OVERVIEW OF EMERGING SAFETY STANDARDS:
ADS & ARTIFICIAL INTELLIGENCE

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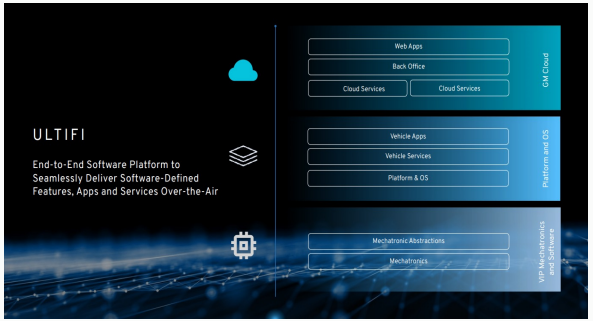
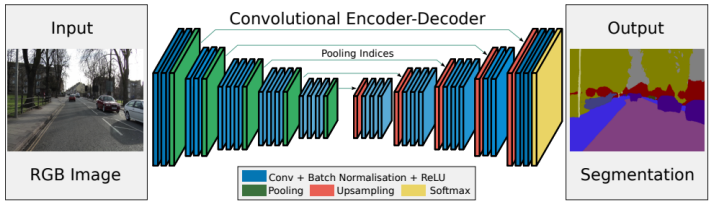
General Motors R&D

NEXT GEN VEHICLES

- SW Defined Vehicles
  - Functionality defined in the hands of the customer
  - Continuous Improvement/Continuous Deployment
  - Incremental Functionality transported Over-The-Air (OTA)
- Automated Driving System
  - SAE Level 3 and 4
  - Complex features
  - Perception & Planning
- AI and Machine Learning Enabled
  - Deep Neural Networks for perception tasks
  - Reinforcement Learning for planning

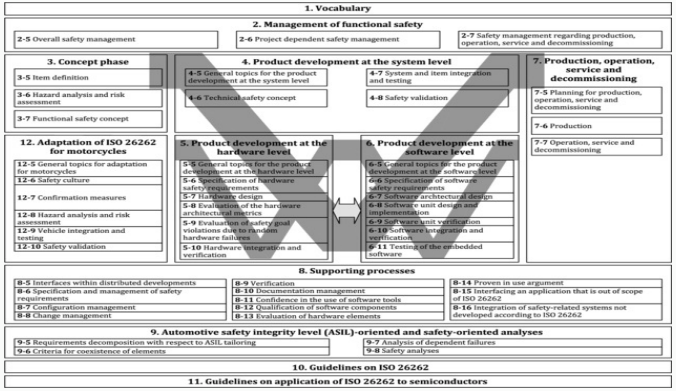


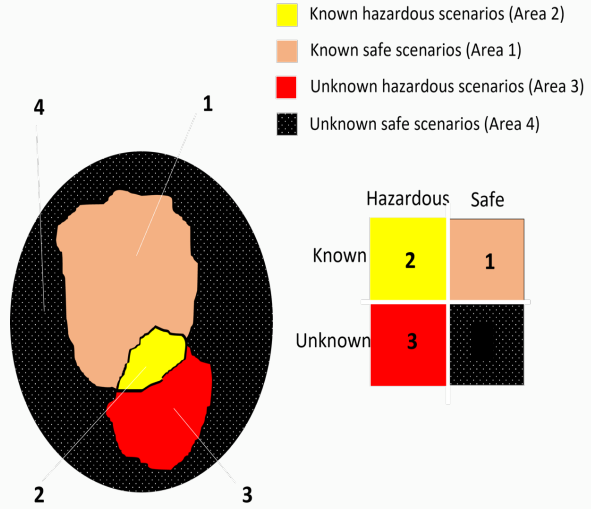
SAE level	Name	Narrative Definition	Execution of Steering and Acceleration/ Deceleration	Monitoring of Driving Environment	Fallback Performance of Dynamic Driving Task	System Capability (Driving Modes)
***Human driver* monitors the driving environment**						
0	No Automation	the full-time performance by the *human driver* of all aspects of the *dynamic driving task*, even when enhanced by warning or intervention systems	Human driver	Human driver	Human driver	n/a
1	Driver Assistance	the *driving mode*-specific execution by a driver assistance system of either steering or acceleration/deceleration using information about the driving environment and with the expectation that the *human driver* perform all remaining aspects of the *dynamic driving task*	Human driver and system	Human driver	Human driver	Some driving modes
2	Partial Automation	the *driving mode*-specific execution by one or more driver assistance systems of both steering and acceleration/ deceleration using information about the driving environment and with the expectation that the *human driver* perform all remaining aspects of the *dynamic driving task*	**System**	Human driver	Human driver	Some driving modes
***Automated driving system* ("system") monitors the driving environment**						
3	Conditional Automation	the *driving mode*-specific performance by an *automated driving system* of all aspects of the dynamic driving task with the expectation that the *human driver* will respond appropriately to a *request to intervene*	System	**System**	Human driver	Some driving modes
4	High Automation	the *driving mode*-specific performance by an automated driving system of all aspects of the *dynamic driving task*, even if a *human driver* does not respond appropriately to a *request to intervene*	System	System	**System**	Some driving modes
5	Full Automation	the full-time performance by an *automated driving system* of all aspects of the *dynamic driving task* under all roadway and environmental conditions that can be managed by a *human driver*	System	System	System	**All driving modes**



Safety Engineering and Road Vehicles

- Has a long history
- Two Standards and subsequent revisions
  - ISO 26262: Functional Safety
  - ISO 21448: Safety of the Intended Function (SoTIF)
- Functional Safety
  - Safety under random failures of HW and systematic failures of SW
  - ASIL and elaborate Design, Verification & Validation Processes
- SoTIF
  - Safety in spite of functional insufficiency or misuse
  - Trigger conditions and Acceptance Criteria
  - Scenario based testing
- Both standards attempt to accommodate the new way of building automotive software – CI/CD



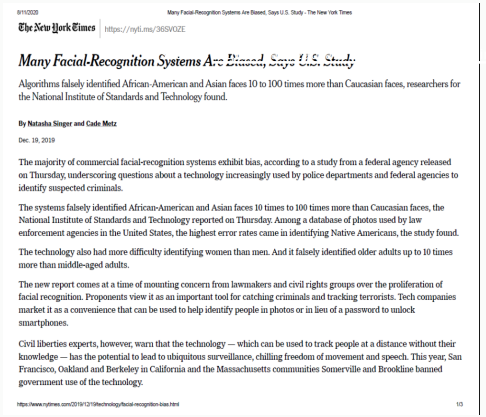


AI/ML

- AI Technology and Machine Leaning (ML) increasingly being used in vehicle applications some of which are safety-related
  - Supercruise uses MobilEye Camera which uses ML function
  - Ultracruise plans to use in-house AI/ML components
- Nature and Development and Validation of such system quite different
  - Black box, probablistic outputs
  - Data based, Training, Validation and Testing cycles
- Lack of guidelines and standards for AI system development and validation
  - OEM & Supplier specific internal guidelines
  - ISO 26262 and ISO 21448 devoted appendices to deal with AI systems
  - Several projects done and ongoing in our group

Robustness & Safety of AI systems

- AI problem is ill-posed
  - Likelihood Estimation, Stochastic
  - Incomplete (Frame Problem)
  - Bias & Uncertainty
- Predictability
  - Flounder in rare or new situations not encountered in training data set
  - Black swan issue
- Functional Safety process assumes a traditional view of development and verification
- Missing safety lifecycle assets and new assets
  - Hardly any requirements
  - Training and Test Data Sets
- Development Lifecyle for AI/ML components is non-traditional
  - Data Intensive



8/11/2020 Many Facial-Recognition Systems Are Biased, Says U.S. Study - The New York Times

The New York Times https://nyti.ms/36SVOZE

Many Facial-Recognition Systems Are Biased, Says U.S. Study

Algorithms falsely identified African-American and Asian faces 10 to 100 times more than Caucasian faces, researchers for the National Institute of Standards and Technology found.

By Natasha Singer and Cade Metz

Dec. 19, 2019

The majority of commercial facial-recognition systems exhibit bias, according to a study from a federal agency released on Thursday, underscoring questions about a technology increasingly used by police departments and federal agencies to identify suspected criminals.

The systems falsely identified African-American and Asian faces 10 times to 100 times more than Caucasian faces, the National Institute of Standards and Technology reported on Thursday. Among a database of photos used by law enforcement agencies in the United States, the highest error rates came in identifying Native Americans, the study found.

The technology also had more difficulty identifying women than men. And it falsely identified older adults up to 10 times more than middle-aged adults.

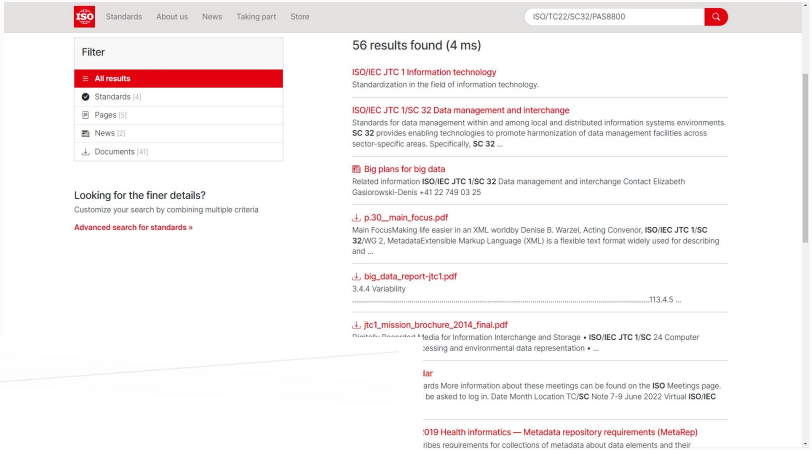
The new report comes at a time of mounting concern from lawmakers and civil rights groups over the proliferation of facial recognition. Proponents view it as an important tool for catching criminals and tracking terrorists. Tech companies market it as a convenience that can be used to help identify people in photos or in lieu of a password to unlock smartphones.

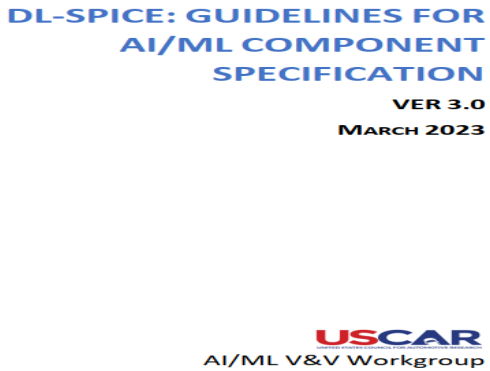
Civil liberties experts, however, warn that the technology — which can be used to track people at a distance without their knowledge — has the potential to lead to ubiquitous surveillance, chilling freedom of movement and speech. This year, San Francisco, Oakland and Berkeley in California and the Massachusetts communities Somerville and Brookline banned government use of the technology.

https://www.nytimes.com/2019/12/19/technology/facial-recognition-bias.html 1/3

The Standards Landscape for AI based systems

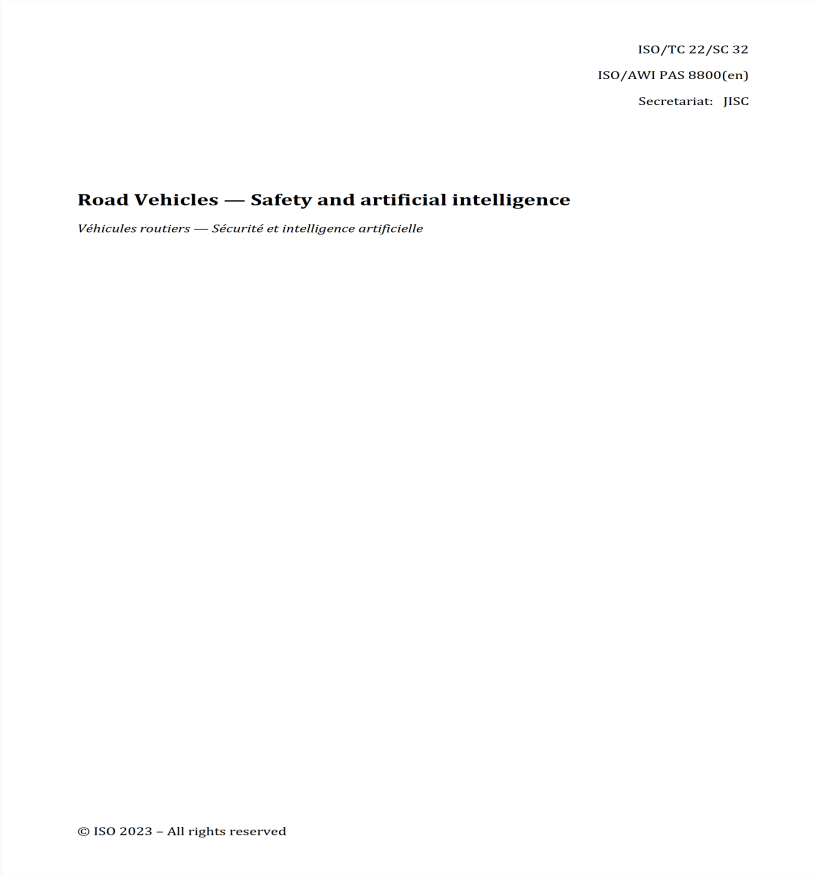
- Quite rich, has been an intense focus for the last few years
- More than 100 guidelines and standards in the general context have come out or under development
  - ISO/TC 22/SC32 – Electrical & Electronic components and general systems aspects
  - ISO/IEC JTC 1/SC42 Artificial Intelli
- Participating in
  - USCAR DL-SPICE Guidelines Docun
  - ISO/AWI TS 5083: Road Vehicles – for ADS – Design, V&V
  - ISP/AWI PAS 8800 Road Vehicles – and AI





ISO/PAS 8800 - Overview

- Industry-specific guidance on the use of AI/ML based systems in safety-related functions of road vehicles
  - Not restricted to specific ML techniques
  - Not restricted to ADS features
    - Annex B of ISO/TS 5083 (under development) adaptation of PAS 8800 for ADS
- Builds on guidance specified in ISO/IEC DTR 5469 (under development)
- Compatible with ISO/IS 26262 and ISO/IS 21448 (SoTIF)
- Harmonizes the concepts in Annex D.2 of ISO/IS 21448



ISO/TC 22/SC 32

ISO/AWI PAS 8800(en)

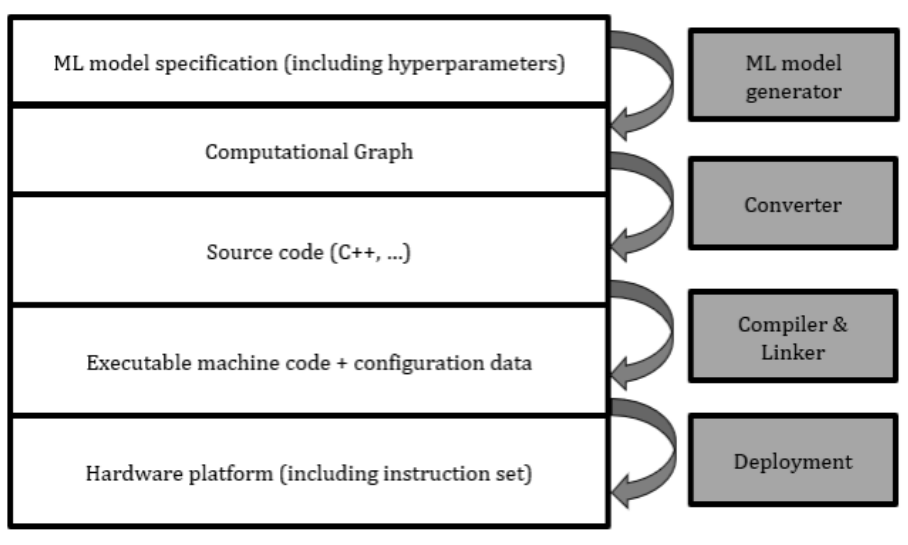
Secretariat: JISC

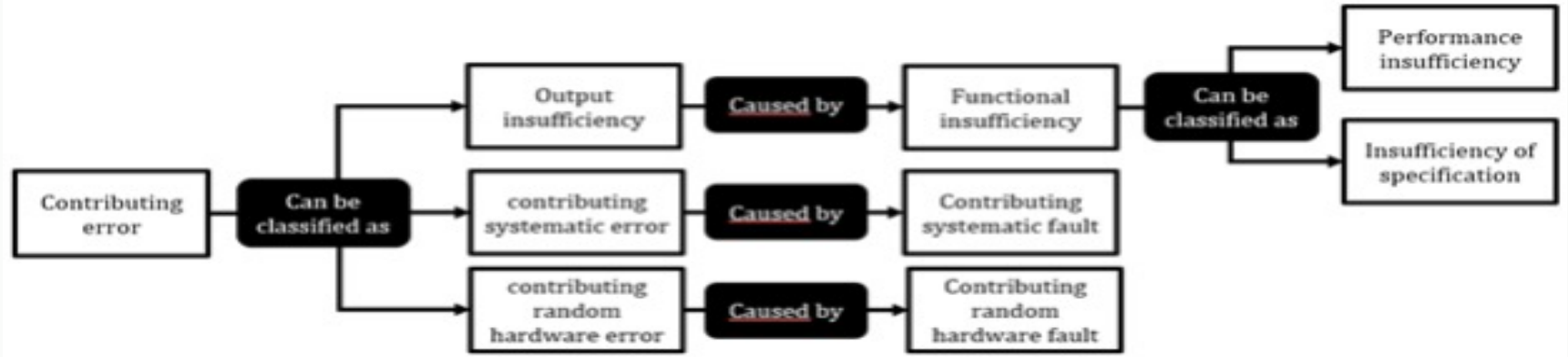
Road Vehicles — Safety and artificial intelligence

Véhicules routiers — Sécurité et intelligence artificielle

ISO/PAS8800 – Focus

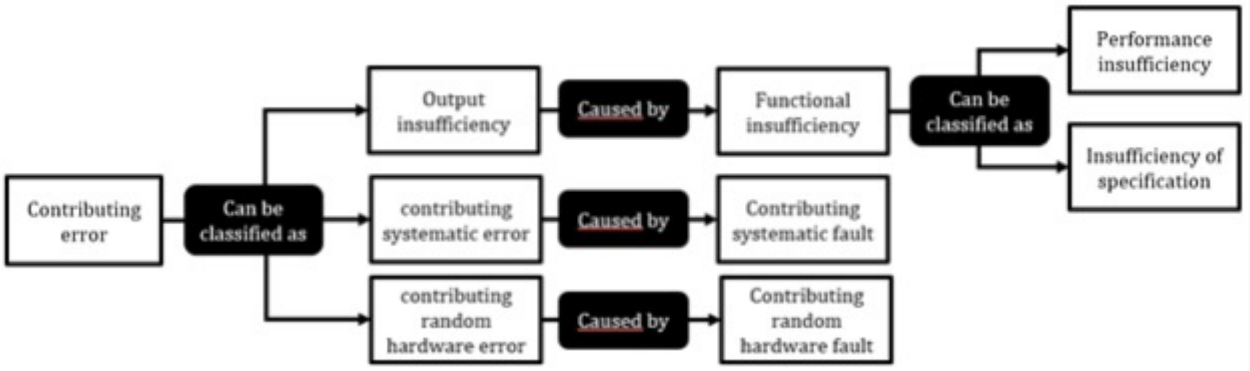
- Items unique to the development of AI component or subsystem
  - SW and HW items along existing standards
- Safety Engineering Process along the lines of ISO 26262 and ISO 21448..
- Errors/Faults and Vehicle level Hazards
- Random HW faults and Systematic SW Faults





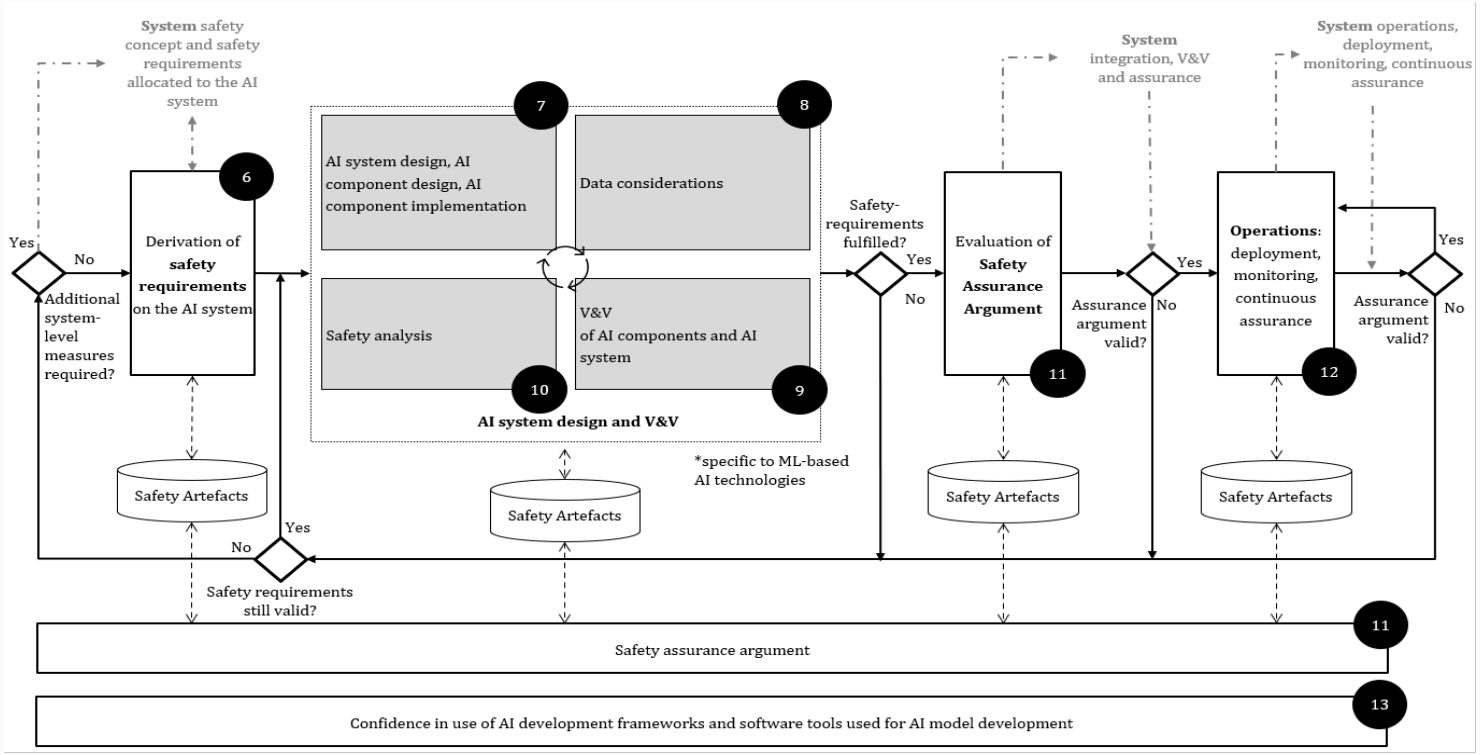
ISO/PAS 8800 – Salient Features

- Supports CI/CD of AI/ML functions (field monitoring and data collection)
- Emphasizes Assurance arguments, besides safety artefacts
- Functional safety-related risks addressed as per ISO/IS 26262
- Performance Limitation risks by extending the concepts and guidance given in ISO/IS 21448
  - Safety requirements are derived by analyzing performance limitations of AI functions.



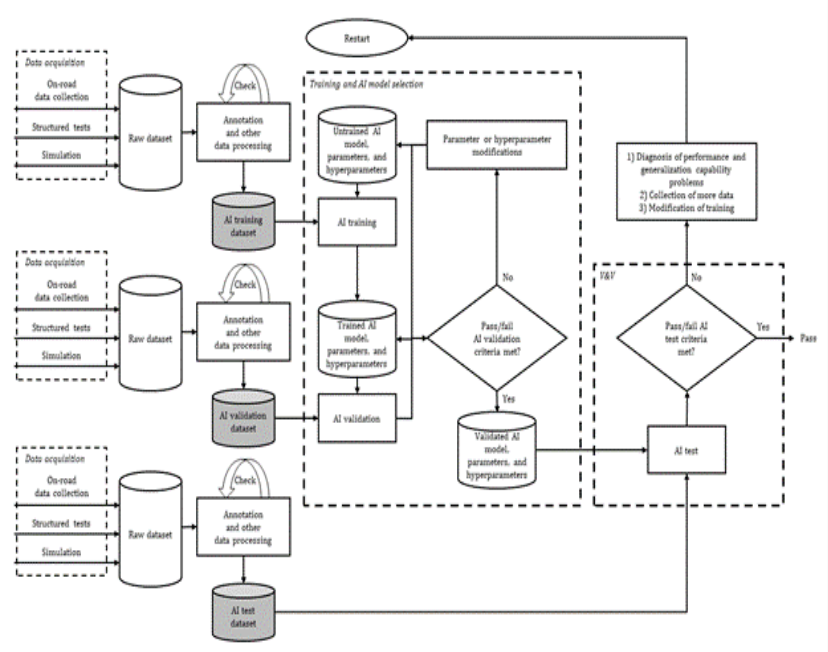
AI Lifecycle

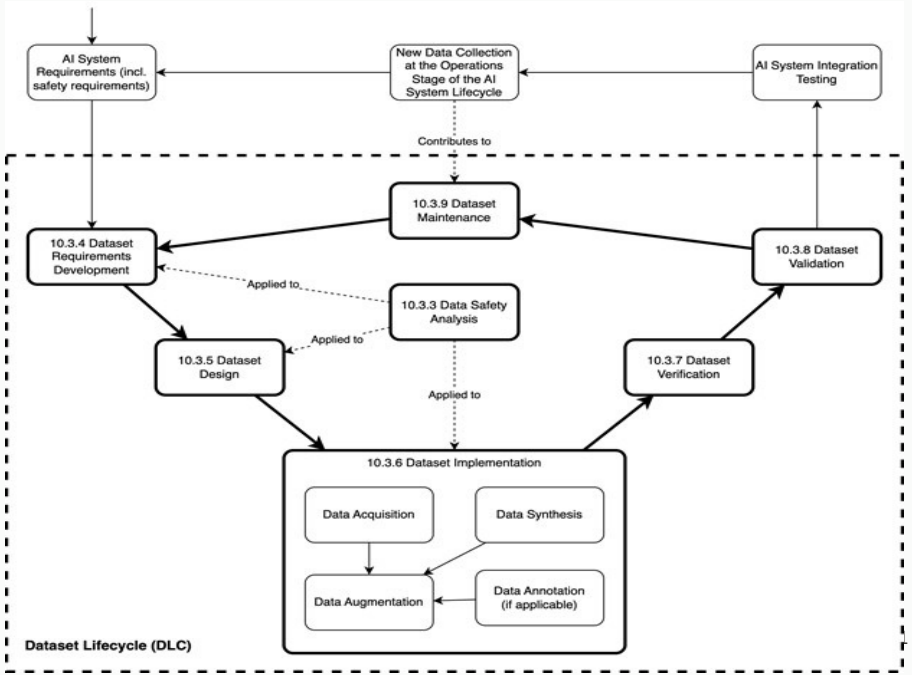
- Exemplary Lifecycle giving rise to Safety Assets forms the basis for the entire document



Data Lifecycle

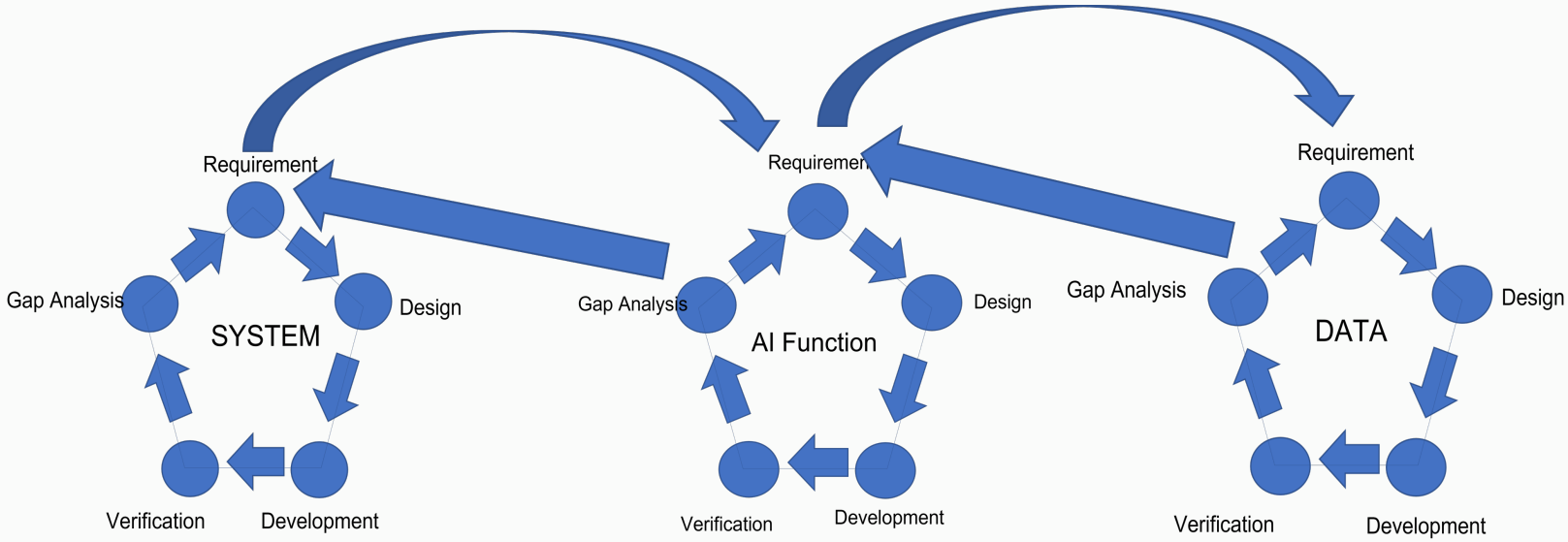
- Data plays a fundamental role in AI system development
  - A dataset lifecycle shall be defined for datasets used in the development of the AI system.
  - The dataset lifecycle shall cover a dataset's requirements development, design, implementation, verification and validation, safety analysis and maintenance.





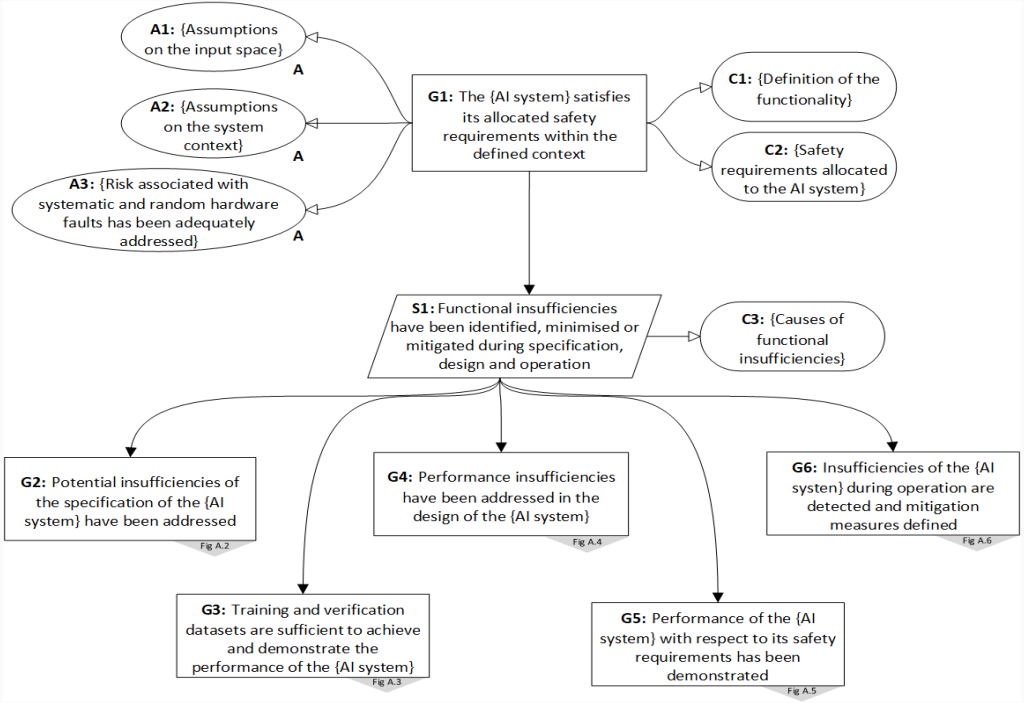
Comprehension of multiple lifecycles

- Traceability from system level to AI and Data life cycle



Assurance arguments

— An assurance argument shall be developed to demonstrate that the AI system fulfils its safety requirements.



Chapters - Current Draft

1 Scope
2 Normative references
3 Terms and definitions
4 Abbreviations
5 AI within the context of road vehicles systems safety engineering
6 Derivation of safety requirements on AI systems
7 Selection of AI-Measures and design-related considerations
8 Data-related considerations
9 Verification and validation of the AI system
10 Risk evaluation and safety analysis
11 Assurance arguments for AI systems
12 Measures during operation and continuous assurance
13 Confidence in use of AI development frameworks and software tools...

Documentation Timing

- *Working Draft 2* preparation: February and March
- *Working Draft 2* commenting: April and May
- *Working Draft 2* observations and updates: June and July
- *Committee Draft* commenting: August and September
- *Committee Draft* observations and updates: October and November
- Preparation for approval ballot: December
- Approval ballot: January and February ‘24

Questions