

The profitability of moving averages in Specific South African Economic Sectors

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Abstract

The Research Paper examines the profitability of simple and exponential Moving Averages on 3 different Johannesburg Stock Exchange Indices from 2006-2017. The research paper tests whether the weak form market hypothesis holds in the south african market. An optimization algorithm is used to determine the best short and long run moving averages in a training period. Forward testing is performed by The results provided two optimal moving strategies that generated profits in the training period and in the validation period.

Keywords: JSE RESI, JSE IND, JSE FIN, Moving Average, chartseries.

1 Introduction

Technical analysis involves using the volume and latest stock prices to determine models and technical trading indicators for a given set of data. Moving averages are the most widely known and used by practitioners and financial traders in the markets. (Sobreiro et al. 2016).

“Moving Average are a type of smoothing method for reducing, or cancelling random variation inherent in data taken over time. When applied properly, this technique reveals more clearly the underlying trend, seasonal and cyclic approach components in the data.” (Okkels 2016) This definition implies that moving average can still be used in understanding the direction of the market even if it cannot be used directly in beating the buy-and-hold strategy. Recent studies suggest that moving averages can be used to generate profits. The paper investigates if moving averages are profitable in the South African market.

The Johannesburg Stock Exchange is the biggest stock exchange in Africa. Performing moving average trend line analysis on the 3 different sectors in the JSE creates a greater opportunity to exploit profits in specific economic sectors. The JSE IND is the top 25 industrial companies index in terms of market capitalization, the JSE RESI is the top 10 resource companies index in terms of market capitalization and the JSE FIN is the top 15 financial companies index in the JSE in terms of market capitalization. The industrial, resource and financial sectors are the 3 biggest sectors in South Africa. The paper aims to test moving average trend analysis in these 3 specific sectors.

Given South Africa’s growing global presence, research on the profitability of moving averages in the South African Market could be utilized for future profits. The paper investigates three different sectors on the Johannesburg Stock Exchange, (JSE). The three sectors are the financial, resource and

industrial sector indices. The paper examines Simple and Exponential moving averages to determine profitability of these three sectors while including transaction costs.

Only one paper has done moving average analysis on South Africa. The paper examined the 5 emerging national economies of Brazil, Russia, India, China and South Africa using technical analysis (BRICS). (Sobreiro et al. 2016) The paper analyzed South Africa but did not cover the South African market extensively.

The importance of the research question can be split up into a number of specific reasons. The South African stock market is described as weak-form inefficient suggesting profitable opportunities. (Appiah-Kusi and Menyah 2003) The lack of information involving moving average trend lines in the South African market and the potential for profits emphasizes the need for Moving Average analysis in the South African Market specific to sectors. There is very little research on moving averages in different sectors.

The Data is explained more in section four. The analysis of each sector could provide crucial information for potential investors and technical traders. Finally, given that the paper is restricted to only simple and exponential moving averages, many of the assumptions underlying technical analysis can be ignored. Section five elaborates more on the assumptions used in the paper and the exact methodology. Section six provides analyses and tests the results. Section seven is the conclusion.

2 Background

2.1 The beginning of technical analysis

Technical analysis was first introduced by Charles H. Dow. Charles H. Dow believed stock market prices could provide information on the overall market. This led to Dow theory which was introduced in the late 1800's starting what is known as technical analysis. (Bessembinder and Chan 1998) Technical analysis is used in the financial industry amongst different market participants. The introduction of the Efficient Market Hypothesis proposed by Fama (1970) led to a decline in the use of technical analysis and the disbelief that profits could be generated from technical analysis. (Sobreiro et al. 2016) In 1992, William Brock, Josef Lakonishok and Blake LeBaron provided significant evidence in justifying the use of technical analysis. (Fong and Yung 2005) The paper found significant evidence that simple trading rules provide economically and statistically significant returns. Moving averages in particular outperformed the buy-and-hold to the resurgence in testing technical analysis in different markets and determining whether it is actually profitable.

2.2 The Efficient Market Hypothesis

The paper rewritten by (Fama 1991) refers to the Efficient Market Hypothesis as the asset prices fully reflecting all available information. The Efficient Market Hypothesis can be split into 3 forms based on the definition of the information:

1. Weak Form EMH: Prices reflect all information in the past price history
2. Semi-Strong Form EMH: Prices reflect all publicly available informa-

tion

3. Strong Form EMH: Prices reflect all information, public and private.
(Park and Irwin 2007)

If all information is available in the prices, then additional information such as moving averages shouldn't provide any financial profitability. The Weak form efficient market hypothesis states that there is no justification for technical analysis. This implies that markets are efficient and arbitrage should not be attainable through technical analysis. (Fama 1991). The weak form of the efficient market hypothesis has been the paradigm in describing the behaviour of prices in speculative market. A paper done in 1988 proved the non-existence of random walks in the stock market using the NYSE-AMEX index. (Lo and MacKinlay 1988) The Efficient Market Hypothesis is one of the main reasons why individuals do not regard technical analysis as useful and has lead market participants to avoid technical analysis. Technical analysis is still used by market participants. Moving averages might not provide exact profits but they do provide other information about the data.

2.3 Sector Analysis

Mebane Fabor (2010) analysed ten sectors by combining different sectors together to try and beat the overall market index. The results provided substantial profits. The paper indicates that analysis of specific sectors and combining sectors could results in profitable opportunities. Resource companies are cyclical companies while industrial and financial companies are only prone to impactful economic events.(Berman and Pfleeger 1997).

2.4 Moving Averages

The simple and exponential moving averages are the most frequently used moving averages and are used in different ways to predict the buy and sell signals in the price history. Exponential moving average gives more weight to recent stock prices while simple provides equal weighting for all stock prices in a given time period. (Okkels 2016). One method is the comparison of short and long run simple moving averages. Short and long run moving averages refers to the specific time frame in which the moving average is calculated. In a paper done on South Asian markets, short is referred to as 10, 20 or 50 day moving averages, while long run is referred to as 50,100 or 200 day moving averages. (Ming-Ming and Siok-Hwa 2006) Moving average techniques require buy and sell signals after comparing short and long run simple or exponential moving averages. (Gunasekarage and Power 2001). The paper discusses two moving average techniques which are: Variable length moving average and fixed length moving average. The variable length moving average rule generates a buy-and-sell signal every day. Variable length moving average signals to buy when short moving average is above the long moving average. The crossing over of Moving averages is referred to as the dual moving average crossover. (Ming-Ming and Siok-Hwa 2006). The variable length moving average is written as $VMA(S,L,B)$ where S is the short moving average, L is the long moving average and B is the band that allows the actual buying and selling signal to occur. B is written as a percentage. B is important as it allows us to know when we should or should not purchase. B is in line with transaction costs(Bessembinder and Chan 1998). The fixed length moving average holds a moving average for a fixed period of time. This is a 10-day holding period in Brock (1992) and Ming-Ming, L & Siok-Hwa, L (2006). During this 10-day period signals do not change the fixed

length moving average till the end of the holding period, where cumulative profits are calculated. Fixed length moving average focuses on crossing over of long and short moving averages. A sell (buy) signal occurs when the short (long) moving average crosses the long (short) moving average from above. Moving Average strategies require:

$$Profit(Aftertransactioncosts) = (meanreturn * numberoftrades) - (transactioncosts)$$

The above strategy provides the profit in percentages. The strategy prevents using exact monetary values and accounts for transaction costs. (Ming-Ming and Siok-Hwa 2006) The strategy is simple to use and requires complete investment of funds into each signal. The research paper will follow a similar strategy in determining the profitability of moving averages.

2.5 Empirical results

Brock (1992) analysed fixed and variable moving averages in the Dow Jones industrial Average Index using 90 years of data. (1897-1986). Both the fixed and variable moving averages provided excess returns compared to the buy-and-hold strategy. These returns did not include transaction costs.

6 equity indices in Asia were analysed and provided results indicating strong predictive power of moving averages in Malaysian, Thailand and Taiwan stock markets. The Results confirm the profitability of moving average trend lines. (Bessembinder and Chan 1998) The study obtained stock price index data from (1975-1989). The returns are computed as changes in log price indices thus preventing autocorrelation. The transaction costs were reported as break even costs in order to calculate the highest percentage of profits available for transaction costs. The average transaction cost is 1.57%

for the whole sample. If the transaction cost were 1% then moving averages would have been profitable.

Ming-Ming & Siok-Hwa used the daily closing price index as the short term moving average. (2006) The long term moving average varies between the time period of 20, 60, 120, 180, 240 days. Multiple long moving averages are compared to the daily closing price index. The results concluded that the best strategy was using a 60-day long moving average and the daily closing price index. The results indicated variable moving averages were more profitable than fixed moving average and both provided significant returns.

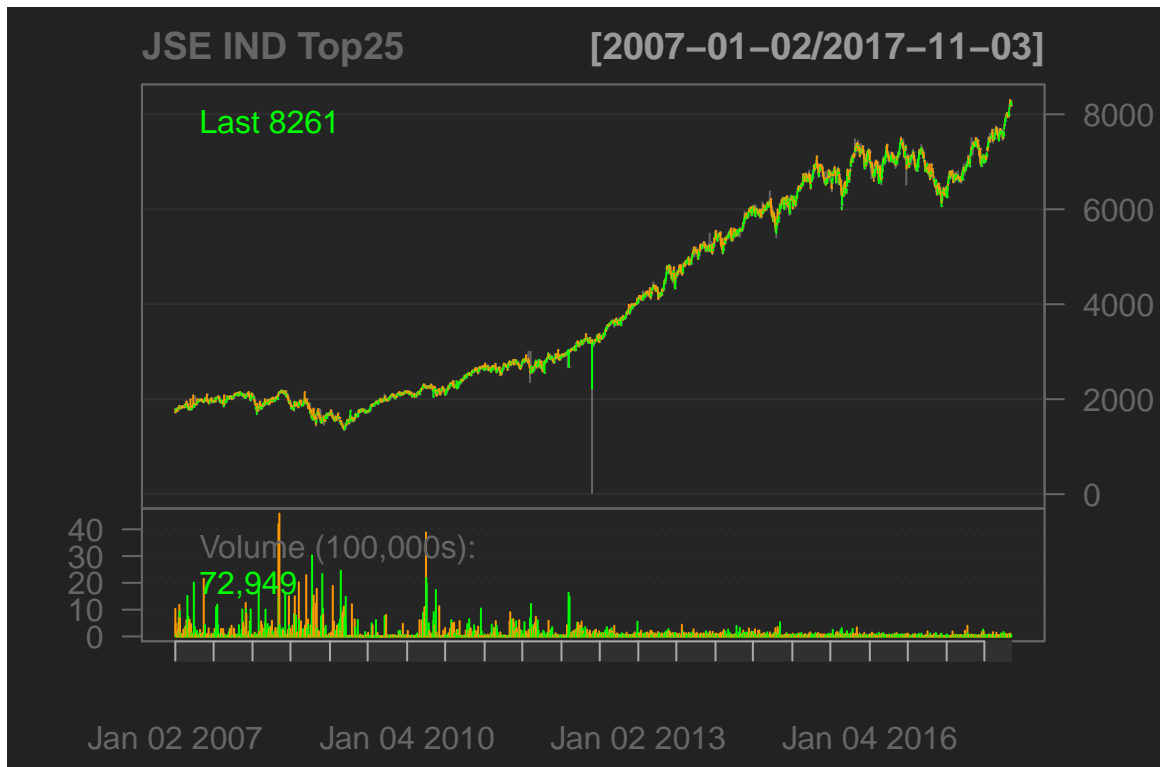
The variable moving average requires multiple buy and sell signals until the short moving average crosses the long moving average. The research paper will only focus on fixed moving averages thus limiting the number of buy and signals.

3 Data

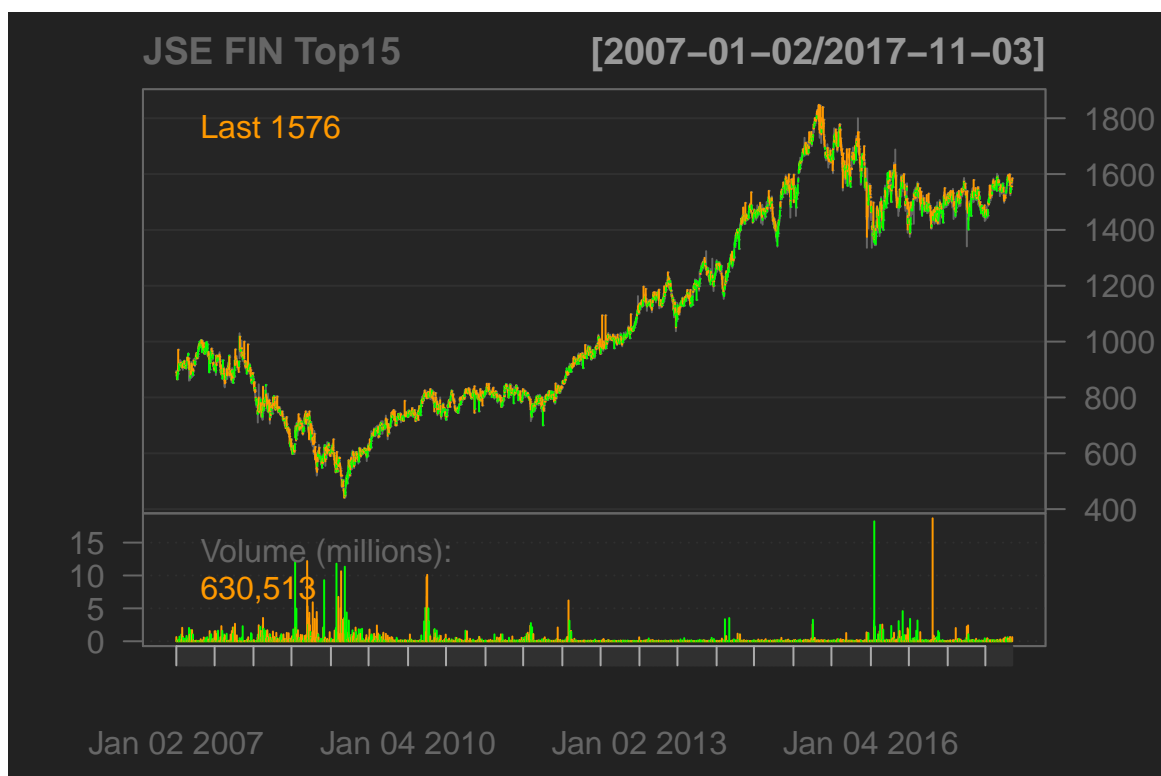
The data chosen for the study involves using closing prices from 2006 to 2017. The closing prices are split into two categories. The first set of closing prices are over an eight year period from 1st of May, 2006 to 31st of April 2014. The first set of closing prices is the training period. This period allows us to generate an optimal moving average strategy. The validation period test will occur between the 1st of May 2014 to 31st of April 2017. The second period allows us to test if our training period moving average strategy is still profitable out-of-sample. All data is taken from Google Finance and focuses on the JSE Top10 Resource companies index, JSE Top25 industrial companies index and the JSE Top15 financial companies index.

The Transaction cost used consists of four aspects:

- Brokerage fee= 0.6%
- Investor protection levy = 0.0002%
- STRATE fee= 0.005459%
- Security Transfer Tax=0.25% The total transaction costs = 0.855659%.(“FNB: FAQs - Buying and Selling” 2017) The closing prices used from Google finance are log prices.The R package quantmod is used to collect the data from google finance and convert prices to log returns.The returns referred to in the paper are the log-returns. To avoid autocorrelation daily log returns were used to calculate profits rather than the actual closing prices.



Chartseries to show the closing prices for the JSE Top25 Industrial companies indice.(2007-2017)



Chartseries to show the closing prices for the JSE Top15 Financial companies indice.(2007-2017)



Chartseries to show the closing prices for the JSE Top10 Resource companies indice.(2007-2017)

The three chartseries provides valuable information on each individual sector. The financial and industrial sector seem to be in a bullish position, the market price is on the rise, for the last 10years and continue to rise. The resource sector data seems highly volatile. Moving average analysis on the JSE Top10 Resource company could provide exploitable opportunities from the multiple fluctuation shown in the chartseries.

4 Methodology

The two Moving Averages used in this research paper are simple moving averages and exponential moving averages.

4.1 Simple Moving Average

Simple moving averages is the sum of latest stock prices divided by the number of stock prices: Simple moving average calculation

$$SMA_n = \frac{1}{k} \times \sum_{t=n-k+1}^n P_t$$

Where: P_t is the closing price of the stock in t period.

n is the relative position of the current period observed; and k is the number of periods included in the SMA calculation;

Simple averages are the easiest to interpret while exponential moving averages provide stronger predictive ability in market prices. Exponential moving averages focus on the most recent values and thus are similar to weighted moving averages.

4.2 Exponential Moving Average

$$EMA_n = \left(\frac{2}{k+1} \right) \times P_{t-1} + \left(1 - \left(\frac{2}{k+1} \right) \right) \times EMA_{n-1}$$

where: P_{t-1} is the closing price of the stock in the previous period.

EMA_{n-1} is the EMA in the previous period.

n is the relative position of the current period observed; and

k is the number of periods included in the EMA calculation;

The Research project will only focus on simple and exponential moving average trend lines. Moving Average Convergence Divergence is the formal name for comparing a short and long run exponential moving average. The simple and exponential moving averages were calculated using the TTR package in R. The Methodology that will be incorporated in answering the research question is the application of dual moving average crossover and moving average convergence divergence. This allows for short and long term

moving averages while comparing simple and exponential moving averages. The literature reviewed suggests using a short moving average of either 5,10, or fifty days and long moving average of either fifty,one hundred or two hundred days.

The exact methodology will be explained in a step by step algorithm how to test the profitability of moving averages in different sectors. The algorithm is used for both simple and exponential moving averages.

Algorithm:

$$\textit{Moving average function} = \textit{function}(\textit{Data}, \textit{short}, \textit{long}, \textit{cost})$$

The variables in the function are defined:

- Data is the actual data from each sector we are analysing and the exact time frame. The data is split into a training period and a validation period.
- Short is the short run moving average
- Long is the long run moving average
- Cost is the actual transaction cost = 0.855659%

4.3 Moving Average Algorithm

The exact methodology of the function will be explained in a step by step process. The function is used seperately for simple and exponential moving averages. The only change is that the short and long moving average must either be a simple or exponential moving average. The research paper does not test short and long run moving averages of different natures.

1. The function compares the short and long run moving averages by generating signals to indicate whether a buy or sell signal should occur. If the short run moving average is greater than the long run moving average a buy signal is generated. A sell signal occurs if the long run moving average is greater than the short run moving average. A signal is also generated to hold our position, if the short and long run moving averages are equal.
2. The function takes these signals and determines the exact days we will trade in our data set. The data now has signals for each day in which either a buy,sell or hold our position will occur. The function also tracks the number of buy and sell signals.
3. The log returns of a buy and hold strategy are then generated for the specific data set.
4. The transaction cost are minused from each buy signals log return generating the complete buy returns. These buy returns are added to the sells signals log returns to generate the total return.
- 5.The function plots a table with the short moving average, long moving average, strategy returns,strategy standard deviation, strategy sharpe ratio, buy and hold return, buy and hold standard deviation and the buy and hold strategy sharpe ration

Simple Moving Average Function = function(JSEFIN, 30, 355, 0.855659)

The Simple moving average function calculates the returns, standard deviation and sharpe ratio for a simple moving average for the JSE Top15 financial companies using a short moving average of 30 days, long moving average of 355 days and a transaction cost of 0.855659%. The function also

calculates the returns, standard deviation and sharpe ratio for the buy and hold strategy in the JSE Top25 financial companies. The function can also be used to calculate exponential moving averages. Section 5 provides tables of the results.

Exponential Moving average function = function(Data, short, long, cost)

The exponential function differs to the simple moving average function in the fact that the short and long run moving average are calculated only using exponential moving averages.

A sequence using the moving average function was run to generate different values for the short and long periods. The short periods ranged from five to fifty-five days in steps of twenty-five days while the long periods ranged from fifty-five to four hundred and five days in steps of fifty. The sequence generates twenty-four moving averages for either simple or exponential moving average. The empirical results will only look at the top six strategies to determine if any strategies are profitable.

The sequence will be run for simple and exponential moving averages in the three different sectors which will result in six tables of data analysis. The main area of risk that need to be analysed is the standard deviation and sharpe ratio. The tables will provide comparative analysis of the optimal strategies and the buy-and-hold strategy.

Back testing is used in the training period to calculate the optimal strategy for each JSE index using simple and exponential moving averages. The Best strategy for each moving average in our training period will be tested in a validation period to ensure the results were not over optimized to match the specific training period. The testing of a strategy in the validation period is forward testing. Forward testing prevents inaccurate results and verifies whether the optimal strategy is profitable out of the initial sample.

5 Empirical results

The empirical results for the best strategies in the training periods are plotted in the tables below. The headings of each column are explained. “short” is the number of days used in the short run moving average. “long” is the number of days used in the long run moving average. “strat_return” is the overall return generated from the strategy of short and long run moving averages. “strat_sd” is the standard deviation for the strategy. “strat_sharpe” is the sharpe ratio for the specific moving average strategy. “bh” refers to the buy-and-hold strategy. “Bh_r” is the buy-and-hold strategy returns. “Bh_sd” is the buy-and-hold standard deviation. “bh_sharpe” is the buy-and-hold strategy sharpe ratio. “Buy” is the number of buys that occur due to the strategy. “sells” is the number of sells that occur due to the strategy throughout the investment period.

Table 5.1: **Table to show the top 6 strategies in the JSE Top15 Financial companies using a simple moving average between 2006-2014**

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
30	205	0.0709	0.1623	0.4369	0.0253	0.2371	0.1066	6	5
55	155	0.0598	0.1601	0.3738	0.0253	0.2371	0.1066	9	8
55	205	0.0586	0.1614	0.3630	0.0253	0.2371	0.1066	7	6
30	155	0.0256	0.1636	0.1564	0.0253	0.2371	0.1066	13	12
55	255	0.0240	0.2181	0.1100	0.0253	0.2371	0.1066	4	3
55	305	0.0126	0.2220	0.0566	0.0253	0.2371	0.1066	4	3

The best strategy is a thirty short day moving average and a two hundred

and five day long run moving average. The profit generated is 4.56% higher than the buy-and-hold strategy. The returns are 1.8 times greater than the buy-and-hold returns. The sequence generated five profitable moving average strategies indicating that multiple moving average combinations could be tested in the validation period. The optimal moving average strategy has less volatility and more return resulting in a sharpe ratio of 0.3378. This is better than the buy and hold sharpe ratio but still is not close to the desired value of 1. The strategy is highly profitable but will have to be tested out of sample in order to be validated.

Table 5.2: Table to show the top 6 strategies in the JSE Top15 Financial companies index using a exponential moving average between 2006-2014

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
55	155	0.0542	0.1605	0.3378	0.0253	0.2371	0.1066	8	7
55	205	0.0476	0.1610	0.2959	0.0253	0.2371	0.1066	7	6
30	205	0.0443	0.1620	0.2731	0.0253	0.2371	0.1066	9	8
55	105	0.0313	0.1518	0.2064	0.0253	0.2371	0.1066	12	12
30	155	0.0291	0.1625	0.1791	0.0253	0.2371	0.1066	13	12
55	305	0.0023	0.2231	0.0101	0.0253	0.2371	0.1066	5	4

The best exponential moving average strategy is a fifty five day short run moving average and a one hundred and fifty five day long run moving average. The strategy provides excess returns of 2.89%. The sharpe ratio is 0.3378 which is similar to the simple moving average sharpe ratio. Both the exponential and simple moving averages provide small returns in the financial companies index however the returns could be leveraged correctly

to generate large profits.

Table 5.3: Table to show the top 6 strategies in the JSE Top25 Industrial companies index using a simple moving average between 2006-2014

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
5	355	0.1420	0.1524	0.9315	0.1367	0.1881	0.7265	3	3
30	305	0.1407	0.1534	0.9168	0.1367	0.1881	0.7265	3	2
5	405	0.1358	0.1851	0.7337	0.1367	0.1881	0.7265	2	1
5	305	0.1320	0.1526	0.8652	0.1367	0.1881	0.7265	8	7
30	355	0.1320	0.1534	0.8602	0.1367	0.1881	0.7265	2	2
55	305	0.1292	0.1562	0.8274	0.1367	0.1881	0.7265	3	2

Table 5.4: Table to show the top 6 strategies in the JSE Top25 Industrial companies index using a exponential moving average between 2006-2014

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
5	305	0.1511	0.1540	0.9809	0.1367	0.1881	0.7265	4	3
30	355	0.1452	0.1540	0.9427	0.1367	0.1881	0.7265	2	2
30	305	0.1432	0.1542	0.9290	0.1367	0.1881	0.7265	3	2
5	405	0.1386	0.1593	0.8698	0.1367	0.1881	0.7265	6	5
5	355	0.1329	0.1546	0.8595	0.1367	0.1881	0.7265	8	8
55	355	0.1329	0.1558	0.8529	0.1367	0.1881	0.7265	2	2

The industrial sector provided the highest returns compared to the fin-

ancial and resource sectors indices. The excess returns for the simple and exponential are 0.53% and 1.44% respectively. The sharpe ratios are close to 1 indicating a considerably good investment for investors. The optimal strategy suggests a long moving average strategy over three hundred and five days indicating a general long term trend. The chartseries in in the data section confirms the bullish trend.

Table 5.5: Table to show the top 6 strategies in the JSE Top10 Resource companies index using a simple moving average between 2006-2014

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
55	105	0.0197	0.2214	0.0888	-0.059	0.318	-0.1854	12	12
30	155	0.0083	0.2118	0.0391	-0.059	0.318	-0.1854	9	9
55	205	-0.0070	0.2110	-0.0331	-0.059	0.318	-0.1854	7	7
30	405	-0.0075	0.1944	-0.0384	-0.059	0.318	-0.1854	6	6
55	255	-0.0094	0.2045	-0.0460	-0.059	0.318	-0.1854	7	7
55	155	-0.0144	0.2142	-0.0674	-0.059	0.318	-0.1854	8	8

Table 5.6: Table to show the top 6 strategies in the JSE Top10 Resource companies index using a exponential moving average between 2006-2014

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
30	105	0.0192	0.2036	0.0942	-0.059	0.318	-0.1854	10	10
55	155	-0.0088	0.2005	-0.0440	-0.059	0.318	-0.1854	7	7
30	55	-0.0118	0.2044	-0.0577	-0.059	0.318	-0.1854	19	19

short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
55	205	-0.0178	0.1986	-0.0897	-0.059	0.318	-0.1854	8	8
55	105	-0.0238	0.2020	-0.1176	-0.059	0.318	-0.1854	9	9
30	205	-0.0248	0.1948	-0.1271	-0.059	0.318	-0.1854	11	11

The Resource sector is the most volatile sector as shown by the chartseries in the data section. The resource sector is a cyclical sector resulting in the excess volatility. The 5.5 and 5.6 indicate this as the Top 10 Resource companies have the highest risk with the top strategies in both simple and exponential moving averages obtaining standard deviations greater than 0.2. The returns are also negative in most strategies. Only the top exponential moving average strategy had a positive return of 1.92% while the top simple moving average had a return of 1.97%. The resource sector is the only sector to have a negative buy and hold return of -5.9%. The results provide evidence that additional risk will provide excess returns. The resource sector experiences strong seasonal changes in demand for specific resources. This also generates additional volatility.

5.1 Final results

The Top 6 strategies were analysed in the validation period from the 1st of May 2014 to 31st of April 2017. The exact same moving average algorithm was used with the new dates generated from the previous moving average strategy sequence. The exact short and long periods were used to calculate the new returns, standard deviations and sharpe ratios for the buy and hold strategy and the optimal moving average strategy. The table below uses the same variables labelled in the beginning of the empirical results section while

including a new column which indicates the specific economic sector and the type of moving average.

Index, MA	short	long	strat_r	strat_sd	strat_sharpe	bh_r	bh_sd	bh_sharpe	buy	sell
JSE FIN SMA	30	205	0.0042	0.1228	0.0341	0.0077	0.1923	0.0402	3	3
JSE FIN EMA	55	105	0.0232	0.1266	0.1835	0.0077	0.1923	0.0402	3	3
JSE IND SMA	5	305	0.0998	0.1457	0.6851	0.1056	0.1567	0.6737	3	3
JSE IND EMA	5	355	0.0906	0.1451	0.6246	0.1056	0.1567	0.6737	2	2
JSE RES SMA	55	105	0.0131	0.1678	0.0783	-0.1426	0.2872	-0.4964	3	3
JSE RES EMA	30	105	-0.0931	0.1975	-0.4712	-0.1426	0.2872	-0.4964	4	3

Table 5.7: Table to show the six optimal moving average strategies tested in the validation period

The Top six strategies in the validation period provided three strategies that outperformed the buy and hold strategy. The exponential moving average in the financial sector and both the simple and exponential moving average in the Resource sector outperformed the buy and hold strategy. The exponential moving average strategy in the JSE Top10 Resource companies indice outperformed the buy and hold strategy by 4.95% however the return on the strategy is -9.31% indicating that this strategy is unprofitable. The exponential moving average in the Top 15 financial companies generated excess returns of 1.55%. The strategy provided the lowest standard deviation for a profitable strategy of 12.66%. The simple moving average strategy for the JSE Top10 Resource companies index generated returns of 1.31% which exceeded the buy and hold strategy by 15.57%. The strategy required additional risk in the form of a 16.78% standard deviation. The best strategy that exceeded the buy and hold returns required a relatively short, long moving average length of 105 days. Due to the volatile prices indicated in the chartseries, it can be seen that a long moving average of 105 days would be

profitable as worthwhile trades are feasible in exploiting abnormal returns. The strategy has enough time to react to a profitable opportunity in the market. The Industrial sector moving averages did not outperform the buy and hold strategy incurring additional costs in the form of transaction costs. The sharpe ratio for the two profitable strategies are both below 0.2 indicating a lack of investor incentive to partake in the investment. The sharpe ratios for the profitable strategies out performed the buy and hold sharpe ratios.

The Final results indicates that moving average analysis is profitable in particular economic sectors using particular types of moving averages. The industrial sector moving analysis generated returns slightly below the buy and hold strategy. The industrial sector results indicate that the weak form efficient hypothesis holds in the industrial sector.



Chartseries to show the closing prices for the JSE Top10 Resource companies indice with a 105 day simple moving average.(2007-2017) Given the steep changes in the trend of the JSE Top10 Resource companies Index, multiple trading opportunities could arise from using the past prices.

6 Conclusion

The results provided for strategies that are unprofitable are not completely useless. The moving average analysis provided understanding of economic sectors in the form of trends. The two strategies that outperformed the buy and hold strategy provide evidence that sectors in the South African market are weak from inefficient. The Simple Moving average strategy in the JSE Top10 Resource companies index provides evidence to confirm the profitability of moving averages in a particular South African economic sector.

Further analysis should be performed on the JSE Top10 Resource companies index and the JSE Top15 Financial companies index to verify the results in the research paper. This study was confined to only three sectors in the JSE and only used two moving averages. More sectors could be analysed to provide more representative results. More assumptions could be kept such as only analysing one specific sector or avoiding sectors such as Resources sectors which are considered highly volatile. Additional research could be performed to test data extending before 2006 and the moving average strategy function could apply more sequences which would require more time and computational burden.

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