# Data Science In FinTech

Landing a data science role hinges on mastering the fundamentals - coding, statistics, and machine learning. These pillars represent about 80% of what it takes to secure a role in data science. For any aspiring data scientist, honing these skills should be the primary focus.

<u>However</u>, once these fundamentals are solid, candidates can shorten their job search by adding another crucial element: **business domain-specific projects**. Tailoring your portfolio to showcase expertise in industry-relevant problems not only highlights your practical application skills but also sets you apart in a competitive job market.

This guide will deep dive into **Data Science projects in the Finance Industry.** 

Finance has multiple sub-fields and we will dive into 6 of the main sub-fields that hire Data Scientists (DS) or Data Analysts (DA). For each sub-field we will briefly cover the **business model** and **metrics** you should be aware of (especially when communicating during your interviews), in addition to **providing project ideas** you can build out.

# **High Frequency Trading**

Utilizes ML to make real-time trading decisions, executing orders at high speeds based on market data.

#### **Business Model**

Companies in this space generate revenue through the spread (difference between buy and sell prices), arbitrage opportunities, and market-making activities. The emphasis is on high volume and low latency trades.

#### **Companies**

- Jane Street: A quantitative trading firm that heavily relies on ML for HFT.
- Citadel Securities: Known for using advanced algorithms in market-making and trading.

• **Two Sigma**: A firm that uses machine learning to predict market movements and execute trades.

#### **Key Metrics**

- Sharpe Ratio: Measures the performance of an investment relative to its risk.
- Execution Speed: The time it takes to execute a trade, crucial in HFT.
- Order Fill Rate: The percentage of trade orders that are successfully executed.

Some of the projects you can consider to stand out are -

## **Predictive Market Making**

- Objective: Develop a model to predict bid and ask prices in real-time, allowing for more efficient market-making.
- Dataset: Kaggle High-Frequency Trading Dataset
- ML Techniques: Time Series Analysis, Reinforcement Learning
- Metrics: Sharpe Ratio, Execution Speed

# **Volatility Prediction for Trading Strategies**

- Objective: Create a model to predict short-term market volatility, helping to optimize trading strategies.
- Dataset: <u>Historical Intraday Stock Market Data</u>
- ML Techniques: GARCH Models, LSTM Networks
- Metrics: Prediction Accuracy, Volatility Forecast Error

## **Arbitrage Opportunity Identification**

- **Objective**: Identify arbitrage opportunities across multiple exchanges or assets using real-time data.
- Dataset: <u>Crypto Arbitrage Dataset</u>
- ML Techniques: Regression Models, Anomaly Detection

Metrics: Profit per Trade, Execution Latency

# **Credit Scoring and Risk Management**

ML models are used to assess creditworthiness, predict defaults, and manage risk portfolios.

#### **Business Model**

Revenue is generated through lending (interest on loans), credit products, and risk assessment services.

#### **Companies**

- FICO: Provides credit scoring and analytics for risk management.
- **Zest AI**: Uses machine learning to offer AI-powered credit underwriting.
- **Kabbage**: A fintech company that uses AI to offer small business loans with rapid credit decisions.

#### **Key Metrics**

- **Default Rate**: The percentage of loans that default.
- Credit Score Accuracy: How accurately the model predicts creditworthiness.
- Loss Given Default (LGD): The amount of loss a lender incurs when a borrower defaults.

Some of the projects you can consider to stand out are -

#### **Credit Risk Prediction**

- Objective: Build a credit scoring model to predict the probability of default for loan applicants.
- Dataset: LendingClub Loan Data
- ML Techniques: Logistic Regression, Boosted Trees, SHAP/PDP/ICE for explanation
- Metrics: AUC-ROC, Accuracy, F1 Score

# **Dynamic Credit Limit Adjustment**

- **Objective**: Develop a model to dynamically adjust credit limits based on user behavior, or predict risk profile, or predict default risk of customers.
- Dataset: Credit Card Data
- ML Techniques: Logistic Regression, Boosted Trees, SHAP/PDP/ICE for explanation
- Metrics: Default Rate, Customer Retention Rate

## **Loan Portfolio Risk Optimization**

- Objective: Optimize the risk-return profile of a loan portfolio using machine learning.
- Dataset: Home Credit Default Risk Dataset
- ML Techniques: Portfolio Optimization, Monte Carlo Simulations
- Metrics: Sharpe Ratio, Expected Shortfall

# **Fraud Detection and Prevention**

ML models are deployed to detect and prevent fraudulent activities, especially in transactions.

#### **Business Model**

Companies in this space earn through transaction fees, subscriptions for fraud detection services, and penalties recovered from fraud attempts.

#### Companies

- **Stripe**: Uses machine learning to detect and prevent payment fraud.
- Feedzai: Provides fraud prevention solutions powered by Al.
- PayPal: Employs advanced ML models to secure transactions and prevent fraud.

#### **Key Metrics**

- False Positive Rate: The percentage of legitimate transactions incorrectly flagged as fraudulent.
- Fraud Detection Rate: The percentage of fraudulent activities accurately identified.
- Chargeback Rate: The proportion of transactions that are disputed by customers due to fraud.

Some of the projects you can consider to stand out are -

#### **Real-Time Transaction Fraud Detection**

- **Objective**: Build a real-time fraud detection model to flag suspicious transactions as they occur.
- Dataset: Credit Card Fraud Detection Dataset
- ML Techniques: Anomaly Detection, Autoencoders, Isolation Forest
- Metrics: Precision, Recall, False Positive Rate

# **User Behavior Profiling for Fraud Prevention**

- **Objective**: Develop user behavior profiles to detect unusual activities that might indicate fraud.
- Dataset: Synthetic Financial Datasets For Fraud Detection
- ML Techniques: Logistic Regression, Decision Trees, Gradient Boosting
- Metrics: AUC-ROC, F1 Score

# Portfolio Management and Robo-Advisors

ML is used to optimize investment portfolios and automate financial advice.

#### **Business Model**

Revenue is typically generated through management fees, advisory fees, and performance-based fees.

#### **Companies**

- Wealthfront: A robo-advisor that uses ML to manage portfolios.
- **Betterment**: Utilizes AI to provide automated, personalized investment advice.
- BlackRock: Uses ML for portfolio management and financial market analysis.

#### **Key Metrics**

- Portfolio Return: The overall return on investment for the managed portfolio.
- Risk-Adjusted Return: Measures portfolio performance relative to the risk taken.
- **Customer Retention Rate**: The percentage of clients that continue to use the service over time.

Some of the projects you can consider to stand out are -

#### Personalized Portfolio Recommendation

- **Objective**: Build a recommendation system that suggests personalized investment portfolios based on user preferences and risk tolerance.
- Dataset: Portfolio Management Dataset (See this notebook for an example)
- ML Techniques: Collaborative Filtering, Matrix Factorization
- Metrics: User Satisfaction Score, Portfolio Performance

# Portfolio Risk Prediction Using ML

- **Objective**: Develop a model to predict the risk associated with different portfolio compositions.
- Dataset: Create your own portfolio combination using <u>Historical Stock Prices</u>
- ML Techniques: Time Series Forecasting, Monte Carlo Simulations
- Metrics: Value at Risk (VaR), Expected Shortfall

# **Automated Rebalancing of Portfolios**

- **Objective**: Implement an automated system to rebalance investment portfolios based on market conditions.
- Dataset: Create your own portfolio combination using <u>Historical Stock Prices</u>
- ML Techniques: Reinforcement Learning, Optimization Algorithms
- Metrics: Portfolio Return, Transaction Costs

# **Financial Forecasting and Analytics**

Involves using ML to forecast financial metrics like stock prices, economic indicators, and company performance.

#### **Business Model**

Companies provide forecasting tools, financial analytics software, and consultancy services.

#### **Companies**

- **Bloomberg**: Uses AI for financial news and market data analytics.
- Refinitiv: Offers financial data and analytics tools powered by Al.
- **Kensho Technologies**: Provides predictive analytics for financial markets.

#### **Key Metrics**

- **Prediction Accuracy**: How closely the forecasts match actual outcomes.
- Mean Absolute Error (MAE): A common metric for evaluating forecast accuracy.
- **Economic Value Added (EVA)**: A measure of a company's financial performance based on residual wealth.

Some of the projects you can consider to stand out are -

#### **Stock Price Prediction**

 Objective: Build a model to predict stock prices using historical data and market indicators.

- Dataset: <u>S&P 500 Stock Data</u>
- ML Techniques: LSTM Networks, ARIMA, Gradient Boosting
- Metrics: Mean Absolute Error (MAE), R-Squared

## **Economic Indicator Forecasting**

- **Objective**: Predict key economic indicators such as GDP growth or unemployment rates using machine learning.
- Dataset: World Bank Economic Indicators
- ML Techniques: Time Series Forecasting, Bayesian Models
- Metrics: Forecast Accuracy, Root Mean Square Error (RMSE)

# **Earnings Call Sentiment Analysis**

- Objective: Analyze the sentiment of earnings calls to predict stock price movements.
- Dataset: Earnings Call Transcripts
- ML Techniques: NLP, Sentiment Analysis, Text Classification
- Metrics: Sentiment Prediction Accuracy, Impact on Stock Prices

# **Customer Personalization and Marketing Analytics**

ML models are used to personalize customer experiences, optimize marketing campaigns, and improve customer segmentation.

#### **Business Model**

Revenue is generated through enhanced customer retention, cross-selling, and personalized product recommendations.

#### **Companies**

• Capital One: Uses machine learning for customer segmentation and personalized offers.

- **JPMorgan Chase**: Employs AI for targeted marketing and customer relationship management.
- American Express: Uses ML to personalize customer interactions and optimize marketing strategies.

#### **Key Metrics**

- Customer Lifetime Value (CLTV): The total value a customer is expected to bring over their relationship with the company.
- **Conversion Rate**: The percentage of marketing efforts that lead to successful customer actions.
- Churn Rate: The rate at which customers stop doing business with the company.

Some of the projects you can consider to stand out are -

#### **Personalized Financial Product Recommendations**

- **Objective**: Develop a recommendation system that suggests personalized financial products based on user behavior.
- Dataset: Bank Marketing Dataset
- ML Techniques: Collaborative Filtering, Content-Based Filtering
- Metrics: Click-Through Rate (CTR), Conversion Rate

# **Churn Prediction in Banking**

- Objective: Build a model to predict customer churn and suggest retention strategies.
- Dataset: Churn Modeling Dataset
- ML Techniques: Classification Models (Random Forest, SVM), Survival Analysis
- Metrics: Churn Rate, Retention Rate

# **Customer Segmentation for Targeted Marketing**

- **Objective**: Segment customers based on transaction data for more targeted marketing campaigns.
- Dataset: Retail Banking Dataset
- ML Techniques: Clustering (K-means, Hierarchical Clustering)
- Metrics: Segmentation Accuracy, Marketing Campaign ROI

Hope this helps! Feel free to reach out to us for everything DS/ML -

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