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# Incentives to Learn Calibration: a Gender-Dependent Impact. 

Marie-Pierre Dargnies* and Guillaume Hollard ${ }^{\dagger}$

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*Paris School of Economics, Université Paris 1 Panthéon-Sorbonne, CES 106-112 boulevard de l'Hopital 75013 Paris. Tel:(0033) 1440782 13. Fax: (0033) 1440782 31. E-mail: marie-pierre.dargnies@univ-paris1.fr
${ }^{\dagger}$ Paris School of Economics, CNRS, CES 106-112 boulevard de l'Hopital 75013 Paris


#### Abstract

Miscalibration can be defined as the fact that people think that their knowledge is more precise than it actually is. In a typical miscalibration experiment, subjects are asked to provide subjective confidence intervals. A very robust finding is that subjects provide too narrow intervals at the $90 \%$ level. As a result a lot less than $90 \%$ of correct answers fall inside the $90 \%$ intervals provided. As miscalibration is linked with bad results on an experimental financial market (Biais et al., 2005) and entrepreneurial success is positively correlated with good calibration (Regner et al., 2006), it appears interesting to look for a way to cure or at least reduce miscalibration. Previous attempts to remove the miscalibration bias relied on extremely long and tedious procedures. Here, we design an experimental setting that provides several different incentives, in particular strong monetary incentives; i.e. that make miscalibration costly. Our main result is that a thirty-minute training session has an effect on men's calibration but no effect on women's.


## Résumé

On désigne par l'anglicisme "miscalibration" le fait que les individus pensent que leur savoir est plus précis qu'il ne l'est en réalité. On mesure typiquement la miscalibration en demandant aux sujets de fournir des intervalles de confiance à $90 \%$ pour une série de questions. Or, de nombreuses études (voir Lichtenstein and Fischhhoff (1977) pour une revue de littérature) montrent que le taux de réponses correctes appartenant aux intervalles à $90 \%$ fournis est toujours bien inférieur à $90 \%$. On parle alors de miscalibration dans la mesure où les individus fournissent des intervalles à $90 \%$ trop étroits. Les liens existant entre la miscalibration et de mauvaises performances économiques (Biais et al. (2005), Regner et al. (2006)) expliquent l'intérêt pour les économistes d'étudier ce biais.
Notre protocole expérimental vise à réduire la miscalibration par le biais d'incitations directes, notamment monétaires. Notre résultat principal est que nos incitations ont un effet sur la calibration des hommes, ceux ayant suivi notre entrainement fournissant des intervalles plus larges, mais aucun effet tangible sur celle des femmes.

JEL Codes: D81, C91.
Keywords: miscalibration, overconfidence, incentives, gender effect.

## 1 Introduction

In the past decades Economists and Psychologists documented a long list of biases, i.e. substantial and systematic deviations from the predictions of standard economic theory ${ }^{1}$. Many economists will argue that these biases only matter if they survive in an economic environment. In other words, if correct incentives are provided subjects should realize that they are making costly mistakes and then change the way they make such decisions in further decision tasks. In this paper we test this claim regarding a particular bias, namely miscalibration. We then create an experimental setting that provides a lot of incentives (decisions have monetary consequences, successful others can be imitated, feedbacks are provided, repeated trials are used, etc). Finally, we test in a subsequent decision task whether subjects still display some miscalibration.

## What is miscalibration and why is it important to economists?

Calibration is related to the capacity of an individual to choose a given level of risk. In a typical experiment designed to measure miscalibration, subjects are asked to provide subjective confidence intervals. For example, if the question is "What was the unemployment rate in France for the first trimester of 2007?" and the subject provides the $90 \%$ confidence interval $[7 \%, 15 \%$ ], it means that the subject thinks that there is a $90 \%$ chance that this interval contains the correct answer. A perfectly calibrated subject's intervals should contain the correct answer $90 \%$ of the time. In fact, a robust finding is that almost all subjects are miscalibrated. On average, $90 \%$ subjective confidence intervals only contains the correct answer between, say, $30 \%$ and $50 \%$ of the time ${ }^{2}$. Glaser et al. (2005) found an even stronger miscalibration using professional traders.

Miscalibration is a bias having important economic consequences, since miscalibrated people suffer losses on experimental markets (Bonnefon et al., 2005, Biais et al., 2005). Furthermore, it is likely that such a pathology affects the behavior of real traders acting on real markets. Therefore, it does make sense for economists to try to reduce miscalibration and to study the

[^0]best incentives to do so.
Lichtenstein and Fischhoff (1980) attempted to reduce miscalibration by providing subjects with feedback on their performance. They proved that 23 sessions, each lasting about an hour, were required to substantially improve subjects' calibration. Several other psychologists have used various techniques to reduce miscalibration (Pickhardt and Wallace, 1974, Adams and Adams, 1958), with little success so far. Miscalibration thus appears to be a very robust bias.

This paper proposes to provide a maximum of incentives to reduce miscalibration. The main result is that our experimental setting succeeds in reducing overconfident miscalibration but only for males.
The remainder of the paper is organized as follows. Section 2 presents the experimental design. Section 3 presents the results and section 4 discusses them. Finally, Section 5 provides some concluding remarks.

## 2 Experimental design

The measure of miscalibration and associated overconfidence relies on a now standard protocol. Subjects have to provide $90 \%$ subjective confidence intervals for a set of 10 quiz questions. On average, perfectly calibrated subjects should catch the correct answer 9 times; if this is not the case, they are miscalibrated. ${ }^{3}$ The subjects are asked to estimate their hit rate. The difference between their estimated hit rate and their actual one is a classical measure of overconfidence. This protocol will thus serve as a benchmark for measuring miscalibration and overconfidence in our experiment.

The experimental subjects were divided into two groups. The subjects of the first group attended a training session and then performed a baseline treatment aiming at measuring their miscalibration according to the standard protocol. The principle of this training session is to offer a whole set of experimental incentives that enhance learning (monetary incentives, tournament, feedback, loss framing). The second group, the control group, performs the baseline treatment only.

[^1]At first glance, testing the effect of incentives seems possible by simply providing incentives for the basic miscalibration task used as a benchmark. This seems natural but cannot be implemented since there is no simple incentive scheme that rewards correct calibration. Think, for example, of an incentive scheme that would pay a high reward if the difference between the required percentage of hit rates, say $90 \%$, and the actual hit rate (measured over a set of 10 questions) is small. A rational subject can use very wide intervals for 9 questions and a very small one for the remaining question. He is thus certain to appear correctly calibrated, while he is not. Cesarini et al. (2006) chose to provide incentives for the evaluation of the calibration task only (how many correct answers belong to the intervals provided by the subject and by his peers) and made miscalibrated subjects go through the task again. We chose to consider a task similar to the calibration task in which we can provide the necessary incentives. This task, described in the following section aims at making the subjects realize they have a hard time calibrating the level of risk they wish to take. After having completed this training task, subjects have to complete a standard calibration task for which we only provide incentives for the following evaluation of how subjects did in the calibration task as in Cesarini et al. (2006). A control group who did not go through the training task also completed the calibration task to enable us to measure the effect of the training task.

### 2.1 The training period

In the training period, the participants were asked to answer a set of twenty questions: ten questions on general knowledge followed by ten questions on economic knowledge.

The set of questions used in the training period was composed of ten questions some of which were used in Biais et al. (2005)'s experiment plus 10 questions on economic culture. Half of the subjects had to answer the 20 questions in a given order, the other half saw the questions in reverse order. This enabled us to check for learning effects during the training period.

In this training period, the subjects were provided with a reference interval for each question that they could be $100 \%$ sure the correct answer belonged to. Subjects had to give an interval included in the reference interval. Each player received an initial endowment of 2000 ECUs (knowing that they would be converted into euros at the end of the experiment at the rate of 1 euro for 100 ECUs) before beginning to answer the questions but after
having received instructions. They were told that 100 ECUs were at stake for each one of the twenty questions resulting in a loss framing. The payoffs are expressed in experimental currency (ECU). The payoff rule applied for each question was the following. :
payment $= \begin{cases}-100 * \frac{\text { width of the interval provided }}{\text { width of the interval of reference }} & \text { if the correct answer belongs } \\ & \text { to the interval provided } \\ -100 & \text { otherwise }\end{cases}$
According to this formula, the payoff is maximal and equal to 0 when the interval provided by the subject is a unique value, this value being the right answer to the question. In this case, the subject keeps the total 100 ECUs at stake for the question considered.

The payoff is equal to -100 (the subject loses the 100 points) if the subject provides the reference interval and consequently takes no risk at all.

There is therefore a trade-off between the risk taking and the amount of ECUs a subject could keep if the correct answer fell inside his interval. High risk taking is rewarded by a small loss (the subject keeps most of the ECUs at stake) in the case where the answer belongs to the interval provided. Conversely, a subject who only takes little risk will only keep a few ECUs ( meaning he would lose most of the ECUs at stake) even if the correct answer does belong to his interval.

Subjects had 60 seconds to answer each question, indicated by a timer. We applied this time constraint so as to not make the fastest subjects wait too long before switching to the next question as all subjects had to have answered a question before moving to the next one to enable us to provide feedback about the intervals provided for a given question. Nevertheless, we picked the time limit corresponding to the time it took for most subjects to answer a question in the pilot experiment where there was no time constraint. When time was up, if the subject had not validated his interval, the 100 ECUs at stake were lost and the next question was put.
Subjects received feedback providing them with the intervals chosen by all the participants (including themselves) ranked by width from the narrower to the wider as well as the payoff corresponding to each interval. They could infer from this feedback whether they had taken too much risk compared to the others. They could also see the ranking of everybody's score after each question so as to trigger a sense of competition.

After they had answered all 20 questions, subjects were asked to write a comment about their strategy. They then received general feedback about the first step of the experiment.
People being miscalibrated, we expected them to realize it when they saw that the correct answer fell outside their interval less or more often than they had expected, which resulted in a loss of money. As a result, we expected them to better adjust the level of risk they wished to take for the next questions. For instance, a subject quite confident that he knows the answer who provides in consequence a narrow interval will be likely to be more cautious when he realizes he did not catch the right answer. On the contrary, a subject who decides to be safe and provides a wide interval for a question he thinks he knows the answer to, will tend to be less cautious for the next questions when he realizes he could have kept more ECUs by giving a narrower interval. They could also infer information about the right level of risk to take by looking at what others did and how it paid.

### 2.2 The standard calibration task

In the next stage, the subjects who had participated in the training period were asked to answer a set of ten questions (five questions on general knowledge followed by five questions on economic knowledge) by giving their best estimation of the answer and then by providing $10 \%, 50 \%$ and $90 \%$ confidence intervals. Subjects in the control group had to complete the same task. After the pilot experiment was run, we removed and replaced the most difficult questions for which subjects seemed to have no clue about the answer.

Before the beginning, subjects were explained in detail what were $10 \%$, $50 \%$ and $90 \%$ confidence intervals. They were also told that they would receive remuneration regarding this task but that they would only know how the remuneration was established later. As in Cesarini et al. (2006), their remuneration for the calibration tasks depended on the evaluation the subjects were asked to make afterwards of their and the average subject's performance during the calibration task. There was no feedback between the questions and subjects could proceed at their own pace.

## 3 Results

The experiment took place at the laboratory of experimental economics of the University of the Sorbonne (Paris 1) in July 2007. 87 subjects, most of whom were students, participated in the experiment. 53 students went through the training period before they completed the calibration task, while the control group was composed of 34 subjects. The average subject was 22.42 years old in the control group and 22.71 years old in the trained group. The proportion of men was respectively $41.18 \%$ and $41.5 \%$ in the control and the trained groups. The average earning was 11.16 euros. On average, subjects earned 10.62 euros including a 5 euros show-up fee in the control group and 14.24 euros ( 8.42 for the training period and 5.82 for the calibration task) with no show-up fee for the trained group.

In the following section, we distinguish between two measures of confidence. First, the difference between the actual hit rate and the required hit rate, for $10 \%, 50 \%$ and $90 \%$ confidence intervals. This difference measures the miscalibration. Second, the difference between the subject's estimated hit rate and his actual hit rate. This second difference represents the confidence for the calibration task. It is thus another a measure of overconfidence.

### 3.1 General results on calibration

We find that the subjects from the control group exhibit a high level of miscalibration. Indeed, a lot more than one correct answer out of ten belong to the $10 \%$ intervals while fewer than five correct answers out of ten fall inside the $50 \%$ confidence intervals and far fewer than nine correct answers out of ten fall inside the $90 \%$ intervals. The average hit rates in the control group at the $10 \%, 50 \%$ and $90 \%$ levels are respectively $2.03,3.32$ and 4.81 while the corresponding median hit rates are respectively 2,3 and 5 . T-tests show that the observed hit rates significantly ( $\mathrm{p}<0.001$ for the 3 tests) differ from the expected hit rates (respectively 1, 5 and 9 at the $10 \%, 50 \%$ and $90 \%$ levels).
At the $10 \%$ level, people are found to be under-confident, meaning that they provide too wide intervals. As a result, the correct answer belongs too often to the $10 \%$ intervals. This result was expected by Cesarini et al. (2006).

At the $50 \%$ and $90 \%$ levels conversely, subjects display overconfidence as their intervals are too narrow, this is all the more the case for $90 \%$ confidence intervals. The fact that far fewer than $90 \%$ of correct answers belong to the
$90 \%$ confidence intervals of the subjects is in line with the results of Glaser et al. (2005).
A surprising feature is that, when asked to evaluate how many correct answers belong to their intervals, the average answers are respectively at the 10\%, $50 \%$ and $90 \%$ levels: $3.47,5.56$ and 8.04 for the control group; subjects exhibit overconfidence for the calibration task, thinking that they were more cautious than they actually were (see Figure 1). Let us, nevertheless, observe that subjects do predict that their calibration is far from being perfect, otherwise their evaluations would have been 1,5 and 9 .


Figure 1. Expected, Actual and Estimated Hit Rates in the Control Group

These results indicate that not only are people unable to adjust the width of their intervals to the risk level indicated (they are miscalibrated) but they are also unable to predict their bias correctly (they are over or underconfident).
To sum up, people seem to overestimate their underconfidence and underestimate their overconfidence.

### 3.2 The effect of training on miscalibration and confidence in calibration

### 3.2.1 The general picture

The main purpose of this paper was to see whether a training period during which several incentives aiming at improving people's calibration as well as decreasing overconfidence were provided would be efficient.
Trained subjects have only slightly higher hit rates at the $10 \%, 50 \%$ and $90 \%$ level than subjects from the control group (see figure 2). The differences in hit rates between the control and the trained group are not significantly different at any reasonable level. ${ }^{4}$


Figure 2. Hit Rates: Control vs Trained Group

We find that the median $10 \%$ interval width is larger for the trained group than for the control group for 7 questions out of ten. For the 3 remaining

[^2]questions, the median width of intervals is equal across treatments. Note that this goes in the sense of a worsening of the underconfident miscalibration observed at $10 \%$ as people tend to provide too wide intervals at $10 \%$. One reason why we may find such a result is that subjects may not consider the underconfident miscalibration as a bias and consequently, they may not try to correct it.

The same result is found when we compare median widths of $50 \%$ intervals (wider intervals in the trained group than in the control group for 7 questions, the reverse for 1 question and equal median intervals across treatments for the 2 remaining questions). As for $90 \%$ intervals, for six questions out of ten the interval width is larger for the trained group while the control group provided wider intervals than the trained group for 1 question. ${ }^{5}$ We report the regressions of the interval width of the $10 \%, 50 \%$ and $90 \%$ intervals on the sex of the subject, a treatment dummy ( $=1$ if the subject was in the trained group), the interaction between sex and treatment, the age of the subject, his level of education, dummies for the different questions and the interactions between each question and the treatment (see Table 1). We only find the interaction terms between questions 9 and 10 and the treatment to be significant and positive in explaining the $50 \%$ and $90 \%$ interval widths and only the interaction between question 9 and the treatment for the $10 \%$ interval width. The treatment makes subjects provide wider $50 \%$ and $90 \%$ intervals for the last two questions only and it only has a significant and positive effect for the $10 \%$ interval width of the 9 th question. Note that the level of education has a positive and significant impact on the $10 \%$ interval width, meaning that more educated subjects tend to provide wider $10 \%$ intervals and, by doing so, make their underconfident miscalibration worse. We ran logistic regressions of the dummies "the $10 \%$ interval contains the

[^3]correct answer" (ICA10), "the 50\% interval contains the correct answer" (ICA50), "the $90 \%$ interval contains the correct answer" (ICA90) on the same variables (see Table 2). We observe that the treatment significantly increases the probability for the correct answer to fall in the $50 \%$ and $90 \%$ intervals provided for almost all of the questions (the interaction terms between the questions and the treatment are always positive and almost always significant). It is true but to a smaller extent for the $10 \%$ intervals. If anything, our treatment seems to make subjects provide wider intervals (even if this result is far from always reaching significance) and it significantly helps subjects catch the correct answer in their confidence intervals more often. Consequently, the incentives we provided during a short training period decrease the overconfident miscalibration we observe for the $50 \%$ and $90 \%$ intervals but, to a smaller extent, makes the underconfident miscalibration noticed for the $10 \%$ intervals worse.

Table 1. Regression of the Interval Width of the 10, 50 and $90 \%$ intervals


[^4]Table 2. Logistic Regression of "the 10, 50, $90 \%$ interval contains the correct answer" (ICA10 ICA50 ICA90)

| Variable | ICA10 | ICA50 | ICA90 | Variable | ICA10 | ICA50 | ICA90 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Intercept | -0.7402 | 0.4416 | 1.3924 | q8 | -1.1865 | -1.8656 | -1.4668 |
|  | (0.2799) | (0.4948) | (0.0392) |  | (0.0656) | (0.0019) | (0.0171) |
| Sexe | 0.0412 | 0.3924 | 0.0738 | q9 | -1.7349 | -2.3641 | -2.3985 |
|  | (0.8915) | (0.1391) | (0.7711) |  | (0.0190) | (0.0002) | (0.0002) |
| treatment | -0.5951 | -1.5940 | -1.1804 | q10 | -1.7381 | -1.9947 | -1.9420 |
|  | (0.2896) | (0.0048) | (0.0485) |  | (0.0188) | (0.0011) | (0.0020) |
| sextreatment | 0.1398 | -0.0941 | 0.2821 | q2t | 0.8536 | 1.7735 | 0.0282 |
|  | (0.7124) | (0.7764) | (0.3801) |  | (0.2290) | (0.0152) | (0.9776) |
| Age | 0.00302 | 0.00324 | -0.00832 | q3t | 0.8063 | 2.5884 | 1.5503 |
|  | (0.9108) | (0.8895) | (0.7117) |  | (0.3400) | (0.0018) | (0.0441) |
| Education | 0.0507 | 0.0773 | 0.0350 | q4t | 0.7633 | 2.1252 | 1.5607 |
|  | (0.4971) | (0.2327) | (0.5801) |  | (0.2941) | (0.0032) | (0.0345) |
| q2 | 0.2337 | -0.3636 | 1.2972 | q5t | 0.7561 | 2.3399 | 1.6269 |
|  | (0.6766) | (0.5381) | (0.1441) |  | (0.4115) | (0.0014) | (0.0279) |
| q3 | -1.3719 | -2.8147 | -2.3541 | q6t | -0.7081 | 1.2329 | 0.6230 |
|  | (0.0447) | (<.0001) | (0.0003) |  | (0.4120) | (0.1041) | (0.3992) |
| q4 | -0.4377 | -1.5713 | -1.3416 | q7t | 1.3228 | 2.1902 | 1.2944 |
|  | (0.4481) | (0.0074) | (0.0291) |  | (0.0935) | (0.0034) | (0.0833) |
| q5 | -1.7691 | -1.8656 | -1.4668 | q8t | 1.0284 | 2.2127 | 1.5161 |
|  | (0.0167) | (0.0019) | (0.0171) |  | (0.1977) | (0.0026) | (0.0404) |
| q6 | -0.7728 | -1.8656 | -1.3416 | q9t | 1.0319 | 2.3218 | 2.5139 |
|  | (0.1985) | (0.0019) | (0.0291) |  | (0.2509) | (0.0027) | (0.0010) |
| q7 | -1.1124 | -1.9534 | -1.7502 | q10t | 1.5328 | 2.3593 | 1.6794 |
|  | (0.0856) | (0.0015) | (0.0052) |  | (0.0804) | (0.0015) | (0.0246) |

Note: p-values are in brackets.

### 3.2.2 A different impact between genders

This general picture masks some strong heterogeneity across subjects. We can control for several sources of heterogeneity. However, the gender variable captures almost all of it. We observe indeed that there is virtually no improvement in women's calibration especially when we compare the median hit rates between the treatments while men increase their median hit rate
by 0.5 point at the $50 \%$ level and by 1 point at the $10 \%$ and $90 \%$ levels (see Figure 3).


Figure 3. Hit Rates: Gender Differences

The difference in interval width between the control and the training treatments seems to be larger for men than for women, indicating that men learned more than women to reduce their overconfidence.Using a Wilcoxon-Mann-Whitney test, we find that $10 \%$ confidence intervals are significantly wider for the trained group respectively for five questions out of ten and zero question out of ten for men and women. Let us notice that in the trained group both men and women had more than one correct answer inside their $10 \%$ intervals exhibiting underconfident miscalibration. As a result, an increase of $10 \%$ intervals causes an aggravation of underconfidence. For $50 \%$ intervals, the width increases significantly between the control and the training treatments respectively for two and six questions out of ten. Finally, concerning $90 \%$ intervals, the difference is significant in three cases and four cases out of ten respectively for women and men.

### 3.3 What happened during the training session ?

The training period had an impact on men but almost no effect on women. It could be interesting to use the results from the training period to get an insight into the nature of the learning process that arose. In order to be able to measure learning during the training period, the order of the 20 questions was reversed for half of the subjects.

It appears that during the training process, some learning took place. We measured learning at this stage by comparing the width of the intervals provided for the same question by subjects from the two groups corresponding to the two orders of appearance of the questions. We found that there was a significant difference in the width of the intervals between the two groups for seven questions out of twenty, each going in the sense of longer intervals for the group who answered the question later in the training session. For example, the intervals provided for question 18 were wider for the group who had the regular order of questions than for those who had the reversed order (for whom question 18 was actually the third one they had to answer). It seems noteworthy that six out of the seven questions which subjects with more training answered with wider intervals were economic knowledge questions.

We regressed a variable equal to the interval width chosen over the interval width of the interval of reference on the intercept, a dummy indicating the gender ("sex"),the age, the level of education, a dummy indicating whether the question appeared early during the training session ("exp"), the interaction between "exp" and "sex" ("expsex"), the ranking announced to the subject after he had answered the previous question ("Rank-1"), the gap between the midpoint of the interval provided and the correct answer as a proxy for the ignorance ("gap") and dummies for the different questions (See Table 3). We added "gap" in the regressors so as to take in the effect of knowledge on the choice of the interval width. Any residual effect of "Rank-1" can therefore be attributed to competition, ie the effect of the announced rank on the decision to take more or less risk.

Women are found to provide significantly ( p -value $<0.0001$ ) narrower intervals than men. "Age" is also highly significant and negative. "Exp" is positive and significant ( p -value $=0.0622$ ) indicating that subjects who answered a question later in the training period tend to provide wider intervals. We found that the coefficient of "gap" is positive and highly significant showing that the less people knew the answer, the wider the interval they provided. The coefficient of "Rank-1" was found to be positive and highly significant

Table 3. Regression of IW/IWref and ICA)

| Variable | IW/IWref | ICA | Variable | IW/IWref | ICA |
| :--- | :---: | :---: | :--- | :---: | :---: |
| Intercept | 41.12421 | 1.2527 | q 8 | -1.68538 | -0.0629 |
|  | $(<.0001)$ | $(0.2431)$ |  | $(0.8203)$ | $(0.9491)$ |
| Sexe | -6.99129 | 0.1719 | q9 | -19.85722 | 0.8560 |
|  | $(<.0001)$ | $(0.4594)$ |  | $(0.0076)$ | $(0.3943)$ |
| Age | -0.46804 | -0.00749 | q10 | -11.83878 | 0.3414 |
|  | $(0.0011)$ | $(0.7437)$ |  | $(0.1108)$ | $(0.7303)$ |
| Education | 0.31946 | -0.0545 | q11 | 5.27825 | 0.1627 |
|  | $(0.4400)$ | $(0.3976)$ |  | $(0.4797)$ | $(0.8694)$ |
| exp | 2.47820 | 0.4154 | q12 | -6.28333 | -0.6213 |
|  | $(0.0622)$ | $(0.0473)$ |  | $(0.3983)$ | $(0.5272)$ |
| expsex | -0.12668 | -1.0866 | $q 13$ | -5.66640 | 1.0783 |
|  | $(0.9512)$ | $(0.0007)$ |  | $(0.4491)$ | $(0.3119)$ |
| gap | 1.95204 | -0.8905 | $q 14$ | -6.67250 | -0.0545 |
|  | $(<.0001)$ | $(<.0001)$ |  | $(0.3686)$ | $(0.9558)$ |
| rang_1 | 0.80443 | -0.0154 | q15 | -16.97609 | 0.5789 |
|  | $(0.0018)$ | $(0.6979)$ |  | $(0.0223)$ | $(0.5612)$ |
| q2 | -23.23853 | 5.4903 | q16 | -11.00649 | 0.4695 |
|  | $(0.0041)$ | $(<.0001)$ |  | $(0.1385)$ | $(0.6355)$ |
| q3 | 5.05854 | 0.0202 | q17 | -8.28099 | 0.2700 |
|  | $(0.4957)$ | $(0.9836)$ |  | $(0.2645)$ | $(0.7845)$ |
| q4 | 2.15180 | 0.4241 | q18 | -14.72562 | 1.6759 |
|  | $(0.7724)$ | $(0.6688)$ |  | $(0.0476)$ | $(0.1118)$ |
| q5 | 3.09015 | 0.1231 | $q 19$ | -20.02148 | 0.9432 |
|  | $(0.6770)$ | $(0.9007)$ |  | $(0.0071)$ | $(0.3492)$ |
| q6 | -1.74616 | 0.7506 | q20 | -9.39033 | 2.4406 |
|  | $(0.8140)$ | $(0.4522)$ |  | $(0.2106)$ | $(0.0269)$ |
| q7 | -17.73803 | 2.4098 |  |  |  |
|  | $(0.0180)$ | $(0.0253)$ |  |  |  |

Note: p-values are in brackets.
too. It seems that the announcement of a bad ranking leads subjects to take less risk and provide wider intervals.

The logistic regression of the probability to catch the correct answer in one's interval (ICA) on the same variables (See Table 3) reveals that while
"sex" "age" and "rank-1" are not significant, "gap" is negative and highly significant showing that the more ignorant the subject was about the answer, the less chance for the correct answer to belong to his interval. "exp" is found to be positive ( p -value=$=0.0473$ ) and "expsex" negative ( p -value= $=0.0007$ ) indicating that, overall, having answered a greater number of questions previously tends to increase one's chances to catch the correct answer in his interval but the opposite is actually true for men. We tried to search for explanations for the fact that women failed to learn from our training period while men did. It seems likely that the explanation lies in what happens in the training period. The different possible explanations are: a stronger reaction to competition for men than for women, a longer time for decision, the fact that money is a stronger incentive for men... Unfortunately, the too scarce data available to us made it impossible to reach a definitive conclusion. Our results could indicate that men used more the training session to experiment different strategies and took more risk to go up in the ranking and, as a result, benefited more from the training period than women who were more cautious. This is in line with the idea of Gail Osten, author of "What Can Male Traders Learn from Successful Women...And Vice Versa" who says that "women, particularly when starting out, often are more timid in trading and more conservative in the use of their money". ŞMen, on the other hand, seem not to have the same reservation or feeling of guilt regarding their initial funding or the price of tuition, $\check{\mathrm{T}}$ according to Barb Magio, a trader, educator and moderator in woodiescciclub.com. ŞThey take these losses merely as part of the learning process and seem to feel less guilt or necessity to explain why instant profitability is lacking. $\bar{T}$

To conclude, providing monetary incentives helps reduce men's overconfident miscalibration but leaves women's miscalibration unchanged.

## 4 Discussion

This paper contributes to a literature interested in cognitive biases having economic consequences. We focus on miscalibration, a very robust bias correlated with losses on experimental financial markets and bad entrepreneurship. In line with the existing literature on miscalibration, our subjects strongly suffer from the miscalibration bias, their $50 \%$ and $90 \%$ intervals being too narrow (overconfident miscalibration). We find that subject's $10 \%$ intervals are too wide (underconfident miscalibration). These results are widespread in
the population according to the literature and there are very few exceptions. Furthermore, subjects overestimate their underconfidence and underestimate their overconfidence. The fact that people overestimate their underconfident miscalibration could mean that they do not consider it as a bias. Maybe being too cautious is seen as a good thing. Previous attempts to reduce miscalibration relied on very long and repetitive training periods.
We find that men's calibration can be improved by a thirty-minute training period punishing miscalibrated behavior by money losses, while women's cannot. The incentives we implemented had no effect on women. This difference in the impact of monetary incentives between genders is a key interest of Niederle and her coauthors (?Niederle and Vesterlund, 2007, Niederle and Yestrumskas, 2007) who show that it can be detrimental to welfare. Indeed, they highlight the fact that highly able women do not enter tournaments as often as they should while low performing men enter too often. To overcome this issue, Niederle et al. (2007)studied the effect of affirmative action in favor of women and found that it increased the number of women willing to enter the tournament and decreased the number of men, more than what would be predicted solely by the change in the probability of winning. There are probably other incentives one could think of that would have a stronger effect on women than on men and which could therefore benefit welfare.

## 5 Conclusion

We find that people who went through the training session provide wider intervals at 10,50 and $90 \%$ than those who did not. This result is not always significant but it is quite robust as subjects from the control group never provided significantly wider intervals than trained subjects. Moreover, our training significantly increases a subject's chance to catch the correct answer in his interval. This results in an improvement of calibration at the $50 \%$ and $90 \%$ levels but the underconfident miscalibration observed at the $10 \%$ level is made worse by the training. Nevertheless, men seem to have learned more from the training than women, as the increase in interval width between the treatments is greater for men than for women in most cases. As a result, the difference in hit rates between the control and the trained group is greater for men, who become more cautious and increase their hit rates at the three levels while women's hit rates are virtually the same across treatments.

Some consequences can be drawn. It is unlikely that miscalibration disap-
pears in a market environment, since we provided the kind of incentives that are expected on real markets. According to our results, real traders are likely to underestimate the risk they take when they think they invested in a very secure asset. Symmetrically they take less risks than they think when they invest in risky assets. So, the overall effect of miscalibration on real markets is ambiguous. Our results also suggest that men may be more successful in learning calibration. Women traders may need a longer king of training which would give them more time to get rid of their overcautiousness.

## Author's Affiliations

Marie-Pierre Dargnies(Corresponding author)
Université Paris 1 Panthéon-Sorbonne,
Paris School of Economics
Address: Centre d'Economie de la Sorbonne,
106-112 Boulevard de l'Hôpital 75013 Paris, France.
Tel:(0033) 144078213.
Fax: (0033) 144078231.
E-mail: marie-pierre.dargnies@univ-paris1.fr
and
Guillaume Hollard
Paris School of Economics, CNRS.

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## Appendices

## A Instructions

## A. 1 Trained group

You are about to participate to an experiment aiming at evaluating your ability to calibrate risk. This experiment will be divided in several steps.

## A.1.1 First step

In this first step, you will have to answer a set of twenty questions by providing an interval for each question. At the beginning of this step, you will be endowed with 2000 points which will be converted to euros at the end of the experiment. For each of the 20 questions, 100 points will be at stake. You will have to keep the more points you can.
For each question, you will be provided with an interval of reference including the correct answer. The interval you will provide will have to be contained in the interval of reference. Your payoffs will be determined as follows:

- If the correct answer does not belong to the interval chosen, you will lose the 100 points at stake for the question. They will be withdrawn from your endowment.
- If the correct answer does belong to the interval chosen, the narrower the interval you chose, the more points you will keep. Your payoffs will depend on the difference between the length of the interval chosen and the length of the interval of reference given the following formula:
payment $= \begin{cases}-100 * \frac{\text { width of the interval provided }}{\text { width of the interval of reference }} & \text { if the correct answer belongs } \\ & \text { to the interval provided } \\ -100 & \text { otherwise }\end{cases}$
In consequence, the wider the interval chosen, the more chances for the correct answer to belong to your interval but the fewer the points you will get to keep if your interval contains the correct answer.

Symmetrically, the narrower the interval chosen, the less chances for the correct answer to belong to your interval but the more points you will get to keep if your interval contains the correct answer.

After each question, you will see the correct answer, the intervals chosen for the same question by the subjects present in the lab (ranked from the narrower to the wider) as well as the number of points they kept.

Example: For the question, "How old was John Fitzgerald Kennedy when he died?", if the interval of reference is [30;80]:

- If you give the interval [49;54], your potential loss is 10 : if the correct answer belongs to the interval [49;54], you will keep 90 points out of the 100 points at stake for this question. Here, your actual loss would be 100 points as the correct answer, 46, does not belong to your interval. Hence, you would have lost the 100 points at stake for this question.
- If you give the interval [40;65], your potential loss is 50 and it corresponds to your actual loss as the correct answer belongs to your interval. In this case, you would have kept 50 points out of the 100 points at stake for this question.


## A.1.2 Second step

In this second step, you will also be compensated but you will only be informed of the details of the remuneration afterwards. You will have to answer to a set of 10 questions by providing your best estimate of the answer and confidence intervals, ie a lower and an upper bound corresponding to a certain level of confidence that the correct answer falls between these 2 values, knowing that:

- The narrower the interval you will provide, the more chances for the correct answer to fall outside.
- The wider the interval you will provide, the more chances for the correct answer to fall inside.

For each question, you will have to give intervals corresponding to 3 different levels of confidence ( $10 \%, 50 \%$ and $90 \%$ ). To help you calibrate the risk, here are 3 different and equivalent ways to understand what a $10 \%$ interval is: A $10 \%$ interval corresponds to a lower and an upper values such that:

1. You are $10 \%$ confident that the correct answer lies between these 2 values.
2. For 10 questions, 1 correct answer on average belongs to the interval provided and 9 correct answers out of 10 on average fall outside of their interval.
3. You think there are 9 chances out of 10 for the correct answer to fall outside your interval.

Example of question: What was the year of Vincent Auriol's election as President?
If you give the value 1927 as your best estimation of the answer and the intervals $[1921,1930]$ at $10 \%,[1915,1935]$ at $50 \%$ and $[1915,1949]$ at $90 \%$, it means that:

- Your best estimate of the year Vincent Auriol was elected is 1927.
- You are $10 \%$ confident he was elected between 1921 and 1930.
- You think there is 1 chance out of 2 that he was elected before 1915 or after 1935.
- You are $90 \%$ sure that his election happened between 1915 and 1949 .

Considering the 10 questions you just answered, please evaluate:

- The number of correct answers falling inside the $10 \%$ intervals you provided (please enter a number between 0 and 10).
- The number of correct answers falling inside the $50 \%$ intervals you provided (please enter a number between 0 and 10).
- The number of correct answers falling inside the $90 \%$ intervals you provided (please enter a number between 0 and 10).

As well as:

- The number of correct answers falling inside the $10 \%$ intervals provided by the average subject (please enter a number between 0 and 10).
- The number of correct answers falling inside the $50 \%$ intervals provided by the average subject (please enter a number between 0 and 10).
- The number of correct answers falling inside the $90 \%$ intervals provided by the average subject (please enter a number between 0 and 10).

For each correct evaluation, you will earn 100 points.
Finally, you will have to make 2 choices each time between 2 bets. For each of these 2 choices, you will earn 300 points if what you bet on happens.

## A. 2 Control group

You are about to participate to an experiment aiming at evaluating your ability to calibrate risk.

You will be compensated but you will only be informed of the details of the remuneration afterwards. You will have to answer to a set of 10 questions by providing your best estimate of the answer and confidence intervals, ie a lower and an upper bound corresponding to a certain level of confidence that the correct answer falls between these 2 values, knowing that:

- The narrower the interval you will provide, the more chances for the correct answer to fall outside.
- The wider the interval you will provide, the more chances for the correct answer to fall inside.

For each question, you will have to give intervals corresponding to 3 different levels of confidence ( $10 \%, 50 \%$ and $90 \%$ ). To help you calibrate the risk, here are 3 different and equivalent ways to understand what a $10 \%$ interval is: A $10 \%$ interval corresponds to a lower and an upper values such that:

1. You are $10 \%$ confident that the correct answer lies between these 2 values.
2. For 10 questions, 1 correct answer on average belongs to the interval provided and 9 correct answers out of 10 on average fall outside of their interval.
3. You think there are 9 chances out of 10 for the correct answer to fall outside your interval.

Example of question: What was the year of Vincent Auriol's election as President?
If you give the value 1927 as your best estimation of the answer and the intervals $[1921,1930]$ at $10 \%,[1915,1935]$ at $50 \%$ and $[1915,1949]$ at $90 \%$, it means that:

- Your best estimate of the year Vincent Auriol was elected is 1927.
- You are $10 \%$ confident he was elected between 1921 and 1930.
- You think there is 1 chance out of 2 that he was elected before 1915 or after 1935.
- You are $90 \%$ sure that his election happened between 1915 and 1949.

Considering the 10 questions you just answered, please evaluate:

- The number of correct answers falling inside the $10 \%$ intervals you provided (please enter a number between 0 and 10).
- The number of correct answers falling inside the $50 \%$ intervals you provided (please enter a number between 0 and 10).
- The number of correct answers falling inside the $90 \%$ intervals you provided (please enter a number between 0 and 10).

As well as:

- The number of correct answers falling inside the $10 \%$ intervals provided by the average subject (please enter a number between 0 and 10).
- The number of correct answers falling inside the $50 \%$ intervals provided by the average subject (please enter a number between 0 and 10).
- The number of correct answers falling inside the $90 \%$ intervals provided by the average subject (please enter a number between 0 and 10).

For each correct evaluation, you will earn 100 points.
Finally, you will have to make 2 choices each time between 2 bets. For each of these 2 choices, you will earn 300 points if what you bet on happens.

## B Questions

Questions of the Training Session:

1. How long, in months, does the gestation of an asian elephant last? 22) [2,50]
2. What is the diameter of the Moon in kilometers? (3476) [10,150000]
3. What is the distance (in Kilometers) between London and Tokyo? (9559) [300,40000]
4. What is the depth of the deepest point in the ocean? (11033) [10,65000]
5. What was the age at death of Einstein? (76) [10,100]
6. How many countries are members of NATO? (26) [2,200]
7. What is the number (in millions) of inhabitants of Norway? $(4,6)$ [0.5,150]
8. In which year was Mozart born? (1756) [1300,1980]
9. How high (in meters) is the Eiffel tower? (324) [2,4000]
10. How high (in meters) is Mount Blanc? (4808) [1000,10000]
11. How much (in euros) does the school education until high school graduation (without repeating) of a student cost? (87730) [500,300000]
12. What is the gross monthly income of the french Prime Minister? (20206) [1000,80000]
13. What is the percentage of french households accountable for the "Impôt sur la Fortune"? (1.7\%)
[ $0 \%, 30 \%$ ]
14. What is the french poverty line (monthly euro amount such that anyone earning less is considered poor)? (645) [30,1500]
15. What was the unemployment rate in France for the first trimester of 2007? (8.7\%)
[0\%,40\%]
16. What is the after-tax monthly income of a CAPES-holder teacher who has been teaching for 10 years? (1859) [600,10000]
17. How much is the "Revenu Minimum d'Insertion" (Minimum insertion outcome) for a single person with no child? (440.86) [30,1500]
18. What is the after-tax monthly income of a beginning university lecturer and researcher? (1655)
[600,10000]
19. What is the after-tax monthly income of a beginning police officer? (1235)
[600,10000]
20. What was the per inhabitant GDP in 2004 in France? (26788) [100,500000]

Questions of the Calibration Task:

1. What was the age at death of Martin Luther King? (39)
2. How many countries are members of OPEC (Organization of the Petroleum Exporting Countries)? (11)
3. What is the maximal length in meters of a whale? (33)
4. What was the year of Ariane rocket's first launch? (1979)
5. What was the year of JS Bach's birth? (1685)
6. What is the average after-tax monthly income in France? (1903)
7. What was the unemployment rate in France in 1970? (2.5\%)
8. What percentage of the GDP do the taxes and social security contributions represent? (45\%)
9. What is the after-tax monthly income of a french congressman? (5177.66)
10. What is the average annual cost for the school system of the education of a high-school student? (10000)

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[^0]:    ${ }^{1} \mathrm{~A}$ list of almost a hundred of such biases can be found at http: //en.wikipedia.org/wiki/List_of_cognitive_biases
    ${ }^{2}$ see Lichtenstein and Fischhhoff (1977) for a survey and (Klayman et al., 1999) for variables that affect miscalibration

[^1]:    ${ }^{3}$ Note that we cannot say of a single subject who only catches the correct answer in his $90 \%$ intervals, say, 6 times out of 10 that he is miscalibrated. Nevertheless, we can say of a population of subjects for which on average 6 correct answers out of 10 belong to the $90 \%$ intervals provided and with no subject catching 9 correct answers or more that it is globally miscalibrated.

[^2]:    ${ }^{4}$ The hit rates are respectively at the $10 \%, 50 \%$ and $90 \%$ levels $2.03,3.32$ and 4.81 for the control group and $2.40,3.80$ and 5.33 for the trained group.

[^3]:    ${ }^{5}$ If we compare average interval widths, which seems less relevant as averages are sensitive to extreme values, we find that for 7 (6) questions out of ten the average width of $10 \%$ and $50 \%$ (respectively $90 \%$ ) intervals are larger for the trained subjects, while for the remaining 3 (respectively 4) questions, the opposite is true.
    Checking for the significance of these results with a T-test, we find significantly larger intervals for the trained group than for the control group only for the ninth question, all other differences being not significant. However, as variances of interval widths are often very different across the control and trained group and as a way of eliminating the influence of extreme values, we ran a Wilcoxon-Mann-Whitney test. We found that the $90 \%$ interval widths are significantly different (either at the $1 \%, 5 \%$ or $10 \%$ levels) for 5 questions out of ten while $10 \%$ and $50 \%$ intervals widths are significantly different respectively for 3 and 6 questions out of ten.

[^4]:    Note: p-values are in brackets.

