## Firm Uncertainty and Household Spending<sup>\*</sup>

Iván Alfaro<sup>†</sup> Hoonsuk Park<sup>‡</sup>

December 10 2019 First version: June 25 2019

### Abstract

We map rich micro-data from financial accounts of US households to employers listed in the US stock market. Using banking and credit card transaction data, we find that households adjust their spending in response to labor income uncertainty, as proxied by employer-specific option-implied volatility. Households reduce average monthly consumption growth by 1.28 percentage points in response to a one standard deviation increase in firm uncertainty. This negative second moment firm uncertainty effect is larger than the positive first moment effect of firm stock returns. The employer-specific effect is robust to both industry- and aggregate-level volatility effects. The intensity of the spending response increases in the forecast horizon window and lasts up to a year. It is more pronounced for low-income households and for households that work at firms that recently had low employee growth, high CAPM  $\beta$ , low return on assets, and low Tobin's Q. Household spending shows strong asymmetric response to 'good' and 'bad' uncertainty.

<sup>\*</sup>We thank Nick Bloom for helpful comments, our formal discussants Hyun-Soo Choi, Markku Kaustia, Vladimir Mukharlyamov, and Neil Pearson, and conference and seminar participants at the 2019 Australasian Finance & Banking Conference, BI Norwegian Business School, 2019 Conference on Asia-Pacific Financial Markets, Korea University, Nordic Finance Network, NTU 2019 Finance Conference, 2019 Stanford Institute for Theoretical Economics (SITE), Stockholm Business School, University of Melbourne. We thank Francisco Barbosa for exceptional research assistance and the financial company for generously providing the data set.

<sup>&</sup>lt;sup>†</sup>Department of Finance, BI Norwegian Business School, Nydalsveien 37, N-0484 Oslo, Norway. e-mail: ivan.alfaro@bi.no

<sup>&</sup>lt;sup>‡</sup>Division of Banking & Finance Nanyang Business School Nanyang Technological University S3-B1B-67, 50 Nanyang Avenue, Singapore 639798 e-mail: hoonsukpark@ntu.edu.sg

### 1 Introduction

There is an increasing body of literature addressing the question of whether fluctuations in uncertainty affect economic behavior (Bloom (2014) provides a thorough discussion). Uncertainty is a key component of buffer stock models of consumption of Deaton (1991) and Carroll (1997) and is a key driver of aggregate asset pricing models, such as Bansal and Yaron (2004) who model income uncertainty in a long-run risk framework. Despite the surge in interest in uncertainty after the Great Recession and the increased availability of data to proxy for uncertainty, micro-level evidence of household-level response to uncertainty remains largely undocumented.<sup>1</sup> This paper aims at closing this gap by using rich high-frequency banking and credit and debit card transaction data for thousands of US individuals.<sup>2</sup>

By mapping this household financial data to employers publicly listed in the US stock market (with Compustat, CRSP, and OptionMetrics data), we create a rich employeeemployer panel data to examine the micro-level response of households to *income-labour uncertainty*, as proxied by forward-looking employer-specific volatility. To the best of our knowledge, our paper is the first to do this.<sup>3</sup> Moreover, the large cross-section of publicly listed employers in our sample (780+ unique firms) allows us to classify households by recentyear firm characteristics (e.g., employee growth, stock return, Tobin's Q, investment rates, profitability, etc.,) and determine whether firm characteristics further matter in the response of households to uncertainty. Our findings show that not only do households pay attention to uncertainty confronted by their publicly listed employers but are also attentive to the fundamental characteristics of their employers relative to other firms.

The motivation for the response of consumption to uncertainty is a classical precautionary savings motive, in which risk-averse households adjust their consumption downward upon

<sup>&</sup>lt;sup>1</sup>primarily because of the lack of household-level data to measure both consumption and income sources <sup>2</sup>which allows us to see and classify consumption transactions from daily purchases, such as at Starbucks, groceries at Walmart, online at Amazon.com, etc., This type of data has only until recently been made more widely available due to the development of fintech and big data

 $<sup>^{3}</sup>$ by using firm option-implied volatility - an object largely exogenous for households- helps overcome concerns of endogeneity present in using households' own consumption volatility or income to measure uncertainty

an increase in uncertainty about their future income and consumption streams. As long as jumps in employer volatility capture increases in the likelihood of households observing potential negative shocks to their income streams (e.g., layoffs, Chapter 7 and 11 bankruptcy, increased doubts in receiving performance bonuses or option payments, etc.), rational households should respond negatively to employer-specific uncertainty.<sup>4</sup> Our regression analysis below provides strong support for this response.

One possible concern with our methodology is that households may not be aware of uncertainty shocks that affect their employers. For example, it might be unrealistic for low-wage employees at, say, Walmart to be fully aware of the changes in the option-implied volatility at their employer. We attempt to address this concern by showing robust results to realized stock return volatility and by splitting our sample of households by firm-characteristics observed by households, such as firm employment growth. Compared to option-implied volatility, the volatility of the firm's stock is much more visible to employees (albeit the two measures are highly correlated). Moreover, by classifying households by firm characteristics observable by employees such as recent employment growth (e.g., layoffs of colleagues vs hirings) or by firm performance in the stock market (e.g., losers vs winners), we attempt to tease out where most of the effect of uncertainty stems from.

Our main findings can be summarized as follows. First, household (employee) consumption responds negatively to firm (employer) uncertainty shocks. A two standard deviation shock to firm uncertainty decreases future consumption growth of households employed by that firm by 2.56 percentage points. This result is robust to controlling for the firms' stock return (as a first moment control to disentangle from the second moment effect of uncertainty), household indebtedness, and household income shocks. This effect is economically meaningful and comparable to the size of the effect attributed to a similar lagged negative

<sup>&</sup>lt;sup>4</sup>Atkeson, Eisfeldt, and Weill (2017) theoretically argue that firm equity volatility is closely tied to the distance to insolvency and distance to default of firms. Empirically they use firm volatility to identify the degree of financial distress of firms in meeting their financial obligations. Moreover, in real business cycle models of uncertainty with capital adjustment costs (e.g., Alfaro, Bloom, and Lin (2017)), it's optimal for firms to cut investment and employment due to real option effects

shock to the income of the household. Moreover, for comparison, an aggregate drop in US consumption growth by 2 points is massive. Moreover, we find an offsetting positive effect in firm stock returns, but ranges in magnitude between 1/4 to 1/2 of the effect attributed to uncertainty shocks. These results are robust to different measures of uncertainty (e.g., option-implied vs. realized) and a battery of different regression specifications (e.g., multiple dimensions of fixed effects and clustering of standard errors).

Second, we find that the consumption response to uncertainty shocks is robust in the forecast horizon window and lasts up to 12 months. The effect grows in economic magnitude from short horizons of one and three months and peaks at nine months. Our findings suggest that the adjustment behavior of households is not the strongest immediately following an uncertainty shock but rather builds over time. Moreover, our estimates at different horizons are economically larger when using forward-looking volatility measures from options than realized volatility.

Third, we document strong asymmetric responses of household spending to 'good' and 'bad' uncertainty innovations. In particular, household spending responds positively to 'good' uncertainty, but negatively and more strongly to 'bad' uncertainty. These results provide micro-level support for the aggregate-level evidence of good and bad uncertainty in Segal, Shaliastovich, and Yaron (2015). Fourth, we examine the sensitivity of household spending to industry- and aggregate-level uncertainty and stock returns. We find that a one standard deviation increase in industry uncertainty innovations results in a 1.43 percentage point decrease in household consumption growth, while an increase in industry-level stock returns increases household consumption growth by 1.02 points. However, when we include the employer-specific volatility shock and return, we find that the effect of industry uncertainty shocks is no longer significant, while employer-specific uncertainty remains so. This suggests that the household idiosyncratic labour-income uncertainty channel is strong above and beyond industry uncertainty. Moreover, we also find that aggregate uncertainty from the VIX (i.e., volatility implied by S&P 500 index options) does not have a negative and significant effect on household monthly consumption growth, yet household monthly spending is positively related to the S&P 500 return.

Fifth, by classifying consumption expenditures by categories, we find that retail purchases exhibit the strongest response relative to groceries and restaurants. Moreover, when we look at transactions strictly categorized as durable consumption related to automobile and home improvements, we find a robust negative response that is also stronger than that of groceries and restaurants (yet weaker than broad retail which largely is a mix of durables and nondurables). Sixth, when we split our sample of households by groups according to incomelevels, we find that low-income households respond more intensively than high-income, where for durables, the response from low-income is 7 times as large.

Seventh, we document novel dynamics in the intensity of responses across households classified by the characteristics of their employers. In particular, using common company fundamental and financial data from Compustat we classify households by the characteristics of firms in the preceding year. This allows us to examine whether households that work for firms that recently experienced, say, low employment growth (e.g., firms with layoffs) respond differently to uncertainty than households whose employers experienced recent high employment growth (e.g., hiring expansions). We document that the intensity is highly pronounced amongst households that work for firms with recent low employment growth, high intangible investment (arguably seen by employees as undertaking risky projects), low investment opportunities (as proxied by Tobin's Q), high covariance with the market portfolio return (i.e., risky high CAPM beta firms), and low stock returns (i.e., past "loser" firms as in the momentum literature in asset pricing).

Lastly, we perform falsification tests where we randomly map households to false placebo firms also listed in the stock market. Using the placebo firms' volatility shocks instead of the true employers' volatility (in 50 multiple iterations), we find no response of household consumption to placebo employer uncertainty. This largely validates that our findings in the paper arise from the novel link we construct in our data between firms' option-implied volatility and household consumption.

To our knowledge, our paper is the first to examine the household consumption response to firm uncertainty shocks using both detailed financial administrative data to measure consumption and market-driven firm volatility to proxy for labor-income uncertainty. Whereas the literature largely utilizes the household's own subjective expectations of future outcomes to proxy for uncertainty (Dominitz and Manski (1997), Guiso, Jappelli, and Pistaferri (2002), Jappelli and Pistaferri (2000)), none of the variables used in our paper are reported by the household. The consumption data is as how the financial institutions record the transactions on their books, and the uncertainty shocks are firm-specific shocks largely driven by covariances with aggregate variables and not endogenously influenced by the household employees in our data. That is, in contrast to, say, measuring income uncertainty based on the time-series standard deviation of household income (largely an endogenous choice for the household), we rely on a largely exogenous object to the household when measuring income uncertainty: the employers' option-implied volatility.

The closest related paper using similar financial administrative data to examine consumer spending is Baker (2018). In a similar setting linking employers listed in the stock market to employees, the paper examines the effect of household income shocks on spending. Moreover, household debt levels are shown to influence the spending response to income shocks further. We differ in that we control for income shocks (a fist moment effect) and focus on the uncertainty surrounding future income shocks as proxied by firm volatility shocks (a second moment effect). We show that forward-looking employer-specific volatility shocks have an economically large effect on future household spending above and beyond both household income shocks (both current and lagged) and employer stock returns. Therefore, our paper complements Baker (2018) by focusing on the role of second moment income uncertainty rather than 1st moment realized income shocks.

Another related paper is Fagereng, Guiso, and Pistaferri (2017), who using Norwegian population data, also match consumers with their employers to test a precautionary savings

motive in consumption. However, in contrast to our high-frequency transaction data used in measuring and classifying consumption, Fagereng et al. (2017) do not observe consumption directly but rather infer it from the annual-frequency budget constraint of households (so called "imputed" consumption). As Baker, Kueng, Meyer, and Pagel (2018) argue discrepancies can occur between imputed and actual spending between two annual snapshots. Another related paper is Ben-David, Elyas, Kuhnen, and Li (2018), who using household-level survey data find that households with more uncertain expectations about the future indicate their *intention* to reduce their future consumption, which is consistent with our results. However, we connect households to firms, and instead of relying on expectations about future consumption behavior, we measure it directly from realized transactions. Moreover, Knotek and Kahn (2011) and Fulford (2015) find that uncertainty does not have an important role in influencing household consumption. Our paper differs from these papers in that we are testing the consumption response at the household level instead of at the aggregate level, as is in Knotek and Kahn (2011), and that we are able to track the consumption response to firm uncertainty shocks using administrative data instead of survey data as is in Fulford (2015). Another related paper is Agarwal, Aslan, Huang, and Ren (2019) who find that households reduce their stock market participation after shocks to political uncertainty. We differ in that our uncertainty measure is not aggregate and that we explore consumption behavior and how it differs across employer characteristics.

The paper proceeds as follows. Section 2 presents the data and empirical methodology, section 3 the results, and section 4 concludes.

### 2 Data and Empirical Methodology

The household banking and credit card transaction data comes from an online account aggregator. This online service helps households manage their budgeting, bill payments, savings, and investments in a convenient fashion. Households provide their login information of the various banks and credit card services that they are using to the website, and in turn, the website retrieves the information from each financial institution for the household. The data used in this paper is the same as Baugh, Ben-David, and Park (2018). Recent papers that use similar data include Baker (2018), who provides an extensive overview of the characteristics of this type of data.

The data contains the details of daily transactions for approximately 2.7 million households from June 2010 to May 2015. For each transaction, we are able to observe the date, the amount, whether the transaction was an inflow or outflow, the categories provided by the online aggregator, and the transaction description. It is similar to looking at a bank or credit card statement. Since we are able to observe bank transactions, we observe income that comes into the household's bank account from its employers.

For many of these income transactions, we can identify the names of the employers, which allows us to link the household to both private and public firms. This study focuses on the link from households to publicly listed firms in the US stock market, for which we exploit forward-looking option-implied volatility of firms to proxy for labor-income uncertainty. We use a fuzzy matching algorithm to match the employer names of the household data to the company names on Compustat. Table 1 and Figure 1 show the mapping. In the first row of Table 1, we show the number of households that are matched to Compust throughout our sample period. In total, we can identify 90,307 households that we can link to Compustat firms. The universe of Compustat firms is larger than that of firms in CRSP and Option-Metrics, from which we use stock returns and option-implied volatilities, respectively. After dropping households with only limited daily transaction information and after merging the household employees to their employers listed in the US stock market (having Compustat, CRSP, and Optionmetrics data), we are left with 52,228 unique households and 784 unique publicly listed firms. This mapping comprises the sample used in our regression analysis. The reason that the number seems low compared to the 2.7 million households in the sample as a whole is that, for many households, the income description only contains the word

"payroll" or "direct deposit" and does not have any information on the employer. Other households work for private firms, non-profits, or the government, which we cannot link to Compustat. Finally some households do not link the income-receiving bank account to the online account aggregator.

Nonetheless, the resulting household number is large, data-rich, and largely representative of the US population. In Figure 2, we show the distribution of income in our sample, compared to the distribution of income in the 2010 US Census. The income in our sample is similar to the distribution in the US Census, but our measure of income is after withholdings such as tax and contributions. In that context the income for our sample should be considered to be larger than what is shown on the figure, which in turn further helps our data resemble the US Census. Moreover, the matched public firms in our sample are not restricted to small firms; rather they show a nice distribution in characteristics. In Figures 3, 4, and 5, we show how the matched firms are distributed along market equity, number of employees, and book-to-market equity ratios. Our sample includes firms that are large and small, as seen in terms by market capitalization and number of employees. We also have a good distribution of both growth and value firms, according to book-to-market equity ratios.

In Table 2, we provide the summary statistics of the variables used in this paper. We define our baseline household consumption variable using expenditures at retailers, restaurants, and grocery stores. We observe potentially multiple of these transactions per household every day. We aggregate the US\$ dollar consumption transactions to the monthly level every month for each household. In identifying the consumption transactions, we use the transactions that we can identify at major retailers and grocery stores from a list of the top 100 retailers during the sample period.<sup>5</sup> For restaurants, we also use an equivalent list for the top 100 restaurants.<sup>6</sup> We augment this list by searching for relevant keywords such as burger, taco, pizza, grill, steak, and etc. These types of transactions comprise our main measure of consumption in the paper. We prefer this measure of consumption because it is

<sup>&</sup>lt;sup>5</sup>http://www.stores.org/2012/Top-100-Retailers

<sup>&</sup>lt;sup>6</sup>http://nrn.com/us-top-100/top-100-chains-us-sales

based on a clean set of transactions that are likely not misclassified. The average monthly consumption is \$966. For our measure of durable consumption, we use the categorization provided by the online account aggregator, which includes categories for automobile-related expenditures, home improvement, and home maintenance. We find similar results when we use an alternative measure of durables based on keyword searches in transactions at top 10 hardware stores (which include Lowe's, Home Depot, etc.,). Our measures of durables can further be expanded to include items such as clothing, toys, etc.<sup>7</sup>, which only strengthens the results in our baseline durable measure. Therefore, our baseline results for durables are likely a lower bound.

The reason that we use the keyword search measure of consumption as the main result is that the data provider's classification tends to contain many misclassified transactions. For example, we find many transactions that are credit card payments being classified into retail payments because the credit card was issued by the retailers. In contrast, using the keyword searches, we can choose to include the transactions that are very likely to be the consumption transactions that we are interested in. One advantage of using the data provider's classification, however, is that we can classify more of the household's transactions as consumption. There is a trade-off between accurately observing a smaller subset of the household's consumption and inaccurately observing a larger subset of the household's consumption. In this paper, we choose the former of accurately observing a smaller subset, those we make use of the data provider's classification in looking at the household's durable consumption.

Our baseline measure of uncertainty uses the option-implied volatility of firms from OptionMetrics. In particular, our measure of implied volatility of firms follows Alfaro et al. (2017) and is measured as the 252-trading-day average of daily implied volatility values from at-the-money 365-day forward call options, from OptionMetrics. Moreover, we also measure uncertainty using realized stock return volatility from CRSP, where realized volatility is the annualized standard deviation of daily CRSP cum-dividend stock returns within a 365-day

 $<sup>^{7}</sup>$ which are clearly more durable items than groceries and restaurant transactions, but perhaps less so than strict hardware items

window.<sup>8</sup> As shown below, we document robust results to either measure, but stronger using implied volatility. We find similar results when using option-implied volatilities from at-the-money 91-day forward call options.

As controls, we include the firm's cumulative 6 month stock returns (a 1st moment control to disentangle from 2nd moment uncertainty effects), household income shocks (both current and lagged), the households' mortgage payment-to-income ratio (that accounts for debt effects in households' budget constraint), and a local cost-of-living measure that control for local economic shocks unrelated to the shocks affecting the firm. This cost-of-living measure is calculated following Baugh et al. (2018) by computing the mean expenditures of gas, restaurant, groceries, and retail for each city, for every month from the unmatched 2.7 million households in our underlying household data.

In our regressions, most variables are measured in terms of growth rates. For the growth, we follow Davis and Haltiwanger (1992), where for any variable  $x_t$ , the growth is calculated as  $\Delta x_t = (x_t - x_{t-1})/(\frac{1}{2}x_t + \frac{1}{2}x_{t-1})$ . This growth measure has the nice feature of being bounded between -2 and 2 for positive values of x (such as volatility and US\$ dollar consumption values). The variables that do not use this measure are the CRSP stock return, the continuous mortgage-to-income ratios, and the levels in firm volatility. Table 2 shows summary statistics for the variables in our regression sample. All regression variables are winsorized at the 1 and 99 percentiles every month.

Our main regression specifications test whether an increase in the option-implied uncertainty of the firm for which a household works for is associated with future downward adjustments in household consumption. Given that households may take some time in gradually adjusting their monthly spending after rises in uncertainty, our baseline regressions are forecast of changes in average monthly consumption from a 6-months period to the next 6-months. However, we show below that the results are robust to decreasing or increasing the window length in measuring changes in average monthly consumption. Therefore, our

 $<sup>^{8}\</sup>mathrm{to}$  annualize we multiply the realized volatility by the square root of 252 (average number of trading days in a year)

main regression specification is as follows:

$$\Delta \text{Consumption}_{i,t} = \beta_0 + \beta_1 \times \Delta \text{Volatility}_{i,i,t-6} + \beta_2 \times 6\text{M Return}_{j,i,t-6}$$

$$\begin{split} &+ \beta_3 \times \text{Mortgage-Income}_{i,t-6} + \beta_4 \times \Delta \text{Income}_{i,t} + \beta_5 \times \Delta \text{Lag Income}_{i,t-6} \\ &+ \text{Cost of Living Index}_{c,i,t} + \alpha_i + \gamma_j + \delta_t + \epsilon_{i,t} \end{split}$$

This regression examines the forecasting effect of firm (employer) uncertainty shocks on future household (employee) consumption growth. The frequency of all variables is monthly.  $\Delta$ Consumption<sub>*i*,*t*</sub> is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household *i* level (employees). For each household, we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into the next 6-months. Our main uncertainty variable, referred to as uncertainty shocks,  $\Delta$ Volatility<sub>*j*,*i*,*t*-6</sub> is the 6-month growth in the option-implied volatility (365-day horizon from OptionMetrics) of the corresponding employer *j* of each household. The timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome.

Moreover, to disentangle between the predictive effect of second moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>*j*,*i*,*t*-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where Mortgage-Income<sub>*i*,*t*-6</sub> is the mortgageto-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>*i*,*t*</sub> and  $\Delta$ Income<sub>*i*,*t*-6</sub>, respectively.  $\alpha_i$ ,  $\gamma_j$ ,  $\delta_t$  are household, firm, and time fixed effects, respectively. Moreover, to account for the effect of cost-of-living differences, all specifications include the time-varying Cost of Living Index at the city level for households. Standard errors are clustered at the employer level. However, we show robustness to double clustering at the employer and industry level and time.

Table 3 presents the results, which we describe below, where the continuous independent

variables are standardized to make coefficients comparable across regressor variables. The results are fully robust to using unstandardized regressors. The coefficients estimates and standard errors are scaled ( $\times 100$ ) so that they imply a percentage point change in the household's consumption growth given a one standard deviation shock in the independent variable.

### **3** Results

### **3.1** Uncertainty and Consumption

Table 3 presents our main results for spending on retail, restaurants, and groceries. In column (1), we find that households reduce consumption growth by 1.49 percentage points in response to a standard deviation shock in firm employer uncertainty (significant at the 1%), as measured by option-implied volatility shocks. Equivalently, the response amounts to a 2.98% drop in consumption growth given a two standard deviation firm uncertainty shock. Column (2) adds the firms' stock return as control variable to disentangle between second moment uncertainty and first moment effects. The household consumption response to uncertainty shocks remains negative and similar in magnitude (-1.33% coefficient), yet the direction of the response to the stock return of the firm is positive - which is consistent with the common theoretical prediction that households would reduce consumption in response to increased income uncertainty, while they would increase consumption in response to increased future expected income.

Looking at the consumption response to the firm's stock returns, we find that a one standard deviation increase in the stock return of the employer results in a 0.51 percentage point increase in household consumption growth. What is perhaps surprising is that the second moment effect of uncertainty shocks on consumption is more than twice as large as the first moment effect of stock returns on consumption. It seems that households are, indeed, risk-averse and care more about uncertainty in their future income than the positive performance gains as captured by the employer's returns.

In columns (3) to (5) we add additional controls, including the households' mortgageto-income ratio and income shocks of households, which a priori could correlate with firm uncertainty shocks and subsume its effect. Differences in debt effects across households are controlled by the mortgage-to-income ratio. Our baseline specification with full set of controls is in column (5), where we find that households reduce consumption growth by 0.95 percentage points when the mortgage-to-income ratio increases by one standard deviation, and reduce consumption growth by 5.8 percentage points in response to a standard deviation decrease in the households' current income growth and reduce consumption growth by 2 percentage points in response to a standard deviation decrease in lagged income growth. Even after controlling for current and past changes in income and debt, the effect of lagged uncertainty shocks remains significant at the 1% level.

In all, Table 3 shows that the effect of uncertainty on household consumption is significant and large in magnitude, e.g., much larger than the first moment effect of the firm's stock returns and comparable with the direct impact to lagged household income. For comparison, an aggregate drop in US consumption growth of 2 percentage points is massive. Our microlevel evidence suggests that a one standard deviation shock to uncertainty in column 5 translates to a 1.28 point drop in household-level consumption growth. Thus, the effect is economically meaningful. Moreover, an increase in uncertainty combined with a decrease in returns (e.g., double negative shock as in the financial crisis) further combine to negatively affect household level (from the coefficients in columns (5)) given a two standard deviation increase to uncertainty and returns.

In Table 4, we implement a battery of robustness tests to see if our results hold under different specifications. Panel A on the left uses option-implied volatility from OptionMetrics as in Table 3, and panel B on the right uses realized volatility from CRSP. Column (1) in Table 4 replicates the baseline regression with the full set of controls in column (5) of Table 3. In column (2), we cluster the standard errors by time as well as by firm and find that the results remain almost identical. In column (3), we use industry fixed effects instead of firm fixed effects, and again results are similar. In column (4), we cluster by household and time instead of firm and time and find the results to be much more strongly significant than in the baseline specification. In column (5), we cluster standard errors by industry (3-digit Standard Industry Classification codes) and time, which is a strong test accounting for error clustering at a high industry dimension, and find that uncertainty remains significant at the 5% level.

In Panel B, on the right of Table 4, we use the change in realized volatility of the firm's stock returns as a proxy for the shocks to the firm's uncertainty. This is to address some concerns in the uncertainty literature, whether option-implied volatility is a better measure of uncertainty than realized volatility. Our results on household consumption are robust regardless of how uncertainty is measured, yet the coefficients are smaller in magnitude when using realized volatility (e.g., column (1) vs. (1A) coefficients of -1.28 and -0.968 both significant at the 1% level, respectively). The smaller coefficient results for realized uncertainty relative to implied uncertainty is consistent with the findings in Alfaro et al. (2017) on the causal effect of uncertainty shocks on firm investment and capital structure outcomes. Table 4 shows that uncertainty effects on household consumption are robust across all ten specifications explored in the Table.

One concern could be whether it matters if we measure the effect of uncertainty in shocks or in levels. Another question is if uncertainty only matters when measured in lags and not contemporaneously to consumption growth. We address both questions in Table 5, where we look at the effects of the levels in uncertainty on household consumption growth. That is, instead of looking at the shocks to uncertainty as in the results so far, we examine the effect of cross-sectional high and low levels of uncertainty on household consumption. We find that levels of uncertainty are just as important as the shocks to uncertainty, and in fact are even stronger than those documented in Tables 3 and 4. For instance, in column (2) of Table 5, we find that a one standard deviation increase in the lagged 6-month level of uncertainty faced by firms leads to households reducing their consumption growth by 1.58 points. When looking at concurrent levels of uncertainty in column (4), we find that a one standard deviation increase in the contemporaneous uncertainty level leads to a reduction in consumption growth by 2.91 points. Using realized volatility in columns (1A), (2A), (3A), and (4A), gives similar inferences as using implied volatility, yet smaller in magnitude.

## 3.2 Uncertainty and the horizon of Consumption Growth Forecasts

In Table 6, we adjust the horizons in the forecast of household consumption growth predicted by lagged firm uncertainty shocks. We do so to examine whether the precautionary savings motive effect of uncertainty kicks-in at shorter horizons and whether it's more pronounced at longer horizons. To examine the different time horizons, we adjust the time intervals in which the consumption and uncertainty are measured. For example, for the one-month intervals, we measure the change in consumption over one month and measure the change in uncertainty over the preceding month.

We find that at the high frequency of one-month intervals in consumption growth, the effect of uncertainty on consumption is much smaller at -0.16 points and only significant at the 10% level (in column 1). However, from a 3-month interval onward, the results become more pronounced, and the effects grow monotonically at longer horizons up to 9 months. At a full 1-year ahead in column (5), we find that a one standard deviation increase in firm volatility leads to a -1.11 point drop in household consumption growth. From the results in Table 6 we document that firm employer uncertainty appears to have an increasingly long-lived impact on household consumption, lasting up to a year. Moreover, results in Table 6 also show that the offsetting first moment effects of stock returns have a reversed U-shaped pattern over time on consumption growth, peaking at six months. It seems to take about three months for households to start responding to the uncertainty shocks in a significant

way, and the effects of uncertainty are long-lasting to at least twelve months. Using realized volatility in columns (1A), (2A), (3A), (4A), and (5A), gives similar inferences as using implied volatility, yet smaller in magnitude.

We plot the results from Table 6 in Figure 7, which show the increase in the point estimates at longer horizons for both uncertainty and the U-shaped pattern for first moment stock return effects. The red line represents the consumption response to the employer's stock returns, and the blue line represents the consumption response to uncertainty shocks. The confidence intervals are also shown as vertical lines.

### **3.3** Good and bad uncertainty

Results to this point document a strong negative and average effect of employer-specific uncertainty shocks on household spending. However, one might ask whether the effects of downside and upside uncertainty are symmetric? Segal et al. (2015) decompose aggregate uncertainty into 'good' and 'bad' uncertainty associated with positive and negative innovations to macroeconomic growth, and document that good uncertainty predicts an increase in future economic activity, such as consumption, output, and investment, while bad uncertainty forecasts a decline in economic growth and depresses asset prices. We use our rich micro-level data to test whether households respond asymmetrically to good and bad uncertainty in this section.

In particular, the direction of the growth in employer-specific uncertainty,  $\Delta Volatility$ , can be naturally separated into 'bad' ( $\Delta$ Volatility> 0) and 'good' ( $\Delta$ Volatility<= 0) uncertainty shocks. In Table 7, we examine whether households respond asymmetrically to uncertainty shocks by interacting the absolute value of the employer uncertainty shock,  $|\Delta$ Volatility|, with an employer indicator that takes value  $D^{Bad} = 1$  if the shock is bad,  $\Delta$ Volatility> 0, and zero otherwise.

For comparison of the size of the effects, column (1) present the baseline results for the average effect of uncertainty shocks on 6-month ahead household spending growth (i.e., column (5) Table 3). Column (2) shows the results of the interaction terms that test for asymmetry in the response to good and bad uncertainty shocks. The coefficient on the absolute value of the uncertainty shock,  $|\Delta \text{Volatility}|$ , captures the effect of 'good' uncertainty, while the coefficient on the interaction term,  $|\Delta \text{Volatility}| \times D^{Bad}$ , captures the difference between 'good' and 'bad' uncertainty effects. The sum of the coefficients captures the total effect of 'bad' uncertainty.

We find strong asymmetric sensitivity of households to good and bad uncertainty innovations. Using option-implied volatility in column (2) indicates that there is a difference in the effect between 'good' and 'bad' uncertainty of -2.27 percentage points on household monthly consumption growth in response to a standard deviation increase in innovations to uncertainty. We find similar results when looking at 9 month horizon forecasts in column (4), and for realized volatility. Interestingly, for realized volatility we find a positive and statistically significant consumption response to 'good' uncertainty (columns (2A) and (4A)) that is not present implied volatility. The response to 'bad' uncertainty shocks is negative and significant across all specifications in Table 7. Therefore, using our rich-micro data we document strong asymmetric responses to 'good' and 'bad' uncertainty shocks, consistent with the aggregate results in Segal et al. (2015) - however, we emphasize the channel of labor-income uncertainty as the mechanism at work (e.g., as in workhorse models that focus on income uncertainty as in Bansal and Yaron (2004)).

### 3.4 Industry and aggregate uncertainty

In Table 8, we examine if households adjust their consumption growth in response to industry and aggregate uncertainty shocks. Moreover, we test whether the strong negative effects of employer-specific uncertainty shocks remain significant after controlling for industryand aggregate-level uncertainty shocks and stock returns. As in the previous tables, column (1) shows the result for the baseline regression (i.e., column 5 Table 3). In columns (2) and (3), we show the effect of industry-level uncertainty shocks and stock return, measured as the within 3SIC monthly cross-sectional means of firm volatility shocks and 6-month compounded stock return of firms in the same industry of the employer of the household. In column (2), we find that a one standard deviation increase in industry uncertainty innovations results in a 1.43 percentage point decrease in household consumption growth, while an increase in industry-level stock returns increases household consumption growth by 1.02 points. However, in column (3), we include the employer-specific volatility shock and return, and find that the effect of industry uncertainty shocks is no longer significant, while the industry-level return remains highly significant. Employer-specific uncertainty shocks remain significant, suggesting that the household idiosyncratic labour-income uncertainty channel is strong above and beyond industry uncertainty.

In columns (4) through (7), we look at the effects of aggregate uncertainty and stock market return on household consumption growth. For aggregate uncertainty we use the changes (columns 4-5) and levels (6-7) of the VIX (i.e., a measure of volatility implied by S&P 500 index options), and the 6-month compounded return of the S&P 500 index (columns 4-7) to measure for first moment stock market effects. To avoid collinearity between the S&P 500 aggregate variables and month-year fixed effects, columns (4-7) include quarter-year fixed effects. We find that the VIX uncertainty does not have negative or significant effects on household consumption growth, yet household spending is positively and strongly related to the S&P 500 return. The effect of the employer-specific uncertainty shock remains significant on household spending regardless of controlling for the return and uncertainty related to aggregate stock market.

# 3.5 Uncertainty and Retail, Restaurant, and Grocery Consumption

The results so far are consistent with a precautionary savings motive that induces riskaverse households to cut down consumption in response to uncertainty. However, the consumption response might differ in intensity depending on the characteristics of purchases, such as the durability of purchased goods. For example, when households are facing uncertainty shocks, they may find it easier to reduce spending or delay spending for large ticket items and durable goods such as electronics, while they may not be able to reduce the money spent on buying groceries. In Table 7, we look at the three consumption categories that comprise our baseline measure of household consumption: consumption at retail, restaurant, and groceries. While spending at restaurant and grocery stores tend to be non-durable, spending at retailers are more likely to be items that are durable in nature, such as electronics, clothing, and other household items. We examine durable consumption using a different definition in the next section, but here we find that the consumption response to uncertainty tends to be in items that are more durable in nature.

For implied volatility, we find that a one standard deviation increase in firm volatility leads to a -1.32 percentage point decrease in retail spending. The decrease in spending at restaurants and groceries are much smaller, by -0.381 and -0.430, respectively. Furthermore, the effect of uncertainty on spending on restaurants and groceries are only significant at the 10% level. For sensitivity to the firm's stock returns, we find that only retail spending is statistically significant at the 10% level, where households increased spending growth in retail by 0.33 percentage points. For realized volatility, we find no significant change in restaurant spending in response to changes in firm volatility, while households reduced grocery spending in response to changes in firm volatility, which was significant at the 5% level. Specifically, we find that households reduce consumption growth in groceries by 0.50% in response to a one standard deviation increase in firm uncertainty. For other variables, we find the results to be similar to those found when measuring uncertainty using implied volatility.

### 3.6 Durable Consumption

In Table 8, we further examine the response of durable goods to uncertainty shocks using a different measure of durable consumption. In the previous table, we used a measure of consumption that was generated by using keyword searches of retailers, restaurants, and groceries. Though retailers usually sell goods that are more durable in nature compared to restaurants and grocery stores, the measure is different from what is traditionally considered to be durable goods in the literature. In this table, the measure of durable consumption is generated using the data provider's categorization of the household's spending transactions. We include the categories classified as automobile expenses, home improvement, and home maintenance by the data provider. This measure of durable consumption is more similar to the durable consumption often found in the literature, which includes consumption related to automobiles.

We document that durable consumption is responsive to uncertainty shocks. A one standard deviation uncertainty shock forecasts a 0.667 percentage point drop in durable consumption growth. In comparison, a one standard deviation increase in the employers' stock return increases durable growth by 0.308 percentage points (column (3)), though the results for returns are not significant. These directional responses to second and first moment effects are in line with our baseline measure of consumption growth examined to this point in Tables 3 to 6. However, the economic magnitude of the average response of durable goods is smaller than that of the baseline consumption measure that includes groceries, restaurant, and retail (e.g., coefficient of -0.667 in column (3) of durables in Table 8 vs. -1.28 in column (5) in Table 3). We note, however, that the results for durables are stronger if we expand its definition to include retail purchases of clothing, toys, etc (which are more durable than groceries, but perhaps less so than automobile-related). In addition, the consumption variable is generated from the data provider's classification of spending transactions, which tend be noisier than the keyword search method. Thus, Table 8 results are likely a lower bound on the response of durable consumption to uncertainty shocks. In addition, results in Table 8 show a U-shaped pattern for volatility and a reversed U-shaped pattern for returns over time.

### 3.7 Placebo Tests

The results so far show a robust response of household spending to employer uncertainty shocks. Our findings suggest a strong precautionary savings response to labor-income uncertainty. However, despite the controls in our regression, there could still be concerns that our results are driven by other factors, such as regional consumption trends. To validate that our results are driven by a truly idiosyncratic income-labor uncertainty channel between employers and employees, we perform falsifications test in this section. In addition, these results also validate the matching of the households in our sample to their true employers listed in the stock market.

In this placebo test, we conduct an experiment where we replace the true employer of the household in our sample with a placebo employer. If households are perfectly insured against uncertainty shocks uniquely related to placebo employers we should not find any response of household spending to placebo firm's uncertainty shocks, - e.g., no reason why an employee of Microsoft (ticker MSF) would care about the firm-specific uncertainty of, say, fashion clothing firm Abercrombie & Fitch Co (ANF), shoe store The Foot Locker, Inc. (FL), or restaurant Chipotle Mexican Grill, Inc. (CMG). We conduct the experiment in Table 9, where we show the results from 50 iterations of random mapping of households to placebo firms. In columns (2), (4), (6), (8), and (10), we show the average coefficients and standard errors from 50 regressions based on random matches (with different seeds and with replacement from a pool of over 1,700 placebo firms with required data in our sample). For all the dependent variables of interest, we find that the placebo regression coefficients are not significant on average, and their economic magnitude is much smaller and is mostly in the opposite direction, which the exception of restaurant spending. Moreover, we also report the number of times from the 50 placebo regressions where we observed significant (at the 5%) negative coefficients on the placebo vol shock and at the same time, positive coefficients on the placebo stock return. This occurs zero times in our placebo regressions, which validates that not even by random chance did any of the placebo regressions give us the directions and significance obtained in our main regressions.

### 3.8 Intensity of Response Across Household Income Levels

So far, we have examined how households in our sample respond to uncertainty shocks. In the next two tables, we look at the cross-section of households to see how different households respond to uncertainty shocks. One variable of interest is household income. We may expect low-income households to respond differently to uncertainty shocks compared to high-income households. For example, if many low-income households work in temporary jobs, then they may be more exposed to corporate layoffs that follow uncertainty shocks. On the other hand, high-income households may have sufficient precautionary savings that allow them to be less affected by changes in uncertainty.

In Table 10, we classify households into terciles by their average income levels and examine the response to uncertainty shocks for each household sub-sample. The left panel is for the baseline measure of consumption examined in Table 3, which includes retail, restaurant, and groceries, and the right panel is for the durable consumption examined in Table 8, which includes expenditures on automobile-related expenses, home improvement, and home maintenance. We find that low-income households are more sensitive to uncertainty shocks compared to higher-income households. Low-income households reduce their consumption growth by 1.44 percentage points in response to a one standard deviation shock to uncertainty, whereas high-income households reduced their consumption growth by only 0.83 percentage points. For the first moment stock returns, we find that low-income households increase their consumption growth by 0.6 percentage points, while high-income households only increased their spending by 0.39 percentage points, which as also only significant at the 10% level.

The results for durable consumption is similar to the results for our main consumption. However, for durable spending, the response for low-income households is 7.3 times as large as the response to uncertainty shocks for high-income households. While low-income households reduced durable consumption growth by 0.95 percentage points in response to a one standard deviation shock to uncertainty, high-income households only reduced their growth by 0.13 percentage points, which was also statistically indifferent from zero. This is intuitive in that low-income household behave much like less wealthy individuals, who are highly reliant on their jobs to sustain their livelihoods, and thus much more responsive to income risk. In contrast, high-income households are much like wealthy individuals that may have other means and/or buffers (i.e., asset holdings) to mitigate potential negative shocks to their income streams.

### **3.9** Intensity of Response Across Firm Characteristics

In Table 11, we classify households into terciles based on the characteristics of the firm that employ them. In particular, using common company fundamental and financial data from Compustat, we classify households by the characteristics of firms in the preceding year. This allows us to examine whether households that work for firms that recently experienced, say, low employment growth (e.g., firms with layoffs) respond differently to uncertainty than households whose employers experienced recent high employment growth (e.g., hiring expansions). We look at 12 firm characteristics: (1) the change in the number of employees at the firm, (2) investment - defined as capital expenditures over lagged plant, property, and equipment, (3) return on assets, (4) Tobin's Q, (5) CAPM  $\beta$ , (6) sales, (7) past 12 month returns, (8) change in intangibles expenses, (9) the Whited-Wu financial constraints index, (10) the Sales-Age financial constraints index, (11) 12 month implied volatility, and finally (12) 12 month realized volatility.

We find that households that work in firms that previously had low employment growth are the ones most sensitive to firm uncertainty shocks. Indeed, it seems that households that work for firms that have seen layoffs become more attentive to uncertainty shocks than households working for firms with recent increased hiring rates. Households that worked at firms with low past employment growth reduced consumption growth by 2.27 percentage points, while households that worked at firms with high employment growth had consumption growth that was not statistically different from zero. Moreover, the same group of households with employers that have seen layoffs show sensitivity to the firm's stock returns. For stock returns, we also find that the middle group of households in column (2) also showed high sensitivity to stock returns, though they did not show statistically significant sensitivity to uncertainty shocks. Households that worked for firms with high past employment growth did not show statistically significant consumption sensitivity to either uncertainty shocks or stock returns. These results highlight potential avenues for rich heterogenous-agent models, in which agents respond differentially to employer characteristics, where, for some households the primary concern is a firm first moment shock rather a second moment shock, and vice versa.

This result also helps us address potential concerns about the mechanism in which the rise firm uncertainty translates into a reduction in household consumption. As argued previously, our results are not based on the assumption that households pay attention to the implied volatility of the firm in which they are employed. The finding that most of our results are based on the firms which experienced previous low employment growth shows that the change in firm employment is the channel in which households become concerned about future income risk and therefore reduce consumption as a consequence. Households that worked at firms with low employment growth are more likely to have experienced layoff risks and other threats to their income, which drive the reduction in consumption growth.

When we divide households based on the CAPM  $\beta$  of their employers, we find that the households that work for firms with high risk are the ones most sensitive to both uncertainty shocks and firm returns. Households that worked at high  $\beta$  firms reduced their consumption growth by 1.84 percentage points in response to a one standard deviation shock in uncertainty, while households that worked at low  $\beta$  firms did not show a statistically significant response to uncertainty shocks. The coefficient for these households was also very close to zero, at 0.008 percentage points. For investments, the firm's uncertainty shocks and stock returns appear to have a differential impact on household consumption growth. Households that work for firms that had high and mid-levels of investment, as measured by capital expenditures over lagged plant, property, and equipment, show a consumption response to uncertainty, while households that work for firms that had low and mid-levels of investment showed consumption growth sensitivity to stock returns. Households that worked at firms with high levels of investment reduced consumption growth by 0.64 percentage points in response to a one standard deviation shock to uncertainty, while households that worked at firms with low levels of investment increased consumption growth by 0.79 percentage points in response to a one standard deviation increase in stock returns. If we look at intangible investment (potentially seen as risky investment by employees), we see a clearer pattern on the response to uncertainty, where there is a monotonic increase in the response from low to high R&D firms. In terms of investment opportunities, households working for firms with low Tobin's Q are the most responsive to uncertainty shocks, while high Q firm employees mostly respond to firm returns.

When looking at size, we find that households that work for medium and large firms are sensitive to uncertainty shocks, while households that worked for smaller firms are more sensitive to the firm's past returns. When looking at the firm's past 12 month returns, we find that households that work at firms with poor stock market performance are most sensitive to uncertainty shocks, but not to firm returns. When looking at the financial constraints measures of Whited and Wu (2006) and Hadlock and Pierce (2010) - WW and Size&Age indexes-, we find that the households that work for firms that are moderately financially constrained are the most sensitive to uncertainty shocks. Finally, when splitting households by the level of uncertainty of their employers in the past year, we find the strongest response to uncertainty shocks for employees at the highest uncertainty levels, as expected.

### 4 Conclusion

We map rich microdata from linked financial accounts of US households to employers listed in the US stock market. We use this employer-employee panel, comprising 784 listed firms and 52,288 households over a 4.5-year period, to examine detailed household consumption responses to labor income uncertainty, as proxied by employer-specific option-implied volatility.

We document that households robustly reduce their spending in response to second moment firm uncertainty shocks above and beyond firm first moment effects. Our forwardlooking option-implied results are robust to using realized volatility from CRSP stock returns. With regard to timing, it takes about three months for the firm uncertainty shocks to influence household consumption dynamics, the impact is more pronounced at longer horizons and lasts up to twelve months in the future. The negative uncertainty effects on consumption are not subsumed by the positive first moment effect of employer stock returns nor by household-specific income shocks (both current and lagged). Durable consumption is also highly responsive to uncertainty shocks.

Moreover, we find differences in intensity across households classified by income-levels and by firm characteristics. Low-income employees are more responsive to employer uncertainty shocks than high-income. In addition, the intensity to uncertainty shocks is highly pronounced amongst households that work for firms with recent low employment growth, high intangible investment (firms arguably undertaking risky projects), low investment opportunities (as proxied by Tobin's Q), high covariance with the market portfolio return (i.e., risky high CAPM beta firms), and low performing firms in the stock market.

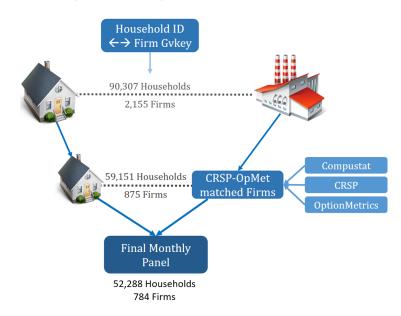
### References

- Agarwal, Vikas, Hadiye Aslan, Lixin Huang, and Hongling Ren, 2019, Political uncertainty and household stock market participation, *Working Paper*.
- Alfaro, Iván, Nicholas Bloom, and Xiaoji Lin, 2017, The finance uncertainty multiplier, *Working Paper*.
- Atkeson, Andrew G., Andrea L. Eisfeldt, and Pierre-Olivier Weill, 2017, Measuring the financial soundness of U.S. firms, 1926–2012, *Research in Economics* 71, 613–635.
- Baker, Scott, Lorenz Kueng, Steffen Meyer, and Michaela Pagel, 2018, Measurement error in imputed consumption, *Working Paper*.
- Baker, Scott R., 2018, Debt and the response to household income shocks: Validation and application of linked financial account data, *Journal of Political Economy* 126, 1504–1557.
- Bansal, Ravi, and Amir Yaron, 2004, Risks for the long run: A potential resolution of asset pricing puzzles, *The Journal of Finance* 59, 1481–1509.
- Baugh, Brian, Itzhak Ben-David, and Hoonsuk Park, 2018, Can taxes shape an industry? evidence from the implementation of the "amazon tax", *The Journal of Finance* 73, 1819–1855.
- Ben-David, Itzhak, Fermand Elyas, Camelia M. Kuhnen, and Geng Li, 2018, Expectations uncertainty and household economic behavior, *Working Paper*.
- Bloom, Nicholas, 2014, Fluctuations in uncertainty, *Journal of Economic Perspectives* 28, 153–176.
- Carroll, Christopher D., 1997, Buffer-stock saving and the life cycle/permanent income hypothesis, *The Quarterly Journal of Economics* 112, 1–55.
- Davis, Steven J., and John Haltiwanger, 1992, Gross job creation, gross job destruction, and employment reallocation, *The Quarterly Journal of Economics* 107, 819–863.
- Deaton, Angus, 1991, Saving and liquidity constraints, *Econometrica* 59, 1221–1248.
- Dominitz, Jeff, and Charles F. Manski, 1997, Using expectations data to study subjective income expectations, *Journal of the American Statistical Association* 92, 855–867.
- Fagereng, Andreas, Luigi Guiso, and Luigi Pistaferri, 2017, Firm-related risk and precautionary saving response, American Economic Review 107, 393–397.
- Fulford, Scott L, 2015, The surprisingly low importance of income uncertainty for precaution, European Economic Review 79, 151–171.
- Guiso, Luigi, Tullio Jappelli, and Luigi Pistaferri, 2002, An empirical analysis of earnings and employment risk, *Journal of Business & Economic Statistics* 20, 241–253.

- Hadlock, Charles J., and Joshua R. Pierce, 2010, New evidence on measuring financial constraints: Moving beyond the kz index, *The Review of Financial Studies* 23, 1909–1940.
- Jappelli, Tullio, and Luigi Pistaferri, 2000, Using subjective income expectations to test for excess sensitivity of consumption to predicted income growth, *European Economic Review* 44, 337–358.
- Knotek, Edward S., and Shujaat Kahn, 2011, How to households respond to uncertainty shocks?, *Federal Reserve Bank of Kansas City, Economic Review* 96, 5–34.
- Segal, Gill, Ivan Shaliastovich, and Amir Yaron, 2015, Good and bad uncertainty: Macroeconomic and financial market implications, *Journal of Financial Economics* 117, 369–397.
- Whited, Toni M., and Guojun Wu, 2006, Financial constraints risk, *The Review of Financial Studies* 19, 531–559.

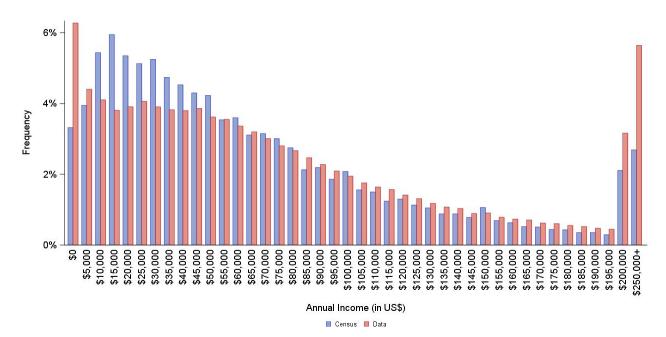
### Figure 1. Mapping of households to public firms

This figure shows the number of unique households (employees) in the online account aggregator data that are mapped to unique publicly listed firms (employers) having financial reports (Compustat), returns (CRSP), and option-implied volatilities (OptionMetrics). The resulting panel is after applying filters to our data.



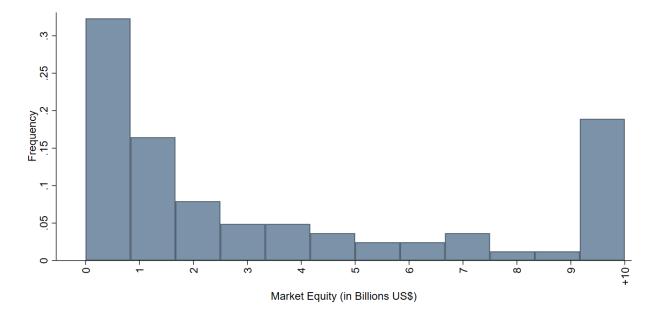
### Figure 2. Distribution of annual income

This figure compares the distribution of annual income for households in our sample (red) to the 2010 U.S. Census (blue). Note that income in our sample is after withholdings, such as income taxes, healthcare contributions, and retirement contributions. These omissions understate the actual household income, before withholdings. Nonetheless, our sample is largely representative of US household income.



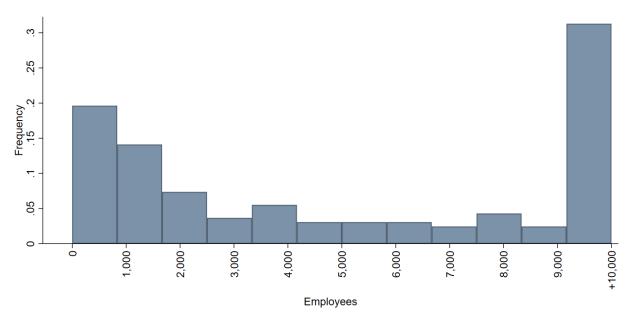
### Figure 3. Distribution of firm market capitalization

This figure shows the distribution of public firms in our regression sample according to their market equity (in Billions of US\$).



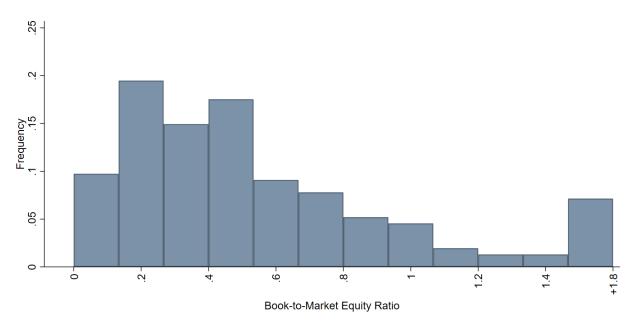
### Figure 4. Distribution of firm employees

This figure shows the distribution of public firms in our regression sample according to their number of employees.



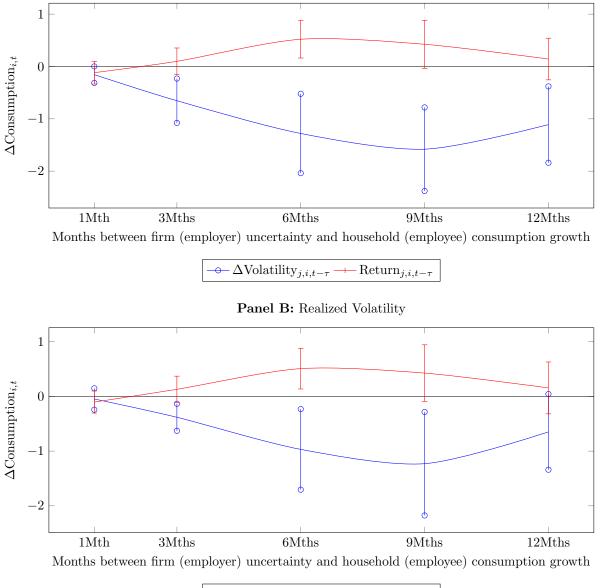
### Figure 5. Distribution of firm book-to-market equity

This figure shows the distribution of public firms in our regression sample according to their book-to-market equity ratios.



### Figure 6. Forecasting employee consumption with firm uncertainty

This figure shows the coefficients at different forecast horizons of regressing (employee) household consumption on firms' (employer) uncertainty shocks and stock return. The point estimates are from Table 6. Panel A shows the coefficients for implied volatility and Panel B the coefficients for realized volatility. The y-axis is in percentage points and the x-axis is forecast horizon in months. The negative effect of firm volatility on future consumption growth is in blue, while the positive offsetting effect of the firms' stock return is in red. The vertical lines above and below the coefficients represent 95% confidence intervals. The response of consumption to firm uncertainty is more intensive at longer horizons up to 9 months, and last up to 12 months.



Panel A: Implied Volatility

 $\rightarrow \Delta$ Volatility<sub>j,i,t-\tau</sub>  $\rightarrow Return_{j,i,t-\tau}$ 

### Table 1. Mapping of households to public firms

This table shows the number of unique households and firms that are matched each year to create our employee-employer panel data. Households and firms are matched based on a textual fuzzy matching algorithm that uses Compustat company names and the household income descriptions that identifies the employer's company name. Our baseline regression panel further uses firm data from CRSP and Option-Metrics for returns and implied volatilities, respectively. This gives a final mapping to 784 unique listed firms (employers) in our sample. We perform a manual inspection and filtering of resulting mapped firms based on the textual matching.

	Jun 2010	Dec 2011	Dec 2012	Dec 2013	Dec 2014	May 2015	Unique
Household ID $\leftrightarrow$ Firm Gvkey	59,029	84,927	86,328	84,364	79,969	70,565	90,307
CRSP - OpMet Firms Matched Households	$592 \\ 30,749$	$680 \\ 50,899$	$698 \\ 51,841$	710 50,708	$724 \\ 48,625$	$678 \\ 41,332$	$875 \\ 59,151$
Firm in Baseline Households in Baseline		$588 \\ 39,031$	$630 \\ 45,114$	$647 \\ 45,319$	$674 \\ 43,242$	$630 \\ 39,099$	784 52,288

### Table 2. Summary statistics

This table shows the summary statistics of the variables used in the main regression analysis. Frequency of all variables is monthly.  $\Delta \text{Consumption}_{i,t}$  is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household *i* level (employees in our sample). For each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months. Similarly,  $\Delta \text{Durables}_{i,t}$  is the 6-month growth in durable consumption.  $\Delta \text{Volatility}_{j,i,t}$  is the 6-month growth in the option-implied volatility (365-day horizon from OptionMetrics) of employers *j* in our sample.  $\Delta \text{Realized Volatility}_{j,i,t}$  is the 6-month growth in the firm annual (365 day) realized volatility from CRSP. 6M Return<sub>*j*,*i*,*t*</sub> is the 6-month CRSP compounded cumdividend stock return of sample firms. Mortgage-Income<sub>*i*,*t*</sub> is the mortgage-to-income ratio of the household.  $\Delta \text{Income}_{i,t}$  is the 6-month change in average household income. Volatility<sub>*j*,*i*,*t*</sub> and Realized Volatility<sub>*j*,*i*,*t*</sub> are the levels of option-implied and realized volatility (annualized) of firms, respectively.

	Obs.	Mean	S.Dev	Min	P1	P25	P50	P75	P99	Max
$\Delta \text{Consumption}_{i,t}$	1,364,114	0.061	0.563	-1.536	-1.435	-0.261	0.053	0.387	1.468	1.681
$\Delta \text{Durables}_{i,t}$	1,117,418	0.004	0.777	-1.768	-1.714	-0.499	0	0.511	1.730	1.803
$\Delta$ Volatility <sub>j,i,t</sub>	$1,\!364,\!114$	-0.034	0.091	-0.285	-0.223	-0.090	-0.041	0.016	0.239	0.333
$\Delta$ Realized Volatility <sub>j,i,t</sub>	$1,\!364,\!114$	-0.040	0.186	-0.628	-0.490	-0.157	-0.041	0.056	0.534	0.718
6M Return <sub><math>j,i,t</math></sub>	$1,\!364,\!114$	0.086	0.213	-0.730	-0.447	-0.025	0.090	0.202	0.678	1.433
$\Delta \text{Income}_{i,t}$	$1,\!364,\!114$	0.009	0.363	-1.459	-1.088	-0.145	0.005	0.166	1.087	1.288
Mortgage-Income <sub><math>i,t</math></sub>	$1,\!364,\!114$	0.162	0.315	0	0	0	0	0.238	1.814	2.784
Volatility <sub><math>j,i,t</math></sub>	$1,\!364,\!114$	0.319	0.114	0.157	0.168	0.238	0.287	0.378	0.720	0.888
Realized Volatility $_{j,i,t}$	$1,\!364,\!114$	0.309	0.142	0.127	0.141	0.209	0.265	0.370	0.770	0.932

### Table 3. Employer uncertainty shocks and future household consumption

This table shows the forecasting regression effect of firm (employer) uncertainty shocks on future household (employee) consumption growth. Frequency of all variables is monthly.  $\Delta \text{Consumption}_{i,t}$  is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household *i* level (employees). For each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months.  $\Delta$ Volatility<sub>i i t-6</sub> is the 6-month growth in the option-implied volatility (365-day horizon from OptionMetrics) of the corresponding employer i of each household. The timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>*i,i,t*-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where  $Mortgage-Income_{i,t-6}$  is the mortgage-to-income ratio of the household.  $\Delta$ Income<sub>i,t</sub> is the 6-month change in average household income, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>*i*,*t*</sub> and  $\Delta$ Income<sub>i,t-6</sub>, respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

$\Delta Consumption_{i,t}$	(1)	(2)	(3)	(4)	(5)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-1.49***	-1.33***	-1.33***	-1.26***	-1.28***
•, ,	(0.407)	(0.396)	(0.396)	(0.388)	(0.387)
$6M \operatorname{Return}_{j,i,t-6}$		$0.511^{***}$	$0.511^{***}$	$0.559^{***}$	$0.520^{***}$
		(0.189)	(0.189)	(0.184)	(0.183)
Mortgage-Income <sub><math>i,t-6</math></sub>			$-0.334^{*}$	-1.15***	$-0.947^{***}$
			(0.178)	(0.166)	(0.165)
$\Delta \text{Income}_{i,t}$				4.71***	$5.80^{***}$
				(0.387)	(0.350)
$\Delta \text{Lag Income}_{i,t-6}$					$2.01^{***}$
					(0.134)
Cost of Living Index	Yes	Yes	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes
Observations	$1,\!363,\!246$	$1,\!363,\!246$	$1,\!363,\!246$	$1,\!363,\!246$	1,363,246
$\mathbb{R}^2$	0.135	0.136	0.136	0.142	0.142

schold consumption, alternative specifications
consumption,
household
hocks and future hous
ncertainty shocks a
Table 4. Ur

baseline specification (1) includes all controls specified in column (5) in Table 3. Frequency of all variables is monthly.  $\Delta$ Consumption<sub>i,t</sub> is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household i level (employees). For each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into columns (1A), (2A), (3A), (4A), (5A),  $\Delta$ Volatility<sub>j,i,t-6</sub> is the lagged 6-month growth in the firm annual (365 day) realized volatility of the firm's CRSP stock return. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the This table shows robustness of the forecasting effect in Table 3 of firm (employer) uncertainty on household (employee) future consumption. The to the next 6-months. In columns (1), (2), (3), (4), (5),  $\Delta$ Volatility<sub>j,i,t-6</sub> is the 6-month growth in the *option-implied* volatility (365-day horizon from OptionMetrics) of the corresponding employer j of each household, over the six month period preceding the consumption growth. Similarly, in lagged stock return of the employer, 6M Return<sub>j,i,t-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where Mortgage-Income $_{i,t-6}$  is the mortgage-to-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>i,t</sub> and  $\Delta$ Income<sub>i,t-6</sub>, respectively. To account for restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors (×100) are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. The specifications below include a combination of firm, industry (3-digit Standard Industry Classification), household, and time fixed effects. Likewise the specifications explore robustness to clustering standard errors at one or multiple the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, different dimensions. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

		ImJ	Implied Volatility	lity			Rea	Realized Volatility	ility	
$\Delta { m Consumption}_{i,t}$	(1)	(2)	(3)	(4)	(5)	(1A)	(2A)	(3A)	(4A)	(5A)
$\Delta  ext{Volatility}_{i,i,t=6}$	-1.28***	-1.28***	-1.27***	-1.28***	-1.28**	-0.968**	-0.968**	-0.918**	-0.968***	-0.968**
	(0.387)	(0.425)	(0.413)	(0.265)	(0.495)	(0.376)	(0.389)	(0.381)	(0.208)	(0.467)
$6M \operatorname{Return}_{i,i,t-6}$	$0.520^{***}$	$0.520^{***}$	$0.511^{***}$	$0.520^{***}$	$0.520^{**}$	$0.509^{***}$	$0.509^{***}$	$0.508^{***}$	$0.509^{***}$	$0.509^{**}$
	(0.183)	(0.189)	(0.184)	(0.149)	(0.199)	(0.190)	(0.189)	(0.185)	(0.134)	(0.201)
$Mortgage-Income_{i,t-6}$	-0.947***	-0.947***	$-0.943^{***}$	-0.947***	-0.947***	-0.939***	-0.939***	$-0.935^{***}$	-0.939***	-0.939**
	(0.165)	(0.160)	(0.161)	(0.175)	(0.144)	(0.163)	(0.158)	(0.158)	(0.170)	(0.142)
$\Delta \mathrm{Income}_{i,t}$	$5.80^{***}$	$5.80^{***}$	$5.79^{***}$	$5.80^{***}$	$5.80^{***}$	$5.81^{***}$	$5.81^{***}$	$5.81^{***}$	$5.81^{***}$	$5.81^{***}$
~	(0.350)	(0.354)	(0.354)	(0.156)	(0.407)	(0.336)	(0.343)	(0.342)	(0.157)	(0.394)
$\Delta { m Lag~Income}_{i.t-6}$	$2.01^{***}$	$2.01^{***}$	$2.01^{***}$	$2.01^{***}$	$2.01^{***}$	$1.98^{***}$	$1.98^{***}$	$1.97^{***}$	$1.98^{***}$	$1.98^{***}$
)	(0.134)	(0.138)	(0.137)	(0.130)	(0.145)	(0.131)	(0.135)	(0.135)	(0.128)	(0.144)
Cost of Living Index	Yes	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	Yes	$\mathbf{Y}_{\mathbf{es}}$	Yes
<b>Fime FE</b>	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	Yes	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Household FE	$\mathbf{Yes}$	Yes	$\mathbf{Yes}$	Yes	Yes	$\mathbf{Yes}$	$\mathbf{Yes}$	Yes	Yes	$\mathbf{Yes}$
Firm FE	$\mathbf{Yes}$	$\mathbf{Yes}$	$N_{O}$	Yes	Yes	$\mathbf{Yes}$	$\mathbf{Yes}$	$N_{O}$	$\mathbf{Yes}$	Yes
Industry FE	$N_{O}$	$N_{O}$	$\mathbf{Y}_{\mathbf{es}}$	$N_{O}$	$N_{O}$	$N_{O}$	$N_{O}$	$\mathbf{Y}_{\mathbf{es}}$	$N_{O}$	$N_0$
3E Clustering - Firm	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$N_{O}$	$N_{O}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$N_{O}$	$N_{0}$
SE Clustering - Household	$N_{O}$	$N_{O}$	$N_{O}$	Yes	$N_{O}$	$N_{O}$	$N_{O}$	$N_{O}$	$\mathbf{Y}_{\mathbf{es}}$	$N_{O}$
SE Clustering - Industry	$N_{O}$	$N_{O}$	$N_{O}$	$N_{O}$	$\mathbf{Y}_{\mathbf{es}}$	$N_{O}$	$N_{O}$	$N_{O}$	$N_{O}$	$\mathbf{Y}_{\mathbf{es}}$
SE Clustering - Time	$N_{O}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$N_{O}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Observations	1,363,246	1,363,246	1,363,246	1,363,246	1,363,246	1,429,048	1,429,048	1,429,049	1,429,048	1,429,048
3.2	0.142	0.142	0.141	0.142	0.142	0.144	0.144	0.142	0.144	0 144

### Table 5. Uncertainty measured in shocks and in levels, lagged and current

This table shows the effect of firm (employer) uncertainty when measured either in shocks or levels on either future or contemporaneous household (employee) consumption. The baseline specifications of uncertainty shocks in (1) and (1A) include all controls specified in columns (5) and (5A) in Table 4. Frequency of all variables is monthly.  $\Delta Consumption_{i,t}$  is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household i level (employees). For each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months. In columns (1), (2), (3), (4), volatility is from the employer firms' option-implied volatility (365-day horizon from OptionMetrics) and in columns (1A), (2A), (3A), (4A), volatility is from the firm's annual (365 day) realized volatility of the firm's CRSP stock return.  $\Delta$ Volatility<sub>*j*,*i*,*t*-6</sub> is the 6-month growth in volatility over the six month period preceding the LHS consumption growth outcome. Volatility i,i,t-6 is the level of firm volatility lagged by 6-months, and Volatility i,i,t is the volatility level measured at the same month t as the LHS outcome. To disentangle between the effect of 2nd moment uncertainty and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>*i*,*i*,*t*-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where  $Mortgage-Income_{i,t-6}$  is the mortgage-to-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta \text{Income}_{i,t}$  and  $\Delta \text{Income}_{i,t-6}$ , respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

		Implied	Volatility			Realized	Volatility	
$\Delta \text{Consumption}_{i,t}$	(1)	(2)	(3)	(4)	(1A)	(2A)	(3A)	(4A)
$\Delta$ Volatility <sub>j,i,t-6</sub>	$-1.28^{***}$ (0.387)				$-0.968^{**}$ (0.376)			
Volatility <sub><math>j,i,t-6</math></sub>	× /	$-1.58^{**}$ (0.665)			· · · ·	$-1.98^{**}$ (0.788)		
$\Delta$ Volatility <sub>j,i,t</sub>			$-0.856^{***}$ (0.308)				$-0.472^{*}$ (0.262)	
Volatility <sub><math>j,i,t</math></sub>			. ,	$-2.91^{**}$ (1.15)			. ,	$-1.76^{***}$ (0.640)
6M Return <sub><math>j,i,t-6</math></sub>	$0.520^{***}$ (0.183)	$0.760^{***}$ (0.238)	$0.587^{***}$ (0.211)	$0.585^{***}$ (0.190)	$0.509^{***}$ (0.190)	$0.519^{***}$ (0.196)	$0.605^{***}$ (0.229)	$0.454^{**}$ (0.196)
Mortgage-Income <sub><math>i,t-6</math></sub>	$-0.947^{***}$ (0.165)	$-0.945^{***}$ (0.166)	$-1.01^{***}$ (0.171)	$-1.00^{***}$ (0.170)	-0.939*** (0.163)	$-0.933^{***}$ (0.162)	-0.998*** (0.168)	$-1.00^{***}$ (0.167)
$\Delta \text{Income}_{i,t}$	$5.80^{***}$ (0.350)	$5.78^{***}$ (0.347)	$5.86^{***}$ (0.361)	$5.85^{***}$ (0.361)	$5.81^{***}$ (0.336)	$5.81^{***}$ (0.338)	$5.88^{***}$ (0.350)	$5.88^{***}$ (0.350)
$\Delta$ Lag Income <sub><i>i</i>,<i>t</i>-6</sub>	$2.01^{***}$ (0.134)	$2.00^{***}$ (0.136)	$2.04^{***}$ (0.137)	$2.04^{***}$ (0.139)	$1.98^{***}$ (0.131)	$1.98^{***}$ (0.133)	$2.02^{***}$ (0.136)	$2.02^{***}$ (0.135)
Cost of Living Index	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations $\mathbb{R}^2$	$1,363,246 \\ 0.142$	$1,374,289 \\ 0.142$	$1,284,141 \\ 0.141$	$1,289,816 \\ 0.141$	$1,429,048 \\ 0.144$	$1,431,767 \\ 0.144$	$1,339,388 \\ 0.142$	$1,339,388 \\ 0.142$

# Uncertainty and the horizon of consumption growth forecasts Table 6.

this growth horizon, where in column (1) "1Mth" consumption growth is the growth from one month to the next, in column (2) "3Mths" growth is volatility (at 365-day horizon from OptionMetrics) and is measured with a  $\tau$ -month lag from 1- to 12-months. Similarly, realized volatility shocks uncertainty and forecasted household consumption growth. Frequency of all variables is monthly.  $\Delta$  Consumption<sub>i,t</sub> is the growth in average monthly consumption of retail, restaurant, and groceries. We construct the growth at different horizons in this table. For instance, our baseline specifications in Tables 3, 4, 5, measure 6-month growths in average monthly household consumption, where for each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months. We vary from a 3-month average monthly consumption to the next 3-month window, and in column (5) "12Mths" is the average monthly consumption growth forecast a full 1-year ahead. To disentangle between the predictive effect of 2nd moment uncertainty and first moment effects, we control for the lagged stock return of the employer, which uses an analogue window as to the forecasting growth of consumption. For instance, for column (2) the return is the 3-month compounded stock return of the firm lagged by 3 months with respect to the LHS outcome, and in (5) the return is the 12-month compounded return lagged by a full year. The main forecasting variable  $\Delta$ Volatility<sub>*j*,*i*,*t*- $\tau$  is the employer *j*'s 6-month growth in the option-implied</sub> are the lagged growths in the firm annual (365 day) realized volatility of the firm's CRSP stock return. We further control for household debt effects, where Mortgage-Income $_{i,t-\tau}$  is the mortgage-to-income ratio of the household. To account for the effect of cost-of-living differences all specifications Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and This table shows the effect of firm (employer) uncertainty shocks on household (employee) consumption using different time periods in between include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

		Imp	Implied Volatility	lity			Reali	Realized Volatility	ity	
$\Delta { m Consumption}_{i,t}$	$1 \mathrm{Mth}$ $(1)$	3Mths $(2)$	$\begin{array}{c} 6 \mathrm{Mths} \\ (3) \end{array}$	$\begin{array}{c} 9 \mathrm{Mths} \\ (4) \end{array}$	$\begin{array}{c} 12 \mathrm{Mths} \\ (5) \end{array}$	$1 \mathrm{Mth}$ $(1 \mathrm{A})$	$3 \mathrm{Mths}$ $(2 \mathrm{A})$	6Mths (3A)	$\begin{array}{c} 9 \mathrm{Mths} \\ (4 \mathrm{A}) \end{array}$	$\begin{array}{c} 12 \mathrm{Mths} \\ (5 \mathrm{A}) \end{array}$
$\Delta \mathrm{Volatility}_{j,i,t-\tau}$	$-0.156^{*}$ (0.0802)	$-0.654^{***}$ (0.216)	$-1.28^{***}$ (0.387)	$-1.58^{***}$ (0.407)		-0.0468 (0.0995)		$-0.968^{**}$ (0.376)	$-1.23^{**}$ (0.483)	$-0.648^{*}$ (0.353)
$\operatorname{Return}_{j,i,t-\tau}$	-0.118	0.0985	$0.520^{***}$	$0.424^{*}$		-0.102		$0.509^{***}$	0.427	0.156
Mortgage-Income_ $i,t- au$	Ŷ	(0.130) - $0.667^{***}$	$(0.183)$ - $0.947^{***}$	(0.234)-1.09***		$(0.103) - 0.378^{***}$		$(0.190) - 0.939^{***}$	(0.264)-1.13***	(0.242)-1.05***
Ĭ	$\smile$ .	(0.128)	(0.165)	(0.230)		(0.0863)		(0.163)	(0.223)	(0.312)
$\Delta \mathrm{Income}_{i,t}$	$3.55^{***}$ $(0.231)$	$4.68^{***}$ (0.337)	0.350	0.368		$3.57^{\pm\pm\pm}$ $(0.225)$		0.336	$0.09^{***}$	0.378
$\Delta \text{Lag Income}_{i,t- au}$	$1.89^{***}$	$1.37^{***}$	$2.01^{***}$	$0.618^{***}$		$1.89^{***}$		$1.98^{***}$	$0.619^{***}$	0.287
	(0.132)	(0.143)	(0.134)	(0.132)		(0.126)		(0.131)	(0.129)	(0.200)
Cost of Living Index	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	Yes		Yes	Yes	
	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$		$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	
Household FE	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$		$\mathbf{Yes}$	$\mathbf{Yes}$	
Firm FE	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	${ m Yes}$		$\mathbf{Yes}$	$\mathbf{Yes}$	
$Observations$ $R^2$	$\begin{array}{ccc} 1,638,463 & 1,\\ 0.029 \end{array}$	$1,459,070\\0.069$	$1,363,246\\0.142$	$938,170 \\ 0.263$	611,085 0.420	$1,715,714\\0.029$	$1,465,744\\0.069$	$1,429,048\\0.144$	982,127 $0.264$	$638,462 \\ 0.420$

# Table 7. Good and bad uncertainty

in columns (1A), (2A), (3A), (4A),  $\Delta$ Volatility<sub>j,i,t- $\tau$ </sub> is the lagged by  $\tau = \{6, 9\}$  months growth in the firm annual (365 day) realized volatility of riod in columns (3),(4),(4),(4A), obtain the average monthly consumption over this span, and construct the growth into to the next 6- or 9-months accordingly. In columns (1), (2), (3), (4),  $\Delta$  Volatility  $_{j,i,t-\tau}$  is the  $\tau = \{6,9\}$  month growth in the *option-implied* volatility (365-day horizon from the firm's CRSP stock return. Columns (1) and (1A) report the baseline average effects of uncertainty shocks on 6-month ahead future consumption horizons, respectively. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, Return<sub>j,i,t- $\tau$ </sub>, defined as the CRSP compounded  $\tau$  month cum-dividend stock return. We further control for household debt effects, where Mortgage-Income<sub>i,t- $\tau$ </sub> is the mortgage-to-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by  $\tau$  months household income growth,  $\Delta$ Income<sub>i,t</sub> and  $\Delta$ Income<sub>i,t-\tau</sub>, respectively. To account for the effect show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the  $\Delta Consumption_{i,t}$  is either the 6- or 9-month growth in average monthly consumption of retail, restaurant, and groceries at the household *i* level growth as reported in columns (3) and (3A) in Table 6, while columns (3) and (3A) report the stronger average effects at 9-months ahead reported in columns (4) and (4A) in Table 6. Columns (2), (2A) and (4), (4A) test for asymmetry in the responses to uncertainty shocks at 6 and 9 month This table examines the effect of good and bad uncertainty shocks on future household consumption. We test for asymmetric responses to 'bad'  $(\Delta \text{Volatility} > 0)$  and 'good' ( $\Delta \text{Volatility} <= 0$ ) employer uncertainty shocks, by interacting the absolute value of the employer uncertainty shock,  $\Delta$ Volatility, with an employer dummy that takes value  $D^{Bad} = 1$  if  $\Delta$ Volatility> 0, zero otherwise. Frequency of all variables is monthly. OptionMetrics) of the corresponding employer j of each household, over the  $\tau = \{6, 9\}$  month period preceding the consumption growth. Similarly, of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

		Implied Volatility	Volatility			Realized Volatility	Volatility	
$\Delta { m Consumption}_{i,t}$	Baseline 6Mths (1)	Good & Bad 6Mths (2)	9 Mths (3)	$\begin{array}{c} Good \& Bad \\ 9 Mths \\ (4) \end{array}$	Baseline 6Mths (1A)	$\begin{array}{c} Good \ \& \ Bad\\ 6Mths\\ (2A)\end{array}$	9 Mths (3A)	$\begin{array}{c} Good \ \& \ Bad \\ 9 \ Mths \\ (4A) \end{array}$
$\Delta { m Volatility}_{j,i,t- au}$	$-1.28^{***}$ (0.387)		$-1.58^{**}$		$-0.968^{**}$		$-1.23^{**}$ (0.483)	
$ \Delta  ext{Volatility}_{j,i,t- au} $		0.370		0.420		$0.555^{***}$		$0.606^{**}$
$ \Delta  ext{Volatility}_{j,i,t- au}   imes  ext{D}_{j,i,t- au}^{Bad}$		(0.260)-2.27***		(0.288)-2.62***		(0.193)-1.56***		(0.250)-2.23***
$\mathrm{D}^{Bad}_{s,i+\tau}$		(0.664) -1.69***		(0.661) -2.79***		(0.457) -1.32**		(0.652) -1.96**
		(0.502)		(0.729)		(0.623)		(0.829)
$\operatorname{Return}_{j,i,t- au}$	$0.520^{***}$	$0.446^{***}$	$0.424^{*}$	0.289	$0.509^{***}$	$0.494^{***}$	0.427	0.366
	(0.183)	(0.172)	(0.234)	(0.214)	(0.190)	(0.187)	(0.264)	(0.246)
Controls	$\mathbf{Y}_{\mathbf{es}}$	Yes	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$
Cost of Living Index	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$
Time FE	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Household FE	${ m Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Firm FE	${ m Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Observations	1,363,246	1,363,246	938, 170	938, 170	1,429,048	1,429,048	982, 127	982, 127
$\mathbb{R}^2$	0.142	0.142	0.263	0.263	0.144	0.144	0.264	0.264

### Table 8. Firm, industry, and aggregate uncertainty

This table examines the response of household consumption to uncertainty and stock returns at the industry and aggregate levels, and tests whether the effects of employer-specific uncertainty remain after controlling for industry and aggregate effects. Frequency of all variables is monthly. Column (1) presents the baseline specification with controls presented in column (5) of Table 3.  $\Delta$ Consumption<sub>*i*,*t*</sub> is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household i level (employees). For each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months. Employer volatility shocks  $\Delta$ Volatility<sub>*i,i,t*-6</sub> is the 6-month growth in the firms' option-implied (OptionMetrics 365 day) volatility over the six month period preceding the LHS consumption growth outcome. Industry-level (3-digit Standard Industry Classification) volatility shocks and stock returns in columns (2) and (3) are the within 3SIC monthly cross-sectional means of firm volatility shocks and 6-month compounded stock return of firms in the same industry of the employer of the household. Columns (4-7) use the 6-month compounded S&P500 stock return and either 6-month shocks to the monthly VIX in columns (4) and (5) or its level in (6) and (7). To avoid collinearity with month-year fixed effects, columns (4-7) include quarter-year fixed effects. To disentangle between the effect of 2nd moment uncertainty and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>j,i,t-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where Mortgage-Income<sub>i,t-6</sub> is the mortgage-toincome ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta \text{Income}_{i,t}$  and  $\Delta \text{Income}_{i,t-6}$ , respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included and specified at the bottom of the Table. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

$\Delta Consumption_{i,t}$	(1)	(2)	(3)	(4)	(5)	(6)	(7)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-1.28***		-1.07***		-0.823**		-0.804**
- 0 / /	(0.387)		(0.297)		(0.381)		(0.381)
$6M \operatorname{Return}_{j,i,t-6}$	0.520***		0.290		0.472**		0.477**
<i></i>	(0.183)		(0.180)		(0.198)		(0.199)
$\Delta$ Ind Volatility <sub>j3SIC,i,t-6</sub>	· · · ·	-1.43**	-0.616		· /		× ,
		(0.708)	(0.722)				
Ind 6M Return <sub><math>j3SIC,i,t-6</math></sub>		1.02***	0.971***				
<i>JJJJJJJJJJJJJ</i>		(0.291)	(0.301)				
$\Delta \text{VIX}_{t=6}$			· · · ·	-0.0673	-0.153		
				(0.287)	(0.289)		
S&P500 6M Return $_{t-6}$				1.34***	1.08***	1.54***	1.30***
				(0.180)	(0.233)	(0.107)	(0.168)
$VIX_{t-6}$						$0.565^{*}$	0.428
						(0.291)	(0.304)
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Cost of Living Index	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Month-year FE	Yes	Yes	Yes	No	No	No	No
Qtr-year FE	No	No	No	Yes	Yes	Yes	Yes
Observations	1,363,246	1,363,246	1,363,246	1,363,246		1,363,246	1,363,246
$\mathbb{R}^2$	0.142	0.142	0.142	0.140	0.140	0.140	0.140

### Table 9. Uncertainty and spending at retail, restaurant, grocery stores

This table shows the forecasting regression effect of firm (employer) uncertainty shocks on future household (employee) retail, restaurant, and grocery consumption growth. Frequency of all variables is monthly.  $\Delta \text{Consumption}_{i,t}$  is the 6-month growth in average monthly consumption of retail, restaurant, and groceries at the household *i* level (employees). For each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months. In columns (1), (2), (3), volatility is from the employer firms' option-implied volatility (365-day horizon from OptionMetrics) and in columns (1A), (2A), (3A), volatility is from the firm's annual (365 day) realized volatility of the firm's CRSP stock return.  $\Delta$ Volatility<sub>j,i,t-6</sub> is the 6-month growth in the option-implied volatility (365-day horizon from OptionMetrics) of the corresponding employer j of each household. The timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome. Similarly, realized volatility shocks are the lagged 6-month growths in the firm annual (365 day) realized volatility of the firm's CRSP stock return. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>j,i,t-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where  $Mortgage-Income_{i,t-6}$  is the mortgage-to-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>i,t</sub> and  $\Delta$ Income<sub>i,t-6</sub>, respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

	Imj	olied Volatili	ty	Rea	alized Volatil	ity
$\Delta Consumption_{i,t}$	Retail (1)	Restaurant (2)	Grocery (3)	Retail (1A)	Restaurant (2A)	Grocery (3A)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-1.32***	-0.381*	-0.430*	-1.03**	-0.081	-0.500**
	(0.422)	(0.208)	(0.226)	(0.448)	(0.158)	(0.239)
$6M \operatorname{Return}_{j,i,t-6}$	$0.336^{*}$	0.163	0.0470	$0.343^{*}$	0.151	0.005
	(0.204)	(0.140)	(0.162)	(0.203)	(0.138)	(0.160)
Mortgage-Income <sub><math>i,t-6</math></sub>	$-0.748^{***}$	-0.192	-0.337	-0.774***	-0.183	-0.297
	(0.195)	(0.171)	(0.216)	(0.193)	(0.168)	(0.213)
$\Delta \text{Income}_{i,t}$	$5.43^{***}$	$5.31^{***}$	$2.67^{***}$	$5.45^{***}$	$5.35^{***}$	$2.67^{***}$
	(0.357)	(0.310)	(0.210)	(0.342)	(0.298)	(0.205)
$\Delta Lag Income_{i,t-6}$	$1.69^{***}$	$2.05^{***}$	$1.03^{***}$	$1.70^{***}$	$2.05^{***}$	0.982***
	(0.169)	(0.165)	(0.182)	(0.163)	(0.158)	(0.178)
Cost of Living Index	Yes	Yes	Yes	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	1,213,647	$1,\!120,\!955$	647,601	1,273,263	$1,\!176,\!864$	678,223
$\mathbf{R}^2$	0.138	0.106	0.152	0.139	0.106	0.152

d durable consumption
dur
ousehol
shocks and future h
and
shocks
Uncertainty s
Table 10.

Frequency of all variables is monthly.  $\Delta Durables_{i,t}$  is the 6-month growth in average monthly expenditures on automobile-related expenses, home is the 6-month growth in the option-implied volatility (365-day horizon from OptionMetrics) of the corresponding employer j of each household. The are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are improvement, and home maintenance at the household *i* level (employees). For each household we measure durable consumption every month over a 6month period, obtain the average monthly durable consumption over this span, and construct the growth into to the next 6-months.  $\Delta$  Volatility  $_{j,i,t-6}$ timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome. Similarly, realized volatility shocks are the lagged 6-month growths in the firm annual (365 day) realized volatility of the firm's CRSP stock return. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return  $j_{i,i,t-6}$ , defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where  $D_{i,t-6}^{Mortgage-Income}$  is a lagged household indicator variable equal to one if the 6-month average mortgage-to-income ratio of the household is equal to or greater than 10%, zero otherwise, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>i,t</sub> calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are This table shows the forecasting regression effect of firm (employer) uncertainty shocks on future household (employee) durable consumption growth. and  $\Delta \text{Income}_{i,t-6}$ , respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%. 5%, and 10% levels, respectively.

		Imp	Implied Volatility	ity			Reali	Realized Volatility	lity	
$\Delta \mathrm{Durables}_{i,t}$	$1 \mathrm{Mth}$ $(1)$	3Mths (2)	6 M ths (3)	$\begin{array}{c} 9 \mathrm{Mths} \\ (4) \end{array}$	$12 \mathrm{Mths}$ (5)	$1 \mathrm{Mth}$ $(1 \mathrm{A})$	3Mths (2A)	6Mths (3A)	9Mths (4A)	$12 \mathrm{Mths}$ $(5 \mathrm{A})$
$\Delta \text{Volatility}_{j,i,t-6}$	-0.005 (0.100)	$-0.352^{**}$ (0.167)	$-0.667^{***}$ (0.229)	$-0.597^{*}$ (0.334)	-0.258 (0.384)	$0.298^{**}$ (0.122)	0.136 (0.181)	$-0.431^{*}$ (0.254)	$-0.689^{**}$ (0.319)	-0.452 (0.350)
$6M \operatorname{Return}_{j,i,t-6}$	0.038 (0.161)	(0.126)	0.308 (0.190)	0.161 (0.280)	-0.258 (0.289)	0.028 (0.148)	0.010 (0.117)	$0.381^{**}$ (0.187)	0.314 (0.258)	-0.103 (0.287)
Mortgage-Income $_{i,t-6}$ -0	-0.804***	-1.03***	-1.46***	-2.36***	-2.87***	-0.785***	-1.04***	-1.40***	-2.32***	-2.83***
AIncome: 4	(0.122)	(0.157) 3 41***	(0.314) 4.13***	(0.385) 3.76***	(0.597) 4 65**	(0.121) 2.43***	(0.153) 3 40***	(0.285) 4.16***	(0.374) 3 82***	(0.578) 4 66***
-	(0.207)	(0.305)	(0.329)	(0.308)	(0.422)	(0.203)	(0.299)	(0.320)	(0.305)	(0.410)
$\Delta Lag Income_{i,t-6}$	$1.29^{***}$	$1.12^{***}$	$1.86^{***}$	$0.846^{***}$	$0.616^{*}$	$1.26^{***}$	$1.08^{***}$	$1.87^{***}$	$0.760^{***}$	0.473
	(0.160)	(0.136)	(0.229)	(0.217)	(0.351)	(0.153)	(0.132)	(0.224)	(0.218)	(0.361)
Cost of Living Index	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	Yes	Yes	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$		$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$
Time FE	${ m Yes}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$		$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Household FE	${ m Yes}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$		$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Firm FE	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$		$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Observations	1,040,428	1,387,864	1,116,247	744, 795	484,108	1,088,685	1,454,345	1,170,085	779,333	505,528
$\mathrm{R}^2$	0.021	0.039	0.087	0.179	0.340	0.021		0.087	0.179	0.340

# Table 11. Placebo tests

the average coefficients and standard errors from 50 regressions based on random matches (with different seeds and with replacement from a pool of increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles This table shows results from placebo falsification tests. We replace the true employer of the household in our sample with a placebo employer, where the null is that there is no response from household consumption to placebo employer uncertainty shocks. Under placebo columns, we show the average results from 50 iterations of random mapping of households to placebo firms. In particular, in columns (2), (4), (6), (8), and (10), we show over 1,700 placebo firms with required data in our sample). In row "Count: Vol & Return", we also report the number of times from the 50 placebo regressions where we saw significant (at the 5%) negative coefficients on the placebo vol shock and at the same time positive coefficients on the placebo stock return. Frequency of all variables is monthly.  $\Delta$  Volatility  $_{j,i,t-6}$  is the 6-month growth in the option-implied volatility (365-day horizon from Option Metrics) of the corresponding employer j of each household. The timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return  $_{j,i,t-6}$ , defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where  $D_{i,t-6}^{Mortgage-Income}$  is a lagged household indicator variable equal to one if the 6-month average mortgage-to-income lagged by 6-months household income growth,  $\Delta \text{Income}_{i,t}$  and  $\Delta \text{Income}_{i,t-6}$ , respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation ratio of the household is equal to or greater than 10%, zero otherwise, and household income shocks, where we include both the contemporaneous and each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

	Consumption	mption	Durables	bles	Re	Retail	Restaurant	urant	Grocery	cery
	$\begin{array}{c} \text{Sample} \\ (1) \end{array}$	Placebo (2)	Sample (3)	Placebo (4)	$\begin{array}{c} \text{Sample} \\ (5) \end{array}$	Placebo (6)	Sample (7)	Placebo (8)	Sample (9)	Placebo (10)
$\Delta  ext{Volatility}_{j,i,t-6}$	-1.28***	0.061	-0.667***	0.027	-1.32***	0.064	-0.381*	-0.011	$-0.430^{*}$	0.057
	(0.387)	(0.337)	(0.229)	(0.250)	(0.422)	(0.371)	(0.208)	(0.171)	(0.226)	(0.212)
$6M \operatorname{Return}_{j,i,t-6}$	$0.520^{***}$	0.064	0.308	0.022	$0.336^{*}$	0.063	0.163	0.037	0.047	0.048
	(0.183)	(0.222)	(0.190)	(0.210)	(0.204)	(0.254)	(0.140)	(0.145)	(0.162)	(0.161)
$Mortgage-Income_{i,t-6}$		-0.945***	-1.46***	-1.47***	-0.748***	-0.747***	-0.192	-0.193	-0.337	-0.343
	(0.165)	(0.166)	(0.314)	(0.312)	(0.195)	(0.195)	(0.171)	(0.171)	(0.216)	(0.216)
$\Delta \mathrm{Income}_{i,t}$	$5.80^{***}$	$5.81^{***}$	$4.13^{***}$	$4.14^{***}$	$5.43^{***}$	$5.44^{***}$	$5.31^{***}$	$5.31^{***}$	$2.67^{***}$	$2.68^{***}$
	(0.350)	(0.349)	(0.329)	(0.328)	(0.357)	(0.355)	(0.310)	(0.310)	(0.210)	(0.209)
$\Delta Lag \ Income_{i,t-6}$	$2.01^{***}$	$2.02^{***}$	$1.86^{***}$	$1.86^{***}$	$1.69^{***}$	$1.69^{***}$	$2.05^{***}$	$2.05^{***}$	$1.03^{***}$	$1.02^{***}$
	(0.134)	(0.136)	(0.229)	(0.229)	(0.169)	(0.168)	(0.165)	(0.166)	(0.182)	(0.182)
Cost of Living Index	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	Yes	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$
Time FE	$\mathbf{Yes}$	$\mathbf{Yes}$	Yes	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Household FE	$\mathbf{Yes}$	$\mathbf{Yes}$	Yes	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Firm FE	$\mathbf{Y}_{\mathbf{es}}$	${ m Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	${ m Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$	$\mathbf{Yes}$
Count: Vol & Return		0/50		0/50		0/50		0/50		0/50
Observations	1,363,246	1,363,246	1,116,247	1,116,247	1,213,647	1,213,647	1,120,955	1,120,955	$647,\!601$	647,601
$\mathrm{R}^2$	0.142	0.142	0.087	0.087	0.138	0.138	0.106	0.105	0.152	0.152

### Table 12. High and low-income households, consumption response to uncertainty shocks

This table shows the difference in intensity of the response of future household (employee) consumption to firm (employer) uncertainty shocks by household income levels. Frequency of all variables is monthly. The left panel is the baseline measure of consumption in Table 3 ( $\Delta$ Consumption<sub>*i*,*t*</sub>), which includes retail, restaurant, and groceries expenditures at the household i level (employees). The right panel is for durable consumption presented in Table 8 ( $\Delta$ Durables<sub>*i*,*t*</sub>), which includes expenditures on automobile-related expenses, home improvement, and home maintenance. We classify households into quartiles by their average income levels, from low-income columns (1) and (1A) to high-income (3) and (3A). Growth in the dependent variables are measured as the 6-month growth in average monthly expenditures for the corresponding consumption categories, where for each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months.  $\Delta$ Volatility<sub>*i.i.t*-6</sub> is the 6-month growth in the option-implied volatility (365-day horizon from OptionMetrics) of the corresponding employer j of each household. The timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>*i.i.t*-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where  $Mortgage-Income_{i,t-6}$  is the mortgage-to-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>i,t</sub> and  $\Delta$ Income<sub>i,t-6</sub>, respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

	ΔΟ	Consumpti	$\mathrm{on}_{i,t}$	$\Delta \text{Durables}_{i,t}$			
	Low-income (1)	(2)	High-income (3)	Low-income (1A)	(2A)	High-income (3A)	
$\Delta$ Volatility <sub>j,i,t-6</sub>	-1.44***	-1.07***	-0.830**	-0.952**	-0.860**	-0.130	
- 07 7	(0.507)	(0.385)	(0.376)	(0.432)	(0.365)	(0.288)	
$6M \operatorname{Return}_{j,i,t-6}$	$0.600^{**}$	0.363	$0.386^{*}$	$0.776^{**}$	-0.131	0.211	
•, ,	(0.284)	(0.225)	(0.232)	(0.331)	(0.281)	(0.332)	
Mortgage-Income <sub><math>i,t-6</math></sub>	$-1.59^{***}$	-1.14***	-0.521*	-2.29***	-1.32***	$-1.36^{***}$	
	(0.373)	(0.252)	(0.280)	(0.795)	(0.416)	(0.289)	
$\Delta \text{Income}_{i,t}$	8.14***	$4.79^{***}$	$4.66^{***}$	$4.50^{***}$	$4.68^{***}$	$4.19^{***}$	
	(0.402)	(0.231)	(0.259)	(0.481)	(0.281)	(0.460)	
$\Delta Lag Income_{i,t-6}$	$1.15^{***}$	$0.805^{***}$	$1.90^{***}$	0.487	0.800***	$2.61^{***}$	
	(0.203)	(0.179)	(0.223)	(0.360)	(0.305)	(0.320)	
Cost of Living Index	Yes	Yes	Yes	Yes	Yes	Yes	
Time FE	Yes	Yes	Yes	Yes	Yes	Yes	
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	
Observations	454,348	$454,\!460$	$454,\!433$	$334,\!049$	$373,\!448$	408,739	
$\mathbf{R}^2$	0.178	0.138	0.124	0.104	0.085	0.082	

### Table 13. Consumption response to uncertainty shocks, by firm characteristics

This table shows the difference in intensity of the response of future household (employee) consumption to firm (employer) uncertainty shocks by employer characteristics. In particular, using common company fundamental and financial data from Compustat we classify households by the characteristics of firms in the preceding year. This allows us to examine whether households that work for firms that recently experienced, say, low employment growth (e.g., firms with layoffs) respond differently to uncertainty than households whose employers experienced recent high employment growth (e.g., hiring expansions). We look at 12 firm characteristics: (1) the change in the number of employees at the firm, (2) investment - defined as capital expenditures over lagged plant, property, and equipment, (3) return on assets, (4) Tobin's Q, (5) CAPM  $\beta$ , (6) sales, (7) past 12 month returns, (8) change in intangibles expenses, (9) the Whited-Wu financial constraints index, (10) the Sales-Age financial constraints index, (11) past calendar year 12 month implied volatility, and finally (12) past calendar year 12 month realized volatility. Frequency of all variables is monthly. The dependent variable is our baseline measure of consumption in Table 3 ( $\Delta$ Consumption<sub>*i*,*t*</sub>), which includes retail, restaurant, and groceries expenditures at the household i level (employees). Consumption growth is measured as the 6-month growth in average monthly spending, where for each household we measure consumption every month over a 6-month period, obtain the average monthly consumption over this span, and construct the growth into to the next 6-months.  $\Delta$ Volatility<sub>j,i,t-6</sub> is the 6-month growth in the optionimplied volatility (365-day horizon from OptionMetrics) of the corresponding employer j of each household. The timing of these employer uncertainty shocks is lagged by a full 6-months with respect to the LHS outcome. To disentangle between the predictive effect of 2nd moment uncertainty shocks and first moment effects, we control for the lagged stock return of the employer, 6M Return<sub>*i,i,t*-6</sub>, defined as the CRSP compounded 6-month cum-dividend stock return. We further control for household debt effects, where Mortgage-Income $_{i,t-6}$  is the mortgage-to-income ratio of the household, and household income shocks, where we include both the contemporaneous and lagged by 6-months household income growth,  $\Delta$ Income<sub>i,t</sub> and  $\Delta$ Income<sub>i,t-6</sub>, respectively. To account for the effect of cost-of-living differences all specifications include a time-varying Cost of Living Index, calculated from the mean expenditures of gas, restaurant, groceries and retail for each city, for every month. The continuous independent variables are standardized to make coefficients comparable and show the effect of a standard deviation increase. Coefficients and standard errors  $(\times 100)$  are reported for each independent variable. Variables are winsorized at the 1 and 99 percentiles each month. Firm, household, and time fixed effects are included. The standard errors are clustered at the firm level and are reported in parentheses. \*\*\*, \*\*, and \* denote statistical significance at the 1%, 5%, and 10% levels, respectively.

	Employment growth			Investment			Return on assets		
	$\begin{array}{c} \text{Low} \\ (1) \end{array}$	(2)	$\begin{array}{c} \text{High} \\ (3) \end{array}$	$\begin{array}{c} \text{Low} \\ (4) \end{array}$	(5)	High (6)	Low (7)	(8)	High (9)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-2.27***	-0.828	-0.150	-0.266	-1.02**	-0.640*	-1.42***	-0.381	-1.13*
	(0.598)	(0.616)	(0.425)	(0.433)	(0.487)	(0.365)	(0.488)	(0.672)	(0.591)
6M Return <sub><math>j,i,t-6</math></sub>	$0.736^{**}$	$0.937^{**}$	0.167	0.792***	$0.556^{**}$	$0.421^{*}$	0.193	0.389	$0.716^{**}$
	(0.351)	(0.408)	(0.295)	(0.296)	(0.259)	(0.215)	(0.296)	(0.284)	(0.351)
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	444,613	447,290	441,392	$445,\!258$	443,357	438,676	447,854	443,553	445,197
$\mathbb{R}^2$	0.193	0.253	0.224	0.226	0.193	0.165	0.141	0.177	0.185

		Tobin's Q			CAPM #	3		Sales	
	$\begin{array}{c} \text{Low} \\ (1) \end{array}$	(2)	$\begin{array}{c} \text{High} \\ (3) \end{array}$	Low (4)	(5)	High (6)	$\frac{\text{Low}}{(7)}$	(8)	High (9)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-2.19***	-0.262	-0.609	-0.008	-0.246	-1.84***	-0.311	-0.655**	-2.36**
	(0.531)	(0.369)	(0.424)	(0.428)	(0.428)	(0.476)	(0.379)	(0.329)	(1.01)
6M Return $_{j,i,t-6}$	0.269	0.290	$0.526^{**}$	0.241	0.276	$0.695^{***}$	$0.530^{**}$	0.156	0.379
	(0.284)	(0.308)	(0.232)	(0.412)	(0.271)	(0.235)	(0.242)	(0.210)	(0.635)
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	$445,\!559$	$449,\!548$	441,381	458,892	432,143	$445,\!031$	445,678	448,825	441,823
$\mathbb{R}^2$	0.158	0.210	0.166	0.214	0.229	0.173	0.163	0.150	0.167
	12 1	month ret	urns	Inta	ngible In	vest.	WW-inc	lex Finar	n. Const.
	Low		High	Low		High	Low		High
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-1.30**	-0.493	0.0930	0.090	-1.20*	-1.47*	-0.653	-1.83**	-0.206
	(0.577)	(0.374)	(0.453)	(0.400)	(0.724)	(0.769)	(0.420)	(0.841)	(0.366)
6M Return <sub><math>j,i,t-6</math></sub>	0.322	$0.693^{***}$	0.411	0.498	$1.01^{**}$	0.436	$0.715^{**}$	0.299	$0.644^{***}$
	(0.326)	(0.246)	(0.325)	(0.449)	(0.492)	(0.359)	(0.324)	(0.445)	(0.229)
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	450,292	442,712	442,315	169,881	163, 181	166,408	448,179	435,998	441,000
$\mathbb{R}^2$	0.268	0.245	0.288	0.259	0.261	0.234	0.120	0.174	0.169
	SA-index Finan. Const.			Implied volatility			Realized volatility		
	Low		High	Low		High	Low		High
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
$\Delta$ Volatility <sub>j,i,t-6</sub>	-0.196	-2.60**	-1.03**	-0.632	-0.647	-1.52***	0.407	-0.338	-1.62***
	(0.368)	(1.11)	(0.430)	(1.01)	(0.523)	(0.513)	(0.611)	(0.607)	(0.525)
6M Return <sub><math>j,i,t-6</math></sub>	$0.590 \\ (0.358)$	$0.519^{***}$ (0.190)	0.432 (0.278)	-0.328 (0.254)	$0.621^{**}$ (0.279)	$0.600^{**}$ (0.257)	0.183 (0.334)	$0.579^{**}$ (0.291)	$0.592^{**}$ (0.262)
Controls	Yes	Yes	Yes	Yes	Yes	. ,	Yes	Yes	Yes
Time FE	Yes	Yes	Yes	Yes	Yes	Yes Yes	Yes	Yes	Yes
Household FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	476,327	423,530	436,784	444,774	445,047	444,451	454,662	435,699	445,147
$R^2$	0.173	0.159	0.159	0.210	0.243	0.185	0.234	0.277	0.210
10	0.110	0.103	0.103	0.210	0.240	0.100	0.204	0.411	0.210

Table 13.Consumption response to uncertainty shocks, by firm characteristics(Continued)