

MEET THE

MODERN DATA TEAM

Successful strategies for leaders
navigating AI transformation

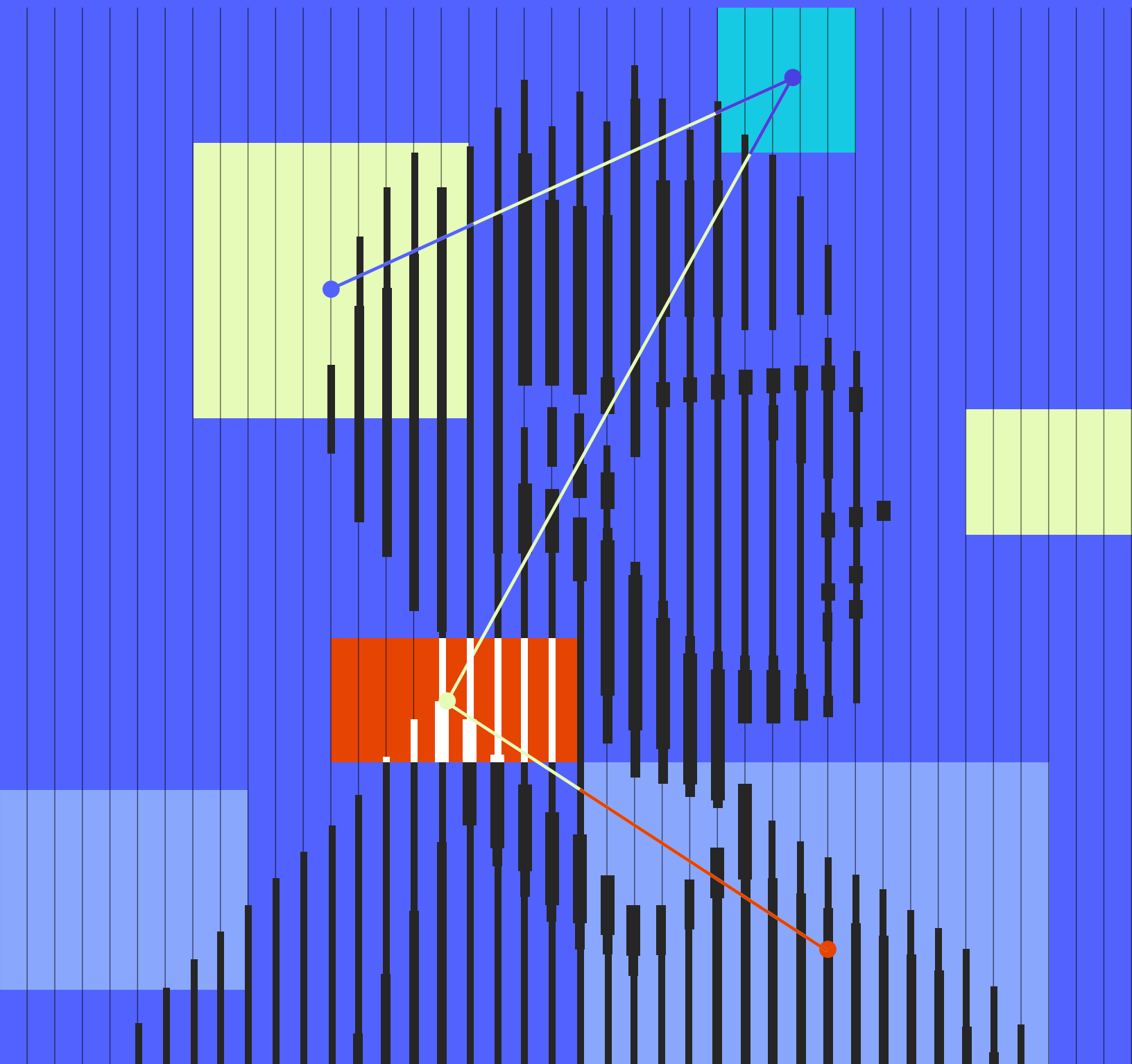


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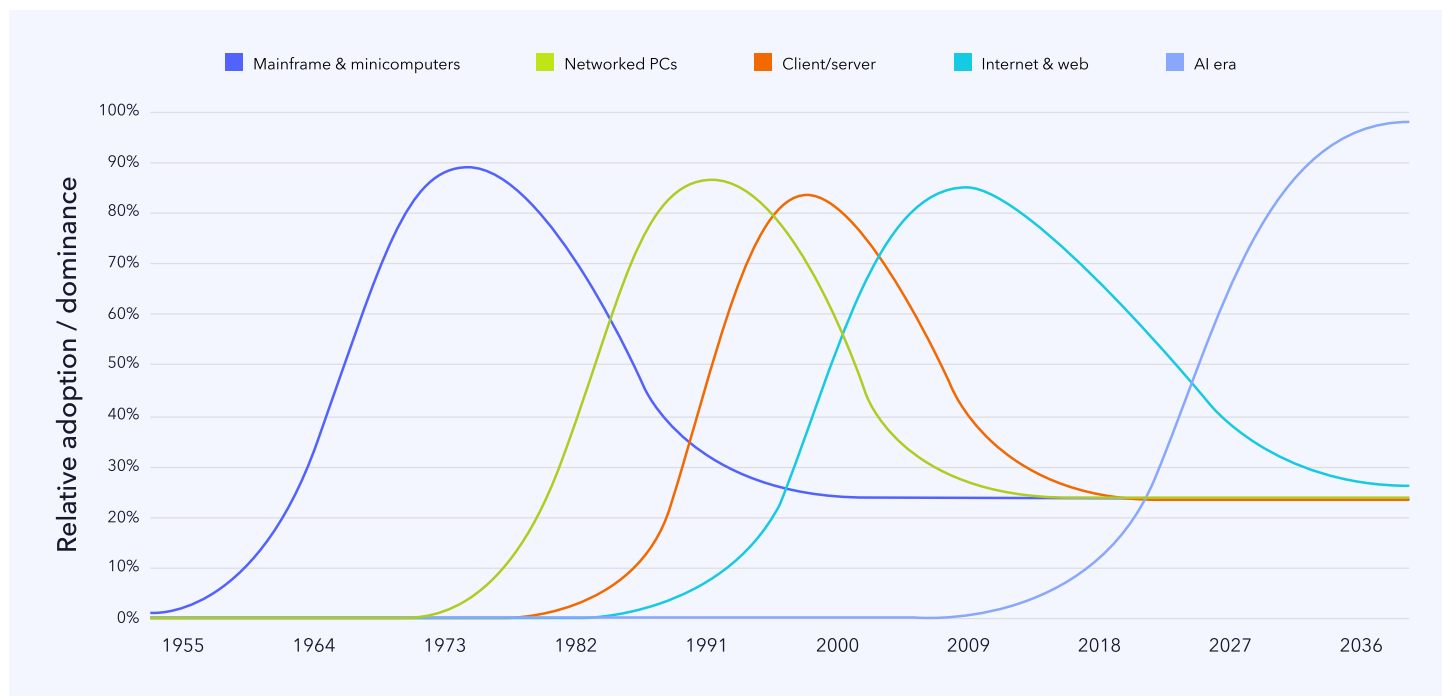
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The new dynamics of AI analytics:

Transforming people, process, and technology

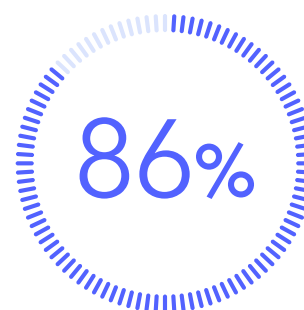
We've seen this before. AI is forcing the same reckoning that mainframes, client/server computing, and the cloud each triggered in their turn. During these times of transformation, technology takes the headlines. But the real work – and the real advantage – belongs to the leaders who develop the talent and processes to advance the strategic business objectives powered by these new platforms.



As AI ushers in a new wave of transformation in data and analytics, it's forcing a similar talent and technique recalibration on data teams and those who lead them. But this time, something is different: the pace and scale of transformation far outweighs anything that came before.

Our survey of 201 enterprise data, AI, and analytics leaders makes that clear: 86% say AI has fundamentally changed their team's core responsibilities, and 50% say it's done so to a dramatic extent.

To fulfill the promise of AI analytics, leaders must rethink their stacks, workflows, and team structure – a sentiment that's echoed in the survey data, qualitative interviews, and first-hand experiences captured in this report. From these accounts, you'll gain a clear picture of what's changing, understand the challenges and opportunities ahead, and take away a blueprint for delivering accurate analytics at enterprise scale.



Say AI has fundamentally changed their team's core responsibilities.

Key findings

01

Change brings real opportunity

Nearly all of the leaders in our survey are experimenting (37%) or actively using (56%) AI in analytics, and over half are already experimenting with general-purpose LLMs like Claude and ChatGPT for analytics.

93%

Are using or exploring AI for data and analytics



56%

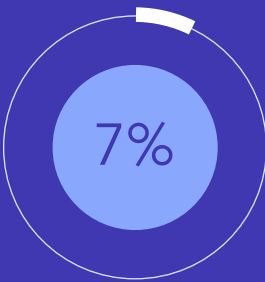
Are experimenting with general-purpose LLMs for analytics



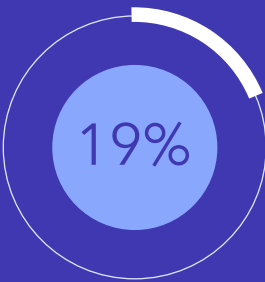
02

But transformation isn't easy

Fewer than 10% have org-wide AI analytics adoption. The rest are facing limited adoption or stuck in proof-of-concept purgatory. Of those who are actively deploying AI analytics, only 19% are fully confident in their AI-generated answers.



Have org-wide AI analytics adoption



Are fully confident in their AI-generated answers

03

Leaders look to context to close the gap

While the majority of leaders do not yet have a complete, shared definition of enterprise context, 94% plan to invest in new context management systems over the next 18 months. And all of the survey respondents have either hired (75%) or plan to hire (25%) context-related roles over the next two years.

94%

Plan to change how they store and manage enterprise context

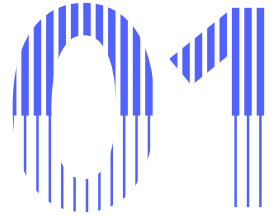
100%

Have hired or plan to hire context-related roles

CHAPTER 1

AI is changing how businesses consume data

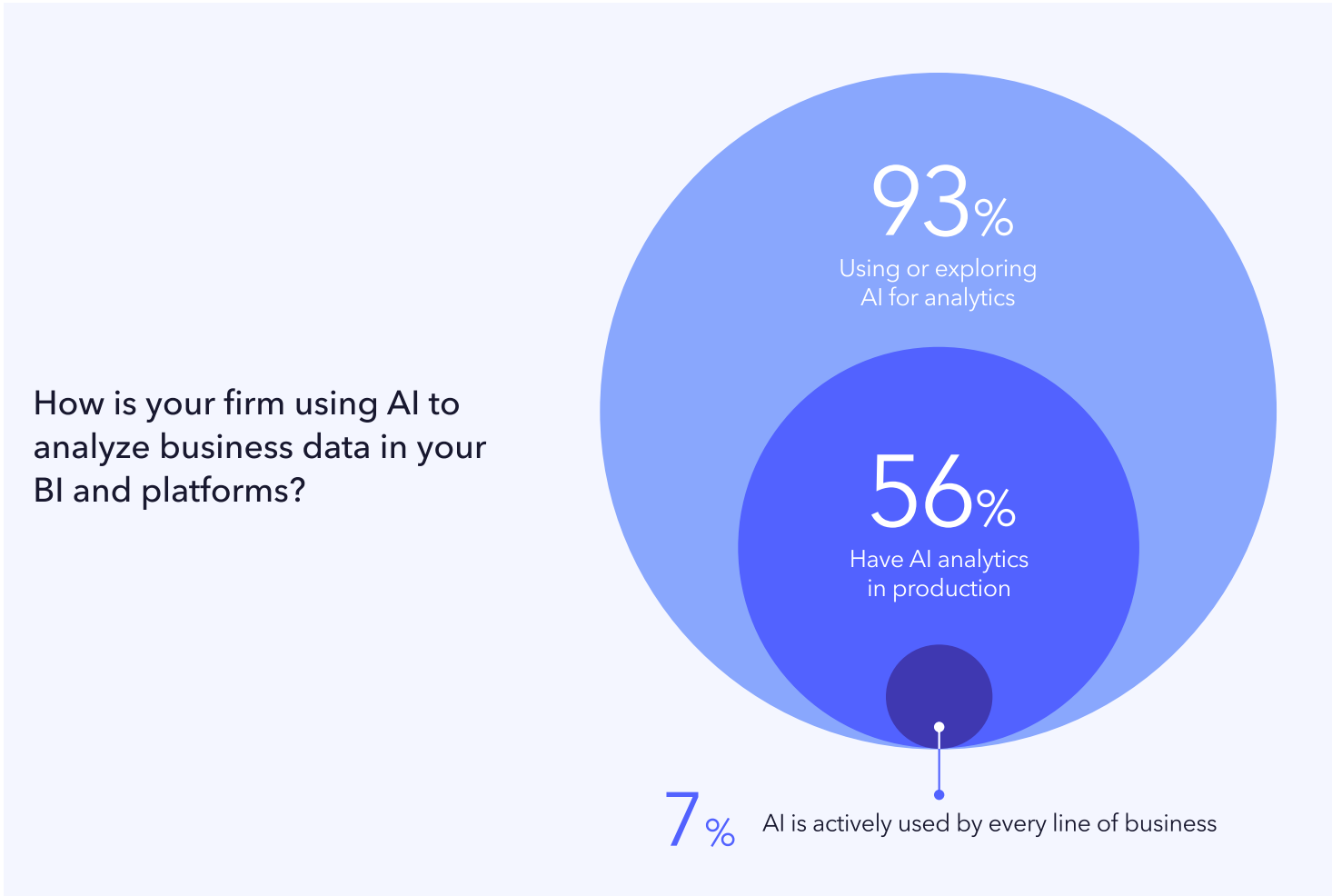
AI is changing how businesses consume data



Since OpenAI put generative AI in front of a mass audience in late 2022, enterprise adoption has moved faster than any previous technology wave, which took a decade or more to reach maturity.

After only four years in the market, almost all (93%) of the surveyed leaders are using or exploring AI for analytics. However, few teams have managed to scale AI to deliver on its promise of enterprise-wide time efficiency and workflow productivity.

While 56% of data leaders have AI analytics in production, only 7% say AI is actively used by every line of business today.

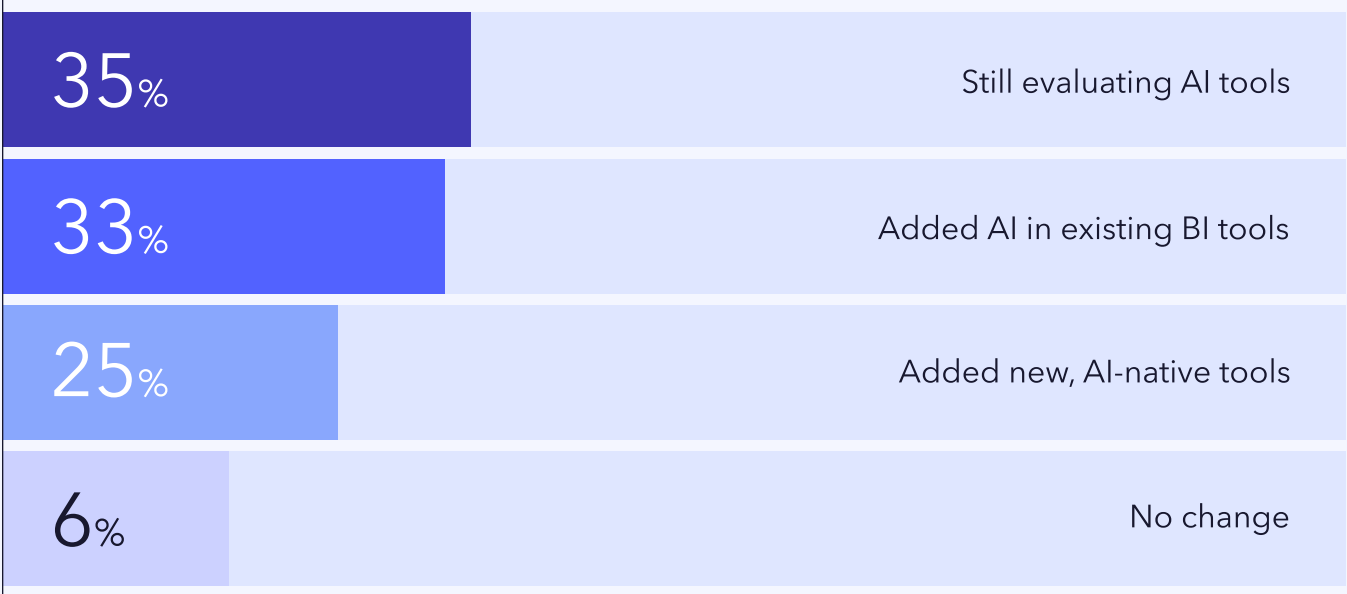


Where is AI analytics happening?

Over 9 out of every 10 respondents said their organizations are using or exploring new AI functionality in their existing BI tools, and/or opting to use other AI tools alongside their existing stack.

This suggests that most companies are hedging their bets – sticking with tried-and-true BI tools while the market matures. More importantly, they’re waiting to see which solutions can actually deliver the reliable analytics enterprise transformation requires.

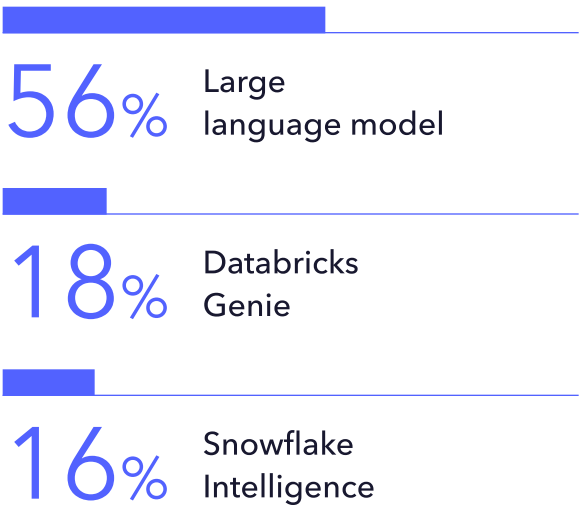
How have recent advancements in AI and data analytics impacted your BI and analytics tooling over the last three years?



Leaders aren't waiting idly

Traditional BI is facing steep competition as already 56% of leaders say their teams are using LLMs, such as Claude and ChatGPT, to directly analyze business data.

Meanwhile, 34% have adopted natural-language interfaces within their enterprise data warehouse (i.e., Snowflake Intelligence or Databricks Genie).



How Cisco is navigating the Modern Data Stack collapse

Cisco's finance team lived the version of the "modern data stack" most enterprises still run today: hundreds of dashboards, each requiring its own pipeline, semantic definitions, governance sign-off – leaving weeks between a question and an answer.

As Raj Kamra, Head of AI Strategy, Cisco Finance said, the problem wasn't simply a missing dashboard – it was the architecture itself:

"We had a legacy data architecture, lack of data governance, and hundreds of dashboards, which sometimes would have different numbers for the same KPIs. This would create more confusion and frustration for everyone."

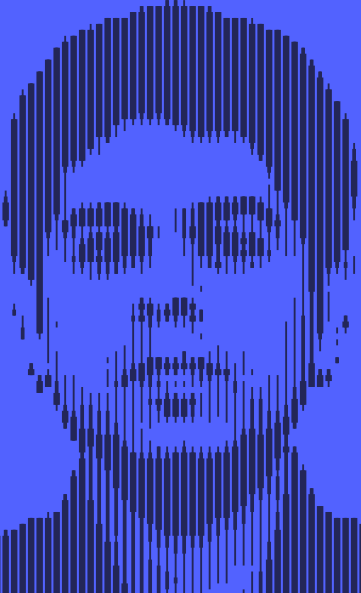
The instinct, he adds, is to patch the old model rather than replace it – bolt on a few more dashboards to reconcile the ones you already have. Kamra's team chose not to. Instead of building another layer on top of the stack, they chose to tear the stack down.

"The traditional approach is you start with a very specific use case, then you design the ETLs and everything around it – your models, your frameworks, your algorithms. It's very specific to that use case, so it doesn't scale," Kamra explains. "If your set of questions is ever-growing... how do you provide relevant context at scale? That's a very hard problem."

In its place, Kamra's team built an entirely new data stack and operating model – one where semantics, governance, and institutional knowledge stop living in separate systems, owned by separate teams, and start operating as a single layer the AI can reason against directly.

"You have to provide structure to the reasoning model so it can go deeper into cross-domain intelligence."

The result: Cisco procurement team's time to insight improved by more than 90%. As Kamra puts it, "That's a quotable number."



"Dashboards are static insights, and the insights we want to deliver are more dynamic in nature – because your context changes, your priorities change, and value changes over time."

Raj Kamra

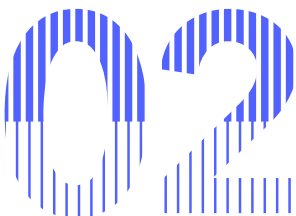
Head of AI Strategy, Cisco Finance



CHAPTER 2

Surface-level AI use cases deliver shallow outcomes

Surface-level AI use cases deliver shallow outcomes



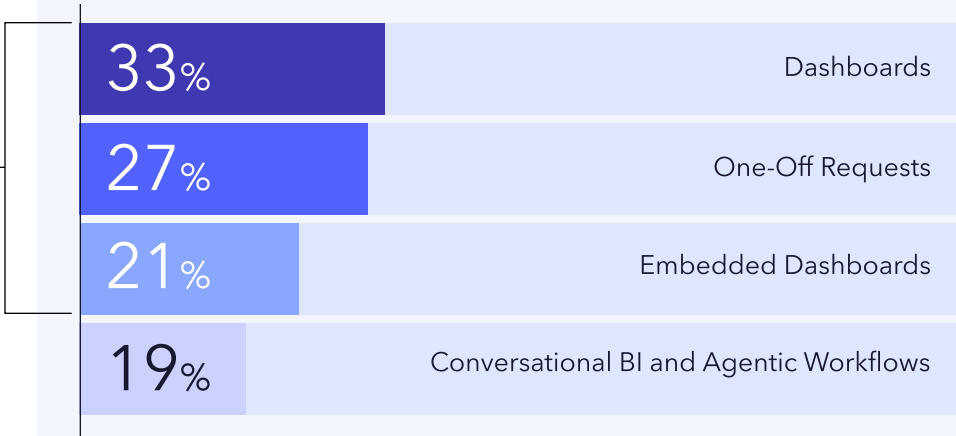
While overall AI adoption may have surpassed that of previous transformation waves, most companies are sticking to the shallow end of the use case pool.

IN FACT...

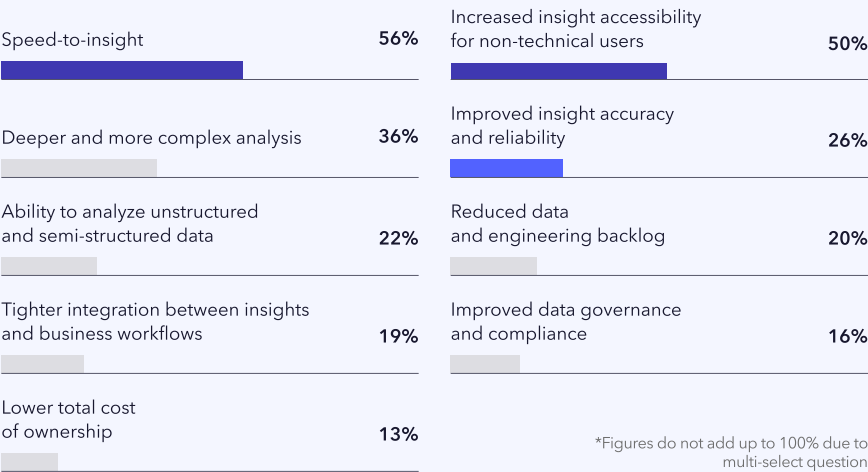
81%

of all surveyed leaders still rely on dashboards and one-off requests as their primary insight delivery method.

How do business users at your organization primarily find or engage with analytics insights?



What are the biggest operational benefits you've noticed since using AI in your BI and analytics platform?*



In the 113 organizations where AI analytics is in production, leaders say use cases are concentrated in AI-automated data visualization (54%) and natural-language summaries (43%) – the easiest problems to solve but not necessarily the highest-value. Multi-step reasoning, proactive agents, and insight-to-action workflow automation remain rare.

The same pattern shows up in the reported benefits:

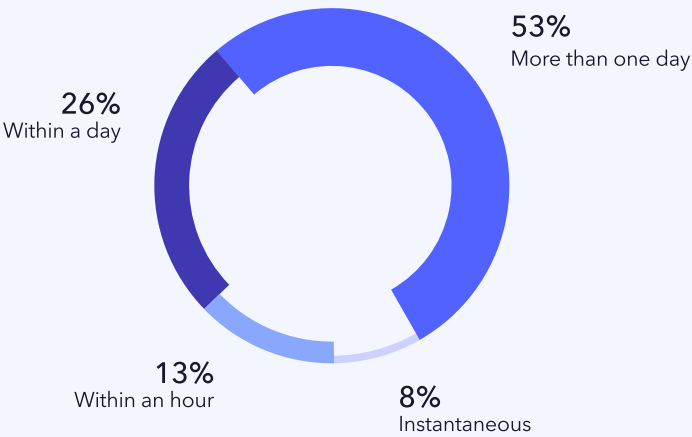
Speed-to-insight (56%) and access for non-technical users (50%) top the list, while gains in accuracy and reliability (26%) lag well behind.

The tools have changed, but the time-to-insight hasn't

At first glance, improved speed-to-insights seems like a huge win. However, when pressed on the topic, leaders report that most insight and dashboard requests still take a day or longer to fulfill.

Only about one in five organizations can deliver an answer within the hour, underscoring the reality that there's still a large gap between most enterprises and the AI's instant-insight promise.

Average time to insight



What are the biggest challenges with your existing BI and analytics tools?*

Lack of insight actionability	35%	Fragmented analytics workflows and tooling	35%
Slow speed to insights	33%	High total cost of ownership	32%
Technical skill requirements and usability challenges	27%	Low accuracy, reliability, and trust	24%
Inability to analyze semi- and unstructured data	22%	Insufficient security, governance, and compliance	20%
Data latency and / or limited source connectivity	18%	Visualization limitations	10%

*Figures do not add up to 100% due to multi-select question

Leaders rank the lack of insight actionability, fragmented analytics workflows and tooling, slow speed to insight, and high total cost of ownership as the four biggest problems with their existing BI tools, in that order.

Each of these data challenges directly highlights the cost of a disconnected, outdated architecture – one that requires data to be moved, modeled, and re-modeled before anyone can ask it a question.

Taken together, we see a picture of an enterprise data team that has superficially adopted AI analytics – the chat box, summary, and visualization – without yet solving the harder and deeper problem: Fast, connected, trusted AI that gives the business confidence to take action.

How Arm delivers outsized AI results

Arm, a nearly \$5 billion supplier of high-performance chipsets and software, uses AI analytics to deliver faster time-to-decision, deeper analysis, and true self-service across its procurement function.

Specifically, AI has empowered their team to widen the insights aperture beyond structured data, helping procurement managers answer key questions across their workflow – from overall spend and supplier activity through operating metrics and purchase order tracking. This insight accessibility is fundamentally changing how the team operates.

For instance, procurement managers shifted from asking whether termination for convenience clause exists for a particular contract, to now understanding precisely what the clause means. Previously, “that was such a heavyweight manual process,” Tom Smith, Procurement Governance, Policy, and Reporting Lead at Arm, recalls. By replacing human effort with a context-aware AI analytics platform, Smith says his team immediately cut reporting time in half.

Smith mentions:

“It’s moving from answering questions which was our old reporting functionality – to the agentic model proactively identifying the issues before anyone’s even thought to ask.”

This transformation gives the data team opportunity to focus on more constructive, meaningful work beyond tending to SQL, dashboards, and reports. Smith says:

“It’s freeing up the team’s time to improve data quality, rather than just producing reports.”

To deliver this type of proactive, self-service data exploration, Smith’s team is focused on giving AI the right context. Smith estimates that the data team is now spending a quarter or more of their time on developing this context, stating that “like any other business asset – context needs to be maintained.”

He attributes the success of the program to the acceptance that governance and quality are part of the improvement, rather than a prerequisite.



“If I was waiting for perfect procurement data, we wouldn’t have ever started... What matters is having enough structure to allow the AI to interpret the data correctly. And that comes down to business context.”

Tom Smith

Procurement Governance, Policy, and Reporting Lead

arm

CHAPTER 3

Trust stunts AI's scale

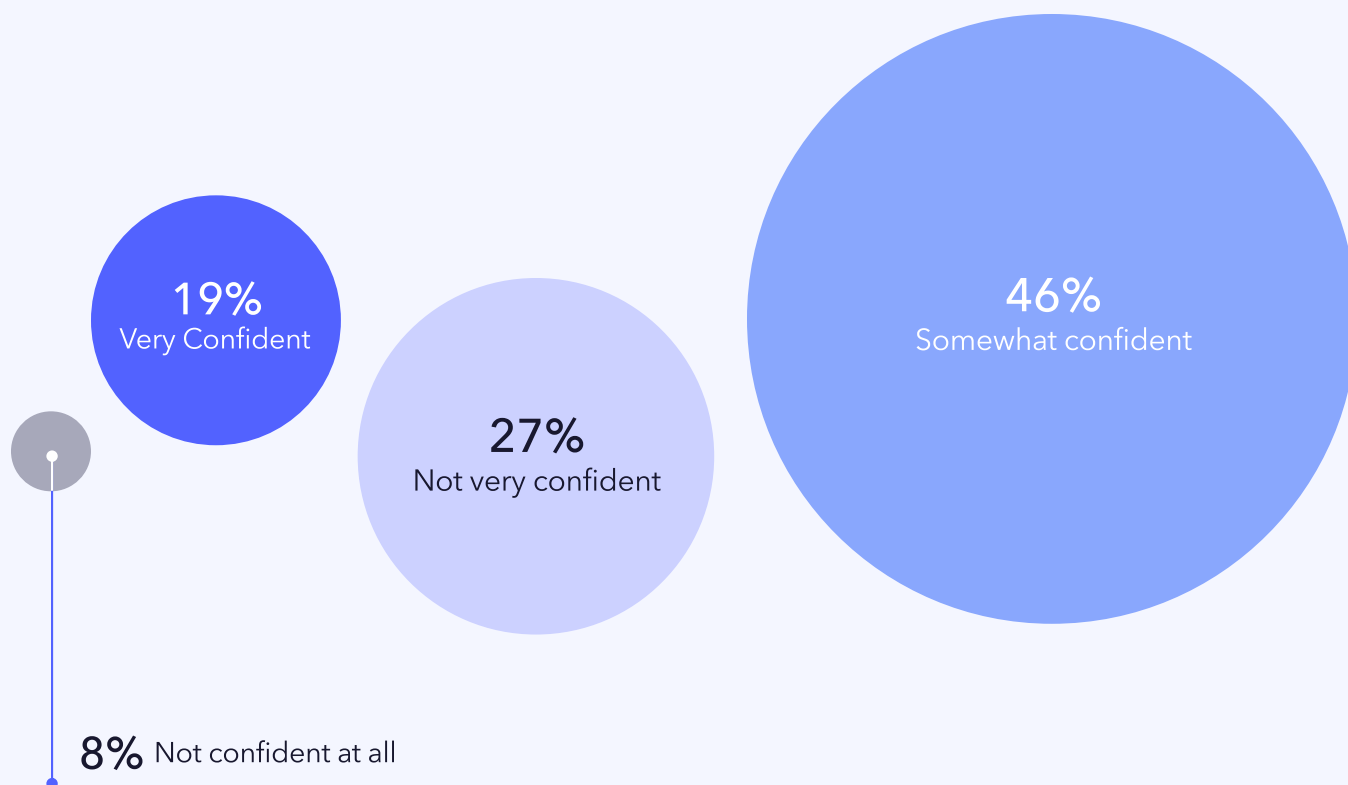
Trust stunts AI's scale

To build trust in AI analytics, leaders must see the challenge for what it is – an enterprise context problem – and quickly course correct.

Less than one in five data leaders are very confident in their AI answers.

For a large majority of enterprises, this means every AI-generated insight still needs a human sanity-check before it reaches a decision. That strategy doesn't scale.

How confident are you in your AI-generated answers?

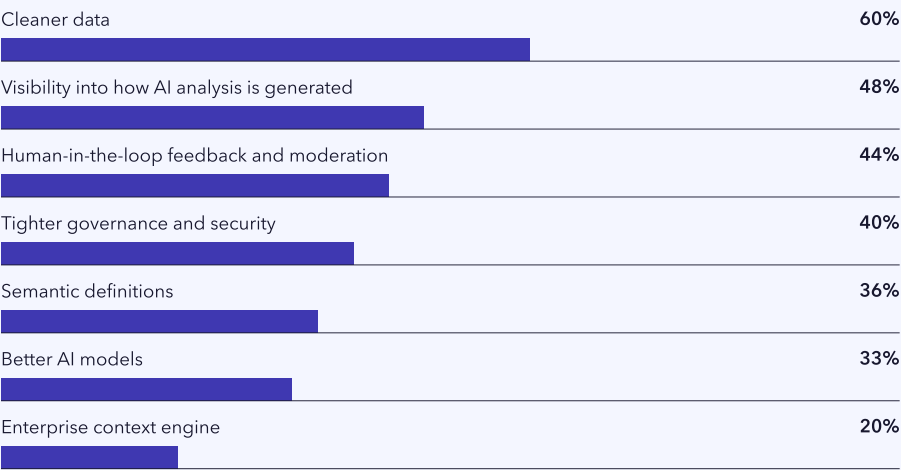


Improving accuracy isn't as straightforward as it looks

Most leaders (60%) are looking to cleaner data as the solution. Yet, the pursuit of perfect data is as old as the mainframe, and there is little consensus on how to solve it.

Given the fact that less than half of surveyed leaders have a shared definition for AI context (see page 19), it's not entirely surprising that only 20% of respondents say a context engine will improve AI accuracy. In reality, all of these technological improvements – from better AI visibility to semantic definitions – can be accomplished within a sophisticated context engine.

What technology solutions do you think would improve your AI accuracy in data and analytics?*

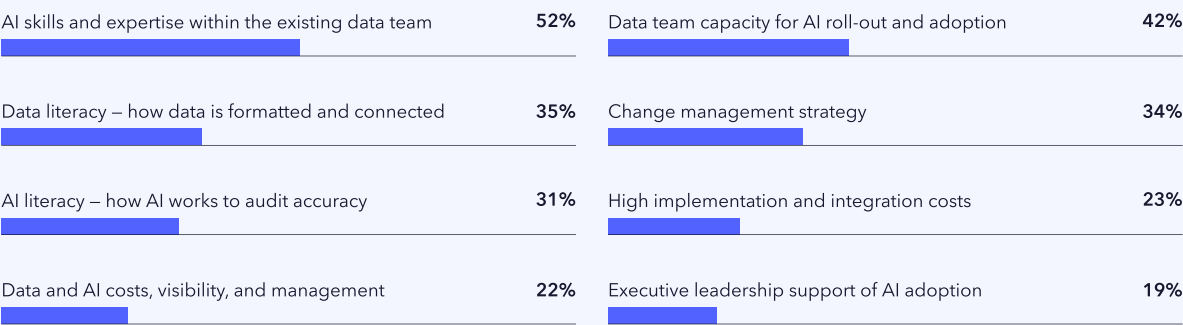


*Figures do not add up to 100% due to multi-select question

AI skills and data team resourcing are also commonly cited gaps. In fact, over half (52%) of respondents whose organizations use AI-powered analytics said their companies face talent challenges. Specifically, they said their data teams lack the requisite AI skills and expertise.

Another major challenge is data literacy. Approximately 35% of respondents said their organization doesn't broadly understand how data is formatted and connected to insights. Without this, business users cannot trust their insights enough to take action.

Which of the following represent significant people and organizational challenges for your organization when using AI in your BI and analytics platform?*



*Figures do not add up to 100% due to multi-select question

The antidote for AI inaccuracy

Tom Smith, Procurement Governance, Policy, and Reporting Lead at Arm, says:

"Any business function at any company has its own nuance, exceptions, and process complexity. All that bears out in the data. So, we have to give AI agents that context and the operational understanding they need to succeed."

The operational understanding Smith describes is context. Without it, AI cannot produce accurate answers consistently, no matter how capable the underlying model is.

Consider this scenario: An LLM trained on publicly available data can discern the meaning of "customer churn" and apply that global definition to your connected enterprise data.

Here's what AI doesn't inherently know:

- **Data context:** Which tables (with ambiguous names) apply to this calculation
- **Business context:** How the company specifically defines and applies the term
- **Historical context:** How that metric definition changed last month after a board meeting and subsequent Slack conversation
- **Presentational context:** How sales, marketing, and finance each have their own definition for the same term

To ensure AI analytics is viable at scale, teams need to treat context as infrastructure and build adaptive context systems that scale with analytics consumption surfaces.

Here's what that looks like in real enterprise deployments:

arm

Arm initially assumed more context would produce better answers. "What we've actually found it's the opposite," Smith says. "Less context – very specific, very refined context – is key."

Arm now runs a formal review process before any new context enters the system – treating it as seriously as any other governed business asset.



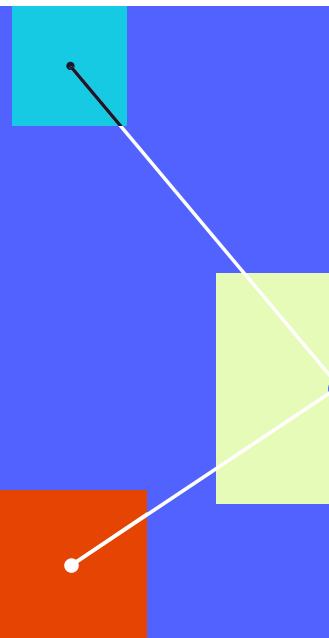
"A whole lot of context is hidden within the business," says Raj Kamra, Head of AI Strategy at Cisco Finance.

That reasoning is rarely captured, but Kamra's team is changing that with a centralized context repository. 12 months in, Kamra says this process is still evolving, highlighting the adaptive nature of context management.



Volkan Oktem, VP of Analytics at Property Finder, measures context with two metrics: fulfillment rate and accuracy.

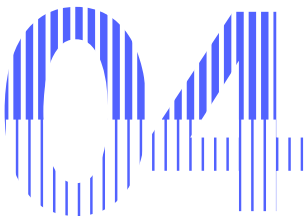
An AI agent continuously monitors and triages unfulfilled answers. "We look at it to understand if that question was an ad hoc unique case, or if there is a change in business need that requires new context."



CHAPTER 4

Leaders invest in context to close the trust gap

Leaders invest in context to close the trust gap

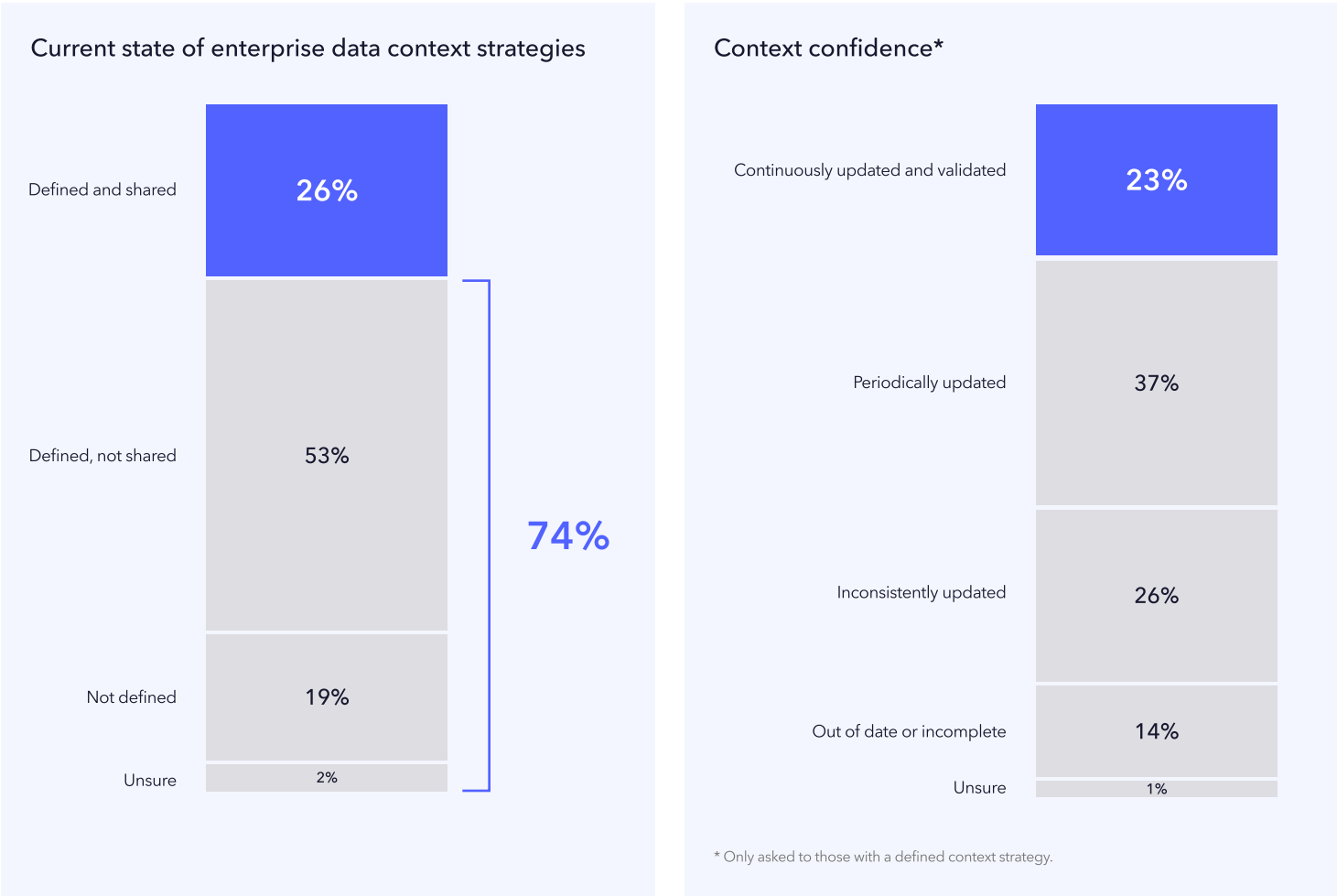


To build trust in AI analytics, leaders must see the challenge for what it is – an enterprise context problem – and quickly course correct.

The problem is structural

Nearly three-quarters of respondents (74%) say they do not have a complete and shared definition of enterprise context across their organization.

Of those who have defined enterprise context, only 26% have a shared definition, and even less (23%) are “very confident” that their context is continuously updated and validated.



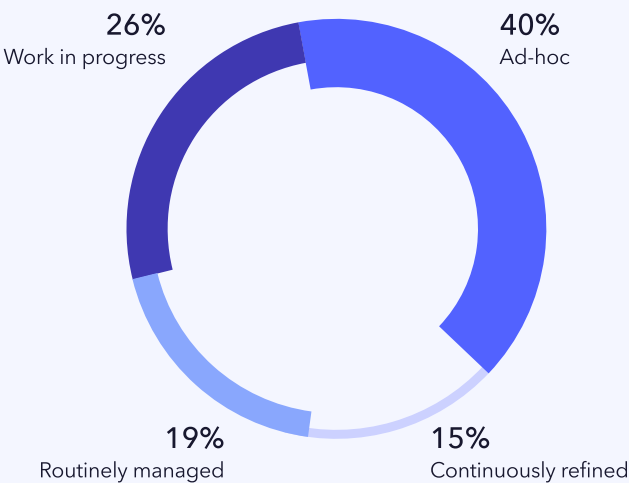
Where context does exist, it's scattered

Of those with defined enterprise context, about 40% admit they're still managing that context on an ad hoc, case-by-case basis. More than half (56%) say their enterprise context lives embedded inside individual AI or BI tools, and roughly the same number of leaders (53%) point to tribal knowledge – undocumented, sitting in someone's head or a Slack thread instead of a system of record.

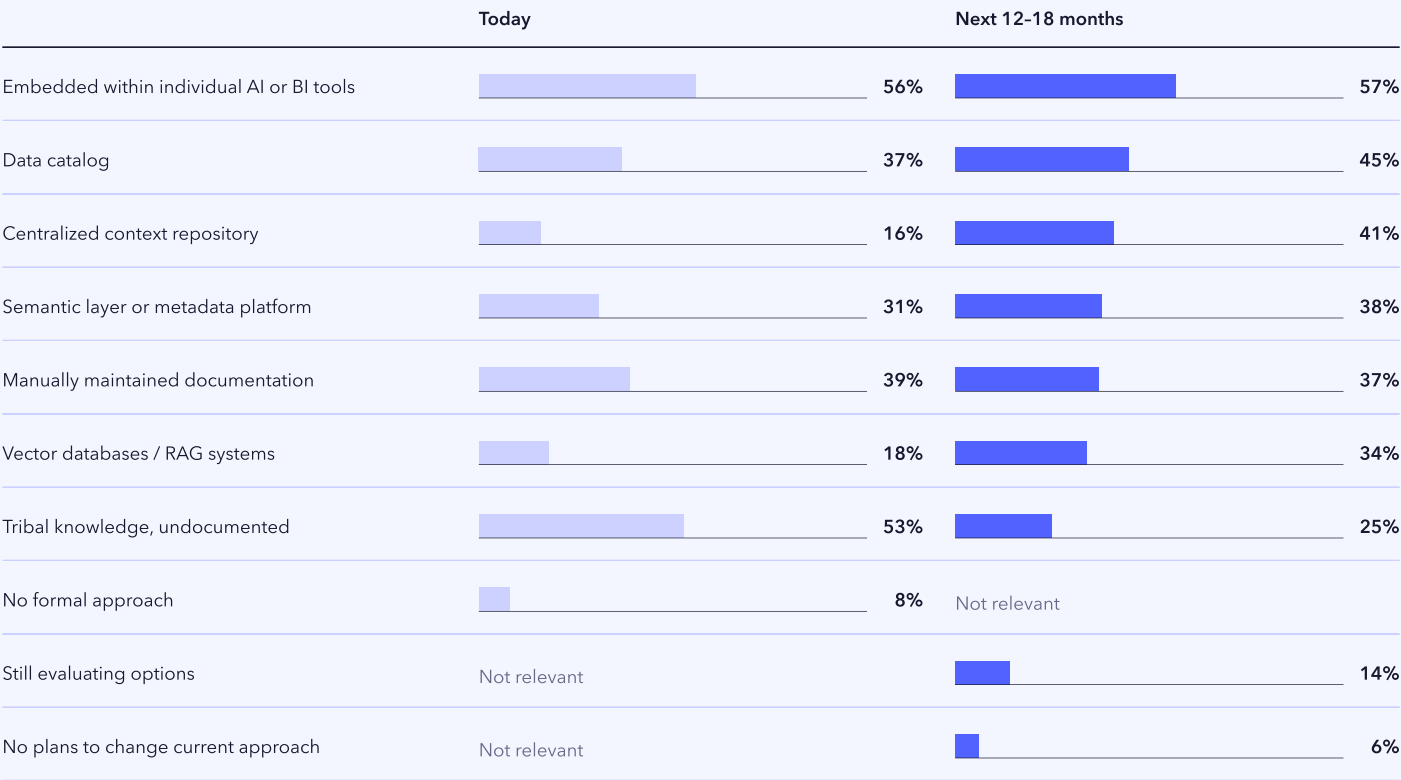
Semantic layers and data catalogs – the “modern data stack” tools built to solve this problem in the pre-GenAI era – are in use at fewer than 4 in 10 organizations. Still, data leaders are betting big on more comprehensive context-management systems.

In fact, 94% of surveyed leaders plan to change how they store and manage enterprise context in the next 12 to 18 months. The majority are heading toward the same destination: centralizing tribal knowledge into AI and BI tools, data catalogs, or dedicated context repositories.

How do you manage enterprise context?



Where enterprise data context is stored and managed*



*Figures do not add up to 100% due to multi-select question

How Property Finder democratizes data by centralizing context

Property Finder, a leading real estate platform services provider, realized that self-service analytics is hyper-critical to hypergrowth. Their rationale: everyone in the 600-person-plus company has a role that involves some data. However, not everyone has the equal ability or aptitude to properly find and apply those insights.

While assessing general-purpose LLMs, Volkan Oktem, Vice President of Analytics, realized the company needed, “a solution where context training does not require heavy effort, especially not heavy engineering effort.” None of the available LLM solutions met the firm’s expectations for a tool that was as “simple as asking another individual,” Oktem recalls.

Instead, he selected WisdomAI, an Agentic Intelligence Platform that lets the analytics team focus on more complex and higher impact strategic endeavors.

Today, roughly 80% of the organization uses the platform, asking an average of 10 questions a week. Accuracy is around 98%, Oktem says, which is building trust and scaling usage among users. And Property Finder’s analytics team is now twice as productive as it was pre-AI analytics.

Oktem explains:

“We are utilizing our resources in more complex and higher impact areas, which we would not be able to do without this productivity increase.”

One of those new focus areas is context, specifically updating and extending the existing context to remain aligned with the business’s evolving priorities. By assigning a designated context owner for each domain – commercial, product, and marketing – the data team is able to avoid conflicting context and maintain domain consistency.



“Business users don’t get into the complexity of how context is provided. At the end of the day, what they care about is: ‘When I ask a question, am I getting a good enough answer in terms of reasoning and accuracy?’”

Volkan Oktem

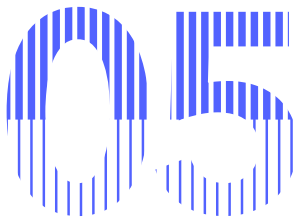
Vice President of Analytics



CHAPTER 5

The data team is redefined, not replaced

The data team is redefined, not replaced

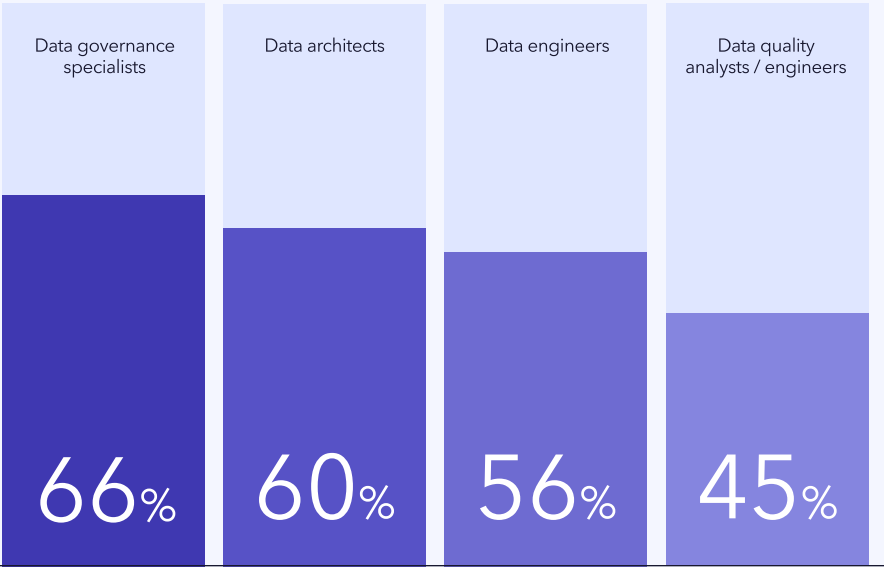


The rapid infusion of AI throughout business is fundamentally altering the roles and responsibilities of the data team. While AI-induced role replacement may be in the headlines, data teams are seeing a different shift play out in real time.

Context work is hiding in plain sight

At first glance, the data team's org chart looks the same as it did two years ago. But the work underneath it doesn't. As new emphasis is placed on enterprise context, those responsibilities are being distributed among more traditional data team titles, including data governance specialists and architects.

Who on the data team is responsible for managing enterprise data context?*



*Figures do not add up to 100% due to multi-select question

FOR REFERENCE

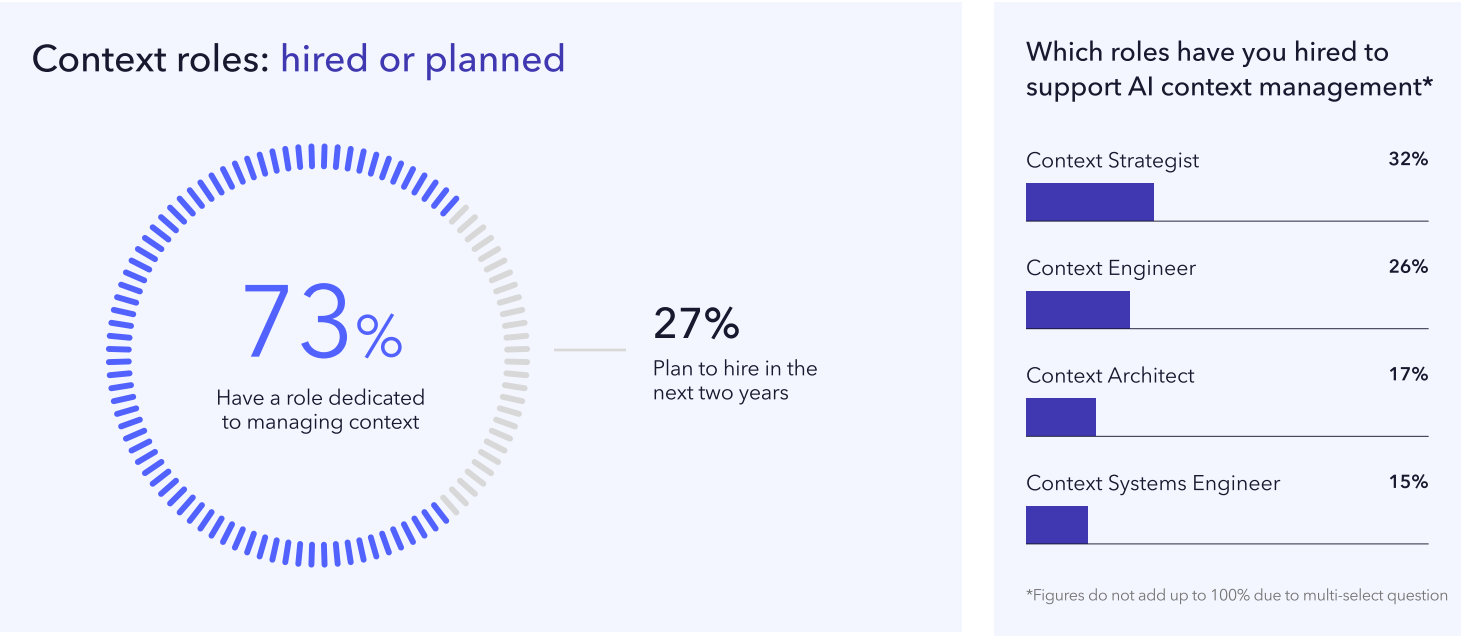
Which roles are currently on your data team?*

Data analysts	78%
Database administrators	65%
Data quality analysts	58%
Data engineers	57%
Business intelligence analysts	53%
Data architects	52%
Data scientists	47%
Machine learning developers	46%
ETL developers	37%
Data governance specialists	37%
Data product managers	32%
Data integration engineers	32%
ML Ops / AI Ops engineers	26%
Prompt engineers	21%
AI context engineers	10%

*Figures do not add up to 100% due to multi-select question

Data teams need designated context managers

Still, roughly three quarters say they have a role on their data team dedicated to owning and managing AI context. Of respondents whose companies do not have anyone in this role (55 of the 201 companies surveyed), 33% currently have open positions. The rest plan to hire within the next two years. This suggests the role and responsibilities of the AI Context Engineer are top of mind.



Retraining and upskilling for AI analytics is the main focus

The majority of respondents (57%) said they are looking to upskill or reskill the workforce to add new AI-specific skills, including context engineering. About 20% said they are looking outside the organization to hire additional resources.



A tale of two data teams

The survey findings offer a clear signal for data team hiring: leaders aren't debating whether context needs to be managed as its own discipline – they're focused on the fastest time to AI viability. As the market converges on a centralized approach to managing context, there's an outsized upskilling and repositioning opportunity for data practitioners looking to advance their careers.

Here's an in-depth look into how two leaders are approaching this within their own data teams:

arm

Tom Smith, Arm's Procurement Governance, Policy, and Reporting Lead, says that historically, people on his team "would spend their time building reports and answering data requests, developing dashboards for people, collecting requirements" – the traditional data engineering work of connecting systems and preparing data for reporting. "That's the easy part now," he says.

Instead, the highest-value work has shifted to "helping AI understand our business" – reviewing how the AI is used, examining the underlying queries, and identifying where context has broken down. That's "a totally different problem from just building dashboards," he says.

Context management now takes up roughly a quarter of his team's time, and Smith estimates it's "possibly going to push to more" as adoption deepens.

Above all, Smith says, the most important trait for this new kind of role is curiosity:

"You need somebody that's interested in how organizations and processes work."

A bit of skepticism matters too, he adds, describing what he calls the successful AI Context Engineer's "virtuous cycle" – ask, dig, pry, and then challenge. "And when you've got that, then solve it, and do it again."



Property Finder is upskilling its data team with people who understand what metrics are needed by the business – someone who can anticipate what business users might need to know. Volkan Oktem, Property Finder's Vice President of Analytics, is finding these skills in a potentially unlikely place: people who come from consulting backgrounds. He says:

"In an ideal world, and I don't think we're far from it, maybe in a year or two... 10% to 20% of the team will still need to do queries and maintain the data, but 80% of the team will no longer require any technical skills."

Oktem's rationale: For starters, WisdomAI's Adaptive Context Engine makes managing context easy – and can be used with other AI tools to help develop, extend, and maintain that context.

Second, Oktem's analytics team already understands key metrics, can anticipate what business users may ask next, and has critical data and table knowledge.

This enables Property Finder's data team to more easily review the queries the tool generates and make contextual improvements that elevate accuracy and relevance.

A blueprint for delivering AI analytics at scale

As data and analytics leaders race to adopt AI, new challenges are piling up just as fast. The fix isn't a better model or cleaner data – it's the bigger picture: infrastructure, people, and process. And at the center of all three sits enterprise context, which most organizations still treat as an afterthought instead of the foundation.

That's the gap. Here's how to close it:

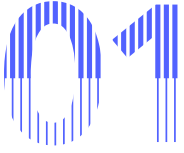
01 Stop treating clean data as the finish line

02 Manage context like a system, not a one-time project

03 Meet your data where it is

04 Invest in the right context engine, not a better AI

05 Build the team around function, not job titles



Stop treating clean data as the finish line

Over half of leaders still reach for cleaner data first when accuracy falls short – it's the fix every generation of data leader has reached for long before AI. But clean data can only tell AI what the values are and how they are connected. It doesn't tell the AI what they mean, why they changed, or which definition applies to which team. That's what enterprise context captures.

WHAT TO DO

By focusing on closing the context gap, rather than chasing the cleanest data in the business world, you can finally deliver AI accuracy your team can base decisions on. Stop routing every accuracy complaint back to the engineering team by default. Instead, start auditing each data domain's context, figure out what's missing, and then build that context directly into your analytics stack.



Manage context like a system, not a one-time project

Roughly a quarter of organizations have a context definition that's fully defined and shared across the business. Of those, a large share are still managing it ad hoc – on a case-by-case basis, with no tooling or process to keep it current. That's the same failure mode as the old semantic layer, just with a new name: something gets uploaded once, and nobody revisits it until it's proven wrong.

WHAT TO DO

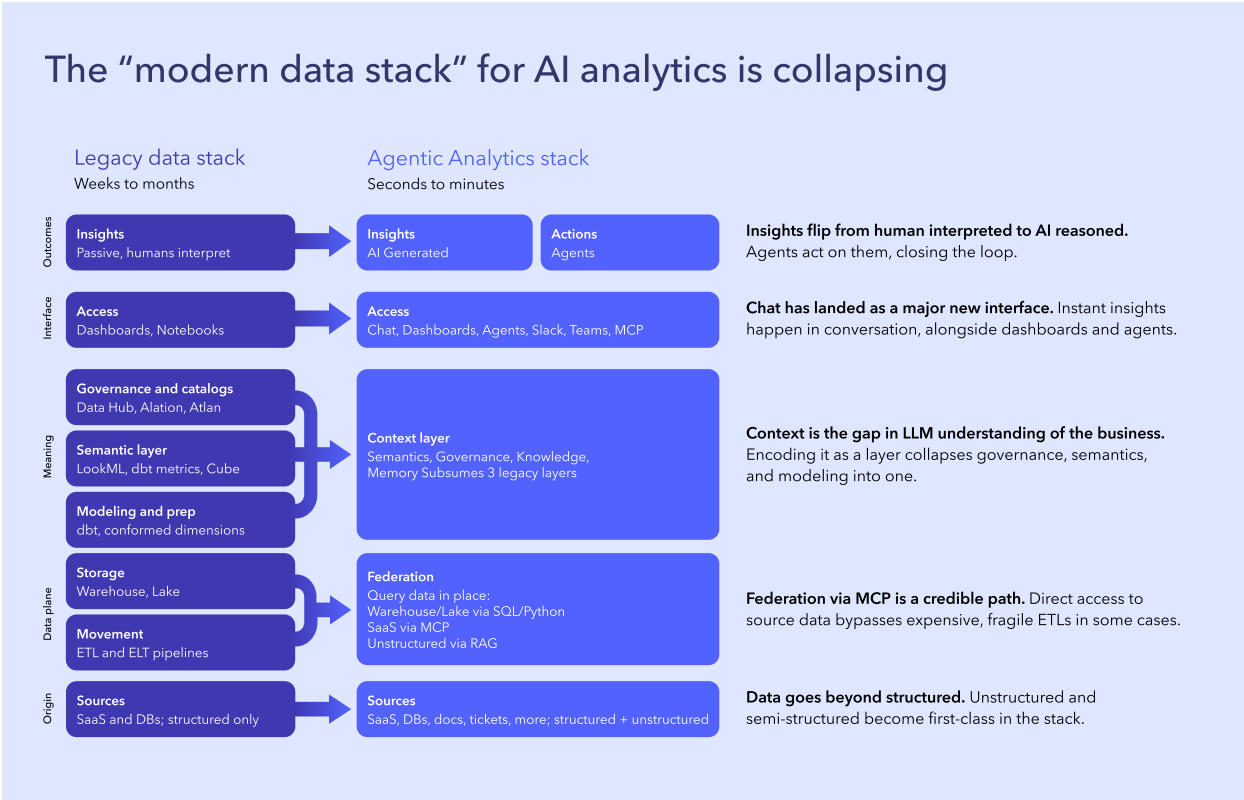
The organizations further ahead treat context more like code. Each domain has an owner with an update schedule, process, and versioning – not a static document. Don't let your context strategy end at "we defined it." Assign an owner, build a review cycle, and track whether it's actually still accurate – the same discipline you'd apply to any other system your business depends on.



Meet your data where it is

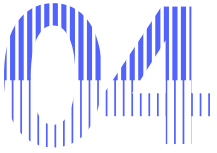
Leaders rank fragmented tooling, slow speed to insight, and high total cost of ownership among the biggest problems with their existing BI stack. These architecture problems all stem from the same root cause: an assumption that data has to be extracted, moved, and remodeled into a warehouse before anyone's allowed to ask it a question.

Every layer in the old modern data stack – the pipeline, warehouse, and semantic layer – existed because a human needed data translated into something a dashboard could display. AI agents don't have that constraint. They can reason over a question the moment it's asked, wherever the data actually sits, as long as the context is there to support it.



WHAT TO DO

Before you fund another migration or a bigger warehouse, ask the harder question: are you trying to consolidate data, or are you trying to make it usable? Those used to be the same project, but they aren't anymore. The fix isn't moving more data into one place. It's building the context layer that lets your AI reach it wherever it already lives.



Invest in the right context engine, not a better AI

Accuracy isn't a model problem. Without the right context, even the most advanced AI will still give a confidently inaccurate answer. Semantic layers and data catalogs – the modern data stack's answer to this exact problem – are in use at fewer than 4 in 10 organizations. And even where they exist, context still ends up scattered across individual tools or trapped as tribal knowledge nobody wrote down.

WHAT TO DO

Leaders are already realizing this gap and allocating funds to address it, with 94% of leaders planning to change how they store and manage context in the next 12 to 18 months. Don't get left behind. Start auditing your stack today. Figure out where context is getting lost in translation, and look for fully integrated solutions that help you manage it – not bolted-on quick fixes that were never designed for the AI era.



Build the team around function, not job titles

AI has already changed the data team's core responsibilities, but that's not a bad thing. The analyst's value used to be measured in shipped reports. Now it's measured in whether the business can trust what the AI just told them. Yet, most leaders are bolting context-related responsibilities onto pre-existing roles without redesigning the workflow around it, which is exactly how it ends up neglected.

WHAT TO DO

To get ahead, start by auditing your existing data team to understand where the skills and capacity exist. Decide whether you need new, designated context roles, or if existing roles can evolve to adapt the Context Development Lifecycle. Then, ensure these new roles and responsibilities are built into your process roadmap – not as an add-on to someone's job but as a fundamental requirement.

Methodology

In May 2026, WisdomAI conducted an online survey of 201 Senior AI, Analytics & Data Leaders at North American organizations with at least \$1 billion in revenue to understand the people, process, and platform implications of AI-powered business intelligence and analytics platforms.

201

Senior AI, Analytics & Data
Leaders surveyed

Respondents by representative title

	N	%
Chief Data Officer	41	20.4
Chief Analytics Officer	30	14.9
Chief Data & Analytics Officer	30	14.9
VP / Director, Data Analytics	27	13.4
VP / Director, Business Intelligence	24	11.9
VP / Director, AI Transformation	21	10.4
VP / Director, Data Engineering	13	6.5
VP / Director, Data Science	11	5.5
VP / Director, Data Governance	4	2.0
Total	201	100

Respondents by primary industry

	N	%
Healthcare	37	18.4
Banking, Financial Services & Insurance	34	16.9
Consumer Goods / Retail	33	16.4
Hi-Tech	28	13.9
Life Sciences	27	13.4
Energy & Resources	22	10.9
Industrial	20	10.0
Total	201	100

Note: Percentages throughout this report may not add up to 100% due to rounding.



About WisdomAI

WisdomAI is the leader in a new generation of AI analytics and business intelligence, built agent-first for the enterprise. Founded in 2023, WisdomAI delivers the first Agentic Intelligence Platform: a single, trusted surface that replaces a fragmented stack of legacy BI, single-user chatbots, and rule-based automation.

At its core is the Adaptive Context Engine, a living system that continuously learns the metrics, rules, and vocabulary your business runs on. It grounds every chat, dashboard, and analytics agent in shared context, reaching human-level accuracy across structured and unstructured data no matter where it lives. The result: insights you can trust, and agents that act on them.

Looking for tools to support your Context Development Lifecycle?

WisdomAI delivers over 95% accuracy in real enterprise environments. See how we can help you.

Scan the QR code to schedule a demo today.

