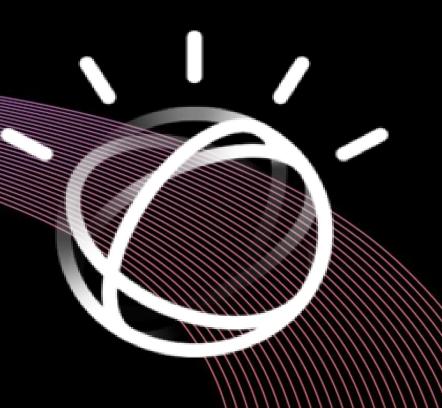
Trustworthy AI with IBM



Brian Snyder Sr Data Science Technical Specialist bsnyder@us.ibm.com





Rising concerns on trust in AI decisions

YouTube sued for using AI to racially profile content creators

They claim YouTube's algorithms discriminate against black users

BlackRock shelves unexplainable AI liquidity models

COVID-19?

Risk USA: Neural nets beat other models in t Data science during COVID-19: Some reassembly required

> Most likely, the assumptions behind your data science model or the patterns in your data did not survive the coronavirus pandemic. Here's how to address the challenges of model drift

Can AI models respond to black swan events like

Over-Segmenting In Financial Services Is So Over - Bye, Bye

The Washington Post Sections = Get 1 year for \$29 Democracy Dies in Darkness Apple Card algorithm sparks gender bias allegations against Goldman Sachs

Amazon scraps secret AI recruiting tool that showed bias against women

EFF to HUD: Algorithms Are No Excuse for Discrimination

BY JAMIE WILLIAMS, SAIRA HUSSAIN, AND JEREMY GILLULA | SEPTEMBER 26, 2019

...and regulators are catching up

USA

SR 11–7 requires model risk management for all models in financial services

2019—Proposal for Algorithmic Accountability Act

2021 – US Govt. National AI Initiative Act

Canada

models

2017—National AI Strategy launched. Impact Analysis

2020—All public agencies must do an impact analysis for AI **European Union**

2021 – Draft regulation for trust in AI development

2019—Guidelines for AI development

Mexico

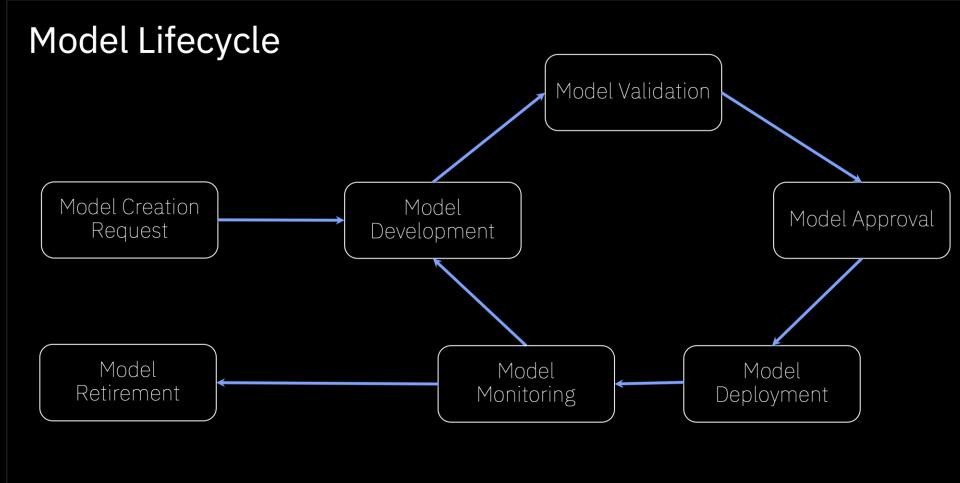
2018—General principles for AI development in the government

Partnerships on AI

Partnership between tech companies to study best practices and impact of AI

AI Now Institute

NYU research center focused on social implications of AI



Aspects of Trustworthy AI



Fairness

Impartial and addressing bias

Are **privileged groups** at a systematic advantage compared to other groups?



Robustness

Models need to perform well across the lifecycle, handle exceptions effectively, enable confidence in systems outcomes

Are relevant performance **metrics** monitored over time?



Drift

Changes in input data cause model to make inaccurate decisions.

Do anomalies exist between training data and data ranges or combinations seen in real life?



Explainability

Easy to understand outcomes/decisions



Transparency

Open to inspecting facts and details

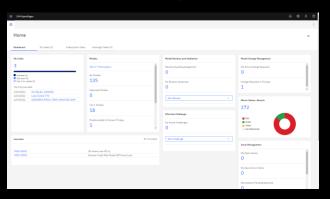
Why did the AI arrive at an outcome? At what point would the outcome have been different?

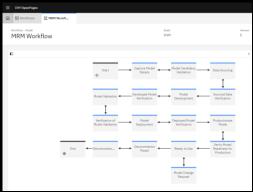
Can we increase understanding of why and how AI was created?

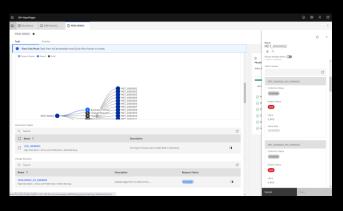
- **Common AI Use Cases**
- Lending: Loan Origination, Loan Default
- Collections
- Claims Processing
- Underwriting

- Targeted Marketing Campaigns
- Segmentation
- Customer Management
- HR

Trustworthy AI Demo





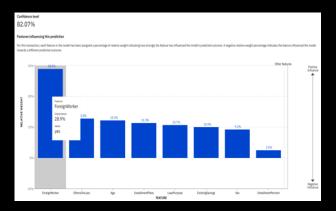


Enterprise Inventory Dashboard

Enterprise Workflow

Variety of Risk Metrics Captured For Reporting

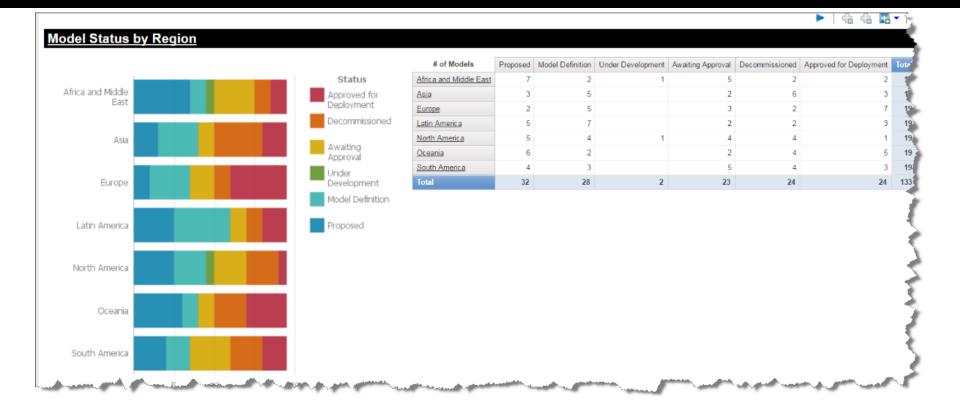




Drift in data consistency, Drift in accuracy, Bias Detection

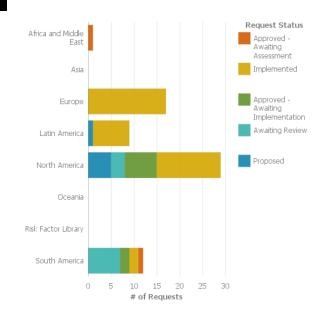
Local and contrastive explanations

Model Status by Region



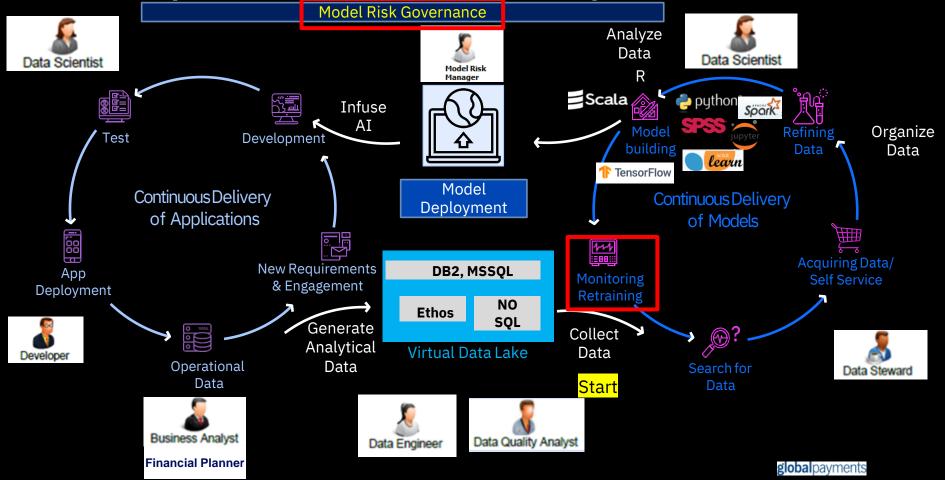
Change Request by Region

Change Requests by Region



# of Requests	Proposed	Awaiting Review	Approved - Awaiting Implementation	Implemented	Approved - Awaiting Assessment	Total
Africa and Middle East					1	1
Europe				17		17
Latin America	1			8		9
North America	5	3	7	14		29
South America		7	2	2	1	12
Total	6	10	9	41	2	68

ModelOps Data Science and Trustworthy AI as a Team - DEMO



US-based Multinational Bank

Upon movement of all proof-of-concept projects into production, the bank will have the ability to govern all AI projects using their existing technology and skill investment so that existing business units do not need to change their current systems and reskill their employees.

*IBM Cloud Pak for Data provides organizations a transversal and centralized view of the Al lifecycle through an integrated platform that covers the three key functions of model build, model deploy, and model management.

Business problem

This bank uses many tools and systems for model development and deployment, making model governance challenging. They were unable to ensure models comply with enterprise policies, identify inefficiencies, provide standardized regulatory reporting, learn and scale best practices

Solution

IBM used its leading enterprise insights platform (Cloud Pak for Data) with its open architecture enabling a smooth integration of models developed and deployed on other platforms from a governance perspective

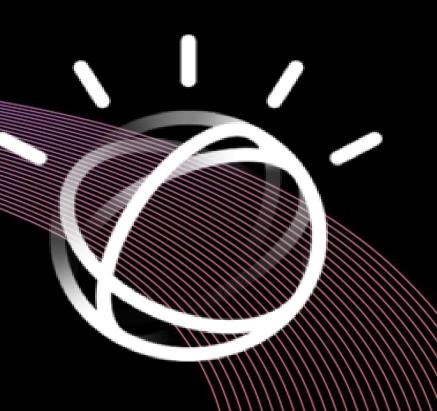


Trustworthy AI with IBM



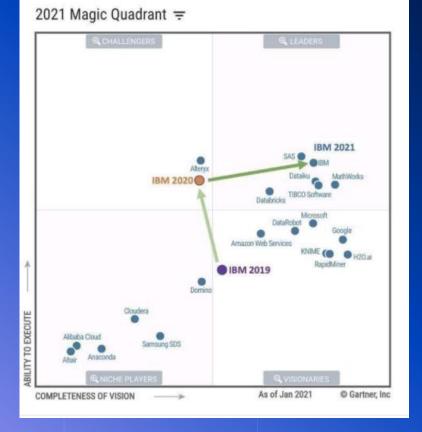
Brian Snyder
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bsnyder@us.ibm.com

THANK YOU!



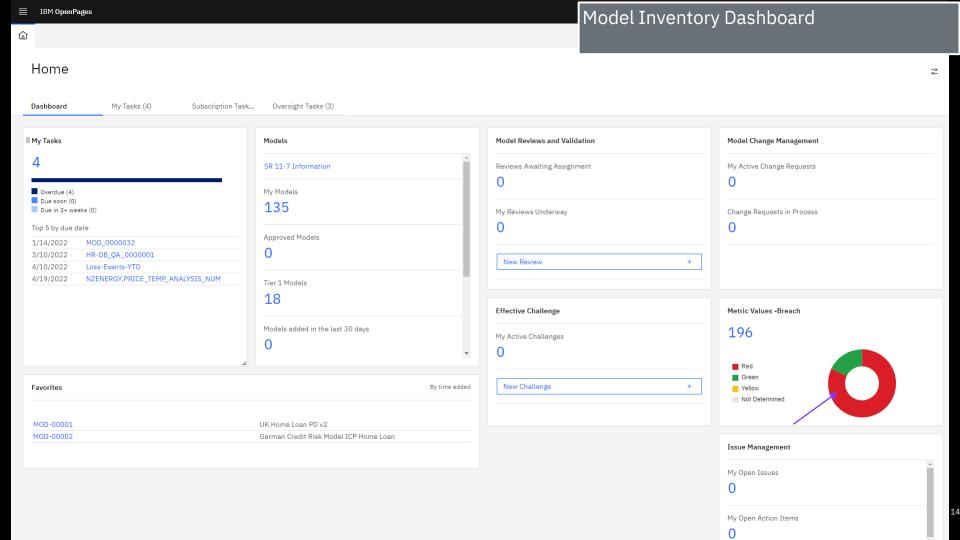
Demo Snapshots

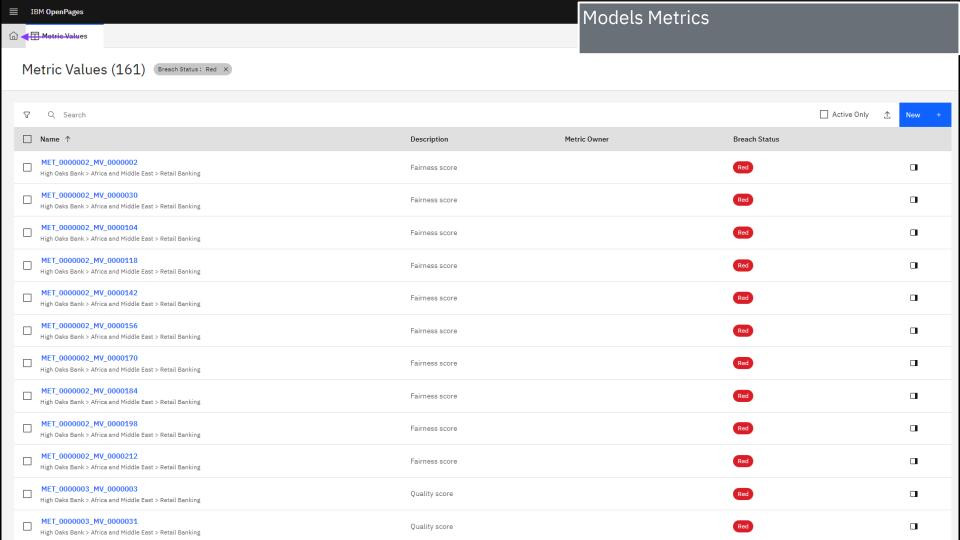
Gartner Magic Quadrant for **Data Science** and Machine Learning **Platforms**



Supports multiple tasks across the data science life cycle, including:

- Problem and business context understanding
- Data ingestion
- Data preparation
- Data exploration
- Feature engineering
- Model creation and training
- Model testing
- Deployment
- Monitoring
- Maintenance
- Data and model governance
- Explainable artificial intelligence (XAI)
- Business value tracking
- Collaboration





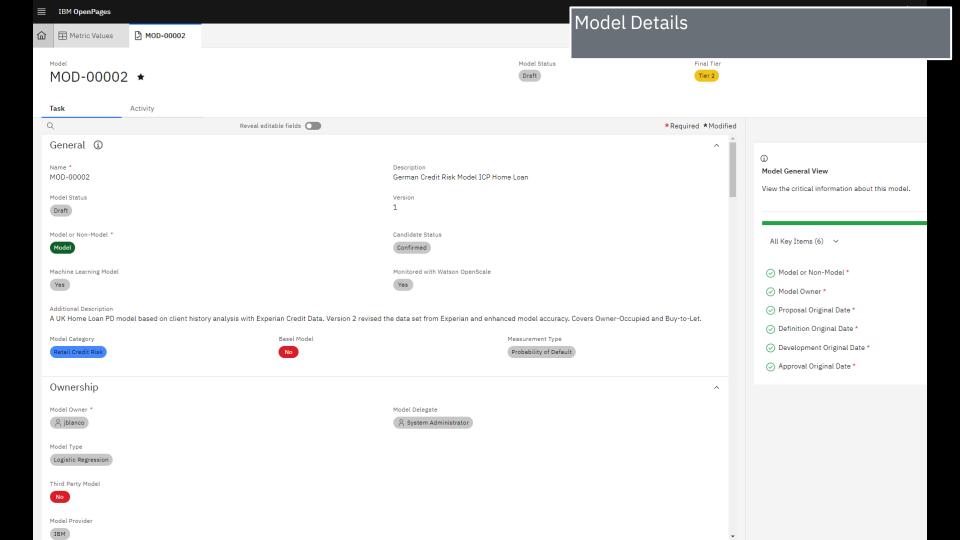
Choose Favorite Model

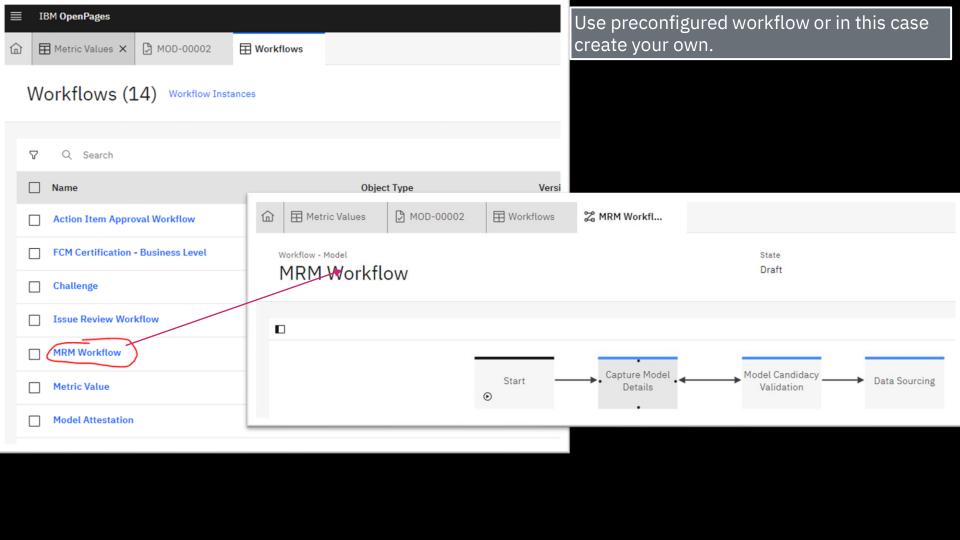
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Home

My Tasks (4) Subscription Task... Oversight Tasks (3) Dashboard **∷** My Tasks Model Change Management Models Model Reviews and Validation My Active Change Requests SR 11-7 Information Reviews Awaiting Assignment My Models Overdue (4) Due soon (0) 135 Due in 2+ weeks (0) My Reviews Underway Change Requests in Process Top 5 by due date Approved Models 1/14/2022 MOD_0000032 HR-DB_QA_0000001 3/10/2022 New Review 4/10/2022 Loss-Events-YTD 4/19/2022 NZENERGY.PRICE_TEMP_ANALYSIS_NUM Tier 1 Models 18 Effective Challenge Metric Values -Breach Models added in the last 30 days 196 My Active Challenges Red Green New Challenge Favorites By time added Yellow Not Determined MOD-00001 UK Home Loan PD v2 MOD-00002 German Credit Risk Model ICP Home Loan Issue Management My Open Issues My Open Action Items

10





Fairness and Artificial Intelligence in Banking

Use of AI/ML creates many risks – bias, lack of explainability, operational, ... – with implications for both compliance and risk management

Regulators issued guidance regarding these risks, but the regulatory landscape continues to evolve as new evidence of consumer impact emerges







Complexity



Explainability

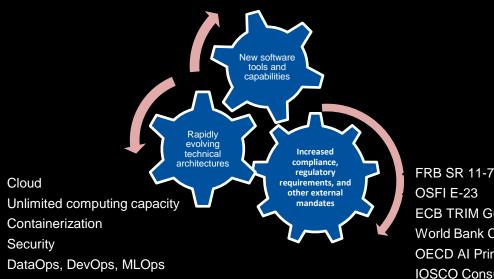


Alternative Data

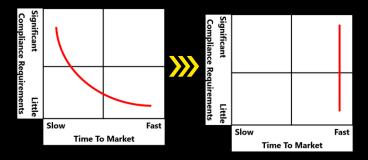
Very specific challenges relate to the nature of a data science model.

> The technology and regulations, and how they are applied, are constantly evolving.

Traditional (SAS, Matlab, IBM, etc.) ML-centric (IBM Watson, Datarobot, RapidMiner, H2O, etc.) New entrants (Algorithmia, ModelOp, Modzy, etc.)



How to ensure the necessary velocity and quality in a changing, multi-system and multi-source environment?

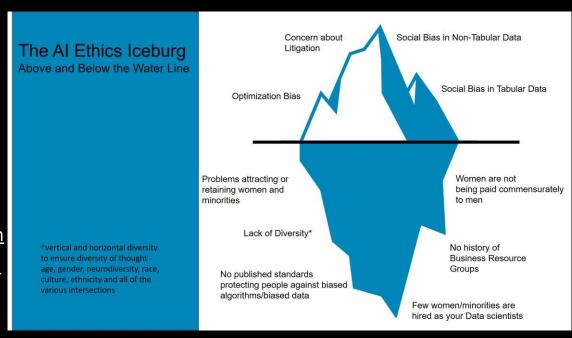


OSFI E-23 **ECB TRIM Guideline** World Bank Credit Scoring Guidelines **OECD AI Principles** IOSCO Consultation report on AI **CA Consumer Privacy Act**

Operationalizing AI requires organizations to demonstrate responsible, explainable use of AI



Driven by data → validity <u>not guaranteed</u>
Models → not code, often <u>unexplainable</u>
Probabilistic → non-deterministic, <u>uncertain</u>
Not just focus on version control, not an <u>assembly line</u> → Model development MUST be part of lifecycle management



Model drift, bias and risk can pose significant liabilities and damage



AI fairness is a corporate social responsibility

