Trusted Al and Al governance

Discussion paper

Building a better working world



AI defined

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Al defined

What is AI?

The application of computational tools to build models from examples, data and experience, rather than following preprogrammed rules.

Programs that enable computers to understand text and spoken words in much the same way human beings can

Programs that allow machines to learn from data and make decisions/ predictions on their own

Large language models

- ► Al assistants (e.g., Apple Siri)
- Search engine (e.g., Google)
- ► Google translate

Computer vision

- ► Self-driving cars
- ► Face detection (phone cameras)



process, analyze and interpret visual data (e.g., digital images or video)

Machine learning (ML)

ML techniques:

- Supervised learning (e.g., credit adjudication)
- Unsupervised learning (e.g., customer segmentation)
- Reinforcement learning (e.g., hedging)

Deep learning

- Product recommendation
- Suggested reads or podcasts

Programs that attempt to simulate the behaviour of the human brain by learning from large amounts of data.

What is NOT AI?

Computer software for which every action and outcome is defined or explicitly programed by humans



1. All is a continuously evolving field and the list is meant to be illustrative.

2. Examples are meant to be illustrative and not exhaustive.

Promise and Risks of Al



Opportunities*

- Develop new product (e.g., enhancing weight management products)
- Transform business (e.g., targeted marketing, streamline supply chain)
- Improve customer experience & operational efficiency (e.g., intelligent recommendations, AI chatbot)

In the absence of proper controls, adoption of AI may expose the organization to potential failures and mishaps

Early experience with AI emphasizes the need for sound governance and risk management to safeguard against regulatory, reputational and business risks

November 2018: BlackRock shelves neural net-based Al liquidity models due to their inability to explain the results to senior management. March 2016: **Tay, Microsoft's Al chatbot, gets a crash course in racism from Twitter.** Hours after launch, Tay makes racial and controversial tweets. October 9, 2018: Amazon scraps Al recruiting tool that showed bias against women. Late 2016: **Uber car runs six red lights**. Uber's self-driving Al technology relies on a highly complex system of Al. May 24, 2018: Amazon Echo sent the conversation of a family to a random person in their contacts.

AI Regulatory Landscape

Rising global guidelines/regulations on trustworthy AI signal urgency



Al ecosystem, risks and response



JJ.

Unbiased Is the AI system free from prejudiced assumptions and intended to drive positive social impact?



Resilient Is the data feeding into the AI system secure from unauthorized access that may lead to incorrect outcome? The Principles of Trustworthy AI



Explainable Can a human understand, challenge, and validate the inner workings and results produced by the AI system?



Transparent Do end users have knowledge and control on what data is being captured and how it is used?



Performance Are the results from the AI system meeting stakeholder expectations consistently?

How do we manage these risks?





The objective is to create **awareness** and **alignment** on key components of the AI governance framework across analytics, technology, risk, ethics and compliance, and privacy teams. This will shape how an organization builds out the 3 pillars of **processes**, **people** and **technology** to enable its AI (data and analytics) transformation journey with a focus on **trusted AI**. AI governance encompasses:



► Model validation guideline (for prioritized model types)



Al model lifecycle

Initiation	Model development ීි සි	Pre-implementation review	Model app	roval	Model implementation
 Initiation procedure Identify the business problem (initiative) Identify the need for AI model 	 Model development procedure Model design Model selection and build-out Model testing and assessment Ethics, compliance and privacy assessment 	 Pre-implementation review procedure 1. Determine model risk tiering 2. Determine ongoing monitoring/review requirements 3. Model evaluation and verification 4. Review ethics, compliance and privacy assessments 	 Model approve to existin Approve Approve Approve remediati issues an 	val procedure new model/changes g model model risk tiering timelines for ion action for model d exceptions	 Model implementation procedure 1. Implement the model as validated 2. Set up adequate post- implementation change management and control
Model change & deco	ommissioning දීදී Ond	going monitoring & review		Post-implementation	on review
 Model change procedure 1. Change assessment and actions 2. Change approval 3. Change implementation & monitoring Model decommissioning procedure 1. Model retirement assessment 2. Retirement approval 		 ngoing monitoring procedure Check if the performance metrics are acceptance range ngoing review procedure Review results from ongoing monitori Perform benchmark analysis, sensitiv etc. to assess appropriateness of the Determine if model change or decommended 	in the ing ity analysis, model missioning	 Perform post-imp the model is imple Ensure that appro as planned 	lementation testing to ensure emented as intended opriate controls are implemented

Illustrative roles and responsibilities matrix (RACI) for AI governance

R = Responsible (executing the task) **A** = Accountable (owner of the task) **C** = Consulted (key stakeholder who should be included in decision or work activity) **I** = Informed (needs to know of decision or action)

Tasks associated with AI risk management		Roles associated with AI Risk Management											
			1	Lst line of defe	ence (risk own	ers)	2nd line of defence (oversight and governance)						
Al lifecycle phase	Activity group	Activities	Al model develop- pers	IT ops (GTS)	Al model managers	Al model business owners	Al model validators	Al model validation owners	Ethics and compliance	Legal (privacy) counsel	Cyber- security	Al model risk committee	Enterprise risk management (ERM)
Initiation	Purpose	Identify the business problem				AR							
	identification	Identify the need for AI model	С	С	С	AR							
	Model design	Design the AI model (scope and objective)	R		А	С							
	Inventory management	Register AI model	R		А	I							
		Determine an appropriate methodology for use	R		А	I							
	Model selection	Determine appropriate data to be used for model development	R	I	А	I			I	I			
	and build-out	Select an appropriate model development platform	R	С	А	I					С		
Model		Program the necessary code for model implementation	R		А								
Development	Model testing	Perform appropriate testing to see if the underlying assumptions are plausible and that the model is programmed correctly	R		А	I							
		Assess the appropriateness of a model's use for the intended purpose	R		А	I							
	Ethics,	Complete Ethics, compliance and privacy assessment	R	I	А	I			С	С			
	compliance,	Review the questionnaire submission	С	I	С	1			С	AR		<u> </u>	<u> </u>
	assessment	Remediate ethics, compliance and privacy concerns	R	I	А	1			С	С			
	Model risk	Determine model risk tier based on risk factors	С	I	С	С	R	А	С			- I	l I
	tiering	Determine requirements for initial validation and ongoing model monitoring/review based on risk tier	С		С	С	R	А				I.	
Pre- implementatio	Model evaluation and	Execute model validation, i.e., assess if all three model components – inputs, computation processes and outputs – are working as intended	С	С	I	I	R	А			С		
n Review	vermcation	Provide recommendations regarding model approval	С		С	С	R	А				l I	I
	Review ethics	Assess if the ethics and privacy assessments were completed accurately by the model developers	С	I	С	С	R	А	С	С		I.	
and	assessments	Verify if the remediation actions were appropriate to address the ethics and privacy concerns	С	I	С	С	R	А	С	С		I.	

Illustrative roles and responsibilities matrix (RACI) for AI governance (cont.)

R = Responsible (executing the task) A = Accountable (owner of the task) C = Consulted (key stakeholder who should be included in decision or work activity) I = Informed (needs to know of decision or action)

Tasks associated with AI risk management		Roles associated with AI risk management											
			1 st line of defence (risk owners)				2 nd defence (oversight and governance)						
Al lifecycle phase	Activity group	Activities	Al model develop- pers	IT Ops (GTS)	Al model managers	Al model business owners	Al model validators	Al model validation owners	Ethics and compliance	Legal (privacy) counsel	Cybe- rsecurity	Al model risk committee	Enterprise risk management (ERM)
Model approval	Model approval	Approve AI model and the assigned model risk tiering	С		с	С	С	С	С	С	С	AR	I.
Model	Model	Select an appropriate model deployment platform	R	С	А	С					С		
implementation	implementation	Execute implementation and performance test	R	С	А	С					С	l I	
		Document model	R		А	С					С		
Post-	Post- implementation	Perform post-implementation testing to ensure the model was implemented as intended	С	С	С	С	R	А				l.	
review	testing and review	Check if appropriate controls are implemented	С	С	С	С	R	А			С	l.	
	Ongoing	Execute performance tests to check if the model performance is in the acceptable range	R	С	А								
	performance	Report performance test results	R		А	I	I	I				l I	
Ongoing	monitoring	Remediate performance test issues	R	I	А	С	I	1			I	l I	
monitoring & review		Execute model review to assess if the model is still fit for the intended purpose	R	С	А	С			С	С	С		
	Ongoing model	Report model review results	R		А	С	I	I				l I	
	Teview	Review the model review results and recommend if any model change/decommission needed	С	I	С	С	R	А			I	l.	I.
Model change &	Model change & decommission-	Approve any recommended model change/decommissions	С		С	С	С	С				AR	I.
decommissioning ning	ning	Execute model change/decommissioning	R	I	А	С	I	I			I	l I	l I
Oversight &	Maintenance &	Escalations and exceptions management and arbitration	С	С	С	С	С	С	С	С	С	AR	I.
governance reporting	reporting	Model risk management (MRM) adoption oversight and enhancements	С	С	С	С	С	С	С	С	С	AR	I.

Chief Risk Officer (Chair)

Leader, Enterprise Risk Management (ERM) Leaders, Ethics and Compliance

Leaders, Privacy Leaders, Data & Analytics, Al Leaders, Cybersecurity

Model Risk

Committee

Al model risk tiering - model classification and assessment

AI model risk classification and assessment requirements						
Model risk tiering	Initial validation	Ongoing monitoring	Ongoing review frequency			
High risk	Yes (high priority)	Monthly	Annual			
Medium risk	Yes	Quarterly	Every two years			
Low risk	Reduced scope validation	Annually	Every three years			

Notes

- Initial validation refers to pre- and post-implementation review. Its objective is to review the conceptual soundness of the model methodology, model testing and verification, available documentation, etc.
- Ongoing monitoring is to confirm that the model continues to perform as expected over time after implementation, and to help identify the need for changes and enhancement should performance deterioration emerge over time (e.g., changes in market conditions or business activities, etc.)
- Ongoing review refers to periodic reviews of the model. Its objectives include reassessment of the continued appropriateness, performance of the model and any benchmarking analysis, back-testing metrics, sensitivity analysis, etc.; updating the model documentation accordingly.
- Reduced scope validation refers to a validation with reduced scope, with focus on documentation, conceptual soundness and performance, excluding thorough sensitivity analyses and benchmarking
- While the technical teams in the 1st and 2nd lines of defence are primarily responsible for the activities in the validation/monitoring/review stage, other teams, including ethics and compliance, privacy, IT/cybersecurity, etc., also play crucial roles in supporting these activities. More details on roles and responsibilities throughout the model lifecycle will be discussed in upcoming sessions.

Risk tiering framework - illustrative template

 Illustrative 3-tier risk matrix (low/medium/ high) to assess the performance risk tier of Al use cases based on complexity and business impact.

 The depth of initial validation and frequency of ongoing monitoring and ongoing review are adapted based on the model risk tier.

Derformen en rick tier		Business impact				
Performan	Performance risk tier		Medium	High		
ty	Low	Tier 3	Tier 3	Tier 2		
mplexi	Medium	Tier 3	Tier 2	Tier 1		
ŭ	High	Tier 2	Tier 1	Tier 1		

Model risk tiering	Initial validation	Ongoing monitoring	Ongoing review frequency
Tier 1	Yes (high priority)	Monthly	Annual
Tier 2	Yes	Quarterly	Every two years
Tier 3	Reduced scope validation	Annually	Every three years

Complexity assessment - illustrative template

Domain	Que	estions	Answers	Score	Max	Total score	Complexity
1. How many data sources does the Alives case			A. One	0		0 to 4	Low
1. Ho pulls	1. How many data sour	rces does the AI use case	B. Two to three	1 2	2	5 to 9	Medium
Data source	,		C. Four or more	2	2		High
	2 Doos the Aluse cos	a use external data?	A. No	0	1		
	2. DOES THE AT USE Case	e use external uata?	B. Yes	1	1		
		A. Yes		2			
	3. Does the Al use	B. No	A. 1 to 20	0	2		
Input Data	unstructured data?	How many features does it use?	B. More than 20	1	L		
	4. Was the AI use case	trained with imbalanced	A. No	0	0		
	dataset?		B. Yes	1	L		
			A. Highly interpretable algorithms (e.g., rule-based systems, linear/logistic regressions, decision trees etc.)	g., rule-based systems, linear/logistic regressions,			
Ę	5. What type of AI algorithms is used by the AI use case?		B. Moderately interpretable algorithms (e.g., support vector machines, random forests, etc.)	2	4		
Methodology			C. Less interpretable algorithms (e.g., deep neural networks, support vector regression, gradient boosting etc.)	4			
Methodology	6. Are parameters dyn	amically adjusted (e.g.,	A. No	0	2		
	active learning, reinfor	cement learning)?	B. Yes	2	2		
	7. Does the Al use case	e employ transfer	A. No	0	1		
	learning?		B. Yes	1	1		
			A. None	0			
Output	8. To what extent is ad required on the output	lditional processing of the AI use case?	B. Somewhat	1	1 2		
			C. Significant	2			
	Total score	2			15		
	Complexity	/					

Business impact assessment - illustrative template

Domain	Questions	Answers	Score	Max	Total score	Business impact
		A. No	0		0 to 3	Low
Revenue	1. Does the AI use case aid in revenue-generating activities of the bank?	B. Indirectly	1	2	4 to 6	Medium
		C. Directly	2		7 to 12	High
		A. None	0			
Operational efficiency	2. To what extent will the bank's operational efficiency be impacted if the AI	B. Somewhat	1	2		
operational enterency	use case does not perform as expected?	C. To a large extent	2	-		
		A. No	0			
Customer experience	3. Is the AI use case impacting external clients or customers, as opposed to internal users?	B. Somewhat	1	2		
		C. Major Impact	2			
		A. No	0			
Compliance	4. Is the AI use case used for control processes, such as internal/regulatory compliance?	B. Yes, internal compliance	1	2		
Compliance		C. Yes, regulatory compliance	4	-		
		A. None	0			
Reputational risk	5. What would be the impact to the bank's reputation if the AI use case does	B. Moderate	1	2		
	not perform as expected:	C. Major	2			
	Total score			12		
	Business impact					

Al model inventory

Al model inventory is intended to capture and track organization-wide Al assets on an ongoing basis.



Table of contents for the AI policy

Scope					
Entity and geographical scope Mention the entity and geographies that are within the scope of this policy.	Definition This will be Examples	finition of AI model is will be based on the AI model definition discussed in the workgroups amples will be provided for guidance			
Model risk management					
Model risk principles and requirements 1. Model risk involves adverse financial and reputational consequences 2. Control on model risk requires an oversight authority for model approval			Model lifecycle Risk is managed through model lifecycle – model initiation, development, independent review, approval, implementation, ongoing monitoring, model change and decommission		
Two lines of defence 1. First line of defence owns and manages the risk, i model development, implementation and monitoring 2. Second line of defence identifies emerging risks, model review and assessment	ncluding J including	Model risk management fra 1. Model initiation and mod 2. Model review and impler 3. Monitoring and governa	mework lel development nentation nce	Summary of roles and responsibilities Management of model risk is a joint responsibility of model owner, developer, user and independent risk management function.	
Policy violationMaterialityPolicy violations can occur through either model exceptions or due to non-compliance of existing models.Materiality in a model in financial, strategic and			in a model is the potential impact of model, including rategic and operational Impact of model, including		
Model Risk Reporting and Execution					
Model risk appetite Risk appetite is amount of risk the organization is wi assume to meet its desired objectives	illing to	Escalation Al Risk Committee will esca Committee as aligned with t	late matters to Management Risk the risk escalation guideline	Committee table List the committees involved in the process and their roles outlined	
Use of external resource Model development and model validators can use external help to complete their tasks as long as they abide by the organization's policies, and they supervise and approve the work			Training and policy attestation Annual training with respect to model risk management must be completed by individuals carrying out model-related roles and responsibilities		

Table of contents for the AI procedure

Model initiation and development		
Model Initiation1. Identify the business problem (initiative)2. Identify the need for AI models	 Model development Model design Model selection and build out Ethics and privacy assessment Model testing and assessment 	Model documentation Model development document captures model design, processing, evidence of industry practice, academic theories, rationale for data choice, model limitation and aligned with risk
Model review and implementation		
 Pre implementation review Determine model risk tiering Determine ongoing monitoring/review requirements Model evaluation and verification Review ethics and privacy assessments 	Model risk tiering Model owners are required to quantify model risk as per risk tiering guidelines, and model reviewers evaluate it	 Model approval Approve new model/changes to existing model Approve model risk tiering Approve timelines for remediation action for model issues and exceptions
 Model implementation 1. Implement the model as validated 2. Set up adequate post-implementation change management and control 	 Post implementation review Perform post-implementation testing to ensure the model is implemented as intended Ensure appropriate controls are implemented as planned 	
Monitoring and governance		
 Ongoing monitoring 1. Check if the performance metrics are in the acceptance range 	 Ongoing review Review results from ongoing monitoring Perform benchmark analysis, sensitivity analysis etc. To assess appropriateness of the model Determine if model change or decommissioning needed 	 Model change 1. Change assessment and actions 2. Change approval 3. Change implementation & monitoring
Model decommissioning1. Model retirement assessment2. Retirement approval	Vendor model management A vendor model is managed similarly as an internal model. The owner is responsible for documentation, etc., and a validator for review	Model inventory management Procedures for maintaining the centralized inventory for all models. The second line of defence will do periodic inventory reconciliation
Appendix		
RACI Model inventory		

EY | Building a better working world

EY exists to build a better working world, helping to create long-term value for clients, people and society and build trust in the capital markets.

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