

J-AISI Business Demonstration WG Data Quality sub-working group

IT Promotion Agency, Japan(IPA)
Digital Infrastructure Center
Teppei Sakamoto

Jan. 15, 2026

Hiroshima Global Forum for Trustworthy AI



Digital Infrastructure Center,
Information-technology Promotion Agency,
Japan(IPA) | Japan AI Safety Institute(J-AISI)

Teppei Sakamoto

2018: NTT DATA Group Corporation

- AI/ML, Data Science, Statistics, etc.
- [XAI (Explainable AI)] (RIC TELECOM, 2021)
co-author,

2025: Researcher, IPA [Current]

Japan AI Safety Institute (J-AISI)

As of Jan. 2026

**AISI Steering
Committee**

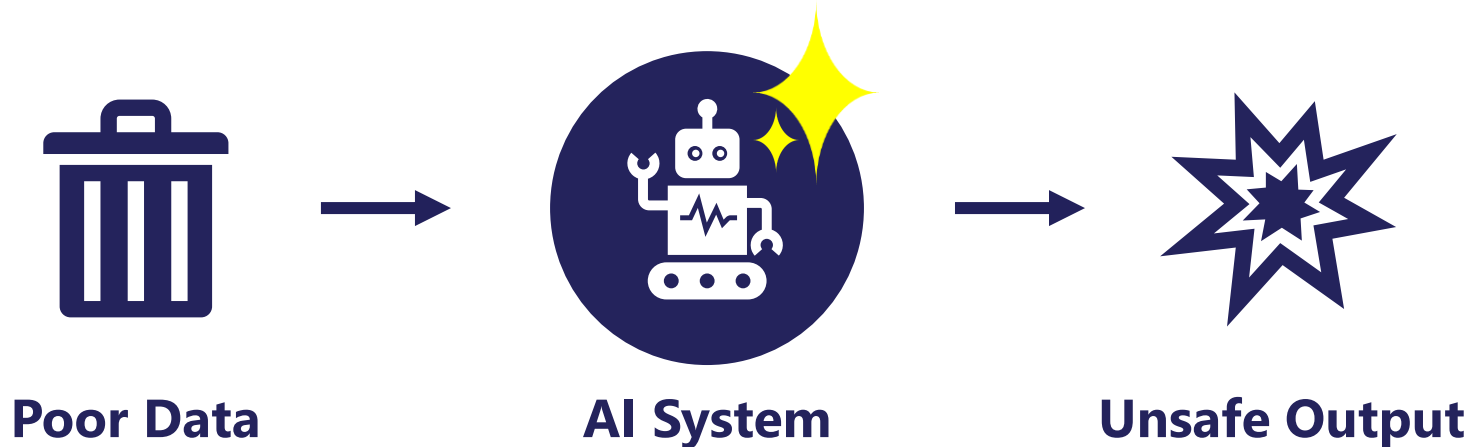
**Business
Demonstration WG**

...

Data Quality SWG

- **Information-technology Promotion Agency, Japan (Leader)**
- Fujitsu Limited
- Kanematsu Corporation
- National Institute of Advanced Industrial Science and Technology
- National Printing Bureau
- NEC Corporation
- NTT DATA Japan Corporation
- NTT DATA Value Engineer Corporation
- Preferred Networks, Inc.
- PricewaterhouseCoopers Japan LLC
- SAS Institute Japan Ltd.
- WingArc1st Inc.

“Garbage in, Garbage out”



- Data is the foundation—models and training depend on it
- Poor data quality leads to unreliable and unsafe AI outputs
- Understanding data quality challenges is the first step

Typical data quality challenges in the AI era— making AI smarter won't fix these.

- ◆ **Outdated or poorly maintained data**

- AI outputs become unreliable when data is not updated or curated.

- ◆ **Lack of data provenance and transparency**

- Users cannot judge whether results are reliable if the origin and processing of data are unclear.

- ◆ **Bias embedded in data**

- Biased data leads to biased AI outputs, even when models are well designed.

- ◆ **Low machine readability**

- Data stored in formats that AI cannot properly interpret causes incorrect or misleading results.

Operational layer

lead

Unsafe
Output

- ◆ **Gap between standards and practice**

- While many data quality standards exist, practitioners struggle to apply them in real projects.

- ◆ **Immature data quality management in the AI era**

- International consensus takes time, but AI deployment—especially generative AI—moves much faster.

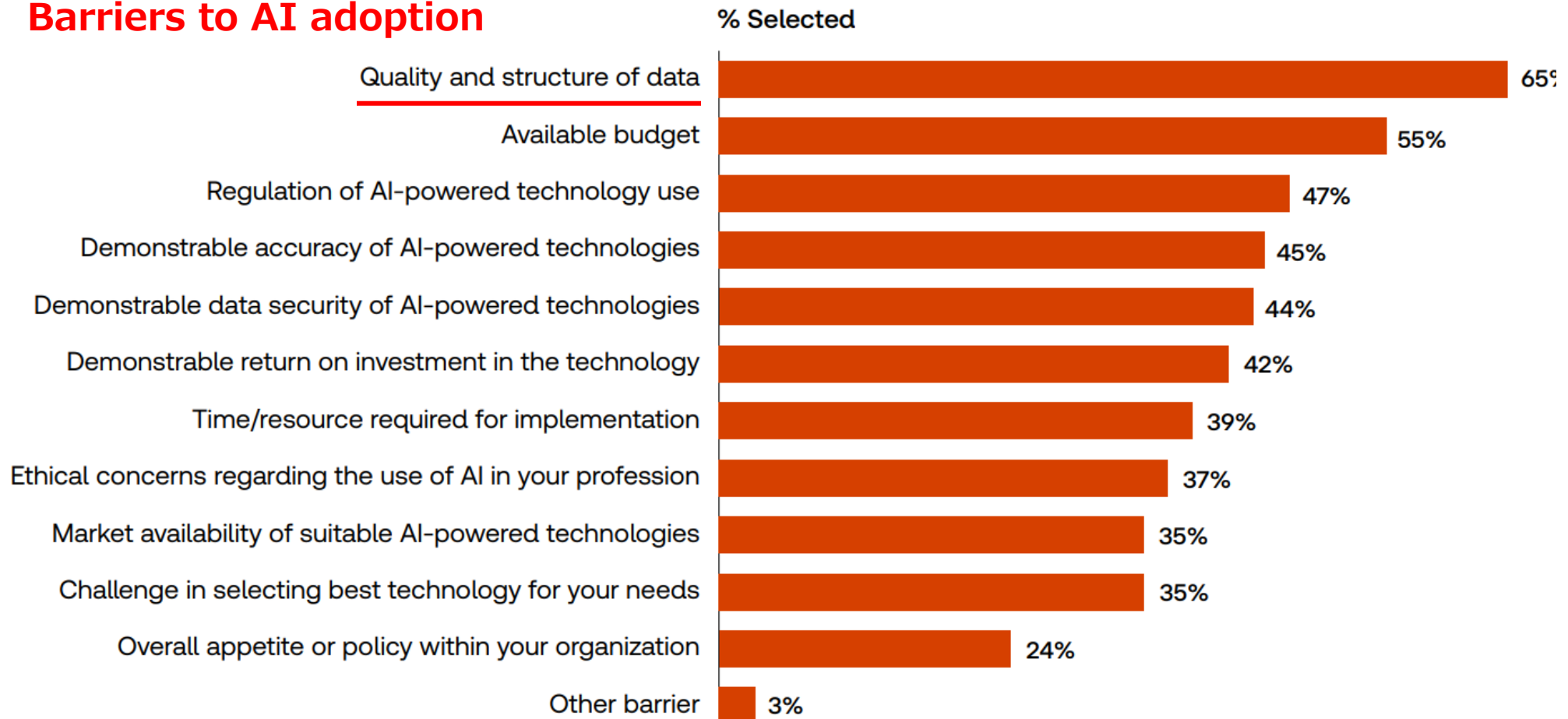
- ◆ **Low organizational prioritization of data quality**

- Compared to AI applications, data quality is often seen as a supporting function and receives insufficient investment

Management & governance layer

Many companies are adopting AI,
but **the most significant challenge is data.**

Barriers to AI adoption



- ♦ Published **AI data quality management** guidebook
- ♦ By viewing data for AI from three perspectives—**Lifecycle**, **Gateway**, and **Governance**—we can reassess data quality from multiple angles.
- ♦ This guide consolidates a wide range of **data-quality standards into a practical, easy-to-apply framework.**

Data Quality Management Guidebook

2025-3-31

AISI Japan AI Safety Institute IPA

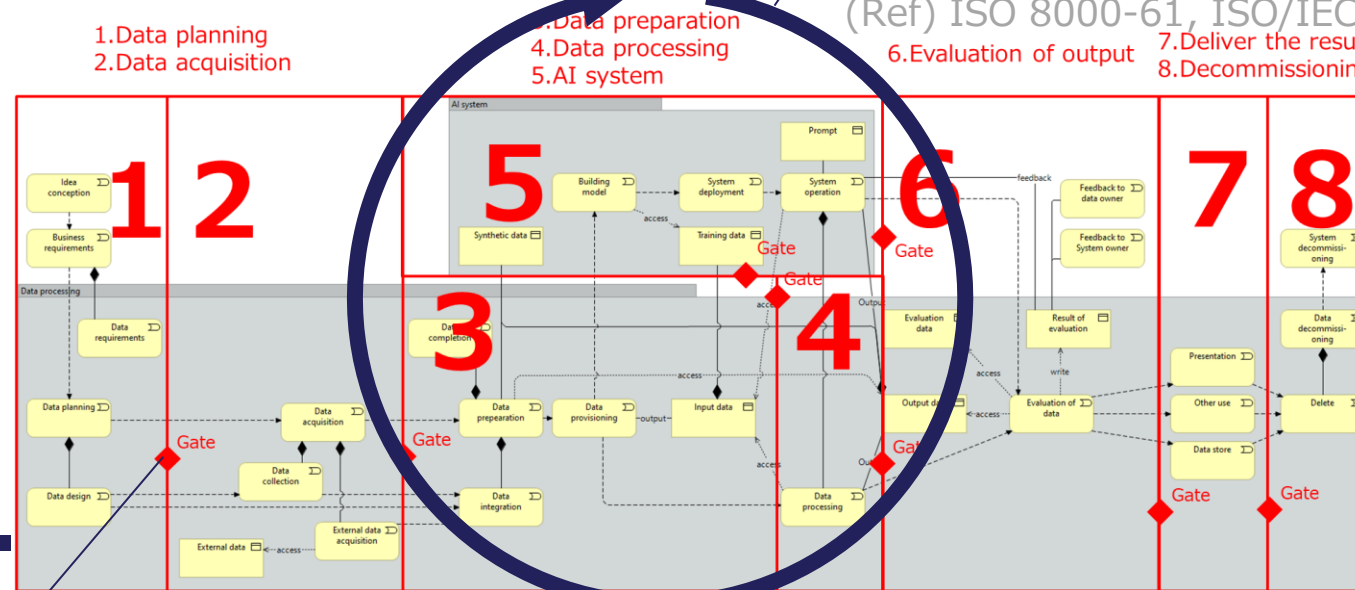
• Lifecycle view

Track data quality from collection to AI deployment and decommissioning
(Ref) ISO/IEC 25024, ISO/IEC 8183, ISO/IEC 5259-1, etc.

Governance cycle view

Ensure the organization has a sustainable structure.

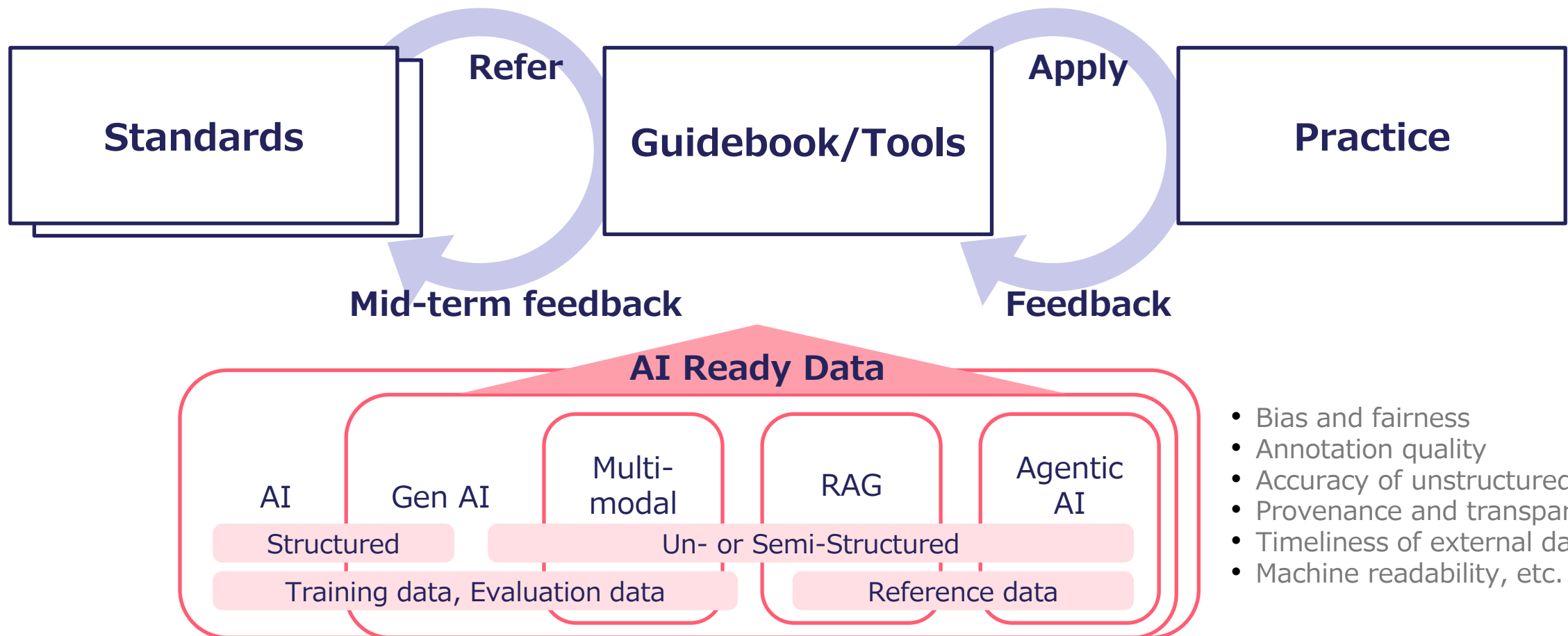
(Ref) ISO 8000-61, ISO/IEC 38505-1, etc.



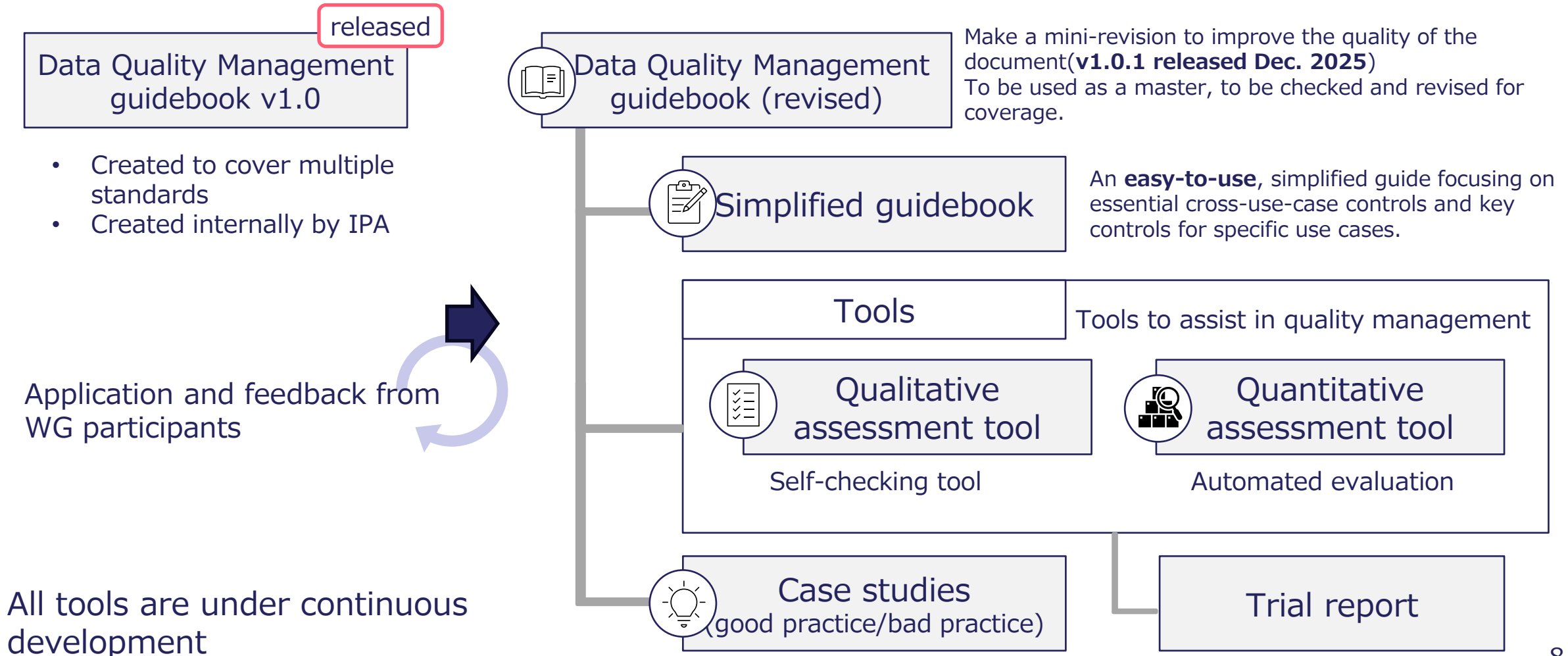
• Gateway view (Characteristics)

Define quality criteria at key decision points
(Ref) ISO/IEC 25012, ISO/IEC 5259, etc.

We complement existing data quality standards with practice-oriented guidance and tools, **bridging standards and real-world practice.**



We are incrementally updating the guidebook and developing supporting tools based on real-world use.



The guidebook was **well received conceptually**, but practitioners found it **difficult to apply directly in daily operations**.

✓ What was positively evaluated

- ♦ **Clear core message**
 - Emphasizes “*Garbage in, Garbage out*,” highlighting that AI success fundamentally depends on data quality.
- ♦ **Alignment with international standards**
 - Grounded in ISO and other international standards, ensuring credibility and legitimacy.
- ♦ **Comprehensive coverage**
 - Provides a broad, high-level view of data quality across governance, processes, and characteristics.

→ What needs improvement

- ♦ **Gap between standards and practice**
 - Still insufficient practical guidance for real-world projects.
- ♦ **Clarity of value and use cases**
 - Need clearer ROI of data quality and concrete, use-case-specific examples.
- ♦ **AI-era specific challenges**
 - Need guidance on data–AI version management and modality-specific issues (text, image, video, etc.).
- ♦ **Document usability**
 - Improve navigation, terminology, and overall readability for non-experts and busy practitioners.

Guidebook

[illegible]

- Data Quality SWG
- Domain-specific SWGs

Feedback

First round of pilot application is currently underway

- 10

Using Japan's Official Gazette as an example of high-value public data, we observe **how data quality levels affect AI response and safety**.

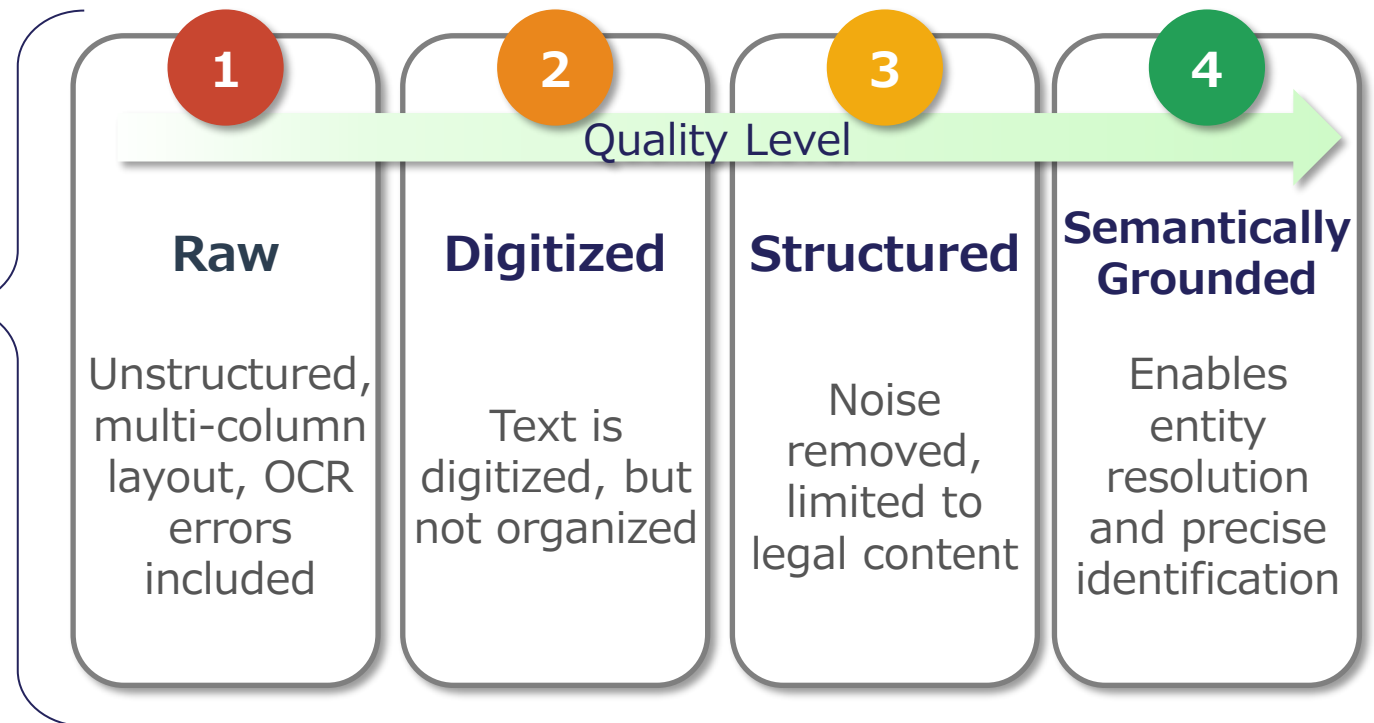


Which laws were amended?

User

AI

AI response



What this examination aims to clarify

- Accuracy and risk of AI responses depending on data quality level (*Effectiveness & Safety*)
- Trade-off between accuracy and preparation cost (*Cost-effectiveness*)

This examination does not validate the effectiveness of the checklist itself. Rather, it illustrates one of the key issues addressed by the checklist— machine readability and data structuring—through a concrete example. This case study is for working-group level discussion.

We aim to contribute **Japan-originated, practice-driven** frameworks to global AI discussions, reflecting **Japan's culture of quality**.

Short-term Initiatives (FY 2025)	Mid-term Initiatives (FY 2026–2027)	Long-term Initiatives (Future Vision)
<ul style="list-style-type: none">• Revise guidebook• Develop and test checklists• Validate applicability through pilot use	<ul style="list-style-type: none">• Enhance assessment tool capabilities• Cross-domain models• Address new data and usage types (e.g., multimodal, multi-agent)	<ul style="list-style-type: none">• Establish continuous data quality models• Update guidebook and tools• Contribute Japanese frameworks to international discussions

- ♦ **Data quality is the foundation of AI safety.**
- ♦ We are developing a **data quality management guidebook**.
- ♦ We are collecting **practitioner feedback** for real-world use.
- ♦ We are building an actionable **checklist** to support practice.
- ♦ We are illustrating the approach through **case study using Official Gazette** data.

How you can engage

- ♦ Access our resource, share your cases and discuss with us.
https://aisi.go.jp/output/output_framework/data_quality_management_guidebook/

Thank you!

AISI

Japan AI Safety Institute