## atlan

Data Governance for the Lakehouse Era: Community-Led and Bottom-Up

Prukalpa Sankar Cofounder, Atlan



# Data governance is pretty misunderstood...

## Data governance is all about control and rules

What is the first thing that comes to your mind when you hear "Data Governance"?

You can see how people vote. Learn more

Rules and Policies 30%

Having Control on Data 29%

Collaboration and Agility 23%

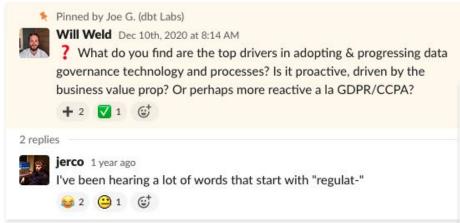
Processes or Management 19%

210 votes · Poll closed



#### **MYTH #2**

## Data governance = data security or protection





#### **MYTH #3**

## Data governance slows you down

"Dear Laura.

We've been "agile light" for a while across our IT teams. Data Governance started that transition with the project work we are responsible for, and, so far, it's working out pretty well. But what keeps tripping us up is the day-to-day questions about data, Data Quality, or just general Data Governance concerns. What do we do with those?

Concerned in California"



People see data governance as a **monarchy** with bureaucratic, ineffective rules dropped down from on high...



**Most governance programs today are ineffective**. The issue frequently starts at the top, with a C-suite that doesn't recognize the value-creation potential in data governance.

As a result, it becomes a set of policies and guidance **relegated to a support function** executed by IT and **not widely followed** — rendering the initiatives that data powers equally ineffective.





## Or they don't even know what data governance actually is





"Dear Laura.

Recently, our Data Governance leader quit. Apparently, she decided that living in Costa Rica and teaching yoga was more fun and interesting than leading a Data Governance effort, but I digress. Now I have an empty hole where my Data Governance leader used to be and within a few weeks, the whole thing has fallen apart. Nobody seems to know exactly what's happening, what we should be doing, and most disturbingly, what the value of the program was in the first place. I lead a large effort and I don't have time to fill the gap myself for governance efforts. What do I do now?

Stranded in San Antonio"



braunk 10 months ago

though Governance is such a massive area it has taken me about 2 months to wrap my head around it.

Last week, I participated in a roundtable during a conference in Paris organized by the French branch of DAMA, the data management international organization. During the question/answer part of the conference, it became clear that most of the audience was confusing data management with data governance (DG). This is a challenge my Forrester colleague Michele Goetz identified early in the DG tooling space. Because data quality and master data management embed

## I'll admit it...

# For a while, I had no idea what data governance was.



## Hi, I'm Prukalpa 👋

Lifelong data practitioner

Co-founder of at an

Pioneering the **Active Metadata** category, the "collaboration layer" for the modern data stack

Powering teams like **Postman, Plaid, WeWork, News Corp, Unilever, and Juniper** 

Named a Gartner Cool Vendor in DataOps and IDC Innovator





## We started as a data team ourselves using data science for social good











## Every day was **chaos**. I didn't know this had anything to do with **governance**.



#### **Data discovery**



Shilpa, Data Scientist Hey @richa I made a request for the data 14 days ago. Any ETA on when the team will share it?

## #team-frontend

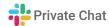
### **Data visibility**



#### Carson, Data Engineer 7.27 AM

@hanna @richa @carson The dashboard widget is not rendering because half the data is in DD/MM/YYYY format while the other is in YYYY-MM-DD. There is also data missing for 721 geographies. Not sure what to do:/

### Human tribal knowledge





Hanna, Data Analyst

@shilpa What does variable column\_xy881 stand for in the data set sales mm blr 2919.csv? Can you please clarify?

#### Data governance





Richa, Project Manager 1:55 PM

@shilpa Please ensure that analysts only get access to the data for the geography they're working on. The client is very cautious about sharing PII data!

The dreaded question... 😓

## "That number doesn't look right..."

## That's how we started the **Assembly Line Project**.

## We tried to buy a solution.



## We Failed to Set Up a Data Catalog 3x. Here's Why.

We thought it would be easy enough to figure this out, but we couldn't have been more wrong.

Our team became **6X more agile**.

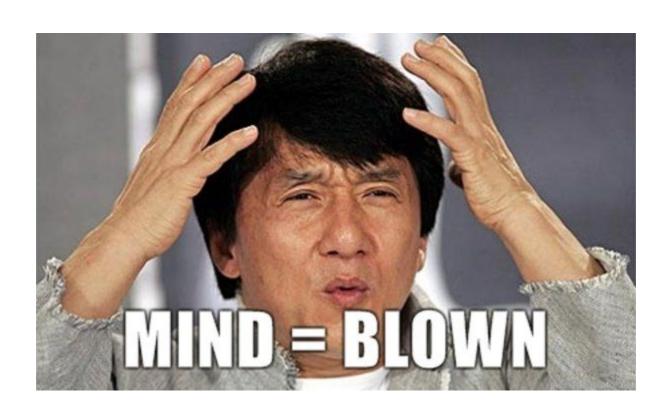


Building The World's Largest Government Data Lake - DISHA Platform

12 8
months member team

12 3.5b 42
master data dynamic data hierarchies points data connected

## After a demo someone told me, "Oh, you are a modern data governance tool!"



# Data governance is changing...

... because our world is changing



## People expect purpose and autonomy in their work

"With Generation Y coming into the business, hierarchies have to disappear. Generation Y expects to work in communities of mutual interest and passion — not structured hierarchies. Consequently, people management strategies will have to change so they look... less like the pyramid structures we are used to."

pwc

"Millennials at work: Reshaping the workplace." PwC, 2011.

-Vineet Nayar, Vice Chairman and CEO, HCL Technologies

"Today's young workers have shifted toward interests in doing valuable work and finding meaning in their day-to-day job functions.

Leaders and managers are the ones who have the power to help foster that connection of meaningful work...

There's a giant risk for employers if they don't help employees have a sense of purpose and a sense of well-being and engagement."



"The Effect of Respect." LaGree, Houston, Duffy, Shin. Sage Journals, May 2021.

## **Consumerization of enterprise**

### aka: people expect work tech to be as cool as personal tech

"It used to be enough to provide tools that improved productivity for the business in a measurable manner like enhancing processes and workflows.

Now, every tool that is used by employees must provide a world-class user experience. **Employees will not adopt tools without a memorable experience.** 

These tools and technologies will eventually be phased out of the organization because enterprise budget-holders cannot realize productivity without adoption."

**Forbes** "The Consumerization Of Enterprise Technology", Dec 2017

- ✓ Intuitive experience
- Anytime, anywhere
- Heavily personalized
- A sense of community
- Multiple modalities
- Quick and snappy
- Alive and changing

## The rise of automation in software

Hyper-automation industry forecasted to reach



by 2022, said Gartner.



#### **Robotic Process Automation (RPA)**

The automation of repeatable and redundant, rule-based human action through software bots



### **Cognitive Process Automation (CPA)**

The ability for bots to replicate decisions requiring human judgment



### Intelligent Automation (IA)

The automation of nonroutine tasks through artificial intelligence



#### Low-Code Automation (LCA)

Rapid application delivery with minimal coding and less reliance on developers/engineers

## Protecting data → Getting value from data

99%

of companies report that they are investing in data initiatives

24%

of these companies say they've actually become data-driven

49%

of these companies say they're actually driving innovation with data





# So what should the future of data governance look like?

## The evolution of governance

Governance 1.0

Governance 2.0

Governance 3.0



Monarchy

Rule from above by "the one" (king/queen)



**Aristocracy** 

Rule by "the few" (a group of elites) with input from above





**Democracy** 

"The many" get a say in the policies governing them

The rise of traditional data warehousing

Data lakes gain prominence as the architecture of choice

The modern data stack goes mainstream with key capabilities like pay as you go, elastic compute, and 30-minute quick start





databricks

**DATA GOVERNANCE 1.0** 

**DATA GOVERNANCE 2.0** 

**DATA GOVERNANCE 3.0** 



Fundamentally built for IT users, acted as a "data inventory"



"Data Stewardship" tools built for top-down governance programs



## Build a data community, not a data governance program

## Lead with your "why"

Ask your team...

What do we want our data culture to look like in 12 months?

## DELHIVERY

1.2 TB

of data ingested per day

66k

Events per second in their data pipeline

1 mil

Packages fulfilled per day, 365 days a year

#### "What do we want our team to look like in 12 months?"

- Become a fully self-organized team: all your data, learnings, experiments, and projects should be reusable, transparent, and easily accessible.
- Create an environment of trust in your data and the decisions you drive.
- Build a collaboration-first culture: everyone should feel empowered and included, despite fundamental diversity in your team.

## Rally your team around a "data product" mindset

Services Product

## Rally your team around a "data product" mindset

#### **Services**

- Start at a client problem, and engage a team to fully solve/ implement that problem.
- Get **paid by the client** for implementation.

McKinsey & Company









#### **Product**

- Build one solution that can be reused by multiple customers/users to solve a problem.
- Get **paid by customers** for **"usage"** of the product.









## Rally your team around a "data product" mindset

	Data Services	Data Product
Success Criteria	Successful implementation — i.e. did we deliver on time?	Successful usage — i.e did it solve the problem for users, and do they use it regularly?
Reusability	Single use: Build once, for use by one client	Scalability & reusability — build once, for use by many
Requirements & Scoping	Build what the customer asks you to build	Understand commonalities in problems across the customer base and build accordingly
Gratification	Instant gratification — get paid on day zero	Delayed gratification for higher rewards — takes longer to get to usage ("Product Market Fit"), but when it does, it can scale incredibly quickly
Investment	No up-front investment necessary	Up-front investment necessary (time and resources)

## Fundamental principles in treating data as product



Reusable



Reproducible



Well-documented



Accessible



Enables selfservice for end users



Scalable (built for more than one user)



Focused on impact not inputs (end user adoption)

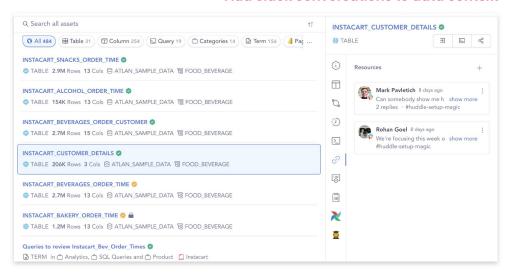
# 2 Collaboration, not control

## Embed collaboration in daily work, rather than creating another siloed tool or workflow

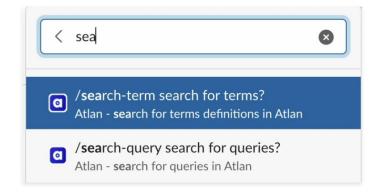


## Activate metadata to collaborate where you work all day

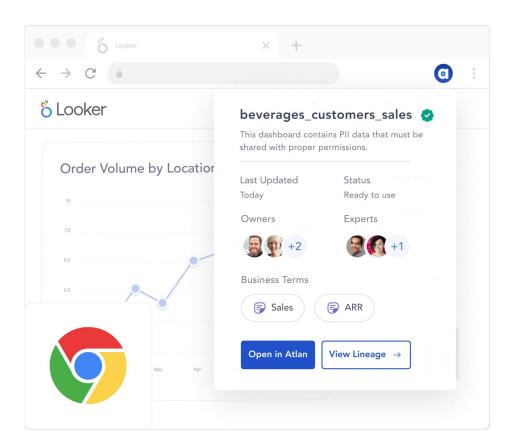
#### Add Slack conversations to data context



#### Search for metrics in Slack



## Bring context into BI to show the value of governance



# 3

# Automate wherever possible









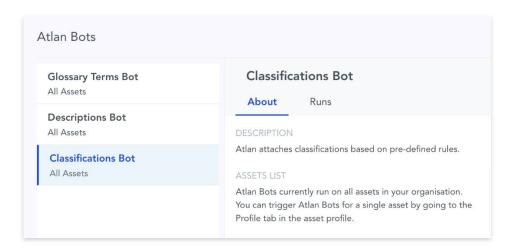
## Auto-assigning owners to assets Auto-attaching column descriptions

Delhivery deploys a bot to automatically scan query log history & custom metadata to find the best owner for every asset.

This helps developers at Delhivery drive documentation volume and standards, and reduce time-to-ownership for assets.

90%

of column descriptions were automatically deduced by the bot



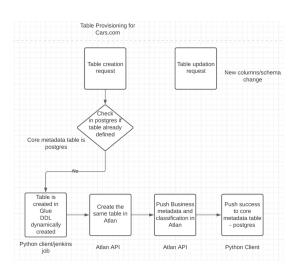
#### **Custom classifications bot**

Delhivery uses a bot to auto-classify its assets based on PII and GDPR restrictions.

**53**% of users reported saving time with Atlan as bot-automation eliminated routine tasks.



## Atlan **integrated into CI/CD pipeline** for auto-creation and enrichment of assets





#### Integrated Atlan into their development workflow to programmatically generate standardized developer documentation (automated templates) via APIs

layout	title	parent	
home	Readme Template	tools/data_catalogue/index.md	
Defa	ult Table Re	adme Template	
Depre	cated		
If the table	e is deprecated or in the	ne process of deprecation, list the status	and reason here. If not remove this section.
Introdu	uction		
		information in this table and its purpose.	1
insights	_production . items_t	y_phone_number , a fitting description mi	ght be
Rows i	n this table map pla	id specific identifiers such as item	ID to a peppered hash of a consumer phone numb
Primar	y Use Cases		
	the way the table is my	eant to be used in the context of analytics	s. For instance using the previous example table
		y_phone_number , the use case would be	

# Go from data governance to DataOps

## Every other domain in our organizations has a focused enablement function

SalesOps & Sales
Enablement

Focused on improving enablement, productivity, ramp time and success of the sales team

Sales Rep Ramp Time Win Rate

## DevOps & Developer Productivity Eng.

Focused on improving collaboration between software teams, and productivity of developers

Developer Productivity
Deployment Time

## ProductOps & Agile

Focused on improving collaboration between product, GTM teams, and customers

Velocity
Cycle Time



DataOps is a collaborative data management practice focused on improving the communication, integration and automation of data flows between data managers and data consumers across an organization.



Lean



Agile





### **Gartner**

	Learning	Applying to Data Governance
Lean Manufacturing	Centered around "value streams", and minimising waste through process mapping	How can we align to "value" for end users and the business?
🏃 Agile	Moved development process from waterfall to <i>Agile</i>	Can we ship "data products" like "software products"? How do we ship fast and involve end users in implementation?
<b>DevOps</b>	Went from siloed teams of software development (shipping software) and ITOps (maintaining software) to integrated dev proces	How do we integrate governance into how we work with data on a daily basis? How should DataOps teams be structured?

DataOps enables the rest of the organization to become data-driven.

This function doesn't actually execute data or analytics projects. Instead, it focuses on **Tools**, **Processes & Culture** that will make the rest of organization more data-driven.

### **Stakeholders of DataOps**

#### 1. Data team

Analysts, Analytics Engineers, Scientists, Data Engineers

#### Impact:

- Improve productivity of data team
- Increase time to value / speed of delivery
- Reduce ramp time of a new joinee
- Reduce attrition

#### 2. Data Consumers

Executives, Business Users, Product Managers, Compliance, Finance etc.

#### Impact:

- Enable self service
- Reduce dependencies on data team
- Improve speed of decision making

### Team work makes the dream work!

### Central DataOps Function



## **Emily Lazio**

**DataOps Enablement** 

- 1) Masters in Information and Library Sciences Understands taxonomy and structure
- 2) Children's Librarian
  Energetic, extroverted, great at bringing people together
- 3) Information Architect in WeWork's Design team Understands data ecosystem and understands user research



**Yong Lu** 

**DataOps Engineering** 

- 1) Masters in Computer Science Understands data and technology
- 2) Engineer & Data Management Leader Systems thinker, great at simplifying complex problems
- 3) Data Engineering Lead in WeWork's Engineering Internal data "guru", able to identify patterns for automation

#### All other Data Pods

#### **Data Teams**

Analysts, Analytics Engineers

#### **Data Product Managers**

Liaison between data team & data consumers

#### **Data Consumers**

Business users, etc.

Create data products for data consumers

# Make Data Governance 3.0 a reality with

atlan + 🗟 databricks

## atlan + 🗟 databricks

Data Governance 3.0 for the Lakehouse Era

- Discovery & catalog
- Data asset 360°
- -- Column-level lineage
- Embedded collaboration
- Glossary

## Attend our joint product demo with a databricks

More context, less chaos: How Atlan and Unity Catalog power column-level lineage and active metadata

(7) 11:30 am PT



We'll walk through a live product demo to show how Atlan and Databricks Unity Catalog work together to power modern data governance, cataloging, and collaboration.





**Prukalpa Sankar,** Co-founder, Atlan







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LI: Prukalpa

metadataweekly.substack.com

www.atlan.com



