



WEBINAR

Governance of AI & ML in Financial Services

January 17, 2024 | 11:00 AM



Aron Johnson, CFA, FRM, FDP
Senior Examiner, Federal Housing Finance Agency



Kathryn Wilkens, PhD, CAIA
Founder of Pearl Quest LLC,
FDP Curriculum Advisor

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Governance of AI & ML in Financial Services

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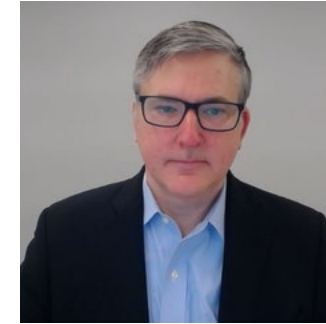


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Introductions



Kathryn Wilkens, Ph.D., CAIA
Founder,
Pearl Quest LLC

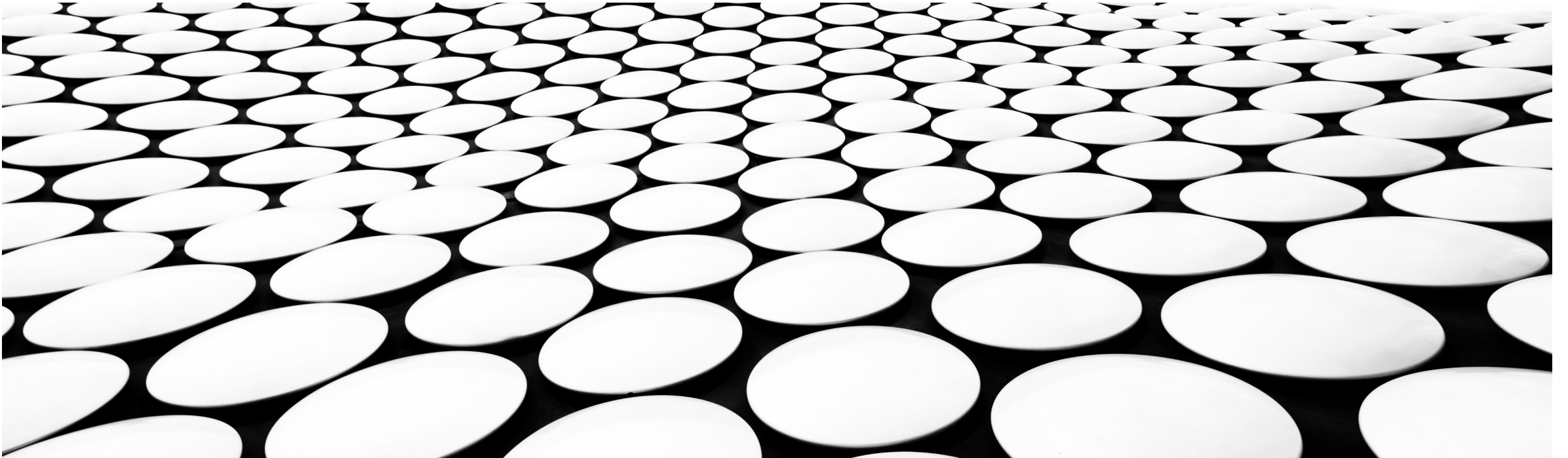


Aaron Johnson, CFA, FRM, FDP
Senior Examiner,
Federal Housing Finance
Agency

Today's Topic:
**Governance of Artificial Intelligence & Machine Learning
In Financial Services**

GOVERNANCE OF ARTIFICIAL INTELLIGENCE IN FINANCIAL SERVICES

AARON JOHNSON  JANUARY 17, 2023





CONTENTS

- Some preliminary definitions
- Risks from AI
- Trustworthy AI
- Some authoritative statements on AI governance
- Practical considerations in AI governance
- Parting thoughts
- Appendix



DISCLAIMER

- The views expressed in this presentation should not be construed to be those of my current employer, the Federal Housing Finance Agency, nor of any prior employer. Nothing in this presentation should be construed to be legal or financial advice.

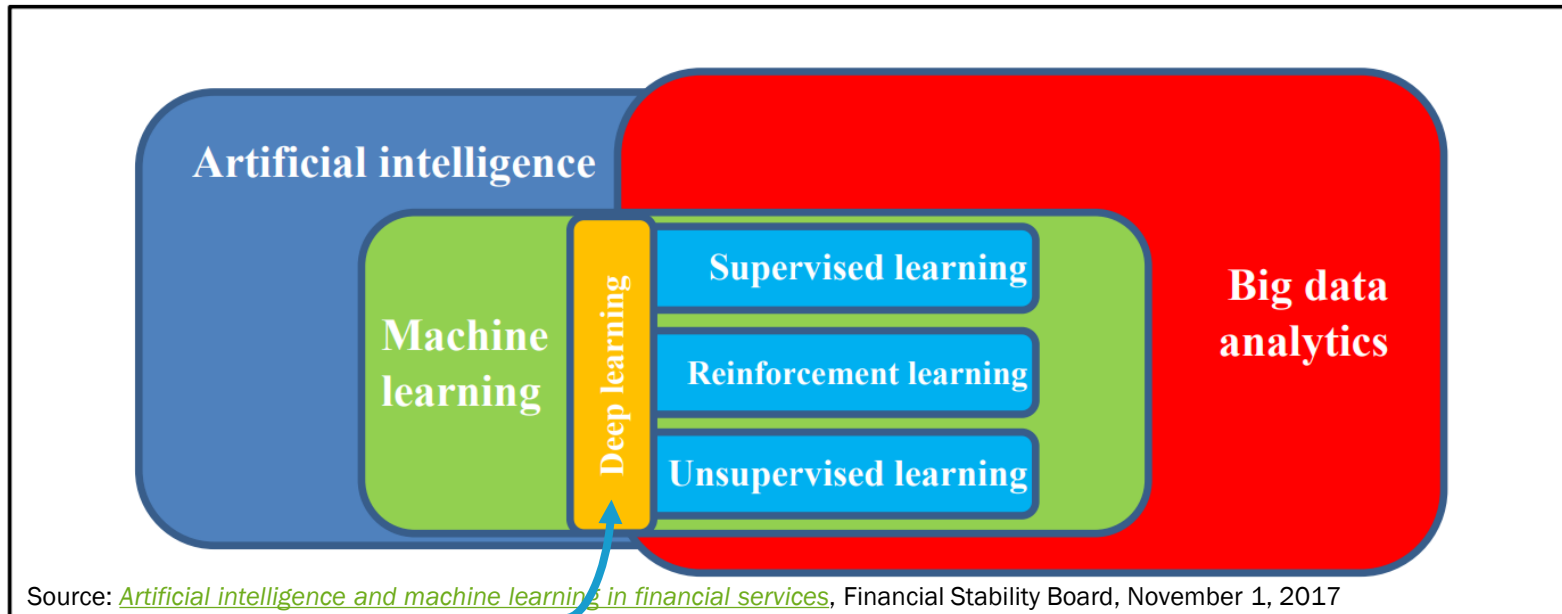


SOME PRELIMINARY DEFINITIONS

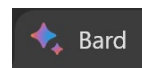
MODELS, ARTIFICIAL INTELLIGENCE, MACHINE LEARNING, ETC.



A SIMPLE CATEGORIZATION OF AI



- **Artificial Intelligence:** “the theory and development of computer systems able to perform tasks that traditionally have required human intelligence” ([FSB, 2017](#)).
- **Machine Learning:** “a method of designing a sequence of actions to solve a problem, known as algorithms, which optimise automatically through experience and with limited or no human intervention” ([FSB, 2017](#)).
- **Deep Learning:** “a form of machine learning that uses algorithms that work in ‘layers’ inspired by the structure and function of the brain” ([FSB, 2017](#)).
- **Large Language Models:** “a machine learning model that aims to predict and generate plausible language” ([Google, 2023](#)).



THE FINANCIAL INDUSTRY AND REGULATORS HAVE CONVERGED ON THE DEFINITION OF A MODEL. WHAT ABOUT AI/ML?

- Regulators have largely converged on the definition of model from first promulgated by the Fed and OCC in 2011 as [SR 11-7](#) and [Bulletin 2011-12](#), respectively. The FDIC adopted identical guidance in 2017 ([FIL-22-2017](#)). (See FHFA's [AB 2013-07](#), Bank of England PRA [SS1/23](#), Canda's Office of Superintendent of Financial Institutions draft [E-23](#))
- Model (SR 11-7): “a quantitative method, system, or approach that applies statistical, economic, financial, or mathematical theories, techniques, and assumptions to process input data into quantitative estimates.”
- Components of a model: Information input + processing + reporting
- SR 11-7 covers qualitative and quantitative approaches, and often covers AI/ML (see p. 7 of the [Model Risk Management chapter](#) in the OCC's *Comptroller's Handbook*).
- A good first guess is that all AI/ML techniques are models for the purposes of governance. But because calling something a model means it must be validated, people sometimes try to have their model not called a model.
- Whether called a model, tool, service, or something else, use of AI/ML must be governed.



RISKS FROM AI

CAPABILITIES AND PREVALENCE OF AI CREATES RISKS FOR FINANCIAL INSTITUTIONS



RISKS FROM AI NOW APPROACHING RISKS OF EMPLOYING PEOPLE

	Non-AI/ML Model	AI/ML	Generative AI	People
Data	<ul style="list-style-type: none"> Data typically not large 	<ul style="list-style-type: none"> 3 Vs: velocity, volume, variety increased governance required 	<ul style="list-style-type: none"> Significant data needed (e.g., the Pile's 825 GB) for training How to govern? 	<ul style="list-style-type: none"> Can exfiltrate, poison, and delete data
Opacity, explainability	<ul style="list-style-type: none"> Often easy to understand (e.g., regression weights, decision trees) 	<ul style="list-style-type: none"> Often opaque (e.g., random forest, neural networks) Explainability techniques (e.g., SHAP, LIME) often required 	<ul style="list-style-type: none"> Significant efforts to finetune are required (e.g., reinforcement learning from human feedback) 	<ul style="list-style-type: none"> Preference falsification, lying, forgetting, lack of understanding of own or business's true motivations
Discrimination, Bias, and Regulatory Compliance	<ul style="list-style-type: none"> Often easy to assess discrimination risks 	<ul style="list-style-type: none"> Increased opacity increases fair lending and other risk of harm, such as privacy, employment discrimination 	<ul style="list-style-type: none"> Even more opaque Large text and image datasets of unknown or uneven quality can perpetuate bias May be cost-prohibitive to clean data or train model to limit bias 	<ul style="list-style-type: none"> Subtle discrimination, bias, unconscious bias, group think, forgetfulness, anchoring, availability bias, base rate neglect, confirmation bias, loss-aversion bias
Privacy and Security	<ul style="list-style-type: none"> Risks from using/storing sensitive data Relatively low demands on IT infrastructure Even without AI movement to cloud may make sense 	<ul style="list-style-type: none"> Automated and intelligent detection of vulnerabilities are balanced by increased capacity of threats Increased complexity and expense of IT 	<ul style="list-style-type: none"> Generative AI can be tricked into revealing private information Deep fakes enhance phishing attacks 	<ul style="list-style-type: none"> Leaking to newspapers or activist organizations, shoulder surfing, document security at home
Performance	<ul style="list-style-type: none"> Some risk of overfitting p-hacking Replication crisis 	<ul style="list-style-type: none"> Increased risk of overfitting; bias/variance trade-off 	<ul style="list-style-type: none"> Unknown risk of overfitting 	<ul style="list-style-type: none"> 80/20 rule in organizations Excellent management is hard

RISK MATRIX FOR AI/ML USE CASES

Lower overall risk

- Google search
- Autocomplete
- Voice-to-text transcription
- Sentiment analysis
- OCR for remote deposit
- Biometric authentication (passkeys)
- Microsoft 365 Copilot
- Virus detection

Medium overall risk

- Automated home valuation
- Resume scanning
- Internal chatbots
- Commercial loan underwriting
- Financial crimes detection
- Automated pair coder (e.g., GitHub Copilot)

High overall risk

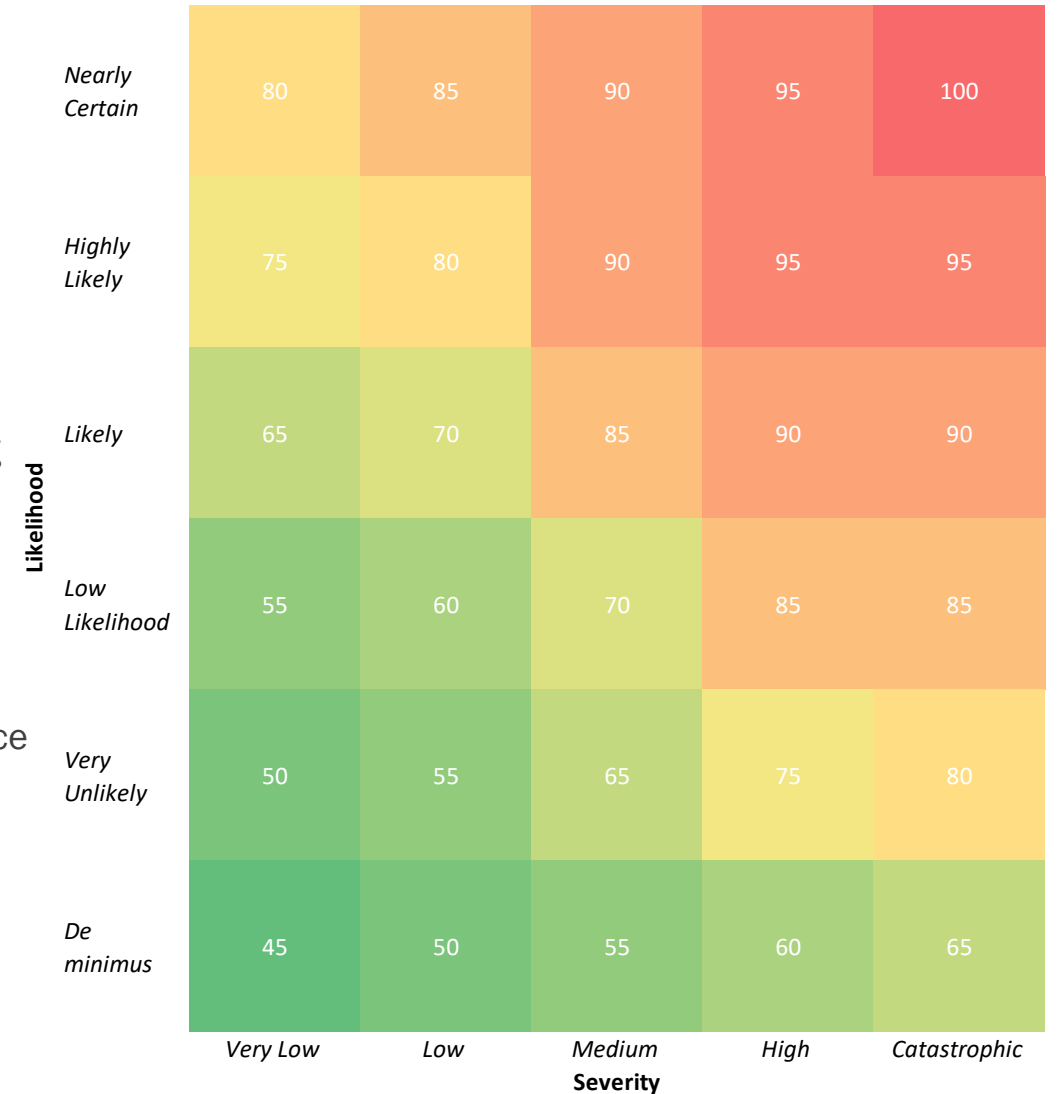
- Consumer-facing chatbots
- Robo-advisor
- Consumer credit underwriting (e.g., credit cards, mortgages) with ML

Unacceptable overall risk

- Consumer credit underwriting using GenAI

To be determined but technology exists

- Automatic summarization of laws, regulation, and guidance for rapid assessment of regulatory compliance
- Automatic model documentation
- Automatic model validation reports
- LLM-driven consumer marketing
- Internal document summarization
- Writing suspicious activity reports





TRUSTWORTHY AI

SOCIETAL AND INDUSTRY CONVERGENCE ON THE GOAL OF AI GOVERNANCE



PRINCIPLED ARTIFICIAL INTELLIGENCE

A Map of Ethical and Rights-Based Approaches to Principles for AI

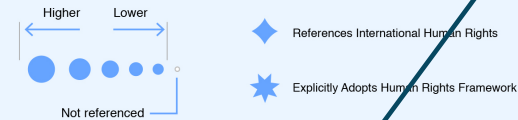
Authors: Jessica Fjeld, Nele Achten, Hannah Hilligoss, Adam Nagy, Madhulika Srikumar

Designers: Arushi Singh (arushisingh.net) and Melissa Axelrod (melissaaxelrod.com)

HOW TO READ:

Date, Location
Document Title
 Actor

COVERAGE OF THEMES:



The size of each dot represents the percentage of principles in that theme contained in the document. Since the number of principles per theme varies, it's informative to compare dot sizes within a theme but not between themes.

The principles within each theme are:

Privacy:

- Privacy
- Control over Use of Data
- Consent
- Privacy by Design
- Recommendation for Data Protection Laws
- Ability to Restrict Processing
- Right to Rectification
- Right to Erasure

Accountability:

- Accountability
- Recommendation for New Regulations
- Impact Assessment
- Evaluation and Auditing Requirement
- Verifiability and Replicability
- Liability and Legal Responsibility
- Ability to Appeal
- Environmental Responsibility
- Creation of a Monitoring Body
- Remedy for Automated Decision

Safety and Security:

- Security
- Safety and Reliability
- Predictability
- Security by Design

Transparency and Explainability:

- Explainability
- Transparency
- Open Source Data and Algorithms
- Notification when Interacting with an AI
- Notification when AI Makes a Decision about an Individual
- Regular Reporting Requirement
- Right to Information
- Open Procurement (for Government)

Fairness and Non-discrimination:

- Non-discrimination and the Prevention of Bias
- Fairness
- Inclusiveness in Design
- Inclusiveness in Impact
- Representative and High Quality Data
- Equality

Human Control of Technology:

- Human Control of Technology
- Human Review of Automated Decision
- Ability to Opt out of Automated Decision

Professional Responsibility:

- Multistakeholder Collaboration
- Responsible Design
- Consideration of Long Term Effects
- Accuracy
- Scientific Integrity

Promotion of Human Values:

- Leveraged to Benefit Society
- Human Values and Human Flourishing
- Access to Technology

Further information on findings and methodology is available in *Principled Artificial Intelligence: Mapping Consensus in Ethical and Rights-Based Approaches* (Berkman Klein, 2020) available at cyber.harvard.edu.



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NIST'S AI RMF LIKELY TO BECOME STANDARD ACROSS INDUSTRIES

- The National Institute of Standards and Technology (NIST) released its [AI Risk Management Framework \(RMF\)](#) in 2023. Other work at NIST includes the [US AI Safety Institute](#), [managing bias](#), [explainability](#), and [security](#).
- Leading technology companies have similar principles, typically under the term “responsible AI”: [Microsoft](#), [Google](#), [Meta](#), [OpenAI](#), [IBM](#), and [Amazon](#).



Source: [National Institute of Standards and Technology, Artificial Intelligence Risk Management Framework \(AI RMF 1.0\)](#), p. 12



SOME AUTHORITATIVE STATEMENTS ON AI GOVERNANCE

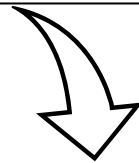
LAWS, REGULATIONS, GUIDANCE, STATEMENTS, ETC.



AN INCOMPLETE AND OPINIONATED AI GOVERNANCE TIMELINE

Pre-2020 (Models and “normal” AI/ML)

- Apr 2011 [SR 11-7/ Bulletin 2011-12](#) Guidance on MRM
- Jun 2017 [FDIC adopts FRB, OCC MRM Guidance](#)
- Nov 2017 [FSB AI/ML in Financial Services](#)
- Feb 2019 [Executive Order 13859 on American Leadership in AI](#)



2020 – 2021 (Increasing focus on risks from AI/ML)

- Feb 2020 [European Commission Whitepaper on AI](#)
- Dec 2020 [Executive Order 13960 on Trustworthy AI](#)
- Jun 2020 [Artificial Intelligence in the Securities Industry \(FINRA\)](#)
- Mar 2021 [AI/ML RFI \(FRB, OCC, FDIC, NCUA, CFPB\)](#)
- Apr 2021 [European Commission proposes AI Act](#)
- Jun 2021 [GAO Accountability Framework](#)
- Aug 2021 [Model Risk Management, Controller’s Handbook \(OCC\)](#)

2022 – 2023 (Cambrian explosion after ChatGPT)

- Feb 2022 [FHFA AB 2022-02](#) and [OMWI SL on AI/ML](#)
- Mar 2022 [Identifying and Managing Bias in AI \(NIST\)](#)
- Oct 2022 [White House Blueprint for an AI Bill of Rights](#)
- Jan 2023 [NIST Risk Management Framework](#)
- Apr 2023 [Measures for Management Generative AI \(China\)](#)
- Nov 2023 [Bletchley Declaration](#)
- May 2023 [SS1/23 – MRM Principles for Banks \(BoE\)](#)
- Jun 2023 [European Parliament Approves Position on AI Act](#)
- Jun 2023 [Chatbots in Consumer Finance \(CFPB\)](#)
- Jul 2023 [Predictive Analytics Proposed Rule \(SEC\)](#)
- Sep 2023 [Guidance on Credit Denials using AI \(CFPB\)](#)
- Oct 2023 [Draft TC260 on generative AI \(China\)](#)
- Oct 2023 [Executive Order 14110 on Trustworthy AI](#)
- Dec 2023 [Questionnaire on AI/ML, Quantum Computing \(OSFI\)](#)
- Dec 2023 [Provisional Agreement on EU’s AI Act](#)



FHFA'S 2022 GUIDANCE ILLUSTRATES CURRENT REGULATORY THINKING

- The FHFA's guidance, [Advisory Bulletin 2022-02](#), contains four parts. This should be read in conjunction with a [Supervisory Letter](#) from the Office of Minority and Women Inclusion.
- This AB is the sole guidance on AI/ML from a US Federal financial regulatory agency, although regulators have made statements, issued requests for information, and updated handbooks.

I. Governance

- AI/ML Core Ethical Principles
 - Transparency
 - Accountability
 - Fairness and equity
 - Diversity and inclusion
 - Reliability
 - Privacy and security
- AI/ML Definitions and Taxonomy
- AI/ML Inventory
- Roles and Responsibilities
- Policies, Standards, and Procedures

II. Risk Identification and Assessment

- Model risks
- Data risks
- Other operational risks
- Regulatory and compliance risks

III. Control Framework

- Model controls
- Data controls
- Other operational controls
- Regulatory and compliance controls

IV. Risk Monitoring, Reporting, and Communication



PRACTICAL CONSIDERATIONS IN AI GOVERNANCE

WHAT STEPS CAN ORGANIZATIONS TAKE NOW AND IN THE NEAR TERM?

DEVELOP AI STRATEGY AND ETHICAL PRINCIPLES

- **Develop an AI Strategy** to ensure alignment across the organization in terms of how AI will be acquired, used, and governed. “The essence of strategy is choosing what not to do.” Michael Porter.
 - Risk tolerance/appetite and risk taxonomy
 - Buy or build? Lead or follow? Which products?
 - Ranking use cases and risks
 - Reference to complementary frameworks (e.g., model risk)
 - Clarify roles, responsibilities, and authorities
- **Develop AI Ethical Principles**
 - Are existing company statements adequate, perhaps with some tweaks?
 - Don't just copy/paste. Customize based on strategy, risks, and benefits.
 - Good starting point are transparency, trustworthiness, safety, human in the loop, and non-discrimination



ENSURING EFFECTIVE CHALLENGE OF AI/ML

- The guiding principle is ensuring effective challenge for AI/ML, including GenAI. This may require:
 - Updating the RCSA and doing a risk assessment.
 - Updating the model inventory and coordinating with or rationalizing this and other inventories, like IT. How to handle ubiquitous AI/generative AI?
 - Reconsidering the definition of model. Include all AI/ML?
 - Revising roles and responsibilities (e.g., committees, working groups, business unit leadership, model owners, developers).
 - Communicating updates and conducting role-based training.
 - Updating other frameworks and controls across the organization (e.g., credit risk, vendor, software development lifecycle, data, vendor).
 - Revising management and board reporting and risk limits to address AI and GenAI risks specifically.



ENSURING EFFECTIVE CHALLENGE OF AI/ML, ADDITIONAL STEPS

- Additional steps to ensure effective challenge may include:
 - Considering incremental risks from AI across the model lifecycle.
 - Easing compliance burden and increasing consistency of governance with enablers (e.g., templates, testing scripts).
 - Carefully considering governance for anything deemed not a model under company policy but containing AI/ML, including GenAI.
 - Revising model risk tiering with AI in mind and scaling governance to risk.
 - Managing vendor relationships with a view to validation, ethical principles, etc.



ILLUSTRATIVE PLAN FOR ADOPTION OF GENERATIVE AI



Initiate

- Set project plan and goals
- Get Board and senior management buy-in
- Identify executives to own outcomes
 - Sketch roles and authorities
 - Determine whether consultants are needed



Communicate

- Set up working groups and committees
- Establish communication plan for Board updates, working groups, committees, and employees



Engage Resources

- Engage and consultants as needed
- Develop hiring plan for strategic areas in AI/ML
- Engage technology vendors to understand capabilities



Revise Governance

- Develop AI strategy
- Identify top use cases
- Identify unacceptable risks, use cases, and technologies
- Update policies, procedures
- Write AI ethical principles



Run Pilot Program

- Start with small set of most promising use cases
- Configure sandbox & identify staff
- Run tests
- Socialize results



Assess Pilot & Iterate

- Track lessons learned
- Optimize pilot process (people, process, technology) for higher throughput
- Move to next set of use cases



Approve and Roll Out

- Finalize review and testing from pilot (engage MRM & others per governance)
- Put into production



PARTING THOUGHTS

PRINCIPLES AND SPECULATIONS

HIGH-LEVEL PRINCIPLES FOR NAVIGATING AI

- If it was illegal before AI, it is illegal with AI.*
 - A corollary: if it was required before AI, it is required with AI.
 - A good example is CFPB's [2023 statement](#) that adverse action notices in consumer credit are required regardless of technology used to make credit decisions.
- While general principles from NIST AI RMF and other organizations are helpful, financial services is already highly regulated. Separate frameworks for AI may not be needed provided existing frameworks are updated intelligently.
- It is impossible to plan for every eventuality. With AGI potentially happening within our lifetimes, people and organizations need to be resilient – meaning sound procedures, good communication, good infrastructure, etc.
- Beware the precautionary principle, panic-driven policies, and box-checking bureaucratization.
- Consider trade-offs in every decision. There are risks from using AI but also risks from not.
- Governance of models and AI is likely to change very rapidly in the near term. (See next slide.)

* I am not a lawyer. This is not legal advice. Seek advice of competent counsel.

SPECULATION ABOUT THE NEAR FUTURE

- FIs are burdened with regulatory compliance and expensive employees. With AI now or soon able to perform as well as the [best available human](#) in many tasks, expect FIs to use AI to build, document, implement, and validate models.
 - Pair coding ([Amazon CodeWhisperer](#), [GitHub Copilot](#), [Replit](#)) + LLM trained on banking regulations, guidance, model documentation, and model validation reports + AutoML tools ([Amazon SageMaker](#), [DataBricks](#), [Dataiku](#)) → BankerGPT that builds and validates models?
- What are we humans to do when AI is at or near human-level performance?
 - Find a government job, preferably one with a union (e.g., IRS, OCC, FHFA, FDIC).
 - Become a super-star model developer or validator who understands the business, can code, knows the math, and can communicate clearly (especially writing) so that your output is the input for training AI.
 - Search the “[jagged technological frontier](#)” to identify ways that AI fails (hallucinates convincingly). Become the human in the loop, i.e., leverage your domain expertise to make sure the AI is operating as designed.



APPENDIX



GLOSSARY

- AB Advisory Bulletin (from FHFA)
- AGI Artificial general intelligence
- AI/ML Artificial intelligence/machine learning
- AVM Automated valuation model (in residential real estate)
- BoE Bank of England
- CFPB Consumer Financial Protection Bureau
- ECOA Equal Credit Opportunity Act
- EU European Union
- FDIC Federal Deposit Insurance Corporation
- Fed The Board of Governors of the Federal Reserve System
- FHFA Federal Housing Finance Agency
- FI Financial institution, e.g., bank, broker-dealer, insurance company
- FIL Financial Institution Letter (from the FDIC)
- FINRA Financial Industry Regulatory Authority
- FRB See Fed
- FSB Financial Stability Board
- GAO Government Accountability Office
- GenAI Generative artificial intelligence, e.g., LLMs like ChatGPT
- GPT Generative pre-trained transformer
- IT Information technology
- LLM Large language model
- MRM Model risk management
- NCUA National Credit Union Administration
- NIST National Institute of Standards and Technology
- OCC Office of the Comptroller of the Currency
- OMWI Office of Minority and Women Inclusion (at FHFA)
- OSFI Office of the Superintendent of Financial Institutions
- PRA Prudential Regulatory Authority
- RCSA Risk and control self-assessment
- RFI Request for information
- RMF Risk Management Framework (from NIST)
- SEC Securities and Exchange Commission
- SL Supervisory Letter (from FHFA)
- SR Supervision and Regulation Letter (from the Fed)

FURTHER READING – WHERE TO START

Model Risk

- [Guidance on Model Risk Management \(FRB, OCC, FDIC\)](#)
- [Model Risk Management Guidance \(FHFA\)](#)
- [S1/23 – MRM Principles for Banks \(BoE\)](#)

AI/ML Risk Management

- [AI/ML Risk Management \(FHFA\)](#)
- [FSB AI/ML in Financial Services](#)
- [Request for Comment on AVMs \(Multiple US Agencies\)](#)
- [Circular on ECOA Requirements with Black-Box AI \(CFPB\)](#)
- [Identifying and Managing Bias in AI \(NIST\)](#)
- [Risk Management Framework \(NIST\)](#)
- [Chatbots in Consumer Finance \(CFPB\)](#)

- [Predictive Analytics Proposed Rule \(SEC\)](#)
- [Guidance on Credit Denials using AI \(CFPB\)](#)

NGOs and Other

- [Institute for Human-Centered AI](#)
- [FinRegLab](#)
- [Ethan Mollick, One Useful Thing \(Substack\)](#)

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Kathryn Wilkens, PhD, CAIA
Founder of Pearl Quest LLC, FDP Curriculum Advisor

Dr. Stylianos Kampakis
CEO, The Tesseract Academy

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