

From Silos to Success: Modernizing **Analytics with Data** Governance and GenAl

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What we've heard



Ideally, organizations want to have.....



Platform to actionable Insights to the business



Robust data governance



Ability to increase the value of hidden data



To spend less time preparing data



Improve operational efficiency

Why do organizations struggle with data?

MIT Technology Review Insights Survey

92%

Culture, people, process

8%

Technology

What does good look like?

Reduced duplication	47%
Ease of data access	38%
Data engineering	36%
Data quality	31%

What holds companies back?

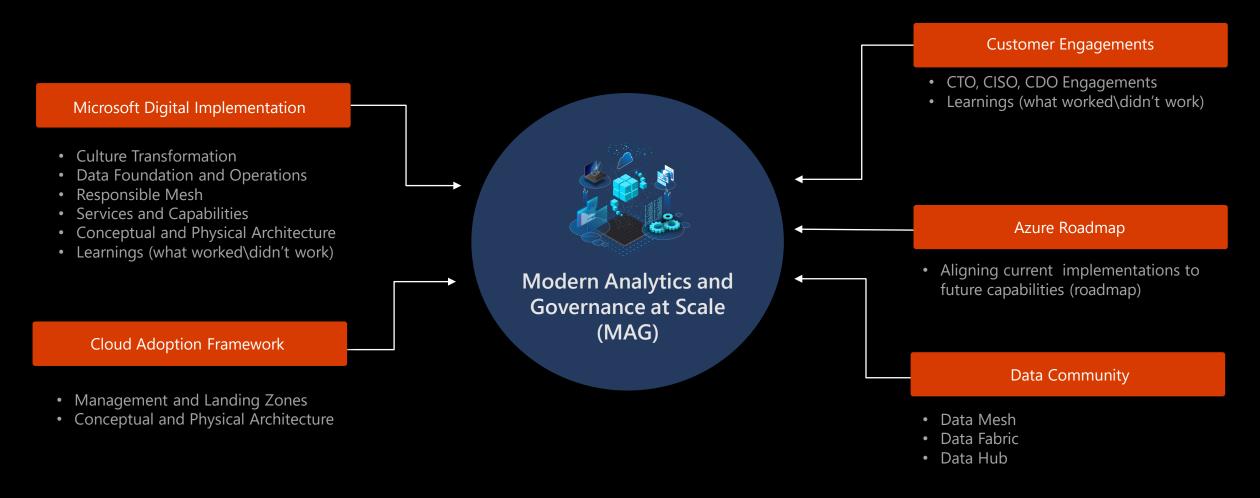
Lack of data management	44%
Slow data processing	39%
Lack of collaboration on analytics	29%
High data duplication	22%

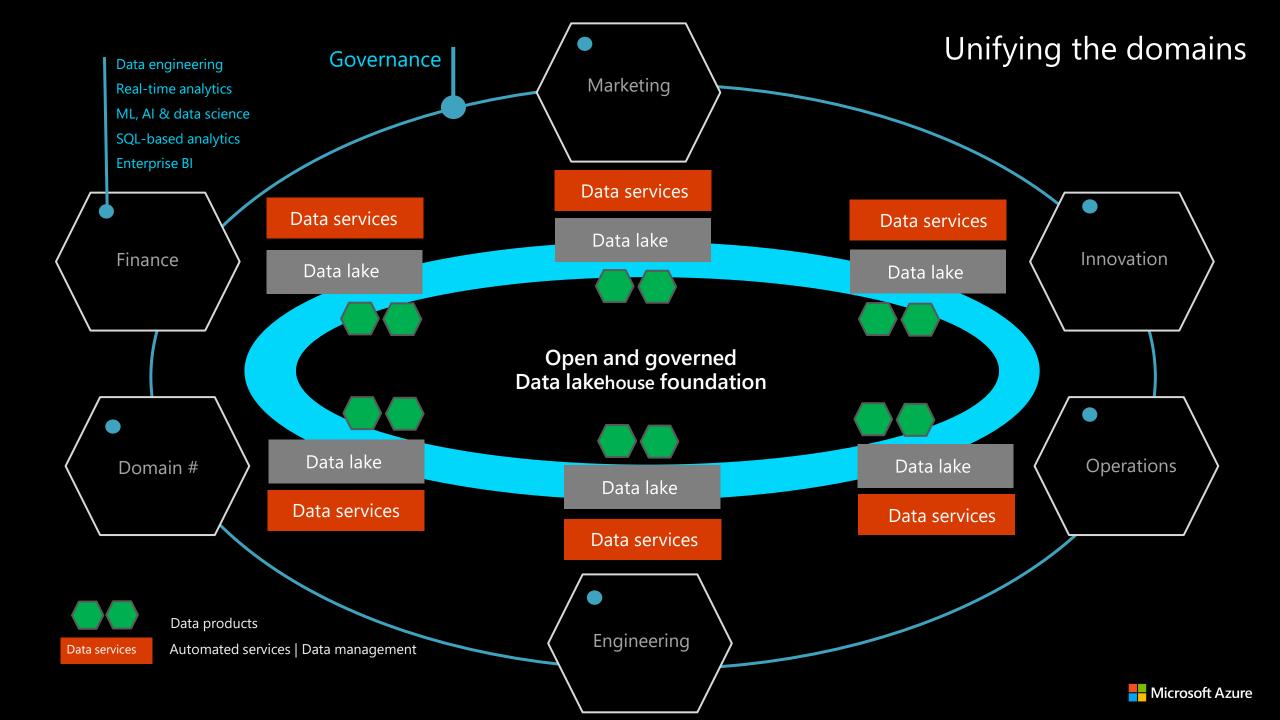


^{*} Building a high-performance data and AI organization | MIT Technology Review

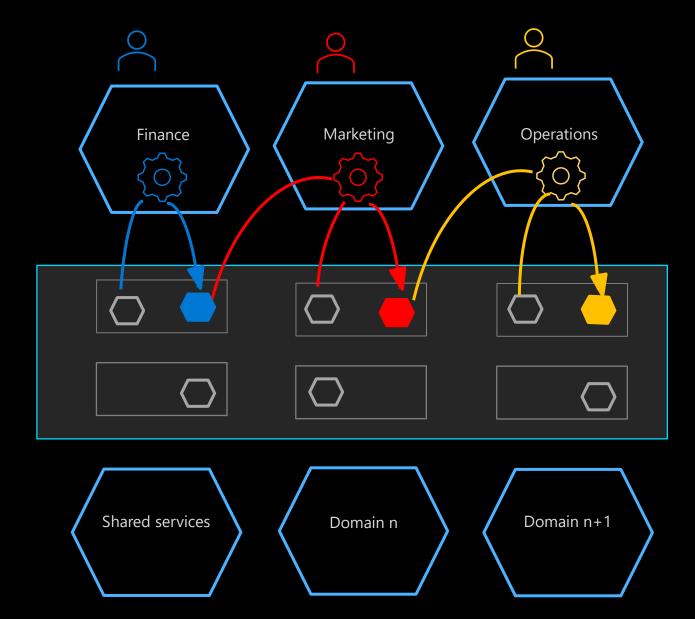
Modern Analytics and Governance at Scale (MAG)

Empowering business units without compromising enterprise requirements





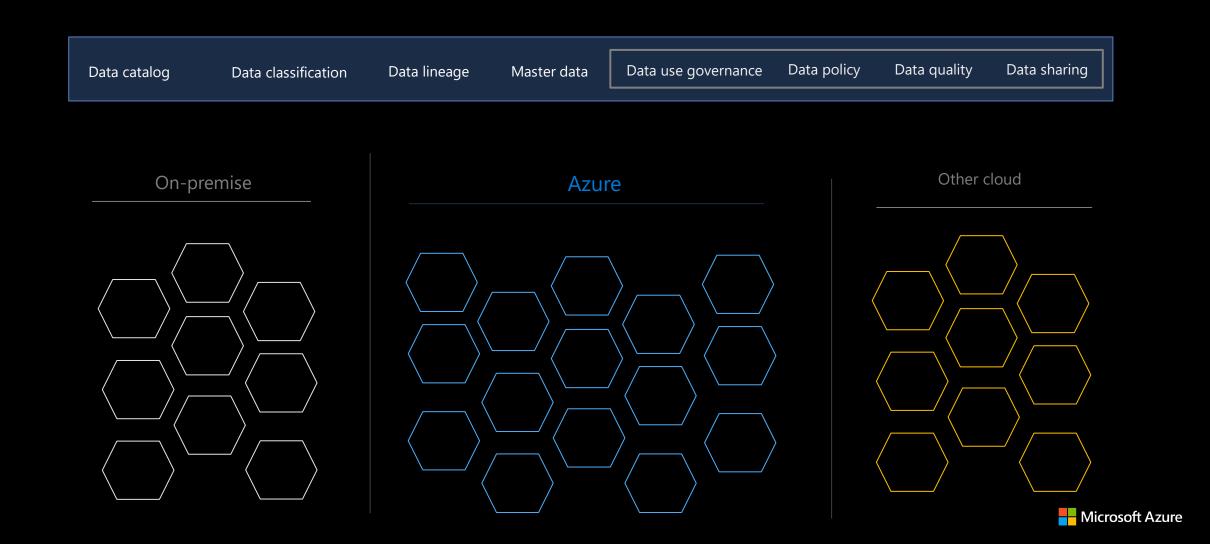
Enable Lines of businesses



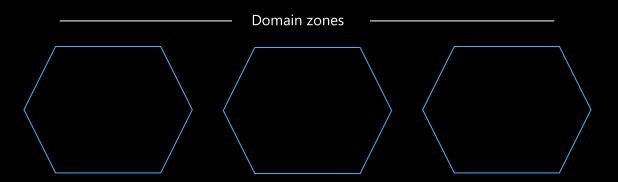
- Self-serve analytics
 - Empower LOBs to implement their own analytics projects
 - Democratize data and analytics across LOBs
- Accelerate cross-business unit collaboration
- Leverage LOB SMEs for business analytics
- Re-use Data products across domains
 - Reduce data engineering
 - Improve data agility

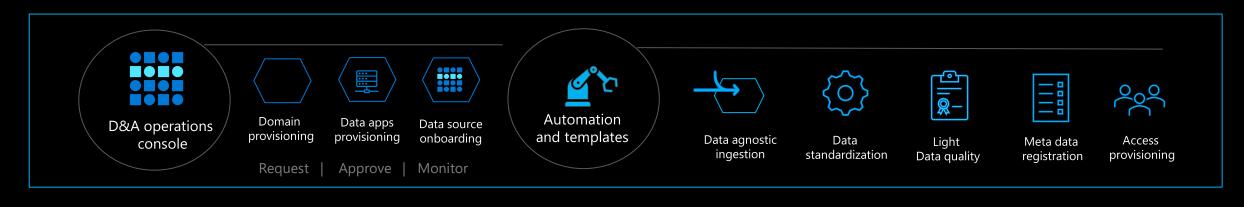
Enterprise data governance

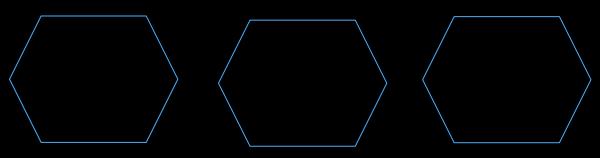
Automated capabilities to implement frictionless governance



Automated data services







Analytics patterns and practices per domain



Aligning business use-cases and analytics patterns

Type 0 Ingestion and curation

- Ingesting data from data sources
- Data engineering

Type 1 Discovery/onboarding

- Project onboarding
- Data exploration, Inventory and profiling
- Ad-hoc queries

Type 2 Descriptive analytics

- Data products creation
- Reliable and trusted data for business analysis
- Reporting and dashboarding
- Ad-hoc quires

Type 3 Predictive analytics

- Data products creation
- AI/ML labs to build, train and deploy models
- Data wrangling

Type 4 Data exchange

- Exchange trusted data to and from vendor and partners
- Data exchange between domains

Note: These are top analytics patterns prioritize by organizations, but implementations are not limited to this list.



MAG Implementation stages

Planning

Current state vs Future State

Current State

- Architecture
- Services
- Deployment
- New use-cases\projects

Evolution to Future State

Leverage MS and Partner lps\Accelerators

Recommendations

- MVPs and prioritization
- Roadmap alignment

Phase 1

Deploy the open and governed foundation services including data operations zone and domain, and onboarding one data product to the platform.

Foundation

- Domain zones (template)
- Data lake
- Template creation

Data governance:

- Data definition, glossary, and Catalog (Purview)
- PII detection (Purview GA)
- PII governance

 Data discovery: data access management (Purview, in private preview)

Ingestion:

- Metadata-driven ingestion framework
- Data standardization

Data provisioning

Host data to server use-cases

Phase 2

Onboarding more data products, optimize UI and operation to be ready for full production.

Data governance:

 Attribute-based control (ADLS Gen2, SQL, Purview Policy)

Data discovery:

- Shared with me (customer build)
- Custom attributes (customer build)
- Catalog extension to other UIs (Purview Q2 CY22)

Phase 3

Enhancements

Data governance

- Data standardization
- Standardized data models
- Master reference data for data cleansing
- Programmable data quality rules

Discovery

- Data exploration with data profile:
- Ingestion tags
- Data usage dashboard

Ingestion

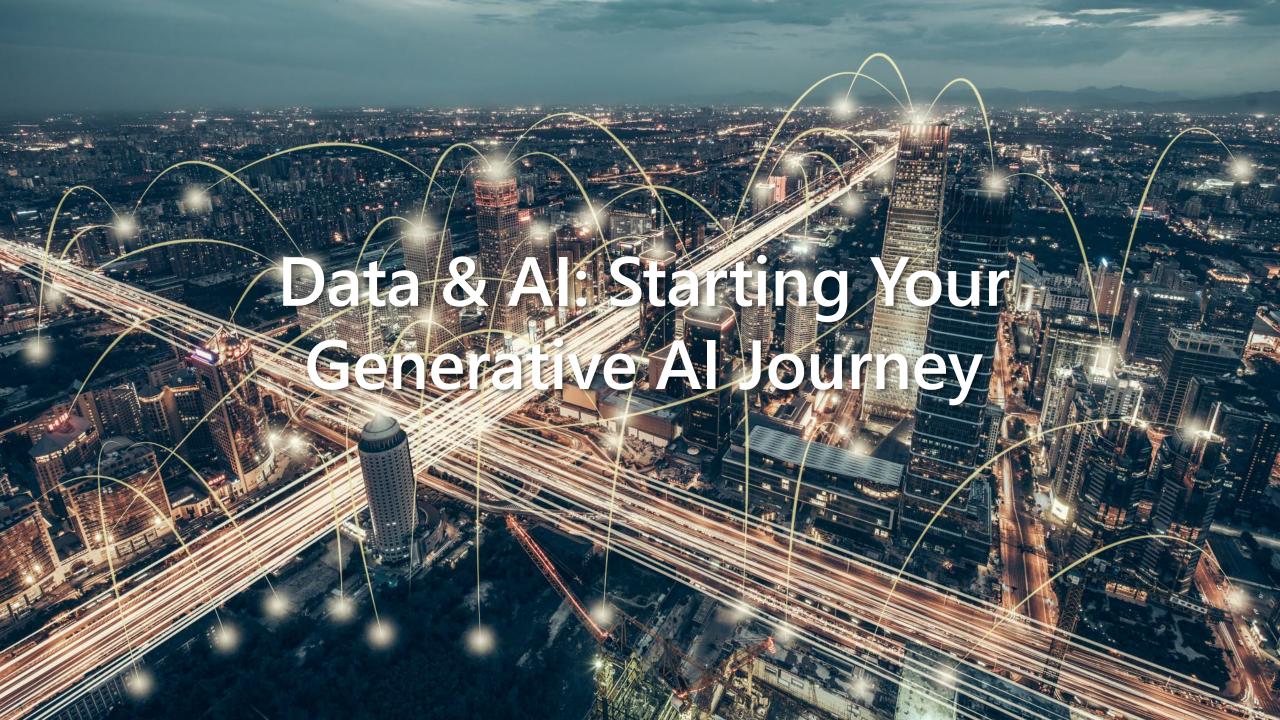
Micro batch/streaming (customer build)

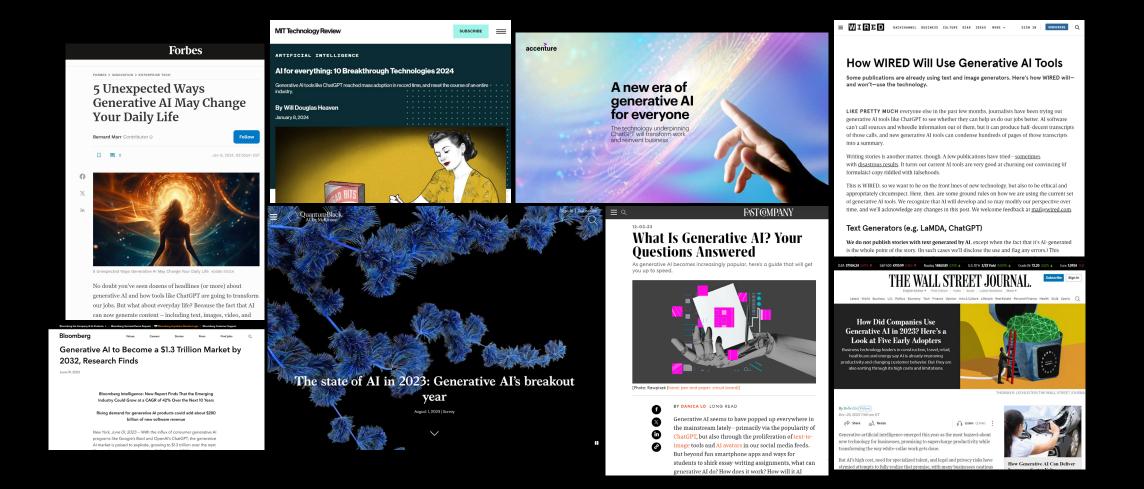
Data management

- Metadata management
- Master data management

Artificial Intelligence







Generative Al is a game changer ... now it's all about implementation!

To maximize value, consider use cases that serve unmet user needs

LEARNING

Transform internal and external search functions into **human-centered insight hubs** for collaboration and marketing

1

Chat with your data

TRAINING

Empower customer service with insight and encouragement that **helps customers** while supporting talent

Intelligent call centers

2

What are your employee and customer pain points?

CREATIVITY

Bring generative assistance into your custom app workflows with your data for any user

Your own copilots

REASONING

Aggregate and reason over multiple data sources to **enrich decisions** and customer interactions

4

Information discovery

TAILORING

Create recommendation engines that **enhance products and services** or launch new offerings

Hyper-personalization

5

The core dimensions of implementing generative AI effectively

Model choice

Select the right models **for your use case**, benchmark and test them with your data

Experience quality

Build high-quality LLM-based applications with **information retrieval** and prompt engineering



Total trust

Innovate confidently on top of data privacy, security, compliance, and critical **content safeguards**

Streamlined lifecycles

Operationalize and scale the **management** of your GenAl apps as they evolve over time

"Through the incorporation of a lease report generator into our fleet Al system...we have revolutionized a time-consuming task that previously took 4 hours, reducing it to just 5 minutes."



大成 DENTONS

Sam Chen, Dentons Legal Al Adoption Manager

Chat with your data

THE CHALLENGE

Expedite summarization of legal contracts and documentation for the largest global law firm in the world

THE INNOVATION

fleetAl, an internal chatbot that helps lawyers conduct research, generate content and identify arguments fast

THE TECHNOLOGY

- Azure OpenAl Service GPT-4
- Meta Llama 2

Approaching your optimal model selection

Content Generation-

Empower your users with Al-generated content based on natural prompt commands

> Summarization Text generation Image generation Natural language to code

Multimodality

Increase **user engagement** with rich interactions that integrate vision, speech, and text

> Chat Image to text Text to image Video to text

Fine-tuning

Enhance model performance using your own data and adapt nimbly to feedback

Question answering Text classification Token classification Zero-shot image classification

Build Capabilities Across Model Families



Microsoft Research **Model Family**



Azure OpenAl **Model Family**



Meta Llama 2 **Model Family**



Hugging Face Model Family



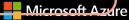
Mistral Al



NVIDIA Model Family



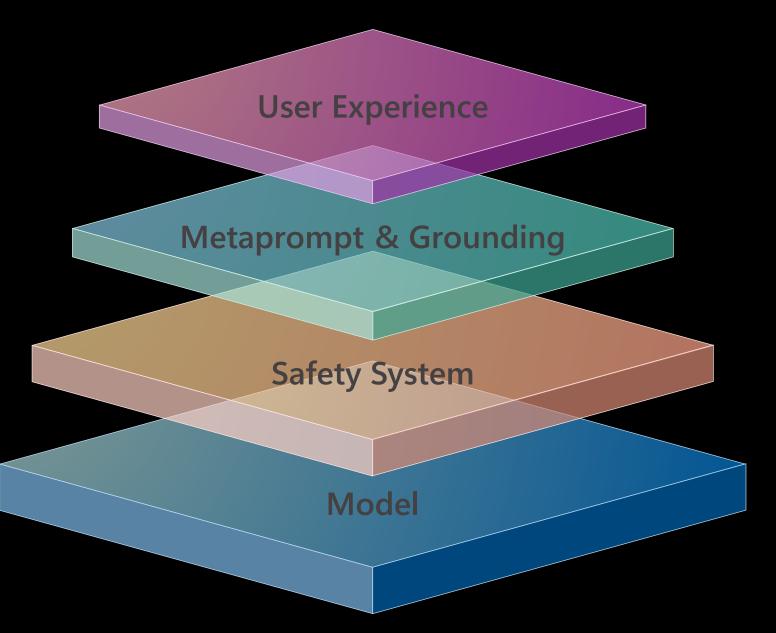
Deci Al **Model Family**





Model Family

Mitigation layers



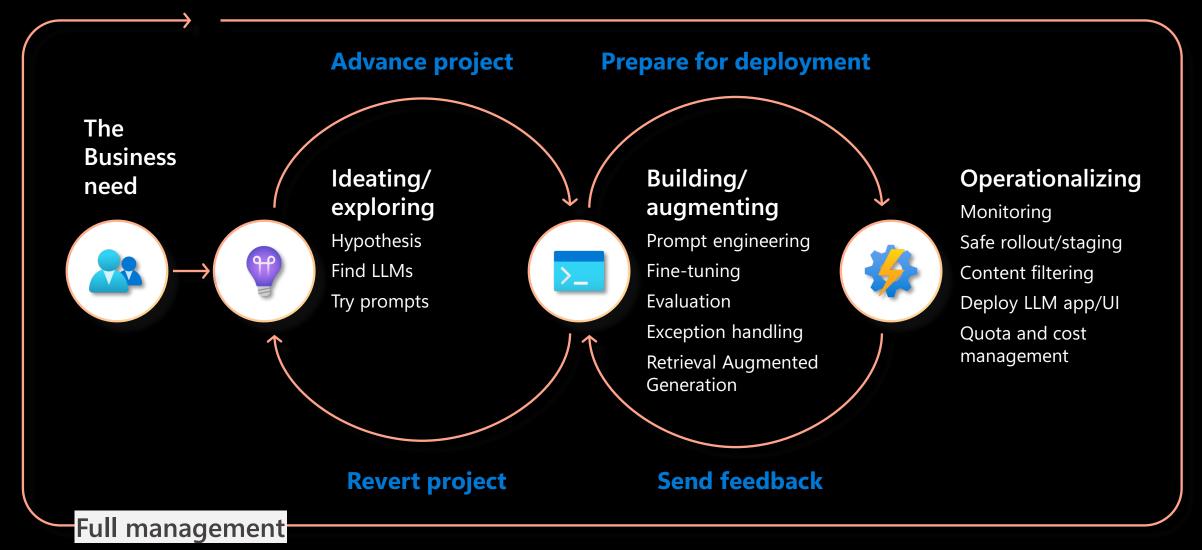


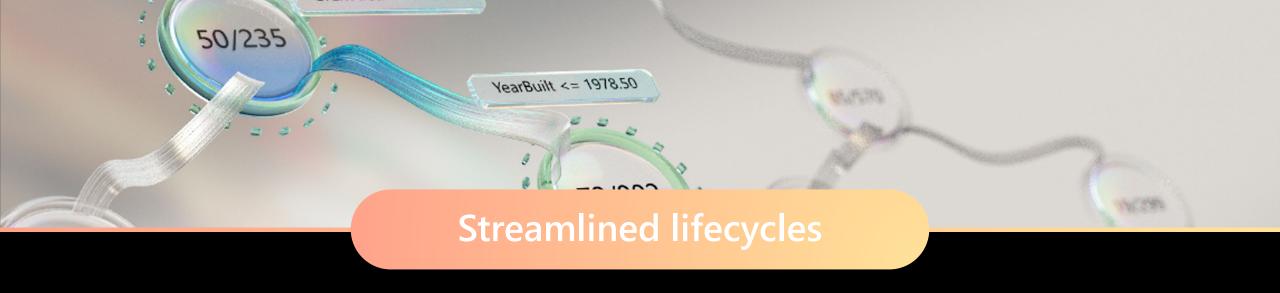


Responsible Al



An LLM lifecycle in the real world





Unlock repeatable value and operationalize Al



Streamline the entire development process with intuitive **prompt flows** for iteration and collaboration



Continue to refine your LLM apps with flexible **prompt engineering** that helps you adapt to human feedback



Apply <u>responsible Al</u>
throughout the
lifecycle to build fair,
explainable, and highly
performant apps



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Operationalize and scale the **management** of your GenAl apps as they evolve over time

Generative Al-based experiences will elevate all aspects of digital life





Thank You!

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