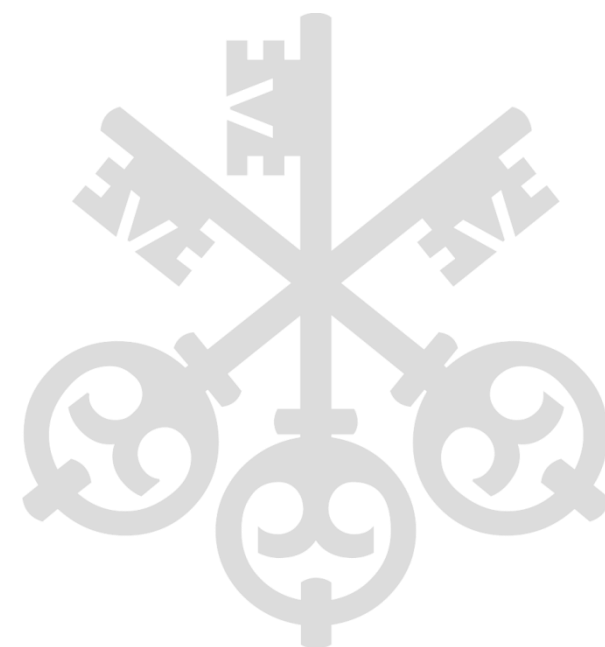


From Risk Model Governance to Model Risk Governance

Risk Day 2021

Michael Amrein, Dr. sc. ETH Zürich, Aktuar SAV
Head AI, Monitoring & Surveillance Models Validation
Model Risk Management & Control



Disclaimer

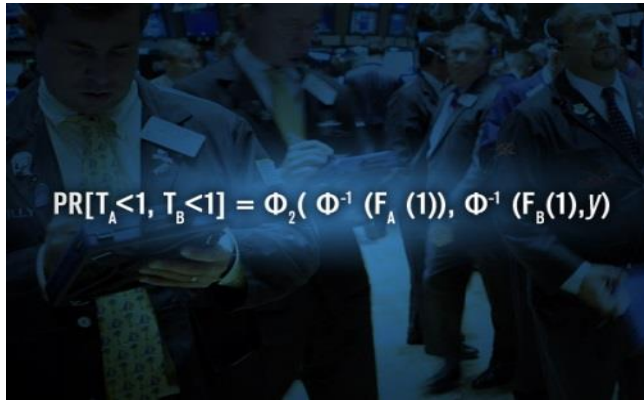
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Model Risk

Growing relevance of model risk in the financial industry due to increasing reliance on models.



Recipe for Disaster: The Formula That Killed Wall Street

(WIRED, Feb 23, 2009)

Relevance: Rating agencies and banks underestimated both the probability and magnitude of stress losses as well as default correlation.

Impact: Large U.S. and European banks lost more than \$ 1tn on toxic assets between 2007 and 2009.



Software Testing Lessons Learned From Knight Capital Fiasco

(CIO, Aug 14, 2012)

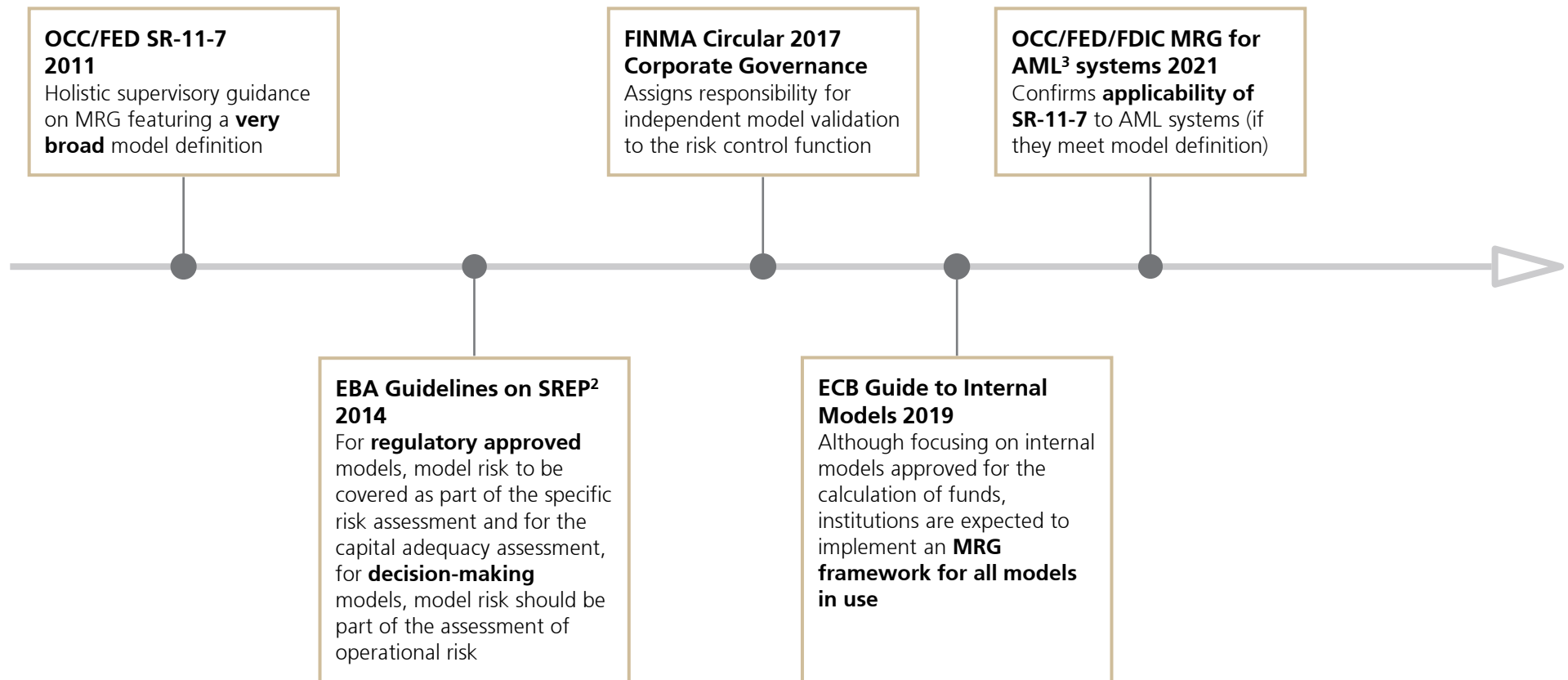
Relevance: Inadequate testing before production release of a trading algorithm. "It was a software bug [...]. It happened to be a very large software bug." Knight Capital CEO Thomas Joyce.

Impact: One defect in a trading algorithm caused Knight Capital to lose \$ 440m in about 30 minutes.

Model risk is the risk of adverse consequences resulting from decisions based on incorrect or misused model outputs and reports.

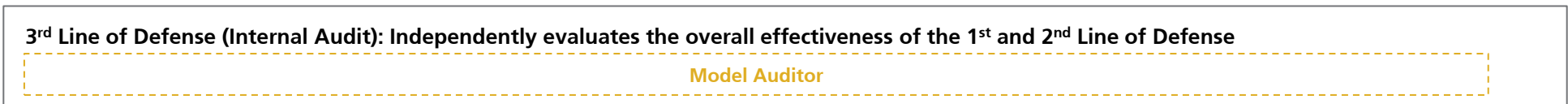
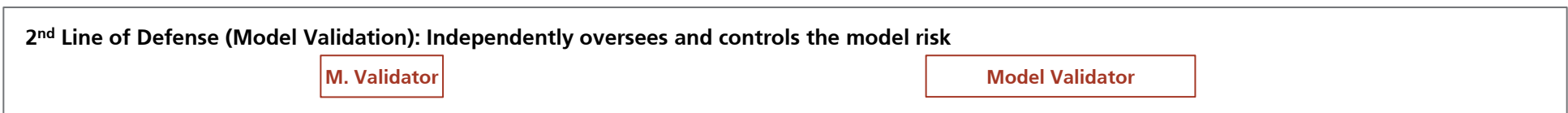
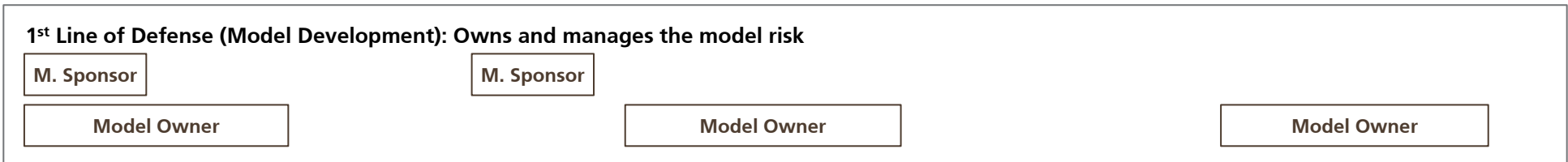
Some MRG¹ Regulatory Milestones

Over the past years, regulatory bodies confirmed that global banks need to implement a comprehensive MRG framework with coverage beyond the traditional models (minimum and economic capital, stress testing, liquidity, valuation).



MRG Processes, Roles and Responsibilities in a Nutshell

Like any other risk, model risk should be managed via three lines of defense over the model's lifecycle.



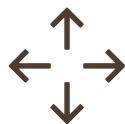
- Has budget authority to commission models for development to meet business needs
- Has approval for use authority for models approved by the 2nd Line of Defense

- Responsible for model risk and its management throughout the model's lifecycle
- Ensures that model development (including documentation and implementation), use and ongoing monitoring is performed in line with policies
- Remediates issues identified by the 2nd Line of Defense

- Provides independent assessment and effective challenge of model risk along a model's lifecycle
- Performs independent model reviews and raises issues (if applicable) in line with policies

Key Aspects of the MRG Enhancement

Extension to new areas, holistic model risk coverage including risk appetite setting and enhanced reporting capacity.



Onboarding existing systems to MRG

Extension of MRG framework to existing, non-traditional models, used for example in the context of Monitoring & Surveillance of Operational Risks and Algorithmic Trading



Artificial Intelligence and Machine Learning

Increasing adoption of Artificial Intelligence and Machine Learning across the bank, especially also in areas not using models previously



Holistic coverage of model risks at individual model level

Development and independent reviews moved away from a methodology-centric approach to holistic coverage of input, methodology, implementation and use including ongoing performance monitoring



Model Risk Measurement and Appetite

Model Risk Measurement and formulation of Model Risk Appetite not only on an individual model level but also in the aggregate



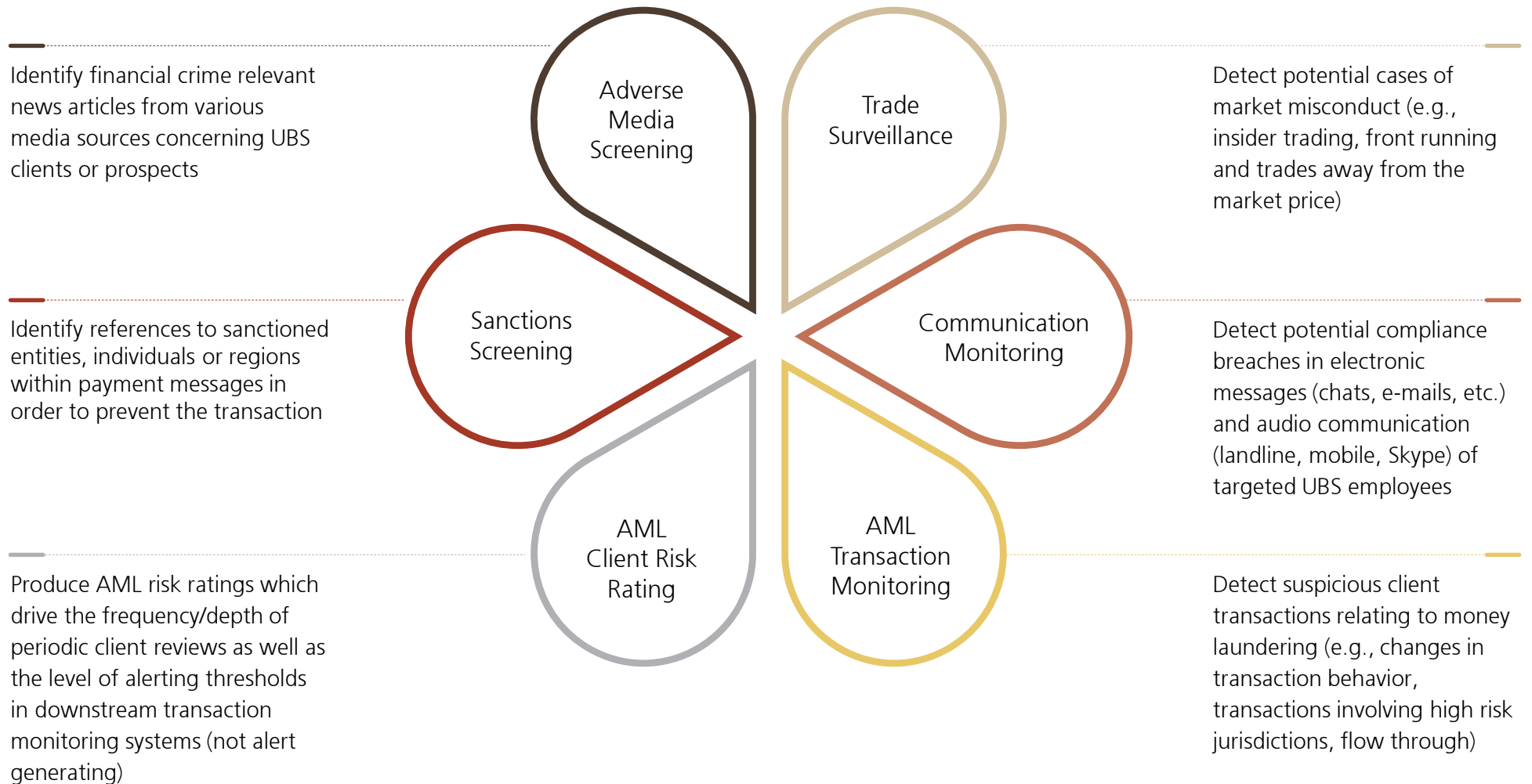
Reporting & Monitoring

Reporting & monitoring of the governance status of individual models as well as in the aggregate for senior management, auditors and regulators

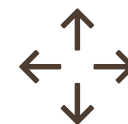
Onboarding Existing Systems: M&S¹ Models – Model Landscape



M&S models are typically alert generators to identify various types of suspicious activities and patterns. Alerts are reviewed by experts for potential escalation.



Onboarding of Existing Systems: M&S Models – Characteristics & Model Risk



Use	Input data	Methodology	Implementation
<ul style="list-style-type: none">• Many models monitor key operational risks (Financial Crime, Market Conduct)• Alerts go through an expert review process and might ultimately lead to a regulatory filing	<ul style="list-style-type: none">• Large amounts of data (trades, orders, text, audio, transactions, payments, client data), typically sourced from core systems• Processing is usually automated	<ul style="list-style-type: none">• Monthly, daily or event based execution of the alerting logic• Many submodels based on rules with many tunable parameters, statistical anomaly detection and/or Machine Learning	<ul style="list-style-type: none">• Inhouse built systems as well as on- and off-premise vendor solutions• Implementation under responsibility of the IT department



Key model risk are false negatives (Type II error)

The model does not produce an alert when it should have (false alerts "only" lead to extra effort and are well controlled through alert review)



Key testing / controls

Regular reviews of non-alerting cases / Below-the-Line testing
Regular testing with synthetic data
Regular coverage assessments (regulatory/internal requirements)

Artificial Intelligence and Machine Learning: Newly Emerging Use Cases - Challenges



There are challenges to adapt the framework to AI/ML¹ as it enables applications in formerly MRG-remote areas rather than introducing fundamentally new model risks².

MRG Awareness

Increase awareness and understanding of the model governance process, the roles it defines, the model lifecycle requirements and its benefits in AI/ML development teams of solution owners

Model Identification

Identify AI/ML models in areas of the bank that have no standing collaboration with Model Risk Governance & Control units

Delineation to risks controlled by other functions (Compliance, Legal, Information Security)

Are the below model risks?

- Logic embedded in a trading algorithm places disadvantage to certain clients or manipulates the market
- Predictions of employee misconduct resulting in investigations prior to any wrong-doing
- Deployment of self-adapting software whose behavior changes without a release
- Algorithmic support in job candidate screening tools resulting in a preference for a particular nationality/gender/ethnicity

Artificial Intelligence and Machine Learning: Newly Emerging Use Cases – Model Landscape

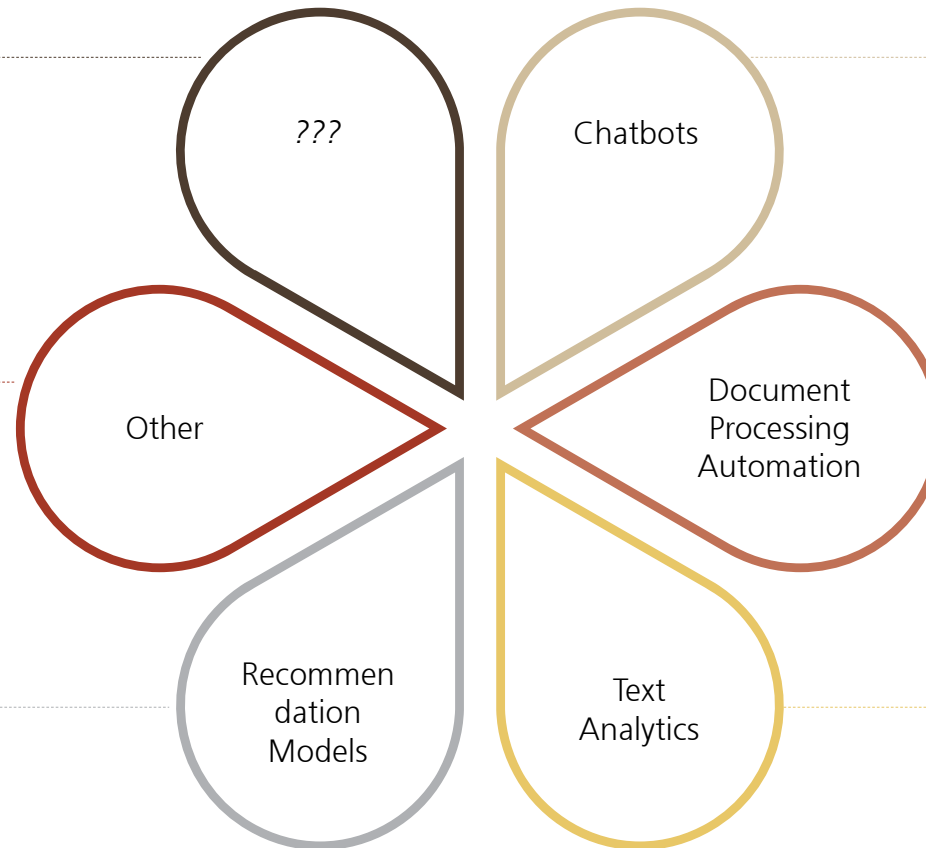


Many new AI models currently used in production present minimal model risks due to their business use.

The AI/ML area is in its infancy and further use cases are investigated

Various models utilizing AI for forecasting, classification, or anomaly detection (e.g., predict work volumes for Group Operations, identify abnormal behavior in log patterns)

Suggest to a client advisor products that might be of interest to their clients based on client attributes, portfolio characteristics, market information, and other sources



Consume user input and trigger actions based on a prediction of the user's intent (retrieve pre-defined text answers or create tickets for the relevant response team).

Scan documents to determine whether and where specific information is located therein (e.g., ISIN of a security or the name of the counterparty that a contract refers to)

Use natural language processing for classification and prediction (e.g., e-mail prioritization, HR general request triage)



Holistic Coverage of Model Risks

Model risks should be assessed consistently in all model areas and holistically along various model risk factors (see an example of a model risk taxonomy below).

Dimensions	Factors	Risk Description
Inputs	1.1 Input Automation	Input data preparation/collection not automated hence posing significant operational risks
	1.2 Input Data Quality	Use of unreliable, unrepresentative or poor-quality data sources, high risk or non-validated feeder models
	1.3 Input Data Appropriateness	Use of data which is unsuitable for the model's intended purpose, reliance on questionable proxies
	1.4 Documentation and Governance	Missing key information on model inputs, insufficient oversight and controls of input data
Methodology	2.1 Analytical Assumptions	Use of analytical assumptions which are inappropriate with respect to the model purpose
	2.2 Expert Assumptions	High dependency of model outcome on expert judgment
	2.3 Conceptual Framework	Ineffective or flawed model design (including input, processing and output)
	2.4 Calibration and Parametrization	Inappropriate, unstable or insufficiently justified calibration
	2.5 Complexity	High or inappropriate model complexity leading to increased potential for undetected issues
	2.6 Documentation and Governance	Missing key information (e.g., assumptions) or unclear responsibilities regarding model development and design
Implementation	3.1 Implementation Soundness	Implementation or coding errors in the production environment
	3.2 Operational Stability	Unauthorized changes, insufficient level of access control, insufficient information on how to operate model
	3.3 Documentation and Governance	Missing key information or unclear responsibilities for model implementation
Model Use	4.1 Alignment with Purpose	Model inappropriately used outside its intended purpose and/or validated scope of applicability
	4.2 Performance Monitoring	Poor model performance
	4.3 Ongoing Monitoring	Inappropriate design of ongoing performance monitoring
	4.4 Reporting	Inaccurate, unreliable or unintuitive reporting of model outputs
	4.5 Documentation and Governance	Missing key information (e.g., limitations, restrictions) or unclear responsibilities regarding model use

Model Risk Measurement & Appetite: Individual Model Level



Inherent Risk Rating (IRR)

Drives frequency and depth of regular independent reviews

Combines

- Model Materiality, the impact of model failure
- Model Complexity, the likelihood of model failure induced through complexity of inputs, methodology and implementation

		Complexity		
		High	Medium	Low
Materiality	High	High	High	Medium
	Med	High	Medium	Low
	Low	Medium	Low	Low
	Imm	Immaterial		

Independent Review

Systematic identification of issues along various model risk factors, including issue severity ratings

Risk dimensions

Input

Input Automation
Input Data Quality
Input Data Appropriateness
Documentation and Governance
Analytical Assumptions
Expert Assumptions
Conceptual Framework
Calibration and Parametrization
Complexity

Methodology

Documentation and Governance
Implementation Soundness
Operation Stability
Documentation and Governance

Use

Alignment with Purpose
Performance Monitoring
Ongoing Monitoring
Reporting
Documentation and Governance

Residual Risk Rating (RRR)

Factors in the outcome of independent reviews and mitigating controls, starting from the IRR as baseline

Risk reducing factors may be:

- Independent reviews are performed in line with prescribed cycles
- Issues are mitigated by compensating controls
- Assumptions, limitations and weaknesses of the model are transparently communicated
- Overarching controls (ongoing performance monitoring, output reviews and sign-offs)

Risk Appetite is formulated in terms of issue severity

- Fundamental issues or conceptual flaws lead to a rejection of the model
- ▨ Severe issues need to be remediated timely and mitigated by compensating controls, or remediated before go-live
- ▶ Moderate issues can be accepted for a longer time horizon

Model Risk Measurement & Appetite: Model Portfolio Level



There are various metrics that could be used to monitor the aggregated model risk. Trigger breaches should be investigated, escalated and addressed.

Metric class	Examples
Model Riskiness	Percentage of models in use with a high Residual Risk Rating Percentage of models in use with a high Inherent Risk Rating Percentage of models in use with open severe issues Percentage of not yet validated models in use ...
Regulatory Matters	Number of regulatory matters related to Model Risk ...
Model Risk Control Framework	Percentage of models in use with significantly overdue revalidation Percentage of models with overdue remediation of validation issues Percentage of models in use with material model inventory data quality issues ...
Model Risk Concentrations	Percentage of models with severe issues in a particular model risk dimension Percentage of models in a particular model area with severe issues ...



Reporting & Monitoring

Central model inventory should provide high quality model data for key reports.

Model Risk Report

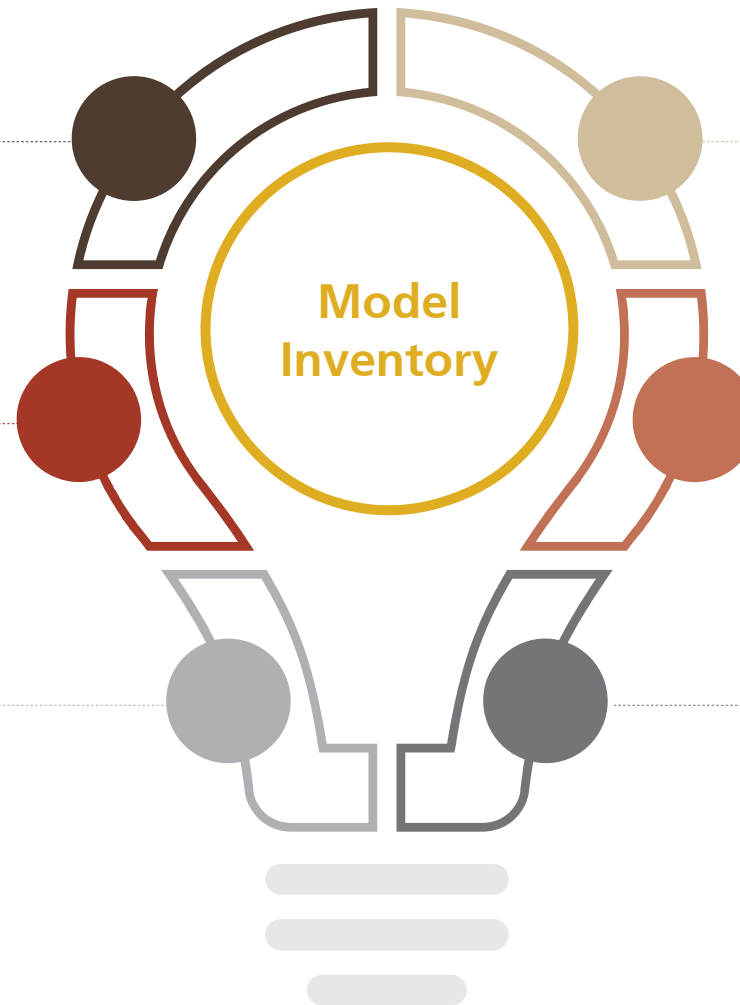
Provides senior management, auditors and regulators with all relevant model portfolio information, supplemented by corresponding narratives

Risk Appetite Monitoring

Daily monitoring of model risk appetite metrics

Model Type Reports

Dedicated model type reports are available to management, sliced and diced by model type, regulatory purpose and/or legal entity



Data Quality Monitoring

Various online reports allow to monitor inventory data quality, including a case manager for issue resolution

Model Reports

Single-model specific reporting providing an overview of the current model status

Stakeholder Dashboards

Online dashboards allow stakeholders to oversee his/her portfolio and to ensure that warning flags are adequately managed

How to Cope with the Enhanced Expectations?

Various measures were implemented to remain effective and efficient.



People & Talent

Significant personnel increase in core locations
Campus recruiting at India's top universities






IT Infrastructure

Common platform for development, independent validation & production for classical risk models
Central model inventory supporting process workflows and storage all model data and documents
Enhanced reporting capabilities



Validation Report
Structuring &
Automation

Reports are built automatically from standardized blocks (supplemented with narratives where needed):
 Analysis: risk factor, objective, design, results, conclusion, identified model risk
 Issues: risk factor, description, criticality & justification, opened on, closed on, reason for closure
 Assumptions, Limitations & Weaknesses
Establishment of analysis menus for well known model groups



Model Tiering

Depth of (independent) testing and revalidation frequency depends on Inherent Risk Rating
Immaterial use cases are not independently reviewed



Simplification of
Model Landscape

Unify inputs, methodology and implementation for similar use cases
Merge similar models

Questions?

Please don't hesitate to ask

