



# **Invisible risk: Algorithms, Implications and the New Scope of Personal Data**

**Tuesday, 11 November**

08:00–09:00 PDT

11:00–12:00 EDT

17:00–18:00 CEST

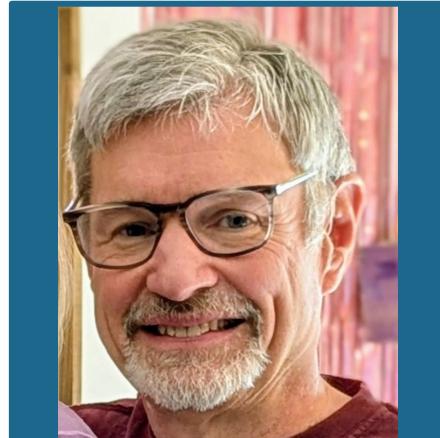


# Welcome and Introductions

## Panelists



**Emmi Bane**  
Principal – Data Ethics and  
Emerging Technologies  
HP



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Principal Privacy Architect  
HP

# Agenda

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# Introduction: La Soupe



Image credit Pablo Picasso via the Art Gallery of Ontario

# Why Personal Data is not Enough

- Lack of full visibility into the entire data ecosystem
- Non-identifying data can yield personal insights
- Privacy Safeguard Gaps – Process and Technology
- Focus on single data elements ignoring combinatorial effects



- Under GDPR, means “any information relating to an identified or identifiable natural person”
  - Refers to “ANY” information that implies personal attributes or behaviors without explicit or even indirect identification, but which is linkable or inferred,
- Results in artificial definitions of personal data: direct, indirect, and quasi-identifiers. These are based on the following,
  - “who can be identified, directly or indirectly, in particular by reference to an identifier such as a name, an identification number, location data, an online identifier or to one or more factors specific to the physical, physiological, genetic, mental, economic, cultural or social identity of that natural person;”
    - Suggests non-personal traits or actions have the potential to be related to an identified or identifiable a natural person
  - These artificial categories do not completely cover the “ANY” in the definition.
  - Places focus on individual data elements ignoring combining data sets and context of the data.
    - Does not capture the full expanse of personal data under GDPR and ignores combinatorial and contextual factors

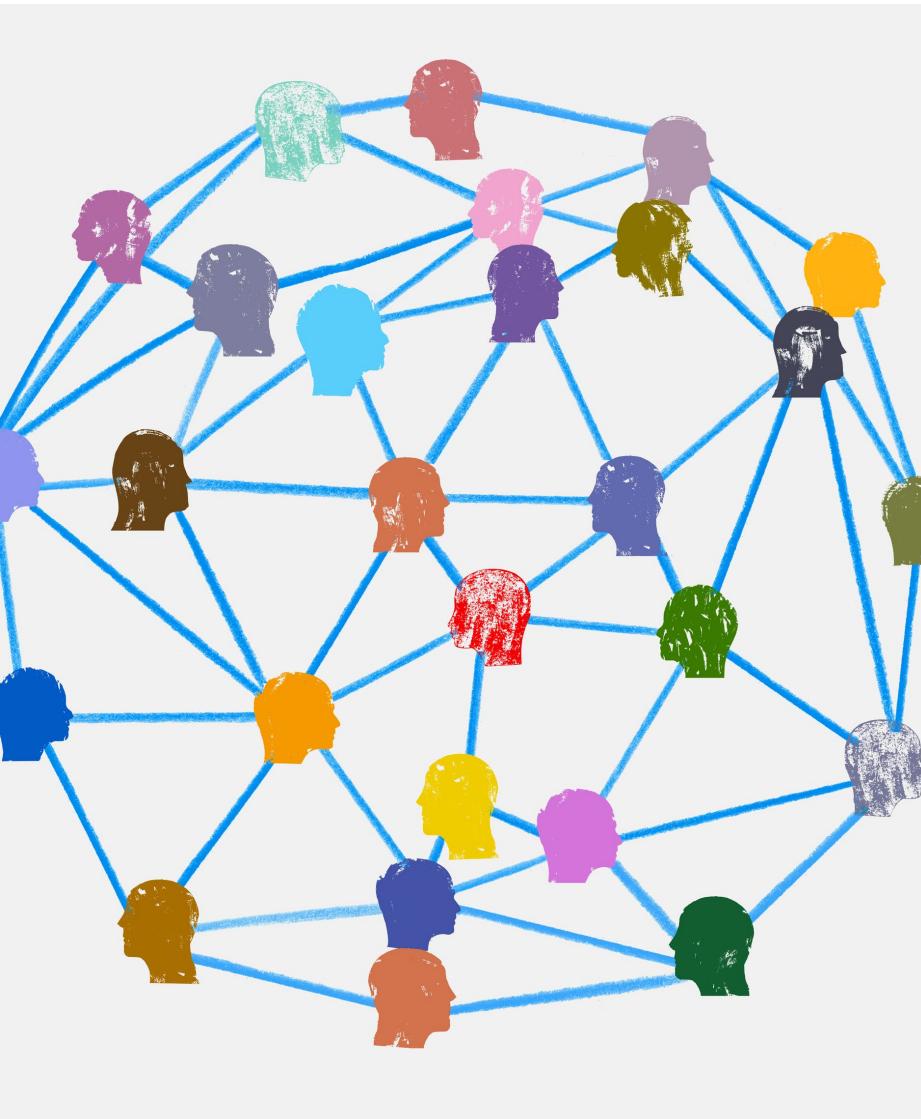
## Risks of the Narrow Governance approach

- Reidentification Risks
- Ethical Risks
- Contractual Risks
- Compliance Risks



## What is Implicative Data?

- Implicative data refers to information that, while not directly or indirectly identifying individuals, can influence perceptions, decisions, or outcomes related to them.



## Benefits of Implicative Data

- Extends data governance to entire data ecosystem with no artificial boundaries, where any data has the potential to be personal
- Surface previously unknown privacy risks at the individual, group, and organization levels
- Align with the level of data protection mandated by GDPR and other data protection regulations and laws

# Consequences of ignoring Implicative Data

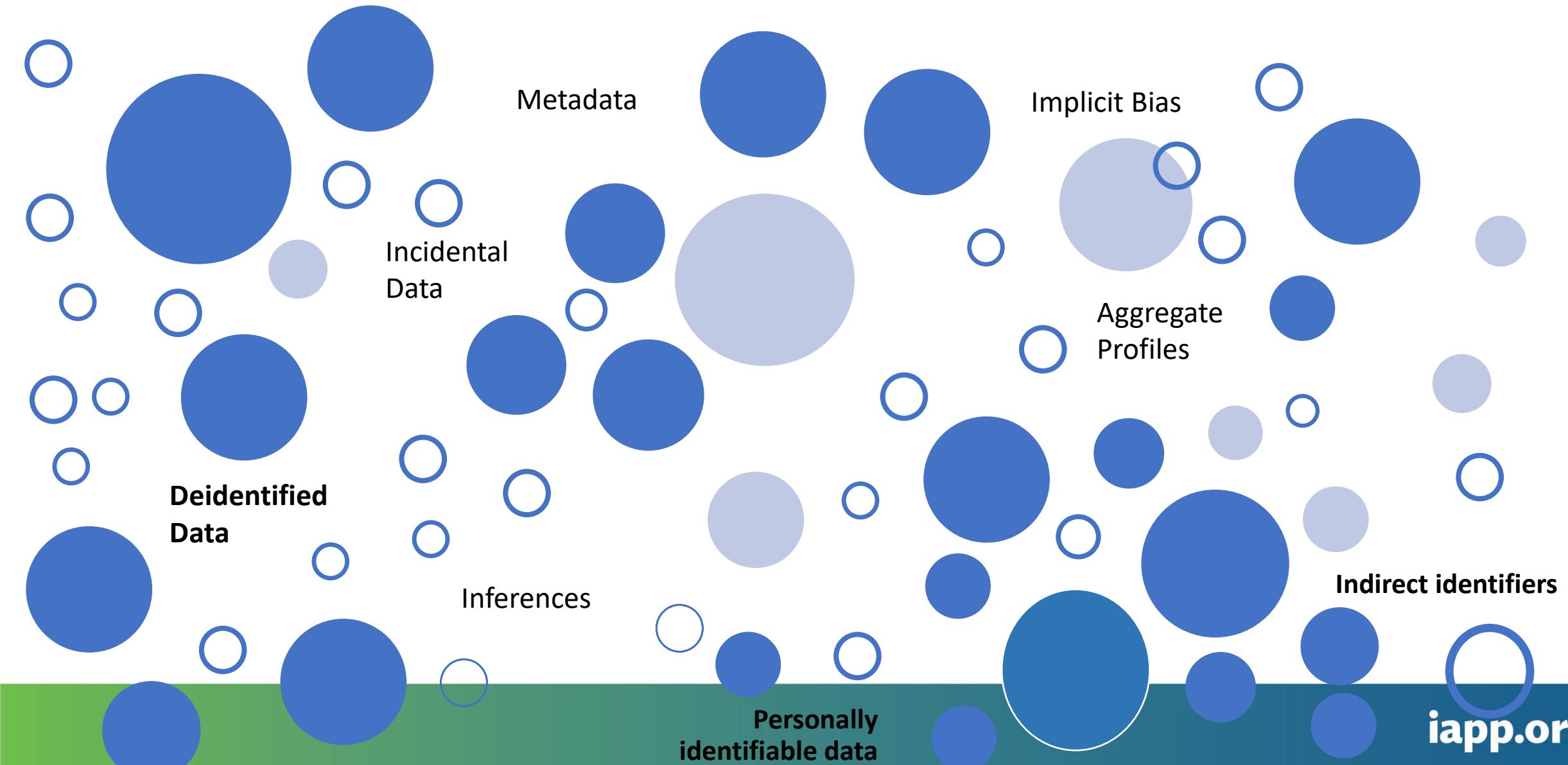
- Failure to trigger a privacy or security review when it is needed
- Inadvertent Revelation and/or Creation of Personal Information
- Bias, Erroneous Conclusions, and Misleading Narratives
- Regulatory Non-Compliance
- Larger Unprotected Data Landscape for Attack



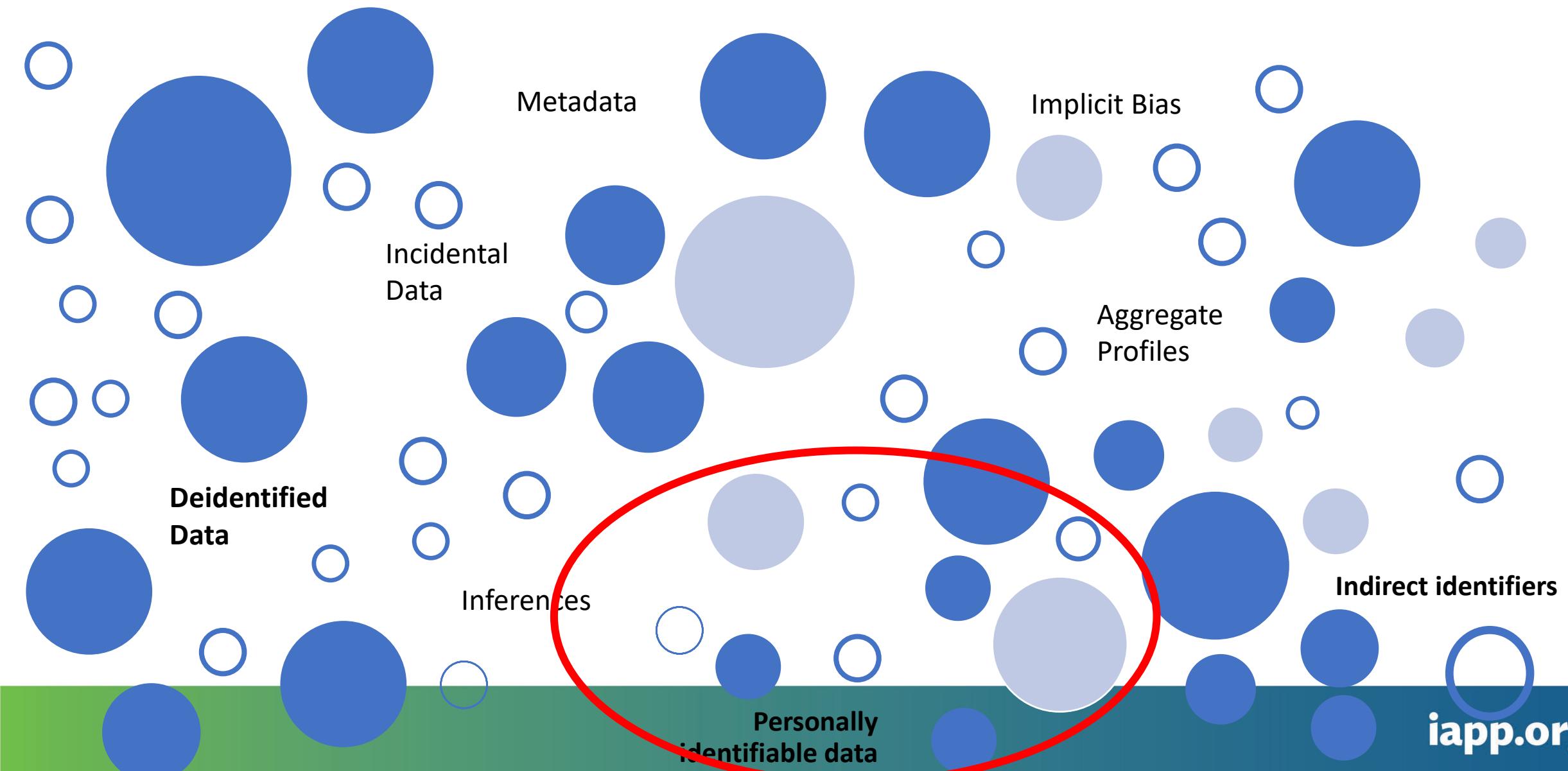
# Identifying individuals can be like detective work



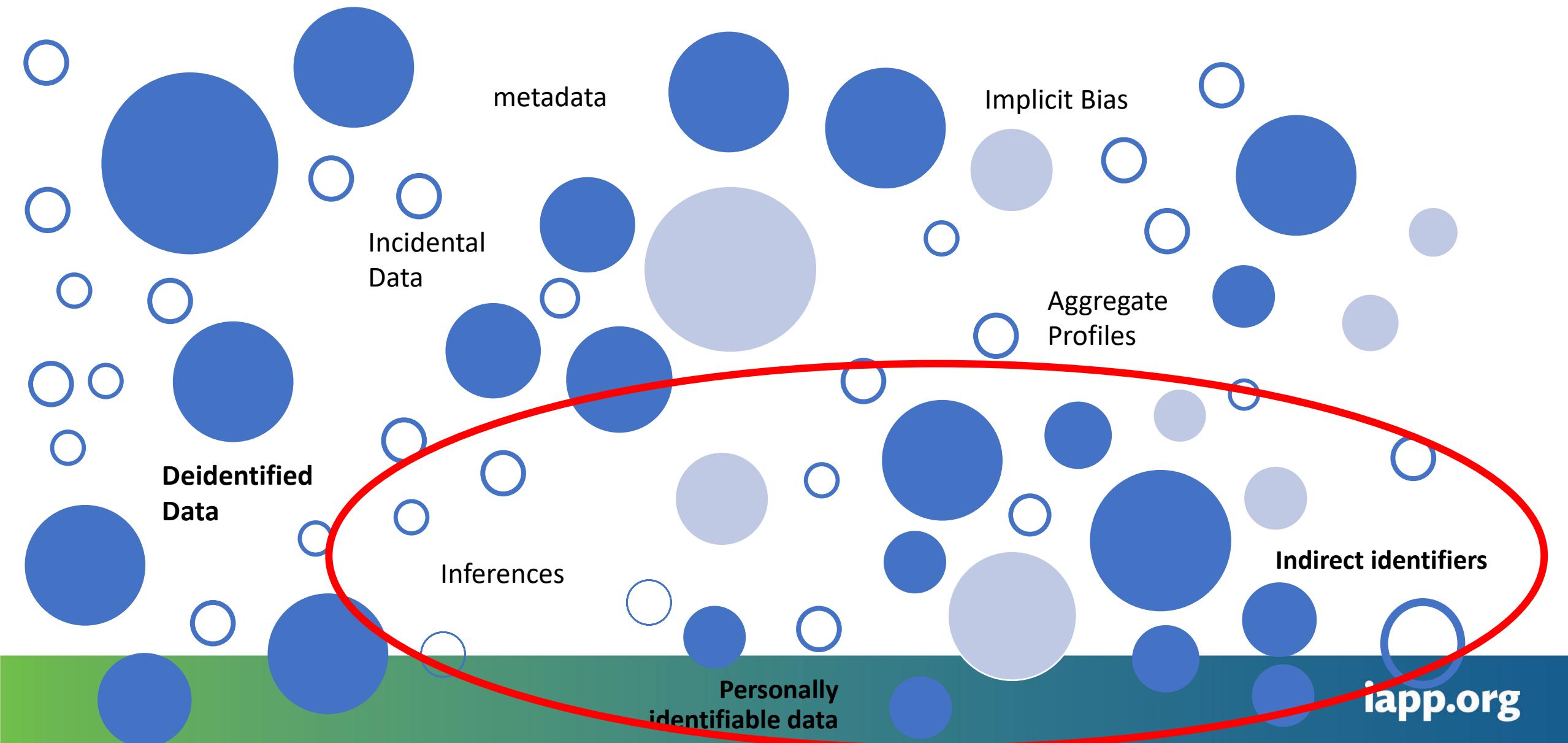
# Implicative data



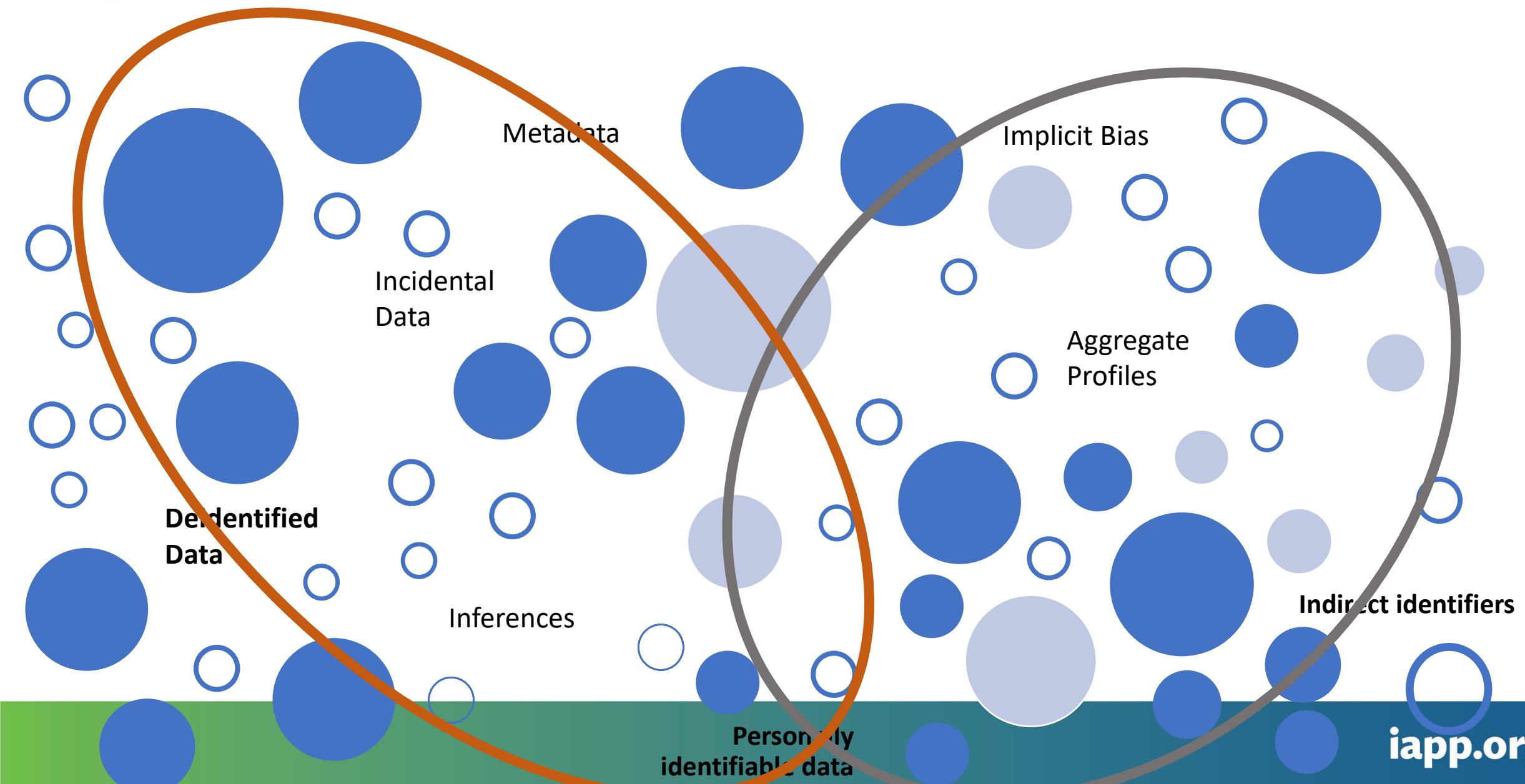
# iapp Personally Identifiable data



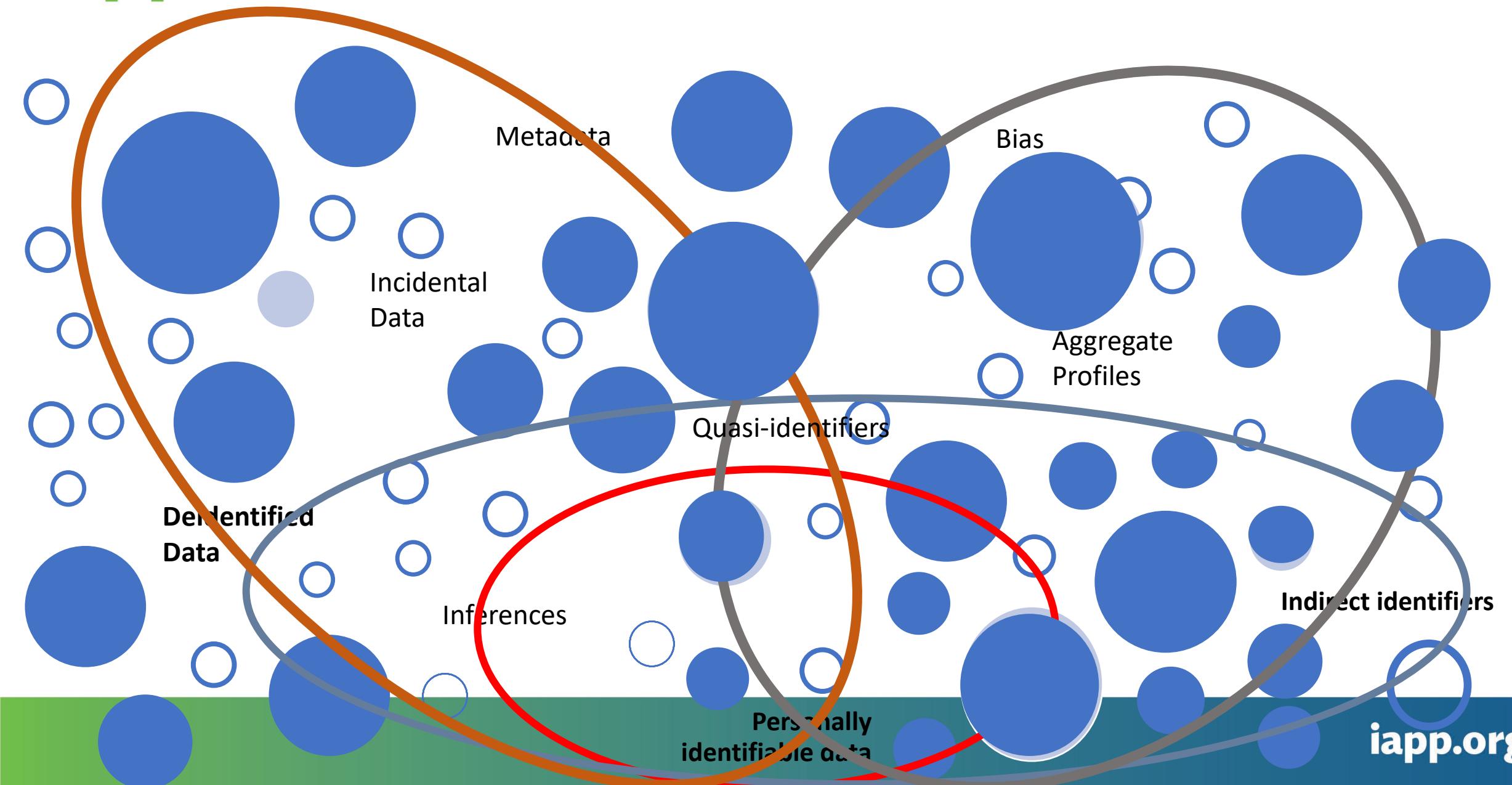
# iapp “Personal” data



# Other Types of Data



# iapp Implicative data – connecting the dots

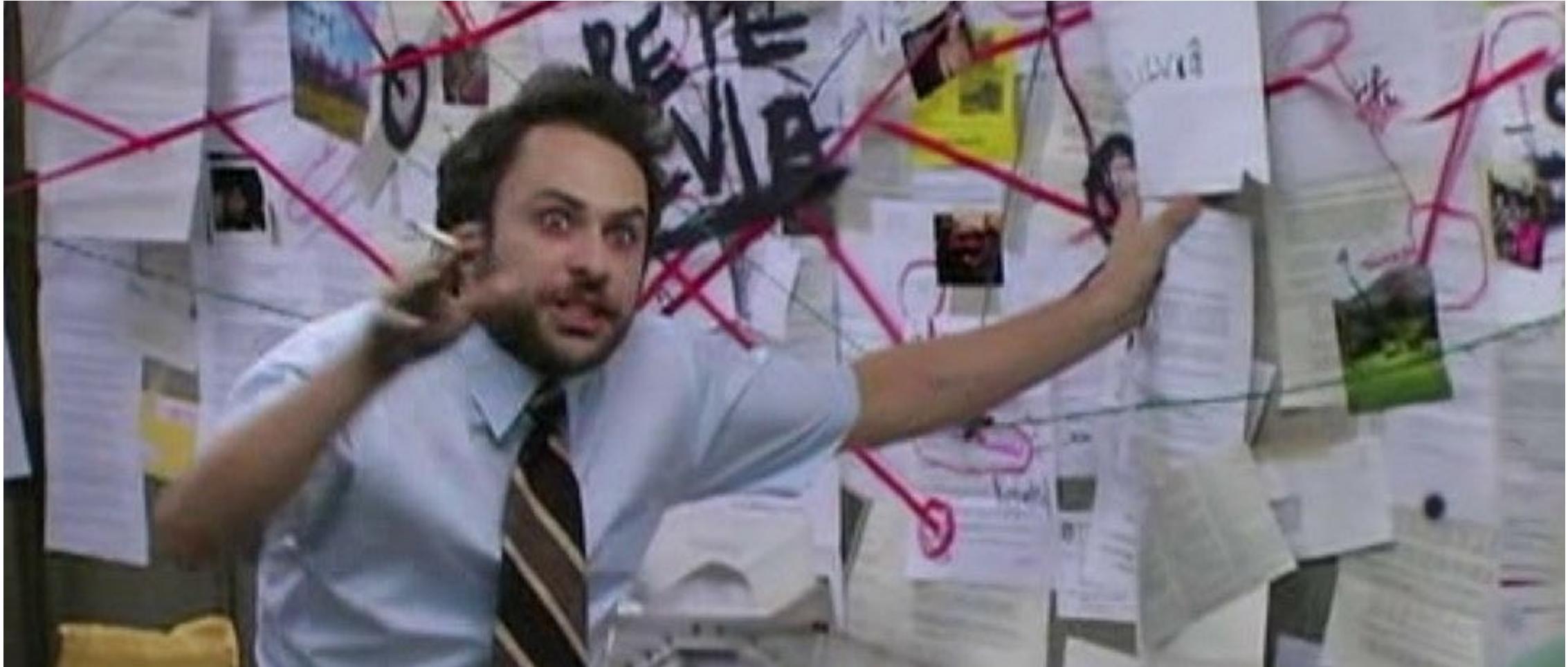


**Examples:  
Calendar Entries,  
GPS Metadata,  
Behavioral  
Profiles and  
Beyond**



**The Scoring of America  
World Privacy Forum  
Report, 2014**

**Voices in the code  
David G Robinson, 2022**



It's all implicative!

# Foundations of Implicative Data Governance



Contextual-based Risk Assessment



Data Minimization Principle



Context-aware by Design Privacy  
Engineering

# Applications: Privacy Context

# Definition

Full set of conditions that shape how personal data is collected, used, and governed at HP. It includes:

- **Source of data:** Where and how data was collected (e.g., app, region, notice).
- **Applicable rules:** Legal, regulatory, and policy frameworks tied to the data.
- **Context:** The purpose and environment in which data is used.
- **User and data collection context:** Who the data relates to and under what conditions it was gathered.

This context is captured through metadata, tagging, and system intelligence to enable precise privacy controls and avoid one-size-fits-all approaches

- **Precision in Controls:** High-context data enables tailored privacy protections, reducing both overreach and under-compliance.
- **Ethical Alignment:** Contextual privacy respects social norms and expectations, especially when handling sensitive or implicative data.
- **Global Consistency:** Privacy commitments transcend borders, requiring context-aware governance to meet and exceed legal standards worldwide.
- **AI and Emerging Tech:** As AI integrates across operations, contextual privacy ensures responsible innovation by aligning data use with ethical and legal expectations.



## Benefits of Privacy Context

- Enables **automated decisioning** based on context.
- Supports **risk-based reviews** and **fast-track lanes** for low-risk use cases.
- Forms the foundation for **Privacy Statement as Code** and future-proof governance architectures.

# Operationalization of Implicative Data

# iappWhat is Context Operationalization?

**Information operationalization** is the act of defining what "information" means in a specific context and determining how it can be communicated and exchanged, understood, and interpreted. This underpins the establishment of norms and expectations.

**Technical operationalization** refers to the process of defining privacy context in terms of a specific data model with associated technical procedures or systems that allow for consistent measurement and analysis at scale.

# Issues

## Informational

Information flow

Ethical assessment

Norms and expectations

Transparency/Interpretability

## Technical

Technical definition of context

Dynamic vs Overlapping context

Context Partitioning: Micro/Macro

# Elements of Consent: Analog to Digital

## Belmont Report (1976)

- **Voluntary**
- **Informed**
- **Comprehended**

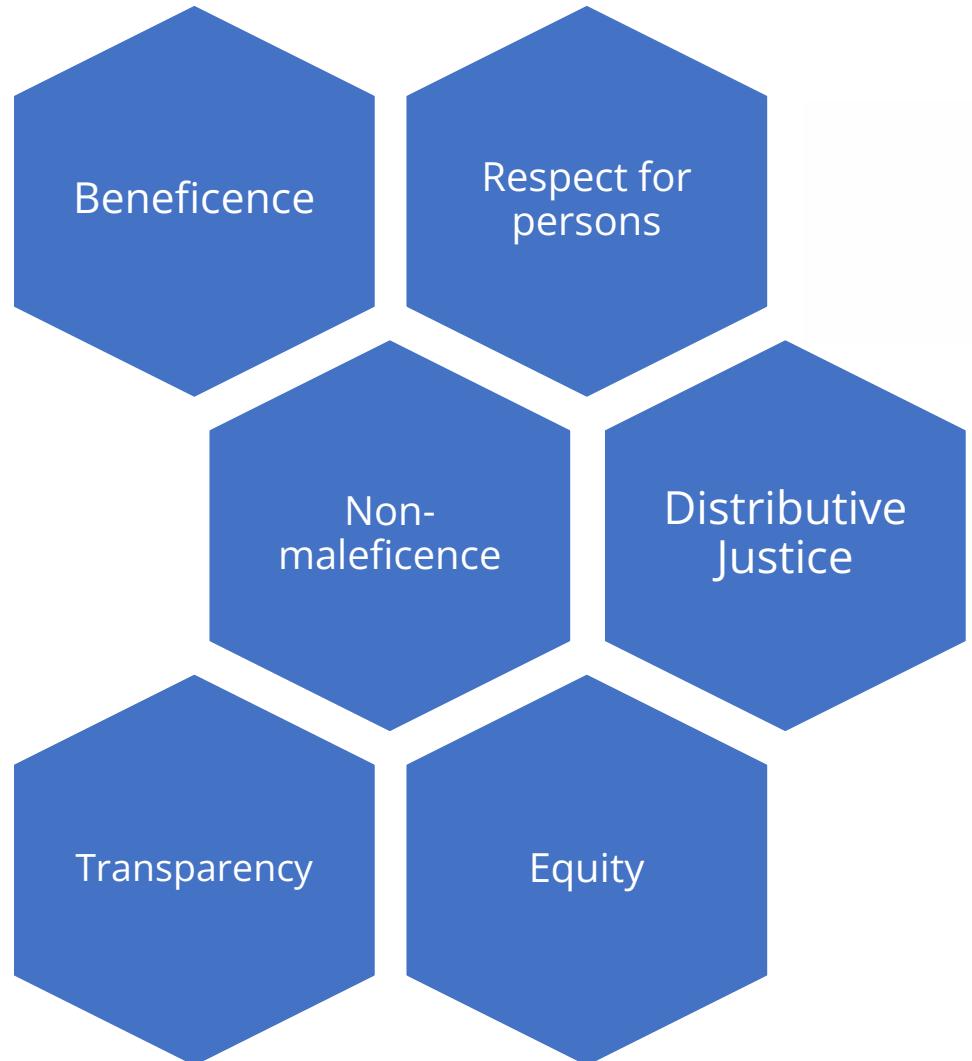


## GDPR (2016)

- **Freely given**
- **Informed**
- **Specific**
- **Documented**
- **Retractable**

# Reality evolves because of context

- What is gained and lost?
- How can we compensate for the limitations of our medium?
- How can we retain (and augment!) the integral function of consent: to respect persons?
- How can we meet changing expectations for expedience and comprehension?
- What is exceptional about emerging technologies?



# Technical

**Technical definition of context**

**Dynamic vs Overlapping context**

**Context Partitioning: Micro/Macro**

# Areas of the Future

# Technical

## Dynamic vs Overlapping Contexts

### Defining and Detecting Contexts:

What exactly constitutes a “context” in practice? Real-world contexts are fluid, overlapping, and can change dynamically.

## Context Partitioning

### Embed CI reasoning into AI

**assistants:** Context aware AI models honor context-specific norms – effectively giving them a built-in “ethical guardrail” for privacy utilizing machine reasoning (to infer norms) and memory architectures that tag data with context.

## Dynamic and Overlapping Contexts

### Declarative Policy Language:

Information flows expressed as temporal-logic formulas that can capture contextual constraints. This formalism allows reasoning about whether a sequence of actions violates or complies with contextual norms

## Technical definition of Context

### Context Taxonomy and Graph

**Database:** Create a contextual element taxonomy and use a graph database to determine relationship and strength of the relationships. GraphRAG and RAG can be leveraged for this purpose.

## Technical Definition of context

### Standardize context for engineering:

Standard definitions of contexts and data flows that engineers can reference (much like standards that exist for data protection terms).

## Context Partitioning

### Contextual Privacy Rules (CPP):

Adaptive policies that change based on context. Rather than a static privacy notice, a CPP would dynamically enforce rules depending on variables like who is requesting data and why.

# Conclusion

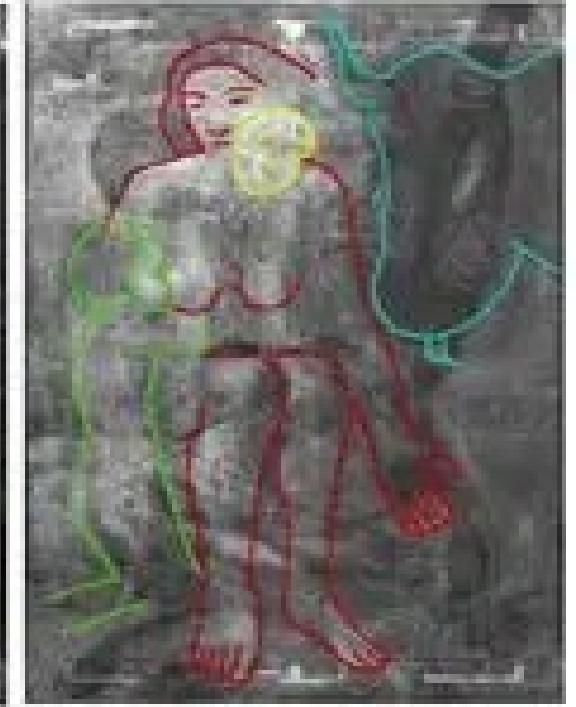


**Addressing  
Implicative  
Data**

**Robust,  
Holistic  
Governance  
Approach**

**Contextual  
Awareness Helps  
Meet User  
Expectations**

**Declarative  
Policy  
Language**



## Connect with us!

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