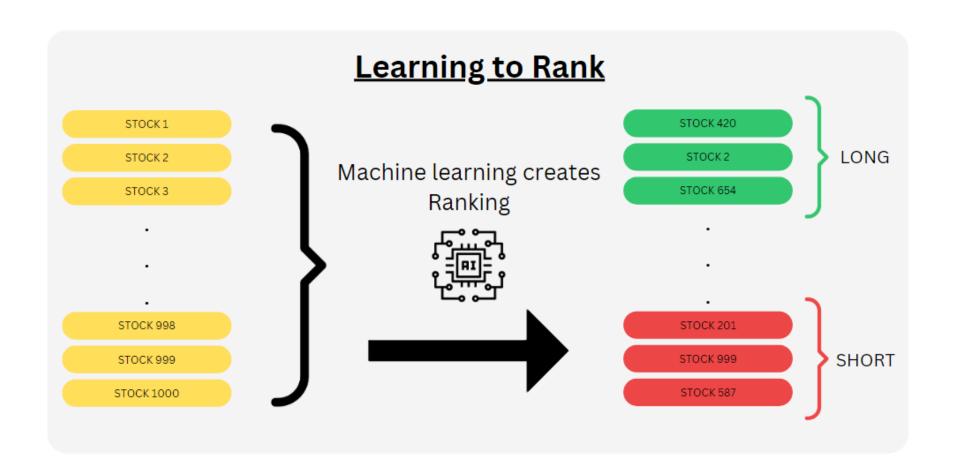
LEARNING TO RANK

A Machine Learning Approach to Rank Assets

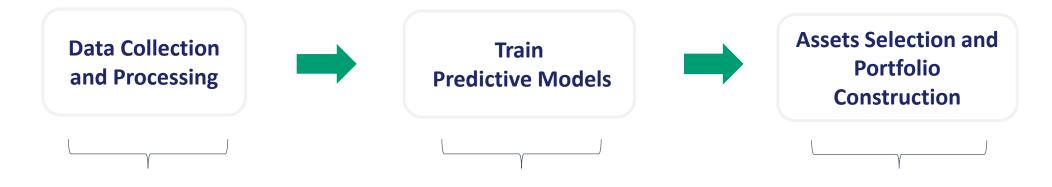
- What is Learning to Rank (LTR)?
- How to use Learning to Rank for trading strategies
- Results
- Next Steps



Learning to Rank in Finance



Trading Strategy Steps



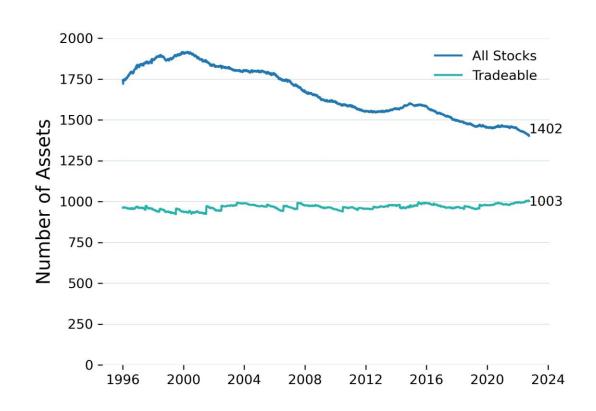
- Collect and clean/check data
- Create input features
- Data transformations
- Data exploration

 Train predictive models to get predictions

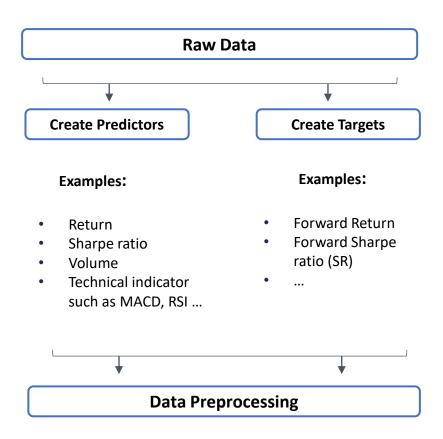
- Construct portfolios using the predictions
- Position sizing

Data Description

- Universe: constituents Russell 1000 from 01/01/1995 – 04/10/2022
- Daily pricing data for 3250 stocks:
 - Open, High, Low, Close, Volume, Market Cap
- Price and volume derived features over several periods
- Over 12.250.000 rows and over 100 columns
- **Tradeable:** stocks considered to be part of Russell 1000 at time *t*



Data Transformation and Preprocessing



- Cross-sectional ranking
- Handling outliers
- Handling missing values

- Features computed on several horizons
 - - e.g. Past Returns over 5, 10, 20, 60 days
- Missing values: cross sectional imputation with the mean
- Normalization: cross-sectional ranking
 - Automatically handles outliers

Backtesting Setup

Models Presented

Benchmark:

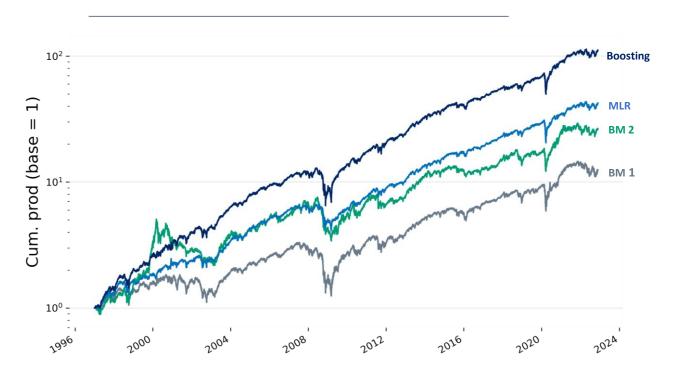
- equal weighted of all tradeable stocks in the universe (Russell 1000)
- One-Factor momentum strategy based on the return over 1, 3, 6, 12 months
- Learning to Rank Models (LTR)
 - Multiple Linear Regression (MLR)
 - Regression Trees (Reg Trees)
 - No Hyper parameter tuning

Configurations

- Universe: constituent Russell 1000 at time t
- **Prediction:** relative 10 day forward Sharp Ratio (SR)
- Expanding walk forward:
 - 1996 2022
 - Minimum train period: 1 year
 - Test period: 1 year
- Rebalancing Period and Allocation process:
 - Russell 1000: bi-monthly and equal weighted
 - One-Factor momentum: monthly and equal weighted
 - · LTR models: bi-monthly and rank-vol weighting
 - Max weight per stock 0.025

LTR Outperforms Benchmarks

Models Performances without Fees



Strategy	BM1	BM 2	MLR	Boosting
Return (%)	10.76	13.96	15.54	19.97
Volatility (%)	20.8	23.22	15.62	16.98
Sharpe Ratio	0.52	0.60	0.99	1.17
Max DD (%)	62.43	57.02	37.96	49.41
Max TUW (Days)	865	1400	404	345

Key Takeaways

Strong consistent performance after transaction costs

- Signal is consistent over time
- More return
- Lower risk in terms of volatility and drawdown
- Returns: 19,97% p.a. / Sharp Ration: 1,17

Multifunctional and Generalizable

- Applicable on other targets (returns, volatility...)
- Applicable to other universes and asset classes (European, Asian... market)
- Intuitive to incorporate new predictive features or add more assets