

A Bibliometric Analysis

The Impact of
Artificial Intelligence and Machine Learning
on the Banking Sector



Research Questions

1. How have AI/ML been applied in the banking sector to address key challenges?
2. What are the emerging trends in AI/ML research in banking?
3. What gaps exist in the current literature on AI/ML applications in banking?
4. How can bibliometric and thematic analyses contribute to identifying future research directions?

Research Questions

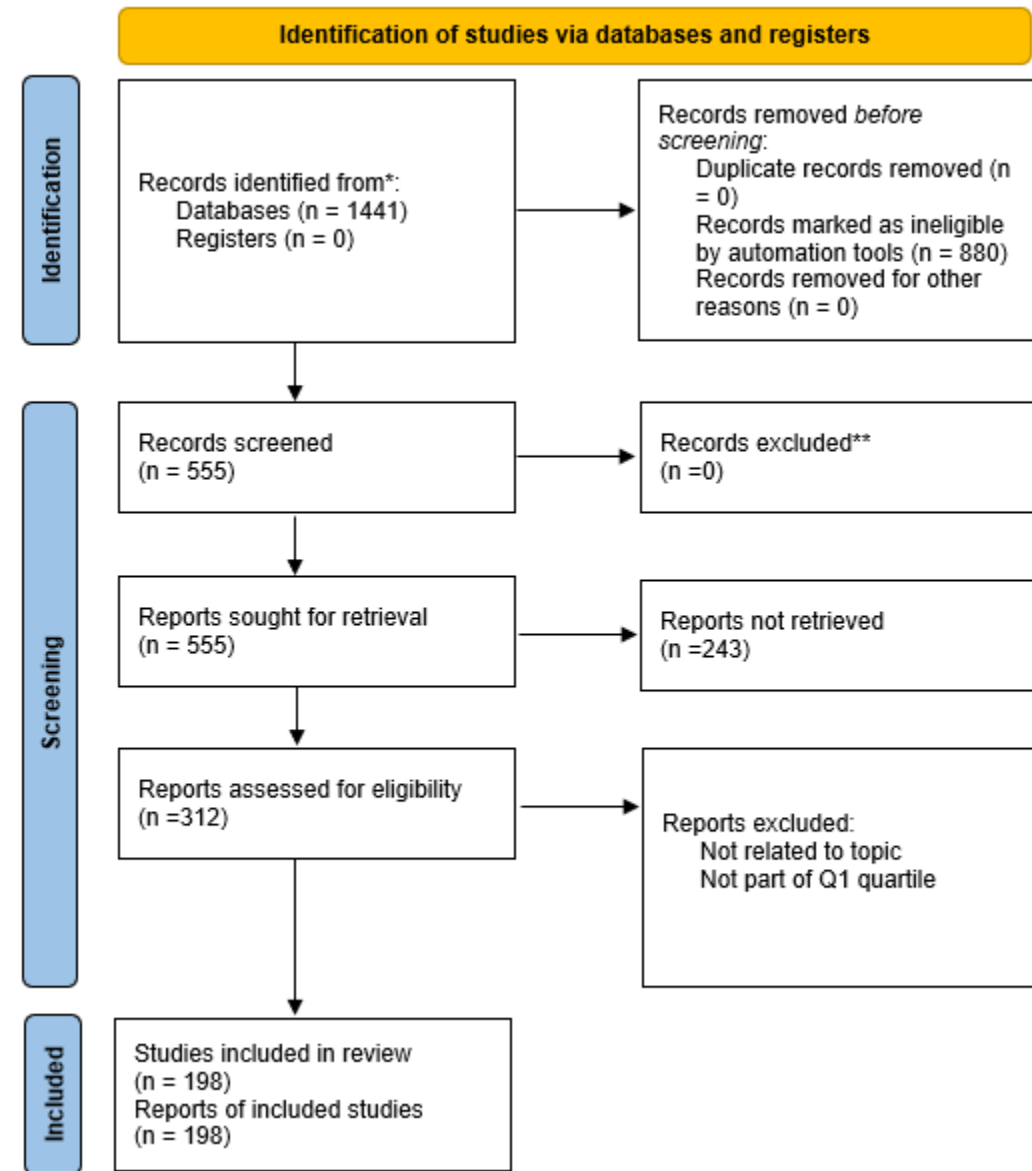
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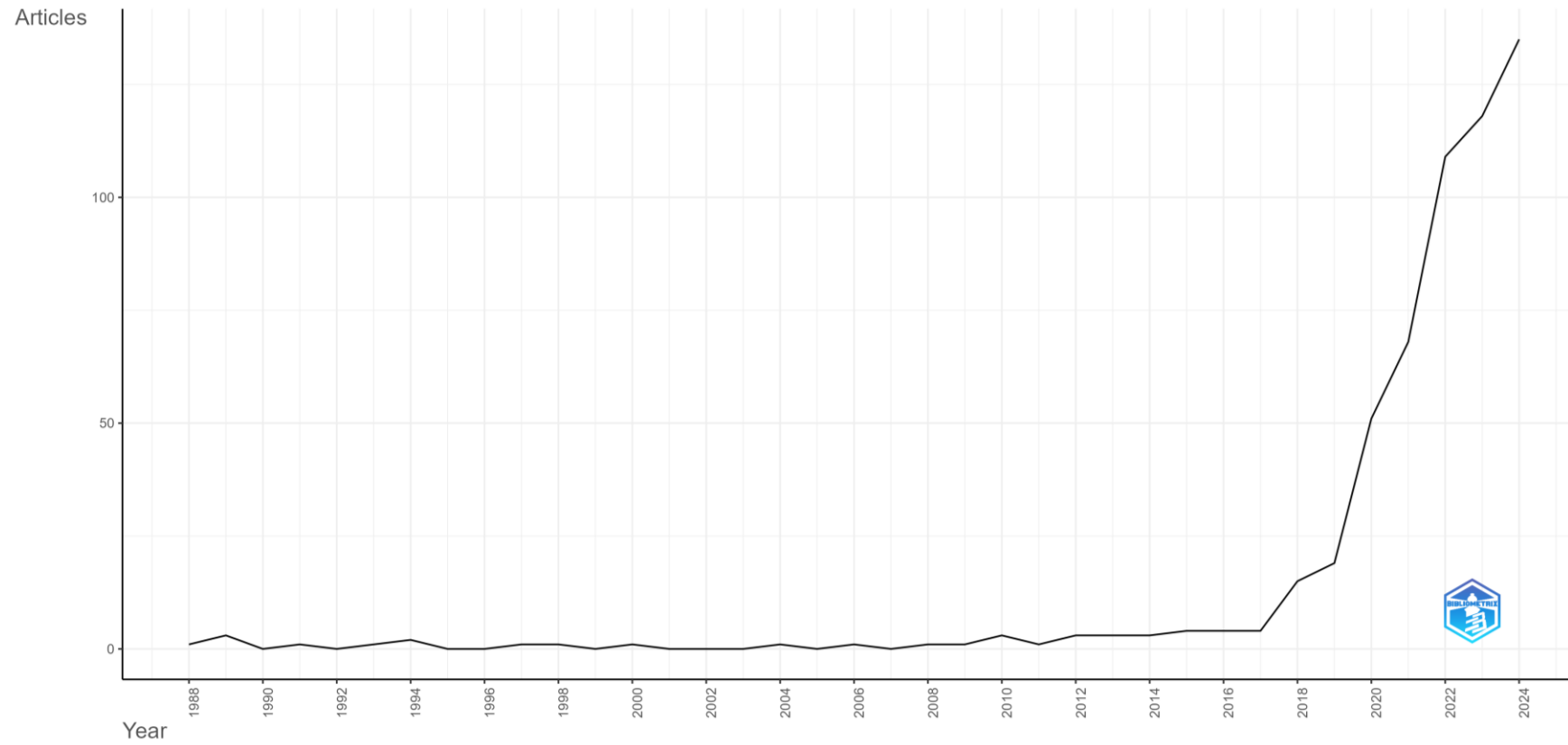
Study Selection Process (PRISMA Flow Diagram)

- Initial pool: 1,441 records identified.
- Automated screening removed 880 irrelevant/duplicate entries.
- 243 records unavailable, leaving 312 reports for eligibility assessment.
- Final selection: 198 high-quality studies based on topic relevance and Q1 quartile.



Annual Scientific Production

- Research output remained low and stable from 1988 to 2016.
- A sharp increase in publications from 2017 onward.
- Reflects growing interest and advancements in AI/ML for banking applications.

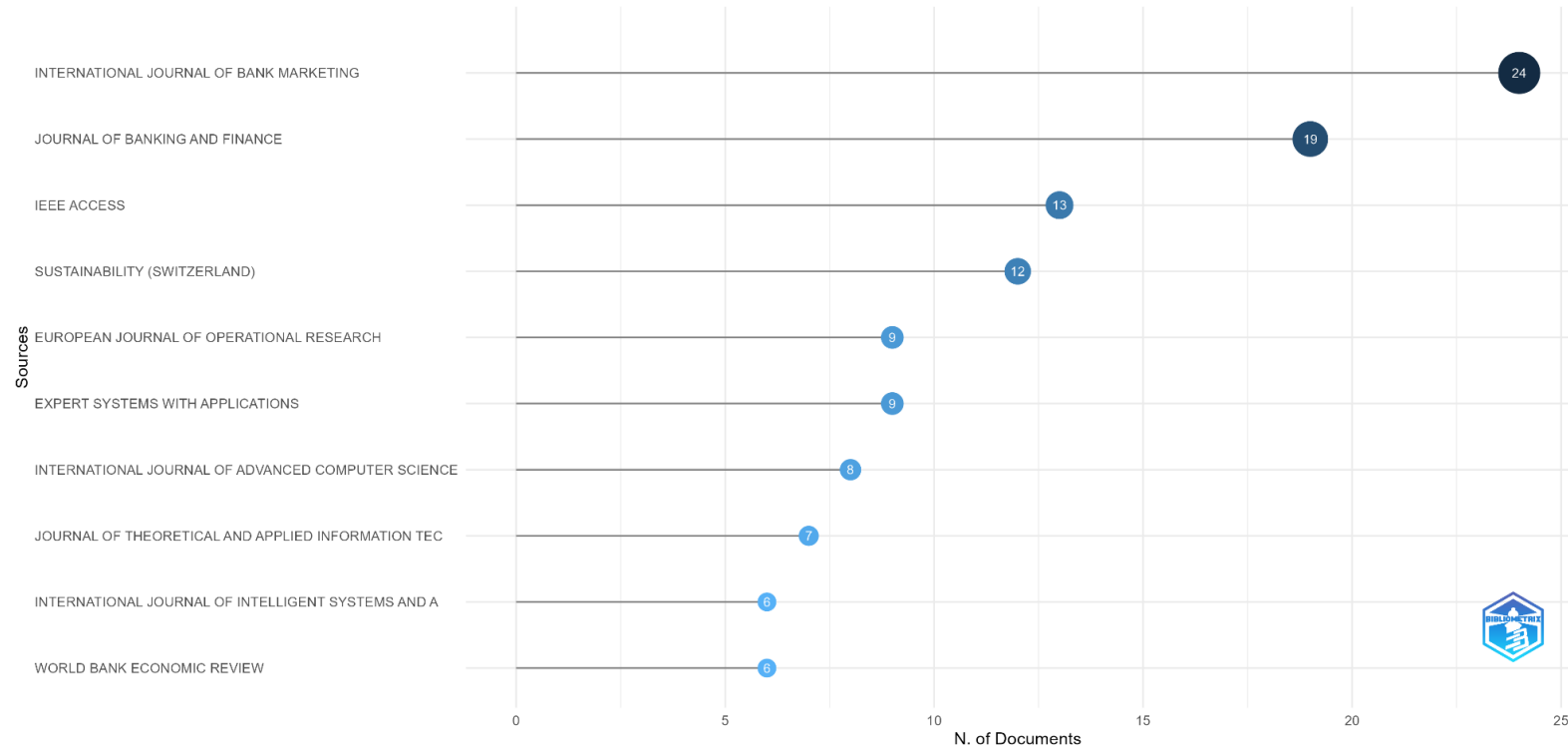


Most Relevant Sources

•Top journals publishing AI/ML research in banking:

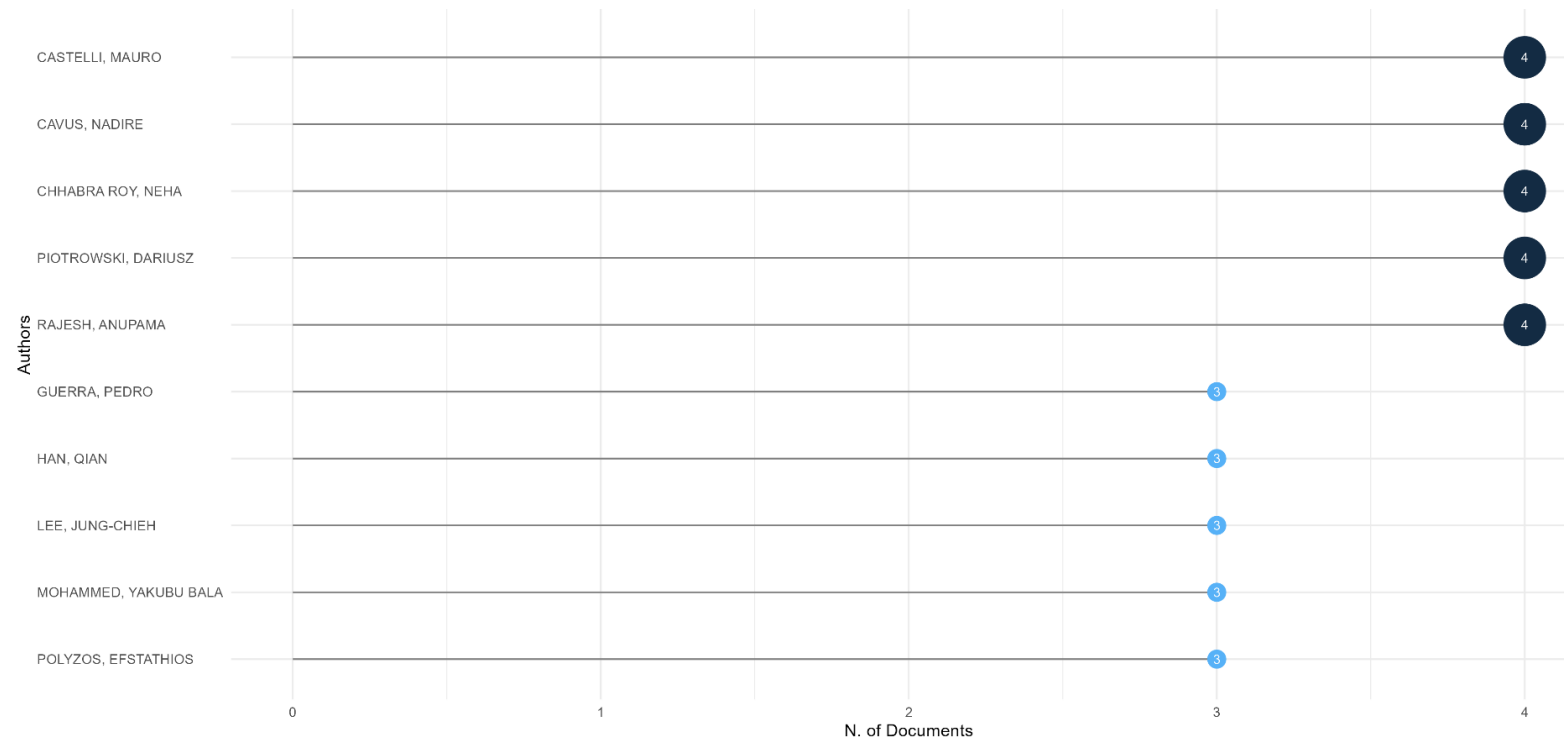
- International Journal of Bank Marketing (24 articles).
- Journal of Banking and Finance (19 articles).
- IEEE Access (13 articles).

•Highlights the interdisciplinary nature of research.



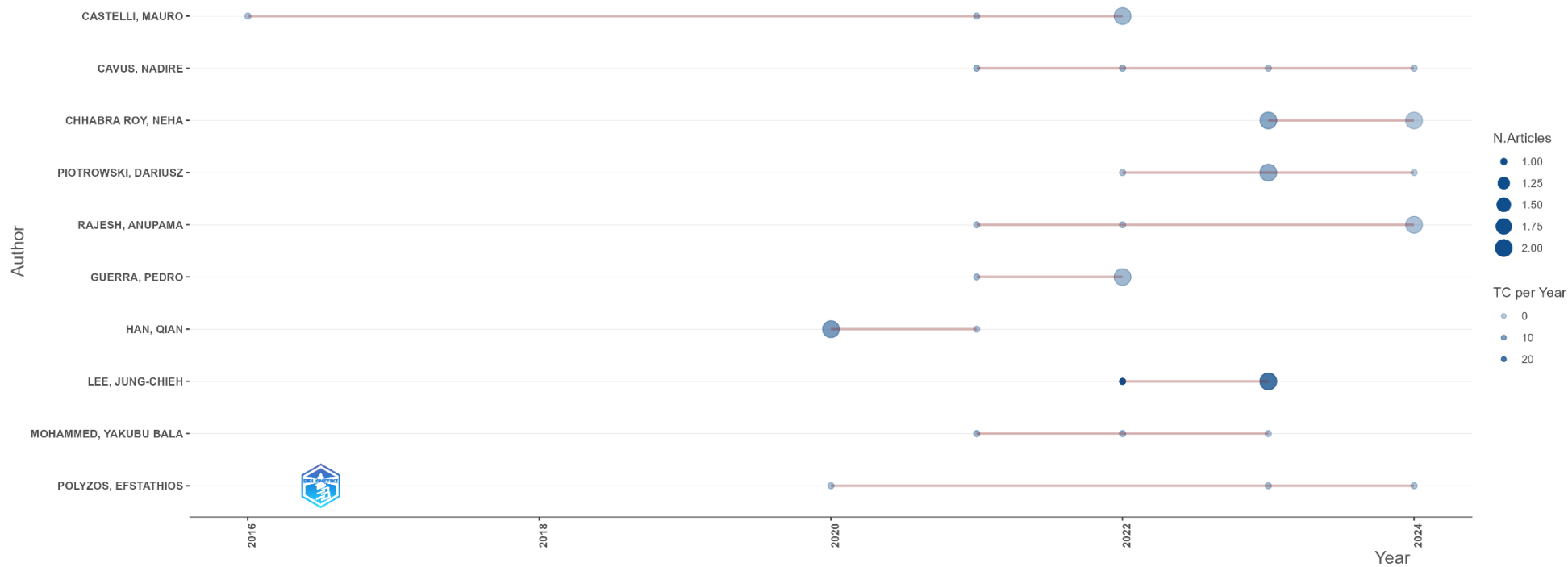
Most Relevant Authors

- Leading contributors include Castelli Mauro, Cavus Nadire, and Rajesh Anupama.
- Work spans fraud detection, risk management, and customer personalization.
- Demonstrates strong interdisciplinary collaboration.



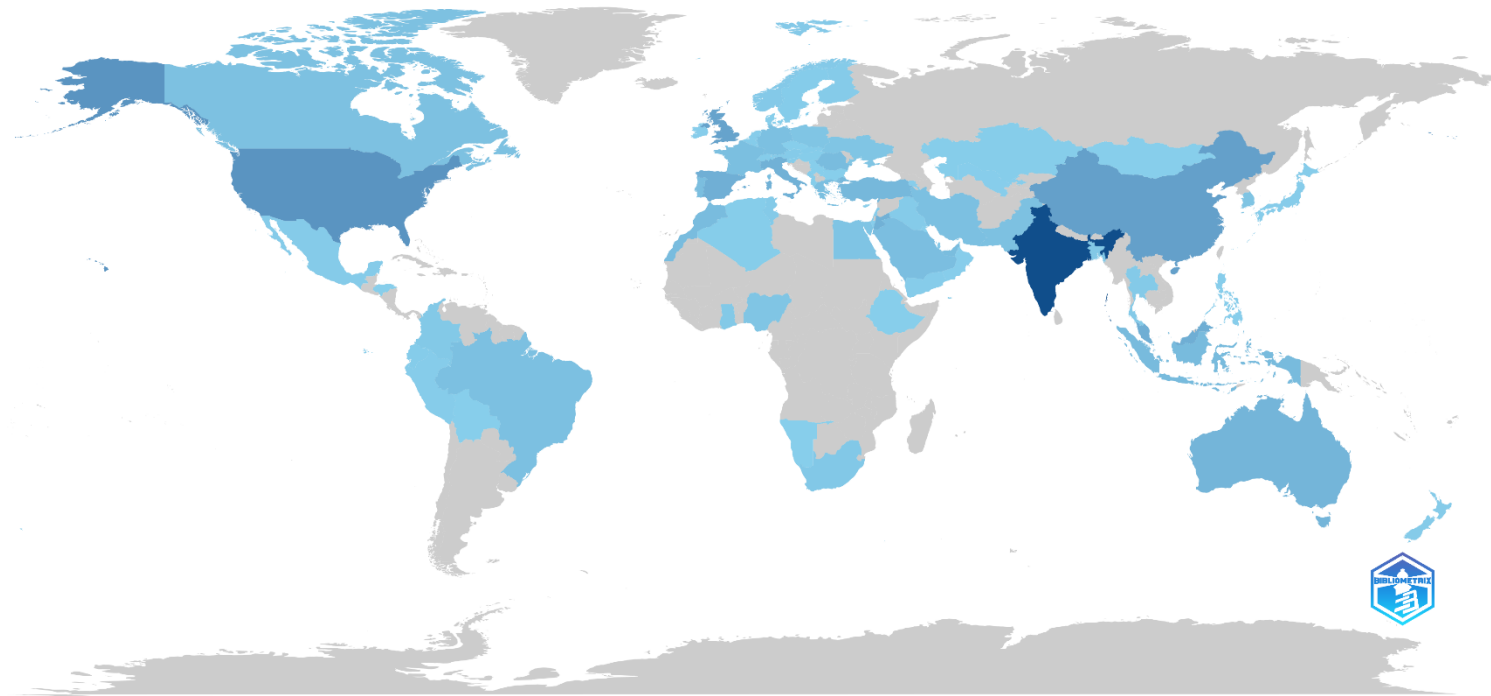
Authors' Production Over Time

- Identifies sustained contributions by key researchers over the years.
- Increased author activity post-2017, aligning with technological advancements.
- Highlights ongoing engagement in AI/ML-driven banking research.



Country Scientific Production

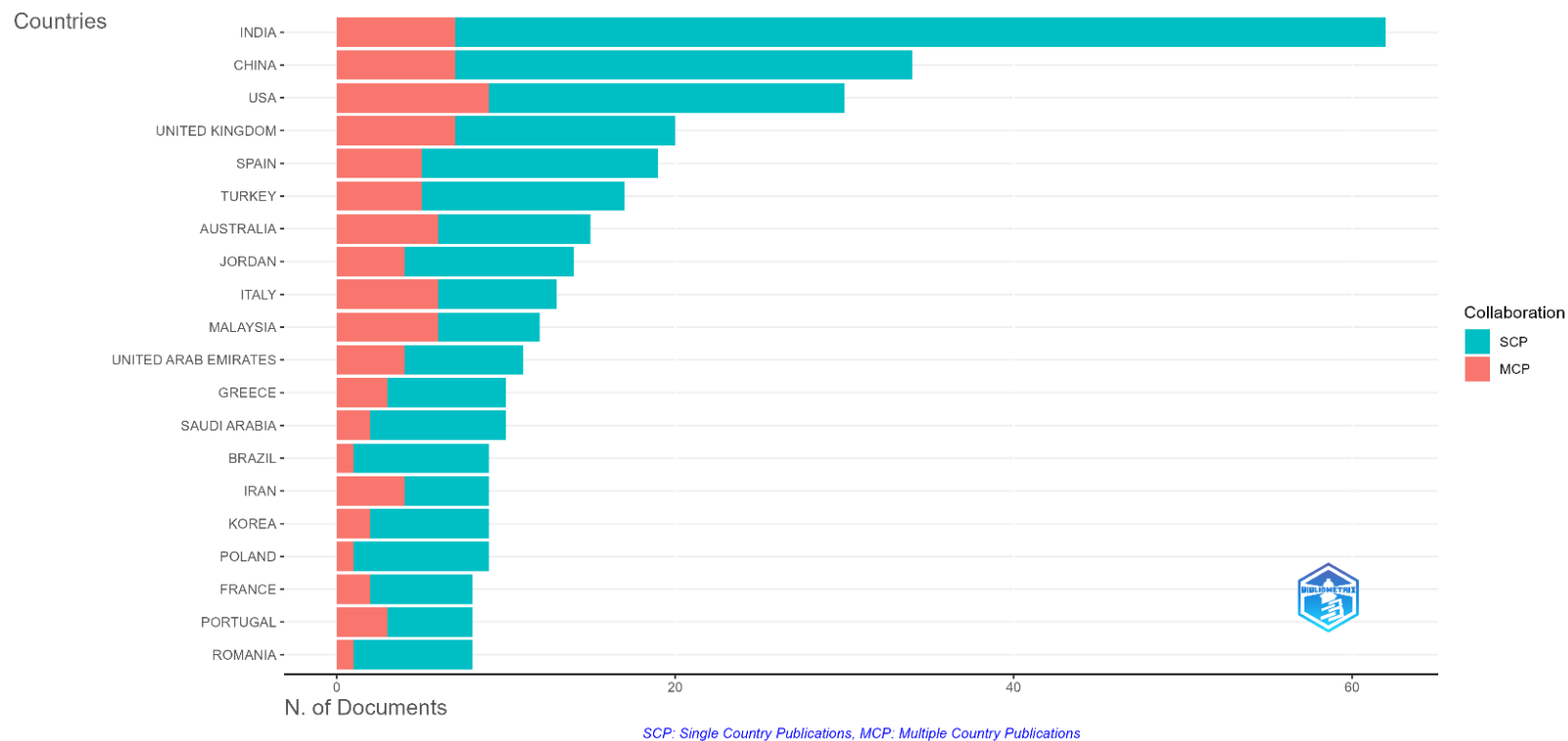
- Visual representation of country-wise research output.
- Highlights the global interest in AI/ML applications in banking.
- Emphasizes contributions from both developed and emerging economies.



Corresponding Author's Countries

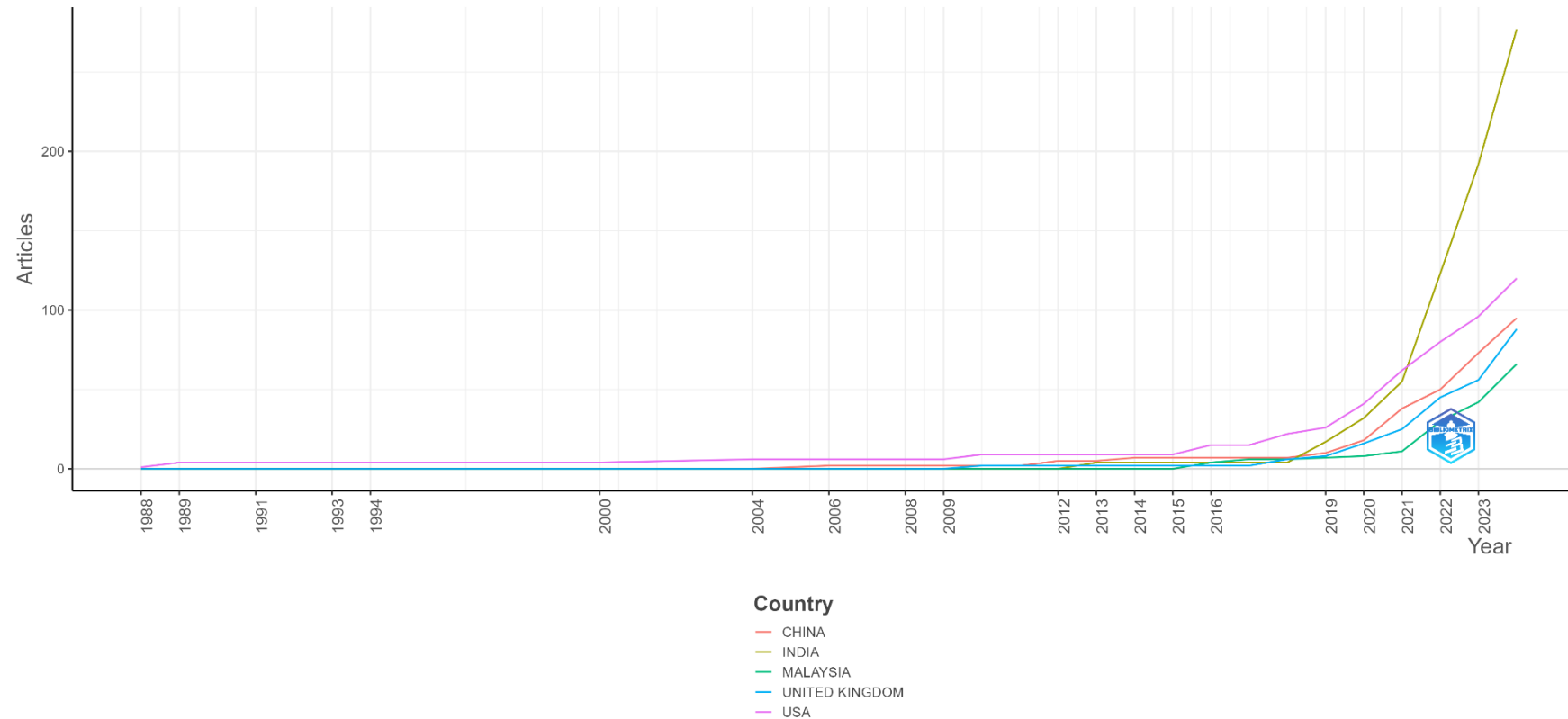
•Top contributing countries:

- India, China, USA, and UK dominate AI/ML research in banking.
- India and China show strong domestic research efforts.
- USA and UK exhibit balanced collaboration with multiple-country publications.



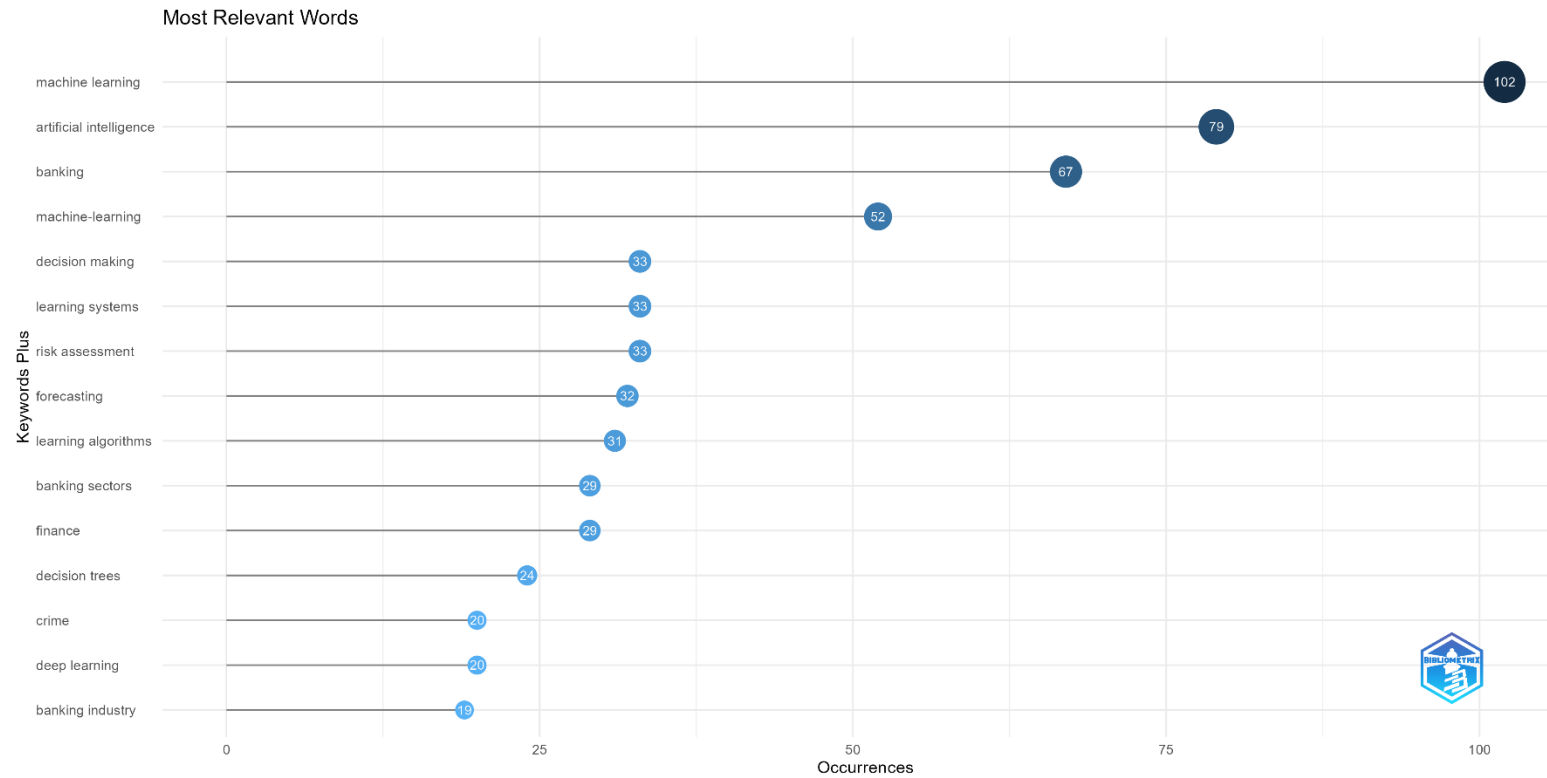
Country Production Over Time

- Notable research growth from China, India, Malaysia, UK, and the USA post-2016.
- Correlates with the increasing integration of AI/ML in banking operations.
- Demonstrates research expansion in response to technological advancements.



Most Relevant Keywords

- Top keywords in AI/ML banking research:
 - “machine learning” (102 occurrences), “artificial intelligence” (79 occurrences), banking (67 occurrences)



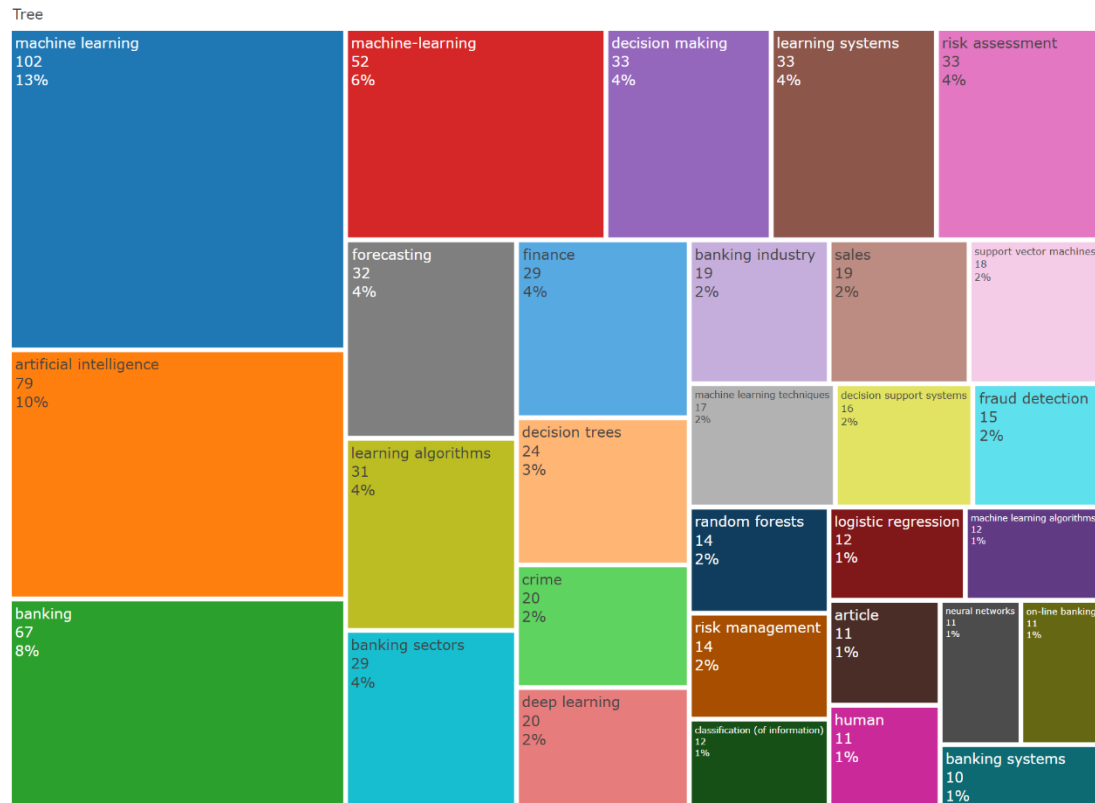
Word Cloud

- Visual representation of frequently used research terms.
- Larger font sizes represent dominant research topics.
- Reinforces the importance of AI/ML in risk management and financial decision-making.



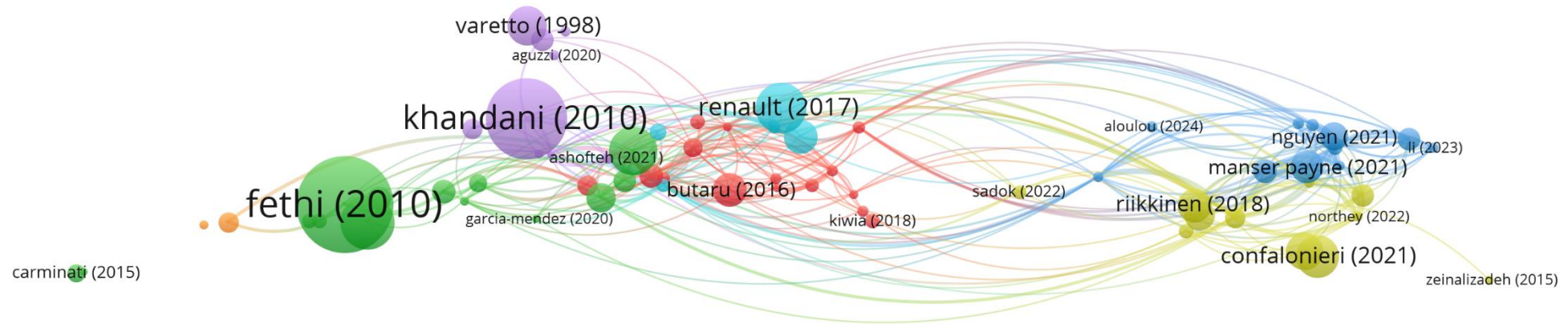
Word Treemap

- Comparative visualization of key research themes.
- Highlights major topics like fraud detection, machine learning models, and risk assessment.
- Identifies secondary research areas such as cybersecurity and support vector machines.



Bibliographic Coupling Analysis

- Illustrates connections between highly cited publications.
- Foundational works such as Khandani (2010) and Fethi (2010) form key clusters.
- Highlights major themes like credit risk modeling and operational efficiency.

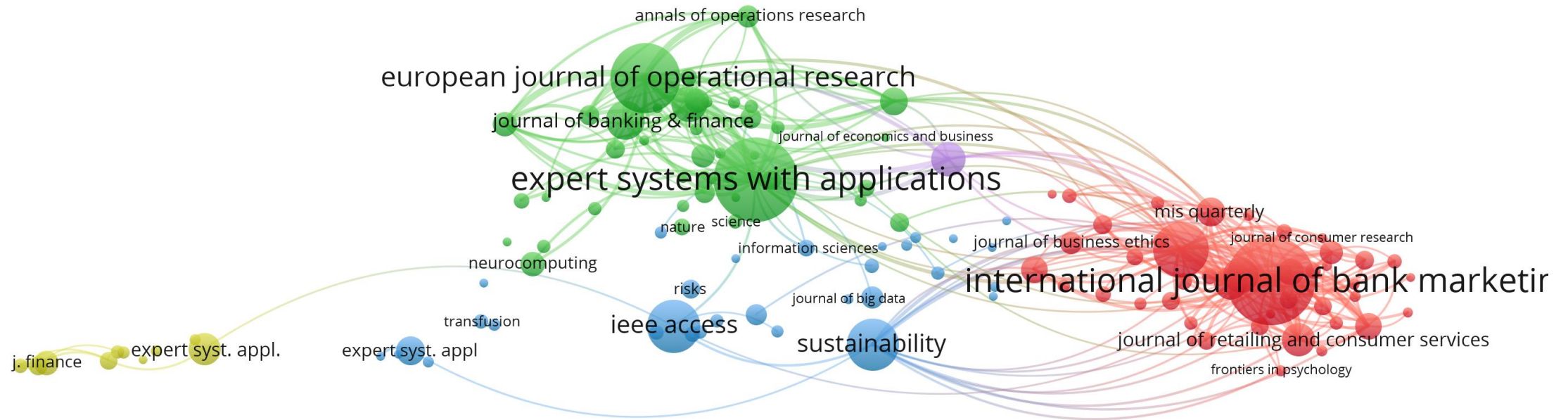


Co-Citation Analysis of Journals

- **Most influential journals in AI/ML banking research:**

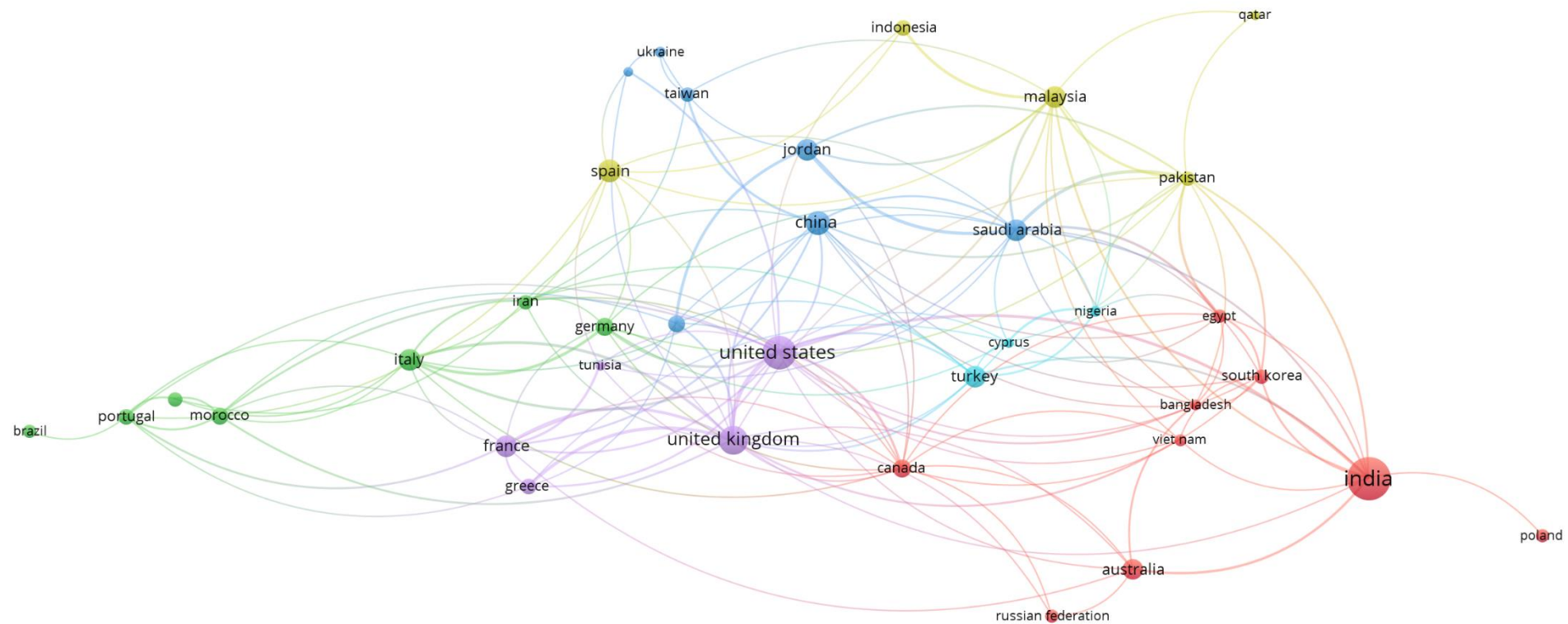
- International Journal of Bank Marketing, Expert Systems with Applications, European Journal of Operational Research.

- Strong connections to IEEE Access indicate growing interest in AI ethics and sustainability.



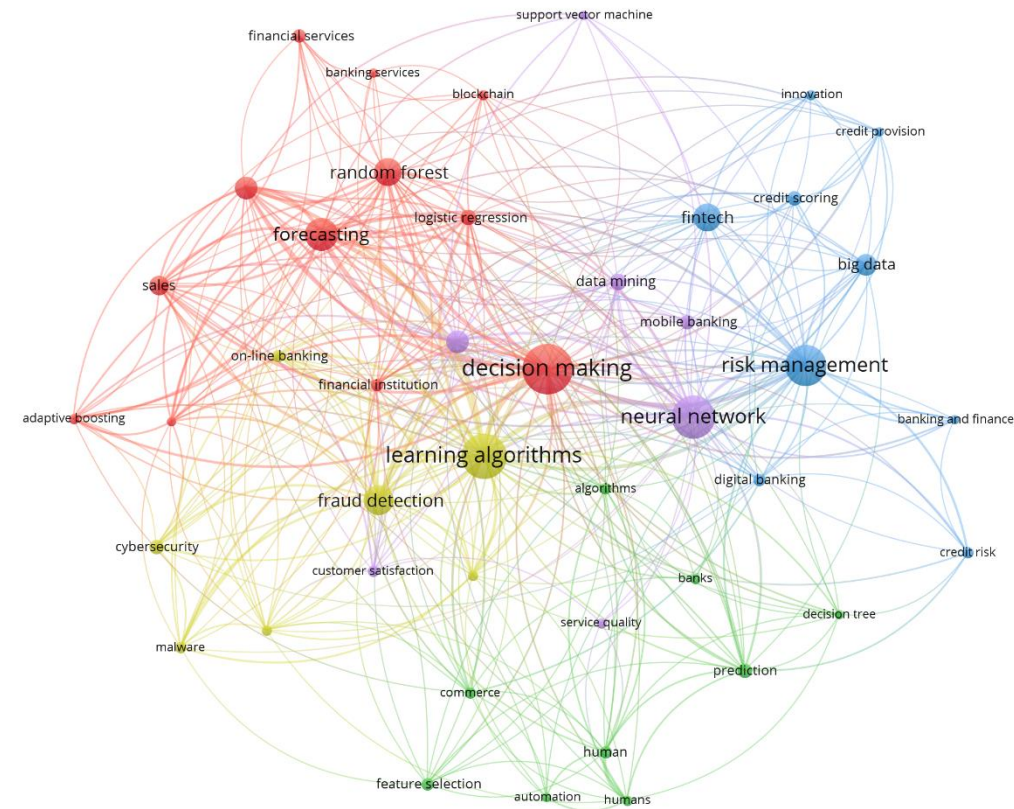
Co-Authorship Network by Country

- India, USA, and UK are central hubs in research collaboration.
- Thick links indicate strong co-authorship relationships, such as India-USA partnerships.
- Reflects the global nature of AI/ML-driven banking research.



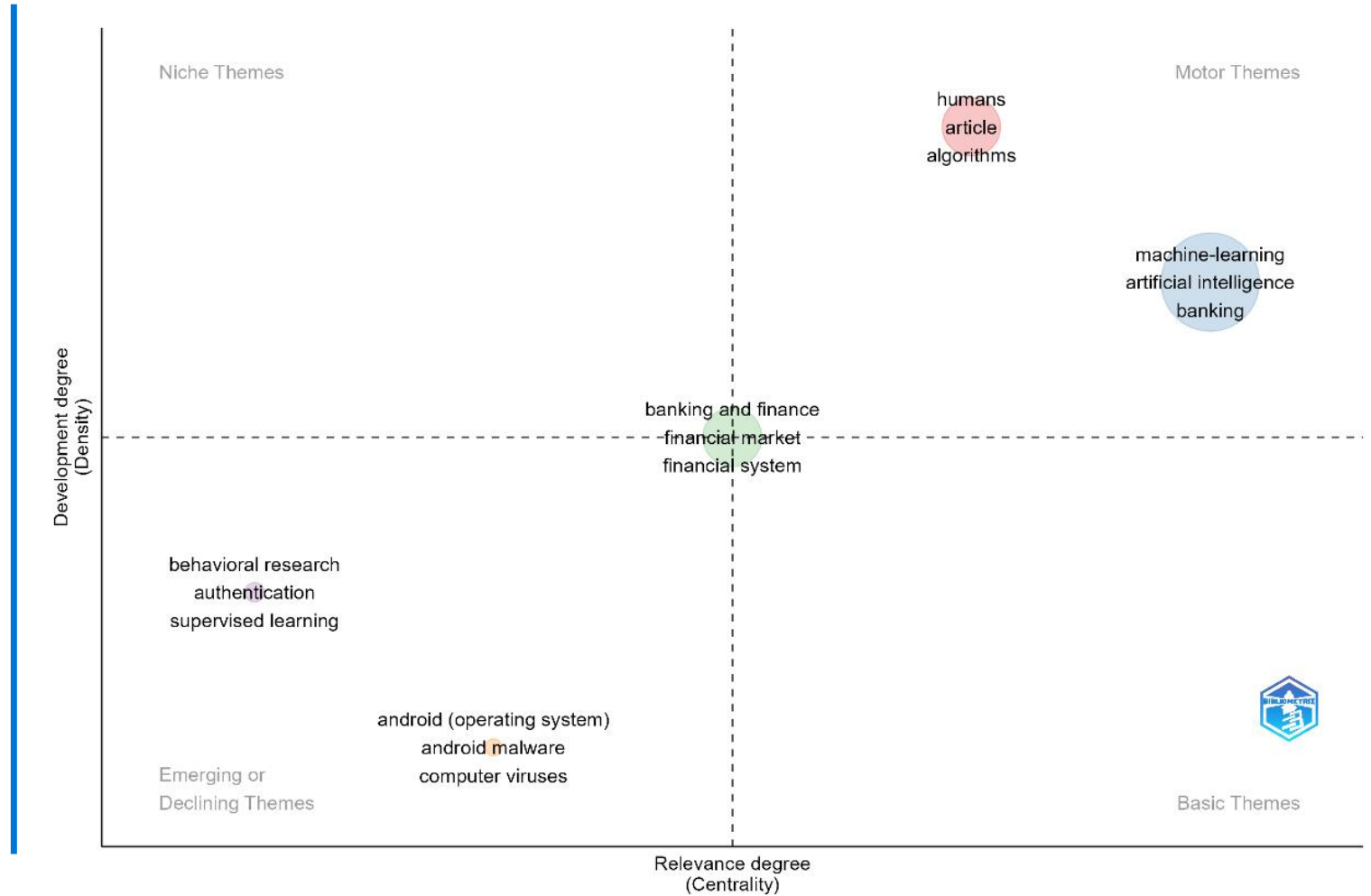
Co-Occurrence Network of Keywords

- Shows keyword relationships in AI/ML banking research.
- Central themes include "decision-making," "risk management," and "fraud detection."
- Emerging topics highlight AI applications in cybersecurity and financial analytics.



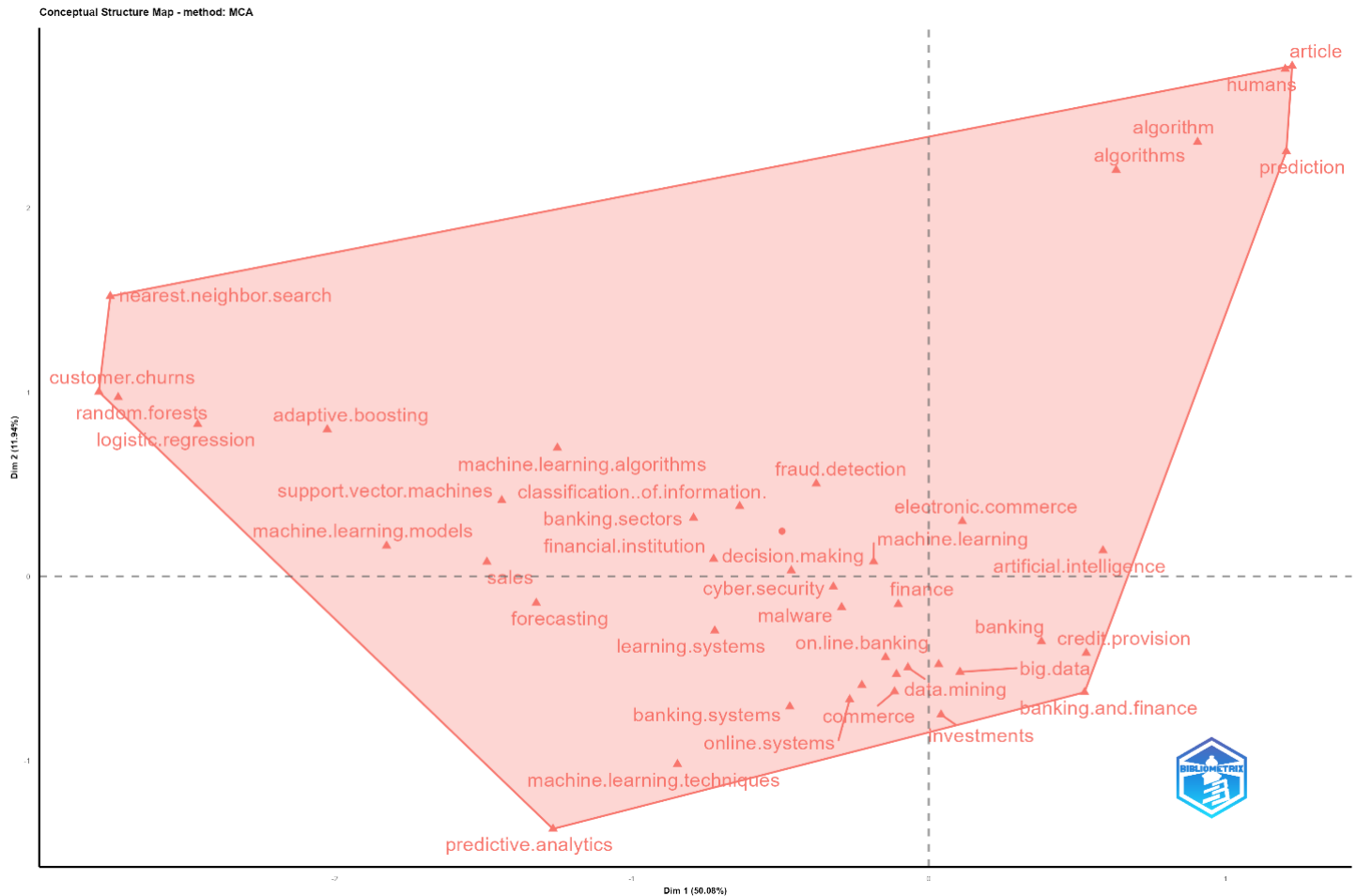
Thematic Map

- Categorizes research themes based on centrality and density.
- **Core themes:** Decision-making, learning systems, and risk assessment.
- **Emerging topics:** Biometrics, natural language processing, and cybersecurity.



Conceptual Structure Map

- Visualizes AI/ML themes in banking using Multiple Correspondence Analysis (MCA).
- Clusters include predictive analytics, decision-making, and financial cybersecurity.
- Identifies key research areas such as big data applications and online banking.



Summary of Trending Topics in AI/ML in Banking

- **Decision Making:** AI/ML is being used to improve decision-making in various banking functions, including credit scoring, fraud detection, and customer service.
- **Learning Systems:** The use of machine learning algorithms, such as neural networks and random forests, is increasing.
- **Risk Assessment:** AI/ML is being used to assess and mitigate various risks, including credit risk, liquidity risk, and operational risk.
- **Fraud Detection:** AI/ML is being used to detect fraud in real-time, particularly in areas like credit card transactions and money laundering.
- **Customer Churn Prediction and Retention:** AI/ML is being used to predict which customers are likely to leave and to develop targeted retention strategies.
- **Technology Adoption:** The adoption of AI/ML technologies in banking is being driven by factors such as trust, perceived usefulness, and ease of use.

Summary of Trending Topics in AI/ML in Banking

- **Bank Failure Prediction:** AI/ML is being used to predict bank failures by analyzing macroeconomic indicators and firm-specific attributes.
- **Customer Service Automation and Personalization:** AI/ML is being used to automate customer service and provide personalized financial advice.
- **Credit Scoring and Risk Assessment:** AI/ML is being used to improve the accuracy of credit scoring and risk assessment.
- **Financial Crime Detection and Cybersecurity:** AI/ML is being used to detect financial crimes and improve cybersecurity.
- **Bank Performance Management:** AI/ML is being used to improve bank performance by optimizing operations and decision-making.
- **Portfolio and Wealth Management:** AI/ML is being used to provide personalized investment advice and manage customer portfolios.

Summary of Trending Topics in AI/ML in Banking

- **Liquidity Risk Management:** AI/ML is being used to assess and manage liquidity risk.
- **Predictive Analytics and Market Forecasting:** AI/ML is being used to predict customer behavior and market trends.
- **Operational Risk Management:** AI/ML is being used to identify and predict operational risks.
- **Systemic Risk Management:** AI/ML is being used to assess and manage systemic risk in the banking sector.
- **Operational Efficiency and Process Automation:** AI/ML is being used to automate tasks and improve operational efficiency.
- **Regulatory Compliance and Risk Management:** AI/ML is being used to ensure regulatory compliance and manage risks.

Summary of Trending Topics in AI/ML in Banking

| Topic | Methods Used | Dependent Variable | Research Gaps |
|---------------------|---|---|---|
| Credit Risk | Support Vector Machines, Neural Networks, Decision Trees, Hybrid models | Creditworthiness, Loan default risk | Data imbalance, Model bias, Explainable AI |
| Customer Churn | Neural Networks, Statistical models, Rule-based algorithms | Customer churn, Customer retention | Data heterogeneity, Model scalability, Explainable AI |
| Technology Adoption | Quantitative surveys, Hybrid models, Machine learning | Technology acceptance, User adoption | Generational differences, Societal impacts, Ethical AI |
| Bank Failure | Statistical models, Machine learning, Early warning systems | Bank failure, Financial distress | Data imbalance, Model accuracy, Real-time prediction |
| Fraud Detection | Imbalanced classification, Machine learning, Deep learning | Fraud detection, Anti-money laundering | Dynamic fraud patterns, Ethical concerns, Real-time analytics |
| Customer Service | AI, Case-based reasoning, Machine learning | Customer satisfaction, Service efficiency | Ethical AI, Legacy systems, Human-centric service |
| Credit Scoring | Machine learning, Explainable AI | Credit score, Default prediction | Dataset imbalance, Model interpretability, Fairness |
| Cybersecurity | Behavioral analysis, Machine learning | Threat detection, Fraud prevention | Insider threats, Ethical practices, Sustainability |
| Bank Performance | Data analytics, Machine learning | Bank efficiency, Profitability | Data quality, Analytics talent, Regulatory conditions |
| Liquidity Risk | ANNs, BNs, Ensemble models | Liquidity risk, Early warning systems | Cultural considerations, Dynamic risk, Model interpretability |
| Market Forecasting | Hybrid models, Machine learning | Transaction forecasting, Cash demand | Model accuracy, Data heterogeneity, Feature extraction |
| Operational Risk | AI/ML, Bayesian networks | Operational risk, Risk events | Model validation, Bias in training data, Explainable AI |
| Systemic Risk | Theoretical frameworks, Machine learning | Systemic risk, Contagion effects | Dynamic and unpredictable events, Procyclicality, ESG integration |

Summary of Trending Topics in AI/ML in Banking

| Topic | No. of Articles | Keywords |
|---|-----------------|---|
| Bank Performance Management | 18 | Bank efficiency, Profitability, Data quality, Analytics talent, Regulatory conditions |
| Customer Churn Prediction and Retention | 10 | Customer churn, Customer retention, Data heterogeneity, Model scalability, Explainable AI |
| Customer Service Automation and Personalization | 19 | Customer satisfaction, Service efficiency, Ethical AI, Legacy systems, Human-centric service |
| Technology Adoption | 22 | Technology acceptance, User adoption, Generational differences, Societal impacts, Ethical AI |
| Bank Failure Prediction | 12 | Bank failure, Financial distress, Data imbalance, Model accuracy, Real-time prediction |
| Fraud Detection and Anti-Money Laundering (AML) | 13 | Fraud detection, Anti-money laundering, Dynamic fraud patterns, Ethical concerns, Real-time analytics |
| Credit Scoring and Risk Assessment | 9 | Credit score, Default prediction, Dataset imbalance, Model interpretability, Fairness |
| Financial Crime Detection and Cybersecurity | 9 | Threat detection, Fraud prevention, Insider threats, Ethical practices, Sustainability |
| Portfolio and Wealth Management | 6 | Personalized investment advice, Customer portfolios |
| Liquidity Risk Management | 4 | Liquidity risk, Early warning systems, Cultural considerations, Dynamic risk, Model interpretability |
| Predictive Analytics and Market Forecasting | 6 | Transaction forecasting, Cash demand, Model accuracy, Data heterogeneity, Feature extraction |
| Operational Risk Management | 4 | Operational risk, Risk events, Model validation, Bias in training data, Explainable AI |
| Systemic Risk Management | 4 | Systemic risk, Contagion effects, Dynamic and unpredictable events, Procyclicality, ESG integration |
| Operational Efficiency and Process Automation | 4 | Automated tasks, Operational efficiency |
| Regulatory Compliance and Risk Management | 4 | Regulatory compliance, Risk management, Explainable AI |
| Cross-selling | 5 | Cross-selling, Customer relationship management (CRM), Marketing |
| Technology Critique | 16 | Technology adoption, Risk management, Regulatory compliance, Operational efficiency |

Conclusion: Responses to Research Questions

- How have AI/ML been applied in the banking sector to address key challenges?
 - AI/ML significantly improves fraud detection, credit risk management, and customer service.
 - Machine learning models such as anomaly detection outperform traditional fraud prevention techniques, enabling real-time fraud prevention.
 - AI-driven predictive analytics enhance customer churn management by offering retention strategies based on behavior patterns.
 - Chatbots and virtual assistants improve customer engagement, reducing response time and operational costs.

Conclusion: Responses to Research Questions

- What are the emerging trends in AI/ML research in banking?
 - Increasing adoption of cybersecurity-focused AI to protect sensitive financial data.
 - The rise of explainable AI (XAI) to improve transparency and trust in AI-driven financial decisions.
 - AI-based regulatory compliance automation helps banks reduce penalties and meet legal requirements efficiently.
 - Predictive analytics are advancing, allowing financial institutions to anticipate customer behavior and mitigate market risks.

Conclusion: Responses to Research Questions

- What gaps exist in the current literature on AI/ML applications in banking?
 - Limited use of **unsupervised learning** in banking, despite its potential for fraud detection and risk analysis.
 - Ethical concerns, such as biases in AI-driven credit scoring and loan approvals, require further exploration.
 - Existing AI models lack real-time adaptability, highlighting the need for **real-time learning systems**.
 - **Explainability and interpretability** of AI models remain a challenge, necessitating further research into transparent AI decision-making.

Conclusion: Responses to Research Questions

- How can bibliometric and thematic analyses contribute to identifying future research directions in AI/ML for banking?
 - Bibliometric analysis reveals key research trends, influential authors, and collaboration patterns.
 - Thematic analysis highlights critical research areas such as **fraud detection**, **predictive analytics**, and **customer relationship management (CRM)**.
 - The study identifies gaps in **cybersecurity**, **real-time AI applications**, and **regulatory compliance**, shaping future research priorities.
 - Tools like **Biblioshiny** and **VOSviewer** provide structured insights into AI/ML applications, helping refine research strategies.