Supplementary Appendix: Representation Behind Closed Doors: The Effect of Electing Women Mayors on Domestic Violence

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December 11, 2024

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1 Appendix A: Security Plan

Municipality security plans were processed in two stages using Python. First, the *pypdf* package¹ was used to clean for stop words, tokenize, and export into a bag-of-words data frame. Some files could not be processed this way due to being scanned images, so the second stage utilized Python OCR. Python OCR is a technology that extracts text from images, such as scanned documents and photos, using Python.² This process was completed using the open-source OCR engine Tesseract. Security files that could not be processed in the first stage were treated as images and processed through OCR, resulting in another bag-of-words data frame.

Table A1 shows the keywords of interest for this analysis in Spanish (left-hand table) and English (right-hand table). The most common mentions are "mujeres" or "women", closely followed by the singular of this keyword and "intrafamilial". The "vif" keyword refers to intrafamilial violence, which is also common across the documents.

Keyword	Mentions	Keyword	Mentions
femicidios	5	femicides	5
femicidio	6	femicide	6
intrafamiliar	853	intrafamilial	853
mujer	858	woman	858
vif	668	vif	668
genero	7	gender	7
mujeres	895	women	895
violencia	125	violence	125

Table A1: Keyword Mentions in Security Plans

Table A2 lists the municipalities and years for security reports, totaling 115 security plans from Chilean municipalities spanning 2011 to 2024.

¹For documentation on usage, see https://github.com/py-pdf/pypdf.

²For documentation on this process, see https://builtin.com/data-science/python-ocr.

Alto Del Carmen2018, 2019, 2020Mejillones2023Ancud2022Melipilla2019Antofagasta2022Molina2022Arica2021, 2024Mulchen2021, 2022Buin2019Nueva Imperial2017, 2021MunicipalityCabo De Hornos2017Nunoa2016, 2017San Antonio200Cabrero2023Osorno2016San Fabian200Calbuco2018Ovalle2022San Javier200Castro2020Padre Hurtado2022San Javier200	ears)15, 2021)23)22
Built2019Nueva Imperial2017, 2021MuncipartyItCabo De Hornos2017Nunoa2016, 2017San Antonio20Cabrero2023Osorno2016San Fabian20Cabuco2018Ovalle2022San Fernando20Castro2020Padre Hurtado2022San Javier20)15, 2021)23)22
Cabo De Hornos2017Nunoa2016, 2017San Antonio20Cabrero2023Osorno2016San Fabian20Calbuco2018Ovalle2022San Fernando20Castro2020Padre Hurtado2022San Javier20)15, 2021)23)22
Cabrero2023Osorno2016San Fabian20Calbuco2018Ovalle2022San Fernando20Castro2020Padre Hurtado2022San Javier20)23)22)22
Castro2018Ovalle2022San Fernando20Castro2020Padre Hurtado2022San Javier20)22)22
Castro 2020 Padre Hurtado 2022 San Javier 20	177
	22
Cerro Navia 2022 Padre Las Casas 2023 San Joaquin 20)22
Chillan 2022 Paillaco 2022 San Miguel 20)21
Cholchol 2019 Paredones 2018 San Pedro De Atacama 20)21
Chonchi 2017, 2023 Parral 2011 San Vicente De Tagua Tagua 20)22
Colina 2021, 2022 Pelluhue 2022 Santiago 20)19, 2023
Collipulli 2022 Penaflor 2022 Sierra Gorda 20)19, 2023
Conchali 2017. 2022. Penalolen 2022. Tagua Tagua 20)22
Continue 2017 2022 Pichilemu 2017 2022 Talca 20)22
Covhainote 2022 Fitmfanen 2021 Talcahuano 20)15
Curento 2022 Providencia 2021-2023 Temuco 20)22
El Tabo 2022, 2025 Pucon 2019 Teno 20)22
Frairing 2017 Puente Alto 2022 2023 Teodoro Schmidt 20	122
Future And 2017 Future And 2022, 2023 Food Schmidt 20)22
Huaduu 2022 Fuctor Monte 2023 Focopria 20)17
Tucchurada 2022 Fuchovarias 2022 Fucaper 20)))))))))
iquique 2020 Fullandue 2010 Valutria 20 La Guiz 2022 Dutando 2003 Valutria 20)17
La Ciuz 2022 Futacitudo 2025 Valparaiso 20)17
La Florida 2022 Quilicura 2022 Vichuquen 20)22
La Pintana 2020 Quinta De Tilcoco 2022 Victoria 20)23
La Reina 2017 Quintero 2022 Villa Alegre 20)21, 2023
Lautaro 2023 Quisco 2021 Villa Alemana 20)22
Limache 2023 Rancagua 2016 Vina Del Mar 20)22
Lo Padro 2019 Recoleta 2020	
Los Alamos 2018 Renca 2017	
Los Angeles 2017, 2021 Rengo 2022	
Los Lagos 2023 Rinconada 2023	
Los Vilos 2017 Saavedra 2017	

Table A2: Security Plans for Municipality-Years

2 Appendix B: Alternative Explanations

An alternative explanation for substantive representation is that women are simply better mayors, leading to a general increase in reports of other offenses that affect people's quality of life. Women mayors might improve reporting avenues overall, resulting in more complaints about various types of crimes. To test this alternative explanation, we focus on two common offenses in Chile that are typically reported by citizens: disturbing the peace and public intoxication. If women mayors are enhancing communication between citizens and the government, we would expect to see an increase in reports of these offenses.



Figure A1: Average effect of having a woman mayor on reports of alternative offenses. A length of exposure of -1 refers to the period before the first exposure, 0 to the first exposure, and 1 to the second exposure. The overall treatment effect is reported with coefficients, standard errors in parentheses, and (*) denoting p-values lower than 0.1. N = 4,729 (municipality-year observations).

We do not find evidence that women mayors have either an overall or dynamic effect on offenses typically reported by citizens, such as public intoxication and disturbance of the peace. This finding strengthens the interpretation of substantive representation, suggesting that women mayors specifically facilitate the reporting of violence that disproportionately affects women.

Additionally, we investigate whether women mayors enhance the security performance of mu-

nicipalities, which could have two implications. First, improved security might explain changes in reporting. Second, and more importantly, it could influence our primary outcome: cases found by the police. In other words, if women mayors improve security measures, our benchmark may no longer serve as a true baseline or reference for reports but instead become an outcome or consequence of electing women mayors. To explore this possibility, we examine the impact of electing a woman mayor on two security-related outcomes: the number of security cameras and the number of security booths.



(a) Number of security cameras (logged)

(b) Number of security booths (logged)

Figure A2: Average effect of having a woman mayor on reports of alternative offenses. A length of exposure of -1 refers to the period before the first exposure, 0 to the first exposure, and 1 to the second exposure. The overall treatment effect is reported with coefficients, standard errors in parentheses, and (*) denoting p-values lower than 0.1. N = 4,729 (municipality-year observations).

We do not find evidence that electing women mayors affects the number of security cameras or security booths. We interpret this as a lack of support for the explanation that our main findings are driven by improved security performance. Additionally, this provides extra support for our benchmark outcome—cases found by the police—since women mayors do not appear to improve security performance.

3 Appendix C: Length of Effects

Figure A3 shows that electing a woman mayor increases reports of domestic violence around five years after her election, but this impact completely disappears ten years later. Why do these effects diminish over time? There are two plausible explanations for this: (i) municipalities controlled by men catch up with those controlled by women following a highly salient national discussion about domestic violence in 2010, which culminated in Congress passing a law on femicides (Vásquez Mejías, 2015); or (ii) policies that facilitate reporting fail to address the structural dynamics of violence against women and therefore may not have long-term consequences for reports of violence (Franceschet, 2010).

To determine which of these explanations is supported by the data, we disaggregate the dynamic difference-in-differences effects by the year of first exposure. Four groups are analyzed: never-treated, first exposed in 2009, first exposed in 2013, and first exposed in 2017 (with data spanning from 2005 to 2020). This allows us to compute the effects for three different groups: those first exposed in 2009, 2013, and 2017.



Figure A3: Average effect of having a woman mayor on violence against women by length of exposure and by group (first exposure in 2009, 2013, and 2017). A length of exposure of -1 refers to the period before the first exposure, 0 to the first exposure, and 1 to the second exposure. N = 4,353, 4,321, and 4,251 (municipality-year observations).

Figure A3 provides consistent results across the three subgroups. For units exposed in 2009, we observe an increase in reports but then they completely disappear ten years after first exposure. For units exposed in 2013, we also observe an increase but since our data do not extend beyond 2020, we cannot compute effects beyond seven years after exposure for this group. Finally, for units exposed in 2017, we do not observe an effect within the first three years, unlike the previous groups.

Given that patterns are similar for groups first exposed in 2009, 2013, and 2017, suggesting that the length of exposure to a woman mayor, rather than the year itself, explains the effects. This undermines the idea that men-controlled municipalities are catching up following the 2010

law categorizing femicides. Finally, we note that the effects are not immediate (as seen in the first, second, and third plots) and vanish after a few years (first plot). This pattern aligns with the second explanation, suggesting that the reforms lose power over time rather than being driven by contextual factors, such as men-controlled municipalities catching up with women-controlled municipalities. We encourage further research to have a better understanding of the long-term effects of electing women.

4 Appendix D: Regression Discontinuity Design

As an alternative empirical strategy, we employ a regression discontinuity design (RDD) in close electoral races, comparing municipalities where a woman narrowly wins over a man to those where a man narrowly wins over a woman. In this RDD setup, the unit of analysis is the municipality-year. Each municipality has a score based on the margin in the previous local election, and treatment is assigned if the score exceeds a particular cutoff. The treatment is defined as having a woman mayor (with a man as the runner-up), while the control is defined as having a man mayor (with a woman as the runner-up). The score represents the vote share difference between women and men candidates (the margin of victory). The cutoff is set at zero; thus, when the score is positive, the winning candidate is a woman, and when the score is negative, the winning candidate is a man.

It is important to note that RDD estimates the effect of electing women at the cutoff, meaning the results are local in nature and apply to closely contested races.

We use the same time frame as the difference-in-differences design used in the manuscript (from 2005 to 2020). To estimate the effects of electing a woman, we use local linear regressions, relying on an MSE-optimal bandwidth and a triangular kernel. The following estimation equation is used for this analysis:

$$Y_{it} = \alpha + \beta_1 T_{it} + \beta_2 M_{it} + \beta_3 T_{it} * M_{it} + \sigma_X + \varepsilon_{it}$$
(1)

Y is domestic violence in municipality *i* and year *t*. *T* depicts the treatment (units above the cutoff). *M* describes the margin of victory. The interaction between *T* and *M* allows the regression function to differ on both sides of the cutoff point; σ_X corresponds to a pre-treatment measure of local human development and a year indicator.

Table A3 shows the results using equation 1, which allows us to observe the effect of electing a woman mayor (at the cutoff) on domestic violence. Only coefficient β_1 from equation 1 is reported.

The conclusions align with those from the difference-in-differences analysis. On the one hand, we observe a significant increase in reports made by citizens. On the other hand, we do not find

	Domestic Violence	
	Reported by citizens	Found by the police
	(1)	(2)
Woman mayor	0.336*	-0.382
	(0.105)	(0.221)
Observations	1,396	1,396
Note:		*p<0.05

Table A3: RDD results

evidence of a significant impact in cases found by the police. Importantly, the RDD estimates only the impact of electing a woman mayor at the cutoff, only for competitive elections, and does not provide insights into how these effects evolve over time or whether they vary during her first, second, or third year in office.

5 Appendix F: Main Results in Table Formant

Event time	Estimate	Std. Error	[95% Simult. Conf. Band]
-11	-0.1179	0.2656	-0.8257, 0.5899
-10	0.1357	0.1776	-0.3375, 0.6088
-9	-0.3115	0.5119	-1.6754, 1.0523
-8	0.2051	. 0.3320	-0.6797, 1.0898
-7	0.1560	0.1629	-0.2781, 0.5901
-6	-0.0197	0.1685	-0.4686, 0.4292
-5	0.0302	0.1317	-0.3207, 0.3812
-4	-0.0328	0.1012	-0.3026, 0.2370
-3	-0.0572	0.1169	-0.3687, 0.2544
-2	0.0727	0.1276	-0.2675, 0.4128
-1	-0.0859	0.0774	-0.2921, 0.1203
0	0.1031	0.0709	-0.0858, 0.2921
1	0.0926	0.0659	-0.0831, 0.2683
2	0.1930	0.0992	-0.0715, 0.4574
3	0.2031	0.1019	-0.0683, 0.4746
4	0.3550	0.1137	0.0521, 0.6578
5	0.3932	0.1451	0.0066, 0.7798
6	0.4149	0.1476	0.0217, 0.8081
7	0.3034	0.1781	-0.1711, 0.7779
8	0.2882	0.2169	-0.2897, 0.8662
9	0.1859	0.2233	-0.4091, 0.7808
10	0.0233	0.2359	-0.6054, 0.6520
11	0.1132	0.2391	-0.5240, 0.7503

In this section, we present the main results from Figure 3a and 3b in table format.

Table A4: Figure 3a

Event time	Estimate	Std. Error	[95% Simult. Conf. Band]
-11	-0.2519	0.1061	-0.5431, 0.0394
-10	0.0891	0.1800	-0.4050, 0.5832
-9	-0.2014	0.2368	-0.8513, 0.4485
-8	-0.2477	0.2908	-1.0459, 0.5504
-7	-0.0289	0.0804	-0.2495, 0.1917
-6	0.0198	0.0820	-0.2052, 0.2448
-5	0.0373	0.0825	-0.1891, 0.2637
-4	-0.0530	0.1308	-0.4120, 0.3061
-3	-0.0744	0.0745	-0.2789, 0.1300
-2	0.0449	0.0867	-0.1931, 0.2829
-1	-0.0764	0.1034	-0.3603, 0.2075
0	-0.1263	0.0787	-0.3423, .0896
1	0.0359	0.0654	-0.1436, 0.2153
2	0.0946	0.1120	-0.2129, 0.4020
3	-0.0176	0.0914	-0.2683, 0.2332
4	0.1165	0.1043	-0.1697, 0.4027
5	0.1327	0.0874	-0.1071, 0.3726
6	0.0943	0.1301	-0.2627, 0.4514
7	0.1150	0.1066	-0.1774, 0.4075
8	0.2009	0.2011	-0.3510, 0.7528
9	0.0844	0.2477	-0.5954, 0.7642
10	-0.0018	0.1496	-0.4125, 0.4088
11	0.2020	0.2565	-0.5018, 0.9059

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Table A5: Figure 3b

References

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