



Technical challenges in the effective safety analysis

Global Aviation Data Management (GADM), IATA
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To represent, lead and serve the airline industry



Introduction to GADM

➤ Global Aviation Data Management (GADM)

GADM is a data management platform integrating several sources of operational data received from different IATA programs.

To provide the industry with comprehensive, cross-database analysis and to support a proactive data-driven approach for advanced trend analysis and predictive risk mitigation.

Accidents



FDX



GDDB



STEADES

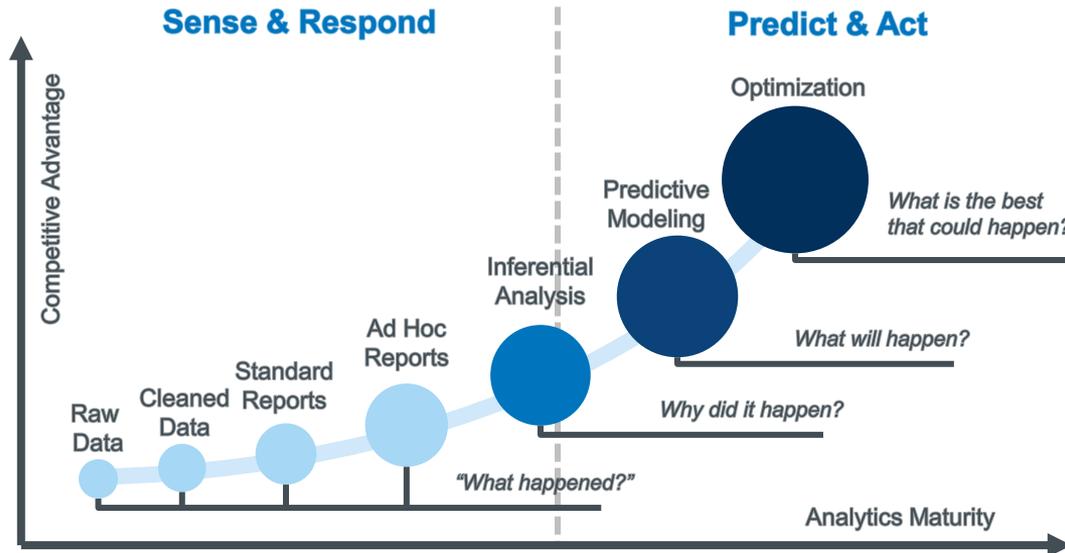


- More than 470 organizations
- 15 + years of successful delivery
- Multiple databases for comprehensive analysis:
 - Accident / Incident reports
 - Ground damage reports
 - Flight data

Contents

- **Background**
- **Challenge**
- **Approach**
- **Next Step**

Evolution of Safety Analysis



➤ Descriptive Analysis

"What is happening?"

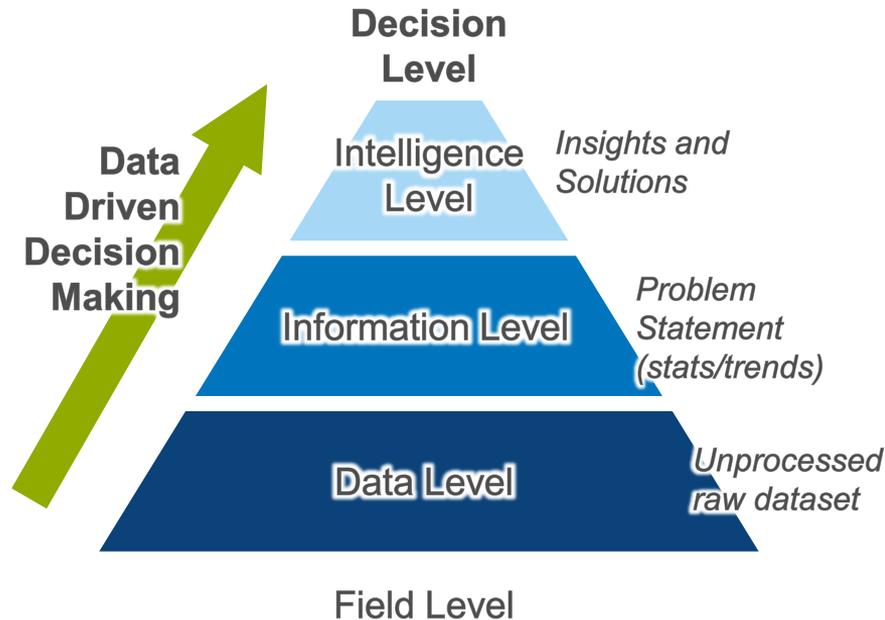
➤ Inferential Analysis

"What is the possible explanation for what is happening now?"

➤ Predictive Analysis

"How can future decisions be made (and how can finite resources be allocated) based upon what is happening now?"

Data-Driven Decision Making

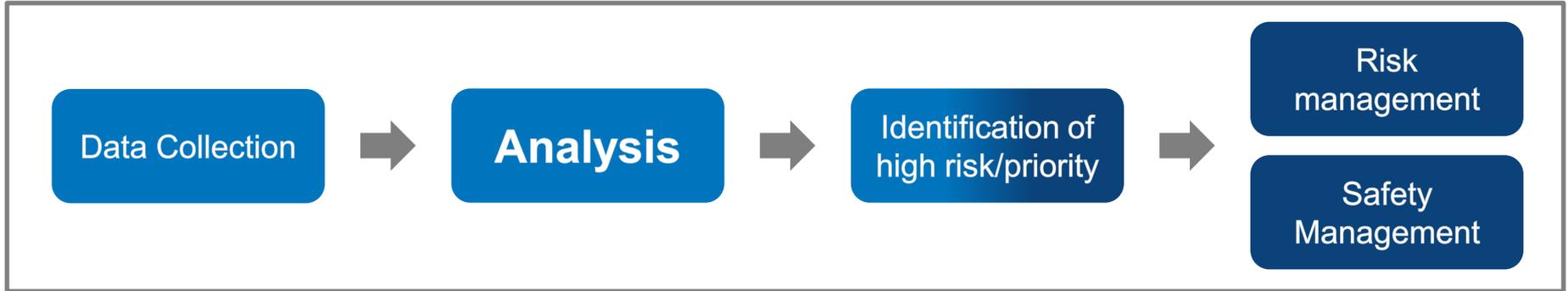


- **Data-driven decision making:** Decisions with the data and quantifiable evidences, rather than intuition or based on observation or experience alone.
- including the most effective and efficient allocation of resources.

Using safety data / information to identify and to mitigate safety risks

Data Driven Decision Making (D3M)

Source: ICAO Doc.9859 Safety Management Manual, 4th edition (2018)



Safety Intelligence

- Leveraging safety data and information to develop **actionable insights**.
More than just stacking lots of data!
- A huge amount of safety data, which, when correctly analyzed, can be transformed in what we call “**Safety Intelligence**”, for example, safety intelligence may be able to identify **precursors** to better address and monitor the safety risks.



- **Finding data in the aviation system is easy, but effectively building knowledge with it is difficult.**

Good Analysis: “Right Data at the Right Time.”

➤ Large, complex and various kinds of data

- **Quantitative Data:** Flight parameters (FDA/FOQA), Radar or ADS-B tracks – *sensor captured*
- **Qualitative Data:** Accident, incident and risk reports, Audits – *written in human language*

➤ Human-Machine Interoperability

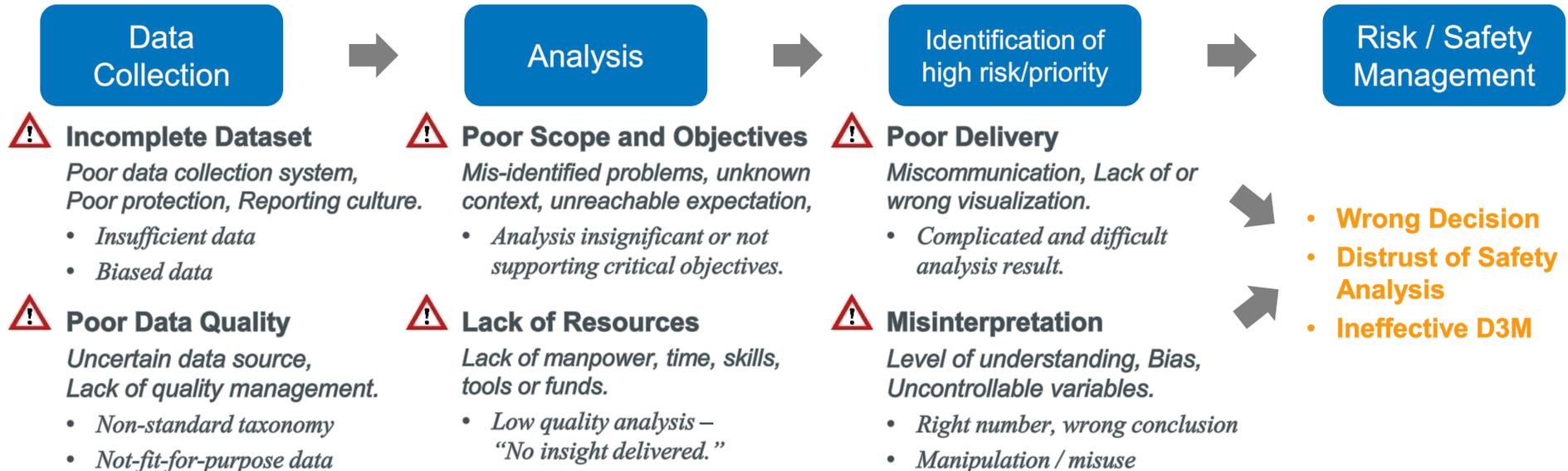
- The data is for the machine, but the analysis is for human.
- Analysis delivery, as well as understanding the context of the analysis, is important to avoid misinterpretation.

➤ Limited Resources

- Clear objectives, scopes, priorities and effective resource allocation, with appropriate competencies, are a must.

Challenges in Safety Analysis

➤ All kinds of reasons how and why safety analysis can be misconducted:



Data Availability: “Can we get the full picture?”

➤ Data is like nutrition for safety analysis: it needs to be sufficient and balanced.



Accident, Mandatory and Voluntary Reports

Undesired Aircraft State,
Operational deviation,
Graded condition

Latent precursors,
Threats and errors

Blind spots,
Unobservables

Conditions that we don't even know.

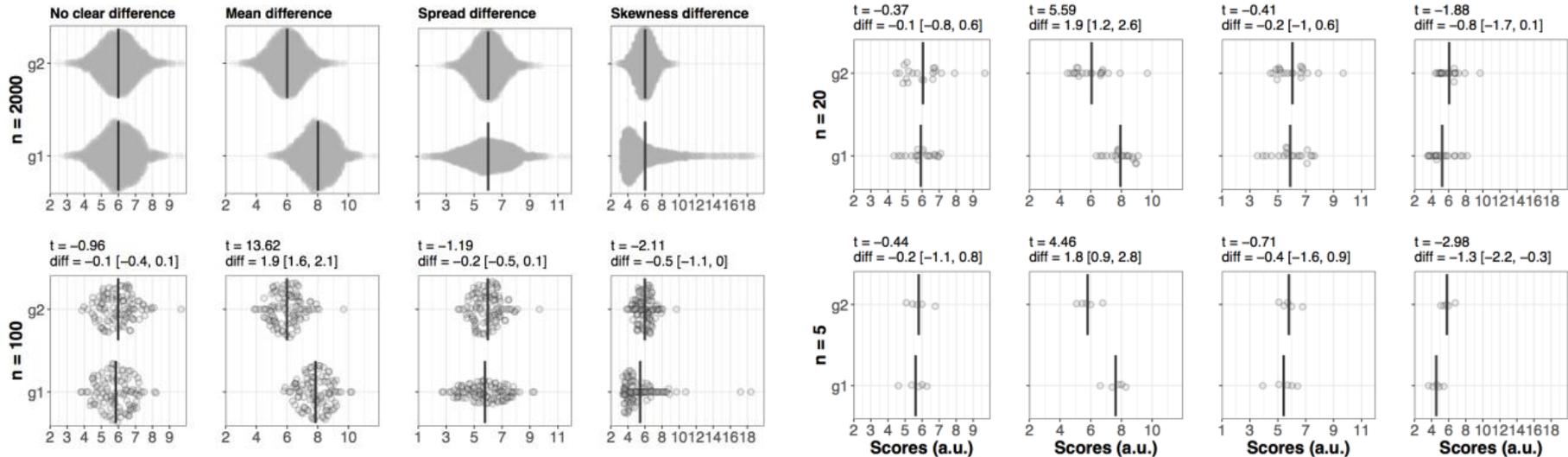
	Data Availability	Data Volume	Processing Cost (Time)	Collected Format
↑ Visible	High (accessible)	Limited	Cheap	Structured
↓ Obscured	Low (difficult)	Large and Complex	Expensive (exponential)	Unstructured

➤ **We will not have data for every known parameter for normal operations.**

- **Insufficient data** – lower confidence level
- **Biased data** – biased conclusion

Sample Size and Uncertainty

➤ Analysis with insufficient sample size has lower confidence level. (Weak Conclusion)



Data Integrity for Data Quality

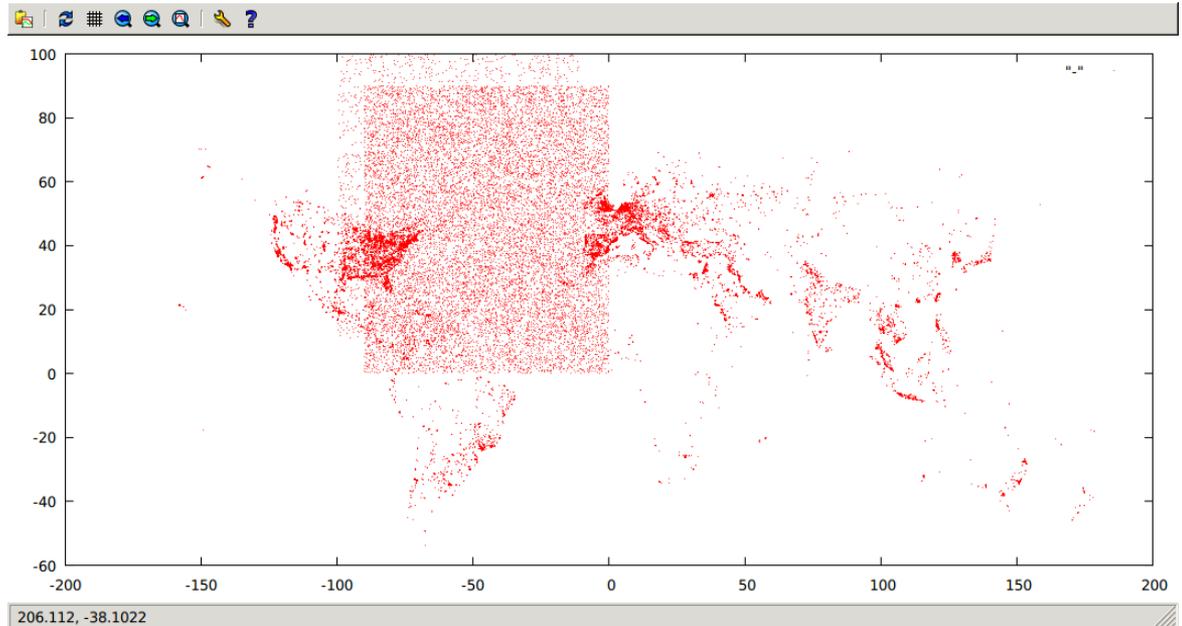
James Lloyd ▶ Skyscanner ✓
August 23 at 2:14pm · ©

Hi Skyscanner. Just wondering what you'd recommend I do during the 47 year layover your website has suggested?

RETURN - Wed, 15 Mar 2017

Emirates	16:55	CHC Christchurch	3h 20
EK419	18:15	SYD Sydney Kingsford Smith	
Connect in airport			
Emirates		SYD Sydney Kingsford Smith	
EK419		BKK Bangkok Suvarnabhumi	
Connect in airport			
Long wait			
413786h 25			
Emirates	02:25 (+1)	BKK Bangkok Suvarnabhumi	6h 40
EK419	06:05	DXB Dubai	
Connect in airport			
Long wait			
8h 30			
Emirates	14:35	DXB Dubai	7h 50
EK9	18:25	LGW London Gatwick	

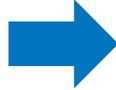
Arrives: Thu, 16 Mar 2017 | Journey duration: 38h 30



Data Integrity for Data Quality

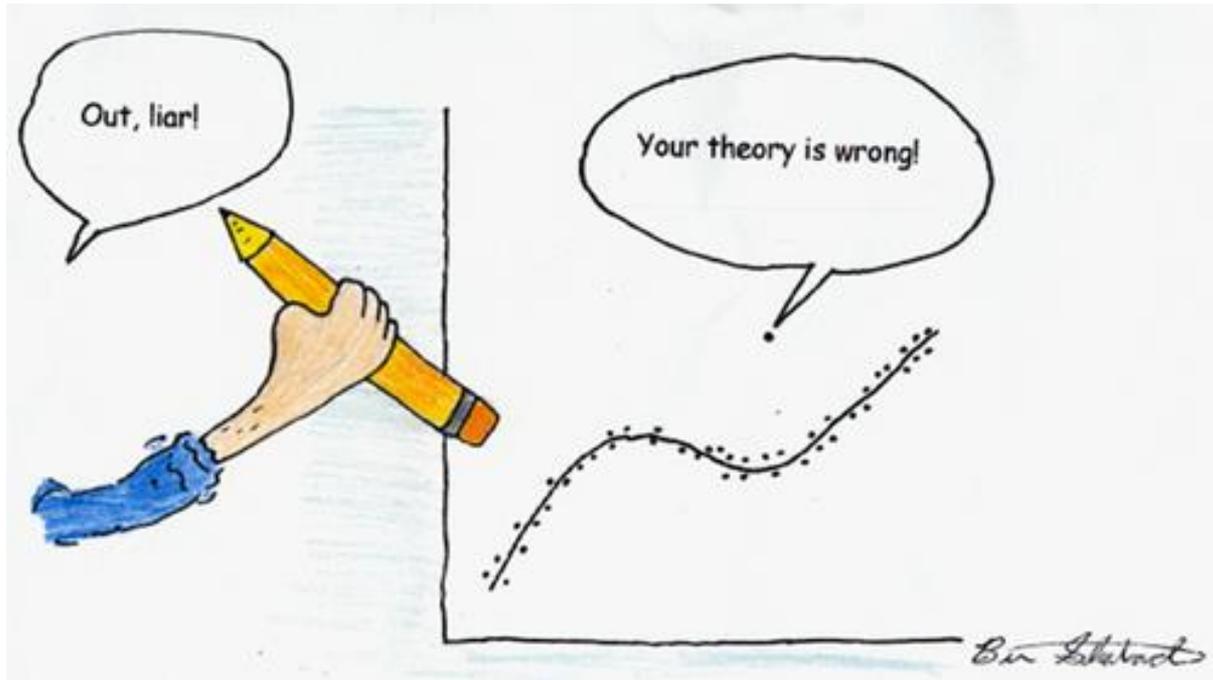
➤ **Data Integrity:** the reliability of the information in terms of its physical and logical validity.

- *Completeness*
- *Uniqueness*
- *Validity*
- *Accuracy*
- *Consistency*



➤ **Data Quality:** the reliability of information to serve an intended purpose including planning, decision making and operations.

“Nearly one third of analysts spend more than 40 percent of their time vetting and validating their analytics data before it can be used for strategic decision-making”



➤ **Multiple poor data points can ruin the whole findings!**

Data Heterogeneity

- Heterogeneous data are any data with high variability of data types and formats.
 - **Syntactic:** two data sources are not expressed in the same language.
 - **Semantic:** the differences in modelling the same domain of interest. (*logical mismatch*)
 - **Terminological:** variations in names when referring to the same entities
 - **Semiotic:** different interpretation of entities by people.

- Heterogeneous data in safety analysis:
 - *Merging global and multi-source data*
 - *Integrating quantitative data (flight data) with qualitative data (written report)*
 - *Especially data written in natural language form (basically all of the safety reports).*

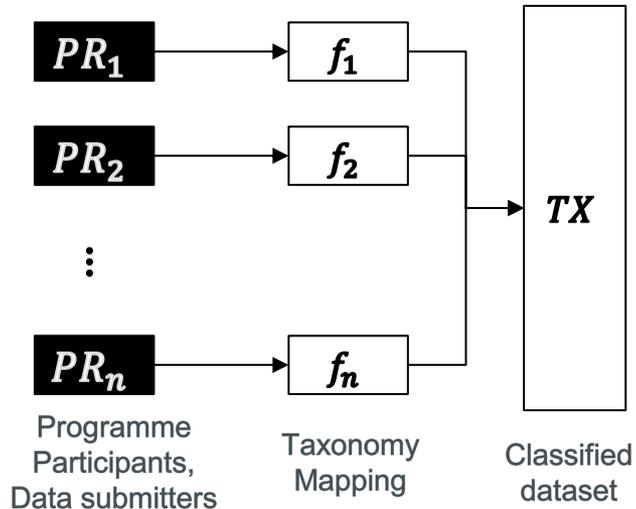
- **Only limited parts of the dataset would be remained as valuable.**

Challenge – Sample size and heterogeneous data

- For aviation safety analysis, high data integrity is required but there is only limited data available.
- ICAO world accident rate: 2.42 per 1 Million departures **(0.000242%)**
Scheduled Commercial flights on airplanes above 5.7 tons only, Source: iSTARS Accident Statistics
- Null Hypothesis: “*there was no accident in 2017*”, ratio of “*correct*” prediction: 99.999758%
Is 99.999 of accuracy is what we can satisfy?
- Things are getting complicated:
 - Is the data point an outlier to carry on, or a rarely observed but critical value?
 - How can we determine that the data point was occurred just by chance or not?
 - How can we clean poor data, if we are not sure about its reliability?
 - How do we maintain the data quality and confidence level if the sample size is small or the sample is extremely skewed?

Approach: Taxonomy

- Defining a structure of classifications of raw data with human supervision

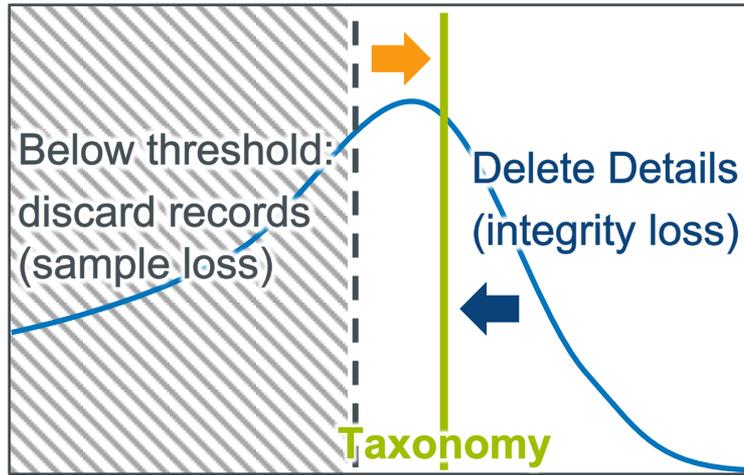


- Can be referred as “data standardization” to achieve, track and manage data quality.
consistent, complete and unambiguous
- Centralized rules and procedures from expert’s knowledge.
- However, human intervention makes data vulnerable to human errors, thus additional quality check and validation process are required.

Taxonomy and Data Standardization

reports

Cleaning Process (Cost)



Lesser Granularity

Greater Granularity

No details

High level description

Low level description

➤ Granularity

How detail the valid information each record contains.

➤ Cleaning Cost

Bringing records with less granularities can be fitted into taxonomy after investing resources.

➤ Quality Threshold

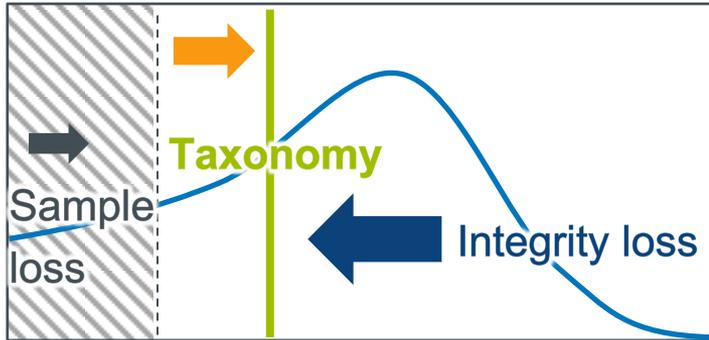
The minimum threshold of granularity for records required to be fitted into taxonomy.

➤ Sample Loss

Records below the threshold cannot be fitted into taxonomy, and thus, such records are deleted for consistency (data quality)

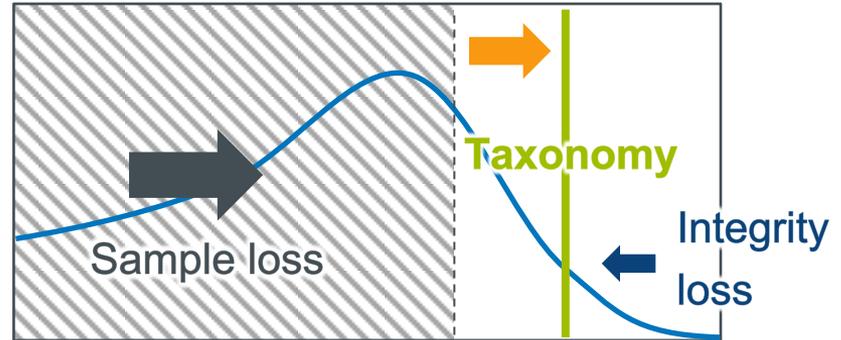
➤ Integrity Loss

Taxonomy cannot hold greater granularity of some records, and thus, such records lost some granularity by being fitted into taxonomy



➤ **Taxonomy in too much high level**

- Less data sample loss
More data integrity loss
- Safety analysis might require greater granularity than taxonomy. (Additional data processing needed).



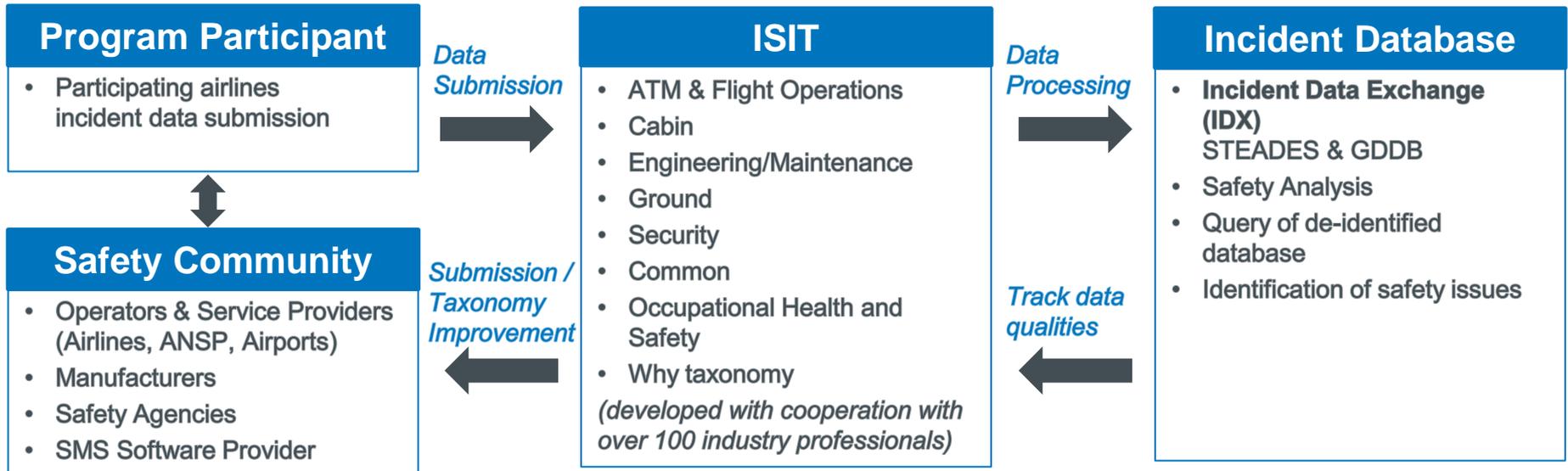
➤ **Taxonomy in too much low level**

- More data sample loss
Less data integrity loss
- Higher processing cost is be needed to keep more sample, but sample loss is inevitable. (Analysis will has lower certainty)

- **Finding the right balance is important: data sample, integrity and resources.**
- **The taxonomy shall keep up with changing industry.**

IATA Safety Incident Taxonomy (ISIT)

➤ Better integrate the data and have a common language among aviation safety communities



Analysis Methodology

➤ Process of building insights from processed dataset

Effective Analysis using right methodologies

Data Level

“What happened”

- Processed data (taxonomy classified data) for analyst to begin safety analysis.

Information Level

“What is going on”

- Using statistics and indicators to show what is going on. (trends)

Intelligence Level

“Why and How did it happened”

- Analyst diagnose data in-depth to identify contributing elements below the surface. (e.g. threat and errors)

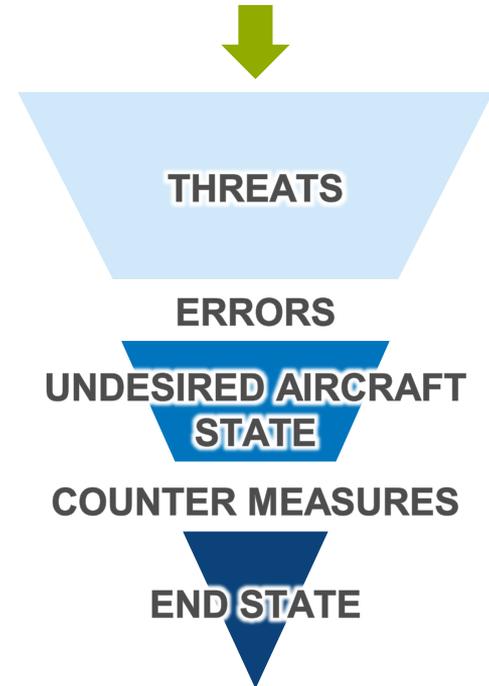
Analysis Methodology

➤ Threat Error Model (TEM)

Threat and Error Model enhances the classification system used by IATA to determine contributing factors in incidents and accidents.

➤ Subject Matter Expert (SME) Interview

SME can provide information or hints to analyst which might be invisible in dataset even that they might not even be conscious of knowing them.



Analysis Methodology

➤ Safety Performance Indicator (SPI)

<u>Period of Review [2018-Q2]</u>	IATA	Non IATA
Rate (Accident per 1 Million Sectors)	0.66	1.96
Number of flights per accident	1,505,636	510,953
Number of Accidents	4	10
Fatalities	0	113
% of sectors flown	54%	46%
% of passengers (Total ASM)	82%	18%

➤ Risk Matrix

Safety Risk		Severity				
		Catastrophic A	Hazardous B	Major C	Minor D	Negligible E
Frequent	5	5A	5B	5C	5D	5E
Occasional	4	4A	4B	4C	4D	4E
Remote	3	3A	3B	3C	3D	3E
Improbable	2	2A	2B	2C	2D	2E
Extremely improbable	1	1A	1B	1C	1D	1E

Note.— In determining the safety risk tolerability, the quality and reliability of the data used for the hazard identification and safety risk probability should be taken into consideration.

➤ Provides strengths that offset the weaknesses of both quantitative and qualitative research, and thus, more complete and comprehensive context.

Visualization for Analysis Delivery

specific measures available about the relative importance of safety, cost, the environment, performance, and convenience for a vehicle fleet or power supply for a factory or business. Questions were asked to assess visitors' views about the likelihood of various future applications of hydrogen technology. The results of the questions, "I don't know" or "I have no opinion" were perfectly acceptable answers. Overall, no direct comparisons between visitors, brands, government, etc. and media (radio, internet, newspaper, etc.) were possible as users have energy technology information is received.

At various stages in their development the survey questionnaires were reviewed by National Hydrogen Association and U.S. Dept of Energy personnel and by management at the DOE EOT office. Final Report before was published, and Office of Management and Budget approval to conduct the survey was obtained, per the Paperwork Reduction Act of 1995.

The general public and selected survey samples were selected by random digit dialing. Potential large-scale and more aware selected by random sampling. Surveys of cost and best government officials and safety and codes officials were of their more target populations (i.e., users for microsystems, they are somewhat targeted). All five surveys were administered by computer-assisted telephone interviewing (CATI). The general public and media surveys were administered in either English or Spanish, at the option of the respondent. For all populations except the safety and codes officials, the length of the survey was less than 15 minutes, including the introduction, opening process, and general information and demographic questions. The average interview length of safety and codes officials was 17 minutes.

Limitations. The biggest data quality limitation of the hydrogen survey data is nonresponse bias. Table E5.1 shows 2004 and 2006-2009 response rates (percentage) by survey component population. Response rates were for the most part very slightly lower for the 2006-2009 surveys. This is somewhat mild, tends toward self-selection, bias, and increasing use of other IT.

Population	2004	2006-2009
General public	74.2%	74.2%
Business	77.2%	76.7%
Environmental agencies	74.2%	76.8%
Cost/price	76.2%	77.0%
Safety and codes officials	82.4%	77.7%

In 2006, the position of safety and codes officials was changed.

We are willing to accept nonresponse bias because we believe that it is not severe enough to invalidate the survey and because all reasonable measures were taken to minimize it. Careful and aggressive callbacks, adjustments to sampling weights, and other efforts to reduce nonresponse bias will not fully address the challenge of nonresponse bias. We also expect that changes in response rates will not obscure measurements of change in knowledge, awareness of, and attitudes toward hydrogen. An issue involving telephone surveys is the possibility of undercoverage because of cell phone-only households. While sampling weights provide a partial correction, it was not feasible to fully address the cell phone-only issue in the 2006-2009 hydrogen surveys.

Strengths. The data analysis focuses on the main data elements and survey objectives. Answers to the technical questions are compiled into technical knowledge scores. Opinions about safety,

Hydrogen Knowledge and Opinions Survey 100 April 27, 2010



➤ Clear Understanding

➤ Enables Story Telling

➤ Stronger Delivery



Dashboard: Self-Service Analysis

➤ User Customized – Interactive Benchmark and Query Tool

- Users can perform taxonomy-based analysis easily, with their objectives and intentions.
- Basic analysis by end-users: better information accessibility with less analysis cost.
- How do we prevent misinterpretation?
User Experience Design



Effective Use of Safety Intelligence

- **Confidence by decision makers** in the accuracy of a safety analysis is the key element for its effective use.
If decision makers do not believe in the analysis, for any reason, they will not use it.
- It is critical to assure that the analysis result is clear to decision makers, and that there is no chance for **misinterpretation**.
- These risks can be mitigated by involving end-users (analysis customer and/or decision makers) from the beginning to the end of the analysis cycle.
 - Inception Interview: to better identify what end-users want.
 - Validation: to ensure that the analysis progress is aligned with the original objective and scope of users.
 - Customer Feedback: to improve the analysis cycle and thus, better reliability of the deliverables.

Quality Management System



➤ ISO 9001:2015 implemented with PDCA Cycle

- **Plan:** Design or revise business process components to improve results
- **Do:** Implement the plan and measure its performance
- **Check:** Assess the measurements and report the results to decision makers
- **Act:** Decide on changes needed to improve the process

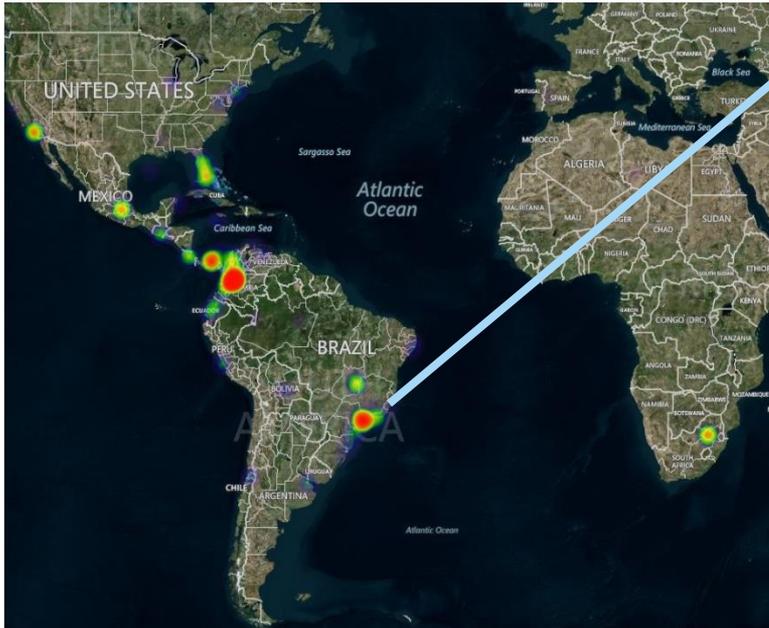
Proactive management to achieve high standards of quality, clarity and reliability of deliverables, so that it can understand and fulfill the needs of our customers.

Process Capability & Maturity

Continuous process improvement (incremental and innovation) Common causes of variation are identified and improved Processes are agile and "best in class"		Optimising
Targets, standards and measures are used Special causes of variation are identified and corrected	Quantitatively Managed	Performance is predictable
Organisation-wide focus Value chains are identified	Defined	All processes are documented Measurements are defined
Departmental & Team focus	Managed	Some organised processes Performance is repeatable
Initial	No organised processes Ad hoc and reliant on "heroics" Performance is not repeatable	

Success Analysis Story

Sao Paulo TMA



- IATA + airlines presented to Brazilian authority (RAISING AWARENESS)
- Analysis of TCAS points and SIDs/STARs
- SIDs/STARs modified
- Key point: issue known previously by individual airlines, but aggregate data drove change

Another Problem: Overwhelming Data

- Big Data Era
- Data volume beyond human capabilities
(especially where human intervention is required: taxonomy classification and validation process)
- Proposed solution: data-driven automation tools

Next Step: Automation and A.I.

- Automation – replace repetitive works
- Intelligence Augmentation (IA) – support decision-making without calculating everything!
- A.I. experiment in GADM

<https://projector.tensorflow.org/?config=https://raw.githubusercontent.com/HTjung/embeddings/master/config.json>



GADM
GLOBAL AVIATION DATA MANAGEMENT

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To represent, lead and serve the airline industry

