EARLY INTERVENTION IN THE REAL WORLD



Check for updates

WILEY

A public early intervention approach to first-episode psychosis: Treated incidence over 7 years in the Emilia-Romagna region

Correspondence

Martino Belvederi Murri, Institute of Psychiatry, Department of Neuroscience and Rehabilitation, University of Ferrara, Via Fossato di Mortara 64a (44121), Ferrara, Italy. Email: martino.belvederimurri@unife.it

Funding information

Regione Emilia Romagna; Delibera Giunta Regionale

Abstract

Aim: To estimate the treated incidence of individuals with first-episode psychosis (FEP) who contacted the Emilia-Romagna public mental healthcare system (Italy); to examine the variability of incidence and user characteristics across centres and years. Methods: We computed the raw treated incidence in 2013–2019, based on FEP users aged 18–35, seen within or outside the regional program for FEP. We modelled FEP incidence across 10 catchment areas and 7 years using Bayesian Poisson and Negative Binomial Generalized Linear Models of varying complexity. We explored associations between user characteristics, study centre and year comparing variables and socioclinical clusters of subjects.

Results: Thousand three hundred and eighteen individuals were treated for FEP (raw incidence: 25.3 / 100.000 inhabitant year, IQR: 15.3). A Negative Binomial location-scale model with area, population density and year as predictors found that incidence and its variability changed across centres (Bologna: 36.55; 95% Crl: 30.39–43.86; Imola: 3.07; 95% Crl: 1.61–4.99) but did not follow linear temporal trends or density. Centers were

This is an open access article under the terms of the Creative Commons Attribution-NonCommercial-NoDerivs License, which permits use and distribution in any medium, provided the original work is properly cited, the use is non-commercial and no modifications or adaptations are made.

© 2023 The Authors. Early Intervention in Psychiatry published by John Wiley & Sons Australia, Ltd.

¹Institute of Psychiatry, Department of Neuroscience and Rehabilitation, University of Ferrara, Ferrara, Italy

²Department of Psychiatry, Yale School of Medicine, New Haven, Connecticut, USA

³Department of Mental Health and Substance Abuse, AUSL Piacenza, Piacenza, Italy

⁴Department of Mental Health and Pathological Addiction, AUSL Parma, Parma, Italy

⁵Department of Biomedical, Metabolic and Neural Sciences, University of Modena and Reggio Emilia, Modena, Italy

⁶Dipartimento di Salute Mentale e Dipendenze Patologiche, AUSL, Modena, Italy

⁷Department of Mental Health and Pathological Addiction, AUSL Imola, Imola, Italy

⁸Department of Mental Health and Pathological Addiction, AUSL Ravenna, Ravenna, Italy

⁹Department of Mental Health and Pathological Addiction, AUSL Cesena, Cesena, Italy

¹⁰General Directorate of Health and Social Policies, Emilia-Romagna Region, Bologna, Italy

¹¹Department of Medical and Surgical Sciences, Bologna Transcultural Psychosomatic Team (BoTPT), University of Bologna, Bologna, Italy

¹²Department of Mental Health and Pathological Addiction, AUSL Rimini, Rimini, Italy

associated with different user age, gender, migrant status, occupation, living conditions and cluster distribution. Year was associated negatively with HoNOS score (R=-0.09, p < .001), duration of untreated psychosis (R=-0.12, p < .001) and referral type.

Conclusions: The Emilia-Romagna region presents a relatively high but variable incidence of FEP across areas, but not in time. More granular information on social, ethnic and cultural factors may increase the level of explanation and prediction of FEP incidence and characteristics, shedding light on social and healthcare factors influencing FEP.

KEYWORDS

first-episode psychosis, incidence, schizophrenia, untreated psychosis

1 | INTRODUCTION

Psychotic disorders are relatively rare, but have a profound, long-lasting impact on the individual trajectory. Thus, patients with the first episode of psychosis (FEP) need early detection and specific clinical management. Analysing data from a large geographical area where this approach is widely implemented, such as the Emilia-Romagna region, may provide useful information.

When a person shows the first signs of a psychotic disorder, a timely, proactive and appropriate specialist care is crucial to increase the odds of reaching satisfactory outcomes. Ample evidence supports this approach. Early intervention programs improve clinical outcomes and prevent severe psychosocial deterioration to a greater extent than usual care (Correll et al., 2018; Shah et al., 2022). The call for early intervention on psychosis has been received across countries worldwide which provide tailored, community-based, multidisciplinary psychosocial intervention to promote personal and vocational recovery (McDonagh et al., 2022). This approach takes into account that for each individual psychosis is shaped by an idiosyncratic array of factors spanning the familial, peer, work and community social domains (Heinz et al., 2013).

Similar considerations can be made at the population level. Knowing specific population characteristics should make it possible to tailor public mental healthcare for the specific local needs in terms of detection, prevention and management (Kirkbride, 2015). Indeed, the distribution and combination of risk factors for FEP can be extremely variable. Similarly, there is high variability in the incidence, course and outcomes of psychosis across different communities or countries (Jongsma et al., 2018). Not surprisingly, studies have identified a number of associations between patterns of sociodemographic characteristics, substance use and social deprivation, and changes in the incidence, clinical features or outcomes of individuals with FEP (Anglin et al., 2021; Heinz et al., 2013; Jongsma et al., 2021; Leighton et al., 2019; Mascayano et al., 2020; Suvisaari et al., 2018). More recently, studies have begun to translate epidemiological knowledge on risk factors into models that predict the incidence (or features) of psychosis at the individual level, or within specific catchment areas (Lee et al., 2022; McDonald et al., 2021; Suvisaari et al., 2018). For instance, the incidence of psychosis in various areas of England is accurately forecasted using data on age, sex, ethnicity, social deprivation and population density (Kirkbride, 2015). Another model uses individual clinical data and area-level information to predict the outcomes of patients with FEP during treatment (Leighton et al., 2019). Similar instruments may be of crucial help for policy-makers to plan resource allocation (Aceituno et al., 2019; Campion et al., 2019) as well as for clinicians to identify high-risk individuals/populations (Belvederi Murri et al., 2022).

Building actionable epidemiological or risk prediction models for psychosis requires that one identifies what causes the variability of FEP incidence and patient clinical heterogeneity. Examining data from a homogenous healthcare system has the advantage of reduce confounding due to healthcare-related variables (e.g. detection of caseness and pathways to care) and some heterogeneity due to social factors (e.g. cultural and economic characteristics). The first sensible step towards the characterization of local FEP services is represented by the analysis of the heterogeneity of incidence and user characteristics across geographic subareas, or time fluctuations (March et al., 2008; Singh et al., 2000). Then, further investigations may pinpoint the specific contribution of individual risk factors.

The aim of this study was to estimate the treated incidence, and describe the user characteristics of individuals with FEP that had contact with the Emilia-Romagna regional healthcare system. This study makes use of data from the Emilia-Romagna Region Program for First-Episode Psychosis (RER-FEP) that has been implemented since 2012 (Belvederi Murri et al., 2020; Ferrara et al., 2019). The RER-FEP is one of the few large-scale public healthcare initiatives in Italy to address this condition in a coordinated way. In particular, we sought to analyse the variability of incidence across the region catchment areas, over the first 7 years since the establishment of the RER-FEP, including both users who entered the program and those who did not (Ajnakina et al., 2017).

2 | METHODS

2.1 | Setting and organization of the study

The Emilia-Romagna region is one of the largest in Northern Italy (4.5 million residents), with a high density (about 200 inhabitants per square km). The region presents with a varied demographic and

-Wileyli

economic census: a rather old population and steadily increasing immigration fluxes (ISTAT Italian Institute of Statistics, 2020). It is among the top 10 regions for the highest average gross income in Europe (Annuario Statistico Regionale, 2020), and has an overall low prevalence of social deprivation (Regione Emilia Romagna, 2021). However, it ranks high for substance use in Italy. The region comprises nine administrative provinces: Bologna (approximately one million inhabitants), Modena (700 000), Reggio Emilia (530 000), Parma (450 000), Forlì-Cesena (400 000), Ravenna (390 000), Ferrara (350 000), Rimini (340 000) and Piacenza (290 000). They correspond to 11 Local Health Trusts (Forli and Cesena have separate Trusts as well as Imola, a city in the province of Bologna), from which 10 final groupings resulted.

Since the 1978 reform, mental healthcare in Italy is provided by publicly funded Departments of Mental Health (DMH) within the Local Health Trusts. Currently, 45 early intervention services for psychosis are active within the national healthcare system although they are still unevenly distributed in the territory (Cocchi et al., 2018). The Emilia Romagna regional system has been one pioneer delivering evidence-based care for FEP in the Italian public sector. The RER-FEP represents the first region-wide coordinated program to provide routine early intervention for individuals aged 15-35 manifesting a psychiatric condition pertaining the clinical high risk for, or the onset of psychosis (Regione Emilia Romagna, 2016). Various aspects of care for individuals for FEP, and of the RER-FEP program have been previously described for the provinces of Bologna (Tarricone et al., 2012, 2016), Modena (Ferrara et al., 2019), Reggio Emilia (Pelizza et al., 2019), Parma (Leuci et al., 2020) and Ferrara (Belvederi Murri et al., 2020).

The RER-FEP program allows each DMH to tailor the implementation and the delivery of the FEP service (FES) according to the local organization and workforce availability (Regione Romagna, 2016). Each FES is organized either as follows: (i) FEP stand-alone service with full-time specialists dedicated to FEP patients only and geographically distant from the general CMHC, (ii) dedicated staff, but not-exclusive for FEP, or (iii) 'specialist within generalist service', corresponding to FEP-trained professionals within the individual general CMHC (Ferrara et al., 2019). A FES usually includes a minimum of three professional categories: a psychiatrist, a psychotherapist, and a case manager (nurse or rehabilitation therapist). Over the past decade, the composition of each FES and the caseload has varied due to changes in local policies, and staff turnover (Meneghelli et al., 2023).

2.2 | Detection of FEP cases in adults

The region applies a coordinated approach to the detection of adult individuals with FEP (Regione Emilia Romagna, 2016). Guidelines eligibility for the RER-FEP program in adults is based on the following inclusion criteria: aged 18–35 years; Duration of Untreated Psychosis (DUP) shorter than 2 years; the absence of prior treatment with antipsychotic medications, fluency in the Italian language. Using the ICD-9 system, the following diagnostic codes are admissible:

affective psychosis (295.34, 296.24, 296.44, 296.14, 296.54, and 296.64) and non-affective psychosis (295.0–295.95, 299.9, 297, and 298). Subjects are also excluded from the multi-component intervention if they present severe intellectual disability. Subjects who are not eligible for the RER-FEP interventions because of longer DUP, substance-induced psychosis, or intellectual disability are referred to the CMHC for Treatment As Usual (TAU). Patients are also excluded if they lack linguistic fluency, as communication in Italian is necessary to access the cognitive behavioural treatment, a key component of FEP service. However, CMHCs are equipped with translators for routine visits.

Pathways to care to the RER-FEP programme or to TAU include self-referrals, referrals from other services within the DMH, such as psychiatric inpatient units, eating disorders units and addiction units, referrals from the general hospital, particularly the emergency room (ER). General practitioners (GPs) in particular were trained in the recognition of early psychosis and encouraged to refer patients to mental health services, in the context of a liaison programme (Curcetti et al., 2005). In addition, local services such as social service and local police can sometimes be the first who come into contact with people with FEP advising a referral to the ER or to the CMHC.

Ethical approval was obtained by the Area Vasta Emilia Centro Ethical Committee (CE-AVEC); the study conforms to principles expressed in the Declaration of Helsinki. Participants gave their informed consent prior to their inclusion.

2.3 | Data collection

Incidence data was available for 7 years (January 1st, 2013 to December 30th, 2019) at the level of 10 Local Health Trusts: Bologna, Ferrara, Forlì-Cesena, Modena, Parma, Piacenza, Ravenna, Reggio Emilia, Rimini, and Imola. We obtained data on the size of the population at risk (18–35 years old) in each catchment area, per year from the ISTAT census (ISTAT Italian Institute of Statistics, 2020). Population density was calculated as the total number of inhabitants per square kilometre, and log-transformed.

Data for sociodemographic and clinical variables was obtained during clinical consultations and recorded on the electronic or physical charts. Additional information was collected by clinicians on migrant status (either first- or second-generation), current substance misuse, and personality disorder in comorbidity (ICD-9 diagnosis of personality disorder in comorbidity). For descriptive purposes, we report data on the Italian version of the Health of the Nation Outcome Scale (HoNOS) (Lora et al., 2001) which was used to rate the severity of behavioural, impairment, social problems and symptoms over the previous 2 weeks (Starace & Mazzi, 2012).

2.4 | Data analyses

First, we report the raw estimates of incidence of psychosis, calculated as the number of the newly treated cases per 100 000 persons/year in the region and each departmental catchment area.

on Wiley Online Library for rules of use; OA articles

are governed by the applicable Creative Commons License

7517893, 0, Downloaded from https://onlinelibrary.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Library on [24/05/2023]. See the Terms and Conditions (https://onlinelibrary.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Library on [24/05/2023]. See the Terms and Conditions (https://onlinelibrary.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Library on [24/05/2023]. See the Terms and Conditions (https://onlinelibrary.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Library on [24/05/2023]. See the Terms and Conditions (https://onlinelibrary.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Library on [24/05/2023]. See the Terms and Conditions (https://onlinelibrary.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Library.wiley.com/doi/10.1111/eip.13437 by Cochraneltalia, Wiley Online Lib

Second, we estimated the incidence of FEP using a Bayesian modelling framework. The Bayesian approach assumes that observed data are a noisy indicator of a "true" generating process: it estimates both the parameter(s) and its level of uncertainty, expressed as a posterior probabilistic distribution. Different from null hypothesis testing, Bayesian models provide the probabilities a whole range of parameter values, thus conveying richer information. When data are measured from multiple centres from the same population it is recommended to use generalized hierarchical models (a.k.a. "mixed" models) because they reduce the sensitivity to noise in data. In such models, differences of incidence between centres may depend on different distribution of unmeasured risk factors (McElreath, 2018). By accounting for the level of uncertainty and by combining information from multiple centres (partial pooling), Bayesian models proved particularly successful for the prediction and forecasting of the incidence of rare events such as FEP (McDonald et al., 2021).

In this study, we used different Generalized Linear Models of increasing complexity to estimate the distribution and time-related changes of FEP incidence rate. In particular, we modelled data using both a Poisson and Negative Binomial distribution. The latter assumes there could be overdispersion of data. Poisson models have one outcome parameter, the ratio of incident cases over the at-risk population (offset). This main single parameter represents the mean count of events: its variance is constrained to be equal to the mean. In the first model, the population density and study year served as the predictors. In the subsequent models, we used a hierarchical approach, allowing: (1) the intercept; (2) the intercept and longitudinal slope, to vary by centre. Then, we used a Negative-Binomial distribution, where the mean incidence and its variance (a.k.a. shape) are distinct parameters that serve as dependent variables. Each can be associated with an individual set of predictors. For instance, such models may be able to detect if time is associated with increase or decreases of the variability of incident cases, even if the mean incidence remains the same (Williams et al., 2019). We explored the different incidence rates between centres by comparing posterior predictions and overlaps in their 95% credible intervals. Default, non-informative priors were used for all models. Models were compared evaluating based on their out-of-sample predictive accuracy using Leave-One-Out Cross Validation, where higher values of the expected log pointwise predictive density (ELPD-LOO) indicate better predictive accuracy. Influential cases were identified using the pareto-k-diagnostic. However, we also provide Bayesian R² values as an intuitive measure of model fit, calculated as predicted variance divided by predicted variance plus error variance (Gelman et al., 2019). Analyses were conducted with the 2.1.16 brms R package (Bürkner, 2017) and reported with the tidybayes 3.0.2 (Kay, 2021).

Third, we compared user sociodemographic and clinical characteristics between centres and study years. Chi-square and non-parametric analyses were used to compare single categorical and continuous variables, respectively, while the supplement reports detailed plots for each variable for visual aid.

Fourth, we re-examined the association between user characteristics, study centre and year using clusters of patients based on a basic

set of sociodemographic features (age, gender, migrant status and substance use). Cluster analysis was based on Gower distance and the Partitioning Around Medoids (PAM) method; the number of clusters was identified examining silhouette width. This analysis was performed with the *cluster* R package (Maechler et al., 2022).

3 | RESULTS

3.1 | Raw incidence of treated cases

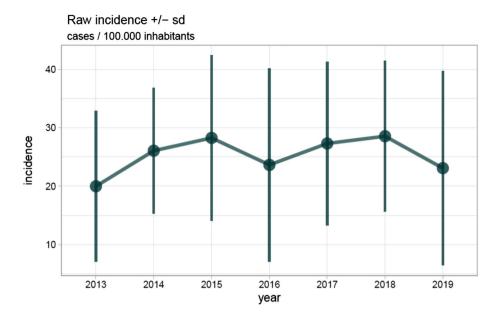
The program included 1454 individuals with FEP (Figure S1). This corresponds to a grand mean incidence of 25.3 cases/100 000 inhabitants per year, Interquartile Range (IQR) = 15.3, using the reference 2016 census size of the population at risk (18–35 years old, 781.172 inhabitants).

The raw incidence of FEP in the regional program did not reveal specific longitudinal temporal patterns. There was relatively lower incidence in 2013, 2016 and 2019 (Figure 1, Table S1). There were large differences in the average incidence of each centre (from 11.76 Imola to 39.54 / 100.000 inhabitants per year in Parma, Table S2). However, by examining the temporal changes within each centre (depicted as diamonds in Figure 2) no clear temporal pattern emerged.

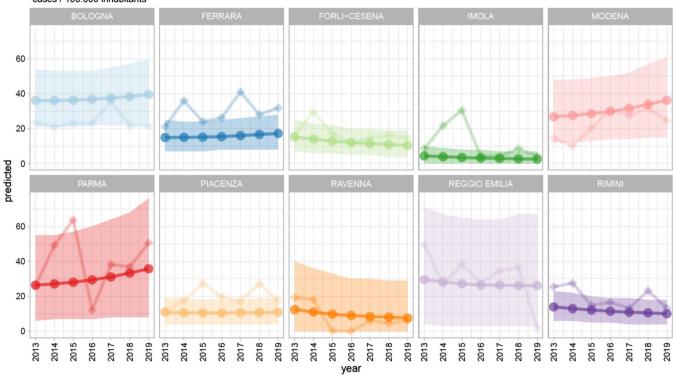
3.2 | Bayesian models of FEP incidence

The incidence data were modelled using a series of Bayesian models. Based on the out-of-sample predictive accuracy, a Hierarchical model based on the Negative Binomial distribution resulted as the most accurate (Table S3). The model used population density and study year to predict the mean incidence rate, nesting temporal trends within study centre (grouping factor). In addition, the dispersion of incidence rates was allowed to vary by study centre. The model converged well, with good posterior predictive checks and rhat values equal to 1.00. Accounting for both random and fixed effects, the model explained 64.3% of the variance in incidence rates (Bayesian R²; Error: 6.6%, 95% Crl: 49.5%-75.3%); based only on fixed effects, that is, not accounting for centre variability, it explained 37.2% (Error: 9.9% 95% Crl: 17%-6%-56.5%). The model did not detect evidence of linear temporal trend in the general regional population. Similarly, there was no evidence of an effect of population density, with a very weak trend for a negative effect on incidence (Table S3. 'fixed' effects and Figure S5). Variability of FEP incidence rates could be largely attributed to study centre, rather than year (Figure 1, S3, S4, and S6). For instance, in 2016 Imola had the lowest incidence (incidence: 3.07; 95% Credible Intervals: 1.61-4.99), followed by Ravenna (8.15; 95% Cr.I.: 2.95-17.31), Piacenza (10.41; 95% CI: 7.73-13.41), Rimini (11.36; 95% Cr.l.: 8.56-14.60), Forlì-Cesena (11.98; 95% Cr.l.: 8.92-15.25), Ferrara (15.38; 95% Cr.I.: 12.11-19.11), Reggio Emilia (25.81; 95% Cr.l.: 15.66-39.33), Parma (29.04; 95% Cr.l.: 18.18-40.18), Modena (29.39; 95% Cr.I.: 22.53-37.50) and Bologna (36.55; 95% Cr. I.: 30.39-43.86). Imola, Rimini, Piacenza were characterized by low

cases per 100 000 inhabitant year.



Raw incidence per year/center cases / 100.000 inhabitants



Raw and modelled incidence of treated cases per 100 000 inhabitant year. Observed values are diamonds; circles and shaded areas represent mean values predicted by the model with 95% Credible Intervals.

degrees of uncertainty around the median (Figure S3). Compared posterior predictions and 95% Credible Intervals across centres for the year 2016: the model predicted that Bologna had a larger incidence of FEP than Ferrara, Forli-Cesena, Imola, Piacenza and Rimini. Imola had a lower incidence than Ferrara, Modena and Parma (the full list of comparisons is reported in Table S5). In addition, Rimini had a larger dispersion than Ravenna and Reggio Emilia.

User characteristics 3.3

The entire sample (1318 individuals) comprised 69% of males; the median age of 22 and 21% of the sample were migrants (Table S6). The vast majority was single, with a middle- or high-school education level and living with the family of origin (62%). The median DUP was 5 months.

We compared user characteristics across centres. Users had significant differences in mean age (p < .001), gender (p < .001), migrant status (p < .001), occupation (p = .009) and living conditions (p < .001), but not in education levels (p = .13) or civil status (p = 0.24) (Table 1 and Figures S7–S13). Considering clinical characteristics, there were significant differences in the percentage of users displaying personality disorders, substance use, as well as HoNOS total and subscale scores. Moreover, users differed across centres in terms of referrals to the program and length of the DUP (Table 2 and Figures S14–S18).

The same comparisons were drawn by study year: the differences were not significant in mean age, gender, migrant status, occupation, education levels, civil status, personality disorder and substance use but they were in terms of living conditions (p=.01), referral type (p=.006), HoNOS total and subscale scores (p<.01) and DUP length (p=.002; Tables S7 and S8). In addition, there was a significant negative association between study year and HoNOS total score ($R=-0.09,\ p<.001$) and between study year and DUP length ($R=-0.12,\ p<0.001$) (Figures S19 and S20).

Cluster analysis identified a seven-cluster solution as the one with the highest silhouette width. Clusters were differentiated by almost all possible permutations of gender, migrant status and substance use, with smaller differences in age (Table S9). They comprised: Male users (M), not migrant (migrant-), not substance users (substance-) (n=343), M, migrant+, substance+ (n=304), Females (F), migrant-, substance- (n=229), M, migrant+, substance+ (n=95), M, migrant+, substance- (n=89), F, migrant-, substance+ (n=72), F, migrant+, substance- (n=69). The distribution of user clusters displayed a significant variability by study centre (chi square = 123.6, df. = 54, p < .0001) but not by year (chi square = 41.2, df = 36, p- value = .253) (Figure 3 and Figure S21).

4 | DISCUSSION

This study describes the incidence and user characteristics of FEP in Emilia-Romagna during the first 7 years when the Regional Program for FEP was active. The treated incidence rates and user characteristics were rather variable across different departments, although deriving from the same region. Whereas, clinical severity of FEP and modality of access to the program were the few factors that may have displayed some variability in relation to the study year. This study is strengthened by a rigorous methodological approach and a large, representative sample.

The incidence of FEP in our study varied between 20 and 30 cases per 100 000 inhabitants in 7 years. This is in line with other estimates from the literature (Kirkbride et al., 2006). The more appropriate comparison is with the European EUGEI study, which adopted similar enrolment criteria and also included the Center of Bologna and other two Italian centres, over a period of 48 months (Jongsma et al., 2018; Tarricone et al., 2016). Our estimates are higher than those from the Bologna and the other Italian centres (Palermo and

Veneto), as well as that of the region of Sicily (Mulè et al., 2017), but they are lower than those from South London and Amsterdam, urbanized areas with high prevalence of risk factors, especially migration and cannabis (Di Forti et al., 2019). Our estimates are also much lower if compared to a recent Australian estimate (102.4 per 100 000 population at-risk) (Pignon et al., 2021). Finally, our estimates are similar to those of a recent meta-analysis (Jongsma et al., 2019) which detected extreme levels of heterogeneity in the incidence of psychosis across countries, and also failed to detect time-related trends. The meta-analysis of Jongsma and colleagues suggested that methodological factors may have influenced incidence rates in that larger studies, especially population-registers of whole countries or cohort studies similar to ours may generally detect higher incidence than first contact studies (Jongsma et al., 2019).

Our study was conducted within one region and assumes some level of homogeneity in sociocultural characteristics, case-selection criteria and healthcare approach that derive from the coordination between departmental areas and from the adoption of shared guidelines (Regione Emilia Romagna, 2016). Nonetheless, we observed large variability in the incidence, sociodemographic and clinical characteristics of individuals with psychosis. This variability was almost entirely attributable to the study centre, while we did not detect an effect of population density, as some studies (Kelly et al., 2010; Kirkbride et al., 2017; March et al., 2008), but not all did (Jongsma et al., 2018). Hence, factors other than population density may explain the differences we observed. Compared with other regions in Italy, the Emilia-Romagna region, in fact, has a relatively high and homogeneous population density. However, its population displays the highest use of substance in Italy (Di Forti et al., 2019), a higher presence of migrants - also internal migrants for work or study reasons (Tarricone et al., 2016), and a high average pro-capita income. The former two factors are associated with higher incidence of FEP, while the latter is expected to reduce the incidence of psychosis, although income inequality would need to be taken in account (Kirkbride et al., 2014). Further analyses of these data will clarify whether, and how the distribution of these factors contributes to shape the risk of psychosis. Based on the previous literature, one should expect non-linear patterns and complex interactions between different factors to shape incidence. For instance, the role of migrant status may interact with ethnic density (Baker et al., 2021; Bosqui et al., 2014), negative personal experiences (Tarricone et al., 2021) and social dynamics (Termorshuizen et al., 2022). Network analyses may aid the process of modelling complex interactions at the clinical and sociocultural levels (Amore et al., 2020).

This study is strengthened by a large sample, representative of well-defined regional catchment area, and ample coverage of cases of FEP. However, it should be considered preliminary, given that data with greater level of detail has still not been collected. We argue that more granular information on social, ethnic and cultural factors may allow to increase the level of explanation and prediction

Sociodemographic characteristics. TABLE 1

Characteristic	$\begin{array}{l} \text{Piacenza,} \\ N=71^1 \end{array}$	Parma, $N=232^1$	Reggio emilia, $N=215^1$	$\begin{array}{l} Modena, \\ N = 205^1 \end{array}$	Bologna, $N=261^1$	Imola, $N=19^1$	Ferrara, $N=113^{1}$	Ravenna, $N=36^1$	Forlices ena, $N=84^{1}$	Rimini, $N=82^1$
Age* N/No. obs. % not missing	70/71 99%	228/232 98%	206/215 96%	ı	ı	18/19 95%	107/113 95%	35/36 97%	ı	
Mean (SD)	22.8 (5.0)	24.8 (5.4)	22.4 (4.7)	24.0 (4.9)	23.4 (4.5)	22.9 (4.2)	22.2 (4.3)	21.3 (4.1)	24.8 (5.4)	23.0 (5.0)
Sex*										
ш	22 (31%)	66 (28%)	83 (39%)	65 (32%)	61 (23%)	8 (42%)	24 (21%)	10 (28%)	36 (43%)	18 (22%)
Σ	48 (68%)	162 (70%)	123 (57%)	140 (68%)	200 (77%)	10 (53%)	83 (73%)	25 (69%)	48 (57%)	64 (78%)
Missing	1 (1.4%)	4 (1.7%)	9 (4.2%)	(%0) 0	(%0) 0	1 (5.3%)	6 (5.3%)	1 (2.8%)	(%0) 0	(%0) 0
Education										
Elementary	1 (1.4%)	8 (3.4%)	5 (2.3%)	5 (2.4%)	4 (1.5%)	(%0) 0	3 (2.7%)	(%0) 0	4 (4.8%)	(%0) 0
Middle school diploma	29 (41%)	63 (27%)	67 (31%)	74 (36%)	96 (37%)	8 (42%)	33 (29%)	14 (39%)	21 (25%)	23 (28%)
High school diploma	26 (37%)	83 (36%)	72 (33%)	74 (36%)	124 (48%)	7 (37%)	38 (34%)	9 (25%)	42 (50%)	40 (49%)
University degree	5 (7.0%)	17 (7.3%)	6 (2.8%)	16 (7.8%)	31 (12%)	(%0) 0	4 (3.5%)	1 (2.8%)	11 (13%)	5 (6.1%)
Missing	10 (14%)	61 (26%)	(%08) 59	36 (18%)	6 (2.3%)	4 (21%)	35 (31%)	12 (33%)	6 (7.1%)	14 (17%)
Civil status										
Single	46 (65%)	186 (80%)	141 (66%)	162 (79%)	244 (93%)	13 (68%)	72 (64%)	24 (67%)	(%08) 29	58 (71%)
Stable relationship	5 (7.0%)	13 (5.6%)	6 (2.8%)	13 (6.3%)	16 (6.1%)	2 (11%)	3 (2.7%)	1 (2.8%)	8 (9.5%)	1 (1.2%)
Separated	(%0) 0	2 (0.9%)	(%0) 0	(%0) 0	(%0) 0	(%0) 0	1 (0.9%)	(%0) 0	2 (2.4%)	1 (1.2%)
Widow	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0
Missing	20 (28%)	31 (13%)	68 (32%)	30 (15%)	1 (0.4%)	4 (21%)	37 (33%)	11 (31%)	7 (8.3%)	22 (27%)
Migrant*										
٥N	26 (79%)	158 (68%)	176 (82%)	145 (71%)	194 (74%)	16 (84%)	92 (81%)	28 (78%)	(22 (24)	77 (94%)
Yes	14 (20%)	74 (32%)	38 (18%)	45 (22%)	56 (21%)	3 (16%)	21 (19%)	8 (22%)	19 (23%)	5 (6.1%)
Missing	1 (1.4%)	(%0) 0	1 (0.5%)	15 (7.3%)	11 (4.2%)	0 (0%)	(%0) 0	(%0) 0	(%0) 0	(%0) 0
Occupation*										
Student or other	25 (35%)	86 (37%)	73 (34%)	55 (27%)	103 (39%)	7 (37%)	42 (37%)	13 (36%)	20 (24%)	30 (37%)
Unemployed	18 (25%)	(%8E) 68	52 (24%)	69 (34%)	103 (39%)	6 (32%)	23 (20%)	11 (31%)	39 (46%)	28 (34%)
On disability	(%0) 0	(%0) 0	(%0) 0	2 (1.0%)	(%0) 0	1 (5.3%)	1 (0.9%)	(%0) 0	1 (1.2%)	(%0) 0
Employed	11 (15%)	35 (15%)	23 (11%)	34 (17%)	54 (21%)	3 (16%)	10 (8.8%)	1 (2.8%)	18 (21%)	10 (12%)
Missing	17 (24%)	22 (9.5%)	67 (31%)	45 (22%)	1 (0.4%)	2 (11%)	37 (33%)	11 (31%)	6 (7.1%)	14 (17%)
										(Continues)

Characteristic	$\begin{array}{l} \text{Piacenza,} \\ N=71^{1} \end{array}$	$\begin{array}{l} \text{Parma,} \\ N=232^{1} \end{array}$	Reggio emilia, $N=215^1$	$\begin{array}{l} Modena, \\ N = 205^1 \end{array}$	Bologna, $N=261^1$	Imola, $N=19^1$	Ferrara, $N=113^1$	Ravenna, $N=36^1$	Forlicesena, $N=84^1$	Rimini, $N=82^1$
Living condition*										
Family of origin	39 (55%)	126 (54%)	127 (59%)	0 (%)	180 (69%)	12 (63%)	52 (46%)	22 (61%)	60 (71%)	64 (78%)
Acquired family	6 (8.5%)	12 (5.2%)	11 (5.1%)	131 (64%)	30 (11%)	3 (16%)	5 (4.4%)	1 (2.8%)	7 (8.3%)	(%0) 0
Alone	(%0) 0	13 (5.6%)	2 (0.9%)	8 (3.9%)	16 (6.1%)	1 (5.3%)	6 (5.3%)	(%0) 0	5 (6.0%)	1 (1.2%)
Homeless	(%0) 0	2 (0.9%)	(%0) 0	2 (1.0%)	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0	(%0) 0
Residential	(%0) 0	16 (6.9%)	(%0) 0	1 (0.5%)	(%0) 0	(%0) 0	(%0) 0	1 (2.8%)	1 (1.2%)	(%0) 0
Other	9 (13%)	11 (4.7%)	8 (3.7%)	33 (16%)	35 (13%)	1 (5.3%)	19 (17%)	1 (2.8%)	7 (8.3%)	10 (12%)
Missing	17 (24%)	52 (22%)	67 (31%)	30 (15%)	(%0) 0	2 (11%)	31 (27%)	11 (31%)	4 (4.8%)	7 (8.5%)

Note: Categorical data: n (%); Continuous data: mean (5D). *p < .001 in Chi-squared test or Kruskal-Wallis rank sum test

of psychosis incidence, pathways to care, and clinical characteristics (Fett et al., 2019; Heinz et al., 2013; March et al., 2008). In fact, the distribution of modifiable and non-modifiable risk factors not only changes across countries or at the regional level, but also at the neighbourhood-level (O'Donoghue et al., 2016; Schofield et al., 2022). For example, in our sample educational levels were unevenly distributed across centres, with an overall high level. This might signal some difficulties of access to care by those with a lower socioeconomic level, even within a universal, accessible healthcare system. Thus, in order to reach a youth population with a high risk of psychosis, information campaigns and promotion of referrals should also be strengthened outside the school environment. Heterogeneity across departmental areas, in fact, may crucially depend on factors that are external to the mental health department, such as the efficacy of referrals and the pathways to care, or the knowledge regarding the availability of a specific programme for FEP. In addition, the assumption that the region has homogeneous application of criteria for access to care, and a similar bandwidth of resource allocation, may not entirely hold in clinical practice and should be further investigated (Starace et al., 2018). The present study has not examined the impact of staff resources on the detection of case and their entry into the mental healthcare system. Nonetheless, we have set the grounds for elaborating a robust method of ascertainment of FEP incidence to guide stakeholders on staffing allocation and resources planning (Kirkbride, 2015). In this regard, we did not observe a temporal trend in rising incidence of psychosis at the regional level, however, an increase in number of cases of psychosis due to the COVID-19 pandemic would likely require a modulation of resources (Jauhar et al., 2021).

Other limitations include that we did not collect data of individuals who refused care to minimize interference with routine clinical work. In addition, we did not seek to estimate the rate of subjects with FEP who were treated within other settings than the public DMH (e.g. the private sector or public health agencies outside the Emilia-Romagna region). Thus, the treated incidence estimates of FEP may approximate the true population values, and this prevents to some extent the generalization of our findings from the treated cohort to the general FEP population. However, this latter phenomenon may be marginal, as private practices are generally reluctant to engage with clients with early psychosis. Finally, this study does not distinguish individuals with prodromal schizophrenia, and between the treated incidence of FEP who had access to the RER-FEP program and those who were ineligible for the program (e.g. due to lack of language fluency, low IQ or psychosis due to substance misuse) and received TAU.

In conclusion, the treated incidence of FEP in the Emilia-Romagna regional public healthcare program was relatively high and varied across the different areas, but not according to the population density or time. Further detailed analyses of this data, informed by theory and prior knowledge, may allow examination of psychosocial factors influencing the onset of psychosis and may favour the elaboration of useful prediction models.

Clinical characteristics, referrals and DUP. TABLE 2

	$\begin{array}{l} \text{Piacenza,} \\ N=71 \end{array}$	Parma, N = 232	Reggio emilia, $N=215$	$\begin{array}{l} \text{Modena,} \\ \text{N} = 205 \end{array}$	Bologna, $N=261$	Imola, $N=19$	Ferrara, N = 113	Ravenna, N = 36	Forli- cesena, N = 84	Rimini, N = 82
Personality_disorder*	4 (5.6%)	125 (54%)	83 (39%)	34 (18%)	67 (27%)	2 (11%)	57 (51%)	5 (16%)	8 (9.5%)	11 (13%)
Missing	(%0) 0	(%0) 0	1 (0.5%)	18 (8.8%)	17 (6.5%)	(%0) 0	1 (0.9%)	4 (11%)	(%0) 0	(%0) 0
Substance use*	19 (27%)	82 (36%)	77 (36%)	70 (37%)	121 (49%)	10 (53%)	26 (50%)	15 (47%)	19 (23%)	44 (54%)
Missing	(%0) 0	2 (0.9%)	1 (0.5%)	17 (8.3%)	15 (5.7%)	(%0) 0	1 (0.9%)	4 (11%)	(%0) 0	0 (0%)
HONOS_total*	12.92 (6.17)	23.32 (8.44)	17.35 (5.68)	15.80 (6.57)	13.36 (8.84)	14.11 (6.12)	14.58 (6.89)	14.97 (6.45)	18.23 (7.60)	19.21 (6.71)
Subscales										
Behaviour *	1.80 (2.10)	3.84 (2.53)	2.92 (2.12)	2.43 (2.05)	2.39 (2.60)	1.84 (1.26)	2.74 (2.15)	2.94 (1.97)	2.99 (2.53)	3.16 (2.54)
Impairment*	0.94 (1.23)	3.72 (2.02)	1.51 (1.31)	1.40 (1.44)	0.84 (1.33)	1.21 (1.36)	1.42 (1.24)	1.25 (1.23)	1.60 (1.48)	2.11 (1.41)
Psychological*	5.25 (2.42)	7.25 (2.62)	7.53 (2.51)	5.94 (2.61)	4.85 (3.27)	5.63 (2.99)	5.50 (2.50)	5.36 (2.52)	6.65 (2.25)	7.43 (2.54)
Social*	4.92 (3.06)	8.52 (4.01)	5.39 (2.55)	6.03 (3.06)	5.28 (3.92)	5.42 (2.22)	4.91 (3.55)	5.42 (3.50)	6.99 (3.71)	6.51 (2.83)
Referral*										
Self-referral	10 (18%)	20 (11%)	28 (17%)	17 (9.6%)	49 (23%)	3 (21%)	11 (12%)	8 (38%)	11 (15%)	4 (6.6%)
Primary care	7 (12%)	26 (30%)	57 (35%)	60 (34%)	56 (26%)	1 (7.1%)	27 (30%)	1 (4.8%)	23 (32%)	19 (31%)
Public hospital	20 (36%)	14 (7.6%)	30 (19%)	48 (27%)	(33%)	6 (43%)	35 (39%)	6 (29%)	31 (44%)	30 (49%)
Private health service	1 (1.8%)	54 (29%)	1 (0.6%)	2 (1.1%)	(%0) 0	(%0) 0	2 (2.2%)	1 (4.8%)	(%0) 0	(%0) 0
Other health service	4 (7.1%)	37 (20%)	26 (16%)	38 (21%)	18 (8.5%)	2 (14%)	8 (9.0%)	1 (4.8%)	1 (1.4%)	7 (11%)
Social service/ police/prison	1 (1.8%)	4 (2.2%)	3 (1.9%)	4 (2.3%)	3 (1.4%)	(%0) 0	(%0) 0	1 (4.8%)	1 (1.4%)	(%0) 0
Other referral	13 (23%)	(%0) 0	16 (9.9%)	8 (4.5%)	17 (8.0%)	2 (14%)	6 (6.7%)	3 (14%)	4 (5.6%)	1 (1.6%)
Missing	15 (21%)	47 (20%)	54 (25%)	28 (14%)	49 (19%)	5 (26%)	24 (21%)	15 (42%)	13 (15%)	21 (26%)
DUP (months)*	3.14 (3.65)	7.83 (7.03)	16.69 (18.11)	6.67 (10.03)	7.74 (9.55)	13.16 (21.93)	4.40 (3.80)	9.50 (8.89)	9.47 (16.19)	4.63 (3.44)
N/No. obs. % not missing	ı	1	206/215 95.81%	176/205 85.85%	231/261 88.51%	ı	103/113 91.15%	32/36 88.89%	75/84 89.29%	

Note: Categorical data: n (%); Continuous data: mean (SD). *p < .001 in Chi-squared test or Kruskal–Wallis rank sum test.

FIGURE 3 Distribution of user clusters across centres. F. female; M, male; Mig, migrant; noMig, not migrant; Sub, substance use and noSub. no substance use

The Early Psychosis Working Group includes:

General Directorate of Health and Social Policies: Sandra Ventura, Maristella Salaris:

DMH-PA Piacenza: Silvia Chiesa, Massimo Rossetti, Ornella Bettinardi;

DMH-PA Parma: Emanuela Leuci, Lorenzo Pelizza, Antonella Squarcia;

DMH-PA Reggio Emilia: Lorenzo Berardi, Francesca Fontana, Sara Paderni:

DMH-PA Modena: Francesca Vacca, Nicoletta Nigro, Virginia Guerra;

DMH-PA Imola: Vincenza Giannini, Noemi Alagia, Pietro Papili; DMH-PA Bologna: Caterina Bruschi, Lorenzo Gammino, Stefania Artioli;

DMH-PA Ferrara: Paola Carozza, Luigi Grassi, Spyridon Zotos;

DMH-PA Ravenna: Sara Sternini, Rosaria Rispoli, Francesca Catena Egitto;

DMH-PA Forlì-Cesena: Forlani Claudia, Enrico Meregalli, Elena Soldati:

DMH-PA Rimini: Laura Mulazzani, Elena Vagnoni, Michela Pratelli.

FUNDING INFORMATION

Early Intervention Service Implementation has been funded by Regione Emilia Romagna since 2012 (Delibera Giunta Regionale Nun. 2155 del 2011). Last funding granted by Delibera Giunta Regionale Num. 1170 del 23/07/2018. Seduta n.32. Piano attuativo salute mentale e superamento ex oo.pp. riparto e assegnazione del fondo alle aziende sanitarie per l'anno 2018.

CONFLICT OF INTEREST STATEMENT

The authors declare they have no conflict of interest.

DATA AVAILABILITY STATEMENT

Research data are not shared.

ORCID

Martino Belvederi Murri https://orcid.org/0000-0002-7262-3528 Mattia Marchi https://orcid.org/0000-0003-2970-1276 Ilaria Tarricone https://orcid.org/0000-0002-5786-2520

REFERENCES

Aceituno, D., Vera, N., Prina, A. M., & McCrone, P. (2019). Costeffectiveness of early intervention in psychosis: Systematic review. The British Journal of Psychiatry, 215(1), 388-394. https://doi.org/10. 1192/bjp.2018.298

Ajnakina, O., Morgan, C., Gayer-Anderson, C., Oduola, S., Bourque, F., Bramley, S., Williamson, J., MacCabe, J. H., Dazzan, P., Murray, R. M., & David, A. S. (2017). Only a small proportion of patients with first episode psychosis come via prodromal services: A retrospective survey of a large UK mental health programme. BMC Psychiatry, 17(1), 308. https://doi.org/10.1186/s12888-017-1468-y

Amore, M., Murri, M. B., Calcagno, P., Rocca, P., Rossi, A., Aguglia, E., Bellomo, A., Blasi, G., Carpiniello, B., Cuomo, A., dell'Osso, L., di Giannantonio, M., Giordano, G. M., Marchesi, C., Monteleone, P., Montemagni, C., Oldani, L., Pompili, M., Roncone, R., ... members of the Italian Network for Research on Psychoses involved in this study. (2020). The association between insight and depressive symptoms in schizophrenia: Undirected and Bayesian network analyses. European Psychiatry: The Journal of the Association of European Psychiatrists, 63(1), 1-21. https://doi.org/10.1192/j.eurpsy.2020.45

Anglin, D. M., Ereshefsky, S., Klaunig, M. J., Bridgwater, M. A., Niendam, T. A., Ellman, L. M., DeVylder, J., Thayer, G., Bolden, K., Musket, C. W., Grattan, R. E., Lincoln, S. H., Schiffman, J., Lipner, E., Bachman, P., Corcoran, C. M., Mota, N. B., & van der Ven, E. (2021). From womb to neighborhood: A racial analysis of social determinants of psychosis in the United States. The American Journal of Psychiatry, 178(7), 599-610. https://doi.org/10.1176/appi.ajp.2020.20071091

Annuario Statistico Regionale. (2020). Tabella Prodotto interno lordo. - Principali Regioni europee. ASR Lombardia. https://www.asr-lombardia.it/

- asrlomb/it/13548regioniprodotto-interno-lordo-principali-regionieuropee
- Baker, S. J., Jackson, M., Jongsma, H., & Saville, C. W. N. (2021). The ethnic density effect in psychosis: A systematic review and multilevel metaanalysis. The British Journal of Psychiatry: The Journal of Mental Science, 219(6), 632–643. https://doi.org/10.1192/bjp.2021.96
- Belvederi Murri, M., Bertelli, R., Carozza, P., Berardi, L., Cantarelli, L., Croce, E., Antenora, F., Curtarello, E. M. A., Simonelli, G., Recla, E., Girotto, B., Grassi, L., Callegari, V., Cardelli, R., Emanuelli, F., Garofani, L., Marangoni, C., Mazzoni, P., Nappi, G., ... Zerbinati, L. (2020). First-episode psychosis in the Ferrara mental health department: Incidence and clinical course within the first 2 years. *Early intervention*. Psychiatry, 5, 1–11. https://doi.org/10.1111/eip.13095
- Belvederi Murri, M., Cattelani, L., Chesani, F., Palumbo, P., Triolo, F., & Alexopoulos, G. S. (2022). Risk prediction models for depression in community-dwelling older adults. The American Journal of Geriatric Psychiatry: Official Journal of the American Association for Geriatric Psychiatry, \$1064-7481(22), 949-960. https://doi.org/10.1016/j.jagp.2022.05.017
- Bosqui, T. J., Hoy, K., & Shannon, C. (2014). A systematic review and meta-analysis of the ethnic density effect in psychotic disorders. *Social Psychiatry and Psychiatric Epidemiology*, 49(4), 519–529. https://doi.org/10.1007/s00127-013-0773-0
- Bürkner, P.-C. (2017). {brms}: An {R} Package for {Bayesian} Multilevel Models Using {Stan}. *Journal of Statistical Software*, 80(1), 1–28. https://doi.org/10.18637/jss.v080.i01
- Campion, J., Taylor, M. J., McDaid, D., Park, A.-L., & Shiers, D. (2019). Applying economic models to estimate local economic benefits of improved coverage of early intervention for psychosis. *Early Interven*tion in Psychiatry, 13(6), 1424–1430. https://doi.org/10.1111/eip. 12787
- Cocchi, A., Cavicchini, A., Collavo, M., Ghio, L., Macchi, S., Meneghelli, A., & Preti, A. (2018). Implementation and development of early intervention in psychosis services in Italy: A national survey promoted by the Associazione Italiana Interventi Precoci nelle Psicosi. *Early Intervention in Psychiatry*, 12(1), 37–44. https://doi.org/10.1111/eip.12277
- Correll, C. U., Galling, B., Pawar, A., Krivko, A., Bonetto, C., Ruggeri, M., Craig, T. J., Nordentoft, M., Srihari, V. H., Guloksuz, S., Hui, C. L. M., Chen, E. Y. H., Valencia, M., Juarez, F., Robinson, D. G., Schooler, N. R., Brunette, M. F., Mueser, K. T., Rosenheck, R. A., ... Kane, J. M. (2018). Comparison of early intervention services vs treatment as usual for early-phase psychosis: A systematic review, meta-analysis, and meta-regression. *JAMA Psychiatry*, 75(6), 555–565. https://doi.org/10.1001/jamapsychiatry.2018.0623
- Curcetti, C., Morini, M., Neri, G., & Zappi, A. (2005). Integrazione tra cure primarie e salute mentale-II programma "Giuseppe Leggieri" della Regione Emilia-Romagna. Editrice Compositori.
- Di Forti, M., Quattrone, D., Freeman, T. P., Tripoli, G., Gayer-Anderson, C., Quigley, H., Rodriguez, V., Jongsma, H. E., Ferraro, L., La Cascia, C., La Barbera, D., Tarricone, I., Berardi, D., Szöke, A., Arango, C., Tortelli, A., Velthorst, E., Bernardo, M., Del-Ben, C. M., ... EU-GEI WP2 Group. (2019). The contribution of cannabis use to variation in the incidence of psychotic disorder across Europe (EU-GEI): A multicentre case-control study. The Lancet Psychiatry, 6(5), 427-436. https://doi.org/10.1016/S2215-0366(19)30048-3
- Ferrara, M., Tedeschini, E., Baccari, F., Musella, V., Vacca, F., Mazzi, F., Ferri, M., Srihari, V., & Starace, F. (2019). Early intervention service for first episode psychosis in Modena, northern Italy: The first hundred cases. *Early Intervention in Psychiatry*, 13(4), 1011–1017. https://doi.org/10.1111/eip.12788
- Fett, A.-K. J., Lemmers-Jansen, I. L. J., & Krabbendam, L. (2019). Psychosis and urbanicity: A review of the recent literature from epidemiology to neurourbanism. *Current Opinion in Psychiatry*, 32(3), 232–241. https://doi.org/10.1097/YCO.000000000000486

- Gelman, A., Goodrich, B., Gabry, J., & Vehtari, A. (2019). R-squared for Bayesian regression models. *The American Statistician*, 73(3), 307–309. https://doi.org/10.1080/00031305.2018.1549100
- Heinz, A., Deserno, L., & Reininghaus, U. (2013). Urbanicity, social adversity and psychosis. World Psychiatry, 12(3), 187–197. https://doi.org/10.1002/wps.20056
- ISTAT Italian Institute of Statistics. (2020). Resident population data. In Popolazione Residente Comunale per Sesso, Anno di Nascita e Stato Civile. http://demo.istat.it/
- Jauhar, S., Lai, S., Bonoldi, I., Salazar de Pablo, G., di Forti, M., Alameda, L., Donocik, J., Iacoponi, E., Spencer, T., Haege, B., McLaughlan, D., Taylor, D., Young, A., Thornicroft, G., Gaughran, F., MacCabe, J., Murray, R., McGuire, P., & Fusar-Poli, P. (2021). Early intervention in psychosis during the COVID-19 pandemic: Maudsley recommendations. European Neuropsychopharmacology, 47, 130–135. https://doi.org/10.1016/j.euroneuro.2021.02.005
- Jongsma, H. E., Gayer-Anderson, C., Lasalvia, A., Quattrone, D., Mulè, A., Szöke, A., Selten, J. P., Turner, C., Arango, C., Tarricone, I., Berardi, D., Tortelli, A., Llorca, P. M., De Haan, L., Bobes, J., Bernardo, M., Sanjuán, J., Santos, J. L., Arrojo, M., ... Cristofalo, D. (2018). Treated incidence of psychotic disorders in the multinational EU-GEI study. JAMA Psychiatry, 75(1), 36–46. https://doi.org/10.1001/jamapsychiatry.2017.3554
- Jongsma, H. E., Karlsen, S., Kirkbride, J. B., & Jones, P. B. (2021). Understanding the excess psychosis risk in ethnic minorities: The impact of structure and identity. Social Psychiatry and Psychiatric Epidemiology, 56(11), 1913–1921. https://doi.org/10.1007/s00127-021-02042-8
- Jongsma, H. E., Turner, C., Kirkbride, J. B., & Jones, P. B. (2019). International incidence of psychotic disorders, 2002–17: A systematic review and meta-analysis. *The Lancet Public Health*, 4(5), e229–e244. https://doi.org/10.1016/S2468-2667(19)30056-8
- Kay, M. (2021). Tidybayes: Tidy data and geoms for Bayesian models. *In R Package Version*, 3, 1–1. https://doi.org/10.5281/zenodo.1308151
- Kelly, B. D., O'Callaghan, E., Waddington, J. L., Feeney, L., Browne, S., Scully, P. J., Clarke, M., Quinn, J. F., McTigue, O., Morgan, M. G., Kinsella, A., & Larkin, C. (2010). Schizophrenia and the city: A review of literature and prospective study of psychosis and urbanicity in Ireland. Schizophrenia Research, 116(1), 75–89. https://doi.org/10.1016/j.schres.2009.10.015
- Kirkbride, J. B. (2015). Epidemiology on demand: Population-based approaches to mental health service commissioning. *BJPsych Bulletin*, 39(5), 242–247. https://doi.org/10.1192/pb.bp.114.047746
- Kirkbride, J. B., Fearon, P., Morgan, C., Dazzan, P., Morgan, K., Tarrant, J., Lloyd, T., Holloway, J., Hutchinson, G., Leff, J. P., Mallett, R. M., Harrison, G. L., Murray, R. M., & Jones, P. B. (2006). Heterogeneity in incidence rates of schizophrenia and other psychotic syndromes: Findings from the 3-center ÆSOP study. Archives of General Psychiatry, 63(3), 250-258. https://doi.org/10.1001/archpsyc.63.3.250
- Kirkbride, J. B., Hameed, Y., Ankireddypalli, G., Ioannidis, K., Crane, C. M., Nasir, M., Kabacs, N., Metastasio, A., Jenkins, O., Espandian, A., Spyridi, S., Ralevic, D., Siddabattuni, S., Walden, B., Adeoye, A., Perez, J., & Jones, P. B. (2017). The epidemiology of first-episode psychosis in early intervention in psychosis services: Findings from the social epidemiology of psychoses in east Anglia [SEPEA] study. American Journal of Psychiatry., 174, 143–153. https://doi.org/10.1176/appi.ajp.2016.16010103
- Kirkbride, J. B., Jones, P. B., Ullrich, S., & Coid, J. W. (2014). Social deprivation, inequality, and the neighborhood-level incidence of psychotic syndromes in East London. *Schizophrenia Bulletin*, 40(1), 169–180. https://doi.org/10.1093/schbul/sbs151
- Lee, R., Leighton, S. P., Thomas, L., Gkoutos, G. V., Wood, S. J., Fenton, S.-J. H., Deligianni, F., Cavanagh, J., & Mallikarjun, P. K. (2022). Prediction models in first-episode psychosis: Systematic review and critical appraisal. *The British Journal of Psychiatry*, 220(4), 179–191. https://doi.org/10.1192/bjp.2021.219

- Leighton, S. P., Upthegrove, R., Krishnadas, R., Benros, M. E., Broome, M. R., Gkoutos, G. V., Liddle, P. F., Singh, S. P., Everard, L., Jones, P. B., Fowler, D., Sharma, V., Freemantle, N., Christensen, R. H. B., Albert, N., Nordentoft, M., Schwannauer, M., Cavanagh, J., Gumley, A. I., ... Mallikarjun, P. K. (2019). Development and validation of multivariable prediction models of remission, recovery, and quality of life outcomes in people with first episode psychosis: A machine learning approach. *The Lancet Digital Health*, 1(6), e261–e270. https://doi.org/10.1016/S2589-7500(19)30121-9
- Leuci, E., Quattrone, E., Pellegrini, P., & Pelizza, L. (2020). The 'Parma-early psychosis' program: General description and process analysis after 5 years of clinical activity. *Early Intervention in Psychiatry*, 14(3), 356–364. https://doi.org/10.1111/eip.12897
- Lora, A., Bai, G., Bianchi, S., Bolongaro, G., Civenti, G., Erlicher, A., Maresca, G., Monzani, E., Panetta, B., Von Morgen, D., Rossi, F., Torri, V., & Morosini, P. (2001). The italian version of HoNOS (health of the nation outcome scales), a scale for evaluating the outcome and the severity in mental health services. *Epidemiologia e Psichiatria Sociale.*, 10, 198–204. https://doi.org/10.1017/s1121189x00005339
- Maechler, M., P. R. A. S. M. H. (S, Hornik [trl, K., maintenance(1999-2000)), ctb] (port to R., Studer, M., Roudier, P., Gonzalez, J., Kozlowski, K., pam()), E. S. (fastpam options for, & Murphy (volume. ellipsoid({d >= 3})), K. (2022). cluster: 'Finding Groups in Data': Cluster Analysis Extended Rousseeuw et al. (2.1.4). https://CRAN.R-project.org/package=cluster
- March, D., Hatch, S. L., Morgan, C., Kirkbride, J. B., Bresnahan, M., Fearon, P., & Susser, E. (2008). Psychosis and place. *Epidemiologic Reviews*, 30(1), 84–100. https://doi.org/10.1093/epirev/mxn006
- Mascayano, F., van der Ven, E., Martinez-Ales, G., Basaraba, C., Jones, N., Lee, R., Bello, I., Nossel, I., Smith, S., Smith, T. E., Wall, M., Susser, E., & Dixon, L. B. (2020). Predictors of early discharge from early intervention Services for Psychosis in New York state (Vol. 71, pp. 1151–1157). Psychiatric Services. https://doi.org/10.1176/appi.ps.202000025
- McDonagh, M. S., Dana, T., Kopelovich, S. L., Monroe-DeVita, M., Blazina, I., Bougatsos, C., Grusing, S., & Selph, S. S. (2022). Psychosocial interventions for adults with schizophrenia: An overview and update of systematic reviews. *Psychiatric Services*, 73(3), 299–312. https://doi.org/10.1176/appi.ps.202000649
- McDonald, K., Ding, T., Ker, H., Dliwayo, T. R., Osborn, D. P. J., Wohland, P., Coid, J. W., French, P., Jones, P. B., Baio, G., & Kirkbride, J. B. (2021). Using epidemiological evidence to forecast population need for early treatment programmes in mental health: A generalisable Bayesian prediction methodology applied to and validated for first-episode psychosis in England. *The British Journal of Psychiatry*, 219(1), 383–391. https://doi.org/10.1192/bjp.2021.18
- McElreath, R. (2018). Statistical rethinking: A bayesian course with examples in R and stan. In Statistical rethinking: A Bayesian course with examples in R and Stan (p. 469). Chapman and Hall/CRC. https://doi.org/10.1201/9781315372495
- Meneghelli, A., Ciancaglini, P., Di Domenico, M., Mazzola, A., Ghio, L., & Preti, A. (2023). Implementation of early intervention clinical services within the National Health System in Italy: Third wave survey with focus on structures, resources, and fidelity to the evidence-based model. Early Intervention in Psychiatry. https://doi.org/10.1111/eip.13380
- Mulè, A., Sideli, L., Capuccio, V., Fearon, P., Ferraro, L., Kirkbride, J. B., La Cascia, C., Sartorio, C., Seminerio, F., Tripoli, G., Di Forti, M., La Barbera, D., & Murray, R. M. (2017). Low incidence of psychosis in Italy: Confirmation from the first epidemiological study in Sicily. Social Psychiatry and Psychiatric Epidemiology, 52(2), 155–162. https://doi.org/10.1007/s00127-016-1322-4
- O'Donoghue, B., Lyne, J. P., Renwick, L., Lane, A., Madigan, K., Staines, A., O'Callaghan, E., & Clarke, M. (2016). Neighbourhood characteristics and the incidence of first-episode psychosis and duration of untreated

- psychosis. *Psychological Medicine*, 46(7), 1367–1378. https://doi.org/10.1017/S003329171500286X
- Pelizza, L., Azzali, S., Paterlini, F., Garlassi, S., Scazza, I., Chiri, L. R., Poletti, M., Pupo, S., & Raballo, A. (2019). The "Reggio Emilia At-risk mental states" program: A diffused, "liquid" model of early intervention in psychosis implemented in an Italian Department of Mental Health. Early Intervention in Psychiatry, 13(6), 1513–1524. https://doi.org/10.1111/eip.12851
- Pignon, B., Eaton, S., Schürhoff, F., Szöke, A., McGorry, P., & O'Donoghue, B. (2021). Temporal variation in the incidence of treated psychotic disorders in young people. *Schizophrenia Research*, 231, 221–226. https://doi.org/10.1016/j.schres.2021.03.011
- Regione Emilia Romagna. (2016). Raccomandazioni regionali per la promozione della salute e del benessere in persone all'esordio psicotico. Centro Stampa Della Regione Emilia Romagna. https://salute.regione.emilia-romagna.it/normativa-e-documentazione/linee-di-indirizzo/archivio-documenti-tecnici/linee-guida/raccomandazioni-regionali-esordio-psicotico-2016
- Regione Emilia Romagna. (2021). Grave deprivazione materiale. *Statistica Regione Emilia Romagna*, 1–5. https://statistica.regione.emiliaromagna.it/factbook/fb/benessere/gdm
- Schofield, P., Thisted Horsdal, H., Das-Munshi, J., Thygesen, M., Pedersen, C., Morgan, C., & Agerbo, E. (2022). A comparison of neighbourhood level variation and risk factors for affective versus nonaffective psychosis. *Schizophrenia Research*, S0920-9964(22)00192-X. https://doi.org/10.1016/j.schres.2022.05.015
- Shah, J. L., Jones, N., van Os, J., McGorry, P. D., & Gülöksüz, S. (2022). Early intervention service systems for youth mental health: Integrating pluripotentiality, clinical staging, and transdiagnostic lessons from early psychosis. The Lancet Psychiatry, 9(5), 413–422. https://doi.org/10. 1016/S2215-0366(21)00467-3
- Singh, S. P., Croudace, T., Amin, S., Kwiecinski, R., Medley, I., Jones, P. B., & Harrison, G. (2000). Three-year outcome of firstepisode psychoses in an established community psychiatric service. *The British Journal of Psychiatry*, 176(3), 210–216. https://doi.org/10. 1192/bjp.176.3.210
- Starace, F., & Mazzi, F. (2012). HoNOS-MHCT Scala di esito per soggetti adulti. Mental Health Clustering Booklet v2.01 2011/12 Department of Health. Versione italiana a cura di F. Starace, F. Mazzi AUSL Modena.
- Starace, F., Mungai, F., & Barbui, C. (2018). Does mental health staffing level affect antipsychotic prescribing? Analysis of Italian national statistics. PLoS One, 13(2), e0193216. https://doi.org/10.1371/journal. pone.0193216
- Suvisaari, J., Mantere, O., Keinänen, J., Mäntylä, T., Rikandi, E., Lindgren, M., Kieseppä, T., & Raij, T. T. (2018). Is it possible to predict the future in first-episode psychosis? Frontiers in Psychiatry, 9, 1-15. https://doi.org/10.3389/fpsyt.2018.00580
- Tarricone, I., Boydell, J., Kokona, A., Triolo, F., Gamberini, L., Sutti, E., Marchetta, M., Menchetti, M., Di Forti, M., Murray, R. M., Morgan, C., & Berardi, D. (2016). Risk of psychosis and internal migration: Results from the Bologna first episode psychosis study. *Schizophrenia Research*, 173(1–2), 90–93. https://doi.org/10.1016/j.schres. 2016.02.032
- Tarricone, I., D'Andrea, G., Jongsma, H. E., Tosato, S., Gayer-Anderson, C., Stilo, S. A., Suprani, F., Iyegbe, C., van der Ven, E., Quattrone, D., di Forti, M., Velthorst, E., Rossi Menezes, P., Arango, C., Parellada, M., Lasalvia, A., La Cascia, C., Ferraro, L., Bobes, J., ... Morgan, C. (2021). Migration history and risk of psychosis: Results from the multinational EU-GEI study. *Psychological Medicine*, 1-13, 2972-2984. https://doi.org/10.1017/S003329172000495X
- Tarricone, I., Mimmi, S., Paparelli, A., Rossi, E., Mori, E., Panigada, S., Carchia, G., Bandieri, V., Michetti, R., Minenna, G., Boydell, J.,

Morgan, C., & Berardi, D. (2012). First-episode psychosis at the West Bologna community mental health Centre: Results of an 8-year prospective study. *Psychological Medicine*, 42(11), 2255–2264. https://doi.org/10.1017/S0033291712000335

Termorshuizen, F., van der Ven, E., Tarricone, I., Jongsma, H. E., Gayer-Anderson, C., Lasalvia, A., Tosato, S., Quattrone, D., La Cascia, C., Szöke, A., Berardi, D., Llorca, P.-M., de Haan, L., Velthorst, E., Bernardo, M., Sanjuán, J., Arrojo, M., Murray, R. M., Rutten, B. P., ... Selten, J.-P. (2022). The incidence of psychotic disorders among migrants and minority ethnic groups in Europe: Findings from the multinational EU-GEI study. *Psychological Medicine*, 52(7), 1376–1385. https://doi.org/10.1017/S0033291720003219

Williams, D. R., Zimprich, D. R., & Rast, P. (2019). A Bayesian nonlinear mixed-effects location scale model for learning. *Behavior Research Methods*, 51(5), 1968–1986. https://doi.org/10.3758/S13428-019-01255-9/FIGURES/4

SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

How to cite this article: Belvederi Murri, M., Ferrara, M., Imbesi, M., Leuci, E., Marchi, M., Musella, V., Natali, A., Neri, A., Ragni, S., Saponaro, A., Tarricone, I., Tullini, A., Starace, F., & for the Early Psychosis Working Group (for Group Authorship) (2023). A public early intervention approach to first-episode psychosis: Treated incidence over 7 years in the Emilia-Romagna region. *Early Intervention in Psychiatry*, 1–13. https://doi.org/10.1111/eip.13437