

Fundamental investment research – do US results apply to Australian investors?

The global financial crisis has sparked renewed interest by many investors in value-based investment approaches. Much published research, particularly from the United States, supports the use of fundamental analysis for value-based investment. This paper indicates how to test a well-documented US fundamental investment strategy but, in line with our previously published research, it also finds that these strategies do not translate well to our market.



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THE INVESTMENT MODELS used by most institutions and investors are easily classified as either growth-based or value-based. Typically, growth-based models perform well during bull market periods and perform poorly during bear markets. Value-based models typically underperform during bull market phases, but are less subject to large drawdowns during bear markets.

The objective of value-based approaches is to attempt to buy stocks when they are underpriced, and sell them when they become overpriced. The difficulty, of course, is in trying to determine just how far a stock's price may fall, and at what level it really represents a bargain.

The danger of simply selecting stocks that are trading at very low prices is that there may be some fundamental reason why the stocks are priced so low. For example, the company that issues the stocks may be performing poorly, or facing difficult competition in its market. Alternatively, the stock prices of a company may simply be poorly priced, and the low price may not fully reflect the issuing companies' future potential.

To try and differentiate between these two possibilities, practitioners often rely on fundamental analysis, or fundamental variables. The source of these variables is generally from companies' published financial statements.

Essentially, what value-based investors attempt to do is to screen stocks based on the values of these fundamental variables, to sift out those that do not have good future prospects, as measured by the company results.

One of the main difficulties with this type of approach is that it takes many years of data for different companies before rules can be defined and tested. Due to the large size of the US markets, it is perhaps unsurprising that many of the published fundamental research strategies originate there.

Australian investors often read US published research and implement the same ideas in the Australian marketplace. However, Australian investors do not appear to value the same fundamental characteristics as US investors. This can result in well-documented, well-published US strategies performing poorly, or failing altogether when applied to the Australian market.

This paper demonstrates the process we use to determine whether a specific US-based fundamental strategy can be applied successfully in the Australian market. We conduct this process using the fundamental filters published by Aby et al. (2001a, 2001b) as an example.

Aby et al. (2001b) focus on combining fundamental variables to screen stocks for value. Aby et al. developed portfolios based on four fundamental conditions, namely:

- single valued P/E ($P/E < 10$);
- market price < book value;

- established track record of return on shareholder equity (ROE > 12%); and
- dividends paid out less than 25% of earnings.

They concluded that when the four criteria are used to screen stocks, quality investments seem to result.

It is interesting to note that in earlier work by Aby et al. (2001a), the authors had focused on shares with simply a low P/E and a market price below book value (the first two conditions), and had concluded that this filter method did not produce satisfactory returns.

Methodology

One of the challenges of benchmarking a set of filter rules is that they are selective as to when they enter and exit the market. For this reason, to draw sensible comparisons, it is necessary to use a market benchmark that is constructed to be a suitable comparator.

For this paper, the following methodology is used:

- Create a set of all the transactions (matched buys/sells) as indicated by applying the four filter rules. The starting capital for each transaction is \$10,000.
- For every transaction created by the filter rules, create an identical transaction, which buys and sells on the same dates in the underlying benchmark, the XAO index. The starting capital for each of these transactions is also \$10,000.

In this way, there are two sets of trades – one set of trades produced on those companies identified by the four filter rules, and one matching set of trades on the comparison benchmark. Both sets of trades have covered exactly the same time periods, and (initially) had the same amount of capital exposed.

Next, ANOVA tests can be used to determine whether the two sets of trades are significantly different from each other, and hence reveal whether there is any benefit to be expected from using these filters.

This paper aims to test performance of the Aby et al. filters in the Australian market. In total, there are two sets of trades that need to be created, namely:

- '4FAF' (four-factor Aby filter) – all the trades produced by the four filter rules; and
- 'Market' – the comparable trades produced for the market benchmark, using the methodology described above.

One of the primary purposes of this study is to document the returns to the Aby filters in the Australian stock market. To do this, the data used in the study consists of the members of the ASX Fully Paid Ordinaries, starting in 1992 and ending in 2008. Where possible, this data has been adjusted for delistings, name changes, stock splits etc. The comparison benchmark is the XAO (All Ordinaries Index). All buy/sell signals are taken in the presence of transaction costs (\$50 each way), and an allowance for slippage (0.5%) is made on both buy and sell signals. As this study examines the raw returns to the Aby filters, the effects of dividend payments are not included.

Analysis

The following tables show the results from each of the different tests that were performed.

Table 1 shows some of the basic characteristics of each of the above sets of transactions.

In their original paper, Aby et al. determined that the '4FAF' trades did outperform their benchmark.

Using the individual trade sets created during the formation of '4FAF' and 'Market', ANOVA tests can be performed to determine the answers to the following questions:

1. Does the '4FAF' set of trades (and hence the four filter rules) outperform its benchmark?
2. How 'achievable' are these results?

TABLE 1: Characteristics of trades initiated by filters

Characteristic	'4FAF'	'Market'
Average profit per day	\$ 19.71	\$ 0.84
Average Profit/Loss %	39.28 %	2.29%
% of winning Trades	44.83 %	63.79%
Average Profit/Loss % (Winning Trades)	127.97 %	14.59%
Average Profit/Loss % (Losing Trades)	-32.79 %	-19.38%
Number of transactions	174	174
Average Holding Period	273 days	273 days

TABLE 2: Descriptives for '4FAF' and 'Market'

Set	Mean	95% Interval (Lower) (Upper)	Confidence for Mean
4FAF	16.187	0.462	31.912
Market	-1.085	-2.827	0.656

TABLE 3: Characteristics of trades initiated by filters

Characteristic	'4FAF' with Market Cap filter ('4FAFM')
Average profit per day	\$ 47.93
Average Profit/Loss %	14.15 %
% of winning Trades	33.33 %
Average Profit/Loss % (Winning Trades)	103.52 %
Average Profit/Loss % (Losing Trades)	-30.53 %
Number of transactions	27
Average Holding Period	228 days

TABLE 4: Descriptives: '4FAFM' and 'Market' equivalent

Set	Mean	95% Interval (Lower) (Upper)	Confidence for Mean
4FAFM	26.920	-51.051	104.891
Market equivalent	-7.71	-12.689	-2.736

The ANOVA test allows for a comparison of sets of trades to determine whether they are significantly different from each other. The variable chosen for comparison from each set of trades is the Net Profit per Day (profit after costs/slippage, etc).

For question 1, 'Does the '4FAF' set of transactions (and hence the four filter rules) outperform its benchmark?', the ANOVA test shows that the two sets of trades are very different, specifically $F(1,346)=4.643$, $p=0.032$ ($p<0.05$). In other words, the set of trades produced by the '4FAF' rules are very different from the market trades, and the results show that the '4FAF' trades are much more profitable.

Table 2 provides descriptive information concerning the two sets of trades compared above.

To answer question 2, 'How achievable are these results?', a minimum market capitalisation filter was added. To enable a stock to be 'bought', a minimum market capitalisation of \$50 million was chosen. The tests were then re-run, and the results were then re-analysed.

After examining the trades resulting from the '4FAFM', it is apparent that there is a heavy right tail. The question is whether this is a result of the filter or just random occurrences. The dataset may be biased with respect to fundamental values since it covers the bursting of the dot-com bubble, which saw a return to a traditional valuation approach, at least for some time.

In the absence of a longer time horizon or additional information about the underlying valuation processes, the set of trades was further constrained to those trades within three standard deviations of the mean. This process retains the vast majority of trades while removing what could be outliers, i.e. trades not representative of the true relationship between the filter rules and the expected return. This outlier removal process only removed one trade.

Figure 1 shows the histogram of profit per bar values for the '4FAFM' model before the outliers have been removed. A normal curve is included.

Figure 2 shows the effect that outlier removal has had on the distributions of profit per bar. Again, a normal curve is included.

FIGURE 1: '4FAFM' model with outliers

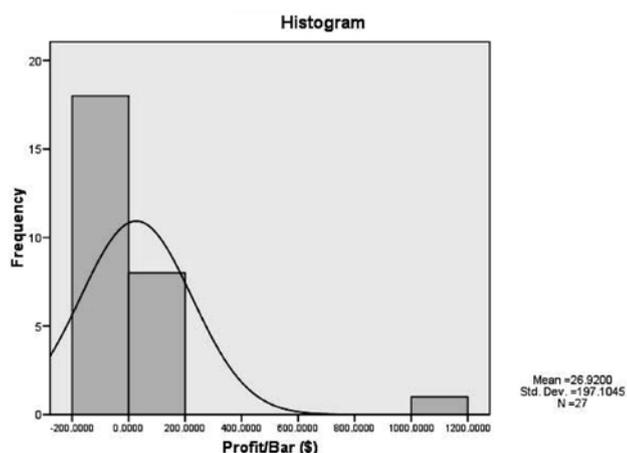


FIGURE 2: '4FAFM' excluding outliers

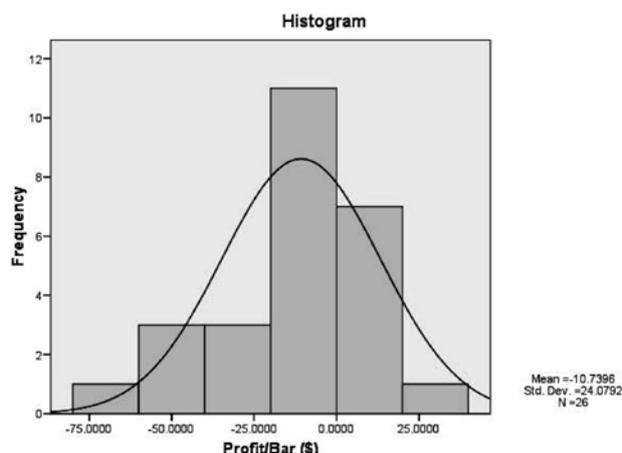


TABLE 5: Model comparisons with and without outliers

Characteristic Model	Mean	Number
4FAFM	26.92	27
4FAFM excl outliers	-10.739	26
4FAFM market equivalent	-7.71	

Table 5 shows the number of trades and the mean of those trades for the '4FAFM' with and without the outlier. Clearly, the removal of the one outlier has significantly affected the results. Now the model is performing worse than the overall market.

Discussion of results

Aby et al. found that their model with four factors performed much better than the overall market. However, in the Australian market, we reach a different conclusion.

In the Australian market, the four-factor model appears to outperform. However, when the universe of stocks is constrained to those that are liquid, and the one massive outlier is removed, it can be seen that the four-factor model significantly underperforms the market.

Conclusion

Although the premise upon which the Aby filters are built seems to be sound, it is clear that the filters underperform in the Australian market, and are therefore unsuitable for Australian investors.

Given the quantity of research that supports the use of each of these fundamental variables, it is possible that the cut-off values applied in the filter rules are too strict for the Australian market. The cut-off values used by Aby et al. are anecdotal values at best, and it is unclear how new cut-off values should be established.

In previously published work concerning fundamental investment strategies (Vanstone and Agrawal, 2006) we benchmarked other fundamental-based techniques, and found that although they appeared successful in the US market in some cases, this success did not translate to the Australian market. ☹

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