



On Transaction-Based Metrics as a Proxy for Profitability of Financial Asset Recommendation

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Financial Asset Recommendation

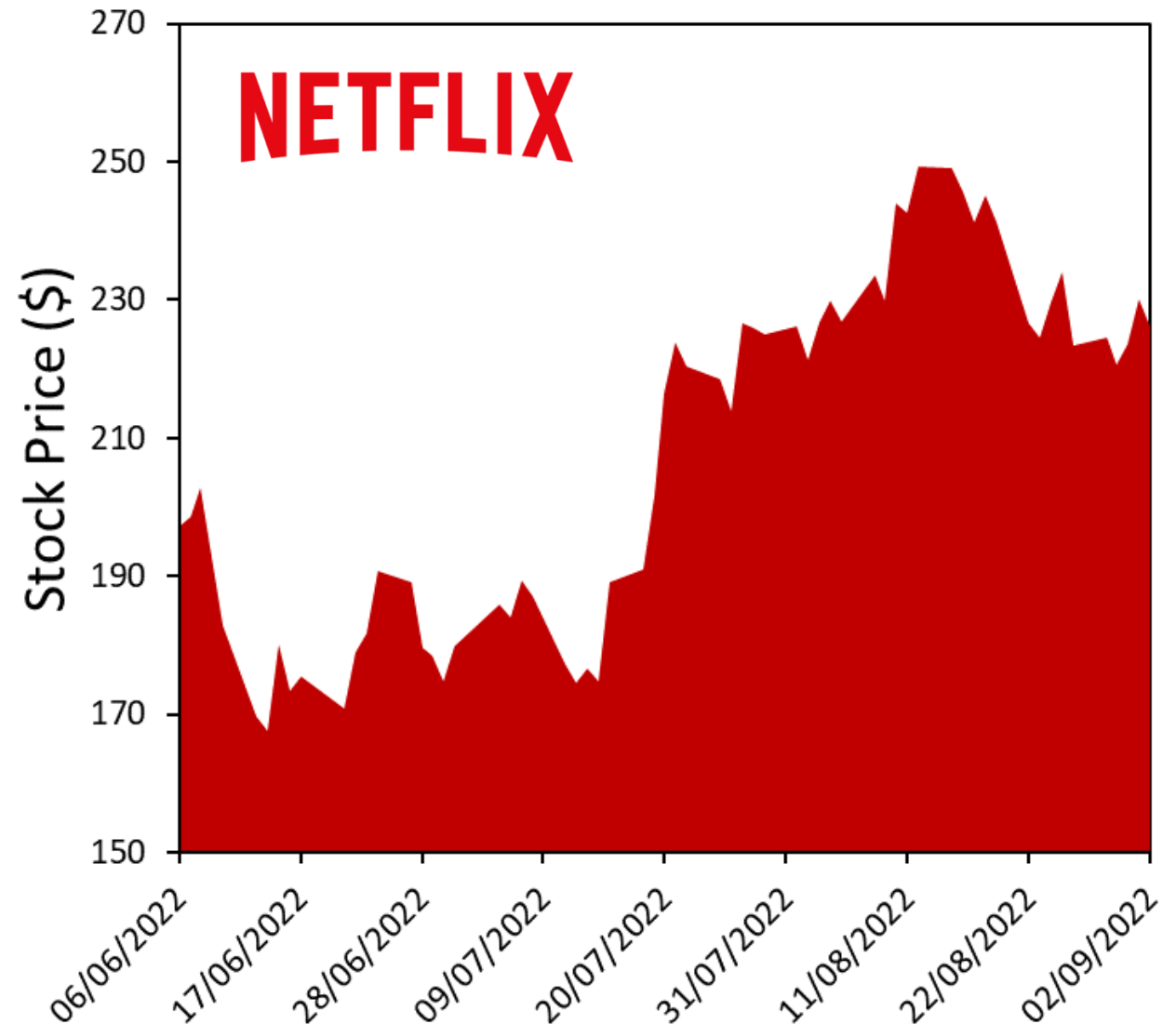
Customer's goal: Earn money

Financial assets

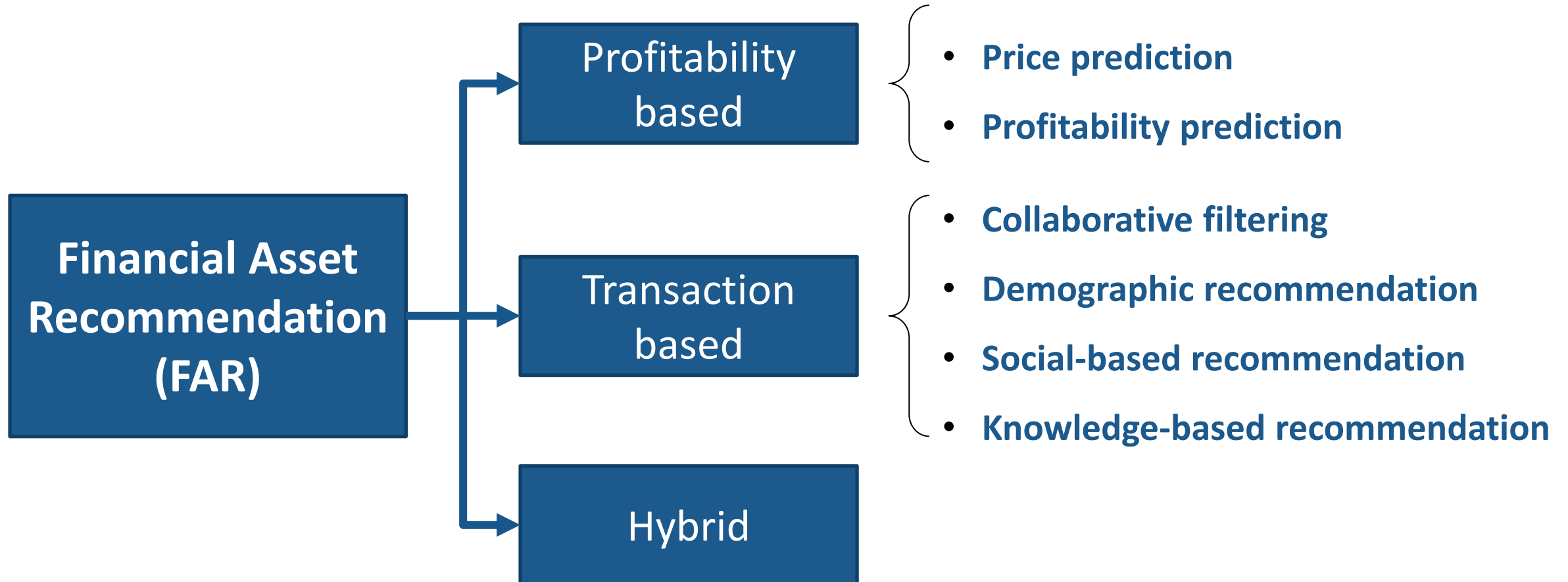
- Stocks
- Bonds
- Mutual funds

Price changes

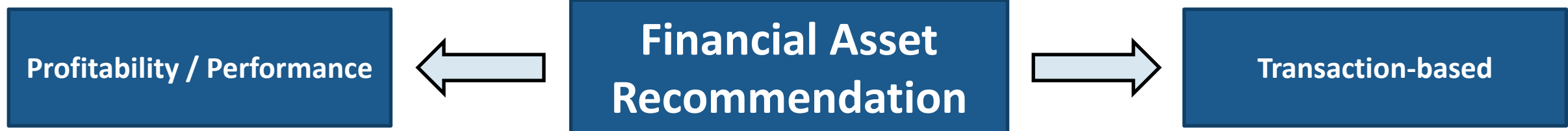
- External factors (market)
- Important for evaluation



Algorithms



Evaluation



Do our customers earn money?

- Aligned with customer interests.
- Ignores customers actual investments.

Can we predict future investments?

- Investment transactions indicate strong preference.
- Relevant transactions: acquisitions.
- Ignores temporal pricing information.

In both cases, metrics look at a fixed time interval

- **Metrics:** Key performance indicators
 - Return on Investment (ROI)
 - Net profit

- **Metrics:** IR / RecSys metrics
 - Precision
 - nDCG

On transaction-based metrics

If customers invest intelligently

- Expected high correlation between transaction-based and profitability metrics.
- In that case, transaction-based metrics should be superior.



They consider customer preferences

But, is that the case?

RQ1. Can we indistinctively use transaction-based and profitability-based metrics for evaluating financial asset recommendations?

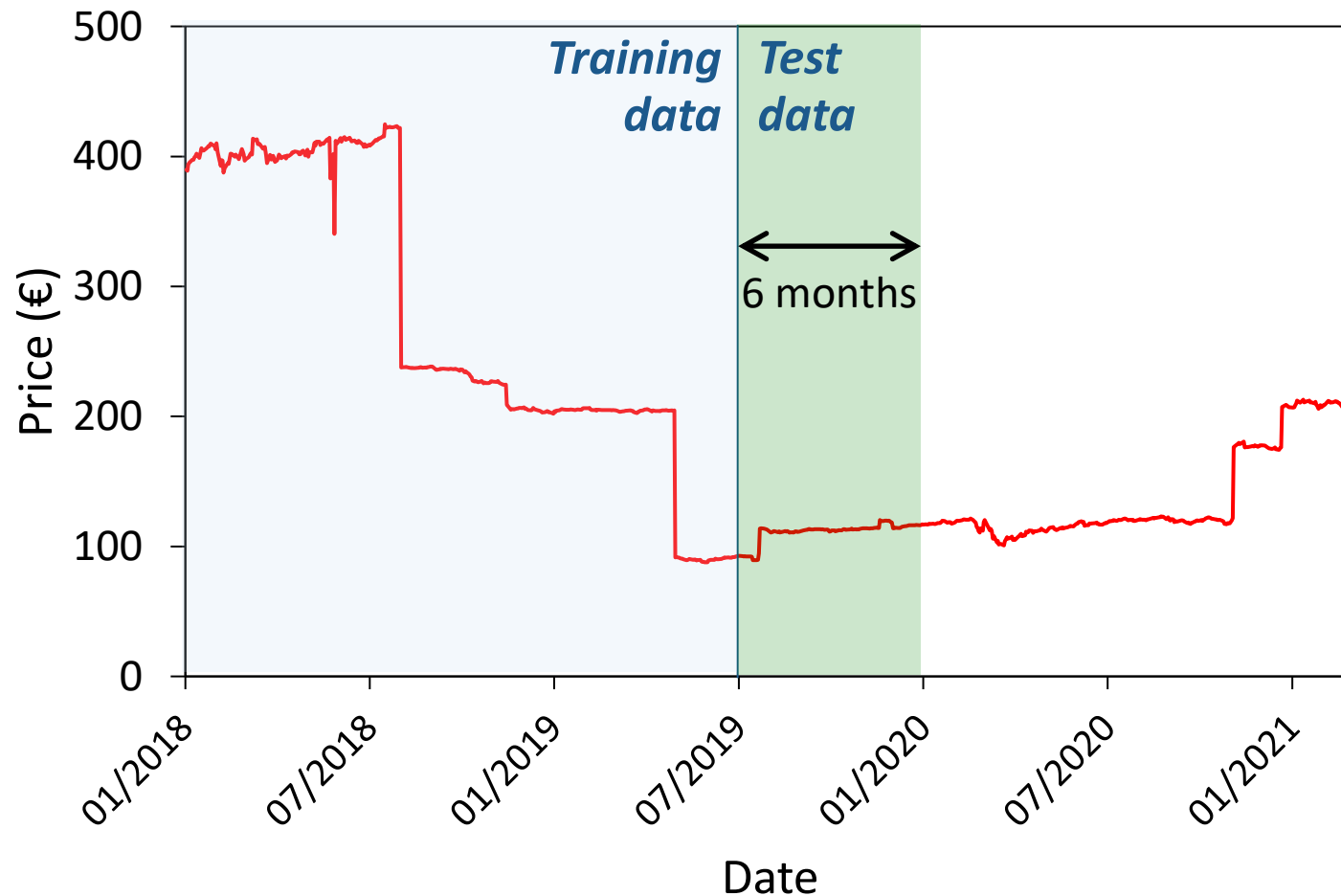
Dataset

- **Greek market:** stock, bonds, mutual funds and other banking products
- **Period:** 1st January 2018 – 21st March 2021
- **Combines**
 - Time series data (pricing information)
 - **Customer investments**
- **Time series data:**
 - 5,371 financial assets (2,025 assets with investments)
 - 1,768,128 data points
- **Customer investments**
 - 52,390 customers
 - 313,004 transactions



Experimental procedure

Avg. Pricing data

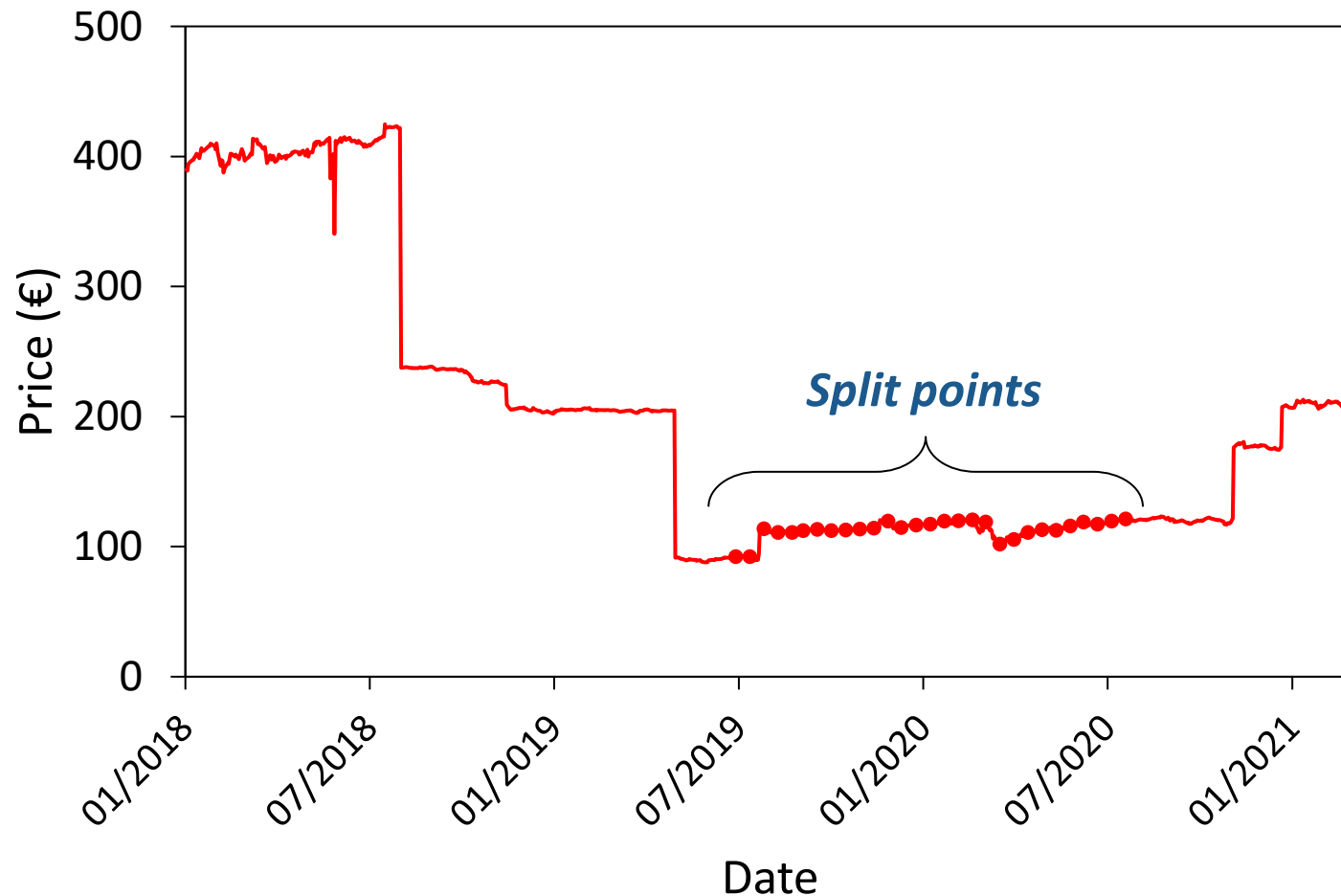


Procedure

1. Select recommendation time t
2. Split into training / test
 - **Training:** From 1st Jan 2018 to t
 - **Test:** From t to $t + 6$ months
3. Train models
4. Execute recommendations at t
5. Evaluate

Experimental procedure

Avg. Pricing data



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29 split points

- One every two weeks
- From: 1st July 2019
- To: 22nd July 2020

Metrics

Profitability-based: return on investment (ROI@10)

- Relative increase w.r.t. the initial investment after some time Δt
- **Initial price:** price at recommendation time
- **Final price:** price at recommendation time + Δt
- Δt : six months

Transaction-based: nDCG@10

- Higher nDCG indicates our model predicts future customer investments.
- Ranking-based IR/RecSys evaluation metric
- Relevant transactions:
 - Asset acquisitions (buys)
 - Up to 6 months after recommendation

Algorithms



University
of Glasgow

Profitability-based regression models

- Support vector regression (SVR)
- Random forest
- LightGBM

Transaction-based models

- **Not personalized:** popularity-based, random
- **Collaborative filtering:** LightGCN, MF, UB kNN, association rule mining
- **Demographic methods:** UB kNN with customer information

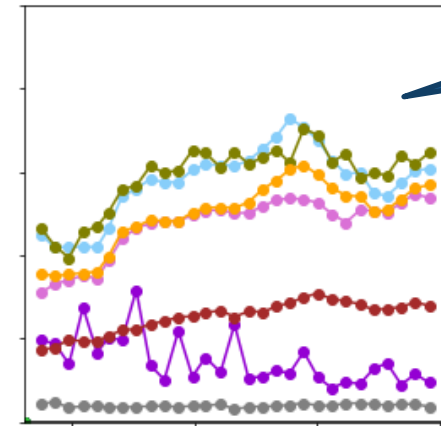
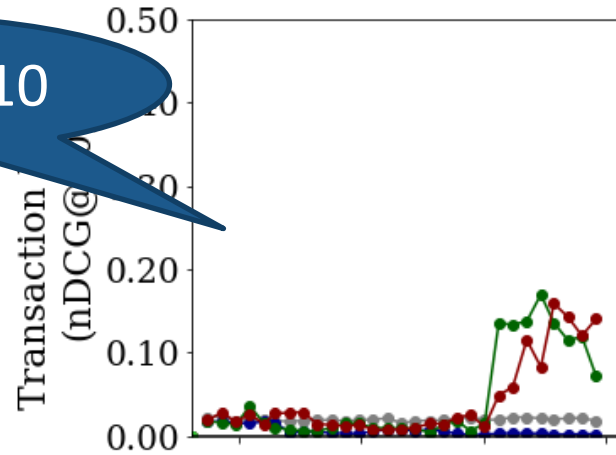
RQ1. Can we exchange transaction-based and profitability-based metrics?

Pricing-based Models

Transaction-based models

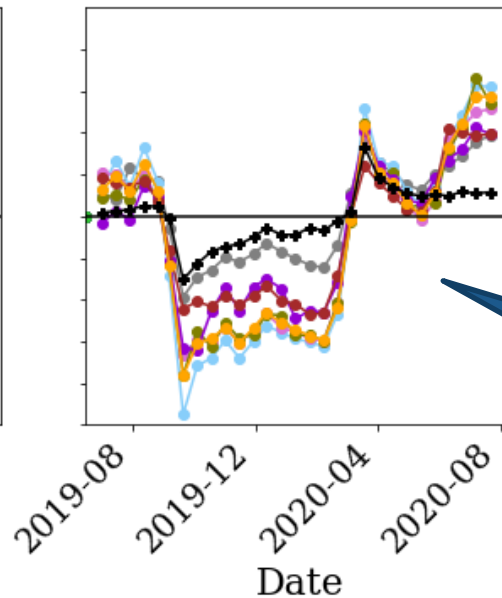
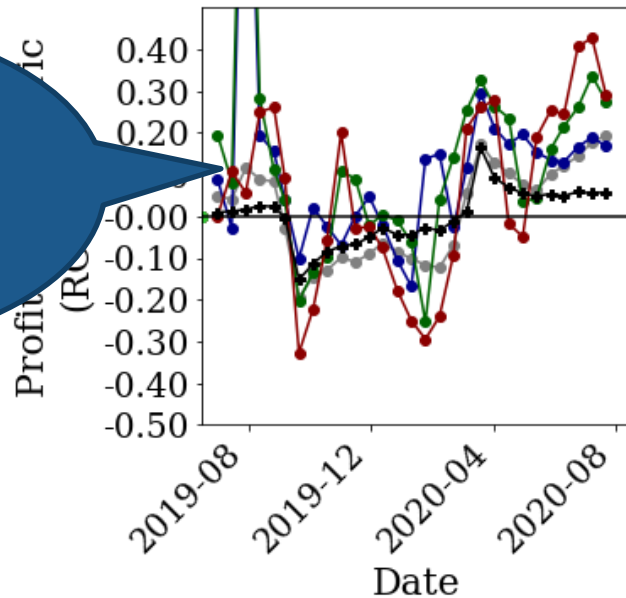
Low nDCG@10

High nDCG@10



- Random
- SVR
- LightGBM
- Random forest
- Popularity
- ARM
- LightGCN
- MF
- UB kNN
- CPS

High ROI@10
Beat market



Low ROI@10
Does not beat market

RQ1. Can we exchange transaction-based and profitability-based metrics?

- We observe many differences between nDCG and ROI.
- But... are they even correlated?
- We take:
 - Average metric values (ROI@10 / nDCG@10) for each algorithm / split point
 - Compute Pearson correlation between both metrics
- Result: **-0.22!**
- Increasing nDCG \Rightarrow money losses!

CONCLUSION: We cannot exchange both metrics. But why?

RQ2. What are the main factors that influence transaction-based metrics?

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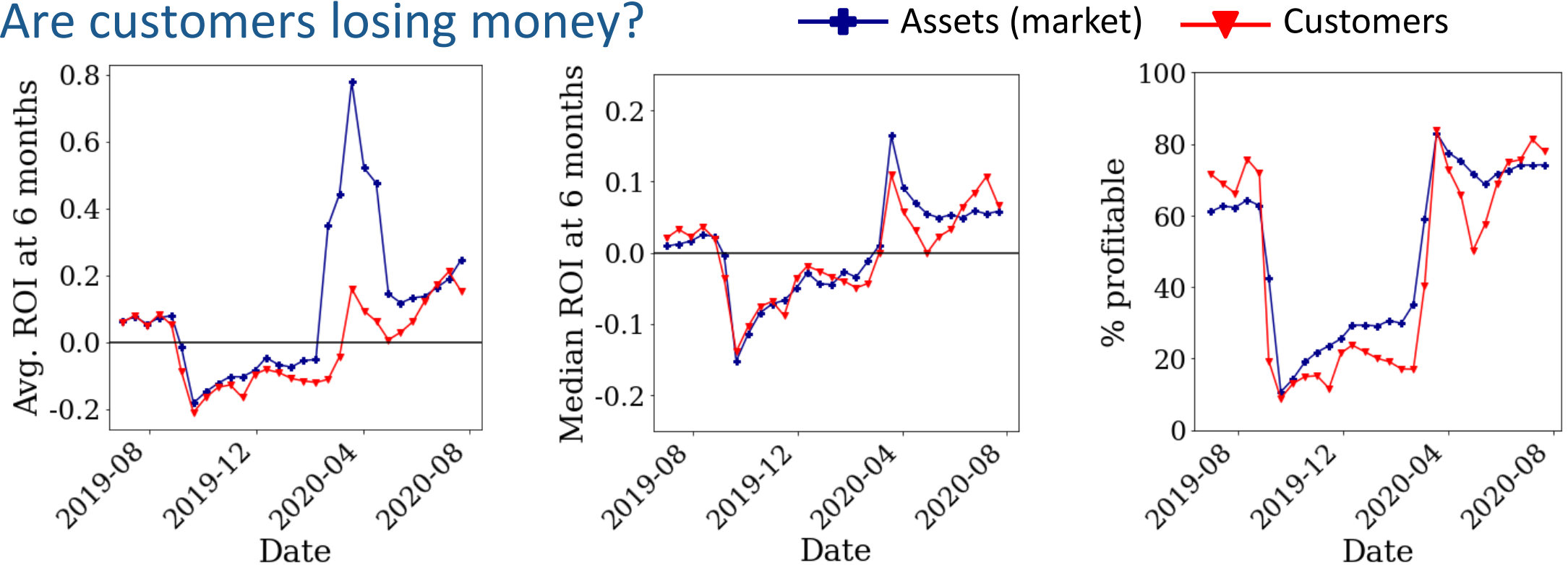


We consider three factors:

- Effectiveness of the customers
 - Do our customers invest well?
- Market changes
 - Pandemics, wars, economic crises, etc. affect market prices
 - Example: Covid-19 sank the markets
- Customer strategy
 - How much time do customers hold assets?

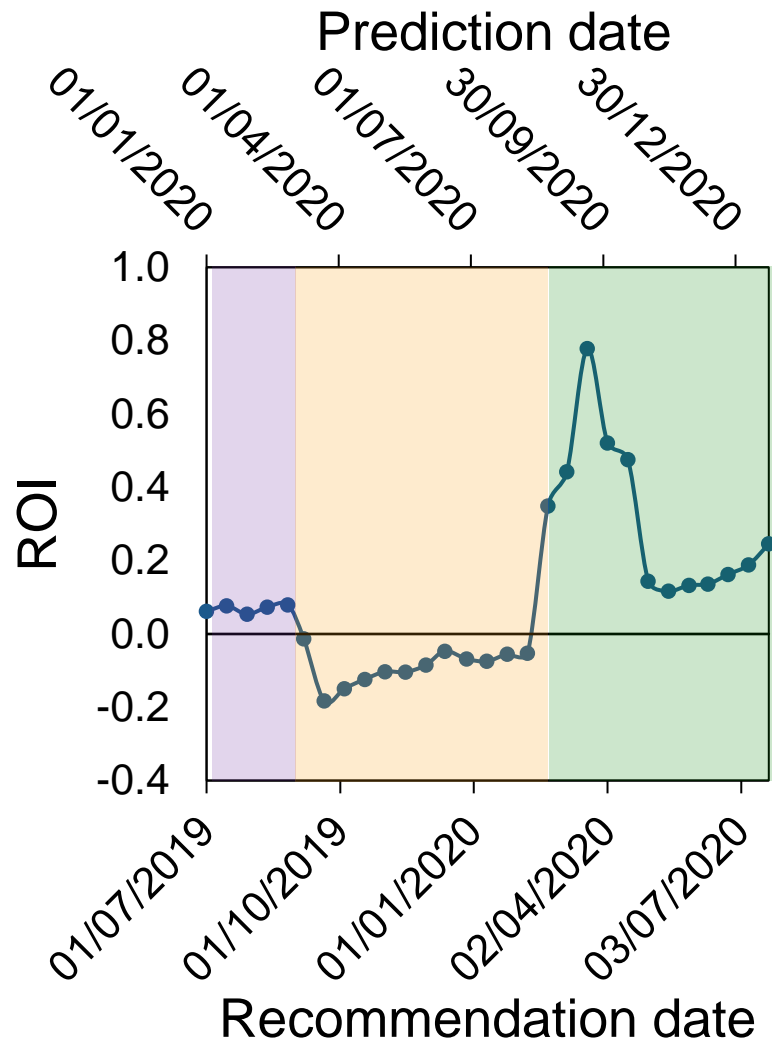
Effectiveness of the customers

- If customers invest intelligently, we would expect positive correlation
- Are customers losing money?



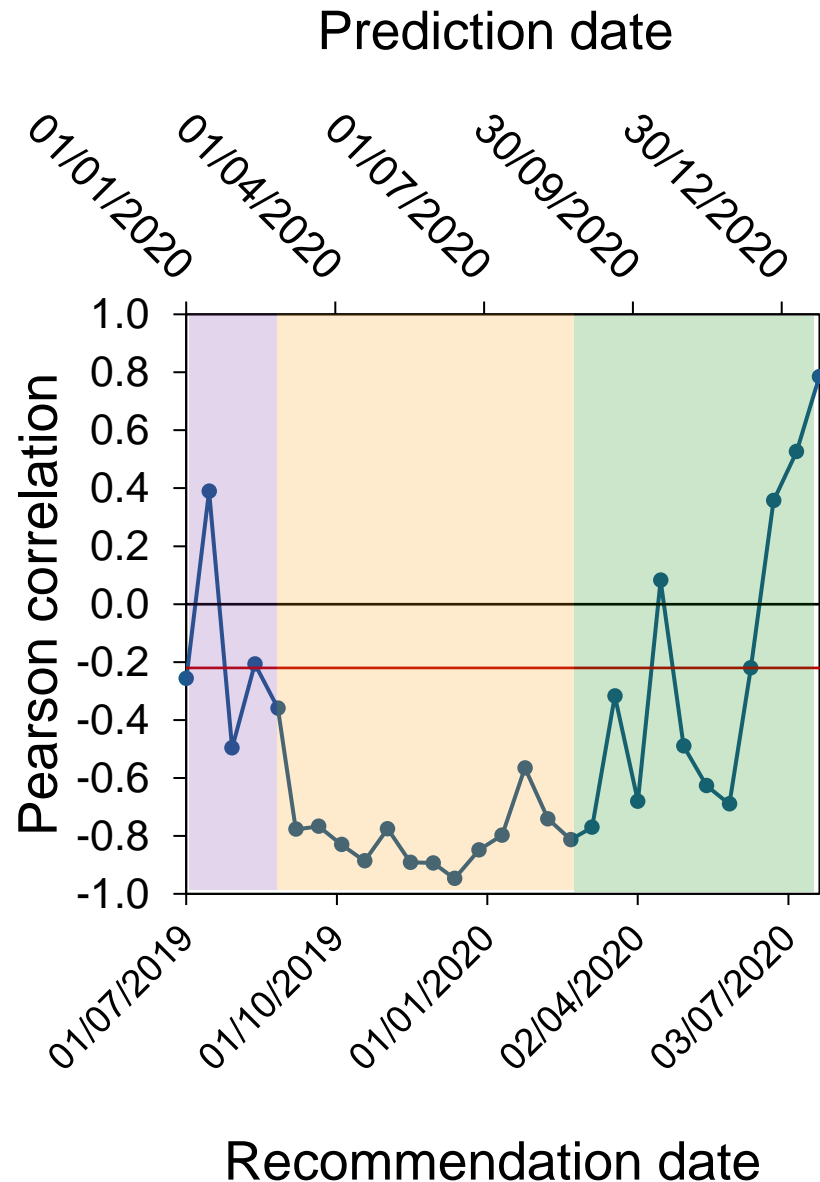
- Customers in our dataset are not particularly good investors.
- This explains lack of correlation between ROI and nDCG

Market changes



- We observe our profitability at 6 months
- Changes in asset profitability over time
- **PERIOD 1:** January 2020 – March 2020
 - Normal period
 - Market growth (slow)
- **PERIOD 2:** March 2020 – September 2020
 - Great loss period
 - Impact of Covid-19 pandemic
- **PERIOD 3:** September 2020 – January 2021
 - Recovery period
 - Great market growth

Market changes (II)



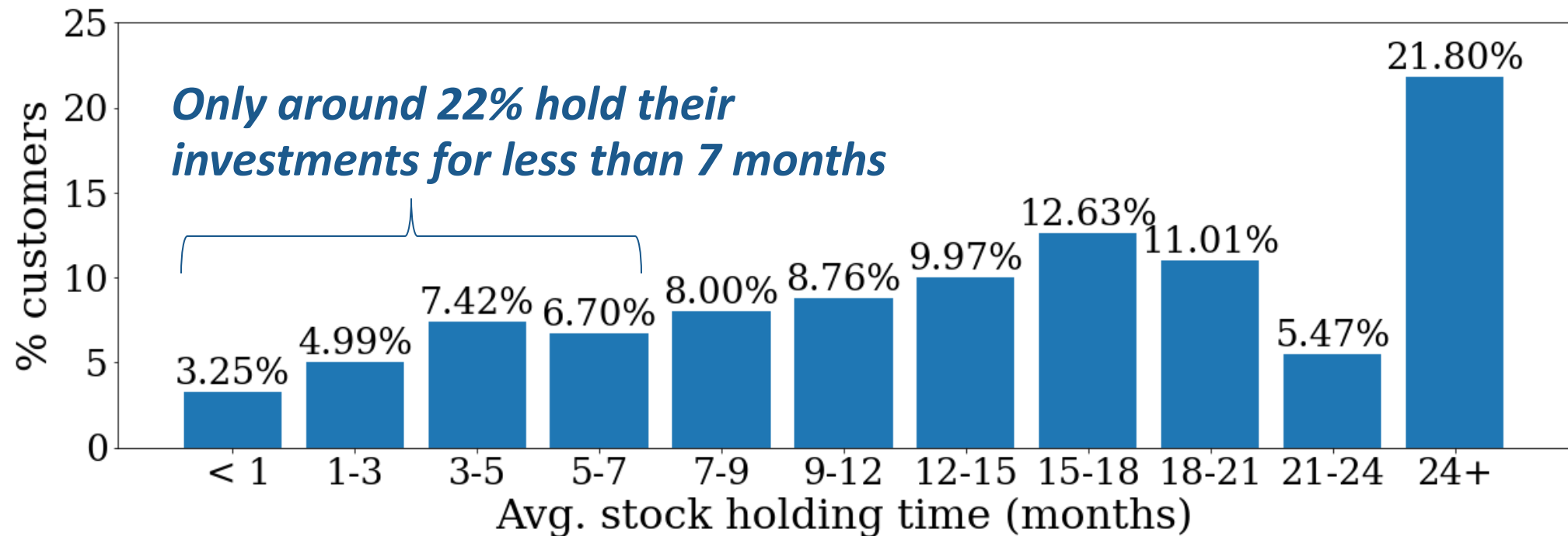
Do these changes affect the correlation between the metrics?

- Correlation between nDCG and ROI at every recommendation point
- Red line: overall correlation (-0.22)
- **PERIOD 1:** Positive to mildly negative correlation (Between -0.5 and 0.5)
- **PERIOD 2:** Very negative correlation (< -0.7)
- **PERIOD 3:**
 - Slow growth of Pearson correlation
 - Ends in high correlation (around 0.7)
- Market conditions affect correlation

Customer strategy

Is six months a reasonable future time target?

- We analyze how long people hold their investments (on average)



- Investments captured by nDCG might not necessarily align with a 6 month investment horizon.

Conclusions

- **We cannot use transaction-based metrics in exchange of profitability-based metrics – negatively correlated.**
- **Reasons**
 - Customer underperform the market average.
 - Global events impact on profitability patterns.
 - Customers might have different investment horizons / strategies.
- **Recommendations**
 - Consider changing market conditions when testing financial recommendation algorithms.
 - Customer strategies might confound our evaluation

Thanks for your attention

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Link to the paper

