

Algorithms & Models used in Banks

The banking sector utilizes a range of algorithms and models to enhance decision-making, improve customer experiences and sell more products to customers. Most of us are not aware of these algorithms but behind the scene there's a lot going on...

Commonly used algorithms and classification models include:

Core Operations Algorithms

- Credit Scoring Models: Predict the probability that a borrower will default on a loan. These models include Logistic Regression, Decision Trees, Random Forest, Neural Networks etc.
- Transaction Classification Models: These models automatically categorize transactions into predefined groups such as groceries, utilities, dining, entertainment, travel, etc. This helps both the bank and the customers in understanding where the money is being spent.
 - Transaction classification models can detect unusual patterns in transaction data and raise alert for transactions that might suggest <u>money laundering activities</u>. They also categorize transactions into different types for reporting and analysis. These models include Naive Bayes, Support Vector Machines (SVM), K-Nearest Neighbors (K-NN) etc.
- Fraud Detection: Machine learning techniques like clustering, neural networks, and anomaly
 detection methods can help identify unusual patterns in transaction data that might indicate
 fraudulent activity. Logistic Regression Decision Trees Neural Networks, Anomaly Detection
 Algorithms are some of the well-known algorithms.
- Asset and Liability Management (ALM) Models: These models help in managing and planning
 the allocation of assets and liabilities to meet certain requirements. They can use linear
 programming, simulation techniques, or stochastic optimization methods.
- Interest Rate Forecasting: Time series models like ARIMA or machine learning methods can be used to predict future interest rates.
- Cash Demand Forecasting: Predict the demand for cash in ATMs to optimize cash loading and reduce outages. These include Time Series Models (e.g., ARIMA, Prophet)

Sales & Client Analytics Algorithms

 Customer Segmentation: Clustering algorithms, especially k-means clustering, are used to segment bank customers into different groups based on their behavior and demographics.



- Churn Prediction: Predictive models, such as logistic regression, decision trees, and neural networks, can forecast which customers are likely to leave the bank's services in the near future.
- Cross-Selling and Up-Selling Models: Predict which products or services a customer is likely to purchase next. These include Association Rule Mining (e.g., Apriori, Eclat) and Collaborative Filtering.
- Recommendation Engines: Suggest products, services, or information to users based on their behavior, preferences, or profile. Collaborative Filtering, Content-Based Filtering, Matrix Factorization are algorithms power Recommendation engines
- Market Basket Analysis: Association Rule Mining (e.g., Apriori) help understand which products or services are frequently purchased together.
- **Customer Lifetime Value Models**: Estimate the total net profit a bank can obtain from a particular customer throughout their relationship.
- Customer Segmentation Models: Segment customers into different groups based on their behavior, preferences, or demographics. These include K-Means Clustering, Hierarchical Clustering, DBSCAN (Density-Based Spatial Clustering of Applications with Noise)

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