

Digital Train Control

International Railway Safety Council Conference

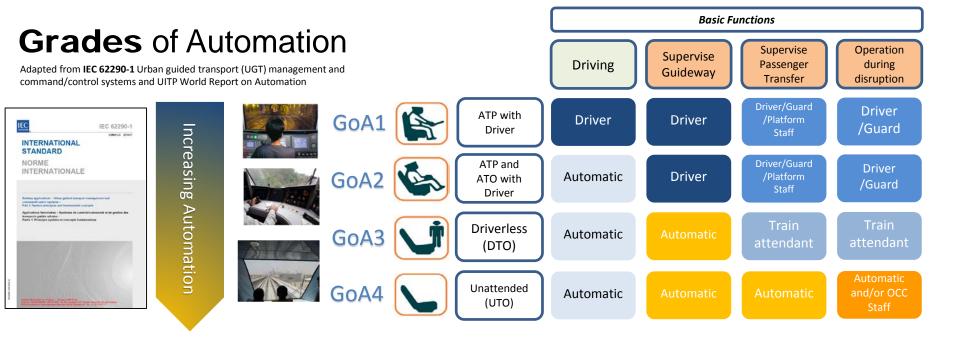
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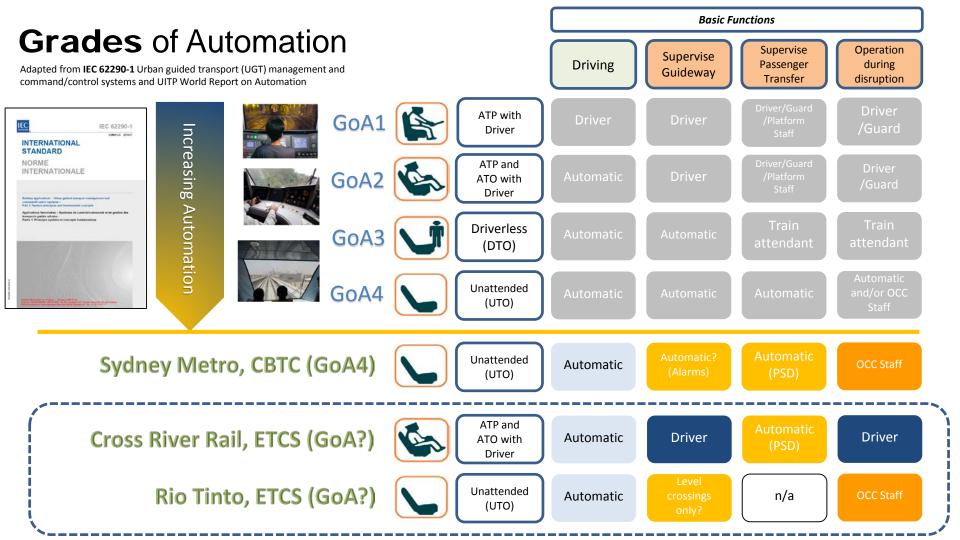
Functional Safety for AI based Systems



GHD Transportation

David Milburn & Mike Erskine





The Future - Automatic versus Autonomous

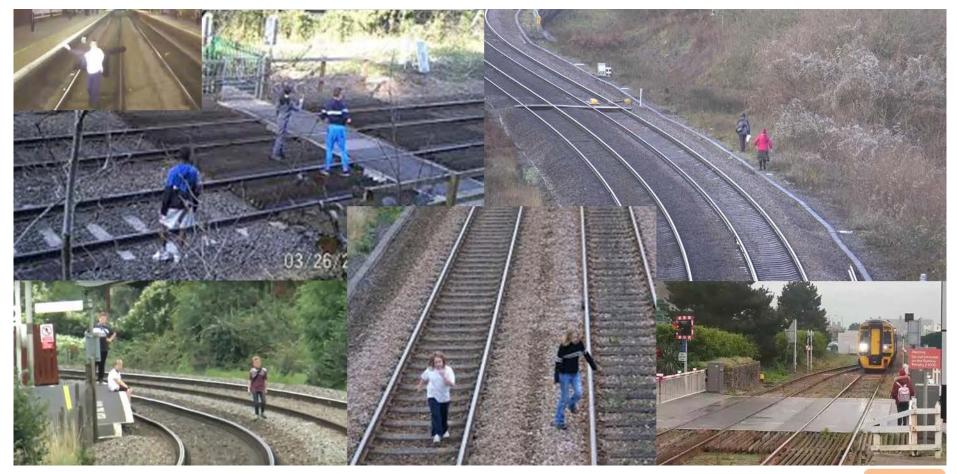
Automatic System: performs task sequences based on pre-defined rules. The information required to understand the environment is provided to enable the system to undertake rehearsed actions. *Predominantly Deterministic*.



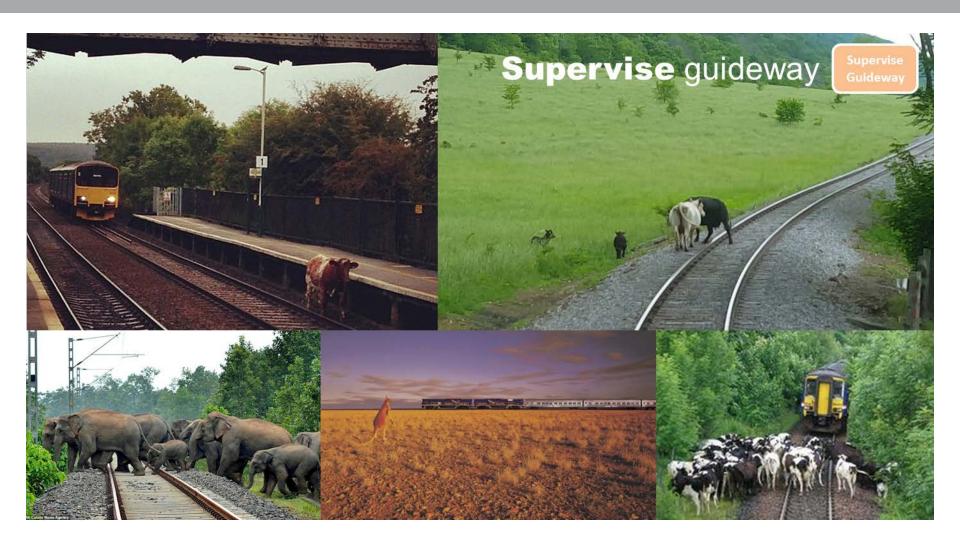
Autonomous System: capable of making independent decisions to respond to all cases in real-time, and in some situations without reference to pre-defined instructions. It must therefore manage the functions of comprehension, environmental awareness, and spontaneous decision making. *Predominantly Non Deterministic*.



DTO/UTO will be delivered by Autonomous Systems?



Supervise Guideway



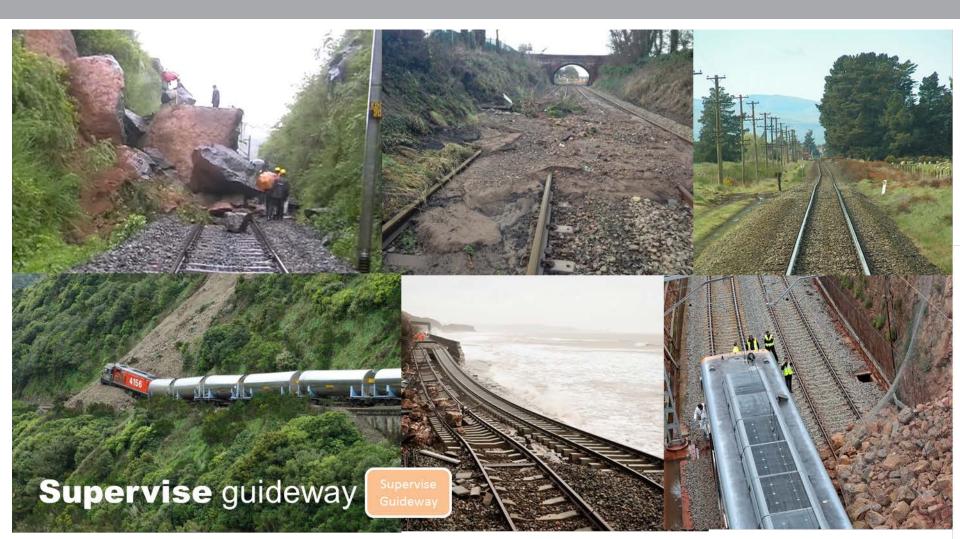
Guideway





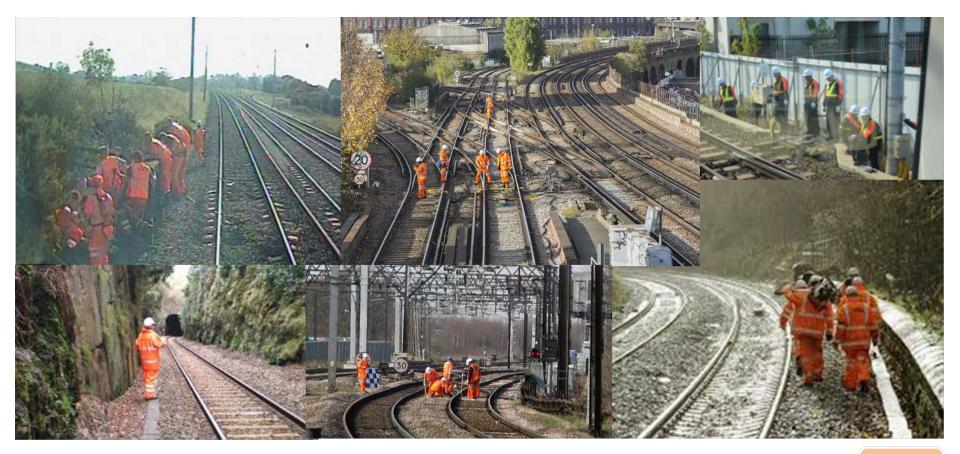








Supervise Guideway



Supervise Guideway









Report rolling stock issues





Monitor Vehicle Health

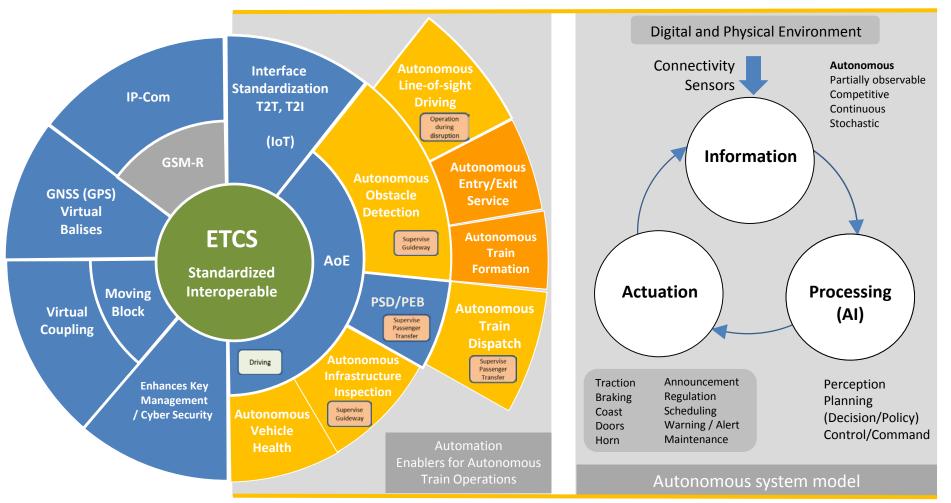




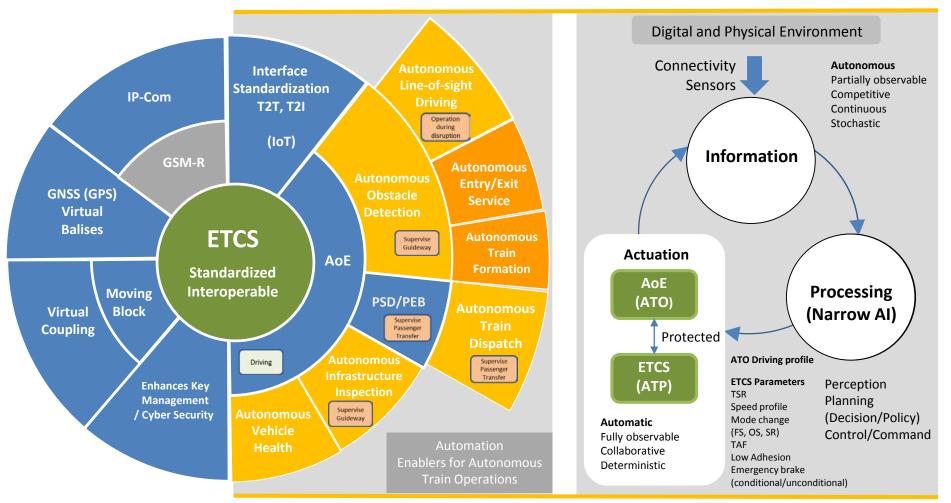




Building on existing technology - ETCS example (context and evolution)



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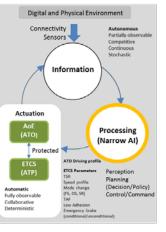


Functional safety for autonomous systems

Autonomous driving systems will use **AI**, this is likely to be based on commercial autonomous car technology (with likely evolution of ISO 26262)

Railway Functional Safety methods will need to adapt

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ICS 29.280; 45.020	Supersedes EN 50120-1:1999	EUROPÄISCHE NORM	Supervedes EN 50128:2001	103 93.100	Supersedes CLC/TR 50451 2007, CLC/TR 50506- 1/2007, CLC/TR 50508-2/2009, EN 50129/2003
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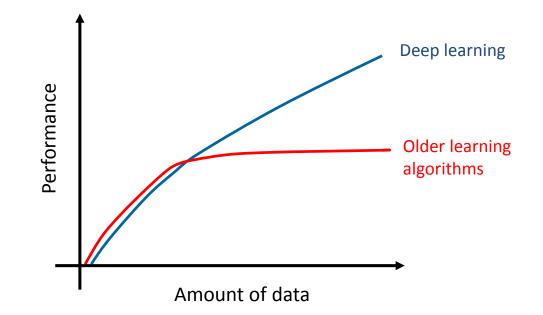
Intelligence

The ability to learn or understand or to deal with new or trying situations.

Artificial Intelligence

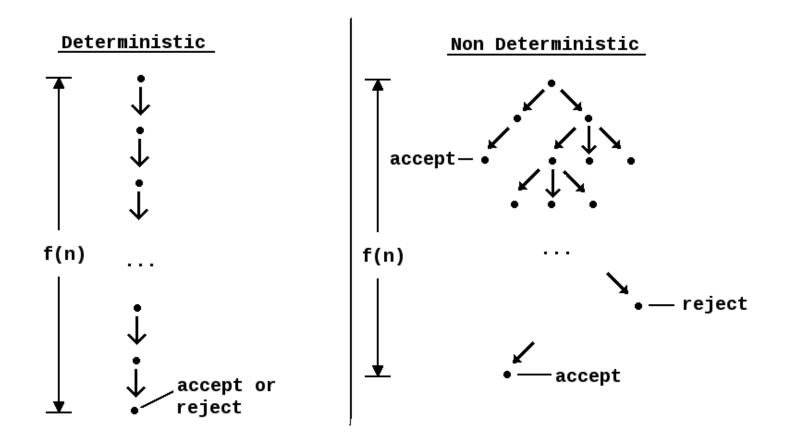
Both the intelligence of machines and the branch of computer science which aims to create it, through "the study and design of intelligent agents" or "rational agents", where an intelligent agent is a system that perceives its environment and takes actions which maximize its chances of success.

Evolution of AI – why deep learning?



How do data science techniques scale with the amount of data?

Deterministic versus non-deterministic

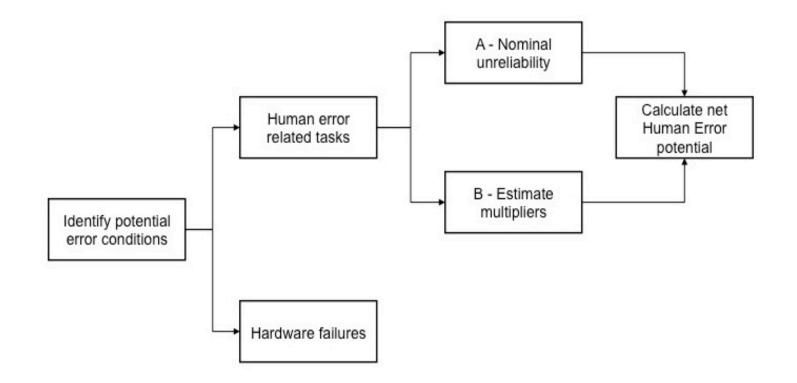


Non-deterministic vs. Stochastic

- Non-deterministic means we know there is uncertainty but do not know the distribution of uncertainty
 - NON-DETERMINISTIC != Uniform distribution
- E.g. The agent may be in one of {s1,....sn}. We don't know which of them are more vs. less likely (we also don't know that they are all equally likely)
- E.g. the action a done in s can lead to {s1..sm}—no information is available on which outcomes are more likely

- Stochastic is Non-deterministic + distribution information.
- So stochastic means more knowledge
 - Note that more knowledge implies more problems are expressible and solvable (but also might mean that the computational burden on the agent increases). With nondeterminism, the agent has to say whether there is a strong vs. weak plan. With stochastic actions, the agent can also talk about plans that satisfy goals with different levels of probability

HEART - Human Error Assessment & Reduction Technique



HEART Error Producing Conditions

Generic Task	Nominal Error Probability
Totally unfamiliar, performed at speed with no idea of likely consequences	0.55 (0.35 – 0.97)
Shift or restore system to a new or original state on a single attempt without supervision or procedures.	0.26 (0.14 - 0.42)
Complex task requiring high level of comprehension and skill.	0.16 (0.12 – 0.28)
Fairly simple task performed rapidly or given scant attention.	0.09 (0.06 – 0.13)
Routine, highly practiced, rapid task involving relatively low level of skill.	0.02 (0.007 – 0.045)
Restore or shift system to original or new state following procedures, with some checking.	0.003 (0.0008 – 0.007)
Completely familiar, well designed, highly practised routine task, oft-repeated and performed by well-motivated, highly trained individual with time to correct failures but without significant job aids.	0.0004 (0.00008 – 0.009)
Respond correctly to system even when there is an augmented or automated supervisory system providing accurate interpretation of system state.	0.00002 (0.000006 – 0.00009)
Miscellaneous task for which no description can be found.	0.03 (0.008 – 0.11)

HEART Error Producing Conditions

Error Producing Condition	Multiplication Factor
Short time available for correction	17
Ambiguity in required standards	5
Poor / ambiguous feedback	4
Little or no independent checking	3
Unclear allocation of responsibility	1.6
Low intrinsic meaning in a task	1.4
High level emotional stress	1.3
Excess team members (per head)	1.03

HEART Generic Violation Behaviors

Generic Violation Behaviours	Nominal error probabilities for females (x 1.4 for males)
Distinctly inconvenient to comply. Potential violator does not feel bound by any implied requirement to comply. Easy to violate. Little likelihood of detection.	0.42
Compliance relatively unimportant. Easy to violate. Little or no inducements to comply.	0.35
Compliance may be fairly important, but chances of detecting violation low. Personal benefits of violating are high and direct.	0.38
Personal benefit in violating, though likelihood of detection is moderate to high. Or else compliance fairly important, but chances of detection low.	0.18
Compliance important, usually legally required, but chances of detection low to moderate.	0.03
No immediate incentive to violate, but likelihood of violation detection moderate to high	0.007
Socially unacceptable, likelihood of detection low and likelihood of unfavourable outcome for violator low.	0.007
Socially unacceptable, chances of detection high and chances of bad outcome high.	0.0001

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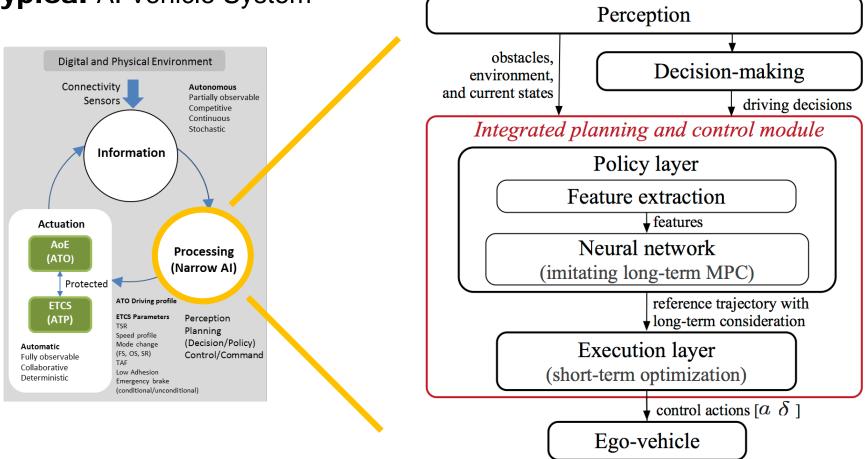
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Yottabyte (10 ²⁴)	Future Global	Logically in this space within about 3-4 years

Typical AI Vehicle System



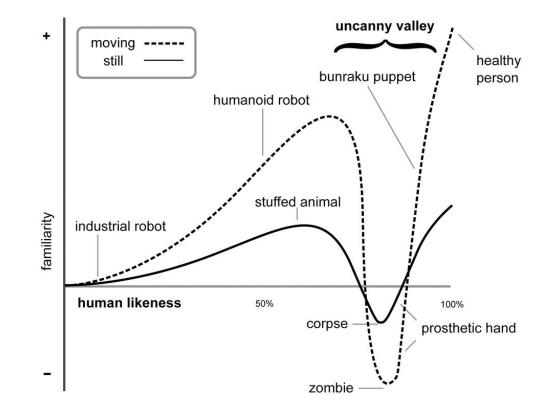
Organisational impacts of AI

- Is your organisation a Gigabyte or a Terrabyte organisation?
- Where is your thinking at?
- Transition to a Petabyte and Exabyte mode of thinking (Facebook, Microsoft, Google and similar territory)
- Stakeholder Messaging
- Education and upskilling
- Organisations need to develop Stakeholder KPI's
 - Journey
 - Content for events for context
 - Pre inform so less infotainment factor for the media

Societal Expectations Regarding Confidence – reference points

- SIL 4 10,000 to 100,000 (10⁴ to 10⁵) for rail and nuclear safety
- Proof of new particle discovery Higgs Boson 5 σ
- Privacy we can be personally lax, but demand very high sigma externally
- Safety Highly trained humans driving cars, planes and trains
- Capability $-1 2 \sigma$ better than we do on average
- Honesty/Integrity Our current human governance systems in organisations and governments with IT and controls (already implicit AI)
- Health Professionals human diagnosis and treatment with extensive IT and machine support.

Uncanny Valley



Deeper Uncanny Thinking

Coherent Extrapolated Volition (CEV) is intended to be what humanity objectively would want, all things considered, but it can only be defined relative to the psychological and cognitive qualities of present-day, unextrapolated humanity.

Our feeling of caution or discomfort is to do with not only the visual appearance, but the lack of knowledge of what similarity it will have to our volition.

Therefore, assurance has to be centred around validation and verification of volition of key values and principles (4V's).

Mountain of Caution



ISO/IEC SC42

Standards for Management of Artificial Intelligence – Early 2018

Aust Stds IT-043 (Mirror)

Standards for Management of Artificial Intelligence – March 2019

ISO/IEC JTC 1/SC 42 - Artificial intelligence

- Formed in 2018
- International response
- June 2019 Australian discussion paper
- Sep 2019 feedback to IT-043 members from Australian survey

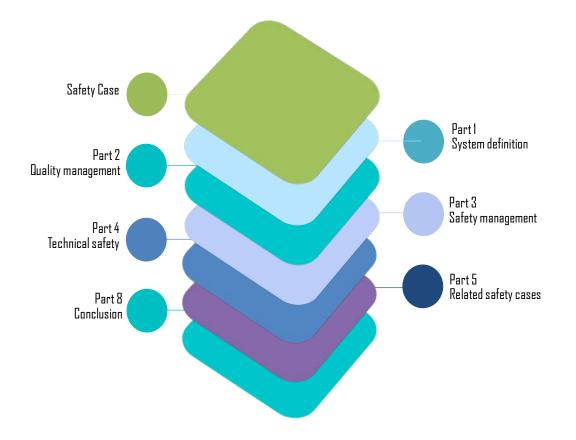


ISO/IEC JTC 1/SC 42

- Artificial intelligence

Duringt	Forms area
Project	Focus area
ISO/IEC AWI TR 20547-1	Information technology – Big data reference architecture – Part 1: Framework and application process
ISO/IEC DIS 20547-3	Information technology – Big data reference architecture – Part 3: Reference architecture
ISO/IEC WD 22989	Artificial intelligence - Concepts and terminology
ISO/IEC WD 23053	Framework for Artificial Intelligence (AI) Systems Using Machine Learning (ML)
ISO/IEC AWI 23894	Information Technology – Artificial Intelligence – Risk Management
ISO/IEC NP TR 24027	Information technology – Artificial Intelligence (AI) – Bias in AI systems and AI aided decision making
ISO/IEC PDTR 24028	Information technology – Artificial Intelligence (AI – Overview of trustworthiness in Artificial Intelligence
ISO/IEC NP TR 24029-1	Artificial Intelligence (AI) – Assessment of the robustness of neural networks – Part 1: Overview
ISO/IEC NP TR 24030	Information technology – Artificial Intelligence (AI) – Use cases
ISO/IEC NP TR 24368	Information technology – Artificial intelligence – Overview of ethical and societal concerns
ISO/IEC NP TR 24372	Information technology – Artificial intelligence (Al) – Overview of computational approaches for Al systems
ISO/IEC NP 38507	Information technology – Governance of IT – Governance implications of the use of artificial intelligence by organizations

Traditional Safety Case for Rail, Oil and Gas, & other Facilities



Intelligence – data and AI engine

Data – suitable for level of performance required in the application

Sensors required based on needs

Rich field of data

Changing nature of data

Different types of statistical bias (including over 40 types of human bias)

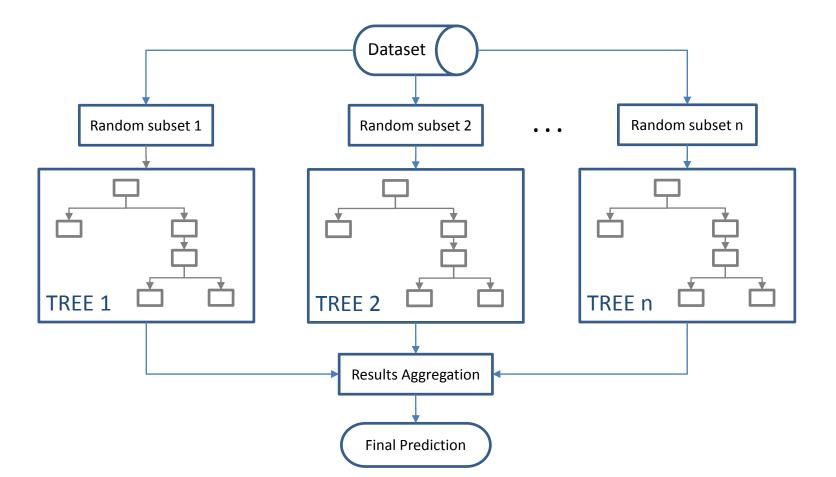
Dropout

Too simple

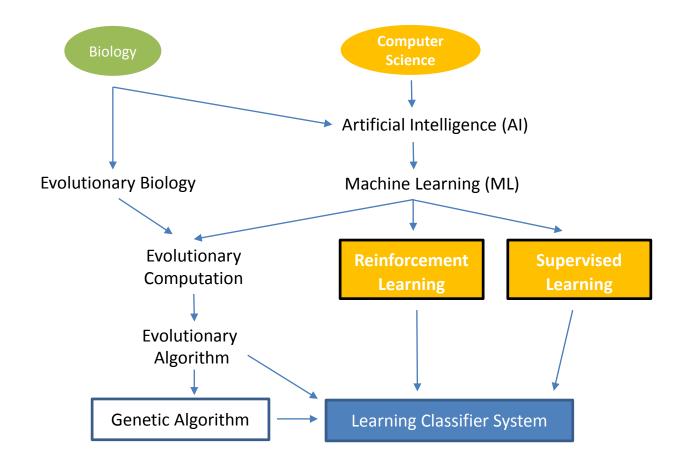
OverFitting, Regularization

Normalization

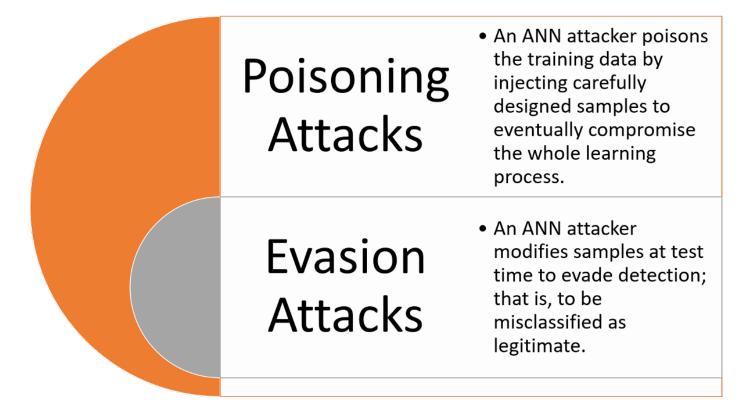
Ensemble Learning – used with AVs



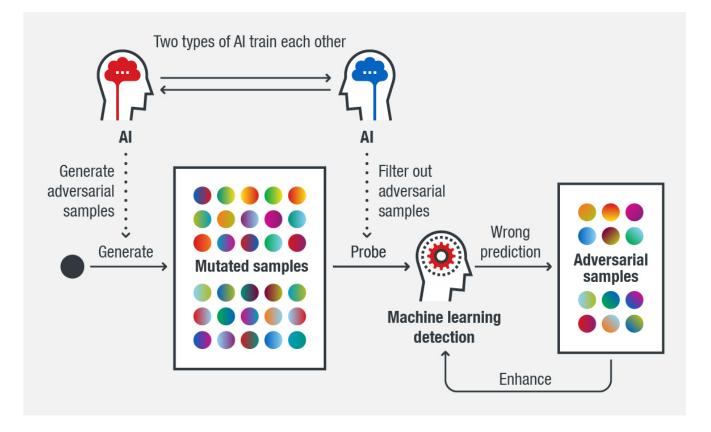
Ensemble Learning with Genetic algorithm – used with AVs



Adversarial Learning



Adversarial Learning

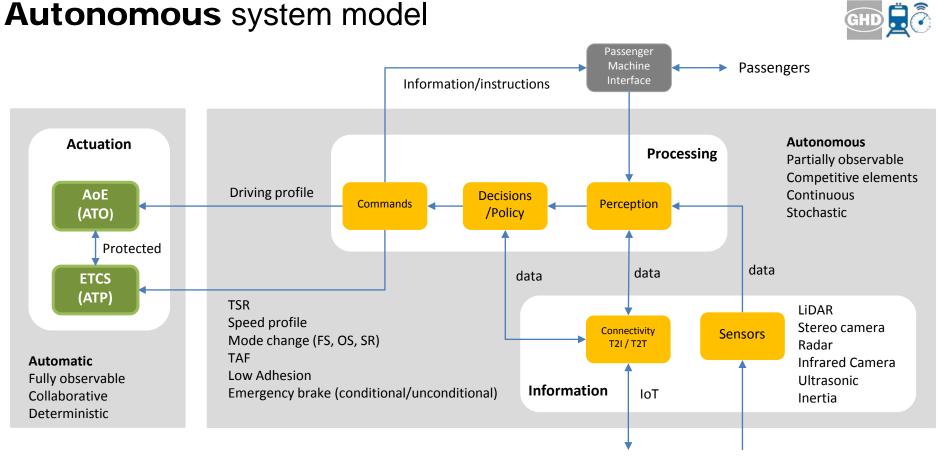


Conclusions



- Al in train automation is happening.
- Potential for safety, operational and financial benefit is high.
- There are also many risks that need to be managed.
- Validation and Verification approaches need updating for non deterministic complex systems.
- Higher standards of safety will likely be expected
- Human Factors, risk management and stakeholder engagement are very important.
- ETCS/CBTC provides foundational benefit for all levels of automation.
- There is potential of new Human Factors risks in a region of higher levels of automation as evidenced from aviation, in vehicles, and similarly for trains.
- Standards need updating, i.e. functional safety, new generation AI standards, IEEE standards,







Digital and Physical Environment