## Macroeconomic Experiences and Risk Taking of Euro Area Households<sup>1</sup>

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## Abstract

This paper studies to what extent the experiences of households shape their willingness to take financial risks. It starts by extending the U.S. evidence of Malmendier and Nagel (2011) to euro area households, and shows that experienced stock market returns matter in a statistically significant and economically substantial fashion: better experiences increase the financial risk that households are willing to take as well as their stock market participation along the intensive and the extensive margin. It finds that more distant experiences receive a somewhat lower (but still substantial) weight than the corresponding findings suggest for the United States. Subsequently, the paper moves on to show that there are additional effects stemming from the experience of extreme stock market crash give substantially more weight to the most recent experience, suggesting that in these countries an even more pronounced underinvestment in the stock market should be expected in the years to come.

**JEL-codes**: D03, D14, D83, G11

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#### 1. Introduction

There is ample evidence that the willingness of economic agents to take financial risks has decreased in the course of the global financial crisis. Such a pattern has been found for financial markets (Bekaert and Hoerova 2014), banks (Bassett et al. 2014) and households (Guiso, Sapienza and Zingales 2013). It can therefore be assumed that the willingness to take financial risks varies over time, and depends on the experiences that economic agents have undergone.

Beyond the immediate reaction to adverse events, a recent paper by Malmendier and Nagel (2011) has shown that U.S. households' risk taking is furthermore affected by their experiences over longer time spans: households that have experienced higher real stock market returns over their lifetime tend to be more willing to take financial risks, have a higher propensity to hold stocks and hold larger amounts of stocks.

Personal experiences shape economic behavior in a variety of ways. Having experienced higher inflation, for instance, tends to lower happiness (Blanchflower 2007), increase inflation expectations (Lombardelli and Saleheen 2003; Malmendier and Nagel 2009) and inflation aversion (Ehrmann and Tzamourani 2012). Having grown up during recessionary times matters for preferences: as Alesina and Giuliano (2011) and Giuliano and Spilimbergo (2014) demonstrate, such individuals are more likely to believe that success in life depends more on luck than on effort, and therefore have a more favorable attitude toward redistributional policies. Beyond these macroeconomic factors, experiences of financial market performance also shape agents' behavior: Kaustia and Knüpfer (2008) show that investors are more likely to subscribe to initial public offerings (IPO) on the stock market if their previous IPO investments have performed relatively well, Koudijs and Voth (2014) find that having been exposed to a substantial risk of losing money in a lending transaction makes investors lend with higher haircuts, and Choi et al. (2009) suggest that investors over-extrapolate from their personal experience when they make their savings decisions.

Of course, also the socio-economic background of an individual affects beliefs and behavior. As reported in Dohmen et al. (2011), the educational background of an individual's parents affects her willingness to take risks. Guiso, Sapienza and Zingales (2004) measure social capital in a region by the electoral turnout and the willingness to donate blood, and find that in high-social-capital regions in Italy, more households invest in stocks, a pattern that even persists if the individual leaves the region. Finally, using data on German households, Alesina and Fuchs-Schündeln (2007) identify persistent effects of communism on attitudes toward the role of the state in providing social services, insurance or redistribution.

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If we accept that individual experiences shape beliefs and behavior, another question is how long these patterns persist. The above-mentioned findings in Alesina and Fuchs-Schündeln (2007) and in Guiso, Sapienza and Zingales (2004) suggest that there is quite some persistence. Malmendier and Nagel (2011), estimating the impact of financial market experience on risk taking, find that more distant experiences are relatively less important than more recent ones, but that their impact remains noticeable for some decades. Their findings also suggest that young individuals are particularly affected by more recent events. Nakov and Nuño (2015) model this setup and show that in such an economy, the stock price exhibits stochastic fluctuations around the rational expectations equilibrium due to successive waves of optimism and pessimism.

The current paper adds to the existing literature in two ways. First, it extends Malmendier and Nagel's (2011) evidence along a geographical dimension by testing it for the case of Europe. Second, it probes further into what type of experiences might be relevant by distinguishing the effects of extreme stock market downturns from those obtained more generally.

For the first contribution, the paper applies the methodology developed by Malmendier and Nagel (2011) to a novel data set on household finances, the Eurosystem Household Finance and Consumption Survey (HFCS). This data set provides information on households' willingness to take financial risks and on participation in financial markets, along with a large number of important control variables, in a harmonized fashion for several countries in the euro area. Our data cover more than 58,000 households in Austria, Belgium, Finland, France, Germany, Greece, Italy, Luxembourg, the Netherlands, Spain and Portugal, i.e. in eleven different countries of the euro area.

While our measure of the willingness to take financial risks varies relatively little, stock market participation<sup>4</sup> is widely different across countries, ranging from an average of 3% in Greece to 22% in Finland. Also, the data show considerable variation in the experienced stock market returns. Within countries, it typically holds that the younger have experienced lower stock market returns than the older. At the same time, the experiences differ substantially across countries, with the young in some countries having experienced higher returns than the old in some other countries. This substantial cross-country variation allows us to identify experience effects separately from age effects despite the fact that only one wave of the survey is currently available.

Our estimates of the effects of lifetime experiences on the willingness to take financial risks and stockholdings among euro area households are fully in line with those identified in Malmendier and Nagel (2011). They are statistically significant and economically substantial. To give just a few examples,

<sup>&</sup>lt;sup>4</sup> Direct stockholdings, or holdings via investments in mutual funds that invest predominantly in equity.

households at the 90th percentile of the distribution of experienced stock returns are 7 percentage points more likely to report that they are willing to take financial risks than households at the 10th percentile, and 11 percentage points more likely to be invested in the stock market. In contrast, when we conduct a placebo experiment (where we randomly assign a different nationality to each cohort in a given country), the pseudo-lifetime experiences are not found to affect stock market participation. This shows that our results are not the artifact of a spurious relation.

While these estimates match those for the United States qualitatively, our evidence suggests that the effect of experienced stock market returns is less persistent in Europe. Still, also in Europe experiences matter for the willingness to take financial risks and stock market participation for several years.

Subsequently, the paper distinguishes the general effects of experiences from those of extreme stock market outcomes (during which nominal stock market returns decline by more than 20% in a given year). The effects of such events are substantial – for each additional experienced event, the tendency to hold stocks shrinks by 2 percentage points. Over the interdecile range of the experience distribution, this amounts to a 9 percentage point difference in stockholdings. Importantly, these effects are observed on top of those triggered by the average experienced stock market returns.

The paper therefore provides further evidence on the relevance of personal experiences for household behavior. These findings have important policy implications. It is a well-known fact that households are generally underinvested in the stock market, a phenomenon that has been dubbed the stockholding puzzle (Haliassos and Bertaut 1995; Campbell 2006). The puzzle is particularly pronounced in Europe, where household stock market participation is even lower than in the United States. This is especially problematic given that households have been made more and more responsible for their own finances after retirement (van Rooij et al. 2011). The findings in the current paper imply that stock market participation will likely be further depressed due to the recent experience of the 2008 stock market crash, suggesting an even more pronounced underinvestment of European households in the stock market in times to come. Policy-makers should therefore monitor developments carefully, and possibly consider policies to encourage stock market participation among the most-affected groups.

The paper proceeds as follows. Section 2 provides more detail on the underlying data and the econometric methodologies that we employ. Section 3 reports the main findings regarding the effect of individuals' stock market experiences on the willingness to take financial risks and stock market participation, and provides the results of several robustness tests. Section 4 expands the evidence by focusing on the consequences of extreme events. Section 5 concludes.

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## 2. Data and methodology

## 2.1 Data

In order to conduct our analysis we combine household-level data from the HFCS and historical data for stock returns. The HFCS provides ex-ante comparable data for 15 euro area countries.<sup>5</sup> Since we could not obtain sufficiently long historical data for the stock market performance of Cyprus, Malta, Slovakia and Slovenia, we discard them from the analysis. Effectively, our data cover therefore more than 58,000 households in 11 euro area countries, namely Austria, Belgium, Finland, France, Germany, Greece, Italy, Luxembourg, the Netherlands, Spain and Portugal.

The HFCS contains information regarding socio-demographic variables, assets, liabilities, income and consumption for a sample of households that is representative both at the national and the euro area level. A set of population weights is provided in order to ensure the representativeness of the sample. All our calculations use these population weights. In section 3.2 we perform unweighted calculations as part of our robustness checks.

Another important feature of the HFCS is that missing observations for all variables necessary to construct wealth and income aggregates (i.e. questions that were not answered by the respondent households) are imputed five times – an issue that we will take into account when assessing the statistical significance of our estimates. The HFCS data refer to the year 2008 in Spain, to 2009 in Finland, Greece and the Netherlands, and to 2010 in all remaining countries. We account for these differences when calculating respondents' lifetime experiences. It is important to note, however, that all the households in our sample have lived through the 2008 stock market crash.

From the HFCS we are going to retrieve our dependent variables and a set of control variables. The variables of interest are the household's willingness to take financial risks, whether it participates in the stock market or not, and the share of liquid assets invested in stocks.

For determining the household's willingness to take financial risks we use the following question: *"Which of the following statements comes closest to describing the amount of financial risk that you (and your husband/wife/partner) are willing to take when you save or make investments?"* The respondent can choose one of the following options: 1. *Not willing to take any financial risk, 2. Take average financial* 

<sup>&</sup>lt;sup>5</sup> For more details on the survey, see <u>http://www.ecb.europa.eu/home/html/researcher\_hfcn.en.html</u>. The results from the first wave are described in detail in Household Finance and Consumption Network (2013a), and the most relevant methodological features of the survey are discussed in Household Finance and Consumption Network (2013b).

risks expecting to earn average returns, 3. Take above average financial risks expecting to earn above average returns, or 4. Take substantial financial risks expecting to earn substantial returns.<sup>6</sup>

For stock market participation, we include direct stockholdings as well as investments in mutual funds which invest predominantly in equity. For the share of liquid assets invested in stocks we define liquid assets – in the same way as Household Finance and Consumption Network (2013a) – as the sum of the value of sight accounts, savings accounts, mutual funds, bonds, ownership of non-self-employment private businesses, shares and managed accounts.<sup>7</sup>

In all our model specifications we will control for gender, age, income, education, the stock of liquid assets, whether the reference person<sup>8</sup> is married, retired, has children or works in the financial sector. The controls follow Malmendier and Nagel (2011), but we added the financial sector affiliation (because it might affect the household's tendency to hold stocks) and gender (since there is an ample literature documenting that risk attitudes differ between men and women). Finally, we also control for country-fixed effects, given that the literature has found cross-country differences in stock ownership to be primarily linked to differences in economic environments and institutions (Christelis et al. 2013). Furthermore, country-fixed effects take account of possible differences in reporting styles across countries.

In order to construct the stock market experiences which the households in our sample have lived through, we use long-term historical time series obtained from Global Financial Data. We use real stock returns (deflated with consumer prices) from 1930 until the year prior to the survey. Since the data do not go back further in time than 1930 (1932 in Portugal), we treat all households born before 1930 as if they were born in 1930 (1932 in Portugal).<sup>9</sup>

<sup>&</sup>lt;sup>6</sup> Unfortunately, this question has not been asked in France and Finland. Also, it has not been imputed for all countries, which somewhat restricts the available sample size. Note that we changed the ordering of this variable relative to the way it is measured in the HFCS to match the measurement in Malmendier and Nagel (2011). Accordingly, high values in the original HFCS data set correspond to low values for our variable, and vice versa.

<sup>&</sup>lt;sup>7</sup> Malmendier and Nagel (2011) also include stocks held in retirement accounts, a variable that is not available for the HFCS. In the robustness section, we will include households that have invested in voluntary pension schemes to get closer to their definition.

<sup>&</sup>lt;sup>8</sup> Throughout the paper, "household" and "reference person" should be seen as interchangeable concepts. For example, when we talk about the age of the household it is understood that we are referring to the age of the reference person. The household reference person is chosen according to the international standards of the so-called Canberra Group (UNECE 2011). This definition uses the following sequential steps to determine a unique reference person in the household: (i) household type, (ii) the person with the highest income, (iii) the eldest person.

<sup>&</sup>lt;sup>9</sup> This affects 3,636 households. Dropping them from the sample does not change the results in any relevant manner – as we will see, experiences before 1930 would anyway get a negligible weight in determining household behavior in current times. For Greece, Global Financial Data extends back only to 1953, but we were able to expand the series back to 1930 using data provided to us by the Bank of Greece.

We furthermore generate a variable that measures how often a household has experienced a substantial drop in stock prices, which we define as an annual return of below -20%. This threshold coincides roughly with a one-standard-deviation event, and it covers around 10% of our year-country observations. Such a decline could occur due to a genuine stock market crash or, alternatively, through a sustained but more gradual decline over the course of a year. Since our data are annual, we cannot distinguish between the two. Of course, we will subject the results to a robustness test where the definition of a stock price drop is altered, to an annual return of below -40%, which roughly amounts to a 1.7-standard-deviation event, and covers around 2.5% of our year-country observations. Note that we base this variable on nominal returns, whereas the overall stock market experiences were calculated using real returns. The reason is that for smaller movements in the stock market, what matters for consumers is the real return they can make with their investment, whereas stock market crashes are typically defined using nominal returns. A robustness test using real returns to define crashes does not alter our results.

#### 2.2 Methodology

We are interested in studying the effect of past experiences on the willingness to take financial risks, and the portfolio-choice decisions of households. Following Malmendier and Nagel (2011), we synthesize the lifetime experienced returns of a household using a weighted average of the annual returns conditional on a weighting parameter  $\lambda$ . The weighting scheme is flexible enough to allow households to give either higher or lower weights to more recently experienced returns. In particular, for each household *i* in country *c*, the experienced return is constructed as follows:

$$A_{ic}(\lambda) = \sum_{k=1}^{age_i - 1} w_i(k, \lambda) R_{T-k}^c, \tag{1}$$

$$w_i(k,\lambda) = \frac{(age_i - k)^{\lambda}}{\sum_{k=1}^{age_i - 1} (age_i - k)^{\lambda}}.$$
(2)

 $R_{T-k}^c$  denotes the stock market return in year *T-k* (where *T* is the reference period of the survey) in country *c*. The weights  $w_i(k, \lambda)$  depend on the age of the household and a weighting parameter  $\lambda$  which determines the shape of the weighting function (in particular, whether the slope is positive, negative or flat), and the steepness of the slope.

To understand the form of the weighting function, Figure 1 depicts possible weights for the example of a 50-year-old household, using different values of  $\lambda$ : -0.2, which corresponds to an increasing weighting function (where the distant past matters more than the recent past); 1, which implies linearly

decreasing weights; and 5, a concavely decreasing weighting function. Generally, a negative  $\lambda$  implies that the household places a larger weight on more distant experiences, whereas a positive  $\lambda$  indicates that returns from the recent past are given a larger weight. As  $\lambda$  increases, the effect of past returns fades away more quickly and recent returns are given a relatively larger weight.

#### Figure 1 here

When calculating lifetime experiences in this manner, we impose a number of assumptions. First, we assume that the relevant horizon extends back to the year of birth. This assumption turns out not to be critical, as we will show by varying the start of the relevant horizon, once to include 10 years prior to birth, and once to start 10 years after birth. A second assumption is that all households "experience" stock market returns, whether they are actually holding stocks or not. Third, we assume that it is the national stock market returns that matter, and thereby implicitly that the reference person did not live abroad or experienced stock market returns in another country by some other means, e.g. by holding an internationally diversified portfolio. While country size might be a relevant factor in this, we think of the latter as a realistic assumption due to the well-known home bias in portfolios, and will subject the former to a robustness test by excluding all households that were not born in the country of residence.<sup>10</sup>

We are going to estimate  $\lambda$  from the data. In general, our regression models will have the following form:

$$y_{ic} = \alpha_c + \beta A_{ic}(\lambda) + \delta x_{ic} + \varepsilon_{ic}, \qquad (3)$$

where  $y_{ic}$  denotes the measure for the willingness to take financial risks, the variable indicating whether a household participates in the stock markets, or the share of stocks in liquid assets.  $\alpha_c$  are the countryfixed effects,  $x_{ic}$  the various control variables and  $\varepsilon_{ic}$  a residual. Since  $A_{ic}(\lambda)$  is a non-linear term, we have to use non-linear estimation techniques, irrespective of the remaining model specification.

Note that this model identifies experience effects via the variation of experiences over age and across countries. In the paper by Malmendier and Nagel (2011), identification was achieved by using several waves of the U.S. Survey of Consumer Finances (SCF), such that experiences vary over age and across waves. In other words, equation (3) simply substitutes their time subscript with a country subscript.

<sup>&</sup>lt;sup>10</sup> IMF data suggest that the share of foreign stocks in overall stock holdings across the countries under study is 9% on average. We assume that these are unevenly distributed across households, with relatively few sophisticated households holding relatively more foreign stocks

The idea of identification is, however, equivalent, and based on a similar amount of variation in the data.<sup>11</sup>

We first look at the effect of experiences on the willingness to take financial risks. Since the dependent variable takes four values, we use an ordered probit model for the estimation. When our dependent variable is the stock market participation decision we use a probit model, and when we look at the share of the portfolio invested in stocks we use a Tobit model.

When the experienced return is our independent variable of interest, we first identify an initial value for  $\lambda$  by estimating the model on a tight grid of given lambdas. The value for  $\lambda$  that achieves the highest likelihood is then used as the initial value in the non-linear estimation. This procedure ensures avoiding local maximums, apart from substantially reducing computation time.

Our other independent variable of interest is the number of stock market crashes experienced. For the model specifications dealing with this independent variable we do not include a weighting function, thereby implicitly assuming that the effects of crashes persist and accumulate. Therefore, it is important to allow for a non-linear effect, which we do by using a quadratic term, such that the model is estimated as follows:

$$y_{ic} = \alpha_c + \beta_1 S_{ic} + \beta_2 S_{ic}^2 + \delta x_{ic} + \varepsilon_{ic}.$$
(4)

All variables are described as in equation (3), and  $S_{ic}$  is the number of experienced stock market crashes.

We use a weighted estimation to account for the fact that the survey does not always represent the same fraction of the overall population across countries. Our weights readjust each observation to reflect their relative importance for the euro area as a whole. In so doing, we also follow Faiella (2010) and Magee et al. (1998), who recommend the use of weights for two similar surveys, namely Italy's Survey on Household Income and Wealth and Canada's SCF. They argue that, in surveys with complex survey design, the use of weights protects against the omission of relevant information, which otherwise would have to be modelled explicitly by incorporating all available geographic and operational variables that determine sampling rates. Another reason for using weights is due to the possibility of endogenous sampling (Solon et al. 2015), since the HFCS oversamples wealthy households, and given that stock market participation varies with wealth.

<sup>&</sup>lt;sup>11</sup> The number of observations is comparable, with around 51,000 for the U.S. and around 58,000 for Europe. There are 19 waves in the U.S. data, and 11 countries in the European data. The mean and standard deviation of stock market experiences for  $\lambda$ =1.5 are also comparable, at 0.091 and 0.022 for the U.S. and 0.087 and 0.020 for Europe.

#### 2.3 Descriptive statistics

Table 1 provides descriptive statistics for households' willingness to take financial risks, stock market participation and the share of liquid assets invested in stocks. The willingness to take financial risks shows little variation, both within and across countries. In eight of the nine countries where this variable is available (remember that this question was not asked in Finland and France), the median household reports the lowest willingness to take financial risks (coded as 1). Italy is the only exception, with a median of 2. The mean figure is 1.4 for the euro area as a whole, and it varies from 1.1 in Portugal to 1.7 in Italy. Overall, these results are not very different from the mean value of 1.8 that was found for U.S. households in Malmendier and Nagel (2011).<sup>12</sup> Still, as we will see, despite the low variability of this variable, it is sufficient to estimate meaningful results.

#### Table 1 here

Participation rates in stock markets are very low (see the second panel of Table 1); only 13% of households report some stockholdings. Importantly, however, there is considerable variation across countries, with participation rates ranging from 3% in Greece to 22% in Finland. Conditional on stock-market participation, euro area households keep 30% of their liquid assets in stocks. But this figure, reported in the third panel of Table 1, also varies across countries. The mean ranges from 24% in Germany and the Netherlands to 38% in Greece. Interestingly, there is also a substantial amount of variation within countries. There are many households with very small amounts of stocks in their portfolios, as shown by the small numbers for the 10th percentile, whereas the 90th-percentile household in several countries holds substantial amounts of stocks (e.g. above 80% in Finland, Greece, Luxembourg and Spain).<sup>13</sup> Taken together, the low participation rates and the small fraction of assets that are held in stocks suggest that households account for a very small fraction of stock market capitalization, thereby making concerns about reverse causality (whereby changes in households' willingness to take financial risks affect stockholdings and thereby stock market returns) less relevant.

#### Table 2 here

Table 2 provides a first look at our main explanatory variables. In the upper panel, we report summary statistics for the experienced stock market returns of households,  $A_{ic}$ . They are calculated using a

<sup>&</sup>lt;sup>12</sup> Looking at previous waves of the Spanish and Italian components of the survey, it turns out that the results have been remarkably stable over the pre-crisis years.

<sup>&</sup>lt;sup>13</sup> The dependent variable in our regressions will not be conditional on stockholdings, i.e. we include households that do not hold stocks in our sample.

weighting factor of  $\lambda$ =4.5, which is close to the estimates that we will report below. There is substantial variability in the experiences across and within countries: they range from 4% in Italy to 13% in Finland. The variation within countries is largest in Greece, where the 10th percentile of the return distribution is 3% and the 90th percentile is 13%.

Figure 2 plots the experienced returns by country and age group. Within countries, it is typically the case that the younger have experienced lower stock market returns than the older. At the same time, the experiences differ substantially across countries, with the young in some countries having experienced higher returns than the old in some other countries. This substantial cross-country variation is important, as it allows us to identify experience effects separately from age effects despite the fact that only one wave of the survey is currently available.<sup>14</sup>

## Figure 2 here

The variability in real stock market returns is largely due to differences in nominal returns, and only to a small extent to differences in inflation rates. Table 3 shows the correlations between each country's nominal stock market returns for the whole sample from 1930 to 2010. Correlations are rarely higher than 0.5, and in a few cases they even take negative values.<sup>15</sup>

## Table 3 here

When we examine the number of protracted stock market declines or genuine stock market crashes that households have experienced (reported in the second panel of Table 2), we once more find substantial variability across and within countries. The mean number of stock market downturns that households have experienced ranges from 3.4 in Austria to 11.6 in Portugal. In many countries, the difference between the 10th and 90th percentiles of the distribution is as large as, or even larger than, six events.

To summarize, the descriptive statistics show that there is substantial variation in our dependent and explanatory variables both across and within countries. We next study how an individual's experience affects the willingness to take financial risks and stock market participation.

<sup>&</sup>lt;sup>14</sup> Disentangling these effects is important, given that age has been shown to affect financing decisions of households (see, .e.g., Agarwal et al. (2009) and Korniotis and Kumar (2011).

<sup>&</sup>lt;sup>15</sup> In addition to the substantial cross-country variation in stock market performance, there has also been substantial crosscountry variation in business cycles, even after the introduction of the euro (Camacho, Perez-Quiros and Saiz 2006, 2008). In the absence of such variation, it would be difficult to disentangle age from experience effects, for instance because lifetime earnings are known to depend on the job market situation when individuals enter the labour market (Kahn 2010).

# 3. The effect of experiences on the willingness to take financial risks and stock market participation

#### 3.1 Benchmark results

Table 4 provides the first set of results. It reports the estimated coefficients of the ordered probit model, explaining the willingness to take financial risks, as well as the average marginal effects for each category. Note that the standard errors take account of the multiply imputed nature of the data, thereby properly reflecting the uncertainty of the imputed values. Several of the control variables are relevant. Higher income and a higher stock of liquid assets tend to decrease the willingness to take financial risks, even though for both variables there are important non-linearities, as suggested by the statistical significance of the squared terms. The retired are somewhat less willing to take financial risks than other households, an effect that is found on top of a decreasing willingness to take financial risks with age (the latter has already been documented in the literature, see Dohmen et al. (2011)). Education also seems to matter, with higher levels of education being associated with a higher willingness to take financial risks. As is well known from the literature (see, inter alia Borghans et al. (2009)), males tend to be more willing to take risks than females, a pattern that is also observed in our data. Our control for respondents who are working in the financial sector is highly statistically significant, and suggests that these individuals are more willing to take financial risks (the average marginal effect suggests that they are 7.5 percentage points less likely to be unwilling to take any financial risk, which makes the financial sector dummy, together with gender, the most influential socio-demographic factor). Finally, the country-fixed effects are estimated to be highly relevant.

#### Table 4 here

Moving to the two main parameters of interest,  $\beta$  and  $\lambda$ , both are statistically significant and have the expected sign. The weighting parameter  $\lambda$  is estimated to be 3.9, considerably larger than Malmendier and Nagel's (2011) corresponding estimate of 1.4 for the United States. This points to a higher decay factor in Europe. To take the example of a 30-year-old individual, a European would assign a weight of 15.6% to the previous year's experience, whereas a U.S. household would give it a weight of only 7.9%. Despite this large initial difference, memories are rather persistent also for the European household, who is estimated to assign a weight of 3.7% to experiences undergone 10 years ago (whereas the number in the United States amounts to 4.7%). Taking the example of an individual with a longer life history, the relevance of past experience becomes even more apparent: according to our estimates, a 50-year-old person would weigh the most recent year with 9.5%, and the experience undergone a decade

ago with 4.3%. Even the stock market returns experienced 20 years ago would enter the weighting function with 1.4%.

As expected, the coefficient estimate for  $\beta$  indicates that higher experienced returns tend to increase the willingness to take financial risks. The average marginal effects show that an increase in experienced returns by 1 percentage point makes households 1.4 percentage points more likely to declare that they are not willing to take any financial risk. Comparing the average of the fitted probabilities at the 90th percentile of the distribution of experienced returns with the average of the fitted probabilities at the 10th percentile yields a difference of 6.7 percentage points. This effect is of substantial magnitude (it is similar to that found for financial sector employees or males), and is comparable to the 10.3 percentage points that were identified by Malmendier and Nagel (2011) for the United States.<sup>16</sup>

The next question is whether there are any repercussions on actual stock market participation. Table 5 reports the results from the probit model explaining the households' participation decision. Once more, a number of control variables appear to be significant. Participation is found to increase for males as well as for households with high liquid assets, high education and those working in the financial sector. Compared to Germany, stock market participation is higher in Belgium and France, and lower in Austria, Luxembourg and Portugal.

#### Table 5 here

As before, parameter  $\lambda$  is significantly estimated, and at 5.2 larger than what was found for the United States (1.3). Once again, however, the parameter still implies that memories persist – for the 30-year old, experiences undergone 10 years ago receive a weight of 2.8%; for a 50-year old, it amounts to 4.1%.

Parameter  $\beta$  is statistically significant. Judging from the marginal effect and the interdecile range reported in Table 5, it is apparent that its magnitude is economically important – a one-percentage-point higher experienced stock return increases the propensity to hold stocks by 2 percentage points, and the difference in stock market participation along the interdecile range of the stock market experiences amounts to 11 percentage points, which is rather close to the 10 percentage points estimated by Malmendier and Nagel (2011), and again similar to the effect of working in the financial sector.

<sup>&</sup>lt;sup>16</sup> The difference between the 90th and the 10th percentile is broadly comparable between the euro area and the United States. At the respectively estimated  $\lambda$ , it amounts to (11.9%-6.2%=5.7%) for the United States, and to (9.3%-4.2%=5.1%) in the euro area.

The third test is conducted on the share of liquid assets invested in stocks. The results, reported in Table 6, are based on a Tobit model, such that the coefficients are now directly interpretable.<sup>17</sup> The share of stocks in the liquid assets held by financial sector employees is 26 percentage points higher than among other households. Furthermore, the share of stocks rises with the stock of liquid assets and education (college graduates have a 19-percentage-point higher share of stock investments than households with less than a high school degree).

#### Table 6 here

As previously, we estimate statistically significant parameters for  $\lambda$  and  $\beta$ .<sup>18</sup> Comparing households along the interdecile range suggests that those at the 90th percentile of the distribution invest 5 percentage points more in stocks than those at the 10th percentile (once more, these numbers are comparable with those for the United States).

#### 3.2 Robustness tests

We have subjected our results to a large number of robustness tests. First, analogous with Malmendier and Nagel (2011), we have also tested whether similar results can be obtained for bond market experiences and their effects on bond holdings.<sup>19</sup> Judging from the descriptive statistics, there is much less variability in bond market returns than in stock market returns. In large part, this is due to the near-complete convergence of government bond yields in the euro area between 1999 and 2010 (Ehrmann et al. 2011). Accordingly, we expect our results to be weaker than for stockholdings. Comparing the estimates for  $\beta$  and  $\lambda$  (reported in Table 7) between the benchmark model in row (1) and those for bond markets in row (2), it is apparent that we estimate a rather similar coefficient for  $\lambda$ , at 3.99 (compared to 5.24 for stocks). Parameter  $\beta$ , in contrast, is only marginally significant for the bond market participation decision.

#### Table 7 here

<sup>&</sup>lt;sup>17</sup> Non-linear least-squares models for the shares conditional on stockholdings (i.e. excluding households with a share of zero) did not lead to any significant results. This suggests that households' experiences mainly affect their participation decision, rather than the amounts held.

<sup>&</sup>lt;sup>18</sup> Our estimates of  $\lambda$  are quite different for the effect of experiences on households' willingness to take financial risks, stock market participation and the share of stocks in liquid assets, whereas they are rather similar across these three models in Malmendier and Nagel (2011). We do not see any reason why they would need to be similar across the three specifications, given that they measure very different concepts, which might be affected differently by previous experiences.

<sup>&</sup>lt;sup>19</sup> Bond returns are calculated for long-term bonds. Since bond returns for Luxembourg are not available prior to 1947, we exclude Luxemburgish households born before 1947. The bond holdings are defined in analogy to the stockholdings as directly held bonds or investments in mutual funds that themselves predominantly invest in bonds.

The remaining robustness tests, reported in rows (3) to (16) of Table 7, go back to explaining the stock market participation decision as a function of stock market experiences. The first of these allows for an additional effect of experienced stock market volatility. For that purpose, we added the experienced stock market volatility (calculated as the weighted standard deviation of the respondents' lifetime experience, using the previously estimated  $\lambda$  as a weighting parameter) to the benchmark regression. As can be seen from row (3) of Table 7, our results remain robust. While the experienced volatility itself lowers stock market participation in a statistically significant manner (a result that has also been found for the United States in Appendino (2013)), the effects of the experienced returns and the weighting parameters are basically unaltered.

Results are also stable for the robustness test in row (4), where we broadened the definition of stockholdings to include not only direct stockholdings and investments in mutual funds that themselves predominantly invest in stocks, but also investments in voluntary pension plans. This change in definition raises the stock market participation rate of euro area households from 13% to 39%. Still, all results go through.

For the subsequent robustness test, we reran our estimations without using population weights. Here, the quantitative results change, but qualitatively remain robust. The experienced stock returns exert a smaller effect on stockholdings, and the weighting parameter is substantially larger, indicating that the more recent experiences matter more. Where do these differences come from? The new set of results treats each observation equally, whereas, before, observations reflected the countries' population shares in the euro area. In Table 1, it is evident that countries such as France and, in particular, Finland receive much more prominence in the new estimation (since they have by far the largest samples in the survey, exceeding their population share), whereas the relevance of, for instance, German observations diminishes when using an unweighted regression (since the approximately 3,500 households representing Germany in the HFCS make up 6% of the overall sample, whereas the German households effectively account for around 29% of the euro area household population). The change in coefficients does therefore point to differences in the economic significance of the effects across the various countries. As we will see below, these differences are tightly related to how severely the countries were hit by the 2008 stock market crash. Finland and France were among the more strongly affected countries compared to Germany, and in the countries with the severest stock market crashes, the most recent experience receives a rather strong weight.

The fifth robustness test includes an additional regressor, namely the bond returns that households have experienced over their lifetimes (keeping the weighting parameter from the robustness test provided in row (2), i.e. when explaining bond market participation with experienced bond returns).

Experienced bond returns themselves exert a significant effect on stockholdings. As one would expect, this somewhat diminishes the quantitative importance of the experienced stock returns, but does not change the picture qualitatively (see row (6) of Table 7).

The next two rows of Table 7 show how our results change if we vary the experience horizon of respondents, by either including 10 years prior to birth (thereby allowing for an influence of the experience of respondents' parents), or starting 10 years after birth. Mechanically, we should expect that a longer horizon leads to a larger estimate of  $\lambda$ , as the additional years need to be weighted down if they are not relevant. In contrast, for the shorter horizon, we should expect  $\lambda$  to fall. This is indeed what we find – however, while the magnitudes of our parameters change, the overall results are not affected qualitatively. Also, shortening the time horizon further (to, e.g., one that starts at the age of 18 years) does not lead to different conclusions.

In row (9), we also show that including the willingness to take financial risks as an additional regressor has barely any impact on the results. While not a definite test, this finding suggests that the effect of experiences on stockholdings works primarily via influencing beliefs rather than preferences, as also argued by Malmendier and Nagel (2011). In row (10), we add the level of a household's real asset holdings, since these could be seen as a substitute to stockholdings. We find our results to be unaltered. Row (11) includes year-of-birth dummies as control variables and it shows that our results are not driven by cohort effects.

Row (12) of the table shows the result for a regression in which we exclude immigrants from the sample. Specifically, we drop all households who were born in a country different from the one they have been interviewed in, since immigrants are more likely to have been exposed to stock market returns in countries other than their country of residence. We exclude France, Spain and the Netherlands, since we do not have information on the country of birth of the household for these households. Again, all our results hold.<sup>20</sup> The subsequent robustness test, reported in row (13), clusters standard errors by country. All results go through.

Finally, as a way to test for possible spurious correlations, we run a placebo experiment.<sup>21</sup> For that purpose, we randomly assign a different nationality to each cohort in a given country (for instance, all 35-year-old households in France are randomly allocated a nationality other than the French one, all 36-year-old French households are independently assigned a random nationality, etc.). With this placebo

 $<sup>^{20}</sup>$  As can be seen in Table 7, the coefficients for this robustness check differ from the ones in the baseline specification, but this is due to the different samples used. When we run the baseline specification excluding France, Spain and the Netherlands, the results are almost identical to those of row (10) in Table 7.

<sup>&</sup>lt;sup>21</sup> We are grateful to Dimitris Georgarakos for suggesting this idea.

allocation of nationalities, we then rerun our estimations. As can be seen from row (14) of Table 7, the pseudo-lifetime experiences are not found to affect stock market participation: they are neither statistically significant nor economically large.

#### 4. Any difference for extreme events?

The experience of the stock market crash in 2008 is still vividly remembered by stock market participants. Many of these have lost substantial amounts of wealth, which in turn has been shown to affect risk taking (Necker and Ziegelmeyer 2013). A natural question is therefore whether extreme events such as stock market crashes influence attitudes and behaviors in a more persistent manner than less-extreme experiences. Related evidence supporting this hypothesis is provided by Ehrmann and Tzamourani (2012), who show that the effect of experienced inflation on inflation aversion typically fades away, whereas memories of hyperinflation tend to stay in people's minds and affect attitudes in a much more persistent manner.

While of interest on their own, studying the effect of crashes on households also provides another robustness test of the previous results. We had assumed that households "experience" regular stock market developments, regardless of whether they actually hold stocks. For the current estimations, we can relax this assumption partially and only need to assume that households take note of stock market crashes, which seems a rather uncontroversial assumption.

#### Table 8 here

Table 8 reports the estimates of the effect of stock market crashes – or protracted stock market declines – on the willingness to take financial risks. Note that this specification does not contain a  $\lambda$  factor, i.e. we simply count the number of such experiences the individuals have made over their lifetimes and enter this as an explanatory variable (thereby already assuming that these experiences remain an important factor in influencing the willingness to take financial risks and stock market participation, and that they are additive). The results indicate that for each such additional experience, the propensity not to take any financial risks increases by 1 percentage point. Looking at the interdecile range, this amounts to a difference of 4.4 percentage points. While this number might not sound overly large, it is important to note that many of the stock market declines were experienced a considerable time ago (more than 70% before 1990, 45% before 1970). These numbers take into account a non-linearity in the effects: the squared number of experienced events enters with a significant positive sign, suggesting that with

increasing numbers of experienced stock market downturns the decrease in the willingness to take financial risks becomes less pronounced.

The propensity to hold stocks is affected in a similar fashion, as can be seen from Table 9. Here, the fitted probabilities along the interdecile range generate a difference in stockholding propensities of 9.1%, i.e. nearly as much as the differences generated by the interdecile range in the experience of stock market returns themselves. In contrast, the share of liquid assets invested in stocks does not seem to be affected by the number of experienced stock market downturns (given that the parameter estimates reported in Table 10 are statistically insignificant, and the difference across the interdecile range is small), suggesting that the effect is more one of whether to hold stocks than how much to hold in stocks.

#### Tables 9 and 10 here

We extended the analysis in several dimensions, focusing in particular on stock market participation. These extensions, as well as a number of robustness tests, are reported in Table 11 – which repeats the average marginal effect obtained in the benchmark estimations in row (1). First, we combined regression models (3) and (4) by including both  $S_{ic}$ , the number of experienced stock market downturns (as well as its squared value), and  $A_{ic}$ , the experienced returns. The results show that the effect of experienced returns and the weighting parameter  $\lambda$  barely change in the new specification compared to the previous results, whereas the number of experienced downturns (reported in row (2) of Table 11) exerts an additional effect on the participation decision.

#### Table 11 here

We have furthermore extended equation (4) by allowing for a separate effect of stock market booms (which we defined in analogy to downturns as nominal annual returns in excess of +20%). We find that booms are much less relevant than downturns (the coefficient, not shown in the table, is insignificant). Even though one might expect that more households are inclined to invest in the stock market during boom times (and stay invested subsequently), this effect is not evident in the data. In contrast, the coefficient estimates for the effect of downturns remain basically unaltered (row (3) of Table 11).

Given that the survey was conducted just after the 2008 stock market crash, all households in our sample have experienced at least one crash. To get at the importance of the most recent crash on household portfolios, we made use of the fact that the 2008 crash was hitting the various countries in rather different ways. Based on the analysis in Bekaert et al. (2014), we split the countries into those that were affected by the crisis somewhat less severely (namely Austria, Belgium, Germany, Spain, Luxembourg and the Netherlands, which on average saw their stock markets decline by 36%), and those

where stock markets were hit particularly badly (i.e. Finland, France, Greece, Italy and Portugal, with an average drop of 52%), and then repeated the analysis of section 3 separately for each country group. The results are provided as the two bottom rows of Table 7. There are remarkable differences across the two groups: whereas our results are robust for the countries that got hit less badly, the weighting parameter  $\lambda$  in the more strongly affected countries is estimated at 10.9. This implies that, in these countries, the experience of the recent crisis overshadows the earlier experiences, which receive a much smaller weight in households' decisions. It can help to explain why  $\lambda$  is estimated to be so much higher in Europe than in the United States (given that Malmendier and Nagel (2011) used several waves of the SCF, therefore also covering the years prior to the recent crisis).

Finally, we subjected our findings in this section to a number of robustness tests, by (i) changing the definition of a downturn to cases where annual nominal stock returns were below -40% (moving us from 10% to around 2.5% of the year-country observations), (ii) including voluntary pension plans in our definition of stockholdings, (iii) estimating the models without using population weights, (iv) including the household's willingness to take financial risks as an additional regressor, (v) adding the level of real asset holdings, (vi) including year-of-birth dummies as control variables, (vii) excluding immigrants from the sample and (viii) clustering standard errors by country. Results are reported in rows (4) to (11) of Table 11. This table shows that for more extreme events, the effects are substantially larger, as well as when we broaden the definition of stockholdings to include voluntary pension plans. The average marginal effect becomes insignificant if we run the regression unweighted and if we drop the immigrants from the sample (which also implies dropping France, Spain and the Netherlands because of data availability).

In addition, we conduct a placebo experiment analogous to the one explained in the previous section (row (12)). Once we randomly assign the number of crashes experienced, the effect of this placebo variable is not significant. This supports the validity of our results.

#### **5.** Conclusions

This paper has studied to what extent the experiences of households shape their willingness to take financial risks, their inclination to participate in stock markets and the amounts that they invest in stocks. It has applied the approach developed by Malmendier and Nagel (2011) and extended the evidence to Europe, using the Eurosystem Household Finance and Consumption Survey, a novel data set on household finances covering more than 58,000 households in eleven different countries of the euro area.

The data show considerable variation in the experienced stock market returns, stock market participation and the invested amounts both within and across countries. Our estimates show that experienced stock market returns exert statistically significant and economically substantial effects on households' willingness to take financial risks and portfolio decisions, even if we find that more distant experiences receive a somewhat lower (but still substantial) weight than the corresponding findings for the United States. In contrast, a placebo experiment shows that randomly assigned pseudo-lifetime experiences do not affect stock market participation. This evidence adds to the literature on time variations in the risk-taking attitudes of households and its determinants, as well as on the factors that shape households' portfolio decisions, emphasizing the importance of personal experiences for the formation of attitudes and economic behavior.

The paper has also tested whether the experience of extreme stock market downturns has a bearing on households' willingness to take financial risks and stock market participation. Here as well, the effects are substantial and – importantly – come on top of the experienced average stock market returns.

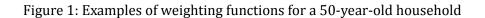
These findings have important policy implications. Households are known to be generally underinvested in the stock market (and more so in Europe than in the United States), especially in light of the fact that they have been made more and more responsible for their own finances after retirement. Particularly, the young and households in countries where the stock market crash in 2008 was very severe tend to give substantially more weight to the recent past when forming their participation decision. This, in turn, implies an even more pronounced underinvestment in stocks among these households in times to come. Policy-makers should monitor developments carefully, and possibly consider policies to encourage stock market participation among the most affected groups.

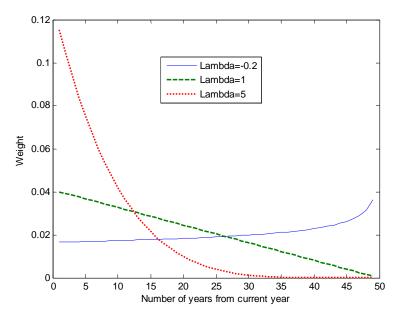
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Note: The figure plots weighting functions for a 50-year-old household according to equations (1) and (2), for different values of  $\lambda$ .

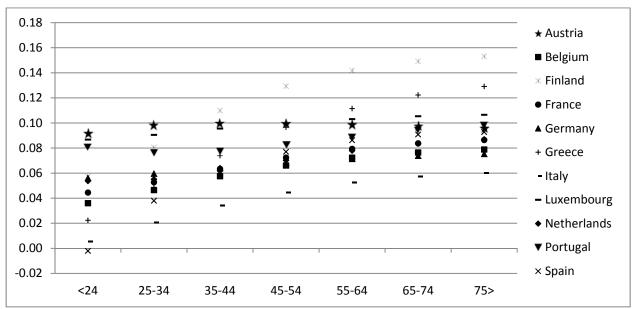


Figure 2: Experienced stock returns across age groups and countries

Note: The figure shows the experienced stock returns (calculated according to equations (1) and (2) with a  $\lambda$  of 4.5) across age groups and countries. Sources: Eurosystem Household Finance and Consumption Survey, own calculations.

Country	Mean	Std. Dev.	p10	Median	p90	Observations
		Willingness	to take fina	ncial risks		
Austria	1.48	0.71	1	1	2	2340
Belgium	1.33	0.60	1	1	2	2307
Finland						0
France						0
Germany	1.39	0.56	1	1	2	3467
Greece	1.31	0.66	1	1	2	2971
Italy	1.70	0.79	1	2	3	7951
Luxembourg	1.28	0.53	1	1	2	950
Netherlands	1.31	0.52	1	1	2	1253
Portugal	1.10	0.38	1	1	1	4365
Spain	1.19	0.47	1	1	2	6197
Euro Area	1.41	0.64	1	1	2	31801
			arket partici			
Austria	0.08	0.28	0	0	0	2380
Belgium	0.20	0.40	0	0	1	2327
Finland	0.22	0.41	0	0	1	10989
France	0.16	0.37	0	0	1	15006
Germany	0.16	0.37	0	0	1	3565
Greece	0.03	0.17	0	0	0	2971
Italy	0.06	0.23	0	0	0	7951
Luxembourg	0.18	0.39	0	0	1	950
Netherlands	0.15	0.36	0	0	1	1301
Portugal	0.05	0.22	0	0	0	4404
Spain	0.11	0.31	0	0	1	6197
Euro Area	0.13	0.34	0	0	1	58041
		Share of liquid				
Austria	0.28	0.26	0.02	0.22	0.71	130
Belgium	0.31	0.27	0.02	0.24	0.72	425
Finland	0.34	0.30	0.03	0.23	0.84	2996
France	0.33	0.28	0.03	0.25	0.79	3161
Germany	0.24	0.23	0.02	0.17	0.59	648
Greece	0.38	0.32	0.05	0.29	0.91	73
Italy	0.34	0.27	0.05	0.27	0.79	427
Luxembourg	0.34	0.30	0.02	0.26	0.86	123
Netherlands	0.24	0.26	0.02	0.13	0.76	170
Portugal	0.29	0.29	0.01	0.21	0.79	213
Spain	0.34	0.30	0.02	0.23	0.84	1387
Euro Area	0.30	0.27	0.02	0.21	0.76	9753

Table 1: Summary statistics: willingness to take financial risks, stock market participation and the share of liquid assets invested in stocks

Note: The table shows summary statistics for households' willingness to take financial risks (top panel), for whether or not households hold stocks (middle panel), and for the share of stocks in liquid assets, conditional on stock ownership (bottom panel). Sources: Eurosystem Household Finance and Consumption Survey, own calculations.

Country	Mean	Std. Dev.	p10	Median	p90	Observations
	E	perienced aver	age real stoc	ck return (λ=4.5)		
Austria	10.62	0.37	10.09	10.65	11.16	2380
Belgium	6.85	1.12	5.11	7.16	8.05	2327
Finland	12.95	2.81	8.28	14.07	15.57	10989
France	7.48	1.26	5.44	7.83	8.84	15006
Germany	8.06	0.15	7.84	8.06	8.26	3565
Greece	8.84	3.83	3.19	10.15	12.73	2971
Italy	3.86	1.39	1.93	4.16	5.38	7951
Luxembourg	10.39	0.37	9.82	10.41	10.86	950
Netherlands	7.50	1.11	5.87	7.67	8.84	1301
Portugal	8.86	0.91	7.60	8.92	10.12	4404
Spain	7.93	1.65	5.64	8.52	9.41	6197
Euro Area	7.32	2.27	4.24	7.94	9.33	58041
	N	umber of stock	market crasl	hes experienced		
Austria	3.39	2.55	1	3	8	2380
Belgium	4.96	1.49	3	5	7	2327
Finland	6.75	2.17	4	6	10	10989
France	7.82	2.49	5	7	12	15006
Germany	5.62	1.98	3	6	8	3565
Greece	10.19	2.69	8	9	14	2971
Italy	10.97	2.49	8	11	14	7951
Luxembourg	4.4	1.82	3	4	8	950
Netherlands	5.06	1.31	3	5	7	1301
Portugal	11.62	2.01	9	12	13	4404
Spain	6.68	2.06	4	6	10	6197
Euro Area	7.37	3.14	3	7	12	58041

Table 2: Summary statistics: lifetime experiences

Note: The table shows summary statistics for experienced stock returns (calculated according to equations (1) and (2) with a  $\lambda$  of 4.5), and for the number of experienced stock market crashes or prolonged downturns (defined as annual nominal returns below -20%). Sources: Eurosystem Household Finance and Consumption Survey, own calculations.

	Austria	Belgium	Finland	France	Germany	Greece	Italy	Luxembourg	Netherlands	Portugal	Spain
Austria	1.00	0.42	0.11	0.44	0.44	0.42	0.46	0.40	0.36	0.10	0.39
Belgium		1.00	0.42	0.58	0.47	0.18	0.46	0.75	0.57	0.14	0.49
Finland			1.00	0.34	0.32	-0.05	0.12	0.50	0.39	0.17	0.40
France				1.00	0.36	0.25	0.44	0.47	0.52	0.05	0.46
Germany					1.00	0.03	0.30	0.46	0.60	0.08	0.27
Greece						1.00	0.39	-0.08	-0.06	0.02	0.28
Italy							1.00	0.38	0.33	0.24	0.45
Luxembourg								1.00	0.51	0.25	0.37
Netherlands									1.00	0.13	0.37
Portugal										1.00	0.28
Spain											1.00

Table 3: Correlations between nominal stock market returns, 1930-2010

Note: The table shows correlations between annual national nominal stock market returns, 1930-2010.

			AME		AME		AME		AME	
	Coefficient	Std. error	Willingness=1	Std. error	Willingness=2	Std. error	Willingness=3	Std. error	Willingness=4	Std. error
Weighting parameter ( $\lambda$ )	3.860 ***	0.569								
Experienced return (β)	4.637 ***	1.463	-0.014 ***	0.005	0.010 ***	0.003	0.004 ***	0.001	0.001 ***	0.000
Log income	-0.186 **	0.092	-0.025 ***	0.008	0.015 ***	0.005	0.008 ***	0.002	0.002 ***	0.001
Log income squared	0.013 ***	0.005								
Number of children	-0.013	0.039	0.004	0.009	-0.003	0.006	-0.001	0.002	0.000	0.001
Number of children squared	0.000	0.013								
Log liquid assets	-0.119 ***	0.015	-0.032 ***	0.003	0.019 ***	0.002	0.011 ***	0.001	0.003 ***	0.000
Log liquid assets squared	0.013 ***	0.001								
Retired	-0.099 **	0.043	0.031 **	0.013	-0.020 **	0.009	-0.009 **	0.004	0.002 **	0.001
College	0.314 ***	0.043	-0.098 ***	0.014	0.062 ***	0.009	0.030 ***	0.004	0.006 ***	0.001
High school	0.192 ***	0.038	-0.060 ***	0.012	0.038 ***	0.008	0.018 ***	0.004	0.004 ***	0.001
Age	-0.020 **	0.008	0.005 ***	0.001	-0.003 ***	0.000	-0.001 ***	0.000	0.000 ***	0.000
Age squared	0.000	0.000								
Married	-0.086 **	0.034	0.027 **	0.011	-0.017 **	0.007	-0.008 **	0.003	-0.002 **	0.001
Financial sector employee	0.238 ***	0.056	-0.075 ***	0.018	0.048 ***	0.011	0.023 ***	0.005	0.005 ***	0.001
Male	0.238 ***	0.030	-0.075 ***	0.009	0.047 ***	0.006	0.022 ***	0.003	0.005 ***	0.001
Austria	0.051	0.058	-0.016	0.018	0.010	0.012	0.005	0.005	0.001	0.001
Belgium	-0.207 ***	0.052	0.065 ***	0.016	-0.041 ***	0.010	-0.019 ***	0.005	-0.004 ***	0.001
Greece	-0.061	0.075	0.017	0.016	-0.011	0.010	-0.005	0.005	-0.001	0.001
Italy	0.914 ***	0.073	-0.287 ***	0.022	0.182 ***	0.014	0.086 ***	0.007	0.018 ***	0.002
Luxembourg	-0.590 ***	0.071	0.184 ***	0.023	-0.117 ***	0.014	-0.056 ***	0.007	-0.012 ***	0.002
Netherlands	-0.239 ***	0.058	0.074 ***	0.018	-0.047 ***	0.011	-0.022 ***	0.005	-0.005 ***	0.001
Portugal	-0.763 ***	0.056	0.239 ***	0.018	-0.152 ***	0.011	-0.072 ***	0.006	-0.015 ***	0.002
Spain	-0.438 ***	0.050	0.136 ***	0.015	-0.087 ***	0.010	-0.041 ***	0.005	-0.009 ***	0.001
Fitted probabilities at p90-p10										
Willingness=1 (low)	-0.067 ***	0.002								
Willingness=2	0.042 ***	0.001								
Willingness=3	0.020 ***	0.001								
Willingness=4 (high)	0.005 ***	0.000								
Cutoff point 1	-0.271	0.478								
Cutoff point 2	0.949	0.479								
Cutoff point 3	1.985	0.482								
Pseudo R squared					0.11					

Table 4: The effect of experienced stock market returns on households' willingness to take financial risks

Note: The table shows estimated coefficients of the ordered probit model according to equation (3), explaining households' willingness to take financial risks, as well as average marginal effects (AME) for each category of the ordered probit. For the effect of experienced returns ( $\beta$ ), the table also reports the average of the fitted probability at the 90th percentile minus the average fitted probability at the 10th percentile of the distribution of experienced returns, for each category of the ordered probit.

	Coefficient	Std. error	AME	Std. error
Weighting parameter ( $\lambda$ )	5.244 ***	0.848		
Experienced return (β)	15.176 ***	3.104	0.022 ***	0.004
Log income	-0.168	0.289	0.021 ***	0.005
Log income squared	0.015	0.014		
Number of children	0.001	0.043	0.000	0.005
Number of children squared	-0.003	0.013		
Log liquid assets	0.301 **	0.123	0.066 ***	0.002
Log liquid assets squared	0.007	0.006		
Retired	-0.061	0.054	-0.009	0.008
College	0.390 ***	0.049	0.057 ***	0.007
High school	0.199 ***	0.044	0.029 ***	0.007
Age	0.006	0.012	-0.002 ***	0.000
Age squared	0.000	0.000		
Married	-0.061	0.041	-0.009	0.006
Financial sector employee	0.662 ***	0.080	0.097 ***	0.011
Male	0.165 ***	0.037	0.024 ***	0.005
Austria	-0.991 ***	0.110	-0.146 ***	0.016
Belgium	0.195 **	0.084	0.029 **	0.011
Finland	-0.209	0.247	-0.030	0.021
France	0.307 ***	0.069	0.045 ***	0.007
Greece	-0.463 **	0.230	-0.067 ***	0.014
Italy	0.183	0.136	0.027	0.020
Luxembourg	-0.750 ***	0.117	-0.110 ***	0.016
Netherlands	0.063	0.095	0.009	0.012
Portugal	-0.380 ***	0.085	-0.056 ***	0.011
Spain	0.110	0.116	0.017 *	0.010
Fitted probabilities at p90-p10	0.113 ***	0.003		
Pseudo R squared		0.31		

Table 5: The effect of experienced stock market returns on stock market participation

Note: The table shows estimated coefficients of the probit model according to equation (3), explaining households' participation in stock markets, as well as average marginal effects (AME). For the effect of experienced returns ( $\beta$ ), the table also reports the average of the fitted probability at the 90th percentile minus the average fitted probability at the 10th percentile of the distribution of experienced returns.

	Coefficient	Std. error
Weighting parameter ( $\lambda$ )	7.558 ***	0.799
Experienced return (β)	3.772 ***	1.048
Log income	0.020	0.202
Log income squared	0.003	0.009
Number of children	-0.010	0.019
Number of children squared	0.003	0.006
Log liquid assets	0.232 ***	0.057
Log liquid assets squared	-0.002	0.003
Retired	-0.015	0.024
College	0.188 ***	0.026
High school	0.131 ***	0.024
Age	0.011 **	0.005
Age squared	0.000 ***	0.000
Married	-0.027	0.019
Financial sector employee	0.260 ***	0.034
Male	0.082 ***	0.018
Austria	-0.343 ***	0.051
Belgium	0.130 ***	0.042
Finland	0.278 ***	0.052
France	0.266 ***	0.031
Greece	0.064	0.069
Italy	0.083	0.071
Luxembourg	-0.251 ***	0.053
Netherlands	0.053	0.053
Portugal	-0.026	0.037
Spain	0.256 ***	0.048
Fitted value at p90-p10	0.045 ***	0.002
Pseudo R squared	0.28	

Table 6: The effect of experienced stock market returns on the share of stocks in liquid assets

Note: The table shows estimated coefficients of the Tobit model according to equation (3), explaining the share of stocks in liquid assets. For the effect of experienced returns ( $\beta$ ), the table also reports the average of the fitted probability at the 90th percentile minus the average fitted probability at the 10th percentile of the distribution of experienced returns.

		Experienced r	eturn (β)	Weighting para	ameter (λ)	Pseudo R-
		Coefficient	Std. error	Coefficient	Std. error	squared
(1)	Benchmark model	15.18 ***	3.10	5.24 ***	0.85	0.31
(2)	Explaining bond holdings with bond returns	27.78 *	14.92	3.99 ***	0.33	0.36
(3)	Adding experienced volatility	16.77 ***	2.91	5.01 ***	0.26	0.31
(4)	Stock holdings include voluntary pension plans	15.54 ***	2.67	5.11 ***	0.24	0.24
(5)	Unweighted estimation	4.72 ***	0.79	9.80 ***	0.73	0.34
(6)	Adding experienced bond returns	10.71 ***	2.35	6.03 ***	0.32	0.31
(7)	Longer experience horizon (10 years before birth)	21.03 ***	3.72	6.41 ***	0.42	0.31
(8)	Shorter experience horizon (10 years after birth)	10.50 ***	2.01	3.81 ***	0.48	0.31
(9)	Adding risk aversion	12.86 ***	2.82	5.84 ***	0.45	0.35
(10)	Adding real assets	15.23 ***	2.67	5.24 ***	0.21	0.35
(11)	Adding year-of-birth dummies	11.50 ***	3.19	4.02 ***	0.25	0.32
(12)	Excluding immigrants	6.74 ***	0.95	9.97 ***	0.90	0.33
(13)	Clustered errors by country	15.18 ***	4.22	5.24 ***	0.30	0.31
(14)	Placebo experiment	0.07	1.10	5.24	[fixed]	0.31
(15)	Countries with a less severe 2008 stock market crash	15.60 ***	3.63	5.47 ***	0.25	0.29
(16)	Countries with a very severe 2008 stock market crash	1.48	1.56	10.77 ***	1.13	0.34

Table 7: The effect of experienced stock market returns on stock market participation; extensions and robustness tests

Note: The table shows estimated coefficients  $\beta$  and  $\lambda$  of the probit model according to equation (3). Row (1) repeats the benchmark results, explaining households' participation in stock markets. Row (2) explains participation in bond markets as a function of experienced bond returns. Rows (3) to (13) explain participation in stock markets. Row (3) adds the experienced stock market volatility. Row (4) is based on a broader definition of stockholdings, also including investments in voluntary pension plans. Row (5) provides unweighted results. Row (6) additionally includes the bond returns that households have experienced over their lifetimes. Rows (7) and (8) vary the experience horizon of respondents, by either including 10 years prior to birth, or by starting 10 years after birth. Row (9) adds households' willingness to take financial risks as additional regressor. Row (10) adds the level of real asset holdings. Row (11) adds year-of-birth dummies. Row (12) excludes immigrants from the estimation. Row (13) clusters standard errors by country. Row (14) reports results from a placebo experiment. Rows (15) and (16) contain split sample estimates, once for countries with less-severe stock market crashes in 2008, and once for the very severely hit countries.

Table 8: The effect of stock market downturns on households' willingness to take financial risks

			AME		AME		AME		AME	
	Coefficient	Std. error	Willingness=1	Std. error	Willingness=2	Std. error	Willingness=3	Std. error	Willingness=4	Std. error
Number of experienced crashes ( $\beta_1$ )	-0.122 **	0.048	0.011	0.007	-0.007 *	0.004	-0.003	0.002	0.000	0.000
Number of experienced crashes squared ( $\beta_2$ )	0.009 **	0.003								
Log income	-0.185 ***	0.092	-0.024 ***	0.008	0.015 ***	0.005	0.008 ***	0.002	0.002 ***	0.001
Log income squared	0.012 **	0.005								
Number of children	-0.009	0.039	0.003	0.009	-0.002	0.006	-0.001	0.003	0.000	0.001
Number of children squared	-0.001	0.013								
Log liquid assets	-0.118 ***	0.015	-0.032 ***	0.003	0.019 ***	0.002	0.011 ***	0.001	0.003 ***	0.000
Log liquid assets squared	0.013 ***	0.001								
Retired	-0.107 **	0.043	0.033 **	0.013	-0.021 **	0.009	-0.010 **	0.004	-0.002 **	0.001
College	0.309 ***	0.043	-0.097 ***	0.014	0.062 ***	0.009	0.029 ***	0.004	0.006 ***	0.001
High school	0.189 ***	0.039	-0.059 ***	0.012	0.038 ***	0.008	0.018 ***	0.004	0.004 ***	0.001
Age	-0.005	0.008	0.003 ***	0.001	-0.002 ***	0.000	-0.001 ***	0.000	0.000 ***	0.000
Age squared	0.000	0.000								
Married	-0.086 **	0.034	0.027 **	0.011	-0.017 **	0.007	-0.008 **	0.003	-0.002 **	0.001
Financial sector employee	0.235 ***	0.056	-0.074 ***	0.018	0.047 ***	0.011	0.022 ***	0.005	0.005 ***	0.001
Male	0.241 ***	0.030	-0.076 ***	0.009	0.048 ***	0.006	0.023 ***	0.003	0.005 ***	0.001
Austria	0.018	0.077	-0.006	0.024	0.004	0.015	0.002	0.007	0.000	0.002
Belgium	-0.258 ***	0.049	0.081 ***	0.016	-0.051 ***	0.010	-0.024 ***	0.005	-0.005 ***	0.001
Greece	0.019	0.051	-0.006	0.016	0.004	0.010	0.002	0.005	0.000	0.001
Italy	0.778 ***	0.069	-0.244 ***	0.021	0.155 ***	0.013	0.074 ***	0.007	0.015 ***	0.002
Luxembourg	-0.508 ***	0.062	0.159 ***	0.020	-0.101 ***	0.012	-0.048 ***	0.006	-0.010 ***	0.002
Netherlands	-0.342 ***	0.071	0.107 ***	0.022	-0.068 ***	0.014	-0.032 ***	0.007	-0.007 ***	0.002
Portugal	-0.730 ***	0.108	0.229 ***	0.034	-0.145 ***	0.022	-0.069 ***	0.010	-0.015 ***	0.003
Spain	-0.439 ***	0.048	0.138 ***	0.015	-0.087 ***	0.010	-0.042 ***	0.005	-0.009 ***	0.001
Fitted probabilities at p90-p10										
Willingness=1 (low)	0.044 ***	0.003								
Willingness=2	-0.028 ***	0.002								
Willingness=3	-0.013 ***	0.001								
Willingness=4 (high)	-0.003 ***	0.000								
Cutoff point 1	-0.556	0.485								
Cutoff point 2	0.664	0.485								
Cutoff point 3	1.699	0.488								
Pseudo R squared	1.055	0.400			0.11					

Note: The table shows estimated coefficients of the ordered probit model according to equation (4), explaining households' willingness to take financial risks, as well as average marginal effects (AME) for each category of the ordered probit. For the effect of experienced crashes ( $\beta_1$  and  $\beta_2$ ), the table also reports the average of the fitted probability at the 90th percentile minus the average fitted probability at the 10th percentile of the distribution of experienced crashes, for each category of the ordered probit.

	Coefficient	Std. error	AME	Std. error
Number of experienced crashes ( $\beta_1$ )	-0.316 ***	0.071	-0.020 ***	0.004
Number of experienced crashes squared ( $\beta_2$ )	0.019 ***	0.005		
Log income	-0.183	0.259	0.020	0.005
Log income squared	0.015	0.012		
Number of children	0.002	0.043	0.000	0.005
Number of children squared	-0.002	0.012		
Log liquid assets	0.293 **	0.122	0.067 ***	0.002
Log liquid assets squared	0.008	0.006		
Retired	-0.091 *	0.055	-0.013	0.008
College	0.377 ***	0.047	0.056 ***	0.007
High school	0.187 ***	0.043	0.028 ***	0.006
Age	0.042 ***	0.009	0.001 ***	0.000
Age squared	0.000 ***	0.000		
Married	-0.065	0.041	-0.010	0.006
Financial sector employee	0.657 ***	0.080	0.097 ***	0.012
Male	0.174 ***	0.038	0.026 ***	0.006
Austria	-1.008 ***	0.124	-0.149 ***	0.019
Belgium	-0.046	0.069	-0.007	0.010
Finland	0.476 ***	0.049	0.070 ***	0.007
France	0.251 ***	0.047	0.037 ***	0.007
Greece	-0.395 ***	0.086	-0.058 ***	0.013
Italy	-0.308 ***	0.087	-0.045 ***	0.013
Luxembourg	-0.485 ***	0.083	-0.072 ***	0.013
Netherlands	-0.342 ***	0.105	-0.051 ***	0.016
Portugal	-0.139	0.140	-0.020	0.021
Spain	-0.012	0.064	-0.002	0.009
Fitted probabilities at p90-p10	-0.091 ***	0.001		
Pseudo R squared		0.31		

Table 9: The effect of experienced stock market downturns on stock market participation

Note: The table shows estimated coefficients of the probit model according to equation (4), explaining households' participation in stock markets, as well as average marginal effects (AME). For the effect of experienced crashes ( $\beta_1$  and  $\beta_2$ ), the table also reports the fitted probability at the 90th percentile minus the average fitted probability at the 10th percentile of the distribution of experienced returns.

	Coefficient	Std. error
Number of experienced crashes ( $\beta_1$ )	-0.034	0.029
Number of experienced crashes squared ( $\beta_2$ )	0.002	0.002
Log income	-0.203	0.153
Log income squared	0.009	0.007
Number of children	-0.011	0.016
Number of children squared	0.004	0.004
Log liquid assets	-0.172 ***	0.056
Log liquid assets squared	0.007 ***	0.003
Retired	0.044 **	0.021
College	0.026	0.020
High school	0.006	0.018
Age	0.006	0.004
Age squared	0.000	0.000
Married	0.002	0.014
Financial sector employee	0.013	0.020
Male	-0.003	0.015
Austria	0.006	0.061
Belgium	0.057 **	0.023
Finland	0.079 ***	0.019
France	0.091 ***	0.017
Greece	0.122 **	0.048
Italy	0.112 **	0.038
Luxembourg	0.097 **	0.039
Netherlands	-0.025	0.047
Portugal	0.030	0.072
Spain	0.081 ***	0.025
Fitted value at p90-p10	-0.012 ***	0.002
Pseudo R squared	0.14	

Table 10: The effect of experienced stock market downturns on the share of stocks in liquid assets

Note: The table shows estimated coefficients of the Tobit model according to equation (4), explaining the share of stocks in liquid assets. For the effect of experienced crashes ( $\beta_1$  and  $\beta_2$ ), the table also reports the average of the fitted probability at the 90th percentile minus the average fitted probability at the 10th percentile of the distribution of experienced returns.

		Coefficient	Std. error	Pseudo R
				squared
(1)	Benchmark model	-0.020 ***	0.004	0.31
(2)	Adding experienced stock returns	-0.011 **	0.004	0.31
(3)	Adding the number of experienced booms	-0.018 ***	0.004	0.31
(4)	Crashes defined as below -40% annual returns	-0.062 ***	0.012	0.31
(5)	Stock holdings include voluntary pension plans	-0.075 ***	0.006	0.24
(6)	Unweighted estimation	-0.003	0.002	0.34
(7)	Adding risk aversion	-0.014 ***	0.005	0.34
(8)	Adding real assets	-0.020 ***	0.004	0.31
(9)	Adding year-of-birth dummies	-0.013 ***	0.004	0.32
(10)	Excluding immigrants	-0.009	0.007	0.36
(11)	Clustered errors by country	-0.020 *	0.010	0.31
(12)	Placebo experiment	-0.000	0.004	0.31

Table 11: The effect of experienced stock market crashes on stock market participation; extensions and robustness tests

Note: The table shows estimated average marginal effects of the experienced stock market crashes on participation in stock markets, based on the probit model according to equation (4). Row (1) repeats the benchmark results. Row (2) adds the experienced stock market returns. Row (3) adds stock market booms. Row (4) changes the definition of a downturn to cases where annual nominal stock returns were below -40%. Row (5) includes voluntary pension plans in the definition of stockholdings. Row (6) provides unweighted results. Row (7) adds households' willingness to take financial risks. Row (8) adds the level of real asset holdings. Row (9) adds year-of-birth dummies. Row (10) reports results for an estimation that excludes immigrants. Row (11) clusters standard errors by country. Row (12) shows the results for a placebo experiment where the number of experienced crises has been assigned randomly across the distribution of households.