

# JRC TECHNICAL REPORT

# EUROLAB: A Multidimensional Labour Supply-Demand Model for EU countries

# JRC Working Papers on Taxation and Structural Reforms 15/2021

Edlira Narazani, Ugo Colombino and Bianey Palma

2021



This publication is a Technical report by the Joint Research Centre (JRC), the European Commission's science and knowledge service. It aims to provide evidence-based scientific support to the European policymaking process. The scientific output expressed does not imply a policy position of the European Commission. Neither the European Commission nor any person acting on behalf of the Commission is responsible for the use that might be made of this publication. For information on the methodology and quality underlying the data used in this publication for which the source is neither Eurostat nor other Commission services, users should contact the referenced source. The designations employed and the presentation of material on the maps do not imply the expression of any opinion whatsoever on the part of the European Union concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries.

Contact information Name: Edlira Narazani Address: Joint Research Centre B2 Fiscal Policy Analysis Unit Edificio EXPO 01/A54, C/Inca Garcilaso 3 41092 Sevilla, Spain E-mail: Edlira.NARAZANI@ec.europa.eu Tel.: +34 954.488.219

EU Science Hub https://ec.europa.eu/jrc

JRC127383

Seville: European Commission, 2021

© European Union, 2021



The reuse policy of the European Commission is implemented by the Commission Decision 2011/833/EU of 12 December 2011 on the reuse of Commission documents (OJ L 330, 14.12.2011, p. 39). Except otherwise noted, the reuse of this document is authorised under the Creative Commons Attribution 4.0 International (CC BY 4.0) licence (<u>https://creativecommons.org/licenses/by/4.0/</u>). This means that reuse is allowed provided appropriate credit is given and any changes are indicated. For any use or reproduction of photos or other material that is not owned by the EU, permission must be sought directly from the copyright holders.

All content © European Union, 2021

How to cite this report: Narazani, Edlira, Colombino, Ugo and Bianey Palma, "EUROLAB: A Multidimensional Labour Supply-Demand Model for EU Countries", European Commission, Seville, 2021, JRC127383.

# Contents

#### Table of contents

| Executive Summary   | 1  |
|---|----|
| Acknowledgements  | 3  |
| Abstract  | 4  |
| 1. Introduction   | 5  |
| 2. Behavioural modelling                                    | 8  |
| 2.1 Random Utility Maximization                             | 9  |
| 2.2 Definition of choice set                                |    |
| 2.3 Decision-making unit                                    | 14 |
| 2.4 Working time and hourly wage                            | 14 |
| 3. Microsimulation model and data                           | 17 |
| 4. Labour demand  |    |
| 5. EUROLAB applications                                     | 21 |
| 5.1 An example using a multidimensional version             | 21 |
| 5.2 A EUROLAB application related to sectoral demand shocks | 23 |
| 5.3 Examples using a one-dimensional version of EUROLAB     | 24 |
| 6. Future extensions of EUROLAB                             | 27 |
| 6.1 Indirect taxes  | 27 |
| 6.2 Wealth taxation   | 27 |
| 6.3 Labour Market Adjustment (LMA)                          | 28 |
| Conclusions   | 29 |
| References  |    |
| Appendix A  | 35 |
| Appendix B  | 43 |

#### **Executive Summary**

**Often, fiscal reforms, intentionally or not, can influence** labour market responses. Such behavioural effects cannot be assessed using the EUROMOD model, a static tax-benefit microsimulation model that simulates tax liabilities and individual and family benefit entitlements based on the fiscal rules in force in each EU Member State. Accounting for behavioural effects is important not only to measure changes in employment, unemployment and inactivity margins, but also to assess the revenue implications of a particular tax reform.

**Although EUROMOD does not take into account** behavioural responses in the assessment of policy effects, its capacity to closely replicate existing and counterfactual fiscal reforms, together with the heterogeneity of underlying data, is an appropriate environment for constructing a behavioural microsimulation model. Several studies show that EUROMOD is likely to develop flexible labour supply modelling for country or cross-country analysis. However, there is a necessity for a behavioural model at the EU level which takes full advantage of the unique feature of comparability between EU countries, as envisaged by EUROMOD.

**To fill this gap, we have developed EUROLAB**, a behavioural microsimulation model based on EUROMOD, which can be used to assess the impact of fiscal or economic policies on labour supply and overall employment in EU countries. EUROLAB is based on the modelling of discrete choice labour supply (Aaberge et al., 1995; Van Soest, 1995) based on the Random Utility Maximization approach (McFadden, 1974). Like other behavioural microsimulation models, EUROLAB estimates a set of structural parameters of the utility function and applies them to predict labour supply behaviour. However, unlike other behavioural models, EUROLAB adopts a multidimensional approach in the construction of choice setting as well as providing a labour demand module. Furthermore, what makes EUROLAB unique in the domain of behavioural microsimulation is its characteristics of covering all EU countries in a comparable way as regards the modelling of preferences and opportunities, as well as income concepts.

**EUROLAB is based on a multidimensional choice** set covering not only the alternatives of working hours (one-dimensional choice set) but also other job peculiarities such as employment arrangement (employment versus self-employment status) or occupational sectors (essential versus non-essential sectors). As a result, wage rates are determined differently according to employment arrangements, taking into account the large differential between them. Second, EUROLAB is designed to take into account involuntary unemployment. SILC data, used as the underlying data of EUROMOD, contains information on individuals' job search efforts. In addition, the choice of unemployment is modelled as one of the choices people face when deciding to enter or leave the labour market, and the related benefits are simulated according to the rules applied in a country. Third, the model addresses the underestimation of part-time work by using information on underemployment by adding terms of part-time penalization in the job density function.

**EUROLAB accounts for the labour demand side** and represents a partial labour market equilibrium in line with Colombino's approach (2013). Behavioural models typically show changes in supplied hours of work and participation as a reaction to a tax reform, often referred to in the literature as "second-order effects". In this way, they produce pure variations in the desired number of working hours, disregarding the demand side of the labour market, which, depending on the elasticity, would lead to a different labour supply when market achieves its equilibrium.

**EUROLAB can be used to simulate** the behavioural effects of policy changes related to personal income tax rates or schedules, employee social security contributions, benefit entitlement and amount, tax credits or allowances. The main contributions of EUROLAB are the assessment of: (i) labour supply elasticities, (ii) changes in the labour participation rate and working hours, and (iii) changes in labour supply when labour demand is taken into account.

**Possible extensions of the model may consist of synergies** with EUROMOD's in-home functionalities, such as the Indirect Tax Tool and the WEALTH Taxation Tool, which aim to simulate reforms in the area of indirect taxation and wealth taxes. In addition, a synergy between EUROLAB and the Labour Market Transition tool, a functionality of EUROMOD which allows for changes in the labour market status, can help automate the construction of unemployment alternatives in EUROLAB as well as assess behavioural responses triggered by monetary compensation schemes or similar policy interventions from a medium and long-term perspective.

#### Acknowledgements

The views expressed in the text are the private views of the authors and may not, under any circumstances, be interpreted as stating an official position of the European Commission. We would like to thank Francesco Figari and Salvador Barrios for useful comments on the paper and colleagues from JRC Fiscal Policy Units, Silvia de Poli, Andreas Thiemann and Tine Hufkens for their contribution to the policy simulation in EUROMOD. We also thank Christian Witneben, Matias Dolls and Andreas Peichl for providing a first version of the labour supply module as well as Estefania Vazquez and Wouter Van der Wielen for useful advises on the technical side of the model. Any remaining errors are ours.

#### Authors

Edlira Narazani Ugo Colombino Bianey Palma

# Abstract

This paper describes EUROLAB, a labour supply-demand microsimulation model that relies on EUROMOD, the static microsimulation model for the European Union countries. EUROLAB is built on a multidimensional discrete choice model of labour supply and accounts for involuntary unemployment. The model estimates individual changes in supplied hours of work and participation as a reaction to a hypothetical or real tax-transfer reform, often referred to in the literature as "second-order" effects. Furthermore, the model allows for the demand-side effects of a labour market that, depending on how elastic it is, would lead to a different labour supply when the market reaches its equilibrium. The model is unique in covering 27 countries under the same specification of preferences, opportunity set representation and the same concept of income and working hours. We illustrate the usefulness of the model by showing several examples of EUROLAB, using both the one-dimensional and multidimensional versions. Potential extensions of the model are also discussed in the paper.

JEL Classification: J22, J23, J640, H310.

Keywords: Behavioural Models, Discrete Choice Modelling, Labour supply, Labour market equilibrium.

# 1. Introduction

EUROMOD is a static tax-benefit microsimulation model that simulates tax liabilities and individual and family benefit entitlements on the basis of fiscal rules in force in each EU Member State (Sutherland and Figari, 2013). This model has been widely used to assess the static distributional and budgetary effects of hypothetical or actual tax-transfer reforms, as well as to assess policy changes over time. However, policymakers often design fiscal reforms with the aim of creating incentives to work or formalizing informal work. For example, countries with a low female employment rate, are often recommended to implement fiscal reforms that incentivise women's participation in the labour market. In addition, reforms that are not aimed at influencing work behaviour may also have an impact on the families and individuals concerned.<sup>1</sup> Like most tax microsimulation models, EUROMOD is not behavioural and, as such, does not consider the possible effects of policy reforms on individual labour supply.<sup>2</sup> However, accounting for behavioural effects is important not only to measure changes in employment, unemployment and inactivity margins but also to assess the revenue implications of a particular fiscal reform.

Although EUROMOD does not take into account behavioural responses in the assessment of policy effects, its capacity to closely replicate existing and counterfactual fiscal reforms, together with the heterogeneity of underlying data, is an appropriate environment for constructing a behavioural microsimulation model. Klevmarken (1997) is the first attempt to consider behavioural responses in EUROMOD. Subsequent studies have shown that EUROMOD is likely to develop flexible labour supply modelling for country analysis (Figari et al., 2019, Figari and Narazani, 2019, Coda Moscarola et al., 2019) or for cross-country analysis (Colombino et al, 2010, Colombino and Narazani, 2015, Bargain et al., 2014 and Vandelannoote and Verbist, 2020). However, there is a need for a behavioural model at the EU level which takes full advantage of the unique feature of comparability between EU countries. To fill this gap, we have developed EUROLAB, a behavioural microsimulation model that is based on EUROMOD, which can be used to assess the impact of fiscal or economic policies on labour supply and overall employment in the EU countries.

The aim of this paper is to explain the modelling approach behind EUROLAB as well as to give examples of the model's usefulness. EUROLAB relies on discrete-choice labour supply modelling (Aaberge et al., 1995; Van Soest, 1995), based on the Random Utility Maximization approach (McFadden, 1974). Like other behavioural microsimulation models, EUROLAB estimates a set of structural parameters of utility function and applies them to predict labour supply behaviour. As such, EUROLAB can be compared to country-specific behavioural models such as MICSIM for the Netherlands (Jongen et al., 2014), MITTS and B-TAXBEN for Australia (Creedy et al., 2002), TAXMOD-B for New Zealand (Mercante and Mok, 2014a, 2014b), SWEtaxben for Sweden (Ericson et al., 2009), and IZAΨMOD (Peichl et al., 2010) for Germany. However, unlike other behavioural models, EUROLAB adopts a multidimensional approach in the construction of choice setting as well as providing a labour demand module. Furthermore, what makes EUROLAB unique

<sup>&</sup>lt;sup>1</sup> For example, an across a board increase in property tax is supposed to increase the tax burden of low-income individuals and trigger changes in their labour behaviour. Similarly, a change in mortgage tax relief may alter labour supply behaviour.

<sup>&</sup>lt;sup>2</sup> Fiscal reforms can trigger behavioural effects not only on labour markets but also on consumption, wealth and capital investment, educational choices or childcare services or services for the elderly. All these behavioural effects are not considered in EUROMOD.

in the behavioural microsimulation domain is its characteristics of covering all EU countries in a comparable way as regards the modelling of preferences and opportunities, as well as income concepts.

EUROLAB is based on a multidimensional choice set covering not only the alternatives of working hours (one-dimensional choice set) but also other job peculiarities such as employment arrangement (employment versus self-employment status) or occupational sectors (essential versus non-essential sectors). As a result, wage rates are determined differently according to employment arrangements, taking into account the large differential between them. Second, EUROLAB is designed to take into account involuntary unemployment. SILC data, used as the underlying EUROMOD data, contain information on individuals' job search efforts (in particular, on active job search in the last four weeks and availability for work). Moreover, the choice of unemployment is modelled as one of the choices people face when deciding to enter or exit the labour market. Third, the model addresses the underestimation of part-time work by using information on underemployment by adding terms of part-time penalization in the job density function. Fourth, the model takes into account financial constraints through the inclusion of possible mortgage liabilities in the utility function.

Behavioural models typically show changes in supplied hours of work and participation as a reaction to a fiscal reform, often referred to in the literature as "second-order effects". In this way, they produce pure changes in the desired number of working hours disregarding the demand side of the labour market, which, depending on the elasticity, would lead to a different labour supply when a market reaches its equilibrium. EUROLAB accounts for labour demand side and models a partial labour market equilibrium in line with Colombino's approach. Finally, the paper discusses the technical feasibility of constructing an EU-wide behavioural model. As noted by Klevmarken, 1997 the use of cross/sectional data and in particular the timing of labour market variables and income variables, both recorded in two different years, represent an important challenge when incorporating behavioural responses in EUROMOD.

EUROLAB can be used to simulate the behavioural effects of policy changes related to personal income tax rates or schedules, employee social security contributions, benefit entitlement and amount, tax credits or allowances. The main contributions of the model are assessing: (i) labour supply elasticities, (ii) changes in labour supply, and (iii) changes in employment, inactivity and unemployment when labour demand is taken into account. However, the EUROLAB model can be also used to assess the short-term and long-term effects of job retention policies and, in particular, to provide differential results depending on whether the job retention policy was implemented before or after the new labour market equilibrium was achieved.

The paper is organized as follows. Section 2 provides a description of the microeconometric model and highlights the way in which the choice set, the decision-making unit and working time are defined. Section 3 discusses the microsimulation model and the issues related to the underlying data. Section 4 describes the theoretical modelling, taking into account labour demand side when estimating labour reactions. Section 5 illustrates an application of the multidimensional version of EUROLAB as well as some examples where the model has been used for assessing the behavioural effects of hypothetical or real fiscal reforms.

Future extensions of the model and synergies with in-home EUROMOD synergies are discussed in last section.

# 2. Behavioural modelling

Building a behavioural model requires two crucial elements: (1) a microeconometric model and (2) a taxbenefit microsimulation model. The first element is needed to estimate a set of utility parameters that capture the individual behaviour over income and working hours under socio-demographic constraints. The second element is required to construct a counterfactual disposable income computed at each job alternative following the tax-benefit rules as simulated through microsimulation models. Based on these elements, a behavioural model predicts the supply of working hours and labour market participation for each individual in the data under the existing fiscal rules and under counterfactual fiscal rules.

EUROLAB follows a discrete choice approach that consists in modelling the budget sets from a set of mutually exclusive and collective exhaustive alternatives of hours or jobs. Under the principle of random utility maximization, discrete choice analysis means that an individual chooses the alternative with the highest utility among those available in the market at the time the choice is made. The concept of random utility and in particular the assumption on the distributional form of the random element of the utility function (McFadden, 1974) has converted the discrete choice approach from a theoretical model to a practical one and is thus usable in a variety of disciplines. Since the mid-1990s, labour supply is one of the disciplines benefiting from the practical advantages of discrete choice analysis (Aaberge et al., 1995; Van Soest, 1995).

A standard discrete choice model of labour supply is normally characterized by a one-dimensional choice set consisting of working hours. There might be circumstances, however, that render the decision of work dependent on factors different from working hours. Consequently, the choice set may result in a combination of these factors and in this case, the choice set is considered as multidimensional. For example, individuals, based on their socio-demographic characteristics, may consider the number of working hours as equally important to other job features such as the occupational sector (essential versus non-essential sectors) and contractual arrangements (permanent contract versus temporary or atypical jobs, wage employment versus self-employment, etc.).<sup>3</sup>

EUROLAB is built on a multidimensional choice set where the decision of work is based on the range of working hours, employment status and occupational sectors. Modelling individual preferences concerning employment statuses, in particular wage employment and self-employment, is often left out of the discrete choice model of labour supply. However, the nature of self-employment is changing and in particular solo self-employment is increasing relative to self-employment with dependent employees. In addition, the COVID-19 crisis has caused further changes in the composition of jobs where alternative work arrangements appear to prevail. Recent studies such as Boeri et al., 2020, based on ad-hoc surveys conducted in the US, UK and Italy have documented that "solo self-employment is substantively different from self-employment with employees, being an intermediate status between employment and unemployment, and for some, becoming a new frontier of underemployment". Datta, 2019 indicates that

<sup>&</sup>lt;sup>3</sup> For example, individuals may have a strong preference for working in the public sector. This preference may often be so strong that it can direct them into inactivity and educational training.

the working situation of the self-employed and in particular the solo self-employed (self-employed without employees) is more prone to being precarious and transiting to unemployment than that of employees; in addition, their wages are also lower. Therefore, to take account of these statuses of employment, EUROLAB treats employment and self-employment statuses as two distinct features of work, similarly to Coda-Moscarola et al. (2019). Furthermore, although it is believed that workers would be less likely to make a voluntarily choice concerning the occupational sector based on factors such as education, age or income, we consider the occupational sector as another dimension of the choice set with the intention of using the model for simulating employment effects of sectoral demand shocks like the one triggered by the Covid-19 pandemic (Narazani & Colombino, 2021). Two main occupational sectors are used in the current version of EUROLAB. The first sector covers an aggregate sector group, referred to as 'non-essential', which includes the hospitality, wholesale, transportation and construction sectors, while the latter covers sectors defined as 'essential' such as health, education, food-processing and financial sector.

The consideration of an unemployment alternative as a separate choice from inactivity is another modelling feature of EUROLAB. To that end, we exploit the information of the SILC survey as to whether an individual is actively looking for a job when reporting to not work. This allows us to identify an individual as being in an inactivity status. Those who declare to be looking for a job can be classified as unemployed and can be distinguished in the modelling of working hours from those who are not looking for a job as they spend some hours searching for a job as well as may be receiving unemployment benefits depending on their contributory history. Accounting for an unemployment choice may be important in discrete choice labour supply modelling. In fact, there are studies (Kalb, 2000) where accounting for unemployment-related welfare payments leads to significant disutility associated with unemployment benefits. Similarly, Bargain et al. (2010), find significant differences in labour supply effects when integrating involuntary unemployment in the labour supply model. Other studies (De Boer, 2015), by contrast, find a relatively small effect on behavioural responses when accounting for involuntary unemployment in labour supply modelling.<sup>4</sup> These findings suggest that accounting for involuntary unemployment in labour supply modelling may lead to better predictions of labour supply in countries where unemployment is an issue such as Southern European countries like Italy, Spain and Greece – but not only there. Accounting for involuntary unemployment helps better estimate the labour supply effects of "making work pay" reforms (Bargain et al, 2008), i.e. policies that target low-wage individuals, such as for example minimum wage reforms.

#### 2.1 Random Utility Maximization

Suppose a household *i* with two decision-makers (couple) or one decision-maker (single) is faced with a set of feasible discrete alternatives or "job" types  $\Omega = (H, S, E)$  with wage  $w_{h,s,e}^i$  given the policy regime  $\tau$  (a vector of parameters characterizing the policy regime). A rational household selects the alternative that yields greatest utility. As utility is assumed as a random variable, we can divide it into two additive parts: the systematic component, *V(.)* and a random component  $\varepsilon_{ij}$  or disturbance.

<sup>&</sup>lt;sup>4</sup> As Boer argues, this insignificant effect on behavioural responses is possibly due to the small share of individuals in involuntary unemployment in the Netherlands.

$$U = V(w_{h,s,e}^{i}, H, S, E, X, Z, \tau; \gamma_{i}) + \varepsilon_{i}$$

Where

*H* (hours of work required by the job), *S* (sector of activity) and *E* (employment statuses) are scalars or vectors depending on whether the household is a single person or a couple, and their opportunity density is a function g(H, S, E).

 $Z_i$  and  $\gamma_i$  are a vector of attributes and parameters that characterize the preferences of household *i*.  $Z_{ij}$  might include observed attributes that affect choice *j* like income and leisure.  $X_i$  is a vector of attributes that characterizes the decision maker *i*.

 $\Omega$  = set of all opportunities available to the household (including non-market and market opportunities). Market work includes all time spent working on main and second jobs, not including time spent commuting to or from work. Among the *S* sectors of activity and *E* employment statuses there is also 'Non-participation' (H = 0, S = 0, E = 0) and 'Unemployment', i.e. looking for a (different) job. If unemployed, the household might receive a 'wage', i.e. unemployment benefits or social assistance.

Assuming a Gumbell distribution for the random component we can obtain the probability that household *i* is willing to accept a job (h, s, e) (e.g. Colombino, 2013):

$$P(w_i, h, s, e, \tau; \gamma_i, \delta_i) = \frac{\exp\{V(w_i, h, s, e, \tau; \gamma_i)\}}{\sum_{s} \sum_{E} \sum_{H} \exp\{V(w_i, h, s, e, \tau; \gamma_i)\}}$$

(1)

Probabilistic choice (1) does not predict labour supply preferences very well, e.g. it over-predicts part-time hours. Van Soest (1995) notes that the over-prediction happens because the model does not take into account the low demand for part-time jobs. In general, the model, as it stands, ignores the density or demand of certain types of jobs. To correct for this biased in prediction, Van Soest (1995) adopts a simple approach by including alternative specific dummies for different types of jobs such as part-time and full-time. Aaberge et al. (1995, 1999) also adopt a similar refinement approach but ground it on a specific structural interpretation of the alternative-specific dummies and their respective coefficients. The idea behind this is that households can face different availability of job types because some labour markets are able to provide more jobs with market hours in a certain interval. We adopt a similar refinement approach and obtain the probability that household *i* is willing to accept a job (*h*, *s*, *e*):

$$P(w_i, h, s, e, \tau; \gamma_i, \delta_i) = \frac{\exp\{V(w_i, h, s, e, \tau; \gamma_i) + D_i(h, s, e)'\delta_i\}}{\sum_S \sum_E \sum_H \exp\{V(w_i, h, s, e, \tau; \gamma_i) + D_i(h, s, e)'\delta_i\}}$$

(2)

As an example, for a single household and an opportunity set containing jobs in two sectors and with three ranges of hours, the vector  $D_i$  might be defined as follows (with 1[.] denoting the indicator function):

$$D_{1,0} = 1[s = 1, h > 5]$$

$$D_{1,1} = 1[s = 1, 16 \le h < 30]$$

$$D_{1,2} = 1[s = 1, 31 \le h < 45]$$

$$D_{1,0} = 1[s = 1, h > 5]$$

$$D_{1,1} = 1[s = 1, 16 \le h < 32]$$

$$D_{1,2} = 1[s = 1, 33 \le h < 42]$$

$$D_{Un} = 1[1 \le h \le 5]$$
(3)

For couples,  $D_{s,h}$  contains two analogous sets of variables, one for each partner. The hour ranges  $[16 \le h < 32]$  and  $[31 \le h < 42]$  correspond to part-time and full-time jobs, respectively, in sector s.  $D_{Un}$  corresponds to the unemployment alternative, where hours in the range  $[1 \le h \le 5]$  denote hours of job search.

However, in some countries, the take up of part-time work may often be a sign of underemployment rather than desired alternative. In other words, a part-time specific dummy would only capture the supply for part-time jobs and not the demand. To correct for this, we include a dummy  $-(D_v)$  — on desired part-time pattern among the job opportunities dummies. This dummy is built based on the information in the SILC data as to whether workers supply less than 30 hours in a usual week because they cannot work more hours.<sup>5</sup> The choice probability, based on this new specification of dummies refinement becomes:

$$P(w_i, h, s, e, \tau; \gamma_i, \delta_i) = \frac{\exp\{V(w_i, h, s, e, \tau; \gamma_i) + D_v(h)'\delta_v + D_i(h, s, e)'\delta_i\}}{\sum_S \sum_E \sum_H \exp\{V(w_i, h, s, e, \tau; \gamma_i) + D_v(h)'\delta_v + D_i(h, s, e)'\delta_i\}}$$

(4)

<sup>&</sup>lt;sup>5</sup> According to the EU-SILC Methodological Guidelines and Target Variables, variable PL120 of EU-SILC refers to the main reason for working less than 30 hours. However, this EU-SILC variable does not capture a fully-harmonized definition of underemployment because this concept might differ by country.

The most-used specifications of the systematic component of utility in the discrete choice modelling are quadratic specification (Van Soest, 1995, Blundell et al., 2000, Colombino et al. 2010), translog (Van Soest) and Boc-Cox (Aaberge et al., 1999, Aaberge and Colombino, 2009, Blundell and Shephard, 2011, Figari and Narazani, 2019). The specification chosen as the default in EUROLAB is a quadratic specification in income and leisure. It is chosen not only for the sake of simplicity and speed in estimation, but also because it does not impose any quasi-concavity restriction to the utility. In EUROLAB, this specification can be easily exchanged with a translog specification.

$$V(j, T_{ij}) = \alpha_C C + \alpha_{CC} C^2 + \alpha_F (T - h_F) + \alpha_{FF} (T - h_F)^2 + \alpha_M (T - h_{MM}) + \alpha_{MM} (T - h_M)^2 + \alpha_{FM} (T - h_F) (T - h_M)$$

The preference parameters assigned to linear terms such as income and leisure are allowed to differ by some individual and household characteristics such as age, age squared, number of children 0-3 years, number of children 3-6 years, number of children and household size in the following way. Additionally, we interact leisure with two dummy variables indicating respectively whether the decision-making unit is a migrant in order to account for labour market integration constraints or holds a mortgage liability to control for other economic constraints like financial ones.

$$\begin{split} \alpha_{C} &= \beta_{C}HH_{size} \\ \alpha_{M} &= \beta_{M1}Children_{03y} + \beta_{M2}Children_{36y} + \beta_{M3}Children_{Total} + \beta_{M4}Children_{36y} \\ &+ \beta_{M5}Age + \beta_{M5}Age^{2} + \beta_{M6}Migrant + \beta_{M7}Mortgage \\ \alpha_{F} &= \beta_{F1}Children_{03y} + \beta_{F2}Children_{36y} + \beta_{F3}Children_{Total} + \beta_{F4}Children_{36y} + \beta_{F5}Age \\ &+ \beta_{F5}Age^{2} + \beta_{F6}Migrant + \beta_{F7}Mortgage \end{split}$$

#### 2.2 Definition of choice set

Let us consider two choice sets *H* and *E* with *h* and *e* elements, respectively. The Cartesian product  $H \times E$  will contain  $h \times e$  elements. For example, a choice set based on two dimensions – four hours of work and three employment statuses – will have a dimension of  $4 \times 3$ . Let further define  $D_n^*$  the set of all elements of the Cartesian product *D* that are not feasible for all individuals. Then, a multidimensional choice set for individual *n* is equal to  $D_n = H \times E - D_n^*$ .

Let us now use these definitions to illustrate the choice set used in the multidimensional choice labour supply model. As an example, for a single household and a choice set composed of three ranges of positive working hours ([15-30], [31-45] and [46-60]), two occupational sectors (s1 and s2) and two employment

statuses (e1 and e2), the choice set will be composed of  $3 \times 2 \times 2$  elements. In addition to these combinations of choices, individuals also face two distinct choices of zero working hours: one of being "inactive" and another one being "unemployed but active". While in the former choice, the individual does not receive any income, in the latter individuals might receive some income in the form of unemployment benefits. Obviously, depending on the working history and social contributions, not all individuals can be eligible for receiving this unemployment income. In total, there are 14 ( $3 \times 2 \times 2 + 2$ ) for singles and 196 alternatives for couples.

Unemployment (as different from non-participation) can also be interpreted as choosing a "job" (Colombino et al. 2010), which pays a "wage" (unemployment benefits or some social security support) and possibly requires some "hours" for job search and or confirmation of willingness to work (e.g. participation in re-training or motivational activities, etc.). The number of available unemployment "jobs" and the level of their "wages" can be explicitly represented in the model in the same way as we do with market jobs and market wages (i.e. through the "dummy" terms). However, they are determined by policy decisions. It is also possible to account for the "mixed" alternative where households hold a market job (possibly short-term) and are simultaneously unemployed (i.e. looking for a different job).

To make it clear, the multidimensional choice set is composed of alternatives h\_s\_e (where h stands for working hours, s stands for sectors and e for employment statuses) modelled in the way that:

When index h takes value 0 and from interval [1-5], the indexes s and e take value 0

When index h takes value from intervals [15-30], [31-45] and [46-60] the index s takes value 1, 2 and the index e takes value 1 and 2

Hours equal to 0 correspond to an inactivity alternative while positive hours lower than 5 corresponds to an unemployment alternative and stand for searching job hours. In case of an unemployment alternative, EUROMOD unemployment policies are switched on to simulate unemployment benefits depending on the rules applied in one country. For example, to simulate unemployment benefits using Italian input data in EUROMOD, we assume that the individual under this alternative is unemployed, with at least 6 months in employment in the previous year, and eligible to receive unemployment benefits for the whole year. To calculate the amount of the benefit, EUROMOD needs a monthly wage earned in the previous year. In case this wage was reported in the data, we use this information to simulate unemployment benefits for the sample of employed individuals. In case no wage is reported from the previous year, we predict a monthly wage using the predicted wage rate for wage employment. Furthermore, we assume that unemployment benefits are received over 12 months. Importantly, we put to zero positive earnings in case of employed or self-employed individuals.

Furthermore, EUROLAB allows the user to sample the counterfactual hours at each hour alternative using three different options: (1) a fixed number of hours, (2) randomly sampled with fixed intervals, and (3) sampled from observed distribution of the working hours. The use of the last option, as shown by Aaberge et al. (2009), can help to improve the model prediction as the distribution of the potential alternatives is

modelled such as to respect the proportion of women observed to work a specific number of hours within each interval.

### 2.3 Decision-making unit

The decision-making unit is the head of unit with and without a partner. In the first case, the decisionmaking unit is made of one person, while in the latter case it is made of two persons who take collective decisions about their participation in the labour market. A household head is defined as the member with highest labour earnings. If the ranking of earnings does not help to qualify a member as household head, the criterion of age is considered. Once the decision-making unit is identified within a household, other criteria are used to build the sample of individuals whose labour supply behaviour is considered as endogenous.

We also do not take into account complex intra-household bargaining processes and consequently do not model the behaviour of other persons who are reported as part of the household. In other words, we treat as exogenous their labour supply behaviour. Labour behaviour of several categories such as students, retired and people with disability are not modelled as their choice set should be expanded on other decision dimensions, e.g. education, pension or early retirement schemes. Age is another selection criteria, by default we model the behaviour of individuals aged 20-64 but selection by age is one of the parameters than can be easily changed in the model.

Based on the decision-making unit, the conditional logit model is estimated separately by household type. In EUROLAB we distinguish between the following household types: (1) Couples, (2) Single women, and (3) Single men. Distinguishing for further types of households is very desirable either for better identification of the sample as well as for performing specific policy analysis. However, the representativeness of additional types of households is not good in SILC data and to overcome this drawback administrative data or merged data over years should be used. Furthermore, accounting for other household types such as adult children living with parents is very desirable for assessing the labour supply behaviour of young adults especially in Southern European countries where parents-adult child co-living patterns are very common. But this would require a complication of the interactions within households moving from a two-agent household type to a three-agent type or higher ranks of interfamily relationships.

#### 2.4 Working time and hourly wage

Most of studies on labour supply modelling employ weekly data and ignore the weeks-per-year dimension. Under this treatment, persons who do not work during a given survey week are defined as not working. However, this approach is incomplete because the weekly hours unit represents just a snapshot of the labour market status at the time of interview. Labour supply is treated in EUROLAB as a combination of two time dimensions or commodities: (1) hours of work (usually worked per week) and (2) annual months of part-time and full-time work. EU-SILC data availability imposes some constraints on this treatment of labour supply unit. EU-SILC records the information on working hours usually worked per week at the time the survey-data collection, or most of the collection, is carried out. The year that corresponds to this period is defined in this survey as "survey year". On the other hand, information on months of work (and income) is recorded on income reference period which is a 12-month period that can be fixed (such as the previous calendar or tax year) or moving (such as the 12 months preceding the interview).<sup>6</sup>

The time lag between survey year and income reference period may affect the match between annual labour patterns and income (which refer to the past employment situation) and weekly working hours (which refer to the present situation). The mismatch can lead to missing information both for individuals who worked in the survey year but not in the income reference period and those who worked in the income reference period but not in the survey year. While the former individuals report positive income and working months but zero working hours, the latter would report positive hours but no income and working months. Therefore, the combination of these time commodities may lead to a missing observation of total annual hours.<sup>7</sup> This is particularly important for individuals in unstable employment, indicating that missing information on working hours at the survey year is not random. The non-random lack of information points to a selection bias, Hanoch (1980). As Hanoch argues, individuals intending to work *K* weeks in the survey year would have on average a probability equal to K/52 to work in that year. Since the selection probability is correlated to the endogenous variable K, labour supply estimates are biased based on these data. To correct somehow for this non-random missing information, instead of tagging individuals with missing weekly hours as not working, we allocate them weekly hours of work using information on gender-specific median hours of part-time and full-time workers, following Brandolini et al. (2010). An advantage of this method is that the two variables on months in employment refer to the income year, hence matching with the information on earnings. However, this method includes a homogenous imputation of working hours and does not account for the potential non-randomness of missing hours of work at the survey year.

The computation of income available on any particular job (*h*, *s*, *e*) requires information on the wage rate of that type of job. As this information is available only for the chosen job, we have to estimate the wage rate for the other types of jobs. The wage equation is specified as a logarithmic function of observed wage rates and depends linearly on a set of conventional explanatory variables such as education, work experience, work experience squared, education level and some regional dummies. To estimate the wage equation over a non-random selected sample, Heckman (1979) provides a standard technique among economists. However, when a selection is made over a number of exclusive choices, multinomial logit or probit specification is normally entrenched into a multiple selection bias correction model.<sup>8</sup> Three different methods can be used in EUROLAB to control for the possibility of non-random selection. The first method follows the Dagsvik and Strom, 2004 approach. Assuming a correlation between the random variables in

<sup>&</sup>lt;sup>6</sup> For example, the survey year for the EU-SILC 2018 wave is 2018, while the income reference period is 2017.

<sup>&</sup>lt;sup>7</sup> Furthermore, the time lag may cause a loss in data quality, especially when the date of interview is far removed from the reference income period, especially for individuals in unstable employment.

<sup>&</sup>lt;sup>8</sup> We consider the wage rates as exogenous, although they can be treated as endogenous by estimating wage functions simultaneously with the utility function parameters and the job opportunity density, all embedded in the maximum likelihood function.

the wage and selection equation to a common latent ability factor, they show how the parameters of the wage equation for sector j can be estimated consistently and asymptotically efficient by OLS on the subsample of women/men that work in sector j by means of the regression equation where selectivity bias is controlled by including logPj as an additional explanatory variable in the wage equation. Pj is the probability of being in sector j, j = 0, 1, 2 (where j = 0 means not working) and is calculated running a multinomial logit model where the dependent variable is the type of sector and employment status.<sup>9</sup> The second method follows the Dubin and McFadden (1984) approach, which is based on two assumptions: (1) a linear relationship between the error terms in the wage selection equations, and (2) a correlation equation between the two error terms sum to zero. The third method follows a modified version of Dubin and McFadden), where the assumption of zero sum of correlation terms is relaxed as suggested by Bourguignon, Fournier and Gurdand (2007).<sup>10</sup> In EUROLAB we can choose between actual wages or predicted wages for the working sample and for the chosen alternative.<sup>11</sup>

<sup>&</sup>lt;sup>9</sup> The details of this approach are given in appendix D of Dagsvik and Strøm (2004).

<sup>&</sup>lt;sup>10</sup> The first method is implemented estimating sequentially the selection equation and the wage equation in STATA. The other two methods are implemented using the selmlog command written by Bourguignon, Fournier and Gurdand (2007).

<sup>&</sup>lt;sup>11</sup> The estimates of the wage equation are available upon request from the author.

### 3. Microsimulation model and data

The model used to construct the counterfactual budget constraint at each alternative of the choice set – EUROMOD – is a tax-benefit microsimulation model that simulates cash benefit entitlements, direct taxes and social insurance contributions and consequently disposable incomes for all EU countries. The model does these simulations based on the information available in the underlying micro-datasets and in line with the country tax-benefit rules.<sup>12</sup> Non-simulated benefits (mainly contributory pensions, due to data constraints), as well as market incomes, are taken directly from the input datasets. For further information on EUROMOD, see Sutherland and Figari (2013).

The model computes the net household income as follows. For each alternative in the choice set characterized by (i) positive working hours, (ii) employment status, and (iii) sectoral choice, a monthly wage is attributed to each individual in the sample eligible for labour supply modelling. To compute the monthly wage, we predict an hourly wage rate for each sectoral choice and employment status using the prediction procedures described in Section 3.4. In the case of the unemployment alternative, unemployment benefits are simulated according to the rules applied in the country. For example, to simulate unemployment benefits using the Italian spine of EUROMOD, an individual would be entitled to such benefits if he or she was in employment for at least 6 months in the previous year. Moreover, in order to calculate the amount of the benefit, EUROMOD requires information on the monthly wage earned in the previous year. If the wage is reported in the data, we use this information to simulate unemployment benefits. Otherwise, a predicted monthly wage is used. In addition, unemployment benefits are assumed to be received over a period of 12 months. In the case of the inactivity alternative, a monthly wage equal to 0 is allocated to the selected individuals. EUROMOD considers this predicted wage of the individual, along with any other source of family income, in deriving the net disposable income of the individual and the household under each alternative, taking into account the whole tax-benefit system and the household characteristics.

EUROLAB runs on the underlying data of EUROMOD that come from European Statistics on Income and Living Conditions (EU-SILC) surveys. EU-SILC surveys are representative samples of EU populations and collect comparable detailed information on socio-demographic characteristics and income from different sources at the individual and household levels.

<sup>&</sup>lt;sup>12</sup> Institute for Social and Economic Research, University of Essex and Joint Research Centre, European Commission, EUROMOD: Version I3.0+ [software], January 2021. The EUROMOD model is maintained and updated by the European Commission Joint Research Centre, for further information see <u>https://euromod-web.jrc.ec.europa.eu/</u> and Sutherland and Figari (2013).

#### 4. Labour demand

While non-behavioural microsimulation models are intended to generate the "first-round" effects of a fiscal reform assuming that individuals do not change their supply of labour, behavioural models are intended to generate the so-called "second-round" effects assuming that individuals may change their behaviour. These effects are often expressed in terms of changes in hours of work and participation and can be triggered by intentional or unintentional reactions of individuals or households to fiscal reforms. However, the "second-round" effects represent pure changes in the desired number of working hours or activity/inactivity status disregarding the demand side of the labour market that, depending on how elastic it is, may lead to a different labour supply and wage rate under a labour market equilibrium.

EUROLAB brings the behavioural modelling one step further by accounting for the demand side of the labour market through a representation of a partial labour market equilibrium in line with Colombino's (2013) approach. To improve the predictability of the model, in particular for part-time work, a set of dummy variables (accounting for part-time and full-time jobs) are now commonly used to calculate the probability of choices. This "dummies refinement" approach has two main interpretations in the literature of discrete choice modelling. The first interpretation of these dummies assumes that they reflect a number of factors such as fixed and search costs, commuting costs, etc.<sup>13</sup> The other interpretation considers these dummies as reflecting availability or density of job types that are not represented by the systematic part of the utility function. Colombino (2013) uses the latter interpretation to link dummies' coefficients to the number of jobs available on the market and to develop a structural model that takes into account labour market equilibrium conditions. The first attempt to model a supply-demand equilibrium using a discrete choice labour supply model can be attributed to Creedy and Duncan (2005)<sup>14</sup> who, through a multi-stage procedure, simulated the labour supply effects of a policy change and aggregated them to construct the demand side of the labour market. Their approach involves the concept of a "supply response schedule". Labour market equilibrium is achieved through wage adjustment. However, such a procedure, as argued by Colombino (2013), would not be consistent with a matching model that simultaneously includes a representation of both labour supply and demand.

The method proposed by Colombino (2013) and recently revised in Colombino and Narazani (2021) expresses the dummies' coefficients or parameters  $\delta_i$  of expression (2) as logarithmic functions of the number of jobs available in the market. Let  $\delta_{s,e,0} = \ln(A_{s,e,0} J_{s,e})$ , with s = 1, 2, 3, e = 1, 2 where  $J_{s,e}$  denote number of jobs in sector *s* and employment status *e* and  $A_{s,e,0}$ , is constant. If we consider the simple case with only one dummy (i.e. one type of employment) and with households which have identical  $\gamma$  and  $\delta$ , we can write  $\delta = \ln J + a$  where  $a = \ln A$ . Let us further assume a current tax-benefit policy  $\tau$  and that the data represent a labour market equilibrium status or in very simple terms that the total of people working have chosen jobs available in the market. The introduction of a new policy  $\tau'$  (e.g. an increase in marginal tax rates) will change the probability (following ex 4, page 13) of taking up a job that must be available to allow for a new match. This entails an equilibrium conditionality between available

<sup>&</sup>lt;sup>13</sup> See the survey by Blundell et al. (2007) for an example of modelling accounting for fixed costs of working.

<sup>&</sup>lt;sup>14</sup> Peichl and Siegloch (2012) adopt Creedy and Duncan (2005) approach.

jobs and desired labour supply, taking into account that the desired labour supply is affected by the available jobs through the term  $\delta$ .

The policy reform will induce a change in the desired labour supply. An equilibrium condition requires that the number of available jobs *J* is equal to the desired labour supply. *J* depends on the wage rates and  $\delta$  depends on *J*. Therefore, the policy will determine a change in the values of *J*,  $\delta$ , of the wage rate and labour supply.

Let  $e^{v}$  be the proportional change in *J*, and  $\delta(v)$  the changed value of  $\delta$ :

$$\delta(v) = \ln(Je^v) + A = \ln J + a + v = \delta + v \tag{5}$$

We can also write the changed value of *J* as J(v):

$$J(v) = Je^{v} \tag{6}$$

By assuming  $J = Kw^{-\eta}$  or  $w = K^{1/\eta}J^{-1/\eta}$  we get the wage rate corresponding to  $Je^{\nu}$ :

$$w = K^{1/\eta} (Je^{\nu})^{-1/\eta} = K^{1/\eta} J^{-1/\eta} e^{-\nu/\eta} = w e^{-\nu/\eta}$$
(7)

where  $\eta$  is the elasticity of labour demand, assumed equal to 0.5 in the current version of EUROLAB.<sup>15</sup> The new values of  $\delta(v)$  and w(v) determine new choice probabilities.

Let  $\sum_{i} \sum_{h,s,e>0} P(w_i(v), h, s, e, \tau'; \gamma, \delta(v))$  be the desired labour supply given the policy  $\tau'$  and the adjustment v. Then the equilibrium value  $v^*$  is such that

$$\sum_{i} \sum_{h,s,e>0} P(w_{i}(v^{*}), h, s, e, \tau'; \gamma, \delta(v^{*})) = J(v^{*})$$
(8)

The left-hand side represents the total desired labour supply in terms of number of jobs that households are willing to accept. The right-hand side represents the available jobs, or labour demand. The equality (equilibrium) determines the effective labour supply. Note that the adjustment in the number of jobs through a change in the level of the wage rates is a movement along the labour demand curve.

<sup>&</sup>lt;sup>15</sup> The next version of EUROLAB will allow the user to change the parameter of labour demand elasticity in the EUROLAB interface.

A similar logic applies if we also consider a case with a dummy representing the level of employment and a dummy representing the density of unemployment opportunities.

Let  $\delta_{u,g}$  be the coefficient assigned to the unemployment dummy.

We have  $\delta_{u,g} = ln(U_g) + \beta_g$  where  $U_g$  = number of available unemployment slots (or hours) and  $\beta_g$  is a constant. The value of the dummy coefficient after the shock will be:

$$\delta_{u,g}(u) = \ln(U_g e^u) + \beta_g = \ln U_g + u = \delta_{u,g} + u$$
(9)

The adjusted total number of unemployment slots (or hours) will be  $U_n(u) = \sum_g U_g e^u$ .

If M(u) = the number of individuals choosing unemployment (or the number of hours spent as unemployed), then the equilibrium value v<sup>\*</sup> is such that

$$M(u^*) = U(u^*)$$
 (10)

Moreover, in principle we might allow for a different equilibrium adjustment  $v_{g,s,h}$  for each specific job type. The equilibrium constraint might also identify specific conditions depending on the sector of activity and/or range of hours. The extent to which we can approximate the above general framework depends on data availability and computational constraints.

#### 5. EUROLAB applications

This section illustrates the usefulness of the multidimensional and one-dimensional versions of the EUROLAB model through a set of examples where the model is used or can be used to simulate the behavioural effects of hypothetical or real tax reforms. The first sub-section illustrates the application of the multidimensional version of EUROLAB to simulate the behavioural effects of hypothetical tax reforms, while the second part examines its usefulness to take into account the sectoral demand shock when analysing the behavioural effects of job retention schemes. The last sub-section looks at concrete examples of cases where the one-dimensional version of EUROLAB has been used by the Joint Research Centre (JRC) to conduct behavioural analyses of various tax-benefit reforms that have been part of the policy debate or have been implemented across the EU countries.

#### 5.1 An example using a multidimensional version

EUROLAB can mainly be used to assess (i) labour supply elasticities, (ii) changes in the labour participation rate and working hours, and (iii) changes in labour supply when labour demand is taken into account. To get a taste of the model, this section presents the results that the model can produce in calculating the elasticity of labour supply and in simulating the behavioural effects of the tax-benefit reform. Subsequently, two hypothetical reforms are simulated to change the income tax bracket rates of the Italian tax system running the Italian module of EUROMOD with the underlying data from the Italian SILC 2018.<sup>16</sup> Specifically, with the first reform (PIT1), we increase income tax rates in the first and second income tax brackets from 0.23 and 0.27 to 0.30 and to 0.32 respectively, and decrease these rates in the third, fourth and last tax brackets from 0.38, 0.41 and 0.43 to 0.34, 0.36 and 0.38. With the second reform (PIT2), we lower income tax rates across the entire income distribution from 0.23, 0.27, 0.38, 0.41 and 0.43 to 0.1, 0.16, 0.22, 0.28 and 0.34 respectively.<sup>17</sup>

First, before calculating the elasticity of labour supply and the behavioural effects triggered by a reform, the EUROLAB model estimates a number of parameters of preference or utility (ex. 5) and a number of job density coefficients (ex. 4) described in subsection 2.1. For the purpose of this example, a version of EUROLAB is run which, in the case of a single decision-making unit, is based on a multidimensional choice set consisting of three ranges of positive working hours ([15-30], [31-45] and [46-60]) and two employment statuses (e1 and e2) more alternatives of inactivity and unemployment. The final choice set established in this case consists of 8 alternatives for singles and 64 for couples. In addition, we use the wages observed for the working sample and the prediction method Dubin and Mcfadden as explained in subsection 3.2. Table 1, Appendix A shows the estimated behavioural parameters and job density coefficients as well as several statistical test on the goodness of fit of the conditional logit model. In addition, in order to assess the ability of EUROLAB to replicate the choices observed, we compute predicted

<sup>&</sup>lt;sup>16</sup> The survey is representative for the national population at the regional level, and is the national component of the EU-SILC carried out annually to collect comparable information on income, poverty, social exclusion and living conditions across EU countries.

<sup>&</sup>lt;sup>17</sup> SILC 2018 data and policy year 2020 have been used for simulating the reforms in EUROMOD.

probabilities for each alternative of labour supply and compare them with the proportions observed in Table 2.A-2.D, Appendix A.

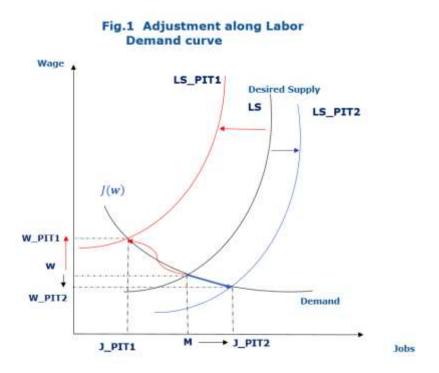
After having been estimated, utility parameters and job density coefficients are kept unchanged and used to compute the average values of labour supply elasticities and behavioural effects in the simulation of a new reform. Labour supply elasticities are calculated at the individual level and broken down by a range of socio-demographic characteristics such as gender, marital status, education attainment, age and presence of children. Three main results emerge from a summary of the average labour supply elasticity values calculated on the Italian sample (Table 1, Appendix B). First, women have higher elasticity of labour supply than men. Second, low-educated people and young people have higher elasticity values than highly educated and older people respectively. If the level of education was used as a proxy for income, this would indicate a higher value of labour supply elasticity for people on low incomes. Thirdly, women with partners have a lower elasticity of labour supply than single women, and single mothers stand out as the most elastic group.

Subsequently, EUROLAB computes the average changes in labour supply triggered by the simulated reforms and Table 1 (Appendix B) gives a summary of labour supply effects for eight demographic groups: (1) men in couple without children, (2) men in couple with children, (3) single men without children, (4) single men with children, and (8) single women without children. For these groups, Table A1 shows the average changes in weekly working hours and participation rates triggered by reforms. Men have the highest participation rates and number of hours worked in the baseline scenario, and this fact should be considered when commenting on percentage changes in reform scenarios. The gender gap in average working hours ranges from 1.5 to 11 hours for demographic groups of single individuals with children and in couple without children respectively, suggesting a similar pattern of work for the first group and a very divergent pattern for the second one. After the reform of PIT1, all demographic groups are estimated to decrease their labour supply, but women stand out as the group with the highest decrease in working hours. On the contrary, the reform of PIT2 leads to an increase in labour supply in all demographic groups, but again women are more responsive to tax-rate reductions. However, both reform simulations indicate that single parents react more than other groups, and this conclusion is robust in the relevant literature.

In Addition, Table 2 (Appendix B) shows the effects of labour supply broken down by income quintile. Men provide a significantly higher number of hours on the labour market and have higher participation rates than women. The gap is higher for the lowest income quintile, narrowed when it moves to higher quintiles and closes to the last income quintile. Moreover, the highest responses under both reforms come from the first part of the income distribution, while changes in labour supply are rather modest.

Another set of outcomes that the EUROLAB model can deliver is linked to changes in employment, inactivity and unemployment when taking into account the demand side of the labour market. Assuming an elasticity of labour demand of -0.5, the equilibrium model (Section 4) applies running an optimisation procedure to

search the value of the change in average wage (parameter v, Section 4) that corresponds to a new labour market equilibrium status under the reform. The optimization procedure is explained in details in Narazani and Colombino, 2021. The results are shown in Table 3 (Appendix B).<sup>18</sup> If equilibrium conditions are not taken into account, the PIT1 reform should shift the desired labour supply curve (Fig.1 LS\_PIT1) to the left and reduce total employment by 0.8%. A new market equilibrium condition that requires consistency between the number of jobs available and the desired labour supply is achieved through a movement along the demand curve and an adjustment of the wage rate. Indeed, the equilibrium condition leads to a 3.5% increase in wages and a 0.56% decrease in employment, mainly offset by a 6% increase in unemployment. The increase in unemployment is mainly due to the generosity of the unemployment benefit amounts which, unlike wage rates, were assumed unchanged in this exercise. It is important to stress that in the equilibrium model, only an employment equilibrium constraint is imposed for ease of computation and the imposition of an equilibrium constraint on unemployment can lead to different results. On the contrary, the reform PIT2 is expected to increase labour supply by 1.44% and shift the curve of labour supply (Fig. 1, LS\_PIT2) to the right. A new labour market equilibrium is achieved, with the wage rate falling by 1% and a lower increase in employment (1.39%).



#### 5.2 A EUROLAB application related to sectoral demand shocks

The multidimensional version of EUROLAB has been developed during the COVID-19 pandemic period, 2020-2021. Restrictive measures taken by EU countries to mitigate the spread of the virus led to the cessation of economic activity and severe job losses, in particular in the second quarter of 2020. The

<sup>&</sup>lt;sup>18</sup> We use the Amoeba, an optimisation routine written in STATA. Amoeba is an efficient (derivative-free) algorithm for optimising a multidimensional function developed by Nelder and Mead. Convergence rules are needed to break the iteration cycle. We set a step size (percentage change in each parameter used to set up Amoeba in the parameter space) equal to 0.1 and a tolerance (the tightness of Amoeba before the algorithm quits) equal to E-06.

structure of labour market has undergone a dramatic change, characterized by a decline in employment and in the total number of hours worked, which has not been felt in the same way in all the sectors of activity. Some sectors of activity, such as non-essential services and production, have been more exposed to lockdown restrictions than others and have therefore experienced a greater reduction in working hours and employment. However, due to the delay in obtaining sufficient data, it is difficult to make a precise assessment of the extent of job losses and forecast of working hours for subsequent months or longer periods. In order to overcome the limitation of real-time data on employment, in particular at sectoral level, as well as the limitation of existing studies that consider exogenous transitions from work to unemployment or inactivity (or vice versa), we have built the multidimensional version of EUROLAB that takes into account differences between occupational sectors and employment statuses and allows for transitions to and from unemployment and inactivity status. Being capable to model the behavioural reactions of individuals within the labour market, depending on their occupational sector or employment status, the multidimensional version can be used to consider behavioural reactions to external sectoral demand shocks in employment.

Narazani and Colombino, 2021, show how the multidimensional choice set of EUROLAB can be used to consider endogenous reactions to labour demand shocks as well as to assess the effectiveness of policy reforms, such as job retention schemes, in absorbing the impact of these shocks on employment under alternative hypothetical scenarios regarding the timing of reforms. In addition they show theoretically how the EUROLAB model can be used to assess the short-term and long-term effects of job retention policies and, in particular, to deliver differentiated results depending on whether the retention policy was implemented before or after the new labour market equilibrium was achieved. To illustrate the theoretical model, they perform a simulation of the effects of a hypothetical wage subsidy using the Italian module of EUROMOD as well as the Italian version of EU-SILC (SILC, 2018) and show in particular how sectoral demand shocks lead to lower wages and employment and an increase in unemployment under equilibrium conditions. In order to assess the potential impact of a demand shock in the presence of a simplified wage subsidy scheme, they simulate an *ex ante* and *ex post* scenario that differ from each other in the timing of when the subsidy is allocated to potential beneficiaries – before or after labour market equilibrium is achieved. In the *ex post* scenario, the model predicts an increase in employment compared to the equilibrium situation in the absence of the wage subsidy, mainly due to the consideration of the recipients as employed. The ex ante scenario predicts a smaller increase in employment which would correspond to a long-term labour market equilibrium. These results imply that the timing of the introduction (or announcement) of the wage subsidy matters when determining the short and long-term impact of such policy.

#### 5.3 Examples using a one-dimensional version of EUROLAB

The EUROLAB model needs to be considered from a holistic perspective, taking into account both the onedimensional and multidimensional versions. Therefore, for the sake of completeness, this subsection discusses a set of examples where the one-dimensional version of EUROLAB that relies on a choice set consisting of working hours has been used. These examples regard fiscal reforms, mostly analysed within the framework of the European Semester<sup>19</sup> or been put forward in policy debates. These reforms, either hypothetical or part of political debate, are designed with the objective of reducing the tax burden in particular on low-income individuals, reducing income inequality and/or increasing labour participation of under-employed segments of the population. Often they are constrained to be budget neutral and may represent a tax shift from labour to tax bases that are less detrimental to growth.

We exemplify the usefulness of a one-dimensional version of EUROLAB by presenting four main types of policy simulations.<sup>20</sup> The first type regards changes in tax bracket, the second one involves tax shifting from labour to property, the third type considers changes in the amount of in-work tax credits, and the last type concerns changes in childcare-related subsidies for working mothers.

The first example regards the simulation of a hypothetical tax reform for Germany (2019 European Semester Country Report – Germany), a country with a relatively high share of revenues from taxes on labour. The hypothetical reform is intended to flatten the "middle-class bulge", a peculiarity of the German income tax regime generated by the sharp increase of the marginal tax rate (from 14% to 24%) in the 2nd tax bracket between EUR 8,653 to EUR 13,669. More specifically the reform consists in extending the upper threshold of this bracket to EUR 16,500. Simulation results indicate that after the reform, women are expected to increase their labour supply more than men, partly because they benefit more from the reform and partly because they have high part-time employment rates, which leaves space for adjustment in working hours. Furthermore, given relatively low levels of inactivity and unemployment in Germany, most labour supply response comes primarily from the intensive margin rather than the extensive margin. However, the highest responses come from women located in the first part of the income distribution due to a lower marginal effective tax rate paid after the reform and a higher labour supply elasticity that characterizes this population segment. The positive effects on labour are very limited for the top income deciles.

The second type of reform involves a hypothetical tax shift from labour to property in Italy (2020 European Semester Country Report – Italy).<sup>21</sup> Italy is another EU country where the tax structure relies heavily on labour, with other tax bases being exploited less. In fact there is scope for exploiting more recurrent property taxes, more specifically those applied to main residences as they were abolished in 2014, leading to a substantial revenue loss. Moreover, although employment continued to increase in 2019, the Italian labour market was still characterized by high shares of involuntary unemployment and inactivity rates as well as underemployment especially among women and young people. In light of this tax structure, JRC simulated a tax shift from labour to property by (1) reintroducing taxes on all residences exempting or not low-value properties and eventually low-income pensioners, and (2) using the additional fiscal revenues to reduce the tax burden, namely, social security contributions paid by employees with an annual income

<sup>&</sup>lt;sup>19</sup> The European Semester is a cycle of economic and fiscal policy coordination within the EU. It is part of the European Union's economic governance framework. Its focus is on the 6-month period from the beginning of each year, hence its name – the 'Semester'.

<sup>&</sup>lt;sup>20</sup> Christl et al., 2019 make use of the unilateral version of EUROLAB to analyse labour supply responses to changes in social assistance. Similarly, Christl and de Poli, 2021 assess behavioural effects of implementing a child tax credit in Austria in 2019.

<sup>&</sup>lt;sup>21</sup> See Box 4.1.1: EUROMOD-QUEST simulation – Shifting taxes from labour to property in Italy.

below EUR 24,600. After the tax shift, the model predicted a rise in the labour force participation by 1.6% and total hours worked by 2.3% for women and 0.8% for men on average, with stronger responses for low-income workers who benefitted most from the tax shifting.

The third example regards an announced reform to the existing in-work tax credit scheme in Italy (2020 European Semester Country Report – Italy), approved later that year in July 2020 by the Italian Parliament. The existing tax credit, targeting workers, amounted to EUR 80 per month for annual income between EUR 8,000 to 24,600, gradually phasing out for incomes of EUR 26,600 or more. The reform was announced to (1) increase the tax credit amount to EUR 100 per month for income up to EUR 28,000, (2) gradually reduce this amount to EUR 80 per month for annual income of EUR 35,000, and (3) phase out completely at EUR 40,000. After the reform, the model predicted (Box 4.1.1) an increase in participation rates by 2.3% for women and 0.9% for men, and total hours worked by 2.6% and 0.9% for women and men, respectively.

The fourth example of behavioural analysis regards a hypothetical reform on rationalising childcare benefits in Italy (2018 European Semester Country Report – Italy). The reform was aimed at raising the low participation of mothers in the labour market. The reforms consisted in replacing four childcare-related bonuses (Voucher babysitting, Bonus bebè, Bonus mamma, Assegno di maternità), which seemed to not properly meet working mothers' needs as well as overlapping in eligibility conditions, with a single in-work benefit, targeting low-income working mothers with children aged under three. The model predicted an increase in the labour market participation of around 17% and the average weekly working hours of around 20%. It is important to note that for this exercise, we adopted the methodological approach described in Figari and Narazani, 2020 to account for the possible interaction between labour supply and childcare choices. For that purpose, the EUROLAB one-dimensional version was modified in order to address properly the labour supply patterns of specific population groups such as working mothers and specific benefitrelated policies such as those related to childcare. More specifically, EUROLAB was enriched with information on childcare fees in public and private sectors in order to endogenously consider the childcare decisions of working mothers. This extension of EUROLAB was also important for assessing the impact of hypothetical reforms of decreasing childcare costs in four EU countries (ESDE report 2019) on labour supply of mothers and the usage of formal child care.<sup>22</sup> The behavioural analysis pointed out that reducing childcare costs by half is expected to increase the use of formal childcare and the labour supply of mothers in countries where childcare costs are relatively high (Finland and the Netherlands). In countries with relatively low childcare cost (Hungary and Lithuania) the increase in formal childcare use is very small, suggesting that policies focused on increasing availability might work better.

<sup>&</sup>lt;sup>22</sup> See Annex 2: Euromod simulations of the impact of the reduction of childcare costs on the use of the service and on the mothers' labour supply decisions, Employment and Social Developments in Europe, Annual Review 2019.

# 6. Future extensions of EUROLAB

The multidimensional modelling of EUROLAB offers several functionalities that can be exploited to extend the model in several directions. This section describes potential extensions of EUROLAB synergies with inhome EUROMOD functionalities such as the Indirect Tax Tool – version 3 (IITv3), the EUROMOD WEALTH Taxation Tool (EWIGE) and the Labour Market Transition tool (LMA).

#### 6.1 Indirect taxes

Reforms on indirect taxation can trigger behavioural effects on the labour supply decisions of households that cannot be captured by EUROMOD or similar non-behavioural microsimulation models. This is also the case with the current version of EUROLAB. For example, increases in VAT, reduced or standard rates have been frequently discussed in EU countries, for example, as part of fiscal consolidation packages taken in prior to the global financial crisis. Such reforms can stand alone or be part of budget-neutral tax-shifting from personal income tax or social security contributions for the whole population or for a part of it that can loose from VAT increase. However, simulations of indirect tax changes cannot be performed using EUROMOD as the SILC data, the underlying data of EUROMOD, do not contain detailed information on household expenditures. For that purpose, data from the Household Budget Survey are used together with the SILC data to develop an extension of the EUROMOD tool IITv3 that allows running reforms on indirect taxes at a highly disaggregated level of consumption.<sup>23</sup> Therefore a potential extension of the EUROLAB model can be towards the behavioural effects of indirect tax reforms or policy mixes between direct and indirect tax tools. In light of this extension, linking EUROLAB with the EUROMOD ITTv3 tool can help in assessing the behavioural effects of potential policy mixes of direct and indirect taxes. The linking mechanism can enable the simultaneously generation of household disposable income (through EUROMOD) and household expenditures in durable and nondurable goods at each counterfactual choice of labour supply.

# 6.2 Wealth taxation

Announcements of changes in wealth taxation parameters can result in labour supply reactions of the potential heirs. For example, decreasing the inheritance tax exemption threshold or increasing the tax rate leads to a lower income from inheritance and therefore a lower dependency on inheritance income which in turn may increase the labour supply of heirs if leisure is a normal good. Therefore, accounting for the labour effects of inheritance tax reforms might be important for assessing net budgetary effects after the reform. EWIGE 2 (Boone et al., 2019) represents the EUROMOD Wealth taxation tool that allows the simulation of wealth-related taxes and fiscal policies often deemed as an optimal tool for reducing inequality. It integrates the Household Finance and Consumption Survey (HFCS) data in EUROMOD, enabling assessment of the distributional effects of different current and hypothetical wealth taxes and policies. Yet, behavioural effects of wealth-related tax reforms are not taken into account by the EWIGE 2. For that purpose, EUROLAB can be linked with EWIGE and help quantify the second and third-round effects on heirs.

<sup>&</sup>lt;sup>23</sup> For more details on ITTv3, see "<u>A new indirect tax tool for EUROMOD Final Report</u>", EUROMOD Note.

The synergy of these models becomes even more important in face of potential tax shifting reforms from labour to wealth.

# 6.3 Labour Market Adjustment (LMA)

Constructing the unemployment alternative and simulating respective unemployment benefits in the EUROLAB model require a set of assumptions on eligibility criteria and calculation of the wage rate. These assumptions may differ country-by-country and setting them manually in EUROLAB may cause computational errors that in turn lead to an over/underestimation of unemployment benefits. Such a potential source of bias can be circumvented through a EUROMOD tool, namely the LMA Add-on, which allows changes in the labour market status of an individual or simulation of hypothetical labour market scenarios.<sup>24</sup> More concretely, the LMA Add-on can be used to construct the unemployment alternative and simulate corresponding benefits in line with the country-specific rules through its capacity of simulating the transition from employment to unemployment, either short-term or long-term. In addition, EUROLAB can benefit from the additional features of the LMA Add-on to relax the assumption of annual transitions to unemployment (or employment) by allowing users to set-up the duration of transitions.

Recently, the LMA Add-on has been enhanced with the intention of simulating the effect of monetary compensation schemes, widely used in EU countries to cushion the income loss of employed and self-employed households from the COVID-19 crisis. The enhancement consists in simulating labour market transitions from work to either unemployment or monetary compensations schemes through the LMA Add-on in order to simulate with EUROMOD the budgetary and distributional impact of these schemes. The LMA Add-on can be linked with EUROLAB to assess behavioural responses triggered by these schemes or similar policy interventions in a medium and long-run perspective using the labour demand module of EUROLAB (chapter 4). A simplistic representation of these monetary schemes is implemented in Narazani and Colombino (2021) but, as the authors argue, actual policies and more recent statistics are needed in order to analyse the extent to which the existence and design of such schemes matters in order to strengthen the resilience of labour markets to economic shocks.

<sup>&</sup>lt;sup>24</sup> For more details, see <u>LMA Add-On Documentation.pdf (europa.eu)</u>.

# Conclusions

The EU-wide microsimulation model, EUROMOD, is a static model that does not account for behavioural responses when assessing policy effects. However, its ability to replicate closely existing and counterfactual fiscal reforms together with the heterogeneity of the underlying data represent an appropriate tool to build a behavioural microsimulation model that can be used to assess the impact of fiscal or economic policies on labour supply and overall employment in EU countries. This paper provides an overview of EUROLAB, the EU labour supply-demand microsimulation model build on EUROMOD, that similarly to other behavioural microsimulation models, estimates a set of structural parameters of utility function and applies them to predict labour supply responses to hypothetical or real policy reforms. However, unlike other behavioural models, EUROLAB adopts a multidimensional approach in the construction of choice setting as well as including a representation of both labour supply and demand accounting in this way for labour market equilibrium achieved through wage adjustment.

After describing the modelling of the multidimensional structure of the choice set and the labour supplydemand mechanism, the paper discusses the potential use of the model and possible extensions through synergies with in-home EUROMOD functionalities such as the Indirect Tax Tool, the EUROMOD WEALTH Taxation Tool and the Labour Market Transition Add-on. For example, a potential synergy between EUROLAB and the Indirect Tax EUROMOD extension can help in assessing behavioural effects of indirect tax reforms like increases in VAT rates for specific goods or even more ambitious reforms like those on green taxation. On the other hand, a potential synergy with the EUROMOD Wealth extension can help in assessing the effects of announced changes in wealth taxation parameters on labour supply of potential heirs, which in turn might be important for assessing the net budgetary effects of the reform. Furthermore, a synergy of EUROLAB with the Labour Market Transition tool, a functionality of EUROMOD that allows changes in the labour market status, can help in automatizing the construction of unemployment alternatives in EUROLAB and the simulation of corresponding benefits in line with the country specific rules through its capacity to simulate the transition from employment to unemployment, either short-term or long-term. In addition, linking the LMA Add-on with EUROLAB can help in assessing behavioural responses triggered by monetary compensation schemes or similar policy interventions in a medium and long-run perspective using the labour demand module of EUROLAB.

EUROLAB proves to be relevant for assessing the effects of policies implemented by Member States in order to mitigate the negative effects of the Covid-19 pandemic on employment. However, its use may go beyond COVID and post-COVID analysis. For example, EUROLAB can be used to predict the aggregated behavioural effects of future changes in the working-age population that are related to internal or external migration flows, ageing of specific population groups or educational composition. For that, data on population projections can be integrated in the micro-data using, for example, a reweighting approach and modifying the individuals' sampling weights in order for them to correspond to the projected working-age population in a given year. On the other hand, another use of EUROLAB can be towards gender-biased reforms, for example, to assess the behavioural effects of changes in the Barcelona targets that ensure suitable childcare provision as an essential step towards equal opportunities in employment between women and men. In this case, a more advanced modelling approach that can embed childcare preferences

and childcare rationing constraints in a labour supply-demand setting would be the appropriate tool rather that a reweighting approach as in the former example.

Optimal taxation is another domain where EUROLAB can provide valuable assistance for policy analysis in the EU area. Assessing quantitatively the optimal system of taxation in order to achieve a desired target of income distribution with the least inefficiency is an important objective of the European Commission. EUROLAB can help in this debate by providing empirical analysis that enhances understanding. For that, first we have to choose a criterion that establishes the optimality of a given policy reform. For example, inefficiency reduction may represent a policy-relevant criterion in case of a minimum income scheme reform targeted to reduce poverty, while targeting inequality reduction might be more appropriate when assessing labour incentive reforms such as in-work tax credits. A combination of both of these evaluation criteria through a social welfare function may provide a more conclusive result in the identification of the optimal reform that targets simultaneously efficiency and equality objectives.

#### References

Aaberge, R., J.K. Dagsvik and S. Strøm (1995): "Labor Supply Responses and Welfare Effects of Tax Reforms", Scandinavian Journal of Economics, 97, 4, 635-659.

Aaberge, R., U. Colombino and S. Strøm (1999): "Labour Supply in Italy: An Empirical Analysis of Joint Household Decisions, with Taxes and Quantity Constraints", Journal of Applied Econometrics, 14(4), 403-422.

Akoğuz, E., Capéau, B., Decoster, A., Sadeleer, L., Güner, D., Manios, K., Paulus and A. T. Vanheukelom (2020): "A new indirect tax tool for EUROMOD", JRC Project no. JRC/SVQ/2018/B.2/0021/OC.

Almeida,V., Barrios, S., Christl, M., De Poli, S. Tumino, A and Van der Wielen, W. (2020): "The impact of COVID-19 on households ' income in the EU". Journal of Economic Inequality 19, 413–431 (2021).

Bargain, O., Caliendo, M., Haan, P., and K. Orsini (2010): "Making work pay' in a rationed labour market. Journal of Population Economics, 23(1), 323–351.

Blundell, R., MaCurdy, T. and C. Meghir (2007): "Labor Supply Models: Unobserved Heterogeneity, Nonparticipation and Dynamics", in Heckman, J. and Leamer, E. (eds.), Handbook of Econometrics, edition 1, volume 6, chapter 69 Elsevier.

Boeri, T., Giupponi, G. Krueger, A. and S. Machin (2020): "Solo Self-Employment and Alternative Work Arrangements: A Cross-Country Perspective on the Changing Composition of Jobs." Journal of Economic Perspectives, 34 (1): 170-95.

Boone, J., Derboven, J., Figari, F., Kuypers, S. and G. Verbist (2019): "EWIGE 2 – Update and Extension of the EUROMOD Wealth Taxation Project", European Commission, Joint Research Centre, Seville, JRC117204.

Bourguignon, F., Fournier, M. and M. Gurgand (2007): "Selection Bias Corrections Based on the Multinomial Logit Model: Monte-Carlo Comparisons". Journal of Economic Surveys, 21, 174-205.

Brandolini, A., A. Rosolia, A. and R. Torrini, (2010): "The distribution of employees' labour earnings in the European Union: Data, concepts and first results", in Atkinson, A. B. and Marlier, E. (eds.), "Income and living conditions in Europe", Eurostat Statistical Books, Publications Office of the European Union, Luxembourg.

Christl M., De Poli S., and J. Vargas (2019): "Reducing the income tax burden for households with children: An assessment of the child tax credit reform in Austria", European Commission, Joint Research Centre, Seville, JRC092019,.

Christl, M., and S. De Poli (2021): "Trapped in inactivity? Social assistance and labour supply in Austria". Empirica 48, 661–696.

Coda Moscarola, F., Colombino, U., Figari, F. and M. Locatelli (2020): "Shifting taxes from labour to property. A simulation under labour market equilibrium", Journal of Policy Modelling 42 (2), 367-384.

Colombino, U. (2013): "A new equilibrium simulation procedure with discrete choice models". International Journal of Microsimulation 6(3), 25–49.

Colombino, U. and E. Narazani (2013). "Designing a Universal Income Support Mechanism for Italy. An Exploratory Tour", Basic Income Studies, 2013, 8(1), 1-17.

Colombino, U., Locatelli, M., Narazani, E. and C. O'Donoghue (2010). "Alternative Basic Income Mechanisms: An Evaluation Exercise with a Microeconometric Basic Income Studies", 2010, 5(1), Article 3.

Creedy, J., Duncan, A.S., Scutella, R. and M. Harris, M.(2002): "Microsimulation Modelling of Taxation and The Labour Market: The Melbourne Institute Tax and Transfer Simulator", Edward Elgar, Cheltenham.

Creedy J. and A. Duncan, (2005): "Aggregating Labour Supply and Feedback Effects in Microsimulation", Australian Journal of Labour Economics, Vol. 8, No. 3, September 2005, pp 277 – 290.

Dagsvik, J. and S. Strøm (2004): "Sectoral Labour Supply, Choice Restrictions and Functional Form", Discussion Papers No. 388, September 2004 Statistics Norway, Research Department.

Dubin, J.A. and McFadden, D.L. (1984): "An Econometric Analysis of Residential Electric Appliance Holdings and Consumption". Econometrica, 52, 345-362.

Ericson, P., Flood, L. and R. Wahlberg (2009): "SWEtaxben: A Swedish Tax/benefit Micro Simulation Model and an Evaluation of a Swedish Tax Reform," Working Papers in Economics 346, University of Gothenburg, Department of Economics.

EU Commission (2019). 2019 European Semester: Country Reports and Communication, Germany. 2019-european-semester-country-report-germany\_en.pdf (europa.eu)

EU Commission (2018). 2018 European Semester: Country Reports and Communication, Italy. 2018-european-semester-country-report-italy-en.pdf (europa.eu)

EU Commission (2020). 2019 European Semester: Country Reports and Communication, Italy. <u>EUR-Lex - 52020SC0511 - EN - EUR-Lex (europa.eu)</u>

Employment and Social Developments Review (2019): "Sustainable growth for all: choices for the future of Social Europe".

Publications catalogue - Employment, Social Affairs & Inclusion - European Commission (europa.eu)

Figari, F. and Narazani, E. (2020). "The joint decision of female labour supply and childcare in Italy under costs and availability constraints," Economia Politica: Journal of Analytical and Institutional Economics, , vol. 37(2), pages 411-439, July.

Heckman, James J. "Sample Bias As A Specification Error," Econometrica, 1979, v47(1), 153-162.

Jongen, E, de Boer, H. and Dekker, P. (2014): "MICSIM - A behavioural microsimulation model for the analysis of tax-benefit reform in the Netherlands", CPB Background Document | 27 November 2014.

Klevmarken, A. (1997). "Behavioral Modeling in Micro Simulation Models. A Survey," Working Paper Series 1997:31, Uppsala University, Department of Economics.

McFadden, D. (1974): "Conditional Logit Analysis of Qualitative Choice Behavior". In: Frontiers in Econometric, Academic Press, New York, 105-142.

Mercante, Joseph and Penny Mok (2014a) "Estimation of wage equations for New Zealand." Wellington, New Zealand Treasury Working Paper 14/09.

Mercante, Joseph and Penny Mok (2014b) "Estimation of labour supply in New Zealand." Wellington, New Zealand Treasury Working Paper 14/08.

Narazani, E. and U. Colombino, (2021). "Modelling sector-specific employment shocks with EUROLAB, a multi-dimensional behavioural model", European Commission, Seville, 2021, JRC12559

Peichl, A., Schneider, H. and Siegloch, S. (2010): "Documentation IZAWMOD: The IZA Policy Simulation MODel". IZA Discussion Paper No. 4865.

Peichl, A. & Siegloch, S., (2012). "Accounting for labor demand effects in structural labor supply models," Labour Economics, Elsevier, vol. 19(1), pages 129-138.

Silvia De Poli, Tine Hufkens, Chrysa Leventi, Andrea Papini (2021): "SUMMARY NOTE FOR EUROMOD LABOUR MARKET ADJUSTMENT ADD-ON" <u>LMA Add-On Documentation.pdf (europa.eu)</u>.

Sutherland, H. and Figari, F. (2013) "EUROMOD: the European Union tax-benefit microsimulation model," International Journal of Microsimulation, 6(1): 4-26.

van Soest, A. (1995): "Structural Models of Family Labor Supply: A Discrete Choice Approach", Journal of Human Resources, 30, 63-88.

Vandelannoote, D. and Verbist, G. (2020). "The impact of in-work benefits on work incentives and poverty in four European countries", Journal of European Social Policy 2020, Vol. 30(2) 144–157.

# Appendix A

# A.1. Conditional Logit results

| clogit dependent variable          | Couples                          | Single Women                   | Single Men                        |
|------------------------------------|----------------------------------|--------------------------------|-----------------------------------|
| Voluntary Part-time - Male         | $2.090^{***}$<br>(5.77)          |                                | 2.758 <sup>***</sup><br>(10.43)   |
| In-work dummy - Male               | 0.253<br>(0.80)                  |                                | 0.230<br>(0.42)                   |
| Part-time dummy - Male             | -0.918 <sup>***</sup><br>(-5.28) |                                | -0.577**<br>(-2.96)               |
| Full-time dummy - Male             | 1.821 <sup>***</sup><br>(17.90)  |                                | 1.706 <sup>***</sup><br>(15.03)   |
| Unemployment dummy - Male          | 0.797 <sup>***</sup><br>(4.39)   |                                | 0.759 <sup>***</sup><br>(6.99)    |
| Voluntary Part-time - Female       | -0.0830<br>(-0.25)               | 1.441 <sup>***</sup><br>(9.06) |                                   |
| In-work dummy - Female             | 0.241<br>(1.06)                  | -1.023*<br>(-2.40)             |                                   |
| Part-time dummy - Female           | 1.408 <sup>***</sup><br>(6.83)   | 1.502***<br>(7.06)             |                                   |
| Full-time dummy - Female           | 2.180 <sup>***</sup><br>(13.93)  | 2.633***<br>(17.56)            |                                   |
| Unemployment dummy - Female        | 0.352 <sup>**</sup><br>(3.23)    | 0.285 <sup>**</sup><br>(2.69)  |                                   |
| Leisure - Male                     | -0.000294<br>(-0.01)             |                                | 0.0148<br>(0.49)                  |
| Leisure square - Male              | -0.000186<br>(-0.77)             |                                | 0.0000334<br>(0.11)               |
| Leisure $\times$ age - Male        | -0.0000621<br>(-0.05)            |                                | -0.000877*<br>(-2.23)             |
| Leisure $\times$ age square - Male | 0.00000835<br>(0.68)             |                                | 0.00000804 <sup>*</sup><br>(2.07) |

| Table 1: | Conditional | Logit | results |
|----------|-------------|-------|---------|
|          | Contaitiona | LOGIC |         |

| Leisure × #children - Male            | -0.00587*<br>(-2.48)              |                                    | 0.00477<br>(0.65)                  |
|---------------------------------------|-----------------------------------|------------------------------------|------------------------------------|
| Leisure × #children < 3 year - Male   | -0.00258<br>(-0.40)               |                                    | 0.0158<br>(0.51)                   |
| Leisure × #children 3-6 year - Male   | 0.00375<br>(0.69)                 |                                    | 0.00548<br>(0.23)                  |
| Leisure × Migrant - Male              | -0.000844<br>(-0.13)              |                                    | -0.0169***<br>(-3.65)              |
| Leisure × Mortgage - Male             | 0.000132<br>(0.67)                |                                    | -0.00143 <sup>***</sup><br>(-3.76) |
| Leisure - Female                      | 0.110 <sup>**</sup><br>(2.59)     | -0.00901<br>(-0.25)                |                                    |
| Leisure square - Female               | -0.000328<br>(-0.97)              | 0.000623<br>(1.83)                 |                                    |
| Leisure × age - Female                | -0.00161<br>(-1.79)               | -0.000897*<br>(-2.28)              |                                    |
| Leisure × age square - Female         | 0.0000159<br>(1.72)               | 0.00000820*<br>(2.13)              |                                    |
| Leisure × #children - Female          | 0.00544 <sup>**</sup><br>(2.99)   | 0.0137 <sup>***</sup><br>(4.80)    |                                    |
| Leisure × #children < 3 year - Female | 0.00988 <sup>*</sup><br>(2.37)    | 0.00909<br>(1.01)                  |                                    |
| Leisure × #children 3-6 year - Female | 0.00493<br>(1.30)                 | 0.00123<br>(0.15)                  |                                    |
| Leisure × Migrant - Female            | 0.00251<br>(0.47)                 | -0.0354 <sup>***</sup><br>(-7.40)  |                                    |
| Leisure × Mortgage - Female           | 0.0000720<br>(0.49)               | -0.000880 <sup>**</sup><br>(-2.64) |                                    |
| Leisure - Male × Leisure - Female     | 0.000778 <sup>***</sup><br>(5.61) |                                    |                                    |
| Net income                            | 0.00975 <sup>***</sup><br>(16.22) | 0.0117 <sup>***</sup><br>(16.48)   | 0.00315 <sup>***</sup><br>(6.22)   |
| Net income square                     | -0.00000177***                    | -0.00000354***                     | -                                  |

|                                    | (-20.67)                 | (-15.12)                          | 0.000000295***<br>(-3.53)          |
|------------------------------------|--------------------------|-----------------------------------|------------------------------------|
| Net income × household size        | 0.000226<br>(1.79)       | 0.000769 <sup>***</sup><br>(3.50) | 0.000612 <sup>***</sup><br>(3.45)  |
| Net income $\times$ Leisure - Male | -0.00000231<br>(-0.46)   |                                   | 0.0000370 <sup>***</sup><br>(7.96) |
| Net income × Leisure - Female      | -0.0000232***<br>(-5.44) | -0.0000102<br>(-1.90)             |                                    |
| Observations                       | 223872                   | 24256                             | 26576                              |
| 11                                 | -10395.6                 | -4505.5                           | -5064.6                            |
| r2_p                               | 0.285                    | 0.285                             | 0.267                              |
| aic                                | 20859.2                  | 9047.0                            | 10165.2                            |
| bic                                | 21210.1                  | 9192.7                            | 10312.6                            |

*t statistics in parentheses* \* *p* < 0.05, \*\* *p* < 0.01, \*\*\* *p* < 0.001

## A.2. Prediction results

| supply allematives | wonnen m Gou               | pie                     |
|--------------------|----------------------------|-------------------------|
|                    | Predicted<br>Probabilities | Observed<br>Frequencies |
| Inactive           | .0054                      | .0054                   |
| Unemployed         | .0076                      | .0076                   |
| Work 15-30 hours   | .0197                      | .0196                   |
| Work 30-45 hours   | .0333                      | .0334                   |
| Work 45-60 hours   | .0029                      | .0029                   |

Table 2.A. The average of predicted probabilities and observed fractions across labour supply alternatives – Women in Couple

Table 2.B. The average of predicted probabilities and observed fractions across labour supply alternatives – Men in Couple

|                  | Predicted<br>Probabilities | Observed<br>Frequencies |
|------------------|----------------------------|-------------------------|
| Inactive         | .0018                      | .0018                   |
| Unemployed       | .0039                      | .0039                   |
| Work 15-30 hours | .0041                      | .0024                   |
| Work 30-45 hours | .046                       | .0475                   |
| Work 45-60 hours | .0096                      | .0097                   |

Table 2.C. The average of predicted probabilities and observed fractions across labour supply alternatives – Single Women

|                  | Predicted<br>Probabilities | Observed<br>Frequencies |
|------------------|----------------------------|-------------------------|
| Inactive         | .0518                      | .0518                   |
| Unemployed       | .0686                      | .0686                   |
| Work 15-30 hours | .0876                      | .0872                   |
| Work 30-45 hours | .3164                      | .3168                   |
| Work 45-60 hours | .0357                      | .0358                   |

|                  | Predicted<br>Probabilities | Observed<br>Frequencies |
|------------------|----------------------------|-------------------------|
| Inactive         | .0376                      | .0376                   |
| Unemployed       | .0816                      | .0816                   |
| Work 15-30 hours | .041                       | .0391                   |
| Work 30-45 hours | .3388                      | .3402                   |
| Work 45-60 hours | .0606                      | .0611                   |

 Table 2.D. The average of predicted probabilities and observed fractions across labour supply alternatives – Single Men

## A.3. Elasticities results

|           |              |       | All       |           |       | Couples   |           |       | Singles   |           |
|-----------|--------------|-------|-----------|-----------|-------|-----------|-----------|-------|-----------|-----------|
|           |              | Total | Extensive | Intensive | Total | Extensive | Intensive | Total | Extensive | Intensive |
| Education | Low level    | 0.159 | 0.134     | 0.025     | 0.131 | 0.108     | 0.023     | 0.191 | 0.164     | 0.027     |
|           | Middle level | 0.145 | 0.118     | 0.027     | 0.121 | 0.090     | 0.030     | 0.176 | 0.152     | 0.024     |
|           | High level   | 0.124 | 0.093     | 0.031     | 0.095 | 0.064     | 0.031     | 0.160 | 0.129     | 0.031     |
| Age       | 20-30        | 0.210 | 0.183     | 0.027     | 0.182 | 0.147     | 0.035     | 0.220 | 0.196     | 0.024     |
|           | 31-40        | 0.159 | 0.130     | 0.029     | 0.140 | 0.106     | 0.034     | 0.180 | 0.156     | 0.023     |
|           | 41-on        | 0.128 | 0.100     | 0.027     | 0.103 | 0.077     | 0.026     | 0.163 | 0.134     | 0.029     |
| Child     | Yes          | 0.153 | 0.129     | 0.024     | 0.121 | 0.093     | 0.028     | 0.167 | 0.144     | 0.023     |
|           | No           | 0.132 | 0.101     | 0.032     | 0.115 | 0.086     | 0.029     | 0.237 | 0.189     | 0.048     |
| Total     |              | 0.144 | 0.116     | 0.028     | 0.117 | 0.088     | 0.029     | 0.176 | 0.150     | 0.026     |

### Table 1. Labour elasticities for All, it

#### Table 2. Labour elasticities for Men, it

|           |              |       | All       |           |       | Couples   |           |       | Singles   |           |
|-----------|--------------|-------|-----------|-----------|-------|-----------|-----------|-------|-----------|-----------|
|           |              | Total | Extensive | Intensive | Total | Extensive | Intensive | Total | Extensive | Intensive |
| Education | Low level    | 0.113 | 0.100     | 0.013     | 0.093 | 0.078     | 0.015     | 0.134 | 0.123     | 0.011     |
|           | Middle level | 0.112 | 0.095     | 0.017     | 0.080 | 0.060     | 0.020     | 0.147 | 0.134     | 0.013     |
|           | High level   | 0.087 | 0.068     | 0.019     | 0.041 | 0.021     | 0.019     | 0.149 | 0.130     | 0.019     |
| Age       | 20-30        | 0.163 | 0.154     | 0.009     | 0.078 | 0.058     | 0.021     | 0.178 | 0.171     | 0.007     |
|           | 31-40        | 0.118 | 0.101     | 0.017     | 0.077 | 0.055     | 0.023     | 0.154 | 0.142     | 0.013     |
|           | 41-on        | 0.094 | 0.078     | 0.017     | 0.074 | 0.058     | 0.017     | 0.126 | 0.110     | 0.016     |
| Child     | Yes          | 0.132 | 0.119     | 0.014     | 0.100 | 0.085     | 0.015     | 0.144 | 0.130     | 0.013     |
|           | No           | 0.067 | 0.047     | 0.020     | 0.064 | 0.044     | 0.020     | 0.124 | 0.107     | 0.017     |
| Total     |              | 0.107 | 0.091     | 0.016     | 0.075 | 0.057     | 0.018     | 0.143 | 0.130     | 0.013     |

|           |              |       | All       |           |       | Couples   |           |       | Singles   |           |
|-----------|--------------|-------|-----------|-----------|-------|-----------|-----------|-------|-----------|-----------|
|           |              | Total | Extensive | Intensive | Total | Extensive | Intensive | Total | Extensive | Intensive |
| Education | Low level    | 0.225 | 0.183     | 0.042     | 0.180 | 0.146     | 0.034     | 0.286 | 0.233     | 0.054     |
|           | Middle level | 0.182 | 0.143     | 0.039     | 0.161 | 0.120     | 0.040     | 0.210 | 0.174     | 0.036     |
|           | High level   | 0.151 | 0.111     | 0.040     | 0.137 | 0.096     | 0.040     | 0.169 | 0.128     | 0.040     |
| Age       | 20-30        | 0.261 | 0.215     | 0.046     | 0.228 | 0.186     | 0.042     | 0.280 | 0.231     | 0.049     |
|           | 31-40        | 0.200 | 0.159     | 0.041     | 0.190 | 0.148     | 0.043     | 0.213 | 0.176     | 0.038     |
|           | 41-on        | 0.164 | 0.125     | 0.039     | 0.137 | 0.100     | 0.036     | 0.202 | 0.160     | 0.042     |
| Child     | Yes          | 0.181 | 0.142     | 0.039     | 0.141 | 0.100     | 0.040     | 0.203 | 0.164     | 0.038     |
|           | No           | 0.185 | 0.144     | 0.041     | 0.166 | 0.128     | 0.038     | 0.253 | 0.201     | 0.052     |
| Total     |              | 0.183 | 0.143     | 0.040     | 0.159 | 0.120     | 0.039     | 0.215 | 0.174     | 0.042     |

Table 3. Labour elasticities for Woman, it

# Appendix B

## **B.1. Behavioural results**

# Table 3.1 : Labour supply changes at intensive and extensive margin, employees

|               |                              | p2020<br>_base | p2020<br>_pit1 | %<br>Change<br>after<br>p2020_<br>pit1 | p2020_<br>pit2 | %<br>Change<br>after<br>p2020_<br>pit2 |
|---------------|------------------------------|----------------|----------------|--|----------------|--|
| Hours of work |                              |                |                |  |                |  |
| Men           | In couple - without children | 39.90          | 39.78          | -0.30%                                 | 40.14          | 0.60%                                  |
|               | In couple - with children    | 38.54          | 38.32          | -0.57%                                 | 38.90          | 0.94%                                  |
|               | Single - without children    | 36.91          | 36.68          | -0.62%                                 | 37.39          | 1.30%                                  |
|               | Single - with children       | 35.48          | 35.20          | -0.78%                                 | 35.97          | 1.40%                                  |
| Women         | In couple - without children | 31.18          | 30.78          | -1.29%                                 | 31.83          | 2.07%                                  |
|               | In couple - with children    | 32.62          | 32.25          | -1.12%                                 | 33.21          | 1.81%                                  |
|               | Single - without children    | 31.62          | 31.19          | -1.35%                                 | 32.28          | 2.10%                                  |
|               | Single - with children       | 33.21          | 32.77          | -1.30%                                 | 33.87          | 1.98%                                  |
| Participation |                              |                |                |  |                |  |
| Men           | In couple - without children | 0.96           | 0.96           | -0.43%                                 | 0.97           | 0.51%                                  |
|               | In couple - with children    | 0.94           | 0.94           | -0.66%                                 | 0.95           | 0.88%                                  |
|               | Single - without children    | 0.91           | 0.90           | -0.79%                                 | 0.92           | 1.12%                                  |
|               | Single - with children       | 0.88           | 0.87           | -0.84%                                 | 0.89           | 1.28%                                  |
| Women         | In couple - without children | 0.89           | 0.88           | -0.96%                                 | 0.90           | 1.25%                                  |
|               | In couple - with children    | 0.92           | 0.91           | -0.82%                                 | 0.93           | 1.10%                                  |
|               | Single - without children    | 0.86           | 0.85           | -1.19%                                 | 0.87           | 1.37%                                  |
|               | Single - with children       | 0.88           | 0.87           | -1.19%                                 | 0.89           | 1.46%                                  |
| Unemployment  |                              |                |                |  |                |  |
| Men           | In couple - without children | 0.02           | 0.03           | 11.39%                                 | 0.02           | -13.47%                                |
|               | In couple - with children    | 0.04           | 0.04           | 10.91%                                 | 0.03           | -14.63%                                |
|               | Single - without children    | 0.06           | 0.07           | 7.79%                                  | 0.06           | -11.08%                                |
|               | Single - with children       | 0.08           | 0.09           | 6.30%                                  | 0.07           | -9.61%                                 |
| Women         | In couple - without children | 0.06           | 0.07           | 8.04%                                  | 0.06           | -10.61%                                |
|               | In couple - with children    | 0.05           | 0.05           | 9.41%                                  | 0.04           | -12.67%                                |
|               | Single - without children    | 0.08           | 0.09           | 7.49%                                  | 0.07           | -8.58%                                 |
|               | Single - with children       | 0.07           | 0.08           | 8.74%                                  | 0.06           | -10.78%                                |

**Inactivity** 

| Men   | In couple - without children | 0.01 | 0.01 | 10.95% | 0.01 | -12.89% |
|-------|------------------------------|------|------|--------|------|---------|
|       | In couple - with children    | 0.02 | 0.02 | 10.36% | 0.02 | -13.82% |
|       | Single - without children    | 0.03 | 0.03 | 7.73%  | 0.03 | -11.12% |
|       | Single - with children       | 0.04 | 0.04 | 6.27%  | 0.03 | -9.64%  |
| Women | In couple - without children | 0.04 | 0.05 | 7.76%  | 0.04 | -9.96%  |
|       | In couple - with children    | 0.04 | 0.04 | 8.53%  | 0.03 | -11.26% |
|       | Single - without children    | 0.06 | 0.06 | 7.33%  | 0.05 | -8.38%  |
|       | Single - with children       | 0.05 | 0.06 | 8.44%  | 0.05 | -10.29% |

Note: Children are defined as son-daughter of the decision-making unit. They are not older than 18 years, or if older, in education.

# Table 3.2: Labour supply changes at intensive and extensive

margin, self-employed

|                      |                              | p2020<br>base | p2020 | %<br>Change<br>after<br>p2020_ | p2020_ | %<br>Change<br>after<br>p2020_ |
|----------------------|------------------------------|---------------|-------|--------------------------------|--------|--------------------------------|
| Hours of work        |                              | _Dase         | _pit1 | pit1                           | pit2   | pit2                           |
| Men                  | In couple - without children | 40.07         | 39.87 | -0.50%                         | 40.37  | 0.75%                          |
|                      | In couple - with children    | 39.00         | 38.71 | -0.73%                         | 39.50  | 1.29%                          |
|                      | Single - without children    | 38.02         | 37.81 | -0.56%                         | 38.37  | 0.92%                          |
|                      | Single - with children       | 35.70         | 35.38 | -0.90%                         | 36.25  | 1.53%                          |
| Women                | In couple - without children | 32.49         | 32.02 | -1.44%                         | 33.19  | 2.17%                          |
|                      | In couple - with children    | 33.89         | 33.49 | -1.18%                         | 34.60  | 2.11%                          |
|                      | Single - without children    | 32.37         | 31.80 | -1.78%                         | 33.19  | 2.53%                          |
|                      | Single - with children       | 34.02         | 33.52 | -1.47%                         | 34.84  | 2.41%                          |
| <b>Participation</b> |                              |               |       |                                |        |                                |
| Men                  | In couple - without children | 0.96          | 0.95  | -0.57%                         | 0.96   | 0.56%                          |
|                      | In couple - with children    | 0.93          | 0.92  | -0.81%                         | 0.94   | 1.14%                          |
|                      | Single - without children    | 0.93          | 0.92  | -0.56%                         | 0.93   | 0.73%                          |
|                      | Single - with children       | 0.87          | 0.87  | -0.90%                         | 0.89   | 1.37%                          |
| Women                | In couple - without children | 0.90          | 0.89  | -1.06%                         | 0.91   | 1.32%                          |
|                      | In couple - with children    | 0.93          | 0.92  | -0.91%                         | 0.94   | 1.28%                          |
|                      | Single - without children    | 0.86          | 0.85  | -1.44%                         | 0.88   | 1.75%                          |
|                      | Single - with children       | 0.89          | 0.88  | -1.32%                         | 0.90   | 1.75%                          |
| <u>Unemployment</u>  |                              |               |       |                                |        |                                |
| Men                  | In couple - without children | 0.03          | 0.03  | 12.48%                         | 0.03   | -12.28%                        |
|                      | In couple - with children    | 0.05          | 0.06  | 10.23%                         | 0.04   | -14.50%                        |

|                   | Single - without children<br>Single - with children  | 0.05<br>0.09                 | 0.05<br>0.09                 | 7.22%<br>6.23%                     | 0.04<br>0.08                 | -9.51%<br>-9.46%                         |
|-------------------|--|------------------------------|------------------------------|------------------------------------|------------------------------|--|
| Women             | In couple - without children<br>In couple - with children<br>Single - without children<br>Single - with children | 0.06<br>0.04<br>0.08<br>0.06 | 0.07<br>0.05<br>0.08<br>0.07 | 9.64%<br>11.84%<br>9.23%<br>10.55% | 0.05<br>0.04<br>0.07<br>0.05 | -11.99%<br>-16.80%<br>-11.22%<br>-13.99% |
| <b>Inactivity</b> |  |                              |                              |                                    |                              |  |
| Men               | In couple - without children   | 0.01                         | 0.02                         | 11.83%                             | 0.01                         | -11.51%                                  |
|                   | In couple - with children  | 0.02                         | 0.02                         | 10.37%                             | 0.02                         | -14.19%                                  |
|                   | Single - without children  | 0.02                         | 0.02                         | 7.19%                              | 0.02                         | -9.40%                                   |
|                   | Single - with children   | 0.04                         | 0.04                         | 6.34%                              | 0.04                         | -9.69%                                   |
| Women             | In couple - without children   | 0.04                         | 0.04                         | 9.50%                              | 0.04                         | -11.82%                                  |
|                   | In couple - with children  | 0.03                         | 0.03                         | 11.22%                             | 0.03                         | -15.80%                                  |
|                   | Single - without children  | 0.06                         | 0.06                         | 9.22%                              | 0.05                         | -11.18%                                  |
|                   | Single - with children   | 0.05                         | 0.05                         | 10.37%                             | 0.04                         | -13.75%                                  |
|                   |  |                              |                              |                                    |                              |  |

Note: Children are defined as son-daughter of the decision-making unit. They are not older than 18 years, or if older, in education.

# B.2. Behavioural results by quintile

Table 4: Labour supply changes by income quintiles

| II<br>V<br>V<br>V               | -11<br>11-1V<br>/-V1<br>/11-<br>/111<br>X-X | 32.53<br>35.67<br>37.55<br>38.54<br>38.25 | 32.15<br>35.36<br>37.34<br>38.39<br>38.20 | -1.15%<br>-0.87%<br>-0.56%<br>-0.38%<br>-0.13% | 33.03<br>36.19<br>37.97<br>38.88 | 1.54%<br>1.46%<br>1.09%<br>0.89% |
|---------------------------------|---|---|---|--|----------------------------------|----------------------------------|
| Men I-<br>II<br>V<br>V<br>V     | - V<br>/-V <br>/  -<br>/                    | 35.67<br>37.55<br>38.54                   | 35.36<br>37.34<br>38.39                   | -0.87%<br>-0.56%<br>-0.38%                     | 36.19<br>37.97<br>38.88          | 1.46%<br>1.09%                   |
| II<br>V<br>V<br>V               | - V<br>/-V <br>/  -<br>/                    | 35.67<br>37.55<br>38.54                   | 35.36<br>37.34<br>38.39                   | -0.87%<br>-0.56%<br>-0.38%                     | 36.19<br>37.97<br>38.88          | 1.46%<br>1.09%                   |
| V<br>V<br>V                     | /-VI<br>/II-<br>/III                        | 37.55<br>38.54                            | 37.34<br>38.39                            | -0.56%<br>-0.38%                               | 37.97<br>38.88                   | 1.09%                            |
| V<br>V                          | /11-<br>/111                                | 38.54                                     | 38.39                                     | -0.38%   | 38.88                            |                                  |
| V                               | /111  |   |   |  |                                  | 0.89%                            |
|                                 |   |   |   |  |                                  | 0.89%                            |
| D                               | X-X   | 38.25                                     | 38.20                                     | -0 13%   |                                  | 0.0070                           |
|                                 |   |   |   | 0.13/0   | 38.38                            | 0.35%                            |
|                                 |   |   |   |  |                                  |                                  |
| Women I-                        | -11   | 26.80                                     | 26.12                                     | -2.55%   | 27.61                            | 3.03%                            |
| 11                              | II-IV                                       | 32.04                                     | 31.24                                     | -2.51%   | 33.33                            | 4.00%                            |
| V                               | /-VI  | 34.39                                     | 33.85                                     | -1.57%   | 35.28                            | 2.58%                            |
| V                               | /11-  |   |   |  |                                  |                                  |
| V                               | /111  | 36.22                                     | 35.96                                     | -0.72%   | 36.74                            | 1.43%                            |
| D                               | X-X   | 34.61                                     | 34.62                                     | 0.03%  | 34.62                            | 0.02%                            |
| <u>Participatio</u><br><u>n</u> |   |   |   |  |                                  |                                  |
|                                 | -11   | 0.85                                      | 0.84                                      | -1.12%   | 0.86                             | 1.28%                            |
|                                 | II-IV                                       | 0.91                                      | 0.90                                      | -0.93%   | 0.92                             | 1.33%                            |
| V                               | /-VI  | 0.93                                      | 0.92                                      | -0.70%   | 0.94                             | 1.07%                            |
|                                 | /11-  |   |   |  |                                  |                                  |
| V                               | /111  | 0.93                                      | 0.93                                      | -0.58%   | 0.94                             | 0.89%                            |
| l)                              | X-X   | 0.91                                      | 0.91                                      | -0.43%   | 0.92                             | 0.50%                            |
| Women I-                        | -11   | 0.72                                      | 0.70                                      | -2.02%   | 0.73                             | 2.05%                            |
|                                 | II-IV                                       | 0.83                                      | 0.81                                      | -2.11%   | 0.85                             | 2.65%                            |
|                                 | /-VI  | 0.90                                      | 0.89                                      | -1.40%   | 0.92                             | 1.85%                            |
|                                 | /11-  | 0.00                                      | 0.05                                      |  | 0.02                             | 2.0070                           |
|                                 | /111  | 0.93                                      | 0.92                                      | -0.85%   | 0.94                             | 1.15%                            |
|                                 | X-X   | 0.89                                      | 0.89                                      | -0.39%   | 0.89                             | 0.14%                            |
| Note: Income qui                |   |   |   |  |                                  |                                  |

reform system

## **B.3. Demand results**

#### Table 1: % Changes in employment, unemployment and inactivity rate

#### pit1 Reform

|              |             |             |             | No          | With        |
|--------------|-------------|-------------|-------------|-------------|-------------|
|              |             | No          | With        | Equilibrium | Equilibrium |
|              | Baseline    | Equilibrium | Equilibrium | (%)         | (%)         |
| Employment   | 15481124.96 | 15356684.14 | 15394522.72 | -0.80%      | -0.56%      |
| Inactivity   | 1339669.173 | 1375306.5   | 1358925.861 | 2.66%       | 1.44%       |
| Unemployment | 1117347.457 | 1206150.944 | 1184693.003 | 7.95%       | 6.03%       |
| v            |             |             |             |             | 3.434135839 |

#### pit2 Reform

|             |             |             |             | No          | With         |
|-------------|-------------|-------------|-------------|-------------|--------------|
|             |             | No          | With        | Equilibrium | Equilibrium  |
|             | Baseline    | Equilibrium | Equilibrium | (%)         | (%)          |
|             |             | 15704434.3  | 15695696.5  |             |              |
| Employment  | 15481124.96 | 2           | 1           | 1.44%       | 1.39%        |
|             |             | 1244382.77  | 1248423.04  |             |              |
| Inactivity  | 1339669.173 | 3           | 6           | -7.11%      | -6.81%       |
| Unemploymen |             | 989324.496  | 994022.033  |             |              |
| t           | 1117347.457 | 7           | 1           | -11.46%     | -11.04%      |
| v           |             |             |             |             | -0.800336563 |

\* Definitions:

Employment = Total number of people in employment, only labour supply endogenous sample Inactivity = Total number of people in inactivity, only labour supply endogenous sample. Unemployment = Total number of people in unemployment, only labour supply endogenous sample.

*v* = Parameter of percentage change in wages.

#### **GETTING IN TOUCH WITH THE EU**

#### In person

All over the European Union there are hundreds of Europe Direct information centres. You can find the address of the centre nearest you at: <u>https://europa.eu/european-union/contact\_en</u>

#### On the phone or by email

Europe Direct is a service that answers your questions about the European Union. You can contact this service:

- by freephone: 00 800 6 7 8 9 10 11 (certain operators may charge for these calls),

- at the following standard number: +32 22999696, or
- by electronic mail via: <u>https://europa.eu/european-union/contact\_en</u>

#### FINDING INFORMATION ABOUT THE EU

#### Online

Information about the European Union in all the official languages of the EU is available on the Europa website at: <a href="https://europa.eu/european-union/index\_en">https://europa.eu/european-union/index\_en</a>

#### **EU publications**

You can download or order free and priced EU publications from EU Bookshop at: <u>https://publications.europa.eu/en/publications</u>. Multiple copies of free publications may be obtained by contacting Europe Direct or your local information centre (see <u>https://europa.eu/european-union/contact\_en</u>).

# The European Commission's science and knowledge service

Joint Research Centre

## **JRC Mission**

As the science and knowledge service of the European Commission, the Joint Research Centre's mission is to support EU policies with independent evidence throughout the whole policy cycle.



EU Science Hub ec.europa.eu/jrc

@EU\_ScienceHub

**f** EU Science Hub - Joint Research Centre

in EU Science, Research and Innovation

EU Science Hub