





Update on Global AI/ML Regulatory and Standards Landscape







The Regulatory Landscape

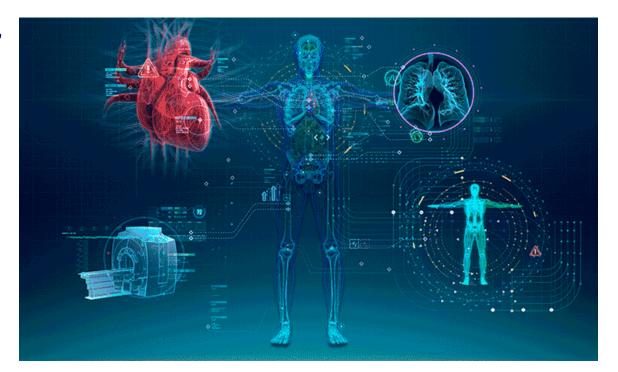
Eric Henry



A Useful Model

"All models are wrong, but some are useful."

-George E.P. Box-



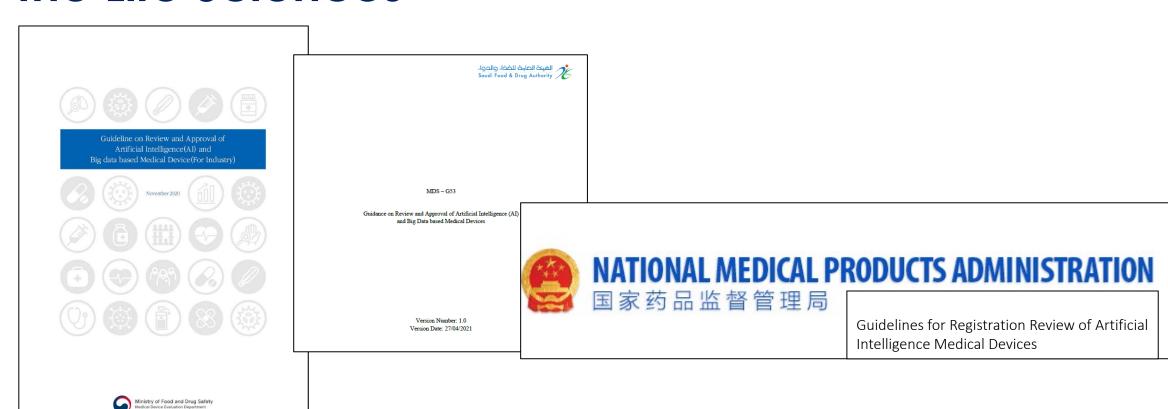








The Current Regulatory Landscape for AI/ML in the Life Sciences









Lots of Thinking





Perspectives and Good Practices for AI and Continuously Learning Systems in Healthcare

Data Quality for AI in **Healthcare**

FEBRUARY 2021

XAVIER HEALTH ↑ ① ○ ○ 74.1% · □ | ☐ ☑ ☑ | **△** Dezember 2020

DIN SPEC 92001-2



ICS 35.080; 35.240.01

Artificial Intelligence -Life Cycle Processes and Quality Requirements -Part 2: Robustness

Künstliche Intelligenz -Life Cycle Prozesse und Qualitäts Teil 2: Robustheit

Intelligence artificielle -Processus du cycle de vie et exig Partie 2: Exigences techniques e

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Generating and Evaluating Synthetic UK Primary Care Data: Preserving Data

Utility & Patient Privacy

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Abstract - There is increasing interest in the potential of synthetic data is which and benchmark functions have been supported by the property of the properties of property of the property group dataset [16]. In addition, anonymisation measures for real data may compromise data utility due to information loss [19].

I. INTRODUCTION

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5) Completeness - It can be difficult to conduct unbiased data (either longitudinally or latitudinally). Synthetic data can supplement real data by either filling gaps or enlarging a sub-

6) Benchmarking capabilities - This is useful when comparing different machine learning methods against a standardized dataset while focusing on a specific set of diseases (e.g. cardiovascular diseases).
Despite all the advantages outlined above, there is no

synthetic data generation and evaluation approach that can be applied to UK primary care data to ensure that the generated data preserve the ground truth (such as sensible biological relationships between variables) whilst ensuring privacy. In this paper, we propose a framework to enable syntheti

truth and the preservation of patient privacy. The following four key requirements are envisioned to be critical for making synthetic primary care data usable:

Preservation of biological relationships. The chosen
data variables and target studies should preserve the correct

dats variables and target indice should preserve the correct underlying relationships (e.g. finality specific diseases should not include male patients) or well-established clinical between the control of the control

Multivariate similarity. Data with multiple dimensions often have correlated structure between data variables. Retaining such correlational structure in synthetic data is crucial to ensure that it is a truthful representation of the real world. In some studies, initial knowledge (for e.g. clinical would in some studies, mind knowledge (for a g. clinical separties of finalizarity with nad would distribution) may be required to inform the development but this would only be formally the studies of concentral carbolic is relatively in 112 annual approach will be a challenge. Freservation of patient privace, Privacy might be a concern seven in the case of fully symbolic data. This can occur when symbolic personnel profices a very nimal fastest in terms of agregated characters to real-world data. For example, a small number (i) of rea desires costs in the









A Deeper Dive into the Chinese NMPA Al Guidance

Locked algorithms only

Methods and tools (including validation evidence) for:

- Data capture
- Data compilation
- Data labeling
- Data augmentation not allowed except for "adversarial sets," which must be detailed

Algorithm name, type, structure, input/output data types, flowcharts, programming framework, operating environment, and basis for selection

Algorithm training methods

Algorithm requirement specifications (including trace analysis)

Comparative verification

Verification and validation evidence

Incorporation of other guidelines

- Cybersecurity
- Mobile medical device
- · Medical device software
- Human factors

Al chip names, models, specifications, manufacturer, performance indicator, and "other information"

Risk management dossier

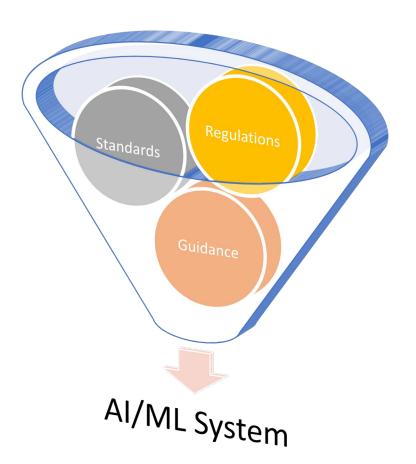
Use of third-party databases for validation (6 requirements)







Using Existing Frameworks



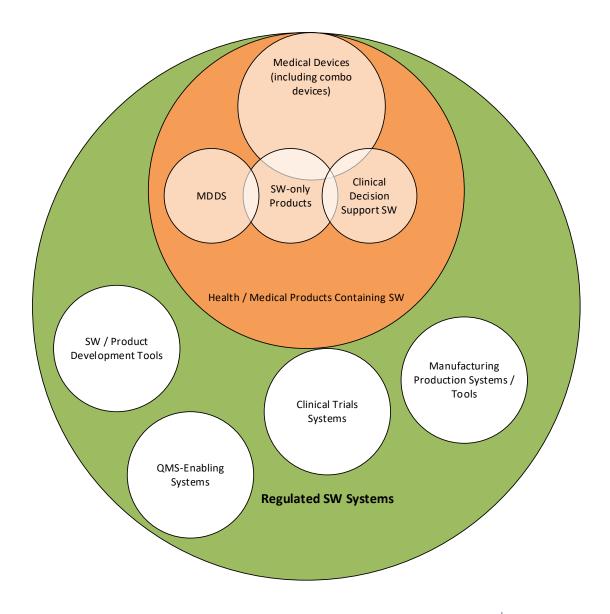








Know Thyself







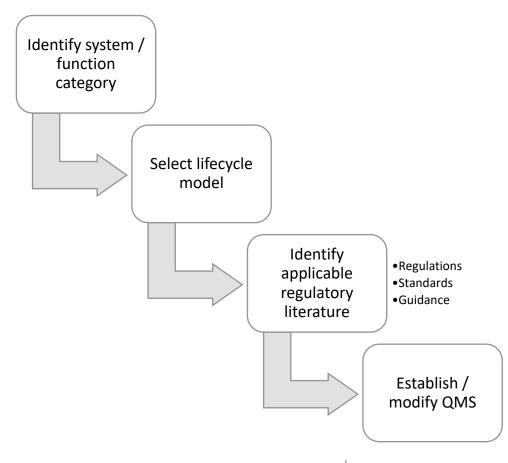


A Multipurpose Tool



Two Primary Purposes:

- Choose a lifecycle model applicable to the system
- Identify applicable regulatory literature as the basis for the Quality Management System (QMS)





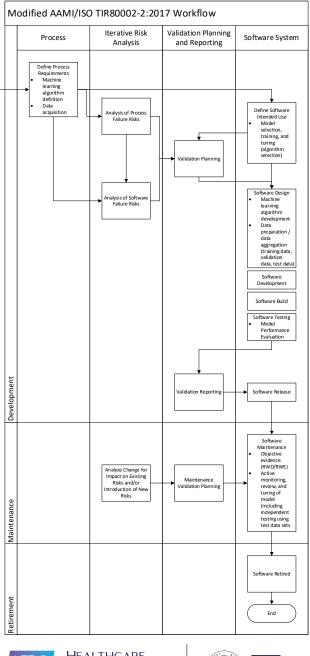






Non-Product Categories: Non-Manufacturing Systems

- Based on AAMI/ISO TIR80002-2:2017
- Clinical Trials Systems
 - Data analytics
- Software / Product Development Tools
 - · AI modeling
 - Defect management
 - NLP tools
- QMS-Enabling Systems
 - Complaint management / analysis
 - CAPA management / analysis

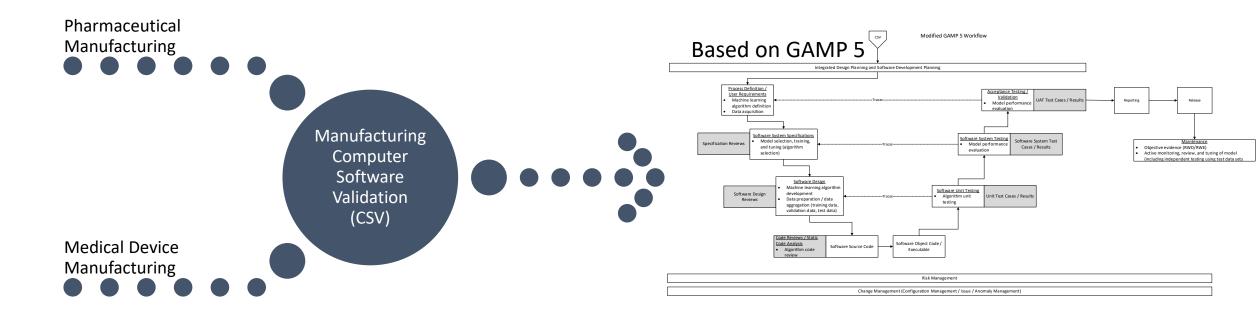








Non-Product Categories: Manufacturing Systems

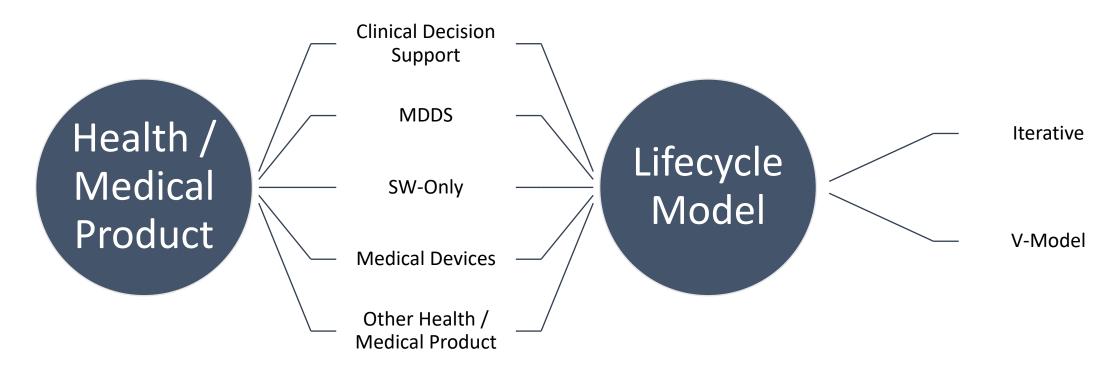








Product Categories









The Standards Landscape

Pat Baird



Why do we need standards? One reason: We forget items that are second nature to us

- Draft UL standard on fully-autonomous vehicles: what does the vehicle do when approaching a stoplight that just turned yellow?
- Defeating facial recognition software
- Defeating speed limit sign

Takeaway: we are in an age of "Narrow AI"



https://www.businessinsider.com/hackers-trick-tesla-accelerating-85mphusing-tape-2020-2





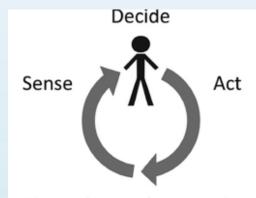


To what degree do we need "humans in the loop"?

First level needs a human to complete the task

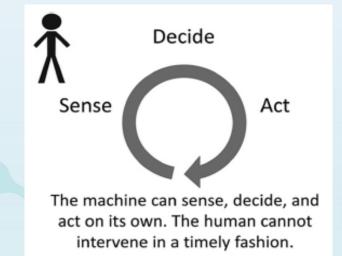
Second level allows for human over-ride

Third level does not allow for intervention



The machine performs a task and then waits for the human user to take an action before continuing. Sense Act

The machine can sense, decide, and act on its own. The human user supervises its operation and can intervene, if desired.











The International Medical Device Regulators Forum published a framework for assessing the risk of (non-AI) SaMD, according to the significance of the software when making decisions, and the criticality of the patient state

For AI, I think that "Treat or diagnose" has several levels, depending on how much freedom the AI is given – does the human have to "approve" before action is taken or does human have the ability to "override".

This has been included in CTA standards, is being considered by WHO, and has been adopted by AMA's payment codes.

State of Healthcare	Significance of information provided by SaMD to healthcare decision			
situation or condition	Treat or diagnose	Drive clinical management	Inform clinical management	
Critical	IV	III	II	
Serious	III	II	I	
Non-serious	II	I	I	

IMDRF "Software as a Medical Device: Possible Framework for Risk Categorization and Corresponding Considerations"

	Significance of information provided by SaMD to healthcare decision					
State of						
Healthcare	Treat or					
situation or	diagnose w/	Treat or	Treat or	Drive clinical	Inform	
condition	no	diagnose	diagnose	Management	clinical	
	intervention	w/Override	w/Approval	Ivianagement	clinical Management	
	possible					
Critical	???	??	IV	III	II	
Serious	??	IV?	III	П	1	
Non-serious	IV?	Ш	П		I	
	Automated/Autonomous Intelligence			Assistive Intelligence		









Standards Overview

There isn't much that is already published, but there are several efforts that I'll be talking about:

- ISO/IEC JTC1, SC42, developing horizontal standards for all industries. Many simultaneous projects and even more are being created. Not likely that these horizontal standards would be required for medical devices, but they may contain ideas that we like and would carry to healthcare.
- ISO/IEC TC215 (health software standards like 62304, 82304, 80001-x series, + 200 more standards) has created Task Force 5 not to write standards, but to provide some support for people wanting to write standards a landscape of various projects, develop use cases, do some education, liaisons, etc.
- IEEE also developing a number of AI standards, but only a few are specific to healthcare.
- CTA is developing general AI standards as well as healthcare-specific AI standards.
- AAMI & BSI have also started an AI standards committee.







ISO/IEC JTC1 SC42 – Standards for all Industries

Structure:

WG1 – Foundational standards (terminology, framewo WG2 – Big Data (vocabulary, reference architecture) WG3 – Trustworthiness (incl. risk, robustness, bias) WG4 – Use cases and applications

WG5 - Computational approaches & characteristics of

JWG1 (SC40) – Governance implications of Al JWG2 (SC7) – Testing of Al-based systems

AG1 – Management Systems Standard

AG2 – Al Systems Engineering AG3 – Al standardization roadmapping









- 4213 Information technology Artificial Intelligence Assessment of machine learning classification performance
- 4213 Assessment of machine learning classification performance
- 5338 Al system life cycle processes
- 5339 Guidelines for AI applications
- 5392 Information technology Artificial intelligence Reference architecture of knowledge engineering

5469 - Functional Safety

- 5471 Artificial intelligence Quality evaluation guidelines for AI systems
- 6254 Objectives and methods for explainability of ML models and AI systems
- 8200 Controllability of automated artificial intelligence systems
- 12791 Treatment of unwanted bias in classification
- 12792 Transparency taxonomy of AI systems
- 20546 Big Data Overview and Vocabulary
- 20547.1 Big Data reference architecture Part 1: Framework and application process
- 20547.2 Big Data reference architecture Part 2: Use cases and derived requirements
- 20547.3 Big Data reference architecture Part 3: Reference architecture
- 20547.4 Information technology Big data reference architecture Part 4: Security and privacy
- 20547.5 Big Data reference architecture Part 5: Standards roadmap
- 22989 Al Concepts and Terminology
- 23053 Framework for Al using ML
- 23894 Risk Management (ISO 31000, not 14971)
- 24027 Bias in AI systems and AI aided decision making
- 24028 Overview of Trustworthiness in Al
- 24029.1 Assessment of the robustness of neural networks Part 1 Overview
- 24029.2 Formal methods methodology
- 24030 Use cases and application
- 24368 Overview of ethical and societal concerns
- 24372 Overview of computations approaches for AI systems
- 24668 Process management framework for Big data analytics
- 25059 Systems and software Quality Requirements and Evaluation (SQuaRE)
- 38507 Governnance implications of the use of AI by organizations.
- 42001 Management system
- 22100-5 Safety of machinery Relationship with ISO 12100 Part 5: Implications of artificial intelligence machine learning
- 5259-1 Data quality for analytics and ML Part 1: Overview, terminology, and examples
- 5259-2 Data quality for analytics and ML Part 2: Data quality measures
- 5259-3 Data quality for analytics and ML Part 3: Data Quality Management Requirements and Guidelines
- 5259-4 Data quality for analytics and ML Part 4: Data quality process framework

ISO/IEC JTC1 SC42 has a lot of projects...

- Some of these topics are not what you think -- note that "risk management" is enterprise-risk, not safety-risk. If you want safety, look at 5469
- Some new projects include "Oversight"; another is exploring the positive use-cases for Al applications.
- Also be aware that some legislators assume that horizontal standards can apply to all industry – after all, they are horizontal!

HEALTHCARE PRODUCTS COLLABORATIVE





IEEE

IEEE also has multiple (horizontal) AI standards, as well as a few specific to healthcare:

P2801 Recommended Practice for the Quality Management of Datasets for Medical Artificial Intelligence Recommendation

P2802 Standard for the Performance and Safety Evaluation of Artificial Intelligence Based Medical Device: Terminology

And although it's not healthcare-specific...

P7003 Algorithmic Bias Considerations











TC215 Task Force 5: AI Stuff!

- In 2019, ISO/TC215 created an Ad Hoc Group (AHG2) on Application of AI in Health Informatics; the resulting report had over 30 recommendations so TC215 formed TF5 to implement the recommendations, including:
 - Create (and maintain) a standards landscape so that TC215 knows what other teams are working on.
 - Setup collaborations with other organizations
 - Develop common definitions for TC215 teams to use
- The idea is not to have TF5 write standards, but for them to provide infrastructure to help with future standards.









IEC TC62 Software Network and Artificial Intelligence advisory Group (SNAIG)

- TC62 formed a group to make recommendations about what standards might need to be updated or created. Their final report will be discussed in an upcoming TC62 Plenary meeting.
- The group came up with a process for identifying and prioritizing future projects – this was not simply a collection of people's opinions.







IEC/TC62 PT 63450 Al-enabled Medical Devices – Methods for the Technical Verification and **Validation**

"This document establishes methods for medical device manufacturers to **verify** and validate artificial intelligence / machine learning-enabled medical devices (AI/ML-MD), i.e. medical devices that use artificial intelligence, in part or in whole, to achieve their intended medical purpose. This includes verification and validation activities for the **model** of the artificial intelligence as well as **selection**, **metrological characterization** and **management** of the **data sets**.

Such activities are implemented at various stages of the medical device lifecycle, especially including design control, monitoring and design change.

This document is also applicable to any hardware or software utilizing artificial intelligence that impacts the intended use of a medical device"



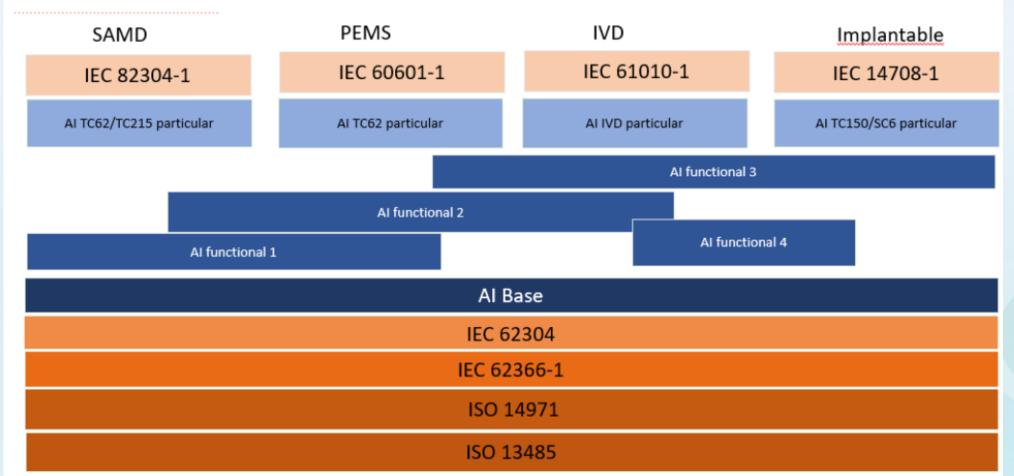








- a common Al Base Level* applying to all Al-enabled medical devices (e.g., bias management, Testing of ML-enabled Medical Devices); and
- Specific Al aspect applying to a group of Al-enabled medical devices (functional*)
 (e.g., image analysis, waveform analysis); and
- Al methods or measures specific to intended purpose (particulars*) (e.g., 2D x-ray for dentistry, tongue image analysis (ISO 20498-1)) documents.



IEC TC62 should consider preparing NPs regarding the following standardization fields in green in Figure 2.

Overview					
Existing standards sufficient; s	some additions for the application to	AI-MD software	different partners		
(all sorts of MD) QMS, RMS, PMS, Security, Privacy	Clarification where additional factors are to be considered - basic standards are basically sufficient		Impact of patient as operator in the private space and cloud are hardly considered at present		
Traditional ME (Hardware + non Al Software)	covered	covered			
SAMD	Extension/addition to 62304 for ML	clinical validation for SAMD			
New standards are necessary;	IEC TC62 is in the lead or works in a J	WG with partne	rs		
(ML – MD, maybe AI - MD in general) Logic component	Quality metrics for external data component	clinical validation (data aspects,	Techno-Vigilance necessary		
	Quality criteria and docu req. for "logic" component	statistical effectiveness)	Clinical utility (effective benefit at the site)		
	Methods for building test sets		Transparency in Usage and operation		
(AI – MD) Bias (Ethics?)	Overarching process standard and consideration in all columns for symbolic AI + ML				
New standards are necessary;	ISO TC215 is in the lead or works in J	WGs; IEC TC 62 o	contributes		
(ML – MD) Data component	Data Lifecycle process standard (incl. Selection, collection, vetting, documentation etc.)				
	Quality standards in design and development	Including changes based on PMS data or continuous learning			
	Quality assurance methods and quality metrics				
	Methods for validation				

Figure 2: Proposed IEC/ISO cooperation for AI/ML Medical Device Standards



Possible projects for TC62 or JWG

- ISO/IEC Develop a standard for documentation requirements providing evidence of the fulfullment of quality criteria
- Methods to build training and test sets from available data
- Clinical validation (might be handled elsewhere)
- Techno-vigilance (post-Market surveillance, but no concrete recommendations right now)
- Transparency (e.g. compliance, explainability, trustworthiness, etc.)
- Bias management (monitor SC42 TR 24027)
- Security







Possible Collaboration with TC210

- Addition to ISO 13485 Quality Management
- Addition to ISO 14971 Risk Management
- Addition to ISO TR 20416 Post-market Surveillance
- Addition to ISO 20417 information supplied by manufacturer
- Addition to IEC 62366-1 Usability
- Collaborative work for bias management...







Possible Collaboration with TC215

- Quality criteria and metrics for the data used
- Terminology





AAMI/BSI Standards Collaboration

After publishing a few whitepapers with BSI, AAMI/BSI started working on another whitepaper regarding AI risk management. Feedback we received on the whitepaper was "why are you doing another whitepaper? A standard or a TIR would be more useful..."

An official AI committee was formed, and the first project is TIR 34971 – Guidance on the Application of ISO14971 to Artificial Intelligence and Machine Learning.

You might remember that the ISO/IEC 24971 provides guidance on how to implement 14971, and includes an annex of special considerations for IVDs. The risk management process is the same, but there are some things you might not have considered when it comes to IVD risk management.

34971 is following that same pattern – the process is the same, but there are new ways to fail. There are different things to look for (e.g. bias), RAPS there are different risk controls to consider.

DRAFT AAMI BSI Joint Report

c BSI 34971/AAMI TIR 34971:2021

DRAFT: Guidance on the Application of ISO 14971 to Artificial Intelligence and Machine Learning

Approved <mark>Day Month Year</mark> by AAMI & BSI

> This strawman document provides a first framework and provisional content for the development of BSI 34971/AAMI 34971. The intent is that this document will be a companion to ISO 14971:2019 for those performing risk management for AI or ML incorporating medical devices





Consumer Technology Association (CTA)

CTA is the trade association for the consumer technology industry (all consumer industries – not just healthcare)

Established AI working group, published two whitepapers in 2018 – general introduction & use cases.

Al standards committee (R13) & Health Care working group (R13 WG1). Publications include:

"Definitions / Characteristics of AI in Health Care (ANSI/CTA-2089.1)"

"The Use of AI in Health Care: Trustworthiness (ANSI/CTA-2090)"

"The Use of AI in Health Care: Managing, Characterizing, and Safeguarding Data"

Current project is about Bias Management





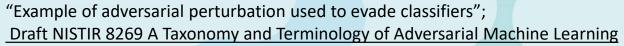




Cybersecurity!

- ML systems have a lot of data. Potentially very attractive data. This data is often handled by multiple stakeholders as it is passed from one system to another.
- Due to the nature of the data & how ML systems work, it might not be obvious that there has been a security issues...











Cybersecurity!

- Standards organizations are starting to recognize that this is an issue and have started some projects to address it.
- CTA has published a whitepaper; this generated enough interest that they are currently working on a standard.
- ISO/IEC SC27 & SC42 have joined forces to work on an international standard for security of ML systems; however, this is a horizontal standard across all sectors – I am hoping that we can have an informative annex that addresses some of the unique considerations that we have in healthcare.











Questions?