# An Impact Evaluation of the Effects of Income Support Benefits on Aggregate Labour Supply

# Preliminary draft Please do not cite or distribute without permission

Javier Martín-Román\*

UNED

jmartin@cee.uned.es

Ángel L. Martín-Román\*\*

Universidad de Valladolid almartin@uva.es

#### Abstract

This paper evaluates the impact of income support benefits on aggregate labor supply, focusing on the Agrarian Unemployment Benefit (AUB) implemented in Andalucía, Spain. Using a quasi-experimental design, the Synthetic Control Method (SCM), we compare the labor force participation rate of Andalucía with a synthetic counterpart. The results indicate that the AUB increased the participation rate by approximately 2 percentage points during the initial years after implementation. However, this rise did not surpass the number of AUB recipients, suggesting partial offsetting effects. The analysis identifies both encouraging and discouraging labor supply responses, highlighting the role of monitoring mechanisms and the macroeconomic environment. This research contributes to the literature by providing a macroeconomic perspective on income support policies, exploring theoretical channels of labor supply decisions, and applying SCM as an innovative tool in this context. The findings have implications for the design of similar policies in regions with structural unemployment.

**Keywords:** Labour supply, Labour force participation, Income support benefits, Synthetic Control Method.

**JEL codes**: J21, J65, H24, K31, C32, C54.

# 1. Introduction

The raising inequality in developed countries in the last decades has encourage governments to pass legislation to support low-income earners (Meyers et al., 2001; Bucci and Jansa, 2021). Among the policies used to improve living conditions of the latter, we might highlight the so-called Minimum Income Scheme (MIS) and a set of special (and more generous) unemployment benefits (SUB). This kind of policies are often linked to a specific behaviour regarding labour supply decisions. Particularly, active job search is required in some cases in order to be eligible to receive the allowance.

At the same time, there is a growing concern about regional unbalances, with thriving regions coexisting together with lagging territories within the same country. This, in turn, triggers central governments to transfer large amounts of financial resources to their lagging regions to compensate such differences. These financial resources are channelled through pensions, education, national health care, and unemployment protection systems that are funded through taxes collected mainly by the central administration. The Italian Mezzogiorno, East Germany, and the southern Spanish region of Andalucía are well studied cases of heavily subsidized regions (Boltho et al., 1997; Sinn and Westermann, 2001; Jofre-Monseny, 2014). Some of the social programs may modify the labour supply behaviour of the persons living in the targeted territories (e.g. more generous SUB in specific territories).

The aim of the paper is to examine the effects of one of these SUB, the Agrarian Unemployment Benefit (AUB), on the aggregate labour supply at the extensive margin. The AUB was intended to help unemployed workers in the agrarian sector of two Spanish regions, Andalucía and Extremadura. We take advantage of the fact that the AUB was not implemented in all Spanish regions to set up a quasi-experimental design by constructing a counterfactual Andalucía and comparing it with the real one.

Thus, we investigate two research questions that are interrelated. The first one is whether significant income support programs such as the AUB have encouraging or discouraging effects on the labour supply at the macro level. More precisely, we attempt to ascertain whether the labour force participation rate increases or decreases derived from the enactment of this income support program. The second research question is conditional on having obtained a positive result in the first one. Put differently, if the participation rate rises after the implementation of the policy, we seek to determine whether the number of "new active persons" outweighs the number of benefit recipients or not. If the former is the case, it would be possible to state that there exist spill-over effects associated to the policy, whereas discouraging effects would be an issue otherwise.

Our approach to addressing these research questions is twofold. First, we build a theoretical model of labour supply decisions, following the lines set up in Martín-Román et al. (2020) and Martín-Román (2022), but adapted to tackle the main questions of interest in this research. Equipped with this theoretical background, we can identify the incentives and disincentives that the AUB creates among the individuals regarding their labour supply choices. Moreover, it allows us to determine the major channels through which such a set of incentives operates and to differentiate among them between microeconomic and macroeconomic effects affecting the working-age population. Second, we test two hypotheses arising straightforwardly from the two research questions by means of the Synthetic Control Methods (SCM) approach. We make use of this empirical methodology since the focus in this research is on the aggregate labour force participation rates. Thus, since we wanted to consider not only the microeconomic factors but also those other effects operating at the macro level, we adopt a macroeconomic perspective instead of using a microeconometric approach.

As regards the results, we obtain evidence indicating that the implementation of the AUB increased the participation rate in Andalucía by about 2 percentage points during the years immediately following the law's approval. This is a rather significant finding because it is often stated that income support benefits tend to discourage active job search, dropping labour force participation rates.

Anyhow, when computing whether this increase in labour force exceeds the total number of AUB beneficiaries, the answer is negative. In other words, we found that the number of "activated" individuals as a consequence of the AUB implementation is positive but less than the number of AUB recipients. This finding entails that what could be called the labour-enhancing effects on the workforce only partially offset the discouraging effects created by that income support program.

The paper contributes to the literature in three ways. First, it adds a macroeconomic perspective to the previous works on the effects of income support programs on labour supply, which mainly adopt a microeconomic approach. This is important since, as will be shown later, even non-beneficiaries might be affected by the AUB. Second, it explores the different theoretical channels through which the AUB might encourage or discourage the labour supply. Hence, our theoretical framework allows us to clearly set up the two hypotheses to be tested. Third, it makes use of the SCM approach to address the issue, which is a novelty within this topic to the best of our knowledge. Moreover, this methodology enables us to test whether the spill-over effects offset the abovementioned discouraging effects, which is not possibly when using a microeconometric framework.

# 2. Institutional Background

#### 2.1. Historical Overview

Since the late decades of the 20th century, Andalucía and Extremadura have faced significant structural challenges related to the dependence on their rural economy and labor precariousness. These regions, characterized by a strong reliance on the agricultural sector, have historically experienced unemployment rates above the national average, high economic inequality, and significant rural depopulation, which have jeopardized their social and territorial cohesion.

In response to this situation, Spain implemented a series of public policies aimed at protecting seasonal agricultural workers, who were particularly affected by income seasonality due to the nature of agricultural work. Among the first initiatives were the Community Employment Programs of 1971 and 1976, which focused on generating temporary employment through basic infrastructure projects. More specifically, the Community Employment System of 1971 and 1976 was an initiative implemented in Andalucía and Extremadura to tackle high rural unemployment and labor precariousness associated with the agricultural sector. This system consisted of creating temporary jobs through local infrastructure and public works projects, with the objective of providing income to seasonal agricultural workers during periods of seasonal inactivity.

Although this initiative had positive impacts by providing temporary income, its transitory nature and certain inefficiencies led to its replacement in 1984 by the Integrated Protection System for Seasonal Agricultural Workers (SIPTEA), which was more structured and provided broader coverage. One of the three pillars of SIPTEA from the outset was the Agrarian Unemployment Benefit in Andalucía and Extremadura (AUB), which established a better-organized system to provide an economic safety net for these workers. This marked the beginning of a landmark social policy that has evolved significantly to the present day.

# 2.2. The Agrarian Unemployment Benefit (AUB)

The public intervention analyzed in this paper is the AUB. Also known as "subsidio agrario", the AUB is a program that has only been accessible to residents of the Spanish regions of Andalucía and Extremadura since 1984. As said above, the examined benefit is one of the three pillars that constitute a wider national policy, the SIPTEA, which aims to protect temporary agricultural workers in the aforementioned territories. The other two pillars that are part of it are the "Planes de Formación Ocupacionales" (PFO) and the "Plan de Empleo Rural" (PER).

The main purpose of the PFO was to promote the qualification and professional integration of unemployed youth aged between 16 and 25

years. Training was provided within the framework of the so-called "Plan Específico de Formación Ocupación Rural."

As for the PER, it should be noted that it was primarily structured based on agreements between the former "Instituto Nacional de Empleo" (INEM)¹ and the town halls of the two regions already cited. The goal was for temporary agricultural workers registered in the REASS to participate, through the corresponding employment contract, in the execution of infrastructure projects that were the responsibility of the municipalities. Recognizing the relevance of the program, it is also important to emphasize that in 1987 a total of 237,700 jobs were accommodated. On average, each infrastructure project generated 30 jobs, each lasting less than a month. Workers could count the days worked on these projects to fulfill the minimum workday requirements, making them eligible for the AUB. Beyond just providing employment, the main objective of the PER program was to empower its workers to qualify for the AUB.

Finally, the third of the axes that make up the SIPTEA, the AUB, is the measure evaluated in this paper, and it is a benefit to which temporary agricultural workers registered in the REASS census could have access. This subsidy is a welfare (Social Security) benefit, and its specific regulation is carried out by REASS. This special Social Security regime was created by Royal Decree 3237/1983, dated December 28, and came into effect on January 1, 1984.

On the other hand, to be eligible as a beneficiary of the AUB, in the general modality<sup>2</sup>, compliance with the following conditions were required: (1) Be unemployed; (2) Work for somebody else, be a temporary worker and be registered in the REASS census, and be registered with a Social Security scheme or in an equivalent situation; (3) Have a residence in a locality in Andalucía or Extremadura; (4) Be 16 years or older and not having reached the minimum age for obtaining a retirement pension; (5) Have no income of any kind, individual or family, at the time of application and during the receipt of the subsidy, that exceeds the legally established maximums on an annual basis; and (6) Have contributed a minimum of 10 days in the REASS<sup>3</sup>.

According to the Royal Decree 2298/1984, the subsidy will have a maximum duration of one hundred and eighty days within a twelvementh period, and its amount will be 75 % of the minimum interprofessional wage in force at any given time for non-seasonal workers (see **TABLE 1**).

<sup>&</sup>lt;sup>1</sup> The public entity INEM disappeared in 2003 and was replaced by the "Servicio Público de Empleo Estatal" (SEPE), an organization that fulfills the same task as the previous institution.

<sup>&</sup>lt;sup>2</sup> There is a special type of subsidy for temporary agricultural workers over 52 years of age whose fundamental difference with the general case is in the longer duration of the right to receive the subsidy.

<sup>&</sup>lt;sup>3</sup> This is the original requirement. Subsequent regulatory reforms have modified this threshold.

TABLE 1. Duration and amount of the AUB (1984)

Total contributed workdays ("peonadas")	Maximum duration of the subsidy (number of days)	Amount
Between 10 and 33	100	
Between 34 and 59	Equal to three times the total contributed workdays	75 % of minimum wage*
60 and more	180	

Note: (\*) Minimum wage in force at all times for non-temporary workers, which also includes the worker's contribution to the REASS during the period receiving the AUB. Source: Royal Decree 2298/1984.

Additionally, it will cover the worker's contribution to the Special Agricultural Social Security Scheme during the period in which the subsidy is received.

Furthermore, Article 6 of the aforementioned Royal Decree sets out the following: *Unemployment benefits are incompatible with:* 

- a) The simultaneous performance of paid work, whether self-employed or employed by others.
- b) Periodic Social Security benefits, except for family protection.
- c) Any other form of unemployment benefit.
- d) The receipt of any kind exceeding the minimum interprofessional wage (SMI), excluding the proportional part of extraordinary payments, as specified in section e) of paragraph 1 of Article 2.

#### 2.3. Evolution in the number of beneficiaries

Another relevant aspect is the evolution in the number of beneficiaries. **TABLE 2** presents this evolution from 1984 to 2000 for Andalucía.

TABLE 2. Evolution of the Number of Beneficiaries in Andalucía (by Province): 1984-2000

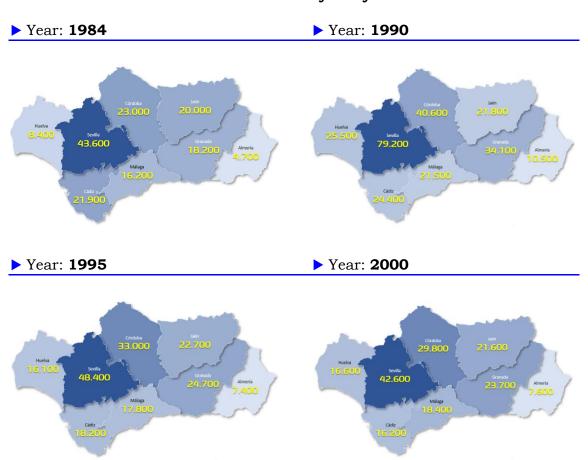
Year	Andalucía	Almería	Cádiz	Córdoba	Granada	Huelva	Jaén	Málaga	Sevilla	
1984	156,000	4,700	21,900	23,000	18,200	8,400	20,000	16,200	43,600	
1985	152,100	4,500	18,200	22,900	17,900	9,800	19,800	14,200	45,100	
1986	193,700	7,100	21,300	27,700	24,700	14,500	24,600	17,900	55,900	
1987	213,900	7,300	23,000	29,800	27,900	18,300	22,900	18,800	65,900	
1988	241,000	8,500	23,800	35,500	31,600	22,900	25,600	20,000	73,000	
1989	256,400	10,400	26,000	40,400	33,400	25,100	23,200	20,700	77,200	
1990	257,600	10,500	24,400	40,600	34,100	25,500	21,800	21,500	79,200	
1991	206,200	8,300	16,600	33,700	28,100	18,600	24,400	20,300	56,100	
1992	173,900	7,400	14,600	31,100	23,700	13,100	23,100	18,200	42,800	
1993	196,300	8,400	17,900	34,300	25,800	15,400	24,500	19,400	50,700	
1994	198,900	7,900	18,700	35,000	25,500	16,500	24,000	18,800	52,500	
1995	188,300	7,400	18,200	33,000	24,700	16,100	22,700	17,800	48,400	
1996	167,400	6,800	16,700	29,800	23,700	13,400	19,900	16,900	40,200	
1997	168,605	7,000	16,500	29,700	23,200	14,100	20,400	18,100	39,700	
1998	177,410	7,700	16,200	31,600	23,000	14,100	22,000	19,000	43,700	
1999	184,040	8,020	16,180	32,820	24,845	14,750	22,050	19,900	45,400	
2000	176,500	7,600	16,200	29,800	23,700	16,600	21,600	18,400	42,600	

Source: Labor Statistics Bulletin.

The total number of beneficiaries in Andalucía shows a significant increase between 1984 and 1990, reaching its peak in 1990 with 257,600 beneficiaries. Subsequently, from 1991 onwards, a general downward trend is observed, with some fluctuations, reaching 176,500 beneficiaries in 2000. A more detailed analysis reveals a general pattern across all provinces: a continuous increase in the number of beneficiaries throughout the 1980s, peaking in 1990. Thereafter, there is a consistent decline until the end of the period. In absolute terms, Seville consistently leads, while Almería and Huelva record the lowest figures.

On the other hand, **FIGURE 1** presents the distribution of beneficiaries across Andalucía's eight provinces using four choropleth maps for the years 1984, 1990, 1995, and 2000.

FIGURE 1. Provincial Distribution of Beneficiaries in Andalucía



Source: Own elaboration based on the Labor Statistics Bulletin.

Another interesting aspect —regarding the number of beneficiaries— is the distribution over the years by sex (see **FIGURE 2**). Since the implementation of the subsidy in 1984, the trend observed in the number of recipients has been upward until 1990, with a particularly significant increase in the number of female receivers of the benefit.

This growth has more than compensated for the reduction in the number of subsidized men. Later, from 1991 onwards, the number of women getting the benefit is even higher than that of men.

This singular evolution of the period 1984-1990 is justified by the changes that the approval of the AUB entailed compared to the previous scheme. The AUB is aimed at REASS members who meet certain requirements, which replaces its previous discretionary character with its current regulated nature. Furthermore, the approval of the AUB makes the subsidy into an individual right, not a family one, which implies the possibility of accumulating more than one benefit in a single family. These changes led to both an increase in membership and a rise in the number of recipients. The social diversification of those who could access their collection was also favored. Especially, in the case of women, who took center stage.

**FIGURE 2** also let us check that, while the number of male beneficiaries of the AUB in Andalucía decreased by about 40 % in the period 1984-2000, data referring to females multiplied by almost 7.

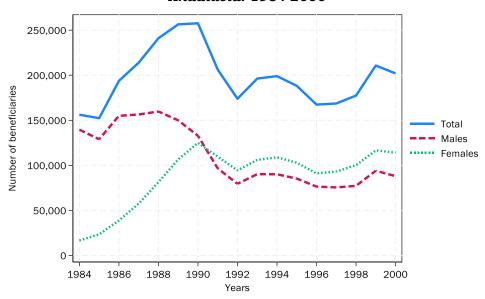


FIGURE 2. Evolution in the number of AUB beneficiaries.
Andalucía: 1984-2000

Source: own elaboration from "Boletín de Estadísticas Laborales".

A final note of interest is the one revealing that, during the period under study, there is evidence suggesting that a significant portion of AUB recipients were *not authentic* agricultural workers. An *ethically questionable* practice involved allocating workdays ("peonadas") to relatives to maximize household-level unemployment benefits (Cansino, 2000).

The particular shift in the sex composition of recipients did not align with changes in the actual agricultural workforce in Andalucía. According to the "Encuesta de Población Activa" (Spanish Labor Force Survey), the male-to-female ratio among agricultural workers in these regions remained around a seventh both in 1984 and 1991 (Jofre-Monseny, 2014).

# 3. A Theoretical Model

## 3.1. Basic theoretical setting

We construct a labour market participation model in order to identify the theoretical channels through which the AUB affects the LFPR. As we are focused on the extensive margin of the labour supply, we assume a fixed working week. Thus, labour supply choices coincide with participation decisions (e.g., Martin-Roman et al., 2019; Martin-Roman 2020). Furthermore, the model developed here is extended to account for the effects of unemployment and the AUB on the LFPR.

According to the distinction made by Rodrik (2015) between critical and non-critical assumptions, the structure of the model comprises of three critical assumptions (i.e., the mechanisms driving our results) and a set of other non-critical assumptions (discussed in **Appendix 1**).<sup>4</sup> The three critical assumptions are listed below:

**Assumption 1.** Persons are eligible to receive the AUB if they live and work in the agricultural sector. To be entitled to collect the benefit, in principle, they need to be unemployed too. Thus, there are three types of individuals, those eligible and entitled, those eligible and not entitled, and those not eligible. All are potentially affected by the implementation of the AUB, though. The variable indicating whether the AUB is effectively implemented is denoted by r. This variable could be thought of as a dummy variable taking the value 1 when the AUB is effectively applied and 0 otherwise.

**Assumption 2.** A positive unemployment rate u exists. That rate determines the likelihood p of finding a job, which is the same for all individuals. This probability also depends positively on the implementation of the AUB. Therefore, p(u,r) with  $[\partial p(u)/\partial u] < 0$  and  $[\partial p(r)/\partial r] > 0$ . Taking into consideration the discrete nature of the variable:  $p(u_0, r = 1) > p(u_0, r = 0)$ .

-

<sup>&</sup>lt;sup>4</sup> In Appendix 1 some variables are defined too.

 $<sup>^5</sup>$  Therefore, unemployment is primarily involuntary. Obviously, the higher the unemployment rate, the lower p. On the other hand, the positive relationship between the AUB and the probability of finding a job may be based on the grounds of macroeconomic considerations. In this vein, if the AUB is effectively implemented, the overall economic activity in the area will be stimulated, and, consequently, the job search will produce better results.

**Assumption 3.** The AUB allowance consists of an unemployment benefit amounting to b, in principle, conditional on looking for a job. However, even though labour agencies attempt to enforce that requirement, they cannot always monitor the entire population rightly. With probability q, an individual who is not effectively looking for a job is caught and loses the entitlement to receive the AUB, and, with probability (1-q), he or she is able to obtain the AUB without looking effectively for a job. The AUB is an increasing function of the variable r,  $[\partial b(r)/\partial r] > 0$  or, in discrete terms: b(r = 1) > b(r = 0) = 0.6

In **FIGURE A2-1**, the set of alternatives for the three types of workers is shown. The levels of consumption and leisure have been replaced, within the utility function, by the corresponding values associated with each decision. Thus, we are already accounting for the budget constraints within the decision-making framework. In **FIGURE A2-1**, *y* is the real non-labour income. Total time has been normalized to 1.

# [FIGURE A2-1]

To better understand the theoretical channels through which the AUB influences the aggregate LFPR, we analyse sequentially the three groups mentioned above and then aggregate them.

#### 3.2. Effects of the AUB on non-eligible persons

To begin with, let us analyse an individual who is not eligible to receive the AUB. The individual has two options (see panel A in **FIGURE A2-1**). Each option is associated with a level of utility, either certain or expected: (1) not participating and (2) participating, which can be formalized, respectively, as:

$$U(y,1) \tag{1}$$

$$p(r)U(w\bar{l} + y, 1 - \bar{l} - s) + (1 - p(r))U(y, 1 - s)$$
(2)

The reservation wage for an individual  $(w_0^R)$  may be defined, as usual, as the value of w equalizing both options:

$$p(r)U(w_0^R\bar{l}+y,1-\bar{l}-s)+(1-p(r))U(y,1-s)=U(y,1) \hspace{1cm} (3)$$

It is easy to calculate from expression (3) that when b=0, then  $w^R$  is positive ( $w^R>0$ ). Focusing first on leisure time, we have that  $1>(1-s)>(1-\bar{l}-s)$ . This would entail that  $w^R\bar{l}+y>y$  in order to attain an equality in (3), which in turn implies that  $w^R_0>0$ .

<sup>&</sup>lt;sup>6</sup> Here, for the sake of simplicity, we are abstracting from other types of unemployment benefits or social allowances. Including those elements into the model would add complexity without providing much insight.

It is worth mentioning, though, that even those workers not eligible for the benefit are affected by the enactment of the AUB law. The reason for that is that the AUB changes the likelihood of finding a job. Thus, even those not directly affected by the AUB are indirectly influenced because of the fiscal stimulus caused by such an allowance. Taking equation (3) and making use again of the implicit function theorem, it is straightforward to compute the effects of changes in p on  $w_0^R$ :

$$\frac{\partial w_0^R}{\partial p} = -\frac{U(w_0^R \bar{l} + y, 1 - \bar{l} - s) - U(y + b, 1 - s)}{p\bar{l}U_C(w_0^R \bar{l} + y)} < 0 \tag{4}$$

The negative sign of (4) is the result of the definition given in (3). First, it is evident that U(y,1) > U(y,1-s). Second, to achieve equality in (3),  $U(w_0^R \bar{l} + y, 1 - \bar{l} - s) > U(y + b, 1) > U(y, 1 - s)$  must be fulfilled. In other words: when p rises (drops),  $w_0^R$  decreases (increases).

Following the aggregation process described in **Appendix 1**, it is easy to conclude that the AUB has an encouraging effect on the participation rate for non-eligible workers. Let us denote  $PR^{NE}$  to such a rate. Holding constant the rest of determinants of  $w_0^R$ , the effect can be formalized through expression (5):

$$\frac{\partial PR^{NE}(r)}{\partial r} = \underbrace{\frac{\partial PR}{\partial w_0^R}}_{(-)} \cdot \underbrace{\frac{\partial w_0^R}{\partial p}}_{(-)} \cdot \underbrace{\frac{\partial p}{\partial r}}_{(+)} > 0$$
 (5)

We can state that  $\partial p/\partial r > 0$  (by definition), that  $\partial w_0^R/\partial p < 0$  (from the discussion in this section), and that  $\partial PR/\partial w_0^R < 0$  (from the concept of reservation wage).

#### 3.3. Effects of the AUB on eligible and entitled persons

The analysis of individuals eligible to receive the AUB is more complex, Furthermore, it depends on if the requirement to be considered jobseekers is enforced or not, as mentioned in Assumption 3. In this case, the expected utility level associated with the option of not participating is given by expression (6):

$$qU(y,1) + (1-q)U(y+b,1)$$
(6)

It will prove to be useful to study separately the cases of perfect monitoring (i.e., q = 1), no monitoring (i.e., q = 0), and partial monitoring (i.e., 0 < q < 1).

Thus, we can define a different reservation wage for each case. Equations (7), (8), and (9) define the reservation wages for perfect monitoring  $(w_1^R)$ , no monitoring  $(w_2^R)$ , and partial monitoring  $(w_3^R)$ , respectively:

$$p(r)U(w_1^R\bar{l}+y,1-\bar{l}-s) + (1-p(r))U(y+b,1-s) = U(y,1)$$
(7)

$$p(r)U(w_2^R\bar{l}+y,1-\bar{l}-s) + (1-p(r))U(y+b,1-s) = U(y+b,1)$$
(8)

$$p(r)U(w_3^R\bar{l}+y,1-\bar{l}-s) + (1-p(r))U(y+b,1-s) = qU(y,1) + (1-q)U(y+b,1)$$
 (9)

It can be established a relationship among the three and relating them to  $w_0^R$ . First, notice that when b=0, both expressions (7) and (8) coincide with expression (3) and, therefore,  $w_0^R=w_1^R=w_2^R$ . Then, we can examine how  $w_1^R$  and  $w_2^R$  change when b varies. By using the implicit function theorem with (7) and (8), respectively, it is quite straightforward to calculate that:

$$\frac{\partial w_1^R}{\partial b} = -\frac{(1-p)U_C(y+b)}{p\bar{l}U_C(w_1^R\bar{l}+y)} < 0 \tag{10}$$

$$\frac{\partial w_2^R}{\partial b} = \frac{U_C(y+b)}{\bar{l}U_C(w_2^R\bar{l}+y)} > 0 \tag{11}$$

The sign of (10) is evident. The positive sign of expression (11) is due to the additivity of the utility function (Assumption A5), which implies that  $U_C(y+b,1-s) = U_C(y+b,1)$ . Thus, it can be concluded that  $w_2^R > w_0^R > w_1^R$  for b > 0.7 These relationships are shown in **FIGURE A2-2**.

# [FIGURE A2-2]

To account for how  $w_3^R$  is related to the other three reservation wages, first note that equation (9) predicts that when q=0 then  $w_3^R=w_2^R$ , and when q=1 then  $w_3^R=w_1^R$ . Moreover, it is easy to compute how  $w_3^R$  evolves when q varies:

$$\frac{\partial w_3^R}{\partial q} = -\frac{U(y+b,1) - U(y,1)}{p\bar{l}U_C(w_3^R\bar{l}+y)} < 0 \tag{12}$$

The negative sign of (12) is due to the decreasing marginal utility of income. In addition, it can also be established that there is a linear relationship between  $w_3^R$  and q (i.e., the slope is constant) as depicted in **FIGURE A2-3**, since  $(\partial^2 w_3^R/\partial q^2) = 0$ . Finally, it is possible to compute a critical value for q (that we denote  $q^*$ ) that equals  $w_3^R$  and  $w_0^R$ .8

$$\frac{\partial^2 w_1^R}{\partial b^2} = -\frac{(1-p)U_{cc}(y+b)}{p\bar{l}U_c\big(w_1^R\bar{l}+y\big)} > 0; \ \frac{\partial^2 w_2^R}{\partial b^2} = \frac{U_{cc}(y+b)}{\bar{l}U_c\big(w_2^R\bar{l}+y\big)} < 0$$

<sup>8</sup> From the definitions in equations (3) and (9) and assumption A5, when  $w_0^R = w_0^R$ , then:

$$U(y,1) - (1-p)U(y,1-s) = qU(y,1) + (1-q)U(y+b,1) - (1-p)U(y+b,1-s) \Leftrightarrow p = q^*$$

-

<sup>&</sup>lt;sup>7</sup> Besides, it is straightforward to prove that  $w_1^R$  is a convex function of b and  $w_2^R$  a concave function of b:

# [Figure A2-3]

The relevance of **FIGURE A2-3** is that it reveals that the AUB encourages or discourages labour participation depending on the level of monitoring. Thus, for a given amount of b, if the level of monitoring exceeds  $q^*$ , the worker is incentivised to participate in the labour market, compared to the situation with no AUB since  $w_3^R < w_0^R$ . Otherwise, when  $q < q^*$ , the individual is discouraged, compared to a scenario with no AUB, since  $w_3^R > w_0^R$ . The economic rationale is based on a comparison between the known without uncertainty loss of leisure time linked to labour participation and the expected gain of receiving the AUB. Specifically, the likelihood of losing the AUB entitlement, if caught not looking for a job, determines such assessment. When  $q > q^*$ , the expected gain outweighs the loss and does not compensate otherwise.

Expression (13) shows this ambiguity formally:9

$$\frac{\partial w_3^R}{\partial b} = \frac{(p-q)U_C(y+b)}{p\bar{l}U_C(w_3^R\bar{l}+y)} \le 0 \tag{13}$$

Therefore, the sign of (13) depends on the difference between the likelihood of finding a job and the likelihood of being caught without searching for a job when claiming for the AUB. It is also evident from (13) that, for relatively high values of q, the sign is negative (i.e.,  $(\partial w_3^R/\partial b) < 0$ ). Hence, the level of monitoring becomes key to account for the encouraging or discouraging effect that the AUB has on the labour supply. The higher the q, the greater the incentives for an individual to participate in the labour market.

Let us now analyse the influence of the AUB on the PR, at the aggregate level, through the first theoretical channel (i.e., changes in b holding constant p). We dub the participation rate for those eligible to receive the benefit as  $PR^E$ . Following the aggregation process described in Appendix 1, the effects of the AUB on the LFPR may be summarized by means of expression (14):

$$\left. \frac{\partial PR^{EE}(r)}{\partial r} \right|_{\bar{p}} = \underbrace{\frac{\partial PR}{\partial w_3^R}}_{(-)} \cdot \underbrace{\frac{\partial w_3^R}{\partial b}}_{(?)} \cdot \underbrace{\frac{\partial b}{\partial r}}_{(+)} \le 0 \tag{14}$$

In (14),  $\partial p/\partial b > 0$  by hypothesis and  $\partial PR/\partial w_3^R < 0$  from the concept of reservation wage. However, as has been discussed,  $\partial w_3^R/\partial b$  does not have a definite sign. It can be stated, though, that the higher the level of monitoring, the more likely the PR increases as a result of the AUB.

$$\frac{\partial^2 w_3^R}{\partial b^2} = \frac{(p-q)U_{CC}(y+b)}{p\bar{l}U_C(w_3^R\bar{l}+y)} \le 0$$

\_

<sup>&</sup>lt;sup>9</sup> This ambiguity affects the concave or convex profile of  $w_3^R$  as a function of b too:

The second theoretical channel through which the AUB can yield effects on  $PR^{EE}$  is via changes in the probability of finding a job. At the individual level, from the definition of  $w_3^R$  in (9), the reservation wage varies as expression (15) shows:

$$\frac{\partial w_3^R}{\partial p} = -\frac{U(w_3^R \bar{l} + y, 1 - \bar{l} - s) - U(y + b, 1 - s)}{p \bar{l} U_C(w_3^R \bar{l} + y)} \le 0$$
 (15)

Unlike the case of non-eligible individuals shown in (4), expression (15) does not have an unambiguous sign. The utility index  $U(w_3^R \bar{l} + y, 1 - \bar{l} - s)$  can be greater or less than U(y + b, 1 - s), and, thus, the numerator may be positive or negative. This is so because U(y + b, 1 - s) can be greater or less than the option of not participating qU(y,1) + (1-q)U(y+b,1), depending on individual's income-leisure preferences. At the aggregate level we have:

$$\frac{\partial PR^{EE}(r)}{\partial r} \bigg|_{\bar{b}} = \underbrace{\frac{\partial PR}{\partial w_3^R}}_{(-)} \cdot \underbrace{\frac{\partial w_3^R}{\partial p}}_{(?)} \cdot \underbrace{\frac{\partial p}{\partial r}}_{(+)} \le 0$$
(16)

Expression (16) confirms that this second theoretical channel (i.e., changes in p holding constant b) also generates a mixed set of incentives for eligible persons.

# 3.4. Effects of the AUB on eligible but not entitled persons

The last case is that of individuals who are eligible to receive the AUB (i.e., they work in the Andalusian agricultural sector) but are not entitled since they have not contributed the minimum number of days yet. In this case, the reservation wage is defined formally by (17):

$$p(r)U(w_4^R\bar{l} + y + b(r), 1 - \bar{l} - s) + (1 - p(r))U(y, 1 - s) = U(y, 1)$$
 (17)

In (17), we assume that when the individual finds a job, he/she earns the labour income (i.e.,  $w\bar{l}$ ) and receives the AUB in the same period. In other words, we consider a time span that comprises the contribution period and an additional period to receive the benefit. The rest of the terms in (17) may be interpreted as before. From (17) and (7), it is easy to prove that  $w_4^R < w_1^R$ . The analysis of the effect of the AUB is quite straightforward from the previous discussion. In mathematical terms, expression (18) and (19) shows how the reservation wage varies when the AUB and the likelihood of finding a job change, respectively: 10

$$\frac{\partial^2 w_4^R}{\partial b^2} = -\frac{U_{CC}(y+b)}{\bar{l}U_C\big(w_4^R\bar{l}+y\big)} > 0$$

\_\_\_

<sup>&</sup>lt;sup>10</sup> It can also be proved that  $w_4^R$  is a concave function of b:

$$\frac{\partial w_4^R}{\partial b} = -\frac{U_C(y+b)}{\bar{l}U_C(w_4^R\bar{l}+y)} < 0 \tag{18}$$

$$\frac{\partial w_4^R}{\partial p} = -\frac{U(w_4^R \bar{l} + y + b, 1 - \bar{l} - s) - U(y, 1 - s)}{p\bar{l}U_C(w_4^R \bar{l} + y)} < 0 \tag{19}$$

The unambiguous negative sign in both expressions highlights the strong incentive to participate in the labour market for this group. As both effects reinforce each other, at the aggregate level, the participation rate for those eligible but not entitled  $(PR^{ENE})$  ought to increase when the law establishing the AUB is passed. Formally:

$$\frac{\partial PR^{ENE}(r)}{\partial r} = \underbrace{\frac{\partial PR}{\partial w_4^R}}_{(-)} \left( \underbrace{\frac{\partial w_4^R}{\partial b}}_{(-)} \cdot \underbrace{\frac{\partial b}{\partial r}}_{(+)} + \underbrace{\frac{\partial w_4^R}{\partial p}}_{(-)} \cdot \underbrace{\frac{\partial p}{\partial r}}_{(+)} \right) > 0 \tag{20}$$

# 3.5. From theory to empirics

Adding up the participation rate of eligible and non-eligible persons, we obtain the total participation rate (PR). If we denote the percentage of non-eligible population by  $\theta$ , the percentage of eligible and entitled population by  $\pi$ , and, consequently, the share of eligible and but not entitled population by  $(1 - \theta - \pi)$ , we can write the total PR as a linear combination of the three groups:

$$PR(r) = \theta \cdot PR^{NE}(r) + \pi \cdot PR^{EE}(r) + (1 - \theta - \pi) \cdot PR^{ENE}(r)$$
 (21)

From the definition given in (21) and the outcomes obtained in (5), (14), (16), and (20), it follows:

$$\frac{\partial PR(r)}{\partial r} = \underbrace{\theta \frac{\partial PR^{NE}(r)}{\partial r}}_{(+)} + \underbrace{\pi \frac{\partial PR^{E}(r)}{\partial r}}_{(?)} + \underbrace{(1 - \theta - \pi) \frac{\partial PR^{ENE}(r)}{\partial r}}_{(+)} \ge 0 \quad (22)$$

In other words, the effect on the labour supply of the AUB for the entire population is not defined without ambiguity from a theoretical standpoint. The group of non-eligible persons is incentivised by the AUB law as the perspectives of finding a job after the search process improve. The group of eligible but not entitled persons is incentivised by the same theoretical mechanism too, and, in addition, they are also incentivised to participate since the financial reward for the participation option is greater. Nevertheless, the group of eligible and entitled persons can be encouraged or discouraged to participate, as argued above. There are theoretical reasons to expect that the LFPR may be affected by the law approving the Agrarian Unemployment Benefit (AUB), either positively or negatively. The sign of such effect is an empirical matter.

To account for the effects of AUB on LFPR, first, we define the Activated Population to Beneficiaries Index (APBI) as the estimated number of persons entering the labour force because of the AUB divided by the number of AUB receivers (AUBR) in each moment of time. To compute the number of Activated Persons (AP), we need to calculate the counterfactual labour force participation rate ( $LFPR^{C}$ ) in the case in which the law of the AUB had not been enacted. To do so, we make use of the Synthetic Control Method (SCM) approach, discussed later. Then we calculate de difference between the real labour force participation rate ( $LFPR^{R}$ ) and  $LFPR^{C}$  and multiply this result by the working-age population (WAP) in each moment of time t. Formally:

$$AP_t = (LFPR_t^R - LFPR_t^C) \cdot WAP_t \tag{23}$$

Thus, APBI can be defined formally in the following way:

$$APBI_t = \frac{AP_t}{AUBR_t} \tag{24}$$

By examining how this index evolves after passing the law, it is possible to assess the effects of AUB on the LFPR. More precisely, we are interested in establishing whether the LFPR increases or decreases after AUB was approved, on the one hand, and, conditional on an increase in APBI, whether such an increase was more proportional or less proportional than the growth in the AUBR.

Let us write two formal hypotheses to better understand the implications of the enactment of the law approving the AUB.

**Hypothesis 1**: The enactment of the law approving the AUB caused an increase in the LFPR. In other words,  $APBI_t > 0 \ \forall t = 1984 \ to \ 2000$ .

This hypothesis implies that AP is positive since AUBR is always positive by definition after the law enactment. Put differently, it indicates that the encouraging effects (i.e., the incentives to enter the labour force) exceed the discouraging effects (i.e., the incentives to quit the labour force) for the whole working-age population at the aggregate level.

**Hypothesis 2:** Conditional on APBI being positive, the enactment of the law approving the AUB caused an increase in AP more than proportional than the increase in AUBR (which is just the number of AUBR since before passing the law, there were no beneficiaries). In other words,  $APBI_t > 1 \,\forall t = 1984$  to 2000.

This second hypothesis attempts to unveil whether there is a multiplier effect within the labour force. It is important to know whether the labour force growth is greater than or less than the rise in AUBR from an economic policy standpoint. If so, that would mean that these types of

policies generate economic activity beyond the group of beneficiaries. On the contrary, if there is an increase in the labour force, but less than proportional than that of the AUBR, we should conclude that, even though the encouraging effects above mentioned outweigh the discouraging effects, these second ones are still playing a role and they should be monitored by the employment agencies.

# 4. The Synthetic Control Methodology (SCM)

#### 4.1. Intuition

To assess the effects of passing the RD 3237/1983 on the activity rate in Andalucía, we apply the Synthetic Control Methodology (SCM). The comparison unit in the SCM is selected as the weighted average of all potential comparison units that best resembles the characteristics of the case of interest during the preintervention period. This technique was originally proposed by Abadie and Gardeazabal (2003) as a means to analyze the effects of terrorism in the Basque Country on GDP per capita, and with Abadie et al. (2010) the generalized application of the methodology was established. Since this work, the method has been widely used to examine effects caused by a broad variety of specific events—see Craig (2015) for a wide review.

The SCM has been applied in numerous studies ranging from the evaluation of the economic impact of natural disasters (Cavallo et al., 2013) to the assessment of the effect of institutional interventions on a population's consumption and welfare (Abadie et al., 2010), among others. Within the framework of public policy evaluation, the SCM has been consolidated as one of the most powerful methodologies for conducting impact evaluations in the last decade.

The most important advantages associated with the SCM are the following: (1) A number of public policy interventions affect aggregate units. The management of and access to macro-level data are more common and simple than the treatment of micro-level data, and there are many series available at that level of aggregation; (2) Regressions applied to samples of large entities, such as countries or regions, have been frequently questioned. Such regressions involve carrying comparisons of entities with potentially different characteristics. In applying the SCM methodology, we resort to data-driven procedures that reduce the discretion in the choice of comparison control units and that allow us to create appropriate comparison groups; (3) The SCM does not involve making strict hypotheses to make precise estimations as with other quantitative techniques such as those of the difference-indifferences approach<sup>11</sup>; (4) Finally, the standard results inform us, not only of which are the units that make up the synthetic unit, but of the particular contribution – the relative importance – of each one of them.

<sup>&</sup>lt;sup>11</sup> See Abadie (2021) for a more detailed explanation.

Among restrictions applied, it is important to point out the ones mentioned below: (1) Some units in the *donor pool* should present both higher and lower values in predictor variables in comparison to that affected by the intervention. Otherwise, it would be impossible to appropriately recreate the unit of treatment; (2) In the preintervention period, units of control should have predictor values comparable to those of the treated unit.<sup>12</sup> In addition, these variables should have an approximately linear effect on the result; (3) It has been recommended that using all preintervention outcomes together with covariates as predictors be avoided (Kaul et al., 2018). Otherwise, one would restrain the predictive power of the remaining covariates. (4) Lastly, the statistical inference procedure is much less formal than those implemented by other quantitative methods and more traditional techniques.

#### 4.2. Formalization

Initially, let us assume that there are J+1 regions where j=1 denotes the region treated (Andalucía, in this case) and j=2,...,J+1 denote untreated or control region (the rest of the *Spanish Comunidades Autónomas*, with the exception of Extremadura<sup>13</sup>). It is thus assumed that a single region is affected by the event considered and that J units are available to contribute to the synthetic control (*donor pool*).

Let us assume that  $Y_{it}^N$  represents the outcome (activity rate in the main results) that would be noticed for region i at time t without the passing of the RD 3237/1983, for units i=1,...,J+1, and time periods t=1,...,T. We also suppose that  $T_0$  is the number of pre-intervention periods, with  $1 \le T_0 < T$ , and  $Y_{it}^I$ , the outcome that would be checked for unit i at time t if unit i is exposed to the event investigated in periods  $T_{0+1}$  to T. 14

Let us consider as well that  $\alpha_{it} = Y_{it}^I - Y_{it}^N$  stands for the effect of the RD 3237/1983 for unit i at time t, and imagine that  $D_{it}$  is an indicator taking value one when unit i suffers the effects of its passing, and value zero otherwise. Then, the observed outcome for unit i at time t could be described as follows:

$$Y_{it} = Y_{it}^N + \alpha_{it} D_{it} \tag{25}$$

Bearing in mind that the only the first region is affected by the legislative norm analyzed, and only when  $t > T_0$ , we can state that:

<sup>&</sup>lt;sup>12</sup> We proceed this way to avoid interpolation bias and overfitting (Abadie et al., 2015; Grier and Maynard, 2016).

<sup>&</sup>lt;sup>13</sup> Extremadura was also affected by the RD 3237/1983. When implementing the SCM, all those units – regions, in this case – that have adopted measures similar to the treatment analyzed must be excluded from the group of potential *donors* or controls. Otherwise, the estimates carried out would not be correctly capturing the impact of the examined policy.

<sup>&</sup>lt;sup>14</sup> We presuppose that there is no effect of the passing of the Law on the outcome of interest before its occurrence, that is,  $Y_{it}^I = Y_{it}^N$  when  $t \le T_0$ .

$$D_{it} = \begin{cases} 1 & if \ i = 1 \quad and \ t > T_0 \\ 0 & otherwise \end{cases}$$
 (26)

Ultimately, we intend to estimate  $\alpha_{1t}$  for  $t > T_0$ . Thus, reordering terms in (1) we get:

$$\alpha_{1t} = Y_{1t}^I - Y_{1t}^N = Y_{1t} - Y_{1t}^N \tag{27}$$

For the region affected by the Law passed (treated unit),  $Y_{1t}^N$  cannot be observed in the post-treatment periods. Data are available for the actual path of the outcome  $(Y_{1t}^I)$ , but it is unknown what would have happened with that trajectory if it had not suffered the effects of the event under study. Therefore, we look for an estimate of  $Y_{1t}^N$  that, following Abadie et al. (2010), is given by a linear factor model. This is necessary to quantify the effect of the event by calculating the difference specified in (27).

To find optimal weights, Abadie and Gardeazabal (2003), for the pre-RD 3237/1983 passing, defined a  $(K \times 1)$  vector  $X_1$  with the values of K predictors of the outcome variable, and a  $(K \times J)$  matrix  $X_0$ , which measures the values of the same variables for the *donor pool*. The vector of optimal weights referring to the control regions,  $W^*$ , is the one that minimizes the following problem:

$$\| X_1 - X_0 W \|_{\nu} = (X_1 - X_0 W)' V (X_1 - X_0 W)$$
 (28)

where  $W^* = (w_1^*, w_2^*, ..., w_{J+1}^*)'$  is a  $(J \times 1)$  vector of non-negative weights that sums to one, and V is a symmetric, diagonal matrix of non-negative components that represents the relative importance of the selected predictors. Once we have obtained the matrix  $W^*(V^*)$  formed by the estimated optimal weights that each region of the control group receives for the design of the synthetic control unit, it is enough to apply these weights in (3) to obtain the estimate of the effect of the RD 3237/1983:

$$\hat{\alpha}_{1t} = Y_{1t} - \sum_{i=2}^{J+1} w_j^* Y_{jt}$$
 (29)

# 4.3. Inference

With the SCM methodology, neither confidence intervals nor statistical significance parameters are calculated, which are typical procedures in an inference analysis. Alternatively, the SCM offers complementary options also known as *falsification* tests. With "in-space" placebos, each region integrating the original *donor pool* is separately conceived as a treated entity and the SCM is applied as if all these regions, individually, were affected by the pass of the Law (Abadie et al., 2010; Abadie et al., 2015).

By applying this iterative mechanism, we obtain a distribution of estimated placebo treatment effects for all regions in which no event occurred. Considering that none of these control regions has been influenced by the Law studied, we should only observe great disparities between these *placebo* regions and their corresponding synthetic control randomly and in sporadic cases. A more accurate mechanism for identifying the significance of the results is based on the Root Mean Squared Prediction Error (RMSPE), which is the index typically used to assess the goodness of fit when applying the SCM. It measures for a given unit of analysis the fit —or lack thereof— between the actual outcome variable and its synthetic counterpart. In other words, it represents the distance or discrepancy between the path drawn by each variable. Formally, it is defined as follows:

$$RMSPE = \sqrt{\frac{1}{T_0} \sum_{t=1}^{T_0} \left( Y_{1t} - \sum_{j=2}^{J+1} w_j^* Y_{jt} \right)^2}$$
 (30)

Ultimately, we must calculate the ratio between the postintervention RMSPE (the average for 1984q1–2020q1) and preintervention RMSPE (the average for 1980q1–1984q1) and determine how many control regions present an effect as large as that observed in the treated one: Andalucía.

# 5. Data

We use quarterly regional-level data (1980q1–2000q1) from the Spanish Statistical Office (INE) for the 17 Spanish regions (*Comunidades Autónomas*). The region considered to be affected by the event analyzed is Andalucía. The rest stand as possible candidates to take part in the control group (*donor pool*). The successful use of the SCM requires an important assumption to be fulfilled: it is essential to dispense with all units suffering the effects of a similar event in some years of the preintervention period – in our case: 1980q1 to 1984q1. If these units were included, they could interfere with and condition the true effects of the intervention examined (Abadie et al., 2010).

Regarding the predictors included in the estimates, we use the following: males in active population (%), actives aged 25-54 years over total actives (%), agricultural employment over total employment (%), construction employment over total employment (%), long-term unemployment (1-2 years) (%), very long-term unemployment (> 2 years) (%), and the lagged outcome variable for several periods preceding the pass of the examined Law – activity rate (1981q2), activity rate (1982q2) and activity rate (1983q2).

With respect to the number of predictors used, it should be underscored that increasing their number does not always improve the fit, and similarly eliminating some of them does not necessarily worsen the fit (McClelland and Gault, 2017).

Additionally, regarding the predictors considered, one of the most common practices in the application of this methodology involves the use of the lagged outcome variable (Abadie et al., 2010). By including several lags of the outcome variable, we measure the effect of other predictors. This strategy somehow mitigates the effects of not incorporating relevant predictors into the analysis. However, there is no consensus on what a suitable number of lags is. Some authors have drawn attention to the desirability of encompassing all outcome lags available as predictors. Furthermore, they believe that including other covariates has hardly any influence on the final estimates (Athey and Imbens, 2006). On the other hand, other scholars claim that only using the lags of the outcome variable is not the best solution (Kaul et al., 2016).

Without any additional predictor, the estimated model cannot be supported by economic theory and does not have any justification. Ferman et al. (2016) recommend working with different specifications, using several combinations of lags and generating all possible results. This latter option is the one we use in this investigation (see Table A.1 in Appendix 2). We determined which models provides a better fit – the ones that present the lowest RMSPE – when selecting a maximum of three lags of the outcome variable from the set of predictors. The three best models are those whose RMSPE is highlighted in bold: specification [1], [3] and [8].

#### 6. Results

We are interested in determining how the activity rate of Andalucía would have evolved in absence of the Law passed in 1984. For this purpose, we use a combination of different Spanish regions to construct a synthetic control unit that resembles as much as possible the actual evolution of the actual activity rate in Andalucía before 1984. The subsequent track of this counterfactual Andalucía, without effects of the "treatment", is then compared to the actual path.

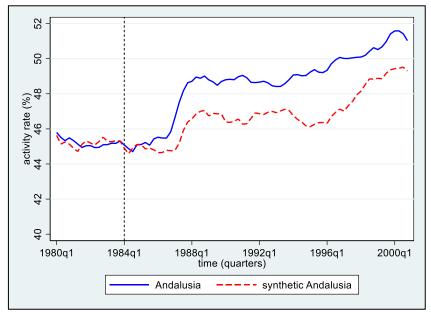
#### 6.1. Main results

Regarding what constitutes a good fit or how to appraise similarities, the most direct and immediate option is to resort to the *eyeball test* **(FIGURES 3a and 3b)** by comparing the evolution of the activity rate in the treatment region (Andalucía) to that of the control group. Our first result is that the evolution of actual Andalucía and its synthetic counterpart practically overlap in the three models analyzed <sup>16</sup>. This is the first requirement to be met if we want to rely on estimates of causal impact.

 $<sup>^{15}\,\</sup>mathrm{We}$  rule out using more lags for the reasons stated above.

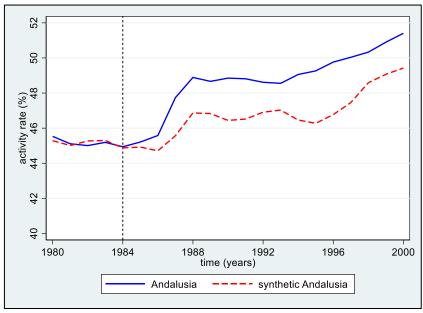
<sup>&</sup>lt;sup>16</sup> We only include the figure corresponding to specification [1], the model we follow henceforth for presenting the most significant results in the post-treatment periods. The other figures are available upon request.

FIGURE 3a. "Eyeball test" (quarterly data)



Source: Own elaboration.

FIGURE 3b. "Eyeball test" (yearly data)



Source: Own elaboration.

This visual inspection allows for a preliminary assessment of how closely the two series follow a similar trajectory over time. While the *eyeball test* is a straightforward and intuitive approach, it remains subjective and can overlook subtle differences. Therefore, complementing this initial evaluation with quantitative measures can provide a more robust and reliable assessment of the model's performance (see *6.2. Inference*).

As for the effects estimated, the gap between the actual rate and that of the synthetic unit reports and quantifies the impact on percentage points. On average (1984-2000), the impact is close to 2 percentage points using both quarterly data (FIGURE 4a) and annual data (FIGURE 4b).

1980q1 1984q1 1988q1 1992q1 1996q1 2000q1 time (quarters)

FIGURE 4a. Impact / effects (quarterly data):

Source: Own elaboration.

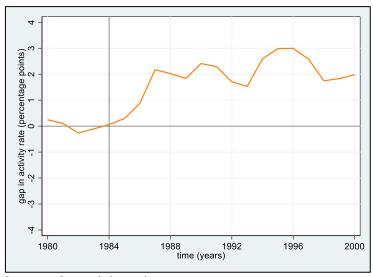


FIGURE 4b. Impact / effects (yearly data):

Source: Own elaboration.

# 6.2. Inference

In the context of the SCM, the ratio of the post-treatment RMSPE to the pre-treatment RMSPE plays a crucial role in the inference and robustness of the estimated treatment effects. This ratio, often referred to as RMSPE post/pre, serves as a diagnostic measure to assess whether the observed deviation of the treated unit from its synthetic counterpart after the policy intervention is significant and meaningful **(FIGURES 5a and 5b)**.

The RMSPE measures the average discrepancy between the observed values of the outcome variable and the corresponding synthetic estimates.

A high ratio suggests that the policy or intervention introduced a notable change in the treated unit's outcome variable, providing strong evidence of a causal effect. In other words, the synthetic control constructed from the donor pool accurately replicated the treated unit's pre-treatment trajectory, but post-intervention, the observed outcome diverged considerably. This divergence is precisely what SCM aims to detect as potential evidence of policy impact.

Post-RMSPE / Pre-RMSPE Quarterly data Andalusia Catalonia Balearic Islands Basque Country Asturias Navarre Aragon Canary Islands Castile and Leon Galicia La Rioja Community of Madrid Cantabria Valencian Community Castilla-La Mancha Region of Murcia

FIGURE 5a. Post-RMSPE / Pre-RMSPE (quarterly data):

Source: Own elaboration.

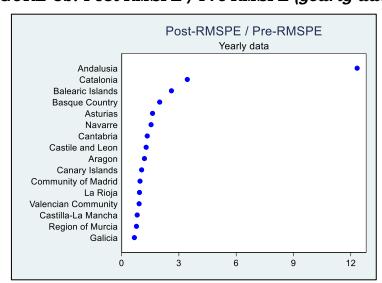


FIGURE 5b. Post-RMSPE / Pre-RMSPE (yearly data):

When considering Andalucía as the unit of treatment, it emerges in first position with a ratio around 8 (with quarterly data). If we manage annual data, the post-event RMSPE is roughly 12 times the RMSPE of the pre-event period. It indicates that the synthetic control accurately captured the pre-treatment dynamics, and that the observed post-treatment divergence is substantial. This metric, combined with placebo tests, provides compelling evidence of a causal policy effect, reinforcing the credibility and reliability of the SCM approach.

This quotient is the analytical result of one of the most well-known resources in the analysis of synthetic controls: the *placebo runs* – an iterative method showing the distribution of the estimated gaps for the regions in which no event occurred (see **FIGURE 6** below).

These placebo runs or placebo tests in space are a critical component of the inferential process. SCM is a data-driven approach that constructs a synthetic counterfactual for a treated unit by optimally combining untreated units. Nevertheless, even when a strong pre-treatment fit is achieved, one must ensure that the observed post-treatment divergence is not simply due to chance or model overfitting. Placebo tests in space provide a systematic way to evaluate whether the estimated effect is genuinely attributable to the intervention or if similar deviations are observed for units that were not treated.

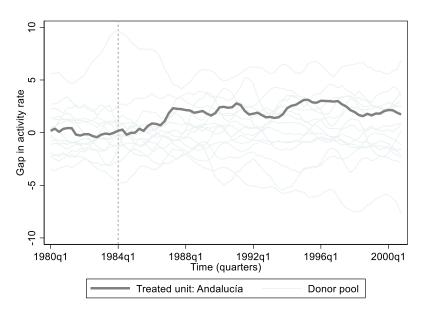


FIGURE 6. Placebo tests

Source: Own elaboration.

The robustness checks implemented, including *placebo tests* and sensitivity analyses, verify the statistical significance of the observed effects.

The *placebo runs* applied to regions without the policy intervention, yield negligible changes, reinforcing the causal interpretation of the results. Sensitivity analyses confirmed the initial impressions and preliminary evidence provided by the *eyeball test*, proving that the good fit shown by the *eyeball test* is not at all a product of chance.

The results also highlight the presence of potential labor supply disincentives, albeit to a lesser extent than initially hypothesized. The moderate magnitude of the increase suggests that the income effect, associated with higher financial security from the subsidy, may have partially offset the substitution effect, which incentivizes additional labor market participation. This nuanced outcome reflects the importance of incorporating both microeconomic and macroeconomic perspectives in the analysis of social policies.

# 7. Discussion and Policy

The evidence presented make available a clear understanding of how the AUB influences labor supply dynamics at the macro level. The increase in the participation rate, though significant, does not exceed the number of beneficiaries, suggesting that the policy primarily functions as an income support mechanism rather than an activator of additional labor supply.

This outcome aligns with theoretical expectations regarding income support policies, which often generate mixed incentives for labor market participation. One relevant aspect observed is the heterogeneous response by sex, with a notable increase in female participation. The theoretical model proposed explains these patterns through variations in reservation wages and monitoring intensity, which play a crucial role in shaping participation decisions.

Additionally, the analysis indicates potential spillover effects, though insufficient to alter the fundamental labor supply structure. These insights emphasize the importance of aligning policy design with monitoring strategies to optimize labor market outcomes.

# 8. Conclusions

The results of the analysis, obtained through the application of the SCM, reveal a noteworthy increase in the labor force participation rate in Andalucía following the introduction of the AUB. Specifically, the participation rate rose by approximately 2 percentage points in the initial years post-implementation. This finding is particularly remarkable given the robustness of the SCM methodology, which allows for a precise

comparison by constructing a synthetic counterfactual from a weighted combination of control regions.

On the other hand, the analysis of the AUB demonstrates the complex relationship between income support benefits and labor supply. The observed increase in labor force participation following the implementation of the AUB reflects both direct and indirect effects. Directly, the subsidy provided temporary financial relief, stabilizing income for seasonal agricultural workers. Indirectly, it may have influenced participation decisions through altered expectations regarding job availability. Nevertheless, the overall effect remains moderate, indicating that such policies should be accompanied by complementary measures that incentivize active labor market engagement.

Future research could explore long-term impacts and the role of structural factors, such as agricultural sector shifts and regional economic disparities. The findings underscore the relevance of macroeconomic perspectives in policy design, providing valuable insights for regions facing similar labor market challenges.

# References

- Abadie, A., & Gardeazabal, J. (2003). The economic costs of conflict: A case study of the Basque Country. *American Economic Review* 93: 113-132.
- Abadie, A., Diamond, A., & Hainmueller, J. (2010). Synthetic control methods for comparative case studies: Estimating the effect of California's tobacco control program. *Journal of the American statistical Association* 105: 493-505.
- Abadie, A., Diamond, A., & Hainmueller, J. (2015). Comparative politics and the synthetic control method. *American Journal of Political Science* 59: 495-510.
- Abadie, A. (2021). Using synthetic controls: Feasibility, data requirements, and methodological aspects. *Journal of Economic Literature*, 59(2), 391-425.
- Athey, S., & Imbens, G. W. (2006). Identification and inference in nonlinear difference-in-differences models. *Econometrica* 74: 431-497.
- Boltho, A., Carlin, W., & Scaramozzino, P. (1997). Will East Germany become a new Mezzogiorno? *Journal of Comparative Economics*, 24(3), 241-264.
- Bucci, L. C., & Jansa, J. M. (2021). Who passes restrictive labour policy? A view from the States. *Journal of Public Policy*, *41*(3), 409-439.
- Cansino, J. M. (2000). El subsidio agrario. Principales magnitudes (1984–1999). Estudios Agrosociales y Pesqueros 189, 11–28.
- Cavallo, E., Galiani, S., Noy, I., & Pantano, J. (2013). Catastrophic natural disasters and economic growth. *Review of Economics and Statistics* 95: 1549-1561.
- Craig, P. (2015). Synthetic Controls: A New Approach to Evaluating Interventions. Working Paper. What Works Scotland.
- Ferman, B., Pinto, C., & Possebom, V. (2016). Cherry Picking with Synthetic Controls. FGV Working Paper 420. São Paulo, Brazil: Sao Paulo School of Economics.
- Jofre-Monseny, J. (2014). The effects of unemployment protection on migration in lagging regions. *Journal of Urban Economics*, 83, 73-86.
- Kaul, A., Klößner, S., Pfeifer, G., & Schieler, M. (2018). Synthetic Control Methods: Never Use All Pre-Intervention Outcomes as Economic

- Predictors. Working Paper. Saarbrücken, Germany: Saarland University.
- Martín-Román, Á. L. (2022). Beyond the added-worker and the discouraged-worker effects: the entitled-worker effect. Economic Modelling, 110, 105812.
- Martín-Román, Á. L., Cuéllar-Martín, J., & Moral, A. (2020). Labor supply and the business cycle: The "Bandwagon Worker Effect". *Papers in Regional Science*, 99(6), 1607-1642.
- McClelland, R., & Gault, S. (2017): "The Synthetic Control Method as a Tool to Understand State Policy." *Washington, DC: Urban-Brookings Tax Policy Center.*
- Meyers, M. K., Gornick, J. C., & Peck, L. R. (2001). Packaging support for low-income families: Policy variation across the United States. *Journal of Policy Analysis and Management: The Journal of the Association for Public Policy Analysis and Management*, 20(3), 457-483.
- Rodrik, D. (2015). Economics rules: The rights and wrongs of the dismal science. New York: WW Norton & Company.
- Sinn, H. W., & Westermann, F. (2001). Two Mezzogiornos. NBER Working Paper Series, 8125.
- Tatsiramos, K., & van Ours, J. C. (2014). Labor market effects of unemployment insurance design. *Journal of Economic Surveys*, 28(2), 284–311.

# > APPPENDIX 1: Technical Notes

# Non-critical assumptions:

**Assumption A1.** Labour is homogeneous. This supposition entails that the wage, denoted by w, is identical for all workers.

**Assumption A2.** Labour contracts last one period. To sign a new contract, an individual needs to spend a fixed amount of time in jobsearch activities, as specified in the next assumption.

**Assumption A3.** Before signing a contract, the worker has to devote s units of time to job-search. Here, s is considered to be a fixed and exogenous sum of time.  $^{17}$ 

**Assumption A4.** The size of the working week, which we denote by  $\bar{l}$ , is fixed and exogenously determined.<sup>18</sup>

**Assumption A5.** The utility function is additive. In other words, if we denote the consumption (or the total income because there is no saving) by C and the leisure time (i.e., total time minus hours of work) by H, this assumption establishes that  $U(C,H) = \Lambda(C) + \Omega(H)$ . As usual, marginal utilities are supposed to be positive and decreasing. 19

# The aggregation process:

If workers have different preferences over consumption and leisure and different non-labour incomes, they will also have different reservation wages. This variety of reservation wages  $w^R \in [0, +\infty)$  might be represented by a cumulative distribution function  $\Phi(w|Z)$ , with Z being the rest of the PR determinants. If those determinants do not change, the aggregate labour supply could be expressed in formal terms according to (A1):

$$L = N \cdot \Phi(w) \tag{A1}$$

where L stands for the labour force, and N stands for total working age population. Therefore, the PR is simply  $\Phi(w)$ , as expressed in equation (A2):

<sup>18</sup> As we are interested in the extensive margin of the labor supply, this assumption allows us to focus on the participation decision.

 $<sup>^{17}</sup>$  Considering s as an endogenous variable is out the scope of the paper. That is the subject of the job-search theory. See Tatsiramos and van Ours (2014).

<sup>&</sup>lt;sup>19</sup> This assumption is less restricting than it seems at first glance. Firstly, it is well known that this sort of utility function generates indifference curves that, typically, decrease and are convex to the origin. Secondly, within the ordinal utility theory, a logarithmic transformation of the very well-known Cobb–Douglas utility function is additive, representing an identical set of preferences.

$$PR = \frac{L}{N} = \Phi(w) = \int_0^w \phi(v) dv \tag{A2}$$

Inasmuch as  $\Phi(w)$  is a cumulative distribution function, by definition, that proportion is increasing in its argument,  $\Phi_w = \phi > 0$ , i.e. the density function is positive. Nevertheless, in order to study the role of the AUB, it is necessary to analyse the influence of other determinants on PR. To incorporate this idea, let us call  $w_M^R$  the reservation wage for the median individual within the cumulative distribution. In this way, a stylized PR function can be described by means of expression (A3):

$$PR = \Phi(w, w_M^R) \tag{A3}$$

As pointed out before,  $(\partial PR/\partial w) > 0$ , by definition. Furthermore, consistently with the concept of reservation wage, we have that  $(\partial PR/\partial w_M^R) < 0$ . Finally, it is worth bearing in mind that  $w_M^R$  is, in turn, a function of some additional arguments. Due to the objective of this paper, we emphasize the dependence of  $w_M^R$  on b and p. In addition, we must point out that both b(r) and p(r) are regarded as functions of the AUB. Thus, we may rewrite expression (A3) as follows:

$$PR(r) = \Phi(w, w_M^R[b(r), p(r)]) \tag{A4}$$

Equation (A4) reveals that PR depends on the AUB through a double channel. The first channel operates via the likelihood of finding a job, which affects not only those eligible to collect the AUB, but also the non-eligible population. Second, the channel working directly through the AUB, b, which modifies the behaviour of the eligible population.

# > APPENDIX 2: Figures and Tables

FIGURE A2-1. Set of alternatives regarding labour participation

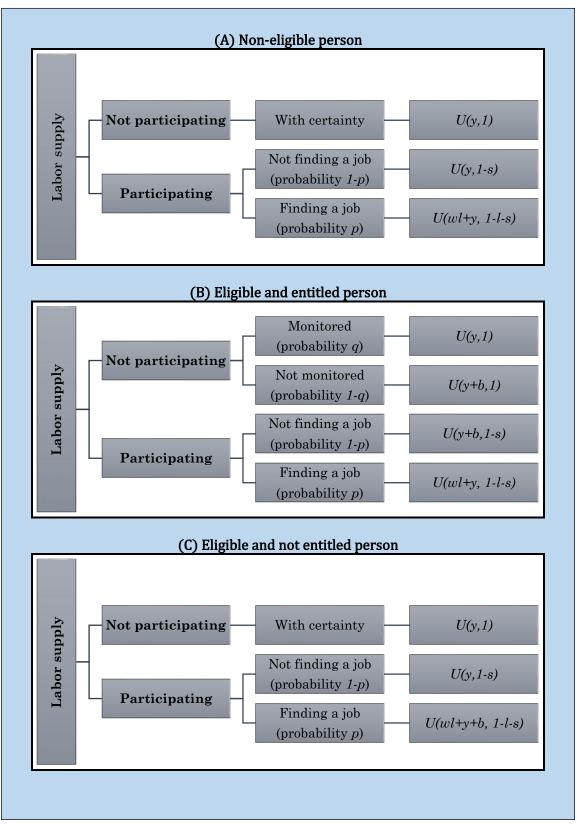
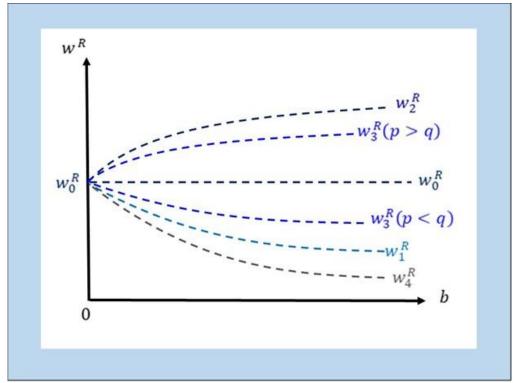


FIGURE A2-2. Reservation wages as a function of "b"



Source: Own elaboration.

FIGURE A2-3. Reservation wages as a function of q

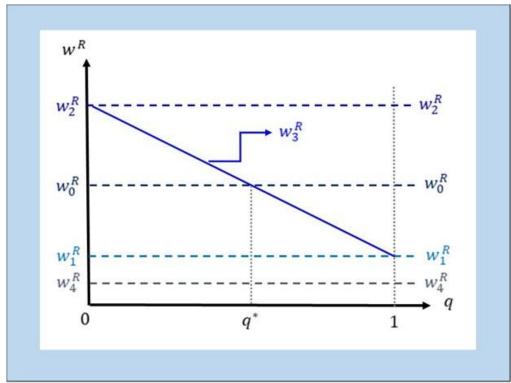


TABLE A2-1. Different specifications

Predictors	Specification							
Predictors	[S1]	[S2]	[83]	[S4]	[85]	[86]	[87]	[88]
Males in active population (%)	✓	✓	✓	✓	✓	✓	✓	✓
Actives aged 25-54 years over total actives (%)	✓	✓	✓	✓	✓	✓	✓	✓
Agricultural employment over total employment (%)	✓	✓	✓	✓	✓	✓	✓	✓
Construction employment over total employment (%)	✓	✓	✓	✓	✓	✓	✓	✓
Long-term unemployment (1-2 years) (%)	✓	✓	✓	✓	✓	✓	✓	✓
Very long-term unemployment (> 2 years) (%)	✓	✓	✓	✓	✓	✓	✓	✓
Activity rate (1981q2)	_	✓	_	_	✓	✓	_	✓
Activity rate (1982q2)	_	_	✓	_	✓	_	✓	✓
Activity rate (1983q2)	_	_	_	✓	_	✓	✓	✓
RMPSE	0.268	0.811	0.258	0.816	0.307	0.334	0.509	0.292

TABLE A2-2. Weights in the synthetic Andalucía (W\*)

G	Composition of the donor pool (synthetic Andalucía)									
Spanish regions	[S1]	[S2]	[S3]	[84]	[85]	[S6]	[87]	[88]		
Aragon	0	0	0	0	0	0	0	0		
Asturias	0	0	0	0	0	0	0	0		
Balearic Islands	0	0	0	0	0	0	0	0		
Basque Country	0	0	0	0	0	0	0	0		
Canary Islands	0	0	0	0	0	0	0	0		
Cantabria	0	0	0.019	0.010	0.050	0.060	0.107	0.044		
Castile and Leon	0	0	0	0	0	0	0	0		
Castilla-La Mancha	1	0.706	0.981	0.692	0.950	0.940	0.893	0.956		
Catalonia	0	0	0	0	0	0	0	0		
Comunidad de Madrid	0	0.019	0	0	0	0	0	0		
Extremadura*	_	_	_	_	_	_	_	_		
Galicia	0	0	0	0	0	0	0	0		
La Rioja	0	0	0	0	0	0	0	0		
Navarre	0	0	0	0	0	0	0	0		
Region of Murcia	0	0.275	0	0.298	0	0	0	0		
Valencian Community	0	0	0	0	0	0	0	0		
RMSPE	0.268	0.811	0.258	0.816	0.307	0.334	0.509	0.292		

Notes: (1) (\*) Conflicting region excluded. (2) The autonomous cities of Ceuta and Melilla have not been included in the analysis.