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*Jukka Ilomäki<sup>1</sup>*  
*Hannu Laurila<sup>1</sup>*  
*Michael McAleer<sup>2</sup>*

<sup>1</sup> Faculty of Management, University of Tampere, Finland

<sup>2</sup> Department of Finance, Asia University, Taiwan

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Burg. Oudlaan 50  
3062 PA Rotterdam  
The Netherlands  
Tel.: +31(0)10 408 8900

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**Jukka Ilomäki**

Faculty of Management  
University of Tampere  
Finland

**Hannu Laurila**

Faculty of Management  
University of Tampere  
Finland

**Michael McAleer**

Department of Finance  
Asia University, Taiwan  
and  
Discipline of Business Analytics  
University of Sydney Business School, Australia  
and  
Econometric Institute, Erasmus School of Economics  
Erasmus University Rotterdam, The Netherlands  
and  
Department of Economic Analysis and ICAE  
Complutense University of Madrid, Spain  
and  
Institute of Advanced Sciences  
Yokohama National University, Japan

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Corresponding author: hannu.laurila@uta.fi

## **Abstract**

Consider using the simple moving average (MA) rule of Gartley (1935) to determine when to buy stocks, and when to sell them and switch to the risk-free rate. In comparison, how might the performance be affected if the frequency is changed to the use of MA calculations? The empirical results show that, on average, the lower is the frequency, the higher are average daily returns, even though the volatility is virtually unchanged when the frequency is lower. The volatility from the highest to the lowest frequency is about 30% lower as compared with the buy-and-hold strategy volatility, but the average returns approach the buy-and-hold returns when frequency is lower. The 30% reduction in volatility appears if we invest randomly half the time in stock markets and half in the risk-free rate.

**Keywords:** Market timing, Moving averages, Risk-free rate, Returns and volatility.

**JEL:** G32, C58, C22, C41, D23.

## 1. Introduction

According to the standard investing separation theorem of Tobin (1958), investors allocate investments between risk-free and risky assets. If the risk-free rate is low (high), the investors shift their wealth to (from) the risky assets. Fama (1972) divides forecasters into two categories, namely macro forecasters (or market timers) and micro forecasters (or security analysts), who try to forecast individual stock returns relative to the market returns.

Merton (1981) defines a market timer to forecast when stocks will outperform (underperform) the risk-free asset, indicating that, when  $r_t^m > r_t^f$  ( $r_t^m < r_t^f$ ), where  $r_t^m$  is average stock market returns,  $r_t^f$  is the risk-free asset,  $r_t^i = r_t^f + \beta^i(r_t^m - r_t^f) + \varepsilon_t^i$ ,  $r_t^i$  is the return for individual stock  $i$  included in the market portfolio  $m$ ,  $\beta^i$  is a positive parameter, and  $E[\varepsilon_t^i | r_t^m] = E[\varepsilon_t^i]$ . That is, a market timer only forecasts the statistical properties of  $r_t^m - r_t^f$ , indicating that their forecasts contain only the differential performance among individual stocks arising from systematic risk in the markets.

Merton (1981) shows theoretically that when investors have heterogeneous beliefs and imperfect information, the value of a random market timing forecast is zero, and if the forecast variable is distributed independently or the forecast is based on public information, its value is zero, too. In fact, Merton shows that the maximum value of skilled market timing is the value of the protective put against buy-and-hold strategy.

Henriksson and Merton (1981) present an empirical procedure whereby correct forecasts can be analyzed statistically. However, if it is assumed that  $\varepsilon_t^i$  follows an approximate normal distribution, this leads to the CAPM of Sharpe (1964) and Lintner (1965).

We use a simple MA rule for the timing aspect for individual Dow Jones Industrial Average (DJIA) stocks with different frequencies. Zhu and Zhou (2009) show analytically that MA trading rules, as a part of asset allocation rules, can outperform standard allocation rules when stock returns are partly forecastable. The standard rule

means investing a fixed proportion of wealth in risky assets and the rest in risk-free assets, with the ratio determined by the risk tolerance of an investor.

This is the well-known reward/risk (or mean-variance) principle in the spirit of Markowitz (1952), Tobin (1958) and Sharpe (1964). Zhu and Zhou (2009) argue that the fixed allocation rule is not optimal if returns are forecastable by using the MA rule. Therefore, assuming that risk tolerance and the forecast performance of stock market returns are constant, the linear combination rule means that, when the MA rule suggests an uptrend (downtrend), the rule suggests that the total weight should be allocated to stock markets (the risk-free rate).

The empirical findings suggest a low volatility anomaly that might be explained by investors' affection to high volatility, as suggested by Baker et al. (2011) and noted in Ang et al. (2009). On the other hand, the reported predictability of risk premia (see, for example, Cochrane 2008, and Fama 2014) can explain why, for instance, MA rules forecast better than using random highs and lows in the stock market (as noted in Jagannathan and Korajczyk 2017). The topic is important as Friesen and Sapp (2007), among others, report that mutual fund investors had negative outcomes, on average, in their timing to invest and withdraw cash from US mutual funds from 1991 to 2004. Munoz and Vicente (2018) report similar results with more recent data in US markets.

The plan of the remainder of the paper is as follows. Section 2 provides a literature review, and alternative model specifications are presented in Section 3. The empirical analysis is conducted in Section 4, while Section 5 gives some concluding comments.

## **2. Literature Review**

In efficient markets, investors earn above average returns only by taking above average risks (Malkiel 2003). Samuelson (1998) conforms with Fama (1972) by noting that market efficiency can be divided into micro and macro efficiency. The former concerns the relative pricing of individual stocks, and the latter, for markets as a whole. The CAPM by Sharpe (1964), and Lintner (1965) argues that beta is a proper definition for systematic risk for stock  $i$ , if unexplained changes in risk adjusted returns for the stock follow approximately normal distribution with zero mean.

Black (1972) states that the slope of the security market line (SML) is flatter if there exists restrictions in borrowing, that is, leverage constraints in the model. Starting from Black et al. (1972), many studies have reported that the security market line is too flat in US stocks compared with the SML suggested by the CAPM version of Sharpe and Lintner.

Ang et al. (2009), Baker et al. (2014), and Frazzini and Pedersen (2014) find that low-beta stocks outperform high-beta stocks statistically significantly. In fact, Frazzini and Pedersen report that significant excess profits in US stocks can be achieved by shorting high-beta stocks and buying low-beta stocks with leverage, but that leverage constraints make them disappear. Using Black (1972), investors often have leverage constraints, thereby making them place too much weight on risky stocks, which results in lower required return for high-beta stocks than would be justified by the Sharpe-Lintner CAPM.

Markowitz (1952) defines portfolio risk simply as the volatility of portfolio returns. Clarke et al. (2010) find that the volatility of stock returns contains potentially an additional risk factor with respect to systematic risk that can be defined in the betas of CAPM by Sharpe and Lintner. Moreover, Ang et al. (2009) report that the total volatility of international stock market returns is highly correlated with US stock returns, thereby suggesting a common risk factor for US stocks.

Baker et al. (2011) suggest that the low-volatility anomaly is due to investor irrational behaviour, mainly because an average fund manager seeks to beat the buy-and hold strategy by overinvesting in high-beta stocks. The explanations include preference for lotteries (Barberis and Huang 2008; Kumar 2009; Bali et al. 2011), overconfidence (Ben-David et al. 2013), and representativeness (Daniel and Titman 2006)), which means that people assess the probability of a state of the world based on how typical of that state the evidence seems to be (Kahneman and Tversky 1974).

Baker and Wurgler (2015) argue that the anomaly is also related to the limits of arbitrage. In fact, the extra costs of shorting prevents to take advantage of overpricing (Hong and Sraer 2016). More importantly, Li et al. (2016) report that the excess returns of low-beta portfolios are due to mispricing in US stocks, indicating that the low-volatility anomaly does not exist because of systematic risk by some rational, stock

specific volatility risk factor. They tested the low-volatility anomaly with monthly data from January 1963 to December 2011 in NYSE, NASDAQ, and AMEX stocks.

Market timing is closely related to technical trading rules. Brown and Jennings (1987) show theoretically that using past prices (like the MA rule in Gartley (1935)) has value for investors, if equilibrium prices are not fully revealing, and signals from past prices have some forecasting qualities. More importantly, Zhu and Zhou (2009) indicate that the MA rules are particularly useful for asset allocation purposes among risk averse investors, when markets are forecastable (quality of signal).

Moskowitz et al. (2012) argue that there are significant time series momentum (TSM) effects in financial markets that are not related to the cross-sectional momentum effect (Jegadeesh and Titman 1993). However, TSM is closely related to MA rules, since it gives a buy (sell) signal according to some historical price reference points, whereas MA rules give a buy (sell) signal, when the current price moves above (below) the historical average of the chosen calculated rolling window measure.

Starting from LeRoy (1973) and Lucas (1978), the literature in financial economics states that financial markets returns in efficient markets are partly forecastable, when investors are risk averse. This leads to the time-varying risk premia of investors, as noted by Fama (2014). For example, Campbell and Cochrane (1999) present a consumption-based model, which indicates that when the markets are in recession (boom), risk averse investors require larger (smaller) risk premium for risky assets. More importantly, Cochrane (2008) notes that the forecastability of excess returns may lead to successful market timing rules.

Brock et al. (1992) test different MA lag rules for US stock markets, and find that they gain profits compared with holding cash. On the other hand, Sullivan et al. (1999) find that MA rules do not outperform the buy-and-hold strategy, if transaction costs are accounted for. Allen and Karjalainen (1999) use a genetic algorithm to develop the best ex-ante technical trading rule model using US data, and find some evidence of outperforming the buy-and-hold strategy. Lo et al. (2000) find that risk averse investors benefit from technical trading rules because they reduce volatility of the portfolio without giving up much returns when compared against the buy-and-hold strategy.



More recently, Neely et al. (2014) use monthly data from January 1951 to December 2011, and report that MA rules forecast the risk premia in US stock markets statistically significantly. Marshall et al. (2017) find that MA rules give an earlier signal than TSM, suggesting better returns for MA rules, but they both work best with large market value stocks.

Moskowitz et al. (2012) use monthly data from January 1965 to December 2009, and report that TSM provides significant positive excess returns in futures markets. However, Kim et al. (2016) report that these positive excess returns produced by TSM are due to the volatility scaling factor used by Moskowitz et. al. (2012).

### 3. Model Specifications

Consider an overlapping generation economy with a continuum of young and old investors  $[0,1]$ . A young risk-averse investor  $j$  invests their initial wealth,  $w_t^j$ , in infinitely lived risky assets  $i = 1, 2, \dots, I$ , and in risk-free assets that produce the risk-free rate of return,  $r^f$ . A risky asset  $i$  pays dividend  $D_t^i$ , and has  $x_t^i$  outstanding. Assuming exogenous processes throughout, the aggregate dividend is  $D_t$ .

A young investor  $j$  maximizes their utility from old time consumption through optimal allocation of initial resources,  $w_t^j$ , between risky and risk-free assets:

$$\max x_t^j \left( \frac{E_t(P_{t+1} + D_{t+1})}{P_t} - (1 + r^f) \right) - \frac{\nu^j}{2} x_t^{j^2} \sigma^2$$

*s.t.*

$$x_t^j P_t \leq w_t^j$$

where  $E_t$  is the expectations operator,  $P_t$  is the price of one share of aggregate stock,  $\nu^j$  is a constant risk-aversion parameter for investor  $j$ ,  $\sigma^2$  is the variance of returns for the aggregate stock, and  $x_t^j$  is the demand of risky assets for an investor  $j$ . The first-order condition is:

$$\frac{E_t(P_{t+1} + D_{t+1})}{P_t} - (1 + r^f) - \nu^j x_t^j \sigma^2 = 0,$$

which results in optimal demand for the risky assets:

$$x_t^j = \frac{E_t((P_{t+1} + D_{t+1})/P_t) - (1 + r^f)}{\nu^j \sigma_i^2}. \quad (1)$$

Suppose that an investor  $j$  is a macro forecaster who allocates their initial wealth,  $w_t^j$ , between risky stocks and risk-free assets according to their forecast about the return of the risky alternative. Then, equation (1) says that the investor invests in the risky stocks only if the numerator on the right hand side is positive.

#### 4. Empirical Analysis

This section presents the empirical results from seven frequencies for the (MA) trend-chasing rules. The data consist of 29 companies included in the Dow Jones Industrial Average (DJIA) index in January 2018. The trading data (daily closing prices) cover 30 years from 1 January 1988 to 31 December 2017. Choosing the current DJIA companies for the last 30 years creates a “survivor bias” in the buy-and-hold results. However, this should not be an issue as we intend to compare the performance of the alternative MA frequency rules.

The rolling window is 200 trading days. The first rule is to calculate MA in every trading day; the second frequency takes into account every 5th trading day (thereby providing a proxy for the weekly rule); the third frequency takes into account every 20th trading day (proxy for the monthly rule); the fourth rule is to calculate MA for every 40th trading day (proxy for every other month); the fifth rule takes into account every 60th trading day (proxy for every third month); the sixth rule takes into account every 80th trading day (proxy for every fourth month); and the seventh rule takes into account every 100th trading day (proxy for every fifth month).

For the 29 DJIA companies, 26 of them have daily stock data available from 27 March 1987, thereby giving 4 January 1988 as the first trading day. The data for Cisco are available from 12 February 1990, for Goldman Sachs from 4 May 1999, and for Visa from 19 March 2008. There are 217 569 observations of daily returns from DJIA stocks. Thus, there are  $217569 \times 9 = 1\,958\,121$  daily returns for the first three frequencies (rules),  $217\,569 \times 4 = 870\,276$  daily returns for the fourth rule,  $217\,569 \times 3 = 652\,707$  daily returns for the fifth rule,  $217569 \times 2 = 435\,138$  daily returns for the sixth rule, and 217 569 daily returns for the seventh rule.

The trading rule for all cases is to use a simple crossover rule. When the trend-chasing MA turns lower (higher) than the current daily closing price, we invest the stock (three-month US Treasury Bills) at the closing price of the next trading day. Thus, the trading rule provides a market timing strategy where we invest all wealth either in stocks (separately, every stock included in DJIA), or to the risk-free asset (three-month U.S. Treasury bill), where the moving average rule advises the timing.

At the first frequency (every trading day), we calculate daily returns for MA200, MA180, MA160, MA140, MA120, MA100, MA80, MA60, and MA40. For example, MA200 is calculated as:

$$\left( \frac{P_{t-1} + P_{t-2} + \dots + P_{t-200}}{200} \right) = X_{t-1}.$$

At the lowest frequency, where every 100th daily observation is counted, MAC2 is calculated as:

$$\left( \frac{P_{t-1} + P_{t-100}}{2} \right) = X_{t-1}.$$

If  $X_{t-1} < P_{t-1}$ , we buy the stock at the closing price,  $P_t$ , thereby giving daily returns as

$$R_{t+1} = \ln\left(\frac{P_{t+1}}{P_t}\right).$$

Tables 1-7 in Appendix 1 show that the annualized average log returns of MA200 - MA40 are **+0.053** after transaction costs (with 0.1% per change of position). Recall that there are 200 closing day prices in the rolling window MA200, whereas MA40 means that there are 40 closing day prices in the window. The respective log returns for MAW40-MAW8 (weekly) are **+0.063**; for MA10 - MA2 (monthly) **+0.071**; for MAD5 - MAD2 (every other month) **+0.078**; for MAT4 - MAT2 (every third month) **+0.084**, for MAQ3 - MAQ2 (every fourth month) **+0.094**; and for MAC2 (every fifth month) **+0.088** after transaction costs.

Tables 1-7 show that, as the frequency decreases until every fourth month frequency (MAQ3 - MAQ2), average returns tend to increase, and decrease thereafter. In comparison, the biased buy-and-hold strategy produces **+0.117** with equal weights among all DJIA stocks, and with **0.295** annual volatility. A random investment (half the time in the risk-free rate, and half in the equally weighted portfolio from 4 January 1988) produces  $(0.117*0.5+0.022*0.5) = \mathbf{+0.070}$  annually, on average, with  $(1-\sqrt{0.5} = 0.293) = 29.3\%$  reduction in volatility, indicating **0.209** annual volatility for that portfolio.

The data are dividend excluded, but the average annual dividend yield in DJIA stocks over the last thirty years has been +0.026, so that the biased buy and hold strategy produces +0.143 annually with equal weights among DJIA stocks before taxes. Thus, the random investment strategy produces +0.083 annually, with survivor bias.

Appendix 1 (that is, the second column of Tables 1-7) also reports the annualized average log returns calculated in the largest sample (full 200 observations) in every category: MA200 **+0.065**; MAW40 **+0.073**; MA10 **+0.079**; MAD5 **+0.083**; MAT4 **+0.089**; MAQ3 **+0.091**; and MAC2 **+0.088** after transaction costs and before dividends. Adding +0.013 produces after dividends and before taxes: MA200 **+0.078**; MAW40 **+0.086**; MA10 **+0.092**; MAD5 **+0.096**; MAT4 **+0.102**; MAQ3 **+0.104**; and MAC2

**+0.101.** These results imply that starting from every fifth trading day frequency, a macro forecaster beats the buy and hold strategy **in returns**.

Figure 1 below illustrates the effects of frequency on the returns to volatility ratio (the second column in Tables 1-7).

**< Figure 1 goes here >**

In Figure 1, the straight line illustrates the return to volatility ratio of portfolios, where wealth is randomly invested in combinations of the three-month Treasury Bill (risk-free rate), with stocks included in the DJIA between 4 January 1988 and 31 December 2017. The red crosses represent the average return/volatility points calculated in the 200-day rolling window with the following frequencies: daily, every five days, every 20 days, every 40 days, every 60 days, every 80 days, and every 100 days (with only the most observations in each frequency giving 200, 40, 10, 5, 4, 3, and 2 observations). The red crosses plot a convex curve that deviates increasingly from the straight return to volatility ratio line, thereby symbolizing superior portfolio efficiency.

Tables 8-14 in Appendix 2 show that the annualized volatility of daily returns read, on average: MA200-MA40 **0.2044**; MAW40-MAW8 **0.205**; MA10-MA2 **0.2091**; MAD5-MAD2 **0.213**; MAT4-MAT2 **0.219**; MAQ3-MAQ2 **0.221**; and MAC2 **0.218**. Thus, there is virtually no difference between the MA frequencies, while the biased buy-and-hold strategy produces **0.295**.

Figure 1 presents the volatilities calculated in the largest sample (full 200 day rolling window in every category, the second column in Tables 8-14). They read MA200 **0.207**; MAW40 **0.208**; MA10 **0.211**; MAD5 **0.213**; MAT4 **0.218**; MAQ3 **0.215**; and MAC2 **0.218** after transaction costs. Investing randomly half of the time in the risk-free rate and the other half in the equally weighted portfolio, produces **0.209**. Thus, the difference between the annual volatilities produced in profitable market timing MA rules (MA10 – MAC2) and random market timing (half and half) ranges from **0.009** to **0.002**.

**< Figure 2 goes here >**

In Figure 2, the straight line again presents the return to volatility ratio of portfolios with random investment in the risk-free rate and the stocks in DJIA between 4 January 1988 and 31 December 2017. The red crosses plot the average return to volatility ratios, calculated by using a 200 day rolling window, with the following frequencies: daily, every five days, every 20 days, every 40 days, every 60 days, every 80 days, and every 100 days. The averages of every lag are reported in Tables 1-14, Appendices 1 and 2. Thus, all daily returns from Tables 1-14 are included.

Comparing Figures 1 and 2, it is clear that using the whole 200 daily observation windows in the MA rules produces more efficient results in market timing. That is, comparing the products of shorter and longer MA rule rolling windows, say, the last two monthly observations compared with ten monthly observations, average realized returns drop from **+0.079** to **+0.059** before dividends, while volatility remains approximately unchanged (from 0.211 to 0.207). This suggests that, in both cases, about half and half is invested in the equally-weighted DJIA portfolios and in the risk-free rate, and the MA rules advice the timing. More importantly, Tables 8-14 in Appendix 2 show that the range in volatilities with all MA rules varies between 0.202 – 0.227 (with 0.02 difference), whereas Tables 1-7 in Appendix 1 show that realized returns vary between 0.096 – 0.033 before dividends (with 0.063 difference).

These results indicate that a macro market timing with 200 days rolling window produces a reduction in volatility from **0.295** (the buy-and hold) to between 0.207-0.218, but the average annualized returns (dividends included) tend to rise as the MA frequency falls (+0.078 with all 200 observations to +0.104 with every fourth month observations). Thus, the results indicate that MA market timing finds long term stochastic trends more efficiently than short term stochastic trends.

The Sharpe ratio of random market timing (half and half) with dividends is **0.292**; for MA200 **0.271**; for MAW40 **0.308**; for MA10 **0.332**; for the MAD5 **0.347**; for MAT4 **0.370**; for MAQ3 **0.381**; and for MAC2 it is **0.362**.

< Figure 3 goes here >

Figure 3 shows that when the volatility changes 1% in the DJIA stocks, then the average returns change is 0.39%. Figures 1 and 2 suggest that the theoretical change should be such that when the volatility changes 1%, then the average returns change is 0.50%, suggesting a flatter SML line in the data. This suggests strongly that DJIA investors have overweight high-beta stocks in the last 30 years.

It is obvious that transaction costs are crucial in MA performance. In the above calculations, the transaction costs are 0.1% per transaction from current wealth. Tables 15 and 16 in Appendix 3 report the transaction costs for the MA200-MA40 and MA10-MA2 rules. In the MA200-MA40 rules, the average annualized transaction costs are **0.0133**, such that the rules have about 13 changes in positions per year. Meanwhile, for the MA10-MA2 rules, the average annualized transaction costs are **0.0032**, suggesting about 3 changes in positions per year.

Allen and Karjalainen (1999) give reasons for a cost of 0.2% per transaction in their sample, but since technological progress has reduced transaction costs since the mid-nineties, 0.1% per transaction should be fair, on average. Nevertheless, a trial with 0.2% transaction costs shows that, for example, the average annualized daily returns become 0.0403 for the MA200-MA40 rules, and 0.0674 for the MA10-MA2 rules. Note that the returns grow 67%, on average, for the MA10-MA2 rules (with about the same volatility) compared with costs of 0.1% per transaction.

Note that the model prohibits short selling since we have only long positions in stocks or investing in the risk-free rate. Then the limits of arbitrage argument of Baker et al. (2015) are consistent with our results.

## **5. Concluding Remarks**

The analysis suggests that a macro forecaster can obtain higher returns with equal volatility (30 % below that of the buy-and-hold strategy) by reducing the frequency used in MA rules. The return to volatility ratio for risk-averse investors with MA market timing significantly outperforms the random benchmark strategy, when the frequency in

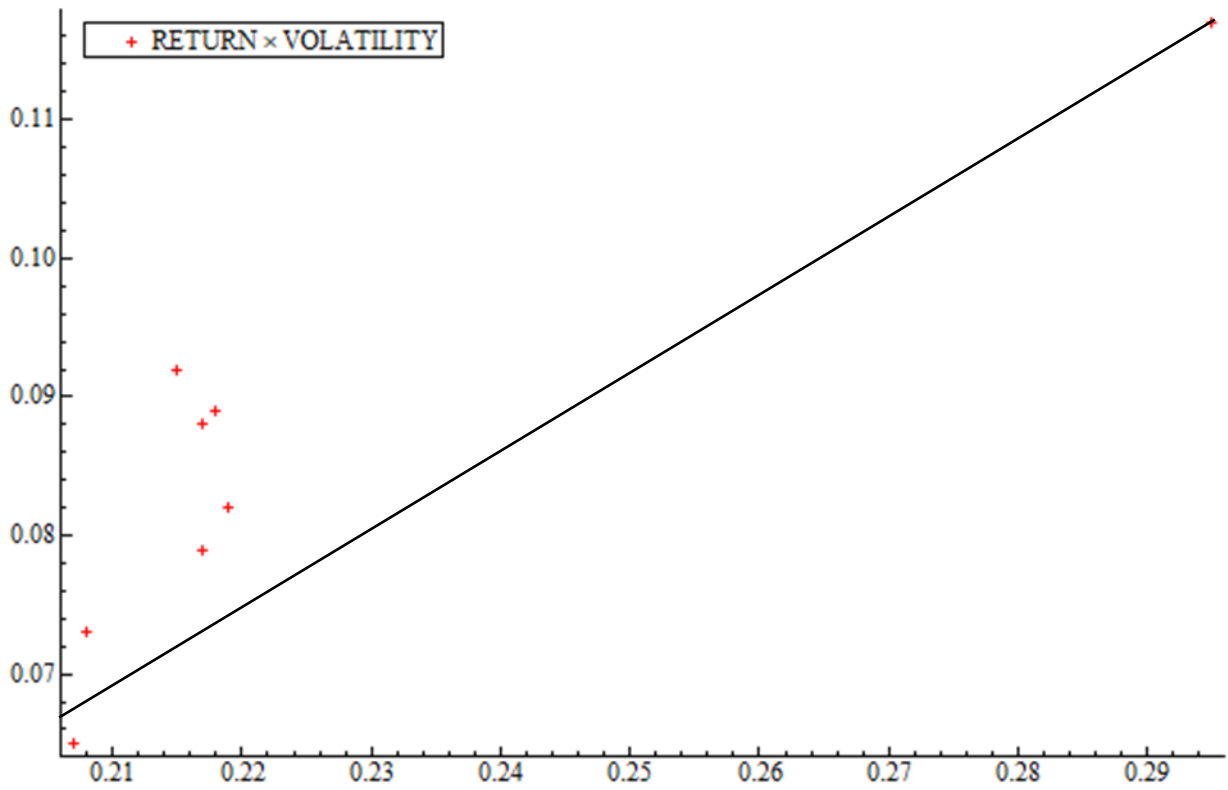
the MA rules is reduced. This indicates that the forecasts are more accurate the longer is the time frame.

The results suggest that a flatter SML in the CAPM can be followed by the irrational preference of investors in high-beta stocks, as suggested by Baker et al. (2011) and Li et al. (2016), since the empirically efficient frontier of portfolios becomes flatter than the theoretically efficient SML (random timing) (see Figure 1). In other words, the empirical results suggests that market timing with the few past observations (for example, every fourth month) in the past 200 rolling window daily prices, have produced significantly better returns to risk ratio for the portfolio of DJIA equally weighted stocks in the past 30 years than random timing. The finding points to the low-volatility anomaly.

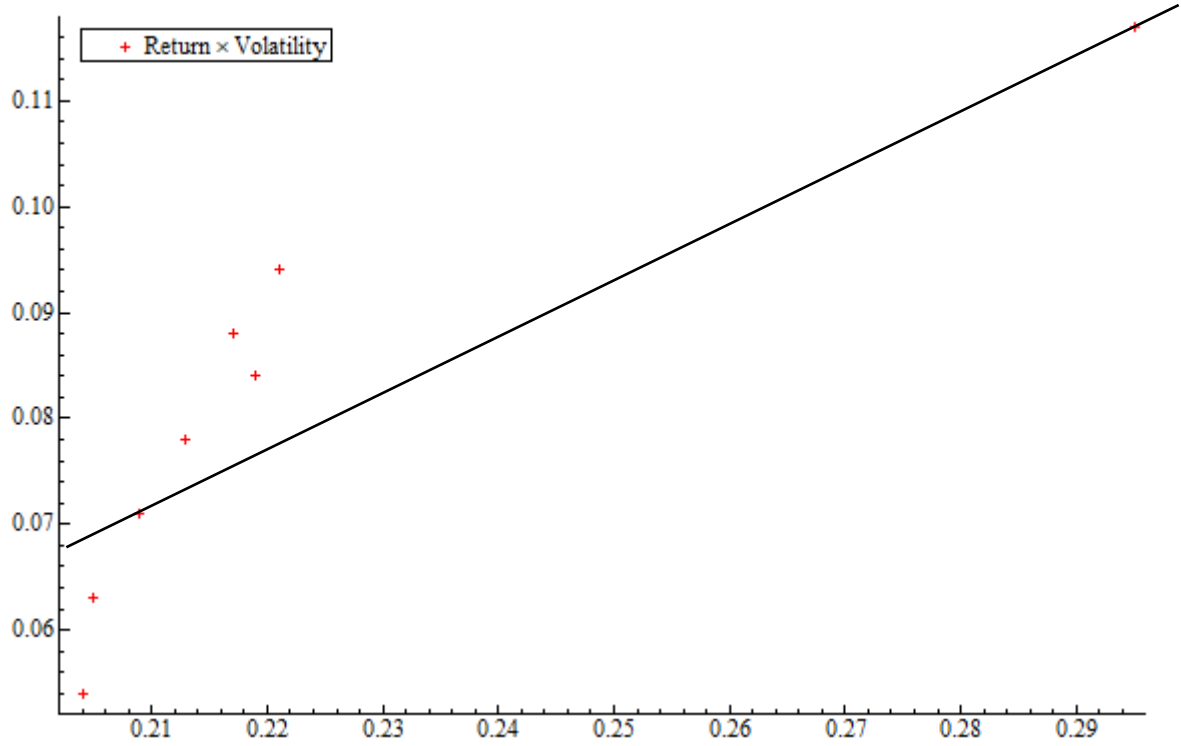
One explanation for the results is that they are due to time-varying risk premiums. This is emphasized by Neely et al. (2014), who claim that MA rules, in effect, forecast changes in the risk premium. If the results are rational products of time-varying risk premiums, the results suggest that investor sensitivity to risk must be extremely high, and their risk premium is larger (smaller) in downs (ups), as suggested by Campbell and Cochrane (1999). As volatility rises (decreases), usually in downs (ups), the results suggest that when volatility is high, investors as a group tolerate significantly more risk (that is, volatility) than in calmer periods.

Consider the following numerical example: Assume that the risk premium is 0.08 in volatile downs, and 0.04 in calm ups, and the variance of returns is 0.03 in downs and 0.09 in ups. Then the risk aversion coefficient must be 0.89 in volatile down periods, and 1.33 in calm up periods. As market timing with MA rules works better in longer periods with few observations, it seems to be more accurate in longer stochastic (up or down) trends.

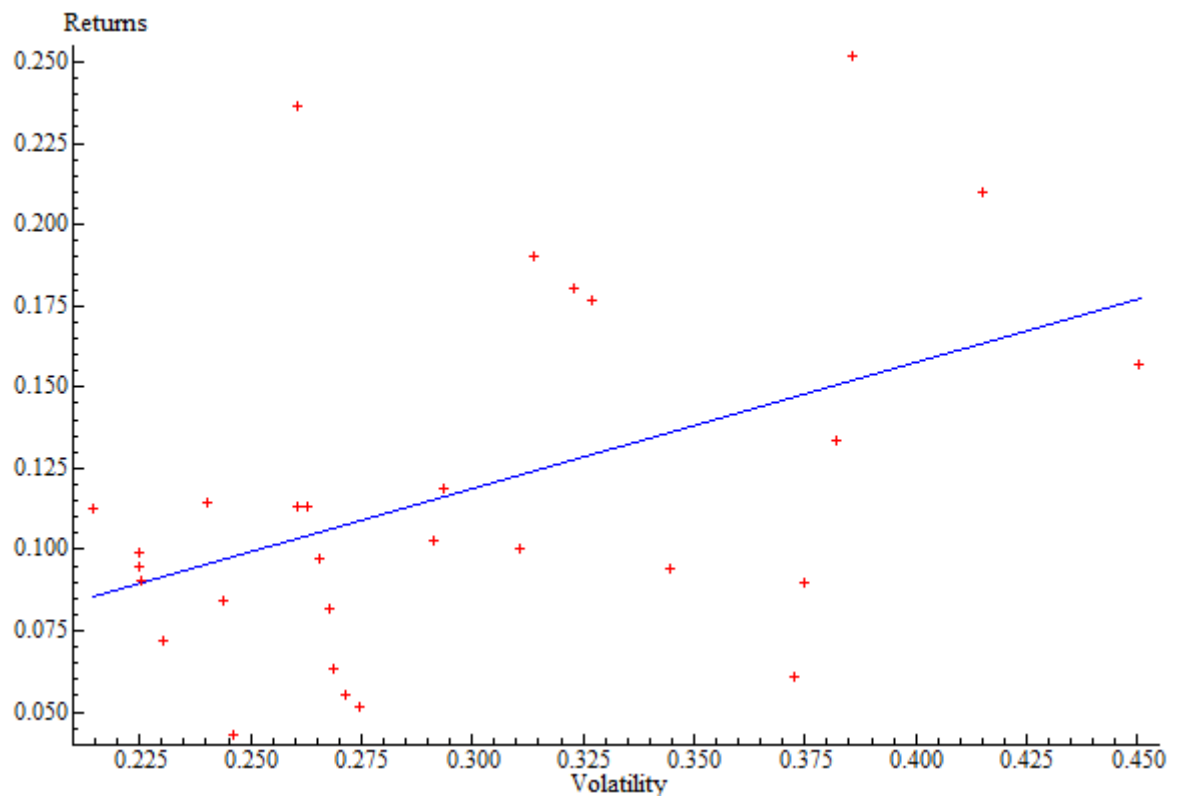




**Figure 1: Returns to volatility ratio in MA200, MAW40, MA10, MAD5, MAT4, MAQ3, MAC2, and the theoretical random timing efficient SML**



**Figure 2: Returns to volatility ratio in MA200-MA40, MAW40-MAW8, MA10-MA2, MAD5-MAD2, MAT4-MAT2, MAQ3-MAQ2, MAC2, and the theoretical random timing efficient SML**



**Figure 3: Returns to the volatility ratio in current DJIA stocks in annual averages from 4 January 1988 to 31 December 2017**

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## **Appendix 1**

**Table 1: Annualized daily returns of MA40-MA200, average annualized returns**

	<b>Buy &amp; Hold</b>	<b>MA200</b>	<b>MA180</b>	<b>MA160</b>	<b>MA140</b>	<b>MA120</b>	<b>MA100</b>	<b>MA80</b>	<b>MA60</b>	<b>MA40</b>
<b>3M</b>	0.090	0.042	0.034	0.017	0.015	0.019	0.014	0.006	-0.009	6E-04
<b>American Express</b>	0.094	0.035	0.037	0.039	0.055	0.039	0.042	0.043	0.041	0.008
<b>Apple</b>	0.157	0.147	0.145	0.147	0.142	0.156	0.149	0.150	0.146	0.164
<b>Boeing</b>	0.119	0.088	0.089	0.060	0.055	0.061	0.061	0.058	0.046	0.048
<b>Caterpillar</b>	0.100	0.075	0.079	0.058	0.058	0.049	0.034	0.028	0.039	0.025
<b>Chevron</b>	0.084	0.005	0.013	0.002	-0.000	-0.000	0.003	-0.01	-0.025	-0.05
<b>Coca-Cola</b>	0.099	0.058	0.055	0.030	0.035	0.039	0.027	0.023	0.009	0.003
<b>Walt Disney</b>	0.103	0.072	0.078	0.079	0.074	0.077	0.074	0.076	0.056	0.048
<b>Exxon</b>	0.072	-0.011	-0.010	-0.020	-0.030	-0.020	-0.025	-0.01	-0.044	-0.05
<b>GE</b>	0.052	0.072	0.071	0.058	0.039	0.039	0.033	0.018	0.013	9E-04
<b>Home Depot</b>	0.190	0.125	0.116	0.102	0.092	0.087	0.076	0.067	0.068	0.058
<b>IBM</b>	0.055	0.016	0.029	0.033	0.028	0.016	0.021	0.031	0.029	0.048
<b>Intel</b>	0.134	0.083	0.082	0.083	0.073	0.091	0.082	0.080	0.077	0.078
<b>Johnson &amp; Johnson</b>	0.113	0.062	0.058	0.053	0.042	0.032	0.044	0.028	0.008	-0.00
<b>JP Morgan</b>	0.090	0.013	0.014	0.003	0.010	0.017	0.013	0.031	0.038	0.025
<b>McDonalds</b>	0.114	0.047	0.048	0.040	0.044	0.040	0.035	0.043	0.030	0.018
<b>Merck</b>	0.063	0.050	0.048	0.044	0.032	0.033	0.029	0.022	0.016	-0.02
<b>Microsoft</b>	0.180	0.117	0.128	0.105	0.102	0.104	0.095	0.090	0.070	0.062
<b>Nike</b>	0.177	0.087	0.093	0.085	0.102	0.108	0.107	0.119	0.133	0.112
<b>Pfizer</b>	0.097	0.059	0.056	0.043	0.042	0.052	0.044	0.040	0.024	0.009
<b>Procter &amp; Gamble</b>	0.095	0.037	0.045	0.037	0.036	0.037	0.029	0.023	0.004	0.017
<b>Travellers</b>	0.082	0.036	0.035	0.038	0.029	0.008	-0.004	-9E-04	-0.001	0.006
<b>United Technologies</b>	0.113	0.051	0.057	0.046	0.059	0.057	0.049	0.049	0.041	0.017
<b>United Health Group</b>	0.252	0.181	0.182	0.157	0.147	0.136	0.130	0.118	0.125	0.076
<b>Verizon</b>	0.043	-0.017	-0.020	-0.010	-0.000	-0.020	-0.020	-0.02	-0.029	-0.02
<b>Wal-Mart</b>	0.113	0.019	0.016	0.010	0.012	0.012	0.016	0.012	0.020	0.024
<b>Cisco</b>	0.210	0.198	0.194	0.210	0.208	0.198	0.205	0.152	0.096	0.085
<b>Goldman Sachs</b>	0.061	0.038	0.029	0.033	0.038	0.050	0.057	0.078	0.076	0.063



<b>Visa</b>	0.236	0.112	0.118	0.129	0.141	0.128	0.132	0.120	0.094	0.085	
<b>Average</b>	<b>0.117</b>	<b>0.065</b>	<b>0.066</b>	<b>0.059</b>	<b>0.058</b>	<b>0.057</b>	<b>0.053</b>	<b>0.05</b>	<b>0.041</b>	<b>0.033</b>	<b>0.054</b>

**Table 2: Annualized daily (every fifth trading day) returns of MAW8-MAW40  
(W = number of weeks), average annualized returns**

	<b>Buy&amp;Hold</b>	<b>MAW40</b>	<b>MAW36</b>	<b>MAW32</b>	<b>MAW28</b>	<b>MAW24</b>	<b>MAW20</b>	<b>MAW16</b>	<b>MAW12</b>	<b>MAW8</b>	
<b>3M</b>	0.090	0.035	0.033	0.020	0.021	0.019	0.012	0.019	0.032	0.026	
<b>American Express</b>	0.094	0.058	0.053	0.062	0.063	0.047	0.046	0.035	0.034	0.015	
<b>Apple</b>	0.157	0.130	0.137	0.143	0.131	0.134	0.131	0.188	0.174	0.144	
<b>Boeing</b>	0.119	0.089	0.079	0.075	0.074	0.080	0.082	0.066	0.074	0.076	
<b>Caterpillar</b>	0.100	0.057	0.062	0.058	0.058	0.061	0.054	0.049	0.043	0.023	
<b>Chevron</b>	0.084	0.005	0.015	3E-04	0.004	0.008	0.009	0.004	0.004	-0.03	
<b>Coca-Cola</b>	0.099	0.055	0.054	0.054	0.041	0.054	0.047	0.047	0.029	0.011	
<b>Walt Disney</b>	0.103	0.071	0.073	0.062	0.080	0.076	0.080	0.078	0.065	0.051	
<b>Exxon</b>	0.072	0.018	0.016	0.007	0.008	0.010	0.013	0.020	0.011	0.005	
<b>GE</b>	0.052	0.061	0.046	0.047	0.047	0.045	0.023	0.018	0.031	0.023	
<b>Home Depot</b>	0.190	0.135	0.133	0.124	0.112	0.110	0.088	0.076	0.096	0.077	
<b>IBM</b>	0.055	0.020	0.037	0.044	0.040	0.051	0.027	0.028	0.008	0.016	
<b>Intel</b>	0.134	0.088	0.091	0.075	0.061	0.075	0.073	0.070	0.076	0.085	
<b>Johnson &amp; Johnson</b>	0.113	0.074	0.079	0.071	0.059	0.050	0.050	0.048	0.042	0.027	
<b>JP Morgan</b>	0.090	0.040	0.036	0.027	0.033	0.033	0.048	0.051	0.042	0.020	
<b>McDonalds</b>	0.114	0.086	0.068	0.059	0.058	0.052	0.052	0.059	0.058	0.044	
<b>Merck</b>	0.063	0.051	0.039	0.029	0.034	0.034	0.030	0.033	0.024	0.029	
<b>Microsoft</b>	0.180	0.128	0.125	0.116	0.116	0.116	0.105	0.099	0.062	0.078	
<b>Nike</b>	0.177	0.087	0.091	0.098	0.093	0.087	0.094	0.102	0.119	0.091	
<b>Pfizer</b>	0.097	0.070	0.061	0.057	0.053	0.063	0.049	0.050	0.044	0.050	
<b>Procter &amp; Gamble</b>	0.095	0.050	0.044	0.050	0.051	0.040	0.043	0.042	0.031	0.033	
<b>Travellers</b>	0.082	0.020	0.006	0.010	0.014	0.006	0.005	0.008	0.017	0.015	
<b>United Technologies</b>	0.113	0.071	0.077	0.062	0.072	0.071	0.056	0.061	0.051	0.053	
<b>United Health Group</b>	0.252	0.171	0.133	0.130	0.151	0.124	0.134	0.123	0.113	0.087	
<b>Verizon</b>	0.043	-0.00	-0.01	0.002	0.006	-0.01	-0.01	-0.01	-0.009	-0.00	
<b>Wal-Mart</b>	0.113	0.050	0.049	0.045	0.038	0.028	0.033	0.026	0.038	0.029	
<b>Cisco</b>	0.210	0.209	0.211	0.219	0.222	0.219	0.204	0.164	0.120	0.094	
<b>Goldman Sachs</b>	0.061	0.050	0.030	0.031	0.040	0.036	0.071	0.089	0.078	0.077	
<b>Visa</b>	0.236	0.143	0.142	0.131	0.171	0.167	0.159	0.113	0.119	0.080	
<b>Average</b>	<b>0.117</b>	<b>0.073</b>	<b>0.069</b>	<b>0.066</b>	<b>0.067</b>	<b>0.065</b>	<b>0.062</b>	<b>0.061</b>	<b>0.056</b>	<b>0.046</b>	<b>0.063</b>



**Table 3: Annualized daily (every 20s trading day) returns of MA2-MA10, average annualized returns**

	<b>Buy and Hold</b>	<b>MA10</b>	<b>MA9</b>	<b>MA8</b>	<b>MA7</b>	<b>MA6</b>	<b>MA5</b>	<b>MA4</b>	<b>MA3</b>	<b>MA2</b>	
<b>3M</b>	0.090	0.033	0.035	0.023	0.023	0.024	0.023	0.038	0.021	0.012	
<b>American Express</b>	0.094	0.086	0.087	0.091	0.107	0.088	0.062	0.062	0.036	0.038	
<b>Apple</b>	0.157	0.057	0.069	0.056	0.076	0.076	0.094	0.069	0.099	0.071	
<b>Boeing</b>	0.119	0.122	0.122	0.102	0.099	0.115	0.110	0.100	0.091	0.077	
<b>Caterpillar</b>	0.100	0.065	0.062	0.071	0.083	0.081	0.063	0.057	0.009	0.051	
<b>Chevron</b>	0.084	0.022	0.021	0.025	0.026	0.019	0.032	0.032	0.013	0.005	
<b>Coca-Cola</b>	0.099	0.083	0.072	0.087	0.071	0.073	0.072	0.069	0.046	0.026	
<b>Walt Disney</b>	0.103	0.061	0.066	0.073	0.077	0.071	0.079	0.081	0.073	0.057	
<b>Exxon</b>	0.072	0.040	0.038	0.028	0.028	0.034	0.020	0.027	0.025	0.026	
<b>GE</b>	0.052	0.079	0.078	0.080	0.072	0.070	0.063	0.018	0.038	0.037	
<b>Home Depot</b>	0.190	0.126	0.133	0.134	0.136	0.120	0.14	0.119	0.118	0.110	
<b>IBM</b>	0.055	0.029	0.033	0.032	0.038	0.036	0.026	0.033	0.026	0.03	
<b>Intel</b>	0.134	0.079	0.080	0.096	0.095	0.085	0.063	0.082	0.110	0.116	
<b>Johnson &amp; Johnson</b>	0.113	0.078	0.076	0.071	0.059	0.057	0.058	0.050	0.052	0.031	
<b>JP Morgan</b>	0.090	0.057	0.051	0.051	0.063	0.046	0.070	0.079	0.067	0.067	
<b>McDonalds</b>	0.114	0.077	0.077	0.057	0.055	0.045	0.056	0.042	0.045	0.033	
<b>Merck</b>	0.063	0.069	0.069	0.054	0.059	0.05	0.045	0.027	0.011	3E-04	
<b>Microsoft</b>	0.180	0.122	0.127	0.123	0.099	0.112	0.093	0.095	0.090	0.108	
<b>Nike</b>	0.177	0.128	0.136	0.130	0.127	0.115	0.111	0.109	0.082	0.089	
<b>Pfizer</b>	0.097	0.070	0.069	0.067	0.068	0.066	0.068	0.056	0.040	0.034	
<b>Procter &amp; Gamble</b>	0.095	0.057	0.060	0.055	0.042	0.043	0.021	0.024	0.038	0.039	
<b>Travellers</b>	0.082	0.045	0.049	0.047	0.041	0.034	0.016	0.009	0.002	0.017	
<b>United Technologies</b>	0.113	0.064	0.062	0.074	0.078	0.063	0.046	0.037	0.050	0.050	
<b>United Health Group</b>	0.252	0.158	0.162	0.167	0.154	0.168	0.176	0.174	0.180	0.158	
<b>Verizon</b>	0.043	0.002	9E-04	0.011	0.017	0.025	-0.00	0.01	-0.00	-0.02	
<b>Wal-Mart</b>	0.113	0.046	0.046	0.040	0.044	0.032	0.041	0.037	0.023	0.038	
<b>Cisco</b>	0.210	0.228	0.227	0.222	0.221	0.191	0.186	0.184	0.160	0.134	
<b>Goldman Sachs</b>	0.061	0.029	0.030	0.020	0.052	0.067	0.065	0.070	0.041	0.068	
<b>Visa</b>	0.236	0.171	0.161	0.162	0.149	0.122	0.113	0.115	0.142	0.097	
<b>Average</b>	<b>0.117</b>	<b>0.079</b>	<b>0.079</b>	<b>0.078</b>	<b>0.078</b>	<b>0.073</b>	<b>0.069</b>	<b>0.066</b>	<b>0.059</b>	<b>0.055</b>	<b>0.071</b>

**Table 4: Annualized daily (every other month) returns of MAD2-MAD2 (D = every other month, and 5,4,3,2 are the numbers of observations in the rolling window), average annualized returns**

	<b>Buy&amp;Hold</b>	<b>MAD5</b>	<b>MAD4</b>	<b>MAD3</b>	<b>MAD2</b>	
<b>3M</b>	0.090	0.062	0.063	0.042	0.049	
<b>American Express</b>	0.094	0.089	0.098	0.052	0.041	
<b>Apple</b>	0.157	0.040	0.042	0.030	0.085	
<b>Boeing</b>	0.119	0.112	0.110	0.102	0.110	
<b>Caterpillar</b>	0.100	0.079	0.09	0.089	0.084	
<b>Chevron</b>	0.084	0.033	0.036	0.026	0.028	
<b>Coca-Cola</b>	0.099	0.093	0.102	0.080	0.078	
<b>Walt Disney</b>	0.103	0.068	0.074	0.080	0.084	
<b>Exxon</b>	0.072	0.022	0.018	0.010	0.009	
<b>GE</b>	0.052	0.067	0.066	0.041	0.033	
<b>Home Depot</b>	0.190	0.174	0.175	0.156	0.160	
<b>IBM</b>	0.055	0.016	0.023	0.017	0.021	
<b>Intel</b>	0.134	0.093	0.098	0.089	0.112	
<b>Johnson &amp; Johnson</b>	0.113	0.083	0.086	0.048	0.071	
<b>JP Morgan</b>	0.090	0.053	0.052	0.048	0.054	
<b>McDonalds</b>	0.114	0.094	0.098	0.071	0.070	
<b>Merck</b>	0.063	0.084	0.067	0.036	0.031	
<b>Microsoft</b>	0.180	0.138	0.136	0.106	0.088	
<b>Nike</b>	0.177	0.140	0.144	0.133	0.122	
<b>Pfizer</b>	0.097	0.062	0.051	0.061	0.059	
<b>Procter &amp; Gamble</b>	0.095	0.048	0.054	0.048	0.034	
<b>Travellers</b>	0.082	0.018	0.015	0.018	2E-04	
<b>United Technologies</b>	0.113	0.066	0.073	0.096	0.060	
<b>United Health Group</b>	0.252	0.181	0.179	0.191	0.207	
<b>Verizon</b>	0.043	-0.018	-0.01	-0.02	-0.02	
<b>Wal-Mart</b>	0.113	0.067	0.065	0.050	0.061	
<b>Cisco</b>	0.210	0.217	0.226	0.207	0.196	
<b>Goldman Sachs</b>	0.061	0.041	0.059	0.060	0.039	
<b>Visa</b>	0.236	0.174	0.173	0.151	0.120	
<b>Average</b>	<b>0.117</b>	<b>0.083</b>	<b>0.085</b>	<b>0.073</b>	<b>0.072</b>	<b>0.078</b>

**Table 5: Annualized daily (every third month) returns of MAT2-MAT4 (T = every third month, and 4,3,2 are the numbers of observations in the rolling window), average annualized returns**

	<b>Buy&amp;Hold</b>	<b>MAT4</b>	<b>MAT3</b>	<b>MAT2</b>	
<b>3M</b>	0.090	0.061	0.055	0.039	
<b>American Express</b>	0.094	0.113	0.091	0.066	
<b>Apple</b>	0.157	0.089	0.073	0.096	
<b>Boeing</b>	0.119	0.127	0.131	0.114	
<b>Caterpillar</b>	0.100	0.070	0.069	0.078	
<b>Chevron</b>	0.084	0.047	0.053	0.037	
<b>Coca-Cola</b>	0.099	0.077	0.078	0.072	
<b>Walt Disney</b>	0.103	0.043	0.042	0.068	
<b>Exxon</b>	0.072	0.055	0.049	0.037	
<b>GE</b>	0.052	0.084	0.080	0.047	
<b>Home Depot</b>	0.190	0.161	0.163	0.128	
<b>IBM</b>	0.055	0.054	0.048	0.028	
<b>Intel</b>	0.134	0.107	0.115	0.072	
<b>Johnson &amp; Johnson</b>	0.113	0.094	0.094	0.074	
<b>JP Morgan</b>	0.090	0.058	0.076	0.007	
<b>McDonalds</b>	0.114	0.080	0.082	0.069	
<b>Merck</b>	0.063	0.062	0.062	0.049	
<b>Microsoft</b>	0.180	0.127	0.128	0.080	
<b>Nike</b>	0.177	0.146	0.151	0.099	
<b>Pfizer</b>	0.097	0.078	0.070	0.056	
<b>Procter &amp; Gamble</b>	0.095	0.068	0.072	0.076	
<b>Travellers</b>	0.082	0.041	0.043	0.025	
<b>United Technologies</b>	0.113	0.077	0.089	0.079	
<b>United Health Group</b>	0.252	0.147	0.161	0.178	
<b>Verizon</b>	0.043	-0.00	-0.00	-0.02	
<b>Wal-Mart</b>	0.113	0.081	0.081	0.083	
<b>Cisco</b>	0.210	0.211	0.217	0.213	
<b>Goldman Sachs</b>	0.061	0.044	0.026	0.030	
<b>Visa</b>	0.236	0.183	0.199	0.177	
<b>Average</b>	<b>0.117</b>	<b>0.089</b>	<b>0.089</b>	<b>0.075</b>	<b>0.084</b>

**Table 6: Annualized daily (every fourth month) returns of MAQ2-MAQ3  
(Q = every fourth month, and 3 and 2 are the numbers of observations  
in the rolling window), average annualized returns**

	<b>Buy&amp;Hold</b>	<b>MAQ3</b>	<b>MAQ2</b>	
<b>3M</b>	0.090	0.056	0.058	
<b>American Express</b>	0.094	0.089	0.094	
<b>Apple</b>	0.157	0.094	0.094	
<b>Boeing</b>	0.119	0.122	0.128	
<b>Caterpillar</b>	0.100	0.064	0.084	
<b>Chevron</b>	0.084	0.060	0.054	
<b>Coca-Cola</b>	0.099	0.083	0.093	
<b>Walt Disney</b>	0.103	0.061	0.062	
<b>Exxon</b>	0.072	0.056	0.064	
<b>GE</b>	0.052	0.069	0.081	
<b>Home Depot</b>	0.190	0.152	0.157	
<b>IBM</b>	0.055	0.048	0.031	
<b>Intel</b>	0.134	0.064	0.070	
<b>Johnson &amp; Johnson</b>	0.113	0.080	0.079	
<b>JP Morgan</b>	0.090	0.085	0.091	
<b>McDonalds</b>	0.114	0.096	0.112	
<b>Merck</b>	0.063	0.056	0.061	
<b>Microsoft</b>	0.180	0.143	0.145	
<b>Nike</b>	0.177	0.181	0.199	
<b>Pfizer</b>	0.097	0.059	0.045	
<b>Procter &amp; Gamble</b>	0.095	0.073	0.077	
<b>Travellers</b>	0.082	0.051	0.051	
<b>United Technologies</b>	0.113	0.080	0.077	
<b>United Health Group</b>	0.252	0.185	0.218	
<b>Verizon</b>	0.043	0.027	0.023	
<b>Wal-Mart</b>	0.113	0.087	0.076	
<b>Cisco</b>	0.210	0.195	0.180	
<b>Goldman Sachs</b>	0.061	0.042	0.056	
<b>Visa</b>	0.236	0.195	0.228	
<b>Average</b>	<b>0.117</b>	<b>0.091</b>	<b>0.096</b>	<b>0.094</b>

**Table 7: Annualized daily (every fifth month) returns of MAC2 (C = every fifth month, and 2 = observations accounting in the rolling window), average annualized returns**

	<b>Buy &amp; Hold</b>	<b>MAC2</b>
<b>3M</b>	0.090	0.076
<b>American Express</b>	0.094	0.088
<b>Apple</b>	0.157	0.132
<b>Boeing</b>	0.119	0.080
<b>Caterpillar</b>	0.100	0.094
<b>Chevron</b>	0.084	0.047
<b>Coca-Cola</b>	0.099	0.094
<b>Walt Disney</b>	0.103	0.044
<b>Exxon</b>	0.072	0.049
<b>GE</b>	0.052	0.048
<b>Home Depot</b>	0.190	0.143
<b>IBM</b>	0.055	0.032
<b>Intel</b>	0.133	0.057
<b>Johnson &amp; Johnson</b>	0.113	0.081
<b>JP Morgan</b>	0.090	0.045
<b>McDonalds</b>	0.114	0.079
<b>Merck</b>	0.063	0.080
<b>Microsoft</b>	0.180	0.094
<b>Nike</b>	0.177	0.141
<b>Pfizer</b>	0.097	0.099
<b>Procter &amp; Gamble</b>	0.095	0.039
<b>Travellers</b>	0.082	0.068
<b>United Technologies</b>	0.113	0.056
<b>United Health Group</b>	0.252	0.152
<b>Verizon</b>	0.043	0.048
<b>Wal-Mart</b>	0.113	0.093
<b>Cisco</b>	0.210	0.225
<b>Goldman Sachs</b>	0.061	0.053
<b>Visa</b>	0.236	0.217
<b>Average</b>	<b>0.117</b>	<b>0.088</b>

## Appendix 2

**Table 8: Annualized daily volatility of MA40-MA200, average annualized volatility**

	Buy & Hold	MA200	MA180	MA160	MA140	MA120	MA100	MA80	MA60	MA40	
3M	0.225	0.164	0.165	0.161	0.161	0.159	0.159	0.158	0.158	0.157	
American Express	0.345	0.227	0.228	0.221	0.225	0.224	0.225	0.224	0.228	0.229	
Apple	0.451	0.317	0.321	0.315	0.315	0.313	0.315	0.315	0.310	0.305	
Boeing	0.294	0.201	0.203	0.199	0.201	0.199	0.198	0.198	0.201	0.204	
Caterpillar	0.311	0.216	0.218	0.216	0.216	0.214	0.215	0.214	0.213	0.215	
Chevron	0.244	0.167	0.168	0.166	0.166	0.166	0.165	0.164	0.167	0.168	
Coca-Cola	0.225	0.164	0.166	0.161	0.160	0.159	0.158	0.158	0.156	0.155	
Walt Disney	0.291	0.196	0.201	0.199	0.200	0.199	0.198	0.203	0.204	0.203	
Exxon	0.230	0.162	0.163	0.159	0.159	0.159	0.157	0.156	0.155	0.157	
GE	0.275	0.174	0.175	0.172	0.173	0.173	0.171	0.168	0.168	0.168	
Home Depot	0.314	0.226	0.228	0.223	0.221	0.221	0.219	0.217	0.217	0.214	
IBM	0.271	0.187	0.189	0.185	0.184	0.181	0.179	0.177	0.176	0.174	
Intel	0.382	0.273	0.275	0.267	0.265	0.263	0.260	0.257	0.256	0.254	
Johnson & Johnson	0.215	0.163	0.164	0.161	0.159	0.157	0.155	0.153	0.152	0.149	
JP Morgan	0.375	0.223	0.226	0.223	0.224	0.227	0.237	0.242	0.245	0.248	
McDonalds	0.240	0.183	0.184	0.18	0.178	0.177	0.176	0.176	0.175	0.174	
Merck	0.269	0.177	0.179	0.173	0.173	0.174	0.172	0.174	0.174	0.177	
Microsoft	0.323	0.248	0.249	0.243	0.241	0.237	0.236	0.233	0.232	0.231	
Nike	0.327	0.243	0.245	0.238	0.236	0.235	0.235	0.232	0.232	0.233	
Pfizer	0.266	0.188	0.19	0.187	0.186	0.187	0.186	0.187	0.187	0.187	
Procter & Gamble	0.225	0.169	0.169	0.164	0.163	0.161	0.158	0.157	0.156	0.156	
Travellers	0.268	0.174	0.175	0.174	0.175	0.178	0.180	0.184	0.182	0.185	
United Technologies	0.261	0.179	0.181	0.179	0.178	0.177	0.177	0.177	0.176	0.173	
United Health Group	0.386	0.290	0.293	0.290	0.290	0.283	0.282	0.282	0.280	0.273	
Verizon	0.246	0.163	0.165	0.164	0.163	0.163	0.163	0.161	0.161	0.163	
Wal-Mart	0.263	0.203	0.204	0.200	0.198	0.195	0.191	0.19	0.189	0.191	
Cisco	0.415	0.300	0.302	0.297	0.295	0.291	0.290	0.285	0.282	0.275	
Goldman Sachs	0.373	0.222	0.226	0.22	0.222	0.223	0.228	0.230	0.227	0.229	
Visa	0.260	0.209	0.212	0.209	0.208	0.212	0.208	0.206	0.205	0.197	
<b>Average</b>	<b>0.295</b>	<b>0.207</b>	<b>0.209</b>	<b>0.205</b>	<b>0.205</b>	<b>0.204</b>	<b>0.203</b>	<b>0.203</b>	<b>0.202</b>	<b>0.202</b>	<b>0.204</b>



**Table 9: Annualized daily (every fifth trading day) volatility of MAW8-MAW40****(W = number of weeks), average annualized volatility**

	<b>Buy&amp;Hold</b>	<b>MAW40</b>	<b>MAW36</b>	<b>MAW32</b>	<b>MAW28</b>	<b>MAW24</b>	<b>MAW20</b>	<b>MAW16</b>	<b>MAW12</b>	<b>MAW8</b>	
<b>3M</b>	0.225	0.165	0.165	0.163	0.163	0.16	0.159	0.157	0.157	0.159	
<b>American Express</b>	0.345	0.227	0.224	0.224	0.227	0.225	0.223	0.228	0.232	0.234	
<b>Apple</b>	0.451	0.316	0.316	0.313	0.318	0.316	0.343	0.317	0.312	0.309	
<b>Boeing</b>	0.294	0.204	0.203	0.204	0.204	0.203	0.203	0.201	0.201	0.206	
<b>Caterpillar</b>	0.311	0.216	0.215	0.215	0.217	0.214	0.215	0.215	0.213	0.214	
<b>Chevron</b>	0.244	0.169	0.168	0.169	0.168	0.168	0.167	0.166	0.168	0.172	
<b>Coca-Cola</b>	0.225	0.165	0.165	0.164	0.162	0.160	0.159	0.159	0.157	0.155	
<b>Walt Disney</b>	0.291	0.195	0.198	0.197	0.197	0.199	0.200	0.202	0.203	0.204	
<b>Exxon</b>	0.230	0.163	0.161	0.160	0.161	0.160	0.157	0.156	0.153	0.158	
<b>GE</b>	0.275	0.174	0.174	0.174	0.175	0.174	0.170	0.169	0.171	0.166	
<b>Home Depot</b>	0.314	0.228	0.228	0.226	0.225	0.222	0.224	0.219	0.219	0.214	
<b>IBM</b>	0.271	0.190	0.188	0.185	0.184	0.183	0.178	0.177	0.178	0.177	
<b>Intel</b>	0.382	0.267	0.267	0.268	0.264	0.263	0.259	0.256	0.259	0.259	
<b>Johnson &amp; Johnson</b>	0.215	0.164	0.163	0.162	0.160	0.158	0.156	0.156	0.152	0.15	
<b>JP Morgan</b>	0.375	0.222	0.225	0.224	0.230	0.236	0.239	0.243	0.241	0.252	
<b>McDonalds</b>	0.240	0.185	0.182	0.181	0.179	0.177	0.177	0.176	0.174	0.171	
<b>Merck</b>	0.269	0.179	0.175	0.174	0.173	0.173	0.172	0.175	0.176	0.175	
<b>Microsoft</b>	0.323	0.250	0.247	0.245	0.244	0.24	0.236	0.236	0.230	0.232	
<b>Nike</b>	0.327	0.244	0.241	0.239	0.240	0.241	0.238	0.235	0.232	0.232	
<b>Pfizer</b>	0.266	0.189	0.187	0.186	0.187	0.188	0.190	0.189	0.189	0.184	
<b>Procter &amp; Gamble</b>	0.225	0.170	0.168	0.167	0.165	0.164	0.161	0.158	0.160	0.156	
<b>Travellers</b>	0.268	0.175	0.175	0.175	0.178	0.177	0.177	0.184	0.184	0.185	
<b>United Technologies</b>	0.261	0.181	0.179	0.178	0.177	0.177	0.177	0.177	0.176	0.172	
<b>United Health Group</b>	0.386	0.292	0.291	0.292	0.291	0.290	0.289	0.287	0.282	0.278	
<b>Verizon</b>	0.246	0.163	0.162	0.162	0.162	0.164	0.162	0.161	0.160	0.159	
<b>Wal-Mart</b>	0.263	0.205	0.202	0.201	0.198	0.194	0.191	0.191	0.190	0.192	
<b>Cisco</b>	0.415	0.307	0.305	0.300	0.296	0.292	0.293	0.288	0.285	0.281	
<b>Goldman Sachs</b>	0.373	0.225	0.223	0.221	0.221	0.220	0.230	0.233	0.241	0.241	
<b>Visa</b>	0.260	0.203	0.210	0.209	0.208	0.210	0.208	0.206	0.203	0.195	
<b>Average</b>	<b>0.295</b>	<b>0.208</b>	<b>0.207</b>	<b>0.206</b>	<b>0.206</b>	<b>0.205</b>	<b>0.205</b>	<b>0.204</b>	<b>0.203</b>	<b>0.203</b>	<b>0.205</b>

**Table 10: Annualized daily (rule in every 20s trading day) volatility of MA2-MA10, average annualized volatility**

	Buy and Hold	MA10	MA9	MA8	MA7	MA6	MA5	MA4	MA3	MA2	
3M	0.225	0.167	0.169	0.162	0.163	0.161	0.161	0.157	0.156	0.156	
American Express	0.345	0.232	0.235	0.222	0.218	0.22	0.219	0.22	0.243	0.235	
Apple	0.451	0.343	0.347	0.342	0.339	0.339	0.338	0.342	0.335	0.331	
Boeing	0.294	0.207	0.210	0.202	0.202	0.199	0.200	0.197	0.207	0.205	
Caterpillar	0.311	0.216	0.220	0.217	0.215	0.214	0.217	0.218	0.221	0.224	
Chevron	0.244	0.169	0.171	0.172	0.17	0.169	0.169	0.167	0.181	0.171	
Coca-Cola	0.225	0.168	0.171	0.169	0.168	0.166	0.161	0.161	0.161	0.156	
Walt Disney	0.291	0.203	0.207	0.202	0.203	0.203	0.210	0.212	0.215	0.211	
Exxon	0.230	0.166	0.167	0.165	0.164	0.163	0.162	0.157	0.161	0.160	
GE	0.275	0.177	0.177	0.175	0.175	0.175	0.172	0.169	0.172	0.180	
Home Depot	0.314	0.234	0.235	0.228	0.221	0.230	0.228	0.233	0.225	0.219	
IBM	0.271	0.194	0.196	0.196	0.196	0.196	0.19	0.194	0.195	0.190	
Intel	0.382	0.273	0.277	0.272	0.272	0.268	0.266	0.266	0.264	0.259	
Johnson & Johnson	0.215	0.168	0.169	0.167	0.167	0.162	0.158	0.158	0.154	0.150	
JP Morgan	0.375	0.222	0.223	0.217	0.220	0.230	0.233	0.234	0.244	0.234	
McDonalds	0.240	0.189	0.189	0.186	0.185	0.185	0.179	0.170	0.171	0.180	
Merck	0.269	0.177	0.178	0.173	0.173	0.174	0.173	0.181	0.182	0.192	
Microsoft	0.323	0.250	0.251	0.247	0.239	0.233	0.235	0.237	0.233	0.234	
Nike	0.327	0.247	0.248	0.244	0.241	0.240	0.235	0.236	0.238	0.248	
Pfizer	0.266	0.188	0.190	0.186	0.186	0.186	0.187	0.187	0.191	0.189	
Procter & Gamble	0.225	0.173	0.174	0.171	0.167	0.165	0.163	0.164	0.158	0.155	
Travellers	0.268	0.171	0.172	0.17	0.169	0.171	0.191	0.186	0.192	0.198	
United Technologies	0.261	0.178	0.179	0.178	0.177	0.177	0.175	0.178	0.176	0.173	
United Health Group	0.386	0.300	0.302	0.299	0.298	0.294	0.289	0.280	0.283	0.275	
Verizon	0.246	0.167	0.167	0.164	0.162	0.160	0.164	0.157	0.160	0.163	
Wal-Mart	0.263	0.208	0.210	0.205	0.199	0.196	0.197	0.198	0.198	0.189	
Cisco	0.415	0.304	0.307	0.301	0.298	0.300	0.292	0.290	0.281	0.278	
Goldman Sachs	0.373	0.230	0.232	0.225	0.232	0.245	0.239	0.253	0.268	0.256	
Visa	0.260	0.204	0.203	0.212	0.225	0.221	0.219	0.217	0.217	0.196	
<b>Average</b>	<b>0.295</b>	<b>0.211</b>	<b>0.213</b>	<b>0.209</b>	<b>0.208</b>	<b>0.208</b>	<b>0.208</b>	<b>0.208</b>	<b>0.210</b>	<b>0.207</b>	<b>0.209</b>

**Table 11: Annualized daily (every other month) volatility of MAD2-MAD2 (D = every other month, and 5,4,3,2 are the numbers of observations in the rolling window), average annualized volatility**

	<b>Buy&amp;Hold</b>	<b>MAD5</b>	<b>MAD4</b>	<b>MAD3</b>	<b>MAD2</b>	
<b>3M</b>	0.225	0.168	0.169	0.162	0.159	
<b>American Express</b>	0.344	0.222	0.226	0.216	0.211	
<b>Apple</b>	0.450	0.351	0.363	0.357	0.338	
<b>Boeing</b>	0.294	0.210	0.216	0.211	0.208	
<b>Caterpillar</b>	0.311	0.218	0.229	0.215	0.211	
<b>Chevron</b>	0.244	0.168	0.175	0.166	0.165	
<b>Coca-Cola</b>	0.225	0.168	0.173	0.165	0.158	
<b>Walt Disney</b>	0.291	0.197	0.200	0.198	0.203	
<b>Exxon</b>	0.230	0.172	0.174	0.159	0.156	
<b>GE</b>	0.274	0.175	0.181	0.176	0.182	
<b>Home Depot</b>	0.314	0.229	0.230	0.221	0.237	
<b>IBM</b>	0.271	0.196	0.199	0.200	0.200	
<b>Intel</b>	0.382	0.274	0.286	0.267	0.265	
<b>Johnson &amp; Johnson</b>	0.215	0.173	0.175	0.165	0.154	
<b>JP Morgan</b>	0.375	0.236	0.241	0.246	0.237	
<b>McDonalds</b>	0.240	0.182	0.186	0.178	0.169	
<b>Merck</b>	0.269	0.185	0.196	0.188	0.199	
<b>Microsoft</b>	0.323	0.245	0.249	0.238	0.250	
<b>Nike</b>	0.327	0.252	0.258	0.253	0.253	
<b>Pfizer</b>	0.266	0.199	0.203	0.191	0.189	
<b>Procter &amp; Gamble</b>	0.225	0.173	0.177	0.169	0.166	
<b>Travellers</b>	0.268	0.176	0.178	0.183	0.191	
<b>United Technologies</b>	0.261	0.182	0.187	0.178	0.177	
<b>United Health Group</b>	0.386	0.313	0.313	0.299	0.305	
<b>Verizon</b>	0.246	0.163	0.171	0.165	0.153	
<b>Wal-Mart</b>	0.263	0.197	0.199	0.194	0.193	
<b>Cisco</b>	0.415	0.312	0.317	0.315	0.285	
<b>Goldman Sachs</b>	0.373	0.229	0.245	0.239	0.265	
<b>Visa</b>	0.260	0.215	0.215	0.225	0.222	
<b>Average</b>	<b>0.295</b>	<b>0.213</b>	<b>0.218</b>	<b>0.212</b>	<b>0.210</b>	<b>0.213</b>

**Table 12: Annualized daily (every third month) volatility of MAT2-MAT4 (T = every third month, and 4,3,2 are the numbers of observations in the rolling window), average annualized volatility**

	<b>Buy&amp;Hold</b>	<b>MAT4</b>	<b>MAT3</b>	<b>MAT2</b>	
<b>3M</b>	0.225	0.172	0.174	0.171	
<b>American Express</b>	0.344	0.230	0.237	0.206	
<b>Apple</b>	0.450	0.345	0.357	0.349	
<b>Boeing</b>	0.294	0.206	0.219	0.200	
<b>Caterpillar</b>	0.311	0.219	0.223	0.214	
<b>Chevron</b>	0.244	0.176	0.182	0.170	
<b>Coca-Cola</b>	0.225	0.177	0.179	0.181	
<b>Walt Disney</b>	0.291	0.220	0.228	0.205	
<b>Exxon</b>	0.230	0.168	0.176	0.158	
<b>GE</b>	0.274	0.178	0.185	0.177	
<b>Home Depot</b>	0.314	0.236	0.251	0.241	
<b>IBM</b>	0.271	0.205	0.209	0.193	
<b>Intel</b>	0.382	0.285	0.296	0.274	
<b>Johnson &amp; Johnson</b>	0.215	0.185	0.188	0.165	
<b>JP Morgan</b>	0.375	0.242	0.248	0.240	
<b>McDonalds</b>	0.240	0.198	0.204	0.192	
<b>Merck</b>	0.269	0.191	0.191	0.180	
<b>Microsoft</b>	0.323	0.257	0.267	0.258	
<b>Nike</b>	0.327	0.264	0.265	0.258	
<b>Pfizer</b>	0.266	0.195	0.206	0.208	
<b>Procter &amp; Gamble</b>	0.225	0.177	0.181	0.168	
<b>Travellers</b>	0.268	0.187	0.188	0.198	
<b>United Technologies</b>	0.261	0.192	0.199	0.187	
<b>United Health Group</b>	0.386	0.300	0.308	0.315	
<b>Verizon</b>	0.246	0.176	0.176	0.160	
<b>Wal-Mart</b>	0.263	0.202	0.208	0.208	
<b>Cisco</b>	0.415	0.310	0.311	0.303	
<b>Goldman Sachs</b>	0.373	0.226	0.232	0.235	
<b>Visa</b>	0.260	0.204	0.215	0.208	
<b>Average</b>	<b>0.295</b>	<b>0.218</b>	<b>0.224</b>	<b>0.214</b>	<b>0.219</b>

**Table 13: Annualized daily (every fourth month) volatility of MAQ2-MAQ3 (Q = every fourth month, 3 and 2 are the number of observations in the rolling window), average annualized volatility**

	<b>Buy&amp;Hold</b>	<b>MAQ3</b>	<b>MAQ3</b>	
<b>3M</b>	0.225	0.168	0.176	
<b>American Express</b>	0.344	0.220	0.226	
<b>Apple</b>	0.450	0.360	0.373	
<b>Boeing</b>	0.294	0.213	0.224	
<b>Caterpillar</b>	0.311	0.222	0.239	
<b>Chevron</b>	0.244	0.167	0.177	
<b>Coca-Cola</b>	0.225	0.173	0.182	
<b>Walt Disney</b>	0.291	0.206	0.218	
<b>Exxon</b>	0.230	0.160	0.176	
<b>GE</b>	0.274	0.180	0.195	
<b>Home Depot</b>	0.314	0.237	0.242	
<b>IBM</b>	0.271	0.194	0.218	
<b>Intel</b>	0.382	0.274	0.293	
<b>Johnson &amp; Johnson</b>	0.215	0.181	0.186	
<b>JP Morgan</b>	0.375	0.218	0.227	
<b>McDonalds</b>	0.240	0.177	0.193	
<b>Merck</b>	0.269	0.204	0.212	
<b>Microsoft</b>	0.323	0.248	0.260	
<b>Nike</b>	0.327	0.258	0.265	
<b>Pfizer</b>	0.266	0.198	0.207	
<b>Procter &amp; Gamble</b>	0.225	0.173	0.174	
<b>Travellers</b>	0.268	0.182	0.192	
<b>United Technologies</b>	0.261	0.181	0.188	
<b>United Health Group</b>	0.386	0.299	0.314	
<b>Verizon</b>	0.246	0.167	0.177	
<b>Wal-Mart</b>	0.263	0.194	0.207	
<b>Cisco</b>	0.415	0.341	0.349	
<b>Goldman Sachs</b>	0.373	0.240	0.260	
<b>Visa</b>	0.260	0.212	0.225	
<b>Average</b>	<b>0.295</b>	<b>0.215</b>	<b>0.227</b>	<b>0.221</b>

**Table 14: Annualized daily (every fifth month) volatility of MAC2 (C = every fifth month, 2 = observations in rolling window), average annualized volatility**

	<b>Buy &amp; Hold</b>	<b>MAC2</b>
<b>3M</b>	0.225	0.176
<b>American Express</b>	0.344	0.226
<b>Apple</b>	0.450	0.323
<b>Boeing</b>	0.294	0.218
<b>Caterpillar</b>	0.311	0.227
<b>Chevron</b>	0.244	0.165
<b>Coca-Cola</b>	0.225	0.168
<b>Walt Disney</b>	0.291	0.206
<b>Exxon</b>	0.230	0.166
<b>GE</b>	0.274	0.187
<b>Home Depot</b>	0.314	0.242
<b>IBM</b>	0.271	0.202
<b>Intel</b>	0.382	0.296
<b>Johnson &amp; Johnson</b>	0.215	0.187
<b>JP Morgan</b>	0.375	0.244
<b>McDonalds</b>	0.240	0.182
<b>Merck</b>	0.269	0.194
<b>Microsoft</b>	0.323	0.250
<b>Nike</b>	0.327	0.249
<b>Pfizer</b>	0.266	0.191
<b>Procter &amp; Gamble</b>	0.225	0.187
<b>Travellers</b>	0.268	0.183
<b>United Technologies</b>	0.261	0.204
<b>United Health Group</b>	0.386	0.298
<b>Verizon</b>	0.246	0.170
<b>Wal-Mart</b>	0.263	0.223
<b>Cisco</b>	0.415	0.333
<b>Goldman Sachs</b>	0.373	0.218
<b>Visa</b>	0.260	0.220
<b>Average</b>	<b>0.295</b>	<b>0.218</b>

### Appendix 3

**Table 15: Transaction costs per year of MA40-MA200, with one transaction costing 0.1% of total wealth, average annualized transaction costs**

	<b>MA200</b>	<b>MA180</b>	<b>MA160</b>	<b>MA140</b>	<b>MA120</b>	<b>MA100</b>	<b>MA80</b>	<b>MA60</b>	<b>MA40</b>	
<b>3M</b>	0.010	0.011	0.010	0.011	0.012	0.013	0.016	0.019	0.022	
<b>American Express</b>	0.011	0.011	0.011	0.012	0.012	0.013	0.016	0.017	0.023	
<b>Apple</b>	0.007	0.008	0.008	0.009	0.010	0.012	0.014	0.015	0.020	
<b>Boeing</b>	0.008	0.009	0.010	0.011	0.011	0.012	0.014	0.015	0.020	
<b>Caterpillar</b>	0.008	0.009	0.010	0.011	0.012	0.013	0.015	0.015	0.019	
<b>Chevron</b>	0.011	0.012	0.012	0.013	0.014	0.016	0.018	0.020	0.024	
<b>Coca-Cola</b>	0.009	0.010	0.011	0.011	0.011	0.012	0.015	0.018	0.022	
<b>Walt Disney</b>	0.007	0.008	0.009	0.011	0.012	0.012	0.013	0.017	0.021	
<b>Exxon</b>	0.011	0.013	0.016	0.017	0.017	0.018	0.019	0.023	0.028	
<b>GE</b>	0.007	0.008	0.009	0.010	0.011	0.012	0.014	0.017	0.023	
<b>Home Depot</b>	0.008	0.009	0.010	0.011	0.013	0.014	0.016	0.018	0.021	
<b>IBM</b>	0.009	0.010	0.010	0.010	0.012	0.012	0.013	0.014	0.019	
<b>Intel</b>	0.007	0.009	0.010	0.010	0.012	0.014	0.014	0.016	0.019	
<b>Johnson &amp; Johnson</b>	0.009	0.008	0.009	0.010	0.012	0.014	0.016	0.020	0.024	
<b>JP Morgan</b>	0.010	0.010	0.011	0.012	0.012	0.014	0.015	0.016	0.020	
<b>McDonalds</b>	0.010	0.011	0.011	0.013	0.012	0.014	0.016	0.018	0.023	
<b>Merck</b>	0.008	0.009	0.009	0.011	0.011	0.013	0.015	0.017	0.022	
<b>Microsoft</b>	0.008	0.009	0.010	0.010	0.011	0.013	0.015	0.015	0.020	
<b>Nike</b>	0.009	0.009	0.010	0.010	0.011	0.012	0.013	0.014	0.019	
<b>Pfizer</b>	0.008	0.010	0.010	0.011	0.011	0.012	0.014	0.017	0.021	
<b>Procter &amp; Gamble</b>	0.010	0.010	0.010	0.011	0.012	0.014	0.016	0.019	0.022	
<b>Travellers</b>	0.010	0.011	0.012	0.012	0.013	0.015	0.016	0.018	0.024	
<b>United Technologies</b>	0.009	0.010	0.011	0.011	0.012	0.014	0.015	0.018	0.021	
<b>United Health Group</b>	0.008	0.008	0.010	0.010	0.011	0.012	0.014	0.017	0.021	
<b>Verizon</b>	0.011	0.011	0.011	0.011	0.013	0.014	0.017	0.018	0.023	
<b>Wal-Mart</b>	0.010	0.010	0.012	0.013	0.013	0.014	0.015	0.019	0.022	
<b>Cisco</b>	0.006	0.006	0.008	0.010	0.009	0.010	0.014	0.017	0.023	
<b>Goldman Sachs</b>	0.008	0.010	0.012	0.012	0.014	0.015	0.022	0.026	0.035	
<b>Visa</b>	0.008	0.008	0.009	0.009	0.008	0.010	0.011	0.014	0.022	
<b>Average</b>	<b>0.009</b>	<b>0.0010</b>	<b>0.010</b>	<b>0.011</b>	<b>0.012</b>	<b>0.013</b>	<b>0.015</b>	<b>0.018</b>	<b>0.022</b>	<b>0.013</b>

**Table 16: Transaction costs per year of MA2-MA10, average annualized transaction costs**

	<b>MA10</b>	<b>MA9</b>	<b>MA8</b>	<b>MA7</b>	<b>MA6</b>	<b>MA5</b>	<b>MA4</b>	<b>MA3</b>	<b>MA2</b>	
<b>3M</b>	0.003	0.003	0.003	0.003	0.003	0.004	0.004	0.005	0.006	
<b>American Express</b>	0.002	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.006	
<b>Apple</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.005	0.006	
<b>Boeing</b>	0.002	0.002	0.002	0.002	0.002	0.003	0.004	0.004	0.006	
<b>Caterpillar</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.005	0.006	
<b>Chevron</b>	0.002	0.003	0.003	0.003	0.003	0.003	0.004	0.005	0.007	
<b>Coca-Cola</b>	0.002	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.006	
<b>Walt Disney</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.003	0.004	0.006	
<b>Exxon</b>	0.002	0.002	0.003	0.003	0.003	0.004	0.004	0.005	0.006	
<b>GE</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.004	0.006	
<b>Home Depot</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.003	0.004	0.006	
<b>IBM</b>	0.003	0.002	0.003	0.002	0.003	0.003	0.004	0.004	0.006	
<b>Intel</b>	0.002	0.003	0.003	0.003	0.003	0.003	0.004	0.004	0.006	
<b>Johnson &amp; Johnson</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.005	0.006	
<b>JP Morgan</b>	0.002	0.003	0.003	0.003	0.003	0.003	0.003	0.004	0.006	
<b>McDonalds</b>	0.002	0.002	0.003	0.003	0.003	0.003	0.004	0.005	0.006	
<b>Merck</b>	0.002	0.002	0.002	0.003	0.003	0.003	0.004	0.005	0.006	
<b>Microsoft</b>	0.002	0.002	0.002	0.003	0.003	0.003	0.004	0.004	0.006	
<b>Nike</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.004	0.006	
<b>Pfizer</b>	0.002	0.002	0.002	0.003	0.003	0.003	0.004	0.004	0.006	
<b>Procter &amp; Gamble</b>	0.002	0.002	0.003	0.003	0.003	0.004	0.004	0.005	0.006	
<b>Travellers</b>	0.003	0.002	0.003	0.003	0.003	0.004	0.004	0.005	0.007	
<b>United Technologies</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.004	0.004	0.006	
<b>United Health Group</b>	0.002	0.002	0.002	0.003	0.003	0.003	0.003	0.004	0.006	
<b>Verizon</b>	0.003	0.003	0.003	0.003	0.003	0.004	0.004	0.005	0.006	
<b>Wal-Mart</b>	0.003	0.003	0.003	0.003	0.003	0.004	0.004	0.005	0.006	
<b>Cisco</b>	0.002	0.002	0.002	0.002	0.003	0.003	0.003	0.005	0.006	
<b>Goldman Sachs</b>	0.002	0.002	0.002	0.003	0.003	0.003	0.003	0.004	0.005	
<b>Visa</b>	0.002	0.001	0.002	0.002	0.002	0.003	0.003	0.003	0.005	
<b>Average</b>	<b>0.002</b>	<b>0.002</b>	<b>0.002</b>	<b>0.003</b>	<b>0.003</b>	<b>0.003</b>	<b>0.004</b>	<b>0.004</b>	<b>0.006</b>	<b>0.003</b>