

# Lily Liu

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## EDUCATION

### University of California, Los Angeles

2022

*PhD in Econometrics and Quantitative Economics (Primary Field: Financial Econometrics)*

GPA: 3.9/4.0

### Peking University

2017

*Master of Finance, National School of Development/China Center for Economic Research (Primary Field: Finance)*

## EXPERIENCE

### Goldman Sachs

2023 – present

*Quantitative Researcher*

- Assessed volatility and PnL of portfolios under different market shock scenarios by monitoring various hedging strategies
- Reviewed performance of portfolio optimization and pricing models; performed sensitivity testing; validated the soundness
- Enhanced asset allocation models by incorporating risk budget constraints on assets via portfolio volatility decomposition

### University of California, Los Angeles

2019 – 2022

*Financial Econometrician*

#### Asset Pricing Models with Machine Learning Algorithms

##### Trading Signals

- \* Identified various predictive signals for stock markets: price trends in the US and liquidity in Chinese long-only markets
- \* Developed optimal alpha strategies based on over 1k signal features, trained tree models, neural networks on 35k stocks
- \* Tracked out-of-sample Sharpe ratio ( $> 3$ ) with large-scale data (over 60 years) by decomposing variance-covariance matrices
- \* Constructed weighted portfolios based on principal component analysis; improved performance by utilizing cross-predictors

##### Model Tests

- \* Corrected bias for Black-Scholes model; reduced variance ( $\approx 25\%$ ) for deep-out-of-the-money options with Monte Carlo
- \* Constructed bias-corrected model selection tests of contingent-claims prices with high dimensional stock heterogeneity
- \* Derived the limiting null distribution of the test statistics, conducted power analysis, and proved bootstrap validity

#### Stock Price Prediction with Web Scraped Data and Natural Language Processing

- \* Predicted stock prices using features from web-scraped data, including job openings, sales, purchases, news, social media
- \* Built predictive models using LSTM and ARIMA by incorporating sentiment features from Twitter as financial lexicons
- \* Applied N-Gram and TFIDF methods to vectorize text; improved accuracy by 10% by incorporating sentiment indicators

## Publications

Hahn, J. and Liu, X., 2022. Jackknife bias reduction for simulated maximum likelihood estimator of discrete choice models. *Economics Letters*, p.110784.

Hahn, J., Ridder, G. and Liu, X., 2023. Estimation of average treatment effects (ATE) for massively imbalanced data. *Econometric Reviews* (forthcoming).

### RAND Corporation

2017 – 2018

*Financial Economist*

#### Estimation of Macroeconomic Policy Effects for Massively Imbalanced Data

- Examined maximum likelihood estimators when events occur with a probability  $< 10\%$  using simulated imbalanced data
- Designed Monte Carlo simulations to generate different distributions of propensity scores for mortgage prepayment
- Reduced bias in estimators of price elasticity in Logistic Regression by 5%, average effects by 2% in simulations
- Trained Neural Networks and Decision Trees to detect fraud transactions with high AUC scores ( $> 90\%$ )

• Performed upsampling minority, downsampling majority, and SMOTE for imbalanced classification datasets

#### Predicting Housing Prices with Supervised Learning Using Time Series Data

- Predicted housing prices using Regularized Regression algorithms, Random Forest, Gradient Boosted Decision Trees
- Engineered informative features with past prices, housing properties, locations, and demographics of the neighborhood
- Fine-tuned the model with grid search and cross validation, the best model has a root mean square error of less than \$50k

## TECHNICAL SKILLS

- Programming Languages: Python, SQL, R, Matlab
- Econometric/Statistical/Machine Learning Methods: Cross-Validation, Synthetic Control, Causal Inference, Latent Dirichlet Allocations, Principal Component Analysis, Regularization, Ensemble Methods (Boosting and Bagging), Linear and Logistic Regression, Naïve Bayes, Decision Tree, Random Forest, Gradient Boosting, K-Nearest Neighbors, K-Means Clustering