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Research Article
Impact of COVID-19 on the health of the general and more vulnerable population and its determinants: Health care and social survey-ESSOC, study protocol

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Abstract: Background: This manuscript describes the rationale and protocol of a real-world data (RWD) study entitled Health Care and Social Survey (ESSOC, Encuesta Sanitaria y Social). The study’s objective is to determine the magnitude, characteristics, and evolution of the COVID-19 impact on overall health as well as the socioeconomic, psychosocial, behavioural, occupational, environmental, and clinical determinants of both the general and more vulnerable population. Methods: The study integrates observational data collected through a survey using a probabilistic, overlapping panel design, and data from clinical, epidemiological, demographic, and environmental registries. The data will be analysed using advanced statistical, sampling, and machine learning techniques. The study is based on several measurements obtained from three random samples of the Andalusian (Spain) population: general population aged 16 years and over, residents of disadvantaged areas, and people over the age of 55. Discussion: Given the current characteristics of this pandemic and its future repercussions, this project will generate relevant information on a regular basis, commencing from the beginning of the State of Alarm. It will also establish institutional alliances of great social value, explore and apply powerful and novel methodologies, and produce large, integrated, high-quality and open-access databases. The information described here will be vital for health systems in order to design tailor-made interventions aimed at improving the health care, health, and quality of life of the populations most affected by the COVID-19 pandemic.
Abstract (en español): Introducción: Este artículo describe la justificación y el protocolo de un estudio de datos del mundo real (RWD) titulado Encuesta Sanitaria y Social (ESSOC). El objetivo del estudio es determinar la magnitud, características y evolución del impacto del COVID-19 en la salud en general, así como los determinantes socioeconómicos, psicosociales, conductuales, ocupacionales, ambientales y clínicos tanto de la población general como en la más vulnerable. Metodología: El estudio integra datos de observación recopilados a través de una encuesta utilizando un diseño probabilístico de panel superpuesto y datos de registros clínicos, epidemiológicos, demográficos y ambientales. Los datos se analizarán mediante técnicas avanzadas de estadística, muestreo y aprendizaje automático. El estudio se basa en varias medidas obtenidas a partir de tres muestras aleatorias de la población andaluza (España): población general de 16 años y más, residentes en zonas desfavorecidas y personas mayores de 55 años. Discusión: Dadas las características actuales de esta pandemia y su futuras repercusiones, este proyecto generará información relevante de forma periódica, a partir del inicio del Estado de Alarma. También establecerá alianzas institucionales de gran valor social, explorará y aplicará metodologías poderosas y novedosas, y producirá bases de datos grandes, integradas, de alta calidad y de acceso abierto. La información aquí descrita será vital para los sistemas de salud con el fin de diseñar intervenciones a medida dirigidas a mejorar la atención médica, la salud y la calidad de vida de las poblaciones más afectadas por la pandemia COVID-19.

Keywords: public health, health determinants, health inequalities, COVID-19, SARS-CoV-2, vulnerable populations, real-world data, surveys, population registries, machine learning

1. Background

1.1. Introduction

Currently underway are a large number of studies investigating the evolution of the 2019 coronavirus disease (COVID-19) and the impact it is having on the number of infected patients, hospital admissions, and deaths[1–6], as well as on the mental health and well-being of the population[7]. Nevertheless, very few reports providing information concerning clinical, contextual and citizens’ general perceptions pertaining to the pandemic’s impact and evolution since its onset are being prepared[8–10].

The Spanish Government officially declared a State of Alarm on 14 March, 2020 (Spanish Royal Decree 463/2020[11]) in the face of the global public health emergency caused by COVID-19. Among other actions, it ordered individuals’ freedom of movement to be limited (Article 7, “Limitación de la libertad de circulación de las personas”), which was subsequently further restricted through other decrees (hereinafter referred to as confinement). These limitations have led to a whole series of - still little-studied - problems in the population. In this sense, recent reviews concerning epidemic outbreaks and
subsequent confinements and have concluded that these actions have very negative and long-term impacts on mental health[12,13]. Likewise, this current pandemic has also seen such negative effects, albeit in the short-term[14–17]. Furthermore, all studies agree on the urgent need for more evidence regarding this kind of impact, especially evidence gathered from among the most exposed populations as well as those populations in a situation of greater vulnerability[18]. In this respect, a study on the Roma population observed that 14% found it difficult to follow the measures aimed at preventing the disease and 7% experienced difficulty accessing medicines[19]. As a result, a deterioration in the social and educational gap was observed in this population, along with them experiencing greater stigmatization owing to certain lifestyle habits[20]. Likewise, a study carried out with minors (infants and adolescents) in foster care revealed a decrease in their quality of life during the pandemic compared with the year 2017[21], while another study involving vulnerable populations demonstrated the existence of a link between experiencing anxiety and the perceived risk of contracting the virus[8].

As for the impact of COVID-19 on the migrant population and ethnic minorities, the results of different studies have revealed a higher incidence of the disease in these population groups in relation to deficient conditions in terms of socioeconomic deprivation, comorbidities, unemployment, and livability[22–24]. As in the case of the Roma population, this situation may be aggravated in the migrant population as they have poorer access to health care because of the deterioration of their socioeconomic and administrative conditions or their difficulty understanding the prevention and health care information provided due to language limitations[25,26].

Regarding the effects of pandemics on the health of people with chronic diseases, and also in relation to the above, the results of some studies have highlighted the need to adapt health care to the confinement situation by introducing measures such as remote consultations[27,28]. Furthermore, these studies revealed that throughout the confinement period the symptoms of some diseases, such as Alzheimer’s[29] or diabetes[30], were exacerbated. That said, this worsening of symptoms was not as severe for patients with chronic obstructive pulmonary disease (COPD)[31]. It is well known that health care can be affected during health emergencies and pandemics as a result of resources being diverted towards more urgent areas and, for example, the care of patients with chronic diseases being restricted and/or surgical procedures being delayed[32]. However, the evidence available indicates that ensuring proper follow-up of patients with chronic diseases should be one of the main goals when managing the COVID-19 pandemic because, if infected, they are at risk of experiencing greater severity[33,34].

With respect to the health impact of COVID-19 in Spain, several studies have been launched, most of which are using transversal designs, non-probabilistic samples, and web surveys[35,36]. These studies represent an agile and simple alternative for collecting a large amount of data quickly and
offering results practically in real time; something which is crucial to responding to situations like the one currently being experienced. However, these surveys have a number of disadvantages and limitations. For instance, they do not allow for the impact on and evolution in certain populations in more vulnerable positions to be determined, nor do they allow valid estimates to be obtained which would, in turn, allow the level of error to be limited to the appropriate standards to be conclusive. The reason for this is because they are not usually processed with a view to solving problems resulting from a lack of coverage and response, or from selection bias in the sample[37].

At a national level, the National Epidemiology Center (CNE, Centro Nacional de Epidemiología) launched the National Seroepidemiology Study (Estudio Nacional de Seroepidemiología) based on the data obtained through a population-based macro-survey that provides estimates down to the provincial level, continuous information on the evolution of the epidemic, and information obtained from patients’ medical records[38,39]. Based on its preliminary results, the study has estimated the prevalence of immunoglobulin G (IgG), antibodies against the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) to be at 9.9% (95% confidence interval [CI]: 9.4%–10.4%) in Spain, with a very similar prevalence among men and women. In addition, while the coastal areas have prevalences close to or lower than 4%, the central core surrounding Madrid has figures close to or above 15%, with the highest prevalence (>16%) being detected among health care personnel and women caring for dependent persons[40].

COVID-19 has had the greatest impact on people over 70 years of age and who account for 38% of confirmed COVID-19 cases in Spain, almost half of the hospitalizations due to the virus, and more than 85% of the deaths[41]. Another study has shown that fatality from COVID-19 among individuals over the age of 80 ranges between 12% and 16% in men and 4.5% and 6.5% in women[39]. Internationally, it has been confirmed that the elderly are at greater risk of hospitalization[42] and developing severe symptoms[43], with people over 60 years of age and those with underlying conditions such as hypertension, diabetes, cardiovascular diseases, chronic respiratory diseases, or immunosuppression, being at higher risk of severe infection and death[44]. Smokers and ex-smokers over the age of 60 are also at greater risk and more susceptible to becoming infected by SARS-CoV-2[45]. These extreme pandemic-related incidence and mortality rates among the elderly population are further aggravated by its psychosocial impact[46], with a higher risk of depressive feelings among these individuals who have mostly had little physical contact with other people during the confinement period[47,48]. This is a clear reflection of the tragic COVID-19 situation that is currently being experienced, and may be even harsher in deprived socioeconomic contexts[9], although these results must be further explored and there is no evidence in this respect in our context. Furthermore,
the situation caused by the pandemic means that coordination between social and health care services, together with the provision of resources, are even more necessary and urgent than ever[49].

1.2. The ESSOC Study Framework

The Health Care and Social Survey (ESSOC, Encuesta Sanitaria y SOCial) research project arises from the need to provide specific, reliable, early, and timely data on the impact of COVID-19, which can then be considered when making decisions to prepare and provide an effective Public Health response in the different populations affected, especially the most vulnerable such as, among others, the elderly, the chronically ill, or those at risk of exclusion. ESSOC focuses on Public Health[50] and relies on the participation of society to obtain information on people’s health and their quality of life in order to be able to intervene both individually and collectively in the face of the pandemic (Table 1). The research project is based on a real-world data[51] design that integrates observational data extracted from multiple sources of differing natures and perspectives, i.e., both probabilistic and administrative. Thus, it will link quality information from differing sources based on longitudinal probabilistic population surveys, (which provide data on the economic situation, state of health and well-being from the perspective of citizens), and sources based on clinical, epidemiological, demographic and environmental records, (i.e., data from an administrative perspective). In addition, the data analyses will be performed using advanced statistical, sampling and machine learning techniques, which allow for new research methods to be developed and implemented[52]. The research follows an Open Science approach[53] in terms of disseminating its results, methodologies, processes, and collected data, which will be distributed, reused, and freely and openly accessible to not only the scientific, academic, clinical, and public health managerial community, but also to society at large and, in particular, those population groups identified as being at greater risk of vulnerability to COVID-19. Its management model is that of collaborative and multidisciplinary research, facilitating the creation of a context of scientific cooperation in the fields of public health, health care, public administration, data science, environmental and demographic sciences, and social sciences (Table 1). Finally, the study region, Andalusia, is the most populated (8.5 million inhabitants) and the second largest in area of the 19 regions in Spain. It is also the fifth most populated region in Europe, and it is as populated as other European countries such as Austria or Switzerland. The ESSOC research project is included in the Oxford Supertracker global directory of policy trackers and surveys related to COVID-19[1].
1.3. Hypotheses

- Perceptions of general health, mental health, and emotional well-being have deteriorated in the short- and mid-term since the beginning of the pandemic, with a greater impact being observed in women, young people, and those diagnosed with COVID-19.
- The socioeconomic, psychosocial, behavioural, occupational, environmental, and clinical determinants of health have deteriorated since the onset of the pandemic.
- Health inequalities have increased along the axes of social class, gender, age, ethnicity, and territory as a result of COVID-19, and have been even greater in the mid-term compared with the short-term.
- Chronicity and resulting disability have increased significantly since the beginning of the pandemic.
- Since the beginning of the pandemic, the care burden has increased significantly for women in the short-term and this has had a highly negative impact on their health and well-being.
- Social and emotional support in the population aged over 55 years has decreased significantly since the beginning of the pandemic, with the greatest differences being observed in single-person households in urban areas.

1.4. Objectives

General:

To determine the magnitude, characteristics, and evolution of the impact of COVID-19 on overall health and its socioeconomic, psychosocial, behavioural, occupational, environmental, and clinical determinants in the general population and that with greater socioeconomic vulnerability.
Specific:
1. To determine the short- and mid-term impact of the COVID-19 pandemic on the health and emotional well-being of the general population of Andalusia.
2. To analyse the evolution of the socioeconomic, psychosocial, behavioural, clinical, and environmental determinants of health in the context of the COVID-19 pandemic in the population under study.
3. To identify health inequalities along the axes of social class, gender, age, ethnicity, and territory and their evolution in the context of the COVID-19 pandemic.
4. To evaluate different research sampling techniques to improve the reliability and precision of estimates obtained through surveys using longitudinal designs.

To compile and systematize the existing evidence regarding the design, sources, methodologies, and topics related to measuring the impact of COVID-19 on health and its determinants through surveys.

2. Materials and Methods

2.1. Study Design

This study employs real-world data design to integrate observational data extracted from multiple sources, including information obtained from different providers based on surveys and clinical, epidemiological, population, and environmental registries. The surveys have an overlapping panel design to ensure there are both cross-sectional and longitudinal estimates [54] and to include population-based probability samples carried out via telephone interviews.

2.2. Geographical, Population, and Temporal Scopes

The geographical scope is the Autonomous Region of Andalusia (Comunidad Autónoma de Andalucía), Spain, and the population scopes are the general population over the age of 16 (ESSOCgeneral), the population residing in disadvantaged areas (ESSOCzones)[55], and the population over the age of 55 (ESSOC+55). Collective households (i.e., hospitals, nursing homes, barracks, etc.) are not considered in this study. That said, the study sample includes families who, as an independent group, reside in these collective establishments (e.g., a director or janitor of a centre). The temporal scopes of each sample (Figure 1) are:

- ESSOCgeneral: five measurements taken between 2020 and 2023, at baseline (beginning of the Spanish State of Alarm), at one month from the first interview, at six months, at 12 months, and at 36 months.
- ESSOCzones: two measurements, taken at baseline (12 months from the beginning of the State of Alarm) and at 24 months from the first interview.

- ESSOC+55: two measurements, taken at baseline (six months from the beginning of the State of Alarm) and at 30 months from the first interview.

![Figure 1. Health Care and Social Survey (ESSOC): Population and Temporal Scope, and Samples](image)

2.3. Sampling Frame

The sampling frame used to extract the ESSOCgeneral and ESSOCzones samples is obtained from the Longitudinal Andalusian Population Database (BDLPA, Base Longitudinal de Datos de Población de Andalucía)[56]. The information consolidated in the BDLPA originates from the integration of information on stocks, flows, and variations extracted from the census coordination system which, together with the data obtained from the Civil Registries with respect to births, deaths, and marriages (i.e., vital statistics [MNP, Movimiento Natural de la Población]), as well as that reported in the population and housing censuses, give rise to an integrated longitudinal frame for population and territorial statistics in Andalusia[57]. The sampling frame is extracted from the BDLPA longitudinal
file as a cross-section with the reference date set as 01 January of the most current year possible in each measurement. The selected samples are linked to the information obtained from the User Database (BDU, Base de Datos de Usuarios)[58] of the Andalusian Public Health Care System in order to obtain the telephone numbers of the selected sample units. The BDU coverage in terms of contact telephone numbers for the selected samples is usually above 96%. On the other hand, the ESSOC+55 sampling frame corresponds to the user population of the Andalusian Guadalinfo public network aged 55 years and over [59].

2.4. Sample Size

During the first measurement (M1), the ESSOCgeneral scope started with a random sample comprised of 5,000 people under the assumptions of maximum variability in the estimate (p = q = 0.5), a design effect of 1.8, a precision of 2.4 percentage points for estimates in Andalusia, a confidence level of 95%, and a non-response rate for the theoretical sample of 40%. The subsequent measurements (M2–M5) are comprised of the longitudinal samples of the previous measurements (n_i_theoretical_longitudinal) and, in addition, of a new sample in each measurement (n_i_theoretical_new). That new sample is selected according to the design of the first measurement, except for M5 which will incorporate a new stratum of ‘residing or not in disadvantaged areas’, where DA1 = non-disadvantaged area and DA2 = disadvantaged area[60]. For the previous measurements (M1, M2, M3 and M4), a post-stratification will be carried out according to disadvantaged areas in order to improve the estimates. Finally, due to non-response, from these two theoretical samples (longitudinal+new) we obtain the total effective sample for these measurements (n_i_effective = n_i_effective_longitudinal + n_i_effective_new). With respect to the theoretical sample size for the longitudinal sample of i measurement, this is defined by n_i_theoretical_longitudinal = n_i-1_effective_longitudinal + n_i-1_effective_new, i=2…5). The aim is to reach an effective sample of 3,000 units per measurement for ESSOCgeneral, 2,750 for ESSOCzones and at least 2,400 for ESSOC+55, assuming a total response rate of 60%.

Thus, the ESSOC is made up of a series of measurements broken down into a new sample and a longitudinal sample for each measurement which, in turn, are categorized into theoretical and effective samples for each population study group (general, zones and +55), with a total of over 22,000 effective interviews being carried out over three years.

2.5. Sample Allocation

Allocation of the new samples (including the first sample) will be mixed. On the one hand, they will be uniform by province (150 sample units for eight provinces) and, on the other hand, proportional to
the population size of each province and degree of urbanization (urban, intermediate-density, and rural area; Table 2)[61]. In addition, for the M5 measurement, the distribution of the new sample will be performed, in the first place, in the two DA strata and, subsequently, as in the case of the previous measurements.

Table 2. ESSOC: Sample allocation of the theoretical sample per province and degree of urbanization (percentages).

| Province | Urban area | Intermediate-density area | Rural area | Total |
|----------|------------|---------------------------|------------|-------|
| Almería  | 2.8%       | 4.3%                      | 1.3%       | 8.4%  |
| Cádiz    | 10.1%      | 4.0%                      | 0.6%       | 14.7% |
| Córdoba  | 3.8%       | 4.1%                      | 1.4%       | 9.3%  |
| Granada  | 3.9%       | 4.7%                      | 2.4%       | 11.0% |
| Huelva   | 1.7%       | 3.4%                      | 1.2%       | 6.2%  |
| Jaén     | 2.0%       | 3.8%                      | 1.8%       | 7.6%  |
| Málaga   | 12.1%      | 5.9%                      | 2.0%       | 20.0% |
| Seville  | 12.9%      | 8.7%                      | 1.3%       | 22.8% |
| Total    | 49.3%      | 38.7%                     | 11.9%      | 100.0%|

2.6. Sample Selection

The selection of the new theoretical sample in each measurement is carried out in a simple random manner within each province and degree of urbanization stratum and, in the case of measurement M5, within each DA strata, thus obtaining self-weighted samples in each stratum. The theoretical longitudinal sample of a measurement is composed of the effective sample of the previous measurement, except for measurement M1 which, being the first one, does not have a longitudinal sample.

In the ESSOC+55 sample, the first measurement was stratified by clusters (Guadalinfo Centers, N = 651), with sub-sampling to 1,200 users. These centres are stratified per Andalusian province and inhabitation level (<10,000 inhabitants, 10,000–19,999, ≥20,000), as well as sex and age quotas (55–64 years, 65–74 years, and >75 years). As in the case of the ESSOC general sample, the ESSOC+55 second measurement will be made up of a longitudinal sample (that of the first measurement) and another new sample until a total of 1,200 interviews are completed.

2.7. Fieldwork

The survey information is collected through a computer-assisted telephone interview (CATI). The management, control, and monitoring of the collection of information for measurement M1 is carried out through Pl@teA, the IECA’s survey collection platform, whereas for the rest of measurements these tasks are performed using the Mobinet-Gandia Integra software. The data collection is carried out by
a team of between eight (in M1) and 12 (rest of ESSOCgeneral measures) interviewers assigned solely to this study. This ensures working team stability, which is of fundamental importance in regular, longitudinal surveys. Before starting the study, the interviewers will receive the necessary training regarding the content of the survey. To this end, in addition to virtual meetings held before starting the fieldwork, they are provided with interviewer’s and questionnaire manuals which, besides explaining the questionnaire and study’s content, describe the survey platform, possible incidences, and the protocol to be followed in each case to guarantee the maximum quality of the samples and the information collected. Prior to engaging in the fieldwork, each interviewer performs several pilot tests to measure times and determine the complicated points of the questionnaire.

The schedule set to conduct the surveys is from Monday to Friday from 10:00 to 21:00, and on Saturdays from 10:00 to 15:00 for the first measurement and for the rest of the measurements from Monday to Saturday from 14:00 to 21:00, although deferred appointments can be scheduled without time limits. Furthermore, a telephone line with a 900 prefix and staffed by telephone agents is made available to the public. This line is provided to the survey holders via text message or through the CSyF website, where the characteristics of this study are also published. This call centre also receives calls from people who, having been contacted by CATI agents, need to confirm the official nature of the survey. In fact, many of these calls become completed interviews.

2.8. Quality Control

For the ESSOCgeneral, quality control measures mean the interviews are cross-checked both internally in the Call Centre itself and in the IECA and EASP. Each interviewer is monitored to ensure that they follow the established protocols and that they use each type of incidence correctly. The intervals elapsed between each call and their duration are also monitored.

In addition to recording the calls made by each interviewer, a listening check is also performed to review both the positive aspects and those to be reinforced in the supervised surveys (i.e., for 10% of the calls performed for the first measurement and 25% of the calls performed for the rest of the measurements). During these checks, aspects, such as the interviewer’s self-presentation, their presentation of the study, the offer of being able to call the 900-prefix telephone number, the confirmation of the place of residence, the correct delivery of the questionnaire’s questions and all response options, are assessed.

Quality control, data cleansing, and data coding are carried out simultaneously with the fieldwork with the aid of the software to be used in the study. Each interviewer is provided with a space on the platform to record observations during the survey being conducted. This then allows the supervision team to cleanse those interviews in which the interviewer detected an inconsistency in the
respondent’s answers, or those in which the interviewer made a mistake when completing the questionnaire. Likewise, the values of the variables are revised, and invalid ones cleansed. Moreover, the coding of the variables corresponding to open-ended questions, such as the respondent’s occupation and educational level, is carried out in tandem with the fieldwork. In the rest of open-ended questions, prior to their coding, their possible answers are cleansed, and the categories deemed to correspond to the majority subsequently coded.

During the telephone surveys, different situations may arise that could result in the inability to complete the survey. This is known as field incidences (Table 3), with the most important types being final incidences, i.e., those that, after several attempts, finally result in the inability to complete the survey.

### Table 3. ESSOC: Interview incidences and protocol to be followed:

| Incidence                                                                 | Description                                                                 | Protocol                        |
|--------------------------------------------------------------------------|-----------------------------------------------------------------------------|---------------------------------|
| Frame incidence (reasons that make it impossible to complete the survey due to problems related to the sampling frame; for example, a telephone number with which to contact the sample person could not be obtained or the housing frame was not sufficiently up to date) | The telephone number does not exist                                         | Direct removal                  |
|                                                                         | Wrong number: the telephone number dialled does not exist, corresponds to a fax, or has restricted calls. |                                 |
|                                                                         | Out-of-date frame: a selected person is living in a different municipality, a telephone frame without a telephone number, a person unreachable through the telephone number/home address provided due to circumstances such as death, divorce/separation, etc. | Direct removal                  |
| No contact                                                               | The household cannot be contacted (e.g., nobody answers the telephone, or the answering machine goes off) | Removal after four attempts performed on two different days, at two different times |
| Absent                                                                   | The selected person cannot be contacted                                      | Removal after four attempts performed on two different days and at two different times |
| Inability to answer                                                      | The selected person cannot complete the survey due to an inability to respond to it because of disability, age, illness, lack of knowledge of the language, or any other circumstance. If possible, the survey should be completed by a close relative. | Direct removal                  |
| Refusal                                                                  | The selected person refuses to complete the survey or refuses to continue it after it has begun. | Direct removal                  |
As this is a longitudinal study, one of the most significant reasons for a lack of response is the potential interviewee identifies the incoming call number and does not answer the phone. To solve this, as much as possible, the telephone number from which each call is made is changed periodically, so that even if a number were to identified and blocked, we could continue to attempt to contact that person by employing a new telephone number.

In addition to the quality of the sample, there are other factors of interest in assessing how fieldwork has developed during a surveying operation. One such factor is to determine how a survey has been carried out in terms of effectiveness and efficiency. The most direct way to measure this is to calculate the number of attempts or calls that had to be made in order to complete each survey. This type of information is also very useful to be able to design strategies aimed at optimizing attempts and, therefore, increasing sample levels in future operations.

2.9. Sampling Weights

The original sampling weight for the new samples is obtained from the inverse values of the effective sampling fractions in each stratum and used to calculate the Hajek estimator[62]. This is subsequently calibrated to obtain more reliable estimates based on the demographic characteristics of Andalusia. To this end, we use a truncated linear calibration method[63], and, as auxiliary information, the marginals of the Andalusian population per sex and age (16–19, 20–24, 25–29,...,75–79, and ≥80 years old), province and nationality (Spanish or dual nationality and foreign), having obtained these data from the Continuous Household Survey (Encuesta Continua de Hogares)[64].

Regarding the non-response bias in longitudinal samples, we can predict non-contact and non-cooperation based on auxiliary information and information already known about the sample subjects. Thus, the original weights used in the estimates of longitudinal sample $M_t$ are corrected during a first phase by modelling the non-response with respect to the longitudinal effective sample obtained in $M_{t-1}$ using machine learning techniques[65]. Said non-response is estimated using a XGBoost model[66], which represents the state-of-the-art in machine learning. Every piece of data and variable from the $M_{t-1}$ sample is used for training, thus the algorithm has all the information available in order to learn. Likewise, the hyperparameters of the model are optimized using cross-validation to ensure reliable estimations. Then, during a second phase, they are calibrated following the method described for the new samples. As auxiliary variables, we use those extracted from the BDLPA (e.g., nationality, age, etc.) and the registers from the ESSOC itself in $M_{t-1}$ (Table 4).
Table 4. ESSOC: Adjustment of the design sampling weight in each measurement

| Sample type (effective) | Type of adjustment 1st phase | 2nd phase |
|-------------------------|-----------------------------|-----------|
| New                     | Non-response adjusted by proxy based on the effective sample size in each stratum. | Representativeness by truncated linear calibration with 0.1 and 10 limits based on the auxiliary variables |
| Longitudinal            | Non-response adjusted using an XGBoost model based on variables from the previous measurement. | |

In addition to these adjustments, other methodological alternatives, not yet explored in this type of sample design, for instance double calibration, will also be investigated by considering different variables in order to model non-responses and, on the other hand, correction of the representativeness bias[67,68] and machine learning techniques, and adjusting non-responses with the aid of the Propensity Score Adjustment (PSA)[69,70].

2.10. Variables

The study variables will mainly be extracted from the following sources: BDLPA; the Andalusian Population Health Database (BPS, Base Poblacional de Salud)[71]; the Andalusian Environmental Information Network (REDIAM)[72]; the Andalusian Epidemiological Surveillance System (SVEA)[73] and the ESSOCgeneral, ESSOCzones, and ESSOC+55 surveys.

The personal data of the participants selected for the interview (name, surname, and telephone number) are extracted from the BDLPA. In addition, the BDLPA is linked annually with a repository of georeferenced buildings so that the postal address and coordinates (250m x 250m grid) in the territory can be extracted. This will allow us to extract geographical factors (urbanization degree and population density, among others) via other IECA registries, and environmental factors (pollution and temperature, among others) via the REDIAM registry from the Andalusian Regional Government’s Department of Agriculture, Livestock, Fisheries, and Sustainable Development (Consejería de Agricultura, Ganadería, Pesca y Desarrollo Sostenible de la Junta de Andalucía).

From the SVEA registry, epidemiological information related to COVID-19, such as the date and result of the diagnostic test for active infection (PDIA), will be extracted to detect the presence of an active SARS-CoV-2 infection, which includes both reverse transcription–polymerase chain reaction (real time RT – PCR) as the antigen (Ag) rapid test; date of the onset of symptoms; close contact of confirmed case with PDIA; local or imported case; occupation as a health or social health professional; need for hospitalization or admission to an intensive care unit; date of admission and discharge.
In addition, the clinical information related to chronic diseases[74], functional and cognitive assessments, health resources (volume and cost), population stratification, and drugs consumed, which is obtained from the BPS, will also be added to the valid samples (Table 5). Further information about the variables and the main features of the abovementioned registers can be found in Supplementary Material 3.

Table 5. ESSOC: Auxiliary sources and variables

| Registry | Description | Information | Variables extracted |
|----------|-------------|-------------|---------------------|
| BDLPA – Longitudinal Andalusian Population Database[71] | Information from the census coordination system and civil registries that give rise to a consolidated framework of the Andalusian population | Personal data | Name, surname, identification health number (NUHSA), geographical coordinates |
| BDU – User Database of the Andalusian Public Health Care System[58] | Contact Information of the Andalusian Public Health Care System | Personal contact information | Telephone numbers |
| BPS - Andalusian Population Health Database[71] | Personal health information from the Andalusian Population Health Database and Healthcare information | Health and healthcare information | Chronic diseases, functional and cognitive assessments, health resources (volume and cost), population stratification, drugs consumed |
| REDIAM - Andalusian Environmental Information Network[72] | Daily averages by collecting/meteorological station and at the census section level | Pollution, temperature | Mean daily values from pollution, air quality and temperature |
| SVEA - Andalusian Epidemiological Surveillance System[73] | Functional organization for health surveillance that collects, among other things, epidemiological information related to SARS-COV-2 infection | Epidemiological information of COVID-19 | PCR result, symptoms date, close contact, healthcare professional, hospitalization unit (specifying ICU), date of admission and discharge, need of mechanical ventilation and clinical data |

For further details, see Supplementary Material 3

With regard to the surveys, each measurement is associated with a questionnaire that coincides, to a significant extent, with previous measurements to enable an analysis of its evolution, and incorporate new information to analyse specific characteristics present at each moment of the pandemic. Repeating unchanging information in subsequent measurements is avoided in the case of longitudinal samples. The questionnaire used for each measurement is organized into blocks of information, as shown in Table 6.
### Table 6. ESSOC: Information blocks and variables entered in each measurement.

| Subject area                                      | 1<sup>st</sup> measurement (M1)                                                                 | 2<sup>nd</sup> measurement (M2)                                                                 | 3<sup>rd</sup> measurement (M3)                                                                 | 4<sup>th</sup> and 5<sup>th</sup> measurements (M4 and M5)                                                                 |
|--------------------------------------------------|-----------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------|------------------------------------------------------------------------------------------------------------------|
| Household and housing characteristics            | Municipality, usual household, type of household, surface area, facilities, household changes, number of cohabitants (<6/<16/>60), type of household, and equipment. | Municipality, usual household, type of household, surface area, facilities, household changes, number of cohabitants (<6/<16/>60), equipment, number of rooms, and number of inhabitants with disabilities or requiring care. | Municipality, usual household, type of household, surface area, household changes, number of cohabitants (<6/<16/>60), number of rooms, and number of inhabitants with disabilities or requiring care. | Municipality, usual household, type of household, surface area, household changes, number of cohabitants (<6/<16/>60), number of rooms, and number of inhabitants with disabilities or requiring care. |
| Time use and cohabitation                        | Household chores, care tasks, daily activities during the confinement period (at home and outside), cohabitation and relationships, and causes for optimism. | COVID-19 diagnosis, severity, diagnosis within the person’s settings, self-perception of general and mental health (current and last year), emotional well-being<sup>a</sup>, difficulty to withstand the confinement, malaise, chronic illness, and change of medication. | COVID-19 diagnosis<sup>b</sup>, severity, diagnostic tests, diagnosis within the person’s settings, self-perception of general and mental health, emotional well-being<sup>b</sup>, cohabitation, difficulty to withstand the confinement, happiness, social and emotional support<sup>b</sup>, malaise<sup>b</sup>, chronic diseases (suffering and limitations), and medication (use and change of use<sup>b</sup>). | COVID-19 diagnosis<sup>b</sup>, severity, diagnostic tests, diagnosis within the person’s settings, self-perception of general and mental health, emotional well-being<sup>b</sup>, happiness, social and emotional support<sup>b</sup>, malaise<sup>b</sup>, chronic diseases (suffering and limitations), and medication (use and change of use<sup>b</sup>). |
| Health and emotional well-being                  | COVID-19 diagnosis, severity, diagnosis within the person’s settings, self-perception of general and mental health (current and last year), emotional well-being<sup>a</sup>, difficulty to withstand the confinement, malaise, chronic illness, and change of medication. | COVID-19 diagnosis, severity, diagnostic tests, diagnosis within the person’s settings, self-perception of general and mental health, emotional well-being<sup>b</sup>, cohabitation, difficulty to withstand the confinement, happiness, social and emotional support<sup>b</sup>, malaise<sup>b</sup>, chronic diseases (suffering and limitations), and medication (use and change of use<sup>b</sup>). | COVID-19 diagnosis<sup>b</sup>, severity, diagnostic tests, diagnosis within the person’s settings, self-perception of general and mental health, emotional well-being<sup>b</sup>, happiness, social and emotional support<sup>b</sup>, malaise<sup>b</sup>, chronic diseases (suffering and limitations), and medication (use and change of use<sup>b</sup>). | COVID-19 diagnosis<sup>b</sup>, severity, diagnostic tests, diagnosis within the person’s settings, self-perception of general and mental health, emotional well-being<sup>b</sup>, happiness, social and emotional support<sup>b</sup>, malaise<sup>b</sup>, chronic diseases (suffering and limitations), and medication (use and change of use<sup>b</sup>). |
| Habits and lifestyle                              | Habit modification (exercising, smoking, alcohol consumption, sleep, and diet).                | Habit modification<sup>b</sup>: exercising, drinking, smoking, sleep, food, daily intake of vegetables and fruit, exercising, weight and height<sup>b</sup>, smoking, alcohol consumption sleep, and flu vaccination. | Habit modification<sup>b</sup>: exercising, drinking, smoking, sleep, food, daily intake of vegetables and fruit, exercising, weight and height<sup>b</sup>, smoking, alcohol consumption sleep, and flu vaccination<sup>b</sup>. | Habit modification<sup>b</sup>: exercising, drinking, smoking, sleep, food, daily intake of vegetables and fruit, exercising, weight and height<sup>b</sup>, smoking, alcohol consumption sleep, and flu vaccination<sup>b</sup>. |
| Economic situation and socio-demographic         | Educational level, employment situation, working from home, type of contract, occupation<sup>a</sup>, cohabitation with a partner, identification of the cohabitant with the greater income (educational level, employment situation, occupation), difficulty in making ends meet, late payments, income, future worries, and degree of confidence in public institutions. | Employment situation, educational level, occupation<sup>a</sup>, development<sup>b</sup>, ability to work, identification of the cohabitant with the greater income (educational level, employment situation, occupation), difficulty in making ends meet, late payments, change in economic situation, parents’ educational level, and future worries. | Employment situation, educational level, occupation<sup>a</sup>, development<sup>b</sup>, ability to work, identification of the cohabitant with the greater income (educational level, employment situation, occupation), difficulty in making ends meet, late payments<sup>b</sup>, change in economic situation, parents’ educational level, and future worries. | Employment situation, educational level, occupation<sup>a</sup>, development<sup>b</sup>, ability to work, identification of the cohabitant with the greater income (educational level, employment situation, occupation), difficulty in making ends meet, late payments<sup>b</sup>, change in economic situation, parents’ educational level, and future worries. |
2.11. Data Analysis

The analyses take advantage of all the information available from the measurements and the auxiliary information sources and will be carried out with the free software environment R\cite{81}, considering the sample design, as well as the calibration and inference methods described in the previous sections. The use of free software will guarantee transparency and facilitate the replicability of the study.

A table will be prepared for each variable of each measurement, together with the variable’s original response categories, including the valid sample size (n), the percentage of lost samples, the population size estimate (N), the relevant statistic (mean or percentage), the 95% CI, and the coefficient of variation (CV), for both the total and the cross-disaggregation per sex and age (16–29/30–44/45–64/65+), as well as per province and urbanization degree. The sample size is recorded for the total and the categories of the segmentation variable. In the case of cells with CV estimates >20%, the CV will be indicated in a footnote to the table.

The variables shared by all measurements are dichotomized based on the results reported in the previous tables, identifying, in each case, the most convenient category to be highlighted based on the previous tables. In addition, a table describing the specific estimates with their corresponding CV will also be created.

Alternatively, to evaluate the changes in each measurement with respect to the first one, both the population affected by the change and the percentage segmented per demographic and territorial variables will be estimated. The $p$ value will be calculated to evaluate the effect of such change and will be indicated in a footnote to the table using three categories: $p <0.001$, $p <0.05$, and $p <0.1$.

In the case of variables that coincide in each pair of consecutive measurements (M2–M1, M3–M2, and M4–M3), the estimated percentage of the difference between one measurement and the previous one will be calculated in addition to the estimate of the population size and the signalling when the CV is greater than 20% and segmented by the demographic and territorial variables.

To analyse factors associated with variables of a given measurement or variables measuring the change between one measurement and another, we will use multivariate explanatory models adapted to the characteristics and nature of the variables and specified as generalized linear mixed models (GLMM) with a family dependent on the type of dependent variable: Gaussian, when the variable is continuous (equivalent to a linear regression); binomial, when the variable is dichotomous (equivalent to a logistic regression); or Poisson, when the variable is discrete (equivalent to a Poisson regression).
Random effects will be included in these models to capture the effects of unobserved confounders. Inferences will be made following a Bayesian perspective and using the integrated nested Laplace approximation (INLA)[82,83]. We will use penalized complexity priors known as PC priors. These priors are robust in the sense that they do not impact the results, and, in addition, they allow for an epidemiological interpretation[84]. The analyses will be carried out using free software R (version 4.0.4 or greater)[81], through the INLA package[82,83,85].

Finally, advanced data visualizations will be used to allow an in-depth exploratory analysis of the evolution of the study variables and a representation of the main results of the produced models. These visualizations will be developed using Python[86] programming language and integrated into software and web solutions that allow for interaction and dissemination.

2.12. Data Management Plan

The data management plan is provided as Supplementary Material 1. The type and format of data that will be collected and generated within the scope of this project is described in this plan, together with the procedure provided to access data (by whom, how, and when it can be accessed), data ownership, repository to deposit data, and procedure planned to guarantee the specific ethic and legal requirements.

Details of the Data Protection Impact Assessment (DPIA) will also be presented in here in accordance with the specific adaptation of this methodology to research projects in the health care sector[87,88,89]. Thus, the need for a DPIA was confirmed from the outset (Table S1). Subsequently, the data lifecycle was defined (Table S2), and the need and proportionality of the processing were analysed (Table S3) and, finally, a risk assessment and action plan developed (Table S4).

2.13. Scoping Review

The final objective will be achieved through a Scoping Review, whereby the existing evidence on survey-based research related to the measurement of the impact COVID-19 (from the outset of the pandemic) has had on health and its determinants are summarized. The review will be carried out using free terms and controlled language in the databases Pubmed, Scopus, WoS, EMBASE, CINAHL, PsyInfo, LiLac, OpenGray, Gray Literature Report and, likewise, through a free search in Google and institutional websites to locate institutional documents, abstracts, conferences or in any other format where studies and research work can be found. In a first phase, the research work will be selected independently and blindly in pairs, and the study populations, sample size and main objective information will be identified in order to collect the methodological characteristics, including elements on epidemiological design, auxiliary sources of information, methodologies, and topics addressed. This review will be carried out at the beginning of the project and updated throughout its duration,
which will allow the development of the rest of activities, especially the identification of hypotheses and the application of other methodologies to be guided, as well as define the lines of research and propose more appropriate methods for future research.

Finally, with the Open Science approach, the results, microdata, reports, codes, and documentation related to the Scoping Review will be accessible through an open web platform. Likewise, scientific articles will be sent to Open Access journals and organised events will be aimed at both the scientific community and the general population. There will be accessible dissemination of the data on the Internet through institutional websites and social-media channels and a virtual practice community will be created (from the Mendeley citation manager) to share references identified by the research team. The scientific community and anyone else who might be interested will have access to all of the above.

3. Discussion

Given the characteristics, current and future repercussions of the current pandemic, developing this research project will make it possible to periodically obtain relevant information for decision-making processes in social and health matters and, therefore, promote a more efficient, reliable, and responsible science for social change (like the one that we are currently experiencing as a result of the pandemic). In addition, it will encourage:

- institutional alliances of great social value between the public administration, health care services, and the scientific and academic community;
- powerful and novel methodologies in the fields of public health, epidemiology, and sampling to reduce potential biases caused by a lack of coverage, response, or non-randomized selection[1, 50];
- large, integrated, quality, and open databases containing information extracted from clinical and non-clinical population registries; data concerning social, economic, and environmental contexts and the perception of the population, along with foreseeing the future incorporation of genomic information [90];
- the systematic review throughout the entire project of the scientific evidence obtained through this type of study;
- training with a view to transmitting the available knowledge and increasing capacities and skills in designs, sources, and methodologies; and
• measuring the short- and mid-term impact of COVID-19 at different times and on different populations since the beginning of the official State of Alarm.

The limitations of the study include those derived from the coverage and quality of the sampling frames, which may cause selection biases. In the case of the BDLPA, its telephone coverage is over 90% and it tends to have low percentages of non-existent telephone numbers (7%–8%) or non-contactable telephone numbers due to the frame’s outdatedness (9%–10%). That said, any potential biases due to such defects will be corrected through the estimator calibration techniques described above. Another limitation is that caused by the longitudinal design of the ESSOC surveys in terms of panel attrition, which could lead to potential biases due to an absence of response. We estimate the response rate for the full longitudinal sample to be 25%–35% (the effective sample in M5 of the effective sample in M1 is divided by the effective sample in M1). To reduce the effect of potential non-response biases, we decided, on the one hand, to use an overlapping panel design as this type of design allows for completing each measurement with new cross-sectional samples that become longitudinal in subsequent measurements and, on the other hand, to make adjustments in each measurement according to the weights of the longitudinal samples.

In addition, this project may have information biases intrinsic to sampling research. Consequently, we have chosen to employ scales and variables widely used in population-based health surveys and that also allow comparisons to be made for most of the key indicators. Finally, as in most of these types of studies, its feasibility and limitations will, to a large extent, depend on the quality and availability of the data from different data holders. Data availability for this project would appear not to be a limitation, given that the request for the processing of these data is in accordance with the policies on data disclosure for processing data in research projects in the public sector. Additionally, the EIPD performed resulted in an acceptable residual risk level (see Supplementary Material 1). This evaluation highlights, among other elements, that the data provided is protected by statistical confidentiality, that it is not misused, that its treatment is anonymous and global at all times, and that indirect identification is impossible. Furthermore, the project results will be beneficial to the general population in a holistic way, thanks to its socioeconomic and environmental context, and its evolution over several years from the onset of the pandemic. Therefore, the risk to the privacy of the study population is minimal compared to the potential benefit of its results. The transfer of data will be carried out between organizations within the Junta de Andalucía, as this will be within the context of a research project exclusively in the public sphere and under the legitimate use of records as research infrastructures in accordance with the Spanish health legislation (see Supplementary Material 1). With respect to data quality, the reproducibility of this research will be conditioned by the way in which
the data have been obtained, a phenomenon common to all RWD studies. Even when these are the best possible, the fact that the data are extracted from official sources does not guarantee their quality. Because of this, the owners of the information will be asked to describe the detailed procedure used to extract the data, as well as any previous processing it may have undergone. It should be noted that individuals collaborating on this project also belong to the work teams of the main sources of information to be used which, therefore, should guarantee success in the interpretation of the original data and the optimization of the extraction strategies.

4. Conclusions

The ESSOC will enable precise and valid analysis of the short- and mid-term impact of the policies applied, and interventions made to, not only of the health of the general population, but also the most vulnerable population, during the pandemic. The study will also determine the evolution of the socioeconomic, psychosocial, behavioural, occupational, environmental, and clinical determinants of health and identify the inequalities in health in all its axes (i.e., social class, gender, age, ethnicity, and territory).

The conceptual approach of this study will encompass all aspects affecting health and so will contribute to an extraordinary increase in the current knowledge concerning the impact the COVID-19 pandemic is having. This knowledge will, in turn, be crucial for health systems in order to design quick and effective interventions aimed at improving the health care, health, and quality of life of the populations most affected by the COVID-19 pandemic. Moreover, the project management model based on collaborative, multidisciplinary, and open research will allow the critical mass needed to be generated to thus achieve the objectives that have been set (i.e., at the populational level, as well as at an individual and disease level, in our case, COVID-19).

5. Declarations

5.1. Ethics approval and consent to participate

Institutional Review Board Statement: The study was approved by the Research Ethics Committee of the Department of Health and Families of the Andalusian Regional Government (protocol code 10/20, dated 07 December 2020). The ESSOC is an activity included in the processing activities registry of the Department of Health and Families of the Andalusian Regional Government and is linked to the Andalusian Health Survey (EAS, Encuesta Andaluza de Salud), an official statistical operation included in the Andalusian Statistical and Cartographic Plan (Plan Estadístico y Cartográfico de Andalucía), with code 02.02.21.
Informed Consent Statement: An informed consent was obtained from all subjects involved in the study. Further information can be found in the Data Management Plan (Supplementary Material 1).

5.2. Availability of data and materials

The data, source codes, and other documentation developed in the context of this study will be available through the web platform http://researchprojects.es/ and from the open source project management tool https://osf.io/.

5.3. Competing interests

The authors declare no conflict of interest and the funders had no role in the design of the study; the collection, analyses, or interpretation of the data; the writing of the manuscript; or in the decision to publish its results.

5.4. Funding

The funding received to date for this research is partly public, obtained from funds of the Andalusian Institute of Statistics and Cartography (IECA) and the Andalusian School of Public Health (EASP), and partly private, obtained from the competitive calls of the SUPERA COVID19 Fund of Santander Universities (SAUN), the Conference of Spanish University Rectors (CRUE, Conferencia de Rectores de Universidades Españolas), and the Spanish National Research Council (CSIC, Consejo Superior de Investigaciones Científicas), in addition to the COVID-19 Competitive Grant Program from Pfizer Global Medical Grants. The private funding sources did not participate in the design or conduct of the study; the collection, management, analysis, or interpretation of the data; or the preparation, review, or approval of the manuscript.

5.5. Authors’ contributions

Conceptualization, A.C.L.; methodology, A.C.L, C.S.C, M.R, R.F, L.C., M.S. and I.E.; resources, I.E., M.F., R.V., N.L.; data curation, C.S.C., R.F. and L.C.; writing—original draft preparation, A.C.L. and C.S.C.; writing—review and editing, M.R, I.E., M.S., M.F., N.L, M.B., R.F., L.C., A.D.C. and R.V.; project administration, A.C.L. and C.S.C; funding acquisition, A.C.L. and M.S. All authors have read and agreed to the published version of the manuscript.

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5.7. Supplementary Materials

The Data Management Plan is available online at supplementary material 1. Questionnaires are available online at supplementary material 2 (Spanish version). Variables from administrative registries are available online at supplementary material 3 (Spanish version).

5.8. Abbreviations

The following abbreviations are used in this manuscript:

- **BDLPA**: Longitudinal Andalusian Population Database
- **BDU**: User Database from the Andalusian Health Survey
- **BPS**: Andalusian Population Health Database
- **CIBERESP**: Network Biomedical Research Centre of Epidemiology and Public Health
- **CSyF**: Department of Health and Families of the Andalusian Regional Government
- **DPIA**: Data Protection Impact Assessment
- **EASP**: Andalusian School of Public Health
- **ESSOC**: Health Care and Social Survey
- **IECA**: Andalusian Institute of Statistics and Cartography
- **MNP**: Natural Movement of the Population
- **REDIAM**: Andalusian Environmental Information Network
- **RWD**: Real-World Data
- **SVEA**: Andalusian Epidemiological Surveillance System

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