# IEC 63521 MLMD - Performance MLMD: Machine Learning-enabled Medical Device

### **Evaluation Process**

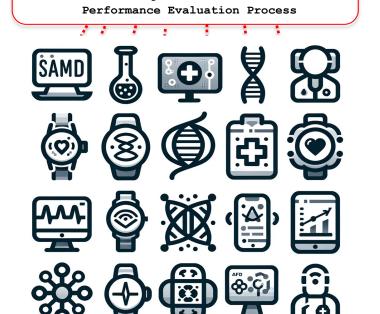
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**IEC 63521** 

Machine Learning-enabled Medical Device -

#### **Contents**



- Definition of MLMD
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  - Three pillar model of Performance (SV, TPV, CPV) & Evaluation flow
  - Safety and Effectiveness of ML

#### Considerations

- Align with baseline standards 62304, 13485, 14971, 62366 ···
- Align with regulatory requirements and International standardization
- GenAl, Foundation model, LLM

### Conclusion



### Simple Curriculum Vitae



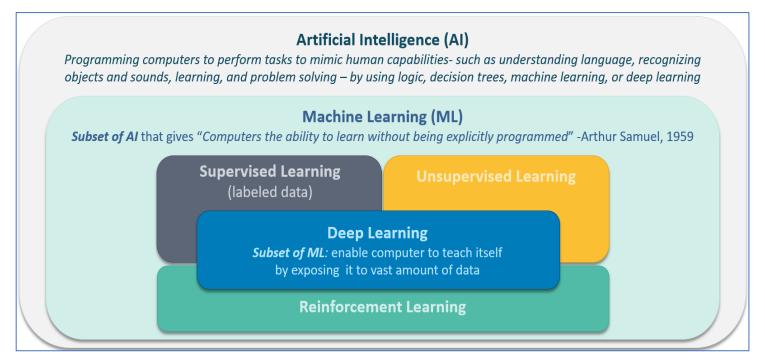
- IEC TC62 A, D (Medical equipment, software and systems)
  - (Convenor) PT 63521 (Project Team 63521) AI/ML-MD performance evaluation
  - (Project leader) IEC 63521 Machine Learning-enabled Medical Device Performance Evaluation Process
  - SNAI (Software Network AI) Adhoc group member
- ISO/IEC JTC 1/SC 42 (Artificial Intelligence)
  - (Project editor) ISO/IEC TS 29119-11 Testing for AI Systems
- IEC TC124 (Wearable Electronics)
  - (Convenor) ahG7 (Future Use Cases for Wearable)
  - (Project Leader) IEC 63203-402-2:2024 (Step counting)
- ISO/IEC JTC 1/WG 12(3D Printing & Scanning)
  - (Project Editor) ISO/IEC 3532-2:2024 Medical image-based Modeling Part2: Segmentation.
  - (Project Editor) ISO/IEC CD 8803 accuracy and precision evaluation process for modeling from 3D scanned data
  - (Proposed NP) Phantom-based evaluation methods for 3D printing modelling software
- IMDRF(International Medical Device Regulators Forum) AIMD WG
  - Chair of Korean mirror committee
- KoSAIM Director of Standardization (1st term: 2018–2020, 2nd term: 2021–2023)
- Head of the Standards and Technology Subcommittee of the Medical AI Technology Standardization Forum



### **Definition of MLMD**



- •IMRDF's MLMD Definition
  - MLMD (Machine Learning-enabled Medical Device )
    - "A medical device that uses machine learning, in part or in whole, to achieve its intended medical purpose"

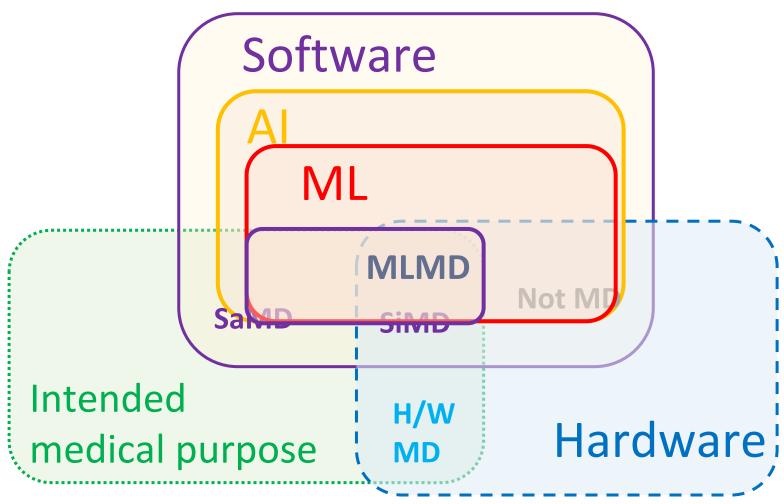


"the ability to learn without being explicitly programmed" → learn from DATA



### **Categorization of MLMD**





SaMD: Software As a Medical Devices

**SiMD**: Software In a Medical Devices (sometimes referred to as "embedded" or "part of")

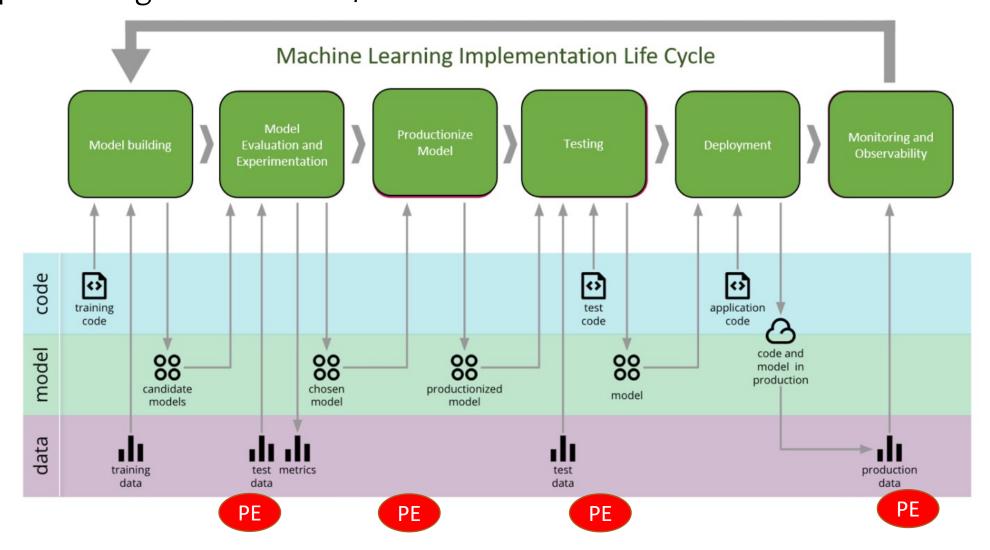
**MLMD**: Machine Learning-enabled Medical Devices



### Why Performance evaluation process?



Performance Evaluating of a AI/ML-MD product is an essential part of implementing an effective AI/ML model





### Why Performance evaluation process?



Performance evaluation of a medical device is a continuous process by which data are assessed and analysed to demonstrate the scientific validity, technical(analytical) performance and clinical performance of that device for its intended purpose as stated by the manufacturer.

The Manufacturer used assessment and analysis of clinical data of a device to verify the clinical safety, performance, and effectiveness of the device.

### <u>Performance evaluation in machine learning</u> is crucial for several reasons, which underscore its necessity:

#### Task processing ability verification

• Performance evaluation allows us to verify how well a model can make task's goal(predictions or classifications). This helps determine if the model is ready for real-world data application.

#### Model Comparison

• When multiple models have been developed, performance evaluation is necessary to compare which model performs better. This enables the selection of the best model.

#### Detecting Overfitting

• It allows for the detection of overfitting, where a model is too closely fitted to the training data and performs poorly on new data. By comparing the performance of a model on training data versus validation or test data, overfitting can be identified.

#### Hyperparameter Tuning

• Performance evaluation is essential during the process of tuning hyperparameters to optimize model performance. By experimenting with various hyperparameter settings and evaluating the outcomes, the optimal configuration can be discovered.

#### Evaluating Generalization Ability

• It's important to ensure that a model not only performs well on training data but also on new, unseen data. Performance evaluation helps assess a model's generalization ability.

#### Achieving Business Objectives

• Performance evaluation is needed to verify if a machine learning model meets business goals or requirements. For instance, specific accuracy or response time may be required, and its achievement needs to be evaluated.

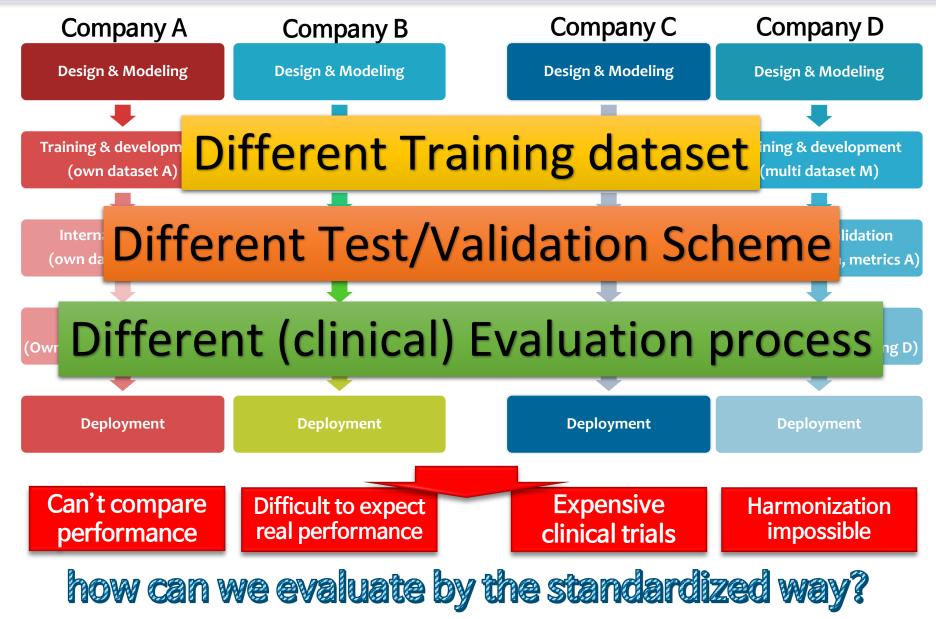
#### Providing Reliability and Transparency

• Systematically evaluating and documenting model performance enhances its reliability and provides transparency to users or stakeholders.



### Why Performance evaluation process?







### History of IEC TC62 PT8 (PT 63521)



- Oct 15,2021 TC62 CIB (62/399/Q, 6 week, by Nov 26)
- Dec 1, 2021 Established PT8 and registered PWI62-3 (according to 62/409e/RQ)
  - Project Leader: Jonghong Jeon (KR)
- Dec 14, 2021 Kickoff meeting of PT8
- Sep 8, 2022 <u>18<sup>th</sup> meeting of PT8 (1<sup>st</sup> phase)</u>
- Sep 15, 2022 NP submit to TC62 plenary
- Nov 11, 2022 CAG presentation and Resolution 62/2022/13
  - TC 62 appreciates the draft for a new work item proposal submitted by PT 8 and requests a further improvement of the document by the end of 2023, in particular by clarifying the terminology and the concepts used. TC 62 hopes to circulate the NP by early 2024.
- May 2, 2023 kickoff meeting of PT8 (2<sup>nd</sup> phase)
- July 17, 2023 9<sup>th</sup> meeting of PT8 (2<sup>nd</sup> phase)
- July 25, 2023 2<sup>nd</sup> NP submit to TC62 plenary
- Aug 11, 2023 Start 62/474/NP ballot (until 11/03)
- Nov 3, 2023 Approved NP, Changed Project Team number to PT 63521
- 1st PT63521 meeting: Nov 29
- 2<sup>nd</sup> PT63521 meeting: Dec 13
- 3<sup>rd</sup> PT63521 meeting: Jan 9, 2024 ~ (bi-weekly)



#### **Current Member of PT63521**

#### ■13 nation

• BE, CH, CN, DE, FR, GB, IN, IT, JP, KR, NG, NL, US

### ■33 experts

Medtronic, Phillips, Elekta, Varian,
 Siemens-healthineers, esaote,
 CFDA, deepeyevision,
 kakaohealth, ETRI ···





List All Experts by NC

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Total number of experts: 33
Total number of records: 33



### Goal of IEC 63521



- 1. It is **based on the existing standard systems of IEC TC62** and should be avoided as much as possible of redundancy.
- 2. It should provide the process (or framework) for verifying the safety/efficacy of AI/ML-MD.
- 3. Should be applicable to various types of Al/ML-MD
- Should be able to support various Al/ML tasks, models, data modalities, evaluation metrics, etc.
- 5. It should be applicable to various types of medical devices (PEMS, SaMD, IVD, Dentistry, Implantable, and medical robots, etc.) that utilize AI/ML technology
- 6. Evaluation requirements, methods, standards, etc. should be **established through the process.**
- 7. The evaluation process should be able to support conformity to global regulations



### Scope of IEC 63521



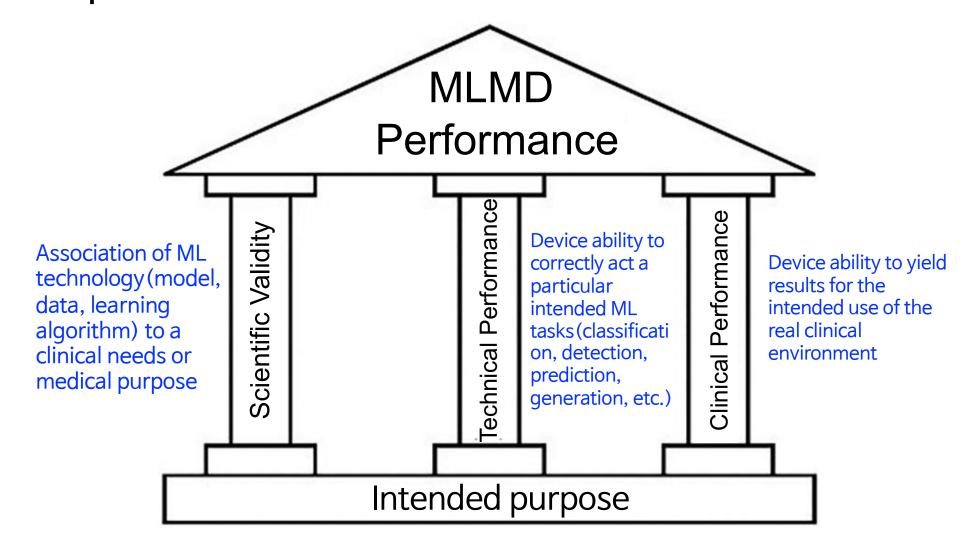
- ■This document <u>defines a standardized performance evaluation process</u> for Machine Learning-enabled Medical Devices. The <u>set of processes</u>, <u>activities</u>, <u>and tasks</u> described in this document establishes a common framework for MLMD performance evaluation processes.
- ■This process is to <u>assist manufacturers to evaluate</u> the ML suitability, the technical and clinical performance of the MLMD. It <u>may</u> be used for evaluation the performance to assure expected performance <u>during post-market monitoring</u>.
- This document is applicable to the <u>performance evaluation of all forms of MLMD</u>, comprising ML components and, where appropriate, <u>the integrated ML component</u> <u>with non-ML components</u>. The most important factor to consider performance evaluation of ML is whether it <u>affects the intended purpose and the safety and effectiveness of medical device</u>.



### Basic [1] - Three pillar model



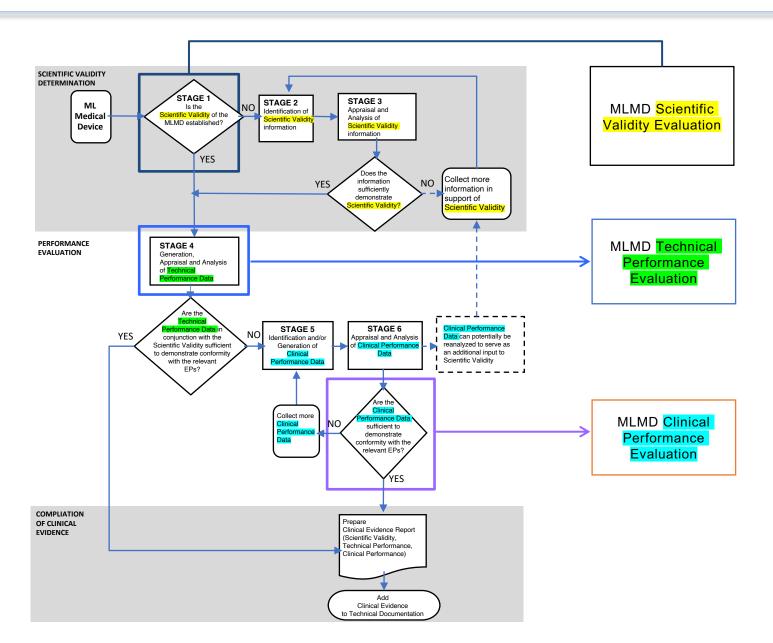
### Three pillar model of Performance





### **Basic [2] - Evaluation flow**







### Basic [3]: Safety and Effectiveness of ML



#### 1. Data quality and management

• Data plays a crucial role in the performance of machine learning medical devices. Ensure accurate data collection, preprocessing, labeling, and annotation.

#### 2. Model validation and verification

• Ensure the machine learning model serves its intended purpose and provides accurate predictions.

#### 3. Algorithm transparency and explainability

• The decision-making process of the machine learning model should be understandable and transparent.

#### 4. Bias and fairness

The machine learning model should provide unbiased predictions for different patient groups and clinical situations.

#### 5. Robustness and generalization

• The machine learning model should be able to adapt to different patient populations, clinical situations, and data quality changes.

#### 6. Personal data protection and data security

· Machine learning medical devices must protect patient privacy and ensure data is secure from external threats.

#### 7. Software and hardware stability

• The software and hardware components of machine learning medical devices must be safe from errors, defects, and failures.

#### 8. User interface and human factors consideration

• The user interface of machine learning medical devices should minimize user errors and enable effective device use.

#### 9. Clinical evaluation and validation

• The safety and effectiveness of machine learning medical devices must be validated in real clinical environments.

#### 10. Benefit and risk

Al should never cause foreseeable or unintentional harm

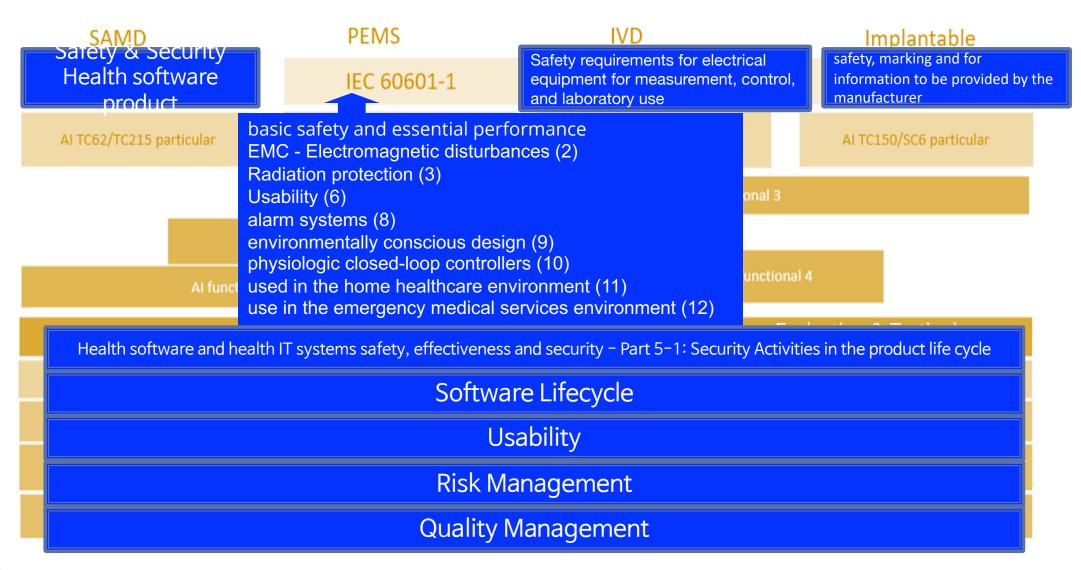


### PE Requirement analysis(...ing)



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						Safety & Effectiveness Category Three pi							model			5. Development process							6. maintenance PROCESS			
No.	category	Candidate Requirements	G	1 2	3	4	5 6	7	8 9	9 10	sv	TPV	CPV	5.1 develop ment planning	ments	ura	ect de	5.4 etailed design	5.5 UNIT implement ation and verificatio n	n and	release	6.1 mainten ance plan	6.2 Problem and modificati on analysis	6.3 Modifica tion impleme ntation	7. RISK MANAGEME NT PROCESS	8. configurati managemer PROCESS.
1	ABSTRACT	A summary shall be provided, detailing objectives, study design, setting, participants, sample size, predictors, outcome, statistical analysis, results, and conclusions.	0			П					О															
2	Background and objectives	The scientific background and rationale for the study shall be explained, including the medical context (diagnostic or prognostic) and references to existing models.	0			П					О															
3		Specific objectives or hypotheses shall be clearly stated.	0	$\neg$	$\top$	$\vdash$	$\top$	$\top$	$\neg$		0						$\neg$									
4	_	Any pre-existing evidence supporting the AI intervention shall be described.	<del>                                     </del>	$\perp$	+	$\vdash$	+	+		)	0		$\vdash$		1	+	+			<b>†</b>		+	+-			
5	┪	The intended use of the Al intervention in the clinical pathway shall be explained, including its purpose and intended users (e.g., healthcare professionals, patients, public).				H			,	+	0															
6		The rationale for the choice of comparators in the study shall be provided.	$\vdash$		+		+	+		)	0				1	+	-			<del>                                     </del>		+	+			<del> </del>
7	clinical problem	The clinical problem for the model's application shall be detailed.	t	$\vdash$	+	$\vdash$	+	+	1	_	0	_	$\vdash \vdash \vdash$		+		+				+	+	+			
8		Characteristics of the cohorts (training and test sets) shall be described in detail.	H	0	0	0	0	+ +	1	_	0	_	$\vdash \vdash \vdash$		+	+	+				+	1	+			
9	cohorts	The representativeness of the cohorts (training and test sets) in real-world clinical settings shall be demonstrated.	-	0		0	0	+ +	(	-	0															
10	Intended use	The targeted medical condition(s) and problem(s), current standard practice, and intended patient population(s) shall be described.		0	,	П		П	-		0															
11	SOTA	The state-of-the-art solution used as a baseline for comparison shall be identified and detailed.	H	0	,	$\Box$	$\top$	$\top$	$\neg$		0					_										
12		De-identification methods shall be described.	$\vdash$	0		$\vdash$	0	+	$\neg$		<del>  </del>	0				+	$\top$						$\overline{}$			<u> </u>
13		The sufficiency of data for training the algorithm shall be evaluated.		0	0	0 (		$\top$	$\neg$		1	0				$\top$										1
14		Data sources shall be identified.	_	0	_	0	0		$\neg$		1	0				_										
15		Data pre-processing steps shall be described.		0	_	0	0	$\rightarrow$	$\top$		T	0			1	+	$\top$					_	$\overline{}$			1
16		The quality of data used for training the algorithm shall be assessed.	_	0	_	0 (	_	+	$\neg$			0			1	+	$\neg$			<del>                                     </del>		1	+			1
17		The extent of data accuracy and bias-free nature shall be evaluated.	_	0	-	0	_	+	$\dashv$		T	0					$\top$						+-			
18		Standardization and interoperability of data shall be confirmed.		0	0	_	0	+	o	$\top$	+	0			1	+	$\top$					1	+			
	Data collection methods	Plans for assessing and collecting trial data shall be outlined, including processes to promote data quality.	_	ō	0	_	ō	+	o	+	+	0	$\Box$		1	+	+						+			
	data independence	Independence between training and test sets shall be confirmed.	_	0 0	_	_	<del>-</del>	+	$\dashv$		+	0			1	+	_			<del>                                     </del>		+	+			<del>                                     </del>
21	· · · · · · · · · · · · · · · · · · ·	Composition and role of the Data Monitoring Committee shall be detailed, including its independence and conflict of interest policies.	-	0	_	0						0														
22		Intended sample size and determination method shall be stated.	$\Box$	0	$\top$	0	十	$\top$	$\neg$		T	0					$\top$									
23		Data assignment to partitions and proportions shall be specified.	$\Box$	0	T	0	$\neg \vdash$	$\top$	$\neg$	$\top$	Т	0	$\Box$				$\top$						1			
24		Definition of the ground truth reference standard shall be sufficiently detailed for replication.		0	0	0 (	0	$\top$	$\top$	$\top$	Т	0	$\Box$				$\top$					1	1			1
25		Rationale for choosing the reference standard shall be provided.		0		0		$\top$	$\neg$		Т	0					$\top$					1	1			1
26		Source of ground-truth annotations and annotator qualifications shall be described.	$\Box$	0	0	0	$\neg \vdash$	$\top$	$\neg$	$\top$	Т	С	_		ľ						1	•	+	-		1
27	•	Annotation tools and measurement of variability shall be detailed.	H	0	0	0		$\Box$			1	С	- 1		_											_
28		Methods for mitigating variability and resolving discrepancies shall be outlined.		0		0	$\top$	$\top$	$\neg$	$\top$	T	С	4						7	Softwar	e Risk Ma	nagemer	nt			
	models	Information on evaluated models and selection criteria shall be provided.	$\sqcap$	0 0	0	0	0	$\top$	$\neg$	$\top$	Т	С	စ္က ၂							Concessar	C .tiak ivid	Permer				_
30	transformation	Data transformations prior to model application shall be described.	$\Box$	0	0	0	$\top$	$\top$	$\neg$	$\top$	Т	С	62304		7						1		5.6	SW	V	
31		Methods for collecting, sharing, and maintaining personal information about participants to protect confidentiality shall be detailed.					0						띩	Develo	SW opment	Require		Archi	3 SW tectural	5.4 SW Detaile	Imple	SW Unit mentation	Integra	tion &	5.7 SW System	5.8 SW Release
32	Consent or ascent	The process for obtaining informed consent or assent from potential trial participants or authorized surrogates shall be specified.					0					С	cobe	Plar	nning /	g Analysis Design Design & Verification Testing Testing										
33		Financial and other competing interests of principal investigators for the overall trial and each study site shall be disclosed.				Ш	0	+				С	ပ္လ						8 Softv	vare Cor	figuration	Manage	ment			
34	Research ethics approval	Plans for seeking research ethics committee/institutional review board approval shall be outlined.	1		- 1	1 1	0			- 1		· ·									Problem I					1

### IEC 63521: Focused on Safety and Effectiveness of ML





### IEC 63521: Focused on Safety and Effectiveness of ML

SAMD atv. & So

Safety & Security
Health software
product

Al TC62/TC215 particular

basic safety and essential performance EMC - Electromagnetic disturbances (2) Radiation protection (3) Usability (6) alarm systems (8) environmentally conscious design (9) physiologic closed-loop controllers (10) used in the home healthcare environment (11) use in the emergency medical services environment (12)

Al TC62 particular

IVD

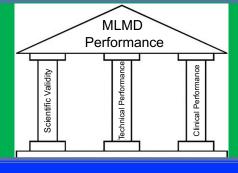
Safety requirements for electrical equipment for measurement, control, and laboratory use

Al IVD particular

**Implantable** 

safety, marking and for information to be provided by the manufacturer

Al TC150/SC6 particular



Safety & effectiveness of ML

Health software and health IT systems safety, effectiveness and security - Part 5-1: Security Activities in the product life cycle

Software Lifecycle

**Usability** 

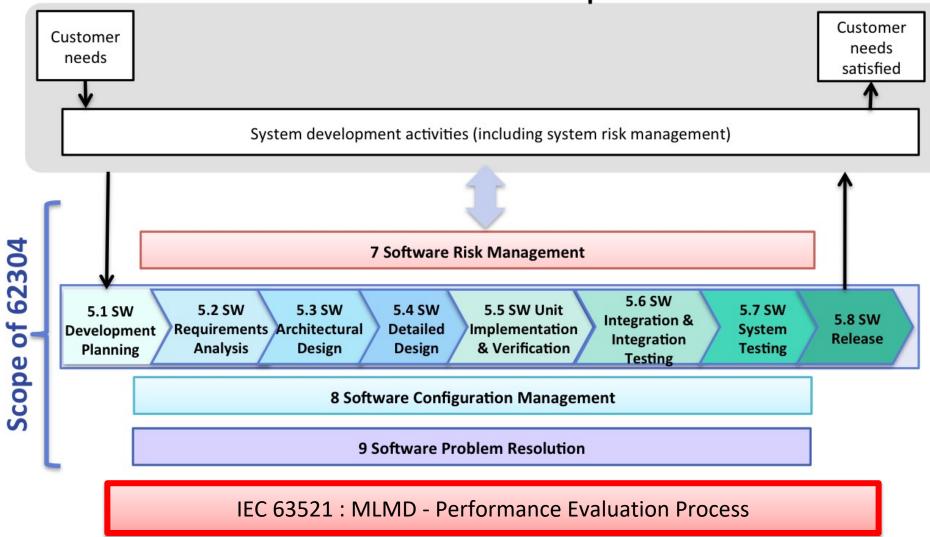
Risk Management

**Quality Management** 



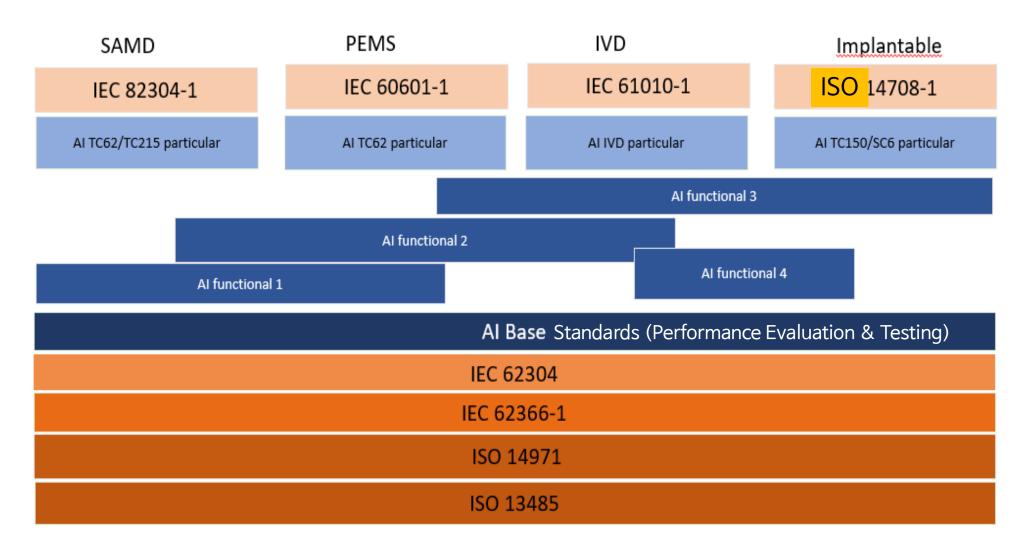
### Considerations 1: Align with IEC 62304 process

### IEC 62304 Software Development Processes





### Considerations 2: IEC TC62's AI standardization roadmap





### IEC TC62's Al standardization roadmap - extend

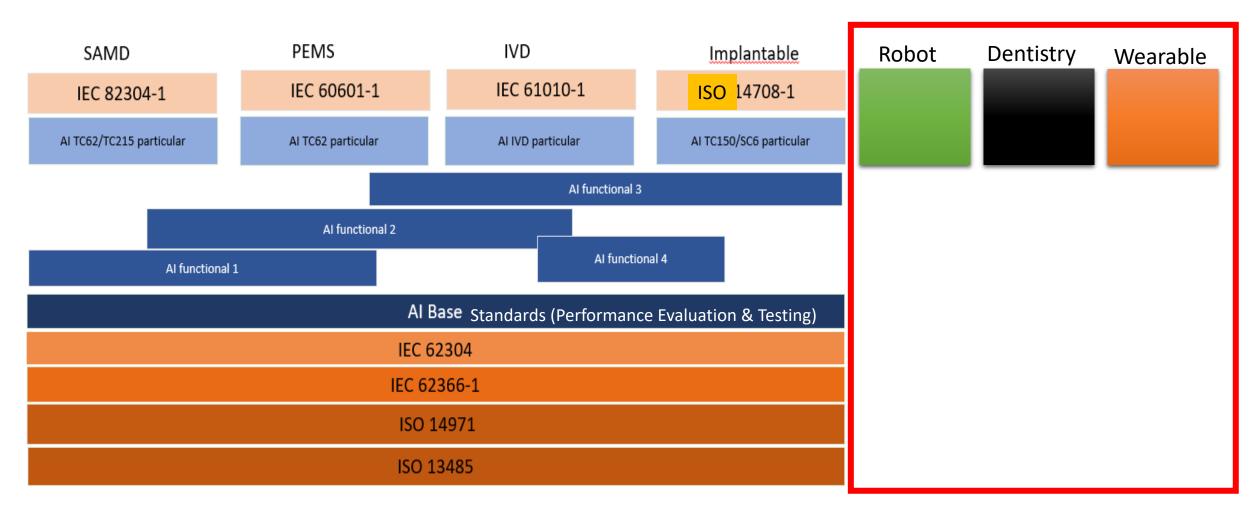


Figure 1 in the 4th SNAIG Report (62/432/INF)



### How to role as Foundational level standard



#### Level 3 - application level

- IEC 63524 Al enabled medical device –
   Computer assisted analysis SW for pulmonary images Algorithm performance test methods
- ... see "list of suitable standards" (1)

#### Level 2 - functionality level

- IEC TR 60601-4-1 Medical electrical equipment and medical electrical systems employing a degree of autonomy
- ... see "list of suitable standards" (1)

#### Level 1 - foundation level

- IEC 63450 Testing of Artificial Intelligence / Machine Learning-enabled Medical Devices
- ISO/IEC TS 4213 Artificial intelligence –
   Assessment of machine learning classification performance
- IEC 63521 Artificial Intelligence/Machine Learning-enabled Medical Device – Performance Evaluation Process
- ... see "list of suitable standards" (1)

#### Level 3 - application level

application requirements, concepts, or processes related to AI/ML technologies relevant for specific use cases

#### Level 2 - functionality level

requirements, concepts, or processes for functionalities realized with Al/ML technologies relevant for various use cases

#### IEC 63521: MLMD - Performance Evaluation Process

Level 1 - foundation level

fundamental requirements, principles, concepts, processes, or methods related to AI/ML technologies

Established "Base Standards" for all medical devices

see "list of suitable standards" (1)

Established general safety standards for medical devices IEC 82304-1 Health software –
Part 1: General requirements for
product safety

SaMD

IEC 60601-1 Medical electrical equipment – General requirements for basicsafety and essential performance

PEMS

IEC 61010-1 Safety requirements for electrical equipment for measurement, control and laboratory use

IVD

ISO 14708-1 Implants for surgery — General requirements for safety, marking and for information to be provided by the manufacturer

Implantable devices



### Align with other standards

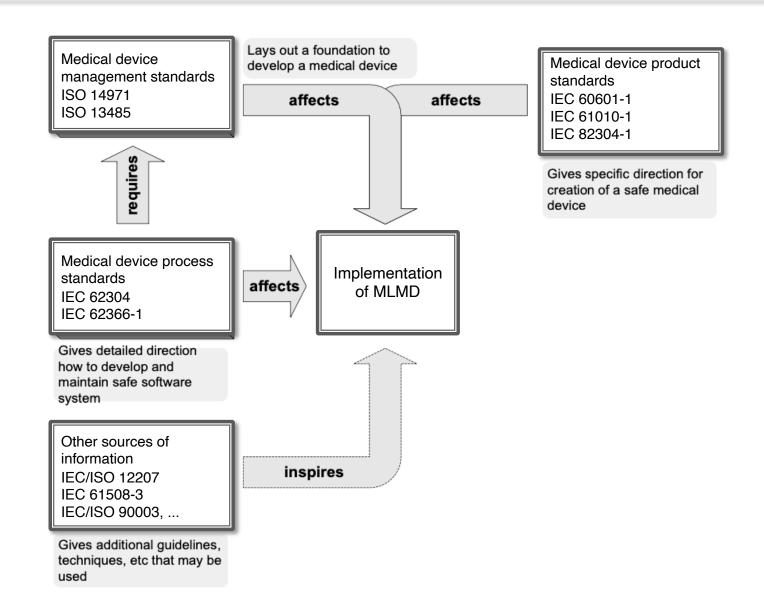


1) Adapted from Figure 1 in the 4th report of IEC Standards architecture for medical devices using artificial intelligence (AI) technology<sup>1)</sup> TC62 SNAIG (62/432/INF) PEMS 4) IVD 5) SaMD 3) Implantable devices IEC 82304-1 Health software -IEC 60601-1 Medical electrical IEC 61010-1 Safety requirements for ISO 14708-1 Implants for surgery -Established sectoral Part 1: General requirements for equipment - General requirements for electrical equipment for General requirements for safety, base standards product safety basic safety and essential performance measurement, control and laboratory marking and for information to be provided by the manufacturer Point Dig Pathology & algor-based image Al implantable particular 1 Al medical device - Computer assisted AI PEMS particular 1 analysis analysis SW for pulmonary images -Algorithm performance test methods Al particular standards Al implantable particular 2 AI PEMS particular 2 Al IVD particular 2 (after publication) Data handling in Al supported 2D Al implantable particular 3 AI PEMS particular 3 Al IVD particular 3 image X-Ray analysis Al functional 3 Al functional 2 Al functional standards IEC TR 60601-4-1 Medical electrical equipment and medical Al functional 4 electrical systems employing a degree of autonomy Al Base standards (ISO/IEC TS 4213, TR 34971,...) After publication also IEC 63450 Testing of Artificial Intelligence / Machine Learning-enabled Medical Devices, Al base standards/ horizontals IEC 63521: MLMD - Performance Evaluation Process 81001-5-1 (...Part 5-1: Security — Activities in the product life cycle) **Point** IEC 62304 (Medical device software – Software life cycle processes) Established medical IEC 62366-1 (Application of usability engineering to medical devices) device base standards/ horizontals ISO 14971 (Application of risk management to medical devices) **AAMI / BSI TS 34971** 3) Software as a medical device ISO 13485 (Medical devices – Quality management systems – Requirements for regulatory purposes) 4) Programmable electrical medical system 5) In vitro diagnostic medical device



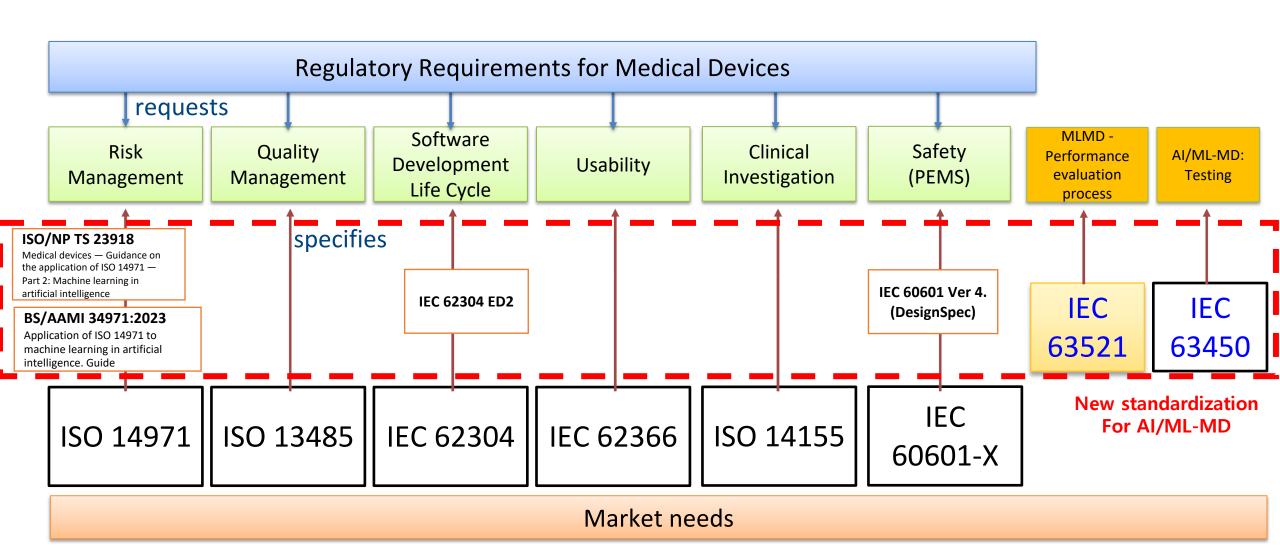
### Align with other standards: point 1





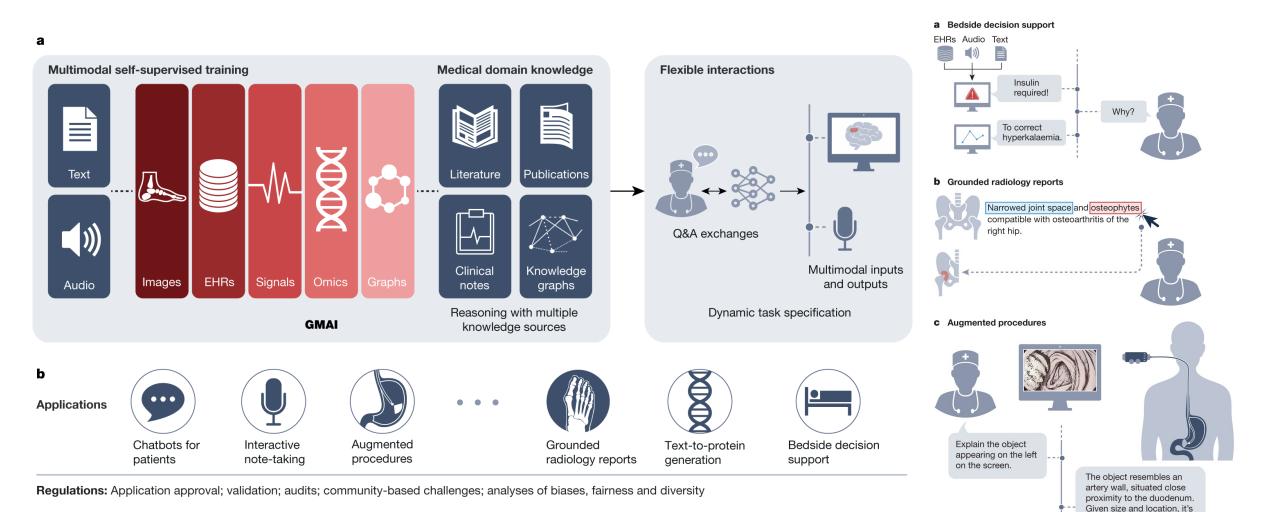


### Considerations 3: Regulation requirements and International standards





### Considerations 4: GenAl, Foundation model, LLM



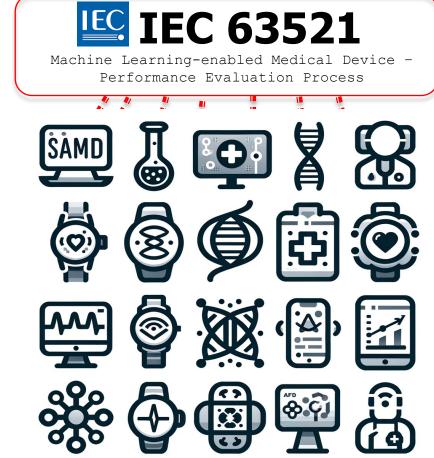


most likely the aorta (aortoduodenal fistula)

### **Conclusions**



- History, Goal, Member, Schedule of PT 63521
- Why Performance Evaluation ?
- Three pillar model of Performance
  - Scientific Validity requirements,
  - <u>Technical (analytical) performance requirements</u>
  - <u>Clinical performance requirements</u>
- Current major discussion topics of PT 63521
  - Safety and Effectiveness of ML
  - TPLC & MLLC & performance evaluation
  - V-model or W-model
  - Align with baseline standards 62304, 13485, 14971, 62366 ···
  - TC 62's AI standardization roadmap & considerations
- Future considerations
  - Global harmonization & AI regulation/law





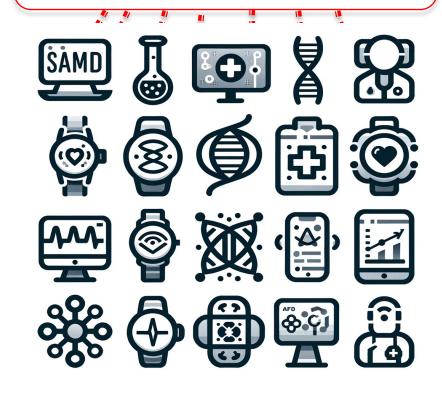
### **Conclusions**



## Please join us



Machine Learning-enabled Medical Device Performance Evaluation Process







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