

# UBS Quantitative Research & Trading - Advanced Chooser Option Pricing Model with Real-World Data & Machine Learning

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Department: Quantitative Research & Trading | Duration: 8 Weeks (2 Months)

## Project Overview

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At UBS, client-driven solutions and proprietary trading strategies often require navigating complex, uncertain market regimes. A **Chooser Option** grants the holder the right to decide, at a future *choice date*, whether the option will be a standard European call or put, with a common strike and final expiration. This instrument is a fundamental building block for:

- **Structured Products:** Creating flexible, cost-efficient portfolios for clients with strong directional views but uncertain about volatility or timing.
- **Risk Management:** Hedging portfolios against asymmetric, regime-dependent risks (e.g., post-earnings announcements, FOMC decisions).
- **Proprietary Strategies:** Serving as a testbed for advanced numerical methods and model calibration, which are core to our QR desk's function.

This project extends the core research of chooser option pricing paper, aiming to **enhance pricing accuracy, adapt to dynamic market conditions, and improve practical applicability** by integrating real-time data collection, feature engineering, and machine learning (ML) optimization. The project bridges academic modeling (Black-Scholes, BSM) with industry practice, empowering interns to build a data-driven, scalable pricing tool for chooser options.

## Project Objectives

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1. Reproduce and validate the original BSM-based chooser option pricing model using updated real-world data.
2. Collect multi-dimensional real-time/historical data to enrich pricing features (beyond the paper's basic parameters).
3. Develop ML models to optimize volatility prediction and option pricing (addressing BSM's "constant volatility" limitation).
4. Compare ML-enhanced models with traditional BSM to quantify performance improvements.
5. Deliver a deployable pricing tool and actionable insights for investors/traders.

# Project Timeline & Core Modules

Week	Core Module	Key Tasks	Deliverables
Week 1	Data Source Design & Initial Collection	<p>1. <b>Data Requirement Specification:</b> Define scope for financial, macroeconomic, and sentiment data</p> <p>2. <b>API Setup &amp; Testing:</b> Configure and test Yahoo Finance, Alpha Vantage, FRED APIs</p> <p>3. <b>Initial Data Pull:</b> Collect JPM stock prices (daily), Treasury rates, VIX for 2018-2024 period</p>	<p>1. Data specification document</p> <p>2. Working API connection scripts</p> <p>3. Initial raw dataset (2018-2024)</p>
Week 2	Data Preprocessing & Feature Engineering	<p>1. <b>Data Cleaning:</b> Handle missing values (interpolation), outliers (IQR), time alignment</p> <p>2. <b>Feature Engineering:</b> - Traditional: Rolling volatilities, daily returns, dividend growth - Advanced: VIX-JPM correlation, interest rate momentum, sentiment scores (0-1)</p> <p>3. <b>Pipeline Development:</b> Build automated preprocessing pipeline with GitHub Actions scheduling</p>	<p>1. Cleaned structured dataset (CSV/Parquet, ≥10 features)</p> <p>2. Automated preprocessing pipeline code</p> <p>3. Feature engineering documentation</p>
Week 3	Original BSM Model Replication	<p>1. <b>Model Implementation:</b> Code the BSM-based chooser option pricing model from the paper</p> <p>2. <b>Parameter Configuration:</b> Set strike=\$150, T2=1 year (matching paper parameters)</p> <p>3. <b>Initial Validation:</b> Compare simulated outputs with paper's reported results</p>	<p>1. Fully functional BSM model code (Jupyter Notebook)</p> <p>2. Model validation notebook with initial comparisons</p> <p>3. Parameter configuration file</p>
Week 4	Baseline Model Performance Evaluation	<p>1. <b>Error Metric Calculation:</b> Compute MAE, RMSE between model predictions and actual CME transaction prices</p> <p>2. <b>Limitation Analysis:</b> Identify failure modes in high-volatility periods, sentiment impact gaps</p> <p>3. <b>Benchmark Establishment:</b> Document BSM model performance baseline for comparison</p>	<p>1. Model validation report with error metrics</p> <p>2. Performance benchmark documentation</p> <p>3. Optimized, well-commented BSM code</p>
Week 5	Machine Learning Model Design & Implementation	<p>1. <b>Two-Approach Architecture:</b> - Approach 1: ML volatility prediction (LSTM/RF/XGBoost) + BSM pricing - Approach 2: End-to-end supervised pricing (Linear Regression/GBDT/NN)</p> <p>2. <b>Feature Preparation:</b> Create time-series split datasets (70%/15%/15%) to prevent look-ahead bias</p> <p>3. <b>Model Framework Development:</b> Build initial ML model pipelines</p>	<p>1. ML architecture design document</p> <p>2. Feature engineering optimization code</p> <p>3. Initial ML model frameworks (both approaches)</p>
Week 6	Model Training, Tuning & Comparison	<p>1. <b>Hyperparameter Optimization:</b> Grid/random search with cross-validation</p> <p>2. <b>Model Training &amp; Evaluation:</b> Train final models, evaluate on test set</p> <p>3. <b>Performance Comparison:</b> Compare ML models vs. BSM baseline across MAE, RMSE, R<sup>2</sup></p> <p>4. <b>Interpretability Analysis:</b> Use SHAP/LIME to explain feature importance and model decisions</p>	<p>1. Trained ML models (pickle files)</p> <p>2. Comparative analysis report with metrics</p> <p>3. Feature importance visualizations (SHAP plots)</p>
Week 7	Advanced Analysis & Tool Development	<p>1. <b>Extended Sensitivity Analysis:</b> - SHAP-based impact quantification of new features (sentiment, VIX) - Extreme scenario testing (50% volatility spike, 2% rate hike)</p> <p>2. <b>Tool Framework Development:</b> Begin building Streamlit/FastAPI application structure</p> <p>3. <b>Real-time Data Integration:</b></p>	<p>1. Comprehensive sensitivity analysis report</p> <p>2. Pricing tool prototype (basic UI/functionality)</p> <p>3. Real-time data integration</p>

		Implement auto-update functionality for market data	module
Week 8	Tool Finalization & Project Delivery	<p>1. <b>Tool Feature Completion:</b> - Implement dual pricing (BSM + best ML model) - Add error margin displays - Complete visualization dashboard (price trends, sensitivity charts, performance metrics)</p> <p>2. <b>Final Report Compilation:</b> Synthesize all findings into 10-15 page report</p> <p>3. <b>Demonstration Preparation:</b> Record tool demo video, create final presentation</p>	<p>1. Fully deployable pricing tool (GitHub repo with README)</p> <p>2. Final project report (PDF)</p> <p>3. Tool demo video (5-10 minutes)</p> <p>4. Final presentation deck</p>

## Key Project Milestones

- **Milestone 1 (End of Week 2):** Complete data pipeline; deliver structured dataset with  $\geq 10$  features
- **Milestone 2 (End of Week 4):** Validate BSM model; establish performance baseline
- **Milestone 3 (End of Week 6):** Complete ML model training; determine optimal pricing approach
- **Milestone 4 (End of Week 8):** Deliver final product suite (tool + report + demo)

## Required Tools

- Data: Alpha Vantage API (free tier), FRED, Yahoo Finance, CME Group Historical Data, Reuters News API (free alternative: NewsAPI).
- Code: Python (Pandas, Scikit-learn, TensorFlow/PyTorch, Streamlit), GitHub (version control + Actions for auto-data collection).

## Reference Papers

Exploration of JPMorgan Chooser Option Pricing (2021)

Rubinstein, M. (1991). *Chooser Options* (original chooser option pricing derivation).

Hull, J. C., & White, A. (1987). *The Pricing of Options on Assets with Stochastic Volatilities* (addressing BSM's constant volatility flaw).

Gu, S., Kelly, B., & Xiu, D. (2020). *Empirical Asset Pricing with Machine Learning* (ML applications in financial asset pricing).

Bakshi, G., Cao, C., & Chen, Z. (1997). *Empirical Performance of Alternative Option Pricing Models* (benchmarking non-constant volatility models).