The Impact of Insider Trading on Forecasting in a Bookmakers' Horse Betting Market

by

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Abstract.

This paper uses a new variable based on estimates of insider trading to forecast the outcome of horse races. We base our analysis on Schnytzer, Lamers and Makropoulou (2008) who showed that inside trading in the 1997-1998 Australian racetrack betting market represents somewhere between 20 and 30 percent of all trading in this market. They show that the presence of insiders leads opening prices to deviate from true winning probabilities. Under these circumstances, forecasting of race outcomes should take into account an estimate of the extent of insider trading per horse. We show that the added value of this new variable for profitable betting is sufficient to reduce the losses when only prices are taken into account. Since the only variables taken into account in either Schnytzer, Lamers and Makropoulou (2008) or this paper are price data, this is tantamount to a demonstration that the market is, in practice, weak-form efficient.

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1. Introduction

Successful forecasting of horse race outcomes requires that the forecaster has a clear understanding of the variables at his disposal. The most common, and arguably important, variables in a horse betting market are the odds of the horses in a race. In the case where bookmakers operate in such a market, it seems reasonable to suppose that the fixed odds they provide would be reasonably unbiased estimators of the horses' winning probabilities. And yet, there is a considerable literature which suggests that this is not so (see, for example, Shin 1991, 1992 and 1993 and Schnytzer and Shilony 2003). What makes bookmakers' odds deviate from winning probabilities is agreed to be the extent of insider trading in the market, even though different authors characterize the mechanism underlying the concomitant distortion and its extent differently.

Accordingly, forecasting of race outcomes should take into account an estimate of the extent of insider trading per horse and of how this extent of insider trading in a bookmakers' horse betting market may be measured. Schnytzer, Lamers and Makropoulovou (2008) [hereafter SLM] have developed a model for measuring the extent of insider trading in horse betting markets with bookmakers.¹ Their paper develops a theoretical framework that examines the optimal price setting by bookmakers in the racetrack betting market and then uses it to measure the extent of insider trading in the market. Bookmakers are faced with the risk that insiders will account for information arriving after the opening odds (which may be assumed to contain most public information) have been set and will thus exploit any mis-pricing by the bookmaker by betting on horses whose price presents an expected profit for the insider. The model is an extension of the model developed by Makropoulovou and Markellos (2007) and applied to the European soccer betting market. The basic intuition underlying the model is that fixed odds² offered by bookmakers at the track are examples of call options and that, while bookmakers hope to offer only net of premium out-of-the-money options, when they err by underestimating a particular horse's true winning probability, they are liable to offer a

¹ Theirs is not the first such method. Shin (1993) developed a method using a very different model.
² For the purposes of this paper, by odds, we mean that odds of, say 5 to 1 represent a net profit of $5 for every $1 bet on the winning horse.
net in-the-money option, which the insider (who is assumed to know her horse's true winning probability) will be glad to snap up.

Throughout this paper, we use the working assumption that the insider knows her horse’s true winning probability and this requires some elaboration. Indeed, it is difficult to come up with a precise definition of an inside trade for which data may ever be available. Thus, in reality, an insider is one who is more familiar with her horse than others and who therefore has an informational edge over outsiders and, ceteris paribus, is in a better position to evaluate the horse’s winning probability. But ceteris is not paribus! There are over- and under-optimistic insiders just as there are different kinds of outsiders. Some people know more about forecasting and some less. And these kinds of differences are never measurable in the kinds of data sets that are available from horse betting markets. Accordingly, we assume what we do about insiders and, with respect to outsiders, we assume that they bet according to the opening odds set by bookmakers, these being the best estimate available of public information prior to the start of betting at the track.

The way in which insiders bet involves the so-called plunge. This is a case where several gents of the insider approach different bookmakers simultaneously and back the same horse at the best odds available. The reason that a single bettor is usually insufficient is that bookmakers are permitted to refuse bets which would leave them with large contingent debts. Accordingly, an insider wishing to place really large sums of money on a particular horse will need to spread the bet across bookmakers. It should be noted that on-course bookmakers are small, independent firms who compete in selling a homogeneous product. Accordingly, competition among them is fierce and the trend in prices during the betting is always downwards unless a horse is plunged. However, since all bookmakers need to determine initial odds and since all in Australia must be members of the relevant state bookmakers’ Association, they save on research costs by obtaining a set of opening odds from the Association. These are not obligatory, but they tend to be

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3 The precise size of the maximum bet which a bookmaker must accept varies from place to place but is rarely above a thousand dollars.
4 For the purposes of this paper, prices are defined in their economic sense as the amount that must be bet on a horse to ensure a total payback (including the initial outlay) of $1. Odds, on the other hand, have their traditional meaning; i.e. If the odds of a winning horse are X to 1, then $1 bet on the winning horse yields a total payback (including the dollar outlay) of X+1.
widely used. The important thing about these prices is they contain a high built-in expected mark-up which serves as a cushion of sorts against insiders. Of course, once a plunge arrives, every bookmaker is on his/her own and the prices of all horses in the race fluctuate freely. For our purposes, any fall in odds (increase in price) is taken as evidence of a plunge and we use this variable as a predictor for the outcome of the race.

We proceed as follows: In Section 2, the data are described and a brief discussion of our forecasting method is provided. The results are presented in Section 3, where it is shown that forecasting on the basis of opening prices only – these prices are readily available around 30 minutes before the race – yields moderate losses. The extent of these losses is reduced when the variable that measures insider trading is added, but the method employed here would be difficult, if not impossible, to implement in practice. Some conclusions are offered in section 4.

2. Data and Methodology

The data set used in this paper contains 45296 horses who ran in 4017 races during the 1997-1998 Australian horse racing season. The data include opening prices (hereafter OP) as set by bookmakers at the start of betting around 30 minutes before each race, middle prices (hereafter MP), which are prices provided in the data set usually, but not always, when there is a change in direction of the horses' odds between OP and odds at the end of the betting. Finally, we have starting prices (SP), the ruling prices at the end of betting. The data set contains all races for which MP are provided. The data were obtained from the CD version of the "Australasian Racing Encyclopedia '98".

SLM estimated several alternative measures of the extent of insider trading in this market and we use three of the estimates for the purposes of forecasting. However, in order to facilitate an understanding of these measures and the differences between them, a summary of the SLM estimation procedure is in order. Bookmakers' odds as initially set (i.e. OP), may be viewed as call options which end in-the-money if the horse wins the race and out-of-the-money otherwise. As inside information enters the market, the odds change and the value of the call options change. As betting continues, the horses' winning probabilities as implied by the odds become more and more accurate until all inside information has entered the market and the betting comes to an end. Assuming that the
inside information enters the market randomly from the point of view of the bookmakers, the dynamics underlying the changing implied winning probabilities may be modeled as a standard Wiener Process.

Using Monte Carlo simulation we are able to derive the option value for each horse. The true winning probability for each horse is simulated in 1000 time steps using a standard Wiener Process. When the simulated probability is larger than the strike price at the 1000th and final step, the option value is this positive difference; otherwise, the option value is zero. For each horse, the option value is calculated as the average value out of 1000 repetitions. In order to calculate the extent of insider trading, a weighting is applied to the option values. The following three different weightings used provide us with our estimates of insider trading for use in this paper.

The first weight used for each horse is the estimated initial winning probability as implied by OP, P(0). The remaining two additional weights are based on plunge behavior in the market and calculated as follows. The first is the relative size of a plunge, called PW: \(\max((\text{MP}-\text{OP})/\text{OP},0) + \max((\text{SP}-\text{MP})/\text{MP},0)\). The second weight is the absolute size of the plunge called PW2: \(\max(\text{MP}-\text{OP},0) + \max(\text{SP}-\text{MP},0)\). Using these weights, the weighted average degree of insider trading for each of the races in the sample is calculated. The simple average of these values is the extent of insider trading in the dataset.

Table I displays the extent of plunges in the data set, where an early plunge is defined as a positive percentage price change from OP to MP and a late plunge is defined as a positive percentage price change from MP to SP. A sustained plunge is defined in the case where the horse in question is subject to both early and late plunges; the extent of the sustained plunge is then the percentage change from OP to SP. It can be seen in Table I that the majority of the 13852 plunges in the dataset are late plunges, suggesting insider trading at MP. However, the average extent of early plunges exceeds that of late plunges.

Table I: The extent of plunges in the dataset

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5 A horse is said to be plunged when its odds suddenly decrease meaningfully owing to large bets placed on the same horse with different bookmakers simultaneously. Schnytzer and Shilony (1995) show that plunges contain inside information.
<table>
<thead>
<tr>
<th>Plunges</th>
<th>Number</th>
<th>Average extent</th>
</tr>
</thead>
<tbody>
<tr>
<td>Early Plunge</td>
<td>1281</td>
<td>21.25%</td>
</tr>
<tr>
<td>Late Plunge</td>
<td>9783</td>
<td>15.72%</td>
</tr>
<tr>
<td>Sustained Plunge</td>
<td>2788</td>
<td>26.33%</td>
</tr>
<tr>
<td>All</td>
<td>13852</td>
<td>18.37%</td>
</tr>
</tbody>
</table>

Note: An early plunge is defined as a positive percentage change from OP to MP. A late plunge is defined as a positive percentage change from MP to SP. When there are both early and late plunges, this forms a sustained plunge.

The simple average of these values is a variable that measures the extent of insider trading as estimated by SLM, is shown in Table II.

Table II: Measures of degree of insider trading for each specification

<table>
<thead>
<tr>
<th>Weight</th>
<th>Degree of insider trading</th>
</tr>
</thead>
<tbody>
<tr>
<td>P(0) – OP</td>
<td>32.68%</td>
</tr>
<tr>
<td>PW</td>
<td>26.38%</td>
</tr>
<tr>
<td>PW2</td>
<td>26.48%</td>
</tr>
</tbody>
</table>

Notes: P(0) – OP is true winning probability at time 0. PW is the \[\max((\text{MP} - \text{OP}))/\text{OP},0) + \max((\text{SP} - \text{MP})/\text{MP},0)\]. PW2 is \[\max(\text{MP} - \text{OP},0) + \max(\text{SP} - \text{MP},0)\].

Armed with opening prices and various measures related to the extent of insider trading for each horse, we proceed to forecast the winners of each race in the data set. We use the generally preferred method of forecasting in the betting literature, namely; the conditional logit model (hereafter CL) of McFadden (1974). We estimate several CL models. The first estimates the probability of horse \(i\) winning race \(j\) based solely on information contained in OP, as follows:

\[
p_{ij}^o = \exp(\alpha_i \text{OP}_i) / \sum_{i=1}^{n_j} \exp(\alpha_i \text{OP}_i)
\]

where \(i = 1, 2, \ldots, n_j\), \(\text{OP}_i\) is the OP of horse \(i\) in race \(j\), \(n_j\) is the number of runners in race \(j\) and \(\alpha_i\) indicates the contribution which OP makes to the horse’s chance of winning race \(j\). We then run four more regressions adding different predictors to OP in turn. These variables are as follows: First, the option value for each horse as estimated in SLM. This
variable is zero for most horses in the sample and positive for one or two in each race. \( \textit{Optionvalue}_{ij} \) is positive if the horse \( i \)'s winning probability in race \( j \) is estimated via Monte Carlo simulation to be greater than the winning probability implied by \( \textit{OP} \) and is measured as the difference. In this case, the model estimated may be written:

\[
p^n_{ij} = \exp(a_2 \textit{OP}_{ij} + b_i \textit{Optionvalue}_{ij}) / \sum_{n=1}^{n_j} (a_2 \textit{OP}_{ij} + b_i \textit{Optionvalue}_{ij})
\]

where \( i = 1,2,\ldots n_j \). The coefficients \( a_1, a_2 \) and \( b_1 \) in regressions (1) and (2) are measured by maximizing the joint probability of observing the winners of all the races in the sample. Next we add \( \textit{EarlyPlunge}_{ij} \), which is equal to the difference between \( \textit{MP} \) and \( \textit{OP} \), when this difference is positive and zero otherwise.\(^6\) Our fourth predictor is the extent of insider trading on horse \( i \) in race \( j \) as measured by SLM (\( \textit{Insidertrading}_{ij} \)). Finally, we add \( \textit{TotalPlunge}_{ij} \), with measures the total extent of early and late plunges on horse \( i \) in race \( j \). We expect, a priori, that all variables would, by themselves, add to a horse's winning probability and should thus receive positive coefficients.

3. Results

Table III shows the results of our five regressions.

<table>
<thead>
<tr>
<th>Specification</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dependent variable</td>
<td>Win</td>
<td>Win</td>
<td>Win</td>
<td>Win</td>
<td>Win</td>
</tr>
<tr>
<td></td>
<td>(50.58)*</td>
<td>(47.09)*</td>
<td>(45.73)*</td>
<td>(44.91)*</td>
<td>(43.88)*</td>
</tr>
<tr>
<td>\textit{Optionvalue}</td>
<td>2.1108</td>
<td>1.5214</td>
<td>-2.7790</td>
<td>-13.7064</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(1.09)</td>
<td>(0.78)</td>
<td>(-1.32)</td>
<td>(-5.63)*</td>
<td></td>
</tr>
<tr>
<td>\textit{EarlyPlunge}</td>
<td>10.824</td>
<td>10.1933</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(6.79)*</td>
<td>(*)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>\textit{Insidertrading}</td>
<td>0.2741</td>
<td>0.2079</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(5.80)*</td>
<td>(4.32)*</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>\textit{TotalPlunge}</td>
<td>9.3020</td>
<td>45296</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
|                    | (9.62)*| (45296)
| \textit{N}         | 45296 | 45296 | 45296 | 45296 | 45296 |
| \textit{Pseudo – R}^2 | 0.1389| 0.1390| 0.1412| 0.1430| 0.1456 |

\(^6\) The winning probabilities, when this and subsequent variables are added, may be estimated by models that follow trivially from (1) and (2) and thus are not noted explicitly.
Notes: Z-scores are reported in parentheses. A * indicates significance up to a 1% confidence level. Optionvalue are the option values generated by SLM. EarlyPlunge is the extent of early plunges as measured by max(MP-OP,0). Insidertrading is the incidence of insidertrading on a specific horse, as generated by SLM. TotalPlunge is the extent of both early and late plunges, as measured by max(MP-OP,0) + max(SP-MP,0).

It is clear on the basis of these results that OP is by far the most important predictor of winning probabilities, both in terms of coefficient size and statistical significance. Given the bookmakers' stake in the outcome of betting, it is clear that OP will reflect as much useful information as possible unless bookmakers deliberately distort prices as a defense mechanism against insiders.\(^7\) With the exception of option value, all other variables receive positive coefficients and these are statistically significant in at least one of the regressions. When option value is used as the sole predictor of winning probabilities, it receives a positive and highly significant coefficient\(^8\), leading us to conclude that the unexpected results here are the result of multicollinearity.

The following table shows the results of betting $1 on each predicted favorite in every race in the sample on the basis of the five regressions shown in Table III.

| Table IV: Betting Simulations |
|-------------------------------|------------------|------------------|------------------|------------------|------------------|
| Betting based on favourite predicted by specification | 1 | 2 | 3 | 4 | 5 |
| Number of races and bets | 4017 | 4017 | 4017 | 4017 | 4017 |
| Profit ($) | -409.83 | -379.40 | -360.20 | -285.00 | -301.50 |
| Rate of Return (%) | -10.20 | -9.44 | -8.97 | -7.09 | -7.51 |

Notes: Betting takes place in all races as in each race there is a favorite as measured by the highest win probability predicted after each regression specification. Betting takes place at SP, the last quoted prices before the race starts.

The results obtained are in line with what the regression results in Table III might have led us to believe. Thus, insider trading seems to influence profits (in this case, losses) in an upward direction, although the tone is clearly set by OP, and betting on the

\(^7\) See SLM, Schnytzer and Shilony (2003) and Shin (1991 and 1992) for more discussion on this point.

\(^8\) Complete results available upon request.
basis of it alone leads to a loss of 10.2 percent. The best performance is achieved by adding option values, early plunges and the extent of insider trading to OP, but this only adds a little over 3 percent to the reduction in losses. However, since SLM rely exclusively on price data in their simulations, these results show that this market is, in practice, weak-form efficient. Further, even if the results are calculated as if betting takes place at the best odds available during the betting (as we would expect insiders to bet), rather than SP, returns are better but remain negative throughout.\(^9\) Finally, it may seem strange that the losses incurred in simulation 5, when all plunges are taken into account along with the other variables, should exceed those in simulation 4, when only early plunges are added to the model. The reason would appear to be the herding on late plunges in this market.\(^10\)

4. Conclusions

In this paper we have shown that variables that measure insider trading, as measured by SLM, only have a moderate impact on the forecasting results. Adding various different measures relating to insider trading by horse to a conditional logit model using only opening prices to predict winning probabilities, reduces moderate losses but does not generate positive profits. Therefore, the relevance of insider trading in this market, in principle, cannot be refuted. However, it should be pointed out that even the small gains in forecasting demonstrated here may be difficult to implement in practice.

It is unlikely that the simulations used by SLM could be carried out in the short time required before each race. Thus, a knowledge of price changes is critical and if the latest prices used in the simulation were to be those ruling in the market five minutes or so before the race start, that would leave less than five minutes for the estimations. Since the simulations carried out by SLM required several days to run, a system based on our estimates could be applied only on a computer far more powerful than is generally available today outside the Pentagon! Furthermore, given the moderate gains generated by the addition of these variables to the basic model, it may be wondered whether it would be worthwhile to struggle for a solution to the computing problem.

\(^9\) Full results available upon request.
So why has the extent of inside information not contributed more dramatically to the forecasts? To the extent that the SLM model provides a reasonable measure of the extent of insider trading, it must be concluded that the reliance on price data alone in forecasting horse races in a bookmakers' market is doomed to failure. On the other hand, perhaps the basic weakness in regression models in forecasting is that they provide predictions on the basis of "on average" results, whereas insiders bet on particular horses in particular races when as many as possible relevant factors unbeknownst to outsiders have been taken into account.
5. References


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