deterministic vs. stochastic outputs in ASPICE assessments

- AI adds an extra wrinkle:
  - If the AI system is **“more” deterministic** (e.g., configured with fixed prompts, data set, parameters) and its outputs are versioned and stored, then the generated work products are more controlled.
  - If the AI output is **stochastic** (e.g., ChatGPT-style free text not saved or not reproducible later), then it is not stable evidence.
- As with the human the AI-generated evidence is only valid if it is **stable, traceable, and under control**.
- To be acceptable, organizations must treat the AI system + configuration + inputs as part of the work product definition (so outputs can be regenerated).

You'll find the foundation in **ISO/IEC 33002** (objectivity, repeatability, reproducibility of assessments), operationalized in **ASPICE PAM 4.0 Information Item Characteristics**, and enforced by **VDA Guidelines** that evidence must be *objective, stable, and verifiable*.

### Possible AI evidence scenarios

- AI outputs as
  - formal evidence (if they meet objectivity + traceability)
  - support material only, needing human-controlled artifacts

Open question: *how far will standardization go?*

# Implications for AI-generated evidence

## Practical recommendations

### Document your AI generation pipeline



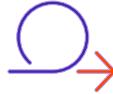
What tool(s) used, versions/configurations, prompt(s), input data, who initiated, who reviewed, timestamps, etc.

### Ensure version and traceability



Store generated outputs in a version-controlled repository; link to earlier process items

### Stability & reproducibility



Capture information that the generation can be repeated or at least to show that the process is deterministic / controlled

### Ownership, status, configuration control



Define owner, status of AI outputs, manage changes

### Accessibility of evidence



Stored and retrievable for project members in appropriate formats

### Define AI process & criteria



Guidelines for acceptance, review and validation steps

# AI limitations in the V-Model lifecycle

## Why engineer remains critical

### Black-box decisions

AI outputs often lack transparency and explainability

### Lack of domain experience

No intuition, no tacit engineering knowledge

### Hallucinations

AI can invent requirements, tests, or links that don't exist

### Memory & performance constraints

Context limits, execution bottlenecks, and scaling issues

### Cascading errors

Small AI mistakes propagate through multiple lifecycle steps

### Conversation & context memory

Long discussions lose precision; context can be dropped

### Incomplete V-Model coverage

AI can assist in tasks, but cannot replace engineers across the full lifecycle

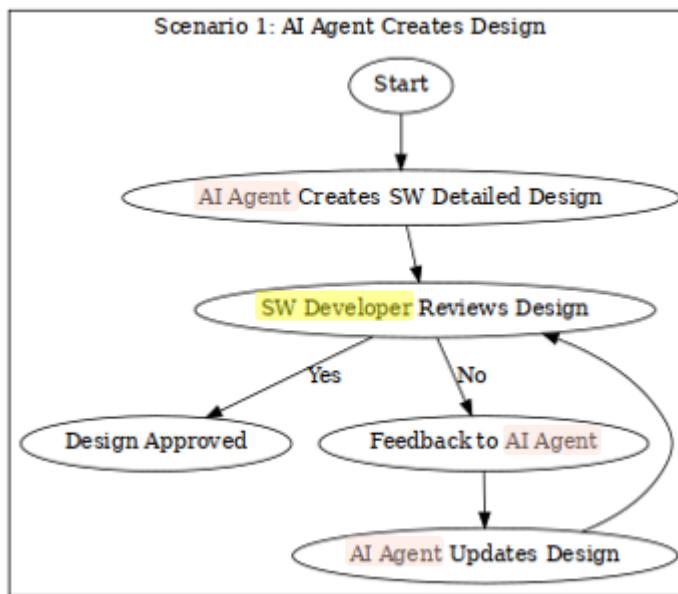
### Accountability

ASPICE and ISO standards demand named, responsible humans

# AI in ASPICE: use-case 1

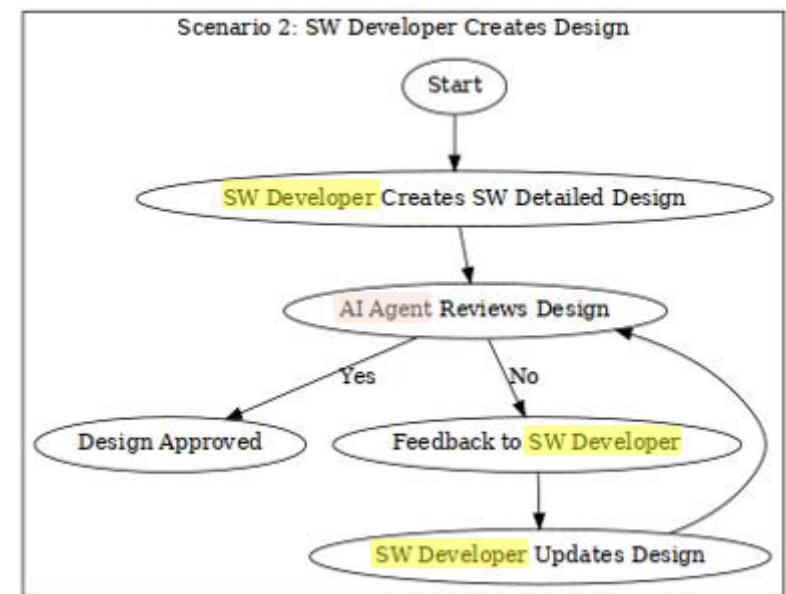
## Process SWE.3 software detailed design and unit construction

Process step	Role	Process step	Role
Create SW detailed design	<b>AI agent</b>	Create SW detailed design	SW developer
Review SW detailed design	SW Dev. / SW unit tester	Review SW detailed design	<b>AI agent</b> / SW unit tester



### Is there any risk in this approach?

*Independent of the approach selected there will be always an accountable person behind the work-product and its content.*



# AI in ASPICE: Use-case 2

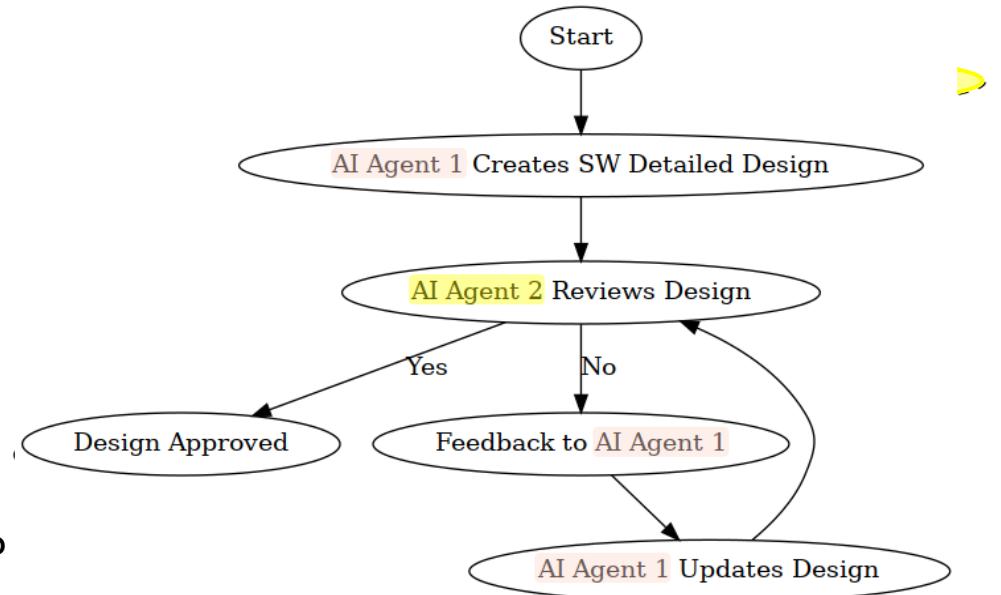
## Process SWE.3 software detailed design and unit construction

Process step	Role
Create SW detailed design	AI agent 1 with company guidelines
Review SW Detailed Design	AI agent 2 with requirement, architecture, review criteria

### Is there any risk in this approach?

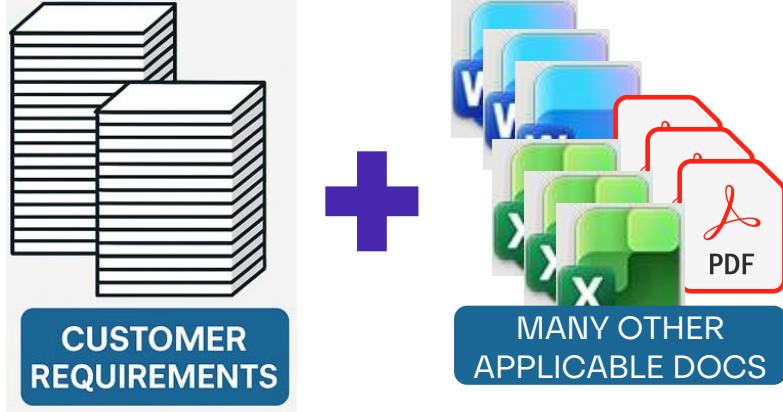
- No human feedback (e.g., functional safety manager, SW developer, SW unit tester).
- Is the risk mitigated by including SW developer accountability?
- What does “observes” really mean?
- How did AI get feedback from Functional Safety and/or SW unit tester?

Accountability: SW Developer (monitoring)?



# AI in ASPICE: Use-case 3

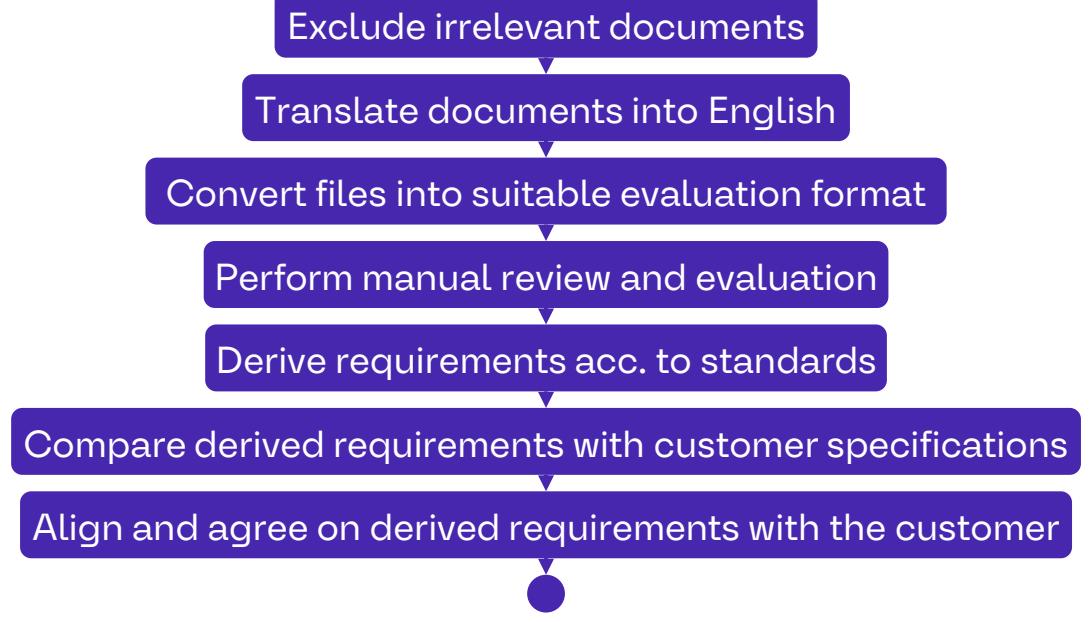
## Process SYS.1 requirements elicitation



### Challenges

- Extremely large volume of documents.
- Multiple file formats complicating traceability.
- Natural language content, including tables with implicit or “hidden” requirements.
- Multilingual documents.
- Complex version control (e.g., legal norms, standards).
- Poorly structured or inconsistent documents.

### Current Approach

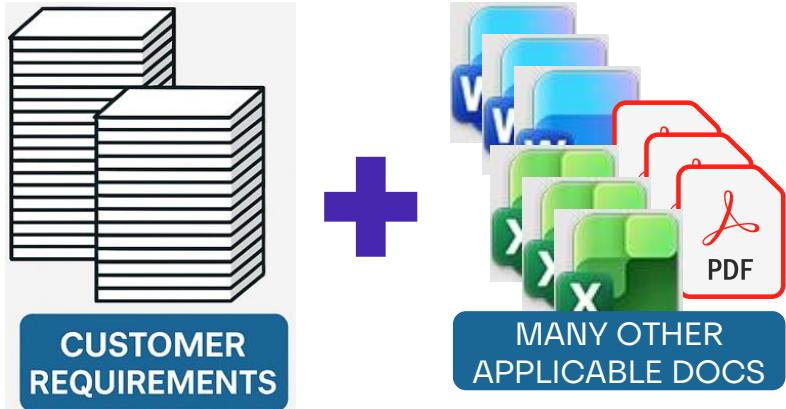


### Remaining Issue

- Quality of requirements still needs improvement.
- Multiple error-prone steps in the process.
- Overall activity is highly time-consuming and sometimes not feasible.

# AI in ASPICE: Use-case 3

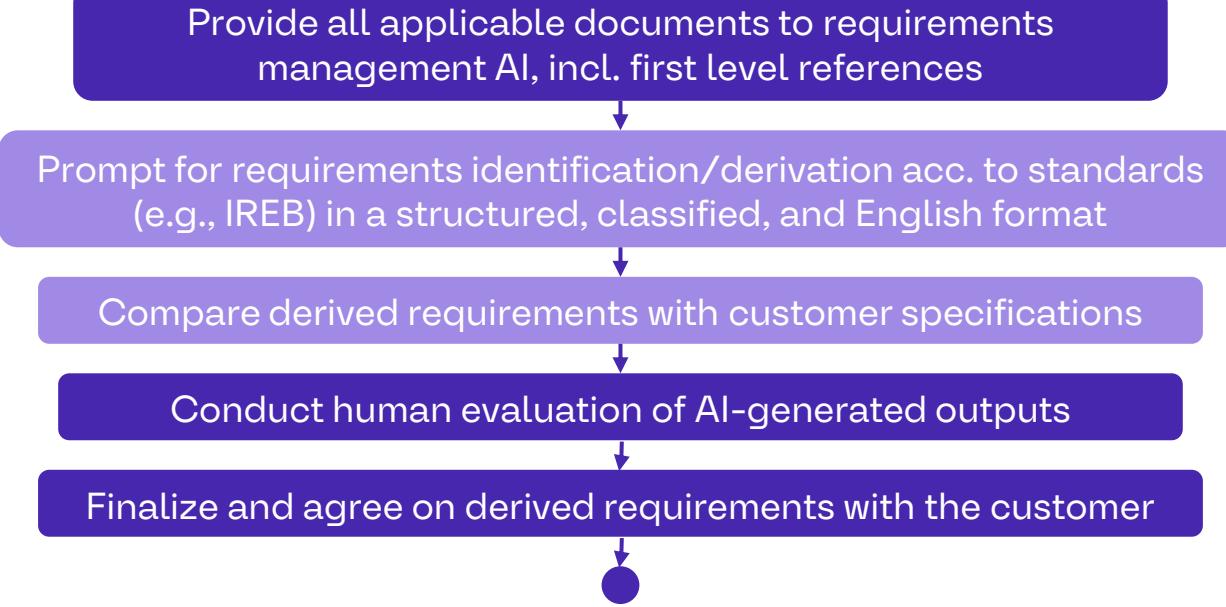
## Process SYS.1 requirements elicitation



### Challenges ✓

- Extremely large volume of documents.
- Multiple file formats complicating traceability.
- Natural language content, including tables with implicit or “hidden” requirements.
- Multilingual documents.
- Complex version control (e.g., legal norms, standards).
- Poorly structured or inconsistent documents.

### Current AI approach



### Key Takeaways

- Significant improvement in requirements quality.
- Enhanced traceability across documents.
- AI reduces requirements elicitation effort by **more than 50%**.

# AI in ASPICE: the road ahead

## From opportunity to accountability

### Outlook

- **Regulatory push** → ASPICE PAM, and VDA may publish guidelines/suggestions
- **Tool qualification pressure** → AI tools could require qualification like safety-related development tools (ISO 26262).
- **Assessment evolution** → Assessors will demand transparency of AI pipelines (inputs, prompts, tool versions, reviewers).
- **New roles emerging** → New AI roles might emerge and become common in engineering organizations.
- **Industry divergence** → Some OEMs may embrace AI faster, while others adopt a wait-and-see approach, leading to inconsistent expectations.
- **Shift from pilots to practice** → AI in ASPICE is moving from experimentation toward operational integration.
- **Cross-standard alignment** → Expect interaction with other frameworks (e.g., ISO/SAE 21434 Cybersecurity, EU AI Act, ISO/IEC 42001 AI Management Systems).

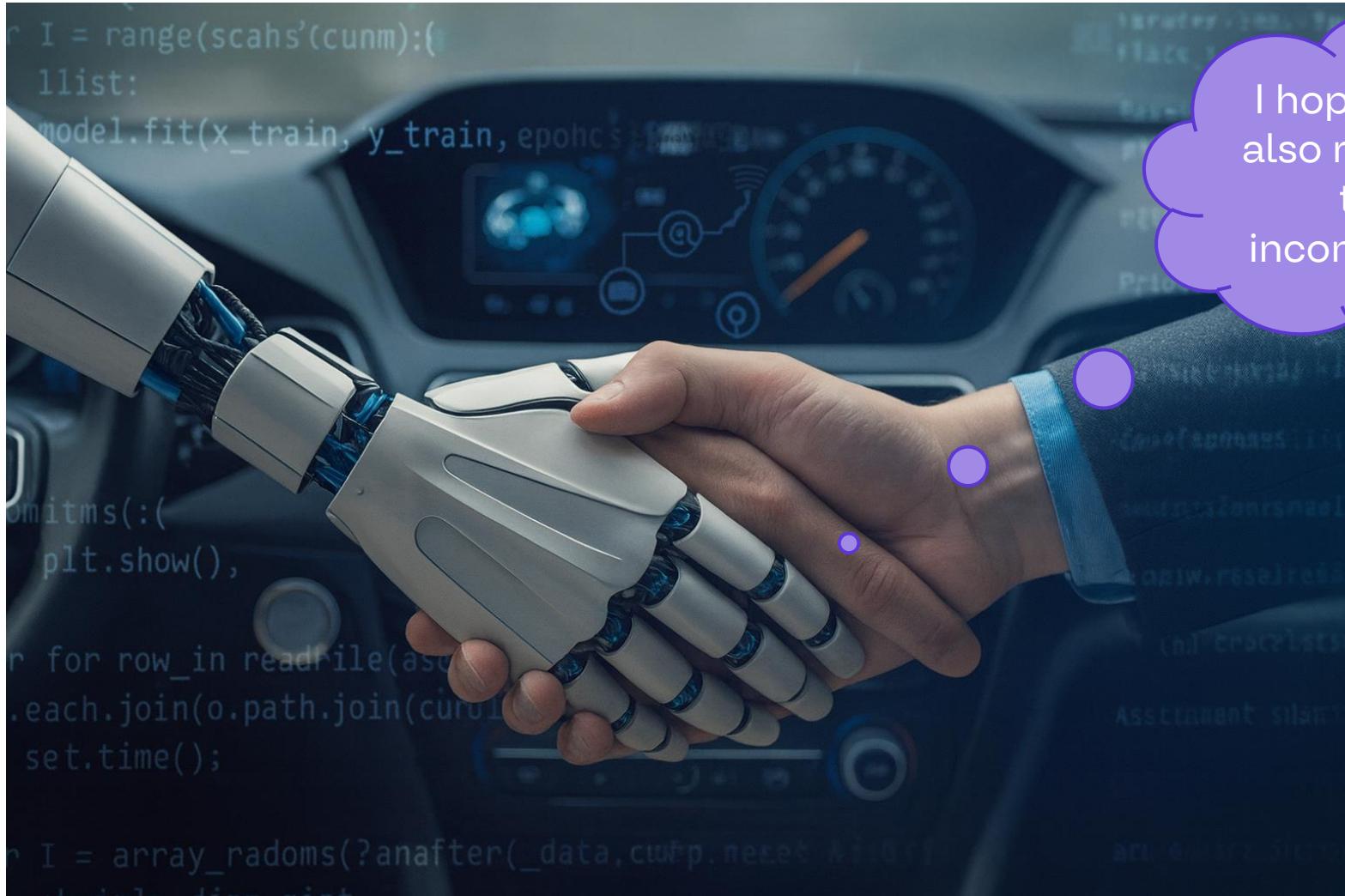
### Key Takeaways

- **AI is touching every ASPICE process - opportunity + compliance risk**
- **Evidence** → Only controlled AI outputs can count as valid assessment evidence.
- **Governance is key** → AI use must be embedded in process definition, policies, and configuration management.
- **AI maturity gap** → Many tools are ahead of standards; organizations must bridge that gap responsibly.
- **Assessor perspective** → Different assessors may interpret AI evidence differently until guidelines stabilize.
- **Cultural shift** → Teams must develop mindset, if “AI helps me” then “I am accountable for AI-supported work.”
- **Competitive edge** → Early adopters that master AI governance will have faster time to market.

The future of ASPICE is not AI versus compliance – it is about making AI evidence trustworthy.

# Closing & Q&A

## From insight to action



# Thank you



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