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The Effect of Growth and Inequality in Incomes on Health Inequality: Theory and Empirical Evidence from the European Panel

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The effect of growth and inequality in incomes on health inequality: Theory and empirical evidence from the European Panel

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Abstract

Europe aims at combining income growth with improvements in social cohesion as measured by income and health inequalities. We show that, theoretically, both aims can be reconciled only under very specific conditions concerning the type of growth and the income responsiveness of health. We investigate whether these conditions held in Europe in the nineties using panel data from the *European Community Household Panel* surveys. We use pooled interval regressions and inequality decompositions to demonstrate that (i) in all countries except Austria, the income elasticity of health is positive and increases with income, and (ii) that income growth was not pro-rich in most EU countries, resulting in little or no reductions in income inequality and modest increases in income-related health inequality in the majority of countries.

JEL Classification: D30; D31; I10; I12

Keywords: income inequality, health inequality, income elasticity of health, Europe

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1. Introduction

Among the fundamental objectives of the European Union are economic and social progress along with improvements in living and working conditions. The EU leaders agreed in Lisbon in 2000 — when setting strategic goals for the current decade — that the Union should strive for economic growth to become “the most competitive and dynamic knowledge-based economy ... with more and better jobs and greater social cohesion” (Atkinson et al., 2002). The Lisbon European Council not only aimed at stimulating economic growth but also at making a decisive impact on the eradication of income poverty and social exclusion and the monitoring of progress towards these goals.² One of the monitoring tools it created was the collection of new sets of comparable longitudinal household level data across all member states, like the *European Community Household Panel* (ECHP) survey, and its successor, the *EU Survey on Income and Living Conditions* (EU-SILC). Social exclusion in the EU is broadly defined. It does not only refer to (lack of) income and employment but includes wider social dimensions like housing, education and health. A set of indicators has been developed for monitoring the degree of inequality in income as well as social indicators like health status.³

An important question, therefore, is to what extent – and under what conditions – the twin goals of income growth and reduction of social inequalities in health are compatible.⁴ This paper focuses on the consequences of income growth, but – as we show further in the paper – these cannot be analysed independently from the effects of evolving income inequalities on social inequalities in health. A second – no less important – question is which countries have managed to achieve these goals, and to what degree. Building on work of Contoyannis and Forster (1999a) and Wagstaff et al. (2003) we develop a decomposition technique that points to the crucial role of the income elasticity of health. If this elasticity is increasing with income, then proportional income growth may – under some conditions – lead to *higher* income-related health inequalities (hereafter denoted as IRHI). If this were the case, then

² There are 8 Lisbon strategic goals: 6 on economic performance and 2 on increasing social inclusion.

³ Among the so-called Level 1 (Laeken) indicators – which consist of a restricted number of lead indicators covering the broad fields of social exclusion – are the ratio of equivalised income of the top and bottom quintile for income inequality (recommendation 15), and the same ratio for the proportion of the population classifying themselves in poor or very poor health (recommendation 23) in Atkinson et al. (2002).

⁴ Note that this paper does not fit into the literature on the role of income growth on poverty reduction (e.g. Kraay, 2006) and reduction of income inequality (e.g. Barro, 2000) since we study the joint income-health distribution.

Europe faces a trade-off between these two goals. If, on the other hand, growth goes hand in hand with a reduction in health inequality by income, then greater social inclusion derives as a windfall profit. It turns out that the degree to which income growth occurs disproportionately at higher or lower incomes, and the degree to which health responds to income changes at varying income levels are both crucial elements for the relationship between income growth and inequality and the degree of income-related health inequality.

The paper also analyzes the empirically observed trends in income (inequality) and health (inequality) in European countries. We do this by estimating regression models of health and by using our decomposition technique to relate trends in income growth and income inequality to changes in income-related inequalities in health on the full 8 waves of the ECHP. Although we use longitudinal data, our approach is different from Hernández-Quevedo et al. (2006) who analysed the difference between long-term and short-term IRHI using an approach proposed by Jones and Lopez (2004). There is also an epidemiological literature on trends in socioeconomic inequalities in self-assessed health (e.g. Dalstra et al, 2005; Kunst et al, 2005) but in contrast to these earlier empirical papers, we attempt to first theoretically identify the role of changes in the level and distribution of incomes on IRHI before empirically testing these relationships for a large set of European countries.

The paper is organised as follows. Section 2 explains our decomposition technique to analyse the consequences of income growth and income inequality upon social inequalities in health. Section 3 describes the ECHP data set and the empirical models used to implement our decomposition technique of section 2. Empirical results on income elasticities of health and on empirical trends in income, health, inequality and our decomposition technique are presented in section 4, while section 5 provides a conclusion and discussion.

2. Decomposition technique for the relation between the distributions of income and health

In this section, we present a decomposition technique to examine the impact of changes in the income distribution on the distribution of health. We focus on two aspects of the distribution of health, i.e. the evolution of average health and of IRHI. Despite the analysis being far more complicated for income-related than for pure health inequalities, we did not consider the latter since policy makers are generally more concerned with the socioeconomic gradient in health. We extend the approach proposed by Wagstaff et al. (2003) by allowing for a non-linear relationship between income and health.

2.1. Decomposing IRHI in a linear framework

Wagstaff et al. (2003) have proposed a method for decomposing IRHI when health is a linear function of a set of determinants. Each individual i is characterised by her health level $h_i \geq 0$, her income level $y_i \geq 0$ and a vector of other characteristics x_i which could include demographics, etc. The linear relationship allows us – without loss of generality – to assume that x_i includes only one variable.

$$(2.1) \quad h_i = \alpha + \beta y_i + \gamma x_i$$

where α, β, γ are parameters.⁵

We are interested in illuminating the role of income growth on the evolution of both mean health and IRHI. First, the relation between mean health and mean income is straightforward:

$$(2.2) \quad H = \alpha + \beta Y + \gamma X$$

where capital letters denote averages. β measures the impact of average income on average

⁵ Note that the methodology in this section can cope with an error term by incorporating it in x_i , but for ease of exposition it is neglected here.

health. Consequently, other aspects of the distribution of income (e.g. income inequalities) do not matter.

Like Wagstaff et al. (1991), we will measure IRHI using the concentration index – which is widely used to measure relative IRHI (see e.g. Van Doorslaer et al., 1997; Van Doorslaer and Koolman, 2004; Bleichrodt and van Doorslaer, 2006). The concentration index $C(h_i|y_i)$ measures the degree of relative inequality and can be written as:

$$(2.3) \quad C(h_i|y_i) = \frac{2 \sum_{i=1}^n h_i R_i}{\sum_{i=1}^n h_i} - 1$$

where $R_i = N^{-1}(i - 0.5)$ denotes the fractional rank of income. Wagstaff et al. (2003) have shown that a factor decomposition in the spirit of Shorrocks (1982) can be obtained for the concentration index.⁶ Combining equation (2.3) and (2.1) gives:

$$(2.4) \quad C(h_i|y_i) = \beta \frac{Y}{H} G(y_i) + \gamma \frac{X}{H} C(x_i|y_i)$$

where $\beta Y(H)^{-1}$ and $\gamma X(H)^{-1}$ can be interpreted as ‘average elasticities’, $G(y_i) = C(y_i|y_i)$ is the Gini index, and $C(x_i|y_i)$ is the concentration index of x_i . Equation (2.4) shows that IRHI are a linear function of the income-related inequalities of its determinants weighted by their respective ‘average elasticity’. The advantage of their approach for our purposes is that it clearly demonstrates that IRHI is related to average income through the income elasticity but it highlights that other aspects of the distribution of income also matter, in particular the Gini index, the effect of average income on the elasticity of x_i (through the effect on H) and the income rank in $C(x_i|y_i)$.⁷

⁶ See also Clarke et al. (2003) and Gravelle (2003).

2.2. Decomposing changes in IRHI in a non-linear equation

While the decomposition presented in the previous section has obvious intuitive appeal, it abstracts from the well documented non-linear relationship between income and health, i.e. that health shows diminishing returns to income (e.g. Smith, 1999; Deaton, 2003; Ecob and Smith, 1999; Gerdtham and Johannesson, 2000; Gravelle and Sutton, 2003; Mackenbach *et al.*, 2005). A common approach has been to log-transform y_i (e.g. Van Doorslaer and Koolman, 2004) or to use a power function of y_i (e.g. Gravelle and Sutton, 2003) in equation (2.1) to preserve linearity in the transformed variable and its decomposition in equation (2.4). This procedure is incapable of informing on the effect of changes in the Gini and in the ‘average income elasticity’ as it only informs on the contribution of the elasticity and the Gini of the transformed variable (e.g. $\ln(y_i)$). We propose to resolve this problem by recomputing the first term of equation (2.4) in a non-linear setting. We show that it is still possible to make (albeit weaker) inferences on the relative importance of the income elasticity versus income inequality.

While our basic interest lies in equations (2.2) and (2.4), we are not interested in average health and IRHI *per se*, but rather in their evolution over time. Therefore, we introduce a decomposition of (discrete) time-differences in section 2.3 to disentangle the effects of proportional income growth and income inequality. Our approach bears some resemblance to equation (8) in Wagstaff et al. (2003), which (i) considers a total differential⁸ and (ii) is formulated in a continuous framework, but consequently has the obvious disadvantage that it is only an approximation – valid for very small changes – whereas our approach is exact. This is important since we intend to analyse large changes in average income. In addition, it is easier to deal with non-linearities in our approach (see below). We further assume that equation (2.1) holds in each time period.

First, we allow for a non-linear income effect in equation (2.1) by adding a time subscript:

⁷ Wagstaff et al. (2003) also propose two methods (an Oaxaca-type decomposition and a differential equation based decomposition) to decompose changes in IRHI. We come back to the latter in section 2.2.

⁸ Wagstaff et al. (2003) also allow for changes in α, β, γ . Although we keep β fixed, the nonlinear relationship of $f(\cdot)$ allows for different income-effects at different income levels, and consequently at different time periods. Fixing of α and γ is less important since we focus on the impact of the distribution of income on the

$$(2.5) \quad h_{it} = \alpha + f(y_{it}; \beta) + \gamma x_{it}$$

We allow the function $f(\cdot)$ and its slope to vary with income and will add additional restrictions below in order to make some analytic predictions. Taking averages, we obtain:

$$(2.6) \quad H_t = \alpha + \frac{\sum_{i=1}^n f(y_{it}; \beta)}{n} + \gamma X_t$$

In contrast to equation (2.2), not only average income, but also the non-linearity of the income profile now matters due to the aggregation of a non-linear income profile.⁹ For example, in the special case of an increasing and concave second order polynomial, one can show that average income increases average health, whereas the variance of income (or income inequality) decreases average health. Doing a similar exercise for IRHI, by introducing a non-linear income profile, changes equation (2.4) in:

$$(2.7) \quad C(h_{it}|y_{it}) = \frac{\sum_{i=1}^n f(y_{it}; \beta)}{nH_t} C[f(y_{it}; \beta)|y_{it}] + \gamma \frac{X_t}{H_t} C(x_{it}|y_{it})$$

Note that the introduction of $f(\cdot)$ removes the exact relationship between $C(h_{it}|y_{it})$ and the income elasticity and the Gini index. Only if $f(y_{it}; \beta) = \beta y_{it}$, equation (2.7) reduces to equation (2.4).

Second, introducing discrete time differences to equations (2.6) and (2.7) gives:

distribution of health. Our approach could be generalised to allow for changes in α, β, γ , but consult section 3.1 for our reasons not to do so.

⁹ Our assumption does not rely on the literature investigating a direct negative effect of income inequality on individual health. Literature surveys (Wagstaff and van Doorslaer, 2000b; Deaton, 2003) did not find convincing evidence for a direct effect of income inequality on individual health. Moreover, Hildebrand and Van Kerm (2005) – who studied the direct effect of income inequality using the ECHP – found a statistically significant, but very small negative effect. This would imply that our findings for the effect of income inequality can – in the worst case – be interpreted as an upper bound.

$$(2.8) \quad H_t - H_1 = \frac{\sum_{i=1}^n f(y_{it}; \beta) - \sum_{i=1}^n f(y_{i1}; \beta)}{n} + \gamma(X_t - X_1)$$

$$(2.9) \quad C(h_{it}|y_{it}) - C(h_{i1}|y_{i1}) = \left\{ \begin{array}{l} \frac{\sum_{i=1}^n f(y_{it}; \beta)}{nH_t} C[f(y_{it}; \beta)|y_{it}] \\ - \frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_1} C[f(y_{i1}; \beta)|y_{i1}] \end{array} \right\} + \left\{ \begin{array}{l} \gamma \frac{X_t}{H_t} C(x_{it}|y_{it}) \\ - \gamma \frac{X_1}{H_1} C(x_{i1}|y_{i1}) \end{array} \right\}$$

Equation (2.8) is relatively straightforward, but the effects of proportional income growth (i.e. a change in average income) and the change in income inequality on IRHI are not so easily inferred from equation (2.9). While it is still straightforward to decompose the effect of a change of x_{i1} to x_{it} on the change in IRHI into an average elasticity and an inequality effect using the methods in Wagstaff et al. (2003)), this is no longer the case for a change of y_{i1} to y_{it} .

2.3. Disentangling proportional income growth from the evolution of income inequality

Our approach consists of the introduction of two hypothetical health levels which allow us to put more analytic structure on the decompositions in equation (2.8) and (2.9). These hypothetical health levels are h_{it}^{pg} (pg for proportional growth) and h_{it}^{ng} (ng for no growth), i.e.

$$(2.10) \quad h_{it}^{pg} = \alpha + f(y_{it}^{pg}; \beta) + \gamma x_{it}$$

$$(2.11) \quad h_{it}^{ng} = \alpha + f(y_{i1}; \beta) + \gamma x_{it}$$

where $y_{it}^{pg} = y_{i1} Y_t (Y_1)^{-1}$. Equation (2.10) presents the hypothetical health level that individual i would have had in period t if her income growth had been equal to the *actual* average growth, but *without any* changes in income inequality between period t and 1. A similar intuition lies behind the introduction of h_{it}^{ng} , but in contrast to equation (2.10), the income distribution remains unchanged. Note that we could also have introduced h_{i1}^{pg} and h_{i1}^{ng} , but we

prefer t as reference period since it seems more natural to evaluate the effects of income changes with the prevailing value of x_{it} .¹⁰ Combining equations (2.8), (2.9), (2.10) and (2.11), we now obtain:

$$(2.12) \quad H_t - H_1 = \frac{\sum_{i=1}^n f(y_{it}; \beta) - \sum_{i=1}^n f(y_{it}^{pg}; \beta)}{n} + \frac{\sum_{i=1}^n f(y_{it}^{pg}; \beta) - \sum_{i=1}^n f(y_{i1}; \beta)}{n} + \gamma(X_t - X_1)$$

$$(2.13) \quad \begin{aligned} & C(h_{it}|y_{it}) - C(h_{i1}|y_{i1}) \\ &= \underbrace{\frac{\sum_{i=1}^n f(y_{it}; \beta)}{nH_t} C[f(y_{it}; \beta)|y_{it}] - \frac{\sum_{i=1}^n f(y_{it}^{pg}; \beta)}{nH_t^{pg}} C[f(y_{it}^{pg}; \beta)|y_{i1}]}_{\text{term 1a}} \\ &+ \underbrace{\frac{\sum_{i=1}^n f(y_{it}^{pg}; \beta)}{nH_t^{pg}} C[f(y_{it}^{pg}; \beta)|y_{i1}] - \frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_t^{ng}} C[f(y_{i1}; \beta)|y_{i1}]}_{\text{term 1b}} \\ &+ \underbrace{C[f(y_{i1}; \beta)|y_{i1}] \left[\frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_t^{ng}} - \frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_1} \right]}_{\text{term 1c}} \\ &+ \underbrace{\gamma \frac{X_t}{H_t} C(x_{it}|y_{it}) - \gamma \frac{X_t}{H_t^{pg}} C(x_{it}|y_{i1})}_{\text{term 2a}} + \underbrace{C(x_{it}|y_{i1}) \left[\gamma \frac{X_t}{H_t^{pg}} - \gamma \frac{X_t}{H_t^{ng}} \right]}_{\text{term 2b}} \\ &+ \underbrace{\gamma \frac{X_t}{H_t^{ng}} C(x_{it}|y_{i1}) - \gamma \frac{X_1}{H_1} C(x_{i1}|y_{i1})}_{\text{term 2c}} \end{aligned}$$

where $C(\dots|y_{it}^{pg}) \equiv C(\dots|y_{i1})$. Equation (2.12) clearly shows that the effects of proportional income growth and changes in income inequality are easily separated, and are unambiguous: average health responds elastically/inelastically/unit elastically if equation (2.5) is convex/concave/linear and increasing with income.¹¹ With respect to equation (2.13) things are less straightforward. Terms 1a-c disentangle the first term of equation (2.9), while terms

¹⁰ Another alternative would be to decompose $C(h_{it}|y_{it}) - C(h_{i,t-1}|y_{i,t-1})$, but it can be shown that the qualitative interpretation of the decomposition is similar. We stick to the comparison with period 1 as it allows for more variation as we decompose over a longer time period.

¹¹ This result was already shown by Contoyannis and Forster (1999a, 1999b).

2a-c disentangle the second term of equation (2.9). We will show that the a-terms are related to the evolution of income inequality, the b-terms to proportional income growth and the c-terms to the evolution of the other determinants of health. In the next subsections we discuss each term in detail.

2.3.1. Proportional income growth: term 1b and 2b

The influence of proportional income growth on IRHI is summarised by terms 1b and 2b in equation (2.13), i.e.

$$(2.14) \quad \frac{\sum_{i=1}^n f(y_{it}^{pg}; \beta)}{nH_t^{pg}} C[f(y_{it}^{pg}; \beta)|y_{i1}] - \frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_t^{ng}} C[f(y_{i1}; \beta)|y_{i1}]$$

$$(2.15) \quad C(x_{it}|y_{i1}) \left[\gamma \frac{X_t}{H_t^{pg}} - \gamma \frac{X_t}{H_t^{ng}} \right]$$

First, note that (2.14) and (2.15) are zero when there is no income growth. Moreover, by definition they are not influenced by changes in income inequality, nor by changes in x_{it} .

Predictions on the sign of (2.14) can be obtained using a result of Contoyannis and Forster (1999a), who considered the special case where income is the sole determinant of health – hence (2.14). They show (in their proposition 6) that proportional income growth increases/decreases IRHI if the income elasticity is monotonically rising/falling with income. The intuition for this result can be grasped by considering the linear version of (2.14), i.e.

$$(2.16) \quad \frac{\sum_{i=1}^n \beta y_{it}^{pg}}{nH_t^{pg}} G(\beta y_{it}^{pg} | y_{i1}) - \frac{\sum_{i=1}^n \beta y_{i1}}{nH_t^{ng}} G(\beta y_{i1} | y_{i1})$$

which, after some algebra, reduces to

$$(2.17) \quad \frac{\beta(Y_t - Y_1)(\alpha + \gamma X_t)}{Y_1 n H_t^{pg} H_t^{ng}} \sum_{i=1}^n [(2R_{i1} - 1)y_{i1}]$$

The term before the sum in (2.17) is a constant for given values of Y_t , Y_1 , X_t and positive if $Y_t > Y_1$.¹² The first term in the sum $(2R_{it} - 1)$ is negative for incomes below the median and positive for incomes above the median. Therefore, (2.17) is positive if $Y_t \geq Y_1$, which establishes the result of Contoyannis and Forster (1999a) for a linear income profile since its income elasticity¹³ always increases with income.

In case of a non-linear income effect, we can generalize the proposition of Contoyannis and Forster (1999a), although the derivation is less intuitive. After some algebra, one can show that (2.14) reduces to:

$$(2.18) \quad \sum_{i=1}^n \left\{ (2R_{it} - 1) \frac{1}{n} \left[\frac{f(y_{it}^{pg}; \beta)}{H_t^{pg}} - \frac{f(y_{i1}; \beta)}{H_t^{ng}} \right] \right\}$$

Taking the partial derivative of the term between square brackets gives¹⁴:

$$(2.19) \quad \frac{\partial \left[\frac{f(y_{it}^{pg}; \beta)}{H_t^{pg}} - \frac{f(y_{i1}; \beta)}{H_t^{ng}} \right]}{\partial y_{i1}} = \frac{\partial f(y_{it}^{pg}; \beta)}{\partial y_{it}^{pg}} \frac{Y_t}{H_t^{pg}} - \frac{\partial f(y_{i1}; \beta)}{\partial y_{i1}} \frac{1}{H_t^{ng}}$$

Multiplying (2.19) by Y_1 again highlights the crucial role of the income elasticity: the expression in (2.18) is positive/negative if the income elasticity (evaluated at the average values Y_t and Y_1) increases/decreases with income.¹⁵ Because the partial derivatives of

¹² It is very plausible to assume that income has a positive effect on health ($\beta > 0$), that average health at income level zero is positive ($\alpha + \gamma X_t > 0$), and that average health and income are positive $Y_1, H_t^{pg}, H_t^{ng} > 0$.

¹³ Elasticity increases with income since for $h = \alpha + \beta y + \gamma x$, we have that $(\partial \varepsilon^y / \partial y) = h^{-2} \beta (h - \beta y) > 0$.

¹⁴ Note that we treat the partial derivatives of H_t^{pg} and H_t^{ng} to y_{i1} as zero, which is justified since we only intend to investigate how the term changes if we move up in the income distribution, without changing the income distribution (in other words, keeping H_t^{pg} and H_t^{ng} fixed).

¹⁵ Previous research (e.g. Smith, 1999; Deaton, 2003; Gravelle and Sutton, 2003; Mackenbach et al., 2005) generally suggests that the marginal effect of income on health is positive $(\partial f(y; \beta) / \partial y) > 0$ and declining with income $(\partial^2 f(y; \beta) / \partial y^2) < 0$ and this is confirmed in (most of) our empirical exercise. Note that concavity does *not* imply that the income elasticity of health ε^y reduces with income. One can see that it is increasing with income if $(\partial \varepsilon^y / \partial y) = (\partial^2 f(y; \beta) / \partial y^2) (y/h) + (\partial f(y; \beta) / \partial y) (h)^{-1} [1 - (\partial f(y; \beta) / \partial y) (y/h)] > 0$, or

$f(y_{it}^{pg}; \beta)$ and $f(y_{it}; \beta)$ are at the individual level, we can establish a generalization of the Contoyannis and Forster (1999a) proof: the income elasticity does not have to increase/decrease *monotonically*. All that is required for (2.18) to be positive/negative is that the sum of $(2R_{it} - 1) \frac{1}{n} \left[\frac{f(y_{it}^{pg}; \beta)}{H_t^{pg}} - \frac{f(y_{it}; \beta)}{H_t^{ng}} \right]$ of incomes above the median is larger/smaller than the identical sum of incomes below the median. Loosely speaking, it is sufficient if the income elasticity (*evaluated at y_{it}*) increases/decreases ‘on average’ and not at each point of the income profile.

The combination of the results in (2.12) and (2.19) is powerful. It implies that proportional income growth leads to a (welfare improving) average health increase and – depending on the slope of the income elasticity – to a (welfare decreasing/increasing) increase/decrease in relative income-related health inequality. Because this result has only limit applicability, as it only refers to proportional income growth, we abstract from proportional income growth in the next section and focus on income inequality changes.

The above discussion illustrates that the income elasticity is a vital element to understand the evolution of IRHI. However, (2.14) assumes that only income affects health, whereas equation (2.13) allows for an additional determinant x_{it} . It follows – as can be inferred from (2.15) – that proportional income growth (through its effect on H) also affects the elasticity of x_{it} . Although one can predict that H_t^{pg} is larger/smaller than H_t^{ng} if the elasticity is rising/decreasing with income, we cannot predict the sign of (2.15), as we do not know the sign of $C(x_{it} | y_{it}) \gamma X_t$ *a priori*.

2.3.2. Changes in income inequality: term 1a and 2a

The effect of changes in income inequality on IRHI is summarised by the terms 1a and 2a in equation (2.13), i.e.

$$(2.20) \quad \frac{\sum_{i=1}^n f(y_{it}; \beta)}{nH_t} C[f(y_{it}; \beta) | y_{it}] - \frac{\sum_{i=1}^n f(y_{it}^{pg}; \beta)}{nH_t^{pg}} C[f(y_{it}^{pg}; \beta) | y_{it}]$$

$(\partial f(y; \beta) / \partial y)(h)^{-1} (1 - \varepsilon^y) > -(\partial^2 f(y; \beta) / \partial y^2)(y/h)$. Moreover, note that the discussed properties imply that

$$(2.21) \quad \gamma \frac{X_t}{H_t} C(x_{it} | y_{it}) - \gamma \frac{X_t}{H_t^{pg}} C(x_{it} | y_{it})$$

We start by noting that both expressions are by definition not influenced by proportional income growth, nor by changes in x_{it} . The only relevant determinant is the evolution of income inequality between period 1 and t . Clearly, without any change in income inequality, both terms are zero.

One can derive from (2.20) – which keeps average income fixed – that a change in income inequality has an immediate direct effect on IRHI and an effect through the income elasticity. Again, we start disentangling (2.20) by first considering when its linear version is positive:

$$(2.22) \quad \frac{G(y_{it})}{G(y_{it}^{pg})} = \frac{G(y_{it})}{G(y_{it})} \geq \frac{\beta \frac{Y_t}{H_t}}{\beta \frac{Y_t}{H_t}} = 1$$

Equation (2.22) clearly shows that an increase in income inequality contributes to IRHI. In the linear case, the effect through the income elasticity is non-existent as the income elasticity does not change due to a change in income inequality.

Using similar algebra as that used to derive (2.18), we arrive at the corresponding non-linear expression, i.e.:

$$(2.23) \quad \frac{1}{n} \sum_{i=1}^n \left\{ 2 \left[R_{it} \frac{f(y_{it}; \beta)}{H_t} - R_{i1} \frac{f(y_{it}^{pg}; \beta)}{H_t^{pg}} \right] - \left[\frac{f(y_{it}; \beta)}{H_t} - \frac{f(y_{it}^{pg}; \beta)}{H_t^{pg}} \right] \right\}$$

It can be seen from the R -terms that – in contrast to (2.18) – income inequality also affects the fractional rank and thus has a direct effect on (2.20) *and* that – similar to (2.18) – the income elasticity matters. If the evolution of income inequality is *on average* in favour of/at the expense of the rich and if the income elasticity is increasing/decreasing with income, one can predict that (2.23) is positive/negative.¹⁶ Intuitively, this can be understood from a

$\varepsilon^y < 1$, and thus that the final term between brackets is positive.

¹⁶ We are grateful to Paul Contoyannis for pointing this out.

comparison between (2.18) and (2.23) which shows that in case of pro-rich (pro-poor) evolving income inequality and an elasticity that increases (decreases) with income, both tendencies reinforce one another. These conditions are sufficient, but not necessary: if these tendencies are opposite, then one cannot *a priori* predict the sign of (2.23).¹⁷ Intuitively, the latter means that an increase/decrease in income inequality is offset by local changes in the income elasticity: which of the two effects dominates is then an empirical question.

It is important to add that (2.21) shows that changes in income inequality have two additional effects, i.e. (i) it affects the concentration indices of x_{it} through differences in the fractional rank and (ii) it affects the ‘average elasticity of the x_{it} -determinant’ through H . The sign of (2.21) cannot be predicted *a priori*.

It is worth emphasizing that the results in this section imply that – contrary to a common belief¹⁸ (e.g. Blakely and Wilson, 2006; Avendano, Glymour and Mackenbach, 2006; Dahlgren and Whitehead, 2006) – reductions in income inequality do *not necessarily* lead to lower IRHI – since there are interactions between the evolution of income inequality and the ‘slope of the income elasticity’ as soon as income affects health in a non-linear way.

2.3.3. The importance of other determinants: terms 1c and 2c

Terms c of equation (2.13) measure the effect of the change from x_{i1} to x_{it} . Term 1c measures the effect of this change through H while term 2c summarizes the effect of this change on the ‘average elasticity’ of x_{it} and the effect that runs via changes in the concentration index of x_{it} .

$$(2.24) \quad C[f(y_{i1}; \beta) | y_{i1}] \left[\frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_t^{ng}} - \frac{\sum_{i=1}^n f(y_{i1}; \beta)}{nH_1} \right]$$

$$(2.25) \quad \gamma \frac{X_t}{H_t^{ng}} C(x_{it} | y_{i1}) - \gamma \frac{X_1}{H_1} C(x_{i1} | y_{i1})$$

¹⁷ Contrary to (2.18) where y_{i1} and y_{it}^{pg} are related through Y_t/Y_1 , there is no obvious relationship between both income terms y_{it}^{pg} and y_{it} .

¹⁸ Contoyannis and Forster (1999a) is a notable exception.

Note again that both expressions drop out from equation (2.13) if x_{it} is constant over time. However, we cannot predict the sign of (2.24) and (2.25) since the signs of γ and of $C(x_{it}|y_{it})$ and $C(x_{it}|y_{it})$ are not known *a priori*. This is not a disadvantage since both expressions only enter equation (2.13) to correct the effect of changes in the income distribution on IRHI for the evolution of other health determinants. In other words, these are just control terms.

In summary, the above approach builds on Wagstaff et al. (2003), but explicitly accounts for a non-linear relationship between income and health, and allows for large income changes. Depending on the restrictions imposed on the non-linear income profile, predictions can be derived on the effects of changes in average income and changes in the variability of individual income. For instance, assuming a concave profile, one can predict a positive effect from proportional income growth and a negative effect from rising income inequality. The picture is more complicated for the effect of changes in the income distribution on the evolution of IRHI. In order to disentangle the effect of proportional income growth from the impact of changes in income inequality, we introduced two hypothetical health levels, i.e. (i) the health level that would prevail in case of a non-changing income distribution and (ii) the health level that would prevail in case of proportional income growth. This enabled us to (i) isolate the effect of changes in the income distribution from changes in the other health determinants, and (ii) to isolate the effect of changes in income inequality from proportional income growth. In both instances there is a direct effect of the change in the income distribution on IRHI, but also an indirect effect through the other health determinants. Building on Contoyannis and Forster (1999a) we showed that the *direct* effect of proportional income growth depends on the slope of the income elasticity. If this elasticity is rising/decreasing with income ‘on average’ (see equation (2.18) for the exact condition), IRHI increase/decline. With respect to the *direct* effect of changes in income inequality, we predict increasing/decreasing IRHI in case of pro-rich/pro-poor evolving income inequality in combination with an income elasticity that increases/decreases with income ‘on average’ (see equation (2.23) for the exact condition). We find that reductions in income inequality do not always lead to reductions in IRHI if income inequality and the elasticity do not move together ‘on average’. In the latter case both have an opposite effect and the net effect is an empirical issue that cannot be resolved *a priori*. The sign of the indirect effects (both for proportional

income growth and income inequality) could not be inferred since these depend on the concentration indices and the elasticity of the other health determinants. Obviously, these are not known *a priori*.

Our empirical analysis has three objectives. First, as introduction, we present estimates of the income elasticity of health and how it varies with income. Because the income elasticity and its slope are very important to determine the consequences of income growth, we use a flexible functional form in the estimation. Secondly, we will examine empirical trends of income growth, income inequality, average health and IRHI in Europe in the nineties. Third, we will use our decomposition technique to isolate the effects of proportional income growth and changing income inequality.

3. Data and empirical model specification

The data used in this paper are taken from the full 8 annual waves (1994-2001) of the *European Community Household Panel User Database* (ECHP-UDB). The ECHP was designed and coordinated by EUROSTAT, and it contains socioeconomic, demographic and health variables, for a panel of households which only includes individuals aged 16 or older. It used a standardised questionnaire which allows for longitudinal analysis. We use all waves that are available for 13 EU member states: Austria, Belgium, Denmark, Finland, France, Germany, Greece, Ireland, Italy, Netherlands, Portugal, Spain and the United Kingdom. We did not analyze the data for Luxembourg (small sample) and Sweden (no panel data in ECHP). For Germany and the UK, we did not use the ECHP (which only ran from 1994 to 1997, i.e. waves 1 to 3) but instead used the *German Socio-economic Panel* (GSOEP) and the *British Household Panel Survey* (BHPS). Austria joined the survey in 1995 (wave 2) and Finland in 1996 (wave 3).

3.1. Estimating the elasticity of health with respect to income

The two key variables for this study are health and income. The ECHP income measure is disposable (i.e. after-tax) household income, which is all net monetary income received by the household members during the *previous* year. It includes income from work (employment and self-employment), private income (from investments and property and private transfers to the household), pensions and other direct social transfers received. No account has been taken of indirect social transfers (e.g. reimbursement of medical expenses), receipts in kind and imputed rent from owner-occupied accommodation. The income variable is (i) converted in Euros by yearly PPPs (see EUROSTAT, 2003) to allow for comparability across countries, and (ii) expressed in constant (1996) prices, i.e. deflated by the harmonised index of consumer prices (HICP), to allow for comparability across waves. The HICP is an overall indicator of price developments in the Euro area and was taken from the European Central Bank (2000, 2003).¹⁹ The income variable was further divided by the OECD modified equivalence scale in order to account for household size and composition (giving a weight of 1.0 to the first adult, 0.5 to the second and each subsequent person aged 14 and over, and 0.3 to each child aged

under 4 in the household).

Self-assessed health (SAH) is measured as the response to an ordered 5-point scale (ranging from very good to very poor) on the question “How is your health in general?” In addition to the mere language differences, the question wording was slightly different in 3 of the 13 countries. For France and Germany, a 6 point, respectively 10 point (health satisfaction) scale was recoded into the common 5 point scale by Eurostat. In the UK, the question wording adds a reference to people of the same age (except for wave 6) (Hernández-Quevedo et al, 2004). Reporting heterogeneity in self-assessed health across cultures and populations is a notable concern (Lindeboom and Van Doorslaer (2004) but as our paper is basically about health trends *within* countries this is less of a concern here. We have adopted the scaling methods based on interval regression proposed by Van Doorslaer and Jones (2003) and used by Van Doorslaer and Koolman (2004) on the ECHP data.²⁰ This approach assumes that there is a stable mapping from the Health Utility Index (HUI) (see e.g. Feeny et al. 2002) to the (latent) variable that determines reported *SAH* and that this applies not only to Canadian but also to European individuals. While the internal validity of this approach was confirmed in the Canadian data (Van Doorslaer and Jones, 2003), it is not possible to test the external validity on the European data. However, sensitivity analysis using other boundaries has shown that the results are almost identical when the imposed thresholds were derived from other (European) generic measures like the Euroqol (Lauridsen et al, 2004; Lecluyse and Cleemput, 2005).

While the decomposition technique of section 2 is applicable on any estimate of equation (2.5), we would prefer to interpret our estimate of the income effect in this equation as causal, as we intend to investigate the impact of proportional income growth and changing income inequality on average health and IRHI. One could apply a simultaneous structural estimation technique to estimate a Grossman type model (e.g. Wagstaff, 1993), but we have opted for a one equation approach because of its transparency, and since we are only interested in the overall income effect rather than the underlying pathways. For the estimation of our health equation, we have included as covariates – besides income – only demographics like age and gender:

¹⁹ We do not use national CPI's since yearly PPP's already eliminate differences in the price evolution between countries. All that remains is a correction for 'average price evolution' in the Euro area, i.e. the HICP.

²⁰ Thresholds are 0, 0.428, 0.756, 0.897, 0.947, and 1

$$(3.1) \quad h_{it}^* = \alpha + f(y_{it}; \beta) + x_{it}'\gamma + \varepsilon_{it}$$

where h_{it}^* is the latent health outcome, α , β and γ are parameters to be estimated, $f(y_{it}; \beta)$ is a non-linear function of income (see below), and $\varepsilon_{it} \sim N(0, \sigma_\varepsilon^2)$. We do not observe h_{it}^* , but we do observe SAH and can impose its interval boundaries derived from HUI scores. As a result, its predictions are contained in the $[0,1]$ interval and can be interpreted as (predicted) health utilities on the HUI scale. The vector of covariates (x_{it}) includes age dummies (categories: 16-29; 30-44; 45-59; 60-69; 70+) for both sexes. We limit the specification to only these covariates on the grounds that these can safely be assumed to be exogenous and that we are mainly interested here in an estimate of the overall income elasticity of health (utility), not in the effects of endogenous variables (like life style or labour choices) that may mediate the effect of income on health. For adults, education would also be an obvious exogenous candidate to include *and* the ECHP records information on the highest level of general or higher education completed. Nevertheless, we had to exclude education from our regression model as EUROSTAT (2003) notes classification problems related to this variable.

Despite the exogenous covariates and the fact that our income measure refers to disposable household income of the previous year – which makes it less prone to reverse causation bias compared to contemporary income – , our approach does not necessarily solve the endogeneity problem between income and health. Therefore, we have also estimated a dynamic version of equation (3.1) which includes SAH for the previous and the first wave (see e.g. Hurd and Kapteyn (2003), Contoyannis et al. (2004) and Jones et al. (2006) for similar approaches).²¹ As the latter specification captures state dependence, removes any correlation between income and initial health from the estimate of β , and models income effects on health transitions, it is less likely to reflect reverse causation. A comparison of the latter estimates with the estimates based on equation (3.1) revealed no major differences except for a smaller income effect. The latter finding is obvious given that the dynamic approach effectively models health transitions. We also repeated all other analyses in section 4 (income elasticities, trends of average income, income inequalities, average health, IRHI,

²¹ Lagged and initial health were included as sets of dummies.

and the decomposition approach explained in section 2) and found no important differences.²² Since the dynamic approach did not alter the qualitative interpretations of our analyses, only captures the short run effect of income on health and since the approach based on equation (3.1) does not have to deal with terms based on lagged and initial health – and is therefore easier to explain – we decided to present the latter.

We have run pooled²³ models on a balanced panel of individuals observed for (up to) 8 waves. We did not consider unbalanced panels for three reasons. First, although the decomposition approach explained in section 2 holds for variable population sizes, it cannot isolate the effects of varying population sizes from the effects of changes in the income distribution. Second, and more importantly, Jones et al. (2006) have shown that health-related non-response in the ECHP hardly affects estimates of income effects in health equations, and therefore our restriction to balanced panels seems acceptable. Our restriction to balanced panels does mean that especially the results on time trends only apply to a cohort of individuals. Third, as with the dynamic specification (see footnote 22), we have repeated all analyses in section 4 using an unbalanced panel and found no important differences.²⁴

We did not include any time dummies on the grounds that these might pick up some average income changes and moreover, in 9 countries the set of time dummies was jointly not statistically significant. We have also kept the β 's fixed across time (see also footnote 8). In 7 countries we could not reject the null that interactions between time dummies and the non-linear income profile are statistically irrelevant. Moreover, in those cases where the null was rejected, almost none of the individual interactions were statistically significant. Note also

²² There is one difference, i.e. the elasticity estimates are lower for the dynamic approach, but reveal a similar pattern. Therefore, we consider it as qualitatively unimportant. Results from the dynamic approach are obtainable from the authors.

²³ We also experimented with random effects panel models in which we parameterised the individual effects as a function of the means of time-varying variables (Chamberlain, 1980). Although the estimated β 's were very similar to those of the pooled models, we prefer the pooled specifications as they impose less stringent exogeneity assumptions (see e.g. chapter 15 in Wooldridge, 200).

²⁴ We found two differences in the analyses based on the unbalanced panel, but consider both as unimportant since the basic messages of this paper are unaffected. Estimates of the evolution of average health show less of a decline and estimates of the trend in IRHI are less increasing. Both findings are not surprising given that we study a cohort in the balanced panel, i.e. there is no healthy refreshment sample and no sample drop out from unhealthy/deceased individuals and the correlation between income and health is plausibly higher within a cohort than across cohorts. We find that the control terms that were explained in section 2.3.3 pick up both effects, while the other terms of our decomposition technique are not affected by using the unbalanced panel. Also noteworthy is our finding that the estimates of income growth and income inequalities based on an unbalanced panel are largely confirmed by those resulting from the balanced panel which suggests that attrition is not the driving determinant of these findings.

that the non-linear income profile without interactions still allows for differences in the income effect across time. The cross-country differences in health, income and demographics are documented in Table 3.1 which presents unweighted means of all variables for the pooled sample.

Table 3.1: Summary statistics by country

	Austria	Belgium	Denmark	Finland	France	Germany	Greece	Ireland	Italy	Netherlands	Portugal	Spain	UK
M1629	0,089	0,052	0,059	0,062	0,064	0,069	0,045	0,069	0,096	0,041	0,080	0,077	0,066
M3044	0,141	0,173	0,163	0,147	0,148	0,167	0,126	0,142	0,141	0,167	0,121	0,128	0,146
M4559	0,129	0,120	0,150	0,139	0,132	0,140	0,128	0,143	0,133	0,143	0,118	0,113	0,124
M6069	0,072	0,059	0,062	0,069	0,064	0,068	0,090	0,072	0,070	0,060	0,078	0,078	0,058
M70+	0,049	0,050	0,047	0,031	0,053	0,032	0,068	0,055	0,046	0,048	0,062	0,067	0,053
F1629	0,076	0,061	0,066	0,075	0,077	0,079	0,069	0,071	0,094	0,058	0,073	0,082	0,078
F3044	0,152	0,203	0,171	0,172	0,170	0,176	0,142	0,157	0,149	0,198	0,136	0,143	0,181
F4559	0,138	0,131	0,162	0,182	0,142	0,140	0,137	0,151	0,137	0,146	0,142	0,131	0,152
F6069	0,080	0,077	0,061	0,072	0,079	0,074	0,099	0,075	0,073	0,075	0,097	0,091	0,065
F70+	0,074	0,073	0,058	0,051	0,070	0,055	0,096	0,064	0,061	0,064	0,091	0,091	0,077
income	14631	16298	15718	12948	14516	15030	7836	11876	10811	14264	7293	9611	14962
sahverybad	0,013	0,006	0,010	0,006	0,034	0,032	0,021	0,005	0,016	0,005	0,039	0,017	0,019
sahbad	0,058	0,035	0,040	0,054	0,038	0,138	0,074	0,022	0,092	0,037	0,194	0,104	0,076
sahfair	0,216	0,216	0,170	0,309	0,338	0,348	0,185	0,165	0,299	0,232	0,343	0,245	0,223
sahgood	0,429	0,531	0,328	0,461	0,473	0,406	0,270	0,367	0,437	0,561	0,398	0,485	0,466
sahverygood	0,285	0,212	0,451	0,170	0,118	0,076	0,450	0,441	0,157	0,165	0,026	0,150	0,216
N	27769	24200	20352	19314	54688	60160	49072	22976	72288	36448	56776	58456	47400

In view of the literature on a non-linear, concave relationship between income and individual health, the discussion on the relevance of equation (2.5) in section 2.2 and the importance of non-linearity for the current paper (i.e. rising versus decreasing income elasticity), we allow for a flexible functional form by implementing polynomial transformations of income.²⁵ This allows for the income elasticity of health to decrease with income in some income ranges and increase in others. The order of the polynomial was determined by first, estimating each model with a fifth order polynomial and then reducing the order of the polynomial until a Likelihood Ratio test (1% significance level) rejected the ‘reduced order’ against the higher order polynomial.

We did not use the Eurostat-provided cross-sectional individual sampling weights to estimate equation (3.1), but we did for the estimation of the elasticity of health with respect to income (see e.g. chapter 24 in Cameron and Trivedi (2005)):

$$(3.2) \quad \hat{\varepsilon}^y = \frac{\partial f(y_{it}; \hat{\beta})}{\partial y_{it}} \frac{y_{it}}{\hat{h}_{it}}$$

²⁵ We have not used a power function transformation (e.g. Gravelle and Sutton, 2003) as this would entail a maximum likelihood grid search procedure which – in the light of the 13 countries and the interval regression technique – would be unduly time intensive.

where \hat{h}_{it} denotes an estimate and \hat{h}_{it} is the predicted value of equation (3.1). We computed equation (3.2) for each individual and calculate the (weighted) mean to obtain the average elasticity over *all time periods*. In order to verify whether the elasticity is increasing/decreasing with income, this procedure was repeated for each income decile for all time periods.²⁶ A comparison between the elasticities of the lowest versus highest deciles provides an indication of whether equation (2.19) holds.

3.2. Estimating trends of average health, income-related health inequalities, average income and income inequality

With respect to the distribution of (predicted) health (utility), we calculate \hat{H}_t and $C(\hat{h}_{it}|y_{it})$ for each t . The concentration index of health $C(\hat{h}_{it}|y_{it})$ is computed using a separate OLS-regression for each wave t . Kakwani et al. (1997) have shown that the point estimate of $\hat{\lambda}_t$ in the following equation equals $C(\hat{h}_{it}|y_{it})$.

$$(3.3) \quad 2\hat{\sigma}_{R_t}^2 \frac{\hat{h}_{it}}{\hat{H}_t} \sqrt{w_i} = \phi_t \sqrt{w_i} + \lambda_t \sqrt{w_i} \hat{R}_{it} + \zeta_{it}$$

where \hat{h}_{it} is the predicted value of equation (3.1), and thus the resulting $C(\hat{h}_{it}|y_{it})$ can be interpreted as income-related inequality in predicted health utility. \hat{H}_t is the weighted average of \hat{h}_{it} in wave t , w_i is the sampling weight²⁷ of individual i in wave 1, ϕ_t and λ_t are parameters to be estimated, and ζ_{it} is an error term with zero mean. \hat{R}_{it} is the estimated weighted fractional rank of income in wave t and $\hat{\sigma}_{R_t}^2 = \left[\sum_{i=1}^n w_i \right]^{-1} \left[\sum_{i=1}^n w_i (\hat{R}_{it} - 0.5)^2 \right]$ is the estimated weighted variance of \hat{R}_{it} in wave t .

²⁶ We also calculated elasticities for each decile in each time period since these are more relevant to get inferences on equation (2.18). However since the results are confirmed by the elasticities for all time periods and since it would overload the paper with additional tables, we decided not to present these results.

²⁷ Due to the restriction to a balanced panel we applied the first period weights to all subsequent periods.

We estimate two characteristics of the income distribution. First, we calculate the weighted average income in each period t . Second, we calculate the Gini index of income in each wave t using equation (3.3) where \hat{h}_{it} and \hat{H}_t have been replaced by respectively y_{it} and \hat{Y}_t .

All four indicators (average income, average health, income Gini and health concentration index) describe the evolution for a single cohort. Especially in the case of average health this may differ from the picture that emerges from repeated cross-section samples where births and deaths are included. In a cohort, the average health will decline.

3.3. The role of proportional income growth and evolving income inequality

We estimate (weighted) versions of (2.14), (2.15), (2.20), (2.21), (2.24), and (2.25) for each country in each wave t , except the first. To compute the terms of our decomposition technique, we need estimates of H_t^{pg} and/or H_t^{ng} which are the weighted averages of \hat{h}_{it}^{pg} and \hat{h}_{it}^{ng} . The latter are obtained by substituting y_{it} by respectively \hat{y}_{it}^{pg} and y_{it} in equation (3.1), and calculating the predicted value of health while keeping the coefficients fixed and the other variables at their actual value. For \hat{h}_{it}^{pg} we need to generate an estimate for \hat{y}_{it}^{pg} , i.e. $y_{it}(Y_t/Y_1)$. The latter estimate allows proportional income growth to differ between each period t and the first period. The sums of (2.14)-(2.15), (2.20)-(2.21) and (2.24)-(2.25) provide an indication of the *total* effect of respectively proportional income growth, income inequality, and the other determinants.

3.4. Statistical inference on income elasticity and trends

For statistical inference on the point estimates of the income elasticity at the various deciles, the trends in average income, the income Gini, average health, the health CI and its decomposition, we use the bootstrap procedure of Mills and Zandvakili (1997). We draw 100 bootstrap samples²⁸ on the level of the individual (i.e. if an individual is drawn in one time period, he is included in all time periods) which corrects the statistical inference for the dependence between time periods, repeat all calculations, and compute 95 percent normal

confidence intervals for the elasticities and all expressions of the decomposition in section 2.3.

²⁸ We did not consider a higher number of bootstrap replications due to the time-intensity of our bootstrap procedure. Nevertheless, we think 100 replications are warranted in our application since we could not reject – at the 5% level – the null hypothesis of a normal distribution for 92% of our bootstrap samples.

4. Empirical findings

We first present results on one of the crucial elements for the evolution of average health and income related health inequalities, i.e. the income elasticity of health, then present the trends in average health, IRHI, average income and income inequalities and discuss the findings on the decomposition approach. Due to space limitations, we present full model estimates of the regression model only in Appendix Table 1. The coefficients for the age-gender dummies show the expected signs and magnitudes, i.e. younger and male persons have higher health than older and female respondents. Income coefficients showed a highly non-linear pattern and were (jointly) significant in all specifications and for all countries.

4.1. Income elasticity of health

A summary of the income elasticity estimates (averages over all time periods) is presented in Table 4.1 while more details are contained in Appendix Table 2. All elasticities are below one – implying a concave income profile – are positive and increasing with income, except for Austria where the confidence intervals of all deciles overlap. In all other countries, the point estimate of the elasticity starts decreasing only at the highest deciles, but the decline is only statistically relevant for Greece and Ireland.

Table 4.1: summary of income elasticity estimates

	Austria	Belgium	Denmark	France	Finland	Germany	Greece	Ireland	Italy	Netherlands	Portugal	Spain	UK
average	0,0268	0,0137	0,0278	0,0301	0,0207	0,0296	0,0211	0,0212	0,0182	0,0152	0,0370	0,0246	0,0283
decile 1	0,0205	0,0065	0,0169	0,0160	0,0130	0,0184	0,0092	0,0160	0,0069	0,0087	0,0156	0,0113	0,0140
decile 2	0,0267	0,0094	0,0222	0,0227	0,0170	0,0260	0,0149	0,0196	0,0121	0,0120	0,0256	0,0177	0,0208
decile 3	0,0285	0,0109	0,0247	0,0260	0,0187	0,0287	0,0183	0,0216	0,0147	0,0132	0,0309	0,0208	0,0242
decile 4	0,0297	0,0123	0,0266	0,0287	0,0201	0,0303	0,0208	0,0233	0,0166	0,0142	0,0350	0,0231	0,0271
decile 5	0,0304	0,0135	0,0283	0,0310	0,0211	0,0317	0,0229	0,0248	0,0185	0,0153	0,0387	0,0254	0,0297
decile 6	0,0306	0,0146	0,0298	0,0331	0,0222	0,0328	0,0249	0,0258	0,0204	0,0163	0,0420	0,0275	0,0319
decile 7	0,0303	0,0158	0,0314	0,0352	0,0233	0,0338	0,0265	0,0259	0,0220	0,0173	0,0453	0,0296	0,0341
decile 8	0,0294	0,0173	0,0331	0,0372	0,0242	0,0342	0,0277	0,0251	0,0236	0,0183	0,0488	0,0315	0,0360
decile 9	0,0268	0,0195	0,0350	0,0390	0,0249	0,0336	0,0279	0,0225	0,0251	0,0191	0,0522	0,0329	0,0376
decile 10	0,0147	0,0173	0,0300	0,0324	0,0220	0,0260	0,0174	0,0077	0,0225	0,0179	0,0362	0,0262	0,0278

In general, we can fairly safely conclude that the income elasticity is positive and non-decreasing with income over most of the income range. The magnitude of the elasticities is rather low; for example, a doubling of income in Austria, results on average in a 2.68 percent increase in health. This may be related to the fact that (good) health has an upper bound while income is unbounded. Nevertheless, the elasticity differences across deciles are highly

relevant. Countries with particularly large differences between higher and lower deciles are Belgium, Greece, Italy, Portugal and Spain.

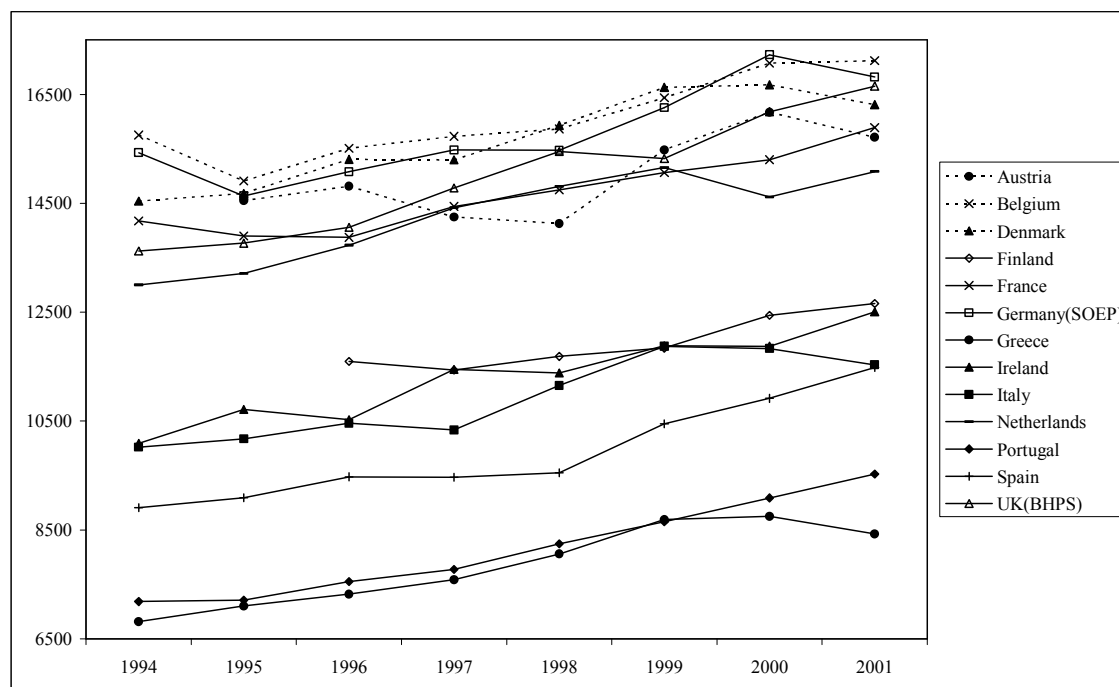
While these elasticities only determine the sign of equations (2.18) and (2.23), the findings suggest that proportional income growth is likely to lead to an increase in average health *and* IRHI. The decrease – if any – at the highest deciles is not problematic since the elasticities above median income are on average much higher than those below the median. The findings further indicate that a pro-rich (c.q. pro-poor) change in income inequality has a negative (positive) effect on average health and that we only have an unequivocal prediction for the evolution of IRHI in the case of a pro-rich evolution of income inequalities.

4.2. Trends in real incomes and income inequality, and trends in average health and income related health inequalities

Detailed information (including statistical significance) on the trends in average income, income inequality, average health and IRHI for all ECHP countries is presented in the appendix in Appendix Table 3-Appendix Table 15. Here we only discuss some summary figures.

First of all, it is obvious from Figure 4.1 that income growth has been unequal across European countries, and that there have been ups and downs in certain periods, but over the entire period (1994-2001), mean incomes have grown in a statistically significant sense in *all* countries in real terms. In percentage terms, mean annual real income growth has been particularly strong in Portugal (4%), Spain (3.6%), Ireland (3%), Greece (3%), UK (2.78%) and the Netherlands (2%), while it was below 2% in the other countries. Recall from section 2.3 that with income elasticities of health (increasing with income), the *direct* predicted effect of positive real income growth is to lead to rising IRHI, *even if income inequality does not rise*.

Figure 4.1: Evolution of average equivalent real income in 1996 Euros, 13 EU countries, 1994-2001



Second, the Gini trends in Figure 4.2 indicate that very few countries have experienced a sustained increase in income inequality over the period 1994-2001. While the trends are by no means monotonic, it is clear from the graph and from the tests (reported in Appendix Table 3-Appendix Table 15) that, on the whole, most countries have experienced either pro-poor income growth (Austria, France, Germany, Greece, Italy, Portugal and Spain) or income inequality has remained fairly stable (Belgium²⁹, Denmark, Ireland, the Netherlands and the UK).³⁰ The sole exception is Finland which shows a statistically significant positive trend: its Gini index was about 10 percent higher in 2001 than in 1996.³¹

While these findings may be seen as somewhat surprising in view of the often reported rising relative income inequality over this period in the OECD context (see e.g. Smeeding, 2002; Kenworthy and Pontusson, 2005), they are consistent with earlier findings reported by Garcia et al. (2004) and Hildebrand and Van Kerm (2005) on the same data and with the series of cross sections compared in Atkinson (2003) and Moran (2005). For example, for the same

²⁹ Lefebure (2005) analysed the data for Belgium in isolation and reported a similar trend.

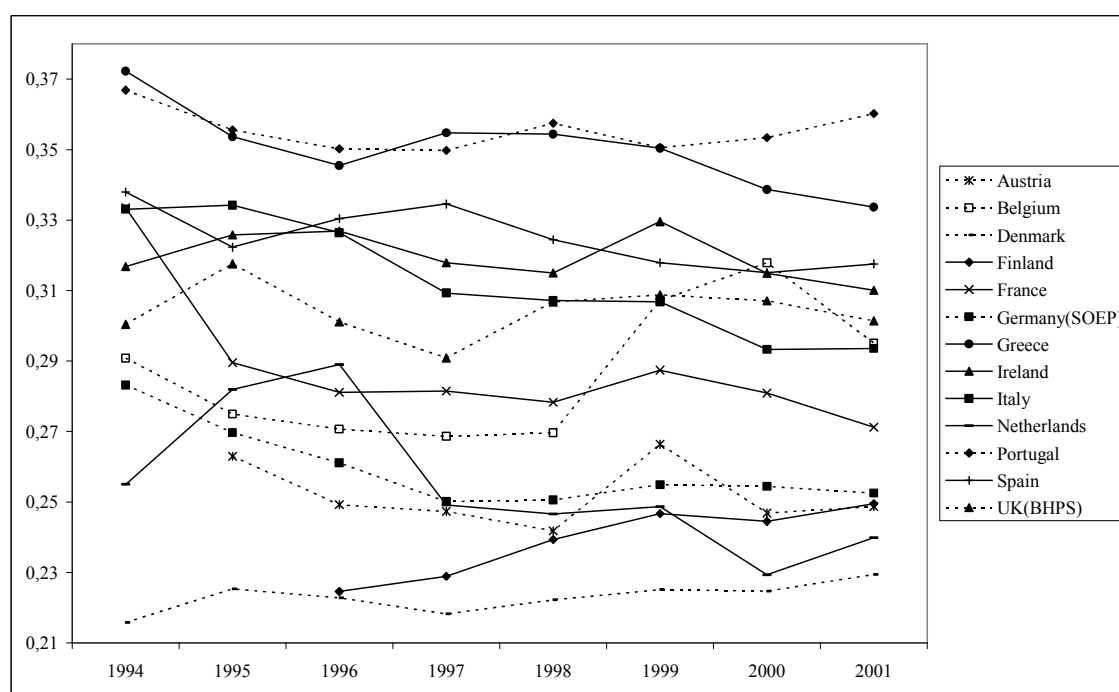
³⁰ Because of the non-monotonic trends of income inequality, it is somewhat arbitrary to subdivide the countries into two groups.

³¹ The increase in Gini indices is in line with the observed reduction of progressivity of the Finnish tax system (see e.g. Jäntti, 2005). We thank Unto Häkkinen for this suggestion.

period, also Atkinson (2003) reports stable Gini indices for the Netherlands, Italy and the UK, a modest rise in Germany and a strong increase only in Finland.

Given the estimated income elasticities and the trends in income growth, this implies that we cannot make *a priori* theoretical predictions on the direct effect of income inequality on IRHI, except for Finland where income growth is combined with an elasticity that increases with income and a pro-rich change in income inequality. For the other countries, we can only apply our decomposition technique of section 2.3 to obtain an empirical answer on the role of proportional income growth and income inequality.

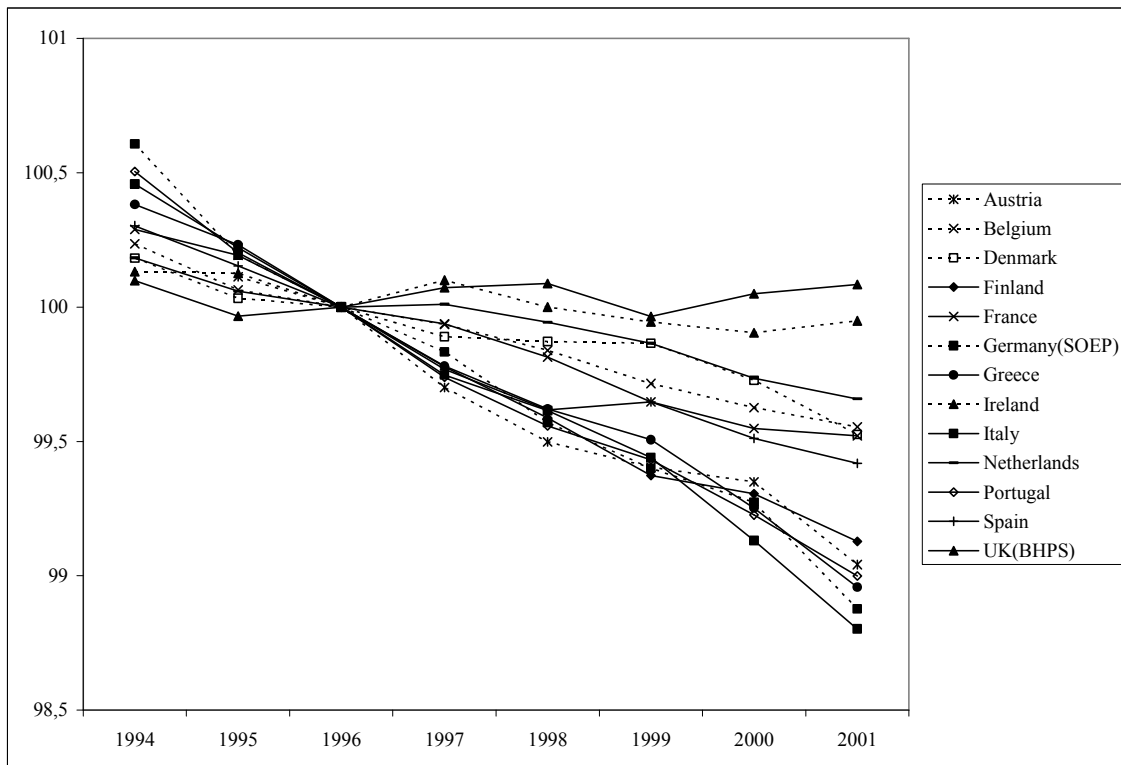
Figure 4.2: Evolution of the Gini index of income, 13 EU countries, 1994-2001



Third, Figure 4.3 describes the trends in average health utility. We have presented the trends as relative deviations from 1996 since reporting heterogeneity in self-assessed health might invalidate comparisons between countries, but probably not within countries. As expected, average health of the ageing cohort decreases but at a slow rate in all countries except the UK³² (see also Appendix Table 3-Appendix Table 15).

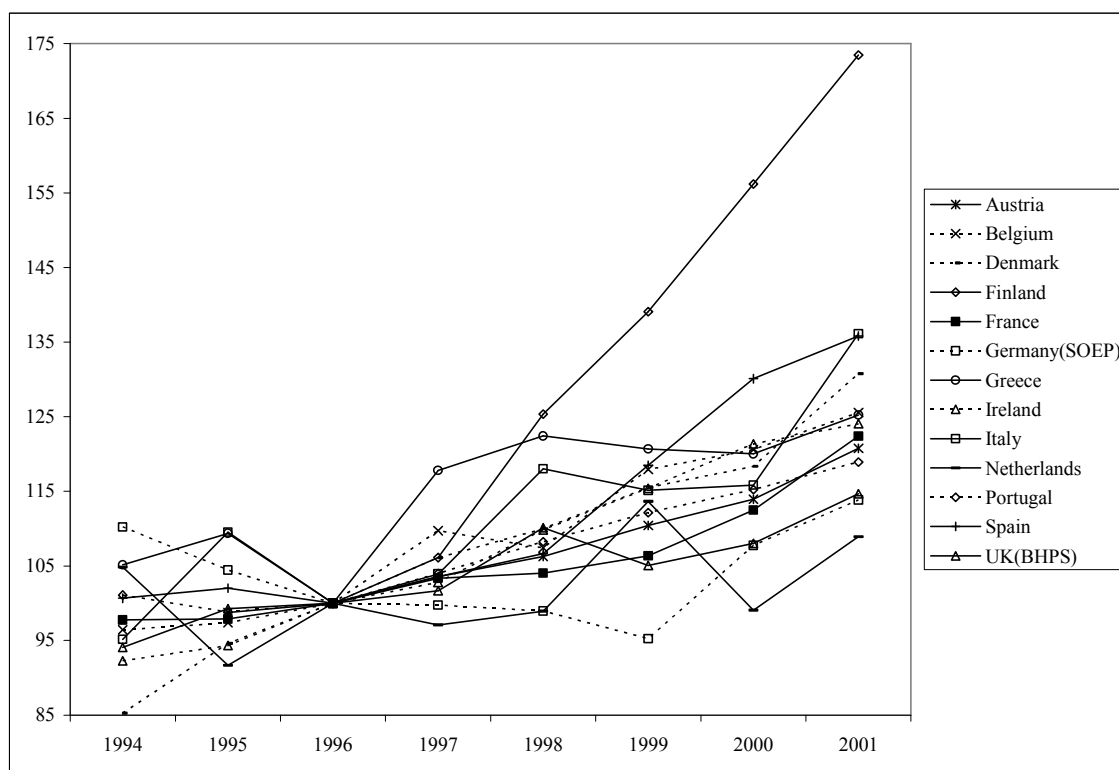
³² It is quite likely that the diverging finding for the UK is influenced by the health assessment 'compared to your own age' (see also section 3.1).

Figure 4.3: Evolution of average predicted health utility, 13 EU countries, 1994-2001, 1996=100



Fourth, in Figure 4.4 we focus on the trends of health concentration indices within countries, with 1996=100. The positive and increasing CI's indicate that relative IRHI favoured the rich in all countries and increased between 1994 and 2001 in all countries, except Germany and the Netherlands. But they have clearly risen much faster in Finland than in any other European country in this period.

Figure 4.4: Trends in health concentration indices, 13 EU countries, 1994-2001, 1996=100



In summary, we find that average income has increased in all countries, whereas relative income inequality remained stable or decreased in all countries except Finland. While average health has deteriorated over time, IRHI has risen more sharply.³³

4.3. Mean health and IRHI: the role of income growth and income inequality

Calculation of equation (2.12) was redundant since one can predict *a priori* the effect of income growth and income inequality on mean health trends. Since the income elasticities reported in section 4.1 are all between zero and one, the income profile is concave. The income elasticities are also increasing over most of the income range. Consequently, average predicted health utility will respond inelastically to proportional income growth and rising/decreasing income inequalities will have a negative/positive effect on mean health. Since all countries (except Finland) experienced stable or decreasing income inequalities, the

³³ It is worth mentioning that for a bounded variable (i.e. with potential range [0,1]) the bounds of the concentration index are determined by the variable mean: the higher the mean, the lower the bounds of the concentration index (see e.g. Wagstaff (2005) and Erreygers (2006)). We used the Wagstaff (2005)

combined health effect of income growth and income inequality was positive between 1994 and 2001.

With respect to the trends in IRHI, the mechanics of our decomposition technique of section 2.3 are somewhat more complex. We will describe the results for all countries but, due to space limitations, we only present the results for the decomposition of the difference between the concentration indices of 2001 and 1994 in Table 4.2. Detailed results for all countries are available in Appendix Table 3-Appendix Table 15.

Table 4.2: Results of our decomposition technique, 1994-2001

	Austria	Belgium	Denmark	Finland	France	Germany	Greece	Ireland	Italy	Netherlands	Portugal	Spain	UK
CI ₁₉₉₄	0,0092	0,0058	0,0067	0,0057	0,0081	0,0088	0,0140	0,0073	0,0053	0,0041	0,0201	0,0094	0,0091
H ₁₉₉₄	0,9042	0,9090	0,9205	0,8909	0,8745	0,8439	0,9092	0,9263	0,8791	0,9056	0,8370	0,8829	0,8860
Y ₁₉₉₄	14546	15754	14539	11596	14177	15431	6817	10089	10020	13000	7183	8911	13625
G ₁₉₉₄	0,2630	0,2908	0,2158	0,2246	0,3335	0,2832	0,3723	0,3168	0,3331	0,2550	0,3669	0,3380	0,3004
CI ₂₀₀₁	0,0112	0,0076	0,0103	0,0099	0,0102	0,0091	0,0166	0,0099	0,0076	0,0043	0,0237	0,0127	0,0111
H ₂₀₀₁	0,8946	0,9028	0,9145	0,8831	0,8678	0,8294	0,8962	0,9246	0,8646	0,9008	0,8245	0,8752	0,8859
Y ₂₀₀₁	15711	17118	16313	12660	15889	16819	8426	12509	11534	15081	9527	11483	16652
G ₂₀₀₁	0,2487	0,2951	0,2294	0,2495	0,2712	0,2525	0,3337	0,3101	0,2936	0,2399	0,3602	0,3176	0,3014
N	3967	3025	2544	3219	6836	7520	6134	2872	9036	4556	7097	7307	5925
CI ₂₀₀₁ -CI ₁₉₉₄	0,00205	0,00175	0,00358	0,00421	0,00206	0,00029	0,00266	0,00252	0,00228	0,00016	0,00355	0,00330	0,00199
ineq_in	-0,00030	-0,00026	0,00042	0,00037	-0,00040	-0,00113	-0,00024	0,00017	-0,00068	-0,00033	-0,00093	-0,00062	-0,00059
elas_in	-0,00004	0,00024	0,00031	0,00015	0,00038	0,00015	0,00030	-0,00012	0,00043	0,00026	0,00146	0,00081	0,00075
past_in	0,00010	0,00003	0,00006	0,00005	0,00012	0,00017	0,00015	0,00004	0,00012	0,00003	0,00035	0,00013	0,00005
ineq_ot	0,00258	0,00179	0,00244	0,00362	0,00197	0,00122	0,00214	0,00254	0,00261	0,00039	0,00249	0,00314	0,00168
elas_ot	0,00000	0,00000	0,00000	0,00000	0,00000	0,00000	-0,00003	0,00000	0,00000	0,00000	-0,00007	-0,00001	0,00000
past_ot	-0,00028	-0,00005	0,00037	0,00002	-0,00001	-0,00012	0,00034	-0,00010	-0,00020	-0,00018	0,00026	-0,00015	0,00011
ineq	0,00228	0,00153	0,00286	0,00399	0,00157	0,00009	0,00190	0,00271	0,00193	0,00006	0,00156	0,00252	0,00109
elas	-0,00004	0,00024	0,00030	0,00015	0,00038	0,00014	0,00027	-0,00013	0,00043	0,00026	0,00139	0,00080	0,00074
past	-0,00018	-0,00002	0,00043	0,00007	0,00010	0,00005	0,00050	-0,00006	-0,00009	-0,00015	0,00060	-0,00002	0,00016

Note: $CI_t = C(h_t | y_t)$; $h_t = H_t$; $y_t = Y_t$; $G_t = G(y_t)$; $CI_{2001} - CI_{1994} = C(h_{2001} | y_{2001}) - C(h_{1994} | y_{1994})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14); **past_in**: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_ terms; **elas**: sum of all elas_ terms; **past**: sum of all past_ terms; **sum**: sum of all terms; shaded: statistically significantly different from zero at 5% level. For Austria and Finland, 2001 is compared with 1995 and 1996 respectively.

The term “elas_in” is an estimate of equation (2.14). It measures the direct effect of proportional income growth on IRHI, i.e. the effect that runs via the income elasticity. If there is positive/negative income growth between period 1 and t , one would expect a positive/negative effect on IRHI if the elasticity increases with income. It is found that equation (2.14) is positive in all periods and countries where there was positive income growth and negative in case of negative income growth, except for Austria, Germany and Ireland where the effects are not statistically significant. Overall, these findings imply that without proportional income growth, the direct effect would be to find smaller IRHI, except in

normalisation to check whether the ‘predetermined’ relationship causes our finding of increasing health inequalities, but we found that it did not.

Austria, Germany and Ireland. Therefore, the estimates of equation (2.18) confirm our expectations based on the income elasticities in section 4.1 and the positive growth of average incomes in all countries in section 4.2.

The second term for which we were able to derive some theoretical predictions is equation (2.20), i.e. the direct effect of a change in income inequality. In case of an income elasticity increasing with income and pro-rich change in income inequality ‘on average’, one can predict that IRHI increases. With opposite configurations of income inequality and the elasticity (note that it is the change in elasticity evaluated at y_{it}^{pg} and y_{it} that matters), offsetting effects will occur and it is an empirical question which of the two dominates. Only for Finland and Denmark (even though the trend in income inequality is not statistically significant in Denmark), the theoretically predicted effects are observed, i.e. *ineq_in* is positive and significant when the Gini becomes larger.³⁴ Furthermore, in the majority of countries, a decreasing Gini has a negative effect on IRHI, which implies that the direct effect of income inequality is more important than the direct offsetting effect through the income elasticity (or that IRHI would have been higher without the concurrent change in income inequality). For some countries both effects seem to balance out, i.e. Belgium, Greece and Ireland. We thus conclude that for the time period considered, income inequality has evolved in the same direction as IRHI (although there are some exceptions) and that only in the case of Finland (and Denmark) we observe the theoretically predictable effect of increasing health inequalities due to the combination of growing income inequality and an elasticity that increases with income. In the majority of countries, opposite effects from proportional income growth and decreasing income inequalities are observed.

Next, we discuss the term “*past_in*” which estimates equation (2.24), i.e. the effect of changes in the age structure on IRHI (through the effect of health on the income elasticity). This can be considered as a ‘control term’ measuring the impact of changes in ‘other variables. It is nonetheless of interest here as it measures the ‘direct’ impact of ageing (in a single cohort). Obviously the ‘direct’ effect of ageing is to *increase* IRHI as it increases the income elasticity through a reduced mean health.

³⁴ For some countries the Gini also increases in some waves, but without being overall ‘pro-rich’, i.e. the changes in the income distribution occur mainly among the rich or among the poor. These countries and waves are Austria (wave 6), Belgium (wave 6-8), Italy (wave 2), the Netherlands (wave 2), UK (all waves except 4). Also in Ireland we find an increase in the Gini but this is combined with a decreasing elasticity.

Next, we discuss the term “elas_ot” (cf. equation (2.15), i.e. the indirect effect of proportional income growth on the elasticities of the age-sex structure. The effect runs through the impact of income growth on average health which affects these elasticities. We could not derive a priori predictions for these terms, but find that they all have a very low contribution (in most cases smaller than 0.00000) and are in many cases statistically insignificant.

The term “ineq_ot” (cf. equation (2.21)) summarizes the indirect effect of changing income inequality. There is an indirect effect on the elasticities of the age-sex dummies (through average health) and an indirect effect through reranking on the concentration indices of the same variables. Again, since no theoretical predictions could be derived, this is an empirical question. We observe in most cases a positive sign, although the effect is not significant for Germany and the Netherlands. Its interpretation is complicated since it is determined by the reranking *and* the effect on the elasticities.³⁵ The effect seems to increase over time, which is not surprising as reranking (and the change in elasticities) becomes a more important phenomenon when a longer period is considered (see Appendix Table 3-Appendix Table 15 for more details).

Finally, the term “past_ot” (cf. equation (2.25)) summarizes the effect of changes in the age-sex dummies on the elasticities (both through the average of the dummies and average health) and on the concentration indices of these variables. Again, this term is only of limited interest since it enters the decomposition as a “control term”. Recall from the discussion of “in_ot” the relevance of this term for issues related to ageing. The term is insignificant for all countries, except the Netherlands.

Table 4.2 (and Appendix Table 3-Appendix Table 15), also shows the sum of all ineq, elas and past terms. These are interesting since they reflect the sum of the direct *and* indirect effects of respectively income inequality, proportional income growth, and ‘ageing’. In the case of opposite indirect and direct effects, they indicate which effects dominate. First, for all

³⁵ Given that the part γX_t is negative in our specification (see Appendix Table 1), this term can only be positive if $\frac{C(x_{it}|y_{it})}{H_t} < \frac{C(x_{it}|y_{i1})}{H_t^{pg}}$. If income inequality decreases, one would expect H_t^{pg} to be smaller than H_t (due to the positive, but decreasing with income, income profile). It follows that the required inequality always arises if

countries, except Denmark, the total effect of 8 years of ageing is insignificant. Although, the direct effect of ageing was to increase the income elasticity of health, the combination of direct and indirect effects is no longer statistically significant. We may conclude that IRHI is hardly influenced by 8 years of ageing in our data. An alternative interpretation is to consider the term ‘past’ as a control term that includes age- and sex-related reporting heterogeneity of self-assessed health.

Second, “elas” is in most cases almost identical to `elas_in`, since the effect of proportional income growth on the elasticity of the other variables is (almost) negligible. This means that positive income growth adds in all countries, except Austria, Germany and Ireland, to an increase in IRHI. This is certainly the most striking finding of our empirical analysis.

Third, the sign of the total effect of income inequality “ineq” is positive, except for Germany and the Netherlands (and Portugal, see Appendix Table 13). This implies that, despite decreasing or stable income inequality, the overall effect has been an increase in IRHI. In other words, the indirect effects seem to dominate for the effect of income inequality. This may not be so surprising given that this term picks up most of the effects of income reranking. Note that for the effect of income growth, the direct effect was most important.

$C(x_{it}|y_{it})$ is not larger than $C(x_{it}|y_{it})$, i.e. if reranking does not increase income-related inequalities in the other determinants, here age and sex.

5. Conclusion and discussion

This paper set out to try and answer the question to what extent Europe's twin goals of income growth and reduction of social inequalities – as formulated in the Lisbon strategic goals – are compatible, both on theoretical grounds and empirically. In particular, we concentrated on the consequences of changes in the income distribution for changes in the distribution of health by income. We developed a decomposition technique to analyse the role of changes in (proportional) income growth and income inequality on expected trends in income-related health inequality. It was then applied to the empirical analysis of these trends in 13 European countries using 8 waves of European panel data.

The theoretical model indicates that – when the relationship between income and health is concave – proportional income growth increases average health and rising income inequality reduces average health. With respect to trends in IRHI, it is more difficult to isolate the role of income growth and income inequality. Our solution was to introduce two hypothetical health levels. Using this method, we found that income growth and income inequality have a direct and indirect effect on IRHI. *A priori* sign predictions could be obtained for the direct effects, but not for the indirect effects. Building on Contoyannis and Forster (1999a), we showed that the expected *direct* effect of proportional income growth depends crucially on the slope of the income elasticity. If this elasticity is rising/decreasing with income 'on average', IRHI increase/decline. With respect to the *direct* effect of changes in income inequality, we predict increasing/decreasing IRHI in case of 'on average' pro-rich/pro-poor evolving income inequality in combination with an income elasticity that increases/decreases with income 'on average'. The signs of the indirect effects (both for proportional income growth and income inequality) could not be inferred from our decomposition technique and are therefore empirical questions.

In our empirical analysis, we first examined how estimates of the income elasticity of health varied with income since it is an important determinant of the consequences of income growth and income inequality. Using a flexible functional form for the estimation, we found that in all countries, the marginal effect of income on health is positive and *decreasing* with income. In other words, the income-health relationship is concave, as expected. But secondly, and

more importantly, the income elasticity of health was nonetheless found to be *rising* with income ('on average').

Second, we presented trends on income growth, income inequality, average health and IRHI. While between 1994 and 2001, all European countries were found to have experienced real income growth, in most countries this growth was not equally distributed by income level. In all EU countries, income growth was found to be either pro-poor or equally distributed, with one exception: only Finland experienced a clear pro-rich growth. Given this combination of income elasticities rising with income and mostly pro-poor growth, no clear pattern of the impact of income inequality upon income-related health inequality could be predicted *a priori*. Only for Finland, a steady rise in the concentration index of health could be anticipated. We also presented evidence on the changing distribution of health. Since we analysed a cohort of individuals, average health deteriorated over time, while IRHI increased.

Third, in order to clarify the role and quantify the contribution of proportional income growth and income inequality in the evolution of average health and IRHI, we used our decomposition technique to disentangle direct from indirect effects. We concluded that proportional income growth leads to better average health, and this is true *a fortiori* when simultaneously income inequality is falling. So economic growth coupled with reduced income inequality is good for mean health levels. However, both the direct and indirect effects of proportional income growth were found to increase IRHI. On the other hand, the direct effect of falling income inequality in Europe led (with a few exceptions) to lower IRHI, while its indirect effect led to increased IRHI. Finally, the model also allows for an analysis of the direct and indirect effects of ageing but we found that the 8 years of ageing in our cohort added next to nothing to IRHI.

So can the twin goals of economic growth and greater social (health) cohesion in the European Union be reconciled? Our analysis suggests that there may be a problem and that Wagstaff's (2002b) hypothesis that developing countries are "swimming against the tide?" by trying to couple growth with reducing relative inequalities may similarly apply to high-income economies like the European. In any case, given the universal observation that everywhere in the 'old' European Union the income elasticity of health rises with income, even proportional growth, leaving income inequality unchanged, will lead to greater IRHI. Had European countries not been able to lower or stabilize their income inequality in the

nineties, then more countries than only Finland would have experienced sharp increases in socio-economic inequalities in health as a result of the economic growth. Obviously, the overall welfare implications of improved mean health coupled with rising relative IRHI depend on the relative weight given to improvements in the mean versus the distribution. This trade-off can be made explicit by using social welfare type functions (cf Wagstaff, 2002a) but the results depend crucially on the prevailing degree and type of societal aversion to health inequality. Little is known on this empirically and more evidence is needed before these measures can be applied for a welfare analysis of health trends.

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Appendix Table 1: health equation estimates

	Austria	Belgium	Denmark	Finland	France	Germany	Greece	Italy	Ireland	Netherlands	Portugal	Spain	UK
M3044	-0.018**	-0.011**	-0.016**	-0.019**	-0.018**	-0.030**	-0.009**	-0.023**	-0.010**	-0.012**	-0.022**	-0.018**	-0.013**
M4559	-0.060**	-0.027**	-0.037**	-0.063**	-0.046**	-0.079**	-0.038**	-0.055**	-0.026**	-0.027**	-0.067**	-0.047**	-0.027**
M6069	-0.078**	-0.040**	-0.052**	-0.087**	-0.069**	-0.103**	-0.087**	-0.090**	-0.045**	-0.044**	-0.120**	-0.087**	-0.041**
M70+	-0.109**	-0.061**	-0.076**	-0.098**	-0.093**	-0.125**	-0.136**	-0.144**	-0.053**	-0.051**	-0.163**	-0.100**	-0.039**
F1629	-0.002	-0.004	-0.007*	-0.007*	-0.002	-0.008*	0.005*	-0.003+	-0.001	-0.008*	0.001	-0.001	-0.011**
F3044	-0.017**	-0.019**	-0.023**	-0.022**	-0.025**	-0.035**	-0.012**	-0.030**	-0.011**	-0.021**	-0.032**	-0.020**	-0.019**
F4559	-0.057**	-0.038**	-0.055**	-0.061**	-0.056**	-0.090**	-0.051**	-0.071**	-0.021**	-0.038**	-0.095**	-0.065**	-0.041**
F6069	-0.089**	-0.052**	-0.062**	-0.080**	-0.079**	-0.112**	-0.109**	-0.116**	-0.037**	-0.053**	-0.156**	-0.114**	-0.032**
F70+	-0.127**	-0.070**	-0.092**	-0.120**	-0.105**	-0.148**	-0.151**	-0.170**	-0.057**	-0.070**	-0.196**	-0.134**	-0.058**
(eqinc/10000)	0.042**	0.011**	0.025**	0.027**	0.031**	0.032**	0.050**	0.024**	0.047**	0.016**	0.069**	0.040**	0.030**
(eqinc/10000) ²	-0.010**	-0.001**	-0.002**	-0.005**	-0.004**	-0.005**	-0.014**	-0.003**	-0.014**	-0.002**	-0.015**	-0.008**	-0.004**
(eqinc/10000) ³	0.001**	0.000**		0.000**	0.000**	0.000**	0.001**	0.000**	0.002**	0.000**	0.001**	0.001**	0.000**
(eqinc/10000) ⁴	-0.000*	-0.000**			-0.000**		-0.000**		-0.000**			-0.000**	-0.000**
(eqinc/10000) ⁵									0.000**				
constant	0.907**	0.920**	0.92**	0.913**	0.880**	0.870**	0.929**	0.911**	0.914**	0.914**	0.861**	0.902**	0.879**
σ_ε^2	0.079**	0.064**	0.078**	0.068**	0.105**	0.114**	0.094**	0.086**	0.060**	0.063**	0.114**	0.090**	0.098**
Observations	27769	24200	20352	19314	54688	60160	49072	72288	22976	36448	56776	58456	47400

Notes: + significant at 10%; * significant at 5%; ** significant at 1%

Appendix Table 2: 95 percent confidence intervals of income elasticity estimates

	Austria		Belgium		Denmark		France		Finland		Germany		Greece		Ireland		Italy		Netherlands		Portugal		Spain	
	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+	95-	95+
average	0.0228	0.0308	0.0093	0.0182	0.0220	0.0336	0.0263	0.0339	0.0162	0.0251	0.0239	0.0352	0.0185	0.0236	0.0178	0.0247	0.0163	0.0202	0.0121	0.0184	0.0337	0.0403	0.0224	0.0268
decile 1	0.0148	0.0263	0.0033	0.0096	0.0130	0.0208	0.0135	0.0184	0.0096	0.0164	0.0134	0.0235	0.0078	0.0107	0.0126	0.0193	0.0059	0.0079	0.0067	0.0107	0.0134	0.0178	0.0097	0.0130
decile 2	0.0202	0.0331	0.0052	0.0136	0.0173	0.0271	0.0194	0.0261	0.0128	0.0212	0.0196	0.0325	0.0128	0.0171	0.0157	0.0236	0.0105	0.0137	0.0093	0.0147	0.0222	0.0290	0.0154	0.0200
decile 3	0.0223	0.0347	0.0063	0.0155	0.0193	0.0301	0.0223	0.0297	0.0142	0.0232	0.0219	0.0354	0.0156	0.0209	0.0175	0.0258	0.0128	0.0165	0.0103	0.0161	0.0270	0.0348	0.0182	0.0234
decile 4	0.0239	0.0356	0.0074	0.0171	0.0209	0.0324	0.0247	0.0327	0.0154	0.0249	0.0235	0.0371	0.0179	0.0238	0.0190	0.0275	0.0146	0.0186	0.0111	0.0173	0.0309	0.0391	0.0204	0.0259
decile 5	0.0250	0.0357	0.0085	0.0185	0.0223	0.0344	0.0268	0.0352	0.0162	0.0259	0.0248	0.0385	0.0198	0.0261	0.0204	0.0291	0.0164	0.0206	0.0120	0.0188	0.0344	0.0430	0.0226	0.0283
decile 6	0.0258	0.0355	0.0095	0.0198	0.0235	0.0362	0.0287	0.0375	0.0172	0.0272	0.0260	0.0396	0.0216	0.0281	0.0215	0.0300	0.0182	0.0227	0.0129	0.0198	0.0376	0.0464	0.0246	0.0304
decile 7	0.0259	0.0348	0.0106	0.0211	0.0248	0.0380	0.0307	0.0398	0.0182	0.0283	0.0270	0.0405	0.0231	0.0298	0.0219	0.0300	0.0197	0.0244	0.0137	0.0210	0.0409	0.0497	0.0267	0.0325
decile 8	0.0249	0.0358	0.0121	0.0226	0.0262	0.0400	0.0326	0.0418	0.0191	0.0294	0.0275	0.0409	0.0243	0.0311	0.0212	0.0291	0.0211	0.0261	0.0145	0.0221	0.0444	0.0532	0.0286	0.0343
decile 9	0.0212	0.0329	0.0142	0.0247	0.0278	0.0422	0.0343	0.0438	0.0197	0.0302	0.0263	0.0409	0.0246	0.0312	0.0183	0.0266	0.0224	0.0279	0.0151	0.0231	0.0475	0.0568	0.0300	0.0358
decile 10	0.0028	0.0266	0.0098	0.0249	0.0221	0.0380	0.0268	0.0380	0.0157	0.0282	0.0111	0.0409	0.0130	0.0219	0.0008	0.0147	0.0174	0.0276	0.0138	0.0220	0.0222	0.0501	0.0202	0.0321

Appendix Table 3: detailed information for Austria

	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE						
CI	0,0092	0,0093	0,0096	0,0099	0,0103	0,0106	0,0112	CI	wave2	wave3	wave4	wave5	wave6	wave7
h	0,9042	0,9032	0,9005	0,8987	0,8978	0,8973	0,8946	wave3	0,0001					
y	14546	14813	14247	14129	15480	16173	15711	wave4	0,0004	0,0003				
G	0,2630	0,2492	0,2473	0,2418	0,2664	0,2468	0,2487	wave5	0,0007	0,0006	0,0003			
N	3967	3967	3967	3967	3967	3967	3967	wave6	0,0011	0,0010	0,0006	0,0004		
CI-CI1		0,00011	0,00044	0,00070	0,00109	0,00141	0,00205	wave7	0,0014	0,0013	0,0010	0,0007	0,0003	
ineq_in		-0,00029	-0,00032	-0,00036	-0,00013	-0,00034	-0,00030	wave8	0,0020	0,0019	0,0016	0,0014	0,0010	0,0006
elas_in		0,00000	0,00000	0,00000	-0,00003	-0,00007	-0,00004	h						
past_in		0,00001	0,00003	0,00004	0,00006	0,00008	0,00010	wave3	-0,0010					
ineq_ot		0,00050	0,00088	0,00103	0,00132	0,00194	0,00258	wave4	-0,0037	-0,0027				
elas_ot		0,00000	0,00000	0,00000	-0,00001	0,00000		wave5	-0,0056	-0,0045	-0,0018			
past_ot		-0,00010	-0,00015	-0,00002	-0,00013	-0,00020	-0,00028	wave6	-0,0064	-0,0054	-0,0027	-0,0008		
ineq		0,00021	0,00056	0,00067	0,00119	0,00160	0,00228	wave7	-0,0069	-0,0059	-0,0032	-0,0013	-0,0005	
elas		-0,00001	0,00000	0,00000	-0,00003	-0,00008	-0,00004	wave8	-0,0097	-0,0087	-0,0060	-0,0041	-0,0033	-0,0028
past		-0,00009	-0,00012	0,00002	-0,00007	-0,00012	-0,00018	y						
								wave3	267					
								wave4	-299	-566				
								wave5	-417	-684	-118			
								wave6	934	667	1233	1351		
								wave7	1627	1361	1927	2045	693	
								wave8	1165	899	1464	1583	231	-462
								G						
								wave3	-0,0137					
								wave4	-0,0157	-0,0019				
								wave5	-0,0211	-0,0074	-0,0055			
								wave6	0,0034	0,0171	0,0191	0,0245		
								wave7	-0,0161	-0,0024	-0,0005	0,0050	-0,0195	
								wave8	-0,0143	-0,0005	0,0014	0,0069	-0,0176	0,0019

Note: $CI = C(h_u|y_u)$; $h = H_t$; $y = Y_t$; $G = G(y_u)$; $CI-CI1 = C(h_u|y_u) - C(h_{n1}|y_{n1})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14);

past_in: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 4: detailed information for Belgium

	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE							
CI	0,0058	0,0059	0,0060	0,0066	0,0065	0,0071	0,0073	0,0076	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7
h	0,9090	0,9075	0,9069	0,9063	0,9054	0,9043	0,9035	0,9028	wave2	0,0001						
y	15754	14909	15510	15728	15865	16437	17074	17118	wave3	0,0002	0,0002					
G	0,2908	0,2749	0,2707	0,2687	0,2697	0,3073	0,3178	0,2951	wave4	0,0008	0,0007	0,0006				
N	3025	3025	3025	3025	3025	3025	3025	3025	wave5	0,0007	0,0006	0,0004	-0,0001			
CI-CI1		0,00006	0,00021	0,00080	0,00066	0,00129	0,00145	0,00175	wave6	0,0013	0,0012	0,0011	0,0005	0,0006		
ineq_in		-0,00007	-0,00029	-0,00008	-0,00013	-0,00012	-0,00033	-0,00026	wave7	0,0015	0,0014	0,0012	0,0007	0,0008	0,0002	
elas_in		-0,00015	-0,00004	0,00000	0,00002	0,00012	0,00023	0,00024	wave8	0,0018	0,0017	0,0015	0,0010	0,0011	0,0005	0,0003
past_in		0,00000	0,00001	0,00001	0,00002	0,00002	0,00003	0,00003	h							
ineq_ot		0,00029	0,00062	0,00099	0,00089	0,00136	0,00165	0,00179	wave2	-0,0016						
elas_ot		0,00000	0,00000	0,00000	0,00000	0,00000	0,00000	0,00000	wave3	-0,0021	-0,0006					
past_ot		-0,00002	-0,00007	-0,00012	-0,00013	-0,00008	-0,00012	-0,00005	wave4	-0,0027	-0,0012	-0,0006				
ineq		0,00022	0,00032	0,00091	0,00075	0,00124	0,00132	0,00153	wave5	-0,0036	-0,0020	-0,0015	-0,0009			
elas		-0,00015	-0,00004	0,00000	0,00002	0,00012	0,00023	0,00024	wave6	-0,0047	-0,0032	-0,0026	-0,0020	-0,0011		
past		-0,00002	-0,00006	-0,00010	-0,00011	-0,00006	-0,00010	-0,00002	wave7	-0,0055	-0,0040	-0,0034	-0,0028	-0,0019	-0,0008	
									wave8	-0,0062	-0,0046	-0,0040	-0,0035	-0,0026	-0,0015	-0,0007
									y							
									wave2	-845						
									wave3	-244	601					
									wave4	-25	819	218				
									wave5	111	956	355	137			
									wave6	683	1528	926	708	571		
									wave7	1320	2165	1564	1346	1209	637	
									wave8	1364	2209	1608	1390	1253	681	44
									G							
									wave2	-0,0159						
									wave3	-0,0201	-0,0042					
									wave4	-0,0222	-0,0063	-0,0020				
									wave5	-0,0212	-0,0053	-0,0010	0,0010			
									wave6	0,0164	0,0323	0,0366	0,0386	0,0376		
									wave7	0,0270	0,0429	0,0471	0,0492	0,0482	0,0106	
									wave8	0,0043	0,0202	0,0244	0,0264	0,0254	-0,0122	-0,0227

Note: $CI = C(h_u|y_u)$; $h = H_t$; $y = Y_t$; $G = G(y_u)$; $CI-CI1 = C(h_u|y_u) - C(h_{n1}|y_{n1})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14);

past_in: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 5: detailed information for Denmark

	wave1 wave2 wave3 wave4 wave5 wave6 wave7 wave8								PAIRWISE TESTS ON DIFFERENCE															
	CI	h	y	G	N	Clt-CII	ineq_in	elas_in	past_in	ineq_ot	elas_ot	past_ot	ineq	elas	past	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7	
CI	0.0067	0.0075	0.0079	0.0084	0.0087	0.0091	0.0093	0.0103								CI	0.0007							
h	0.9205	0.9192	0.9189	0.9179	0.9177	0.9176	0.9164	0.9145								wave2	0.0007							
y	14539	14679	15307	15297	15930	16627	16675	16313								wave3	0.0012	0.0004						
G	0.2158	0.2253	0.2228	0.2182	0.2222	0.2251	0.2247	0.2294								wave4	0.0016	0.0009	0.0005					
N	2544	2544	2544	2544	2544	2544	2544	2544								wave5	0.0019	0.0012	0.0008	0.0003				
Clt-CII		0.00074	0.00116	0.00164	0.00195	0.00238	0.00260	0.00358								wave6	0.0024	0.0016	0.0012	0.0007	0.0004			
ineq_in		0.00005	0.00010	0.00011	0.00028	0.00035	0.00025	0.00042								wave7	0.0026	0.0019	0.0014	0.0010	0.0007	0.0002		
elas_in		0.00003	0.00014	0.00014	0.00025	0.00035	0.00036	0.00031								wave8	0.0036	0.0028	0.0024	0.0019	0.0016	0.0012	0.0010	
past_in		0.00001	0.00002	0.00002	0.00003	0.00004	0.00005	0.00006								h								
ineq_ot		0.00060	0.00081	0.00126	0.00131	0.00146	0.00171	0.00244								wave2	-0.0014							
elas_ot		0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000								wave3	-0.0017	-0.0003						
past_ot		0.00005	0.00009	0.00011	0.00008	0.00018	0.00025	0.00037								wave4	-0.0027	-0.0013	-0.0010					
ineq		0.00065	0.000914	0.001364	0.001594	0.001812	0.001958	0.002855								wave5	-0.0028	-0.0015	-0.0012	-0.0002				
elas		0.00003	0.000141	0.000139	0.000244	0.000347	0.000354	0.000302								wave6	-0.0029	-0.0015	-0.0012	-0.0002	-0.0001			
past		0.00006	0.000105	0.000136	0.000109	0.000223	0.000294	0.000426								wave7	-0.0042	-0.0028	-0.0025	-0.0015	-0.0013	-0.0013		
															wave8	-0.0060	-0.0047	-0.0044	-0.0033	-0.0032	-0.0031	-0.0019		
															y									
															wave2	139								
															wave3	768	629							
															wave4	758	619	-10						
															wave5	1391	1252	623	633					
															wave6	2088	1949	1320	1330	697				
															wave7	2136	1997	1368	1378	745	48			
															wave8	1774	1635	1006	1016	383	-314	-362		
															G									
															wave2	0.0095								
															wave3	0.0070	-0.0026							
															wave4	0.0024	-0.0072	-0.0046						
															wave5	0.0064	-0.0031	-0.0006	0.0040					
															wave6	0.0093	-0.0003	0.0023	0.0069	0.0029				
															wave7	0.0089	-0.0006	0.0020	0.0066	0.0025	-0.0003			
															wave8	0.0136	0.0040	0.0066	0.0112	0.0072	0.0043	0.0046		

Note: CI= $C(h_{it}|y_{it})$; h= H_t ; y= Y_t ; G= $G(y_{it})$; Clt-CII= $C(h_{it}|y_{it}) - C(h_{it}|y_{it})$; ineq_in: equation (2.20); elas_in: equation (2.14);

past_in: equation (2.24); ineq_ot: equation (2.21); elas_ot: equation (2.15); past_ot: equation (2.25); ineq: sum of all ineq_-terms; elas: sum of all elas_-terms; past: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 6: detailed information for Finland

	wave3 wave4 wave5 wave6 wave7 wave8						PAIRWISE TESTS ON DIFFERENCE															
	CI	h	y	G	N	Clt-CII	ineq_in	elas_in	past_in	ineq_ot	elas_ot	past_ot	ineq	elas	past	CI	wave3	wave4	wave5	wave6	wave7	
CI	0.0057	0.0061	0.0072	0.0080	0.0090	0.0099										CI	0.0004					
h	0.8909	0.8889	0.8872	0.8853	0.8847	0.8831										wave4	0.0015	0.0011				
y	11596	11433	11685	11839	12444	12660										wave5	0.0022	0.0019	0.0008			
G	0.2246	0.2289	0.2393	0.2467	0.2445	0.2495										wave6	0.0032	0.0029	0.0018	0.0010		
N	3219	3219	3219	3219	3219	3219										wave7	0.0042	0.0039	0.0028	0.0020	0.0010	
Clt-CII		0.00035	0.00145	0.00224	0.00322	0.00421										wave8	0.0042	0.0039	0.0028	0.0020	0.0010	0.0010
ineq_in		0.00012	0.00029	0.00035	0.00032	0.00037										h						
elas_in		-0.00003	0.00001	0.00004	0.00012	0.00015										wave4	-0.00198					
past_in		0.00001	0.00002	0.00003	0.00004	0.00005										wave5	-0.00368	-0.0017				
ineq_ot		0.00028	0.00107	0.00154	0.00246	0.00362										wave6	-0.00558	-0.0036	-0.0019			
elas_ot		0.00000	0.00000	0.00000	0.00000	0.00000										wave7	-0.00619	-0.00422	-0.00251	-0.00061		
past_ot		-0.00003	0.00006	0.00028	0.00028	0.00002										wave8	-0.00778	-0.0058	-0.0041	-0.0022	-0.00158	
ineq		0.00040	0.00136	0.00190	0.00278	0.00399										y						
elas		-0.00003	0.00001	0.00004	0.00012	0.00015										wave4	-163					
past		-0.00002	0.00008	0.00031	0.00032	0.00007										wave5	89	252				
																wave6	243	406	154			
																wave7	848	1.011	759	605		
																wave8	1.064	1.226	974	820	216	
																G						
																wave4	0.0043					
																wave5	0.0147	0.0104				
																wave6	0.0221	0.0178	0.0074			
																wave7	0.0200	0.0157	0.0053	-0.0021		
																wave8	0.0249	0.0206	0.0102	0.0028	0.0049	

Note: CI= $C(h_{it}|y_{it})$; h= H_t ; y= Y_t ; G= $G(y_{it})$; Clt-CII= $C(h_{it}|y_{it}) - C(h_{it}|y_{it})$; ineq_in: equation (2.20); elas_in: equation (2.14);

past_in: equation (2.24); ineq_ot: equation (2.21); elas_ot: equation (2.15); past_ot: equation (2.25); ineq: sum of all ineq_-terms; elas: sum of all elas_-terms; past: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 7: detailed information for France

	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE							
CI	0.0081	0.0082	0.0083	0.0086	0.0087	0.0089	0.0094	0.0102	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7
h	0.8745	0.8736	0.8719	0.8714	0.8703	0.8689	0.8680	0.8678	CI	0.0000						
y	14177	13902	13872	14442	14749	15061	15300	15889	CI	0.0002	0.0002					
G	0.3335	0.2896	0.2811	0.2815	0.2783	0.2874	0.2809	0.2712	CI	0.0005	0.0005	0.0003				
N	6836	6836	6836	6836	6836	6836	6836	6836	CI	0.0005	0.0005	0.0003	0.0001			
CI-CII	0.00001	0.00019	0.00047	0.00052	0.00072	0.00123	0.00206		CI	0.0007	0.0007	0.0005	0.0002	0.0002		
ineq_in	-0.00010	-0.00028	-0.00017	-0.00023	-0.00035	-0.00025	-0.00040		CI	0.0012	0.0012	0.0010	0.0008	0.0007	0.0005	
elas_in	-0.00007	-0.00008	0.00006	0.00014	0.00021	0.00026	0.00038		CI	0.0021	0.0020	0.0019	0.0016	0.0015	0.0013	0.0008
past_in	0.00002	0.00004	0.00005	0.00007	0.00009	0.00010	0.00012		h							
ineq_ot	0.00028	0.00068	0.00066	0.00065	0.00094	0.00120	0.00197		wave2	-0.0008						
elas_ot	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000		wave3	-0.0025	-0.0017					
past_ot	-0.00012	-0.00017	-0.00013	-0.00011	-0.00017	-0.00009	-0.00001		wave4	-0.0031	-0.0022	-0.0005				
ineq	0.00018	0.00040	0.00048	0.00042	0.00059	0.00095	0.00157		wave5	-0.0041	-0.0033	-0.0016	-0.0011			
elas	-0.00007	-0.00008	0.00007	0.00014	0.00021	0.00026	0.00038		wave6	-0.0056	-0.0048	-0.0031	-0.0025	-0.0015		
past	-0.00010	-0.00013	-0.00008	-0.00004	-0.00009	0.00001	0.00010		wave7	-0.0065	-0.0056	-0.0039	-0.0034	-0.0023	-0.0009	
									wave8	-0.0067	-0.0059	-0.0042	-0.0036	-0.0026	-0.0011	-0.0002
									y							
									wave2	-275						
									wave3	-305	-30					
									wave4	265	540	570				
									wave5	572	847	877	307			
									wave6	884	1159	1189	619	312		
									wave7	1123	1398	1427	858	551	239	
									wave8	1712	1988	2017	1447	1140	828	590
									G							
									wave2	-0.0439						
									wave3	-0.0524	-0.0085					
									wave4	-0.0520	-0.0081	0.0004				
									wave5	-0.0552	-0.0113	-0.0028	-0.0032			
									wave6	-0.0461	-0.0022	0.0063	0.0059	0.0091		
									wave7	-0.0525	-0.0086	-0.0002	-0.0006	0.0027	-0.0065	
									wave8	-0.0623	-0.0184	-0.0099	-0.0103	-0.0071	-0.0162	-0.0098

Note: $CI = C(h_{it}|y_{it})$; $h = H_t$; $y = Y_t$; $G = G(y_{it})$; $CI-CII = C(h_{it}|y_{it}) - C(h_{it}|y_{it})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14); **past_in**: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 8: detailed information for Germany

	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE							
CI	0.0088	0.0083	0.0080	0.0080	0.0079	0.0076	0.0086	0.0091	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7
h	0.8439	0.8405	0.8388	0.8374	0.8352	0.8338	0.8327	0.8294	CI	-0.00046						
y	15431	14637	15077	15482	15475	16257	17229	16819	CI	-0.00082	-0.0004					
G	0.2832	0.2697	0.2611	0.2501	0.2506	0.2549	0.2544	0.2525	CI	-0.0008	-0.0004	0.0000				
N	7520	7520	7520	7520	7520	7520	7520	7520	CI	-0.0009	-0.0004	-0.0001	-0.0001			
CI-CII	-0.00046	-0.00082	-0.00083	-0.00090	-0.00119	-0.00020	0.00029		CI	-0.0012	-0.0007	-0.0004	-0.0004	-0.0003		
ineq_in	-0.00030	-0.00062	-0.00100	-0.00101	-0.00099	-0.00108	-0.00113		CI	-0.0002	0.0003	0.0006	0.0006	0.0007	0.0010	
elas_in	-0.00011	-0.00005	0.00001	0.00001	0.00009	0.00018	0.00015		CI	0.0003	0.0007	0.0011	0.0011	0.0012	0.0015	0.0005
past_in	0.00002	0.00005	0.00007	0.00009	0.00012	0.00014	0.00017		h							
ineq_ot	-0.00007	-0.00024	0.00005	0.00001	-0.00041	0.00062	0.00122		wave2	-0.0035						
elas_ot	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000		wave3	-0.0051	-0.0016					
past_ot	-0.00001	0.00004	0.00004	0.00000	-0.00001	-0.00006	-0.00012		wave4	-0.0065	-0.0030	-0.0014				
ineq	-0.00037	-0.00086	-0.00095	-0.00099	-0.00140	-0.00046	0.00009		wave5	-0.0087	-0.0052	-0.0036	-0.0022			
elas	-0.00011	-0.00005	0.00001	0.00001	0.00009	0.00018	0.00014		wave6	-0.0101	-0.0067	-0.0050	-0.0036	-0.0014		
past	0.00002	0.00009	0.00011	0.00009	0.00011	0.00009	0.00005		wave7	-0.0112	-0.0077	-0.0061	-0.0047	-0.0025	-0.0011	
									wave8	-0.0145	-0.0110	-0.0094	-0.0080	-0.0058	-0.0044	-0.0033
									y							
									wave2	-794						
									wave3	-354	440					
									wave4	51	845	405				
									wave5	44	838	398	-7			
									wave6	826	1620	1180	775	782		
									wave7	1797	2591	2152	1746	1753	971	
									wave8	1388	2182	1742	1337	1344	562	-409
									G							
									wave2	-0.0134						
									wave3	-0.0220	-0.0086					
									wave4	-0.0331	-0.0197	-0.0111				
									wave5	-0.0326	-0.0191	-0.0105	0.0005			
									wave6	-0.0283	-0.0148	-0.0062	0.0048	0.0043		
									wave7	-0.0288	-0.0154	-0.0068	0.0043	0.0038	-0.0005	
									wave8	-0.0307	-0.0173	-0.0087	0.0024	0.0019	-0.0024	-0.0019

Note: $CI = C(h_{it}|y_{it})$; $h = H_t$; $y = Y_t$; $G = G(y_{it})$; $CI-CII = C(h_{it}|y_{it}) - C(h_{it}|y_{it})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14); **past_in**: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 11: detailed information for Italy

									PAIRWISE TESTS ON DIFFERENCE							
	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7
CI	0.0053	0.0061	0.0056	0.0058	0.0066	0.0064	0.0064	0.0076	CI	0.00080						
h	0.8791	0.8770	0.8751	0.8729	0.8717	0.8702	0.8675	0.8646	wave2	0.00027	-0.00053					
y	10020	10174	10461	10336	11152	11873	11832	11534	wave3	0.00049	-0.00031	0.00022				
G	0.3331	0.3342	0.3264	0.3093	0.3072	0.3068	0.2933	0.2936	wave4	0.00127	0.00047	0.00100	0.00078			
N	9036	9036	9036	9036	9036	9036	9036	9036	wave5	0.00111	0.00031	0.00084	0.00062	-0.00016		
CI-CI1		0.00080	0.00027	0.00049	0.00127	0.00111	0.00115	0.00228	wave6	0.00115	0.00035	0.00088	0.00066	-0.00012	0.00004	
ineq_in	-0.00006	-0.00019	-0.00025	-0.00043	-0.00050	-0.00068	-0.00068	-0.00068	wave7	0.00228	0.00148	0.00201	0.00179	0.00101	0.00117	0.00113
elas_in	0.00005	0.00013	0.00010	0.00033	0.00050	0.00050	0.00050	0.00043	h							
past_in	0.00002	0.00003	0.00005	0.00006	0.00008	0.00010	0.00012	0.00012	wave2	-0.00207						
ineq_ot	0.00086	0.00042	0.00085	0.00147	0.00120	0.00146	0.00261	0.00261	wave3	-0.00401	-0.00194					
elas_ot	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	wave4	-0.00622	-0.00415	-0.00221				
past_ot	-0.00006	-0.00013	-0.00025	-0.00016	-0.00018	-0.00023	-0.00020	-0.00020	wave5	-0.00738	-0.00531	-0.00337	-0.00116			
ineq	0.00080	0.00023	0.00059	0.00104	0.00071	0.00078	0.00193	0.00193	wave6	-0.00891	-0.00684	-0.00490	-0.00269	-0.00153		
elas	0.00005	0.00013	0.00010	0.00033	0.00051	0.00050	0.00043	0.00043	wave7	-0.01161	-0.00954	-0.00760	-0.00539	-0.00423	-0.00270	
past	-0.00005	-0.00009	-0.00020	-0.00010	-0.00010	-0.00013	-0.00009	-0.00009	wave8	-0.01449	-0.01242	-0.01048	-0.00827	-0.00711	-0.00558	-0.00288
									y							
									wave2	154						
									wave3	442	288					
									wave4	317	162	-125				
									wave5	1132	978	690	815			
									wave6	1854	1699	1412	1537	721		
									wave7	1812	1658	1371	1496	680	-41	
									wave8	1514	1360	1072	1198	382	-339	-298
									G							
									wave2	0.00106						
									wave3	-0.00673	-0.00779					
									wave4	-0.02385	-0.02490	-0.01712				
									wave5	-0.02591	-0.02697	-0.01918	-0.00207			
									wave6	-0.02634	-0.02740	-0.01961	-0.00249	-0.00043		
									wave7	-0.03978	-0.04084	-0.03305	-0.01593	-0.01387	-0.01344	
									wave8	-0.03948	-0.04054	-0.03275	-0.01563	-0.01357	-0.01314	0.00030

Note: CI = $C(h_u|y_u)$; $h = H_t$; $y = Y_t$; $G = G(y_u)$; CI-CI1 = $C(h_u|y_u) - C(h_u|y_{t1})$; ineq_in: equation (2.20); elas_in: equation (2.14);

past_in: equation (2.24); ineq_ot: equation (2.21); elas_ot: equation (2.15); past_ot: equation (2.25); ineq: sum of all ineq_-terms; elas: sum of all elas_-terms; past: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 12: detailed information for the Netherlands

									PAIRWISE TESTS ON DIFFERENCE							
	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7
CI	0.0041	0.0036	0.0040	0.0038	0.0039	0.0045	0.0039	0.0043	CI	-0.0005						
h	0.9056	0.9044	0.9039	0.9040	0.9034	0.9027	0.9015	0.9008	wave2	-0.0002	0.0003					
y	13000	13209	13730	14419	14808	15157	14617	15081	wave3	-0.0003	0.0002	-0.0001				
G	0.2550	0.2819	0.2890	0.2491	0.2466	0.2487	0.2293	0.2399	wave4	-0.0002	0.0003	0.0000	0.0001			
N	4556	4556	4556	4556	4556	4556	4556	4556	wave5	0.0003	0.0009	0.0005	0.0007	0.0006		
CI-CI1		-0.00052	-0.00019	-0.00031	-0.00023	0.00035	-0.00023	0.00016	wave6	-0.0002	0.0003	0.0000	0.0001	0.0000	-0.0006	
ineq_in	-0.00004	0.00004	-0.00028	-0.00029	-0.00030	-0.00041	-0.00033	-0.00033	wave7	0.0002	0.0007	0.0004	0.0005	0.0004	-0.0002	0.0004
elas_in	0.00003	0.00010	0.00018	0.00023	0.00027	0.00021	0.00028	0.00028	h							
past_in	0.00000	0.00001	0.00001	0.00002	0.00002	0.00002	0.00003	0.00003	wave2	-0.0011						
ineq_ot	-0.00043	-0.00020	-0.00003	0.00003	0.00061	0.00016	0.00039	0.00039	wave3	-0.0017	-0.0005					
elas_ot	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	wave4	-0.0016	-0.0004	0.0001				
past_ot	-0.00009	-0.00014	-0.00019	-0.00021	-0.00024	-0.00022	-0.00018	-0.00018	wave5	-0.0022	-0.0010	-0.0005	-0.0006			
ineq	-0.00046	-0.00016	-0.00031	-0.00027	0.00030	-0.00024	0.00006	0.00006	wave6	-0.0029	-0.0017	-0.0012	-0.0013	-0.0007		
elas	0.00003	0.00010	0.00018	0.00023	0.00026	0.00021	0.00026	0.00026	wave7	-0.0041	-0.0029	-0.0024	-0.0025	-0.0019	-0.0012	
past	-0.00009	-0.00013	-0.00018	-0.00020	-0.00022	-0.00019	-0.00015	-0.00015	wave8	-0.0047	-0.0036	-0.0031	-0.0032	-0.0026	-0.0019	-0.0007
									y							
									wave2	209						
									wave3	731	522					
									wave4	1419	1210	689				
									wave5	1808	1600	1078	389			
									wave6	2157	1948	1426	738	348		
									wave7	1618	1409	887	199	-191	-539	
									wave8	2082	1873	1351	663	273	-75	464
									G							
									wave2	0.0270						
									wave3	0.0341	0.0071					
									wave4	-0.0059	-0.0329	-0.0400				
									wave5	-0.0084	-0.0354	-0.0425	-0.0025			
									wave6	-0.0063	-0.0333	-0.0404	-0.0004	0.0021		
									wave7	-0.0256	-0.0526	-0.0597	-0.0197	-0.0173	-0.0194	
									wave8	-0.0150	-0.0420	-0.0491	-0.0091	-0.0067	-0.0088	0.0106

Note: CI = $C(h_u|y_u)$; $h = H_t$; $y = Y_t$; $G = G(y_u)$; CI-CI1 = $C(h_u|y_u) - C(h_u|y_{t1})$; ineq_in: equation (2.20); elas_in: equation (2.14);

past_in: equation (2.24); ineq_ot: equation (2.21); elas_ot: equation (2.15); past_ot: equation (2.25); ineq: sum of all ineq_-terms; elas: sum of all elas_-terms; past: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 13: detailed information for Portugal

	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE
CI	0.0201	0.0197	0.0199	0.0207	0.0215	0.0223	0.0229	0.0237	CI
h	0.8370	0.8345	0.8328	0.8306	0.8291	0.8281	0.8264	0.8245	wave1
y	7183	7210	7554	7773	8242	8654	9088	9527	wave2
G	0.3669	0.3556	0.3502	0.3498	0.3575	0.3506	0.3534	0.3602	wave3
N	7097	7097	7097	7097	7097	7097	7097	7097	wave4
CI-CI1		-0.00045	-0.00022	0.00058	0.00143	0.00220	0.00282	0.00355	wave5
ineq_in		-0.00035	-0.00067	-0.00074	-0.00064	-0.00087	-0.00081	-0.00093	wave6
elas_in		0.00002	0.00030	0.00047	0.00079	0.00104	0.00127	0.00146	wave7
past_in		0.00005	0.00010	0.00015	0.00020	0.00024	0.00030	0.00035	wave8
ineq_ot		-0.00019	-0.00006	0.00068	0.00103	0.00190	0.00203	0.00249	h
elas_ot		0.00000	-0.00001	-0.00002	-0.00003	-0.00004	-0.00006	-0.00007	wave2
past_ot		0.00003	0.00012	0.00004	0.00008	-0.00006	0.00009	0.00026	wave3
ineq		-0.00055	-0.00073	-0.00006	0.00039	0.00102	0.00122	0.00156	wave4
elas		0.00002	0.00029	0.00045	0.00076	0.00099	0.00121	0.00139	wave5
past		0.00008	0.00022	0.00019	0.00028	0.00018	0.00039	0.00060	wave6
									wave7
									wave8
									y
									wave2
									wave3
									wave4
									wave5
									wave6
									wave7
									wave8
									G
									wave2
									wave3
									wave4
									wave5
									wave6
									wave7
									wave8

Note: $CI = C(h_{it}|y_{it})$; $h = H_t$; $y = Y_t$; $G = G(y_{it})$; $CI-CI1 = C(h_{it}|y_{it}) - C(h_{it}|y_{it1})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14);

past_in: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 14: detailed information for Spain

	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE
CI	0.0094	0.0096	0.0094	0.0097	0.0100	0.0111	0.0122	0.0127	CI
h	0.8829	0.8816	0.8803	0.8782	0.8769	0.8772	0.8760	0.8752	wave1
y	8911	9090	9477	9467	9552	10449	10918	11483	wave2
G	0.3380	0.3223	0.3304	0.3346	0.3244	0.3179	0.3151	0.3176	wave3
N	7307	7307	7307	7307	7307	7307	7307	7307	wave4
CI-CI1		0.00013	-0.00006	0.00027	0.00056	0.00167	0.00276	0.00330	wave5
ineq_in		-0.00027	-0.00037	-0.00012	-0.00027	-0.00044	-0.00065	-0.00062	wave6
elas_in		0.00007	0.00022	0.00022	0.00025	0.00054	0.00067	0.00081	wave7
past_in		0.00002	0.00004	0.00006	0.00007	0.00009	0.00011	0.00013	wave8
ineq_ot		0.00035	0.00014	0.00012	0.00050	0.00164	0.00274	0.00314	h
elas_ot		0.00000	0.00000	0.00000	0.00000	0.00000	-0.00001	-0.00001	wave2
past_ot		-0.00004	-0.00009	0.00000	0.00001	-0.00015	-0.00010	-0.00015	wave3
ineq		0.00008	-0.00023	0.00000	0.00023	0.00119	0.00209	0.00252	wave4
elas		0.00007	0.00022	0.00021	0.00025	0.00054	0.00067	0.00080	wave5
past		-0.00002	-0.00005	0.00006	0.00008	-0.00006	0.00000	-0.00002	wave6
									wave7
									wave8
									y
									wave2
									wave3
									wave4
									wave5
									wave6
									wave7
									wave8
									G
									wave2
									wave3
									wave4
									wave5
									wave6
									wave7
									wave8

Note: $CI = C(h_{it}|y_{it})$; $h = H_t$; $y = Y_t$; $G = G(y_{it})$; $CI-CI1 = C(h_{it}|y_{it}) - C(h_{it}|y_{it1})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14);

past_in: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.

Appendix Table 15: detailed information for UK

	wave1	wave2	wave3	wave4	wave5	wave6	wave7	wave8	PAIRWISE TESTS ON DIFFERENCE							
CI	0.0091	0.0096	0.0097	0.0098	0.0106	0.0101	0.0104	0.0111	CI	wave1	wave2	wave3	wave4	wave5	wave6	wave7
h	0.8860	0.8848	0.8851	0.8858	0.8859	0.8848	0.8856	0.8859	wave2	0.0005						
y	13625	13767	14057	14781	15450	15319	16183	16652	wave3	0.0006	0.0001					
G	0.3004	0.3176	0.3011	0.2909	0.3068	0.3088	0.3071	0.3014	wave4	0.0007	0.0002	0.0002				
N	5925	5925	5925	5925	5925	5925	5925	5925	wave5	0.0015	0.0010	0.0010	0.0008			
CI-CI1		0.00050	0.00057	0.00074	0.00155	0.00106	0.00135	0.00199	wave6	0.0011	0.0006	0.0005	0.0003	-0.0005		
ineq_in	-0.00015	-0.00025	-0.00060	-0.00033	-0.00051	-0.00063	-0.00059	-0.00059	wave7	0.0013	0.0008	0.0008	0.0006	-0.0002	0.0003	
elas_in	0.00004	0.00013	0.00033	0.00049	0.00046	0.00065	0.00075	0.00075	wave8	0.0020	0.0015	0.0014	0.0013	0.0004	0.0009	0.0006
past_in	0.00001	0.00002	0.00002	0.00003	0.00004	0.00004	0.00005	0.00005	h							
ineq_ot	0.00056	0.00060	0.00090	0.00126	0.00095	0.00116	0.00168	0.00168	wave2	-0.0012						
elas_ot	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	0.00000	wave3	-0.0009	0.0003					
past_ot	0.00005	0.00008	0.00010	0.00010	0.00012	0.00013	0.00011	0.00011	wave4	-0.0002	0.0009	0.0006				
ineq	0.00041	0.00035	0.00029	0.00093	0.00044	0.00053	0.00109	0.00109	wave5	-0.0001	0.0011	0.0008	0.0001			
elas	0.00004	0.00013	0.00033	0.00049	0.00046	0.00065	0.00074	0.00074	wave6	-0.0012	0.0000	-0.0003	-0.0010	-0.0011		
past	0.00005	0.00010	0.00012	0.00013	0.00016	0.00017	0.00016	0.00016	wave7	-0.0004	0.0007	0.0004	-0.0002	-0.0003	0.0007	
									wave8	-0.0001	0.0010	0.0007	0.0001	0.0000	0.0011	0.0003
									y							
									wave2	142						
									wave3	433	291					
									wave4	1156	1014	723				
									wave5	1825	1683	1392	669			
									wave6	1694	1552	1261	538	-131		
									wave7	2559	2417	2126	1403	734	865	
									wave8	3028	2885	2595	1872	1203	1334	469
									G							
									wave2	0.0172						
									wave3	0.0007	-0.0165					
									wave4	-0.0095	-0.0267	-0.0102				
									wave5	0.0064	-0.0108	0.0057	0.0159			
									wave6	0.0083	-0.0088	0.0077	0.0179	0.0020		
									wave7	0.0067	-0.0105	0.0060	0.0162	0.0003	-0.0017	
									wave8	0.0010	-0.0162	0.0003	0.0105	-0.0054	-0.0074	-0.0057

Note: $CI = C(h_u | y_u)$; $h = H_t$; $y = Y_t$; $G = G(y_u)$; $CI-CI1 = C(h_u | y_u) - C(h_u | y_{u1})$; **ineq_in**: equation (2.20); **elas_in**: equation (2.14); **past_in**: equation (2.24); **ineq_ot**: equation (2.21); **elas_ot**: equation (2.15); **past_ot**: equation (2.25); **ineq**: sum of all ineq_-terms; **elas**: sum of all elas_-terms; **past**: sum of all past_-terms; shaded: statistically significantly different from zero at 5% level.