

LEADS on Macroeconomic Risks to and from the Household Sector

Jonathan A. Parker

The household sector is both a propagator of shocks to the economy, as wealth is redistributed across households with differing propensities to consume, and an originator of risky claims held in systemically important places, as losses are shifted from households to creditors such as financial institutions. Information about these exposures, like information generally, is conveyed by prices and so is underproduced by markets. Thus increased public collection, analysis, and distribution of information on household exposures to macroeconomic risk factors can potentially lead to better macroeconomic performance, both through better informed private decision making and through better public policy.

This chapter describes a system for monitoring, measuring, and publicizing exposures to and from the household sector. This system, called the LEADS system, is designed to provide market participants, regulators, and households with additional information to understand the reallocation of resources within, from, and to the household sector in response to macroeconomic events. In short, the system is designed to stress test the balance sheets of the households and include the households sector in measurement of systemic risk.

Jonathan A. Parker is the Donald C. Clark/HSBC Professor of Consumer Finance at Northwestern University and a research associate of the National Bureau of Economic Research.

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1 The LEADS system has three components, of which the first is the main
2 focus of this chapter. The first step is the collection of data on LEADS—
3 liabilities, earnings, assets, demographics, and financial sophistication—
4 at the household level. I argue that these categories are the key dimensions to
5 measure, that measurement at the household level is critical for accurate
6 measurement, and that much of this information is available in institutions
7 already subject to government oversight and reporting requirements. The
8 second step in LEADS is the measurement of the exposure of each asset,
9 each liability, and each income stream to macroeconomic risk factors. This
10 step is the subject of much of the field and practice of finance, and describing
11 the vast and evolving set of techniques for this step is beyond this chapter.
12 This component of the system requires historical data on returns (at a mini-
13 mum) and modeling of future exposures.

14 The final step is the analysis and release of information. I propose analysis
15 of the implications of changes in risk factors in four important dimensions:
16 the distributional impacts on both liquid wealth and lifetime wealth; the
17 changes in household demand; the effects of balance sheet adjustments on
18 the prices and payouts of claims on the household sector held by other sec-
19 tors; and the resulting impact on the revenues and liabilities of the govern-
20 ment, through possible transfers and de jure and de facto guarantees to the
21 household sector. Such projections necessarily involve household decisions.
22 This information on exposures across groups of households could then be
23 combined with information on financial sophistication. Where exposures—
24 to or from the household sector—are large and sophistication low, macro-
25 economic risks may be mispriced or amplified by lack of sophistication and
26 knowledge.¹ Finally, the results of the analyses could be made public—
27 potentially pushed in some cases to “unsophisticated” households—and
28 the underlying data made public in a suitably limited form that maintains
29 privacy.

30 Together the LEADS system would uncover information relevant to
31 addressing the following types of questions that are of interest to market
32 participants and policymakers:

- 33 • How risky are loans to the household sector? Is a particular type of
34 aggregate risk concentrated among households with few resources or
35 little sophistication and so represents a potential source of losses for
36 other claims on these households?
- 37 • Is a particular type of macroeconomic or systematic risk held primar-
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41 1. The focus of this chapter on sophistication in the household sector is not intended
42 to single out or incriminate this sector. Systematic risks could be caused or hidden by lack
43 of sophistication, limited knowledge and biases among legislators, regulators, and courts (as
44 well as financial institutions, businesses, etc.). Lack of sophistication outside the household
sector may have been a cause of the recent crisis, and may still cause systemic risk even with
the LEADS information. This chapter simply focuses on measurement in the household
sector.

ily by households with little financial sophistication and so potentially mispriced?

- How exposed is consumer demand to declines in different assets or asset classes? Is a particular type of aggregate risk concentrated in liquid wealth or on households whose demand is highly sensitive to losses?
- How likely are private losses to become public liabilities? Is a particular group of households, like pensioners, holding enough of a specific macroeconomic or systematic risk so that losses might lead to ex post public assistance?

With very narrow exceptions, these questions cannot be answered by analysis of existing data sets.

This chapter—and the NBER Initiative on Systemic Risk and Macro Modeling of which it is a part—are both motivated directly by the credit market disruptions and financial crisis of 2007 and early 2008, the contemporaneous declines in asset values, and the large macroeconomic consequences. At the time, most observers expected that the decline in demand for investment and consumption goods following these events would be both similar to that caused by the stock market decline in the year 2000 and concentrated on the construction sector (e.g., Bernanke 2008).² In fact, consumption demand fell significantly during the Great Recession—more than output—and has been slow to recover after (relative to most previous US recessions).

The financial crisis and recession of 2008 to 2009 illustrate the two main ways in which the household sector is important for measurement of systemic risks. First, household demand is critical for business cycles, and as such the monitoring of households' balance sheets and wealth is a natural part of the monitoring of macroeconomic risks. Second, systemically important institutions hold claims on the household sector, and so understanding the correlated risks of these institutions requires understanding the value of these claims in different macroeconomic scenarios. Section 13.1 contains a discussion of these issues.

Section 13.2 presents the structure of the LEADS data and describes how this structure allows one to measure aggregate risks to and from the household sector. Section 13.3 describes how the LEADS data can be collected and compiled. Current sources of information are disparate, do not cover the same households, lack sufficient detail on asset holdings, and do not measure household sophistication. The data set for the LEADS system can be constructed by merging administrative data on investments and debts

2. There were even reasons to believe the macroeconomic impact would be smaller than in 2000. Housing is consumption, and as house values decline, the (opportunity) cost of housing falls, providing insurance to households that own homes. Further, the structure of mortgages provides households with an option that increases in value when house prices decline, transferring wealth from high-wealth, high-saving households to low-wealth, high-consumption households.

1 with a panel survey of households that focuses on demographics, income,
2 and financial sophistication.

3 Section 13.4 discusses analysis and dissemination. Analysis of the data
4 can use existing tools employed in the study of financial risks and household
5 finances. I sketch a three-step procedure to first measure individual-asset
6 exposures, calculate liquid wealth and lifetime wealth exposures, and map
7 these back to changes in both household demand and the value of claims
8 on households. To allow better management of aggregate risks, the analysis,
9 summary data, and an anonymous random sample of detailed data can be
10 released to the public. Section 13.5 concludes, and an appendix discusses the
11 role of the government in the provision of this type of information.

12 13 **13.1 Why Monitor Household Exposures?**

14 The exposures to and from the household sector are important for moni-
15 toring and measuring aggregate risks for three broad reasons.

16 First, movements in household expenditures amplify and propagate
17 shocks to the economy. This was true in the recession of 2008 to 2009; during
18 and following the recession, sluggish household expenditures have ampli-
19 fied and propagated slowdown. This has also been true more broadly. As
20 Hall (1986 and forthcoming) shows, the volatility of GDP comes primarily
21 from household spending. And long slumps from the Great Depression to
22 the recent recession are arguably amplified by low consumer demand due in
23 part to debt overhang (e.g., Melzer 2010).

24 Since households own firms and are the government, household demand
25 is exposed to macroeconomic risks through changes in aggregate income
26 and aggregate asset values (and gains or losses on net asset positions with
27 foreign countries). Thus, a major part of risk in household demand can be
28 measured from aggregate data on the share of wealth held in different asset
29 classes and an evaluation of the riskiness of each asset class. But household-
30 level data on individual asset holdings and their characteristics can provide
31 a better understanding of the exposure of aggregate demand to asset values.

32 Any decline in any asset value has a disproportionate impact on house-
33 hold demand if it is accompanied by transfers among households of differ-
34 ing propensities to consume. This disproportionate effect can happen, for
35 example, if shocks redistribute resources between middle-aged households,
36 whose behavior is reasonably approximated by the life cycle model, and
37 younger households, whose behavior is better characterized by the buffer
38 stock model of consumption.³ In the recent recession and in the Great
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3. See Carroll (1997) and Gourinchas and Parker (2002). Kaplan and Violante (2011) provide a related model. Most current models of the macroeconomy are largely linear, even those with heterogeneous agents, so that the household-level data is not necessary in these *models* for predicting macroeconomic dynamics. This statement even applies to models with precautionary savings that match the large share of households with low wealth (Krusell and Smith 1998), although these models typically miss the volatility of asset markets.

1 Depression, for example, household demand seems to have been reduced
2 by the concentration of losses among households with leverage.⁴

3 Thus, the household sector has historically been an important proximate
4 source of output volatility and the household sector's response to wealth
5 changes is determined by the distribution of wealth changes both across
6 households and across two measures of wealth: short-term, liquid, finan-
7 cial resources and long-term, illiquid wealth, such as retirement accounts or
8 future income.

9 The second reason to monitor household exposures is that both the gov-
10 ernment and systemically important institutions hold financial claims on
11 the household sector. The ability of households to meet these claims and
12 not default in different macroeconomic scenarios determines the exposure
13 of these assets to systematic risks and thus the extent to which those holding
14 the financial claims are exposed.

15 In the recession of 2008 to 2009, a significant reason for the depth and
16 severity of the recession was large losses on loans to households that were
17 held by systemically important financial institutions. These exposures may
18 well have been smaller had the exposures to aggregate risk factors of the
19 various dimensions of wealth of the households with mortgages been better
20 understood by market participants or regulators.

21 The final reason to monitor household exposures is that, as with groups
22 of banks, the government cannot commit ex ante not to make large transfers
23 to groups of households following adverse outcomes.⁵ And in general, moni-
24 toring is helpful in dealing with the moral hazard problems that accompany
25 insurance. These types of assistance are public risks that expose aggregate
26 growth to the risks born by these households through increased tax rates
27 and decreased future spending and benefits.

28 Examples of governments assisting a subset of households that lose sig-
29 nificant resources include those following natural disasters, and the bailout
30 of the elderly following the Great Depression with the Social Security system
31 (enacted in 1935 and 1939), which paid retirees benefits starting in 1940 to
32 aid seniors whose wealth was wiped out by the Great Depression.⁶ As long
33 as the government ex post makes transfers to households that have suffered

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36 4. See Fisher (1933), Eggertsson and Krugman (2012), Hall (2011). Parker and Vissing-
37 Jorgensen (2009, 2010) find larger consumption declines for high-consumption or high-income
38 households, implying an important role for declines in asset prices and expected future pros-
39 pects of high-income households.

40 5. Farhi and Tirole (2012) clarify the point that it is not just that a single institution can be
41 too big to fail, but also groups of institutions. The Troubled Asset Relief Program provided
42 equity investments not just in large banks but in most small banks also.

43 6. And we may observe it if the Pension Benefit Guaranty Corporation (PBGC) fails or
44 some US states go bankrupt. On the other hand, households facing foreclosure in the current
45 recession were not bailed out. What little assistance that was provided—the Home Affordable
46 Modification Program, which affected less than a million mortgages—focused on overcoming
47 an ex post market inefficiency rather than being a transfer to households. Also, there was no
48 government rescue for the employees of Enron or Arthur Anderson, nor for the victims of the
49 Madoff fraud.

1 large losses, it is optimal for the government, like any private insurer, to
2 monitor these households and react to these risks.⁷

3 These are the reasons that market participants and policymakers benefit
4 from information on household asset holdings, but why does the govern-
5 ment have a role in gathering, analyzing, and publicizing this information?
6 There are two main reasons, addressed in more detail in the appendix. First,
7 as a general principle, in markets the social value of information exceeds
8 its private value, so information is underproduced by the market. Given
9 too little information, mistakes are made relative to the economy with the
10 socially optimal amount of information. And mistakes will tend to be based
11 on the common information that exists, so that misestimation leads to coordi-
12 nated mistakes that are by definition macroeconomic. Second, lack of
13 financial sophistication (such as lack of financial knowledge, limited infor-
14 mation processing abilities, and limited time allocated to financial decisions)
15 can lead some households to misprice aggregate risks, both leading to misal-
16 location by other sectors and exposing household demand and claims on
17 households to “mispriced” aggregate risk factors.⁸ While it is efficient to
18 have the downside economic risk held by those most willing to hold it, it is
19 generically inefficient to have it held by those least able to understand it or
20 who most underestimate it.

21 These considerations imply that data collection and analysis focus on
22 measuring the systematic exposure of two measures of household financial
23 wealth—liquid financial resources over a few years and lifetime wealth—
24 for not just the household sector but also for different households grouped
25 according to consumption response to these two measures of resources,
26 according to their importance for claims on the household sector, accord-
27 ing to the likelihood of losses being born by the government, and according
28 to measures of financial sophistication of the household decision makers.

30 13.2 LEADS Data

31 What information is required to measure exposures to and from the
32 household sector? First, household-level data is necessary so that the com-
33 mon risks of different types of households can be studied across both dif-
34 ferent groups and risks of interest. That is, while the units of analysis will
35 typically be groups of households, to characterize exposures among groups
36 of households with differing propensities to consume, for example, one has
37 to have information for many possible groupings defined by demographics

40 7. In terms of regulatory response, the government can simply disclose these exposures so
41 that government accounting is more informative, or the government can hedge the exposures
42 so that tax rates can be smoothed and market prices better reflect true risks, or finally, it can
43 restrict or intervene to deter the exposures ex ante that lead to bailouts in some states ex post.

44 8. Mispricing relative to a benchmark in which people are not limited in knowledge or
information-processing capacity.

1 and/or financial measures of interest such as liquid wealth or home owner-
2 ship status. Individual-level data allow the study of the history of exposures
3 and behavior of different groups or types of households through different
4 aggregations of historical data. Further, to calculate the exposure of claims
5 on household resources to an aggregate shock, one needs to model the default
6 of each household, which again is most straightforward (and requires the
7 least extraneous assumptions) using household-level data. For example, to
8 predict how much default would accompany a 10 percent decrease in house
9 values for a group of households, it would be useful to know not just that
10 the average loan-to-value ratio of households in that group was 80 percent,
11 but also the distribution of loan-to-value ratios and how correlated other
12 household assets and incomes were with the considered aggregate risks.
13 Finally, the study of household-level data is a useful input to risk calcula-
14 tions, such as under what conditions households default or how different
15 regulations might change household behavior. That is, household-level data
16 allows one to use existing variation in laws, regulation, prices, and so forth
17 across households to study and measure household behavior and thus infer
18 better what losses to and from the sector would occur in response to what
19 aggregate events.

20 In terms of the information on each household, the arguments in sec-
21 tion 13.1 imply that the data contain enough detail on assets, income, and
22 liabilities to accurately measure the extent to which a household's *liquid*
23 *wealth* and *lifetime wealth* are correlated with macroeconomic risk factors.
24 This requires knowledge about the nature of income, of assets held in each
25 asset class, and of credit terms for debt, so that the impact of changes in
26 each on liquid wealth and on total wealth can be calculated. For example,
27 holdings in retirement accounts clearly expose lifetime wealth and not liquid
28 wealth, while temporary shocks to income affect liquid wealth more than
29 lifetime wealth. In sum, a system to monitor systemic risk in the household
30 sector requires data on assets and liabilities both at the household level and
31 in enough detail to assess their roles in liquid wealth and lifetime wealth.

32 What actual financial information about households is needed? As in
33 the monitoring and regulation of the US banking system, one would like
34 to observe sufficient detail about household balance sheets to accurately
35 measure the exposure of each asset and liability to aggregate risk factors of
36 interest. To this end, the data need to contain information not just on the
37 holdings in any asset class, but on the actual details of the securities held.
38 It is insufficient to measure the risk of a class of assets because one group
39 of households may differ significantly from another in the actual securities
40 held within that class, and so actually have quite a different exposure to
41 a macroeconomic risk factor. As examples, among mortgages, the extent
42 to which households default will differ dramatically with the terms of the
43 mortgage. Stocks can have high exposure to aggregate risk or provide insur-
44 ance against aggregate shocks. Hedge funds can be highly levered and lose

1 money in response to credit shock, or provide liquidity in a credit shock and
2 be highly profitable.

3 A similar argument applies to labor and benefit income. One needs to
4 know enough details of the labor income of the household to measure the
5 exposure of labor income to macroeconomic risk factors and to measure the
6 household's ability to avoid default or bankruptcy. Some households have
7 stable labor and benefit incomes and others are highly exposed to business
8 cycles. The actual exposures, as for assets, can be estimated from existing
9 data on historical labor incomes of similar households. And benefits and
10 income amounts from each source are necessary to infer total exposure.

11 Finally, an important part of monitoring banks is the quality of manage-
12 ment and its plans for future contingencies. The measurement of financial
13 sophistication in the household sector is similarly important for measur-
14 ing and monitoring risks and for providing clues as to which risks might
15 be mispriced. In banking regulation, the quality of management informs
16 the regulator about the likelihood that the financial institution can man-
17 age the exposures inherent in the bank's asset and liability positions. In
18 the recent financial crisis, measurement of this dimension for Fannie Mae
19 and AIG Financial Products would have shown poor management prac-
20 tices, been easy to correlate with massive exposures to real estate prices and
21 price impact, and potentially been useful to other investors taking prices as
22 informative about the riskiness of mortgage-backed securities. While poorly
23 managed firms tend to suffer a Darwinian fate, this argument has little bite
24 for households living well above subsistence. And regulation today reflects
25 this: regulations restrict most households from making many investment
26 choices, which are available only to qualified investors.⁹

27 While financial sophistication has many dimensions, the most pertinent
28 to measure is the extent to which households are informed about aggregate
29 risks and their exposures to them. If households are not informed, that
30 is not proof that they are incorrectly exposed, but it suggests that greater
31 information about exposures might change behavior for the better. The mea-
32 sures, discussed later, capture the extent to which households are informed
33 about the financial decisions they are making, the extent to which they have
34 the abilities to make reasonable financial decisions, and the extent to which
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37 9. Would information on sophistication have been helpful before the financial crisis and 2008
38 to 2009 recession? Probably not (although Lusardi [2010] argues to some extent otherwise).
39 This sort of information on households seems unlikely to have been useful in monitoring the
40 systematic risks and in avoiding them, since closely related data was available. The informa-
41 tion in credit ratings bureaus showed that subprime borrowers had very poor credit ratings,
42 which probably correlates highly with lack of financial sophistication. But this does not mean
43 that this type of information will not be useful next time. In the next potential crisis, it might
44 be that households take on more risk than they intend absent LEADS-type information, or
that markets or regulators would find information on the sophistication of the households
driving financial flows to be a useful signal to help interpret price movements and measure
macroeconomic risks.

Table 13.1 An overview of LEADS for households

Data at the household level on:

Liabilities	Measure terms of each borrowing instrument and calculate exposure—collateral information, commitment/term, interest-rate determination, penalties, etc. <i>Details almost completely lacking in current data sets.</i>
Assets	Details on each investment, including restricted accounts like retirement. Examples: name of hedge fund, actual security, house address, etc. <i>Details almost completely lacking in current data sets.</i>
Earnings	Measure of current and past incomes at the household level as well as dynamics. <i>Current data sets strong (PSID, NLSY, CPS, ACS, IRS).</i>
Demographics	(age, family structure, geographic location, occupation, industry, etc.) For grouping households into groups to study exposure, for public data. <i>Details available not tied to data other than earnings and course measures of assets and liabilities.</i>
Financial Sophistication	Measures of households' expectations and subjective probabilities of different scenarios, and responses to tests of understanding of investment choices and consumption smoothing in the markets in which they are operating. <i>Completely lacking in current data sets</i>

they exhibit characteristics correlated with good financial decision making. We would better understand macroeconomic risks if we were always able to observe when credit was increasing to households with low financial sophistication and when households with low financial sophistication were increasing their exposures to macroeconomic risks.

In sum, the needed data is information collected at the household level on the following categories, summarized in table 13.1: *liabilities, assets, earnings, demographics, and financial sophistication (LEADS)*. Table 13.1 also highlights what is missing from current data sets. The two main missing items are (a) the details on assets and liabilities at the household level, and (b) measures of financial sophistication.

The fact that the LEADS data would contain a host of information on household financial positions raises important issues of privacy. It will be necessary to insure the anonymity of households in any summary statistics released to the public or limited data sets released to researchers. I do not address these issues in this chapter, but merely note that these issues are of great import and surmountable. They are important, as the provision of accurate information in part relies on the confidence of the provider that the information will not be misused. That these issues are surmountable

1 is shown by the regulation of the banking sector in which bank regulators
2 have been able to preserve the privacy of confidential bank information.¹⁰
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4 **13.3 How to Collect the LEADS Data**

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6 How could a statistical agency gather the LEADS data? While the details
7 of the system could rival the documentation of the measurement of risk in
8 the traditional banking sector, in short, there are several principles that are
9 necessary to gather this information.

10 First, the collection of data must rely heavily on administrative data.
11 Survey response rates are declining, and administrative data is increasingly
12 detailed, already computerized, and has low rates of error. Even cooperating
13 households with good intentions are likely not to know the details of their
14 financial contracts or holdings. This happens in two ways. The household
15 may simply not know the details of the asset or liability that are available to
16 them, like a household not knowing the aggregate and idiosyncratic risks of
17 the returns on a stock they own, or whether their mortgage gives the lender
18 recourse or not, or the covenants and seniority of a bond they own. Or the
19 household may not have access to this information, as would be the case for
20 a household holding a mutual fund or hedge fund, or having its investment
21 advisor allocating its assets.

22 This information, however, is available through financial institutions, all
23 of which are already (or seemingly will be) covered under the large umbrel-
24 las of financial regulation and reporting. The organizations that sell the as-
25 sets or hold claims on households either understand the details of the pay-
26 ments that must be made in different states of the world or have on file the
27 terms of the mortgage contract, the National Securities Identification Num-
28 ber (or CUSIP) of the security, and so forth. One approach is for the appro-
29 priate regulator to gather reports on all financial holdings by or against a
30 given household by all financial entities.¹¹ The gold standard for information
31 on financial positions—assets and liabilities—is administrative data from
32 the universe of financial institutions, merged by household for a subset of
33 households. The universe of financial institutions would have to include
34 everything from hedge funds to payday lending to limited partnerships.¹²
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38 10. The IRS has also not allowed leaks of confidential tax return documents while allowing
39 researchers to use the data for important social scientific research.

40 11. There are questions of feasibility and privacy issues that this chapter does not address.
41 Presumably one could gather this administrative data with the consent of the surveyed house-
42 hold. Another option would be to require universal reporting by financial institutions for
43 certain types of financial transactions. Household information is gathered by credit bureaus
44 and households wishing to engage in certain financial transactions are also choosing to be
monitored. Tax law requires reporting of all dividends.

12. It is unclear how to handle international holdings, although one could cover a large por-
tion through intermediaries who sell (or lend) from abroad.

1 Even if this is not completely possible, it seems necessary to rely as much as
2 possible on administrative data to avoid significant loss in detail on holdings.

3 Could statistical agencies use private companies that serve households
4 and collect financial information, such as Mint.com? Data from these
5 private money management companies suffers from a serious shortcoming:
6 the sample of the population they cover are not representative of house-
7 holds in the United States. These data cover only households that choose
8 to manage their finances and do so online. Households that use these types
9 of services are not the average household, and in particular they probably
10 display more than average financial sophistication—certainly in the dimen-
11 sion of planning—and so likely have different wealth and incomes and take
12 different financial risks.

13 Second, household demographics are probably most accurately gathered
14 by a household-level survey or possibly even simply gleaned from existing
15 household surveys. Administrative data may provide better coverage, since
16 any household survey will suffer the usual problems of surveys, including
17 potentially low response rates. But much of the basic demographic and even
18 income data already exists for many households in several extant data sets.
19 Using the various surveys of households conducted by the Census Bureau,
20 and combining data across existing data sets within the US government at
21 the household level could yield an accurate picture of household demo-
22 graphics and income. With household permission, the government agency
23 could merge this extant survey data with the financial information provided
24 by the financial sector to create a close-to-ideal data set. A further improve-
25 ment would be to merge with the income and tax return data in the Internal
26 Revenue Service.¹³ There are, of course, many hurdles to these coordinated
27 efforts, including issues of consent, issues of biasing responses and data
28 provision, and issues of interference with the primary missions of the origi-
29 nal data sources.

30 Third, financial sophistication can be measured using a combination of
31 data from financial institutions and survey methods. Financial institutions
32 have information on household choices and responses to financial offers.
33 These responses have been used by economists to measure financial sophis-
34 tication in behavior from dominated choices (see Agarwal et al. 2009, 2010).
35 Detailed holding can also be used to ask to what extent observed behavior
36 conforms to an economic model's views of what optimal behavior ought to
37 be (see, for example, Calvet, Campbell, and Sodini 2007, 2009). In either case,
38 financial institutions have some existing information on the quality of finan-
39 cial management within a given household. However, this information is
40 unlikely to prove sufficient, and the collection of this type of information
41 may distort the incentives for financial institutions to create the information.

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13. While historically these data have been very difficult to use for confidentiality reasons,
the data at least could be used to construct group-level statistics available outside the IRS.

1 Complementarily, many dimensions of financial sophistication can be
2 measured by survey methods, as shown by existing surveys. A notable ex-
3 ample is the Financial Industry Regulatory Authority (FINRA)'s National
4 Financial Capability Study (2009). At this point, however, there is insuffi-
5 cient previous research on the usefulness of different dimensions of sophis-
6 tication to know exactly which questions or information will be most useful.
7 And to some extent flexibility must be maintained in data collection so that
8 measures can evolve with the financial situation and the risks perceived as
9 potentially most interesting.

10 Despite these caveats to measurement of financial sophistication, three
11 types of data are potentially useful indicators of suboptimal responses of
12 household spending or mispricing of claims on the household sector. The
13 first type of information is measures of household contingency plans for
14 macroeconomic events of interest.¹⁴ This information would be directly
15 useful for understanding default on credit instruments and indirectly for
16 modeling whether expectations of actions are inconsistent across house-
17 holds (and potentially financial institutions). The information that everyone
18 plans to run for the same exit in the event of a fire is useful information for
19 households themselves.

20 The second class of useful information concerns whether households
21 understand the financial products they are using. Lack of understanding
22 would suggest suboptimal exposure to macroeconomic risk factors, and
23 potential exploitation leading to increased aggregate exposures. Of course
24 this is far from proof, as illustrated by the analogy of the pool player who
25 plays well but does not understand classical mechanics.¹⁵ But there is strong
26 evidence that measures of this type of financial sophistication are correlated
27 with financial choices.¹⁶

28 Note that the motivation to measure macroeconomic risk exposure is
29 distinct from the motive to protect investors; that is, the goal of the newly
30 legislated Consumer Financial Protection Agency. While the methods may
31 overlap and the data may be of interest for both purposes, the measurement
32 of aggregate risk exposures requires a focus on common misunderstandings
33 that align with measured exposures.

34 Finally, the third dimension of sophistication to measure is general abili-
35 ties and behaviors related to good financial decision making. For example,
36 saving for retirement is much larger among households that report that they
37 plan for vacations (Ameriks, Caplin, and Leahy 2003). While it is probably
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40 14. Brunnermeier, Gorton, and Krishnamurthy (2012) provide a modeling strategy for finan-
41 cial institutions that makes use of these type of questions of financial institutions.

42 15. Note that an important aspect of the analogy is inconsistent with our modeling of agents
43 as pool players. There is no uncertainty in pool. If an *economist's* agent played pool, in theory,
44 the agent would always win by sinking all the balls in the right order on the first shot.

16. See, for example, Lusardi and Mitchell (2007).

1 not realistic to measure IQ for these purposes, IQ does correlate with finan-
2 cial decision making (Grinblatt, Keloharju, and Linnainmaa 2011).

3 Turning to the structure of the data, for accurate measures of systemic
4 risk, it will generally be necessary that all data be in the same data set—that
5 is, that one has all LEADS information for a set of households. Short of this,
6 if groups of interest can be defined based on observables that are measured
7 in two different data sets, then the group averages can be constructed for
8 each group in each data set and combined. For example, if assets, liabili-
9 ties, and income are well measured in one data set with demographics, and
10 financial sophistication is well measured in another data set with demo-
11 graphics, then one can calculate statistics about financial sophistication
12 for any demographic group. What one cannot do is observe if, within any
13 demographic group, it is the less sophisticated households that are holding
14 particular assets held by that group and not other assets. With some loss of
15 accuracy, one can extrapolate from a small sample with complete informa-
16 tion to a larger sample with information on different dimensions for different
17 households.

18 Finally, to what extent is repeated cross-sectional data sufficient, or would
19 the LEADS system be significantly stronger with panel data that follows
20 the same households over time? The risks of assets and liabilities require
21 that one track the performance of these assets, and not that one track the
22 same households over time. To measure income dynamics and correlations
23 with macroeconomic risks requires repeated measures of a household’s
24 income, but there are data sets from which measures could be constructed
25 with long time series already. The Panel Study of Income Dynamics, the
26 Current Population Survey, and the Social Security Earnings Records in
27 the Health and Retirement Study all provide long histories of earnings on
28 households that could be mapped to demographics and then applied to
29 households in the new data set. I expect that the heterogeneity within a
30 demographic group missed by this method would be of little systemic inter-
31 est. That said, repeated cross-sectional data will reduce the power of many
32 measures, particularly when studying household-specific changes to better
33 understand behavior or when tracking the impact of changes over longer
34 periods of time.

35 36 **13.4 Outline of LEADS Data Analysis and Dissemination**

37
38 The LEADS system is designed to allow the measurement of the exposure
39 of the liquid wealth and lifetime wealth of different groups of households
40 to different macroeconomic shocks and the construction of measures of
41 the exposure of both aggregate consumption demand and claims on the
42 household sector to these macroeconomic shocks. This section first out-
43 lines a framework for analysis of three steps: (a) measure the risk exposure
44 of each asset, each liability, and each income stream to macroeconomic

1 risk factors; (b) aggregate exposures to household-level liquid wealth and
2 lifetime wealth and then aggregate exposures to groups of households; and
3 (c) model the exposures of aggregate consumption demand, claims on the
4 household sector and government liabilities. Second, dissemination of both
5 analysis and anonymous raw data is critical. While regulators may find pat-
6 terns of exposure informative, the private sector can also better respond to
7 and price risks when it is better informed of their aggregate consequences, as
8 would be the case if it had access to both the analyses and some of the data.

9 To begin, the measurement of the risk exposure of assets and liabilities
10 can be based on textbook asset pricing and its application, which necessar-
11 ily involves the difficulties of the real world. This part of the analysis of the
12 LEADS system is not novel and is reasonably well understood by academics
13 and practitioners. I simply propose standard modeling of asset and debt
14 cash flows that makes up the bulk of quantitative finance and fundamental
15 analysis.¹⁷ Returns, cash flows, and prices, are described as a sum of expo-
16 sures (betas) times realizations of aggregate risk factors and idiosyncratic
17 or deterministic components. Modeling income risk is similarly reasonably
18 well understood and applied in labor economics. Here “understood” does
19 not mean straightforward. Perhaps the most important assumptions are
20 those about the behavioral responses over the period studied in cases where
21 these responses affect cash flows, which as noted subsequently, may depend
22 on other holdings of a given household.

23 Given estimated exposures, the liquid wealth of a household can be writ-
24 ten as a sum of stochastic cash flows into and out of liquid assets from
25 income, assets, and liabilities, and of terminal prices of liquid assets and
26 liabilities. And the lifetime wealth of a household can be written similarly,
27 but for all assets (not just liquid assets). Thus, we would have measures of
28 the exposure of these two concepts of wealth to variation in aggregate risk
29 factors. Finally, aggregating across households, one calculates the implica-
30 tions of a change in any set of aggregate factors for the liquid wealth and
31 lifetime wealth for any group of households.

32 The third step in analysis is to use a model of the consumption sensitivity
33 of different households to these two types of wealth to measure the expo-
34 sure to aggregate risk factors of the aggregate demand for consumption. As
35 discussed shortly, this requires modeling household behavior. But having
36 modeled household saving, consumption, default, and portfolio behavior,
37 the LEADS system provides a measure of the exposure to default of any set
38 of claims on the household sector.¹⁸ Finally, one can examine under what
39

40
41 17. One danger going forward is that new financial products have limited histories and so are
42 potentially the most subject to misestimation.

43 18. Similarly, given decision rules about portfolio choice, one can evaluate the change in asset
44 demand from the household sector. With assumptions about the potential other buyers of the
asset, one can check the aggregate risk factor exposures are reasonable or reasonably consistent
with the household responses.

1 scenarios there are significant direct or possible effects on the government
2 budget through explicit or implicit guarantees or legislative responses.

3 As noted, the modeling of household behavior is a critical step in the con-
4 struction of both income and asset/debt outcomes as well as household con-
5 sumption, saving and rebalancing in response to events. In short, we need to
6 model how households respond to changes over any horizon considered. In
7 terms of risks to the household sector, most relevant are exposures of house-
8 hold income, where cash flows depend on household labor supply responses,
9 and exposures of household debt, where the household can choose to exer-
10 cise an option to default or change portfolio or saving behavior. In terms
11 of outcomes, the LEADS analysis needs to model household consumption
12 behavior to understand how the demand of different households are more
13 and less exposed to macroeconomic events. Further, to understand claims
14 on households, one has to make assumptions that determine the situations
15 in which households default. Fortunately, many of these behaviors have
16 been or can be measured from past experiences and existing data, so that
17 modeling assumptions can be disciplined by data. Further, one advantage
18 of LEADS data is that, because it represents a significant increase in the
19 quality of available information, it will increase the ability to learn about
20 these behaviors from future events. Nevertheless, given the importance of
21 these assumptions, any analysis will have to carefully evaluate robustness to
22 alternative assumptions.

23 Household financial sophistication may be quite useful in modeling
24 household behavior. Household decision rules may differ importantly
25 with financial sophistication. The second use of financial sophistication is
26 in grouping households for analysis, in combination with information on
27 exposures. Where exposures—to or from the household sector—are large
28 and sophistication low, this is suggestive that macroeconomic risks may be
29 mispriced or amplified by lack of sophistication or information.

30 In sum, the proposed analysis produces information on the exposure of
31 aggregate demand to various risks, information helpful for the private sec-
32 tor and the government. The analysis measures the size of risks emanating
33 from the household sector, and allows them to be evaluated and tracked into
34 other sectors of the economy. Finally, the analysis contains information on
35 the potential costs to the government in terms of likely payouts through
36 automatic means-tested programs and possible payouts through ex post
37 bailouts to subgroups of the household sector.

38 Finally, the LEADS information could be made public, in three ways.
39 First, the results of the analysis—the range of possible impacts on aggregate
40 demand and defaults and asset prices for many aggregate scenarios under
41 any assumptions—could be made public. This dissemination can lead to
42 many benefits, as discussed in section 13.1 and the appendix. Further, if
43 regulators were concerned about the possibility of “unsophisticated” house-
44 holds unintentionally holding too much risk, then particular results about

1 riskiness could be pushed to the type of household that might most benefit
2 from this information. Similarly, if regulators were concerned about the
3 exposures of some market participants, they again could highlight the risks
4 and allow participants to react to the information as they see fit.

5 Second, a data set of aggregated data could be made publically available
6 for analysis and investigation by academics and investors. Ideally this data set
7 would combine analysis and raw data. A large number of government agencies
8 produce detailed tables based on data that they collect through surveys.
9 A reasonable model for the LEADS summary data is the published tables
10 based on the Survey of Consumer Finances (SCF). But the LEADS tables
11 would be better if summary statistics were released by demographic groups
12 of possible interest including financial sophistication, and if the focus were
13 not on current value of holdings but on asset detail and riskiness. Further, it
14 would be useful to observe not just mean holdings, but covariances of hold-
15 ings across assets within a group and quantiles of holdings—both of which
16 would allow a better understanding of the possibility that large shocks lead
17 to large movements for some households in the group. Finally, to be most
18 useful, the statistics should, to the extent possible, convey information about
19 exposure to both liquid wealth and lifetime wealth.

20 Third, a household-level data set could be made accessible to researchers
21 working to improve our understanding and modeling of macroeconomic
22 risks and financial stability. In keeping with standard survey data protocols,
23 this data set can omit enough detail to ensure confidentiality and anonymity
24 of individual agents. The data set would be useful to bring new evidence on
25 the behavioral assumptions inherent in the extant measures of systematic
26 exposures.

27 28 29 **13.5 Conclusion**

30 Macroeconomic shocks can lead to large changes in demand from the
31 household sector and can lead to large changes in cash flows from debt
32 claims on the sector. The LEADS system outlined in this chapter is designed
33 to measure sufficient detail on household liabilities, assets, earnings, demo-
34 graphics, and financial sophistication, to project these large changes.

35 This LEADS system would help households, firms, and policymakers
36 determine what sources of aggregate risk are most pertinent for claims on
37 the household sector and household demand, and when these exposures
38 are likely to be large. More specifically, the system is designed to allow mea-
39 surement of (a) the distributional impacts on both the distribution of liquid
40 wealth and lifetime wealth, and the resultant changes in household demand;
41 (b) the effects of balance sheet adjustments on the prices and payouts of
42 claims on the household sector held by other sectors; and (c) the resulting
43 impact on the revenues and liabilities of the government, through possible
44 transfers and de jure and de facto guarantees to the household sector. While

1 the scope of the LEADS system implies that this chapter can only provide
2 an overview of the system, methodologies exist for most of the individual
3 component tasks.

4 While it is beyond the scope of this chapter to lay out a complete frame-
5 work for macroprudential data analysis and regulation of the household
6 sector, the analogy between banks and households suggests that policy-
7 makers want to consider three tools. First, policymakers might consider
8 capital requirements, such as existed in mortgage markets in practice in the
9 period of the conforming mortgage. A related system is the Social Secu-
10 rity system, which guarantees/imposes a basic standard of living for elderly
11 households by requiring young households to pay while working for health
12 insurance and basic retirement income when elderly. Second, policymakers
13 might consider restrictions on what financial assets households can use. This
14 regulation is in place as restrictions on what investors can invest in unless
15 they are qualified investors. It seems suboptimal to make this determination
16 based on wealth—errors are made both by excluding sophisticated investors
17 from markets and by allowing wealthy unsophisticated investors to invest
18 in any assets. A better regulation might license an investor on the basis of a
19 test, like what is required to get a driver’s license to drive. Finally, the regula-
20 tor may want to inform the public and warn them about specific exposures,
21 or even lean against the wind and try to change prices, limit access, or tax
22 certain investments or strategies that they see as destabilizing. A good data
23 measurement system is an essential guide to evaluate the benefits and, just
24 as important, the costs of any potential active regulation.
25
26
27

28 Appendix

29 *Why the Government has a Role in* 30 *Monitoring Household Systemic Exposures*

31
32
33 There are two benefits to measuring the exposure of the household sector
34 to systemic risk.

35 First, as a general principle, in markets the social value of information
36 exceeds its private value. Because information has some of the features of a
37 public good, markets tend to lead to the production of too little information.
38 This result comes from the power of markets to create efficient allocations
39 given the available information. Prices aggregate and convey information
40 to market participants, and optimizing agents can coordinate behavior to
41 efficiently produce and allocate goods. However, this benefit implies that
42 when an individual produces information, others benefit. Thus the social
43 value of information exceeds its private value.

44 When there is too little information, mistakes are made relative to the

1 economy with the socially optimal amount of information. Mistakes will
2 tend to be based on the common information that exists, so that misestima-
3 tion leads to coordinated mistakes that are by definition macroeconomic if
4 not systemic. Misestimation propagates; misestimation of the positions of
5 the household sector can lead to systemic risk elsewhere.

6 This externality implies that there is a role for a governmental or quasi-
7 governmental agency to produce information. This is not a new role. The gov-
8 ernment does this for a large number of macroeconomic variables. The govern-
9 ment also regulates the production and disclosure of information by firms.
10 The goal of this chapter—and this NBER project—is to modernize data
11 production and provision, updating it to account for the modern financial
12 landscape and redirecting some of its focus from the measurement of means
13 to the measurement of standard deviations—risks, and macroeconomic and
14 systemic risks in particular.

15 There are many potential benefits of more information. Market partici-
16 pants would be able to allocate resources more efficiently across potential
17 outcomes given better information on exposures to aggregate risks in places
18 in the household sector. Better measurement of the risk exposure of con-
19 sumer demand and claims on household resources would also allow the
20 government and the market to better predict consumption demand, and so
21 better understand risks given their observed prices.

22 The second reason that the government has a role in monitoring house-
23 hold exposures is that some households lack financial sophistication, due, for
24 example, to a lack of financial knowledge, limited information-processing
25 abilities, and limited time allocated to financial decisions. Thus, some house-
26 holds can make significant financial mistakes relative to the decisions they
27 would make if they had the ability and took the time to collect and process
28 fully all available information and knowledge. The provision of information
29 about the holdings of agents with different levels of sophistication can make
30 behavior closer to that of the full-information or full-sophistication econ-
31 omy, which can increase the information content of prices, and lead to better
32 private decision making and better management of systematic exposures.

33 When the mistakes of market participants are common, they can lead to
34 noise or bias in prices that other agents are using as sources of information
35 for their choices, and may lead to large a systemic exposure of both house-
36 hold demand and claims on households.¹⁹ In financial markets, financial
37 institutions that all expect to be able to sell the same assets or all draw on
38 credit but fail to recognize the extent to which other institutions have made
39 the same plans create a systematic risk. For households, this misestimation
40 can create a similar dynamic or exposure. In the recent crisis, households
41

42 19. A typical argument is that agents do not make similar mistakes over many time periods,
43 not that many agents do not make similar mistakes at the same time.
44

1 that had planned on using home equity to stabilize consumption across bad
2 shocks found that credit was available only at very high prices and that their
3 collateral values had declined, further limiting the extent of borrowing.

4 Further, when a sophisticated/informed investor observes a price, they
5 must try to infer the information in this price. This inference can involve esti-
6 mating the shares of price movement caused by agents with different levels
7 of motivation and sophistication trading in the market. And so the infer-
8 ence could be improved by information on these shares—whether observed
9 prices are due to the actions of other sophisticated/informed investors or
10 instead the actions of unsophisticated investors or “noise traders,” whose
11 choices are not based on as much information or are motivated by different
12 concerns. For example, a large exposure among a group of households that
13 are (in some sense) making unsophisticated choices can distort prices and
14 pass systematic risk through the household sector into other parts of the
15 economy through default on claims on the household sector.

16 Public provision of information about the sophistication of groups of
17 agents who are holding different assets can partly reveal how informa-
18 tive market prices are, and again can lead to, better private sector decision
19 making.²⁰

20 Finally, while it is efficient to have the downside economic risk held by
21 those most willing to hold it, it is generically inefficient to have it held by
22 those least able to understand it or who most underestimate it. Lack of
23 sophistication can be exploited and there is always the risk that this exploita-
24 tion can cause systemic exposure.²¹ A signal of the size of systemic risk is its
25 price. And regulators interested in regulating systemic risks need to be able to
26 observe clues as to whether the risk is correctly priced and correctly placed.

27 Thus, the LEADS data can be informative about whether downside eco-
28 nomic risk is being shifted to the unsophisticated part of the household
29 sector, or passed through the unsophisticated part of the household sector
30 to other systemically important sectors. This argument implies that data
31 collection should support these types of inference and should be designed to
32 provide the information necessary for market participants and good regula-
33 tory responses. There may be situations where aggregate risk can be reduced
34 by some consumer financial protection, and where regulation of financial
35

36
37 20. This information would allow lenders to these households to observe their sophistica-
38 tion, which can be useful as a signal of the quality of their financial decision making in other
39 areas. Thus, an investor with insufficiently detailed information (or sophistication) to evaluate
40 the exposure of the household balance sheet might still learn from the disclosed abilities of
41 households that are holding certain assets.

42 21. While not a group of households, one can make the argument that AIG financial prod-
43 ucts, by mispricing credit default swaps on mortgage-backed securities, lead market partici-
44 pants to underestimate the exposure of these assets to aggregate risk and lead to its being held
in systemically important places. Unsophisticated households could play a similar role in the
next crisis.

choices may avoid “unintentional” systemic exposures.²² Even without government intervention, information on sophistication is likely to be useful for targeting the provision of information about systemic exposures (or the provision of the implications of analysis of systemic exposures), with the goal of allowing household and firm choices that are more consistent with those that they would have made given complete information.

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22. The field of economics has a long history of taking individual choices as optimal and basing policy advice on arguments that rely not on individual mistakes but on market failures due to externalities, for example. Both of the previous arguments—that markets naturally produce too little information and so there is a government role in the provision of information—are in this spirit. However, the field tends to ignore the possibility that agents differ in sophistication and sophisticated agents can regulate the behavior of the unsophisticated and improve welfare. The general commitment to the sovereignty of the decision maker is the real divide between behavioral economics and the rest of the field. In cases where sophistication is about education and information, in theory, revealed-preference tests and utility constructs could lead to welfare measures that proscribe such regulations as welfare improving in the standard approach. Obviously, we are some distance from this point.

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