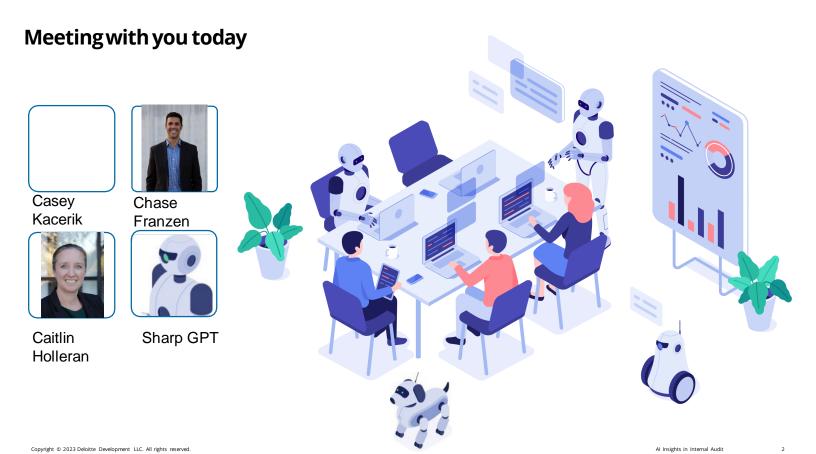


Auditing Artificial Intelligence

CAITLIN HOLLERAN, CHIEF COMPLIANCE OFFICER CHASE FRANZEN, CHIEF INFORMATION SECURITY OFFICER SHARP HEALTHCARE

CASEY KACERIK, SENIOR MANAGER DELOITTE



Agenda & Objectives









Health System



Al Risk Domains



Al Governance & Best Practices – IA Specific View



Discussion, Q&A

By the end of this session, you should:

- ✓ Develop a comprehensive understanding of the Al landscape, including its definition, significance, and current state.
- ✓ Demystify the development and applications of AI, enabling participants to grasp the process and identify common uses.
- Recognize the specific applications of AI in the healthcare industry and understand their benefits and challenges.
- ✓ Gain awareness of the various risk domains associated with Al and understand the importance of Al Governance and best practices.
- ✓ Equip internal auditors with the skills to effectively assess Al-related risks within the organization by learning appropriate questioning techniques.

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Al Insights in Internal Audit

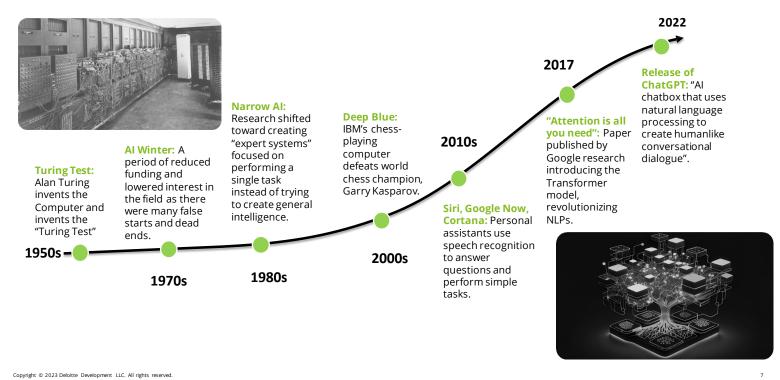
Al Landscape



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Brief History of Al

Timeline demonstrating the advances in artificial intelligence and how it continues to revolutionize various industries.



Al is inherently about augmenting humans with machines to reach greater heights

Al mimics the way humans perceive information, devise insights based on experience, and make decisions accordingly



OXFORD DICTIONARY DEFINITION:

ARTIFICIAL INTELLIGENCE (AI) is the theory and development of computer systems able to perform tasks normally requiring human intelligence

Al encompasses many technologies that work together to build innovative solutions that transform society and business...

Machine Learning

Virtual Assistants

Intelligent Automation Learn

Learning

Robotics Process Automation Natural Language Processing

> Deep Learning

Neural Networks

Predictive Analytics

Speech Recognition

Unsupervised Computer Vision

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Unlocking the Power of Language: Understanding Generative AI

WHAT is Generative AI | artificial intelligence that creates original content across various modalities (e.g., text, images, audio, code, voice, video) that would have previously taken human skill and expertise to create

HOW does it work | Generative Alis powered by foundation models such as OpenAl's GPT-4, NVIDIA's Megatron, and Google's PaLM, which are trained on vast amounts of data and computation to perform a broad range of downstream tasks

WHY now | innovations in machine learning and the cloud tech stack, coupled with the viral popularity of publicly released applications have propelled Generative Al into the zeitgeist

WHO is involved | Big Tech is building—and enabling access to—foundation models; start-ups are developing user applications on these underlying models; and companies are beginning to adopt

POTENTIAL BUSINESS IMPACT | the marginal cost of

producing initial versions of knowledge-intensive content—such as IT code, marketing copy, and creative design—can decrease significantly

EXAMPLE MODALITIES

Text Generation

Prompt: Explain my colleagues the business impact of generative AI in 50 words



Image Generation

Prompt: A bowl of soup that is a portal to another dimension as digital art



Code Generation

Prompt: In python, code a program that predicts the likelihood of customer conversion



Video Generation

Prompt: A teddy bear painting a portrait



Audio Generation

Prompt: Play 'we have to reduce the number of plastic bags' in a sleepy tone

**	Brighton	Sprake Proppt	100	
	Augus	1		
	Street	1		
sent stock the national plants hap	Texas	1		
	tenant			
	Tomate	1		

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Generative AI comes with risks and limitations

There are several limitations to consider when using Generative Al

Bias in; bias out. If the training data is biased (e.g., over/under-representation of a population cohort, sexism, racism), then outputs generated could exhibit biases as well. Bias reductions in the training data and/or human supervision during model training is needed

US \$1000s

Foundation Models generally offer a pay-as-you-go billing mechanism, and the cost per use of sophisticated models is materially significant. Fine tuning the biggest model and running large documents through several times could easily run up a bill of tens of

Is the AI being used in a manner consistent with the purpose of the overall exercise? Is a human being brought into the loop to decide whether the Al's suggestion needs adjustment before actual use? Submitting an Al-generated essay for a high-school assignment may not be ethical

Models might output facts that are factually false. Sources and citations are unavailable for most models. Users should be conscious that outputs could be inaccurate and should perform due diligence to validate generated content.

SaaS-AI companies may save some or all of prompt payloads for future training. Therefore, confidential data will be used to train future versions of the base model how will this affect your organization's competitiveness in the market?

Bias



It is critical to proactively minimize risk from malicious behavior on the network to maintain operations and customer trust. For example, a customer service bot revealing confidential information to a hacker either by prompt or unintentionally

Malicious behavior

Cost

Foundation Models are comprised of billions of parameters (model size) and trained on petabytes of data. In theory, the larger the model, the better the output. Foundation Models take time to produce outputs, which may limit real-time use cases

Model Performance





SaaS-Al companies require to submit text as a payload to users' API call. The data could be crossing borders. Is this in accordance with data privacy laws and with your company's policies? Many cloud service providers offer market-leading controls to manage data privacy of Foundation Models

Privacy



Hallucination



Models are good at understanding text but struggle when the data are in irregular formats, or when the position of the text on the . page (e.g., infographic, PPT presentation slide) is relevant to the context and understanding. Other emphasis generators such as bolded text, font color, etc., don't play a role yet

Text **Formatting**



Most models have a 2k token size limit. Some larger ones can process 4k tokens in a single call. 2k tokens are approximately 2-2.5 pages. This limit makes it difficult to process larger documents

IP Protection

Token Size Limits



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Generative AI Use Cases by function

SALES & MARKETING

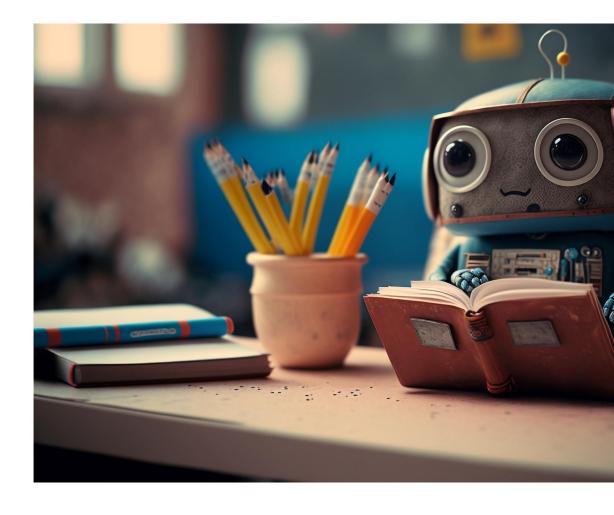
(Mary					
1. Video editing and generation	13. Personal onboarding assistant	25. Demand planning (Consumer Sentiment Analysis)			
2. Metaverse 3D experience	14. Compensation analysis	26. Inventory analysis			
3. Product descriptions and reviews	15. Workforce skill analysis	27. Global trade-logistics analysis			
4. Personalized consumer advertisements	16. 3D avatar creation	28. Contract Adherence & Anomaly Detection			
5. Recommender systems for e-commerce	17. Metaverse 3D workforce experience	29. Scenario simulation			
6. Chatbot / virtual assistant dialogue generation	18. Metaverse 3D workforce upskilling	30. Language translation for global trade			
GOVERNANCE & OPERATIONS 7. Intranet search (knowledge management)	INFORMATION TECHNOLOGY 19. Code generation across languages/frameworks/CSPs	FINANCE & ACCOUNTING 31. Fraud, waste, and abuse prevention			
8. Process analysis	20. Development lifecycle documentation	32. Regulation and oversight analysis			
9. Training for new team members	21. Test automation and test scenario creation	33. Financial report analysis			
10. Document inventory analysis	22. Training on new technologies	34. Proactive value opportunity identification			
11. News and media summaries	23. Peer review for optimized code writing	35. Budget and ROI analysis			
12. Sentiment Analysis for Workforce	24. Legacy code summarization & translation	35. Divestment recommendations			

HUMAN RESOURCES

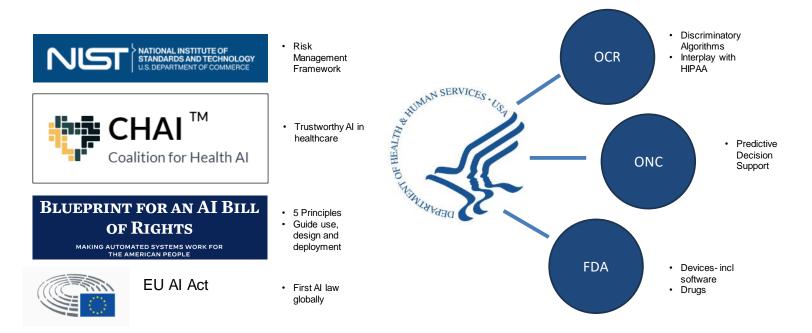
SUPPLY CHAIN & PROCUREMENT

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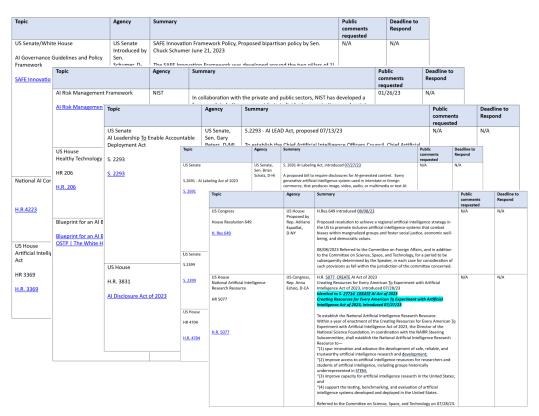
Emerging Al Regulatory Framework



Emerging Regulatory Environment



Federal Regulations



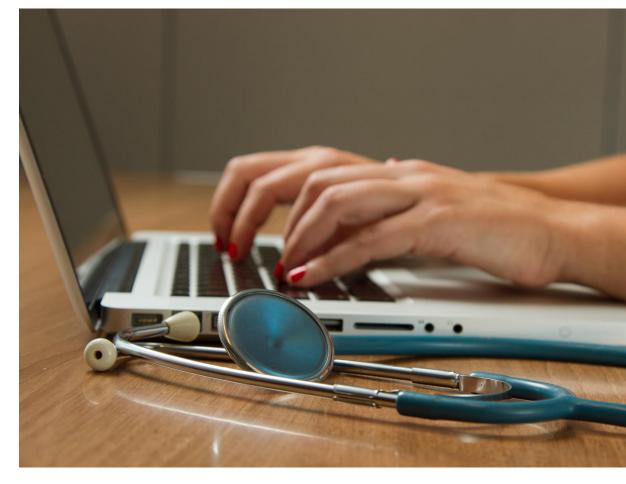
- Frequent additions with federal regulations
- Many agencies are focused on creating frameworks to govern AI, evaluating and reducing risk
- Topics such as trust, security, privacy trend throughout proposed regs
- Some outliers of proposed regs are concerning, H.R.
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California Regulations

Topic	Agency	Summary		Public comments requested	Deadline to Respond			
California Legislature— Intent Bill. Introduced 09/13/23		Presents framework for California to ensure safe development of AI models within state borders.			N/A			
California Legislature — 2023–2024	Торіс		Agency	Summary			Public comments requested	Deadline to Respond
Regular Session – Intent Bill Safety in Artificial Intelligence Act SB 294 California Executive Order signed	California Legislature— 2023–2024 Regular Session. Introduced 02/06/23 Bill Text - SB-313 Department of Technology. Office of Artificial Intelligence: state agency public Interface: use of Al. Ica gozy			Department of Technology that would ov	oses the creation of an Office of Artificial Intelligence within the of Technology that would oversee the use of artificial by state agencies and ensure compliance with state and federal ulations.		SB-313 file notice suspended. 05/18/2023	N/A
09/06/23	Regular Ses 02/16/2023 Bill Text - SE	egislature— 2023–2024 ssion, introduced 3 3-721 California v Al Working Group.	California Civil Rights Council (CRC)	SB-721 California Interagency AI Working • SB 721 proposes the creation of a Ca Group to study the implications of th Legislature with a comprehensive rep two years thereafter until (2030) rega	Majority vote required.	N/A		
CA Executive Order N-12-23	Regular Ses Civil Rights Modification Regulations Decision Sy Employer A Bill Text: CA	egislature – 2023–2024 sion. Introduced 01/30/23 Council Proposed ns to Employment s Regarding Automated- stems II Use Bill AB 331 AB331 2023-2024 sion Amended	California Civil Rights Council (CRC)	AB 31 Employer Use of Automated Decision Tools Bill would impose obligations on employers to evaluate the impact of an ADT, provide notice regarding its use, and provide for formation of a governance program. It would prohibit employers from using an ADT in a way that contributes to algorithmic discrimination. Perform an impact assessment on or before Ian. 1, 2025, and annually threafter, for any ADT that includes: a summary of the type of data collected from individuals and processed by the ADT. an analysis of the potential adverse impacts on the basis of sex, race, color, ethnicity, religion, age, national origin, limited English proficiency, disability, veteran status, or genetic information. a description of the safeguards that are or will be implemented by the deployer to address any reasonably foreseeable risks of algorithmic discrimination arising from the use of the ADT. a description of how the ADT has or will be evaluated for validity or relevance.			Majority vote required. Joint Rule 62(a), file notice suspended. 05/18/2023	N/A
				Bill broadly requires additional technical addition to the CPRA's regulations regard Businesses will need to consider how to i	ing automated decis	ion making.		

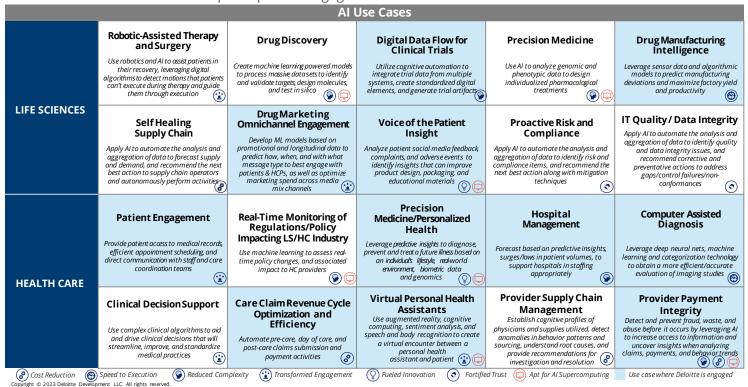
- California home to 36/51 major AI vendors
- State laws can conflict with federal (historical, and with Al framework)
- Some state laws well intentioned but concerning if poorly implemented (SB 294; establishes liability and penalties to damages caused by "foreseeable risk"



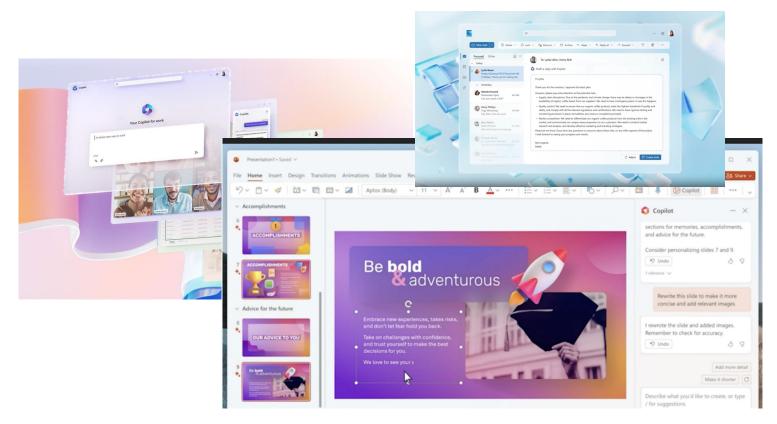


Numerous AI use cases across the industry

Current use cases include AI to improve patient engagement and health outcomes

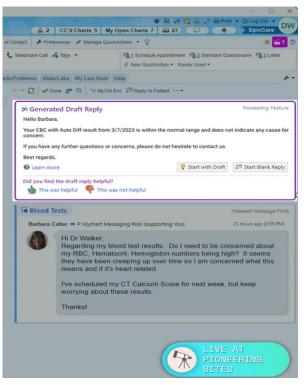


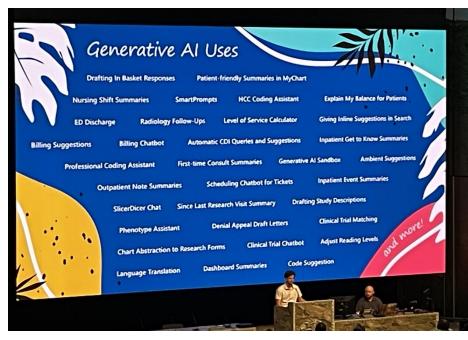
It Is Not Just Clinical - Microsoft 365



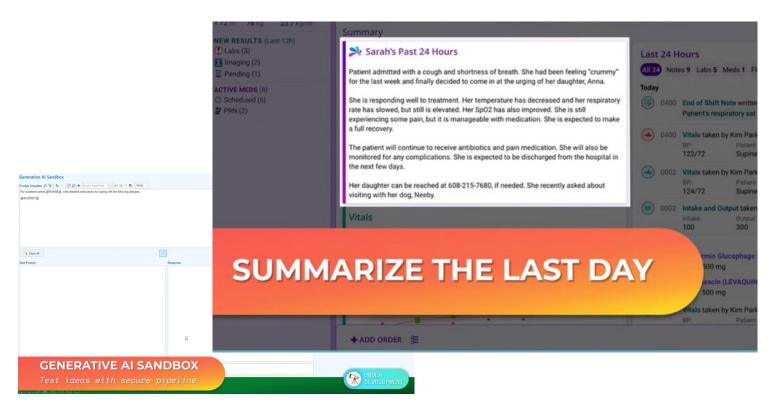
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EMR Vendors - Epic Generative AI





EMR Vendors - Epic Generative AI

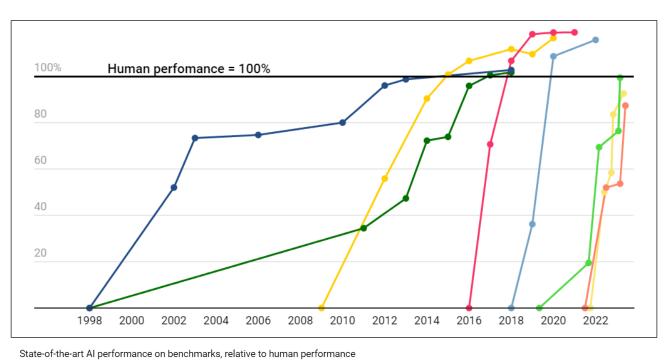




Health System Al

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WHY AI, Why Now?



■ Handwriting recognition
 ■ Speech recognition
 ● Image recognition
 ● Reading comprehension
 ● Language understanding
 ● Common sense completion
 ● Grade school math
 ● Code generation

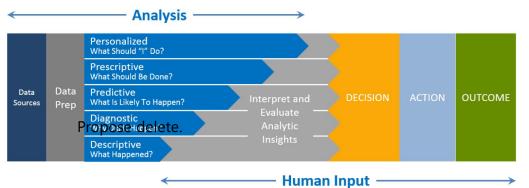


But with Great Technology.... Comes a Need for Great Oversight

A Statement of Purpose – Al Oversight Committee

This Committee will provide oversight of where and how artificial intelligence and data science assets are used within Sharp HealthCare. The Committee is responsible for leveraging individual subject matter expertise to try to anticipate and mitigate unintended consequences of AI. The committee will provide guidance and expertise to develop and implement standards, policies and process around this rapidly evolving discipline of artificial intelligence.

Human AND Machine Together



Combining computer and human analysis to make better decisions and achieve better outcomes.

A Human

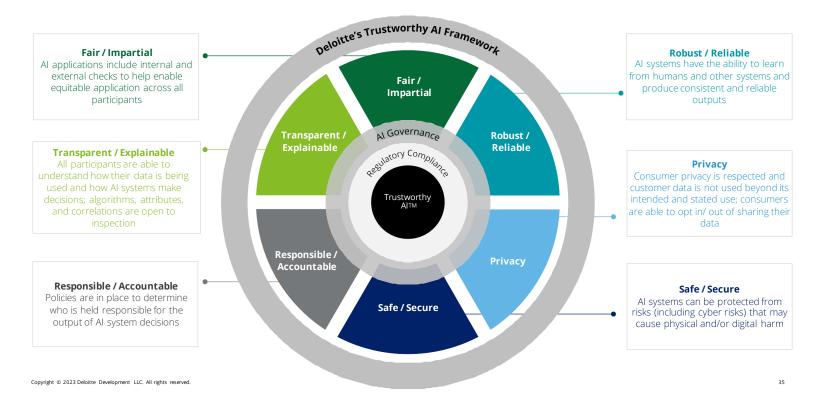


In the Loop ...
Always

Al Risk Domains

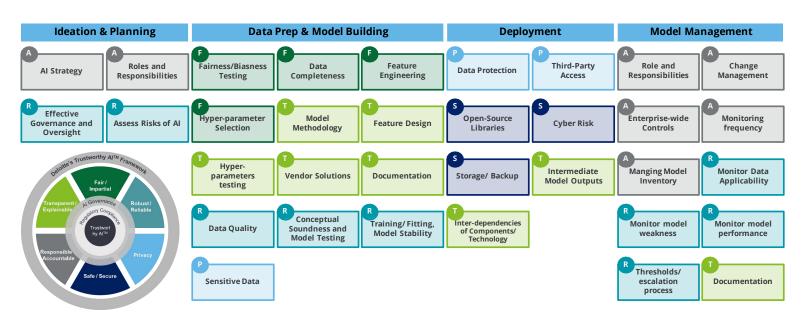


Deloitte's Trustworthy AI framework



Deloitte's Trustworthy AI™ Framework

Below are example areas that our team has identified for Digital Program Assurance and Trustworthy Al™ assessments.





Al Governance and Best Practices

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Internal Audit Considerations throughout the AI lifecycle



Design Process

- Has a business case evaluation been performed (i.e., a broader group weighed in on the need, and pros and cons, for the AI application)?
- Have potential sources of risks (e.g., bias) been identified and addressed/mitigated?
- Is there a process or procedure in place that monitors current and emerging regulations and their applicability to AI implementations?

Data

- Is there a process for how data streams are selected and evaluated?
- Is bias and potential bias with Al implementations, reviewed, evaluated, and documented?
- Are data sets validated to ensure they are representative of the underlying populations and operational environment

Development & Implementation

- Is the complexity of the model commensurate with the use case and benefits of the model?
- Are assumptions and limitations for AI models evaluated?
- Are models testing to ensure they are consistent with goals and objectives set forth in the business case as well as principles to foster public trust

Validation & Review

- Are the test cases comprehensive, with appropriate pass/fail criteria, and is there
 appropriate statistical or other quantitative/qualitative testing of the modeled
 results performed by relevant stakeholders with appropriate expertise?
- Is the applicability and relevance of model policy, procedures and standards reviewed on a regular basis to ensure they are up to date and reflect evolving regulations and corporate requirements?

How Can Internal Audit Support AI?

Ways that Internal Audit can engage with operations in this space

Al strategy, governance and operating model



Provide a strategic cross-functional governance, roadmap and operating model for an effective Al risk management program.

- Design and implement an AI strategy and framework to be supported by processes and controls over governance, deployment, and monitoring of AI based on our Trustworthy AI framework
- Design and implement a strategy for compliance with AI regulations

Al data governance



Establish a data governance and risk management framework to safeguard for the security, privacy, integrity and ethics of data used for Al throughout its lifecycle.

- Provide recommendations for remediation of data and controls over Al based on Trustworthy Al framework
- · Assist stakeholders in the development, design and implementation of controls to address Al-specific risks

Al risk management operations



Build trust and resiliency in Al systems against anomalous activities that could compromise the data, models or outcomes. Develop robust and resilient infrastructure, operations and model development processes.

- Provide independent testing on the design and operating effectiveness of AI controls, findings, and recommendations for deficiencies in the AI environment
- Conduct independent testing of AI models and related datasets for potential adverse outcomes

Establishing an AI risk program and operational constructs in alignment with your business strategy and operations

Implementing a data-centric risk approach and framework for managing risks throughout the Al lifecycle

Analyzing and improving AI technology and related processes, to promote organizational trust for your AI solutions

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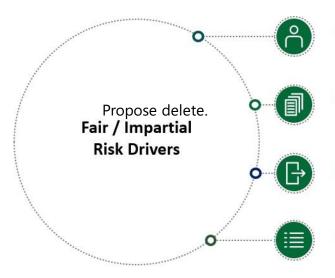
Questions?



APPENDIX

Fair/Impartial

The risk of producing discriminatory bias, or the perception thereof, towards certain subgroups of the populations and thus against the organization's ethical value.



Demographic Data

Unfair or biased outcomes when utilizing certain demographic data that may be correlated with a protected class

Population Completeness

Under/over representation of certain parts of the population when AI training occurs on incomplete data populations.

Data Sources

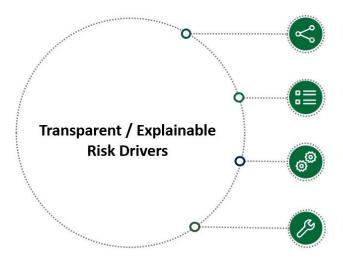
Alternative data sources might be perceived by the public as being unfair or biased, which may raise reputational risk to the organizations.

Feature Engineering

Unfair or biased outcomes for population subgroups when obscure or complex feature engineering yields features with high correlation to protected class.

Transparent/Explainable

The lack of ability to explain a particular behavior of the AI system due to model complexity or feature inexplicability.



Model Methodology

Model complexity and architecture may make model behavior difficult to understand or explain (e.g. multi-layered neural nets).

Feature Design

Complex feature transformation can result in obfuscation of original data attributes yielding difficult to explain outcomes.

Hyperparameters

Difficulty explaining the impact of hyperparameter choices. Hyperparameters are typically derived based on a trial & error or optimization process.

Infrastructure

Potentially difficulty understanding behavior of complex or heavily inter-connected models.

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Responsible / Accountable

Lack of responsibility or accountability at each stage of AI use can lead to increased risk, including regulatory and operational risks.



Al Strategy

- Inability to apply Al across an organization consistently and within the ethical bounds of the organizational values.
- Al usage outside of the risk appetite of the organization.
- Al solutions inconsistent with overarching strategy.

Roles and Responsibilities

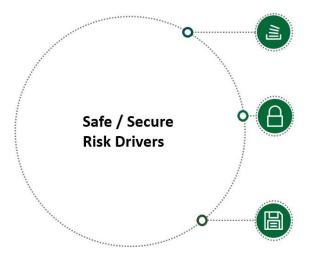
- Poor Al/data/model oversight driven from ambiguous and overlapping roles and responsibilities in the organization.
- Undesired AI outcomes can arise from poorly defined roles and responsibilities and lack of accountability across the 1st, 2nd and 3rd lines of defense
- Undesired outcomes arise from inadequate review.

Change Management

 Uncontrolled changes to models may lead to technology failures or poor model outcomes.

Safe / Secure

The internal and external threats that arise due to lack of consistent and cohesive security of AI systems can lead to multiple risks including loss of business, customers' and regulators' trust.



Open-Source Libraries

- Usage open-source packages or beta versions may pose inherent security risks.
- Open sources packages may transmit data to third party servers (e.g.: Plottly or Dash can transmit data to plot graphs on cloud)

Cyber Risks

- Prototype code on unofficial infrastructure may be vulnerable to cyber attacks resulting in loss of IP.
- Automated systems may expose upstream/downstream AI/IT systems to security threats.
- Fragmented learning systems may be exploited to spread viral/malware infections throughout the AI network.

Storage/Backup

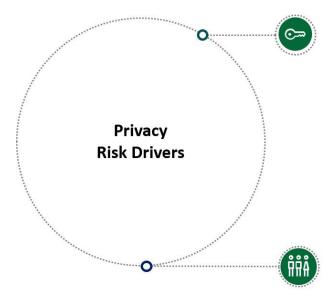
- Large-scale Al systems can take up to weeks of training.
 Any disruption during the training can lead to loss to critical training and compute time.
- Valuable training data collated from years of business experience may accidentally get deleted/lost.
- Lack of formalized storage infrastructure to maintain data and trained models (i.e. desktop storage)

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Privacy

The risk in partial or complete failure to maintain privacy of data used or created within/by the AI systems or non-compliance to internal/external data protection rules.



Data Protection

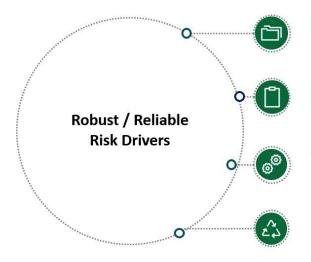
- Non-compliance to local/international data privacy rules (e.g. : GDPR) can lead to:
 - Prohibition of model usage leading to business risks.
 - · Regulatory Penalties
 - Market Cap/ Reputation loss.
- Unauthorized usage of data in Al systems may lead to legal risks (e.g., usage of data beyond permissible/consented time period/applications may be perceived as unauthorized)
- · Unapproved sensitive data used in ML pipelines.
- Lack of data privacy governance in AI strategy may lead to potential sensitive data risks
- Potential privacy risks when working with unstructured data as documents may include various elements of sensitive data.

Third Party Access

- Un-intended usage of data by third party vendors can lead to financial, regulatory and reputational risks.
- Lack of virtual boundaries in cloud infrastructure may leak sensitive data from one AI application into another.

Robust / Reliable

The risk that AI systems are not functioning as designed, resulting in unintended consequences over time or on new data sets.



Data Quality

 Data noise can mislead machine learning algorithms and cause inaccurate generalizations, while live data drift can degrade the accuracy of Al algorithms over time.

Methodology/ Approach

 If an AI model is poorly designed, inadequately justified, or not well understood in terms of its limitations, it can lead to unreliable outcomes.

Training/Fitting

Issues with overfitting or underfitting the data can lead to unreliable Al
outputs, which may also be unstable if the model inputs or parameters
are changed.

Continuous Learning

 Changing data structure can degrade AI models, resulting in poor outcomes, and technological breakdowns can disrupt the continuous learning pipeline.

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