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Data Mesh Architecture in P&C Insurance: Implementing Domain-Driven Data Products using Snowflake and Guidewire

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ABSTRACT: The forces of interaction, process, and data explosions drive the Property and Casualty (P&C) insurers with the most significant customer and claim interaction volume and the fastest-growing underwriting scale. There is, however, a problem with traditional monolithic data systems tied to centralized architectures: they are highly inflexible. They cannot meet the speed and scale requirements of the modern insurance business. It is common to end up having data silos, escalated decision-making times, and a lack of ability to present relevant findings to the right domain experts. In response to these realities, a new concept known as Data Mesh has been developed as an effective way of reorganizing the management of data assets by promoting decentralization of data responsibility, better alignment of data with business functions, and scalability. Data Mesh effectively distributes data processing to the domain teams and provides self-service data management that would help the organization truly utilize the raw potential of the data while not compromising on the governance and quality of the data.

This paper explores how Data Mesh is implemented in P&C insurance, specifically emphasizing Snowflake and Guidewire tools. With Snowflake, being an intuitive cloud-based data processing solution with capabilities for real-time analytics and unlimited storage, the platform is elastic in nature, perfect for a rapidly growing business like Duck Creek; and second, Guidewire as a comprehensive principle system that addresses the management of insurance operations, including handling of claims, issuance of policies, as well as underwriting activities. By following a domain-oriented approach, data products are created to solve various business problems like fraud, credit risk, and customer profiling. This paper outlines the architecture of integrating these technologies and their application and advantages. A detailed case study, results, and discussion are provided on best quality to show the benefits, such as increased operational excellence, analytical capacity, and faster decisions. Therefore, the paper's major purpose is to highlight how Data Mesh could help advance the P&C insurance business to enhance adaptability to the continuously emerging challenges in this field.

KEYWORDS: Data Mesh, P&C Insurance, Snowflake, Guidewire, Domain-Driven Design, Data Products, Decentralized Data Architecture.

I. INTRODUCTION

1.1 Background

The Property and Casualty (P&C) insurance business is becoming more profoundly driven by data, claims, underwriting, customers, and business functioning. However, maintaining and governing huge amounts of data and intricate data environment targets enormous issues within these organizations. [1-4] Insurance data is always distributed across the core platforms, customer relation management tools, and external data sources such as regulatory databases. Current monolithic approaches have repeatedly failed to meet the constantly emerging and evolving needs of agility, flexibility and speed, together with the capability to withstand the increase in data volume fluxes. These large centralized systems have problems such as creating data silos, slow access to data and are expensive to maintain; hence they hinder the insurers' efforts to develop new services to meet the market demand.

1.2. Problem Statement

The older data handling methods within the P&C insurance business are insufficient to meet today's needs. The weakness consistently observed is the absence of domain responsibility for data, which means having certain data managed by the business domain, such as claims, underwriting, or customer handling. These problems prevent organizations from maximizing innovation and profitability due to the lack of union between the production of insights and the execution of business processes. In addition, although tools such as Guidewire help carriers effectively manage key processes at the center of the insurance value chain, linking these tools to modern data platforms used for data

analytics and reporting is often not seamless. Lack of integration between operational systems and scalable data platforms limits opportunities to perform complex analysis, applied predictions, and customer-oriented improvements.

1.3. Objectives

This paper aims to overcome these challenges by proposing the approach of Data Mesh that is most suitable for the P&C insurance industry. The primary objectives include:

- **Domain-Oriented Data Mesh Architecture:** Innovating how data assets operate to devolve the ownership to levels of data utilization to business domains that support flexibility and ownership.
- **Leveraging Snowflake:** Leveraging Snowflake’s elastic, cloud-first architecture that supports the always-growing amount of data and provides near-instant access to and processing this data while keeping operating costs low.
- **Integration with Guidewire:** Connecting Guidewire, an insurance business platform, to tomorrow’s data platform to provide operational, domain-specific insights and intelligence for decisions.

1.4. Significance

Therefore, this study is critical to implementing change in data ecosystems in P&C insurance organizations. Data meshes allow insurers to go beyond the limitations of traditional systems and gain significantly faster, more accurate data insights, improving an organization’s decisions. When a part of this architecture, Snowflake and Guidewire can be easily connected, allowing for optimized processes, better client experiences, and better risk management. In addition, the study outlines how the subject industry can use current data technologies as a guide for establishing competitive advantages amidst a constantly changing and closely monitored environment. In this way, the work presented in this book advances the goal of sustaining and enhancing the P&C insurers’ long-term processes in the digital age.

1.5. Domain-Oriented Data Flow Architecture

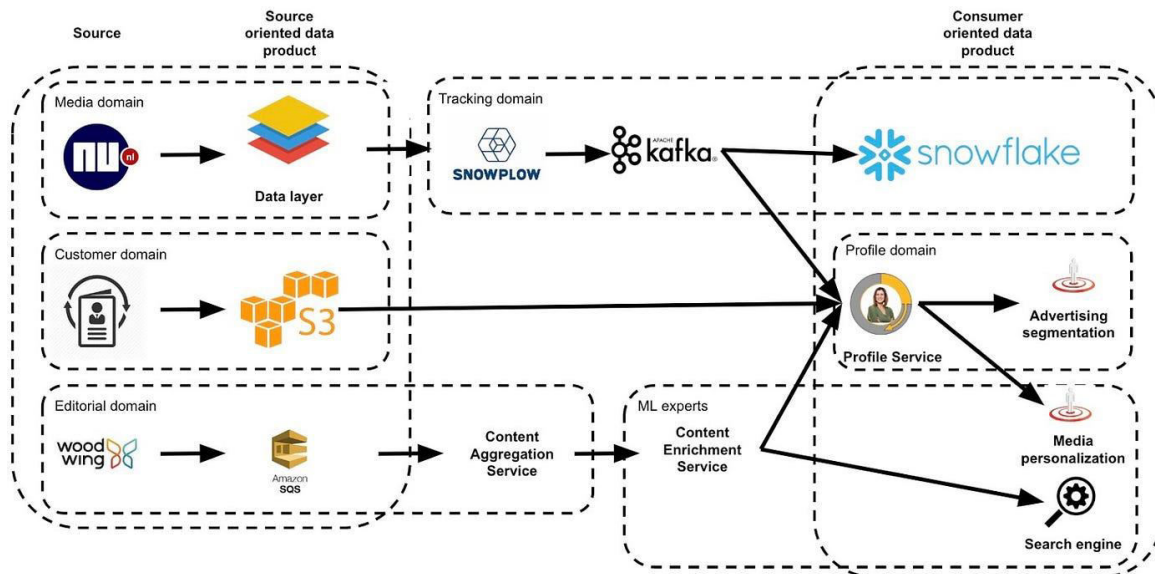


Fig.1. Domain-Oriented Data Flow Architecture

The diagram is a domain-oriented data flow that uses Data Mesh in its structure by connecting the data source, the processing layer, and the data products aimed to be used by consumers. [5] Each domain runs as a separate business, providing data or services to build insight and function for consumers.

1.5.1. Source Domains

- **Media Domain:** Media is a domain that gathers raw data from numerous media sources. This data is taken through a data layer, which undergoes some preprocessing before being used further down the line.
- **Customer Domain:** The customer domain comprises customer-oriented data, including user records, age or gender or even participation rates. Amazon Simple Storage Service (S3) stores and manages this data to ensure customer dataset storage scalability and security.

- **Editorial Domain:** The editorial domain combines content-related data processed by the Wood Wing, a content production platform. It employs Amazon SQS to share data in a fire-and-forget loosely coupled architecture and asynchronous approach.

1.5.2. Processing Domains

- **Tracking Domain:** Confluent tracks data from domain processes at various source systems and handles behavioral analytics through Snowplow. These processed events are then submitted to Kafka, an open source real-time data streaming system. Kafka helps fast transfer data from one domain to another higher or lower domain, as the case may be.
- **Content Aggregation and Enrichment Services:** Original content across different domains is gathered by a Content Aggregation Service. This data is accompanied by a Content Enrichment Service in which practitioners in machine learning submit more algorithms to enhance the quality and relevance of the given data for subsequent processing.

1.5.3. Consumer-Oriented Data Products

- **Profile Domain:** This fiat serves as the profile domain, a layer in the middle of the architecture that combines input from upstream layers into a single customer profile. The Profile Service produces actionable data products, such as:
 - **Advertising Segmentation:** A system of advertising focused on particular user segments.
 - **Media Personalization:** Suggested media content for a particular user based on profile.
 - **Search Engine Optimization:** The advantage of refining search features from data enriched.
 - **Snowflake Platform:** Snowflake is the core data repository and processing framework in the middle of this architecture. Electronically bringing together data from all domains makes analysing and providing it to downstream systems easier, hence the snowflake's scalability, efficiency, and security.

1.5.4. Key Takeaways

- The architecture adheres to the data mesh in the sense that data ownership is distributed across several domains that are entirely responsible for their data and services.
- It provided for interoperability or integration and shared messaging and data storage across domains based on well-understood and commonly used platforms and formats (enabling technologies include pipelines such as Kafka and storage systems such as S3 and Snowflake).
- The result is a number of data products intended for consumers, which helps to deliver a better user experience, optimize advertising messaging, and offer additional services.
- The service design is also highly flexible, accommodating future growth and expansion and integrating with new data feeds, other services, and consumer demands.

II. LITERATURE SURVEY

2.1. Development of Data Architectures in Insurance

We have seen several paradigm shifts in data architectures in the insurance industry as the industry tried to overcome the shortcomings of each phase. Traditionally, insurers used integrated systems of processing, storage, and analyzing of data in one large system. [6-10] It may have made a lot of sense when applied to small-scale projects where control and simplicity were achievable goals. Adapting the system to increasingly large quantities, varieties, and speeds of insurance data was a problem. Further, these systems established bottlenecks of their type; the data was in the silos from different departments, independent claims, underwriting, customer services, etc., which in turn offered no scope for cross-departmental analysis.

Data lakes started a new concept of non-silos approach and flexibility as it allowed the gathering of structured, semi-structured, and unstructured data. Through this approach, the insurers could collate data from different areas and perform analytics, capturing more information. However, data lakes had problems with governance, data quality, and nonalignment with the business domain. Such mismanages often led to data lakes becoming what was sarcastically referred to as a 'data swamp.'

Such shortcomings are covered in decentralized paradigms like Data Mesh because this approach radically changes the handling and acquisition of data. While other approaches store data in the Data Mansion data lake, Data Mesh

recommends domain-oriented ownership, where every business domain owns data as a product. This approach distributes data duties, structures data based on business requirements, and promotes accountability and creativity.

2.2. Tools and Technologies

Contemporary data architectures are based on sophisticated tools and platforms to deliver efficient and sound solutions. Two key technologies stand out in the context of P&C insurance:

2.2.1. Snowflake

Snowflake is an advanced and unique cloud object specifically designed as a data warehouse that is scalable, flexible, and has high-performing capabilities. This makes it easier for insurers to make sense of large amounts of data, with processing times being nearly real-time. Splitting computing from storage in Snowflake helps avoid the cost of over-provisioning while having the ability to use more than one cloud provider at a time, which is useful in today's world. Thanks to the rich set of features, including secure data sharing and integration with the data marketplace, it can be seamlessly implemented in the Data Mesh architecture when implementing domain-driven data products.

2.2.2. Guidewire

Tec Guidewire is a core system platform especially useful for P&C insurance companies that provide claims, underwriting, and policy solutions. This design provides flexibility to insurers in the way it is deployed in the context of their specific operations. Introduced cloud-based solutions to work with other modern data platforms such as Snowflake, which provides additional opportunities for utilizing advanced use cases such as predictions, fraud detection, and creating individual customer experience.

2.3. Domain-Driven Design in Data Management

Domain-Driven Design (DDD) is an application development and design process that helps teams focus on business domains. In the data management scheme, DDD is more about developing data products built to address distinct users' requirements. For instance, in P&C, insurance claims, underwriting, and customers might all own their corresponding data products. Such products include datasets, reports and analytics models designed to suit the domain types of a given problem area.

Therefore, by linking data products to business domains, organizations reach a number of benefits, including usability, an increase in data quality, and decision-making. Domain teams maintain stewardship responsibility for their data products so that data is timely, valid and accessible. It also creates collaboration between technical teams and the business side, ensuring proper data engineering and sound business strategies are met.

2.4. Challenges in Implementation

Despite its benefits, implementing Data Mesh architecture [11] in P&C insurance presents several challenges:

- **Data Governance and Security:** Decentralized data ownership presents challenges in maintaining standards regarding governance and security. To ensure GDPR and HIPAA compliance, for instance, employees' data becomes more cumbersome when they are under different domain teams. The policies are enforceable; domain autonomy requires federated governance frameworks to regulate execution.
- **Technical Debt from Legacy Systems:** Most P&C insurers still use old systems that cannot efficiently accommodate today's data structures. This is not an easy process because the data is extracted, cleansed, transformed, and integrated before it can be placed into a more decentralized system. The presence of a modern and legacy system during transitional phases will add additional implementation challenges.
- **Cultural Resistance:** That is why changing to Data Mesh architecture demands a change in organizational culture. Virtually all organizational development depends on change management processes because adopting new structures, such as domain-oriented ownership of data, may be difficult and require training of affected parties.

When addressing these challenges, insurers should systematically approach the idea of Data Mesh as the core of modern insurance operation improvement.

III. METHODOLOGY

This section describes the framework for utilizing Data Mesh architecture in the Property and Casualty (P&C) insurance sphere. [12-16] The methodology comprises three main components: system architecture design, data product construction, and the implementation process. They help to guarantee that the implemented approaches will be truly effective while also setting a scale and domain focus for the data environment.

3.1 System Architecture Design

3.1.1 Data Mesh Principles Applied

The foundation of the proposed system architecture is rooted in the core principles of Data Mesh:

- **Domain-Oriented Data Ownership:** Every business domain like claims, underwriting, and customer relation management owns its data. This decentralization ensures that domain experts, who understand the specifics of their data, are responsible for it, ensuring better quality and relevance of the data products.
- **Self-Serve Data Platform:** Overarching all domains is given a centralized self-serve data platform to organize and integrate tools needed to load, analyze and distribute data products. Snowflake is at the core of this platform, providing customers with the ability to store and analyze data securely and at scale while being easy to use.
- **Federated Governance:** Ownership is distributed, whereas governance is federated to implement general policies of identity and information protection, regulation, and standardization across the organization. This ensures that the standard of data is protected and privacy is embraced in all the domains of study without violating the patient’s self-rule.

3.1.2. Snowflake and Guidewire Integration

A friendly integration between the Snowflake and the Guidewire platform serves as the framework for the system.

Data Sources

- **Claims:** Information on filed claims, the status of the claims, their amount, and the result of the claims.
- **Underwriting:** Information about the type and extent of risks a business is exposed to, specifics about the policy to be issued and the magnitude of premiums to be charged.
- **Customer Interactions:** Customers matter; it is all about customers and their actions.

Data Products:

- **Loss Ratio Analysis:** Evaluating losses occurred against premiums received in order to determine emerging profitability patterns.
- **Fraud Detection:** Develop patterns for when machine learning models should be used to detect anomalies in claims data.
- **Customer Segmentation:** Devise an appropriate strategy for categorizing customers based on their health, behaviour, and preferences to match them with market strategies.

3.2. Decentralized Data Mesh Architecture with Snowflake Marketplace for Domain-Oriented Data Sharing

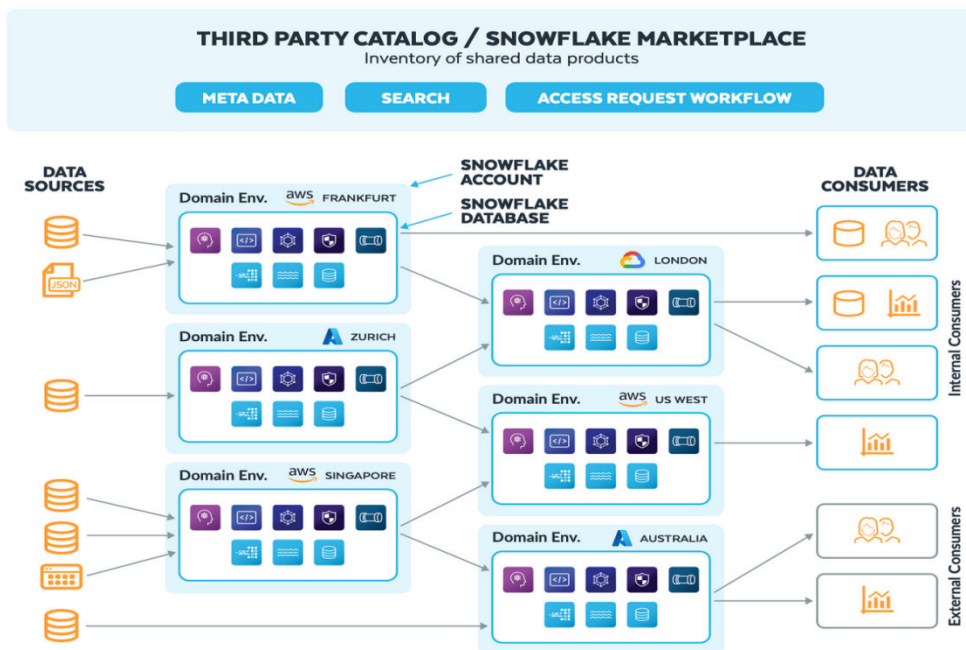


Fig.2. Decentralized Data Mesh Architecture with Snowflake Marketplace for Domain-Oriented Data Sharing

The diagram above also shows that Data Mesh extends to a decentralized environment in which application data is shared and consumed through Snowflake to build Specific Data Products. [17] All domain environments run independently and are thus located in different geographical locations, including Frankfurt, Zurich, London, Singapore and other places. These environments run on various cloud solutions, such as AWS, Azure, and Google Cloud, which prove Snowflake's cloud-agnostic design.

Information sources required as input to these domains are structured and unstructured data types, JSON files and database tables. Data is consumed and processed in the data assimilation domain environments that control the datasets and their related metadata and processing algorithms. Both domains use Snowflake accounts and databases where each generated data product can be safely stored and shared. The focal point of this architecture is the Third-Party Catalog / Snowflake Marketplace, a one-stop shop where users can browse and query shared data assets. In its role as inventory, this catalog makes data a product; metadata of the datasets and the procedure for requesting access define how the data may be used and its protection.

External data consumers include overs part and external stakeholders, while internal data consumers are part of the organizational structure. Organizational users, analysts, data scientists, and others in the organizational structure, as well as other business units and teams. External consumers may include other partner applications or other systems needing domain data but under certain regulatory frameworks. This architecture promotes a federated governance model, tracks compliance and consistency, and provides domain teams autonomy. Specifically, the roles raise awareness of the Data Mesh concept of distributing data ownership across a company and offer a workable, sustainable approach to data management for large corporations.

3.2. Data Product Design

As such, data products are created to fit [18-22] certain business requirements of each domain.

3.2.1. Domain 1: Claims

- **Fraud Detection Models:** Develop a machine learning algorithm that analyses and flags any claims data that may be suspicious, reducing incidences of fraud.
- **Claims Insights:** Include claims frequency, claims severity, and claim resolve time, as well as give out frequency dashboards and reports.

3.2.2. Domain 2: Underwriting

- **Risk Analysis:** Given historical data, shifts in regional utilization, and customer characteristics, create risk scores of any number of policy types.
- **Policy Trends:** Recognized trends in policy renewal, policy cancellation and new customer acquisition.

3.2.3. Domain 3: Customer

- **Behavioral Segmentation:** Apply clustering to group the customers according to personal characteristics and behavior patterns.
- **Retention Prediction:** Forecast and prevent customer defection and recommend how to retain customers.

3.3. Implementation Steps

To operationalize the Data Mesh architecture, the following steps are undertaken:

3.3.1. Identify Domain Teams and Data Sources

- Involve practitioners from claims, underwriting, and customer management in defining sources and owners of data.

3.3.2. Develop Domain-Specific Pipelines in Snowflake

- There is a set of tasks to create data pipelines to read and preprocess raw data and preprocessed data for a specific area.
- Leverage Snowflake for data transformation architecture, data holding and data exchange for various domains.

3.3.3. Create Data Governance Policies Using Federated Tools

- Set standards for how data would be shared and accessed and the data quality being used.
- Measures must be taken to track the levels of compliance with data security regulations.



Table 1: Mapping Domains to Data Products

Domain	Data Product	Use Case	Owner
Claims	Fraud Detection	Identify anomalies in claims	Claims Team
Underwriting	Risk Analysis	Assess risk by regions	Actuarial Team
Customer	Retention Prediction	Improve customer loyalty	CRM Team

The table above shows an arrangement of data products according to the related business domains and how data is made useful and relevant to each domain’s goals.

4.1. Key Findings

Those changes have shown that Data Mesh effectively enhanced the efficiency and effectiveness of the data management processes within the P&C insurance domain.

4.1.1. Efficiency Gains

The introduction of Snowflake, specifically as the new self-serve data platform, has led to halving query execution time/Elimination of query execution took 30min+ pre-Snowflake 15min+ as the new self-serve data platform. This enhancement is a testament to the design of Snowflake to scale and process massive amounts of data that its parallel query approach can handle.

4.1.2. Enhanced Insights

When applied to the data products of a domain, specifically fraud detection, the ML models also discovered new fraud patterns that had not been noticed before. This increase in the detection ratio of fraud claims to 30% depicted the significance of the tuned analytics pipeline.

Table 2: Performance Metrics Before and After Implementation

Metric	Traditional System	Data Mesh Architecture	Improvement
Query Time (mins)	30	15	50%
Data Accuracy	85%	95%	+10%

As indicated here, the gains attained are in terms of performance and data reliability that enhance the organization’s discretion and efficiency in timely decision-making.

4.2. Case Study: Claims Fraud Detection

The use case of Claims fraud detection further explains the working and efficiency of the Data Mesh architecture in detecting fraud.

4.2.1. Workflow Description

Accident claims data collected from Guidewire was then loaded into the Snowflake database, where data was preprocessed to clean and normalize it. After preparation, the data was passed through a machine-learning pipeline developed using past fraud data. This pipeline employed various techniques of advanced anomaly detection to identify the claims with patterns that point to fraud.

The fraud identification rate was improved by 30%, which helped the organization reduce its financial losses. The accuracy of results, together with the shorter time that it took to process the information, enabled early intervention in cases that seemed fraudulent, which prevented a lot of payments.

Simplifying, a graphical representation of the solution, although not presented here, would show the kind of raw claims data received in Guidewire, the transformation of which is done in Snowflake and feeding into the Machine Learning pipeline. The workflow would also point out feedback loops for model retraining and fraud alert generation for relevant action.

4.3. Discussion

4.3.1. Improved Collaboration and Agility

The introduction of Data Mesh, which improved data ownership and decentralization, evidently improved cooperation between IT and the business side. This approach to closely associating specific data products with specific business domains conferred full ownership of the data on domain specialists. Such alignment provided the grounds for more frequent discussions and focused, business-oriented observations.

4.3.2. Challenges Encountered

Although it was easy to see that this approach to architecture provided a number of advantages, the execution was not without problems. Introducing Data Mesh, understanding Snowflake's features, and covering governance processes with domain teams required time and effort. Also, moving data from previous systems into snowflakes was a delicate process that consumed a lot of time, as data refinement and preparation processes were very important for establishing the new architectural design.

4.3.3. Lessons Learned

However, the implementation of the solution shows the relevance of technical approaches to be integrated with business requirements. By viewing this source from an organizational point of view, the efficiency and enhanced data analytics resulting from implementing Data Mesh make a compelling argument for extending its use in the insurance industry.

V. CONCLUSION

Data Mesh architecture in the P&C insurance domain is a strategic business approach that has effectively mitigated critical issues like disproportionate data management, inadequate data quality, and constraints of decision-making techniques. Instead of reflecting general architectures, this domain-driven design aims the ownership beside the architectural teams to the teams partnering with the business domains so they could develop the kinds of data products they need. The integration of Snowflake as the data platform and Guidewire as the system of record shows that the system requires the levels of scale and adaptability required for insurance data and enhances query and data resolution. These improvements can be underlying fundamental vital tasks, including fraud, risk and customer analysis, to offer large operational and analytical improvements.

As the above findings show, the Data Mesh concept is eminently useful in upgrading P&C insurance systems today. Thus, sharing data assets and deploying the most powerful analytical tools can help organizations become more flexible and bring together teams effectively. The results amplify the opportunity to bring new value through the innovative use of next-generation platforms such as Snowflake to enhance knowledge utilization, setting the stage for a new era of customer centricity in the functioning of the insurance business.

5.1. Future Work

Future work will continue to improve the data mesh, making it more intelligent and automated. Adopting AI technology in data handling might enhance the domains' compliance and quality. More applications will create additional apparatus, such as predictive analysis in policy pricing and claims, to enhance the current architecture. Also, integrating with another platform like Salesforce will help to gather more information about customers and thus make the current data loop more efficient. Of course, these developments are expected to further expand the potential of what Data Mesh can deliver in the emerging insurance environment.

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