Forecasting Fama-French Risk Factors: Vs Old School

WHAT

- AI Approach: LSTM (Long Short-Term Memory) network, a type of neural network that can complete a very deep learning task and has <u>outstanding predictive power</u> on time series data because of its "memory" algorithm
- Old School Approach: ARIMA (Autoregressive Integrated Moving Average) model, a <u>benchmark</u> model for time series data analysis
- Fama-French Risk Factors: the risk factors are from the Fama-French five-factor model, a very classic model designed to <u>describe stock returns</u>. The five factors are: SMB, HML, RMW, CMA, and Rm-Rf

WHY

- Background: Deep learning architectures have been applied to a wide range of fields where they have brought many astonishing achievements and improvements. FinTech is a trend of future studies and activities in the finance field
- **Research Gap**: Financial data is a typical class of time series data; however, most studies focused on the traditional forecasting models which rely on a limited number of economic variables.

HOW

- We focus on the AI approach, building a well-designed LSTM network to predict the five risk factors
- The AI and classical finance approaches will be compared: LSTM results vs ARIMA results

GOAL

- To demonstrate the predictive power of LSTM networks when applied to financial time series data
- Shed light on the applications of AI approaches in finance researches
- To further investigate the explanation power of Fama-French risk factors on the expected rate of return



brain

nodes <u>periods</u>

> Day Day ••• Day |

Period: 1990-2010 dataset factors

Hypotheses: LSTM network exhibits better predictive power on financial data than the classical financial approaches regarding two measures – MSE(Mean Square Error) and MAD(Mean Absolute Deviation)

ARIMA

Period: 1990-2010

Date	SMB		Date	Prediction	Real
19900102	-0.67		20110103	0.33	0.57
19900103	0.72	ARIMA	20110104	-0.27	-1.32
19900104	0.44		20110105	0.44	0.61
•••	•••		20110106	0.12	-0.13
20101230	0.13		20110107	0.13	-0.25
20101231	-0.6		•••		•••
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LSTM



"Neural networks are commonly considered as black-box methods."

1	X_1	•••	X_N	Y_1	 Y_5
2	X_1	•••	X_N	Y_1	 Y_5
	X_1	•••	X_N	Y_1	 Y_5
N	X_1	•••	X_N	Y_1	 Y_5



Training Set Data sources: WRDS daily stock return, Fama-French five factor model dataset computing iterations to finalize the X: raw training data – the daily returns of **X**: raw testing data – the daily returns of relationship. all North America firms listed in the all North America firms listed in the The LSTM net will capture the dataset complicated relationship between X and **Y**: raw target data – the daily F-F risk **Z**: Predicted data – the daily F-F risk Y, as well as between previous status and factors current status, that is hidden from **Y**: Raw comparing data – daily F-F risk human sense and the traditional models. factors

Methodology & Hypotheses

H1: MSE(LSTM) < MSE(ARIMA) H2: MAD(LSTM) < MAD(ARIMA)

ARIMA is a very classic and important model for time-series data forecasting Data source: Fama-French five factor model dataset

Pre-test: to dig the characteristics of Fama-French risk factors, then decide the parameters in the ARIMA models

Structure: here we take the factor "SMB" as an example

Measures

 Z_k is the predicted value output from either LSTM or ARIMA Y_k is the actual historical value recorded on the Fama-French website

Comparisons of MSE and MAD will be conducted on five factors

Robustness Test

• Using LSTM to forecast Returns directly • Fama-Macbeth Monte Carlo Analysis

A <u>computing system</u> that processes information by mimicking the way that human neurons

A type of RNN with "gates", which regulate the amount of "memory" inflows and outflows. This benefits the whole process by effectively keeping the useful historical information

Compared with the normal RNN, LSTM has no limitation of lags' length between events, making the model very suitable for processing and predicting time series data

 $MSE = \frac{1}{n} \sum_{k} (Z_k - Y_k)^2$ $MAD = \frac{1}{n} \sum_{k} |Z_k - Y_k|$

Projected Results

- The LSTM network outperforms the ARIMA model with more accurate forecasting on the Fama-French risk factors
- With a better prediction of the risk factors, we look forward to a more trustful return calculation, also validating the Fama-French model

Conclusions

Limitation

The black-box process will not give details about how the results are calculated. The curiosity of WHAT and HOW the stock returns are being influenced still cannot be answered

Further Research

LSTM networks are very diversified, modifications can be applied to many parameters and their structures

Contributions

- The paper differs from previous studies by introducing a welldesigned LSTM neural network as a forecasting vehicle of financial data
- Instead of making predictions directly on return which is the target of all the other researches, we forecast the risk factors which are the essential components of the rate of return.
- This study would confirm that the AI method of investigating financial forecasting has a large potential to be explored.



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