Women Behind Bars: Do Gender-Specific Prisons Reduce Recidivism? *

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Abstract

In 143 countries, incarcerated women serve their sentence in a separate section within prisons that mainly house male inmates, i.e., mixed-gender prisons. In fewer cases, there are prisons exclusively for women, that is, women-only prisons. We investigate whether serving time in a women-only prison, rather than a mixed-gender one, causes a reduction in women's recidivism. We use data from Italy, where both types of prison coexist and where the institutional details of the process allocating women convict to prisons allow us to mimic a quasi-random assignment. Using confidential data on the universe of individual inmates serving time between 2012 and 2022, we find that women-only prisons lower women's recidivism by up to 15 percentage points in the three years after their release. We conjecture, with some supporting evidence, that one mechanism behind this result is that detention structures that accommodate only women can be more easily customized around their needs and allow for more effective rehabilitation initiatives.

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1 Introduction

Around the world, convicted women serve their sentences in different institutional settings. One possibility are mixed-gender prisons, i.e. facilities that predominantly house male convicts and have a separate female section. Another possibility are women-only prisons, i.e., facilities entirely dedicated to women. Currently, in 79 countries, representing 14% of the world's population, there are exclusively mixed-gender prisons. In 56 countries, with 30% of the world population, there are exclusively women-only prisons. In 64 countries, with 54% of the world population, there are both types of prison.[1]

It is not clear why some countries choose one model and others choose a different one. In some cases, it might have been the legacy of historical developments. For example, in the U.S., the so-called reformatory model, more attuned to women's needs, replaced at the beginning of the twentieth century the so-called custodial model, in the wake of social feminism and the purity movement, and left both cultural and infrastructural inheritances favoring the women-only model.[2]

In some cases, cost considerations might have played a role. Given that women represent only about 7% of the global prison population, it would often appear more efficient to ring-fence a few women in an existing building, already occupied by men, rather than building from scratch a separate facility.

In any case, we lack a balanced assessment of the pros and cons of these different models. It is becoming increasingly important to have a better understanding of their consequences, given that, unfortunately, the number of female inmates is on the rise: while the world male prison population witnessed a 22% increase from 2000 to 2022, the number of incarcerated women grew by nearly 60% during the same period, surpassing 740,000 (Penal Reform International, 2023).

In particular, the type of prison, whether it is mixed-gender or women-only, could significantly influence the conditions experienced during the time served by the women inmates and consequently impact their behavior after release. For example, it might be the case that it is easier to structure and organize women-only prisons in such a way as to address the specific needs of women and to offer gender-specific and more effective rehabilitation interventions, thereby reducing female recidivism and the societal costs it entails. Hence, cost considerations need to encompass the possible effect on recidivism.

More broadly, in addition to costs, a less callous approach seems justified to guide

[1] The distribution across the World is shown in Appendix Figure A2. For 12 countries, representing 2% of the world population, we were unable to find information on the type of female prisons.

[2] See Rafter (1983).

the evaluation of the relative merits of the two systems.[3] Widespread concerns raised by NGOs and international organizations have brought attention to the difficult conditions often experienced by female inmates, emphasizing the need for improved standards.[4] To the extent that the two systems are differently equipped to address women's specific needs, such as healthcare, tailored care for pregnant women, education policies, and training programs crafted with a specific focus on women's needs (World Health Organization, 2009), a balanced assessment should also consider these factors.

In this paper, we take a step towards such a balanced assessment by establishing the causal effect on female recidivism of serving a given sentence in a women-only rather than in a mixed-gender one.

To obtain this result, cross-country comparisons are hardly useful, due to multifaceted and difficult-to-control variations in prison systems among countries, compounded with distinct characteristics of the justice system.

The analysis within the same country is also not free from difficulties, mainly because the allocation of inmates across the two types of prisons is typically non-random, and might be correlated with perceived recidivism, thus contaminating with selection the estimated coefficient.

We conduct our analysis using data from Italy, whose prison system contains both mixed-gender and women-only prisons, albeit with a large prevalence of the first kind: 52 prisons housing male inmates with a separated female section, and 4 women-only prisons spread across the country, two in the South (Trani and Pozzuoli), one in Central Italy (Rome) and one in the North (Venice). As a background, it is useful to recall that in Italy, similarly to many other countries, the prison system is characterized by widespread poor conditions, overcrowding, violence and intimidation among inmates (Council of Europe, 2022).

We address the selection problem by leveraging institutional details of the assignment of inmates to prisons that are common to most, if not all, democratic countries, which allow us to mimic a quasi-random allocation of women between prisons of the two kinds.

In Italy, the decision concerning the prison where a convicted woman will serve her sentence is taken by the Prison Administration Department (DAP), balancing two criteria.

On the one hand, the four women-only prisons are considered to be more attuned to

[3]The challenging conditions faced by female convicts were acknowledged as early as 1833, when Auburn's prison chaplain noted that enduring a prolonged period as a female convict would be worse than death, contrasting it with the relatively tolerable conditions for male convicts in the same prison (Rafter, 1983).

[4]Penal Reform International: "Addressing the 105,000 increase in the global female prison population, ten years after the Bangkok Rules were adopted"

the women needs and are in principle the preferred choice.

On the other hand, the allocation tries to minimize the distance between the prison and the residence of the convict, to reduce the disruption of family and social links.[5]

When the (closest) mixed-gender prison is closer to the residence of the convict than the (closest) women-only prison, there is a tension between these two criteria. If the difference between the two distances is not too negative (i.e., when the women-only prison is not much more distant from the residence of the convict), the first consideration tends to prevail and the assigned prison is often a women-only one. If instead the mixed-gender prison is much closer to the residence of the convict, the second consideration tends to prevail, and the assigned prison is often a mixed-gender one.

As a result, the more negative is the difference between the two distances (mixed-gender to residence minus women-only to residence) the lower is the probability of being assigned to a women-only prison.

If the (closest) women-only prison is in fact closer to the residence of the convict than the (closest) mixed-gender one, there is no conflict between the two criteria and the women-only prison tends to be chosen.[6]

The first criterion is likely to imply a selection problem, as inmates who are perceived to be more (or less) likely to recidivate might be more (or less) likely to be assigned to prisons with better environments and treatment opportunities. However, the second criterion is arguably exogenous, being only dependent on the difference between two geographical distances.

We, therefore, use the difference between the two distances (each from the inmate's residence) as an instrument for the likelihood of being assigned to a women-only prison. In this way, we retain only the variability of the assignment that is attributable to an exogenous factor, thus getting close to random assignment. This instrumental variable (IV) approach enables the identification of the causal effect on recidivism stemming from gender-oriented prisons.

A nice feature of our setting is that we can actually test the exclusion restriction, i.e. the assumption that the effect of the instrument on recidivism is only the one mediated by the assignment to the two types of prisons, and not a direct one. This is because, as argued before, when the difference between the two distances is positive, we expect no effect on the assignment, once we control for the actual difference between the residence

[5]This rule is contained in the Penitentiary code (Ordinamento Penitenziario), Art. 14, and, for example, is similar to Title 18 of the U.S. Code §3621 - Imprisonment of a convicted person.

[6]In these cases, the probability of being assigned to a women-only prison is, however, not yet 1, as other considerations, like, for example, the capacity of the prison or the need to sever the social links of high-security convicts, might play a role. We see, however, that the probability is close to 0.8 and fairly constant, irrespective of the relative distance.

and the assigned prison. Hence, if the exclusion restriction is satisfied, we should observe no correlation between the instrument (on its positive range) and recidivism.

Our analysis is based on confidential individual inmate data kindly provided by the DAP, covering the period 2012-2022. We complement these data with annual surveys conducted by the Italian NGO Antigone[7], which provide detailed information on conditions in all Italian prisons.

We find a remarkable reduction in recidivism (measured over the three years following the release from a definitive sentence), ranging from 8 to 15 percentage points, among inmates quasi-randomly assigned to women-only prisons, compared to otherwise similar inmates serving the same sentence in a mixed-gender prison. This effect is present and significant for both Italian and foreign inmates, and it is robust to various specifications and robustness tests detailed below.

We have no conclusive and sharp evidence to identify the mechanism behind this result. Women-only prisons typically exhibit better conditions in terms of improved health services, enhanced social and working spaces, better employment opportunities, and greater autonomy in organizing inmates' time and movements within the facility. It is difficult to disentangle the contributions of these different features to the estimated effect of recidivism. Rather, we are likely dealing with a bundled treatment, with the overall effect arising from the simultaneous improvement in various dimensions.

Interestingly, we observe larger effects when the control group consists solely of women's sections within men's prisons that house very few women. This aligns with qualitative evidence (Palmisano, 2015) suggesting a minimum-scale effect: too few women might not justify activating programs tailored to their specific needs.

A considerable body of research has investigated the effects of imprisonment with a specific focus on deterrence (Abrams, 2012; Drago et al., 2009; Katz et al., 2003; Kuziemko, 2013; Vollaard, 2013) and incapacitation effects (Barbarino and Mastrobuoni, 2014; Buonanno and Raphael, 2013; Owens, 2009; Levitt, 1996). In relation to this study, researchers have explored the influence of alternatives to traditional incarceration policies on recidivism and examined various preventive measures, such as the use of electronic monitoring (Di Tella and Schargrodsky, 2013), rehabilitative practices (Lotti, 2022), better prison conditions (Drago et al., 2011), the contrast between high and minimum security levels in prisons (Chen and Shapiro, 2007), the comparison between new and old facilities (Tobón, 2022), and the distinction between open and closed prisons (Mastrobuoni and Terlizzese, 2022). Collectively, these studies indicate a consistent trend: less severe

[7]Antigone is a leading Italian non-governmental organization specializing in justice, human rights, and prison-related matters. Each year, their independent representatives gain access to Italian prisons, documenting the conditions of inmates and identifying key challenges within the prison system.

prison conditions—whether achieved through electronic monitoring, lower security level placement, or alternative approaches such as open cells—are associated with lower rates of recidivism.

On the contrary, there are no previous studies on the causal effects of gender-oriented prisons.[8] Our contribution is to show the relevance of women-only prisons as a determinant to reduce women's recidivism.

2 Institutional Context

Historically, the imprisonment of women in Italy has been based on a broad criterion of equality in treatment between male and female prisoners, on the separation between genders in prison, and on the protection of the "incarcerated mother". (Agnella, 2023). The approach reflected a gender identity that is narrowly defined by motherhood and traditionally feminine activities, a view that traces back to early modern institutions designed for women, which aimed to preserve or rehabilitate their virtue, often related to sexual deviance. Despite reforms aimed at social reintegration and ensuring formal equality of treatment, the contemporary Italian female prison system still echoes elements of its past. According to Italian NGOs working on this topic,[9] the treatment and rehabilitation opportunities for women are limited, with a focus on stereotypically female jobs and insufficient rehabilitative, educational, and vocational training. Women face inadequate healthcare, limited employment opportunities, and poor support for reintegration, exacerbating the challenges after release.

The prison system includes only four women-only facilities. The women inmates who are not housed in these four are assigned to one of the fifty-two mixed-gender prisons. The heavy reliance on the latter has been criticized for marginalizing female inmates, treating their issues as secondary among the various prison management challenges. A 2008 directive[10] has sought to address the absence of female-specific criminal regulations, introducing standards that recognize and protect gender differences to mitigate the marginalization of women in prison.

The default rule for allocating women inmates is to place them in the closest facility that can accommodate females, which can be either a women-only prison or a mixed-gender one.[11] This aims to minimize the disruption experienced by women inmates, the-

[8] Women recidivism has received relatively little attention and has been explored primarily through descriptive studies like those conducted by (Morash et al., 2017; Cobbina et al., 2012; Mathiassen, 2017).

[9] See "Dalla parte di Antigone: Primo Rapporto sulle donne detenute in Italia (2023)" (in Italian).

[10] *Circolare n.0308268 del 17 settembre 2008.*

[11] *Dispositivo dell'art. 14 Legge sull'ordinamento penitenziario Fonti → Legge sull'ordinamento penitenziario → Titolo I - Trattamento penitenziario → Capo III - Modalità del trattamento.*

oretically allowing for easier access by family and legal representation, and contributing to better overall welfare.

This default rule is not absolute, and the DAP office tasked with the assignment retains some discretion. Convicted women can be assigned to facilities further away than the closest one for security requirements,[12] capacity constraints or specific individual needs that cannot be met by the nearest institution. Most importantly for our purposes, there is widespread awareness, in the office of the DAP tasked with the assignment, that women-only facilities offer better prison conditions and rehabilitating opportunities.[13]

The assignment therefore can be represented as a multi-stage decision tree:

1 is the (closest) women-only prison closer to j's residence than the (closest) mixed-gender one?

- **1.1** if **yes**, is j a security risk inmate, or are there other reasons that prompt the suspension of the default rule?
  - **1.1.1** if *no*, then assign j to the (closest) women-only prison.
  - **1.1.2** if *yes*, then assign j to a prison different from the closest one. This might be another women-only prison, or a mixed-gender one.
- **1.2** if **no**, is j a security risk inmate, or are there other reasons that prompt the suspension of the default rule?
  - **1.2.1** if *no*, does the benefit of assigning j to a women-only prison outweigh the loss of assigning her to a prison farther away from her residence?
    - **1.2.1.1** if ***yes*** then assign j to the (closest) women-only prison.
    - **1.2.1.2** if ***no*** then assign j to the (closest) mixed-gender prison.
  - **1.2.2** if *yes*, then assign j to a a prison different from the closest one. This might be a women-only prison, or a mixed-gender one.

At node 1.1.2 of the decision tree the default rule is suspended, and closeness to j's residence is no longer a factor driving the prison assignment. In principle j could be sent to another women-only prison, but in the data we see that less than 2% of the assignments to a women-only prison involve a prison that is further away than the closest women-only prison. Also at node 1.2.2 the default rule is suspended, and the assignment could be to a women-only prison, if the latter satisfies the requirements that drive the decision when the default rule is suspended.[14] Whether or not this is the case, however,

[12] As a rule, members of organized crime are kept away from their familiar environment, to disrupt their criminal network, and therefore are not sent to the nearest prison.

[13] The presence of such awareness was established through interviews with DAP officials.

[14] Given what we just noted about the data, if this were the case it would be almost surely the women-only prison closest to j's residence.

the comparison between the distances of the two closest types of prison from j's residence is no longer a relevant factor in the decision. In view of these remarks, we slightly simplify the assignment rule by assuming that at node 1.1.2 the assignment is indeed to a mixed-gender prison. At node 1.2.2 we do not specify the trade-offs involved in the assignment decision, rather than assuming that they do not involve the relative distances of j-s residence from the two types of (closest) prisons.

We then formalize the assignment decision as follows.

Let $A(j) \in \{W, M\}$ denote the assignment of j to either a women-only of a mixed-gender prison

Let $B_{ji}, i \in W, M$ be the perceived benefit of sending a specific inmate j to a women-only or mixed-gender prison, with $B_{jW} > B_{jM}$ for all j, though not necessarily equal for all inmates. Let $D_{ji}, i \in W, M$ be the distance between j's municipality of residence and the (closest) women-only and mixed-gender prison. Finally, let α_j denote the probability that j is a security risk inmate, or there are other reasons that might justify a deviation from the default rule, and δ_j denote the probability (independent of the relative distances D_{ji}) that a women-only prison is assigned in the decision node 1.2.2.

Following the (simplified) decision tree, we then have that the probability of assigning j to a women-only prison is:

$$\mathbb{P}(A(j) = W) = \mathbb{1}(D_{jM} - D_{jW} \geq 0)(1 - \alpha_j) + \tag{1}$$
$$(1 - \mathbb{1}(D_{jM} - D_{jW} \geq 0))[(1 - \alpha_j)\mathbb{P}(D_{jM} - D_{jW} > \omega(B_{jM} - B_{jW})) + \alpha_j \delta_j],$$

where ω is a scaling factor to measure the benefit of assignment to either type of prison in the same units as the distance.

To simplify the notation, let ΔD_j denote the difference $D_{jM} - D_{jW}$ and ΔB_j denote the difference $\omega(B_{jM} - B_{jW})$.

Equation (1) then implies that, as long as $\Delta D_j < 0$, that is, as long as the (closest) mixed gender prison is closer to j' residence than the (closest) women only prison, the probability that j is assigned to a women-only prison is an increasing function of ΔD_j: recalling that ΔB_j is strictly negative, the closer ΔD_j gets to 0 the higher is $\mathbb{P}(\Delta D_j > \Delta B_j)$.

Instead, when $\Delta D_j > 0$, the probability that j is assigned to a women-only prison does not depend on ΔD_j.

These two features are important for our empirical strategy.

Most importantly, while ΔB_j (and possibly α_j and δ_j) could reflect selection in the assignment rule, ΔD_j is plausibly independent of any characteristic of j that could be correlated with her recidivism, once we control for a few observables of a geographic nature. Hence ΔD_j is a valid instrument for the assignment to a women-only prison.

Second, the fact that for positive values of ΔD_j the assignment probability is independent of the actual value of ΔD_j implies that we can test whether, in that range, the instrument has no effect on recidivism. In other words, we can test whether the instrument affects recidivism only through its effect on the probability of assignment, and not directly, i.e. we can test the exclusion restriction, which is in general an untestable assumption.

3 Empirical Specification

3.1 Data

The Prison Administration Department (*Dipartimento dell'Amministrazione Penitenziaria*, DAP) of the Italian Ministry of Justice granted us access to a large amount of information on the universe of female inmates who spent any time in prison between 2012 and 2022. The information includes 10,222 incarcerations.

We complement these data with yearly surveys on prison conditions, collected by an Italian NGO, Antigone, across all Italian prisons, as well as yearly data on the prison population by the Ministry of Justice.[15] Since 2019, Antigone collects yearly data specific on women conditions across Italian prisons.

We measure recidivism through re-incarceration within three years (similarly to Mastrobuoni and Terlizzese (2022)) from the release from prison (we also look at shorter time windows, i.e. 1 and 2 years). The data allows us to control for several individual characteristics, like age, marital status, region of residence and region or country of birth, education, the sentence length and sentence status, and the type of crime committed. We can also control for the possible overcrowding of different prisons, at yearly frequency.

Antigone's survey of prison conditions allows us to describe the main differences between women-only and mixed-gender prisons. A few differences stand out (see Table 1). Women-only prisons house on average many more female inmates, 246 against 59. In mixed-gender prisons, it is considerably less likely that inmates are allowed to freely move within the prison (79.8% versus 27.9%). The other notable differences are that inmates in women-only prisons are more likely to be allowed to work outside of the prison, are more likely to have a woman as prison director (97% against 64%) and are considerably more likely to have access to hot water, obstetricians, and gynecologists. In women only prisons there is also a lower share of self-harm episodes per year (12% against 30% in mixed prisons).

It should be noted that the unconditional rate of recidivism is slightly larger among

[15] We collect yearly statistics from Statistiche sulla popolazione detenuta report for the years 2012-2022.

inmates who spent their sentence in a women-only prison, compared to inmates in mixed-gender prisons (about 12% vs. 11.4%), possibly due to an endogenous selection of inmates. At the same time, women-only prisons tend to be localized in or near large cities much more frequently than mixed-gender ones. Given that, for a number of reasons, recidivism is more prevalent in large cities (e.g. Staton-Tindall et al. (2015)), the near equality of the unconditional rates of recidivism is likely to mask important differences between the two types of prison.

3.2 Identification

We aim to estimate the causal effect on recidivism of serving the same sentence in a women-only prison, instead of in a mixed-gender one. In line with most of the recent literature, we use a linear probability model of 3-year recidivism by inmate j (R_j), depending on whether the sentence was served in a women-only prison, captured by a dummy (W_j), on fixed effects for the region of birth ($\gamma_{b(j)}$), the region of residence ($\tau_{r(j)}$) and several controls (X_j):

$$R_j = \beta_0 + \beta_1 W_j + \psi' X_j + \tau_{r(j)} + \gamma_{b(j)} + \eta_j. \tag{2}$$

Given the potential endogeneity of W_j, as argued in the previous section, we exploit the prison administration's decision rule (1), instrumenting the women-only prison dummy W_j using ΔD_j, the difference (in Km) between the (minimal) distances from the residence of inmate j of mixed-gender and women-only prisons. In our main specification, to calculate ΔD_j we use the prison of exit, which typically corresponds to the prison of entry, although exceptions exist. This choice is motivated by the fact that the prison of exit is more likely the one in which the largest part of the sentence was served, whereas the prison of entry might reflect only a temporary assignment. In a robustness test, we calculate ΔD_j according to the prison of entry.

Thus the first stage and the reduced form equations are the following:

$$\begin{pmatrix} W_j \\ R_j \end{pmatrix} = \phi + \theta \Delta D_j + \varphi X_j + \lambda_{r(j)} + \xi_{b(j)} + \epsilon_j, \tag{3}$$

where $\phi, \theta, \varphi, \lambda_{r(j)}, \xi_{b(j)}$ and ϵ_j are column vectors of size 2.

Under the assumption that the relative distance from the two types of prison is conditionally exogenous, we can identify the local average treatment effect: $\hat{\beta}_1 = \hat{\theta}_2 / \hat{\theta}_1$.

Given that the distances between the municipality of residence and the two types of prisons might be correlated with the characteristics of the municipality or with the ease with which the inmate can receive support from the local environment (factors that in turn might affect her post-release recidivism), we control for the distance between the

inmate's place of residence and the actual prison, for whether the inmate resides in a metropolitan area[16], for region of birth and region of residence fixed effects (the latter denoted $\xi_{b(j)}$ and $\lambda_{r(j)}$, respectively). We also control for the overcrowding rate, the inmate's age, age squared, nationality, educational status, marital status, types of crime committed, sentence status, incarceration length and the closing year of incarceration.[17]

A distinctive feature of our setting is that we can do better than buttressing the conditional independence assumption by including a large set of controls. Given the prison administration's decision rule (1), we know that the instrument should only be relevant in the range $\Delta D_j < 0$. This can be tested directly. Moreover, past that threshold, the instrument should not affect the probability of assignment to a women-only prison. A significant reduced-form effect past the 0 threshold would imply a violation of the exclusion restriction. Thus, in summary, we test for and estimate the following coefficients for the different samples:

Relevance: $\theta_1 > 0$ when $\Delta D_j < 0$

Decision Rule: $\theta_1 = 0$ when $\Delta D_j \geq 0$

Reduced-form Effect: $\theta_2 < 0$ when $\Delta D_j < 0$

Exclusion Restriction: $\theta_2 = 0$ when $\Delta D_j \geq 0$

Thus, we can test for something that is usually untestable in instrumental variable settings, the exclusion restriction of the instrument. It is worth noticing that in our setup the monotonicity assumption for a LATE interpretation of the 2SLS parameters should hold as well. There is no reason why inmates should become more likely to be sent to a specific prison as the distance from that prison increases.

3.3 Marginal and Average Treatment Effect

With heterogenous treatment effects and a continuous instrument, our Two-Stage Least Squares (2SLS) estimate of equation 2 provides a variance-weighted average of covariate-specific Local Average Treatment Effects (LATEs). The latter measure the causal effect

[16]We classify metropolitan areas those with population exceeding three million inhabitants, including the three largest cities in Italy: Rome, Milan, and Naples.

[17]We calculate the overcrowding rate as the ratio between the number of inmates and the prison's regulatory capacity. Education status is categorized as follows: no schooling, elementary school, lower secondary, and higher secondary (including high school and university degrees). Marital status is defined within three categories: married or common-law married, divorced/separate or widow, and single. We specify the types of crime as drug-related crimes, property crimes, violent crimes, and other crimes. Lastly, we define sentence status as follows: definitive sentence, under appeal, awaiting trial, mixed-status, and other statuses.

on recidivism of being assigned to a women-only prison for various groups of 'complier' inmates, that is inmates who are assigned to a women-only prison because of the relative distance from their place of residence. The interpretation of the overall IV effect, being an aggregation of LATEs, is a little involved, and might hide potentially interesting patterns of selection induced by changes between specific pairs of values of the instrument.

In our setting, with the binary endogenous assignment (to a women-only or a mixed-gender prison) and a continuous instrumental variable, we can however recover a richer array of treatment effects — the so called marginal treatment effects (MTEs) — that better characterize the selection induced by the instrument and can be aggregated to obtain the average treatment effect (ATE), something that in general would not be possible starting from the LATE estimates.[18]

Specifically, we can estimate a propensity score, as a function of covariates and of the relative distance (our instrument), to characterize the *observed* inducement into the treatment — which in our case is serving the sentence in a women-only prison. The actual assignment then results from balancing out the observed inducement with the *unobserved* resistance to treatment, which captures idiosyncratic (and unobserved by us) motives that might militate against being assigned to a women-only prison. The various LATE estimates (aggregated by the 2SLS procedure) refer to the average effect for compliers who fall in different ranges of the distribution of the unobserved resistance to treatment, while each MTE estimate is the limit as one of these ranges shrinks to a single point. Each MTE estimate then gives the treatment effect for individuals at a given point of the distribution of the unobserved resistance to treatment, in turn identified by a specific value of the propensity score.[19] Since the percentiles of the propensity score are uniformly distributed, a simple average of the MTEs yields the average treatment effect (ATE), that is the average effect in the inmate population, the policy-relevant parameter if one were to decide to treat the entire inmate population.

4 Results

4.1 Balance Tests

We start by looking at whether individual characteristics of inmates assigned to women-only or mixed-gender prisons differ. Figure 1, left panel, reports the coefficients of a linear

[18] MTE were first introduced by Björklund and Moffitt (1987) and Heckman and Vytlacil (1999) and later extended in a series of papers (see, for example, Heckman et al. (2006). For an introduction to marginal treatment effects see Cornelissen et al. (2016)).

[19] Carneiro et al. (2011) has an application of MTE to estimate returns to college education where, similar to our case, distance is used as an instrument.

probability model where the dependent variable is a dummy W_j equal to 1 if j is assigned to a women-only prison (0 otherwise), controlling for regional fixed effects. In line with the potential endogeneity of W_j, we find that several variables display a statistically significant coefficient. In the right panel, we repeat a similar analysis, considering as the dependent variable a dummy equal to 1 if $\Delta D_j < 0$. We see that in this case the observable differences between inmates, depending on whether the relative distance of the two types of prisons from their residence is positive or negative, are almost always statistically insignificant (the joint significance test has a p-value of 0.18, well above the conventional thresholds). The lack of correlation between ΔD_j and inmates' characteristics provides reassurance regarding the exogeneity of our instrument.

Table 2 shows descriptive statistics splitting inmates depending on their relative distance, $\Delta D_j < 0$ and $\Delta D_j \geq 0$. The two samples are generally comparable with some notable exceptions. Women with a positive D are more likely resident in metropolitan areas, where women-only prisons are located (22% for $\Delta D_j < 0$; 65% for $\Delta D_j \geq 0$). In turn, the share of Italians is much lower for $\Delta D_j \geq 0$ as metropolitan areas are more likely to attract migrants (73% for $\Delta D_j < 0$; 48% for $\Delta D_j \geq 0$). Women closer to women-only prisons are also three years younger, 10p.p more likely to be single and 8p.p. more likely to have committed a property crime. The two samples show a marked difference in the three-year recidivism with lower rates for women closer to mixed prisons (9,98% for $\Delta D_j < 0$; 15,13% for $\Delta D_j \geq 0$): this difference might be driven by such imbalances in terms of citizenship, demographics and type of crime.

4.2 Non-parametric Evidence

Figure 2 shows the fraction of inmates assigned to a women-only prison as a function of ΔD_j. We see that the fraction increases with ΔD_j, as long as $\Delta D_j < 0$, while it is broadly constant in the range $\Delta D_j \geq 0$. In plain English, inmates whose residence is only slightly nearer to a mixed-gender prison than to a women-only one are more likely to be assigned to the latter than inmates whose residence is much closer to a mixed-gender prison than to a women-only one. Once the women-only prison is the closer one, the probability of being assigned there is no longer significantly affected by the relative distance.

The figure is a graphical representation of the first stage of our Instrumental Variable (IV) strategy. The increasing part, for values of $\Delta D_j < 0$, highlights the relevance of the instrument. The broadly flat part, for values of $\Delta D_j \geq 0$, shows that the instrument does not predict assignment to a women-only prison for inmates whose residence is closer to that kind of prison, in line with the decision rule described in Section 2.

In Figure 3, the blue line shows the decreasing relationship between our instrument (in the negative range) and recidivism. This is the graphical representation of the reduced

form in our IV strategy, and confirms the relevance of the instrument. The Figure also shows, as a green line, the fairly flat relationship between the instrument and a measure of predicted recidivism. The latter is obtained from a logistic regression model, in which recidivism is projected on all the observables considered in Figure 2.[20] The flat line, implying the absence of a correlation between ΔD_j and (a function of) those observables, confirms that the relationship between the instrument and (actual) recidivism, shown by the blue line, is not capturing the effect of other observables. It is, therefore, reassuring that the exogeneity of the instrument is present.

4.3 Instrumental Variable Regressions

Table 3 presents the estimates of equation 3, with the first stage in panel A, the reduced form in panel B and the IV estimate in panel C. The bottom part of the table reports the Kleibergen-Paap statistics (KP F-Stat) for weak identification and the fitted values of recidivism, under the two alternative assumptions that **all** the inmates are either in a mixed-gender ($\hat{R}_M$) or in a women-only prison ($\hat{R}_W$), leaving all other covariates unchanged. In the first 3 columns the estimates are restricted to the sample of inmates whose residence is closer to a mixed-gender prison than to a women-only one ($\Delta D_j < 0$), in the following 3 columns the sample only includes inmates whose residence is closer to a women-only prison ($\Delta D_j \geq 0$). In column 1 we include no controls nor fixed effects, with the exception of the dummy indicating whether or not the residence of the inmate is within a metropolitan area.[21] Hence the coefficient of interest (the effect on recidivism of being assigned to a women-only prison; β_1 in the notation of equation 2) is identified by the variability that results from all possible reciprocal positions of the inmates residence, the (closest) mixed-gender prison and the (closest) women-only prison, provided that we only compare inmates whose residence is equally within (or equally outside) a metropolitan area (though not necessarily the same). We see that both the first stage and the reduced form are highly significant, confirming the relevance of the instrument (in the notation of equation 3, $\theta_1 > 0$ when $\Delta D_j < 0$), with a large value of KP F-Stat, and the presence of a reduced-form effect. The overall effect is about -7.9%. Hence the fitted value of recidivism would drop from 13.7%, if all inmates were housed in mixed-gender prisons, to 5.9%, if they were all moved to women-only prisons.

In column 2 we include the controls listed in Section 3.2, so we restrict the variability identifying the coefficient of interest, as we compare inmates who differ in terms of the relative distance to the two kinds of prisons, but have the same values for all the included controls. Both the first stage and the reduced form estimates remain significant, with

[20] Appendix Figure A1 shows the high correlation between recidivism and predicted recidivism.

[21] This dummy will always be included in all the specifications discussed in this and in later Sections.

little change in value, and the overall effect becomes slightly more negative, to −9.2%. In column 3 we include, besides the controls, region of birth and region of residence fixed effects. Including the latter greatly restricts the variability of the instrument used to identify the causal effect. In particular, since different regions might be more or less conducive to recidivism,[22] and, at the same time, might be differentially endowed with women-only prisons[23] comparing inmates who live in the same region allows us to clean the effect of being assigned to a women-only prison from the possible (positive) correlation between recidivism and presence of women-only prisons. Since this correlation would partially offset the pure effect of the assignment, we see that the latter is substantially increased (in absolute value), with an estimated value of −16.4%. In this case, therefore, the fitted value of recidivism would drop from 17.8%, if all inmates were housed in mixed-gender prisons, to 1.4%, if they were all moved to women-only prisons.

As mentioned, we can directly test the exclusion restriction, namely the requirement that the instrument affects recidivism only through its effect on prison assignment. Columns 4 to 6 replicate the same specifications of columns 1 to 3 on the sample of inmates whose residence is closer to a women-only prison ($\Delta D_j \geq 0$). The assignment rule described in Section 2 implies that positive values of the variable ΔD_j should have no effect on the prison assignment. Therefore, under the validity of the exclusion restriction, positive values of the variable ΔD_j should also have no effect on recidivism. Hence, in columns 4 to 6, we should observe both an insignificant first stage and an insignificant reduced form. This is precisely what Table 3 shows.

4.4 Robustness Tests

As mentioned before, we compute the instrument ΔD_j on the basis of the prison of exit. In Table A1 we verify that our results are robust to this choice by computing the instrument on the basis of the prison of entry. The results are very similar to the ones in Table 3, with the range of IV estimates from 6 to 14%.

In Table A2 we compute recidivism as re-incarceration within a shorter window, namely 1-year and 2-year after release. We show that the effect of spending the sentence in a women-only prison is already apparent in the shorter time window, with the IV estimate ranging between 7 and 13 percentage points.

A further robustness check concerning the definition of recidivism is presented in Table A3. Instead of including all re-incarcerations after release, irrespective of whether the

[22] For example, regions might differ for the presence of organized crime or for labor opportunities available to released inmates.

[23] For example, it might be the case that women-only prisons were created precisely in those regions where there is more recidivism.

latter occurred after a definitive sentence,[24] we define recidivism as reincarceration after release from a definitive sentence. While this reduces the sample size and yields somewhat less precise estimates, the effect of assignment to a women-only prison are broadly similar to those observed in the main specification.

Finally, we consider two dimensions of potential heterogeneity in our results: foreign status and the type of crime. Our sample includes foreign inmates, whose recidivism might respond differently to the prison assignment. Table A4 replicates the analysis on the sample that only includes Italian inmates. The IV estimates are somewhat larger (in absolute value) than those in Table 3, particularly in the most restricted specification, and highly significant.[25] Table A5 replicates the analysis distinguishing between violent and property crimes and shows that for both types of crime the assignment to a women-only prison reduces recidivism. Depending on the specification, the effect is somewhat stronger for one or the other of the two types of crime, and is in the same ballpark of what found for the entire sample.

4.5 Results for Marginal and Average Treatment Effects

We start the MTE analysis by estimating the propensity to be assigned to a women-only prison using a probit model. The controls are the same used in column 3 of Table 3, although we get similar results when we do not use regional fixed effects. Figure A3 shows the support histogram of the propensity score for inmates in women-only (treated) and mixed-gender (control) prisons and highlights two important facts: the support in the two samples is common, and it covers the entire range $[0, 1]$. These imply that we can use the semiparametric local instrumental variables MTE specification. For robustness, we estimate the MTE function using both a parametric and a semiparametric model.

Figure 4 shows that the semiparametric local linear MTE curve is slightly upward sloping, which means that inmates with a larger unobserved resistance to end up in a women-only prison show smaller reductions in recidivism.

Inmates are thus selected on the basis of unobservables to maximize treatment effects. However, the slope is fairly flat and Table 4 shows that ATEs are only slightly smaller than the main local average treatment effects estimated in Table 3. This means, importantly, that scaling up treatment to the entire prison population, thus increasing the number of women-only prisons by a factor of about 4 (right now about one-fifth of women are housed in four women-only facilities), would drastically reduce recidivism.

[24] The Italian system has two levels of appeal, but a condemned inmate can be in prison while waiting for one of the appeals or because she decided not to appeal.

[25] We do not include a separate test for foreign inmates due to the small sample size.

4.6 Mechanisms

Women-only prisons differ from mixed-gender ones along several dimensions (as documented in Table 1). They typically offer enhanced healthcare services, improved living areas, and a more comprehensive array of educational and vocational opportunities for inmates. We are considering, in other words, a case of bundled treatment, and it is therefore difficult to single out one (or more) of these aspects as being the main one(s) responsible for the effects on recidivism that our analysis identifies. We try to tackle this difficulty in two ways. First, somewhat indirectly, by looking at factors that might enable, or prevent, the design and organization of facilities with these different bundles of aspects. Secondly, arguably more directly, we select some features that we conjecture might be drivers of our results and we repeat our analysis on the effect of two kinds of prisons on recidivism, but instead of contrasting women-only and mixed-gender prisons, we use latent class analysis to repeatedly split the prisons into two groups, according to those features. The idea is to verify whether one (or more) of these features, in and of themselves, and irrespective of whether the prison is a women-only or a mixed-gender one, are associated with a reduction in recidivism.

As to the first approach, one plausible conjecture is that tailoring prison structures to better suit the specific needs of female inmates is both a more pressing concern and easier to implement when women represent a large enough share of the population of inmates housed in a given facility.[26]

To test this conjecture, in Table A6 we compute again the effect on recidivism of being assigned to a women-only prison (as in Table 3) but, in columns 1 and 2, respectively, we restrict the control group to mixed-gender prisons with a relatively high, and relatively low, proportion of women inmates (above and below the median of the distribution). Consistently with the conjecture, the effect on recidivism of the assignment to women-only prisons is more than 3 times larger when the control group only includes prisons with relatively fewer women.[27] Similarly, in columns 3 and 4 we consider the possible role of absolute, instead of relative, size. In column 3 the control group only includes mixed-gender prisons with relatively larger (above the median) female sections, while in column 4 it includes only mixed-gender prisons in which the size of the female section is below the median. Again, the role of size seems confirmed, with the effect on recidivism about 2 times as large when the control group is the one with smaller sized female sections.

[26] According to a report from the Italian Ministry of Interior (Palmisano, 2015), the limited number of women inmates in mixed-gender facilities often leads to neglect and a lack of engagement in activities tailored to their needs, posing challenges for implementing targeted interventions.

[27] For the sake of space, we only report the most restricted specification with all the controls and the fixed effects. The appropriate comparison, therefore, is with column 3 of Table 3.

In columns 5 and 6, we consider another factor, not necessarily alternative, that might facilitate (or hinder) the organization of correctional facilities better suited to the needs of women inmates: the gender of the prison director. The plausible conjecture, in this case, is that a woman director is more attuned to other women needs. In column 5, we use as a control group only mixed-gender prisons with a male director, while in column 6, we focus on mixed-gender prison with a female director. Again, the results lend support to the conjecture. The effect on recidivism of being assigned to a women-only prison is about 2 times as strong when compared with mixed-gender prisons directed by a men. This finding echoes research in political economy highlighting how the presence of female leaders can enhance the welfare of the female population (Chattopadhyay and Duflo, 2004; Pande and Ford, 2009; Bochenkova et al., 2023). However, it is important to note that this result should not immediately advocate for replacing male directors with female ones, as mixed-gender prisons predominantly house male inmates, and we have so far no evidence of the potential impacts of having a female director on the male population.

Moving now to our second approach, we conjecture that the availability of working facilities and opportunities to carry out work outside the prison walls or inside for external firms, as well as the presence and quality of sanitation and health services, might be features that facilitate a smoother return to a normal life after serving the sentence, and therefore reduce recidivism. Based on previous research (see Mastrobuoni and Terlizzese (2022)) and results just reported, we also attribute a similar potential role to social activities withing the prisons, like the opening of prison cells during the day (in excess to the minimum required by law), and to the female gender of the prison director.

Accordingly, we construct four binary classifications of the universe of Italian prisons: based on work facilities and external work (denoted Work), on sanitation and health services (denoted Sanitation), on the social activities (denoted Social), on whether the prison director is a female (denoted Female director). We assign prison to be work-prone, social-prone and health-prone based on a series of observable characteristics. Based on three separate factor analyses, we define proness, based on whether the first factor is above the median.

Figure A4 show that, for each of these four binary classes, female-only prisons are more likely to present the characteristic(s) that define the class.

For each of these four splits we compute, as in our previous analysis, the distances from the residence of each female inmate to the closest prison belonging to each of the two classes identified by the split, and we take the difference between these two distances. We label classes in such a way that a negative difference means that the (closest) prison in the class that we conjecture to be “better", that is more conducive to a reduced recidivism, is farther away for the inmate’s residence than the (closest) prison in the “worse" class. We

then use this difference as an instrument for the effect of the set of features that identify the prison class on recidivism.

Figure A5 shows, for each of the four binary classes, the first stage and the reduced form. The first stage gives the fraction of inmates assigned to the “better" prison class, for various values of the distances difference. The reduced form gives the corresponding fraction of recidivating inmates.

As to the first stage, given the underlying preference by the prison administration of assigning the inmate to the prison closest to her residence, we see two possibilities. If the prison administration has no strong preference for the features that define a given class, we should observe a fairly flat pattern, at a low level, for negative differences, and again a fairly flat pattern, at a much higher level, for positive differences. This is because, absent other considerations, inmates would tend not to be assigned to the "better" class when the latter is farther away form their residence, while would predominantly be assigned to it when it is closer. There would therefore be a sizeable jump at zero. The other possibility is that the prison administration trades off the closeness to residence with some desirable features of the class (as in our earlier analysis of mixed-gender vs female-only prisons). In this case we should observe an increasing pattern to the left of zero, and a fairly flat pattern to the right. Both possibilities are likely to imply a significant first stage, even though the first would suggest a regression discontinuity approach as more appropriate, since away from zero the instrument should have little relevance.

The first stages shown in Figure A5 are all statistically significant. They tend to be of the second type, without jumps. The black colored data point are those which correspond to the range of relevance of the instrument, i.e. those for which a change in the value of the distances difference is associated to a change of the fraction of inmates associated to the “better" class.

Moving to the reduced form, we see that, in the range of relevance of the instrument, there is some evidence of a reduction of recidivism for the classes Work. Overall, tough, there is no single mechanisms that can explain the women-only result.

5 A new prison system

In this section, we present two proposals for the allocation of incarcerated women to women-only prisons. The first proposal is based on the straightforward intuition that, within a given region, existing mixed-gender prisons could be repurposed exclusively for women by reallocating the current male population. This approach would transform selected mixed prisons into women-only facilities. The second proposal involves the construction of new women-only prisons, strategically located to serve areas with high

concentrations of incarcerated women.

5.1 A low hanging fruit: Re-allocation to existing prisons

The official prison capacity for women inmates in Italy, besides the four existing women-only prisons, is of almost 1600 places, and is present in almost all the Italian regions. Setting aside possible technical difficulties, we conduct in this section the following thought experiment: within each region (except the two smaller regions which have no official capacity for women), we imagine to repurpose one or more of the existing mixed-gender prisons, transforming them into women-only facilities, so as to create enough places to host all the official capacity for women currently housed in mixed-gender prisons. In this way, all the women inmates envisaged in the official capacity of the Italian prison system would be housed in women-only prisons.

Clearly, it is only by chance that in a region there is one prison that has (or several ones that in total have) a number of places exactly equal to the total number of places for women inmates in the region, so that the women reallocated would free just enough places to house the males that were originally in the repurposed prison.[28] Nevertheless we verified that in each region the reallocation can be done in such a way that the excess number of places envisaged for male inmates does not exceed 3 percent of the regional residual official prison capacity for males.[29] We believe that, given the limited size of the deficit, and assuming that there could be some reorganization and rationalization of the physical space, the reallocation could be easily accommodated.

Having checked that, at least on the basis of the current distribution of the official prison capacity, the Italian prison system could become a women-only one. We can easily compute the reduction in recidivism that this would entail, given our estimates. The shift to a women-only prison would reduce the yearly number of recidivating women by a minimum of 126 to a maximum of 261 (corresponding to the range of our estimates of the effect on recidivism of women-only prison).

5.2 Re-allocation to new prisons

The above allocation strategy requires a sufficient number of prisons that can be converted and may face additional challenges in practical implementation. For example, the housing conditions of existing prisons may be inadequate for women or it may be infeasible to relocate prison guards. Since policymakers may prefer a solution focused

[28] In the current regional distribution of prison capacity, this coincidence only happens once.

[29] In fact, in most cases the male deficit is smaller than 1 percent. Only in one case, in the sparsely populated region of Sardinia, reaches 11 percent.

on the construction of new dedicated facilities for women, we propose a simple model to determine the optimal number, capacity, and locations of new prisons in the Italian context.

The model addresses a simplified decision-making problem that considers two key factors: geographical constraints and prison capacity constraints. Geographical constraints focus on minimizing the distance between a woman's place of residence and the assigned prisons. The model ensures that each incarcerated woman is placed in a facility within a predefined maximum distance (cutoff) from her municipality of residence, while also respecting the capacity limitations of each prison.

Although determining the exact prison capacities is beyond the scope of this paper, it is important to recognize the trade-offs involved: Larger prisons may generate economies of scale, at the expense of increasing the average distance between the prison and the inmates' places of residence. Since we cannot observe costs, we simplify the trade-off by setting a maximum capacity of each prison at 300, which is in line with the observable largest existing women-only facility.

Second, women are only assigned to prisons within 300 km of their residence, which corresponds to the 90% percentile of such distance in our data.[30]

To estimate the expected number of women to be allocated and their place of residence, we use data on the stock of incarcerated women in 2021 and 2022, which include 2260 women inmates.[31]

The model employs the KMeans clustering algorithm to determine optimal locations for new women-only prisons, based on the geographic distribution of incarcerated women. KMeans is an algorithm that partitions data into clusters by minimizing the distance between data points and their assigned cluster centers. In this context, each cluster center represents a potential prison location, calculated by weighting the coordinates of municipalities according to their population of incarcerated women.[32]

Once the potential prison locations are determined, the model proceeds to assign

[30] These criteria serve as reference points and can be easily adjusted as needed.

[31] A potential limitation of this approach is the possibility that the spatial distribution of women may shift in the future. However, our data indicates that this is unlikely to be a significant issue. Over the years in our data, the correlation between the spatial distribution of women has consistently remained above 0.9, demonstrating strong stability in these geographical patterns.

[32] KMeans aims to minimize the following objective function:

$$J = \sum_{i=1}^{k} \sum_{x \in C_i} \|x - \mu_i\|^2, \tag{4}$$

where k is the number of clusters, x represents the data points (municipality coordinates), μ_i is the centroid of cluster i, and C_i is the set of points assigned to cluster i. The algorithm iterates to find the centroids μ_i that minimize the sum of squared distances between points and their respective centroids.

women from each municipality to the nearest prison that meets both the distance and capacity requirements.[33] Women are assigned as follows:

1. If the nearest prison is within 300 km and has sufficient capacity, the women from that municipality are assigned to that prison.

2. If the nearest prison is full or exceeds the distance cutoff, the model checks the next closest prison for available capacity and distance compliance.

3. If no prison meets these criteria, the women from that municipality remain unassigned, indicating a need for additional prisons. The model iteratively increases the number of prisons (clusters) if some women remain unassigned.

The model offers three main outcomes. First, it determines the optimal number and locations for new women-only prisons using geospatial clustering. Second, it shows how many women are assigned to each prison. Lastly, the model calculates the average distance from each woman's home to her assigned prison.

The figure below illustrates the outcome of this simulation. The model identifies 11 optimal locations for new women-only prisons distributed across Italy. These facilities have an average capacity of 205 women and are, on average, located 75 km away from the women's municipalities of residence. This average distance is shorter than the current actual distance from residence to prison which is 93 km. Appendix Table 5 presents the allocation of women to each of the new prisons, along with the average distance from their municipality of residence to the assigned facility.

6 Conclusions

In 143 countries, incarcerated women serve their sentence in a separate section within prisons that also house male inmates. In this paper, we show that, once we correct for the possible selection and other confounding factors, women that are instead incarcerated in women-only prisons display substantial lower recidivism in the three years after their release.

Our results are based on Italian data, where both women-only and mixed-gender prisons coexist and where the institutional details of the process allocating women convict to prisons allows us to mimic a quasi-random assignment, and thus identify the causal effect on recidivism of the two types of prisons.

[33]The model always accounts for a new prison in Rome, as the municipality hosts a very high number (289, in our exercise) of inmates living in the municipality.

Women-only prisons display better features upon several dimensions, including health, social and work services. Our findings are most likely the result of a bundle effect whereby women-only prisons can be more easily customized around women needs and allow for more effective rehabilitation initiatives. We show that both the size (relative and absolute) of the female inmates within a given facility, and the gender of the director, are relevant factors the difference in recidivism that we estimate.

Despite a nearly 60% increase in the global population of incarcerated women over the past two decades, the issues surrounding women's incarceration and recidivism remain largely neglected. Our findings bring some light on one of those issues, namely whether serving a sentence in a women-only prison leads to a lower recidivism than serving it in a mixed-gender one. They are therefore relevant for governments and international organizations interested in designing policies to reduce crime with a specific focus on women's incarceration. Recognizing and addressing this issue is crucial in fostering safer communities and promoting equitable justice systems worldwide.

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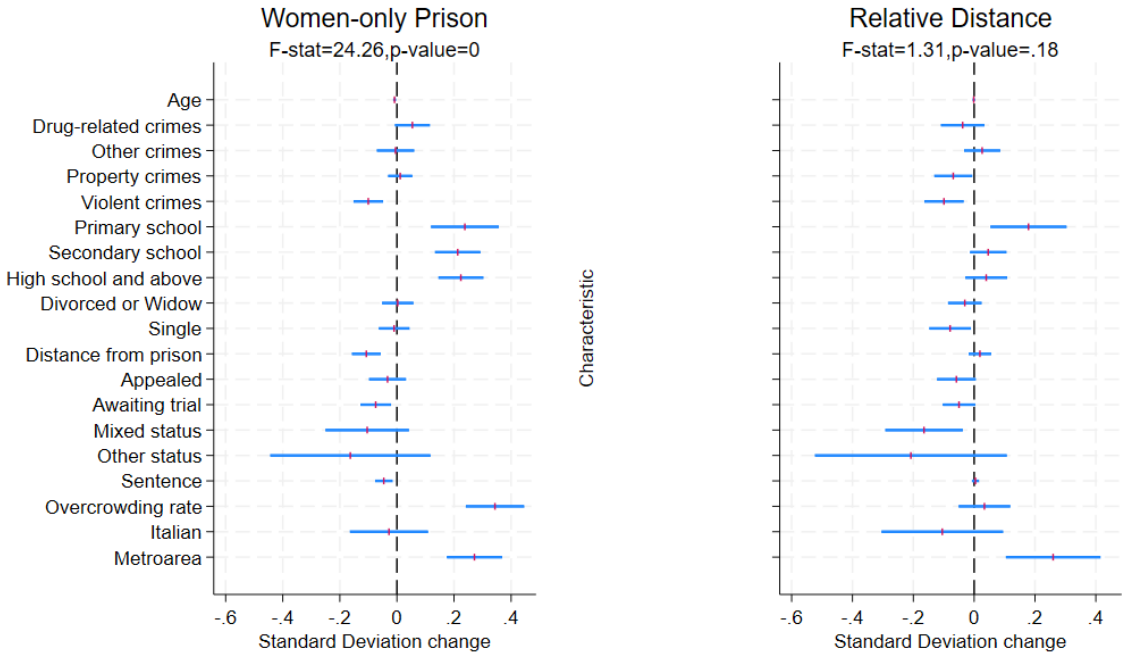
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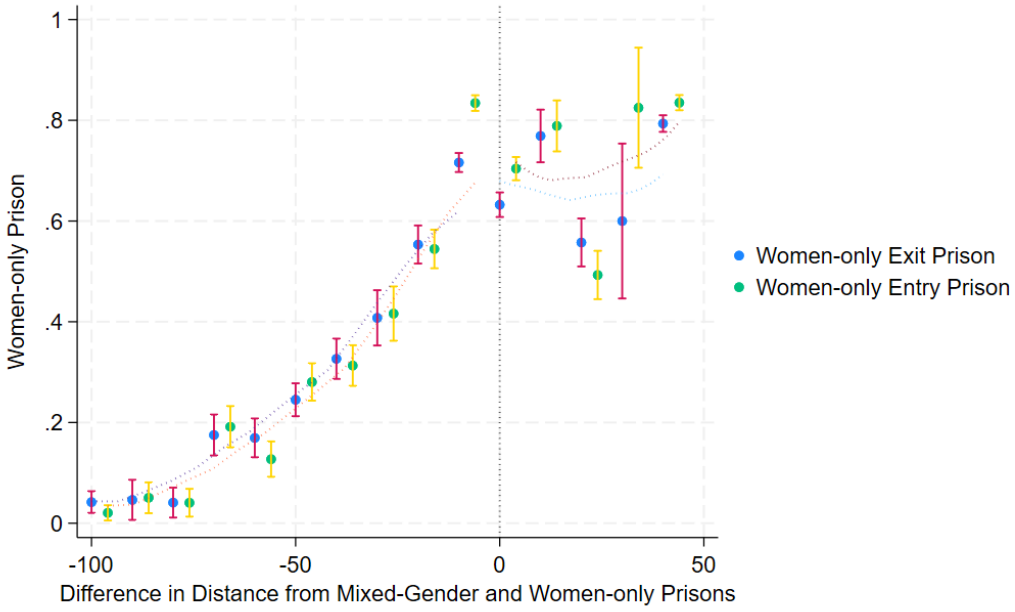
7 Figures and Tables

Figure 1: Balance across prisons and relative distance



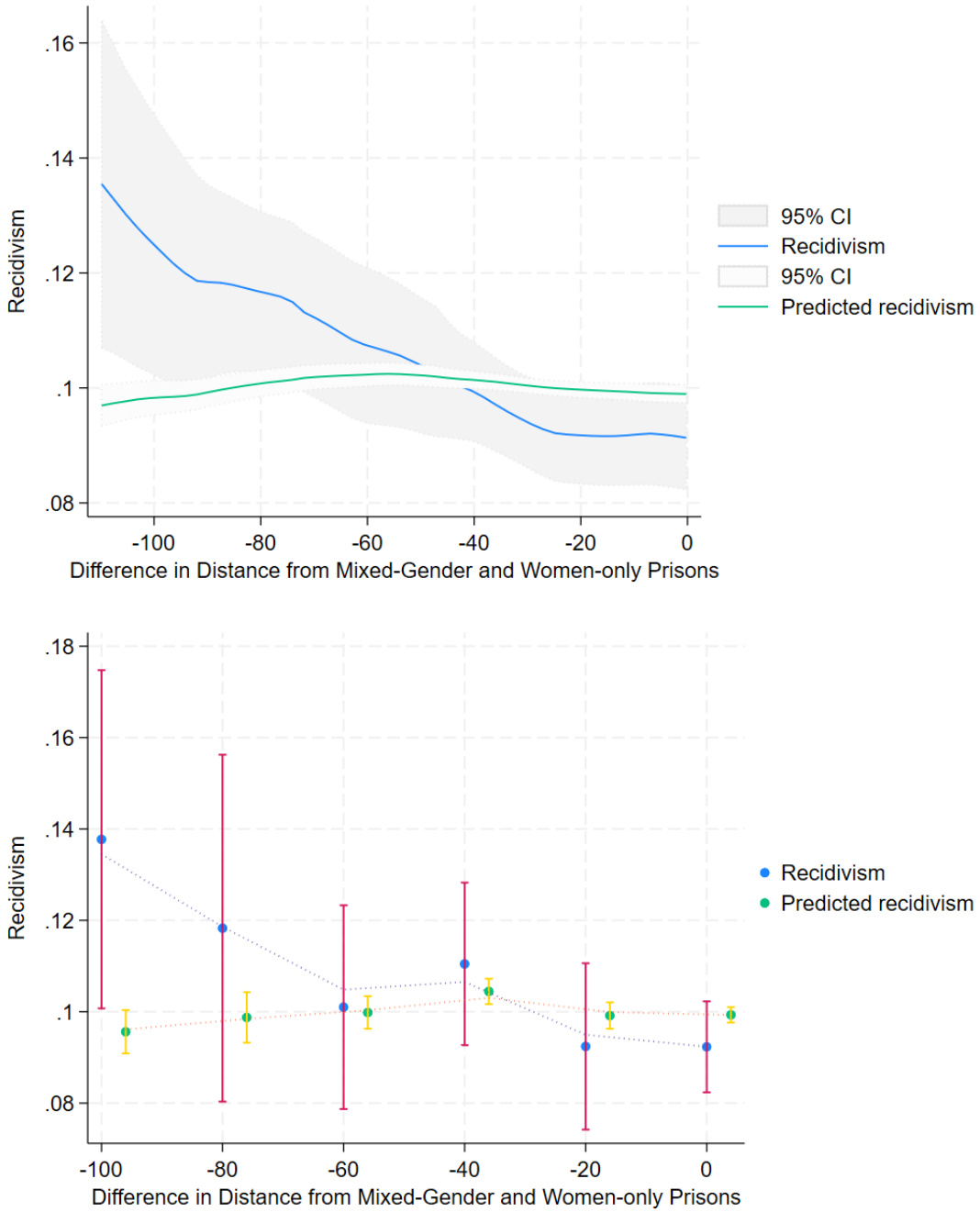
Notes - The figure reports the coefficients of a linear probability model and the corresponding 95% confidence intervals, controlling for regional fixed effects. In the left panel, the dependent variable is the dummy *Women-only Prisons*, while in the right panel, it is the *Prison relative distance*. For both specifications, the joint significance test is reported.

Figure 2: Women-only Prison and difference in distance: First stage



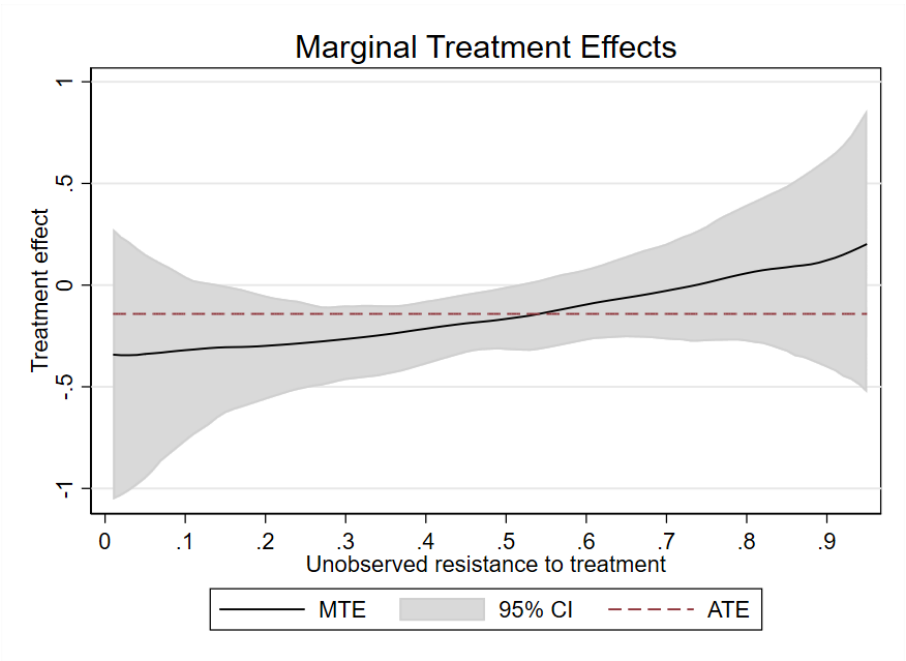
Notes - The figure shows the estimated probability to be assigned to a women-only prison of entry or exit and the corresponding 95% confidence intervals. Distance is divided into groups every 10 km.

Figure 3: Difference in distance and recidivism (or predict residual recidivism.)



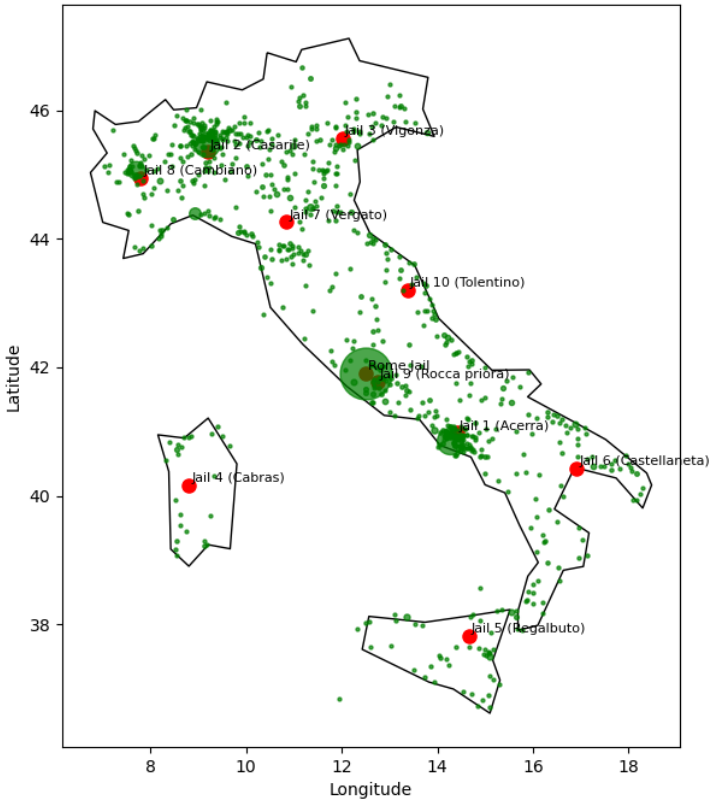
Notes - The figure shows the recidivism rate and predicted recidivism rate (and the corresponding 95% confidence intervals) against the relative distance between the place of residence of inmates and the two types of prison (mixed-gender vs. women-only). The upper panel plots the local linear regression, while in the lower panel averages are computed every 20 km.

Figure 4: Marginal Treatment Effects



Notes - The Figure shows the MTE curve and the Average Treatment Effects (ATE). The propensity score is estimated using a probit model (see Figure A3 for the common support). The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The controls are the same ones used in Column 3 of Table 3. Standard errors are estimated using 200 bootstrap replications.

Figure 5: Simulated New prisons Locations



Notes - The figure shows the outcome of the model simulating the allocation of new women only prisons across Italy.

Table 1: Women-only vs Mixed Gender Prisons

	Women-only	Mixed	Difference	Std. Err.
Health				
Heating	1.000	0.865	0.135	0.001
Hot Water	1.000	0.322	0.678	0.001
Separate WC	0.941	0.964	-0.022	0.000
Shower	0.337	0.278	0.059	0.001
Medical Doctor	0.969	0.943	0.026	0.001
Gynecologist	0.975	0.860	0.115	0.001
Obstetrician	0.270	0.407	-0.137	0.001
Social Spaces				
School Spaces	1.000	0.997	0.003	0.000
Working Spaces	1.000	0.804	0.196	0.001
Non-catholic Religious Spaces	0.049	0.113	-0.064	0.001
Gym	0.718	0.641	0.077	0.001
Sports Facilities	0.994	0.834	0.160	0.001
Hours Spent Outside the Cell	4.738	4.155	0.582	0.003
Freedom of Movement in Prison	0.990	0.474	0.516	0.001
Remote Meetings	0.335	0.414	-0.079	0.001
Web Access	0.000	0.026	-0.026	0.000
Work & Other				
Share Working Women Inside	0.289	0.353	-0.064	0.000
Share Working Women Outside	0.031	0.062	-0.032	0.000
Share Training Women	0.087	0.089	-0.002	0.001
Share Woman in School	0.300	0.262	0.037	0.001
Number of Women	288.851	54.906	233.945	0.096
Prison Occupancy Share	0.734	0.960	-0.227	0.001
Share of Foreigners	0.377	0.329	0.048	0.001
Share on Parole	0.010	0.010	0.000	0.000
Share of Disciplinary Measures	0.136	0.079	0.057	0.000
Self-Harm	0.115	0.200	-0.085	0.001
Female Director	0.994	0.720	0.274	0.001
%Positive description	0.364	0.443	-0.080	0.001
Recidivism Rate	11.958	11.479	-0.479	0.645

Notes: Prison population-weighted mean differences between the two types of prisons. Data come from the Antigone survey for the year 2019. *Positive description* accounts for the fraction of positive words relative to the total number of positive and negative words used by Antigone's representatives to describe prison conditions. *Recidivism Rate* difference is calculated as a mean-comparison test using DAP data.

Table 2: Summary Statistics

Sample	Full		$\Delta D_i < 0$		$\Delta D_i \geq 0$	
	Mean	SD	Mean	SD	Mean	SD
Recidivates	11.75	32.20	10.00	30.01	15.14	35.84
Women-only (Exit)	0.57	0.50	0.48	0.50	0.75	0.43
Relative distance (Entry)	-10.65	43.52	-35.45	30.25	37.39	17.06
Relative distance (Exit)	-10.08	42.69	-34.58	29.14	37.39	17.06
Year of release	2015.23	2.54	2015.25	2.55	2015.19	2.51
Age (10 yrs.)	39.10	23.10	40.07	27.03	37.23	12.13
Italian	0.64	0.48	0.73	0.45	0.48	0.50
Drug-related crimes	0.38	0.48	0.40	0.49	0.33	0.47
Other crimes	0.13	0.34	0.13	0.34	0.13	0.33
Property crimes	0.41	0.49	0.38	0.49	0.46	0.50
Violent crimes	0.33	0.47	0.33	0.47	0.33	0.47
Metropolitan area	0.36	0.48	0.22	0.41	0.65	0.48
Primary school	0.16	0.37	0.18	0.39	0.11	0.32
Secondary school	0.34	0.47	0.34	0.47	0.32	0.47
High school and above	0.18	0.38	0.17	0.38	0.19	0.39
Divorced or Widow	0.16	0.37	0.18	0.38	0.12	0.33
Single	0.44	0.50	0.37	0.48	0.59	0.49
Distance from prison	57.91	111.56	60.48	106.77	52.91	120.15
Appealed	0.14	0.34	0.13	0.34	0.14	0.35
Awaiting trial	0.41	0.49	0.44	0.50	0.37	0.48
Mixed status	0.03	0.17	0.03	0.17	0.03	0.18
Sentence	1.34	1.93	1.43	2.00	1.19	1.78
Other status	0.00	0.06	0.00	0.06	0.00	0.05
Overcrowding rate	1.37	0.55	1.40	0.62	1.32	0.38
Observations	10169		6707		3462	

Notes: This table reports the summary statistics. Data comes from DAP.

Table 3: Effect of Women-only Prison on Recidivism.

	$\Delta D_i < 0$			$\Delta D_i \geq 0$		
	(1)	(2)	(3)	(4)	(5)	(6)
	Panel A: First Stage - Women-only (Exit)					
Relative distance (Exit)	0.007***	0.006***	0.005***	-0.004	-0.004	-0.004
	(0.001)	(0.000)	(0.001)	(0.004)	(0.004)	(0.004)
Mean women-only		0.477			0.747	
	Panel B: Reduced Form - Recidivates					
Relative distance (Exit)	-0.056***	-0.055**	-0.075***	-0.001	0.003	-0.071
	(0.019)	(0.021)	(0.020)	(0.122)	(0.094)	(0.099)
Mean recidivism (in %)		10			15.14	
	Panel C: IV- Recidivates					
Women-only (Exit)	-7.905***	-9.209**	-16.429***			
	(2.902)	(3.738)	(4.826)			
Controls		✓	✓		✓	✓
Region of birth FEs			✓			✓
Region of resid. FEs			✓			✓
Observations		6,707			3,462	
KP F-Stat	188.4	149.6	63.71			
$\widehat{R}_M$ (in %)	13.78	14.40	17.84			
$\widehat{R}_F$ (in %)	5.871	5.189	1.414			

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the exit prison. Columns 1-3 report results for $\Delta D_i < 0$, while columns 4-6 for $\Delta D_i \geq 0$. We report the Kleibergen-Paap for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.

Table 4: Marginal Treatment Effects

	(1)	(2)	(3)
ATE	-0.128	-0.146*	-0.141
(0.0916)	(0.0744)	(0.0907)	
ATT	-0.172**	-0.189**	-0.182*
	(0.0873)	(0.0913)	(0.102)
ATUT	-0.0875	-0.105	-0.0974
	(0.198)	(0.157)	(0.159)
LATE	-0.151***	-0.158***	-0.148**
	(0.0512)	(0.0473)	(0.0614)
Semiparametric			√
Polynomial	3	1	-
Region FEs	√	√	√
Observations	6,653	6,653	6,653

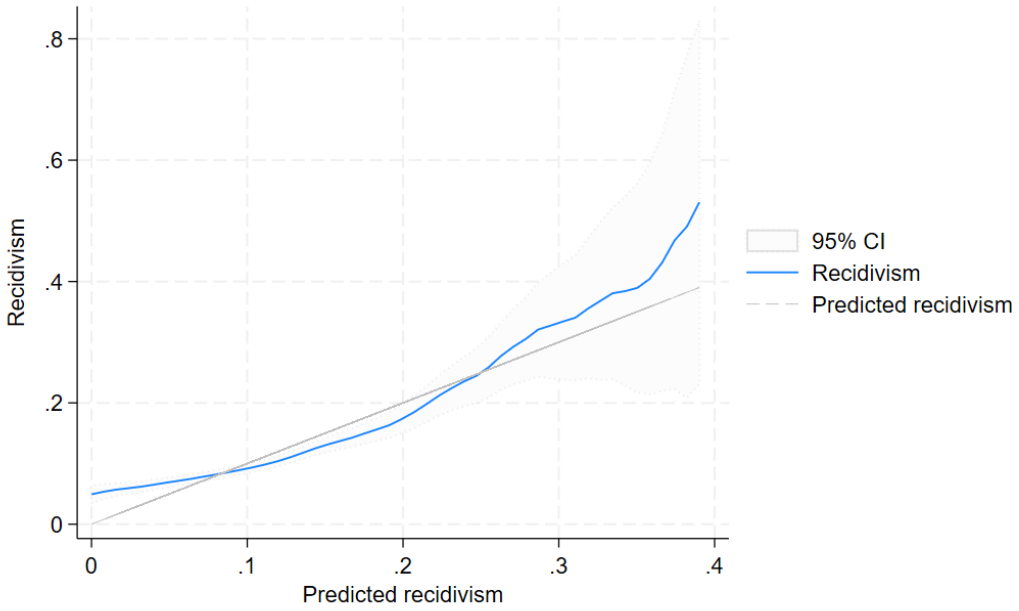
Notes - The Table shows Average Treatment Effects (ATE), Average Treatment Effects on the Treated (ATET) and untreated (ATUT) and Local Average Treatment Effects (LATE) computed using Marginal Treatment Effects estimated with parametric and sempiparametric Local Instrumental Variable specifications. The propensity score is estimated using a probit model (see Figure A3 for the common support). The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only. The controls are same ones used in Column 3 of Table 3. Standard errors are estimated using 200 bootstrap replications.

Jail	**Women Assigned**	**Average Distance (km)**	**Region**
Jail 1 in Acerra	300	37.37	Campania
Jail 2 in Casarile	287	56.80	Lombardy
Jail 3 in Vigonza	169	68.64	Veneto
Jail 4 in Cabras	39	84.01	Sardinia
Jail 5 in Regalbuto	120	95.39	Sicily
Jail 6 in Castellaneta	116	88.99	Apulia
Jail 7 in Vergato	275	129.02	Emilia-Romagna
Jail 8 in Cambiano	251	87.33	Piedmont
Jail 9 in Rocca priora	295	100.47	Lazio
Jail 10 in Tolentino	119	83.33	Marche
Jail 11 in Roma	289	0.00	Lazio

Table 5: Allocation of women and the average distance from each of the new prisons, by region.

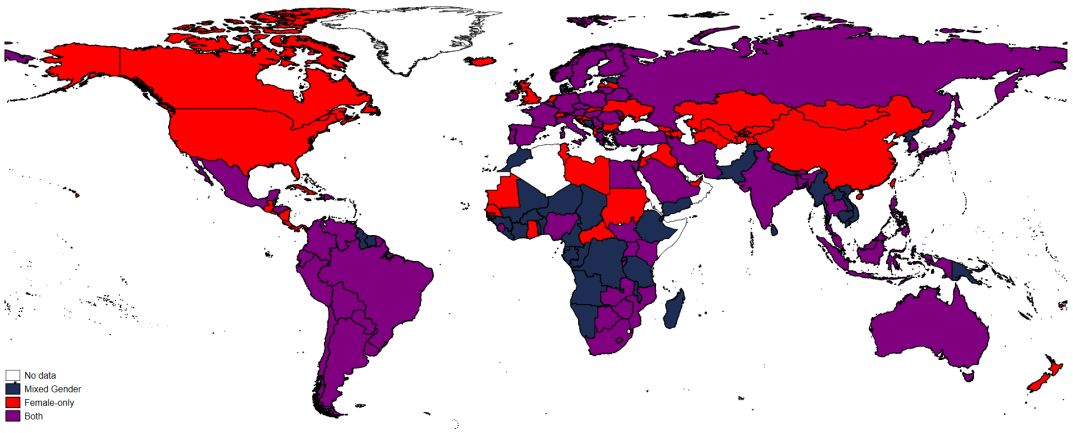
A Appendix

Figure A1: Predicted Recidivism



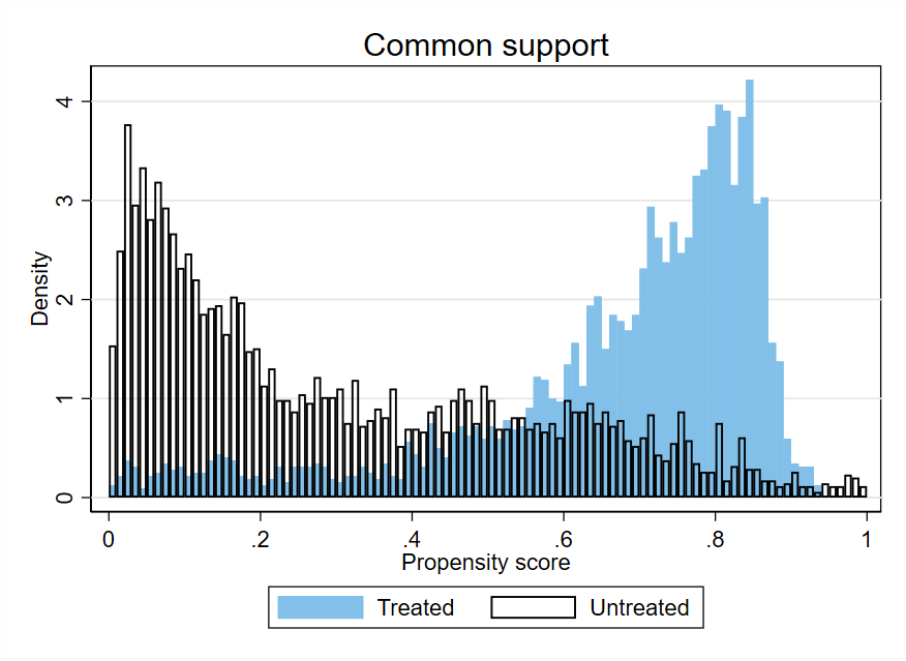
Notes - The figure plots local linear regressions for recidivism against predicted recidivism.

Figure A2: Types of Prisons Across the World



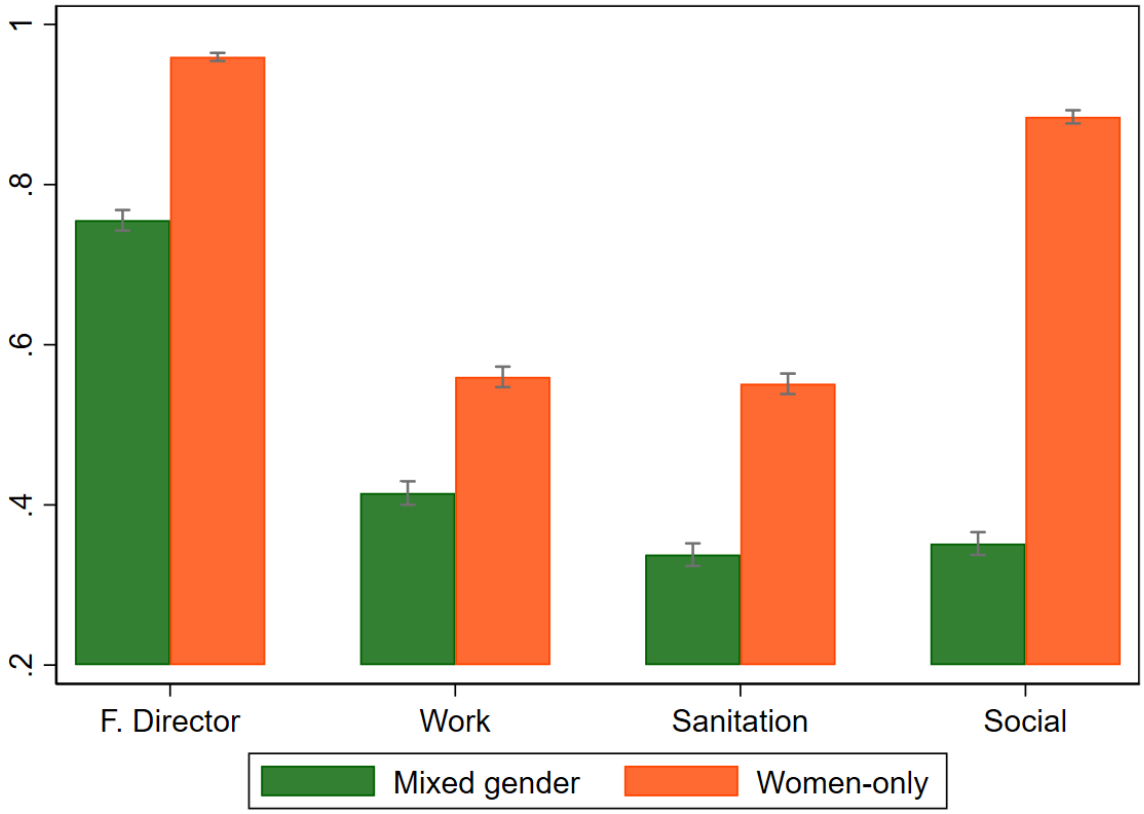
Notes

Figure A3: Propensity Score



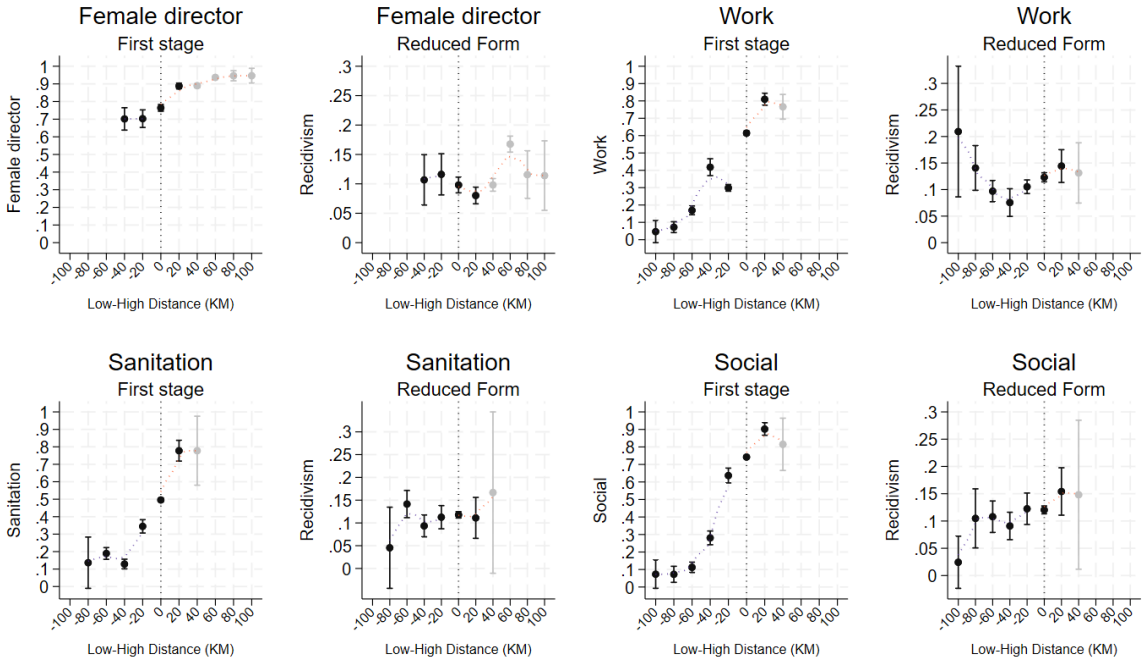
Notes - The figures shows that histogram of the propensity score for inmates sent to the women-only prisons (treated) and mixed-gender prisons (untreated). The propensity score is estimated with a probit model that contain the variables included in column 3 of Table 3.

Figure A4: Mechanisms by Type of Prison



Notes - The figure shows the average characteristic by type of prison, as well as the corresponding 95% confidence intervals. Prisons are classified based on the following characteristics (treatments): whether the director is a woman, work facilities and external work, sanitation and health services, and whether prison cells are open during the day.

Figure A5: Potential Mechanisms



Notes - The figure shows the likelihood of treatment (left) and the treatment effects on recidivism (right), as well as the corresponding 95% confidence intervals, against the relative distance between the place of residence of inmates and the two types of prison. Prisons are classified based on the following characteristics (treatments): whether the director is a woman, work facilities and external work, sanitation and health services, and whether prison cells are open during the day.

Table A1: Effect of Women-only Prison on Recidivism (entry).

	$\Delta D_i < 0$		
	(1)	(2)	(3)
	Panel A: First Stage - Women-only (Entry)		
Relative distance (Entry)	0.008***	0.007***	0.005***
	(0.001)	(0.001)	(0.001)
Mean women-only		0.477	
	Panel B: Reduced Form - Recidivates		
Relative distance (Entry)	-0.053***	-0.052***	-0.072***
	(0.018)	(0.020)	(0.020)
Mean recidivism (in %)		10	
	Panel C: IV- Recidivates		
Women-only (Entry)	-6.660***	-7.786**	-14.705***
	(2.408)	(3.061)	(4.527)
Controls		✓	✓
Region of birth FEs			✓
Region of resid. FEs			✓
Observations		6,707	
KP F-Stat	184.6	146.9	49.36
$\widehat{R}_M$ (in %)	13.47	14.06	17.66
$\widehat{R}_F$ (in %)	6.812	6.272	2.956

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the entry prison. We report the Kleibergen-Paap F statistic for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.

Table A2: Effect of Women-only Prison on Recidivism.

	Recidivism					
	1 year		2 years		3 years	
	Panel A - Reduced Form					
Relative distance (Exit)	-0.049***	-0.061***	-0.044**	-0.062***	-0.056***	-0.075***
	(0.016)	(0.018)	(0.017)	(0.019)	(0.019)	(0.020)
Mean recidivism (in %)	6.471		8.603		10	
	Panel B - IV					
Women-only (Exit)	-7.016***	-13.411***	-6.259**	-13.721***	-7.905***	-16.429***
	(2.473)	(4.097)	(2.591)	(4.265)	(2.902)	(4.826)
Observations	6707					
KP F-Stat	188.4	63.71	188.4	63.71	188.4	63.71
$\widehat{R}_M$ (in %)	9.818	12.87	11.59	15.15	13.78	17.84
$\widehat{R}_F$ (in %)	2.802	-0.541	5.330	1.429	5.871	1.414
Controls		✓		✓		✓
Region of birth FEs		✓		✓		✓
Region of resid. FEs		✓		✓		✓

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the entry prison. We report the Kleibergen-Paap F statistic for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.

Table A3: Effect of Women-only prison on recidivism (final sentence).

	(1)	(2)	(3)
Women-only (Exit)	-4.585	-2.005	-14.635***
	(3.711)	(4.177)	(4.734)
Observations	2,682	2,682	2,682
KP F-Stat	207.4	161.9	106.9
$\widehat{R}_M$ (in %)	11.35	10.13	16.14
$\widehat{R}_F$ (in %)	6.769	8.121	1.500
Controls		✓	✓
Region of birth FEs			✓
Region of resid. FEs			✓

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the entry prison. We report the Kleibergen-Paap F statistic for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.

Table A4: Effect of Women-only prison on Recidivism (Italians).

	(1)	(2)	(3)
Women-only (Exit)	-10.449***	-10.603***	-23.267***
	(2.937)	(3.581)	(4.918)
Observations	4,870	4,870	4,870
KP F-Stat	238.9	165.1	64.95
$\widehat{R}_M$ (in %)	15.83	15.91	22.31
$\widehat{R}_F$ (in %)	5.386	5.310	-0.955
Controls		✓	✓
Region of birth FEs			✓
Region of resid. FEs			✓

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the entry prison. We report the Kleibergen-Paap F statistic for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.

Table A5: Effect of Women-only Prison on Recidivism by Type of Crime.

	Property Crimes		Violent Crimes	
	(1)	(2)	(3)	(4)
	Panel A: First Stage- Women-only (Exit)			
Relative distance (Exit)	0.007***	0.004***	0.006***	0.004***
	(0.001)	(0.001)	(0.001)	(0.001)
Mean women-only	0.498		0.416	
	Panel B: Reduced Form- Recidivates			
Relative distance (Exit)	-0.061***	-0.097***	-0.071***	-0.068***
	(0.022)	(0.021)	(0.025)	(0.030)
Mean recidivism (in %)	10.14		10.46	
	Panel C: IV- Recidivates			
Women-only (Exit)	-8.647**	-22.253***	-11.570**	-15.685**
	(3.295)	(5.527)	(4.490)	(7.234)
Observations	5,002		2,190	
Controls		✓		✓
Region of birth FEs		✓		✓
Region of resid. FEs		✓		✓
KP F-Stat	170.6	56.67	141.5	49.51
$\widehat{R}_M$ (in %)	14.44	21.22	15.27	16.99
$\widehat{R}_F$ (in %)	5.795	-1.035	3.705	1.304

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the entry prison. We report the Kleibergen-Paap F statistic for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.

Table A6: Effect of Women-only prison on recidivism: restrict on control group characteristics.

	(1)	(2)	(3)	(4)	(5)	(6)
Women-only (Exit)	-9.925**	-26.529*	-14.355***	-16.559*	-16.634***	-11.637**
	(4.270)	(13.863)	(4.960)	(8.852)	(6.215)	(4.916)
Observations	5,047	4,860	5,003	4,904	5,887	4,020
KP F-Stat	93.73	10.65	66.48	14.77	39.65	54.08
Mean recidivism	10.26	9.547	10.11	9.706	10.07	9.677
Control group	> median female share	< median female share	> median size	< median size	male director	female director
Controls	✓	✓	✓	✓	✓	✓
Region of birth FEs	✓	✓	✓	✓	✓	✓
Region of resid. FEs	✓	✓	✓	✓	✓	✓

Notes: The dependent variable is whether the inmates recidivades within 3 years. The endogenous variable, *Women-only,* is a dummy equal to one for inmates in one of the four women-only prisons. The instrumental variable, *Prisons Relative distance* is the difference in the distance in km between the municipality of residence of the inmate and the nearest two types of prisons (Mixed-Gender vs. Women-only). The measures are calculated according to the entry prison. We report the Kleibergen-Paap F statistic for weak identification. Controls include age, age squared, nationality, education status, family status, types of crime, sentence status, incarceration length, closing year of incarceration, metropolitan area, true distance, overcrowding. Municipality of residence clustered standard errors in parentheses. * p< 0.10, ** p< 0.05, *** p< 0.01.