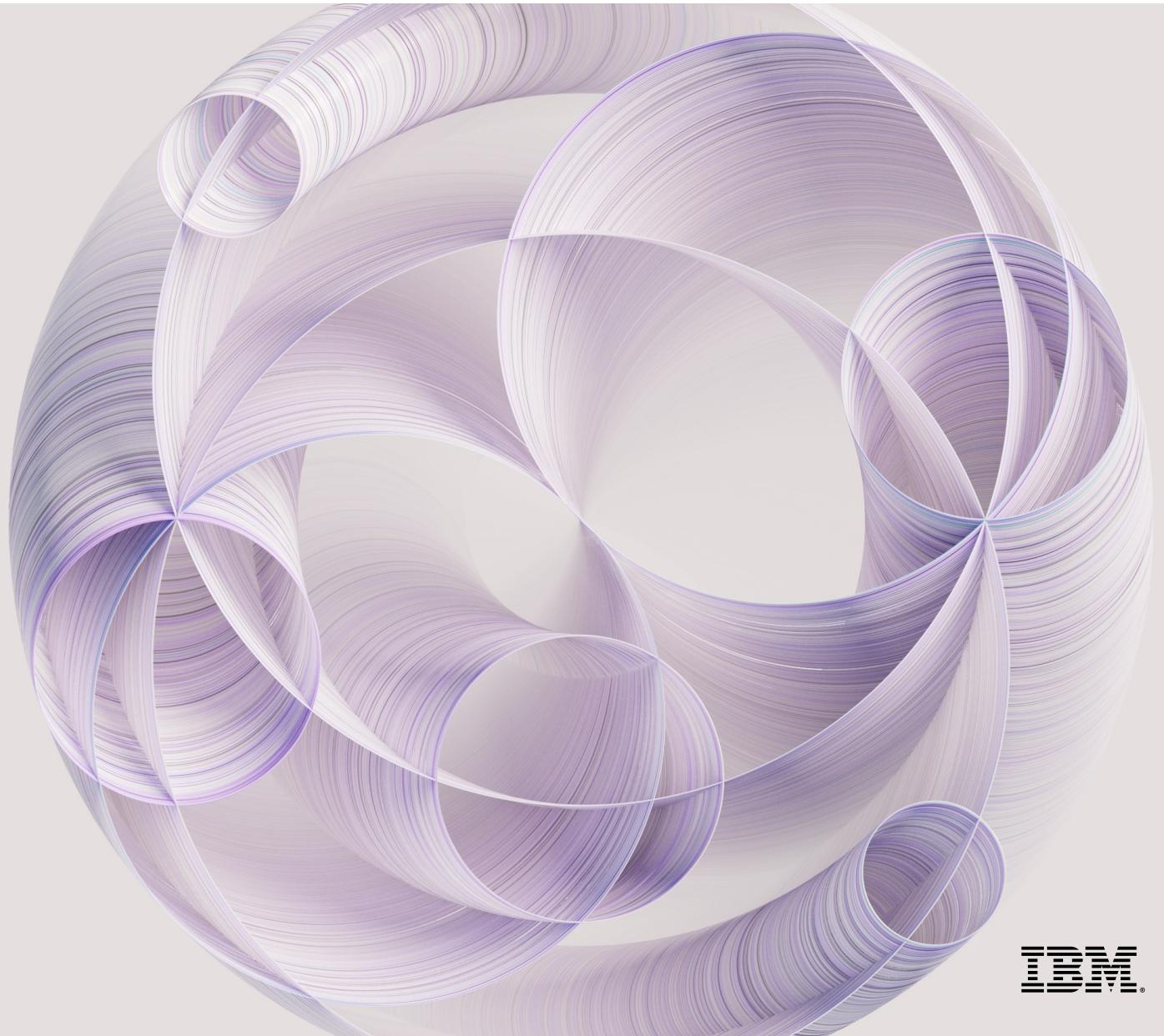


Accelerate  
responsible,  
transparent and  
explainable AI  
workflows



watsonx.governance

IBM®

Foundation Models  
are bringing an  
inflection point in  
AI...

...but how enterprises  
adopt and execute will  
define whether they  
unlock value at scale

## The impact of generative AI |

The opportunity

The speed,  
scope, and scale  
of generative AI  
impact is  
unprecedented

Massive early  
adoption

80%  
of enterprises are working  
with or planning to  
leverage foundation models  
and adopt generative AI

Broad-reaching  
and deep impact

Generative AI could raise  
global GDP by

7% within 10 years

Critical focus of AI  
activity and  
investment

Generative AI expected  
to represent

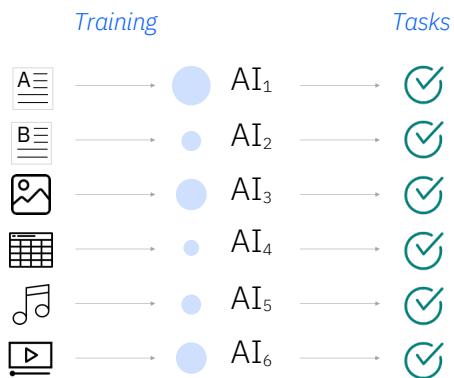
30%  
of overall market by 2025

Sources: Statista; Reuters; Goldman Sachs; IBM  
Institute for Business Value; Gartner. Scale Zeitgeist:  
AI Readiness Report, a survey of more than 1,600  
executives and ML practitioners

## The impact of generative AI | The opportunity

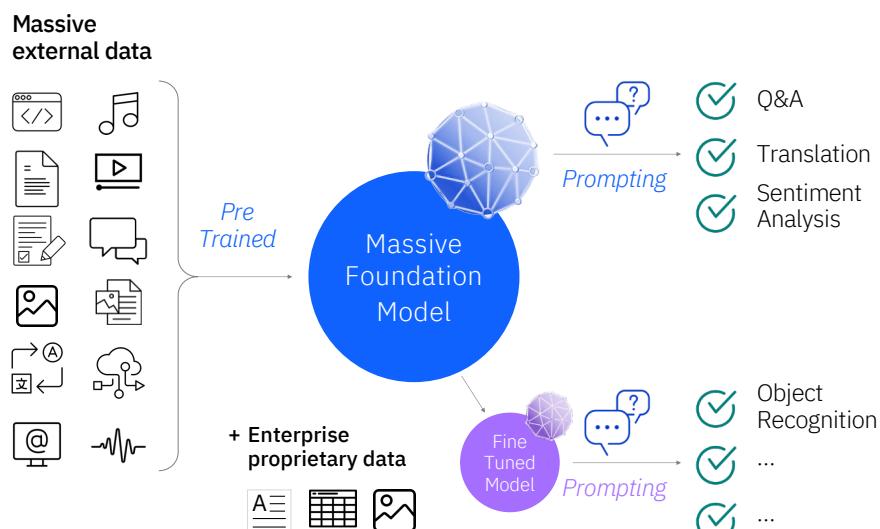
# Foundation models establish a new paradigm for AI capabilities

## Traditional AI models



- Individual siloed models
- Require task specific training
- Lots of human supervised training

## Foundation Models



## Enhanced capabilities

- Summarization
- Conversational Knowledge
- Content Creation
- Code Co-Creation

## Key advantages

- **Lower upfront costs** through less labeling
- **Faster deployment** through fine tuning and inferencing
- **Equal or better accuracy** for multiple use cases
- **Incremental revenue.** through better performance

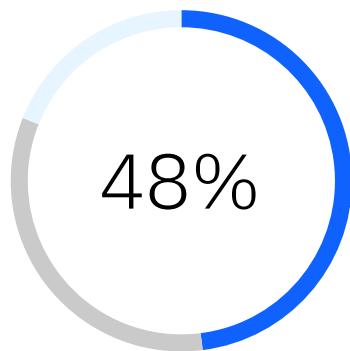
up to **70% reduction**  
in certain NLP tasks

## Enterprise considerations

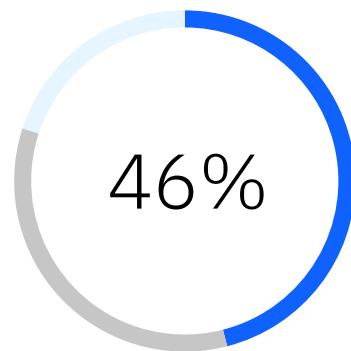
Business leaders face challenges in scaling AI across the enterprise with trust

80% of business leaders see at least one of these ethical issues as a major concern

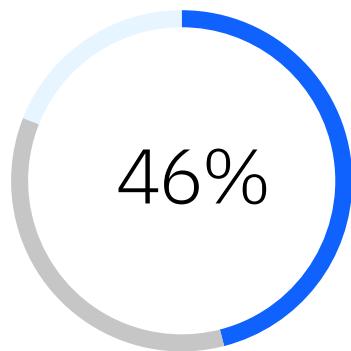
Explainability



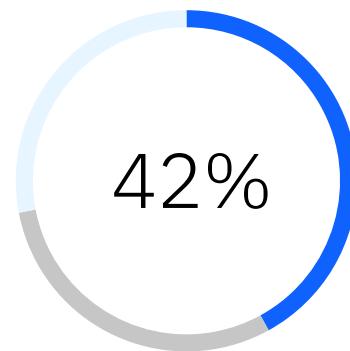
Ethics



Bias



Trust



Believe decisions made by Generative AI are not sufficiently **explainable**.

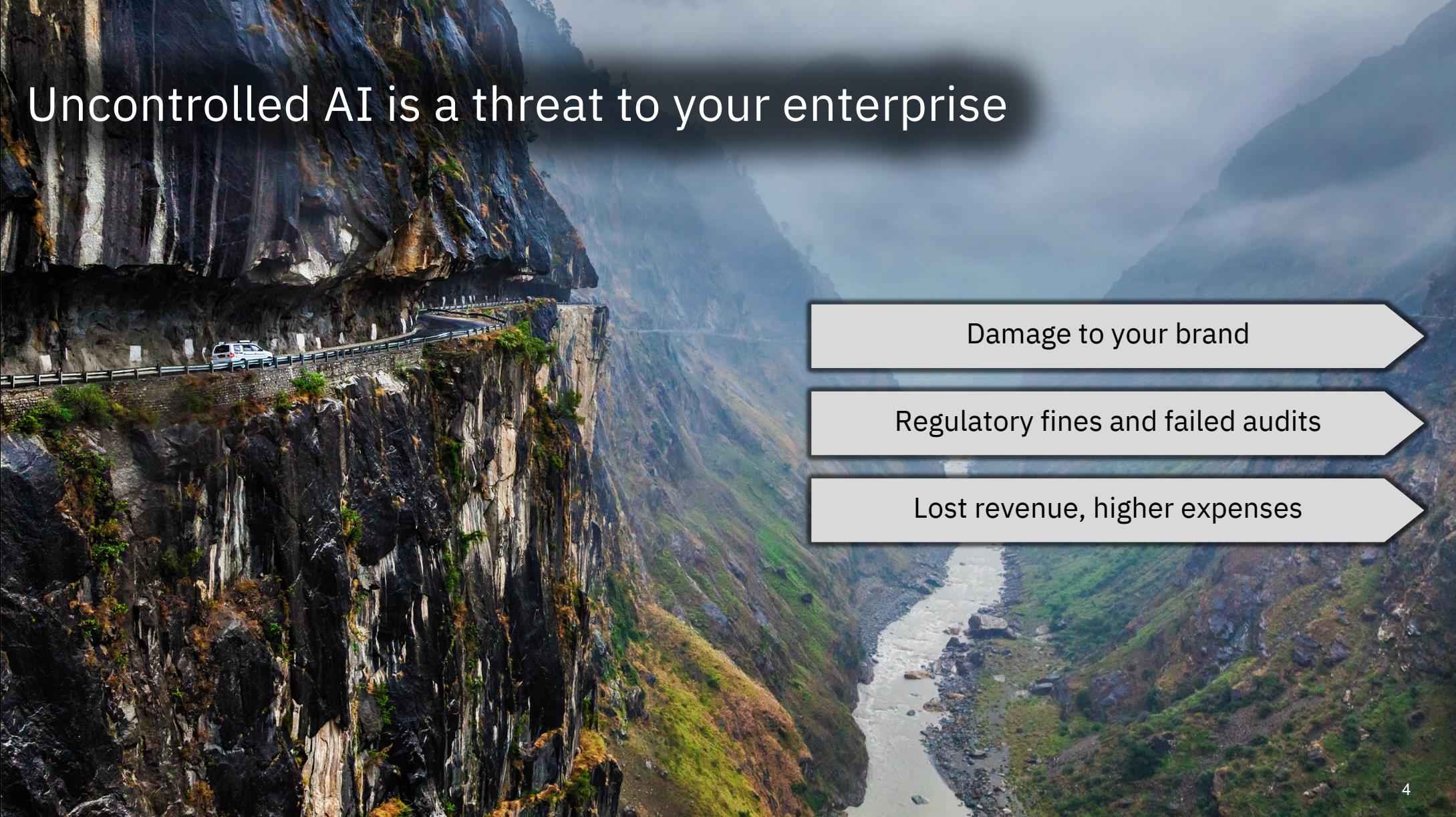
Concerned about the safety and **ethical** aspects of Generative AI.

Believe that Generative AI will propagate established **biases**.

Believe Generative AI cannot be **trusted**.

Source: IBM IBV "Generative AI: The state of the market", June 2023

Agree   Neutral   Disagree



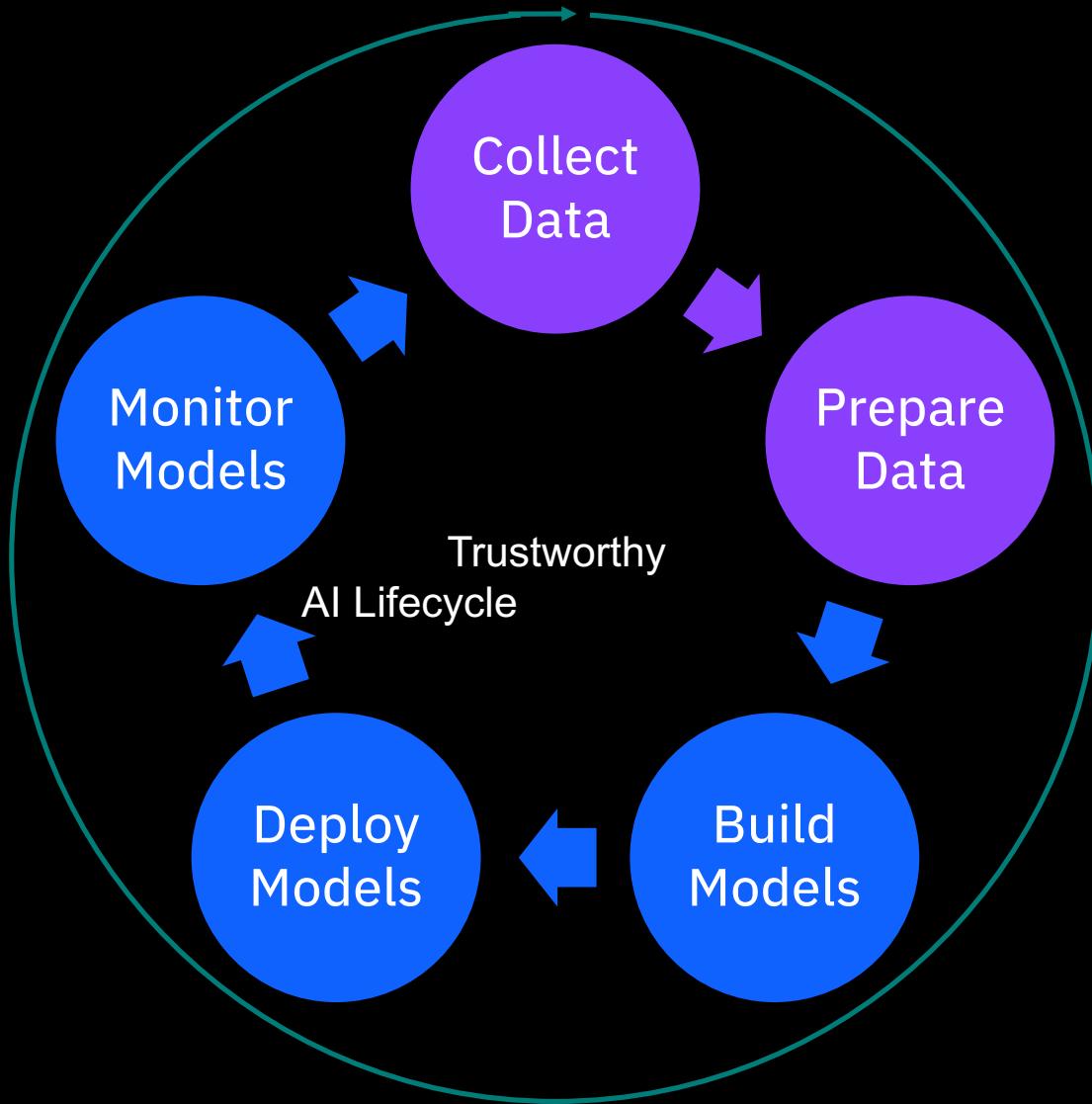
# Uncontrolled AI is a threat to your enterprise

Damage to your brand

Regulatory fines and failed audits

Lost revenue, higher expenses

# MLOps and Trustworthy AI

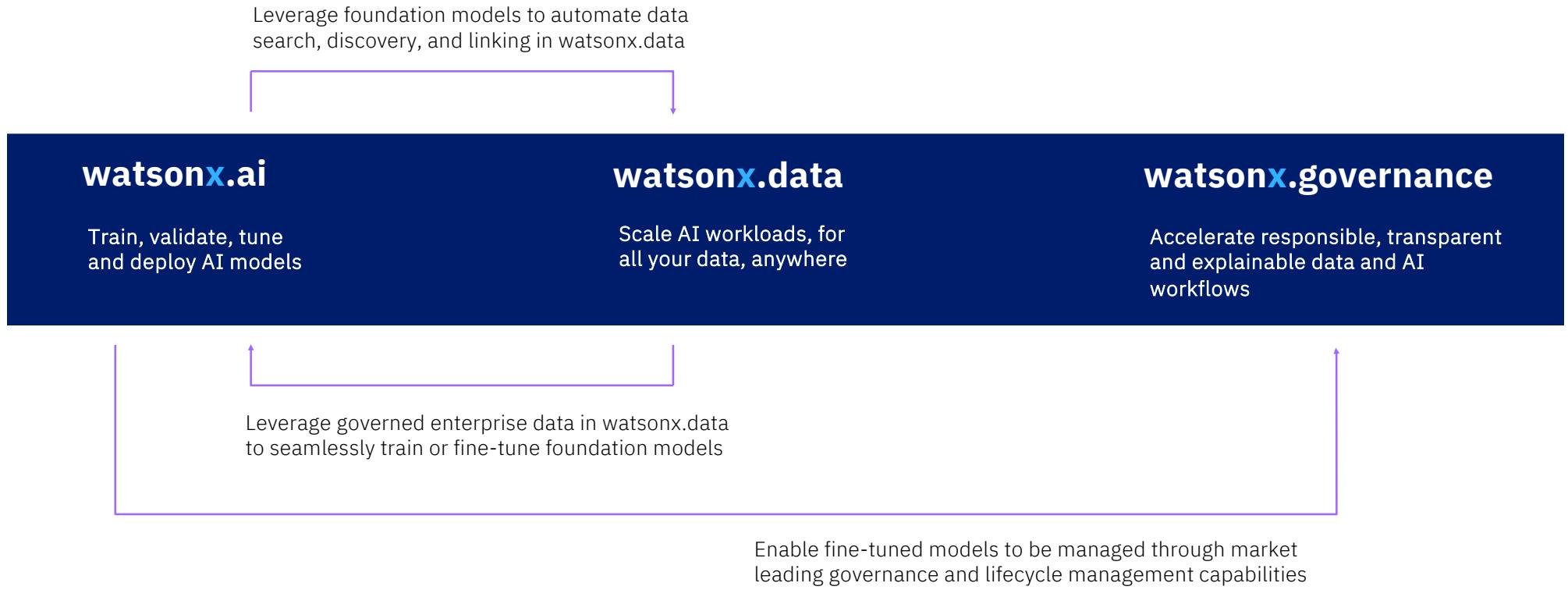


- **Data** : A complete view of quality data that is private, self-served and ready for analysis by multiple personas
- **Model** : MLOps infused with fairness, explainability, and robustness
- **Process** : Automation to drive consistency, efficiency and transparency for AI at scale

AI needs governance –  
the process of directing,  
monitoring and managing the  
AI activities of an organization

# Put AI to work with **watsonx**

Scale and accelerate the impact of AI with trusted data



# IBM watsonx.governance



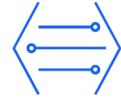
end-to-end automated AI  
lifecycle governance toolkit  
built to mitigate risk and  
improve compliance

## watsonx.governance

Accelerate responsible,  
transparent and  
explainable AI  
workflows



- Govern across the AI lifecycle.
- Automate and consolidate tools, applications and platforms.
- Capture metadata at each stage
- Support models built and deployed in 3<sup>rd</sup> party tools.

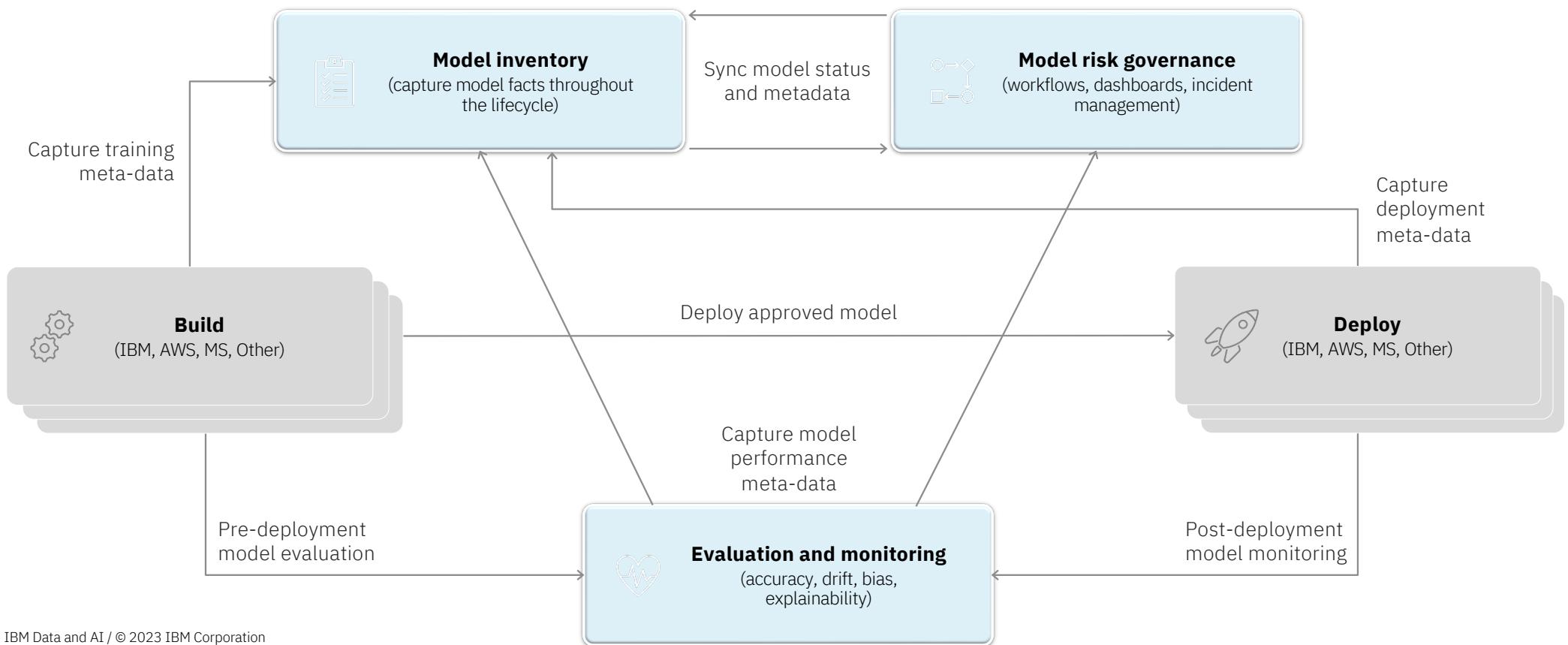


- Manage risk and protect reputation by automating workflows to ensure quality and better detect bias and drift.

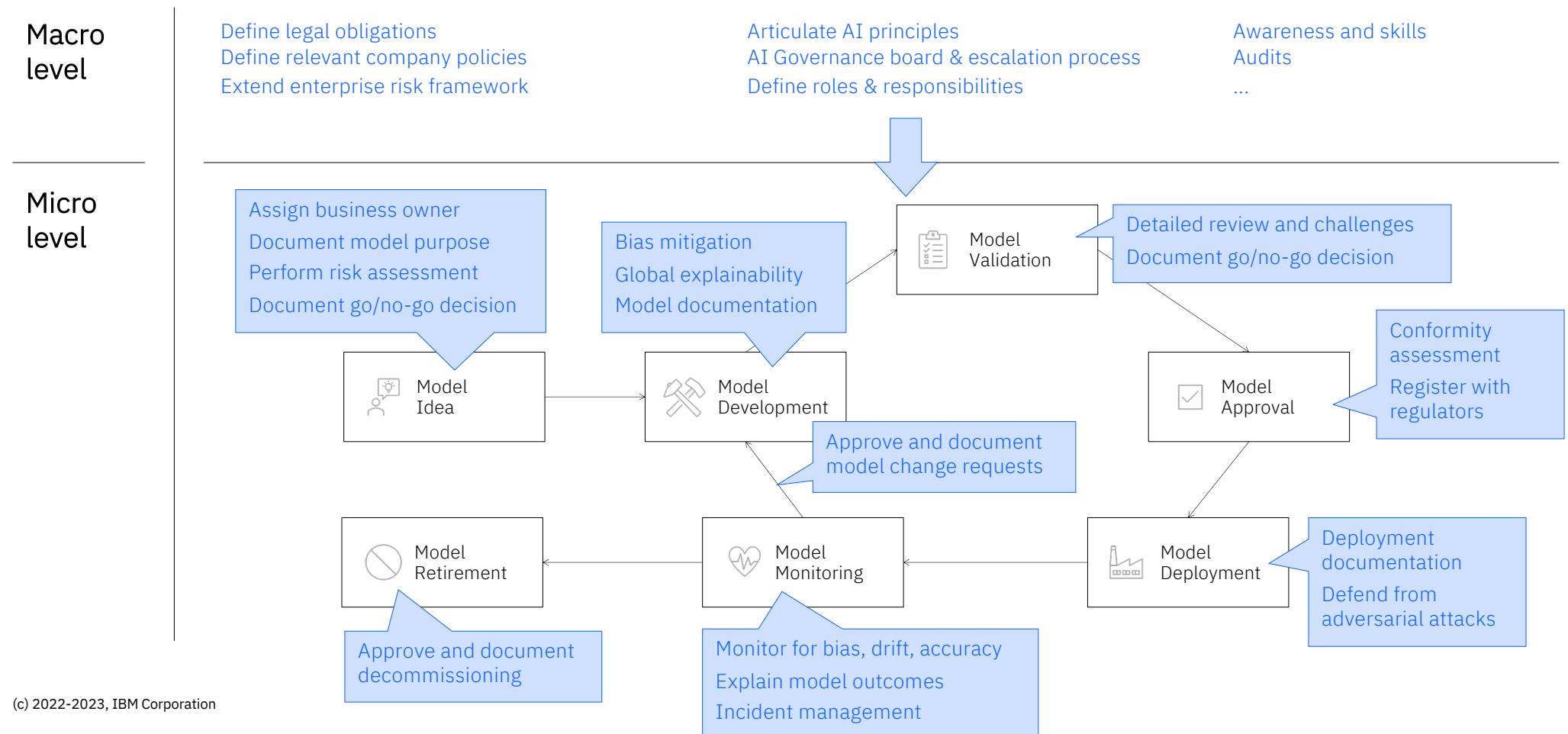


- Adhere to regulatory compliance by translating growing regulations into enforceable policies.
- Helps you to operationalize AI with confidence

# Govern across the AI lifecycle



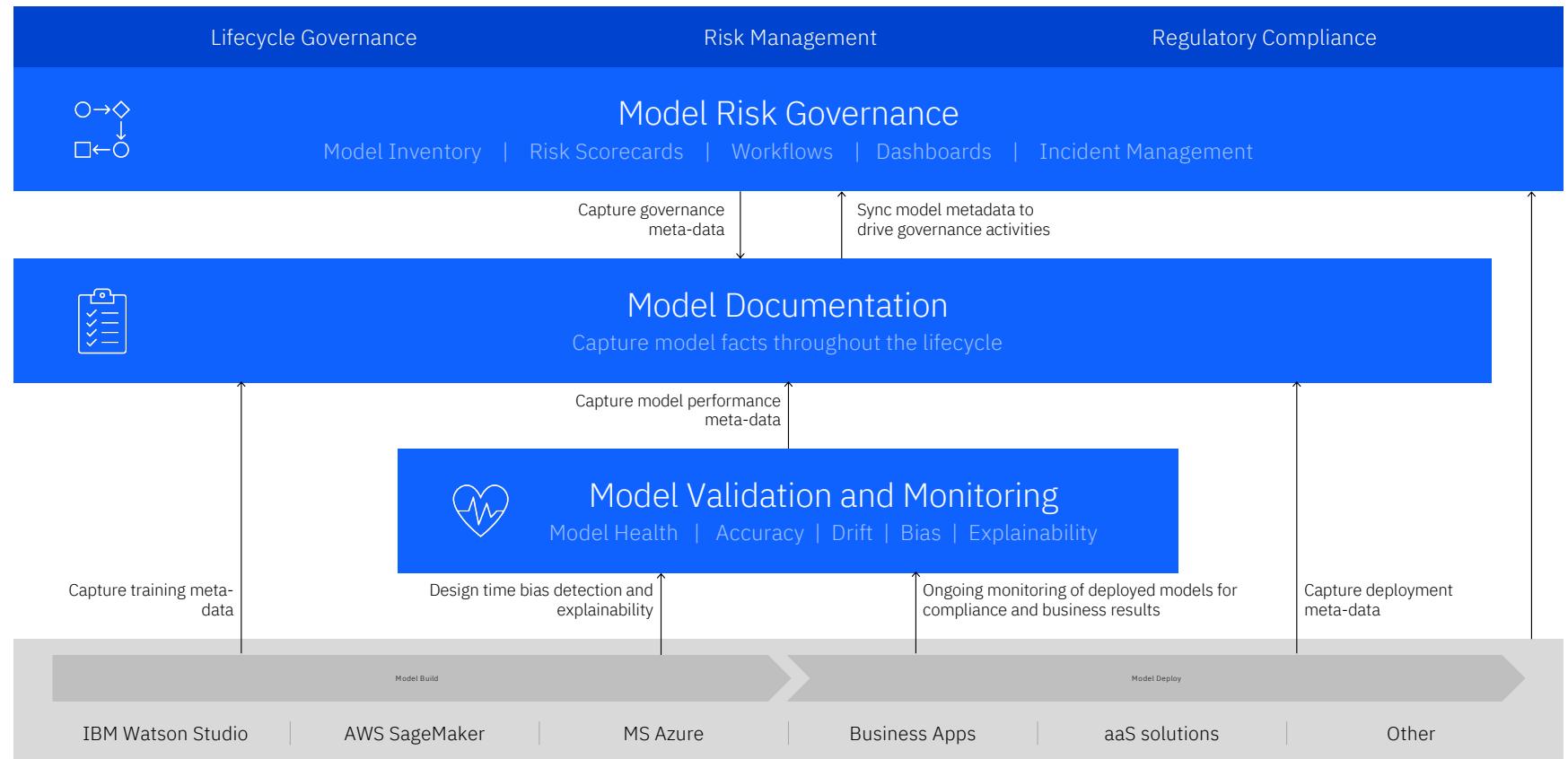
# In practice, that means...



# IBM AI Governance

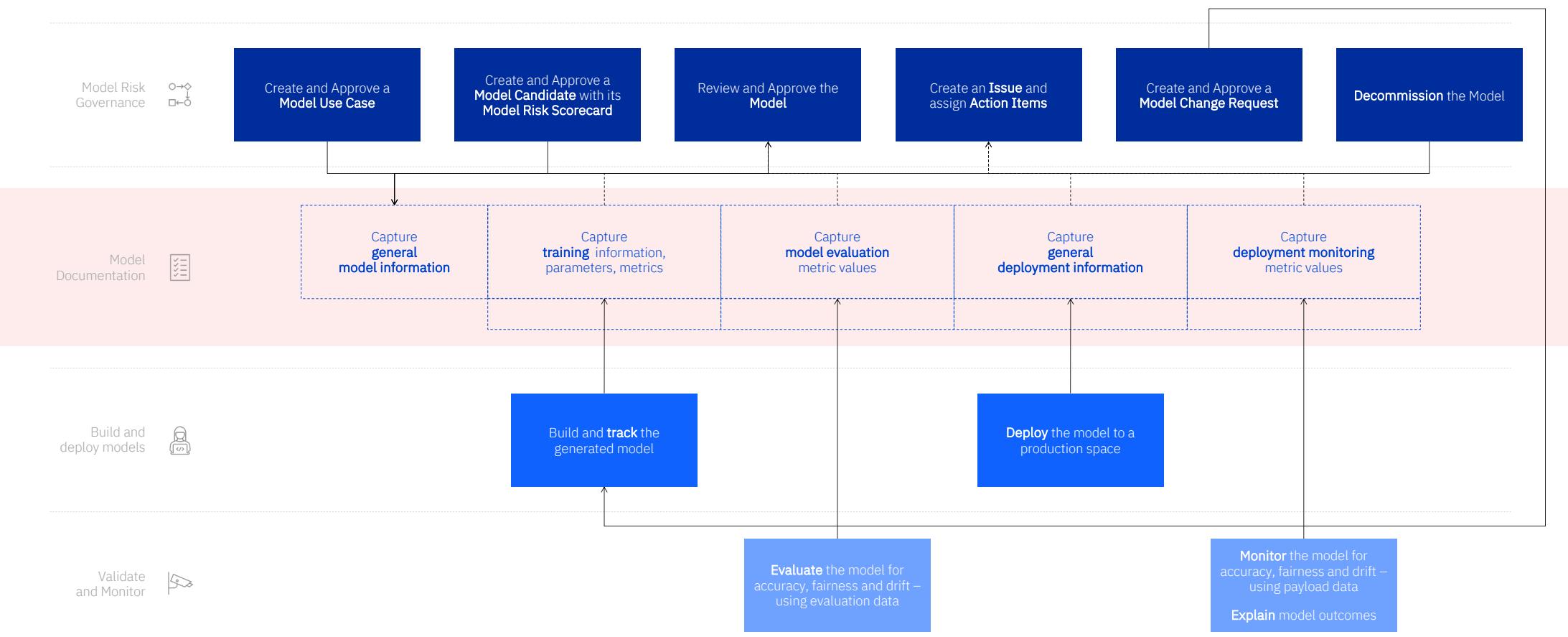


- Model Owners
- Model Validators
- Audit Teams
- Compliance Teams
- Risk Management Teams
- Data Privacy Teams
- Principal Data Scientists



- Data Engineers
- (Citizen) Data Scientists
- MLOps
- ML Engineer

# Lifecycle Governance – Model documentation



# IBM AI Governance

*Role of the solution components in the three use cases*

		Solution components				
		OpenPages		FactSheets	OpenScale	
Use cases	Lifecycle Governance	Objects: •Model •Model Scorecard •Change Request •Review •Challenge	Workflows: •Model Request •Model Risk Assessment •Model Candidate •Model Development •Model Review •Model Change Request •Model Decommission	Dashboards widgets: •Models by Department •Models by Provider •Models by Lifecycle Stage •Model Reviews •Model Change Requests	•Capture training facts •Capture deployment facts •Capture evaluation facts	•OpenScale Python API to evaluate model during development •OpenScale MRM capability to evaluate model during validation •OpenScale production monitoring
	Risk Management	Objects: •Metric •Metric Value •Issue •Action Item	Workflows: •Metric Value Review •Issue Review •Action Item Approval	Dashboards widgets: •Policies •Models by Risk Tier •Model Metric Status •Development Accuracy	•Capture monitoring facts •Investigate metric alerts •Capture training facts for challenger models	•Post-deployment model monitoring •Investigate metric alerts •Pre-deployment evaluation of challenger models •Compare champion and challenger models
	Regulatory Compliance	Objects: •(Sub)Mandate •Policy •Requirement	Workflows: •Model Review on occurrence of major alert •Model checks before deployment to production, etc.	Dashboards widgets: •Mandates •Requirements •Obligations	•Model report export	•Model report export

# What is automated between the AI Governance components?

User action	Automation
Create a new Model Use Case in OpenPages	<ul style="list-style-type: none"><li>• Create a factsheet for the use case</li><li>• Sync general information from OpenPages to Factsheets</li><li>• Create a hyperlink in OpenPages</li></ul>
Train a model in Python	<ul style="list-style-type: none"><li>• Factsheet Python client: Auto-logging model training facts from common open-source frameworks</li><li>• OpenScale Python client: design time bias detection and explainability</li></ul>
Train a model in AutoAI	<ul style="list-style-type: none"><li>• AutoAI automatically saves model training facts in the model object (incl. Fairness metrics if you use that)</li></ul>
Track a model in Factsheets	<ul style="list-style-type: none"><li>• Create a new Model object in OpenPages (or map to an existing one)</li><li>• Sync training metadata to OpenPages</li></ul>
Deploy a model from Watson Studio to WML	<ul style="list-style-type: none"><li>• Capture deployment metadata in Factsheets</li><li>• Create a new Deployment object in OpenPages</li><li>• Sync deployment metadata to OpenPages</li></ul>
Evaluate a model in OpenScale	<ul style="list-style-type: none"><li>• Capture monitoring metadata in Factsheets</li><li>• Sync monitoring metadata to OpenPages</li></ul>
Evaluate a model using custom tech/metrics	<ul style="list-style-type: none"><li>• Capture external metrics in OpenScale</li><li>• Capture monitoring metadata in Factsheets</li><li>• Sync monitoring metadata to OpenPages</li></ul>

Risk is every one's business.

In today's turbulent environment, the need to take on risk with confidence is greater than ever before.

# 90%

of compliance leaders expect evolving business, regulatory, and customer demands to increase compliance-related operating costs by up to 30%.<sup>1</sup>

# 79%

of organizations report that keeping up with the speed of digital and other transformations is a significant risk management challenge.<sup>2</sup>

# 77%

of organizations recognize the need to upgrade their Third-Party Risk Management operating model.<sup>3</sup>

Source: 1. 2022 Compliance Risk Study Report, Accenture

Source: 2. 2022 Global Risk Survey, PwC

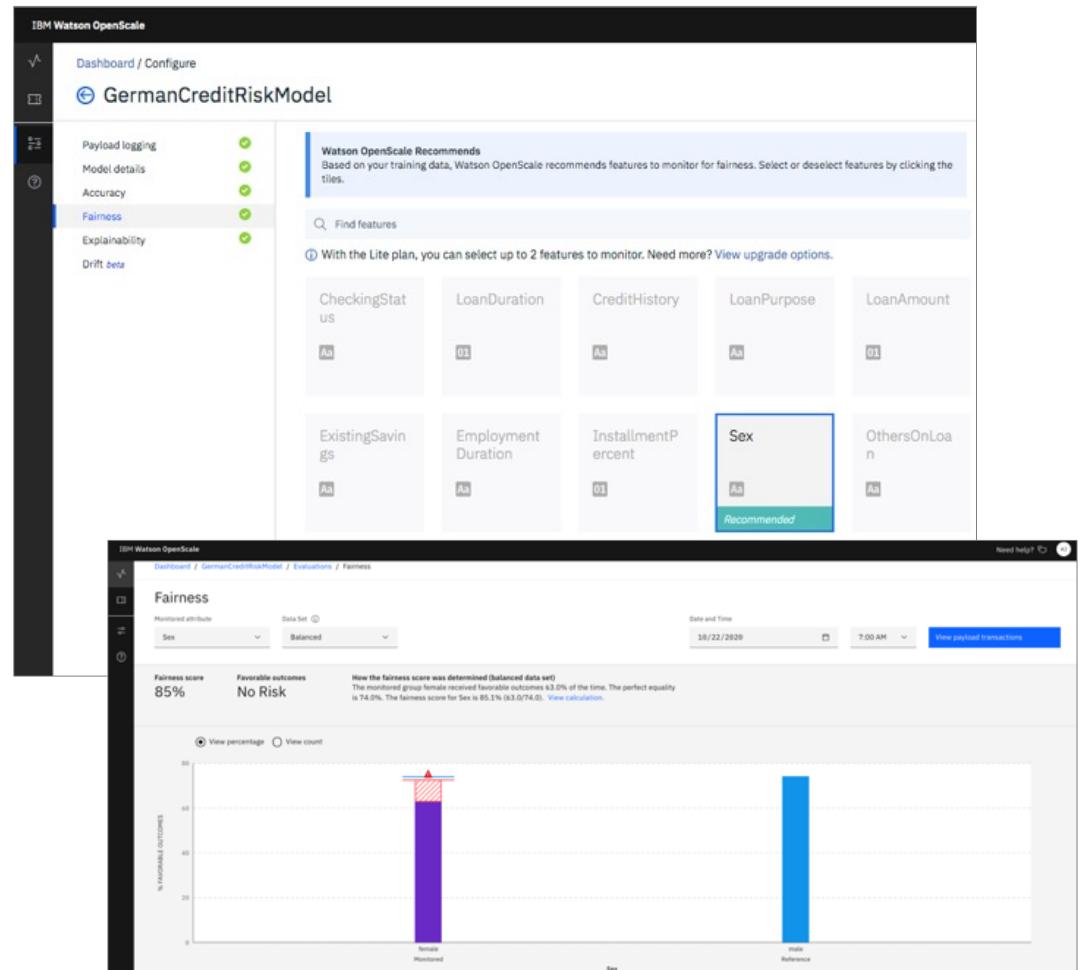
Source: 3. 2022 Third-Party Risk Management Outlook, KPMG

# Bias Detection

Watson OpenScale enables enterprises to enforce fairness in their model's outcome by analyzing transactions in production and finding biased behavior by the model. It pinpoints the source of bias and actively mitigates the biases found in production environment.

## Value:

- Automatically recommend common protected attributes to monitor during production
- Detect biases in runtime in order to catch impacts on business applications and compliance requirements without time consuming, manual data analysis
- Metrics and data to help data scientists further troubleshoot issues in data sets or models
- Mitigate biases in runtime in order to enforce regulatory or enterprise fairness guardrails in real time



# Explainability

Watson OpenScale records every individual transaction and drills down into its working to explain how the model makes decisions.

It provides a simple explanation that is user friendly and interactive

Value:

- Explain individual transaction level decisions made by the model in run time, including details about most important attributes and their values in order to assist in compliance and customer care situations
  - Analyze individual transactions in a what-if manner in order to understand how model behavior will change in different business situations
  - Provide global explanation for decisions made by the model using SHAP



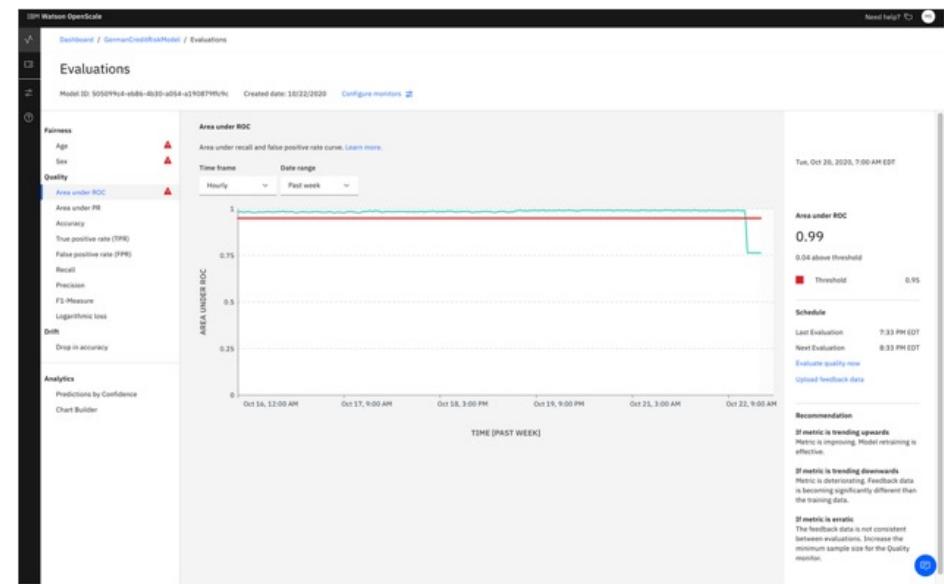
# Quality

Monitor model quality metrics like accuracy, precision, recall, AUC, etc. and get alerts when the value goes beyond the set threshold

Can also define custom metrics to track quality for

## Value:

- Observe and track quality metrics to determine how well your model predicts outcomes
- Use custom metrics to create and track user-defined metrics to assess the quality of models



Quality									
Area under ROC	Area under PR	Accuracy	True positive rate (TPR)	False positive rate (FPR)	Recall	Precision	F1-Measure	Logarithmic loss	Gini
0.76	0.69	0.81	0.62	0.09	0.62	0.78	0.69	0.43	
Prediction									
		No Risk			Risk			Total	
Actual		No Risk	120		12	132			
		Risk	26		42	68			
		Total	146		54	200			

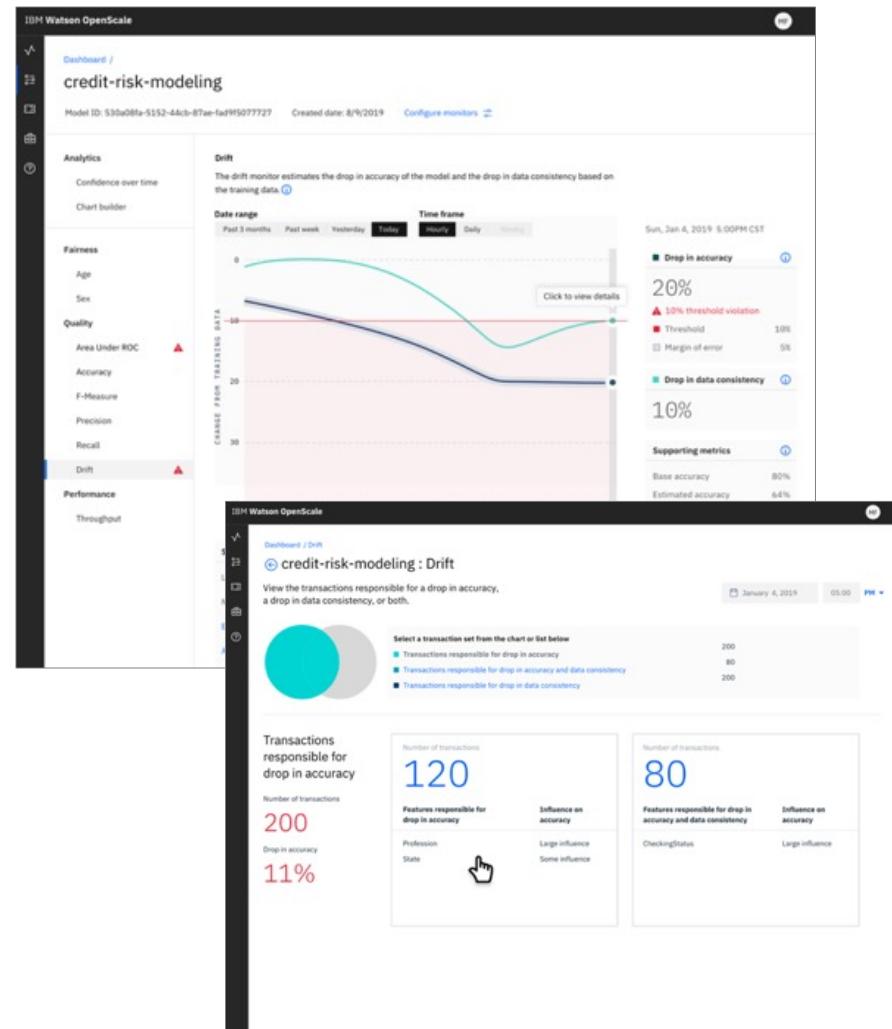
# Drift

Automatically detect drifted transactions and pinpoint datapoints that contribute to drift

Identify drift in Accuracy as well as drift due to data consistency

## Value:

- Business environments are dynamic leading to “drift” in data and cause inaccuracies in model prediction. Monitor drift to make sure model is resilient and updated for such changing conditions
- Drill down into transactions causing drift to identify root cause and take appropriate actions



# Model Validation/Model Risk Management

OpenScale enables enterprises to validate pre-production models before putting them into production to ensure they can be trusted to perform as intended.

- Validate pre-production models and generate reports of outcomes
- Enable customizable tests relevant to AI models
- Compare performance of models
- Automatically configure monitoring of production models to match pre-production settings.
- Synchronize results with Governance, Risk and Compliance (GRC) solutions (Initially with OpenPages Model Risk Governance)
- Integrate with AI Factsheets to capture metadata and track metrics

The screenshot displays two main windows of the IBM Watson OpenScale platform.

**Credit Risk Evaluation Dashboard:** This window shows a summary of model validation results. It includes a circular progress bar indicating 3 tests run (3 passed, 1 failed), a Fairness score of 90% (green), and a Quality score of .99 (green). Below these are detailed tables for Fairness by feature (Age, Sex, Race) and Quality metrics (Area under ROC, Area under PR, Accuracy, True positive rate (TPR)).

**Compare model Overlay:** A modal window titled "Evaluate Compare model" allows users to select a model for comparison. It lists "Credit Risk V2" and "Credit Risk V1" in a dropdown. A grid compares their performance across various metrics: Fairness (Age, Sex), Quality (Area under ROC, Area under PR, Accuracy, True positive rate (TPR)), and Performance measures (Throughput).

**Send to OpenPages Overlay:** A modal window titled "Send to OpenPages" lists various metrics for selection. The "Quality measures" section includes: Area under ROC, Area under PR, Accuracy, True positive rate (TPR), False positive rate (FPR), Recall, Precision, F1-measure, and Logarithmic loss. The "Fairness measures" section includes: Fairness, Age, and Sex. The "Performance measures" section includes: Throughput. The "Drift measures" section includes: Drop in accuracy, Drop in data consistency, Estimated accuracy, and Base accuracy. At the bottom are "Cancel" and "Schedule" buttons.

# Policy Packs

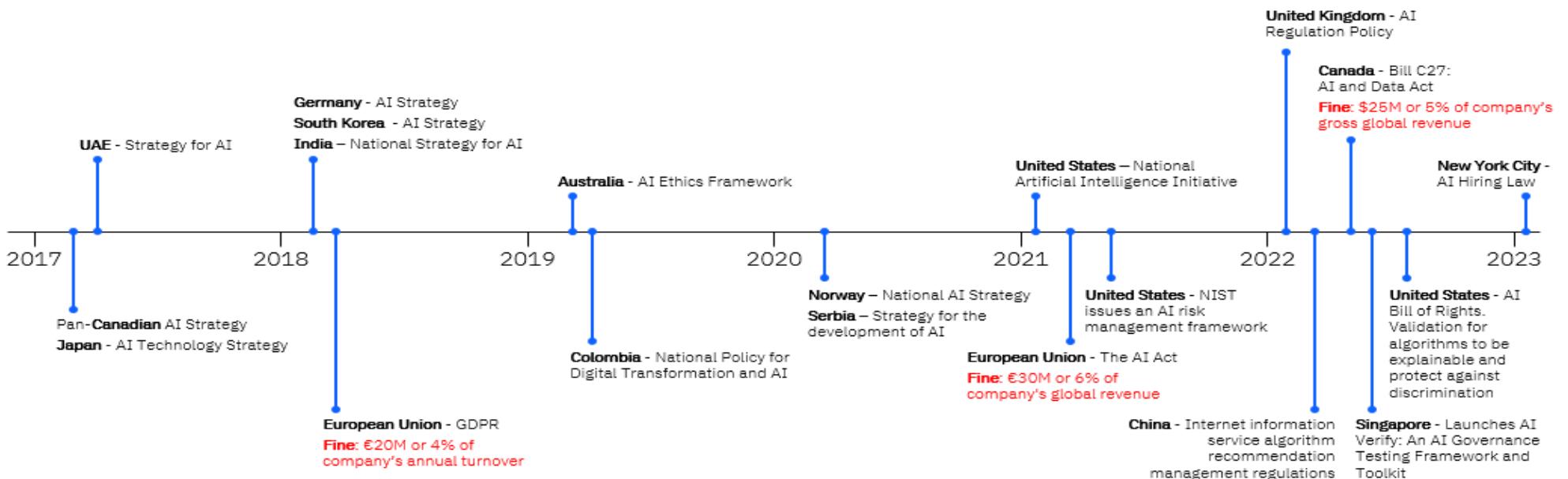
- Pre-built assessments to give clients a “starting point” to aid in regulatory compliance
- Ability to determine **applicability** of model use cases against various regulations
- Receive **guidance** on how to comply
- Content is provided and curated through partnership with **regulatory experts**
- We will support selected regulations (e.g. [NYC Local Law 144](#), SR-11-7, [EU AI Act](#), etc.)



The proposed EU AI Act includes potential monetary penalties of €30M or 6% of company's global revenue<sup>2</sup>

Source:  
2. [EU set to ratchet up AI fines to 6% of turnover](#), Reuters

# Adhere to growing and changing AI regulations



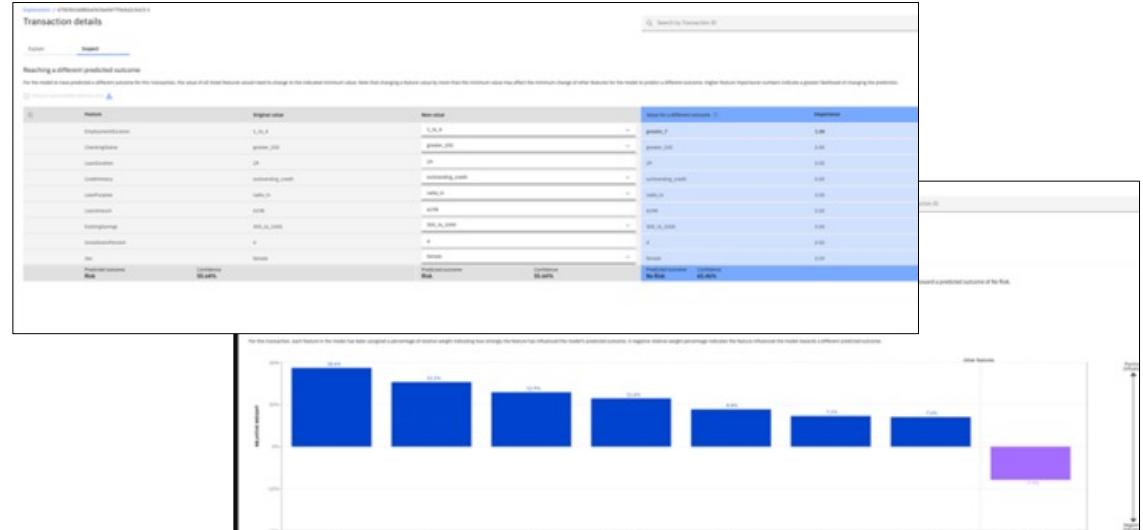
# Adhere to growing and changing AI regulations

## Improve AI regulatory compliance:

- Align processes and actions with your organizations' business goals and external legal and regulatory requirements.
- Automate the translation of AI regulations into enforceable policies and standards, quickly respond to change
- Provide persona-based dashboards and reporting to key stakeholders on the status of compliance

## Document model facts using factsheets

- Support compliance audits, and requests for model details required by auditors, management, stakeholders and customers
- Explain transaction level decisions of models in runtime
- Understand how the model will behave in different business situations
- Facilitate subsequent enterprise validation based on model facts



## Regions Bank – transparent AI with AI governance

### Business challenge:

Regions Bank sought to address revenue loss to attrition among private wealth clients. Using trained models, the bank predicts customer risk to mitigate attrition.

### Solution:

Enhance ML pipelines for Private Wealth Retention models to detect drift in model accuracy and data consistency in production data lake. The solution used IBM Expert Labs, IBM AI Governance, and IBM Watson Studio.

### Results:

- Standardize monitoring of models in the end-to-end ML pipelines
- Continuous read on accuracy of predictions in production
- Automate retraining of models based on monitoring alert
- Identify wealth clients with low confidence predictions



Industry: Banking and Financial Markets

Geography: North America

# Competitive overview

## Traditional data science players

### How is IBM better?

1. Most players lack GRC workflow capabilities such as OpenPages.
2. Automatic fact capture coupled with customizable reporting capability through Factsheets.
3. Monitoring and governance on these platforms tend to work with model developed on their platform only.

## Niche players

### How is IBM better?

1. Most niche players don't have a large or technical enough support and services team to guide customers.
2. Very few niche players focus on the E2E model lifecycle and lack a singular view for all types of model metadata.
3. IBM fuels its AI Governance stack with state-of-the-art algorithms and frameworks from IBM Research.

## Open source

### How is IBM better?

1. Most opensource tools tend to be very technical and tend to ignore the business user such as risk analysts.
2. Although state-of-the-art, opensource tools in the AI Governance space are difficult to maintain, requiring certain skills and resources.

## Consulting firms

### How is IBM better?

1. Consulting firms may develop one-off solutions that are catered to the customers current data science stack leading to heavily customized code, which is difficult to maintain. Leveraging consulting firms for end-to-end technological implementations is costly.

