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# Machine Learning for Supervisory Analytics and Risk Detection Training Course

31 May – 4 June 2027
Munich (Germany)



# Machine Learning for Supervisory Analytics and Risk Detection Training Course

**Ref:** 3255552_201698 **Date:** 31 May – 4 June 2027 **Location:** Munich (Germany) **Fees:** 5900 GBP

Machine learning for supervisory analytics and risk detection is the use of statistical models that learn from reported financial data to flag unusual institutions, transactions and market conditions for supervisory attention. This one-week course is for experienced supervisors, risk analysts and supervisory data specialists, and its final deliverable is a documented early warning model design with an explainability and model risk note ready for internal review.

## Introduction

The course rests on three established method families: supervised learning, where models learn from labelled past outcomes; unsupervised learning, where models find structure and outliers without labels; and model explainability through SHAP values, a technique that attributes each prediction to the inputs that drove it. It is taught at an advanced, practitioner depth for delegates who already read supervisory returns. Learning happens through modelling builds on anonymised reporting data and case studies drawn from banking, payments and money services businesses.

## Course Objectives

- Define the main machine learning tasks used in supervision and match each one to a recurring supervisory question
- Distinguish supervised from unsupervised learning and decide which suits a dataset with few confirmed problem cases
- Compare random forest, XGBoost and hidden Markov models on accuracy, stability and the effort needed to maintain them
- Interpret anomaly scores and SHAP values so that a flagged entity can be explained to a supervisory panel
- Apply text analytics and network metrics to separate risks specific to one firm from risks that spread across a sector
- Justify the data, validation and monitoring choices behind an early warning model in a written design note



## Target Audience

- Supervisors of banks and other financial institutions who must decide which model alerts warrant an on-site follow-up
- Financial stability and risk analysts who must judge whether a stress signal reflects one firm or the wider sector
- Supervisory data and analytics specialists who must select model types and features for reporting datasets
- Model validation and internal audit professionals who must assess whether a supervisory model is reliable enough to use
- Payments and conduct oversight staff who must choose thresholds for flagging irregular transaction patterns

## Course Outline

### Day 1: Supervisory Data and Machine Learning Foundations

- Supervisory Analytics Use Cases: Mapping Model Tasks to Recurring Oversight Questions
- Supervised vs Unsupervised Learning: Labelled Outcomes Compared with Pattern Discovery
- Regulatory Return Data Preparation: Cleaning, Reconciling and Versioning Reported Figures
- Feature Engineering for Prudential Ratios: Turning Balance Sheet Lines into Model Inputs
- Class Imbalance in Failure Data: Handling Rare Distress Events in Training Sets

### Day 2: Classification Models for Early Warning

- Logistic Regression vs Random Forest: Interpretability Weighed Against Predictive Power
- XGBoost Gradient Boosting: Building Stronger Learners from Sequential Decision Trees
- Early Warning Indicator Design: Defining Target Events and Prediction Horizons
- Cross-Validation Over Time: Testing Models Without Leaking Future Information
- Alert Threshold Calibration: Balancing Missed Distress Cases Against False Alarms

### Day 3: Anomaly Detection and Weakly Labelled Data

- Isolation Forest vs Clustering Outliers: Two Routes to Spotting Unusual Entities
- Weakly-Supervised Learning: Two-Layer Approaches When Confirmed Cases Are Scarce
- Transaction Pattern Screening: Irregular Activity in Payments and Money Services Outlets
- Hidden Markov Models: Detecting Shifts Between Calm and Stressed Market Regimes
- Ensemble Methods for Stress Indices: Combining Models to Signal Crisis Episodes

### Day 4: Text Analytics, Network Metrics and Explainability

- NLP Sentiment Analysis: Reading Firm Disclosures and News for Early Signals
- Idiosyncratic vs Systemic Risk Signals: Firm-Level Anomalies Compared with Contagion Paths
- Network Metrics for Interconnectedness: Centrality and Exposure Links Between Institutions
- SHAP Values for Supervisory Review: Attributing Each Alert to Its Driving Inputs
- Model Risk and Governance: Validation, Bias Checks and Ongoing Performance Monitoring



## Day 5: Early Warning Model Design Practice

- Use Case Selection Exercise: Choosing a Supervisory Question and Matching Model Task
- Champion vs Challenger Models: Comparing Two Candidate Designs on Holdout Data
- Alert Triage Simulation: Applying Thresholds and Explanations to a Flagged Entity List
- Deployment and Monitoring Plan: Data Refresh, Drift Checks and Retraining Triggers
- Early Warning Model Design Note: Completing the Explainability and Model Risk Record

## Skills You Will Gain

- Supervisory Feature Engineering
- Imbalanced Classification
- Gradient Boosted Trees
- Anomaly Scoring
- Regime Detection
- Financial Text Sentiment Analysis
- Interconnectedness Analysis
- Model Explainability

## Why Attend This Course

- From reviewing returns line by line after the event to ranking institutions by a model-based distress probability before problems surface
- From relying on fixed ratio thresholds to calibrating alert levels that weigh missed cases against false alarms
- From accepting a model score without explanation to tracing each alert to the inputs that drove it
- From scattered model experiments to a documented early warning model design with an explainability and model risk note

## Conclusion

Supervisory machine learning turns the large volume of data that institutions already report into ranked signals that tell supervisors where to look first. The course makes three distinctions clear: supervised versus unsupervised learning, firm-specific versus sector-wide risk signals, and a model that predicts well versus a model that can be explained and defended. It suits supervisors and analysts who already understand prudential reporting and now need an advanced, hands-on route to building and challenging the models behind early warning.

## Course FAQs

### What is machine learning for supervisory analytics and risk detection?

It is the application of learning algorithms to regulatory returns, transaction records and text so that supervisors can rank institutions by risk, spot irregular activity and receive early warning of distress, rather than relying only on manual review and fixed ratio limits.



## How does supervised learning differ from unsupervised learning in supervisory risk detection?

Supervised learning trains on past cases already labelled as failures, breaches or fraud and predicts the same outcome. Unsupervised learning needs no labels and highlights entities that behave unlike their peers, which helps when confirmed problem cases are rare.

## Which methods does the machine learning for supervisory analytics course follow?

Delegates work with logistic regression, random forest, XGBoost, isolation forest, clustering and hidden Markov models, add sentiment analysis and network metrics, and explain results with SHAP values within a model risk and validation routine.

## Is machine learning for supervisory risk detection suitable for model validation staff?

Yes. Validation and internal audit staff gain a working view of how supervisory models are built, where they fail and which explainability and monitoring evidence to request before a model is relied on.

## What do delegates produce on the machine learning for supervisory analytics course?

Delegates produce an early warning model design note: the supervisory question, data and features, chosen model with its challenger, alert thresholds, SHAP-based explanations and a deployment and monitoring plan.

