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THE LUXEMBOURG HOUSEHOLD FINANCE AND CONSUMPTION SURVEY (LU-HFCS): INTRODUCTION AND RESULTS

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The Luxembourg Household Finance and Consumption Survey (LU-HFCS): Introduction and Results*

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Abstract:

This paper introduces the Luxembourg Household Finance and Consumption Survey (LU-HFCS), presents its background and aim, the field phase, the data treatment, including editing, imputation, and anonymisation, and some basic descriptive findings. The estimated average (median) total net wealth of Luxembourg households is about €733,000 (€403,000) - an extremely high value by international standards, which is likely to be driven by high incomes and correspondingly high prices of real estate in Luxembourg. Average total gross income and total net wealth generally increase with household size, with age up to retirement age and with education. On average, total gross income of Luxembourg national and foreign households do not seem to differ significantly. This cannot be said for total net wealth and net real wealth, which, on average, is more than double as high for Luxembourg households than for foreign households.

Keywords: Household, survey, editing, multiple imputation, wealth, income
JEL Codes: D31, D14, C81, C83

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Résumé non-technique

Les banques centrales de l'Eurosystème ont mis en place l'enquête HFCS¹ afin de collecter des données micro-économiques sur le comportement financier et de consommation des ménages. De telles informations harmonisées à travers les différents pays ne sont pas disponibles par le truchement des sources statistiques existantes. Pourtant, elles présentent une utilité certaine dans plusieurs domaines de compétence des banques centrales, étant donné que les évolutions des agrégats économiques sont déterminées en partie par la distribution, parmi les ménages, de la richesse ou de la dette. Les décisions de consommation et d'épargne dépendent d'ailleurs également de facteurs propres aux ménages individuels (composition, âge, éducation, etc.). Ainsi, par le passé plusieurs banques centrales ont établi des enquêtes régulières auprès des ménages (Federal Reserve Board, Banca d'Italia, Banco de España). Le projet HFCS vise à harmoniser cette pratique et à l'étendre à l'ensemble des pays membres de la zone euro.

Dans le cadre de ce projet, la Banque centrale du Luxembourg en collaboration avec le CEPS/Insee a conduit deux enquêtes, entre le dernier trimestre 2010 et le premier trimestre 2011. Il s'agit d'une part du volet luxembourgeois de l'enquête HFCS – orienté vers les ménages résidents – et d'autre part d'une enquête complémentaire orientée vers les ménages non-résidents dont au moins un membre est employé au Luxembourg. La deuxième enquête reflète l'importance des frontaliers pour l'emploi, la consommation et la production au Luxembourg.

Ce cahier se focalise sur l'enquête auprès des ménages résidents, présentant son contexte et ses objectifs, détaillant la collecte et le traitement des données et résumant certains résultats préliminaires concernant le comportement financier.

Le questionnaire, l'échantillonnage et la collecte des données

Le questionnaire HFCS se compose de deux parties: la première concerne le ménage dans son ensemble et la deuxième s'adresse individuellement à certains membres du ménage. Les questions s'adressant au ménage concernent les actifs réels et leur financement, les autres passifs et contraintes de crédit, les entreprises privées, les actifs financiers, les cadeaux/transferts intergénérationnels et la décision de consommation/épargne. Les questions orientées vers les individus concernent les caractéristiques individuelles (âge, sexe, situation familiale), l'emploi, les droits à la retraite et les autres revenus du travail (les autres sources de revenus sont couvertes au niveau du ménage). Dans l'ensemble, plus de 400 questions différentes sont proposées, dont la plupart proviennent de l'enquête Eurosystème HFCS.

En principe, l'échantillon de l'enquête devrait cibler l'ensemble de la population des ménages résidant au Luxembourg. Cependant, les données du recensement de la

¹ Household finance and consumption survey.

population étaient désuètes au moment de la collecte des données. En conséquence, l'enquête s'est basée sur un échantillon aléatoire stratifié extrait du registre de l'Inspection Générale de la Sécurité Sociale (en date du 31 décembre 2009). Ce registre couvre la quasi totalité de la population résidente (à l'exception des fonctionnaires européens et les résidents de certaines institutions).

Afin de réduire la variance des estimateurs, l'échantillon est stratifié selon la nationalité, l'occupation et le revenu individuel. En accord avec les objectifs de l'enquête, la catégorie de la population présentant une richesse plus élevée a été surreprésentée dans l'échantillon. Il s'agit de prendre en compte la plus grande variabilité de la composition du portefeuille parmi ces ménages et leur contribution plus importante aux agrégats relatifs au secteur des ménages.

L'enquête auprès des ménages résidents au Luxembourg a été conduite par le biais d'entretiens personnels assistés par ordinateur (CAPI)². Les modalités de l'enquête ont été communiquées aux ménages sélectionnés par différents moyens (lettre d'introduction, dépliants d'information, page internet et numéro téléphonique dédié). Un effort considérable a été consacré à la préparation du questionnaire dans les différentes langues véhiculaires au Luxembourg. La collecte des données a commencé fin septembre 2010 et s'est terminée en avril 2011. En moyenne, un entretien a duré 56 minutes (avec une médiane de 51 minutes). Partant de l'échantillon initial de 5,000 ménages résidents extrait du registre IGSS, des entretiens ont eu lieu avec 950 ménages, représentant 2,540 individus. La qualité des réponses fournies est jugée bonne au vu de différentes indications fournies par les enquêteurs (fiabilité des données sur le revenu et la richesse, nombre de documents consultés, intérêt pour l'enquête, compréhension des questions).

Traitement des données

Il est bien connu que des enquêtes aussi complexes sur des sujets particulièrement sensibles conduisent à des incohérences logiques entre les réponses à différentes questions et à des non-réponses à certaines questions. C'est pourquoi les données doivent faire l'objet d'un traitement spécifique avant toute analyse empirique. Dans un premier temps, les données sont éditées afin d'identifier et de corriger d'éventuelles erreurs (incohérences logiques). Ces corrections se basent aussi sur les commentaires de l'enquêteur enregistrés après l'entretien, sur des incohérences institutionnelles, des incohérences logiques entre les réponses à différentes questions ou des erreurs de frappe.

Dans un deuxième temps, la non-réponse à des questions individuelles est traitée. Sans un traitement approprié, la non-réponse peut avoir des conséquences graves, générant un biais des estimateurs et limitant la taille de l'échantillon et donc l'efficacité des estimateurs. La procédure de l'imputation sert à remplacer les non-

² Computer-assisted personal interview.

réponses par des valeurs plausibles. La BCL a eu recours à €MIR³, une suite de routines pour l'imputation multiple, mise à sa disposition par l'Eurosysteme.

Enfin, les données éditées et imputées sont pondérées pour fournir un échantillon représentatif de la population de ménages résidents au Luxembourg (dont au moins un membre était enregistré à l'IGSS fin 2009). La procédure de pondération repose sur les étapes suivantes : ajustement pour la probabilité de sélection (en tenant compte de la surreprésentation des ménages plus favorisés), ajustement pour la non-réponse à des questions individuelles, ajustement selon des sources statistiques externes afin d'approximer au mieux la distribution des ménages et des individus dans la population cible.

Résultats

Pour le ménage résident représentatif (moyenne pondérée des ménages interrogés) le revenu brut annuel est estimé à €83,600. La plus grande part de ce revenu brut provient des salaires (63%), tandis que 7% provient du revenu du travail indépendant et 18% des pensions. Sur l'ensemble des ménages résidents, seulement 71% comptent des salariés et 10% des travailleurs indépendants. Le revenu salarial moyen (par ménage dont au moins un membre est salarié) est de €73,700. Parmi les ménages dont au moins un membre est travailleur indépendant, le revenu moyen émanant de ce travail est de €58,900. Ce contraste s'explique en partie par le fait que bon nombre de ménages combinent un revenu salarial avec un travail indépendant qui génère un revenu plus limité. Ainsi, en se limitant aux ménages dont la personne de référence se déclare travailleur indépendant, cette source de revenu représente €150,000 pour le ménage représentatif.

Environ 35% des ménages résidents reçoivent au moins une pension pour une valeur moyenne de €42,800 par an. Les transferts sociaux concernent 41% des ménages et leur valeur moyenne est estimée à €8,600 par an. Les revenus des investissements immobiliers sont perçus par seulement 13% des ménages, avec une valeur moyenne de €24,400 par an. Les investissements financiers concernent pratiquement 50% des ménages et génèrent un revenu annuel moyen de €2,100.

Pour le ménage représentatif, la richesse nette (après déduction des dettes immobilières et autres) est de €733,000 (médiane €403,000). Cette valeur est très élevée en comparaison internationale. Aux Etats-Unis, la richesse nette du ménage moyen se situe à €425,000 (Bricker et al., 2011), en France à €229,300 (Chaput et al., 2011) et en Allemagne à €140,000 (Frick et al., 2010).

La richesse brute se compose à 11% d'actifs financiers et à 89% d'actifs réels (biens immobiliers, véhicules, etc.). Le ménage représentatif détient €723,000 d'actifs réels et €88,400 d'actifs financiers, pour une richesse brute totale de l'ordre de € 811 000. La résidence principale correspond à 59% des actifs réels du ménage représentatif

³ ECB Multiple Imputation Routines.

(€ 430,000). D'autres biens immobiliers représentent 30% des actifs réels (€ 217,000), tandis que les véhicules et autres objets de valeur ne constituent qu'une fraction de ces actifs. Du côté du passif, les dettes hypothécaires s'élèvent à € 74,000 en moyenne. La dette totale est estimée à € 78,400 pour le ménage représentatif.

On estime que 67% des ménages luxembourgeois sont propriétaires de leur résidence principale (au moins en partie). Sa valeur moyenne est estimée à €640,000 (médiane €500,000). Pratiquement 30% des ménages sont propriétaires d'autres biens immobiliers, d'une valeur moyenne de €770,000 (médiane €300,000). Seuls 6% des ménages possèdent des parts d'entreprise, avec une valeur moyenne de € 760,000 (médiane €190,000). Par contre, 87% des ménages sont propriétaires d'au moins un véhicule, d'une valeur moyenne de €24,000. D'autres objets de valeur sont détenus par 24% des ménages, la valeur moyenne correspondante s'établissant à €39,000.

Le ménage représentatif détient par ailleurs 32% de ses actifs financiers sur un compte d'épargne, 21% dans un organisme de placement collectif, 19% dans le cadre d'un plan de pension privé et 11% sur un compte courant. En moyenne, les ménages luxembourgeois possèdent près de €29,000 sur leur compte d'épargne, €18,000 en parts d'OPC, €17,000 en plan de pension et €10,000 en compte courant. D'autres catégories d'actifs financiers sont moins importantes.

Presque chaque ménage possède un compte courant, mais seuls 75% ont un compte d'épargne et 19% des parts d'OPC. Si on se limite aux ménages qui ont un compte d'épargne, leur montant moyen est de presque €40,000 (médiane €15,000). En se limitant aux ménages qui ont des parts d'OPC, l'investissement moyen est de €95,000 (médiane €27,000). Pour les 4% des ménages qui ont acheté des obligations, leur valeur moyenne est de €120,000 (médiane € 45,000) et pour les 10% qui ont acheté des actions, leur valeur moyenne est de €60,000 (médiane €10,000). Un tiers des ménages souscrit à un plan de pension privé, avec une valeur moyenne de €50,000 (médiane €28,000).

Les caractéristiques spécifiques des différents ménages peuvent expliquer le niveau et la composition de leur richesse nette. En passant des ménages composés d'une seule personne aux ménages composés de deux personnes, on observe une forte augmentation de la richesse totale et de la richesse réelle. Cependant, le passage des ménages à deux personnes aux ménages à trois ou à quatre personnes n'exerce pratiquement aucun effet. Le niveau de richesse augmente à nouveau si les ménages sont composés de plus de quatre personnes.

Le revenu brut, la richesse nette et la richesse réelle augmentent nettement avec l'âge (jusqu'à la retraite) et avec le niveau d'éducation. Par contre, ces variables ont moins d'impact sur la richesse financière. Par exemple, pour les ménages dont la personne de référence ne dispose pas d'un niveau d'éducation primaire, le niveau de richesse nette est de l'ordre de €400,000. Celle-ci augmente jusqu'à €1,075,000 pour les ménages dont la personne de référence présente le niveau d'éducation le plus élevé.

Les ménages dont la personne de référence a la nationalité luxembourgeoise ont un niveau moyen de richesse nette (€939,357) et de richesse réelle (€841,982) constituant pratiquement le double des ménages dont la personne de référence est de nationalité étrangère (respectivement €394,570 et €333,111 en moyenne). Ces différences sont statistiquement significatives. Par contre, en termes de revenu brut, la différence n'est pas significative (€86,499 en moyenne pour le ménage luxembourgeois et €78,963 pour le ménage étranger). Ceci pourrait s'expliquer par les héritages, qui sont vraisemblablement plus élevés pour les ménages luxembourgeois.

Le lien entre le niveau de revenu et le niveau de richesse est bien visible. La richesse nette augmente de façon régulière en passant du premier au deuxième quartile de la distribution des revenus et ensuite du deuxième au troisième quartile. Par contre, une augmentation beaucoup plus importante de la richesse nette se manifeste en passant du troisième au quatrième quartile de la distribution des revenus.

Conclusions

Ce cahier fournit un premier aperçu de l'enquête sur le comportement financier et de consommation des ménages au Luxembourg. Par rapport au revenu et la richesse, les résultats sont cohérents avec les données agrégées quant à la position du Luxembourg en comparaison internationale (OECD, 2011). La valeur ajoutée de cette enquête provient des informations micro-économiques qui permettent de mettre en relation le revenu et la richesse par rapport à des données propres à chaque ménage. De plus, elle permet d'analyser la distribution du revenu et de la richesse à travers la population des ménages. Enfin, la nature harmonisée de l'enquête permettra une comparaison avec la distribution des revenus et de la richesse dans d'autres pays de la zone euro.

1 Introduction

Between the fourth quarter of 2010 and the first quarter of 2011, the Banque centrale du Luxembourg together with CEPS/INSTEAD conducted two household finance and consumption surveys (HFCS). The first survey was targeted at households residing in Luxembourg (LU-HFCS) whereas the second survey was targeted at households residing abroad where at least one member works in Luxembourg, i.e. so called cross-border working households (XB-HFCS). The objective with these two surveys is to obtain representative information on households' wealth and income structure. The resident survey was conducted in close correspondence with similar surveys across Eurosystem member states, initiated by the Eurosystem Household Finance and Consumption Network (HFCN). The cross-border survey was conducted as a complement to the resident survey to take into consideration the importance of cross-border commuters for Luxembourg employment, consumption and output.

This paper introduces the resident HFC survey, presents its background and aim, the field phase, the data treatment and some findings. Section 2 outlines the objectives of the HFCS; section 3 displays the questionnaire and sample design; section 4 discusses the data treatment, including editing, imputation etc...; section 5 shows sample and population characteristics; section 6 presents first results on the income and wealth of Luxembourg resident households; section 7 compares the survey results with results for aggregate sources; finally, section 8 concludes.

2 Objectives

The LU-HFCS is part of a wider Eurosystem HFCS, in which all 17 Eurosystem national central banks (NCBs) together with the European Central Bank (ECB) and some National Statistics Institutes (NSIs) take part. The Eurosystem HFCS was initiated to provide the Eurosystem with micro-level data on euro area households' finances and permit studying in detail their financial and economic behaviour.⁴ Such ex ante harmonized micro data on these topics are not available from other data sources or statistics. As described in detail by the HFCN (2008) the analysis of household behaviour is relevant as most important financial and economic decisions are taken at the household level (as compared to the level of a single individual). As dynamics of economic aggregates are determined not only by macroeconomic variables, but also by household-specific factors (household consumption, savings and investments), detailed information of the latter can provide valuable input into a number of policy areas of the Eurosystem, in particular as these factors appear to be driven to a large extent by expectations, uncertainty with respect to future individual income as well as demographic and social characteristics, which remain hidden in aggregate statistics.

⁴ For more information please consult the documentation on the HFCN available at the ECB website: http://www.ecb.int/home/html/researcher_hfcn.en.html.

Household finance and consumption surveys, or other wealth surveys, have been conducted by several central banks in the past (e.g. the Federal Reserve Board, the Banca d'Italia, the Banco de España). Their research results feed extensively into policy and into their communication with the public. Research results based on these surveys are recurringly consulted for monetary policy analyses and financial stability purposes. For example, the implications of the recent steep increases in household indebtedness in many euro area countries cannot be adequately judged by referral to aggregate data. Information on the households' debt levels and its categories as well as their distribution across various characteristics, such as income, education, and/or age classes obtained from surveys provide crucial information for the assessment of whether or not households are over indebted and whether there exists eventual financial risks associated therewith.

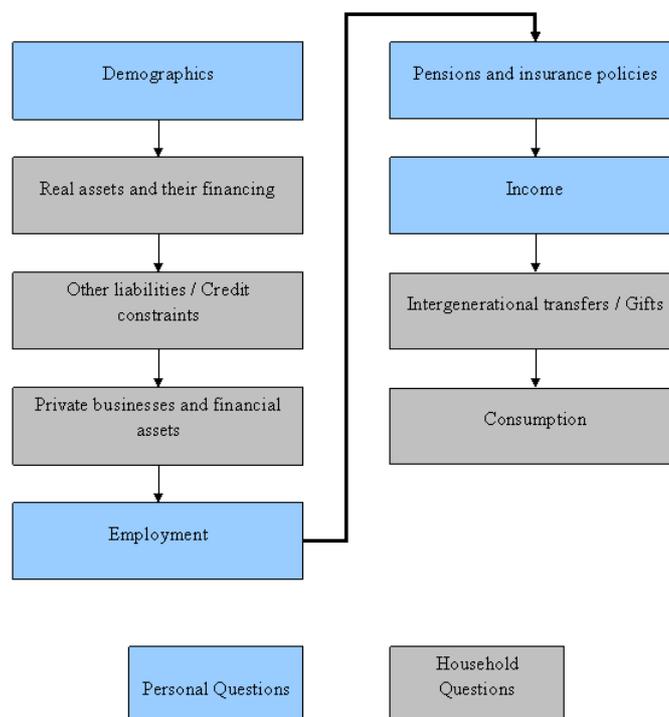
To date the available data did not allow painting a coherent picture on households' finance and consumption behaviour across the euro area. Data was only available for few countries, and methodologies, concepts and questions asked were not ex ante harmonised making adequate inference and aggregation of results a cumbersome and error prone task. Other EU household surveys, such as the European Survey on Income and Living Conditions (EU-SILC) or the Survey of Health, Ageing and Retirement in Europe (SHARE) do exist; however, they have a different focus or target population. The euro area wide HFCS provides Eurosystem policy-makers and researchers with an unprecedented wealth of information on households' financial behaviour and decisions. The Eurosystem HFCS is conducted in a decentralised manner by Eurosystem national central banks and follows an output-oriented approach, which requests participating NCBs to report a set of core output variables. The main aim of the survey is to gather micro-level, structural information on households' assets and liabilities in the euro area, as much as possible in line with aggregates derived from non-financial and financial accounts according to the definitions of ESA95. In addition, the survey collects key information on households' demographic structure, income and consumption expenditures.

3 The questionnaire and sample selection

As shown in Figure 1 and described in detail in the final report of the household finance and consumption network (HFCN 2008) the LU-HFCS questionnaire comprises two main parts: i) questions aimed at the household as such (depicted in grey) and ii) questions aimed at individual household members (depicted in blue). The household level questions include a set of questions on real assets and their financing, other liabilities/credit constraints, private businesses, financial assets, intergenerational transfers/gifts and consumption/savings. The individual level questions concern demographics, employment, future pension entitlements and labour-related income (other income sources being covered at the household level). Together, the questionnaire contains over 400 different questions, the predominant majority of which are common to the Eurosystem HFCS. Other questions are asked in some countries only and few are purely national. For example, due to the relevance of the neighbouring regions for Luxembourg employment and

consumption, in the LU-HFCS, households were asked to provide information on their consumption behaviour in Luxembourg as well as the neighbouring countries.

Figure 1: Structure of LU-HFCS



Source: HFCN (2008)

Prior to the fieldwork, various different versions of the Eurosystem questionnaire had been pre-tested in several countries; the LU-HFCS was pre-tested among a small number of Luxembourg households in July 2010.

In principle, the ideal target population of the LU-HFCS would include all private households resident in Luxembourg at 31st December 2010. In absence of a recent census (at the time of data collection), the Social Security register of Luxembourg (Inspection Générale de la Sécurité Sociale, IGSS) as at 31st December 2009 is used as sampling frame to extract the sample. The IGSS covers approximately all individuals in the Luxembourg resident population with few exceptions.⁵ Table 1 provides an overview of the main characteristics of the sample design for the survey.

To reduce the variance in the estimated wealth components, a stratified sampling procedure, that includes the oversampling of the wealthiest strata in the population, is adopted. Essentially, this means that the sample includes a higher share of households belonging to wealthier strata than actually present in the population.

⁵ The sampling frame does not cover international civil servants and in general individuals not registered in the social security register unless they are part of a household where another member is included in the social security register. In an attempt to integrate international civil servants into the sampling frame a request to the health insurance office of the international civil servants was made but was unsuccessful. CEPS (2011) estimates that 3-5% of the population is not covered.

This is to take account for the higher variability in the portfolio composition of wealthier households. Due to the absence of external statistics on household wealth, the oversampling of wealthy households has been achieved through the oversampling of high income households, assuming a strong link between income and wealth. This is done to take into account the expected higher variability of portfolio composition and higher amounts invested in different assets categories by wealthier households. In this context, households headed by a self-employed individual were also oversampled. To make the sample statistics representative of the underlying population which it represents, the wealthier strata are subsequently down weighted.

Table 1: Sampling characteristics of the LU-HFCS

Sampling frame	IGSS (Inspection Générale de la Sécurité Sociale – Social security register) at 31 December 2009
Reference population	Private household resident in Luxembourg as at 31 December 2010 ^α
Sampling unit	Resident fiscal households
Oversampling of wealthy units	20%
Sample size	950 households
Number of strata	20
Stratification variables	Nationality, employment status, income

Notes: ^α *Excluding institutionalised households, international civil servants and individuals not registered in the social security register in general unless they are part of a household where another member is included in the social security register. The difference in the reference year of the sampling frame (2009) and of the reference population (2010) is addressed via statistical assumptions in the under-coverage adjustment step of the weighting procedure.*

The sampling frame is divided into 20 different strata along the dimensions of three variables (nationality, occupational status and individual income subject to social contributions) describing the characteristics of the head of household. If a household has more than one member, the individual income of the person with the highest wage is used for the determination of the stratum. The final sample size comprises 950 households and 2,540 individuals. To reach this final sample size, an initial sample of 5,000 Luxembourg resident households extracted from the IGSS database was used. As detailed in Section 4.3 the sample is representative of 186,440 private households and 462,618 individuals living in Luxembourg at 31 December 2010 with the exclusion of institutionalized households, international civil servants and in general individuals not entitled to be registered in the social security register unless they are part of a household where another member is included in the social security register. The difference in the reference year of the sampling frame (2009) and of the reference population (2010) is addressed via statistical assumptions in the under-coverage adjustment step of the weighting procedure (see Section 4.3).

3.1 The fieldwork

The survey of Luxembourg resident households was conducted through computer-assisted personal interviews (CAPI). Table 2 presents an overview on the main characteristics of the fieldwork phase. Households selected for participation in the

survey received an information package shortly before the interviewer planned to establish a first contact with the household. The information package contained a letter signed by the Governor of the BCL and the President of CEPS/INSTEAD introducing the survey and describing its different features. Attached to the letter they also received a leaflet further detailing the main features and the aim of the survey (incl. a brief presentation of the BCL, the objectives of the survey, the confidentiality of the answers, the interview procedure, etc...). In order to provide the best possible support households were invited to contact two dedicated telephone numbers, one at the BCL and one at CEPS/INSTEAD, for any needed clarification on the aim, characteristics or legitimacy of the project. A web page on the BCL website contained further information about the survey and made reference to the Household Finance and Consumption Network initiative.

Table 2: Overview over fieldwork

Interview mode	Computer assisted personal interview
Survey company	CEPS/INSTEAD
Pre-test	July 2010
Information material for the household	Introduction letter, leaflet, dedicated web page and phone number
Planned fieldwork	End September 2010 to December 2010
Actual end of fieldwork	April 2011

In order to account for multicultural and multinational aspects of Luxembourg society and to reduce the burden of participating households, great effort was put into translating the LU-HFCS questionnaire into different languages; 60% of the interviews were conducted in French, 38% in German and 2% in English. Although the questionnaire was also available Portuguese, no household chose this language for the interview.

Table 3 shows the number of replies for each month during the field phase of the LU-HFCS. Difficulties encountered during the fieldwork resulted in a low participation rate requiring the extension of the data collection to the end of April 2011.

Table 3: Interview participation by month

	Sep. 2010	Oct. 2010	Nov. 2010	Dec. 2010	Jan. 2011	Feb. 2011	Mar. 2011	Apr. 2011	Total
Number of respondents	52	221	202	111	152	92	71	49	950
Percentage of respondents	5.5	23.3	21.3	11.7	16.0	9.7	7.5	5.2	100

Source: CEPS (2011). Technical Report of the HFCS.

Table 4 reports the duration of the interviews (in minutes). Each interview lasted on average 56 minutes with a median length of 51 minutes; the dispersion of the interview length is sizable and varied with the number of household members, the number of questions a household had to respond to, which depended on the complexity of the structure of its total income as well asset and liability portfolio. The

duration of the interviews shows that households provided serious effort in appropriately responding to the questionnaire, which is indicative of good data quality. This sentiment is also supported by the answers of interviewers provided in the section on paradata below.

Table 4: Interview duration

Minutes	<20	20-29	30-39	40-49	50-59	60-69	70-79	80-89	90-99	100-109	110-119	>120	Total
Number	9	64	169	202	185	122	96	51	23	10	9	10	950
In %	0.9	6.7	17.8	21.3	19.5	12.8	10.1	5.4	2.4	1.1	0.9	1.1	100

Source: CEPS (2011). *Technical Report of the HFCS*.

3.2 Unit non-response

As already mentioned the unit response rate of the survey has been relatively low. Table 5 shows a description of the response process.

Table 5: Response rates of the LU-HFCS

Response rate by category		in %
By nationality	national households	19.7
	non-national households	20.3
by income class	< 7,000 EUR	18.4
	> 7,000 EUR	24.2
Total		20.0

Source: Own calculations based on the gross sample of the LU- HFCS.

The total response rate, calculated as the ratio between the number of “accepted interviews” and the total number of interview requests sent out minus those “out of scope”, is reported in Table 5. In the LU-HFCS, it is not possible to identify a sizable difference in the response rate between different nationalities. Table 5 suggests a positive correlation between the level of personal income and the response rate, which is contrary to expectations and to common findings. Table 6 presents a simple probit regression of unit-non response and its determinants.

The first and third column present marginal effects derived from a probit regression of various household characteristics on the probability of accepting to participate in the survey.⁶ All the covariates are constructed as dummy variables. The dummy “Luxembourg nationality” takes the value 1 if the reference person of the household is of Luxembourg nationality and 0 otherwise. The dummy “High income” takes the value 1 if the personal gross income of the household reference person is higher than €7,000 per month. The base category for the reference person’s employment status is being a pensioner. For simplicity reasons we will refer to the reference person’s characteristics as household characteristics. The sample used for this estimation

⁶ Marginal effects are calculated as derivatives of the response with respect to the single covariate, holding all the other covariates fixed at their mean level. Calculating average marginal effects instead of marginal effects at the mean does neither affect the size nor the sign nor the significance level of the coefficients.

consists of all contacted households whether or not they accepted participating in the LU-HFCS.⁷

Table 6: Unit response model

	(1)		(2)	
	marg. eff.	std. err.	marg. eff.	std. err.
High income	0.071***	0.014	0.182***	0.037
* Luxembourg nationality			-0.018	0.029
* Self-employed			-0.192***	0.044
* Private employee			-0.111***	0.038
* Public employee			-0.086*	0.043
* Other employment			-0.194***	0.067
Luxembourg nationality	-0.024	0.013	-0.022	0.015
Self-employed	-0.083***	0.021	-0.013	0.027
Private employee	-0.01	0.016	0.015	0.018
Public employee	-0.011	0.02	0.007	0.025
Other employment	-0.038	0.026	0.002	0.029
Observations	4530		4530	
Pseudo R-squared	0.008		0.013	

Notes: Marginal effects; base category: Foreign nationality, low income, retired.

* p<0.05, ** p<0.01, *** p<0.001

Source: Own calculations based on the gross sample of the LU- HFCS.

As can be seen in specification (1), having a higher personal income increases the probability of survey participation whereas being self-employed decreases it relative to the base category. The remaining covariates are not statistically significant. Specification (2) of Table 6 presents the estimation of the marginal effects with the high personal income dummy being interacted with the employment status dummies and the nationality dummy. High income remains positively significant, although the effect is now allowed to vary across the employment status of the household and nationality. The negative and significant interaction coefficients indicate that for those employment statuses (relative to the base category) the positive effect of high income on survey participation is diminished and possibly even reversed in case of the self-employed. Thus, the positive effect of income on the probability of participating in the LU-HFCS is mainly driven by high income pensioners' increased participation rate. The very low Pseudo-R² in both regressions is suggestive of a limited sample selection problem, at least in an analysis based on observed characteristics at the stratification level.

3.3 Interviewer information (paradata)

The information interviewers provide after the conduct of an interview can be very useful to assess the overall quality of a survey. Figure 2 shows the distribution of

⁷ Households that are out of scope of the survey (so called non-contacts, i.e. households for which no contact could be established) are the sum of all households in the categories coded as: "No Away / at hospital during survey period", "Inaccessible", "Unable to locate the address", "Non-residential address", "Vacant/empty". They are excluded from this analysis. Their inclusion in the estimation does, however, not affect the presented results.

how interviewers rated the interest of participants in the interview. Only 5% of respondents showed below average interest. 15% showed very high and 36% above average interest. A high interest in the interview is expected to raise the quality of the answers and to reduce the number of questions, which would otherwise be left unanswered due to personal concerns or the response burden (item non-response). Having said this, Figure 2 may also be suggestive of the existence of a sample selection bias; households with a high interest in the topic of the survey are more likely to be included in the final sample. The weighting procedure described below aims to correct for this possibility by taking into account known characteristics of observed and unobserved households.

Figure 2: Interest in the interview

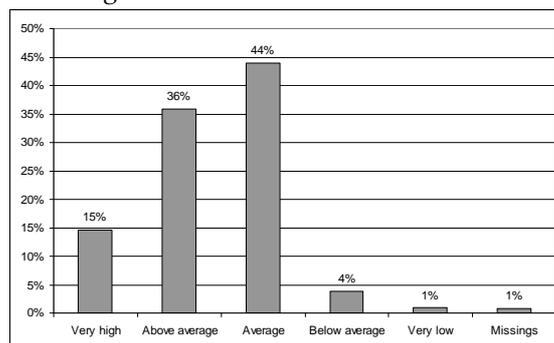
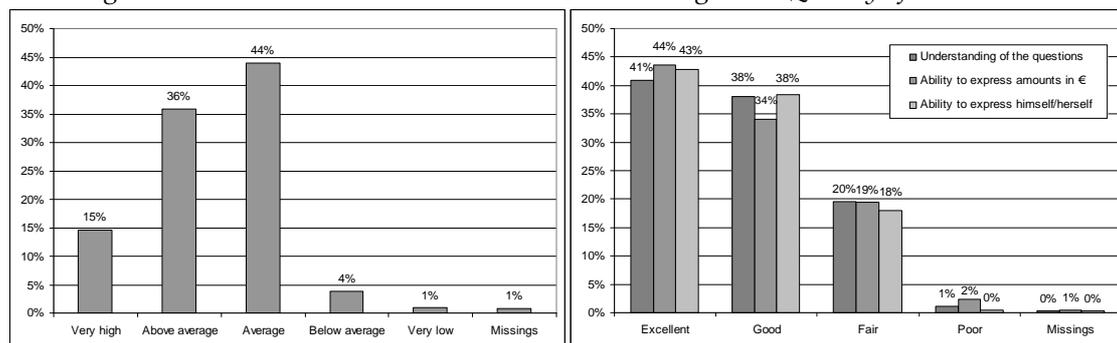


Figure 3: Quality of answers



Source: Own calculations based on LU-HFCS, non-imputed and unweighted.

For the data quality, it is as very important that respondents correctly understand the questions they are confronted with. Moreover, respondents must be able to express themselves verbally or in a mathematical manner providing euro amounts. As Figure 3 shows, only a negligible fraction of 0-2% of respondent households were rated to have a poor understanding of the questions or rated to have difficulties to express themselves. More than 40% of respondents are rated “excellent” by interviewers with respect to these three categories.

It can be relatively difficult to recall exact numbers on income or certain wealth categories. Not surprising, 43% of responding households consulted at least one document during the interview (Figure 4). Conditional on consulting at least one document, on average households consulted 3.2 documents. These figures are encouraging for the data quality since a large fraction of households made the effort to provide exact numbers in case they were not able to recall them by heart.

Finally, interviewers were requested to rate the overall reliability of the income and wealth information provided in the survey (Figure 5). Only 5% of respondent households are rated to provide inaccurate information; 46% are rated to provide fairly reliable information and 48% are rated to provide accurate information. In summary, the information provided by interviewers on data quality suggests that the LU-HFCS can be regarded as reliable source for empirical analysis.

Figure 4: Number of documents interviewee referred to

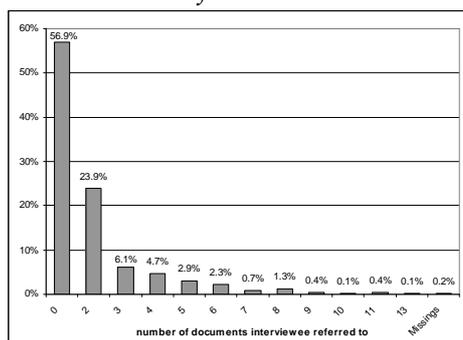
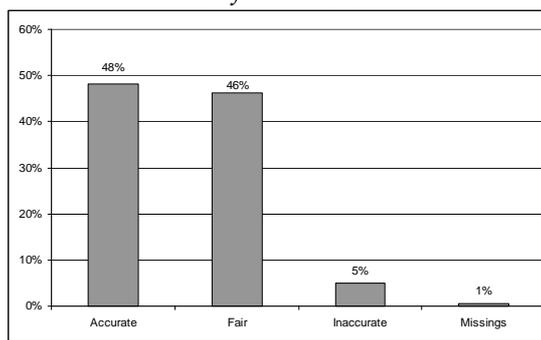


Figure 5: Reliability of income and wealth information



Source: Own calculations based on LU-HFCS, non-imputed and unweighted.

4 Data treatment

4.1 Editing

Logical inconsistencies between questions and item non-response are well known phenomena in complex surveys on sensitive topics, such as income, wealth and consumption, as is the case for the LU-HFCS. In a first step of the data preparation, the data are edited, where editing is “the activity aimed at detecting and correcting errors (logical inconsistencies) in data” (UNECE, 2000). These corrections include changes based on interviewer comments after the interview, institutional inconsistencies, logical inconsistency between questions, or mistyped or erroneous answers (e.g. the digit was wrongly placed or monthly instead of yearly income was provided). Appendix A provides details on which variable and how many entries had to be edited.

4.2 Imputation

Even if households are willing to participate in a survey, not all questions are always answered. It is well known that especially in complex surveys, such as surveys on income and wealth, missing values or so called item non-response can represent a serious problem. Item non-response strongly depends on the survey topic, the survey mode, the phrasing and ordering of questions. In case of a CAPI interview, this being the case for the LU-HFCS, the relationship with the interviewer may have an effect. Despite the HFCN being very careful to phrase questions as clearly as possible, missing values occur since interviewees may not know the answer themselves, are somehow reluctant to give answers on sensitive topics, or may simply not understand the question. Appendix A provides an overview on the number of missing values for each variable. Due to the fact that it might be difficult to recall exact values of certain income components or asset groups, all numeric variables are asked in a three level procedure. First, the interviewer asks for the exact value; second, if the exact value is not provided, the interviewer asks for a lower and upper bound for the specific value; third, if the respondent fails also to provide a own range, the interviewer offers certain pre-defined ranges the respondent can choose from. Appendix A indicates for how many households information in ranges is provided if the exact amount is missing.

The number of missing values and values to be edited varies strongly with the question. Table 7 shows a selection of variables out of Appendix A. It displays the percentage of ‘applicable’ cases, which reflects the number of individuals or households who should have had to reply to these questions. The column ‘unapplicable’ shows the fraction of households who skipped the question due to routing. The ‘undetermined’ cases reflect that it is uncertain whether or not a question should have been answered due to a missing value in a mother variable. The next two columns report the ‘minimum’ and ‘maximum’ fraction of values to be imputed relative to the sample size of applicable cases. The difference between the two is caused by possible missing values in the mother variable. The minimum fraction of imputed values relative to applicable cases is assuming that all undetermined cases are unapplicable; the maximum number assumes that all undetermined cases are applicable. The next two columns show the number of missing values for which bracketed information is available relative to the minimum and maximum number of values to be imputed. The last column shows the fraction of edited values relative to all applicable values (including both replaced and imputed values).

Table 7: Missing and editing rates for some selected variables

Variable name and description		Unapplicable	Applicable	Undetermined	Min. # to be imp. (I)	Max. # to be imp. (II)	Bracket values in % of (I)	Bracket values in % of (II)	Edited
RA0200	gender	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
RA0300	age	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.2%
RA0400	country of birth	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
PA0100	marital status	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
PA0200	highest level of education	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%
PG0100	received employee income	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.6%
PG0110	gross cash employee income	47.0%	52.8%	0.2%	28.3%	28.7%	71.1%	70.1%	1.3%
PG0200	received self-employment income	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.2%
PG0210	gross self-employment income	91.5%	8.3%	0.2%	35.5%	38.5%	66.7%	61.5%	0.0%
HB0300	household main residence (HMR) - tenure status	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
HB0900	current price of HMR	30.0%	70.0%	0.0%	11.7%	11.7%	75.6%	75.6%	0.0%
HB1000	mortgages or loans using HMR as collateral	30.0%	70.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.3%
HB1010	number of mortgages/loans using HMR as collateral	65.5%	34.4%	0.1%	0.3%	0.6%	0.0%	0.0%	0.9%
HB1701	HMR mortgage 1: amount still owed	65.5%	33.7%	0.8%	11.3%	13.8%	58.3%	47.7%	3.1%
HB1702	HMR mortgage 2: amount still owed	94.8%	4.4%	0.7%	11.9%	28.6%	20.0%	8.3%	11.9%
HB4300	ownership of cars	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
HB4400	total value of the cars	10.9%	89.1%	0.0%	6.0%	6.0%	70.6%	70.6%	0.0%
HD1100	household owns sight accounts	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HD1110	value of sight accounts	3.2%	96.8%	0.0%	36.1%	36.1%	64.5%	64.5%	0.0%
HD1200	household owns saving accounts	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HD1210	value of saving accounts	25.6%	74.1%	0.3%	41.2%	41.6%	60.3%	59.7%	0.7%
HD1500	household owns publicly traded shares	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HD1510	value of publicly traded shares	86.5%	13.2%	0.3%	25.6%	28.0%	37.5%	34.3%	0.8%

Source: Own calculations based on LU-HFCS, non-imputed and unweighted.

Socio-demographic information have very low missing rates of normally below 1% as can be seen for example for the variables: gender, country of birth, marital status, and highest level of education. Ownership variables have low missing rates as well: the tenure status of the household main residence, ownership of cars, whether or not households own sight accounts, saving accounts or shares have missing rates of below 1% as well. The number of missing values increases strongly if one asks for the exact amount. The missing rate relative to applicable cases is almost 30% for gross cash employee income and between 36-39% for received self-employment income. More than 70% (62-67%) provided bracketed information if the exact value could not be provided for employee income (self-employment income), which proves to be

very useful for the precision of the imputation procedure. Missing values for amounts range from 6% for the total value of cars to 41% for the value of saving accounts (Table 7).

Item non-response has serious consequences if not properly treated. First and foremost, the missing generating process is most likely not missing completely at random, which means that whether or not a value of a variable is missing depends on observed or possibly unobserved characteristics of the respondent. This might lead to biased estimates. For instance, assuming that wealthy households are more reluctant to provide information about their asset holdings using the observed information on wealth only would result in downward biased aggregate wealth estimates. Second, the sample size and therewith the efficiency of the estimates is reduced depending on the number of variables included in the analysis and their missing rates and structure.

Table 8: Total number of edited and imputed values

Description	In %	Values
Applicable in % of total:	45.0%	231273
Unapplicable in % of total:	54.6%	280726
Undetermined in % of total:	0.5%	2338
Min missings in % of applicable:	2.7%	6346
Max missings in % of applicable:	3.8%	8684
Bracket values in % of min missing values:	23.5%	1493
Bracket values in % of max missing values:	17.2%	1493
Editing: corrected values in % of applicable:	6.6%	15178
Editing: corrected values in % of applicable (without pensions):	0.4%	919
Editing: set to missing in % of applicable:	0.5%	1235
Editing: total in % of applicable:	7.1%	16413

Source: Own calculations based on LU-HFCS, non-imputed and unweighted.

Table 8 summarizes the share and total number of edited and imputed values in the LU-HFCS. It gives a general impression how many values had to be edited and imputed and neglects the strong variation in missing and editing rates across variables. In total, respondents should have answered 45% of questions. 54.6% were unapplicable and should not have been answered and 0.5% were undetermined. The average minimum missing rate as a share of applicable cases is 2.7%. Adding the undetermined values, the average maximum missing rate increases to 3.8%. These missing rates seem to be rather low and might question the very time and resource intensive imputation process applied that follows. However, as pointed out in Table 7 key variables of income and wealth have normally much higher missing rates. The problem will amplify if an aggregate variable is constructed or a multivariate analysis is conducted. For example total net wealth is constructed out of a wide range of variables. If all observations are dropped if one component is missing, 63% of all observations will have to be deleted. On average, bracketed information is available for 24% of the minimum number of missing values. This number is

relatively large given the fact that bracketed information is only asked in case of questions with numerical responses.

Finally, the percentage share of corrected values relative to applicable cases is 6.6%. This relatively high number is mainly due to the editing process of one section concerning public pensions. The public pension section necessitated editing all values due to many inconsistencies between personal characteristics (e.g. employment status) and the institutional framework in Luxembourg. Disregarding the public pension section, the average percentage share of corrected values reduces to 0.4% only. An additional 0.5% had to be deleted and imputed at a later stage since the value could not be replaced with certainty by another one.

One way to deal with item non-response is *imputation*, where “*imputation is a generic term for filling in missing data with plausible values*” (Schafer, 1997, p. 1). Although there are alternative ways to handle missing values like complete case analysis, weighting, or model-based procedures,⁸ there is strong support for multiple imputation. Here, multiple is to mean that missing values are not just replaced by one value but by M different values (normally M=5) to account for the uncertainty in the imputation procedure. First, the advantage is that the imputed dataset can be analyzed with complete-data methods (Rubin, 1996, p. 474). Second, researchers can analyze the same imputed dataset and are spared from the time consuming imputation process. Third, the data provider has normally additional information which is useful for imputation but cannot be released to the researchers and final user (for example due to data confidentiality reasons). In case of the LU-HFCS, the sample design information and information provided by interviewers were for instance used in the imputation models. And finally, as the uncertainty with regard to the imputation algorithm is taken into account and the correlation structure of the dataset can be preserved, multiple imputations have shown to exhibit good properties.⁹

The key assumption for imputation is that the missing mechanism is MAR (missing at random), which means that missingness only depends on components which can be observed and are included in the dataset and not on components which are unobserved (Cameron and Trivedi, 2005, pp. 925-927). The MAR assumption cannot be tested. On that account it is so important to include as many variables and their interactions as possible in the imputation models.

There are a wide range of different imputation programs available. The LU-HFCS makes use of the ECB Multiple Imputation Routines (€MIR), which provides a range of SAS programs designed to perform multiple imputation for the Eurosystem HFCS

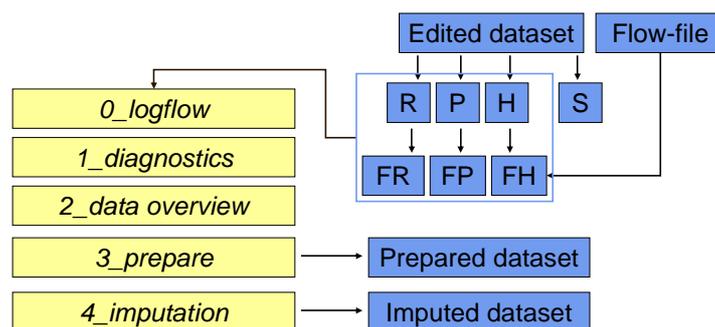
⁸ See Rassler and Riphahn (2006) for an overview of methods to treat item non-response and an application of multiple imputation in this context.

⁹ See for example Schafer et al. (1996), Graham et al. (1996), Graham and Schafer (1999), and Rassler and Riphahn (2006).

(Biancotti, 2010).¹⁰ Figure 6 provides an overview of €MIR. After the dataset is edited as described above, the dataset is separated into different sub-datasets respectively containing the i) household level information (H), ii) basic demographic personal information (R), iii) all other person level information (P), and iv) the sample register file (S). Based on a logical flow file, which contains the filter rules for each variable, the flag files for H, R, and P are generated. Since the S file will be neither imputed nor transmitted to the final user, a flag file is not necessary. Using H, R, P datasets and their flags, €MIR executes the following steps:

- 0.) *Logflow* is a SAS package that performs a series of consistency checks between variable values and status flags and produces tables where inconsistencies of various types are indicated.
- 1.) The output of the *diagnostics* package consists of tables indicating how many missing values of different types exist for every variable in the H, R, and P datasets.
- 2.) *Data overview* displays basic descriptive statistics on all variables, and estimates an item non-response model for each variable. The results of the item non-response models help to identify important predictors of item non-response, which are included in the imputation model at a later stage.
- 3.) The *preparation* module is just a technical step that assembles all relevant files for the imputation process. In case of the LU-HFCS, it combines the H, R, P and S datasets and their flags in one dataset in wide format. Since most of the variables in the HFCS are household variables, the wide format allows taking detailed personal information in imputing household-level variables into account. In addition, the complete dataset is replicated five times as a basis for generating five multiply imputed datasets.
- 4.) The most extensive element is the *imputation* step, which is explained in detail below.

Figure 6: Overview of the ECB Multiple Imputation Routines (€MIR)



Source: Own figure based on Biancotti (2010).

¹⁰ €MIR is a further developed multiple imputation routine based on the Federal Reserve Imputation Technique Zeta (FRITZ) used for the Survey of Consumer Finances (SCF). €MIR has the advantages that it increases the comparability between the imputation algorithms of HFCS countries. Being based on FRITZ, it is known to be a highly flexible imputation tool able to handle complex surveys on household finances.

Figure 7 shows in more detail the procedures within the imputation module. After the production of five replicates of the prepared dataset suitable covariates for the imputation models are generated by variable transformation (e.g. creation of dummy and interaction variables, logarithmic transformations). Three different models depending on the type of variable to be imputed are used.

- 1.) *Hot deck* imputation is applied for the imputation of categorical variables. Respondents and non-respondents for a specific question are classified into adjustment cells with similar observed values. A respondent is randomly drawn out of this cell and the missing value of a non-respondent is replaced. Since only a limited number of variables for cell formation can be used, the selected variables should be highly correlated with the variable to be imputed.
- 2.) *Regression models* (OLS) are used for continuous variables. The following linear relationship is estimated for the observed values of dependent variable (y_{obs}):

$$(1) \quad y_{obs} = X\beta + u \quad \text{with} \quad u|X \sim N(0, \sigma^2 I)$$

where y is the dependent variable, X a vector of independent variables, u is a normally distributed error term, and β is the vector of coefficients. Provided that the missing at random assumption holds, the OLS estimate of β ($\hat{\beta} = (X'X)^{-1}(X'y)$) is unbiased. The used covariate vector X must be complete, which means either observed or imputed. The missing values are imputed by their best linear predicted value plus a normally distributed random variable.

$$(2) \quad \hat{y}_{mis} = X\hat{\beta} + \hat{u}, \quad \hat{u}|X \sim N(0, \hat{\sigma}^2 I)$$

where \hat{u} is drawn from a normal distribution with mean zero and variance equal to the mean squared error of equation 1, which is $\hat{\sigma}^2 = \frac{1}{n-k} (y'y - y'X(X'X)^{-1}X'y)$.

The better the covariates can explain the observed values of dependent variable, the smaller will be the variance of the added random error. In other words, the better the relationship between dependent and independent variables can be modelled, the less uncertain is the imputed value, which is reflected in a smaller variance between the five imputates. The imputation procedure is not proper in the sense of Rubin (1987, pp. 118-119) since it does not properly model the uncertainty with respect to choosing the accurate imputation model. However, adding randomization with respect to coefficients as well adds additional complexity that can almost not be handled in such a complex survey as the LU-HFCS.¹¹ Moreover, the imputation models take provided ranges into account. Provided ranges assure that the imputed value is within this range, which makes the imputation procedure more precise than without any lower and upper thresholds.

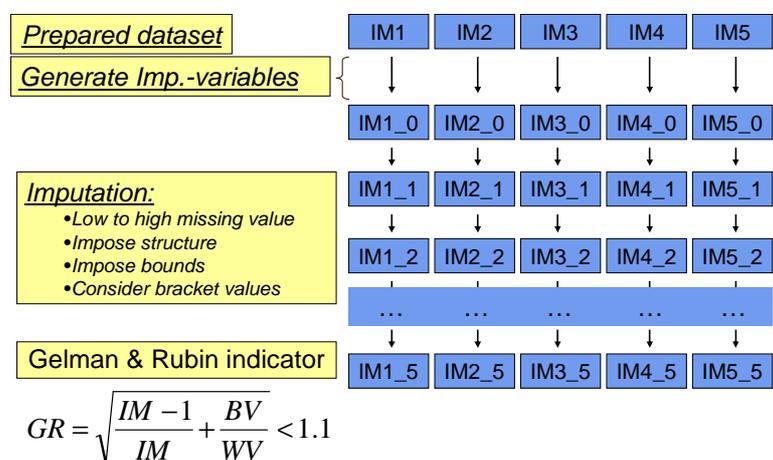
¹¹ As monetary values are usually highly skewed to the right, the dependent variable is often transformed by taking the normal logarithm of this variable. The log transformation reduces, if not eliminates, this skewness. Problematic is the so-called retransformation to the original scale after the missing value is predicted. Ziegelmeier (2012) discusses the retransformation problem and possible solutions.

- 3.) The imputation model used for binary variables - *linear probability model* - is similar to the one outlined in equation (1) and (2). The main difference is that the dependent variable is often an indicator variable of owning a certain assets category or receiving income from a certain source instead of a continuous variable.

In total, 424 variables are imputed with the routing of the questionnaire as additional constraint. To preserve the covariance structure of the questionnaire the maximum number of covariates is included. This normally includes survey design information, variables from the sample register file, predictors of item non-response (see data overview program of €MIR), demographic information of the household head, variables on household composition, various pieces of information on income and wealth and possible interactions for selected variables. In some cases the limited number of observations necessitated the reduction of the number of covariates removing first those with least explanatory power and those likely to cause multicollinearity between the variables.

In a first round, the imputation procedure imputes for each implicate $M \in \{1, 2, \dots, 5\}$ all missing values using the models outlined above to obtain a first complete database without missing values. Generally, the imputation starts with variables having low missing rates and progresses to variables with higher missing rates. The remaining missings are due to the routing of the questions. For each implicate, the imputation procedure is repeated using the imputed values of the first complete dataset as initial value for the next imputation. This process is called "*Markov Chain Monte Carlo Multiple Imputation Procedure*" and it is repeated until certain convergence criteria are fulfilled (Kennickell, 1991, pp. 10-12). For the LU-HFCS, the Gelman and Rubin indicator (Gelman and Rubin, 1992) is used. The indicator is calculated for the mean and median of all continuous variables. The process is stopped after five iterations. This is as the Gelman and Rubin indicator is below the suggested value of 1.1, which indicates that the convergence of the variables of interest is achieved (Gelman, Carlin, Stern, and Rubin, 2004).

Figure 7: Overview of the ECB Multiple Imputation Routines (€MIR)



Source: Own figure based on Biancotti (2010).

A difficult task is to *evaluate* an imputation procedure, which is due to missing values not being observed. In the best case, the expected value of the quantity of interest Q should be equal to the associated complete-data estimate. Usual quantities of interest are the mean and the 25th, 50th, and 75th percentile of the distribution, correlations or regression coefficients. In addition, proper imputation requires the correct estimation of the variance of the quantities Q of interest (Rubin, 1996, pp. 474-475). As true values for missings are unknown, time-consuming and computationally demanding simulation studies are commonly used to evaluate imputation procedures.¹² In large surveys like the LU-HFCS only a few variables could be investigated this way. Instead, the approach followed by the BCL as well as the ECB is to analyze whether imputation leads to the preservation of distributions and correlations between variables by investigating the mean, median, Q1, Q3, P10, P90, standard errors and extreme observations of all the imputed variables.

All descriptive statistics presented in sections 5 and 6 are based on five complete datasets, which only differ in their imputed values. The statistics shown are based on Rubin's rules (Rubin, 1996, pp. 467-477), which calculate the average over the estimated quantities generated by the five different datasets, and adjust standard errors taking both the within-imputation and between-imputation variance into account.

4.3 Weighting

The sampling procedure follows closely the ECB guidelines for the construction of survey weights for the HFCS (HFCN, 2011a). It contains the following steps: i) construction of design weights based on selection probability, ii) non-contact adjustment, iii) over-coverage adjustment, iv) non-response and v) under coverage adjustment. Due to lack of available data it was not possible to calibrate the analytic weights to external data sources.

The reference population of the LU-HFCS comprises all private households living in Luxembourg at 31 December 2010 with the exclusion of institutionalized households, international civil servants and in general individuals not entitled to be registered in the social security register unless they are part of a household where another member is included in the social security register. The sampling frame consists of 248,601 fiscal households¹³ and 466,378 individuals registered in the IGSS database at the end of December 2009 (IGSS data referring to December 2010 were not available at the time of the selection of the sample design). The sample design discussed in Section 3 resulted in selecting 5,000 fiscal households. The difference in the definition of the sample unit (fiscal households) and the target unit (private households as

¹² See Ziegelmeier (2012) for an example using the SAVE survey.

¹³ A fiscal household is composed by one or two individuals directly insured via the Luxembourg social security system and all indirectly insured individuals such as depended spouse or children as indentified in the social security register.

defined in the HFCS Manual of Procedures¹⁴) is addressed in the set up of the weighting procedure. This also holds, as already mentioned, for the difference in the reference year of the sampling frame (2009) and the reference population (2010).

- i) The design weights are calculated as the inverse of the selection probability. The household weight is constructed taking into account the different selection probabilities of each stratum. Having been oversampled the weights of high income and self-employed households are adjusted downward.
- ii) The non-contact adjustment is necessary to take into account that a number of households included in the initial sample could not successfully be contacted by interviewers. A non-contact with a sampled household can arise for a number of reasons such as: inaccessibility of the address, non possibility to locate the selected address, dwelling demolished or under construction, dwelling empty etc. The non-contact adjustment is calculated using the Calmar procedure developed by French National Statistics Institute (INSEE) (e.g. Deville et al., 1993). Auxiliary information necessary for this procedure (at population and sample level) was derived from the IGSS database.
- iii) The over-coverage adjustment addresses that a number of non-eligible households (as non-resident households, or individuals living in collective households, households who moved abroad etc...) may have been accidentally selected in the sampling procedure, given that it was not possible to a priori identify those households by using the information contained in the sampling frame. This step of the weighting procedure aims to exclude non-eligible households either from the selected sample or from the sampling frame. Having established the number of eligible fiscal households present in the sampling frame and in the sample, the number of eligible private households is established linking the two by means of statistical assumptions starting from the calculated number of eligible individuals.
- iv) The non-response adjustment addresses issues related to the fact that sampled households, although having been contacted and eligible, refused to participate in the survey (see Section 3.2). Similar to the non-contact adjustment the weighting procedure uses the Calmar procedure to maintain the representativeness at the population level of the extracted sample. Again, the incorporation of auxiliary information from the IGSS database is necessary for this procedure (at population and sample level).
- v) The last step of the weighting procedure is the under-coverage adjustment. This issue mainly arises because the sampling frame is based on a sample drawn from the IGSS database in December 2009, while the target population consists of all

¹⁴ In the HFCS Manual of procedures (HFCN, 2010 - Annex 1) a household is defined as “*a person living alone or a group of people who live together in the same private dwelling and share expenditures, including the joint provision of the essentials of living*”.

private households resident in Luxembourg at the end of December 2010. Thus, before implementing this adjustment, a small fraction of Luxembourg resident private households, such as new immigrants, or newly born individuals, is not yet covered by the survey. The under-coverage adjustment overcomes this problem via statistical assumptions, making the LU-HFCS representative of all private households living in Luxembourg at 31 December 2010 with the exclusion of institutionalized households, international civil servants and in general individuals not entitled to be registered in the social security register unless they are part of a household where another member is included in the social security register. Table 9 presents an overview of the weighted and unweighted number of observations at household and individual level.

Table 9: Number of observations in the sample

Sample	Unweighted number of observations	Weighted number of observations
Households	950	186,440
Individuals	2,540	462,618

Source: Bienvenue (2012)

Following the ECB guidelines concerning the bootstrap method for variance estimation in the HFCS (HFCN, 2011b), in addition to the analytic weights described in this paragraph, 2,000 sets of replicate bootstrap weights are calculated using the Rao and Wu (1988) method; these sets of weights are necessary in order to allow the final user to properly calculate the variance of the estimates without disseminating sample design information.

The slight variability in the weighted number of individuals in different sets of replicate weights reflects the random component of the statistical assumptions used in the under-coverage adjustment of the analytic weights.

4.4 Anonymisation

It is the aim to make the database publicly available for scientific and research purposes. To this end, a user database (UDB) is created where individual data records are anonymised as to ensure to the extent possible that individuals and households having participated in the LU-HFCS cannot be identified. In this context, *“Anonymised microdata” shall mean individual statistical records which have been modified in order to minimise, in accordance with current best practice, the risk of identification of the statistical units to which they relate* (European Commission, 2002).

The anonymisation of the database is a delicate undertaking; on the one hand, it needs to ensure that the disclosure control of the data is appropriate. On the other hand, if the data is altered and tweaked too much the accuracy and meaningfulness of the data may undermine the purpose of collection in the first place.

Essentially, the following measures are taken to anonymise the LU-HFCS database:

- Removal of direct identifiers
- Removal of some variables (e.g. sample register file)

- Rounding / addition of random noise to continuous variables (e.g. income)
- Top-coding of certain variables (e.g. age, size of residence)
- Regrouping (e.g. education, type of vehicle, number of employees in owned business)
- Non-disclosure of lower levels of disaggregation (e.g. NACE-codes)
- Imputation of selected cases

In general, the measures taken are in line with the guidelines of the ECB (HFCN, 2012) and EU-SILC (Eurostat, 2005, 2007, 2009); additional measures to take into account the special situation in Luxembourg are taken as well.

5 Sample and population characteristics

This section presents household basic descriptive statistics at sample (i.e. unweighted) and population level (i.e. weighted), with a particular focus on the mean, standard error of the mean estimator, including the confidence interval of the mean. Although our main interest is on population representative figures, Table 10 reports even unweighted values in order to show the effect of the sampling and weighting procedure described in the previous sections and to provide some information on the sample characteristics.

The average household in Luxembourg consists of 2.48 individuals, has 0.58 children with an age of below 18 years and in 60% of cases is headed by a male. The modal age group for household heads is 40 to 49 years and the process of population aging becomes evident when analyzing the distribution of this variable with its sizable right tail. Luxembourg is the European country with the highest share of immigrants and this feature is well reflected in the LU-HFCS. At the population level, 62% and 38% of households are headed by a person with Luxembourg and foreign citizenship, respectively.

Comparing the results at sample and population level for this set of variables we observe that the effect of the sampling strategy and weighting procedure on the mean of these variables is relatively small. This is not the case for variables, such as wealth and income or for variables closely related to those, as the education level and employment status. The mean total household income and the total household net wealth are in fact much higher at the sample level than in the population. A comprehensive analysis of income and wealth aggregates will be presented in the next section. These characteristics reflect the effect of the oversampling of the wealthiest households.

With a population share of 38%, the modal household head has completed the education category ISCED 03-04, which refers to upper secondary education and post-secondary non-tertiary education. It is interesting to underline that, at the population level, 26% of household heads successfully completed the ISCED 05-06 level of education (first stage of tertiary and second stage of tertiary education). This value is much higher at the sample level (37%), again, reflecting the effect of the

oversampling of wealthy households and the positive relationship between income and the level of education of the household head. The effect of overweighting the wealthy households also depicted in the statistics on the employment status. Households headed by self-employed individuals account just for 6% of the Luxembourg resident population but they represent 12% of the LU-HFCS sample. It is worth noting that the standard errors of the mean estimator are reasonably low for all variables reported here in this section, which is suggestive of good data accuracy despite the relatively low number of observations in the sample.

Table 10: Household demographic characteristics

Variable	Sample Level (obs=950)				Population Level (obs=186,440)				
	Mean	Std. Err.	[95% Conf. Interval]		Mean	Std. Err.	[95% Conf. Interval]		
Household Size	2.67	0.04	2.59	2.76	2.48	0.00	2.47	2.49	
Number of children < 18	0.58	0.03	0.52	0.64	0.52	0.01	0.50	0.54	
Total income	106,882	3,718	99,585	114,179	83,641	2,358	79,020	88,262	
Total wealth	984,613	78,039	831,462	1,137,765	732,728	61,180	612,817	852,639	
Male*	0.62	0.02	0.58	0.65	0.60	0.01	0.57	0.62	
Under 29	0.07	0.01	0.06	0.09	0.09	0.01	0.07	0.11	
From 30 to 39	0.18	0.01	0.15	0.20	0.18	0.01	0.16	0.21	
From 40 to 49	0.27	0.01	0.25	0.30	0.27	0.01	0.25	0.30	
From 50 to 59	0.19	0.01	0.17	0.22	0.18	0.01	0.16	0.20	
From 60 to 69	0.17	0.01	0.14	0.19	0.13	0.01	0.11	0.14	
Over 70	0.11	0.01	0.09	0.13	0.15	0.01	0.13	0.17	
Natio- nality*	Luxembourg	0.59	0.02	0.55	0.62	0.62	0.01	0.60	0.64
Other	0.41	0.02	0.38	0.45	0.38	0.01	0.36	0.40	
Educa- tion level*	ISCED 01**	0.17	0.01	0.15	0.20	0.24	0.02	0.21	0.27
ISCED 02**	0.10	0.01	0.08	0.11	0.11	0.01	0.09	0.14	
ISCED 03-04**	0.36	0.02	0.33	0.39	0.38	0.02	0.35	0.42	
ISCED 05-06**	0.37	0.02	0.34	0.40	0.26	0.01	0.23	0.29	
Employed	0.53	0.02	0.50	0.56	0.56	0.01	0.54	0.58	
Empl. status*	Self-employed	0.12	0.01	0.10	0.14	0.06	0.01	0.05	0.07
Retired	0.24	0.01	0.22	0.27	0.24	0.01	0.22	0.26	
Other	0.11	0.01	0.09	0.13	0.14	0.01	0.11	0.16	

Source: Own calculations based on the LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights.¹⁵

Notes: * Variables referring to the head of household. ** ISCED 01=primary education or first stage of basic education; ISCED 02 = lower secondary or second stage of basic education; ISCED 03-04= upper secondary education and post-secondary non-tertiary education; ISCED 05-06= first stage of tertiary and second stage of tertiary education.

¹⁵ In situations involving multiple imputation, we use the replicated variance estimation for subpopulations. When one is using the sample design information (strata, clusters, and weights), the possibility arises that the subpopulation is not present for entire strata. In such cases, Stata will drop such strata, which could render the number of observations inconsistent across imputations. This is not the case when using replicated variance estimation.

6 Findings

This section presents preliminary estimates on Luxembourg households' income and wealth using the LU-HFCS data collected in 2010-2011. All values are in nominal euro terms. All income related data refer to 2009, whereas wealth data refer to the actual values at the time of the interview. All descriptive statistics are weighted such that provided information can be regarded representative of the underlying population.

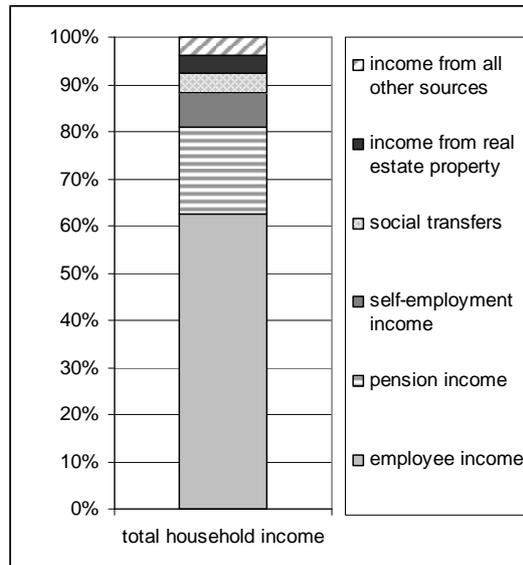
Table 11 presents the mean household gross income, by income category. The estimated average household gross income of Luxembourg households in 2009 was about €83,600. With 63%, the majority of the average gross household income derives from labour income as employee, while 7% and 18% derive from self-employment income and pensions (Figure 8). Other income sources are barely important at the aggregate level.

Table 11: Composition of mean household gross income

Income category	Mean	Std. err.	[95% conf. interval]		In % of total income
employee income	52316	1865	48656	55976	63%
+ self-employment income	6073	719	4661	7485	7%
+ pension income	15340	791	13774	16905	18%
+ social transfers	3601	241	3127	4075	4%
+ income from regular private transfers	394	80	237	550	0%
+ income from real estate property	3230	887	1490	4970	4%
+ income from financial investments	939	181	583	1295	1%
+ income from private business	307	146	20	594	0%
+ income from other income sources	1442	761	-53	2938	2%
= total household income	83641	2559	78618	88664	100%

Source: Own calculations based on the LU-HFCS; data are multiply imputed and weighted; variance estimation based on survey design information.

Figure 8: Composition of mean household gross income



Source: Own calculations based on the LU-HFCS; data are multiply imputed and weighted.

Average self-employment income seems to be rather low, which is due to not all households receiving income from all sources. Employee income and self-employment income are received by 71% and 10% of the resident households, respectively (Table 12). The conditional employee and self-employment income is estimated to be €73,700 and €58,900, respectively. Note however, that the self-employment income includes low incomes of self-employed, for which the self-employment income is not considered the main income source. This partly explains why the income may be lower than the employee income.

Figure 12 additionally presents the income of household reference persons stating that their main employment status is being self-employed. For those the average income is estimated to be €150,000 and thus much higher, thereby indicating that the discrepancy can indeed partly explained by the fact that income from self-employment is not the main income source for a part of the sample.

Table 12: Conditional mean and median household gross income

Income category	Mean	Std. err.	[95% conf. interval]		Subgroup size	Share of total population	Median
employee income	73668	2433	68898	78438	132401	71.0%	58797
self-employment income	58857	6996	45141	72573	19238	10.3%	30200
pension income	43799	1915	39977	47622	65426	35.1%	38766
social transfers	8635	424	7804	9467	77725	41.7%	6983
income from regular private transfers	6724	1179	4412	9036	10919	5.9%	5086
income from real estate property	24359	6240	12129	36590	24838	13.3%	10272
income from financial investments	2095	425	1262	2928	83912	45.0%	497
income from private business	32267	15171	2530	62005	1774	1.0%	11667
income from other income sources	64355	30925	3667	125043	4178	2.2%	20000
total household income	83641	2358	79020	88262	186440	100.0%	64653

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights.

The conditional average household pension income, received by 35% of Luxembourg households, is estimated to be €43,800. 41% of Luxembourg households receive an estimated average of about €8,600 in social transfers. The share of Luxembourg households receiving income from real estate investments is lower at 13% with an average income of about €24,400. In contrast, the estimated income from financial investments, conditional on holding such assets, is rather low at about €2,100. Despite this low value, almost 50% of Luxembourg households receive an financial investment income.

Turning to wealth, the estimated average (median) total net wealth of Luxembourg households is about €733,000 (€403,000) (Table 13). This is an extremely high value by international standards. For example in the U.S., average (median) household total net wealth is estimated to be about €425,000 / \$595,000 (€90,000 / \$125,000) in

2009 (Bricker et al., 2011). In France, mean household net wealth is estimated at €229,300 in 2010 (Chaput et al., 2011); in Germany it is estimated to be around €140,000 in 2002 (Frick et al., 2010).

Table 13: Composition of total net wealth

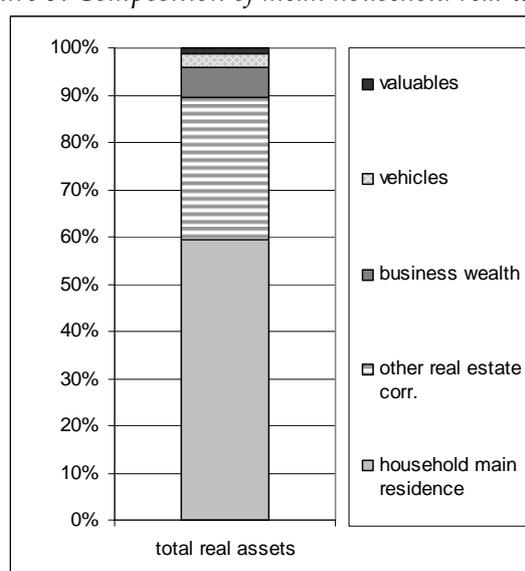
Wealth category	Mean	Std. err.	[95% conf. interval]		Share of total gross wealth	Median
total real wealth	722711	53976	616779	828643	89%	359000
+ total financial assets	88424	7572	73560	103288	11%	26653
= total gross wealth	811135	56231	700780	921490	100%	493937
- total debt	78407	5104	68389	88424	10%	5001
= total net wealth	732728	55646	623521	841936	90%	403103

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on survey design information.

With a share of 89%, the predominant part of average total gross household wealth consists of real wealth. On average, Luxembourg households are estimated to hold about €723,000 in total real wealth and about €88,400 in total financial wealth, resulting in total gross wealth of about €811,000. The average total debt of Luxembourg households is about €78,400.

The main part (59%) of real household wealth consists of the value of the household main residence (Figure 9), which accounts for about €430,000 of the estimated €723,000 total real wealth (Table 14). 30% of households' real wealth derive from other real estate holdings, which are estimated to have an approximate average value of €217,000. Vehicles and other valuables make up only a small fraction of average household real wealth. On the liability side, average total mortgage debt is estimated at €74,000.

Figure 9: Composition of mean household real wealth



Source: Own calculations based on the LU-HFCS; data are multiply imputed and weighted.

Table 14: Composition of mean household real wealth

Wealth category	Mean	Std. err.	[95% conf. interval]		Share of total real wealth
household main residence	429873	30456	370101	489646	59%
+ other real estate corr.	217086	38140	142234	291937	30%
+ business wealth	45681	11539	23033	68328	6%
+ vehicles	20878	1363	18203	23554	3%
+ valuables	9193	1250	6740	11646	1%
= total real wealth	722711	53976	616779	828643	100%
- total mortgage debt	73736	5053	63819	83652	10%
= net real wealth	648975	53374	544225	753726	90%

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on survey design information.

Looking at the respective conditional wealth components drives figures up, as is expected (Table 15). First, an estimated 67% of Luxembourg households own their residence (in full or in part). The average and median value of the household main residence is estimated to be €640,000 and €500,000, respectively. Almost an estimated 30% of households own other real estate properties, which are valued at an estimated €770,000 on average. The median is lower at almost €300,000, indicating a significant amount of right skewness in the data. This is also the case with regard to business wealth. Business wealth is held by an estimated 6% of households, which an average value of almost €760,000, whereas the median is estimated at a much lower €190,000. With 87% the most commonly held real asset category, households' vehicles are estimated to be worth about €24,000 on average. A further 24% of Luxembourg households own valuables with an estimated average value of almost €39,000.

Table 15: Conditional mean and median household real wealth

Wealth category	Mean	Std. err.	[95% conf. interval]		Subgroup size	Share of total population	Median
household main residence	640571	50778	541045	740097	125116	67%	500000
+ other real estate corr.	770852	142921	490728	1050975	52545	28%	299000
+ business wealth	756600	206234	352364	1160835	11257	6%	189685
+ Vehicles	24078	1379	21376	26780	161663	87%	16071
+ Valuables	38601	4954	28890	48311	44402	24%	12251
= total real wealth	722711	59756	605591	839831	186440	100%	445707
- total mortgage debt	190232	10528	169594	210871	72266	39%	127326
= net real wealth	648975	59191	532962	764988	186440	100%	357633

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights.

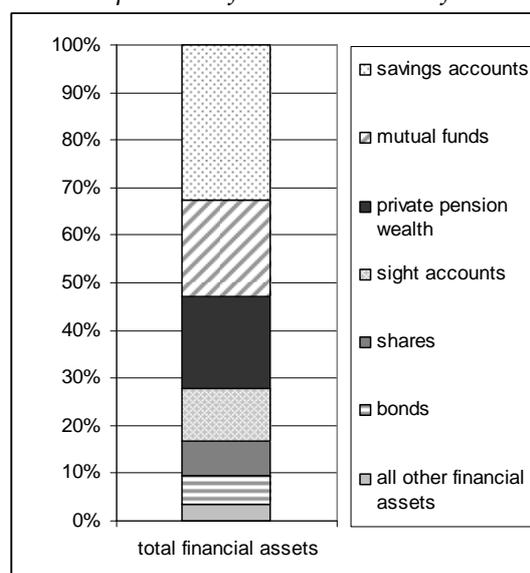
With regard to financial assets, the following figures emerge (Table 16 or Figure 10). With 32%, 21%, 19% and 11% of Luxembourg households' financial wealth is held in form of savings accounts, mutual funds, private pension wealth and sight accounts. On average, Luxembourg households hold almost €29,000 in their savings accounts. Mutual funds amount on average to almost €18,000, private pension wealth to almost €17,000, and sight accounts to around €10,000. Other financial asset components are less relevant.

Table 16: Composition of mean household financial assets

Financial asset category	Mean	Std. err.	[95% conf. interval]		Share of total fin. assets
sight accounts	9973	835	8324	11623	11%
+ savings accounts	28682	2182	24391	32973	32%
+ mutual funds	18143	3595	11055	25230	21%
+ Bonds	5431	2528	470	10391	6%
+ Shares	6365	1998	2444	10286	7%
+ managed accounts	88	72	-54	230	0%
+ amount owned to household	1945	842	293	3598	2%
+ any other financial assets	506	220	60	951	1%
+ value of non-self-empl. not publicly traded businesses	446	239	-26	917	1%
+ private pension wealth	16845	2271	12389	21301	19%
= total financial assets	88424	7572	73560	103288	100%
- non mortgage debt	4671	403	3880	5462	5%
= net financial assets	83753	7578	68876	98630	95%

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on survey design information. Occupational pension wealth is not included in total financial assets due to the fact that more than 50% of those having an occupational pension plan do not have an account on which the exact amount of their savings is stored, e.g. employees who have a defined benefit occupational pension plan have most likely not an account information.

Figure 10: Composition of mean household financial assets



Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted.

Looking at the conditional figures reveals interesting differences (Table 17). Almost every household holds a sight account; still almost three quarters of households have a savings account, while 19% hold mutual funds. The average amount on sight accounts is €10,000 and as expected represents the lowest average value of all components. Luxembourg households with a savings account hold on average almost €40,000 in their savings account. The estimated median is however much lower at almost €15,000. The mean wealth invested in mutual funds is €95,000 and the median €27,000.

Table 17: Conditional mean and median household financial assets

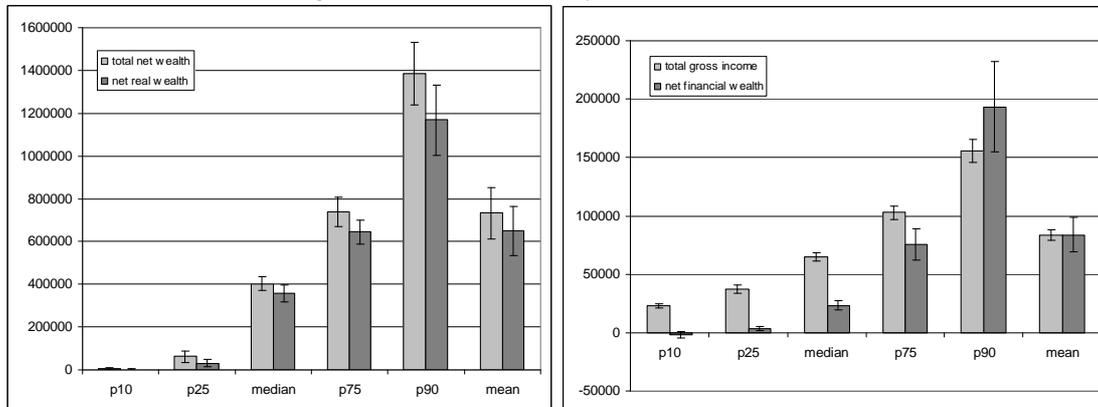
Financial asset category	Mean	Std. err.	[95% conf. interval]		Subgroup size	Share of total population	Median
sight accounts	10358	905	8574	12141	179521	96%	3209
savings accounts	39695	3179	33452	45937	134776	72%	14742
mutual funds	95301	15844	64084	126518	35247	19%	26870
bonds	122984	50657	23690	222277	8189	4%	45773
shares	63874	18055	28485	99262	18503	10%	10800
managed accounts	30032	45434	-59016	119080	546	0%	3000
amount owned to household	27287	10836	6049	48524	13291	7%	3572
any other financial assets	35641	13295	7661	63621	2592	1%	24214
value of non-self-empl. not publicly traded businesses	64320	35722	-6034	134674	1265	1%	6636
private pension wealth	49124	6652	36086	62161	63903	34%	27831
total financial assets	88424	7465	73788	103060	186440	100%	26653
non mortgage debt	12641	962	10756	14526	68797	37%	8092
net financial assets	83753	7474	69100	98405	186440	100%	23394

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights.

Household participation of holding bonds and shares is estimated at 4% and 10% respectively. The conditional mean (median) are estimated to be more than €120,000 (€45,000) and more than €60,000 (€10,000) for bonds and shares, respectively. The conditional mean of private pension wealth (median) (held by about one third of households) is estimated to be almost €50,000 (€28,000).

The distribution of total net and net real wealth reveal skewness, as would be expected (Figure 11, see as well Table 21 in Appendix B for exact figures). For example, total net wealth and net real wealth almost double (1.8) between the 50th and the 75th percentiles. Between the 25th and the 50th percentile, they increase by a factor 6.4 and 11.6, respectively. Gross income in contrast increases by a steady factor of 1.5-1.7 between the 10th and the 25th percentile, the interquartile ranges between the 25th, 50th and the 75th percentile and the 75th and 90th percentile.

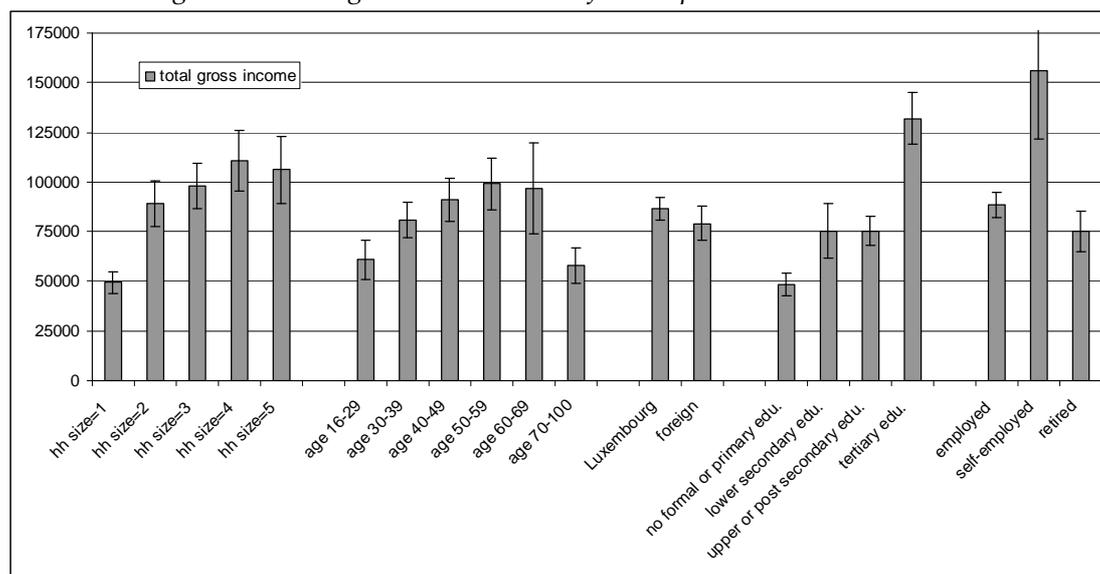
Figure 11: Distribution of income and wealth



Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights; the 95% confidence intervals are included in the figure.

Next, we report some aggregate statistics for total gross income and net wealth over various household characteristics. Average total gross income generally increases with household size, with age up to retirement age and with education (Figure 12, see also Table 22 in Appendix B for exact figures). Furthermore, after reaching retirement age average total gross income is much lower which is due to receiving a pension income. At aggregate level, foreign households do not seem to receive a significantly different average total gross income than Luxembourg households, where nationality is based on the reference person for the household. Lastly, self-employed households, where self-employment is considered the primary income source, earn on average 75% more than employees.

Figure 12: Total gross income over reference persons' characteristics¹⁶



Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights; the 95% confidence intervals are included in the figure.

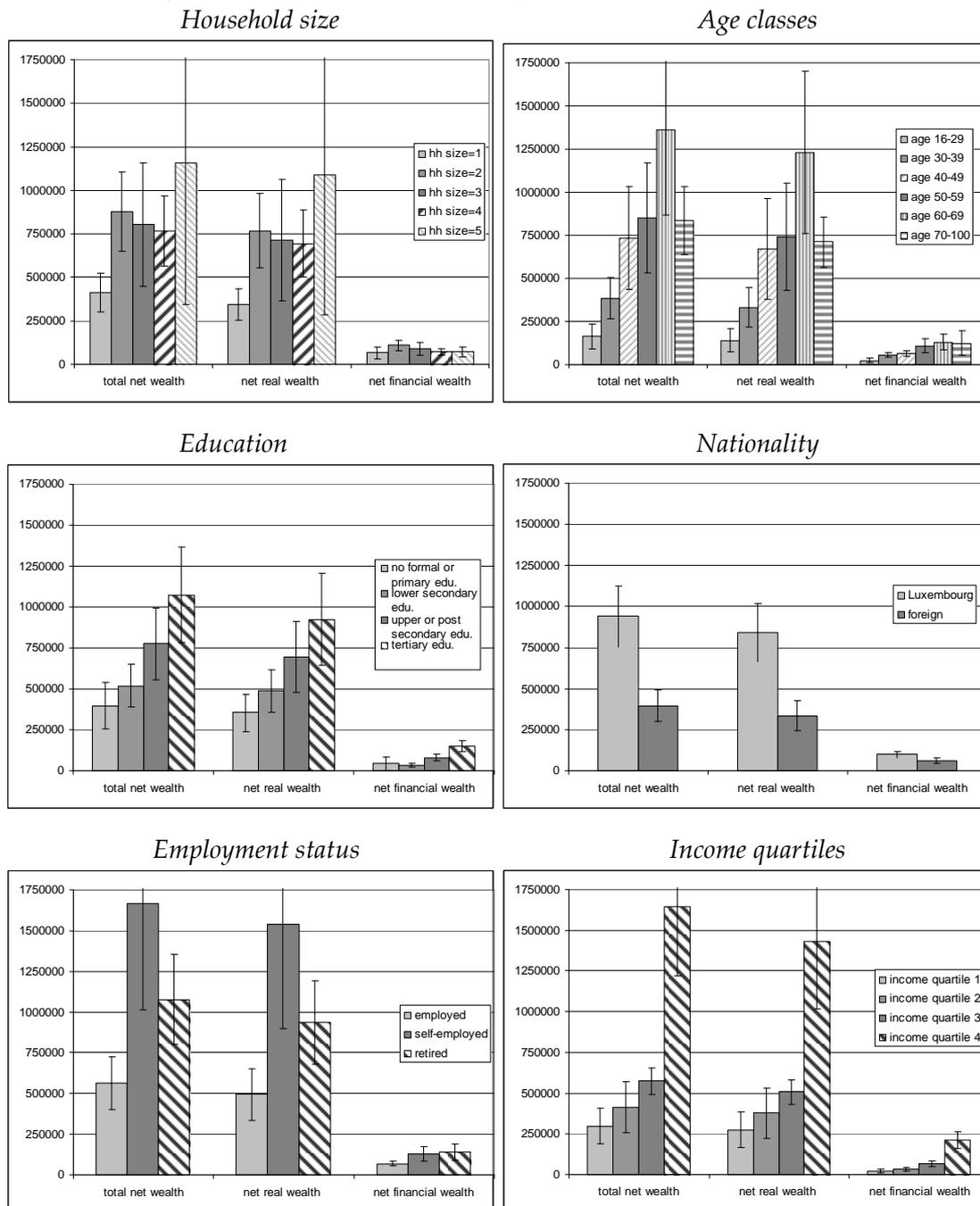
With regard to net wealth and its components across various household characteristics, the figures reveal the following (Figure 13, see also Table 22 in Appendix B for exact figures). Average total net wealth and average net real wealth show a strong increase from a one person to a two person household. No significant difference is observed between a two person household and household having three or four household members. The point estimate increases again for households with a household size of more than four members.

Similarly to gross income, average total net wealth and average net real wealth tend to rise with age (up to retirement) and the education. This is also the case for average net financial wealth, but to a much lower extent. For instance, total net wealth is around €400,000 for households having no formal or primary education. This increases to €1,075,000 for households in the highest education group.

¹⁶ HFCS reference person is the most knowledgeable person on financial matters of the household (or a proxy if the person is not living in the household).

Total net wealth and net real wealth of Luxembourg nationals and foreign households differ, with Luxembourg household holding about more than double as much total net wealth and real net wealth on average. The difference between Luxembourg nationals and foreign households is highly significant and this is despite the fact that no significant difference with respect to gross income is observed. The effect might be mainly driven by bequests, which are expected to be much higher for households of Luxembourg nationality.

Figure 13: Mean net wealth over different household characteristics



Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights; the 95% confidence intervals are included in the figure.

There are significant differences in average total net, net real, and net financial wealth between employed households and self-employed households or retired households, where the larger wealth of the self-employed is driven by business wealth and the higher wealth of retirees by a longer accumulation phase over their working life. The income effect on wealth stands out very clearly in these pictures. Whereas there seems to be an almost equidistant increase in average total net wealth and average net real wealth between the first, second and third income quartile, there is a huge jump for the fourth income quartile, in particular with regard to net real wealth and less so for net financial wealth.

7 Comparison with aggregate statistics

Despite micro data offering great benefits compared to aggregate statistics, such as national or financial accounts, it is well known that total wealth, especially financial wealth, is normally underreported. Often the aggregate statistics are viewed as a benchmark to evaluate the bias in survey statistics. Errors in measurement and errors in estimation based on micro data might cause such a bias. However, linking survey data to aggregate statistics is only partly possible due to different viewpoints of the statistics. Kavonius and Törmälehto (2010) examine the linkage between the HFCS and the integrated Euro Area Accounts (EAA). Based on their consideration and taking the Luxembourg specific situation into account, Table 18 compares total debt, total financial wealth and their components between the LU-HFCS and financial accounts. Statec and BCL published financial accounts data for Luxembourg according to the European system of accounts (ESA95) for the first time in April 2011 (Karpen, 2011). The comparison category by category is sometimes possible only to a limited extend due to the different coverage of certain variables in the LU-HFCS compared to financial accounts.

Table 18: Comparison between the LU-HFCS and financial accounts

variable HFCS	category financial accounts		HFCS	financial accounts		asset category as fraction of	
			in Euro	only pr. hhs in Euro	HFCS in % of fin. acc.	total debt/financial assets HFCS	financial accounts
<i>Total debt</i>	AF.4	Loans	78407	101770	77%	100%	100%
Total mortgage debt	AF.42	Long-term loans (> 1 year)	73736	97867	75%	94%	96%
Non mortgage debt	AF.41	Short-term loans (< 1 year)	4671	3903	120%	6%	4%
<i>Total financial assets</i>			93201	266114	35%	100%	100%
-- n.a. --	AF.21	Currency	-- n.a. --	9942	-- n.a. --	-- n.a. --	4%
Sight accounts & savings accounts & amount owned to household	AF.22+AF.29	Deposits	40600	119587	34%	44%	45%
Mutual funds	AF.4	Loans	18143	26472	69%	19%	10%
Bonds	AF.52	Mutual fund shares	5431	37430	15%	6%	14%
Shares & managed accounts	AF.3	Securities other than shares	6453	11378	57%	7%	4%
Value of non-self-employment not publicly traded businesses	AF.511	Quoted shares	446	22739	2%	0%	9%
Private/occupational pension wealth	AF.512+513	Unquoted equity	21622	38567	56%	23%	14%
Any other financial assets	AF.6	Insurance technical reserves	506	0		1%	0%
	AF.7	Other accounts receivable					

Source: Own calculations based on LU-HFCS and financial accounts data from Statec and BCL. The financial accounts data are from the end of 2010, which is the middle of the field phase of the LU-HFCS. Due to the fact that private households and non-profit institutions serving households (NPISHs) are combined in the financial accounts, BCL estimates are used to obtain the financial accounts for private households alone (pr. hhs). The financial accounts asset categories are divided by 204,900 households (number of resident households in Luxembourg in 2010; Eurostat) to obtain wealth per household.

Total debt is relatively well measured in the LU-HFCS. The average household in the HFCS has a total debt of around €78,000 representing 77% of debt reported in the financial accounts. Comparing the differences between the LU-HFCS and the financial accounts with differences for other countries, the LU-HFCS debt estimates seem to be closer to the financial accounts compared to most other countries (Table 19). The financial accounts report a value for total financial assets of approximately €266,000 per household and the LU-HFCS estimates a €93,000. The covered fraction of 35% seems low and indicates a strong underreporting of financial wealth. Another reason might be the coverage of the LU-HFCS survey, which excludes international civil servants and institutionalized individuals. The discrepancy will decrease if financial wealth holdings of the excluded group are higher. Compared to survey data in other countries (Table 19), a 35% coverage in Luxembourg is among the higher ones. The coverage of single financial asset categories varies strongly. Sight and saving accounts as well as amounts owned to other household cover 34% of the financial account categories “deposits” plus “loans”. Mutual funds and shares (+ managed accounts) are covered relatively well with 69% and 57% respectively. Badly covered are the value of non-self-employment not publicly traded businesses with 2% and bonds with 15%. Private and occupational pension wealth, which includes as well life-insurances, has coverage of 56%. Part of the under-coverage can be explained by the fact that the values of occupational pensions with defined benefit contracts are not included in the survey. As the distribution of the underreporting is relatively unequal over categories, the composition of total financial wealth differs between the financial accounts and the LU-HFCS (see the last two columns in Table 18). The largest part of financial wealth, namely the sight and saving accounts, is relatively well reflected. With 44% in the LU-HFCS and 45% in the financial accounts, the fraction of deposits is relatively large by international standards (BCL, 2011, p. 43).

Table 19: International comparison of wealth in survey data and aggregate statistics

Country	Canada	Italy	UK	USA		Germany	Luxembourg
Survey	SFS	SHIW	BHPS	PSID	SCF	SOEP	HFCS
Year	1999	2002	2000	2001	2001	2002 (A-G)	2010/11
Non-financial assets	87%	65%	91%	95%	117%	98%	-*
Financial assets	16%	18%	13%	25%	38%	34%	35%
Debt	69%	37%	66%	67%	86%	93%	74%
Net wealth	38%	48%	44%	46%	61%	69%	-*

*Source: Table based on Frick et al. (2007, figure 8 and the references therein) and own calculations based on LU-HFCS. Wealth figures provided are per capita in contrast to Table 18 which reports average wealth on a household level. The small differences observed for the fraction of wealth covered in Luxembourg is based on a small difference in average household size between Eurostat (2.45) and the LU-HFCS (2.48). Figures for other countries are from the Luxembourg Wealth Study (α -version for Germany and β -version for other countries). * Due to the lack of appropriate aggregate national accounts data a comparison of non-financial assets to aggregate statistics is not possible.*

Whereas financial account data can be considered as rather precise since it is based on different banking statistics (among others security by security reporting), the balance sheet of the Banque centrale du Luxembourg, statistics of the Bank for International Settlements, insurance statistics and so on, the produced assets of the resident Luxembourg population is calculated by the perpetual-inventory-method

(Schmalwasser and Schidlowski, 2006). Statec uses the perpetual-inventory-method to transform yearly flows of gross fixed capital formation into produced assets using assumptions on average operating time and the functional form of depreciations (Maria and Ciccone, 2006).¹⁷

In principle it is very difficult to compare the estimate of total real wealth to the corresponding macro figure. Produced assets amount to €45.584 billion at the end of 2010.¹⁸ Dividing this value by the number of 204,900 households in 2010 (Eurostat), results in €222,469 for each household (thereafter macro figure). This value is well below the estimated mean total real wealth of households estimated by the LU-HFCS (Table 14), which was around €720,000 (thereafter micro figure). Subtracting vehicles and valuables from this micro figure since durable consumption goods are not included into the macro figure reduces total real wealth to €690,000, still leaving a big part of the difference unexplained. Business wealth is included in both the macro and micro figure. The main reason for the discrepancy between the two figures is that the value of dwellings is included in both figures but the value of land is only included in the micro figure. However, the square meter prices for land in Luxembourg are expensive. Unfortunately, a reliable data source is not available. In the LU-HFCS, housing wealth cannot be separated into value of the dwelling and the land since the survey question refers to the current value of both the dwelling and the land combined. Moreover, renovations and improvements can prolong the average time of use, which is an important increment to calculate the macro figure. Due to a lack of data, these data have to be estimated by Statec. An underestimation of the effect of renovations and improvements cannot be ruled out, which biases the value of produced assets downwards.

The “*Observatoire de l’Habitat*” publishes asking and transaction prices for houses and apartments in Luxemburg. The current value of the household main residence as provided by weighted average of the LU-HFCS is around €640,000 with an average size of 158m² (Table 20). If the household main residence (HMR) is a house, semidetached house or townhouses, the average price is almost €715,000. The average size is 178m². The average value of an apartment is €400,000 and the size is nearly 100m². Comparing the mean price/m² they are almost equal (with €4,023/m² for houses and €4,174/m² for apartments). Table 20 compares these values to the asking prices of the values reported in the “*Observatoire de l’Habitat*” (2011). The mean price of a house is reported to be around €569,000 in 2010, which is substantially lower than the estimate based on the LU-HFCS. The reported house size is more or less equal, which leads into the price of €3,220/m² for houses. The difference in the prices/m² as reported in the LU-HFCS and the numbers presented

¹⁷ Statec assumes an average operating time for houses of 55 years.

¹⁸ The preferred macro figure would have been the stock of fixed assets, which includes produced assets and non-produced assets (land, subsoil assets, non-cultivated biological resources and water resources, as well as intangible non-produced assets). For Luxembourg only produced assets are available, which include tangible fixed assets such as dwellings, other buildings and structures, machinery and equipment, and cultivated assets. In addition it contains produced intangible fixed assets such as mineral exploration, computer software, entertainment, literary or artistic originals.

by the “*Observatoire de l’Habitat*” could reflect a persistent overestimation of house prices by LU-HFCS respondents. Since land is very expensive in Luxembourg, another reason accounting for the part of the discrepancy could be that newly constructed houses are constructed on relatively smaller properties than previously. Thus, household owners of our sample report higher values for their HMR due their larger average property sizes.

Table 20: Comparison of real estate prices

Wealth category	LU-HFCS						Observatoire de l’Habitat			
	Mean	Std. err.	Subgroup size	Population share	% of HMR owners	Sqm	Price / sqm	Mean	Sqm	Price / sqm
Main residence	640571	50778	125116	67%		158	4050			
House	714087	67091	93644	50%	75%	178	4023	569200	175	3220
Apartment	400232	27408	29822	16%	24%	96	4174	350600	88	3990

Source: Own calculations based on LU-HFCS and the *Observatoire de l’Habitat* (2011). Households owning an agricultural household main residence or another kind of household main residence are excluded from the LU-HFCS estimates (1% of population). Data of the *Observatoire de l’Habitat* display the average asking prices for houses and apartments.

This view can be supported with the observation that price/m² of apartments in the LU-HFCS and the mean value provided by “*Observatoire de l’Habitat*” are with €4,174/m² and €3,990/m² respectively much closer to each other. Actual transaction prices are available for apartments only (*Observatoire de l’Habitat*, 2011) and range between €3,715/m² for existing apartments and €4,558/m² for future apartments to be built in 2010.

Gross disposable income (B6G) in Luxembourg first published at the end of 2009 is estimated to be €14.911 billion for 2009 (Eurostat).¹⁹ In order to obtain an aggregate figure which is roughly comparable to household gross income of the LU-HFCS in 2009, the following components are added to B6G: Property income paid (D4), current taxes on income, wealth, etc. (D5), social contributions (D61) (only employee, self-employed and unemployment contributions), net non-life insurance premiums paid (D71) minus non-life insurance claims received (D72), and miscellaneous current transfers paid (D75). This increases the aggregate figure to €19.635 billion (see Appendix C). On average this amounts to €97,205 per household. Based on LU-HFCS data, total household income is estimated to be €83,641 in 2009, which is about €13,500 or 14% less. The difference can be attributed to underreporting in the LU-HFCS (e.g. excluding payments for overtime or holiday pay-outs) and income of non-profit institutions serving households in the national accounts.

In summary, households seem to have a fair notion of their income and debt level. Total financial assets are strongly underreported. The results are in line with the wealth coverage in surveys of other countries. Real estate holdings seem to be over-reported in case of houses as the disparate characteristics of the dwelling and the

¹⁹ The figure combines both private households sector (S.14) and non-profit institutions serving households (S. 15). This leads to an upward bias of private households’ gross income with respect to the aggregate figure.

land are only inadequately taken into account. Estimates of apartment values seem to be close to the one reported in external sources.

8 Final remarks

This paper introduced the Luxembourg Household Finance and Consumption Survey (LU-HFCS), presented some background information, its general objectives, the data collection, data treatment and some first descriptive statistics with respect to Luxembourg households' income and wealth position and their distribution across different household characteristics. With respect to wealth and income the data confirms the results of aggregate statistics, which indicate the high income level in Luxembourg and the impressive net wealth accumulated by resident households (e.g. OECD, 2011). The main results are as follows:

- 1) The estimated average (median) total net wealth of Luxembourg households is about €733,000 (€403,000) and is highly skewed to the right. This is likely to be driven by high incomes and correspondingly high prices of real estate in Luxembourg. Average total net wealth and average net real wealth generally tend to increase with household size, age and education.
- 2) On average, Luxembourg households are estimated to hold about €723,000 in total real wealth and about €88,400 in total financial wealth. The average total debt of Luxembourg households is about €78,400. With a share of 89%, the predominant part of average total gross household wealth consists of real wealth, of which the main part with 59% consists of the value of the household main residence. This is also reflected in the relatively high share of 67% of households owning (in full or in part their) their main residence.
- 3) Almost every Luxembourg resident household holds a sight account and almost three quarters of households have a savings account, while 19% hold mutual funds. 32%, 21%, 19% and 11% of Luxembourg households' financial wealth is held in form of savings accounts, mutual funds, private pension wealth and sight accounts. Household participation of holding bonds and shares is estimated at a rather low 4% and 10% respectively.
- 4) Total gross household income is on average almost €84,000. Similarly to total net wealth and total real wealth it generally increases with household size, with age up to retirement age and with education. After reaching retirement age average total gross income is lower which is due to receiving a pension income.
- 5) Average total gross income of Luxembourg national and foreign households do not seem to differ significantly. This cannot be said for total net wealth and net real wealth, which, on average, is more than double as high for Luxembourg households than for foreign households.
- 6) A comparison with aggregate statistics from the Luxembourg national and financial accounts reveals that households seem to have a fair notion of their income, debt and real estate level. However, total financial assets are strongly underreported, as is the case in surveys of other countries.

The value added of the LU-HFCS is that it allows a detailed breakdown of wealth and income across different household characteristics. Additionally, distributional aspects can be investigated. While having only scratched the surface, it is clear that these kinds of data offer a tremendous wealth of information allowing more detailed analysis of distributional aspects on households' income and wealth that is in the interest of central banks and the public in general. Here, the comparison of the Luxembourg households' income and wealth positions and their distributions with other euro area countries will provide fruitful insights.

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Appendix A: Number of edited and imputed values

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	edited
HB0100	size of household main residence	0.0%	100.0%	0.0%	1.3%	1.3%	0.0%	0.0%	0.1%
HB0200	how long have you been living in the household main residence	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HB0300	main residence - tenure status	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
HB0400	is rent paid for partially owned household main residence	99.3%	0.7%	0.0%	0.0%	0.0%			0.0%
HB0410	amount of rent paid for partially owned household main residence	100.0%	0.0%	0.0%					
HB0500	% of ownership of household main residence	99.3%	0.7%	0.0%	14.3%	14.3%	0.0%	0.0%	0.0%
HB0600	way of acquiring property	30.0%	70.0%	0.0%	0.0%	0.0%			0.0%
HB0700	year of property acquisition	30.0%	70.0%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%
HB0800	property value at the time of its acquisition	30.0%	70.0%	0.0%	5.9%	5.9%	56.4%	56.4%	0.2%
HB0900	current price of household main residence	30.0%	70.0%	0.0%	11.7%	11.7%	75.6%	75.6%	0.0%
HB1000	mortgages or loans using HMR as collateral	30.0%	70.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.3%
HB1010	number of mortgages or loans using HMR as collateral	65.5%	34.4%	0.1%	0.3%	0.6%	0.0%	0.0%	0.9%
HB1101	HMR mortgage \$x: loan refinancing	65.5%	34.4%	0.1%	0.0%	0.3%			0.3%
HB1102	HMR mortgage \$x: loan refinancing	94.8%	4.9%	0.2%	0.0%	4.3%			0.0%
HB1151	HMR mortgage \$x: country of bank or credit institute	65.5%	34.4%	0.1%	0.3%	0.6%	0.0%	0.0%	0.3%
HB1152	HMR mortgage \$x: country of bank or credit institute	94.8%	4.9%	0.2%	0.0%	4.3%			0.0%
HB1201A	HMR mortgage \$x: purpose of the loan	65.5%	34.4%	0.1%	0.0%	0.3%			0.3%
HB1201B	HMR mortgage \$x: purpose of the loan	95.7%	4.2%	0.1%	0.0%	2.5%			0.0%
HB1201C	HMR mortgage \$x: purpose of the loan	99.1%	0.8%	0.1%	0.0%	12.5%			0.0%
HB1201D	HMR mortgage \$x: purpose of the loan	99.5%	0.4%	0.1%	0.0%	25.0%			0.0%
HB1201E	HMR mortgage \$x: purpose of the loan	99.7%	0.2%	0.1%	0.0%	50.0%			0.0%
HB1201F	HMR mortgage \$x: purpose of the loan	99.9%	0.0%	0.1%					0.0%
HB1201G	HMR mortgage \$x: purpose of the loan	99.9%	0.0%	0.1%					0.0%
HB1201H	HMR mortgage \$x: purpose of the loan	99.9%	0.0%	0.1%					0.0%
HB1201I	HMR mortgage \$x: purpose of the loan	99.9%	0.0%	0.1%					0.0%
HB1202A	HMR mortgage \$x: purpose of the loan	94.8%	4.9%	0.2%	2.1%	6.4%	0.0%	0.0%	0.0%
HB1202B	HMR mortgage \$x: purpose of the loan	99.6%	0.2%	0.2%	0.0%	100.0%			0.0%
HB1202C	HMR mortgage \$x: purpose of the loan	99.7%	0.1%	0.2%	0.0%	200.0%			0.0%
HB1202D	HMR mortgage \$x: purpose of the loan	99.8%	0.0%	0.2%					0.0%
HB1202E	HMR mortgage \$x: purpose of the loan	99.8%	0.0%	0.2%					0.0%
HB1202F	HMR mortgage \$x: purpose of the loan	99.8%	0.0%	0.2%					0.0%
HB1202G	HMR mortgage \$x: purpose of the loan	99.8%	0.0%	0.2%					0.0%
HB1202H	HMR mortgage \$x: purpose of the loan	99.8%	0.0%	0.2%					0.0%
HB1202I	HMR mortgage \$x: purpose of the loan	99.8%	0.0%	0.2%					0.0%
HB1301	HMR mortgage \$x: year when loan taken or refinanced	65.5%	34.4%	0.1%	0.6%	0.9%	0.0%	0.0%	0.3%
HB1302	HMR mortgage \$x: year when loan taken or refinanced	94.8%	4.9%	0.2%	2.1%	6.4%	0.0%	0.0%	0.0%
HB1401	HMR mortgage \$x: initial amount borrowed	65.5%	34.1%	0.4%	6.8%	8.0%	54.5%	46.2%	1.5%
HB1402	HMR mortgage \$x: initial amount borrowed	94.8%	4.9%	0.2%	2.1%	6.4%	0.0%	0.0%	2.1%
HB1501	HMR mortgage \$x: additional borrowing on the HMR mortgage	65.5%	34.4%	0.1%	0.3%	0.6%	0.0%	0.0%	0.3%
HB1502	HMR mortgage \$x: additional borrowing on the HMR mortgage	94.8%	4.9%	0.2%	0.0%	4.3%			0.0%
HB1601	HMR mortgage \$x: length of the loan at the time of borrowing/refinancing	65.5%	34.0%	0.5%	0.0%	1.5%			1.2%
HB1602	HMR mortgage \$x: length of the loan at the time of borrowing/refinancing	94.8%	4.9%	0.2%	2.1%	6.4%	0.0%	0.0%	0.0%
HB1701	HMR mortgage \$x: amount still owed	65.5%	33.7%	0.8%	11.3%	13.8%	58.3%	47.7%	3.1%
HB1702	HMR mortgage \$x: amount still owed	94.8%	4.4%	0.7%	11.9%	28.6%	20.0%	8.3%	11.9%
HB1801	HMR mortgage \$x: adjustable interest rate	65.5%	34.4%	0.1%	0.3%	0.6%	0.0%	0.0%	0.3%
HB1802	HMR mortgage \$x: adjustable interest rate	94.8%	4.9%	0.2%	4.3%	8.5%	0.0%	0.0%	0.0%
HB1901	HMR mortgage \$x: current interest loan of the HMR mortgage	65.5%	34.4%	0.1%	12.2%	12.5%	0.0%	0.0%	0.3%
HB1902	HMR mortgage \$x: current interest loan of the HMR mortgage	94.8%	4.9%	0.2%	14.9%	19.1%	0.0%	0.0%	0.0%
HB2001	HMR mortgage \$x: monthly amount of payment made on loan	65.5%	34.2%	0.3%	4.9%	5.8%	50.0%	42.1%	0.9%
HB2002	HMR mortgage \$x: monthly amount of payment made on loan	94.8%	4.7%	0.4%	4.4%	13.3%	0.0%	0.0%	4.4%
HB2100	money still owed on additional HMR loans	98.5%	1.3%	0.2%	8.3%	25.0%	0.0%	0.0%	0.0%
HB2200	monthly amount of payment made on additional HMR loans	98.5%	1.2%	0.3%	9.1%	36.4%	0.0%	0.0%	9.1%
HB2300	monthly amount paid as rent	73.1%	26.9%	0.0%	0.8%	0.8%			1.2%
HB2400	household owns other properties than HMR	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HB2410	number of properties other than household main residence	65.8%	34.1%	0.1%	0.0%	0.3%			0.9%
HB2501	other property \$x: property type	65.8%	34.1%	0.1%	0.0%	0.3%			0.3%
HB2502	other property \$x: property type	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	0.0%
HB2511	other property \$x: In which country is real estate located?	65.8%	34.1%	0.1%	0.0%	0.3%			0.3%
HB2512	other property \$x: In which country is real estate located?	85.8%	14.0%	0.2%	0.0%	1.5%			1.5%
HB2601	other property \$x: for household own use	75.6%	24.3%	0.1%	0.0%	0.4%			0.4%
HB2602	other property \$x: for household own use	91.9%	7.9%	0.2%	1.3%	4.0%	0.0%	0.0%	0.0%
HB2611	other property \$x: for business use	76.0%	23.9%	0.1%	0.0%	0.4%			0.9%
HB2612	other property \$x: for business use	88.4%	11.4%	0.2%	0.0%	1.9%			0.0%
HB2621	other property \$x: leased or rented	78.7%	21.2%	0.1%	0.0%	0.5%			0.5%
HB2622	other property \$x: leased or rented	89.7%	10.1%	0.2%	0.0%	2.1%			0.0%
HB2631	other property \$x: for other use	91.8%	8.1%	0.1%	0.0%	1.3%			1.3%

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	edited
HB2632	other property \$x: for other use	96.0%	3.7%	0.3%	0.0%	8.6%		0.0%	0.0%
HB2701	other property \$x: % of the property belonging to household	65.8%	34.1%	0.1%	1.2%	1.5%	0.0%	0.0%	0.0%
HB2702	other property \$x: % of the property belonging to household	85.8%	14.0%	0.2%	1.5%	3.0%	0.0%	0.0%	0.8%
HB2801	other property \$x: current value	65.8%	34.1%	0.1%	12.7%	13.0%	34.1%	33.3%	0.3%
HB2802	other property \$x: current value	85.8%	14.1%	0.1%	16.4%	17.2%	54.5%	52.2%	0.0%
HB2900	additional properties current value	93.6%	6.2%	0.2%	20.3%	23.7%	25.0%	21.4%	3.4%
HB3000	mortgages or loans using other properties as collateral	65.8%	34.1%	0.1%	0.0%	0.3%		0.0%	0.9%
HB3010	number of mortgages or loans using other properties as collateral	88.2%	11.7%	0.1%	0.0%	0.9%		0.0%	1.8%
HB3021	Location of the first mortgage	88.2%	11.6%	0.2%	0.0%	1.8%		0.0%	0.9%
HB3022	Location of the second mortgage	97.2%	2.7%	0.1%	7.7%	11.5%	0.0%	0.0%	0.0%
HB3101	other property mortgages \$x: loan refinancing	88.2%	11.6%	0.2%	0.0%	1.8%		0.0%	0.9%
HB3102	other property mortgages \$x: loan refinancing	97.2%	2.7%	0.1%	7.7%	11.5%	0.0%	0.0%	0.0%
HB3201A	other property mortgage \$x: main purpose of the loan	88.2%	11.7%	0.1%	1.8%	2.7%	0.0%	0.0%	0.9%
HB3201B	other property mortgage \$x: main purpose of the loan	99.2%	0.7%	0.1%	0.0%	14.3%		0.0%	0.0%
HB3201C	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3201D	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3201E	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3201F	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3201G	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3201H	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3201I	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3202A	other property mortgage \$x: main purpose of the loan	97.2%	2.7%	0.1%	3.8%	7.7%	0.0%	0.0%	0.0%
HB3202B	other property mortgage \$x: main purpose of the loan	99.7%	0.2%	0.1%	0.0%	50.0%		0.0%	0.0%
HB3202C	other property mortgage \$x: main purpose of the loan	99.8%	0.1%	0.1%	0.0%	100.0%		0.0%	0.0%
HB3202D	other property mortgage \$x: main purpose of the loan	99.8%	0.1%	0.1%	0.0%	100.0%		0.0%	0.0%
HB3202E	other property mortgage \$x: main purpose of the loan	99.8%	0.1%	0.1%	0.0%	100.0%		0.0%	0.0%
HB3202F	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3202G	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3202H	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3202I	other property mortgage \$x: main purpose of the loan	99.9%	0.0%	0.1%				0.0%	
HB3301	other property mortgage \$x: year when loan taken or refinanced	88.2%	11.7%	0.1%	2.7%	3.6%	0.0%	0.0%	0.9%
HB3302	other property mortgage \$x: year when loan taken or refinanced	97.2%	2.7%	0.1%	11.5%	15.4%	0.0%	0.0%	0.0%
HB3401	other property mortgage \$x: initial amount borrowed	88.2%	11.6%	0.2%	11.8%	13.6%	38.5%	33.3%	1.8%
HB3402	other property mortgage \$x: initial amount borrowed	97.2%	2.7%	0.1%	15.4%	19.2%	0.0%	0.0%	0.0%
HB3501	other property mortgage \$x: additional borrowing	88.2%	11.7%	0.1%	1.8%	2.7%	0.0%	0.0%	0.9%
HB3502	other property mortgage \$x: additional borrowing	97.2%	2.7%	0.1%	7.7%	11.5%	0.0%	0.0%	0.0%
HB3601	other property mortgage \$x: length of the loan at the time of borrowing/refinancing	88.2%	11.7%	0.1%	3.6%	4.5%	0.0%	0.0%	0.9%
HB3602	other property mortgage \$x: length of the loan at the time of borrowing/refinancing	97.2%	2.7%	0.1%	11.5%	15.4%	0.0%	0.0%	0.0%
HB3701	other property mortgage \$x: amount still owed	88.2%	11.3%	0.5%	15.0%	19.6%	37.5%	28.6%	3.7%
HB3702	other property mortgage \$x: amount still owed	97.2%	2.7%	0.1%	19.2%	23.1%	0.0%	0.0%	0.0%
HB3801	other property mortgage \$x: adjustable interest rate	88.2%	11.7%	0.1%	5.4%	6.3%	0.0%	0.0%	0.9%
HB3802	other property mortgage \$x: adjustable interest rate	97.2%	2.7%	0.1%	11.5%	15.4%	0.0%	0.0%	0.0%
HB3901	other property mortgage \$x: current interest loan of the mortgage	88.2%	11.7%	0.1%	18.0%	18.9%	0.0%	0.0%	0.9%
HB3902	other property mortgage \$x: current interest loan of the mortgage	97.2%	2.7%	0.1%	23.1%	26.9%	0.0%	0.0%	0.0%
HB4001	other property mortgage \$x: monthly payment on loan	88.2%	11.6%	0.2%	9.1%	10.9%	70.0%	58.3%	0.9%
HB4002	other property mortgage \$x: monthly payment on loan	97.2%	2.7%	0.1%	19.2%	23.1%	40.0%	33.3%	0.0%
HB4100	money still owed on additional other property loans	99.2%	0.5%	0.3%	40.0%	100.0%	0.0%	0.0%	40.0%
HB4200	monthly payment on additional other property loans	99.2%	0.7%	0.1%	14.3%	28.6%	0.0%	0.0%	0.0%

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	edited
HB4300	ownership of cars	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
HB4310	number of cars	10.9%	89.1%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HB4400	total value of the cars	10.9%	89.1%	0.0%	6.0%	6.0%	70.6%	70.6%	0.0%
HB4500	ownership of other vehicles	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HB4510A	number of other vehicles	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	0.0%
HB4510B	number of other vehicles	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	0.0%
HB4510C	number of other vehicles	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	0.0%
HB4510D	number of other vehicles	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	0.0%
HB4510E	number of other vehicles	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	0.0%
HB4510F	number of other vehicles	85.8%	14.1%	0.1%	0.7%	1.5%	0.0%	0.0%	2.2%
HB4600	total value of other vehicles	85.8%	14.1%	0.1%	4.5%	5.2%	66.7%	57.1%	0.0%
HB4700	ownership of other valuables	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HB4710	value of other valuables	70.6%	28.8%	0.5%	33.2%	35.0%	57.1%	54.2%	0.4%
HC0100	household has a leasing contract	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HC0110	monthly leasing payments	94.2%	5.7%	0.1%	7.4%	9.3%	25.0%	20.0%	0.0%
HC0200	household has credit line or overdraft	0.0%	100.0%	0.0%	0.7%	0.7%	0.0%	0.0%	0.1%
HC0210	household has outstanding credit line/overdraft balance	43.7%	55.6%	0.7%	0.2%	1.5%	0.0%	0.0%	0.0%
HC0220	amount of outstanding credit line/overdraft balance	90.1%	9.1%	0.8%	12.8%	22.1%	27.3%	15.8%	0.0%
HC0300	household has a credit card	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HC0310	household has outstanding balance on credit cards	14.4%	85.5%	0.1%	0.2%	0.4%	0.0%	0.0%	0.0%
HC0320	amount of outstanding credit cards balance	93.6%	6.1%	0.3%	10.3%	15.5%	50.0%	33.3%	0.0%
HC0400	has any non-collaterised loans	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.3%
HC0410	number of non-collaterised loans	69.3%	30.7%	0.0%	0.3%	0.3%	0.0%	0.0%	1.0%
HC0501A	non-collaterised loan \$x: purpose of the loan	69.3%	30.7%	0.0%	0.0%	0.0%	0.0%	0.0%	1.7%
HC0501B	non-collaterised loan \$x: purpose of the loan	97.8%	2.2%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501C	non-collaterised loan \$x: purpose of the loan	99.5%	0.5%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501D	non-collaterised loan \$x: purpose of the loan	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501E	non-collaterised loan \$x: purpose of the loan	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501F	non-collaterised loan \$x: purpose of the loan	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501G	non-collaterised loan \$x: purpose of the loan	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501H	non-collaterised loan \$x: purpose of the loan	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0501I	non-collaterised loan \$x: purpose of the loan	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502A	non-collaterised loan \$x: purpose of the loan	92.4%	7.5%	0.1%	1.4%	2.8%	0.0%	0.0%	0.0%
HC0502B	non-collaterised loan \$x: purpose of the loan	99.4%	0.5%	0.1%	20.0%	40.0%	0.0%	0.0%	0.0%
HC0502C	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502D	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502E	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502F	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502G	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502H	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0502I	non-collaterised loan \$x: purpose of the loan	99.9%	0.0%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HC0551	non-collaterised loan \$x: Which institution granted the credit?	69.3%	30.5%	0.2%	0.7%	1.4%	0.0%	0.0%	1.0%
HC0552	non-collaterised loan \$x: Which institution granted the credit?	92.4%	7.5%	0.1%	2.8%	4.2%	0.0%	0.0%	0.0%
HC0601	non-collaterised loan \$x: amount initially borrowed	69.3%	30.5%	0.2%	4.1%	4.8%	66.7%	57.1%	1.7%
HC0602	non-collaterised loan \$x: amount initially borrowed	92.4%	7.2%	0.4%	10.3%	16.2%	42.9%	27.3%	4.4%
HC0701	non-collaterised loan \$x: initial length of the loan	69.3%	30.6%	0.1%	5.5%	5.8%	0.0%	0.0%	1.0%
HC0702	non-collaterised loan \$x: initial length of the loan	92.4%	7.5%	0.1%	9.9%	11.3%	0.0%	0.0%	0.0%
HC0801	non-collaterised loan \$x: outstanding balance of loan	69.3%	30.3%	0.4%	8.7%	10.1%	60.0%	51.7%	1.7%
HC0802	non-collaterised loan \$x: outstanding balance of loan	92.4%	7.5%	0.1%	12.7%	14.1%	22.2%	20.0%	0.0%
HC0901	non-collaterised loan \$x: current interest rate of loan	69.3%	30.6%	0.1%	25.8%	26.1%	0.0%	0.0%	1.0%
HC0902	non-collaterised loan \$x: current interest rate of loan	92.4%	7.5%	0.1%	21.1%	22.5%	0.0%	0.0%	0.0%
HC1001	non-collaterised loan \$x: monthly payment on loan	69.3%	30.6%	0.1%	7.2%	7.6%	38.1%	36.4%	1.0%
HC1002	non-collaterised loan \$x: monthly payment on loan	92.4%	7.5%	0.1%	11.3%	12.7%	0.0%	0.0%	0.0%
HC1100	total amount owed for additional non-collaterised loans	98.3%	1.6%	0.1%	6.7%	13.3%	0.0%	0.0%	0.0%
HC1200	monthly payment on additional non-collaterised loans	98.3%	1.6%	0.1%	13.3%	20.0%	50.0%	33.3%	0.0%
HC1300	has applied for loan/credit	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	2.3%
HC1310	was refused credit	58.6%	39.1%	2.3%	0.3%	6.2%	0.0%	0.0%	5.9%
HC1320	re-applying for credit	95.5%	2.1%	2.4%	0.0%	115.0%	0.0%	0.0%	110.0%
HC1400	not applying for credit due to perceived credit constrain	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HD0100	investments in businesses not publicly traded	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HD0200	investments in self-employment businesses	90.3%	9.6%	0.1%	0.0%	1.1%	0.0%	0.0%	0.0%
HD0210	how many self-employment businesses	91.2%	8.7%	0.1%	0.0%	1.2%	0.0%	0.0%	0.0%
HD0301	business \$x: NACE	91.2%	8.7%	0.1%	0.0%	1.2%	0.0%	0.0%	0.0%
HD0302	business \$x: NACE	97.9%	2.0%	0.1%	5.3%	10.5%	0.0%	0.0%	0.0%
HD0401	business \$x: legal form of the business	91.2%	8.7%	0.1%	0.0%	1.2%	0.0%	0.0%	0.0%
HD0402	business \$x: legal form of the business	97.9%	2.0%	0.1%	0.0%	5.3%	0.0%	0.0%	0.0%
HD0501	business \$x: number of employees	91.2%	8.7%	0.1%	0.0%	1.2%	0.0%	0.0%	0.0%
HD0502	business \$x: number of employees	97.9%	1.7%	0.4%	0.0%	25.0%	0.0%	0.0%	18.8%
HD0601A	business \$x: household members working in the business	91.2%	8.7%	0.1%	2.4%	3.6%	0.0%	0.0%	20.5%

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HD0601B	business \$x: household members working in the business	98.1%	1.8%	0.1%	0.0%	5.9%		0.0%	100.0%
HD0601C	business \$x: household members working in the business	99.6%	0.3%	0.1%	0.0%	33.3%		0.0%	100.0%
HD0601D	business \$x: household members working in the business	99.8%	0.1%	0.1%	0.0%	100.0%		0.0%	100.0%
HD0601E	business \$x: household members working in the business	99.9%	0.0%	0.1%				0.0%	
HD0601F	business \$x: household members working in the business	99.9%	0.0%	0.1%				0.0%	
HD0602A	business \$x: household members working in the business	97.9%	2.0%	0.1%	21.1%	26.3%	0.0%	0.0%	21.1%
HD0602B	business \$x: household members working in the business	99.5%	0.4%	0.1%	0.0%	25.0%		0.0%	100.0%
HD0602C	business \$x: household members working in the business	99.9%	0.0%	0.1%				0.0%	
HD0602D	business \$x: household members working in the business	99.9%	0.0%	0.1%				0.0%	
HD0602E	business \$x: household members working in the business	99.9%	0.0%	0.1%				0.0%	
HD0602F	business \$x: household members working in the business	99.9%	0.0%	0.1%				0.0%	
HD0701	business \$x: % of household ownership	91.2%	8.6%	0.2%	1.2%	3.7%	0.0%	0.0%	2.4%
HD0702	business \$x: % of household ownership	97.9%	1.9%	0.2%	0.0%	11.1%		0.0%	5.6%
HD0801	business \$x: value of the business	91.2%	8.7%	0.1%	48.2%	49.4%	32.5%	31.7%	0.0%
HD0802	business \$x: value of the business	97.9%	1.9%	0.2%	33.3%	44.4%	16.7%	12.5%	5.6%
HD0900	value of additional businesses	98.9%	0.9%	0.1%	33.3%	44.4%	0.0%	0.0%	0.0%
HD1000	non-self-employment not publicly traded businesses	90.3%	9.6%	0.1%	1.1%	2.2%	0.0%	0.0%	0.0%
HD1010	value of non-selfemployment not publicly traded businesses	98.6%	1.2%	0.2%	45.5%	63.6%	40.0%	28.6%	0.0%
HD1100	household owns sight accounts	0.0%	100.0%	0.0%	0.0%	0.0%			0.0%
HD1110	value of sight accounts	3.2%	96.8%	0.0%	36.1%	36.1%	64.5%	64.5%	0.0%
HD1200	household owns savings accounts	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HD1210	value of saving accounts	25.6%	74.1%	0.3%	41.2%	41.6%	60.3%	59.7%	0.7%
HD1300	household owns investments in mutual funds	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.5%
HD1310A	types of mutual funds	75.4%	24.0%	0.6%	3.5%	6.1%	0.0%	0.0%	1.3%
HD1310B	types of mutual funds	75.4%	24.0%	0.6%	2.6%	5.3%	0.0%	0.0%	1.3%
HD1310C	types of mutual funds	75.4%	24.0%	0.6%	2.2%	4.8%	0.0%	0.0%	0.4%
HD1310D	types of mutual funds	75.4%	24.0%	0.6%	2.2%	4.8%	0.0%	0.0%	0.4%
HD1310E	types of mutual funds	75.4%	24.0%	0.6%	1.3%	3.9%	0.0%	0.0%	0.4%
HD1310F	types of mutual funds	75.4%	24.0%	0.6%	2.2%	4.8%	0.0%	0.0%	5.7%
HD1310G	types of mutual funds	75.4%	24.0%	0.6%	2.2%	4.8%	0.0%	0.0%	0.0%
HD1320A	market value of mutual funds	83.1%	15.5%	1.5%	34.0%	43.5%	54.0%	42.2%	2.0%
HD1320B	market value of mutual funds	91.4%	7.4%	1.3%	38.6%	55.7%	29.6%	20.5%	1.4%
HD1320C	market value of mutual funds	95.4%	3.5%	1.2%	48.5%	81.8%	56.3%	33.3%	0.0%
HD1320D	market value of mutual funds	98.0%	0.8%	1.2%	62.5%	200.0%	40.0%	12.5%	0.0%
HD1320E	market value of mutual funds	98.2%	0.8%	0.9%	12.5%	125.0%	0.0%	0.0%	0.0%
HD1320F	market value of mutual funds	93.8%	4.7%	1.5%	24.4%	55.6%	63.6%	28.0%	28.9%
HD1320G	market value of mutual funds	0.0%	0.0%	100.0%				0.0%	
HD1400	household owns bonds	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%
HD1410A	kind of bonds owned	93.2%	6.3%	0.5%	3.3%	11.7%	0.0%	0.0%	0.0%
HD1410B	kind of bonds owned	93.2%	6.3%	0.5%	0.0%	8.3%		0.0%	0.0%
HD1410C	kind of bonds owned	93.2%	6.3%	0.5%	1.7%	10.0%	0.0%	0.0%	0.0%
HD1410D	kind of bonds owned	93.2%	6.3%	0.5%	10.0%	18.3%	0.0%	0.0%	1.7%
HD1420	market value of bonds	93.2%	6.3%	0.5%	26.7%	35.0%	12.5%	9.5%	3.3%
HD1500	household owns publicly traded shares	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HD1510	value of publicly traded shares	86.5%	13.2%	0.3%	25.6%	28.0%	37.5%	34.3%	0.8%
HD1520	any shares issued by foreign companies	86.5%	13.2%	0.3%	0.0%	2.4%		0.0%	0.0%
HD1600	household owns managed accounts	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HD1610	managed accounts - assets not already recorded	96.8%	2.7%	0.4%	3.8%	19.2%	0.0%	0.0%	0.0%
HD1620	value of additional assets in managed accounts	99.2%	0.3%	0.5%	33.3%	200.0%		16.7%	0.0%
HD1700	does anyone owe money to household	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HD1710	amount owned to household	92.4%	7.5%	0.1%	5.6%	7.0%	50.0%	40.0%	0.0%
HD1800	investment attitudes	0.0%	100.0%	0.0%	1.7%	1.7%	0.0%	0.0%	0.0%
HD1900	any other financial assets	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HD1910	specification of other assets	97.6%	2.1%	0.3%	10.0%	25.0%	0.0%	0.0%	0.0%
HD1920	value of the other assets	97.6%	2.1%	0.3%	35.0%	50.0%	71.4%	50.0%	0.0%
HG0100	received income from public transfers	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.3%
HG0110	gross income from regular social transfers	57.9%	42.0%	0.1%	10.8%	11.0%	74.4%	72.7%	0.8%
HG0200	received income from regular private transfers	0.0%	100.0%	0.0%	0.0%	0.0%			0.1%
HG0210	income from regular private transfers	94.6%	5.4%	0.0%	13.7%	13.7%	85.7%	85.7%	0.0%
HG0300	received income from real estate property	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HG0310	gross rental income from real estate property	82.4%	17.5%	0.1%	16.3%	16.9%	44.4%	42.9%	0.0%
HG0400	received income from financial investment	0.0%	100.0%	0.0%	1.8%	1.8%	0.0%	0.0%	0.0%
HG0410	gross income from financial investments	48.9%	49.3%	1.8%	44.9%	48.5%	52.9%	48.9%	0.2%
HG0500	received income from private business other than self-employment	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HG0510	gross income from private business other than self-employment	98.3%	1.5%	0.2%	35.7%	50.0%	20.0%	14.3%	0.0%
HG0600	received income from other income source	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HG0610	gross income from other sources	97.3%	2.3%	0.4%	13.6%	31.8%	0.0%	0.0%	0.0%
HG0620	specification of other income source received	97.3%	2.3%	0.4%	4.5%	22.7%	0.0%	0.0%	0.0%
HG0700	is income 'normal' in reference period	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HG0800	future income expectations	0.0%	100.0%	0.0%	3.9%	3.9%	0.0%	0.0%	0.0%

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	edited
HH0100	any substantial gift or inheritance received	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.1%
HH0110	no of gifts/inheritances received	68.8%	30.5%	0.6%	1.4%	3.4%	0.0%	0.0%	0.7%
HH0201	gift/inheritance \$x: year gift/inheritance received	68.8%	30.3%	0.8%	1.0%	3.8%	0.0%	0.0%	0.7%
HH0202	gift/inheritance \$x: year gift/inheritance received	89.7%	9.3%	1.1%	2.3%	13.6%	0.0%	0.0%	1.1%
HH0301A	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301B	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301C	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301D	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.7%	3.1%	0.0%	0.0%	0.3%
HH0301E	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301F	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301G	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301H	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	0.3%	2.8%	0.0%	0.0%	0.3%
HH0301I	gift/inheritance \$x: what kind of assets received	68.8%	30.4%	0.7%	2.1%	4.5%	0.0%	0.0%	0.3%
HH0302A	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302B	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302C	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302D	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302E	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302F	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302G	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	1.1%	12.5%	0.0%	0.0%	0.0%
HH0302H	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	0.0%
HH0302I	gift/inheritance \$x: what kind of assets received	89.7%	9.3%	1.1%	3.4%	14.8%	0.0%	0.0%	0.0%
HH0401	gift/inheritance \$x: value	68.8%	30.4%	0.7%	24.6%	27.0%	50.7%	46.2%	0.7%
HH0402	gift/inheritance \$x: value	89.7%	9.3%	1.1%	12.5%	23.9%	45.5%	23.8%	1.1%
HH0501	gift/inheritance \$x: type of transfer (gift/inheritance)	68.8%	30.5%	0.6%	0.0%	2.1%		0.0%	0.3%
HH0502	gift/inheritance \$x: type of transfer (gift/inheritance)	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	1.1%
HH0601	gift/inheritance \$x: from whom received	68.8%	30.5%	0.6%	0.0%	2.1%		0.0%	0.3%
HH0602	gift/inheritance \$x: from whom received	89.7%	9.3%	1.1%	0.0%	11.4%		0.0%	1.1%
HH0700	expect to receive inheritance in the future	0.0%	100.0%	0.0%	2.4%	2.4%	0.0%	0.0%	0.0%
HI0100	amount spent on food at home	0.0%	99.9%	0.1%	7.1%	7.2%	82.1%	80.9%	0.8%
HI0101	amount spent on food at home: Luxembourg	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HI0102	amount spent on food at home: Belgium	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HI0103	amount spent on food at home: Germany	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HI0104	amount spent on food at home: France	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HI0105	amount spent on food at home: Others	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HI0102	amount spent on food at home: reason	0.0%	100.0%	0.0%	1.5%	1.5%	0.0%	0.0%	0.0%
HI0200	amount spent on food outside home	0.0%	100.0%	0.0%	4.7%	4.7%	80.0%	80.0%	0.7%
HI0201	amount spent on food outside home: graphical breakdown	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	6.8%
HI02012	amount spent on food outside home: Luxembourg	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	6.8%
HI02013	amount spent on food outside home: Belgium	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	6.8%
HI02014	amount spent on food outside home: Germany	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	6.8%
HI02015	amount spent on food outside home: France	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	6.8%
HI0202	amount spent on food outside home: Others	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	6.8%
HI0300	Makes private transfers to individuals out of household/charities (y/n)?	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%
HI0301	amount spent on non-durable consumption	0.0%	100.0%	0.0%	0.0%	0.0%			100.0%
HI03021	amount spent on non-durable consumption: Luxembourg	0.0%	100.0%	0.0%	1.7%	1.7%	0.0%	0.0%	0.0%
HI03022	amount spent on non-durable consumption: Belgium	0.0%	100.0%	0.0%	1.7%	1.7%	0.0%	0.0%	0.0%
HI03023	amount spent on non-durable consumption: Germany	0.0%	100.0%	0.0%	1.7%	1.7%	0.0%	0.0%	0.0%
HI03024	amount spent on non-durable consumption: France	0.0%	100.0%	0.0%	1.7%	1.7%	0.0%	0.0%	0.0%
HI03025	amount spent on non-durable consumption: Others	0.0%	100.0%	0.0%	1.7%	1.7%	0.0%	0.0%	0.0%
HI0303	reason	0.0%	100.0%	0.0%	3.1%	3.1%	0.0%	0.0%	0.0%
HI0310	amount given as private transfers per month	86.5%	13.4%	0.1%	3.9%	4.7%	60.0%	50.0%	2.4%
HI0400A	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400B	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400C	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400D	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	edited
HI0400E	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400F	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400G	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400H	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400I	purpose of saving	0.0%	100.0%	0.0%	2.5%	2.5%	0.0%	0.0%	0.0%
HI0400J	purpose of saving	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HI0400K	purpose of saving	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HI0400L	purpose of saving	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.2%
HI0500	comparison of last 12 months' expenses with average last 12 month expenses were below/above income	0.0%	100.0%	0.0%	1.2%	1.2%	0.0%	0.0%	0.0%
HI0600	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700A	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700B	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700C	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700D	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700E	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700F	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	0.0%
HI0700G	source of extra income to meet expenses	91.4%	7.9%	0.7%	8.0%	17.3%	0.0%	0.0%	4.0%
HI0800	ability to get financial assistance from friends or relatives	0.0%	100.0%	0.0%	3.8%	3.8%	0.0%	0.0%	0.0%
HNB0100	Farms/lf household main residence is a farm HMR mortgages: work for institution granting the loan	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
HNB1401	HMR mortgages: work for institution granting the loan	65.5%	34.4%	0.1%	0.0%	0.3%	0.0%	0.0%	0.3%
HNB1402	HMR mortgages: work for institution granting the loan	94.8%	4.9%	0.2%	0.0%	4.3%	0.0%	0.0%	0.0%
HNB3000	Reasons for moving	80.8%	18.7%	0.4%	0.6%	2.8%	0.0%	0.0%	0.0%
HNC0125	Late or missed payments on loans	69.3%	30.4%	0.3%	1.7%	2.8%	0.0%	0.0%	1.0%
HNC0126	Any outstanding overdue payments	98.3%	1.2%	0.5%	0.0%	45.5%	0.0%	0.0%	0.0%
HNC0210A	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210B	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210C	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	3.0%	6.1%	0.0%	0.0%	0.0%
HNC0210D	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210E	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	3.0%	6.1%	0.0%	0.0%	0.0%
HNC0210F	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	3.0%	6.1%	0.0%	0.0%	0.0%
HNC0210G	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210H	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210I	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210J	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	0.0%	3.0%	0.0%	0.0%	0.0%
HNC0210K	Reasons for not applying for credit due to perceived credit constrain	96.4%	3.5%	0.1%	9.1%	12.1%	0.0%	0.0%	3.0%
HND3000	Largest asset in HH balance sheet	0.0%	98.4%	1.6%	2.2%	3.9%	0.0%	0.0%	1.6%
HND3001	Second largest asset in HH balance sheet	12.5%	85.2%	2.3%	1.1%	3.8%	0.0%	0.0%	0.1%
HND3002	Third largest asset in HH balance sheet	27.6%	69.2%	3.3%	3.0%	7.8%	0.0%	0.0%	0.2%
HND3010	Portfolio shifts last two years?	12.5%	85.2%	2.3%	0.5%	3.2%	0.0%	0.0%	0.1%
HND3020	Portfolio shifts last two years: money out	83.7%	13.6%	2.7%	0.0%	20.2%	0.0%	0.0%	0.8%
HND3030	Portfolio shifts last two years: money in	83.7%	13.6%	2.7%	0.8%	20.9%	0.0%	0.0%	0.8%
HND3040	Would not invest again?	12.5%	85.2%	2.3%	3.6%	6.3%	0.0%	0.0%	0.1%
HND3050	Assets HH would not invest again	78.7%	15.8%	5.5%	0.0%	34.7%	0.0%	0.0%	1.3%
HND3100	Net worth past two years	0.0%	98.4%	1.6%	1.4%	3.0%	0.0%	0.0%	1.6%
HND3200	Net worth next two years	0.0%	98.4%	1.6%	3.6%	5.2%	0.0%	0.0%	1.7%
HNI0700	More or less savings in the next year	0.0%	100.0%	0.0%	6.3%	6.3%	0.0%	0.0%	0.0%
HNI0800	General price expectations	0.0%	100.0%	0.0%	1.4%	1.4%	0.0%	0.0%	0.0%
HP0100	items difficult for interviewee	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HP0200	items missed by the interviewee	0.0%	99.9%	0.1%	1.1%	1.2%	0.0%	0.0%	0.1%
HP0300	interviewee additions	0.0%	99.8%	0.2%	0.1%	0.3%	0.0%	0.0%	0.2%
HR0100	Language of the interview	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
HR0200	Dwelling - interior conditions	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%
HR0300	Interviewee - suspicious before the interview	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HR0400	Interviewee - suspicious after the interview	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR0500	Interviewee - understanding of the questions	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HR0600	Interviewee - reliability of income and wealth information	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%
HR0700	Interviewee - ability to express amounts in EUR	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%
HR0800	Interviewee - easiness in responding	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HR0900	Interviewee - ability to express himself/herself	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HR1000	Interviewee - interest in the interview	0.0%	100.0%	0.0%	0.8%	0.8%	0.0%	0.0%	0.0%
HR1100A	Other persons present during the interview	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1100B	Other persons present during the interview	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1100C	Other persons present during the interview	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1100D	Other persons present during the interview	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1100E	Other persons present during the interview	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1100F	Other persons present during the interview	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%
HR1200	Persons providing information Interviewee - frequency of consulting documentation	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1400A	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1400B	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%
HR1400C	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	bracket values in % of (I)	bracket values in % of (II)	edited
HR1400D	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400E	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400F	Documents interviewee referred to	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400G	Documents interviewee referred to	0.0%	100.0%	0.0%	0.4%	0.4%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400H	Documents interviewee referred to	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400I	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400J	Documents interviewee referred to	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400K	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400L	Documents interviewee referred to	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400M	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400N	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400O	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400P	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400Q	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400R	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400S	Documents interviewee referred to	0.0%	100.0%	0.0%	0.3%	0.3%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400T	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400U	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400V	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1400W	Documents interviewee referred to	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1500	Interviewer's comments - missing/misreported items	0.0%	100.0%	0.0%	87.1%	87.1%	0.0%	0.0%	0.0%	0.0%	0.0%
HR1600	Interviewer's comments - conduct of the interview	0.0%	100.0%	0.0%	87.5%	87.5%	0.0%	0.0%	0.0%	0.0%	0.0%
PA0100	marital status	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%
PA0200	highest level of education completed	0.0%	100.0%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100A	labour status	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.3%	0.0%
PE0100B	labour status	94.8%	5.2%	0.0%	2.9%	2.9%	0.0%	0.0%	0.0%	1.0%	0.0%
PE0100C	labour status	99.5%	0.5%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100D	labour status	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100E	labour status	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100F	labour status	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100G	labour status	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100H	labour status	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0100I	labour status	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PE0200	status in employment	44.9%	55.0%	0.1%	0.1%	0.3%	0.0%	0.0%	0.0%	0.4%	0.0%
PE0300	job description / ISCO	44.9%	55.0%	0.1%	0.5%	0.7%	0.0%	0.0%	0.0%	0.3%	0.0%
PE0400	main employment - NACE	52.8%	47.0%	0.1%	1.0%	1.4%	0.0%	0.0%	0.0%	0.4%	0.0%
PE0500	type of contract	52.8%	47.0%	0.2%	0.2%	0.6%	0.0%	0.0%	0.0%	0.4%	0.0%
PE0600	hours working a week - main job	44.9%	55.0%	0.1%	0.6%	0.8%	0.0%	0.0%	0.0%	0.4%	0.0%
PE0700	time in main job	44.9%	54.9%	0.1%	1.0%	1.3%	0.0%	0.0%	0.0%	0.4%	0.0%
PE0800	currently more than one job/employers	44.9%	54.9%	0.1%	0.1%	0.4%	0.0%	0.0%	0.0%	0.3%	0.0%
PE0810	type of other work	97.8%	2.0%	0.2%	0.0%	9.8%	0.0%	0.0%	0.0%	4.9%	0.0%
PE0900	ever been employed	55.0%	44.9%	0.0%	0.2%	0.3%	0.0%	0.0%	0.0%	0.0%	0.0%
PE1000	total time in employment	13.9%	86.0%	0.1%	1.1%	1.3%	0.0%	0.0%	0.0%	3.9%	0.0%
PE1100	at what age expect to retire	20.3%	75.4%	4.3%	21.1%	26.8%	0.0%	0.0%	0.0%	10.6%	0.0%
PE9020	respondent of the employment section	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PF0100	has public pension plans	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0110	number of public pension schemes	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0200	main public plan - % of current gross earnings contributed	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0300	main public plan - years contributing to pension plan	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0400	additional public plans: % of gross earnings	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0500	has public/social security plan with account balance	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0510	current value of all social security plans that have an account	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	100.0%	0.0%
PF0600	has occupational pension plan	0.0%	100.0%	0.0%	1.4%	1.4%	0.0%	0.0%	0.0%	0.0%	0.0%
PF0610	number of occupational pension plans	88.4%	10.2%	1.4%	0.5%	14.4%	0.0%	0.0%	0.0%	0.5%	0.0%
PF0700	occupational plan has an account balance	88.4%	10.3%	1.4%	12.0%	25.4%	0.0%	0.0%	0.0%	0.0%	0.0%
PF0710	current value of all occupational pension plans that have an account	94.1%	3.3%	2.6%	52.9%	130.9%	13.9%	5.6%	1.5%	1.5%	0.0%
PF0800	occupational plan has regular benefit in retirement	88.4%	10.3%	1.4%	6.7%	20.1%	0.0%	0.0%	0.0%	0.0%	0.0%
PF0900	has voluntary pension scheme	0.0%	100.0%	0.0%	0.6%	0.6%	0.0%	0.0%	0.0%	0.2%	0.0%
PF0910A	type of voluntary pension scheme	74.8%	24.6%	0.6%	0.6%	3.0%	0.0%	0.0%	0.0%	1.0%	0.0%
PF0910B	type of voluntary pension scheme	74.8%	24.6%	0.6%	0.4%	2.8%	0.0%	0.0%	0.0%	1.0%	0.0%
PF0920	voluntary pension schemes - value of accounts	74.8%	24.5%	0.7%	50.4%	53.2%	28.2%	26.7%	2.0%	2.0%	0.0%
PF0930	monthly contributions to voluntary pension schemes	74.8%	24.6%	0.6%	8.6%	11.0%	30.2%	23.6%	2.2%	2.2%	0.0%
PF9020	respondent of the pensions section	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
PG0100	received employee income	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%	0.6%	0.0%
PG0110	gross cash employee income	47.0%	52.8%	0.2%	28.3%	28.7%	71.1%	70.1%	1.3%	1.3%	0.0%
PG0200	received self-employment income	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.2%	0.0%
PG0210	gross self employment income	91.5%	8.3%	0.2%	35.5%	38.5%	66.7%	61.5%	8.0%	8.0%	0.0%
PG0300	received income from public pensions	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%	0.2%	0.0%
PG0310	gross income from public pensions	78.6%	21.2%	0.1%	20.4%	21.1%	67.0%	64.8%	1.4%	1.4%	0.0%
PG0400	received income from private and occupational pension plans	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%
PG0410	gross income from occupational and private pension plans	98.6%	1.2%	0.2%	29.2%	45.8%	14.3%	9.1%	0.0%	0.0%	0.0%
PG0500	received income from unemployment benefits	0.0%	100.0%	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%
PG0510	gross income from unemployment benefits	98.3%	1.6%	0.1%	3.0%	9.1%	0.0%	33.3%	0.0%	0.0%	0.0%
PG9020	respondent of the personal income section	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%

variable name	description	unappli- cable	appli- cable	undeter- mined	min. # to be imp. (I)	max. # to be imp. (II)	bracket values in % of (I)	bracket values in % of (II)	edited
PNA0500A	RP's mother alive	22.4%	77.6%	0.0%	0.3%	0.4%	0.0%	0.0%	0.1%
PNA0500B	partner's mother alive	22.4%	77.6%	0.0%	0.3%	0.4%	0.0%	0.0%	0.1%
PNE0500	RP's/partner's father alive	52.8%	47.0%	0.2%	0.6%	1.0%	0.0%	0.0%	0.5%
PNE0500	Private-public organization	52.8%	47.0%	0.2%	0.6%	1.0%	0.0%	0.0%	0.5%
PNE0600	Number of employees - main employer	52.8%	46.9%	0.2%	3.5%	4.0%	0.0%	0.0%	0.4%
PNE2200	Total time in full-time employment	13.9%	86.0%	0.1%	1.2%	1.4%	0.0%	0.0%	0.4%
RA0100	relationship to reference person	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.4%
RA0200	gender	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
RA0300	age	0.0%	100.0%	0.0%	0.2%	0.2%	0.0%	0.0%	0.2%
RA0400	country of birth	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%
RA0500	how long have you been living in the country	61.1%	38.9%	0.0%	0.5%	0.5%	0.0%	0.0%	0.0%
RNA0200	citizenship	0.0%	100.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%

Source: Own calculations based on LU-HFCS, non-imputed and unweighted.

The column 'applicable' cases reflects the number of individuals or households who should have had to reply to these questions. The column 'unapplicable' shows the fraction of households who skipped the question due to routing. The 'undetermined' cases reflect that it is uncertain whether or not a question should have been answered due to a missing value in a mother variable. The next two columns report the 'minimum' and 'maximum' fraction of values to be imputed relative to the sample size of applicable cases. The difference between the two is caused by possible missing values in the mother variable. The minimum fraction of imputed values relative to applicable cases is assuming that all undetermined cases are unapplicable; the maximum number assumes that all undetermined cases are applicable. The next two columns show the number of missing values for which bracketed information is available relative to the minimum and maximum number of values to be imputed. The last column shows the fraction of edited values relative to all applicable values (including both replaced and imputed values).

Appendix B: Distribution of income and wealth

Table 21: Distribution of household income and wealth

Total net wealth, including private pension wealth				
Statistic	Point estimate	Std. err.	[95% conf. interval]	
p10	5674	2177	1397	9951
p25	62553	13249	36534	88572
median	403456	16776	370575	436338
p75	738134	35719	668076	808191
p90	1383543	75036	1235743	1531344
mean	732728	61180	612817	852639

Net real wealth				
Statistic	Point estimate	Std. err.	[95% conf. interval]	
p10	2400	1050	340	4460
p25	30821	8428	14280	47362
median	357633	20436	317566	397701
p75	644731	28302	589255	700208
p90	1167760	83736	1003566	1331955
mean	648975	59191	532962	764988

Net financial wealth				
Statistic	Point estimate	Std. err.	[95% conf. interval]	
p10	-2183	1326	-4782	416
p25	3397	889	1652	5142
median	23394	2189	19089	27698
p75	75826	6592	62487	89165
p90	193330	19650.2	154679	231981
mean	83753	7474	69100	98405

Yearly gross income				
Statistic	Point estimate	Std. err.	[95% conf. interval]	
p10	22900	1030	20879	24921
p25	37493	1756	33991	40995
median	64653	1806	61112	68194
p75	102884	3026	96939	108829
p90	155626	5004	145748	165504
mean	83641	2358	79020	88262

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights.

Table 22: Mean income and wealth by households' characteristics

	Gross income	Total net wealth	Net real wealth	Net financial wealth	Subgroup size	In % of total population
<i>Household size</i>						
hh size=1	49439	411239	345133	66106	55932	30%
hh size=2	88995	877723	769091	108632	52203	28%
hh size=3	97845	802976	712435	90540	31695	17%
hh size=4	110729	767574	694659	72914	29830	16%
hh size=5	106006	1158619	1086998	71620	16780	9%
<i>Age class</i>						
age 16-29	60844	162588	139775	22813	16943	9%
age 30-39	80916	385151	331209	53942	34098	18%
age 40-49	91019	733500	669572	63928	50469	27%
age 50-59	99060	849746	740982	108764	33573	18%
age 60-69	96807	1360412	1231086	129326	23355	13%
age 70-100	57987	835756	711094	124662	28001	15%
<i>Nationality</i>						
Luxembourg	86499	939357	841982	97375	115726	62%
foreign	78963	394570	333111	61460	70714	38%
<i>Education (ISCED)</i>						
no formal or primary edu.	48296	396903	353685	43218	45106	24%
lower secondary edu.	75257	519143	486211	32932	21236	11%
upper or post secondary edu.	75385	775099	695720	79379	71430	38%
tertiary edu.	131982	1074628	924838	149790	48669	26%
<i>Employment status</i>						
employed	88619	560954	493242	67713	104623	56%
self-employed	155904	1665658	1538658	127000	10793	6%
retired	75128	1077320	937528	139791	45375	24%
<i>Income quartile</i>						
income quartile 1		297887	276345	21542	45998	25%
income quartile 2		412802	378809	33993	46485	25%
income quartile 3		573116	506792	66324	46194	25%
income quartile 4		1643395	1431003	212392	47763	26%

Source: Own calculations based on LU-HFCS; data are multiply imputed and weighted; variance estimation based on replicate weights.

Appendix C: Household gross income calculation based on national accounts

Table 23: Household gross income calculation based on national accounts

National account component	in Mio. euro
B6G: Gross disposable income	14,911
+ D4: Property income	384
+ D5: Current taxes on income, wealth, etc.	2,351
+ D6: Social contributions (only employee, self-employed and unemployed)	1,712
+ D62: Social benefits other than social transfers in kind	0
+ D71 - D72: remove all items related to non-life insurance operations	48
+ D75: Miscellaneous current transfers	229
= Total	19,635

Source: Own calculations based on data from Eurostat and Statec. Figures are for Luxembourg in 2009. The sector includes both private households and non-profit institutions serving households. D6 is an estimate based on the assumption that 38% of the work force in 2009 are cross-border workers.



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