# Technical Infrastructure Architecture Deck

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# **Technical Architecture Document: Al-Augmented System Transformation**

# Overview

This document outlines the current technical landscape at BBB and the transformation toward a unified, Al-augmented architecture. It is intended for technical teams, architects, data engineers, and cross-functional stakeholders who are contributing to the planning and development of the future-state platform.

The proposed system emphasizes scalable infrastructure, intelligent automation, unified data layers, and Retrieval-Augmented Generation (RAG) for contextual, explainable AI outputs.

# 1. Current System Workflow (As-Is)

Fragmented and Manual Workflow Landscape

### **Observations**

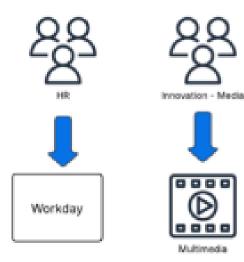
- Siloed systems across HR, Innovation, Core Services, Account Management, Sales, and Marketing
- Manual inputs via Outlook, <u>Monday.com</u>, Aircall, RingCentral, and HubSpot
- Disparate tools and disconnected integrations (e.g., NMI, Gravity Payments, Mandrill)
- Survey and response processing via InMoment and Email Services

# **Technical Challenges**

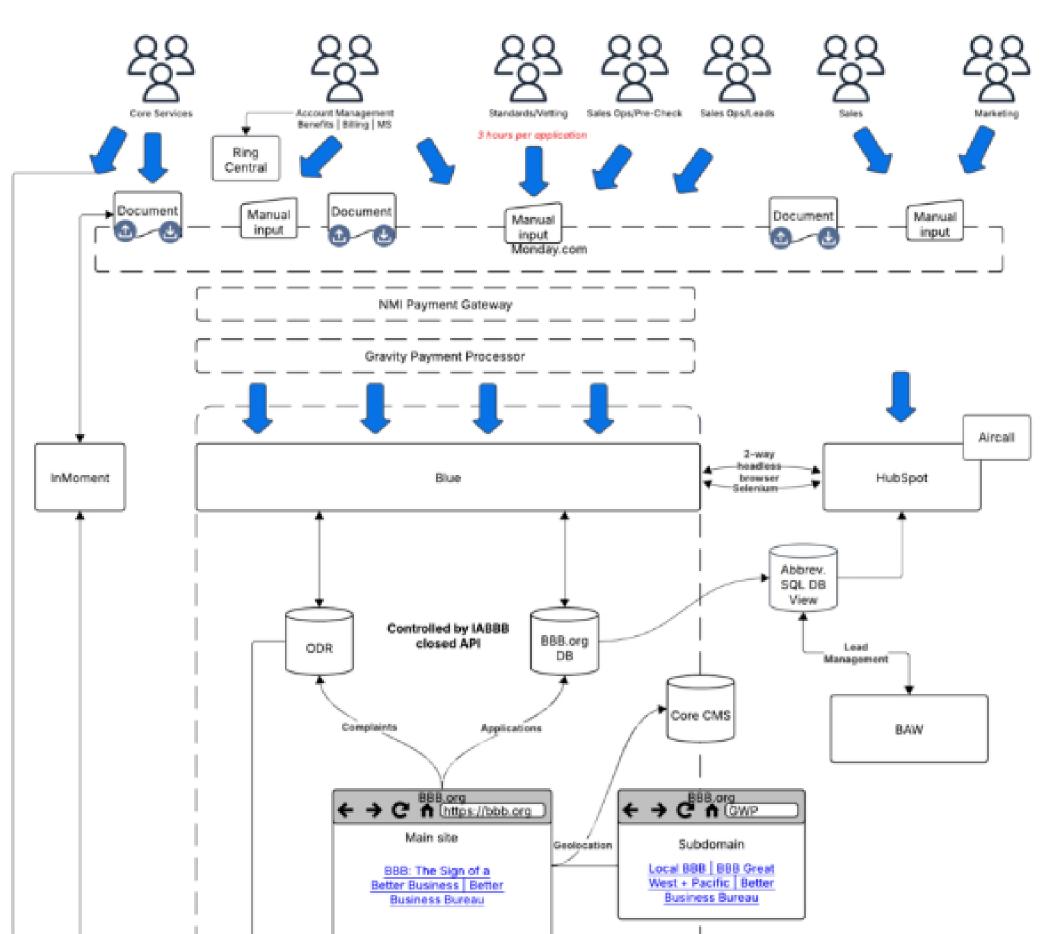
- Redundant data entry and duplicate work
- Lack of centralized data synchronization
- Fragmented customer and internal touchpoints
- High dependency on human input for execution

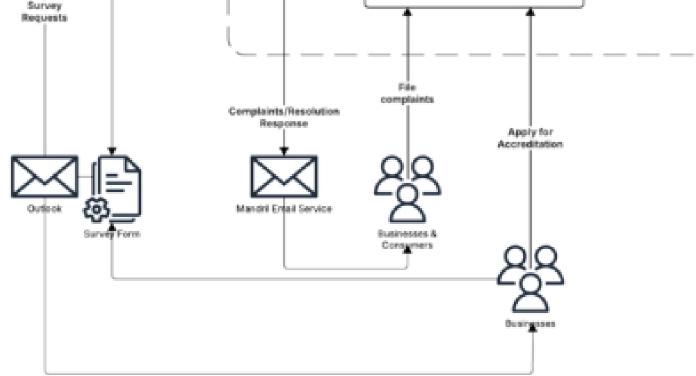
#### CURRENT SYSTEM WORKFLOW

Adding an LLM layer on top of a segregated semantic layer and non-integrated data will result in poor quality output, hallucinations, and lack of trust.



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# 2. Current Data Model (As-Is)

Segregated Data with Limited Semantic Integration

#### **Observations**

- Data resides in spreadsheets, Blue SQL DB, Azure SQL DB, and HubSpot
- Analytics tools (Power BI, HubSpot Analytics) operate in silos
- No consolidated semantic layer or centralized processing

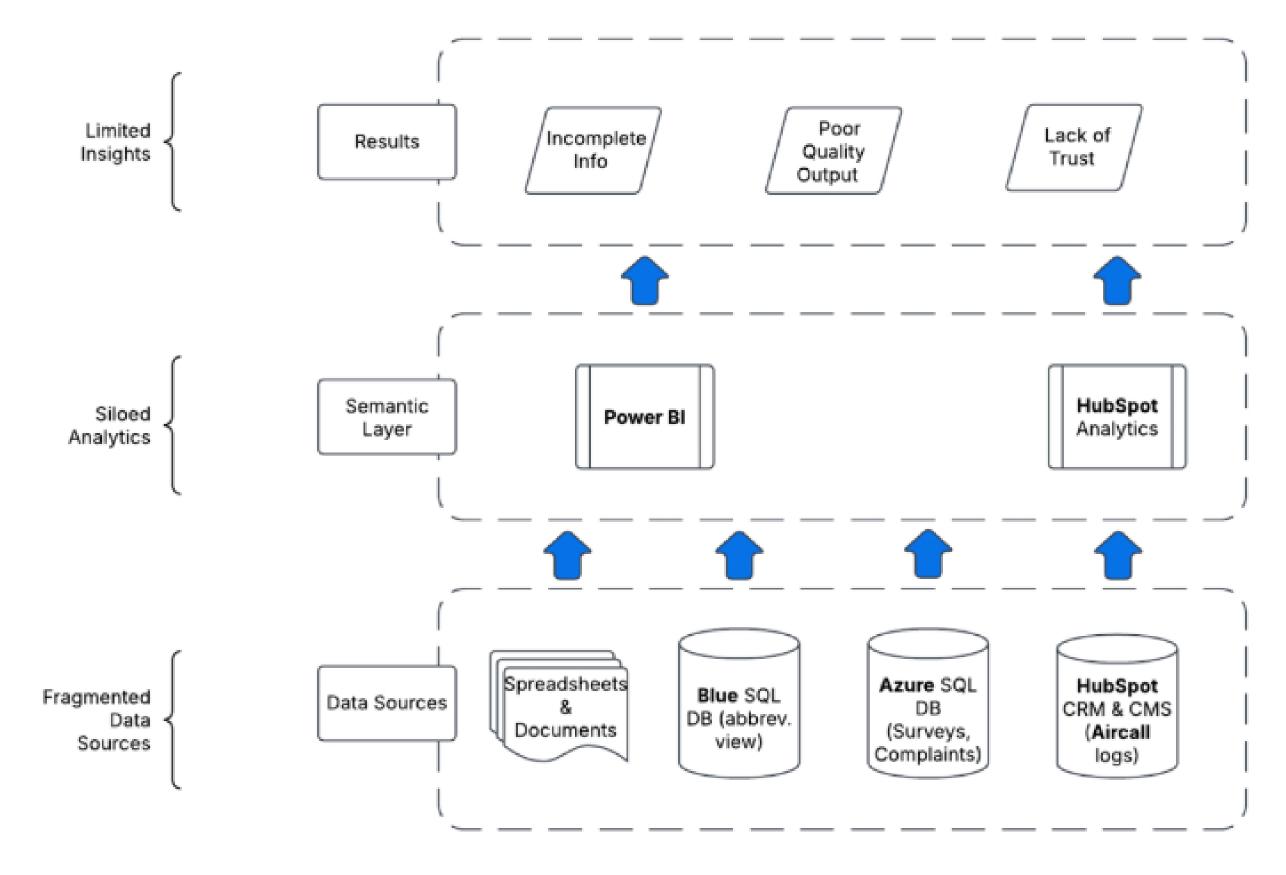
#### **Technical Risks**

Incomplete data = inaccurate insights 

- Poor lineage tracking & trust issues
- Difficult to enable machine learning/AI with inconsistent structure

#### CURRENT DATA MODEL

Adding an LLM layer on top of a segregated semantic layer and non-integrated data will result in poor quality output, hallucinations, and lack of trust.



# 3. Target Data Model (To-Be)

Unified, Explainable Semantic Data Architecture

#### Features

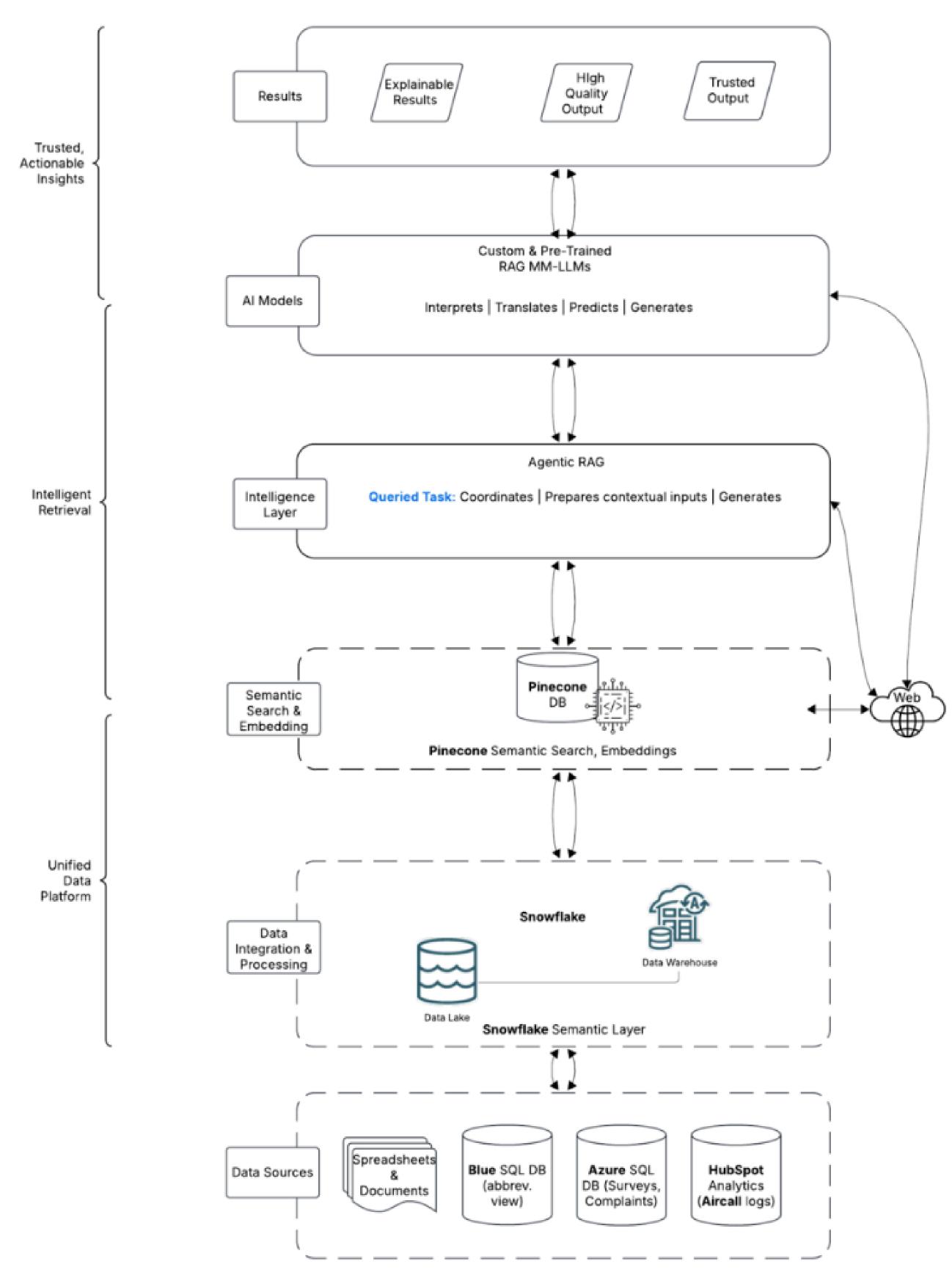
- Snowflake: Acts as both data lake and warehouse with semantic modeling
- Pinecone: Embedding-powered semantic search
- Intelligence Layer: Agentic RAG layer to interpret tasks and orchestrate AI queries

#### **Benefits**

- Real-time, explainable, high-quality outputs
- Stronger AI integrations and knowledge retrieval
- Supports intelligent agents and automated workflows

#### DESIRED DATA MODEL

Adding an LLM layer on clean, organized data with a semantic layer yields in high quality, trusted, and explainable results.



# 4. Target System Workflow (To-Be)

Al-Driven, Department-Specific Automation

#### Features

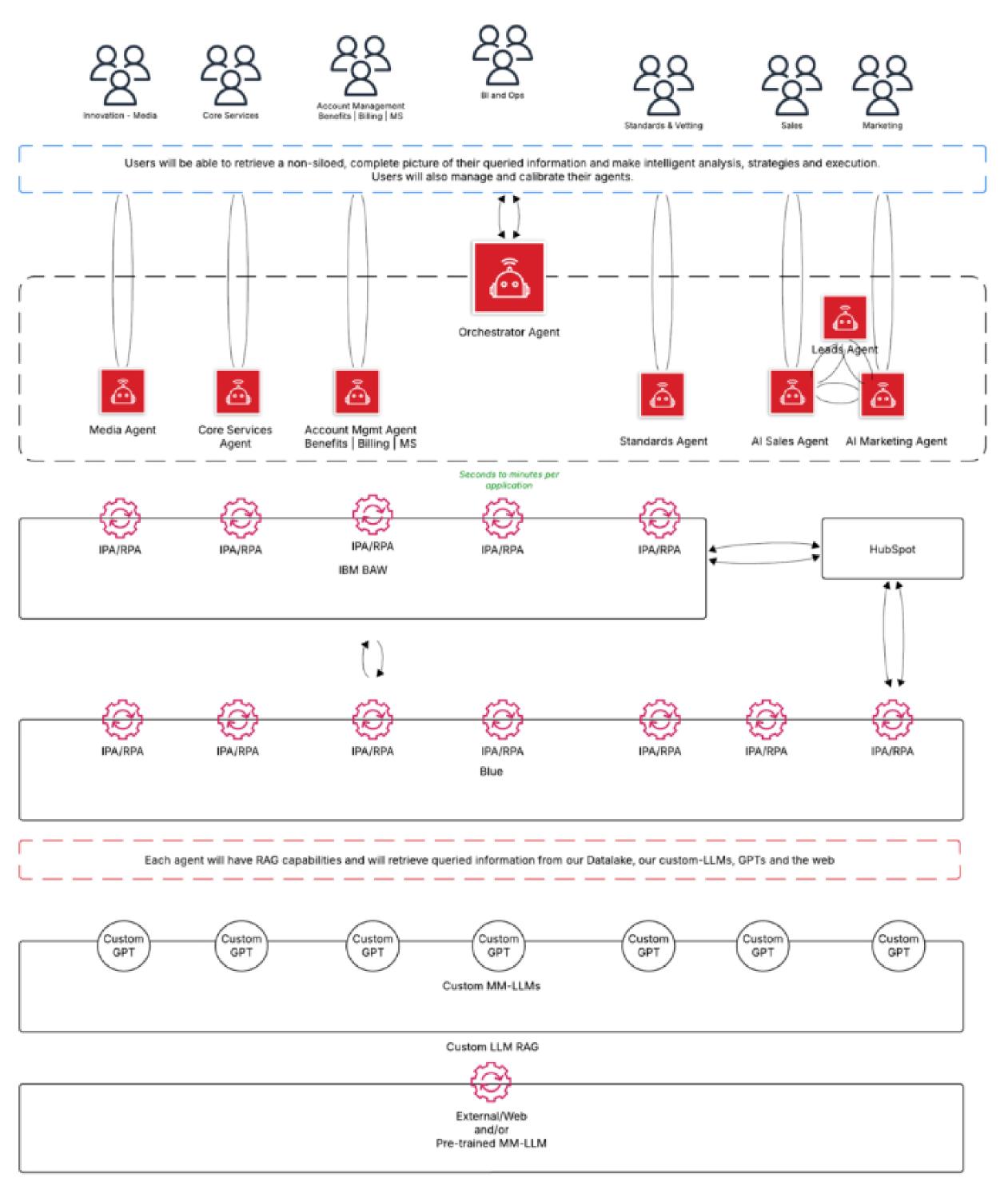
- Department-specific agents (e.g., AI Sales Agent, AI Marketing Agent, etc.)
- Orchestrator Agent coordinates workflows across roles
- IPA/RPA for back-end integrations with Blue, IBM BAW, HubSpot, etc.

# Impact

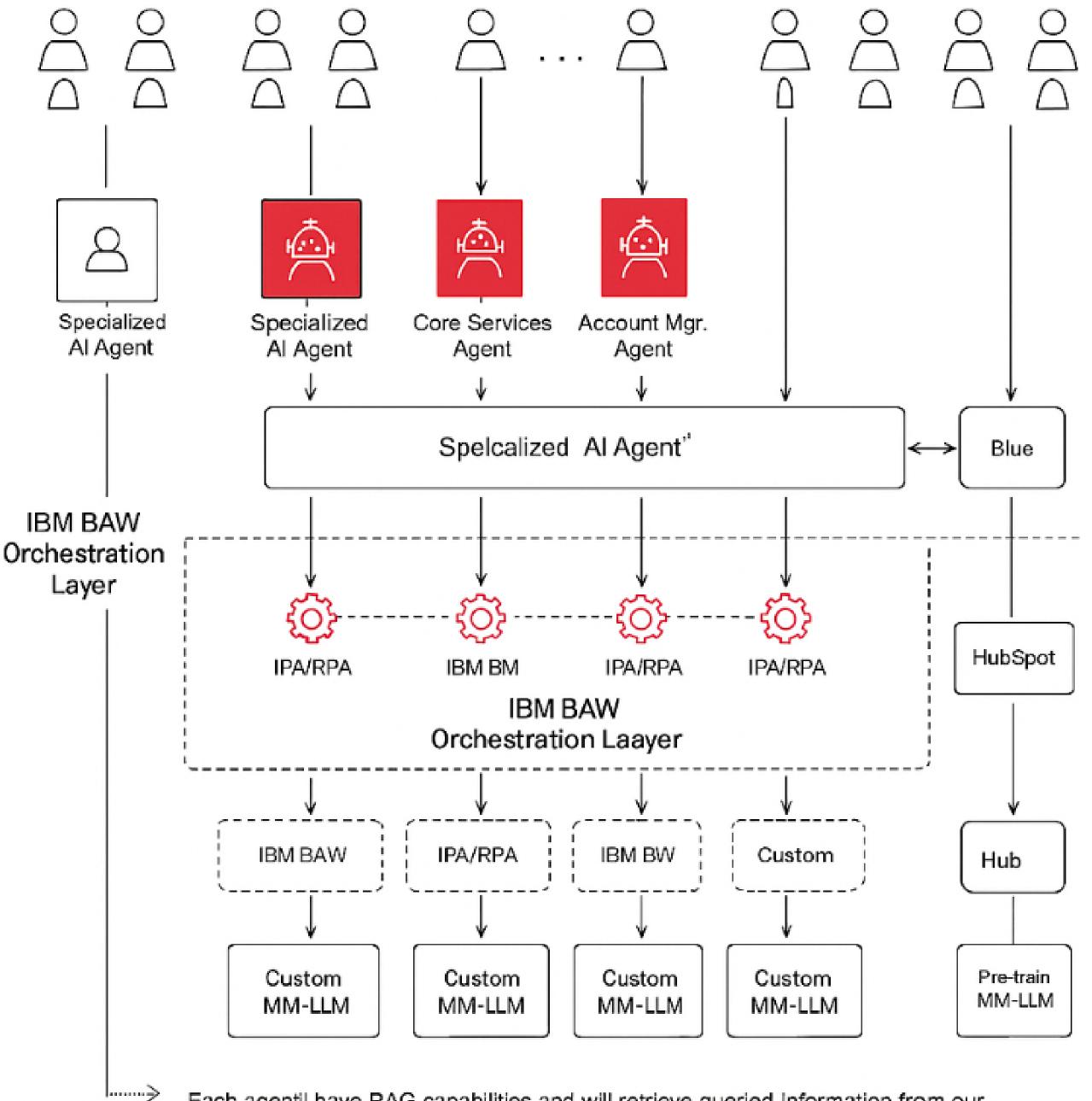
- Reduced time-to-action
- Higher autonomy for departments with less manual input
- Unified experience across departments

#### DESIRED SYSTEM WORKFLOW

Al agents along with RPA/automation will reduce the heavy manual workflows. Al Agents retrieve, analyze and provide well-researched information to their human counterparts.



Users will be able to retrieve a non-siloed, complete picture of their company and make intelligent analysis, strategies and execution.



Each agentil have RAG capabilities and will retrieve queried Information from our

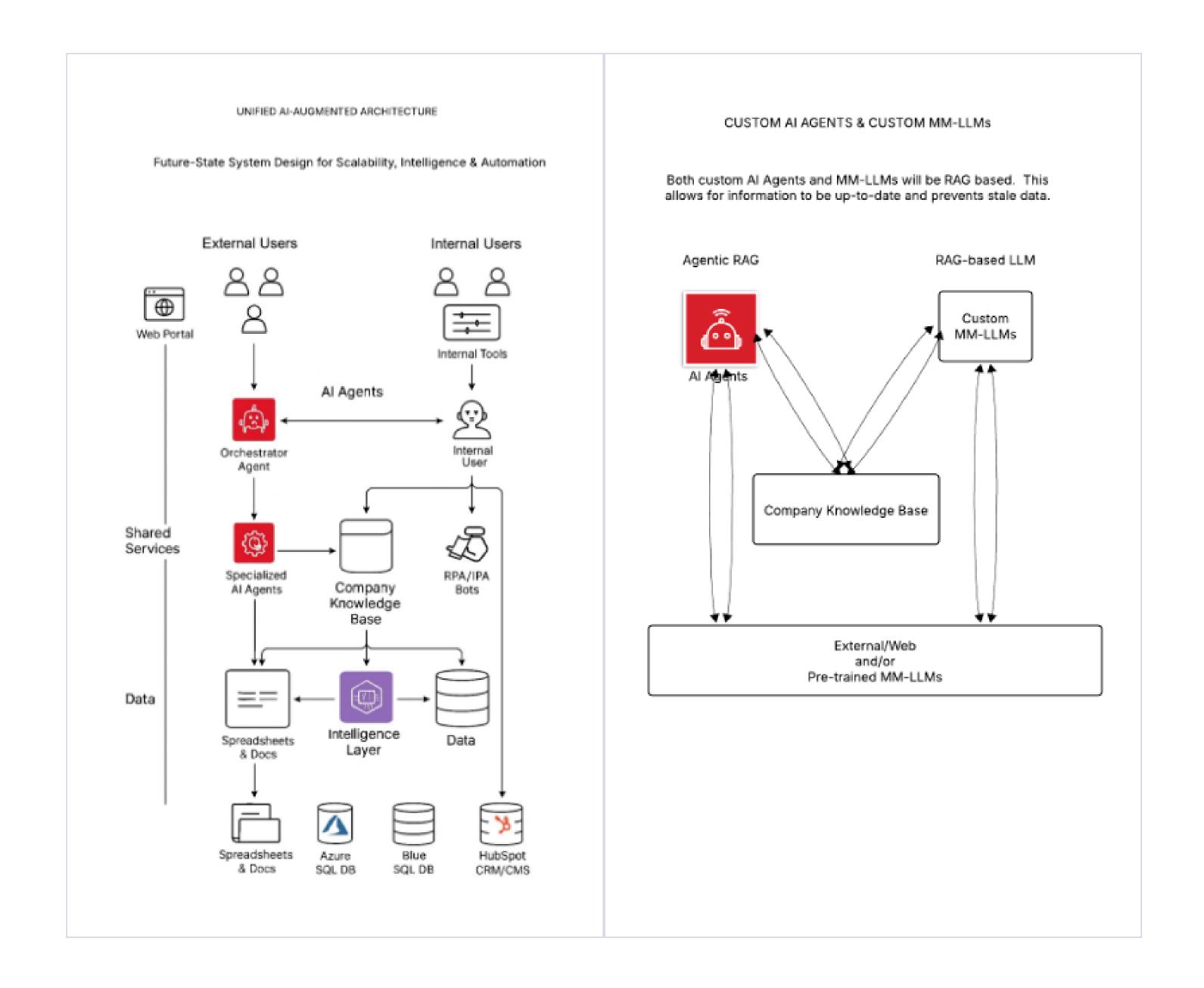
# **5. Unified Architecture Interaction Model**

**AI-Augmented Interactions Across External & Internal Users** 

#### Layers

- 1. **External/Internal Users** Customers and employees interfacing via web portals or internal tools
- 2. Al Agent Layer Orchestrator + task-specific agents
- 3. Company Knowledge Base Core to Al decision-making

- 4. Intelligence Layer Powers contextual query interpretation (Agentic RAG)
- 5. Data Integration Connects semantic search (Pinecone) and structured data (Snowflake, SQL DBs)



# 6. RAG Technical Architecture

**RAG Orchestration for Real-Time Enterprise AI** 

## Components

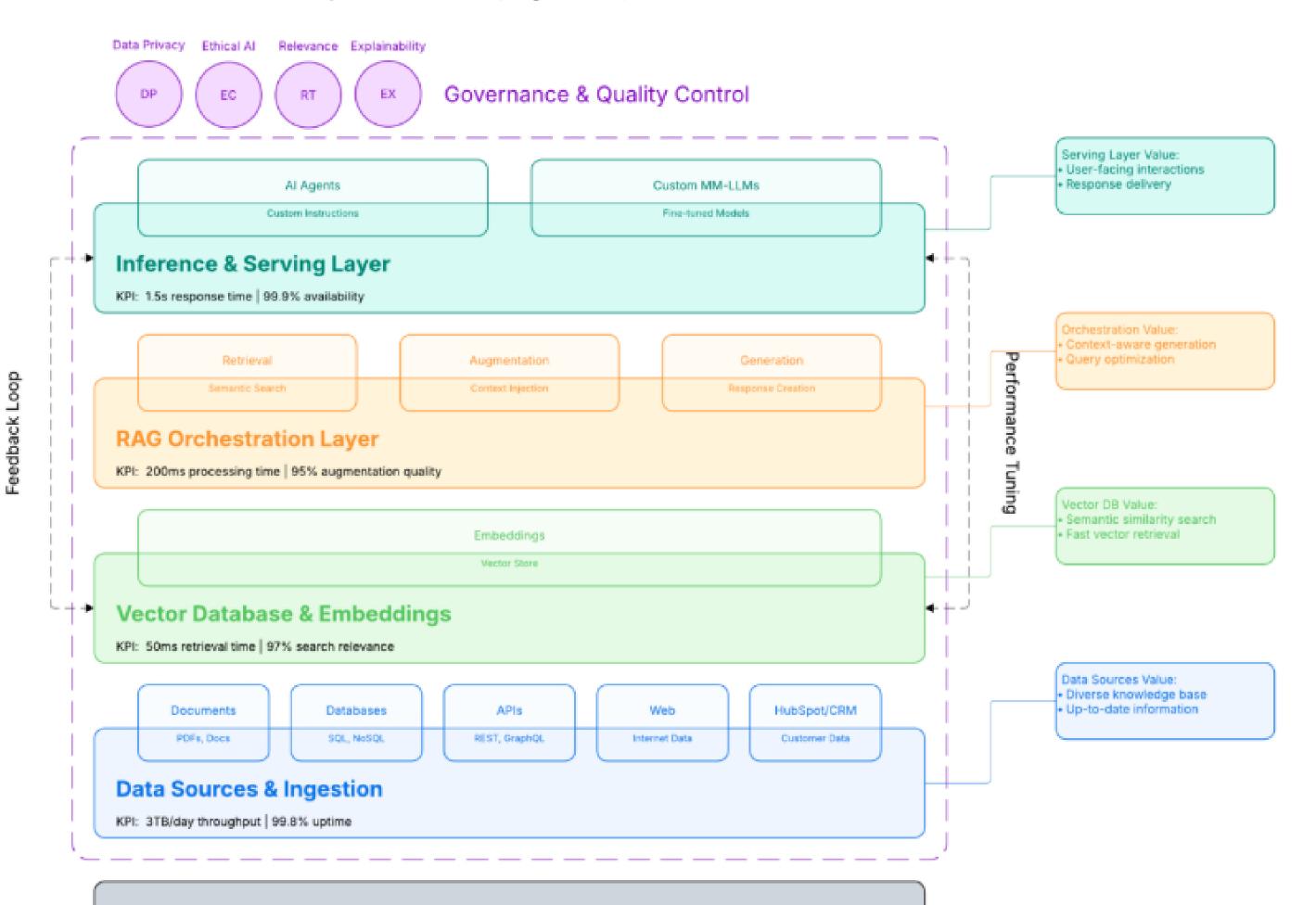
- Inference Layer: MM-LLMs, GPTs, and AI Agents
- **RAG Layer**: Retrieval, Context Augmentation, Generation
- Vector Store: Pinecone for fast semantic similarity
- Data Sources: Snowflake, CRM (HubSpot), SQL, Web, Docs

### **Governance Focus**

- Privacy, Relevance, Explainability, and Quality Control
- KPIs: Sub-1s response times, 95% augmentation quality, 3TB/day ingestion

# RAG ARCHITECTURE

Optimized for Retrieval, Augmentation, & Generation



REST gRPC Kafka SOAP

API Layer & Integration Capabilities

# 7. Summary and Next Steps

**Enabling Intelligent Transformation** 

### Summary

- Current systems are siloed and manual
- Future state enables intelligent agents, unified data, explainable Al
- RAG architecture allows scalable AI interactions using trusted data

# **Action Items**

- Finalize unified data schemas
- Define and build ETL/real-time data pipelines
- Implement Pinecone + Snowflake stack
- Pilot Al agent(s) by function