

```
< $total_pages) ? $total_pages : 1) ? $page - 1 : false;  
< $total_pages) ? $page + 1 : false;  
<=$finish_page; $i++)  
  
$i['NUM'] = $i;  
$i['ACT'] = ($i==$page) ? true : false;
```



```
$page = 1, $total_records, $finish_page;
```

# Modernization Success Story

## Building Telecom Operational Intelligence on Palantir Foundry and AIP

**Performed by:** Cadmus (delivered by **Ventura**, acquired and fully integrated into Cadmus in 2024)

**Period of Performance:** 2020-Present

**Client Organization:** Large Telecom Provider Construction & Engineering (C&E) Team

### Executive Summary

A large U.S. telecom provider's Construction & Engineering (C&E) organization launched a multi-year initiative to modernize its analytics ecosystem and shift from on-prem, warehouse-centric reporting to a cloud-based, Data Mesh-aligned model. **Cadmus** played a critical role in migrating legacy data (10+ Petabytes) and processes; re-engineering complex datastores to Snowflake; ETL from SQL Server/SSIS to scalable Palantir Foundry pipelines; and transforming the user experience from static "canned" reports to near-real-time, self-service analytics.

The program combined deep engineering (distributed transformations, orchestration, performance tuning, and operational hardening) with change management (training, adoption enablement, and a pragmatic reporting transition). Most recently, we extended the platform with AI-powered capabilities, including a GenAI-enabled conversational analytics chatbot using Palantir AIP, now in production and expanding into operational workflows.

# CADMUS

## Background & Starting Point

The C&E organization's legacy environment had grown over time and included:

- **120+ SQL Server-based legacy databases** and data marts
- **Cognos** for standardized reporting and “pixel-perfect” outputs
- **SSIS** and SQL-driven transformations (approximately **100+ SSIS jobs**)
- A reporting culture centered on **once-a-day refresh cycles** and pre-defined outputs

This model constrained agility, increased operational burden, and limited the organization's ability to ask new questions quickly. The target state was a modern platform that could support scalable processing, improved governance, decentralized ownership, and self-service insights, moving teams from “request a report” to “answer it yourself,” with fresher data.

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## Goals

Cadmus aligned delivery to the C&E organization's modernization goals:

- Define **data governance standards** (quality, ownership, lineage, and consistency)
- Modernize data transformation design and build using **PySpark**, **Snowflake**, **Snowpark**, and **SnowSQL**

“ A defining constraint early in the program was the lack of meaningful vendor support during implementation.

- **Democratize access** to data and analytics tools
- Improve data quality through **decentralized data ownership** aligned to Data Mesh principles
- Enable collaboration and **self-serve analytics** through structured training and documentation
- Move from **static, once-a-day reporting to near real-time, self-service analytics**

## A Major Challenge We Overcame: Minimal Vendor Support During Implementation

A defining constraint early in the program was the lack of meaningful vendor support during implementation. When we started in 2020, there was limited structured training and limited hands-on guidance for how to translate a complex legacy data environment into Palantir Foundry at scale.

We closed this gap through deliberate internal capability building:

- **POC-led upskilling:** targeted proofs-of-concept to validate ingestion patterns, transformation approaches, orchestration, and performance strategies before scaling
- **Self-training and internal playbooks:** codified repeatable engineering standards, templates, and runbooks
- **Pattern-driven migration:** standardized

“ This approach reduced vendor dependency and created a sustainable internal delivery capability for the client.

approaches for pipeline design, testing, partitioning, and production hardening

- **Enablement at scale:** “build-with-us” sessions to transition users from dependence to independence

This approach reduced vendor dependency and created a sustainable internal delivery capability for the client.

## Technical Solution and Delivery Approach

### 1) Ingestion and Data Foundation

We migrated application data from SQL Server to Snowflake, so our ingestion approach used a Snowflake-replicated copy of the application database as the primary source. Snowflake served as an intermediary/source system for multiple consumers, while Foundry became the core platform for processing and self-service analytics enablement.

Key actions included:

- Ingesting Snowflake and other sources into the platform with consistent dataset contracts
- Curating governed datasets suitable for downstream analytics and operational use
- Establishing governance-aligned dataset standards (schema discipline, ownership, documentation, and lineage)

### 2) ETL Modernization: SSIS/SQL to Spark-Based Pipelines

The largest technical lift was re-engineering legacy

transformation logic originally implemented in SSIS and SQL into a cloud-scale execution model. We rebuilt transformations primarily using **PySpark**, and leveraged **Snowpark/SnowSQL** where Snowflake-native processing was the best fit.

Core engineering outcomes:

- Re-implemented hundreds of complex **SSIS jobs** and related SQL logic as scalable pipelines
- Rebuilt orchestration and scheduling with dependency management, automated runs, and production monitoring
- Designed pipelines for reliability: idempotent runs, modular stages, and consistent failure handling

This was not a lift-and-shift. It was a re-platforming effort that changed how transformations executed, scaled, and were operated.

### 3) Data Mesh Architecture

With the investment in modern data architecture, the client upgraded data operations standards. A **Data Mesh** approach provided the framework to deliver features quickly at scale by treating curated datasets as governed, reusable **data products** owned by domain teams.

As part of the Data Mesh modernization, Cadmus built foundational capabilities that made the platform scalable, trustworthy, and easier to adopt:

- **Data Contracts:** Defined dataset contracts so consumers had a consistent view of schema, definitions, and expected refresh frequency.

“ This was not a lift-and-shift. It was a re-platforming effort that changed how transformations executed, scaled, and were operated.

- **Access Control:** Implemented role-based and attribute-level access so consumers could access only the data and fields they needed.
- **Health Checks:** Strengthened data quality controls and observability across critical pipelines through automated checks and actionable signals.
- **Discoverability:** Enabled consumers to easily discover the feeds and attributes they needed, reducing reliance on tribal knowledge and accelerating onboarding.

These capabilities reduced operational drag and made self-service analytics viable at scale.

#### 4) Near Real-Time Enablement and Operational Readiness

A key objective was moving from once-per-day reporting to near real-time analytics. We designed ingestion and transformation pipelines to support faster refresh cycles and operational use cases, including:

- **Near real-time data loading** patterns for timely insights
- **Operational alerts** and exception monitoring
- Foundations for **predictive modeling**
- A **GenAI-powered chatbot** that reduces friction for common analytic questions

To accelerate delivery and iteration speed, we leaned on Foundry's developer tooling:

- **Code Workbooks** for rapid development and testing of transformations
- **Repositories** for versioned, collaborative

engineering workflows

- **DIY Syncs** to quickly synchronize, validate, and evolve datasets across environments

To improve confidence and operational maturity, we leveraged:

- **Lineage** to provide transparent, auditable dependency tracking end-to-end
- **Health Checks** to strengthen data quality controls and observability across critical pipelines

#### 5) Performance Engineering and Production Hardening

Once end-to-end flows were established, we focused heavily on performance and stability:

- Partitioning and layout optimization for large bucket-file datasets
- Join/shuffle tuning and refactoring "hot path" transforms
- Monitoring, operational runbooks, and predictable release processes

In a high-volume telecom environment, even small delays or instability quickly erode confidence. We focused on performance and reliability, so the platform stayed fast, dependable, and trusted for day-to-day decision making.

## Self-Service Analytics Transformation and Change Management

### 1) Contour Enablement: From Canned Reports to Self-Service

The client's intent was not only data processing modernization, but also self-service analytics adoption. Foundry's **Contour Dashboards** were central to shifting behavior.

Our adoption approach:

- Built an initial set of Contour assets aligned to high-value reporting needs

“ In a high-volume telecom environment, even small delays or instability quickly erode confidence.

“ Users gained control without requiring developers for every new insight, and teams moved closer to real-time, decision-support analytics.

- Conducted working sessions and hands-on training so users learned by building
- Provided time-boxed support for teams as they transitioned to self-service
- Over time, the environment grew to **hundreds of Contour lineages**, reflecting broad user-driven analytics creation and reduced dependency on developers

## 2) Power BI as a Practical Bridge to Overcome Resistance

Contour adoption faced early resistance, especially from teams used to familiar reporting tools. We avoided disruption by introducing **Power BI as a transitional layer**, providing continuity for key stakeholders while training and adoption matured. As self-service usage increased, reliance on Power BI declined and the organization shifted to Foundry-native analytics.

## 3) Quiver for Flexible, Intuitive Exploration

To improve usability and analytic flexibility, we also used **Quiver** alongside Contour. This strengthened self-service by giving users more intuitive exploration and interaction patterns while maintaining governance.

Result: users gained control without requiring developers for every new insight, and teams moved closer to real-time, decision-support analytics.

## Lightweight Operational Use Cases in Foundry

In addition to analytics, the client used Foundry for small-scale operational workflows that benefit

from governance and traceability but are not high-concurrency transactional systems, including:

- **Reference data management**
- **Metadata management** and controlled edit workflows

These solutions fed downstream systems while maintaining auditability and controlled updates.

## AI/GenAI Innovation: Conversational Analytics with AIP

Most recently, we implemented conversational analytics using **Palantir AIP**:

- Delivered as a POC and moved into production
- **Live in production ~3 months**
- Enables users to ask questions in natural language and receive governed, data-backed analytic responses grounded in enterprise datasets and semantics

We are actively expanding AIP to integrate conversational analytics and predictive insights into operational workflows, especially around **project health monitoring**, so insights show up where work happens rather than staying trapped in dashboards.

## Outcomes and Impact

This engagement delivered transformation across platform, operations, and user behavior:



### Explore Our People Driven, AI-Empowered Approach.

Learn how we help unlock the power of your data, providing intelligent solutions that drive innovation and operational excellence.



### Engineering and Platform Outcomes

- Modernized a complex legacy footprint including **120+ databases** and **100+ SSIS jobs** into scalable, production-grade pipelines
- Improved operational trust through lineage transparency, health checks, and observability-driven support practices
- Enabled near real-time patterns for analytics and operational readiness (alerts, predictive foundations, AI enablement)

### User and Adoption Outcomes

- Shifted teams from stale daily reports toward **near real-time, self-service analytics**

- Reduced dependence on centralized reporting development through Contour/Quiver enablement and training
- Used Power BI strategically as a bridge to maintain continuity and accelerate adoption rather than forcing disruption

### Innovation Outcomes

- Delivered a GenAI-enabled conversational analytics capability using AIP (production and expanding)
- Established a foundation for embedding predictive and conversational insights into operational workflows (project health monitoring and beyond)

## Why This Past Performance Stands Out

This program reflects the full arc of enterprise modernization done the hard way—and done right:

- Deep technical migration (SSIS/SQL – PySpark/Spark and cloud-native processing)
- Data Mesh-aligned governance and decentralized ownership enablement
- Near real-time and operational analytics readiness (alerts, predictive foundations)
- Real change management at scale (Contour adoption, Power BI bridge, training and documentation)
- AI/GenAI enablement in production (AIP conversational analytics)
- Execution resilience (minimal vendor implementation support, capability built through POCs and self-training)

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