

Professionals' conviction of behavioral finance: Does it impact their perception of markets and themselves?

Lukas Menkhoff and Marina Nikiforow
Leibniz Universität Hannover, Germany

Abstract

This paper provides evidence on Hirshleifer's hypothesis that many behavioral finance patterns are so deeply rooted in human behavior that they are difficult to overcome by learning. We test this on a target group which has undoubtedly very strong incentives to learn efficient behavior, i.e. fund managers. We split this group into proponents and non-proponents of behavioral finance. Proponents do, indeed, view markets differently as they regard stronger influences from behavioral biases. However, when it comes to the perception of one's own behavior the conviction of behavioral finance becomes almost meaningless, even though proponents otherwise do adapt behavior to their conviction.

JEL codes: G 10 (general financial markets), D 83 (learning, knowledge, belief)

Keywords: behavioral finance, fund managers, biases

February 28, 2007

We thank all participating fund managers for their time and commitment. We gratefully acknowledge research assistance by Torsten Brozynski.

menkhoff@gif.uni-hannover.de and nikiforow@gif.uni-hannover.de, both: Leibniz Universität Hannover, Königsworther Platz 1, D-30167 Hannover, Germany

Professionals' conviction of behavioral finance: Does it impact their perception of markets and themselves?

1. Introduction

The history of many anomalies in financial markets has shown that they disappear over time (Fama, 1998). This has raised the suspicion that markets may need time to recognize such anomalies – which largely motivate behavioral finance – but that they react consequently afterwards. According to this view one may assess behavioral finance being largely concerned with transitory phenomena. Others argue that many behavioral finance patterns are so deeply rooted in human behavior that they are difficult to overcome by learning (Hirshleifer, 2001). Obviously, these two views make contrary predictions on the permanency of behavioral phenomena. If rational learning is dominant then one could expect that insights into behavioral finance impact one's own behavior. If, however, learning mechanisms are weak in this respect the insights into behavioral finance will have a minor impact on one's own behavior. We provide a test of these competing views and find evidence in support of Hirshleifer's hypothesis.

Our research aims for extending literature in psychology which has clearly revealed the “bias blind spot” (Pronin et al., 2002), i.e. the conviction that one's own judgments are less susceptible to biases than the judgments of others.¹ We extend this research by addressing an important objection that economists might voice against evidence based on student surveys: The probability that behavioral biases hold depends on how strong incentives are to learn and thus to overcome entrenched behavior. One can easily imagine circumstances where people adhere to behavioral patterns as forecasted by behavioral finance just because their welfare is not at stake. It seems therefore advisable for an empirical examination of Hirshleifer's hypothesis to choose a target group that has undoubtedly strong incentives to learn. We have thus targeted professional fund managers. Their investment performance is on the one hand negatively affected by behavioral biases (see e.g. Shefrin and Statman, 1985, Coval and Shumway, 2005, Biais and Weber, 2006) and on the other permanently monitored and linked

¹ This asymmetry arises from the stance of “naive realism” as well as from different strategies people apply to detect bias in own judgments versus the judgments of others (Pronin et al., 2004, see also Ehrlinger et al, 2005). Recent research in this line includes adding individual's self-image to the utility function (Johansson-Stenman and Martinsson, 2006).

to high performance-related bonuses. Among these fund managers we differentiate between “proponents” of behavioral finance and others, which we call “non-proponents”.

Proponents of behavioral finance are those fund managers who believe that the approach of behavioral finance truly reflects decision behavior in fund management (and who know the key messages of behavioral finance well). In contrast to them, there are non-proponents, i.e. fund managers who are not that much convinced about the relevance of behavioral finance. We have surveyed more than 100 fund managers in Germany and classify them according to their self-assessment into these two groups. We then analyze whether fund managers’ conviction of behavioral finance, i.e. being a “proponent”, impacts their perception of markets and themselves: Do these groups respond in the same way to questionnaire items addressing, first, their perception of general fund managers’ behavior and, second, their perception of their own behavior with respect to issues being raised by behavioral finance research?

We find a revealing split in responses: Whereas proponents recognize significantly stronger behavioral finance effects in other fund managers’ behavior than non-proponents, the perception of their own behavior is largely unaffected by their insights. When proponents’ are asked about their own behavior with respect to items being linked to behavioral finance, such as hindsight bias or disposition effect, they answer as non-proponents do. However, there is one exception to this pattern: Proponents show less miscalibration with respect to forecasting the interval of a stock index. As less miscalibration here also means more correct answers, this suggests that proponents’ conviction of behavioral finance may increase awareness for respective distortions and can improve decisions to some extent.

In a final exercise, we analyze whether proponents’ assessment of their own behavior may reflect the fact that behavioral finance does not influence their decisions in any respect. Therefore, we ask whether proponents differ from non-proponents regarding two further items of investment behavior, i.e. their preferred information sources and investment strategies. We find that both groups seem to differ clearly in their use of information sources although not to a statistically significant extent. The difference becomes significant, however, with respect to preferred investment strategies as proponents rely more on momentum and contrarian strategies. We conclude that proponents do not generally behave like non-proponents, making the above found indifference between both groups with respect to their self-assessment more credible.

Thus, we come back to Hirshleifer's hypothesis that many behavioral patterns are difficult to overcome by learning. Our findings provide support for his view as insight into behavioral finance and thus into the behavior of others does not easily change one's own behavior. We regard it as an important and innovative aspect of our research that this result has been found among fund managers because this target group has strong incentives to improve behavior. In this respect we also find a partial justification for Fama's (1998) optimism into learning processes in that the conviction of behavioral finance reduces miscalibration.

The remaining study is structured into four sections. Section 2 describes data of the questionnaire survey. The following sections present our analyses, starting with views on the market's behavior (Section 3), then views on one's own behavior (Section 4) and, finally, consequences on investment behavior (Section 5). Section 6 concludes.

2. Data

This study is based on a questionnaire survey that addressed all relevant fund management companies in Germany between August 15 and December 12, 2002 and yielded a useful and largely representative sample.

In total, out of the 59 relevant fund management companies 35 participated in the survey, with at least one appropriate questionnaire each. This resulted in a response rate of 59% concerning participated fund management companies.²

To ensure the reliability of responses many intensive interviews with fund managers were conducted in advance of the survey. These interviews served to formulate questions in an appropriate manner. Furthermore, in later stages the questionnaire was used in a pre-test with several fund managers as a final check of its acceptance and appropriateness. Feedback indicates that the response is useful for our research purpose.

The descriptive statistics in [Table 1](#) provide a picture of personal characteristics of the respondents in this survey. Accordingly, the mean of responding fund managers shows an age of about 35 years, has professional experience of ten years, is of male gender, earns a bonus payment of 25%, has a university degree, works rather in a non-governing position, practices active fund management and manages stocks rather than bonds. This data is consistent with the information from similar surveys in Germany such as Menkhoff (1998) or Arnsward (2001).

² The structure of this response is largely similar to the industry's structure. Our sample is representative in this sense (see [Appendix 1](#)).

As basis for the formation of groups our questionnaire contains two useful statements on behavioral finance, i.e. conviction and knowledge of behavioral finance. These statements are given in [Table 2](#) as BF1 and BF2, both of which are assessed by 6 answering categories, ranging from “complete approval” (coded as 1) to “complete contradiction” (coded as 6). The main information we use from these statements is the degree of conviction of behavioral finance as it is expressed in the response to BF1. Due to the distribution of answers (see the contingency table in [Table 2](#)) we define proponents by comprising answering categories 1 and 2, whereas category 3 seems to characterize an intermediate stance in comparison to the sample. We complement this definition by information from BF2, i.e. we define that proponents should also know the key messages of behavioral finance well (corresponding to answering categories 1 and 2 again).³ Due to this definition based on answers to BF1 and BF2, the group of proponents consists of 37 fund managers whereas the group of non-proponents consists of 67 fund managers; 3 persons are not considered further in the analysis because they say to be convinced of behavioral finance but have only limited knowledge about it.⁴

As we aim for a comparison of proponents with non-proponents we use the descriptive items given in [Table 1](#) in order to find out about differences in personal characteristics. Interestingly, the results in [Table 3](#) don’t disclose any such difference to be statistically significant or even close to significance. Thus, the conviction to behavioral finance is not closely related to personal characteristics, such as being older or holding a better position etc. In order to test the robustness of this result we also run a binary probit regression where the variable “proponent vs. non-proponent” is “explained” by available personal characteristics. The last two columns in [Table 3](#) show that there is also no significant relation in this multivariate approach. As proponents of behavioral finance do not represent a particular sub-group among fund managers, we do not control for personal characteristics in our further research.

³ Thus non-proponents share the belief that behavioral finance does not reflect decision behavior in fund management quite well (statement BF1). However, two groups may be distinguished with respect to statement BF2, i.e. their knowledge about behavioral finance. Splitting responses into those non-proponents who know behavioral finance well (categories 1 and 2) and the others (categories 3 and more) gives groups with 32 and 35 fund managers, respectively. We have thus repeated all analyses with three groups (proponents and two kinds of non-proponents) in order to see whether the behavioral finance knowledge of non-proponents may be a driving force in understanding them. But results remain almost unchanged and are thus robust in this respect. Only the level of significance is sometimes negatively affected by smaller sample size.

⁴ There are another 7 fund managers who did not answer BF1 or BF2 in a useful way so that sample size is reduced from 114 to 107.

3. Views on market's behavior

We ask fund managers to express their view on the behavior of other fund managers, i.e. the market, with respect to effects stated by behavioral finance. We find consistently that proponents observe these effects significantly stronger than non-proponents.

In order to systemize the variety of human behavior biases Hirshleifer (2001) assigns them either to the process of judgment or to the process of decision-making. Analogically, we ask for fund managers' response to statements [1] and [2] that are judgment biases and their response to statements [3] to [5] that are rather decision biases. [Table 4](#) presents the five statements we are analyzing and which we motivate in the following before going into results.

3.1 Motivation

Statement [1] in [Table 4](#) addresses the *house money effect* caused by the mechanisms of heuristic simplification. Simplification becomes active because of the restraints in human information-processing capabilities due to which the complexity of the available basis of information is reduced by filtering out apparently irrelevant facts. By thus simplifying the information-input for decision-making perception of reality gets biased. This can result in suboptimal decisions. In this context, house money effect describes the phenomenon that investors become less loss-averse and more likely to take risks when reinvesting recently made profits. The psychological explanation for this bias is that the perceived discomfort in case of a loss of the recently earned money is diminished by the fact that it was a gain before (Thaler and Johnson, 1990).

Statement [2] aims at the *confirmatory bias* which is attributed to the natural mechanism of self-deception. After a decision people tend to collect decision-confirming information thereby ignoring contrary evidence (Pronin, 2007). The confirmatory bias describes the observation that people are prone to interpret even ambiguous information in favor of their earlier decision. These natural mechanisms of selective information perception interfere with rational learning, including the recognition of and learning from one's own mistakes. The motivation for the confirmatory bias is to preserve self-respect of the decider as well as the aspiration for emotional consonance after decision-making (Festinger, 1957, Nickerson, 1998).

Statements [3] to [5] mention decision biases. Statement [3] serves to detect the *reflection effect* (Kahneman and Tversky, 1979), grasped here by risk seeking behavior in case of loss positions. The reflection effect describes the dependence of investors' risk attitude on the sign of the outcome, i.e. whether the outcomes are gains or losses. It is included in prospect theory by the s-shape of the value function: concave for gains indicating risk aversion and convex for losses indicating risk seeking. This bias partly accounts for the disposition effect. The preference to invest in near located markets, the *home bias* (statement [4]), is explained – in the behavioral finance approach – by the human preference for familiarity, which evokes the feeling of possessing more relevant information about the home market and thus being more competent to forecast (Kilka and Weber, 2000, Lütje and Menkhoff, 2007). The third decision bias addressed here with statement [5] is *herding*. It is understood as the phenomenon that fund managers may imitate the investment decision of others and thereby ignore their own valuable information (see Bikchandani and Sharma, 2001).

3.2 Results

Regarding the results in Table 4 we detect significant differences between proponents and non-proponents concerning the intensity of perception of biases in their colleagues' behavior and thereby in the market. This finding applies to the two judgment biases and to the three decision biases as well. In all five cases proponents recognize the biases significantly stronger than non-proponents, even though the latter mostly concede the existence of these behavioral finance effects. Regarding the house money effect (statement [1]), for example, proponents approve this with a mean answer of 2.59, whereas non-proponents give a mean answer of 3.13 – indifference between approval and contradiction would occur at a value of 3.50. The difference between both groups is significant at the 5% level.

Remarkably, the perception of herding is by far strongest compared to other biases. All fund managers agree with a mean of 2.02 to statement [5]. Again, the difference between both groups is statistically significant as proponents perceive herding among fund managers stronger than non-proponents.

The results in Table 4 indicate that the major fraction of professional fund managers observes behavioral finance effects in their daily working environment. It does not seem surprising that proponents of behavioral finance perceive these effects stronger than non-proponents. Against the background of this finding the next section examines the impact

from conviction of behavioral finance on fund managers' self-assessment. Will proponents be able to control and quell their psychologically motivated biases to rationalize their behavior in order to optimize investment performance?

4. Views on one's own behavior

The view on one's own behavior provides a stunning contrast to the view on the market (see Section 3): We do not find any significant differences between proponents and non-proponents in our sample, with one exception, i.e. the issue of miscalibration.

The answers to the following five items provide pretty consistent results about fund managers' self-assessment. Just as for the view on markets, we follow Hirshleifer (2001) and distinguish the self-assessment items in those of judgment and decision-making (see [Table 5](#)). The first part of this section shortly motivates these items, the second part gives empirical results.

4.1 Motivation

We refer to judgment biases with statement [6], question [7] and request [8] which serve to analyze the three facets of overconfidence as they have been discussed in the literature, e.g. in Glaser and Weber (2005) and Menkhoff et al. (2006). Theory assumes that overconfidence often appears in situations with rare or ambiguous feedback (Hirshleifer, 2001). However, portfolio manager frequently get unmistakable feedback to their decisions and forecasts and they cannot establish themselves in the market with bad performance. In this context it seems plausible to expect that portfolio manager learn from their experience, thus being able to estimate the precision of their knowledge and abilities more adequate (Deaves et al., 2005). But, as indicated by the self-deception theory, the driving forces behind overconfidence may be stronger than rational learning. Many cognitive mechanisms distort an adequate self-assessment, thereby rather supporting overconfidence to keep self-esteem and respect of others after wrong decisions. Glaser et al. (2005) show in this respect that professional investors are even more overconfident than laymen in all the three interpretations of overconfidence.

The first dimension of overconfidence is the *illusion of control* (Langer, 1975) implying the following aspects: The belief of being able to control random variables, such as market trends, more than it is really possible, an unrealistic high estimation of one's own success

probability as well as an exaggerated optimism concerning the future (Glaser and Weber, 2005). The psychological motivation for the illusion of control is the fact that people feel more comfortable in situations they can control. An aspect of this bias is the illusion of being able to explain events retrospectively which is also regarded as the origin of hindsight bias. Thus, we refer to overconfidence as illusion of control by statement [6]. Hindsight bias hinders an efficient information procession and derogates rational learning. Empirical studies show that despite the knowledge of its origins and effects, the hindsight bias cannot be easily reduced and that also professionals are subject to this bias (Biais and Weber, 2006).

One further facet of overconfidence is the unrealistically positive self-evaluation, the so called *better-than-average-effect* (Taylor and Brown, 1988) which we address by question [7]. Especially concerning desirable skills people tend to estimate themselves above average. A prominent example for the better-than-average self-estimation is the study of Svenson (1981), in which 82% of participating students rank themselves to 30% of the safest drivers. Obviously, biased perception concerning one's own skills is in high gear. From one's own experience only others are subject to cognitive deception and behavioral biases. Thereby successes are referred to one's own abilities whereas failures are ascribed to external circumstances (Pronin et al., 2002, Pronin, 2007).

Finally, overconfidence is often regarded as *miscalibration* which we test by using request [8]. Miscalibration is a systematic overestimation of the precision of one's own knowledge. In consequence, people underestimate the probability distribution of a random variable (Lichtenstein et al., 1982). Experimental psychology adopted the method of confidence intervals to measure miscalibration. Investors signalize their miscalibration by stating too narrow confidence intervals for estimations of stock or stock index developments.

In the following we refer to two decision biases. When approaching disposition effect and home bias that we have already touched to assess the market (Section 3), we do not rely on the same items for self-assessment in order to prevent strategic answers by respondents.

Shefrin and Statman (1985) explain the *disposition effect* by the core messages of prospect theory: loss aversion and reflection effect (Kahneman and Tversky, 1979, Weber and Camerer, 1998, Langer and Weber, 2005). Thus, loss aversion is responsible for risk seeking in case of losses and consequently for the tendency to holding loss positions too long. This behavior is attributed to the hope that loser assets will rise or even outperform the actual winners in the future. Contrary, due to risk aversion in the domain of gains the realization of a

a secure profit is preferred to further holding a “winning” asset. This leads to selling winners too soon. Thus, investors change their risk attitude depending on the type of an outcome, the above elucidated reflection effect. We approach the disposition effect by statement [9], thereby in particular concerning risk aversion in the domain of gains.

Statement [10] refers to the *home bias* although with a different item than statement [4] from Table 4 above. We use the hypothetical question on international asset allocation here because fund managers are subject to restrictions of their clients who are not necessarily well informed about an optimal portfolio allocation or may be subject to regulation restricting international diversification. Thus, managed portfolios might differ from that ones preferred by the fund managers themselves. To avoid such a contortion, we ask fund managers for their personally preferred portfolio allocation, explicitly mentioning to ignore their funds’ restrictions. According to the IAPM (International Asset Pricing Model) the optimal portfolio share of a country corresponds to the ratio of its market capitalization to world capitalization. German investors’ portfolios should thus contain 4% of German stocks and 8% of German bonds (Lütje and Menkhoff, 2007).

For all three biases – overconfidence, disposition effect and home bias – there is clear evidence that they reduce risk-adjusted investment performance. In a series of studies the dismal effect of overconfidence has been revealed by Barber and Odean (2000, 2001) (see also Odean, 1999, Biais and Weber, 2006). Regarding the disposition effect, Odean (1998) discloses it as a source of suboptimal performance for individual investors. Shapira and Venezia (2001) find that both professionalism and experience diminish but do not eliminate the preference to realize profits too soon. Literature warns against the consequences of the disposition effect and proposes greater self-discipline as solution to reduce the costs of loss or regret aversion and relative unsuccessfulness (Coval and Shumway, 2005, Locke and Mann, 2005). Finally, there is evidence that international portfolio diversification may increase portfolio returns (depending of course on the benchmark country return) although emphasis is put on the advantage of international diversification due to imperfect return correlation between national markets (see Levy and Sarnat, 1970).

4.2 Results

We now turn to results which are presented in the main part of Table 5. Starting with the bias “illusion of control”, is there any impact from being a proponent of behavioral fi-

nance on the degree of hindsight bias? By assessing statement [6] with a mean of 3.32 proponents of behavioral finance find the majority of economic news slightly more surprising than non-proponents (3.19). Illusion of control is not significantly lower for proponents in this case although this tendency of information perception may seem desirable from the viewpoint of behavioral finance (as well as from the efficient market hypothesis). Thus, our result supports the persistence of the hindsight bias.

Question [7] sheds light on overconfidence as unrealistically positive self-evaluation. In absence of overconfidence it is to expect that fund managers evaluate their own performance compared to that of their colleagues on average with 3, which is the coding for “equally good”. The results in Table 5 show, however, that fund managers evaluate themselves by a mean of 2.34, thus clearly signaling overconfidence. The difference between proponents and non-proponents in this respect is negligible.

Request [8] serves to analyze the relation of miscalibration and the conviction of behavioral finance. In this case respondents had to estimate a 90% confidence interval for the quote of Germany’s main stock index DAX in one month ahead. Interestingly, proponents state a significantly wider interval than non-proponents (means of 1165 and 769 points respectively).⁵ The share of correctly made forecasts is 66.7% in the group of proponents and 47.2% in the group of non-proponents. As the share of wrong estimations exceeds 10% we confirm miscalibration among professional fund managers in our sample (see e.g. Biais et al., 2005, Glaser et al., 2005) but show that miscalibration is significantly lower for proponents of behavioral finance.

Turning to the disposition effect, i.e. statement [9], our groups do not differ in their assessment of the degree to be subject to the disposition effect themselves. They both rather deny it to hold too long on losing positions (means of 4.03 and 4.07 respectively). In contrast, they observe a related bias - the reflection effect- much clearer in the behavior of others (see Section 3, statement [3]).

Finally, also the result for statement [10] does not differ between our two groups. Both groups equally strongly overweigh the home country in their portfolios. This finding is contrasted by results on statement [4] in Table 4, whereupon home bias is observed in fund management.

⁵ Excluding the extreme 10% answers from the distribution, leaving the inner 90%, the Mann-Whitney U-test also shows a significant difference between the answers of the groups in the miscalibration task. The median in the group of proponents is 760 points and 625 points in the group of non-proponents.

All this raises the question why overconfidence as miscalibration seems to be a bias that is easier to overcome than other behavioral patterns? Psychologists claim that people want to think well of themselves and generally succeed in doing so, which is particularly difficult to overcome by learning when asking for a direct comparison between self versus others (Pronin et al., 2002, 2004). Thus, e.g. overconfidence as better-than-average effect or the denial of the disposition effect in one's own behavior can be regarded as consequence of the wish to see oneself in a better light upon which the knowledge of these biases has no significant influence. The contrasting finding on the calibration task could thus be due to its framing. If we regard the ambition to guard self-esteem as motivation to perform well in self-assessment, then in case of miscalibration proponents of behavioral finance are better informed how to attain a good result in this task, i.e. to forecast correctly. Accordingly, due to their greater awareness of miscalibration proponents state significantly wider (and thereby better) confidence intervals than non-proponents.

In summary, we do not observe an impact of the conviction of behavioral finance on one's own behavior. Rather, both groups – proponents and non-proponents – negate behavioral biases like the disposition effect in their own investment behavior. This creates a discrepancy between the view on the market and that on oneself, i.e. a clear disclosure of the bias blind spot. Thus, fund managers do not behave principally different from other people, despite their strong performance control and bonus incentives to rationalize behavior.⁶ Thus, rational learning is difficult in case of self-perception (Ehrlinger et al., 2005) – here shown for fund managers.

5. Consequence on investment behavior

The purpose of this section is to analyze whether the conviction to behavioral finance has any effect on practical investment behavior. As we find such effects, this serves as a credibility test of the bias blind spot – no impact on perception of own behavior – because proponents obviously do behave differently in accordance with their conviction.

In this section investment behavior is grasped by two items, addressing information sources and investment strategies. It is known in this respect that fund managers primarily

⁶ Pronin et al. (2002) as well as Ehrlinger et al. (2005) reveal for “normal” people that even with knowledge of the relevant biases their perception in the behavior of others is always stronger than their recognition in one's own behavior. Furthermore, test persons fail to agree to this biased self-perception and hold on to judge their self-assessment as objective and accurate.

base their decision on fundamental information (e.g. Menkhoff, 1998, Arnswald, 2001). However, due to the increasing recognition of behavioral effects on price formation in financial markets (DeBondt and Thaler, 1985, Jegadeesh and Titman, 2001, Barberis and Thaler, 2003, Shiller, 2003) the importance of non-fundamental information, such as technical analysis and technical trading strategies (e.g. momentum trading) may be even growing (early evidence in Grinblatt et al., 1995, Keim and Madhavan, 1995). Because of their stronger perception of behavioral anomalies in financial markets we expect proponents trying to profitably exploit the observed anomalies and thus to rely more on non-fundamental sources of information and investment strategies.

In detail, [Table 6](#) gives request [11] with the four important sources of information as revealed by fund managers in ex-ante interviews. From the same set of interviews we derive most popular investment strategies as addressed by question [12] in [Table 7](#).

[Table 6](#) shows at first sight the superior role of fundamental facts as source of information for all fund managers. Compared to non-proponents, the group of proponents relatively stronger uses technical indicators (mean of 2.84 vs. 3.26), whereby it assesses colleagues (mean of 3.38) and other market participants (mean of 3.93) as relatively less relevant sources of information. Even if these differences are not significant, the tendency to reject the no-difference hypothesis is relatively obvious in some cases, e.g. regarding the p-value of the Mann-Whitney U-Test of 0.125 for technical indicators. Colleagues and other market participants can be interpreted as sources of external confirmation for own decisions (Arnswald, 2001). Their relatively minor importance for proponents might result from the fact that the latter relatively stronger perceive the behavior of other market participants as biased ([Table 4](#)). Therewith, the quality of these sources of information is lower for proponents. In this sense their conviction of behavioral finance has a recognizable influence, although it is not statistically significant.

[Table 7](#) contains results about the use of the popular investment strategies in fund management. We see that the three main strategies as identified in interviews, i.e. buy-and-hold, momentum and contrarian trading, are of almost equal importance to our sample of fund managers. When we look at proponents only, however, they prefer the technical trading strategies momentum and contrarian trading to a significantly higher degree than non-proponents. This is consistent with their market assessment in [Section 3](#) ([Table 4](#)) where they significantly stronger regard markets as being influenced by behavioral finance effects. In

particular, the more intensive use of the momentum strategy seems plausible regarding the significantly stronger perception of herding in fund management (statement [5] in Table 4).

6. Conclusion

Recent empirical research in the field of behavioral finance has shown that professional investors are (just as well as private ones) subject to irrational, psychologically motivated biases in their investment decisions (e.g. Glaser et al., 2005, Haigh and List, 2005, Menkhoff et al., 2006). Even though professionalism may reduce biases to some degree it does not eliminate them entirely (Shapira and Venezia, 2001, 2006). This paper offers another perspective on the role of behavioral biases in finance by examining the impact from conviction of behavioral finance on the perception of markets' and one's own behavior.

We provide evidence on this issue by examining survey data of more than 100 German fund managers. Distinguishing the respondents in two groups, proponents and non-proponents of behavioral finance, we test for differences in their answers with respect to perceived market-wide biases, self-assessment, information processing and investment strategies. Interestingly, there is no difference in personal characteristics between the two groups. Our findings show that the conviction of behavioral finance has a significant impact on the professionals' perception of markets, but that it hardly influences the view on one's own behavior. This finding holds although proponents do behave differently from non-proponents in the expected direction with respect to preferred information sources and investment strategies.

So far our findings clearly support Hirshleifer's (2001) hypothesis that many behavioral biases are difficult to overcome by learning, even though the fund managers analyzed here have very strong incentives to learn efficient behavior. However, results do not only feed pessimism about the capabilities of financial professionals. One indication of learning is the significantly lower degree of miscalibration shown by proponents. Moreover, their stronger reliance on technical analysis and trading strategies may not conform to expectations derived from the efficient market hypothesis, however, this behavior seems consistent with their conviction of behavioral finance.

The results of our study demonstrate the persistence of behavioral patterns even with knowledge of their existence and thus their dominance over rational learning (in this sample). The failure to recognize own biases prevents from striving to correct them (Pronin, 2007).

The challenge for professional fund managers, in particular for those with insight in behavioral finance, consists therefore in a more critical assessment of own investment behavior in order to discipline and rationalize it.

References

- Arnsward, Torsten (2001): Investment Behaviour of German Equity Fund Managers: An Exploratory Analysis of Survey Data, *Deutsche Bundesbank Discussion Paper* 08/01, Frankfurt.
- Barber, Brad M. and Terrance Odean (2000): Trading Is Hazardous to Your Wealth: The Common Stock Investment Performance of Individual Investors, *Journal of Finance*, 55:2, 773-806.
- Barber, Brad M. and Terrance Odean (2001): Boys Will Be Boys: Gender, Overconfidence, and Common Stock Investment, *Quarterly Journal of Economics*, 116:1, 261-292.
- Barberis, Nicholas and Richard Thaler (2003): A Survey of Behavioral Finance, in: G.M. Constantinidis, M. Harris and R.M. Stulz (eds.), *Handbook of the Economics of Finance*, Vol.1B, Amsterdam et al.: Elsevier, 1053-1123.
- Biais, Bruno, Denis Hilton, Karine Mazurier and Sébastien Pouget (2005): Judgmental Overconfidence, Self-Monitoring and Trading Performance in an Experimental Financial Market, *Review of Economic Studies*, 72:2, 287-312.
- Biais, Bruno and Martin Weber (2006): Hindsight Bias and Investment Performance, *Working Paper*, University of Mannheim.
- Bikhchandani, Sushil and Sunil Sharma (2001): Herd Behavior in Financial Markets: A Review, *IMF Staff Papers*, 47:3, 279-310.
- Coval, Joshua D. and Tyler Shumway (2005): Do Behavioral Biases Affect Prices?, *Journal of Finance*, 60:1, 1-34.
- Deaves, Richard, Erik Lüders and Michael Schröder (2005): The Dynamics of Overconfidence: Evidence from Stock Market Forecasters, Center for European Economic Research (ZEW), *ZEW Discussion Paper* No. 05-83.
- De Bondt, Werner F.M. and Richard Thaler (1985): Does the Stock Market Overreact?, *Journal of Finance*, 40:3, 793-805.
- Ehrlinger, Joyce, Thomas Gilovich and Lee Ross (2005): Peering Into the Bias Blind Spot: People's Assessments of Bias in Themselves and Others, *Personality and Social Psychology Bulletin*, 31:5, 680-692.
- Fama, Eugene F. (1998): Market Efficiency, Long-Term Returns, and Behavioral Finance, *Journal of Financial Economics*, 49:3, 283-306.
- Festinger, Leon (1957): *A Theory of Cognitive Dissonance*, Stanford, CA: Stanford University Press.
- Glaser, Markus, Thomas Langer and Martin Weber (2005): Overconfidence of Professionals and Lay Men: Individual Differences Within and Between Tasks?, *Working Paper* No. 05-25, SFB 504, University of Mannheim.
- Glaser, Markus and Martin Weber (2005): Overconfidence and Trading Volume, *Working Paper*, University of Mannheim (update of CEPR Discussion Paper 3941).
- Grinblatt, Mark, Sheridan Titman and Russ Wermers (1995): Momentum Investment Strategies, Portfolio Performance, and Herding: A Study of Mutual Fund Behavior, *American Economic Review*, 85:5, 1088-1105.

- Haigh, Michael S. and John A. List (2005): Do Professional Traders Exhibit Myopic Loss Aversion? An Experimental Analysis, *Journal of Finance*, 60:1, 523-534.
- Hirshleifer, David (2001): Investor Psychology and Asset Pricing, *Journal of Finance*, 56:4, 1533-1597.
- Jegadeesh, Narasimhan and Sheridan Titman (2001): Profitability of Momentum Strategies: An Evaluation of Alternative Explanations, *Journal of Finance*, 56:2, 699-720.
- Johansson-Stenman, Olof and Peter Martinsson (2006): Honestly, why Are You Driving a BMW?, *Journal of Economic Behavior and Organization*, 60:2, 129-146.
- Kahneman, Daniel and Amos Tversky (1979): Prospect Theory: An Analysis of Decision Under Risk, *Econometrica*, 47:2, 263-292.
- Keim, Donald B. and Ananth Madhavan (1995): Anatomy of the Trading Process, Empirical Evidence on the Behavior of Institutional Traders, *Journal of Financial Economics*, 37, 371-398.
- Kilka, Michael and Martin Weber (2000): Home Bias in International Stock Return Expectations, *Journal of Psychology and Financial Markets*, 1, 3/4, 176-192.
- Langer, Ellen J. (1975): The Illusion of Control, *Journal of Personality and Social Psychology*, 32, 311-328.
- Langer, Thomas and Martin Weber (2005): Myopic Prospect Theory vs. Myopic Loss Aversion: How General is the Phenomenon? *Journal of Economic Behavior and Organization*, 56:1, 25-38.
- Levy, Haim and Marshall Sarnat (1970): International Diversification of Investment Portfolios, *The American Economic Review*, 60:4, 668-675.
- Lichtenstein, Sarah, Baruch Fischhoff and Lawrence D. Phillips (1982): Calibration of Probabilities: The State of the Art to 1980, in: *Judgement under Uncertainty: Heuristics and Biases*, ed. D. Kahneman, P. Slovic and A. Tversky, Cambridge University Press, 306-334.
- Locke, Peter R. and Steven C. Mann (2005): Professional Trader Discipline and Trade Disposition, *Journal of Financial Economics*, 76:2, 401-444.
- Lütje, Torben and Lukas Menkhoff (2007): What Drives Home Bias? Evidence from Fund Managers' Views, *International Journal of Finance & Economics*, 12:1, 21-35.
- Menkhoff, Lukas (1998): The Noise Trading Approach – Questionnaire Evidence from Foreign Exchange, *Journal of International Money and Finance*, 17, 547-564.
- Menkhoff, Lukas, Ulrich Schmidt and Torsten Brozynski (2006): The Impact of Experience on Risk Taking, Overconfidence, and Herding of Fund Managers: Complementary Survey Evidence, *European Economic Review*, 50:7, 1753-1766.
- Nickerson, Raymond S. (1998): Confirmation Bias: Ubiquitous Phenomenon in Many Guises, *Review of General Psychology*, 2:2, 175-220.
- Odean, Terrance (1998): Are Investors Reluctant to Realize Their Losses?, *Journal of Finance*, 53:5, 1775-1798.
- Odean, Terrance (1999): Do Investors Trade Too Much?, *American Economic Review*, 89:5, 1279-1298.

- Pronin, Emily (2007): Perception and Misperception of Bias in Human Judgement, *Trends in Cognitive Sciences*, 11:1, 37-43.
- Pronin, Emily, Thomas Gilovich and Lee Ross (2004): Objectivity in the Eye of the Beholder: Divergent Perception of Bias in Self Versus Others, *Psychological Review*, 111:3, 781-799.
- Pronin, Emily, Daniel Y. Lin and Lee Ross (2002): The Bias Blind Spot: Perceptions of Bias in Self Versus Others, *Personality and Social Psychology Bulletin*, 82:3, 369-381.
- Shapira, Zur and Itzhak Venezia (2001): Patterns of Behavior of Professionally Managed and Independent Investors, *Journal of Banking and Finance*, 25, 1573-1587.
- Shapira, Zur and Itzhak Venezia (2006): On Timing and Herding: Do Professional Investors Behave Differently than Amateurs?, *Working Paper*, European Financial Management Symposium 2006.
- Shefrin, Hersh and Meir Statman (1985): The Disposition to Sell Winners Too Early and Ride Losers Too Long: Theory and Evidence, *Journal of Finance*, 40:3, 777-790.
- Shiller, Robert J. (2003): From Efficient Markets Theory to Behavioral Finance, *Journal of Economic Perspectives*, 17:1, 83-104.
- Svenson, Ola (1981): Are We All Less Risky and More Skilful Than Our Fellow Drivers?, *Acta Psychologica*, 47, 143-148.
- Taylor, Shelley S. and Jonathan D. Brown (1988): Illusion and Well-Being: A Social Psychology Perspective on Mental Health, *Psychological Bulletin*, 103:2, 193-210.
- Thaler, Richard H. and Eric J. Johnson (1990): Gambling with the House Money and Trying to Break Even: The Effects of Prior Outcomes on Risky Choice, *Management Science*, 36:6, 643-660.
- Weber, Martin and Colin F. Camerer (1998): The Disposition Effect in Securities Trading: An Experimental Analysis, *Journal of Economic Behavior and Organization*, 33, 167-184.

TABLE 1. Descriptive statistics of respondents

Item asked	<i>Responses (in per cent)</i>			Number of responses
Age (in years)	<35: 50.9	35-45: 46.5	>45: 2.6	114
Professional experience (in years)	<5: 22.3	5-15: 59.8	>15: 17.9	112
Gender	Male: 92.1	Female: 7.9		114
Share of bonus payment	Mean: 25.8	Std. dev.: 14.5		85
University degree	Yes: 84.8	No: 15.2		112
Governing position	Yes: 36.9	No: 63.1		103
Kind of fund management (primarily)	Active: 93.7	Passive: 6.3		111
Kind of securities managed*	Stocks: 66.5	Bonds: 33.5		115

Note: *4.2% of the respondents managed stocks and bonds to the same degree. These respondents were added with half weight to stocks as well as bonds, so that the sum adds up to 100%.

TABLE 2. Proponents and non-proponents

[BF1] Statement: “The research findings in behavioral finance reflect decision behavior in fund management”.

[BF2] Statement: “I’ve already concerned myself with behavioral finance, the key messages are well known to me”.

		Distribution of responses (frequencies)						
		BF1: Research findings reflect decision behavior in fund management						
		complete approval			complete contradiction			
		1	2	3	4	5	6	Σ
BF2: Key messages of behavioral finance are well known to respondent	1	5	20	7	2	0	-	34
	2	1	11	17	4	2	-	35
	3	0	2	9	5	5	-	21
	4	0	1	4	5	4	-	14
	5	0	0	2	0	1	-	3
	6	-	-	-	-	-	-	-
	Σ	6	34	39	16	12	-	107

TABLE 3. Characteristics of proponents and non-proponents

Statements encoding in []	Response category (a)	Responses			H ₀ : No difference (b)	Binary probit regressions (c)	
		<i>All fund managers</i>	Proponents	Non- proponents			
Age in years	Mean	2.04	2.08	2.09	-0.031	-0.208	-0.151
<35 [1], 35-45 [3], >45 [5]	N	114	37	66	(0.975)	(0.281)	(0.349)
Professional experience in	Mean	2.89	3.00	2.82	-0.725	0.260	0.158
years <5 [1], 5-15, [3], >15 [5]	N	112	34	67	(0.468)	(0.158)	(0.285)
Gender	Mean	1.08	1.05	1.08	-0.437	1.019	0.150
male [1], female [2]	N	115	37	65	(0.662)	(0.209)	(0.802)
Share variable	Mean	25.78	28.35	25.53	-0.728	7.62*10 ⁻⁵	-
compensation in %	N	85	30	49	(0.467)	(0.995)	
University degree	Mean	1.15	1.17	1.12	-0.604	0.211	0.213
yes [1], no [2]	N	112	36	65	(0.546)	(0.667)	(0.593)
Governing position	Mean	1.62	1.63	1.63	-0.077	0.130	-
yes [1], no [2]	N	105	30	64	(0.938)	(0.745)	
Kind of fund management	Mean	1.06	1.03	1.06	-0.718	-0.891	-0.603
(primarily) active [1], passive [2]	N	110	35	65	(0.473)	(0.305)	(0.409)
Kind of securities managed	Mean	1.34	1.30	1.40	-0.986	-0.210	-0.233
stocks [1], bonds [2]	N	114	35	67	(0.324)	(0.556)	(0.434)
R ²						0.062	0.024
[N]						67	93

(a) Responses in respective answering categories with analogical encoding in the first column.

(b) H₀ states that there is no difference between the groups of proponents and non-proponents. The figures given are the z-value of the Mann-Whitney U-test and the p-value in parenthesis.

(c) The table gives the coefficient of the binary probit regression and the p-value in parenthesis.

Stars refer to level of significance: * 10%, ** 5%, *** 1%.

TABLE 4. Assessment of other fund managers' behavior

- [1] Statement: "After several profitable investments fund managers tend to take on additional positions".
 [2] Statement: "My colleagues pay particular attention to confirmatory news/information after having made an investment decision".
 [3] Statement: "In case of loss positions other fund managers tend to increase their willingness to take risks".
 [4] Statement: "Fund managers prefer to invest in near located markets".
 [5] Statement: "Also fund managers exhibit herding behavior".

Statements	Response category (a)	Responses			H ₀ : No difference (b)
		<i>All fund managers</i>	Proponents	Non-proponents	
[1] <i>HOUSE MONEY EFFECT</i>	Mean	2.92	2.59	3.13	-2.107 ^{**}
	N	115	37	67	(0.035)
[2] <i>CONFIRMATORY BIAS</i>	Mean	2.66	2.35	2.91	-2.483 ^{**}
	N	115	37	66	(0.013)
[3] <i>REFLECTION EFFECT</i>	Mean	3.44	3.00	3.71	-2.639 ^{***}
	N	114	37	65	(0.008)
[4] <i>HOME BIAS</i>	Mean	3.34	3.03	3.54	-1.745 [*]
	N	116	37	67	(0.081)
[5] <i>HERDING</i>	Mean	2.02	1.81	2.19	-1.841 [*]
	N	116	37	67	(0.066)

^(a) There are 6 answering categories, ranging from "complete approval" (coded as 1) to "complete contradiction" (coded as 6). Thus, a mean of 3.5 or less indicates rather approval to the statement. N is the number of responses.

^(b) H₀ states that there is no difference between the groups of proponents and non-proponents. The figures given are the z-value of the Mann-Whitney U-test and the p-value in parenthesis.

Stars refer to level of significance: * 10%, ** 5%, *** 1%.

TABLE 5. Assessment of one's own behavior

- [6] Statement: “The majority of economic news is not surprising for me”.
 [7] Question: “How do you evaluate your own performance compared to other asset managers?”
 [8] Request: “Give an estimation of the DAX in one month. Determine a lower und an upper bound such that the quote of the DAX in one month from now will be inside the resulting interval with a probability of 90%”.
 [9] Statement: “I prefer to take profits when I am confronted with unexpected liquidity demands”.
 [10] Request: “Please, allocate an amount of 10 million € to the following markets so that the shares sum up to 100% (thereby ignore your funds' restrictions)”. Germany, Europe (without Germany), USA and Canada, Asia, Emerging Markets.

Statements	Response category (a)	Responses			H ₀ : No difference (b)
		<i>All fund managers</i>	Proponents	Non-proponents	
[6] <i>ILLUSION OF CONTROL (HINDSIGHT BIAS)</i>	Mean	3.20	3.32	3.19	-0.353
	N	115	36	67	(0.724)
[7] <i>BETTER-THAN-AVERAGE</i>	Mean	2.34	2.29	2.37	-0.621
	N	111	35	66	(0.534)
[8] <i>MISCALIBRATION</i>	Mean	887	1165	769	-2.816***
	N	111	37	64	(0.005)

[9] <i>DISPOSITION EFFECT</i>	Mean	3.99	4.03	4.07	-0.014
	N	116	37	67	(0.989)
[10] <i>HOME BIAS (GERMANY)</i> (c)	Mean in %	14.21	14.60	14.30	-0.230
	N	111	36	64	(0.818)

(a) There are 6 answering categories, ranging from “complete approval” (coded as 1) to “complete contradiction” (coded as 6). Thus, a mean of 3.5 or less indicates rather approval to the statement.

Statement 7: 5 answering categories: much better (coded as 1), slightly better (coded as 2), equally good (coded as 3), slightly worse (coded as 4), much worse (coded as 5). N is the number of responses.

(b) H₀ states that there is no difference between the groups of proponents and non-proponents. The figures given are the z-value of the Mann-Whitney U-test and the p-value in parenthesis.

(c) Concerning the weighting of other countries there are also no significant differences between the groups.

Note: All managers' mean answer for statements [9] and [10] lies not between the means of the comparison groups due to adjustment of the group “proponents” by 3 answers (Table 2.).

Stars refer to level of significance: * 10%, ** 5%, *** 1%.

TABLE 6. Assessment of information sources

[11] Request: “Please, assess the importance of the following sources of information for you”.
 Answering categories: “Fundamental facts about the company / market”, “Technical indicators”,
 “Colleagues from the own company”, “Other market participants, not from the own company”.

Sources of information	Response category (a)	Responses			H ₀ : No difference (b)
		<i>All fund managers</i>	Proponents	Non-proponents	
Fundamental facts	Mean	1.82	1.86	1.87	-0.588
	N	115	37	67	(0.557)
Technical indicators	Mean	3.02	2.84	3.26	-1.533
	N	114	37	66	(0.125)
Colleagues	Mean	3.17	3.38	3.07	-1.147
	N	115	37	67	(0.251)
Other market participants	Mean	3.73	3.93	3.61	-1.258
	N	115	37	67	(0.208)

(a) There are 6 answering categories, ranging from “high importance” (coded as 1) to “no importance” (coded as 6). N is the number of responses.

(b) H₀ states that there is no difference between the groups of proponents and non-proponents. The figures given are the z-value of the Mann-Whitney U-test and the p-value in parenthesis.

Stars refer to level of significance: * 10%, ** 5%, *** 1%.

TABLE 7. Assessment of investment strategies

[12] Question: “How intensive do you use several strategies? Please, allocate 100%.”

Answering categories: “___% Momentum strategy”, “___% Buy-and-Hold strategy”,
 “___% Contrarian strategy (Value strategy)”, “___% Others”.

Investment strategies	Response category (a)	Responses			H ₀ : No difference (b)
		<i>All fund managers</i>	Proponents	Non-proponents	
Momentum	Mean in %	26.27	29.49	23.92	-2.033**
	N	108	36	62	(0.042)
Contrarian	Mean in %	29.31	34.28	26.96	-2.395**
	N	108	36	62	(0.017)
Buy-and-Hold (c)	Mean in %	30.73	23.54	32.96	-1.159
	N	108	36	62	(0.243)
Others	Mean in %	13.69	12.69	16.16	-0.113
	N	108	36	62	(0.910)

(a) N is the number of responses.

(b) H₀ states that there is no difference between the groups of proponents and non-proponents. The figures given are the z-value of the Mann-Whitney U-test and the p-value in parenthesis.

(c) The standard deviation for the buy-and-hold strategy is much larger than that for momentum and contrarian strategies, which explains the insignificant difference of groups' answers in this case.

Stars refer to level of significance: * 10%, ** 5%, *** 1%.

APPENDIX 1. Comparison of the sample with the industry's structure

<i>Structure of the German asset management industry:</i>			
- by assets under management			
<i>Structure of data sample:</i>			
- by assets under management	H ₀ : No difference ^(b) -0.325 (0.745)		
- by company size ^(b)			
	Small player	Medium player	Big player
	[12]	[24]	[8]
- by company size ²⁾	5	13	8
H ₀ : no difference ^(c) 2.981 (0.225)			
Participation in the survey:	Pearson correlation coefficient with company size (by asset under management, p-value in parentheses)		
- by number of questionnaires per company	0.472 ^{***} (0.001)		

The market data is based on the annual report of the German investment and asset management association (BVI) as of 2002.

^(a) The table gives the z-value of the Mann-Whitney U-test with the p-value in parentheses.

^(b) Here we focus on the company size of BVI member firms (ex those managing special funds only). The groups are clustered as follows: small players (mutual fund market share < 0.1%), medium players (mutual fund market share <= 2.5%), and big players (mutual fund market share > 2.5%) with numbers given in squared bracket.

^(c) The table gives the test statistic of the chi-square test with the p-value in parentheses.

Stars refer to level of significance: * 10%, ** 5%, *** 1%.