

A Work Project, presented as part of the requirements for the Award of a Master Degree in
Economics from the NOVA – School of Business and Economics

**IS THE EQUITY PREMIUM PUZZLE JUST A LACK OF FORESIGHT?
THE IMPACT OF TARGETING ON MYOPIC LOSS AVERSION**

A Laboratory Experiment

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Abstract

Imagine an individual facing three identical investment decisions in a row. Each time she decides on how much to invest in a risky asset or save. Also, imagine the same individual deciding about three consecutive investments at once. Equal for rational investors, when suffering *Myopic Loss Aversion*, the latter scenario is perceived differently though: More is invested when payoffs are evaluated over a greater time horizon. Based on the theory of reference points I proposed a novel method – investment targets – to shift attention to longer-term goals. I find that exogenously proposed targets eliminate *Myopic Loss Aversion* in an experiment.

Keywords: *Laboratory Experiment, Myopic Loss Aversion, Reference Point, Evaluation Period*

1. Introduction

Following Mehra and Prescott (1985), the *Equity Premium Puzzle* describes historical higher real returns of stocks over governmental bonds in several developed countries. Though this is supposed to reflect the relative risk of stocks compared to “risk-free” government bonds, the rather sizeable premium still requires an unreasonably high level of risk aversion by investors. The most common explanation, termed *Myopic Loss Aversion*, was introduced by Benartzi and Thaler (1995) and is based on a combination of two other behavioral concepts *Loss Aversion* (Tversky and Kahneman 1992) and *Mental Accounting* (Kahneman and Tversky 1984), discovered in studies following Kahneman and Tversky's (1979) *Prospect Theory*. When influenced by *Myopic Loss Aversion*, investors are distinctly more sensitive to short-term losses than to short-term gains.

After an initial experiment by Thaler et al. (1997), Gneezy and Potters (1997) developed the probably most renown laboratory design that confirmed the existence of *Myopic Loss Aversion* in a repeated lottery setup which will be modified for the purpose of my thesis: Individuals have to make an investment decision over an endowment in a lottery with a slightly positive expected

return where some of them have to decide from lottery to lottery (and hence also evaluate in discrete steps) while others have to decide over three consecutive lotteries at once. The experimenters show in line with *Myopic Loss Aversion* a higher investment when individuals commit to three lotteries.

Besides describing an economic phenomenon, it might also be of interest whether there exist mechanisms that help to overcome such biases, especially if agents profit from an increase in expected utility. Based on evidence from a field experiment conducted by Camerer et al. (1997) among cab drivers in New York, one intervention could be the introduction of a prospective reference point in form of an income target, as it might help individuals to contribute rationally to an investment schedule by maintaining self-control. Despite not being computationally easier than fixed savings, it allows a tracking of the invested amount and how much contribution is still needed.

Following the experiment tradition on *Myopic Loss Aversion* initiated by Gneezy and Potters (1997), I conducted their initial laboratory design at Nova SBE in Portugal and extended it by introducing prospective reference points in form of exogenous and endogenous targets. I extended the existing standard theoretical model for the research purpose of this thesis and predict that an artificial prolonging of the evaluation period helps to surpass *Myopic Loss Aversion*. When proposing an exogenous target, I find evidence that the impact of myopic behavior is eliminated. Letting individuals only select an endogenous target is not enough though to overcome *Myopic Loss Aversion* if they were not provided with an exogenous target previously. Those who had such information before are aiming higher and are more likely to reach their endogenous target. Additionally, women overcome an otherwise persistent loss aversion when committing to the latter.

In the following, section two recalls the experimental literature on reference points and *Myopic Loss Aversion*. Section three describes the proposed experiment and predicted behavior. Section four presents the experiment results. Section five concludes and discusses policy implications.

2. Literature review

Despite the fact that the theory of *Myopic Loss Aversion* relies on reference points – losses or gains are always based on benchmarks – the literature on reference points has largely been separate from *Myopic Loss Aversion*. The following review presents the most relevant contributions of both.

Stemming from the *Equity Premium Puzzle* (Mehra and Prescott 1985) and motivated by the discovery of *Myopic Loss Aversion* (Benartzi and Thaler 1995), a series of laboratory experiments has dealt with the phenomenon in a repeated investment framework. In Gneezy and Potters' (1997) initial design, individuals invested a self-selected part of an endowment in a series of twelve sequential lotteries, aware of the underlying winning probabilities. Participants were split into two different information feedback groups, where those treated with high-frequency feedback invested in every lottery separately while those under low-frequency invested ahead into a bundle of three lotteries. *Myopic Loss Aversion* was assumed to be present as those treated by the latter invested on average more per lottery, making a risky option under a longer evaluation period more attractive.

Modifications of the experiment performed in recent years can be summarized in three main findings: Firstly, disentangling the effect of information feedback from the effect of decision flexibility leads to ambiguous results with Bellemare et al. (2005) arguing oppositely to Langer and Weber (2008) that *Myopic Loss Aversion* is driven by the former rather than the latter. Fellner and Sutter (2009) claim an equal effect of both with individual preferences for frequent feedback.

Secondly, when analyzing retrospective reference points, Hopfensitz and Wranik (2008) added reflections on emotions and expectations and find a stronger *Myopic Loss Aversion* for individuals with low self-confidence or previously suffered losses. Hopfensitz (2009) confirms an impact of previous earnings on *Myopic Loss Aversion* and the underlying of a hot hand fallacy. Van der Heijden et al. (2012) find that impatient individuals suffer more from *Myopic Loss Aversion*.

Thirdly, when changing the scope of decision making Haigh and List (2005) and Eriksen and Kvaløy (2010a) show that financial professionals are more *myopic loss averse* than students. Using real financial assets, Gneezy et al. (2003) confirm their previous results (1997) while Beshears et al. (2017) cannot find evidence for *myopic loss averse* in this context. With changed decision rights, Sutter (2007) argues that investing in teams reduces *Myopic Loss Aversion* while Eriksen and Kvaløy (2010b) confirm *Myopic Loss Aversion* when deciding over other agents' endowment.

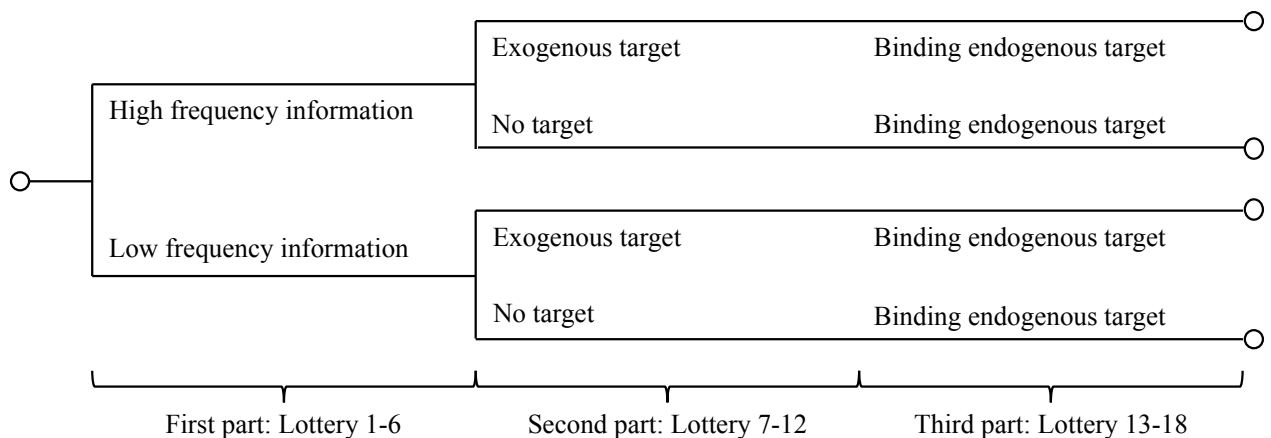
As mentioned above, *Loss Aversion* always has to be related to a reference point (Kőszegi and Rabin, 2006), usually in form of possessed endowment, as a retrospective measure, but also in form of future endowment, as a prospective measure or target. For a long time, individual targeting has concerned economists not from an investment but rather a work income perspective when decisions about labor supply are made under certainty. Camerer et al. (1997), Kőszegi and Rabin (2006), Farber (2005; 2008), Crawford and Meng (2011) and Doran (2014) study the behavior of New York cab drivers and show that younger drivers are loss averse around their daily income target and that a wage raise leads to a reduction in labor supply. Fehr and Goette (2007) argue that Swiss bike couriers reduce their work hours per shift in order to work more shifts and receive a higher salary. Pope and Schweitzer (2011) observe that golf professionals adjust their tournament strokes to payment targets. Abeler et al. (2011) show in a laboratory experiment that participants work longer and earn more if they set higher targets. Gill and Prowse (2012) conduct a real effort, competitive laboratory experiment and find individual loss aversion around endogenous targets.

After showing that individuals are responsive to prospective reference points and that myopic behavior is sensitive to manipulation, we might find positive results when introducing targets to a *Myopic Loss Aversion* experiment. My thesis explores the research question “*Is the equity premium puzzle just a lack of foresight?*” by conducting this novel extension of Gneezy and Potters (1997).

3. Methodology: Laboratory experiment

To assess the impact of targeting on *Myopic Loss Aversion*, a laboratory experiment was designed, planned and conducted at Nova SBE. The experiment consisted of eighteen lotteries structured in three parts and is an extension of Gneezy and Potters' (1997) above described design, which is probably the most well-known and repeated experiment in the context of *Myopic Loss Aversion*. To benchmark this thesis with the current state of research, the first part focuses on replicating *Myopic Loss Aversion* within the experiment. Participants were therefore split into one high-frequency and one low-frequency feedback group. They then received a fixed endowment per lottery over which they made a discrete investment decision with a positive expected return. Moving beyond existing literature, in a second and third part of the experiment two additional treatments were introduced. In part two, half of the participants were informed about the average expected return as an exogenous target and in part three all of them were given the choice to commit themselves to an endogenous, self-selected target. Compared to field experiments which had to infer reference points from behavior (Camerer et al., 1997) and previous laboratory experiments that explicitly ask for personal targets (Abeler et al., 2011; Gill and Prowse, 2012), in my design, I introduced exogenous reference points to the participants. By not relying on self-selected reference points his modification helped to overcome endogeneity issues of the previous targeting literature.

Figure 1: Experiment structure



3.1. Participants

Four experimental sessions were conducted from April 2nd to April 5th, 2018 at Nova SBE with a total of 79 students participating, 50.6% of which were female. Students were invited through a posting on “Facebook”. For each master’s program, an individual group existed on this social media platform, hence targeting every master’s student at Nova SBE equally by publishing the invitation. Participants were aged between 18 and 35, with a mean of 23.5 years. 88.6% of the participants already graduated from their bachelor while only 11.4% were undergraduates and took notice of the experiment through personal referrals. Due to the nature of the undertaking, two-thirds of the participants were economics students, while only one-third studied management or finance. Additionally, 27.8% of the participants attended a behavioral economics class during their studies. Further, 58% of the participants were Portuguese, while the rest consisted of diverse nationalities, including mainly German, Italian, Brazilian and Norwegian students. Table 6 in the appendix displays the population composition per treatment. Based on a Kruskal-Wallis distribution test, randomization of participants was only close to failing for study majors and nationalities.

3.2. Experimental design

After the start of the experiment, participants had to decide on their individual investment in a sequence of eighteen identical and independent lotteries structured in three parts of six lotteries. In every round, participants received an endowment of 10 experimental credit units (ECUs), on which they had to decide in discrete steps (only full ECUs) which amount they wanted to invest in each lottery. The part of the endowment not invested was added to their payoff after the experiment. As it was not allowed to accumulate endowments over several lotteries, the maximum investment per lottery was 10 ECUs. Every lottery had a 2/3 probability of losing this amount and a 1/3 probability of winning 2.5 times the investment, creating a positive expected return. Participants were informed before the experiment about these winning and losing probabilities and the corresponding payoffs.

The first twelve lotteries of the experiment focused on two different manipulations that have been tested in four different treatments: Treatment H (“high-frequency information”) vs. treatment L (“low-frequency information”) was introduced before the first lottery, whereas treatment T (“suggestion of an exogenous target”) vs. treatment N (“no suggestion of an exogenous target”) was added before lottery seven. In the first manipulation under treatment H, the participants played the lotteries one by one. Before lottery one, they had to choose how much of their endowment of 10 ECUs to bet in the lottery and were informed about the realization of the lottery afterward. Next, they could decide on how much of their second endowment of again 10 ECUs to bet for lottery two, and so on. Hence, in this treatment participants made six betting decisions over the course of the first part of the experiment. In treatment L, participants played the lotteries in blocks of three with investments restricted to be equal in order to mimic a single decision. At the beginning of lottery one, subjects decided how much of their 10 ECUs endowment to bet in lottery one, two, and three. After participants committed to their investments, they were informed about the realizations for this bundle of lotteries and were thus not able to assign a gain or loss to a particular lottery, but rather to the combined result. This was repeated for lotteries four, five and six. The idea behind the design of the two treatments was to manipulate the evaluation period: In treatment L, both the frequency of choice and the information feedback were lower than in treatment H and thus participants were expected to evaluate the outcome of their allocation more aggregated. If they were *myopic loss averse*, they should have refused to invest more in the high-frequency lotteries.

In the second part of the experiment, half of the participants from both information frequency treatments were informed before lottery seven, that the expected return over an evaluation period of six lotteries was 70 ECUs when betting the entire endowment. Participants were told that they could consider following this exogenous target but whether or not they reached it had no further consequences. The other half of the participants served as a control group and didn’t receive this

information, leading to the four different treatment groups HxT (“high-frequency information and exogenous target”), HxN (“high-frequency information and no target”), LxT (“low-frequency information and target”) and LxN (“low-frequency information and no target”). Assuming rational investment strategies, neither information frequency nor the provision of an exogenous target should affect individual behavior and participants would bet their full endowment every lottery. However, both the theory on reference points, where individuals evaluate gains and losses not based on their absolute value but rather related to a usually self-selected benchmark as well as the empirical results following Camerer et al. (1997) let us expect that participants would be affected by targets and decide differently. Advantageous over their applied endogenous reference points, an exogenous target might affect individual behavior efficiently and free of endogeneity issues. Further, the effect of an extension of the evaluation period could potentially be stronger than the impact *Myopic Loss Aversion*, making participants invest more (less) of their endowment in the remaining lotteries, should they realize a short-falling (overshooting) of this target.

Finally, before the third part of the experiment, all participants were also encouraged to commit themselves to an endogenous, self-selected target for lotteries thirteen to eighteen. They were informed that if they selected a target, hitting or overshooting it will reward them with a bonus of 5% on exactly the target amount, while failing to do so will not result in a reward but leave the rest of their return untouched. The bonus size itself was chosen to be relevant to make participants actually think about what they could potentially reach yet sufficiently small in order to not distort their behavior and incentivize them to invest more in every lottery just to reach the bonus payment. Additionally, with an endogenous target in the third part, the impact of a previous exogenous target on the former can be tested. Assuming rational investment strategies, no change in behavior should be observed and participants with perfect foresight should set their endogenous target equivalent to an exogenous target of 70 ECUs while adopting a corresponding investment strategy (i.e. invest

the full endowment of 10 ECUs every lottery). Assuming differences across participants (based on their experience, demographics and information frequency), the size of the endogenous target might vary and could have a different impact compared to the exogenous target in part two.

3.3. Remuneration and participant incentives

Unlike the initial experiment by Gneezy and Potters (1997), the earned ECUs were not directly converted into real money after the experiment. Instead, they worked as lottery tickets to win a monetary reward with chances strictly increasing in the ECUs earned relatively to the other participants. 25€ were therefore available for each of the four treatments groups, as in accordance with the experiment's theory the payoffs between the treatment groups were expected to be different. Participants were made aware of the following scheme before the experiment:

$$E(X) = \frac{\Sigma ECU_{individual}}{\Sigma ECU_{allparticipants}} * 25€$$

In order to further increase individual engagement, drinks were prepared as a show-up bonus and ECUs could be performance dependent converted into sweets and fruits after the experiment: A chocolate bar, an apple or a banana were available for 60, a cookie or a peach for 20 and five gummy bears or nuts for 5 ECUs. Therefore, the total ECUs were rounded to the next full five.

3.4. Experimental procedure

Four experiment sessions (one per sub-treatment HxT, HxN, LxT, LxN) were administrated in a classroom by pen and paper with participants seated sufficiently far apart. An assistant supported the conduct of the experiment by distributing the instructions and related questionnaires to the participants upon their arrival. After the procedure was explained by the experimenter, participants could examine the instructions privately, record their ID on the questionnaire and ask questions. Participants were then instructed to record their first bets. Win letters randomly placed on top of

the questionnaire determined their win or loss in every lottery with an almost equal number of participants assigned to win letter A, B or C. To determine the lottery outcome, three balls marked with A, B or C were drawn from a bag. After participants recorded their lottery investment, the ball which determined the winning letter for every lottery was drawn. If a participant's win letter did (did not) match the letter on the ball, he or she won (lost) the lottery. As only one of the three balls matched a participant's win letter the probability of winning (losing) a lottery was 1/3 (2/3). For the high-frequency treatment, the selected ball was put back before the draw for the next lottery happened. The same procedure was repeated three times in a row for the low-frequency treatment.

After every lottery under high-frequency and every third lottery under low-frequency treatment participants calculated and recorded their earnings on the questionnaire. These calculations were checked by the experimenter in order to avoid cheating and to ensure that everyone understood the procedure. Before each of the three experiment parts, participants were informed about changes in the modalities. After lottery eighteen, participants were asked to sum up the three parts and their bonus (in case they reached their endogenous target). When setting their own target for the third part, participants had to note it down on their questionnaire before lottery thirteen. For those affected, the exogenous target was printed on the questionnaire stating the optionality to follow.

3.5. Predicted behavior: Theoretical model

Assuming that they are affected by *Myopic Loss Aversion*, participants weight a loss of x higher than an equal gain of x , illustrated by a weight of $\lambda > 1$ in accordance with Kahneman and Tversky (1979) and Gneezy and Potters (1997). An individual investment utility function can thus be:

$$U(x) = \begin{cases} x & \text{if } x \geq 0 \\ -\lambda x & \text{if } x < 0 \end{cases}$$

With $E(U_i(x))$ as an aggregated distribution value of all independent lottery draws in treatment i , the expected utility for a participant under the high-frequency treatment for an investment x is thus:

$$E(U_H(x)) = \frac{1}{3} * (2.5x) - \lambda_H \frac{2}{3} * (x)$$

which is for a positive x and any $\lambda_H \leq 1.25$ greater than 0. When comparing to the low-frequency information treatment where the outcome is evaluated only every third period, the expected utility:

$$E(U_L(x)) = \frac{1}{27} * (7.5) + \frac{6}{27} * (4) + \frac{12}{27} * (0.5) - \lambda_L \frac{8}{27} * (3)$$

of the invested part of the endowment is again greater than 0 for a positive x and any $\lambda_L \leq 1.56$. When half of the participants are asked in the second part of the experiment to follow the expected return as an exogenous target, the evaluation period for both the affected high and low-frequency information treatments extends to six lotteries and hence the expected investment utility becomes:

$$E(U_T(x)) = \frac{1}{729} * (15x) + \frac{12}{729} * (11.5x) + \frac{60}{729} * (8x) + \frac{160}{729} * (4.5x) + \frac{240}{729} * (x) - \lambda_T \left(\frac{192}{729} * (2.5x) + \frac{64}{729} * (6x) \right)$$

which is again greater than 0 for a positive x and any $\lambda_T \leq 1.84$, allowing even more loss aversion. According to this theoretical model, a longer evaluation period tolerates a higher loss aversion that still allows a positive investment in every lottery. A necessary assumption behind this theory is, that Langer and Weber (2008) and not Bellemare et al. (2005) holds and that *Myopic Loss Aversion* is driven by the evaluation period and rather than commitment. Assuming sufficient computational skills, participants that evaluate their decisions over a prolonged period could thus diminish their loss averse behavior and profit from an increase in expected return (since the latter is positive).

When participants set their own target in part three, they evaluate again the λ -adjusted expected return over six lotteries but now also receive a potential bonus q on their endogenous target T :

$$E(U_C(x)) = \begin{cases} E(U_T(x)) + qT & \text{if } E(x) \geq T \\ E(U_T(x)) & \text{if } E(x) < T \end{cases}$$

This is done by a small fixed percentage addition of $q\%$, that should be of a real size yet still considerably small, in order to make the participants reveal their true target and actually follow it

but not externally incentivize them to target higher they feel reasonable just by an additional effect of the reward itself. Applying a presumably non-distorting 5% bonus on the exact commitment, the expected utility with a target around an expected mean return is again a positive value for all $\lambda_C \leq 1.88$, allowing only for a marginally higher loss aversion than under the exogenous target.

After considering loss aversion under different evaluation periods, the question arises how much of their endowment participants are expected to invest in each lottery. Under risk neutrality, the entire endowment would be invested due to the lottery's positive expected return. Not observing this in the previous experiments, it is most plausible to compare only the investment difference between treatments. Based on the previously determined loss aversion weights λ_i , projections are:

Figure 2: Projected change in invested endowment for different degrees of *Loss Aversion*

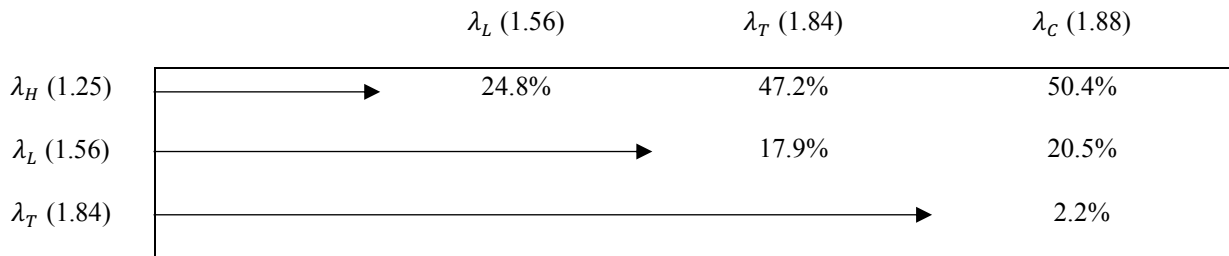


Figure 2 can be interpreted along the following figurative example: Assume that there is a population of 100 individuals with loss aversion coefficients of different magnitude and uniformly distributed from 1.00 to 2.00. Then 25% ($\lambda_H = 1.25$) of the overall endowment of this population will be invested under the high-frequency treatment, 56% ($\lambda_L = 1.56$) will be invested under a low frequency-treatment, and 84% ($\lambda_T = 1.84$) or 88% ($\lambda_C = 1.88$) will be invest with an exogenous or endogenous target. Yet the invested endowment increases sizeable when changing from a high-frequency to a low-frequency treatment (24.8% or from either one an exogenous target (47.2% and 17.9% respectively), the change within the two target types is at 2.2% rather neglectable and in practice probably non-distortive. If the real λ_i is smaller than λ_L , we should not see an effect of targeting on the low-frequency participants as only the high-frequency participants will catch up.

4. Experimental results: The impact of targeting on *Myopic Loss Aversion*

In the following descriptive analysis, the endowment invested per experiment part and in blocks of three lotteries, is compared across the four different sub-treatments. Table 1 presents the average invested endowment per treatment as well as the related Mann–Whitney distribution test results.

Table 1: Average ECUs of the endowment invested per lottery on the treatment level

	Invested Endowment				Mann-Whitney z					
	(ECUs with <i>STDEV italic below</i>)				(two-tailed significance levels with <i>p-values italic below</i>)					
	LxT	LxN	HxT	HxN	LxT LxN	HxT HxN	LxT HxT	LxN HxN	LxN HxT	LxT HxN
Part 1	5.974	5.976	4.517	4.184	0.273	0.253	1.704	2.173	1.740	2.270
	<i>2.870</i>	<i>2.981</i>	<i>3.187</i>	<i>2.875</i>	<i>0.785</i>	<i>0.800</i>	<i>0.088</i>	<i>0.030</i>	<i>0.082</i>	<i>0.023</i>
Lotteries 1-3	5.789	5.905	4.117	3.947	0.222	0.155	-1.816	2.137	1.811	-2.148
	<i>2.955</i>	<i>3.177</i>	<i>3.405</i>	<i>2.976</i>	<i>0.825</i>	<i>0.877</i>	<i>0.069</i>	<i>0.033</i>	<i>0.070</i>	<i>0.032</i>
Lotteries 4-6	6.158	6.048	4.917	4.421	-0.027	-0.634	-1.172	1.742	1.245	-1.808
	<i>3.060</i>	<i>3.025</i>	<i>3.156</i>	<i>3.071</i>	<i>0.978</i>	<i>0.526</i>	<i>0.241</i>	<i>0.082</i>	<i>0.213</i>	<i>0.071</i>
Part 2	6.289	6.333	6.458	4.474	0.123	1.917	-0.142	2.052	-0.211	1.698
	<i>3.114</i>	<i>2.763</i>	<i>3.275</i>	<i>2.661</i>	<i>0.902</i>	<i>0.055</i>	<i>0.887</i>	<i>0.040</i>	<i>0.833</i>	<i>0.090</i>
Lotteries 7-9	5.947	6.048	6.317	4.281	0.206	-2.005	0.483	1.686	-0.238	-1.422
	<i>3.358</i>	<i>3.201</i>	<i>3.132</i>	<i>2.542</i>	<i>0.837</i>	<i>0.045</i>	<i>0.629</i>	<i>0.092</i>	<i>0.812</i>	<i>0.155</i>
Lotteries 10-12	6.632	6.619	6.600	4.667	-0.028	-1.727	0.115	2.100	-0.186	-1.879
	<i>2.967</i>	<i>2.519</i>	<i>3.752</i>	<i>3.008</i>	<i>0.978</i>	<i>0.084</i>	<i>0.908</i>	<i>0.036</i>	<i>0.853</i>	<i>0.060</i>
Part 3	6.289	6.071	6.15	4.316	-0.343	1.790	0.086	1.725	0.013	1.890
	<i>3.501</i>	<i>2.874</i>	<i>3.356</i>	<i>2.695</i>	<i>0.732</i>	<i>0.074</i>	<i>0.932</i>	<i>0.085</i>	<i>0.990</i>	<i>0.059</i>
Lotteries 13-15	6.053	6.190	6.250	4.228	0.028	-1.949	0.115	2.002	-0.093	-1.441
	<i>3.763</i>	<i>3.172</i>	<i>3.310</i>	<i>2.706</i>	<i>0.978</i>	<i>0.051</i>	<i>0.909</i>	<i>0.045</i>	<i>0.926</i>	<i>0.150</i>
Lotteries 16-18	6.526	5.952	6.050	4.404	-0.622	-1.580	-0.418	1.413	-0.158	-1.883
	<i>3.747</i>	<i>3.138</i>	<i>3.662</i>	<i>3.018</i>	<i>0.534</i>	<i>0.114</i>	<i>0.676</i>	<i>0.158</i>	<i>0.874</i>	<i>0.060</i>
N	19	21	20	19	40	39	39	40	41	38

In order to draw comparisons between and within the three experiment parts that go beyond descriptive results, repeated OLS estimations are performed: Part-based with variables aggregated over six lotteries (Table 2) and block-based with an aggregation over three lotteries (Table 7 in the appendix). While the former is the focus of the analysis, the latter serves as a minimum aggregation level as participants under low-frequency feedback are forced to invest three lotteries ahead. OLS is preferred over fixed effects with regressors varying within the experiment and further necessary due to an otherwise low validity of a panel data model with only 79 participants. Yet a Hausman test rejects heteroskedasticity (1% level), estimations use clustered standard errors on a participant level to account for the fact that individual investments are not independent between lotteries.

Table 2: Impact of feedback frequency and targeting on the investment per part

Dependent variable: Investment I_{it}	OLS Estimations					
	(ECUs with standard errors italic below; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$)					
	Information Frequency Only			+Targeting (Future RP)		
	Part 1	Part 2	Part 3	Part 1	Part 2	Part 3
Low-frequency information	1.620** <i>0.669</i>	0.821 <i>0.678</i>	0.919 <i>0.709</i>	1.620** <i>0.669</i>	1.860** <i>0.858</i>	1.496 <i>0.929</i>
Interaction term low-frequency information with exogenous target	-	-	-	-	-2.029 <i>1.335</i>	-1.396 <i>1.393</i>
Exogenous target in part 2	-	-	-	-	1.985** <i>0.953</i>	1.362 <i>1.020</i>
Endogenous target size in part 3	-	-	-	-	-	0.035* <i>0.019</i>
Constant	4.355*** <i>0.481</i>	5.491*** <i>0.499</i>	5.256*** <i>0.505</i>	4.355*** <i>0.481</i>	4.474*** <i>0.610</i>	2.488* <i>1.202</i>
Retrospective RP controls	No	No	No	No	No	No
Demographic controls	No	No	No	No	No	No
R-squared	0.0719	0.0187	0.0213	0.0719	0.0728	0.1013
Dependent variable: Investment I_{it}	+Retrospective Reference Points			+Demographics Control		
	Part 1	Part 2	Part 3	Part 1	Part 2	Part 3
Low-frequency information	1.620** <i>0.699</i>	1.588* <i>0.922</i>	1.616* <i>0.954</i>	1.515** <i>0.688</i>	1.453 <i>1.026</i>	1.543 <i>1.014</i>
Interaction term low-frequency information with exogenous target	-	-1.775 <i>1.389</i>	-1.505 <i>1.407</i>	-	-1.833 <i>1.541</i>	-1.234 <i>1.518</i>
Exogenous target in part 2	-	1.998** <i>0.963</i>	1.329 <i>1.033</i>	-	2.182** <i>1.114</i>	1.646 <i>1.179</i>
Endogenous target size in part 3	-	-	0.035* <i>0.019</i>	-	-	0.028 <i>0.021</i>
Female	-	-	-	-1.383** <i>0.688</i>	-1.548** <i>0.690</i>	-0.944 <i>0.739</i>
Constant	4.355*** <i>0.481</i>	4.562*** <i>0.894</i>	1.860 <i>1.775</i>	8.138** <i>3.411</i>	6.578 <i>4.032</i>	1.779 <i>4.327</i>
Retrospective RP controls	Yes	Yes	Yes	Yes	Yes	Yes
Demographic controls	No	No	No	Yes	Yes	Yes
R-squared	0.0719	0.0777	0.1073	0.1527	0.1732	0.1906

4.1. First part: Testing of Myopic Loss Aversion

In the first part, I replicate Gneezy and Potters (1997) by comparing average investments under high-frequency (HxT, HxN) and low-frequency information (LxT, LxN). The investment within the same feedback frequency is not significantly different (Table 1), rejecting the distribution test null hypothesis, both, by part and in blocks of three lotteries. When comparing investments between feedback frequencies, the null hypothesis can be rejected at the 10% level in all four part-based tests as well as in six of eight block-based tests, which might be attributable to the small sample.

For empirical testing, the invested endowment I_{it} of individual i in part t , is regressed on low frequency treatment dummy L , a time constant vector of demographic controls ζ_i (age, gender, nationality, studies, behavioral economic experience) and in the block-based case also on R_{it} , a set of retrospective reference points (successful previous lotteries and overall accumulated earnings):

$$I_{it} = \beta_0 + \beta_1 L + \beta_2 R_{it} + \beta_3 \zeta_i + u_{it} \quad (\text{I})$$

When including low-frequency information as the only explanatory variable, the result of the point estimator is significant at the 5% level for the first part only (Table 2), while the constant is significant at the 1% level for all three parts, thus matching the descriptive results. Though size and direction of both the constant and the low-frequency treatment are matching the descriptive results, the explanatory power of the model is rather low for all three experiment parts. Testing block-based (Table 7), the effect of low-frequency information is larger and more significant for lotteries one to three than four to six with the explanatory power of the model remaining unchanged. Based on the clear results of the first part, *Myopic Loss Aversion* is assumed to be evident in this experiment.

4.2. Second part: The impact of an exogenous target on *Myopic Loss Aversion*

After introducing the exogenous target in the second part, the invested endowment increases strongly for the affected high-frequency information participants compared to the previous part and the other groups. This is best described as a “complete catching up” of participants affected by high-frequency information feedback with their low-frequency information peers. These results also support Langer and Weber (2008) who argue that *Myopic Loss Aversion* is driven by the evaluation period rather than by commitment. Yet this should be carefully evaluated since they use a slightly different experimental design. As there is no increased investment observed for the low-frequency feedback group, the true loss aversion λ_i of the theoretical model might actually be smaller than λ_L (1.56). An implication of this insight could be that individual foresight actually goes beyond a preassigned lottery evaluation period. Differences in the investment size between and within treatments are mostly significant at the 10% level for both part and block-based testing.

Empirically, regression I is extended by the effect of the exogenous target β_4 . Block-based, the impact of a potential short-falling β_6 is additionally included, with S_{it} indicating the earnings from lottery seven to nine (in case of significance, loss aversion around a target might be underlying):

$$I_{it} = \beta_0 + \beta_1 L + \beta_2 R_{it} + \beta_3 \zeta_i + \beta_4 T + \beta_5 TL + \beta_6 (T - S_{it}) + u_{it} \quad (\text{II})$$

For the second part of the experiment, a significant effect (5% level) of the exogenous target is observed, making affected participants invest on average almost two ECUs more than their peers. After adding the exogenous target, size and significance (5% level) of the low-frequency feedback treatment, as well as the explanatory power of the model for part two, also increase. Testing block-based, the exogenous target effect is prevalent only for lotteries seven to nine, while the impact of feedback frequency remains unchanged. Yet both impact and size of the low-frequency treatment increased for part two, this should not be overinterpreted, as the size of its interaction term with the target is indicative that this effect is reversed for the “LxT” treatment group. Without a significant distance-to-target effect, loss aversion around an exogenous target cannot be supported (Table 7).

4.3. Third part: The impact of an *endogenous* target on *Myopic Loss Aversion*

In the third part, when participants are encouraged to follow a self-selected, endogenous target, the average investment does not change significantly compared to the previous parts. Comparing the size of the endogenous target set between high frequency (58.3) and low frequency (62.2) participants, no significant difference is observed between them, though it is at the 5% level when comparing between those who had an exogenous target before (55.3) and those who have not had (65.3). 76 of 79 participants did not feel any discouragement and selected an exogenous target, of which 46 reached their goal and received the corresponding bonus. Yet differences between the treatments do not change significantly to the second part, when testing part-based, more significant differences are lost when testing block-based. With most significance losses happening in the last lottery block, the two types of targets might work differently, not least as an endogenous target is apparently not enough for those treated with “HxT” to catch up with their peers in the third part.

In the last step, regression II is extended by the endogenous target size, while keeping the exogenous target T as a control. In the block-based case, similar to regression II, from lotteries sixteen to eighteen, the model is again extended by the impact of a potential target short falling β_7 :

$$I_{it} = \beta_0 + \beta_1 L + \beta_2 R_{it} + \beta_3 \zeta_i + \beta_4 T + \beta_5 TL + \beta_6 C_i + \beta_7 (C_i - S_{it}) + \varepsilon_{it} \quad (\text{III})$$

Part-based testing shows that on a 10% significance level, the investment per lottery in the third part increases by 0.035 ECUs for each ECU higher endogenous target size chosen. Similar to the exogenous target, this effect is stronger and more significant (5% level) for lotteries thirteen to fifteen but insignificant for lotteries sixteen to eighteen when testing block-based. One possible explanation for both significance losses could be that individuals lose sight on their target when being influenced by realized outcomes and thus should be remembered repeatedly about them. An additional beneficial trait of the endogenous target is also a gender-specific effect, as an otherwise persistent higher female loss aversion disappears in part three. Despite not entirely replicating the descriptive results without applying additional controls, results are confirmative that an exogenous target helps to overcome *Myopic Loss Aversion* while an endogenous target alone is not sufficient.

With a significant impact of the endogenous target size on the investment, it should also be of interest what determines the commitment. Equation IV regresses the endogenous target size C_i on the feedback frequency L , the exogenous target T , previous earnings Π_i and demographics ζ_i :

$$C_i = \beta_0 + \beta_1 L + \beta_2 \zeta_i + \beta_3 T + \beta_4 \Pi_i + \varepsilon_i \quad (\text{IV})$$

Yet, the exogenous target does not have a direct impact on the investment in the third part under part-based testing, there is however an impact at the 1% significance level on the endogenous target size but no significant impact of the feedback frequency (Table 3). Those provided in part two with an exogenous target on average aim over nine ECUs higher in the third part, and thus come closer to the exogenous target size of 70 ECUs. Significant at the 1% level, women target almost eleven ECUs less than their male peers, which will be further discussed in the robustness check section.

Table 3: Selected endogenous target size

Dependent variable: Endogenous target size C_{it}	OLS Estimations (*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$)	
	Point Estimator (absolute ECUs)	Robust Standard Error
Low-frequency information	2.915	4.070
Exogenous target in part 2	9.323**	3.612
Female	-10.550***	3.636
Constant	81.792**	37.236
Retrospective RP controls	Yes	
Demographic controls	Yes	
R-squared	0.2507	

Besides a lucky streak, the probability of reaching the endogenous target ER_i might also depend on further individual traits, including most of the previous dependent and independent variables. This question is analyzed by a probit model and benchmarked with a linear probability model:

$$ER_i = \beta_0 + \beta_1 L + \beta_2 \zeta_i + \beta_3 T + \beta_4 \Pi_i + \beta_5 C_i + \beta_6 \bar{I}_i + \varepsilon_i \quad (V)$$

Interpreting the marginal effect at means of the probit model (Table 4), chances for those treated by high-frequency feedback reaching the endogenous target increase by 24% when provided with an exogenous target before but decrease by 10% (24%-34%) for those affected by low-frequency feedback previously, which could be related to the different evaluation horizons of the two groups. Also, participants with higher earnings in the previous parts of the experiment are less likely to reach their target. This could be explained by overconfident behavior caused by previous success.

Table 4: Probability of bonus reached

Dependent variable: Probability of bonus reached ER_i	OLS Estimations (*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$)		Probit Estimations (MEAM) (*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$)	
	Point Estimator	Robust SE	Point Estimator	Robust SE
Average investment in part 3	-0.012	0.020	-0.009	0.016
Low-frequency information	0.047	0.156	0.007	0.159
Low-freq. information/exog. target	-0.374	0.229	-0.339*	0.198
Exogenous target in part 2	0.217	0.181	0.241*	0.140
Endogenous target size in part 3	0.002	0.003	0.002	0.003
Total previous earnings	-0.006***	0.001	-0.009***	0.002
Constant	1.504**	0.670	-	-
Retrospective RP controls	Yes		Yes	
Demographic controls	Yes		Yes	
R-squared	0.2614		0.2527 (pseudo-R-squared)	

4.4. Robustness checks

The following section discusses the robustness of feedback frequency and reference points for different controls. While the early literature (Gneezy and Potters, 1997; Bellemare et al., 2005) mainly argued based on descriptive results, later authors (Haigh and List, 2005; Fellner and Sutter, 2009; Hopfensitz and Wranik, 2008) control for robustness by OLS and fixed effect estimations.

4.4.1. Retrospective reference points

As a first robustness check, retrospective reference points have been included. Participants might not only be affected by their expectations but also by their previous experiences. Hopfensitz (2009) finds in the context of *Myopic Loss Aversion* evidence for a *Gamblers Fallacy*, a behavioral bias when individual investment is increased after favorable results of previous lotteries. In my experiment though, I cannot detect an impact of previously accumulated earnings on any testing level. Surprisingly I observe a *Reversed Gamblers Fallacy* (Table 7), with significantly lower investments in the first part after successful previous lotteries: It might be that economics students assess the probability of consecutive successful lotteries differently or that individual targeting behavior actually exists, and individuals prefer riskless savings when the target is within reach. Yet they are mostly insignificant, controlling for retrospective reference points improves the model's explanatory power while decreasing size and significance of the feedback frequency for lotteries four to twelve, underlining the variety of effects that determine individual investment decisions.

4.4.2. Demographic characteristics

With most of the literature applying fixed effect models, little is known about the impact of demographic traits on the investment. Following Table 6 in the appendix, individual characteristics are randomly distributed across treatments. When accounting for them, one of the most interesting insights of my research is that women not only target lower before the third part, as already mentioned but also invest around 1.5 ECUs less over the first two parts while only catching up with

their male peers in part three. Empirically, a higher female loss aversion is among others confirmed by Schmidt and Traub (2002). Decreasing female loss aversion is probably the biggest advantage of the exogenous target over the endogenous one. Further, the model is robust for other demographic traits and increased in explanatory power. Not least, when testing block-based after applying all controls, with a significant distance-to-endogenous target effect, results are speculative for a loss aversion around an individual endogenous target, supporting the proposal of Camerer et al. (1997).

4.4.3. Endogeneity issues in the third part

As participants that invest more logically also aim for a higher return, endogeneity caused by reverse causality between the endogenous target and the investment size might be observed. A solution at hand for this is an instrumental variable approach using the exogenous target from the second part to exactly identify the former. Since there is no significant impact of the exogenous target on the investment size in the third part, exogeneity of the instrument is evident. Knowing from previous results (Table 4) that the exogenous target significantly impacts the endogenous target size, the necessary instrumental relevance is also presumed. Coefficients, however, remain insignificant (though they change in size) when performing the 2SLS estimation (Table 5). Further, a Durbin-Wu-Hausman test fails to reject exogeneity null hypothesis of the endogenous target size, hence the endogenous target size is – against the meaning of its name – not entirely endogenous. Despite different target sizes, a “common sense” between the treatments might be not too far apart.

Table 5: IV of the third part of the experiment (endog. target size instrument by exog. target)

Dependent variable: Investment I_{it}	2SLS Estimations (*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$)		OLS Estimations (*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$)	
	Absolute ECUs	Robust SE	Absolute ECUs	Robust SE
Low-frequency information	0.762	1.122	1.543	1.014
Low-freq. information exog. target	-0.346	1.539	-1.116	1.528
Exogenous Target in part 2	<i>Instrument for the endogenous target size</i>		1.683	1.116
Endogenous target size in part 3	0.168	0.111	0.027	0.022
Constant	-8.818	11.040	2.243	4.327
Retrospective RP controls	Yes		Yes	
Demographic controls	Yes		Yes	
R-squared	-		0.1788	

5. Conclusion

This thesis presented a behavioral economic laboratory experiment on the impact of prospective reference points on *Myopic Loss Aversion* and was conducted with 79 students at Nova SBE. Following a design based on Gneezy and Potters (1997), I added insights on targeting behavior discovered by Camerer et al. (1997) to the original experiment. In line with the literature, evidence for *Myopic Loss Aversion* was found through a treatment with different information frequencies.

Assigning the exogenous target only to a part of the participants, the goal was to assess whether an artificially prolonged evaluation period helps individuals to invest more of their endowment in the lottery. Consequently, I showed that those affected by high-frequency feedback and provided with an exogenous target in the second part managed to completely catch up with the low-frequency feedback group. Against predictions of the theoretical model, the latter did not increase their investment, giving room to speculate that there might be some kind of individual foresight. My results can also carefully be interpreted to support Langer and Weber's (2008) findings, who show that in fact the evaluation period drives *Myopic Loss Aversion* and not the commitment horizon.

Those who previously have had an exogenous target also selected a higher endogenous target and are more likely to reach it, supporting the hypothesis that prospective reference points reduce *Myopic Loss Aversion*. Yet an endogenous target is not enough to overcome myopic behavior, it helps to surpass an otherwise persistent female loss aversion, though they still target less than men. After including demographics and retrospective reference points, I find evidence supporting loss aversion around the endogenous target, which is – against the implications of its name – exogenous.

Though the effect of targeting is pointing towards the hypothesized direction, it is important to note that like in all laboratory experiments, issues of external validity might remain with both the endowment type and the participant group not being representative, leaving room to repeat the setup with more participants, a real monetary endowment or in a context outside of the university.

If the proposed effects are valid and *Myopic Loss Aversion* is truly reduced by prospective reference points, agents invested in assets with varying evaluation periods should be encouraged to assess their investment strategy rather based on its contribution to a long-term goal than on recent developments. This insight could help not only managers to guide their investment professionals in the evaluation of financial returns but also policymakers to provide such targets to their citizens when designing private pension and healthcare saving plans. The results of this experiment suggest that the equity premium puzzle can be partially solved by applying more individual foresight.

6. Literature

- Abeler, Johannes, Armin Falk, Lorenz Goette, and David Huffman. 2011. "Reference Points and Effort Provision." *American Economic Review* 101 (2): 470–92.
<https://doi.org/10.1257/aer.101.2.470>.
- Bellemare, Charles, Michaela Krause, Sabine Kröger, and Chendi Zhang. 2005. "Myopic Loss Aversion: Information Feedback vs. Investment Flexibility." *Economics Letters* 87 (3): 319–24.
<https://doi.org/10.1016/j.econlet.2004.12.011>.
- Benartzi, Shlomo, and Richard Thaler. 1995. "Myopic Loss Aversion and the Equity Premium Puzzle." *Quarterly Journal of Economics* 110 (1): 73–92.
<https://doi.org/10.2307/2118511>.
- Beshears, John, James J. Choi, David Laibson, and Brigitte C. Madrian. 2017. "Does Aggregated Returns Disclosure Increase Portfolio Risk Taking?" *Review of Financial Studies* 30 (6): 1971–2005. <https://doi.org/10.1093/rfs/hhw086>.
- Camerer, Colin, Linda Babcock, George Loewenstein, and Richard Thaler. 1997. "Labor Supply of New York City Cabdrivers: One Day at a Time." *The Quarterly Journal of Economics* 112 (2): 407–41.
<https://doi.org/10.1162/003355397555244>.
- Crawford, Vincent P, and Juanjuan Meng. 2011. "New York City Cabdrivers' Labor Supply Revisited: Reference-Dependent Preferences with Rational-Expectations Targets for Hours and Income." *American Economic Review* 101 (5): 1912–32.
<https://doi.org/10.1257/aer.101.5.1912>.
- Doran, Kirk. 2014. "Are Long-Term Wage Elasticities of Labor Supply More Negative than Short-Term Ones?" *Economics Letters* 122 (2). Elsevier B.V.: 208–10.
<https://doi.org/10.1016/j.econlet.2013.11.023>.
- Eriksen, Kristoffer W., and Ola Kvaløy. 2010a. "Do Financial Advisors Exhibit Myopic Loss Aversion?" *Financial Markets and Portfolio Management* 24 (2): 159–70.
<https://doi.org/10.1007/s11408-009-0124-z>.
- . 2010b. "Myopic Investment Management." *Review of Finance* 14 (3): 521–42.
<https://doi.org/10.1093/rof/rfp019>.
- Farber, Henry S. 2005. "Is Tomorrow Another Day? The Labor Supply of New York City Cabdrivers." *Journal of Political Economy* 113 (1): 46–82. <https://doi.org/10.1086/426040>.
- . 2008. "Reference-Dependent Preferences and Labor Supply: The Case of New York City Taxi Drivers." *American Economic Review* 98 (3): 1069–82.
<https://doi.org/10.1257/aer.98.3.1069>.
- Fehr, Ernst, and Lorenz Goette. 2007. "Do Workers Work More If Wages Are High? Evidence from a Randomized Field Do Workers Work More If Wages Are High? Evidence from a Randomized Field Experiment." *The American Economic Review* 97 (1): 298–317.
<https://doi.org/10.1257/aer.97.1.298>.
- Fellner, Gerlinde, and Matthias Sutter. 2009. "Causes, Consequences, and Cures of Myopic Loss Aversion—An Experimental Investigation." *The Economic Journal* 119 (537): 900–916.
<https://doi.org/10.1111/j.1468-0297.2009.02251.x>.
- Gill, David, and Victoria Prowse. 2012. "A Structural Analysis of Disappointment Aversion in a Real Effort Competition." *American Economic Review* 102 (1): 469–503.
<https://doi.org/10.1257/aer.102.1.469>.
- Gneezy, Uri, Arie Kapteyn, and Jan Potters. 2003. "Evaluation Periods and Asset Prices in a Market Experiment." *Journal of Finance* 58

- (2): 821–37. <https://doi.org/10.1111/1540-6261.00547>.
- Gneezy, Uri, and Jan Potters. 1997. “An Experiment on Risk Taking and Evaluation Periods.” *The Quarterly Journal of Economics* 112 (2): 631–45. <https://doi.org/10.1162/003355397555217>.
- Haigh, Michael S, and John A List. 2005. “Do Professional Traders Exhibit Myopic Loss Aversion? An Experimental Analysis.” *The Journal of Finance* 60 (1): 523–34. <https://doi.org/10.1111/j.1540-6261.2005.00737.x>.
- Heijden, Eline van der, Tobias J. Klein, Wieland Müller, and Jan Potters. 2012. “Framing Effects and Impatience: Evidence from a Large Scale Experiment.” *Journal of Economic Behavior and Organization* 84 (2). Elsevier B.V.: 701–11. <https://doi.org/10.1016/j.jebo.2012.09.017>.
- Hopfensitz, Astrid. 2009. “Previous Outcomes and Reference Dependence : A Meta Study of Repeated Investment Tasks with and without Restricted Feedback.” *TSE Working Paper Series*. Vol. 87. <http://econpapers.repec.org/paper/tsewpaper/22194.htm>.
- Hopfensitz, Astrid, and Tanja Wranik. 2008. “Psychological and Environmental Determinants of Myopic Loss Aversion.” *MPRA Working Paper*. http://mpra.ub.uni-muenchen.de/9305/1/MPRA_paper_9305.pdf.
- Kahneman, Daniel, and Amos Tversky. 1979. “Prospect Theory: An Analysis of Decision under Risk.” *Econometrica: Journal of the Econometric Society* 47 (3): 263–91. <https://doi.org/10.1111/j.1536-7150.2011.00774.x>.
- . 1984. “Choices, Values, and Frames.” *American Psychologist*, no. 39: 341–50. <https://doi.org/10.1037/0003-066X.39.4.341>.
- Köszegi, Botond, and Matthew Rabin. 2006. “A Model of Reference-Dependent Preferences.” *Quarterly Journal of Economics* 152 (4): 1133–65. <https://doi.org/10.1093/qje/qjt005>.
- Langer, Thomas, and Martin Weber. 2008. “Does Commitment or Feedback Influence Myopic Loss Aversion?. An Experimental Analysis.” *Journal of Economic Behavior and Organization* 67 (3–4): 810–19. <https://doi.org/10.1016/j.jebo.2006.05.019>.
- Mehra, Rajnish, and Edward C. Prescott. 1985. “The Equity Premium: A Puzzle.” *Journal of Monetary Economics* 15 (2): 145–61. [https://doi.org/10.1016/0304-3932\(85\)90061-3](https://doi.org/10.1016/0304-3932(85)90061-3).
- Pope, Devin G., and Maurice E. Schweitzer. 2011. “Is Tiger Woods Loss Averse? Persistent Bias in the Face of Experience, Competition, and High Stakes.” *American Economic Review* 101 (1): 129–57. <https://doi.org/10.1257/aer.101.1.129>.
- Schmidt, Ulrich, and Stefan Traub. 2002. “An Experimental Test of Loss Aversion.” *The Journal of Risk and Uncertainty* 25 (3): 233–49. <https://doi.org/10.1023/A:1020923921649>.
- Sutter, Matthias. 2007. “Are Teams Prone to Myopic Loss Aversion? An Experimental Study on Individual versus Team Investment Behavior.” *Economics Letters* 97 (2): 128–32. <https://doi.org/10.1016/j.econlet.2007.02.031>.
- Thaler, Richard, Amos Tversky, Daniel Kahneman, and Alan Schwartz. 1997. “The Effect of Myopia and Loss Aversion on Risk Taking: An Experimental Test.” *The Quarterly Journal of Economics* 112 (2): 647–61. <https://doi.org/10.1162/003355397555226>.
- Tversky, Amos, and Daniel Kahneman. 1992. “Advances in Prospect-Theory - Cumulative Representation of Uncertainty.” *Journal of Risk and Uncertainty* 5 (4): 297–323. <https://doi.org/10.1007/Bf00122574>.

7. Appendix

7.1. Additional statistical tables

Table 6: Distribution of demographic characteristics over the four treatment groups

	LxT (STDEV)	LxN (STDEV)	HxT (STDEV)	HxN (STDEV)	K-Wallis χ^2 (p-value)
Female (100%)	0.47 (0.51)	0.48 (0.51)	0.50 (0.51)	0.57 (0.51)	0.4 (0.937)
Economic Studies (100%)	0.89 (0.31)	0.52 (0.51)	0.50 (0.51)	0.74 (0.45)	6.1 (0.107)
Graduate degree (100%)	0.84 (0.37)	1.00 (0.00)	0.85 (0.36)	0.84 (0.37)	1.1 (0.777)
Portuguese (100%)	0.84 (0.37)	0.42 (0.51)	0.60 (0.50)	0.47 (0.51)	6.0 (0.114)
Age	22.7 (2.2)	23.7 (1.2)	23.1 (2.5)	23.5 (3.6)	3.0 (0.398)
Behav. Econ. Course (100%)	0.37 (0.50)	0.19 (0.40)	0.35 (0.49)	0.21 (0.41)	1.5 (0.682)

Table 7 (next page): Impact of feedback frequency and targeting on the investment per block

OLS Estimations						
<i>(robust standard errors in italic; *** p<0.01, ** p<0.05, * p<0.1)</i>						
Information frequency only						
Dependent var.: Investment I_{it}	L 1-3	L 4-6	L 7-9	L 10-12	L 13-15	L 16-18
Low-frequency information	1.816*** (0.698)	1.424** (0.685)	0.675 (0.702)	0.967 (0.706)	0.860 (0.741)	0.977 (0.769)
Constant	4.034*** (0.506)	4.675*** (0.494)	5.325*** (0.481)	5.658*** (0.561)	5.265*** (0.506)	5.248*** (0.548)
Retrospective controls	No	No	No	No	No	No
Demographic controls	No	No	No	No	No	No
R-squared	0.0810	0.0532	0.0118	0.0240	0.0172	0.0206
+Targeting (Forward-looking RP)						
Dependent var.: Investment I_{it}	L 1-3	L 4-6	L 7-9	L 10-12	L 13-15	L 16-18
Low-frequency information	1.816*** (0.698)	1.424** (0.685)	1.766* (0.910)	1.952** (0.888)	1.617 (0.980)	1.356 (0.955)
Low-freq. information/ex. target	-	-	-2.136 (1.383)	-2.011 (1.420)	-1.866 (1.444)	-0.964 (1.502)
Exogenous target in part 2	-	-	2.036** (0.911)	1.189 (1.654)	1.393 (1.005)	1.319 (1.104)
Distance to exogenous target	-	-	-	0.020 (0.032)	-	-
Endogenous target size in part 3	-	-	-	-	0.047** (0.021)	0.005 (0.023)
Distance to endogenous target	-	-	-	-	-	0.032 (0.021)
Constant	4.034*** (0.506)	4.675*** (0.494)	4.280*** (0.582)	4.667*** (0.694)	1.791 (1.257)	3.415** (1.310)
Retrospective controls	No	No	No	No	No	No
Demographic controls	No	No	No	No	No	No
R-squared	0.0810	0.0532	0.0651	0.0820	0.1250	0.0977
+Retrospective Reference Points						
Dependent var.: Investment I_{it}	L 1-3	L 4-6	L 7-9	L 10-12	L 13-15	L 16-18
Low-frequency information	1.782** (0.698)	0.715 (1.004)	2.074 (1.277)	2.058 (1.448)	1.571 (1.102)	0.507 (1.128)
Low-freq. information/ex. target	-	-	-2.491 (1.748)	-2.171 (1.401)	-1.610 (1.700)	-0.132 (1.566)
Exogenous target in part 2	-	-	2.110** (0.983)	-0.145 (1.861)	1.905* (1.051)	1.116 (1.074)
Distance to exogenous target	-	-	-	0.062 (0.041)	-	-
Endogenous target size in part 3	-	-	-	-	0.058** (0.026)	-0.031 (0.030)
Distance to endogenous target	-	-	-	-	-	0.093** (0.040)
Success in previous lottery	-0.897 (0.700)	-1.831* (1.109)	0.221 (1.461)	0.540 (1.384)	0.258 (1.007)	0.002 (0.011)
Constant	4.471*** (0.628)	5.727*** (1.085)	3.520 (2.379)	3.331 (2.377)	0.349 (1.884)	1.810 (2.384)
Retrospective controls	Yes	Yes	Yes	Yes	Yes	Yes
Demographic controls	No	No	No	No	No	No
R-squared	0.1006	0.0976	0.0705	0.1043	0.1536	0.1454
+Demographic Controls						
Dependent var.: Investment I_{it}	L 1-3	L 4-6	L 7-9	L 10-12	L 13-15	L 16-18
Low-frequency information	1.736*** (0.719)	0.368 (1.187)	1.970 (1.359)	1.883 (1.520)	1.775 (1.089)	0.604 (1.145)
Low-freq. information/ex. target	-	-	-2.286 (1.898)	-2.393 (1.515)	-1.718 (1.688)	0.124 (1.502)
Exogenous target in part 2	-	-	2.223* (1.152)	0.963 (2.215)	2.672*** (1.122)	1.548 (1.184)
Distance to exogenous target	-	-	-	0.035 (0.046)	-	-
Endogenous target size in part 3	-	-	-	-	0.050* (0.021)	-0.015 (0.026)
Distance to endogenous target	-	-	-	-	-	0.070* (0.038)
Success in previous lottery	-1.235* (0.744)	-2.307* (1.234)	0.903 (1.522)	0.312 (1.507)	0.257 (1.070)	-1.277 (1.112)
Female	-1.419** (0.722)	-1.390* (0.727)	1.673*** (0.770)	-1.563** (0.759)	-0.675 (0.807)	-0.935 (0.820)
Constant	9.762** (4.178)	9.548** (4.801)	5.990 (0.529)	4.041 (4.550)	3.900 (4.740)	0.179 (4.803)
Retrospective controls	Yes	Yes	Yes	Yes	Yes	Yes
Demographic controls	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.1973	0.1941	0.1735	0.1845	0.2371	0.2517

7.2. Additional figures

Figure 3: Development of the experimental history on *Myopic Loss Aversion*

Authors	Country (town) & method	Observations High /Low	#Rounds	€ value of the endowment/lottery	Main MLA findings
Gneezy and Potters (1997)	Netherlands (Tilburg); paper and pencil	41/42 students	12	0.90€	Baseline experiment, proof of the underlying of <i>Myopic Loss Aversion (MLA)</i>
Gneezy et al. (2003)	Netherlands (Tilburg, Amsterdam); Computer	40/40 traders	15	0.90€	Under <i>MLA</i> market prices of risky financial assets are significantly higher
Bellmare et al. (2005)	Netherlands (Tilburg); computer	44/44 students	9	0.70€	<i>MLA</i> is persistent even if the information frequency is disentangled from decision making
Haigh and List (2005)	US (University of Maryland); paper and pencil	32/32 students 27/27 financial markets trader	9	0.80€ (students) 3.20€ (traders)	Professional traders exhibit <i>MLA</i> to a greater extent than students
Sutter (2007)	Germany (Jena); computer	64/62 students 28/28 teams	9	1.00€	Team decision making reduces <i>MLA</i> but cannot completely overcome it
Langer and Weber (2008)	Germany (Mannheim); computer	54/53 students	30	25€ for the entire experiment (self-allocation/lottery)	If the feedback is separated from commitment, only the latter affects <i>MLA</i> strongly
Hopfensitz and Wranik (2008)	Switzerland (Geneva); Computer	38/39 students	15	0.90€	<i>MLA</i> stronger when initial investment round lead to loses
Fellner and Sutter (2009)	Germany (Jena); computer	30/30 students	18	0.5€	Both investment horizon and feedback frequency contribute equally to <i>MLA</i>
Eriksen and Kvaøy (2010a)	Norway (Stavanger); computer	80/80 students 24/26 financial advisors	9	0.25€ (students) 1.00€ (fin. advisors)	Financial advisors exhibit <i>MLA</i> to a greater extent than students
Eriksen and Kvaøy (2010b)	Norway (Stavanger); computer	205/205 students	9	0.25€	<i>MLA</i> also exists over other participants endowment
Van der Heijden et al. (2012)	Netherlands (countrywide); computer	551/551 citizens (representative for population)	3	2.00€	More impatient people are stronger affected by <i>MLA</i>
Beshears et al. (2017)	US (Harvard University); computer	160/160 students	18	1.60€	<i>MLA</i> is not persistent if the experiment is conducted with real financial assets
Own Work Project (2018)	Portugal (Lisbon); paper and pencil	39/40 students	18	0.20€ (approximate material value)	Exogenous targets can help to overcome <i>MLA</i> and set a higher endogenous target

7.3. Experiment instructions

Dear Student,

Thank you again for coming today and welcome to this experimental session! For a smooth procedure of the experiment, I would kindly ask you to carefully read the instructions below and follow them as strictly as possible.

Just be yourself: Take the time to think and answer to the best of your capabilities. There is no wrong answer and there are no tricks. All your answers will be treated with the uttermost care and all your responses will be completely anonymous!

During the experiment, it is not allowed to communicate with other participants. If you have a question, please raise your hand and we will approach you in silence. If a question is relevant for other participants, we will answer it aloud for everyone.

You should have 4 pages, something to write and a calculator. Please fill in your 5-digit student ID and your name on top of the sheet.

Part one (*read out loud before lottery seven*)

This experiment is structured as a betting game with 18 successive lotteries and is divided into three parts with six lotteries each. In each lottery, you will start with an amount of 10 Experimental Credit Units (ECUs). You must decide which part of this amount (between 0 and 10 ECUs) you wish to bet in each of the following lotteries. You have a chance of $2/3$ (67%) to lose the amount you bet and a chance of $1/3$ (33%) to win two and a half times the amount you bet. You are requested to record your choice on your questionnaire. Suppose that you decide to bet an amount of X ECUs ($0 \leq X \leq 10$) in the lottery. Then you must fill in the amount X in the column headed “Bet amount” for every lottery. The remaining amount $(10-X)$ you note in “Saved amount”.

Only for the low-frequency treatment: Also, you fix your choice for the next three lotteries. Thus, if you decide to bet an amount X in the lottery 1, then you also bet an amount X for the following lotteries 2 and 3. Therefore, three consecutive lotteries are joined together on the questionnaire.

Whether you win or lose in the lottery depends on your personal win letter. This letter is indicated on top of your questionnaire. Your win letter can be A, B or C, and is the same for all 18 lotteries. In any round, you win the lottery if your win letter matches the lottery letter randomly drawn by the assistant (each with a probability of $1/3$), and you lose vice versa if your win letter does not match. Please always record the letter drawn by the assistant in the column “Outcome of lottery”.

Further instructions for the high frequency-treatment:

The lottery letter is determined as follows: After you have recorded your bet for the next lottery, the assistant will, in a random manner, for each lottery pick one ball from a bag containing three letters: A, B, and C. The letter is drawn, clearly shown to all participants and put back in the bag. Then the bag is shaken again, and the next draw follows. The letter drawn is the lottery letters for the present lottery. The letters are drawn after you have made your bet every lottery. If the lottery letter matches your win letter, you win the lottery, otherwise, you lose. The chance of winning a given lottery is $1/3$ (33%) and the chance of losing is $2/3$ (67%). Your earnings in the lottery are thus determined as follows: If you have decided to put an amount of X ECUs in the lottery, your earnings in the lottery are equal to 0 for each lottery letter that does not match your win letter (i.e. you lose the amount bet for the lottery) and equal to $3.5X$ for each lottery letter that matches your win letter (i.e. you win two and a half times the amount bet for the round).

Further instructions for the low-frequency treatment:

The lottery letter is determined as follows: After you have recorded your bet for the next three lotteries, the assistant will, in a random manner, for each of the next three lotteries pick one ball from a bag containing three letters: A, B, and C. The letter is drawn, clearly shown to all participants and put back in the bag. Then the bag is shaken again, and the next draw follows. The three letters drawn are the lottery letters for the present three lotteries. The letters are drawn after you have made your bet every round. If the lottery letter matches your win letter, you win the lottery, otherwise, you lose. The chance of winning a given lottery is $1/3$ (33%) and the chance of losing is $2/3$ (67%). Your earnings for the three lotteries are thus determined as follows: If you have decided to put an amount of X ECUs in the lottery, your earnings in the lottery are equal to 0 for each lottery letter that does not match your win letter (i.e. you lose the amount bet for the lottery) and equal to $3.5X$ for each lottery letter that matches your win letter (i.e. you win two and a half times the amount bet for the round).

You need to record your earnings in the lottery in the column “Return on betted amount” for every lottery. For every lottery calculate the sum of “Saved amount” and “Return on bet amount” in the column “Result of the lottery” After lottery six has been completed, you have to sum up the column “Total earnings of the lottery” in the field “Total”. This amount determines your total earnings for part 1 of the experiment. Then the instructions for part 2 of the experiment will be announced.

High-frequency treatment: Fill-in example, supposed you are group A

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 1	10	8	2	A	7	15
Lottery 2	10	5	5	B	0	5
Lottery 3	10	10	0	C	0	10
Lottery 4	10	3	7	C	0	3
Lottery 5	10	9	1	B	0	9
Lottery 6	10	4	6	A	21	25
Total						67

Sum up
Multiply by 3.5 if letters match

Though this scheme only refers to the first part of the experiment (lotteries 1 to 6), it also applies in the same way to the next two parts. Remember that the bet amounts are just exemplary and shall not be seen as a strategic suggestion!

Low-frequency treatment: Fill-in example, supposed you are group A

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 1	10	8	2	A	7	15
Lottery 2	10			B	0	8
Lottery 3	10	3	7	C	0	8
Lottery 4	10			C	0	3
Lottery 5	10			B	0	3
Lottery 6	10			A	24.5	27.5
Total						64.5

Sum up
Multiply by 3.5 if letters match

Though this scheme only refers to the first part of the experiment (lotteries 1 to 6), it also applies in the same way to the next two parts. Remember that the bet amounts are just exemplary and shall not be seen as a strategic suggestion!

- Note: Please stop reading here! We will continue after the first part of the experiment! Do not unfold the questionnaire until I tell you! -

Part two (read out loud before lottery seven)

For the exogenous target treatment group: For the next six lotteries, please be aware that the overall expected result should you bet every lottery your entire endowment of 10 ECUs is 70 ECUs. If you like, you can try and reach this target! Other rules do not change!

For the control group (no exogenous target): Part one of the experiment will now be repeated for the second part. No rules will be changed.

Part three (read out before lottery thirteen)

You now have the chance to set your own target for your total earnings in ECU for the final six lotteries. You receive a bonus of +5% on this target if you reach it and are deducted -5% if you fail to do so. Then practically, in the former, you multiply the result of row “Total” in part three by 0.05 and -0.05 in the latter.

After lottery 18, you also sum up all the “Totals” of all three parts, including your bonus (if you selected a target for the last part) in part three.

Fill-in example

Part one

Total	67
-------	----

Part two

Total	64.5
-------	------

Part three (assuming you targeted 50)

Total	70
Bonus	2.5
Overall	204

Pay-off after the experiment

After the experiment, you are allowed to convert the ECUs into the sweets offered as rewards. You can have a chocolate bar, a banana or an apple for 60 ECUs, a cookie for 20 ECUs and a gummy bear for 5 ECUs. Therefore only, your ECUs will be rounded up to the next full five ECUs.

Additionally, the ECUs earned will serve as a ticket in the lottery for the 25€ monetary reward. Therefore, the total number of ECUs you have earned will be divided by the number of ECUs earned by all participants. Weighted by this the chances for you winning the monetary reward thus increase with your total number of ECUs as the probability of one of yours drawn is simply higher.

7.4. Experiment questionnaire

Questionnaire 1: Treatment Group High-Frequency Information/No Exogenous Target (HxN)

Name _____

Student ID _____

Part 1

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 1	10					
Lottery 2	10					
Lottery 3	10					
Lottery 4	10					
Lottery 5	10					
Lottery 6	10					
Total						

Part 2

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 7	10					
Lottery 8	10					
Lottery 9	10					
Lottery 10	10					
Lottery 11	10					
Lottery 12	10					
Total						

Part 3

You now have the chance to set your own target for your total earnings in ECU for the final six lotteries. You receive +5% on this target if you reach it and get no bonus if you fail to do so.

Yes (Target: _____)

No

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 13	10					
Lottery 14	10					
Lottery 15	10					
Lottery 16	10					
Lottery 17	10					
Lottery 18	10					
Total						
Bonus						
Overall						

Questionnaire 2: Treatment Group Low-Frequency Information/No Exogenous Target (LxN)

Name _____

Student ID _____

Part 1

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 1	10					
Lottery 2	10					
Lottery 3	10					
Lottery 4	10					
Lottery 5	10					
Lottery 6	10					
Total						

Part 2

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 7	10					
Lottery 8	10					
Lottery 9	10					
Lottery 10	10					
Lottery 11	10					
Lottery 12	10					
Total						

Part 3

You now have the chance to set your own target for your total earnings in ECU for the final six lotteries. You receive +5% on this target if you reach it and get no bonus if you fail to do so.

Yes (Target: _____)

No

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 13	10					
Lottery 14	10					
Lottery 15	10					
Lottery 16	10					
Lottery 17	10					
Lottery 18	10					
Total						
Bonus						
Overall						

Questionnaire 3: Treatment Group High-Frequency Information/Exogenous Target (HxT)

Name _____

Student ID _____

Part 1

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 1	10					
Lottery 2	10					
Lottery 3	10					
Lottery 4	10					
Lottery 5	10					
Lottery 6	10					
Total						

Part 2

For the next six lotteries, please be aware that the overall expected result when betting every lottery your entire endowment is 70 ECU's. If you like, you can try and reach this target!

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 7	10					
Lottery 8	10					
Lottery 9	10					
Lottery 10	10					
Lottery 11	10					
Lottery 12	10					
Total						

Part 3

You now have the chance to set your own target for your total earnings in ECU for the final six lotteries. You receive +5% on this target if you reach it and get no bonus if you fail to do so.

Yes (Target: _____)

No

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 13	10					
Lottery 14	10					
Lottery 15	10					
Lottery 16	10					
Lottery 17	10					
Lottery 18	10					
Total						
Bonus						
Overall						

Questionnaire 4: Treatment Group Low-Frequency Information/Exogenous Target (LxT)

Name _____

Student ID _____

Part 1

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 1	10					
Lottery 2	10					
Lottery 3	10					
Lottery 4	10					
Lottery 5	10					
Lottery 6	10					
Total						

Part 2

For the next six lotteries, please be aware that the overall expected result when betting every lottery your entire endowment is 70 ECU's. If you like, you can try and reach this target!

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 7	10					
Lottery 8	10					
Lottery 9	10					
Lottery 10	10					
Lottery 11	10					
Lottery 12	10					
Total						

Part 3

You now have the chance to set your own target for your total earnings in ECU for the final six lotteries. You receive +5% on this target if you reach it and get no bonus if you fail to do so.

Yes (Target: _____)

No

	Endowment (ECUs)	Saved amount	Bet amount	Outcome of lottery	Return on bet amount	Total earnings of the lottery
Lottery 13	10					
Lottery 14	10					
Lottery 15	10					
Lottery 16	10					
Lottery 17	10					
Lottery 18	10					
Total						
Bonus						
Overall						