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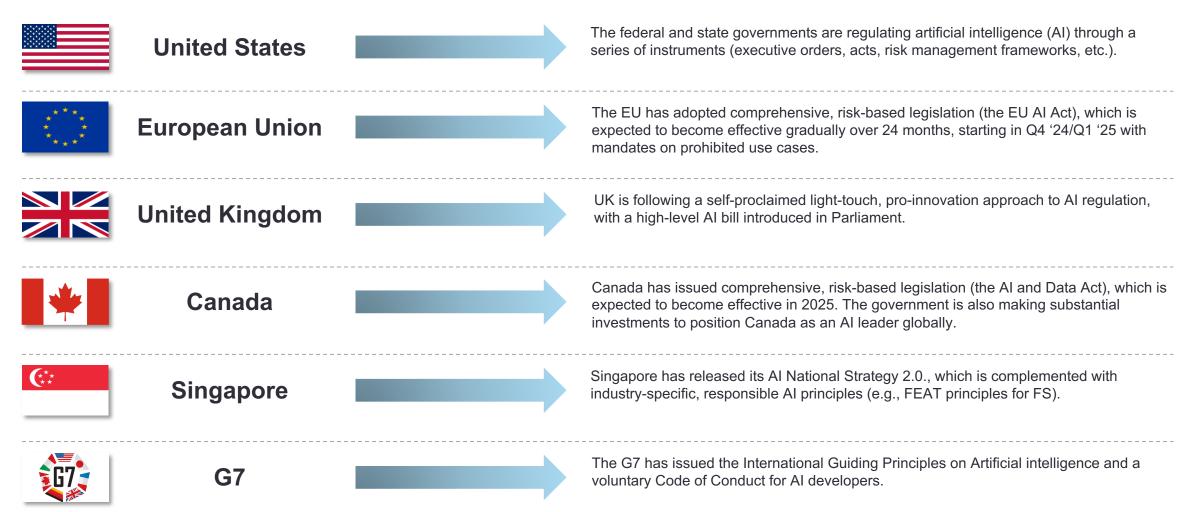
02 Emerging compliance trends

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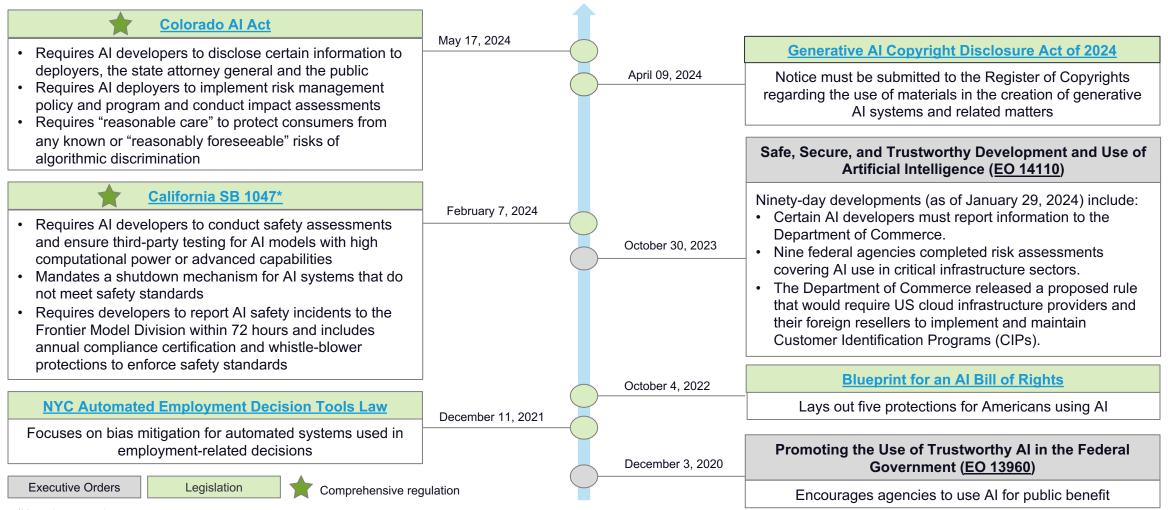
# Summary of Al Regulatory and Policy Developments





### **Key US AI Regulatory and Policy Development**





Yet to be enacted

## Al Regulatory Trends and Expectations for US Banks





#### **Leading practices**

Global jurisdictions

Professional organizations

Sector & usecase specific



#### **US** national regulatory actions

USTD

CFPB

FHFA, HUD

FINRA SEC



**Alignment with NIST** 

#### The What



Al governance framework



**Al inventory** 



Reporting on Al-specific risks

### **Generative Al Risk Taxonomy Continues to Evolve**



#### **Heightened Risks for Generative AI (NIST AI 600-1)**



#### Confabulation

Confidently stated but erroneous or false content



#### Data privacy

Leakage and unauthorized disclosure or de-anonymization of sensitive data

Intellectual property

infringement

Toxicity, bias and

homogenization GenAl models may produce toxic or

biased content due to inherent biases

in training data or feedback loops



#### Information integrity

Generation, exchange and consumption of content that may not be vetted, accurate, etc.



#### Value chain

Non-transparent or untraceable integration of upstream



components across the Al lifecycle



#### **<..>**

#### Security

Data capability

Existing data capabilities (e.g., data

modeling, storage, processing) and

data governance (e.g., lineage and

traceability) may not be sufficient for

fine-tuning and business use of GenAl

Training data and trained GenAl model may be leaked out of the institution or vendor platform due to cyber attack or adversarial prompt engineering



**Risk Carried Over from Existing Al Models** 

#### **Bias/fairness**

**Technology capability** 

GenAl adoption increases the

computational needs and therefore

potentially impacts the current use of

infrastructure by other business use

Large volume of training data used in pre-training may introduce bias and unfairness

Complex model and training process make it hard to identify and control



#### **Environmental**

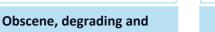
Negative ecosystem impact due to high resource usage in training models

abusive content

GenAl may intensify the spread of

harmful and explicit content.

heightening privacy/safety risks



#### Dangerous or violent recommendations

GenAl eases the ability for bad faith actors to generate violent, inciting or threatening content



### <u>≅</u>-Human-Al configuration

Arrangement or interaction of humans and AI systems that can result in algorithmic aversion, automation bias, overreliance, anthropomorphization, etc.



#### **CBRN** information

Eased access to nefarious information related to chemical. biological, radiological or nuclear (CBRN) weapons



#### **Business continuity**

Heavy reliance on third-party pretrained complex GAI, may aggravate the business continuity



#### **Explainability**

Large models can make the GenAl a black box, which lacks explainability

Source: NIST AI 600-1 publication



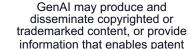








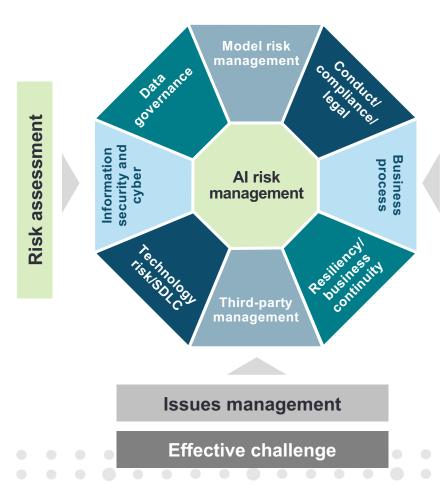
GenAl-powered threats and increased available attack surface for targeted cyber attacks

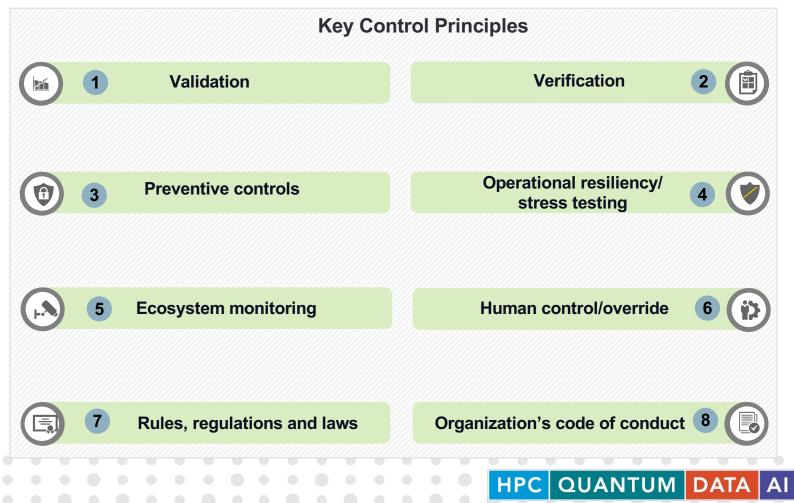


# Effective and Responsible Al Governance Requires Enterprise-wide Coordination



Key AI control principles establish the foundational principles for AI agnostic to the underlying use case/technique.





## Need for Tollgates at the Onset and Throughout the Generative Al Development Lifecycle



Tollgate 1 Tollgate 2 Tollgate 3 Tollgate 4 Tollgate 5

Initialization Data Solution development Validation Implementation

- Design and infrastructure feasibility
- Confirm organizational strategy and governance requirements
- Define situations/thresholds
- Input data assessment –
  solution specific and broader
  RAI (toxicity, PII personally
  identifiable information leakage, etc.)
- Red-teaming and benchmark evaluation

- Task-specific metrics: faithfulness, answer relevancy, golden data set F1
- Task-agnostic metrics: metacognition, harmlessness, logical robustness

- · Benchmark evaluation scores
- Red-teaming evaluation scores
- Re-evaluation of solution testing outputs

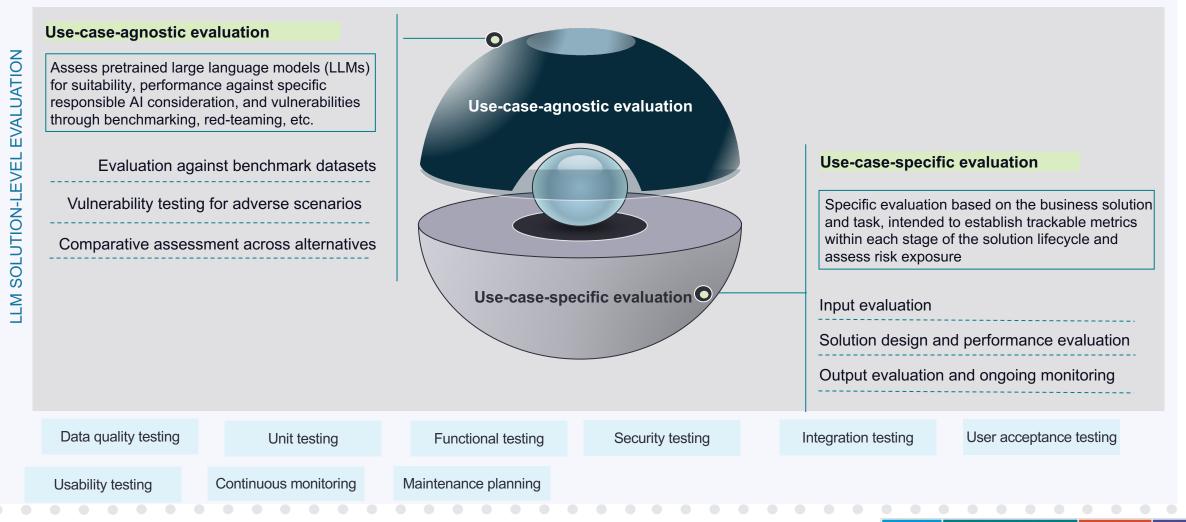
- Deployment testing and protocols
- Configure solution monitoring metrics
- Software development lifecycle (SDLC) testing



Ongoing monitoring/recommendations

### **Evaluation Considerations for Generative Al Solutions to Ensure Performance, and Identify and Mitigate Risks**





## **Use-case Specific Solution Design** and Evaluation Consideration



Input Evaluation		Solution Engineering		Human Oversight and Monitoring
Performance evaluation and monitoring vertical				
	) Zelo×			
Input query	User guardrailing	RAG/LLM solution	Output alignment	Ongoing monitoring
	Prompt library (prompt techniques)  Zero-shot prompt Few-shot prompt	Solution evaluation	Output evaluation	Output logs
•	Chain of thought (CoT) Chain of density (CoD) Chain of verification (CoVe)	LLM Generated Metrics :     faithfulness, answer relevancy     and context relevancy, logical     correctness	Metrics: Rouge, Bleu, BERT- f1, Perplexity (require ground truth)	<ul> <li>Consolidating logs for ongoing model evaluation</li> <li>Using various queries and corresponding embeddings for performance analysis</li> </ul>

## Select Prompting Techniques to Build Robust LLM Solutions



#### **Zero-shot prompting**

#### **Few-shot prompting**

#### **Chain-of-thought prompting**

#### **Chain-of-density prompting**

#### **Prompt**

Write a summary of the following news article. Article: X

Output: Summary: ...

#### **Prompt**

Write a summary of the following news article. Article: X

Here is a sample:

Article: "A study says eating chocolate weekly lowers heart disease ... eaten in moderation." Summary: The study links chocolate consumption ... best choice, but moderation is key

Output: Summary: ...

#### Prompt

Write a summary of the following news article.

Article: X

Instructions: Outline the thought process step by step

Output:

Step 1: Fact X is essential ...

Step 2: Following reasoning depends on fact ... Summary: ...

#### Prompt

Generate concise, entity-dense summaries of the Article: X

Identify up to three informative entities missing from the previous summary. Write a new, denser summary covering all entities and details from the previous summary plus the missing entities.

Output:

Summary 1: ...
Summary 2: ...

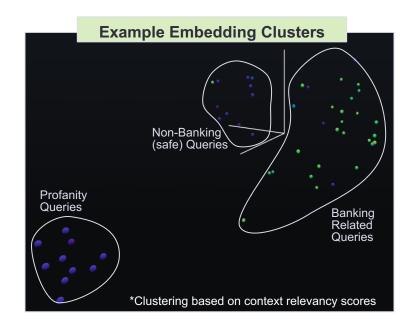
## Ongoing Monitoring of Developed Solution to Ensure Performance



Accessible visualizations and statistics can be employed for ongoing monitoring & human-oversight. These methods can help adding efficiencies to the manual review requirements for tracking the solution post launch.

#### **Approach**

- Create vector embeddings of the knowledge base, query, and response
- Reduce dimensionality of the vector representation using UMAP\* for enhanced visualization
- Perform data clustering using HDBSCAN\*\* to automatically cluster data points
- Introduce additional metrics to enhance visualization capabilities
- Easily surface up and filter on clusters which exhibit unique characteristics for ongoing monitoring and human-oversight



\*UMAP: Uniform Manifold Approximation and Projection

\*\*HDBSCAN: Hierarchical Density-Based Spatial Clustering of Applications with Noise



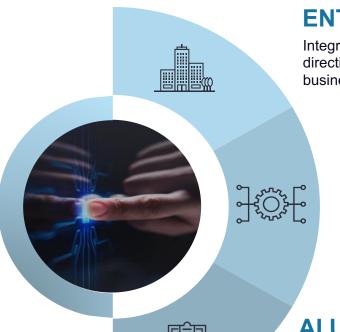
### **Driving Value Through Al**



The evolving regulatory and compliance landscape necessitates robust AI governance at enterprise, process and use case levels to mitigate risks, enforce controls and sustain value creation through AI.

#### AI VALUE CREATION

Responsible AI governance across levels drives economic, social and organizational benefits by enabling ethical compliance, fostering innovation, and supporting sustainable, long-term value creation in line with societal expectations.



#### **ENTERPRISE LEVEL**

Integrate an AI governance framework at the enterprise level to set strategic direction and policies for AI utilization, facilitating ethical practices and alignment with business strategy.

#### PROCESS LEVEL

At the process level, it is essential to identify risks and implement controls to maintain the integrity and governance of Al operations.

#### AI USE CASE LEVEL

When developing AI use cases, it's crucial to design safe AI solutions with built-in monitoring protocols and to incorporate independent validation checks to maintain solution integrity and enable responsible use of AI.