



The Urgency of Action: Focused, Aligned and Ready

SIGNAL Innovation Showcase

# Data Enrichment using LLM Text Analytics

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- Chief Data Scientist at Hitachi Vantara Federal (HVF)
  - Explores the "Art to the Science" of solution architectures orchestrating data, APIs, algorithms and applications enabling operational data fabric and data management solutions
  - Focus on Federal Government domain data for ~20 years
  - AFCEA Member since 2018
- Ph.D. in Computational Sciences and Informatics, GMU
  - Machine Learning Galaxy Redshifts using Color and Morphology
  - Dataset : Sloan Digital Sky Survey (SDSS)
- Enjoys cooking, travel and photography

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Text Analytics is a branch of machine learning (ML) combined with natural language processing (NLP) which extracts key characteristics from unstructured text documents. The additional processing cycles can improve trust for data assets by expanding their accessibility. Improved search of the content through correlation statistics, sentiment analysis, entity recognition and summarization can be easily automated.

Using Large Language Models (LLMs) such as OpenAI's ChatGPT, this talk will explore the potential data engineering operationalization and its impact on the associated metadata management.

**Keywords** LLM, ChatGPT, Text Analytics, Data Engineering, Unstructured Data, Sentiment Analysis, Summarization, Metadata Management, Feature Engineering, Hitachi Lumada DataOps

<https://events.afcea.org/AFCEACyber23/Public/SessionDetails.aspx?FromPage=Sessions.aspx&SessionID=10230&SessionDateID=702>



## Protection

Safeguard and protect your most valuable asset – data.



## Recovery

**Detect anomalies** and recover more quickly to keep mission continuity a priority.



## Insight

**Build intelligence** across your entire data ecosystem to mitigate risks, control governance, and improve data quality and trust.

- Decision intelligence multiplier from content, structure and patterns of textual unstructured/semi-structured data and metadata
- Leverages natural language processing (NLP), machine learning (ML), and computational linguistics to process and analyze text data at scale

- ✓ Automated extraction of information
- ✓ Identification of key concepts
- ✓ Sentiment analysis
- ✓ Topic modeling
- ✓ Entity recognition
- ✓ Other linguistic and statistical analysis

Unstructured data



Structured insights



Hitachi Vantara Federal @HVFederal · Apr 28  
Join us next week at [#TechNetCyber](#) in Baltimore, MD! Stop by Booth 2419 and get a glimpse into our framework for [#cyberresiliency](#) and [#dataprotection](#). Learn more: [hitachivantarafederal.com/about/events/t](https://hitachivantarafederal.com/about/events/t)



1. [arXiv:2304.13841](#) [pdf] [CS.AI](#) [CS.CY](#)

AI-based Predictive Analytic Approaches for safeguarding the Future of Electric/Hybrid Vehicles

Authors: Ishan Shivansh Bangroo

**Abstract:** ...and longevity. AI may improve EHV energy efficiency, emissions reduction, and sustainability. Remote hijacking, security breaches, and unauthorized access are EHV **cybersecurity** vulnerabilities addressed in the article. AI research and development may help make transportation more sustainable, as may optimizing EHVs and charging infrastructure. [View More](#)

Submitted 26 April, 2023; originally announced April 2023.

Comments: 7 pages, 3 figures, 1 table

2. [arXiv:2304.13505](#) [pdf] [CS.NI](#)

Techno-Economic Assessment in Communications: New Challenges

Authors: Carlos Bendicho, Daniel Bendicho

**Abstract:** ...of compliance for several use cases: 5G and beyond, software-defined wide area network (SD-WAN), secure access service edge (SASE), secure service edge (SSE), and cloud **cybersecurity** risk assessment. The authors also present TEA extensibility to request for proposals (RFP) processes and other industries, to conclude that there is an urgent need for agile and... [View More](#)

Submitted 5 January, 2023; originally announced April 2023.

Comments: 18 pages, 1 figure, 2 tables

ACM Class: C.2.0; C.2.1; C.2.3; C.2.6; C.2.m

3. [arXiv:2304.11960](#) [pdf, other] [CS.CR](#) [CS.CL](#) [CS.LG](#)

ThreatCrawl: A BERT-based Focused Crawler for the **Cybersecurity** Domain

Authors: Philipp Kuehn, Mike Schmidt, Markus Bayer, Christian Reuter

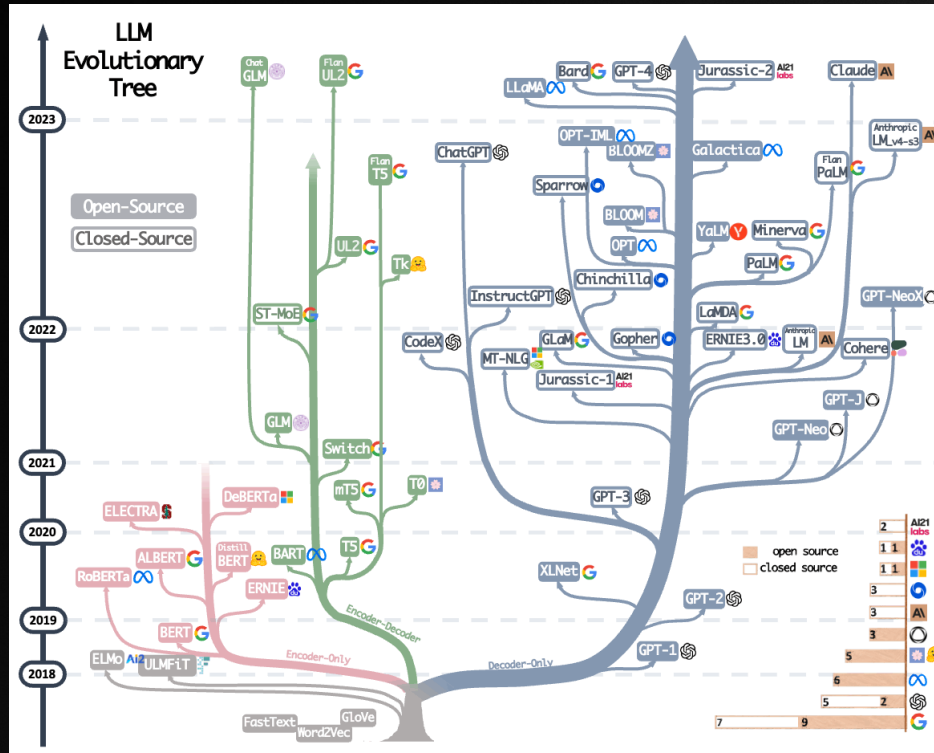
**Abstract:** Publicly available information contains valuable information for Cyber Threat Intelligence (CTI). This can be used to prevent attacks that have already taken place on other systems. Ideally, only the initial attack succeeds and all subsequent ones are detected and stopped. But while there are different standards to exchange this information, a lot of it is shared in articles or blog posts in non-... [View More](#)

Submitted 26 April, 2023; v1 submitted 24 April, 2023; originally announced April 2023.

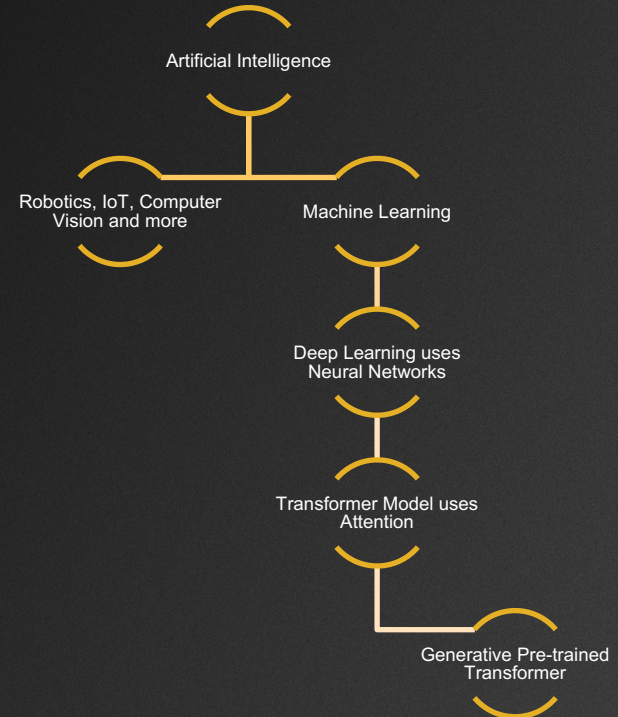
Comments: 11 pages, 9 figures, 5 tables



# Large Language Models (LLM)



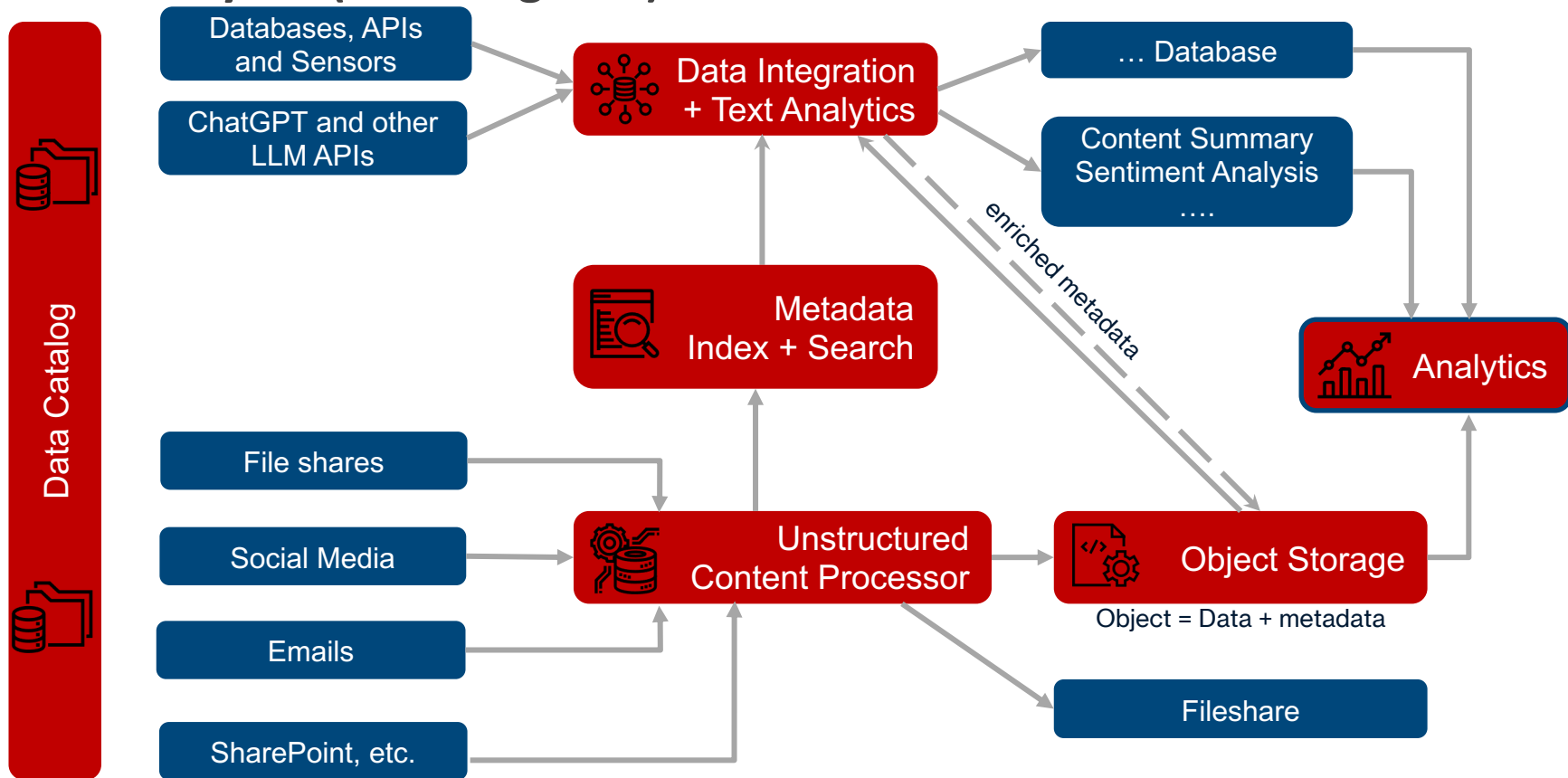
<https://github.com/Mooler0410/LLMsPracticalGuide>



# Text Analytics (including LLM) Dataflow

Master Data Management : Single source of "Truth"

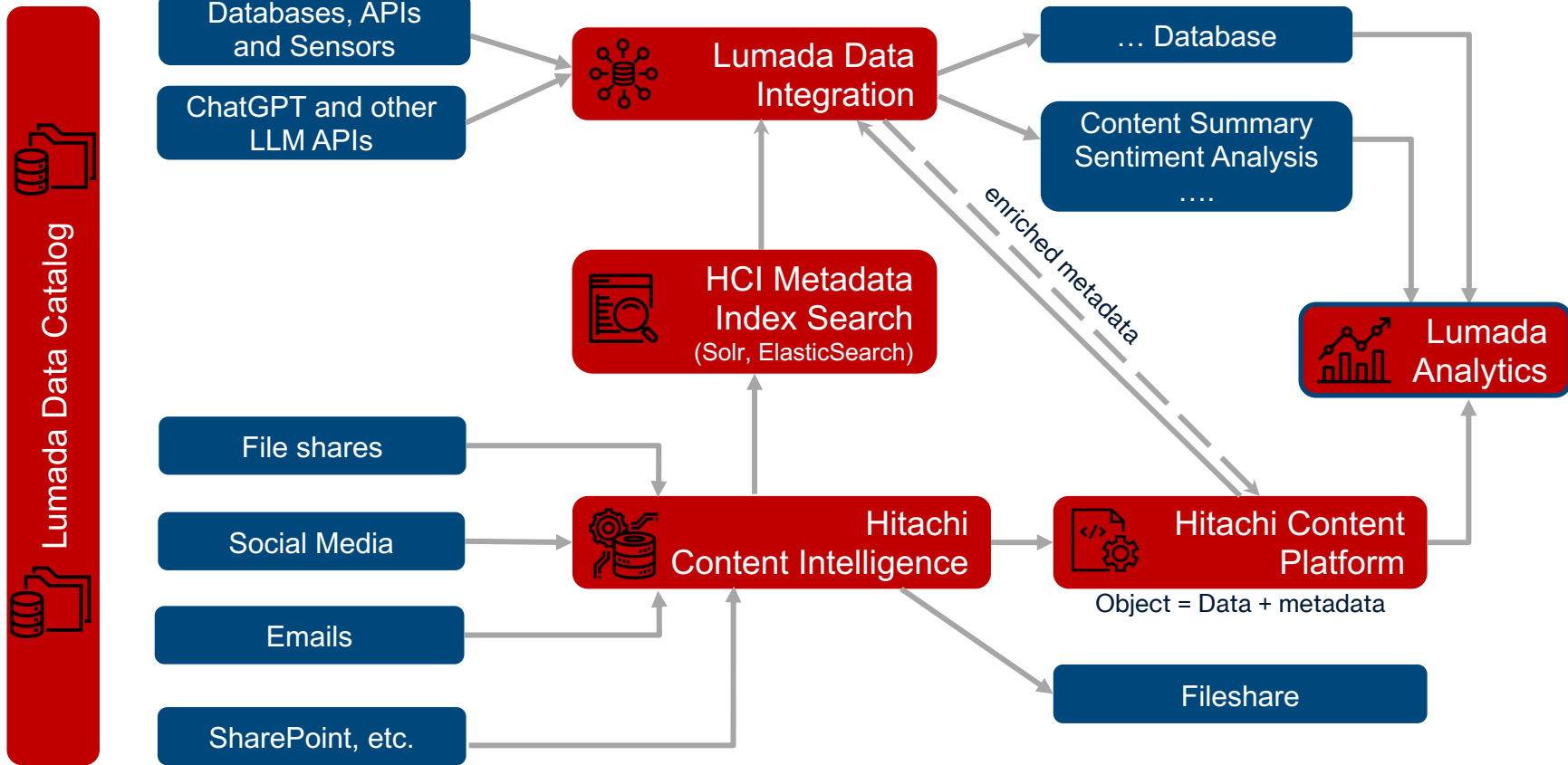
Data Lineage and Provenance



# Text Analytics (including LLM) Dataflow with Hitachi

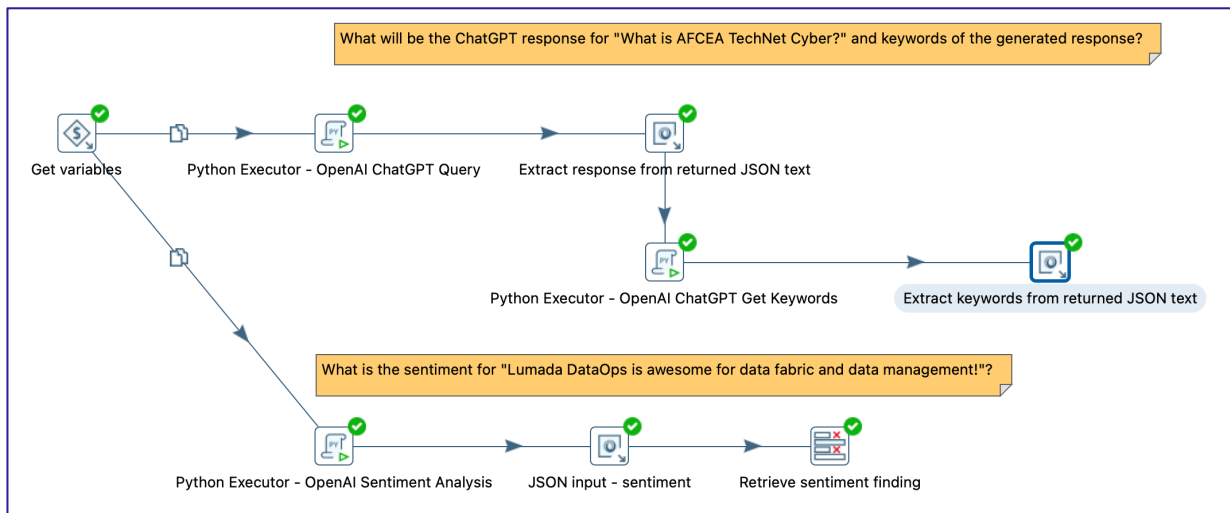
Master Data Management : Single source of "Truth"

Data Lineage and Provenance





# Lumada Data Integration Text Analytics



## Keywords:

1. Cyber Security
2. Information Assurance
3. Emerging Technology
4. Senior Government
5. Industry Leaders

Metadata / Tags /  
Business Terms

**Prompt:** What is AFCEA TechNet Cyber?

**Response:** AFCEA TechNet Cyber is an annual event hosted by the Armed Forces Communications and Electronics Association (AFCEA) that brings together experts in cyber security, information assurance, and emerging technology fields to discuss the latest developments in the field and to develop ways to support mission success. The event typically includes seminars and interactive workshops, as well as keynote presentations from senior government officials and industry leaders.


Freshness of information is limited to the knowledge that was available at the time the LLM was trained (e.g., September 2021 for GPT-4)

# LLM Example App - Synthetic Data Generation

- **Real-world data** is collected from various sources such as surveys, observations, sensors, and other data collection methods. This data reflects the true characteristics of the phenomenon being measured or observed.
- **Synthetic data** is generated based on a set of assumptions, models, or rules that define the patterns and characteristics of the data.
- Contrast with **Fake data** which is intentionally falsified or manipulated data, with the intention of deceiving or misleading the audience. It may be created with the intention of perpetrating fraud, spreading disinformation or misleading individuals or organizations.
- **Pros**
  - ✓ Data generated at scale and is not limited by the availability of real-world data.
  - ✓ Valuable resource for machine learning, artificial intelligence, and data analytics applications
- **Cons**
  - May not fully replicate the complexities and nuances of real-world data

PR

I want you to act as a synthetic data generator. I need a dataset that has 10 rows and 5 columns: id, first name, last name, country, email



Sure, I can generate a synthetic dataset for you. Here's an example dataset with 10 rows and 5 columns:


id	First Name	Last Name	Country	Email
1	Jack	Smith	USA	<a href="mailto:jack.smith@example.com">jack.smith@example.com</a>
2	Emma	Brown	Canada	<a href="mailto:emma.brown@example.com">emma.brown@example.com</a>
3	Liam	Johnson	UK	<a href="mailto:liam.johnson@example.com">liam.johnson@example.com</a>
4	Olivia	Davis	Australia	<a href="mailto:olivia.davis@example.com">olivia.davis@example.com</a>
5	William	Garcia	Spain	<a href="mailto:william.garcia@example.com">william.garcia@example.com</a>
6	Sophia	Rodriguez	Mexico	<a href="mailto:sophia.rodriguez@example.com">sophia.rodriguez@example.com</a>
7	James	Martinez	USA	<a href="mailto:james.martinez@example.com">james.martinez@example.com</a>
8	Isabella	Hernandez	Brazil	<a href="mailto:isabella.hernandez@example.com">isabella.hernandez@example.com</a>
9	Benjamin	Smith	UK	<a href="mailto:benjamin.smith@example.com">benjamin.smith@example.com</a>
10	Charlotte	Green	Australia	<a href="mailto:charlotte.green@example.com">charlotte.green@example.com</a>

**Prompt:** I want you to act as a synthetic data generator. I need a dataset that has 10 rows and 5 columns: id, first name, last name, country, email

# LLM Example App - Dataset Search

PR

Show me a dataset on recent natural disasters.



Here is an example of a dataset on recent natural disasters:

👍

🗨

The Emergency Events Database (EM-DAT) is an international database that compiles information on natural disasters and technological accidents. The following dataset provides information on the most significant natural disasters that occurred in the world from January 2021 to March 2023.

Dataset: EM-DAT Natural Disasters (January 2021 - March 2023)

Columns:

- Year: The year in which the disaster occurred
- Month: The month in which the disaster occurred
- Country: The country affected by the disaster
- Disaster Type: The type of natural disaster (e.g., flood, earthquake, hurricane)
- Deaths: The number of deaths resulting from the disaster
- Injuries: The number of injuries resulting from the disaster
- Damage (USD): The estimated cost of damage in US dollars

Example entries:

yaml

Copy code

Year	Month	Country	Disaster Type	Deaths	Injuries	Damage (USD)
2021	March	Australia	Flood	0	0	1,000,000,000
2021	August	Haiti	Earthquake	2,248	12,763	1,500,000,000
2022	May	India	Cyclone	173	1,223	500,000,000
2022	July	Germany	Flood	181	5,000	5,000,000,000
2023	March	United States	Tornado	25	150	50,000,000

Regenerate response

Example entries:

yamlCopy code

Year	Month	Country	Disaster Type	Deaths	Injuries	Damage (USD)
2021	March	Australia	Flood	0	0	1,000,000,000
2021	August	Haiti	Earthquake	2,248	12,763	1,500,000,000
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2022	July	Germany	Flood	181	5,000	5,000,000,000
2023	March	United States	Tornado	25	150	50,000,000

Please note that this is just an example dataset, and there are many other sources of data on natural disasters. Additionally, the actual dataset may contain additional columns or data points, depending on the specific source and purpose of the data.

**Prompt:** Show me a dataset on recent natural disasters.

# Focused, Aligned and Ready

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- Orchestrate structured, semi- and unstructured data
- Data integration across hybrid environments – on-premise, multi-cloud, hybrid cloud, edge-core-cloud
- Machine learning algorithm evaluation, model selection, drift analysis for performant predictive/prescriptive analytics including LLM-based text analytics seamlessly part of data pipeline
- Metadata-enriched API-driven analytics enablement without the inconvenience of manual handshakes across apps requiring file extract, events monitoring, etc.



# Thank You

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## Data Checklist

Is your data ready for  
the future of cyber?

Volume  
Voracity  
Veracity  
Value

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