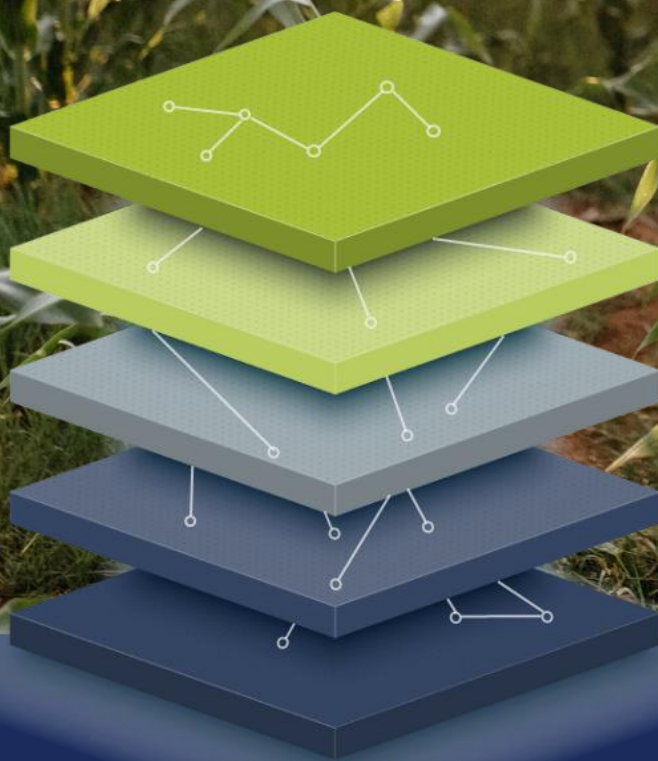


# AXIOM™ Technology



# Introduction



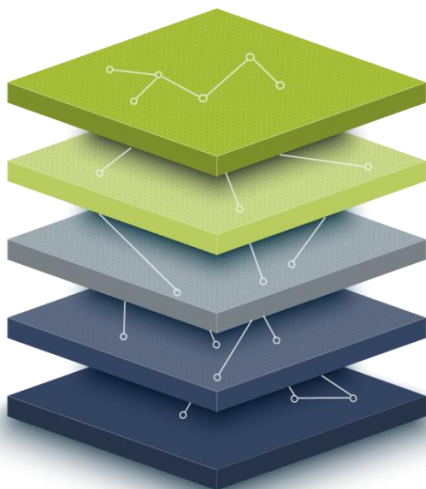
## What is “Axiom”?

Axiom is a comprehensive data platform that spans the entire data lifecycle in agriculture—from ingestion and storage through analytics, modeling (including AI/ML and LLM-based approaches), and delivery of insights.

It leverages multiple data structures and services, enabling advanced agriculture digital twin simulations (with tools like APSIM or DSSAT), remote sensing models, and other proprietary developed ML/AI modeling services.

By incorporating robust data governance, advanced security, and scalable infrastructure, Axiom ensures high-performance analytics and future-proof innovation for agronomic stakeholders.

**Figure 1:** The largest field level ag data knowledge base and next-generation data technology



-  **Predictive Models**  
Accelerate data driven ML models development and validation
-  **Data Fusion & Migration**  
Migrate historical data, standardize and enrich for analysis
-  **LLM Models**  
Leverage data assets for LLM training and finetune for any purpose
-  **Data Discovery & Insights**  
Analysis and discovery directly on the data without data preparations

# Key Components of Axiom

By combining cloud-native scalability with a graph-based approach and a robust agronomic ontology, Axiom enables rapid analytics and AI-driven insights across diverse, ever-evolving agronomic datasets. The result is a holistic platform that caters to data ingestion, semantic curation, advanced analytics, and ML/AI capabilities, ultimately driving faster, more informed decision-making in agriculture.

Key components of Axiom include:

## 1. SaaS Cloud Infrastructure

- **Security & Compliance:** Ensures data protection with role-based access, encryption at rest and in transit, and adherence to relevant data privacy standards.
- **Scalability & Flexibility:** Leverages cloud-native services that dynamically adjust compute and storage resources to accommodate fluctuating data volumes and workloads.
- **Cost-Effectiveness:** Utilizes pay-as-you-go models, optimizing expenses by scaling down during off-peak periods or low data ingestion times.
- **High Availability & Resilience:** Employs multi-region deployments and automated failover mechanisms to minimize downtime and ensure consistent platform uptime.

## 2. Graph DB Technology

- **Adaptive Data Structure:** Uses a graph-based model where it makes sense to represent complex relationships across crops, soils, locations, treatments, and other agronomic entities.
- **Integration & Enrichment:** Incorporates external data sources (e.g., weather, satellite imagery) via flexible APIs, enriching the agronomic graph with additional context.
- **Domain Dynamics:** Easily evolves with changing research needs or new agronomic concepts, avoiding the rigid schemas typical of relational databases.

### 3. Agmatix Data Model, Semantic Layer, & Ontologies

- **GUARDS (Growing Universal Agronomic Data Standard):** An ontological framework that standardizes and unifies all agronomic data, ensuring consistency and interoperability across different datasets.
- **Semantic Layer:** Applies domain-aware logic to validate, categorize, and interpret data according to agronomy-specific taxonomies, units of measure, and controlled vocabularies.
- **Unified Data Curation:** Facilitates cross-trial and cross-domain analytics by enforcing consistent data structures, making it easier to integrate with third-party tools or models.

### 4. Advanced Analytics Layer

- **Ready-to-Use Statistical Analytics:** Offers prebuilt modules for basic statistical operations, comparative analysis, and trend identification—enabling quick insights without extensive coding.
- **Descriptive Reporting:** Generates intuitive dashboards and visualizations that distill large agronomic datasets into actionable summaries (e.g., yield comparisons, treatment efficacy).
- **Collaborative Exploration:** Allows agronomists, data scientists, and stakeholders to collaborate on the same datasets, ensuring everyone works from a single source of truth.

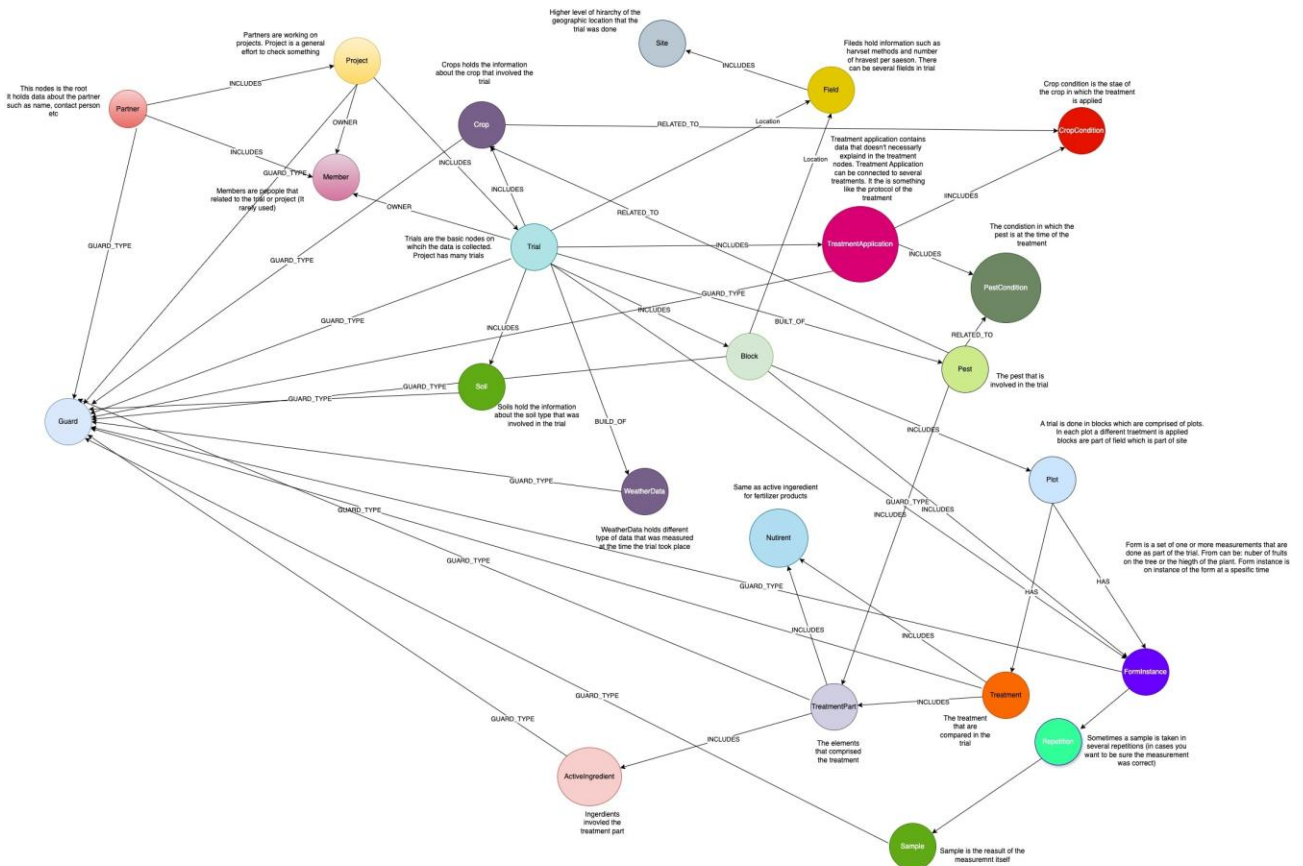
### 5. ML & Generative AI Data Capabilities

- **Data-as-a-Service (DaaS):** Provides AI-driven data services, from raw ingestion to enriched, model-ready data, enabling users to harness advanced analytics quickly.
- **Digital Twins & Synthetic Data:** Leverages simulation technologies (e.g., DSSAT integration) and generative models to project scenarios or create artificial data for robust model training.
- **ML Models-as-a-Service:** Delivers end-to-end machine learning solutions—from custom predictive models (e.g., yield forecasting) to generative AI agents specialized for agronomic workflows.
- **Scalable AI Infrastructure:** Integrates seamlessly with the underlying SaaS cloud infrastructure for scalable AI deployments, ensuring performance even with large or complex data streams.

# Transforming Data Structure into a Knowledge Graph & More

By evolving from a graph database to a full-fledged Knowledge Graph, Agmatix harnesses both the structural and semantic aspects of agronomic data. This approach delivers richer insights, faster model development, and more accurate analytics, fueling innovation across agricultural research, field trials, and decision-support solutions.

Agmatix's platform uses a graph database to represent the complex and ever-evolving relationships within agronomy—linking data points like soils, crops, treatments, and environmental factors. This node-and-edge model naturally captures multidimensional relationships that traditional relational databases struggle to handle.



Agmatix goes beyond merely storing data in a graph and includes:

### **1. Ontologies & Semantic Layer**

- Through GUARDS (Growing Universal Agronomic Data Standard) and other ontological models, the platform embeds meaning into the graph. Each node (e.g., a crop variety) and each relationship (e.g., “is grown in”) is tied to a domain-specific definition, ensuring consistent interpretation across datasets.

### **2. Contextual Data Integration**

- External sources (e.g., weather, satellite imagery) are seamlessly linked via graph-based APIs, reinforcing the contextual web of agronomic knowledge. This transformation—where raw data is combined with domain knowledge—turns the graph database into a Knowledge Graph.

A robust Knowledge Graph opens powerful capabilities for data scientists and agronomists:

### **1. Advanced Relationship Discovery**

- Data scientists can easily traverse the graph to uncover hidden correlations, such as how certain soil types interact with specific crop varieties under particular climatic conditions.

### **2. Feature Engineering & Model Accuracy**

- The semantic relationships encoded in the Knowledge Graph provide rich features that significantly boost the performance of machine learning models—reducing time spent manually wrangling or engineering variables.

### **3. Holistic Data Exploration**

- Instead of siloed datasets, stakeholders explore integrated information—streamlining tasks like yield prediction, disease detection, or climate risk assessment across diverse agronomic domains..

### **4. Scalable AI & LLM Integrations**

- With a well-structured Knowledge Graph, LLMs and other AI services can tap into precise, domain-validated data, improving the relevance and reliability of predictive or generative outputs.

# Key Axiom Capabilities in Scope

## 1. Data Ingestion & Integration

- **Multi-source Ingestion via integration Framework:** Ability to capture data from diverse sources (e.g., sensors, satellite imagery, APIs) in various formats (structured, semi-structured, unstructured).
- **Data Pipelines to Migrate Historical Data:** Automatic and semi-automatic data pipelines to ingest legacy agronomic data from Excel/CSV formats including automated data quality checks, validation rules, and transformations to ensure consistent, reliable information across systems.

## 2. Data Storage & Curation

- **Agronomic Data Lake:** Axiom provides the ability to store raw or lightly processed data (e.g., satellite images, sensor feeds, trials data, ground truth data, our SaaS data) in a secure, cost-effective way with multiple data structures that best fit for purposes.

## 3. Analytical, Predictive, and AI-Driven Modeling

- **AI/ML Models:** Built-in or pluggable AI and ML models for tasks like yield prediction, nutrient optimization, churn predictions, and disease risk identification.
- **LLM Integration:** Leveraging large language models (LLMs) for natural language queries, data summarization, or generating insights/recommendations.
- **Digital Twins with APSIM/DSSAT:** Integration with advanced agronomic simulation and digital twin platforms—such as APSIM or DSSAT—for scenario-based modeling of crop systems, soil conditions, and climate interactions.
- **Data Discovery:** Search and data exploration features to identify hidden relationships (e.g., synergy between crop rotation and soil conditions).
- **Aggregated and Anonymized Data:** Enable the users to see trends, analysis and benchmarks over anonymized and aggregated data.

## 4. Microservices and APIs (“Stat-Calc Layer” advanced)

- **Service-Oriented Architecture:** A set of modular microservices (or APIs) that expose data, analytics, and model functionalities to internal teams, partners, or external applications.

# What are the True Differences Of Axiom vs. Off-the-shelf Solutions?

Axiom's specialized architecture, comprehensive ontology, and tailored data pipelines offer a level of agronomic alignment that off-the-shelf solutions cannot match without significant engineering overhead.

The main reasons are:

## 1. Domain Focus: Field Trials & Agronomy

Off-the-shelf Solutions	Where Axiom Shines
Designed to be general-purpose (big data storage, analytics, etc.).	Purpose-built around field trials, including the largest known trials ontology.
Require significant customization to capture trial-specific data models.	Predefined data structures and vocabularies tailored for trial design, crop treatments, and environmental variables.
No built-in agronomy concepts or experiment management features.	Capable of representing the entire agronomic world through an extended ontology, reducing the need for manual schema design.

## 2. Data Ingestion & Standardization

Off-the-shelf Solutions	Where Axiom Shines
Typically provide generic ETL or data integration tools (e.g., AWS Glue, Azure Data Factory).	Dedicated ingestion pipelines that standardize trial data into a single format, automatically applying agronomic ontologies.
May store data in relational, NoSQL, or graph—but rely on users to define schemas.	Tailored graph data structure pre-configured for relationships between fields, treatments, outcomes, and environmental parameters.
Data cleaning/validation often requires additional tools or custom scripts.	Built-in data validation for trial-specific formats, ensuring high quality and consistent outputs.

## 3. Curation & Potential for LLM Training

Off-the-shelf Solutions	Where Axiom Shines
Can handle large data volumes, but typically lack domain-specific curation.	Houses an extensive, standardized dataset of field trials—ideal for training agriculture-focused LLMs.
Data is rarely pre-labeled or richly annotated for agronomy.	Rich ontology and curated metadata give LLMs domain context, boosting relevance and accuracy.

## 4. Integration Framework & Predefined APIs

Off-the-shelf Solutions	Where Axiom Shines
Often provide a set of rigid APIs or connectors that may not address specialized agronomic data.	Offers a flexible integration framework capable of connecting with any data source or external tool.
May lock users into a specific vendor ecosystem.	Remains vendor-agnostic, enabling cross-platform data sharing and analytics.

## 5. Analytics, Machine Learning, & AI Modeling

Off-the-shelf Solutions	Where Axiom Shines
Need to invest to customize their structure to enable domain specific modeling.	Predefined agriculture model that requires little customization and cleaning before modeling starts.
Require data preparations, building tables and data pre-processing ready for analysis.	Allow direct data discovery on the data with no pre-processing for modeling purposes.
General-purpose digital twin offerings (e.g., for IoT, manufacturing).	integrates directly with DSSAT (open-source crop modeling) for agronomic simulations.
Generic structure even in graph cases requires user to build the knowledge graph and the meaning between nodes.	If we build the knowledge graph on top of the graph structure we can reduce the need to develop it and bring the knowledge graph ready to use.

Axiom's domain-centric design, automated data standardization, and integration with specialized agronomic tools culminate in faster, more accurate, and more insightful analytics and AI modeling for field trials (and beyond).

Here are the main reasons why:

- **More Accurate Models:** Thanks to standardized, high-quality data and direct integration with proven agronomic models, customers can build robust, trustworthy analytics and AI.
- **Faster AI Insights:** Ready-to-use data ingestion pipelines and a massive, well-labeled dataset reduce the time from project start to actionable insight.

- **360° Querying & Analysis:** A graph-centric approach unlocks multidimensional queries—relating soil conditions, genetics, treatments, and outcomes in ways rigid relational systems can't.
- **Scalability & Futureproofing:** Axiom's flexible integration framework and extensible ontology mean it can grow with an organization's needs, from small trials to global multi-crop operations.
- **Streamlined Collaboration:** Domain-specific consistency fosters smoother communication among agronomists, data scientists, and business teams, minimizing data mistranslations or rework.

Outcomes	Axiom	Business Value
More Accurate Models	By enforcing a standardized data ontology, we dramatically reduce noise and boost reliability in analyses.	<p>Accurate models influence critical decisions in Ag which can translate to major yield improvements, cost savings and increased sales on a global scale.</p> <p>Having a consistent, accurate dataset for AI can separate market leaders from followers, especially as climate and resource constraints intensify</p>
Faster AI Insights	Ready-to-use data ingestion pipelines and a large, well-labeled dataset reduce months of labor-intensive efforts to days or even hours.	<p>Rapidly deploying predictive models or new product ideas can shorten R&amp;D cycles, helping agribusinesses and researchers innovate faster and respond quickly to emerging challenges</p> <p>By accelerating insights, we enable continuous experimentation, turning agriculture—often a slow industry—into a more agile, data-driven field.</p>
360° Querying & Analysis	A graph-centric data model captures relationships natively, making it straightforward to run rich, multi-angle queries	<p>Seeing how one variable (like soil pH) intersects with multiple treatments and outcomes provides deeper insights that can uncover hidden correlations.</p> <p>It's easier for agronomists, breeders, and data scientists to ask cross-disciplinary questions that integrate biology, climate, and economics—a critical step toward integrated, sustainable solutions.</p>

Outcomes	Axiom	Business Value
Scalability & Futureproofing	A flexible integration framework and an extensible ontology accommodate diverse data, from small local trials to global multi-crop operations.	Customers won't have to replace or drastically overhaul their data platform when they expand into new crops, regions, or research areas.  A solution that can handle the scale of multinational agribusinesses is more likely to push industry standards and become the standard for data management in Ag.
Streamlined Collaboration	Domain-specific standards and consistent data structures create a common language, so agronomists, data scientists, and business stakeholders can easily share and interpret information	A single, universal data standard cuts down on repeated data cleaning, minimization of error, and rework.  When researchers and commercial teams can quickly align on data insights, breakthroughs move faster from hypothesis to field implementation.

Are those outcomes and possible impacts marginal or transformative ?

Marginal improvements typically focus on small boosts in efficiency or minor feature enhancements while Axiom's proposition is broader and more systemic:

- Raises the overall quality of agronomic data and AI;
- Speeds up innovation cycles in an industry where data science can be slow.
- Reveals deeper relationships across complex datasets.
- Enables large-scale, global collaboration.
- Serves as a foundation for robust, domain-specific AI that generic solutions cannot easily replicate.

These factors have the potential to reshape agronomic decision-making and improve resource use at scale—contributing to everything from yield gains to sustainability goals.



[agmatix.com](https://agmatix.com)