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JEL Codes:

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Does Holding Elections during a Covid-19 Pandemic Put the Lives of Politicians at Risk? *

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Abstract

We estimate the impact of French town hall elections held in mid-March 2020 on the mortality of 170,000 male candidates aged above 60. Their excess mortality during March and April was similar to the general population. We compare candidates in cities with two candidate lists to those in cities with only one list, as elections are more intense in contacts in the former group. We also use a regression discontinuity design and investigate mortality in 2020 depending on how candidates fared in the 2014 election. We cannot detect any causal effect of active participation in the 2020 elections on mortality.

Keywords: Covid-19 pandemic; French elections; individual-level data; regression discontinuity design.

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1 Introduction

In early 2020, the Covid-19 pandemic has forced most of the planet to stop any non-essential activity involving large groups of people until disease transmission is eventually controlled. Whether one should consider elections an essential activity in this context has become a matter of intense debates leading to dramatically different decisions across the globe. Some governments have postponed elections (local elections in Austria, Italy & UK, Ohio primaries), some have switched to mail-in voting (Poland, Alaska primaries), and others have decided to hold the elections as planned with further sanitary precautions (Israel, South Korea, Wisconsin primaries). The empirical evidence to guide public policy is very scarce, owing to how sudden the Covid-19 outbreak has been and to the small number of elections held since it began.

This paper seeks to bridge this gap: we measure the impact of holding the first round of French town hall elections on the mortality of the individuals it exposed the most to Covid-19, i.e. politicians. These elections provide an unusual setting as they were held at the apex of the epidemic in France – on Sunday the 15th of March 2020 –, one day after the nationwide closure of non-essential retail businesses was ordered and two days before a shelter-at-home order.

Our analysis is based on individual death records from the French institute of statistics (Insee) matched with a public registry of candidates in town hall elections. The resulting dataset enables us to compare the mortality rates of male candidates in the 2014 and 2020 municipal elections relative to the general population as well as depending on the degree of participation into the 2020 election.

The correlations we can estimate between political activity and mortality using this new dataset do not necessarily reflect a causal relationship. In particular, citizens might choose to become candidates only after careful consideration of their health status and election participants may adapt their contact behavior to their current knowledge about the epidemic. To deal with these threats to identification, we implement two distinct strategies to estimate the causal effect of active participation in the election. First, turnout and campaign intensity are much lower when only one candidate list is running for election, so we compare the mortality rates of candidates when there are either one or two lists competing for entrance in the city council. Second, we examine whether, among the full population of male candidates in the previous election held in 2014, being either an incumbent councilor or a candidate in 2020 caused an increase in the probability of death. We implement a regression discontinuity design in the spirit of Lee (2008), since electoral success in the prior election has a large impact on active participation in the 2020 election. This strategy identifies the impact of running in the 2020 election at the individual level.

Three main sets of findings emerge from our empirical exercise. First, we compute the mortality rates in March and April 2020 among candidates in the March 15th election. This is a group of around 900,000 individuals, representing 1.8% of the adult French population. They were among the most likely to contract the disease as they actively participated in the campaign and closely monitored the voting process on election day. Many of them also belong to demographic groups with a high risk of dying from Covid-19: 172,000 candidates were

men aged between 60 and 79 years old, representing 2.8% of their category in the general population. We compare their mortality rates with overall mortality rates among individuals of same gender and same age. We find that, across March and April 2020, candidates died significantly less than members of the general population from the same age bracket. The weekly mortality rate among men aged between 60 and 79 years old from March to April was 4 for 10,000 in the general population against 1 for 10,000 among town hall candidates. This could reflect selection into candidacy based on good health. However, we do not find that this mortality "advantage" was smaller in areas where Covid-19 had the greatest incidence, i.e. the Ile-de-France and Grand-Est regions. The difference was also of similar size around the onset of the epidemic, suggesting the excess mortality of candidates due to Covid-19 was in fact similar to the general population.

Second, we compare the mortality of candidates in cities where two candidate lists compete with the mortality of candidates in cities where there is only one candidate list. In the data, the number of competing candidate lists is a very strong determinant of turnout. Yet, we do not find any difference in mortality among candidates between the two types of cities, even though they otherwise do not differ much in observable characteristics. We confirm that these findings from the pool of candidates also hold among city residents in general.

Third, in order to go beyond estimating citywide effects of the election on mortality, we investigate the mortality patterns of the people who were candidates in the previous municipal elections, held in 2014. We find that, among them, those who decided to run again in 2020 were not more likely to die from March to April 2020. Using a regression discontinuity design, we observe that candidates in 2014 are far more likely to participate again in 2020 if their candidate list had won the election in 2014. However, we do not detect any significant effect of winning the election in 2014 on mortality in March-April 2020. This further confirms, using an entirely different identification strategy, that intense participation in the 2020 elections did not generate any specific mortality risk.

This is to our knowledge the first paper to document the mortality patterns of politicians in the middle of a Covid-19 pandemic. It is of interest in its own right as politicians are key decision-makers whose good health is of paramount value in times of crisis. By the nature of their occupation, politicians may also be more likely to contract the disease. Examples of how this has hampered policy-making abound: at certain times in the epidemic, the UK government was leaderless while the French National Assembly had to stop its proceedings once an internal cluster developed.¹ The main research hurdle is to obtain a large enough sample of politicians to analyze. In this respect, our setting, which includes more than 170,000 politicians with significant exposure to mortality from Covid-19, provides a unique opportunity.

Our research also innovates by investigating the effects of elections on the diffusion of Covid-19 using individual data. Several studies have by now investigated this issue at higher levels of aggregation and for the general population. Some of this research (Zeitoun *et al.*, 2020; Cassan and Sangnier, 2020; Cotti *et al.*, 2020) has focused on Covid-19 outcomes measured at

¹See "How the National Assembly became a new cluster", *LeMonde*, March 21, 2020. Details on newspaper articles cited in the paper can be found in Appendix F.2.

the level of each county and related them to turnout at the same level of aggregation, with conflicting results. Those analyses seek to explain and predict the full development of the epidemic. This comes at the cost of making strong assumptions on the behavior of the disease and its diffusion patterns. This is why we focus instead on a narrower goal of identifying individual transmissions of the disease that were directly caused by the election.

Bertoli *et al.* (2020) estimate the citywide impact of electoral competition on mortality among the general population. They find that mortality rates were higher in cities with more electoral competition. In the Online Appendix (Section E), we run a placebo experiment and show that, under the specification and data used by Bertoli *et al.* (2020), electoral competition in 2020 had a positive effect on excess mortality generated by a heatwave taking place 17 years earlier. Using death records with more precise geographical information, we fail to detect a significant impact of electoral competition on mortality in the general population.

The rest of the paper is organized as follows. Section 2 documents the context of the French election as well as the data we use. Section 3 discusses the hindrances to a causal identification and the solutions we implement. Section 4 provides and discusses the results of our analysis of mortality rates among candidates. Section 5 concludes.

2 Context and Data

The 2020 town hall election. The town hall elections in France had long been planned to be held on March 15th, 2020, with potential run-off elections to be held one week later. City governments traditionally hold substantial powers and represent most of local government employment and spending. Moreover, town halls in France are very numerous (less than 1 for every 2,000 French citizens), organized in large councils headed by a powerful mayor. As a result, town hall elections are grassroots events involving roughly 2% of the adult population as candidates and typically generate a lot of turnout: in 2014, 64% of registered voters went to the voting booth for those elections.

As the Covid-19 epidemic started developing, there were only very few calls to cancel the election up until significant lockdowns were announced by the central government in the evenings of March 12th (school closures) and March 14th (restaurant and non-essential shop closures). The government therefore went on to organize the first round with stricter sanitary guidelines issued to town halls organizing the election. Yet, turnout in the first round was much lower than usual, at 45%, and criticisms over the decision to keep the election built up so much that the government decided on the day after the first round to postpone the second round of the election until further notice.

It is now widely considered in French public opinion that holding the first round was one of the big mistakes made by the central government in managing the sanitary crisis.² The French media have collected ample anecdotal evidence of individuals being contaminated by

²According to a poll by YouGov on the 17th and 18th of March, 2020 for the Huffington Post, 81% of respondents thought the first round of the elections should have been cancelled, and 74% believed the results to be biased by the Covid-19 pandemic. Source: link.

Covid-19 because of this election.³ Most of the gathered evidence is on cases of candidates, city councilors and mayors, since they were deeply involved in the campaign, they monitored the voting process on election day and they celebrated the results on the evening of the election. Furthermore, importantly, many of them belong to high-risk categories. Incidentally, the first reported fatality from Covid-19 contracted on French soil was on February 26th, one day before the call for candidates closed, and it was a city councilor aged 61. However, it remains unknown to this day whether this pile of anecdotal evidence represents a significant shock to the mortality rate of this category of the French population. The aim of this paper is precisely to shed statistical light on this issue.

2.1 Data

Data on the deceased. We obtain information on the identity of all the deceased in France from the Fichier Nominatif des Décès (henceforth, FND) held by the National Institute of Statistics (Insee). The dataset includes the names, surnames of birth, date and location of birth, date and location of death of all deceased French residents. Crucially, the dataset is updated monthly by Insee, which centralizes death certificates from all around the country. It usually takes less than a month for deaths to be reported in the dataset, so that the last batch of the data published on June 16th 2020 has close to full coverage of all deaths taking place until mid-May, well after the peak of mortality linked with the first wave of Covid-19 in France. Another useful feature of the dataset is that it is available on the French government open data website, the reason being that information on the deceased is not deemed to be personal data under French laws.⁴

Data on candidates. Based on name and birth information, we match the FND dataset with the registry of candidates in town hall elections held by the Ministère de l'Intérieur. The registry includes current name and surnames, date and location of birth of all candidates in the 2014 and 2020 town hall elections, on top of the number of votes obtained by each candidate. We match the candidate registry with the FND dataset on name, surname, date and location of birth. Since women often change surnames over their lifetime, we restrict our analysis to male candidates. More details on the matching procedure are provided in the Appendix F.1.

Additional data. We gather data on the number of confirmed cases in each county as of March 14th from the website https://github.com/opencovid19-fr/data (for which the primary sources are the regional health agencies). Demographic data (including population density, age structure, class structure and mortality rates) for each county and city are derived from Census data available on the Insee website.

³Stories have been featured in several national or regional newspapers, online media or on the website of 24-hour news channels. See Appendix F.2 for more details.

⁴Under GDPR regulations, information on the deceased is not considered as personal data, which is why the FND data is publicly available with full identities disclosed. Because we also use some demographic information on living candidates, we have registered our treatment of personal data for this paper and we provide information on personal data access and correction rights in the Appendix (see Section F.3).

2.2 Descriptive statistics on nationwide excess mortality

Using the assembled data, we now provide a description of the excess mortality observed at the national level for male candidates in the municipal elections. The comparison of the excess mortality of candidates in the municipal elections with the excess mortality in the overall population already gives indications on the impact of participation in the municipal elections one might expect to find.

Figure 1 shows mortality over time from the beginning of March 2020 to mid-May for the whole population of mean aged between 60 and 79 (in blue) and for male candidates in the same age category (in red), depending on whether they were in hotspots (Île de France and Grand Est regions), represented with solid lines and circles, or in the rest of France, displayed with dotted lines and diamonds. The figure shows that there was indeed an excess mortality of election candidates concentrated in the hotspots during the peak of the Covid-19 epidemic. However, this excess mortality does not appear to be starkly higher than the one observed for the whole population.

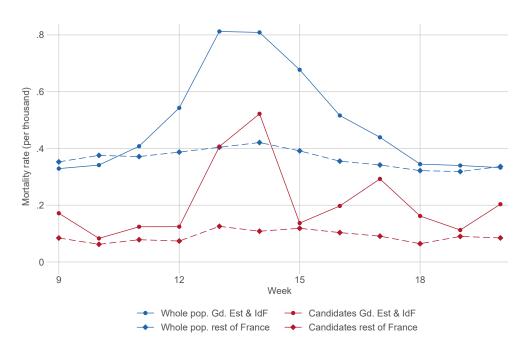


Figure 1: Aggregate mortality rates for candidates and overall population

NOTES: This graph presents mortality rates for men aged between 60 and 79 for the whole French population and for the pool of candidates in this demographic category, in Grand Est and Île-de-France (where most Covid-19 cases gave been located during the first wave of the epidemic) and the rest of France. Mortality rates are expressed per thousand inhabitants. Results are weighted by the population in the area.

While suggestive, this absence of correlation between candidacy in 2020 and excess mortality does not rule out a causal link. This is why we implement a more structured empirical design in the rest of the paper.

3 Empirical Design

In this section, we present the challenges associated with measuring the impact of active participation in the election on excess mortality, and then detail the solutions we implement in order to address them.

3.1 Empirical challenges

Challenges in disentangling different effects caused by the elections. First of all, the election may affect the diffusion of the disease through at least three different channels: the campaign in the weeks prior to election, the meetings between voters and organizers at the voting location, and the celebrations between campaign members on the evening of the election. There is unfortunately no direct measure of how intense each of those mechanisms were in each locality. This makes it in particular very difficult to use the timing of confirmed cases, hospitalizations and deaths in order to identify the impact of the election as opposed to the effects of other important shocks such as lockdowns. Our approach will not make any specific assumption on the timing of fatalities caused by the election.

Challenges in the measurement of epidemic diffusion. The measures researchers can access are usually aggregated at the level of a *département* (the denomination for counties in French) or a city, while the underlying behavior is individual. This leads to a statistical issue known as the ecological inference problem⁵: the estimates of the impact of election intensity on the diffusion of Covid-19 may vary widely and even change sign depending on the level of aggregation at which campaign intensity is measured. This motivates our focus on individual measures of both participation in the election and transmission of the disease.

On top of this aggregation problem, the measure of the diffusion of Covid-19 into the territory is itself problematic. There is to this day no systematic and randomized testing of the French population, which would be ideal for our purposes. Instead, we have on one hand incidence measures and on the other hand measures of tail events such as deaths: the former are naturally higher where testing is more frequent, while the latter are less likely biased but also, thankfully, much less frequent outcomes. In this paper, we use mortality as an outcome but we are aware that this is at the expense of statistical power. We therefore pay careful attention to maximizing the number of observations we have at our disposal.

Reverse causality: evidence of avoidance behavior at city-level. People have been increasingly aware of the diffusion of the virus in the days and weeks prior to the election and might have adjusted their behavior accordingly, through lower participation in the campaign and lower turnout. Crucially, this type of behavioral adjustments might have been more common in areas where the epidemic was developing more intensely. As a result, places where ob-

⁵See King (2013) for an overview of the issue.

served campaign intensity and turnout have been low could display high mortality with the causality running from the latter to the former.⁶

In the absence of reliable data on local diffusion at the city-level, we can nevertheless document strategic avoidance behavior by correlating turnout during the 2020 election with city-level factors that are predictive of the potential severity of exposure to the coronavirus, such as age. In Appendix C, we show that a high share of older residents (60+) at the city level is associated with a larger increase in abstention between the 2020 and 2014 elections, in dense cities – where the virus was actively circulating as of March 2020.

To avoid such strategic avoidance biasing our results, we rely on the same intuition as in Bertoli *et al.* (2020) and use proxies for election intensity that were largely predetermined at the time Covid-19 started its diffusion on the French territory. At the city-level, we use various measures of electoral competition, while at the individual level we examine the decision to be a candidate in the election.

Omitted variable bias Lastly, the epidemic is likely to spread faster in areas with specific characteristics. For instance, Gascard *et al.* (2020) show that excess mortality rates are higher in denser areas across the French territory. As knowledge on Covid-19 progresses, the scientific literature is likely to identify other characteristics of territories that enhance its diffusion. The issue for our purpose is that some of those characteristics are likely to be correlated with turnout and campaign intensity, so any attempt to derive causation from observed correlations between turnout and epidemic outcomes will likely be affected by omitted variable biases. In order to tease out causation, we rely on the existence of discontinuities in drivers of campaign intensity and of participation to the election as a candidate.

3.2 Identification strategies

Variation in the number of lists in the election Our first strategy aims at estimating the effect of electoral competition intensity on the citywide mortality rate among candidates. In order to circumvent the issues exposed above, we use the stark discontinuity of electoral competition between one and two lists as an exogenous source of variation in exposure to the disease for candidates running in the municipal election. The identifying assumption underlying this strategy is that the number of lists in the municipal election in 2020 can only affect mortality through its influence on the intensity of electoral competition.

Out of 9836 municipalities with electoral lists,⁷ 3812 had only one list in the first round of the 2020 election while 3852 had two lists. As Figure 2a shows, there is a sharp drop in the average abstention rate between municipalities with candidates in the municipal elections from only one list, who will therefore be elected no matter how high the electoral participation, and the ones with two lists, where the stakes of the election are much higher. In contrast,

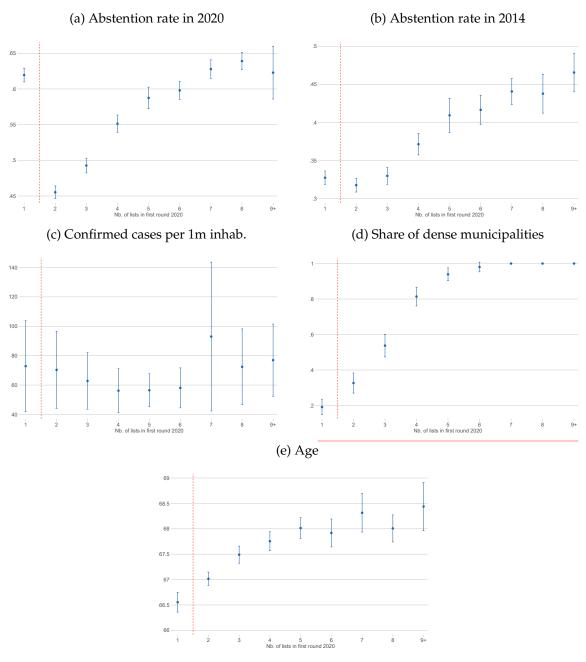
⁶Zeitoun *et al.* (2020) and Cassan and Sangnier (2020) look for such strategic avoidance behavior using county-level data and do not find evidence for it. However, at this level of aggregation, the analysis may lack statistical power, which is why we prefer to analyze turnout at the level of each city.

⁷Municipalities with less than 1,000 inhabitants hold the election with the *scrutin plurinominal* rule, where people can select individually the candidates they vote for.

as Figures 2, B2 and B5 show, the differences in relevant characteristics between municipalities having one versus two lists are very small. In particular, the number of confirmed cases per 1 million inhabitants (at county-level) is extremely similar, and so is the share of dense municipalities. Interestingly, the abstention rates in the previous municipal election are also very similar, indicating that there is low persistence in the fact of having only one list across municipal elections.

A drawback of this strategy is that municipalities with either 1 or 2 lists tend to be located in low density areas: as Figure 2d shows, only around 10% of these municipalities are denser than 2000 inhabitants per square kilometer. This limits the external validity of our approach since the Covid-19 epidemic developed less strongly in low density areas. Therefore, obtaining null or weak results of the electoral participation on covid-related mortality could reflect heterogeneity in the effects of elections according to population density. We could in principle extend the comparison to cities with more than 2 lists. Panels b to e in Figure 2 however illustrate why the full support of the number of lists cannot be used to obtain causal estimates of elections on mortality: higher competition is associated with systematic differences in characteristics relevant to both virus diffusion / lethality and electoral participation, such as confirmed cases at the time of the election, population density, and the average age of candidates 60 or more. To circumvent this issue, we implement an alternative strategy which allows inference along the whole population density support.

Figure 2: Number of lists in the first round of the election and observable municipal characteristics.



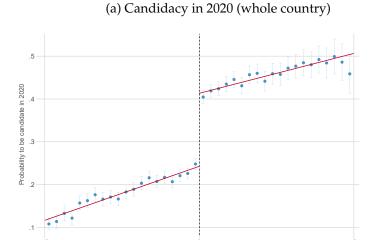
NOTES: This figure presents the number of lists in the first round of the 2020 French municipal election confronted to (a) the absention rate in 2020, (b) the abstention rate in 2014, (c) the number of confirmed cases on March 14, 2020 per million of inhabitants, (d) the share of municipalities with population density higher than 2000 inhabitants per square kilometer (e) age of candidates. Dots represent averages, brackets 95% CI using standard errors clustered at the *départment*-level. Cities with 9 or more lists are binned together (the full distribution of the number of lists is presented in Table A1)

Discontinuity in the win margin in the 2014 municipal elections In an alternative strategy, we exploit the fact that candidates in the 2014 winning list are disproportionately more likely to be candidates again in 2020 than candidates in 2014 losing lists (as in Lee, 2008, for the U.S.). The winners of 2014 are also much more likely to be incumbent city councilors by late 2019 and

to participate in the organization and monitoring of the 2020 elections. The sample comprises all first-past-the-post municipal elections held in 2014: this includes all run-off elections, and elections in which only two lists faced each other in the first round in a city with more than 1,000 inhabitants. We adapt Lee (2008) to a setting with more than two competing lists, as in Eggers and Spirling (2017). Doing so requires defining a win margin for each candidate, which is the vote share margin between the candidate's list and its best competitor in 2014.⁸

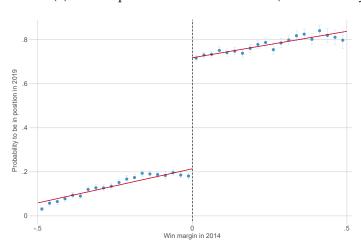
As Figure 3 shows, there is a very clear discontinuity in the relationship between the probability to be a candidate again in 2020 and the victory margin in 2014 just around the winning threshold, both for the entire country (Figure 3a) and for places most strongly affected by Covid-19 (Figure 3b).

Figure 3: First stage – Probability of being a candidate in 2020 / elected official in 2019 according to the win margin in 2014



(b) Elected position at the end of 2019 (whole country)

Win margin in 2014



NOTES: This figure presents averages and confidence intervals of the probability of being (a) a candidate in the 2020 election or (b) an incumbent city councilor in late 2019 for males older than 60 y.o., depending on the win (or loss) margin of the list the candidate belonged to in 2014.

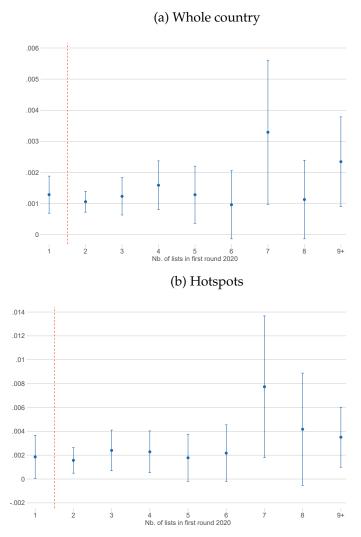
⁸Thus, when the candidate wins in 2014, the running variable is the difference in vote share between his/her list and the second-place list; otherwise, it is the difference in vote share between his/her list and the winning list.

4 Results

4.1 Electoral competition

Here, we compare cities which had only one list in the first round of the 2020 election, implying low competition and low turnout, to the ones which had two lists and thus exhibited much higher political participation – as explained in section 3.

Figure 4: Mortality during the Covid-19 pandemic by number of lists in the first round of the 2020 municipal election.



NOTES: These graphs present the probability of death of candidates in the election by number of lists in the first round of the 2020 municipal election, (a) for the whole country, and (b) in Île de France and Grand-Est. the sample includes all male candidates aged 60 and above. Cities with 9 or more lists are binned together (the full distribution of the number of lists is presented in Table A1).

Figure 4 shows differences in mortality during the peak of the epidemic, between Feb. 26 and April 28, 2020, depending on the number of lists present in the first round of the election, for the whole country (panel a) and focusing on hotspots (panel b). This figure shows no excess

mortality on average in cities with two lists relative to the ones with one list only: if anything, mortality is even slightly higher in cities with one list in hotspot areas.

We find a broadly similar pattern among the general population of males 60 and older. The results are presented in the Appendix in Figure B5. Computing mortality rates and excess mortality rates with respect to 2018 and 2019, we find no differences for the whole country (panels B5a and B5d) or focusing on the most affected counties (panels B5a and B5d).

As noted in section 3, this strategy mechanically focuses on low density and therefore low Covid-prevalence places, and thus carries the risk of ignoring the areas where excess mortality due to Covid-19 was effectively high. We address this problem by studying the mortality of all candidates in 2014 to gain as much statistical power as we can, and then by exploiting a discontinuity in the win margin in 2014 to eliminate potential biases in this approach.

4.2 The causal effect of candidacy on death: RDD results

In this section, we present the RDD results regarding the causal effect of being on a winning list in 2014 on the probability of dying between February 26 and April 28, 2020. This specification is a reduced form and we consider that any causal effect of being on a winning list in 2014 on 2020 mortality will run through the probability of either running as candidate or being an incumbent councilor in 2020 – which as we saw in Figure 3 reacts strongly to being on a winning list in 2014. Figure 5 presents the results. Whether for the whole country (subfigure 5a) or just when focusing on hotspots (subfigure 5b), we find no evidence of an effect of being on a 2014 winning list on mortality in early 2020.

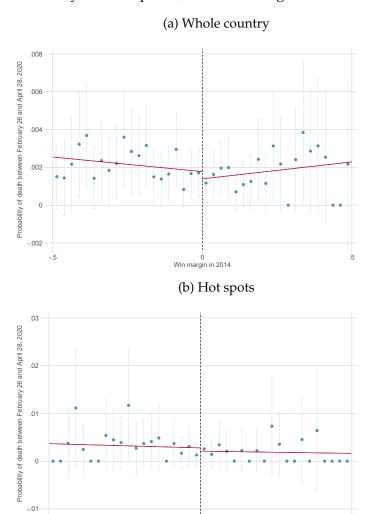
These graphical results are confirmed by a formal RDD estimation of the ITT effect. The results of the main specification are presented in Table A2 in the Internet Appendix. The RDD estimator suggests that, in response to being part of a winning list in 2014, there is a large increase in the probability of being a candidate in 2020 (+15.7 pp) or of being an elected councilor in 2019 (+53.1 pp) but there is no significant effect on mortality in 2020.

In the Internet Appendix, we run various robustness checks, including falsification tests using candidates' personal characteristics that were predetermined prior to the 2014 election (age) and prior to the 2020 pandemic (probability of dying prior to 2020) – see Table A2. We also implement the density test suggested by Cattaneo *et al.* (2018) – see Figure B3. Finally, we also assess the sensitivity of our results to the specification of the local polynomial used in the estimation (Table A4 and A5). These tests all support the validity of our identification strategy and the robustness of our results.⁹

One limitation of our approach is that it may lack statistical power to detect substantial effects of participation in the election on mortality. In the Internet Appendix, we provide RDD estimates and standard errors corresponding to Figures 5a and 5b in Tables A2 and A3 respectively. The standard error of the reduced form effect of winning the election in 2014 on mortality in March and April 2020 is equal to the effect of being 3.5 years older, where the

⁹In particular, we do not find that being among the winners of an election reduces mortality in the following six years. This goes against Borgschulte and Vogler (2019), who identify a mortality benefit of winning elections, but our focus (short-term mortality vs. longevity) and context (municipal vs. gubernatorial elections) are very different

Figure 5: Intention-to-treat – Probability of death of males older than 60 y.o. between February 26 and April 28, 2020 according to the win margin in 2014



NOTES: This figure presents averages and confidence intervals of the probability of being a candidate in the first round of the 2020 municipal election using the win (or loss) margin of the list to which the candidate belonged in 2014 as the running variable, (a) for the whole country, and (b) in Île de France and Grand-Est. The sample is composed of all male candidates in the 2014 municipal election born before 1960.

Win margin in 2014

effect of age on mortality during the pandemic is estimated in a separate OLS regression (see Table A6 in Appendix section D). To obtain more precise estimates, we fit a simple OLS regression of mortality on the decision to be a candidate in 2020 and control for age and the rank of entrance into the city council in 2014. Once rank is controlled for, this delivers a negligible estimate of the effect of being a candidate in 2020 on mortality that is strikingly similar to our RDD estimate: -0.000402 for OLS and -0.000410 for the RDD.¹⁰ The associated standard error is equal to the effect of being one year older. Overall, this makes us confident that substantial effects of participation in the 2020 election on mortality can be ruled out.

¹⁰See Column (3) in Table A6 in Appendix Section D for the OLS coefficient and Column (3) of Table A4 for the RDD.

5 Conclusion

In times of social distancing and bans of non-essential activities, it is crucial to know more about the risks implied by elections. The French context allows an in-depth analysis of this question: in March 2020, after heated debates, the French government decided to maintain the municipal elections on March 15th, two days before the countrywide lockdown.

The key contribution of this paper is to investigate the health impact of an election held in time of pandemic using a very large sample of individual observations and exploiting quasi-random exposure to campaigning activities. We fail to detect any positive effect of active participation in the 2020 elections on mortality among politicians aged 60 and above.

Our results do not, of course, fully exhaust the debate on whether elections should be held in those times and in which format. The elections we have analyzed may have spared the candidates in part because voters strategically did not turn out. This lack of voter participation is concerning as it reduces the legitimacy of voting results. For this reason, a better understanding of turnout decisions in pandemic times and their long-term consequences should be on the research agenda for the years to come.

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A Additional tables (for online publication only)

Table A1: Distribution of number of lists per city in 2020.

| | Unweighted fraction | Weighted - | |
|-------------------|---------------------|------------|--|
| | (in %) | (in %) | |
| Num. lists = 1 | 38.76 | 14.68 | |
| Num. lists = 2 | 39.16 | 32.93 | |
| Num. lists = 3 | 12.75 | 19.01 | |
| Num. lists = 4 | 4.53 | 10.78 | |
| Num. lists = 5 | 2.00 | 6.65 | |
| Num. lists = 6 | 1.04 | 4.53 | |
| Num. lists = 7 | 0.62 | 3.56 | |
| Num. lists = 8 | 0.46 | 2.89 | |
| Num. lists = 9 | 0.26 | 1.77 | |
| Num. lists = 10 | 0.22 | 1.64 | |
| Num. lists = 11 | 0.15 | 1.06 | |
| Num. lists = 12 | 0.01 | 0.16 | |
| Num. lists = 13 | 0.02 | 0.16 | |
| Num. lists = 14 | 0.01 | 0.18 | |
| Total | 100.00 | 100.00 | |
| N | 9836 | 9836 | |

NOTES: This table presents the marginal distribution of the number of lists by city. Column (1) presents the unweighted distribution while column (2) present the distribution weighted by the number of candidates.

Table A2: Main RDD specification: full sample

| | Outcomes | | | Falsification | | |
|--------------|------------------|------------------|------------------------------------|---------------|---------------------|--|
| | (1) | | | (4) Age | (5) | |
| | Candidate in '20 | Councilor in '19 | Councilor in '19 Death MarApr. '20 | | Death bef. Feb. '20 | |
| RD_Estimate | 0.157*** | 0.531*** | -0.000410 | 0.00757 | 0.00364 | |
| | (0.0123) | (0.0107) | (0.000733) | (0.165) | (0.00430) | |
| Observations | 111552 | 111552 | 111552 | 111552 | 111552 | |

NOTES: This table presents RDD estimates using the win (or loss) margin of the list to which the candidate belonged in 2014 as the running variable. The sample is composed of all male candidates in the 2014 municipal election born before 1960. Bias-corrected confidence intervals are constructed following the method proposed by Calonico $et\ al.\ (2014b)$ as implemented in Calonico $et\ al.\ (2014a)$. The local polynomial used to construct the point estimate is of order 1 and is fitted weighing observations with a triangular kernel. The local polynomial used to construct the bias correction is of order 2. Robust standard errors clustered at the county (département) level are reported in parentheses. *p < 0.1, **p < 0.05, ***p < 0.01.

Table A3: Main RDD specification: hot spots

| | Outcomes | | | Falsification | | |
|--------------|----------------------|----------------------|-----------------------|-------------------|---------------------|--|
| | (1) (2) (3) | | (4) | (5) | | |
| | Candidate in '20 | Councilor in '19 | Death MarApr. '20 | Age | Death bef. Feb. '20 | |
| RD_Estimate | 0.169*** (0.0211) | 0.520*** (0.0210) | 0.000755 (0.00174) | -0.175 (0.296) | 0.0117 (0.00855) | |
| Observations | 19589 | 19589 | 19589 | 19589 | 19589 | |

NOTES: This table presents RDD estimates using the win (or loss) margin of the list to which the candidate belonged in 2014 as the running variable. Bias-corrected confidence intervals are constructed following the method proposed by Calonico *et al.* (2014b) as implemented in Calonico *et al.* (2014a). The local polynomial used to construct the point estimate is of order 1 and is fitted weighing observations with a triangular kernel. The local polynomial used to construct the bias correction is of order 2. Robust standard errors clustered at the city level are reported in parentheses. *p < 0.1, *** p < 0.05, **** p < 0.01.

Table A4: RDD specification checks: local polynomial of order 2 (all of France)

| | Outcomes | | | Falsification | | |
|--------------|------------------|--|------------|---------------|---------------------|--|
| | (1) | (2) (3) | | (4) | (5) | |
| | Candidate in '20 | a '20 Councilor in '19 Death MarApr. '20 | | Age | Death bef. Feb. '20 | |
| RD_Estimate | 0.142*** | 0.525*** | -0.000788 | 0.0831 | 0.00255 | |
| | (0.0168) | (0.0139) | (0.000972) | (0.193) | (0.00491) | |
| Observations | 111552 | 111552 | 111552 | 111552 | 111552 | |

NOTES: This table presents RDD estimates using the win (or loss) margin of the list to which the candidate belonged in 2014 as the running variable. Bias-corrected confidence intervals are constructed following the method proposed by Calonico *et al.* (2014b) as implemented in Calonico *et al.* (2014a). The local polynomial used to construct the point estimate is of order 2 and is fitted weighing observations with a triangular kernel. The local polynomial used to construct the bias correction is of order 2. Robust standard errors clustered at the county (département) level are reported in parentheses. *p < 0.1, **p < 0.05, ***p < 0.01

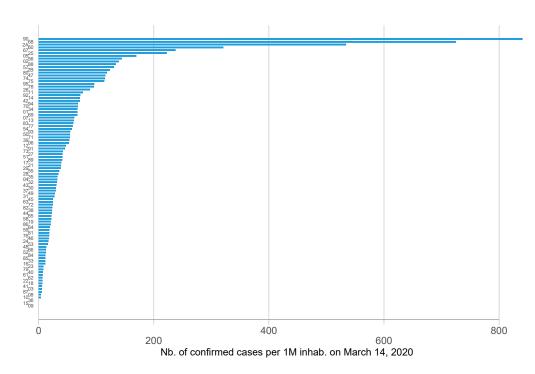
Table A5: RDD specification checks: local polynomial of order 2 (hotspots)

| | Outcomes | | | Falsification | | |
|--------------|------------------|------------------|-------------------|---------------|---------------------|--|
| | (1) | (2) | (3) | (4) | (5) | |
| | Candidate in '20 | Councilor in '19 | Death MarApr. '20 | Age | Death bef. Feb. '20 | |
| RD_Estimate | 0.161*** | 0.516*** | 0.000329 | -0.287 | 0.0112 | |
| | (0.0280) | (0.0271) | (0.00227) | (0.380) | (0.0109) | |
| Observations | 19589 | 19589 | 19589 | 19589 | 19589 | |

NOTES: This table presents RDD estimates using the win (or loss) margin of the list to which the candidate belonged in 2014 as the running variable. Bias-corrected confidence intervals are constructed following the method proposed by Calonico *et al.* (2014b) as implemented in Calonico *et al.* (2014a). The local polynomial used to construct the point estimate is of order 2 and is fitted weighing observations with a triangular kernel. The local polynomial used to construct the bias correction is of order 2. Robust standard errors clustered at the city level are reported in parentheses. *p < 0.1, *** p < 0.05, **** p < 0.01.

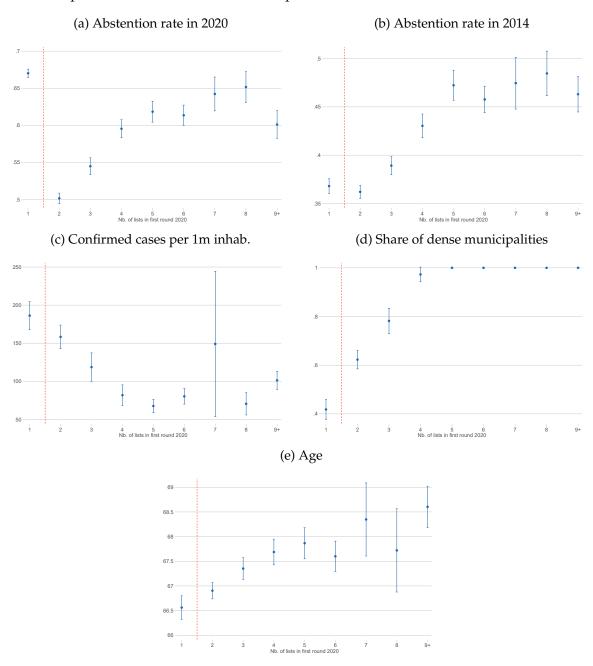
B Additional figures

Figure B1: Number of confirmed Covid-19 cases on March 14, 2020 by county (département)



NOTES: This figure presents the number of confirmed cases on March 14, 2020 by *département*. The y-axis indicates the *département* identifier in the French nomenclature.

Figure B2: Number of lists in the first round of the election and observable municipal characteristics – focus on hotspots.



NOTES: This figure presents the number of lists in the first round of the 2020 French municipal election in hotspots (Île de France and Grand-Est) confronted to (a) the absention rate in 2020, (b) the abstention rate in 2014, (c) the number of confirmed cases on March 14, 2020 per million of inhabitants, (d) the share of municipalities with population density higher than 2000 inhabitants per square kilometer (e) age of candidates. Dots represent averages, brackets 95% CI using robust standard errors clustered at the *city*-level. Cities with 9 or more lists are binned together (the full distribution of the number of lists is presented in Table A1).

1.5

1.5

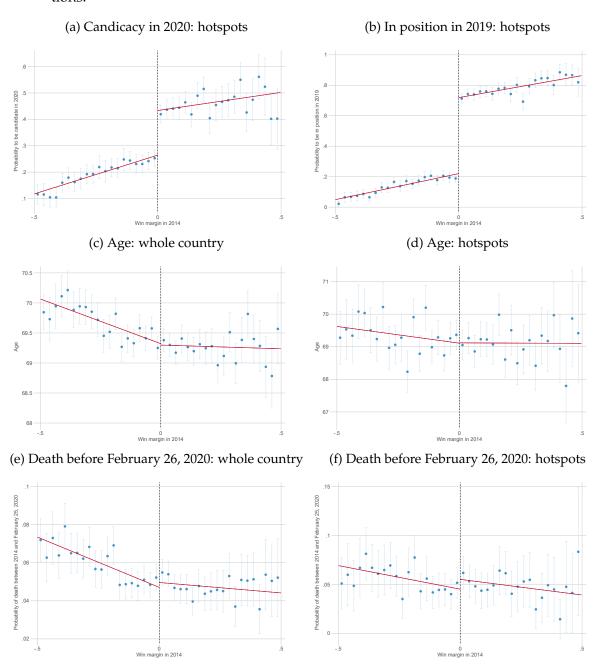
Output

Win margin bin

Figure B3: Distribution of win margins

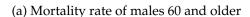
NOTES: This figure displays the density function of the win margin. The variable is discretized into intervals of equal width (corresponding to 1 pp). Graphically there appears to be no obvious sign of manipulation. This finding is confirmed by a manipulation test. The implemented test is based on density discontinuity which is estimated using local-polynomial density estimators following Cattaneo $et\ al.\ (2018)$. The statistic is constructed using a cubic polynomial density estimator with triangular kernel, with different bandwidths above and below the cutoff chosen for an unrestricted model with quadratic local polynomial. The asymptotic variance is estimated using the jackknife estimator. It cannot reject the absence of manipulation (P-value = 0.43).

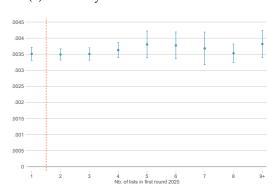
Figure B4: Discontinuity according to the win margin in the 2014 municipal elections.



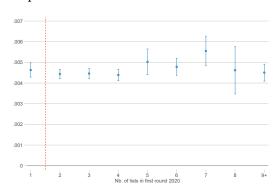
NOTES: This figure presents averages and confidence intervals of the probability of being in position in the city council in 2019 (a and b), the average age (c and d), and the probability of dying between 2014 and February 26, 2020 (e and f) using the win (or loss) margin of the list to which the candidate belonged in 2014 as the running variable, for the whole country (a, c and e), and for Île de France and Grand-Est (b, d and f).

Figure B5: Mortality outcome and age in the general population among 60+ males, depending on number of lists in the first round of the election.

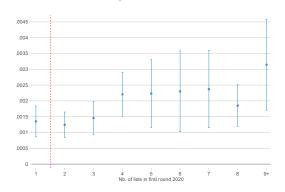




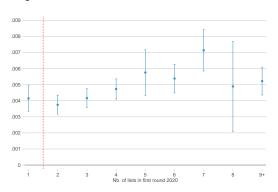
(b) Mortality rate of males 60 and older, in hotspots



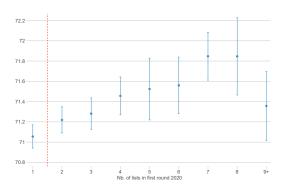
(c) Excess mortality rate of males 60 and older



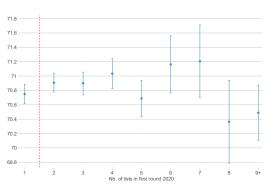
(d) Excess mortality rate of males 60 and older, in hotspots



(e) Av. age of males who are 60 or older



(f) Av. age of males who are 60 or older, in hotspots

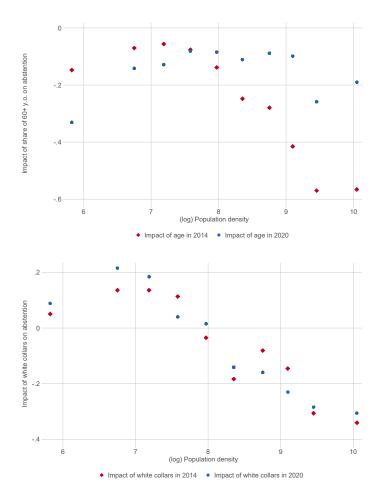


NOTES: This figure presents the number of lists in the first round of the 2020 French municipal election confronted to (a), (b) the mortality rate in 2020 from week 7 to 19, (c), (d) the mortality rate over the same period in excess of the average of 2019 and 2018, (e), (f) average age of males who are 60 or older as of 2016. Dots represent averages, brackets 95% CI using standard errors clustered at the *départment*-level when considering the entire country and at the city-level when considering hotspots only. Cities with 9 or more lists are binned together (the full distribution of the number of lists is presented in Table A1). The data used to compute mortality rate based on place of residence can be found at: https://www.insee.fr/fr/statistiques/4487988?sommaire=4487854 (data published on 06/26/2020). Average mortality and excess mortality rates are weighted by the relevant denominator (here, the number of male residents aged 60 and above).

C Evidence on avoidance behavior at the city level

Figure B6 documents how the age structure of the population correlated with abstention at the city for different levels of density. In particular, the upper panel of Figure B6 displays the coefficient of a regression of abstention on the share of the population (at the city-level) who is 60 and above. The coefficients are estimated separately for each decile of the distribution of density, for the 2014 (red diamonds) as well as 2020 elections (blue circles). We see that for 2014 the share of older people in the local population is consistently associated with lower abstention - a well-documented pattern in empirical studies on individual predictors of turnout (Buisson and Penant, 2017). Moreover, we notice that this association is much stronger in denser cities. We see that this association has changed in 2020 in particular among dense cities where the share of individuals aged 60 or more is associated with smaller (in absolute value) coefficients. This pattern suggests that avoidance took place in dense cities among voters most vulnerable to the ongoing epidemic. The lower panel of Figure B6 shows the same type of regression with another city-level predictor of abstention: the share of white-collar (cadres) workers. We see that the statistical association between the white-collar share and abstention in 2014 changes sign across density deciles: the OLS coefficient is positive at the bottom of the density distribution and becomes negative in the upper half. Interestingly, and somewhat reassuringly, we see that the patterns between 2014 and 2020 are quite similar, unlike what we find for age. As white-collar status is a predictor of voting probability but not a risk-factor with respect to Covid-19, this result is expected and we view it as confirming our interpretation that the differences between 2014 and 2020 according to the share of people older than 60 is indicative of avoidance behavior.

Figure B6: Impact of the share of people older than 60 and of the share of white collar workers on abstention, by decile of population density.



NOTES: This graph presents coefficients of linear regressions of abstention on (a) the share of people older than 60 and (b) the share of white collar workers, by decile of population-weighted density, for the 2014 election (blue dots) and the 2020 election (red diamonds).

D OLS estimates for all 2020 candidates

In this section, we run simple OLS regressions on the whole sample of male candidates in the 2014 municipal elections using mortality between February 26 and April 28, 2020 as the dependent variable and the fact of being a candidate in 2020 as our main covariate of interest. We restrict the sample to individuals alive on February 25th, 2020 – unlike in the RDD sample. It shows a small but negative association between the fact of being a candidate in 2020 and the probability of dying during the Covid-19 mortality peak. The magnitude of this effect is however quite small, it is significant only at the 10% level and 2.5 times lower in magnitude when age and the rank with respect to the last elected candidate are introduced, and the effect is insignificant in hotspots.

Table A6: Coefficients of OLS regressions of Covid-related mortality on being a candidate in 2020 in the whole sample of candidates in 2014.

| | Whole country | | | Hotspots | | | |
|--------------------------|---------------------------|-------------------------------|----------------------------|--------------------------|----------------------------|----------------------------|--|
| | (1) Dep. ' | (2) Var.: Probabili | (3) ity of death be | (4) etween Feb. 2 | (5) 26 and Apr. 28 | (6) 3, 2020 | |
| Candidate in 2020 | -0.00103*** (0.000195) | -0.000658*** (0.000192) | -0.000402* (0.000210) | -0.000846* (0.000484) | -0.000273 (0.000485) | 0.000206 (0.000556) | |
| Age | | 0.000206*** (0.0000227) | 0.000203*** (0.0000225) | | 0.000334*** (0.0000622) | 0.000327*** (0.0000615) | |
| Rank w.r.t. last elected | | | -0.000894*** (0.000242) | | | -0.00155*** (0.000592) | |
| Constant | 0.00237*** (0.000147) | -0.0120*** (0.00150) | -0.0118*** (0.00149) | 0.00306*** (0.000330) | -0.0201*** (0.00417) | -0.0197*** (0.00413) | |
| Observations | 238257 | 238257 | 238257 | 44024 | 44024 | 44024 | |

NOTES: This table presents OLS coefficients of regressions of the probability of dying between Feb. 26, 2020 and Apr. 28, 2020 on the fact of being a candidate in the 2020 municipal election, the age of the candidate, and its rank in the list relative to the last candidate elected. The sample is composed of all male candidates in the 2014 municipal election born before 1960 and alive on February 25, 2020. Robust standard errors clustered at either the county (département) level or the city level (in the hotspot-only samples) are reported in parentheses. * p < 0.1, ** p < 0.05, *** p < 0.01.

Beyond the coefficients, these regression results are quite useful to get a sense of the magnitude of the effects we can exclude. When the whole country is considered and control variables are introduced, as shown in column 3 of the table, the standard errors associated with being a candidate in 2020 are equal to 0.00021, which is in the same range of the effect associated to being one year older. This means than with 95% confidence, we can reject the hypothesis that being a candidate in 2020 would expose individuals to higher excess risks of mortality during the Covid-19 epidemic than being two years older in this period.

E Comparison with existing studies

Our goal in this appendix section is to understand whether differences in the measurement of the local death rate might be important in driving the differences between our results and that of Bertoli *et al.* (2020). Therefore we start by replicating results in Bertoli *et al.* (2020). We first define excess mortality rate for i in week t of 2020 with respect to week t of years 2010 to 2019

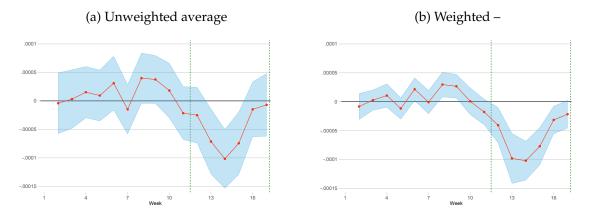
for individuals age 60 or more as:

excess mortality_{it}
$$\equiv \frac{1}{N_i} \left[\text{deaths count}_{it2020} - \frac{1}{10} \sum_{y=2010}^{2019} \text{deaths count}_{ity} \right]$$
 (1)

where N_i is the population of the age group 60+ and deaths count_{ity} is the number of deaths among residents of city i in year y that are 60 or more. Again following Bertoli et al. (2020), we use as a dependent variable the inverse hyperbolic sine of the excess mortality, that is we defined:

$$y_{it} = \operatorname{arcsinh}(\operatorname{excess} \operatorname{mortality}_{it})$$

Figure B7: Measurement error average over weeks (cities with no hospitals).



NOTES: This figure shows averages of the gap between excess of mortality rate measured at the city of death versus at the city of residence. 95% CI intervals computed based on standard errors clustered at by *département* are displayed. Excess mortality is computed as in equation (1) except that only years 2018 and 2019 are used to compute the baseline.

Figure B7 shows the difference between the weekly mortality rate measured at the city of death and the weekly mortality rate measured at the city of residence. The left panel displays unweighted weekly averages across all municipalities without a hospital, while the right panel displays weekly averages weighted per the size of the population above 60 living in each city. The measurement error has substantial variance but it is usually close to zero on average except, crucially, during the episode of excess mortality generated by Covid-19. In times of excess mortality, many more deaths take place outside the city of residence so the local mortality rate measured at the city of death is artificially low. Importantly, this bias is particularly precise using weighted regressions. What this suggests is that, in contrast with Bertoli *et al.* (2020), it is critical that any analysis of excess mortality rates with city-level data a) uses death certificates registered at the city of residence and b) is weighted by population size.

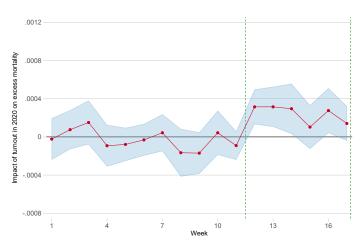
The basic specification used to identify the impact of elections on mortality – see equation (2) of Bertoli *et al.* (2020) – writes as:

$$y_{it} = \alpha_t \ln \left(\text{turnout}_{i2020} \right) + \beta_t \ln \left(\text{turnout}_{i2014} \right)$$
$$+ \mu \ln \left(\text{density}_i \right) + d_{\text{dept}(i)} \times \lambda_{\text{dept}(i),t} + \epsilon_{it}$$
 (2)

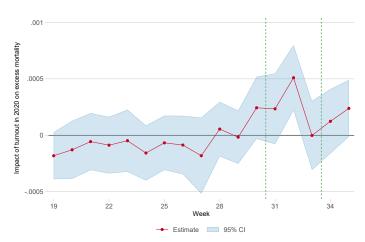
with t = 1, ..., 16.

Figure B8: Estimated impact of 2020 city hall elections turnout on excess mortality: replication and placebo estimates.

(a) Specification as in Bertoli et al. (2020)



(b) 2003 heatwave mortality episod



NOTES: These graphs present coefficients of 2SLS regressions studying the impact of turnout instrumented by electoral competition on mortality during the 2020 Covid-19 pandemic, using (a) the specification from Bertoli *et al.* (2020), and (b) our specification.

As explained in Bertoli *et al.* (2020) and following the logic explained in section 3, we use a measure of local electoral competition as instrument for turnout_{i2020}. This measure is the (logarithm of) the ratio of the number of candidates on the number of local councilors seats to be filled, i.e. competition_i = $\ln \left(\frac{\# \text{candidates}_{i,2020}}{\# \text{seats to be filled}_{i,2020}} \right)$.

The first stage equation writes as:

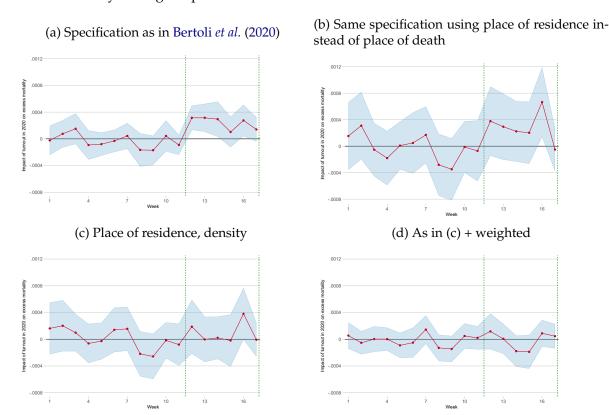
$$\ln(\mathsf{turnout})_{i2020} = \alpha_t^1 \mathsf{competition}_{i,2020} + \beta_t^1 \ln\left(\mathsf{turnout}_{i2014}\right) \\ + \mu \ln\left(\mathsf{density}_{i}\right) + d_{\mathsf{dept}\left(i\right)} \times \lambda_t^1 + \epsilon_{it}^1 \tag{3}$$

Figure B8 replicates the main result of Bertoli *et al.* (2020): it regresses the weekly death rate of residents of age 60 and above on turnout in the first round of the 2020 municipal elec-

tions instrumented with their measure of electoral competition, following the above equation. Figure B8a shows the estimates and confidence intervals of weekly estimates of the effect of turnout on mortality using the specification of Bertoli *et al.* (2020), which does display a very strong increase in mortality after the week of the elections.

The lower panel B8b presents estimates very similar to B8a, but focuses on the 2003 heat wave, another very stark excess mortality period in recent French history. It shows that the same pattern of excess mortality in places with high electoral competition in the 2020 municipal elections is in fact found whenever a mortality outbreak occurs.

Figure B9: Estimated impact of turnout in the 2020 municipal elections on excess mortality during the pandemic.



NOTES: These graphs present coefficients of 2SLS regressions studying the impact of turnout instrumented by electoral competition on mortality during the 2020 Covid-19 pandemic, using (a) the specification from Bertoli *et al.* (2020), (b) a specification with deaths registered at the city of residence rather than at the city of death, (c) a specification with deaths registered at residence and time-varying density controls, and (d) our final specification, with deaths registered at residence, time-varying density controls and weights for population size.

This placebo test is a first suggestion that the data and specification used by Bertoli *et al.* (2020) are inadequate. In order to investigate this more directly, we rerun their main IV specification in B9a and then show how results evolve cumulatively once:

1. one uses deaths registered at residence rather than location of death (see B9b);

- one controls for the logarithm of density interacted with weekly dummies, to allow for the possibility that density affects mortality more during periods of excess mortality (see B9c)¹¹;
- 3. one uses regressions weighted by city population size (see B9d).

What this analysis clearly shows is that the main result in Bertoli *et al.* (2020) gradually loses economic and statistical significance and eventually disappears due to each of those reasonable modifications to their data and specification.

F Data appendix

F.1 Matching procedure between the FND and the file on candidates

In this appendix, we describe the procedure implemented to match the files on candidates in the 2014 elections (referred to as FC2014), the file on the candidates in the 2020 elections (referred to as FC2020) with the Fichier national des décès (referred to as FND). We start by describing the list of variables in the FC2014 and the FND that are used in the matching. In doing so, we will distinguish two versions of the FND which differ in terms of what variables they contain as well as what period they cover.

Variables in FC2014:

- 1. First and last name: The last name tends to be the use name (*nom d'usage*) and is consequently hard to match in the FND for female candidates as the FND only contains the last name at birth (maiden name). This motivates our focus on male candidates.
- 2. Date of birth (DoB)
- 3. City where the candidate is running (CoC): this is likely to correspond with the city of residence in an overwhelming majority of cases.
- 4. *Département* where candidate is running (DoC): we will consider this variable to be the same as the *département of residence* (DoR)

Variables in FC2020:

- 1. First and last name: The last name tends to be the use name (*nom d'usage*) and is consequently hard to match in the FND for female candidates as the FND only contains the last name at birth (maiden name). This motivates our focus on male candidates.
- 2. City of birth (CoB)
- 3. Date of birth (DoB)

¹¹Gascard *et al.* (2020) indeed show that excess mortality during March and April 2020 was particularly pronounced in denser cities.

- 4. City where the candidate is running (CoC): this is likely to correspond with the city of residence in an overwhelming majority of cases.
- 5. *Département* where candidate is running (DoC): we will consider this variable to be the same as the *département of residence* (DoR)

Variables in FND version 1: available since 1970

- 1. First and last name: The last name here is the last name at birth (maiden name).
- 2. Date of birth (DoB)
- 3. Date of death (DoD)
- 4. *City of death (CoD)*: in a large fraction of cases, this city will not coincide with the city of residence given that many people die in hospitals that are unlikely to be located in their city of residence.

Variables in FND version 2: available in the first semesters of 2018, 2019 and 2020.

- 1. Date of birth (DoB)
- 2. Département of residence (DoR)
- 3. City of residence (CoR)
- 4. City of death (CoD)
- 5. Gender
- 6. Date of death (DoD)

Matching FC to FND. Our variable of interest at the individual level is whether someone died during the period stretching from late February to late April 2020.

We start by merging the two versions of the FND for 2020 based on: DoD, DoB, CoD and Gender. This match is exact as DoD, DoB, CoD and Gender uniquely identifies almost all observations in each dataset. We call the resulting dataset FND2020.

We then proceed to match FC2020 with FND2020 based on three sets of variables and consider a candidate as having passed away if he is matched with FND2020 based on a least one set of matching variables.

The three sets of variables as follows: (1) DoR, DoB, Gender, First name, CoB; (2) DoR, DoB, Gender, Last name, CoB; (3) DoR, DoB, Gender, First name, Last name. ¹²

We then proceed to match FC2014 with FND2020 based on three sets of variables and consider a candidate as having passed away if he is matched with FND2020 based on département of residence, date of birth, gender, first name, and last name.

¹²We clean the names from accents when using the names.

Documenting candidacy in 2020 among 2014 candidates. We are able to match FC2014 and FC2020 based on a large set of characteristics (city and département of residence, gender, last name, first name and birth date) which is likely to result in minimal measurement error when documenting whether a 2014 decided to run in 2020.

F.2 Newspaper articles

1. "Coronavirus: how the National Assembly became a new cluster", *Le Monde* (main national newspaper):

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Coronavirus: comment l'Assemblée nationale est devenue un nouveau cluster", March 21, 2020, https://www.lemonde.fr/politique/article/2020/03/21/coronavirus-comment-l-assemblee-nationale-6033906_823448.html, accessed on March 25, 2020. Cited in footnote 1.
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2. "Election night, the other black spot in the first round of the municipal elections.", *Mediapart* (online national media):

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"Les soirées électorales, l'autre point noir du premier tour des municipales" May, 15, 2020, https://www.mediapart.fr/journal/france/150520/les-soirees-electorales-l-autre-point-noir-du-prem accessed June 25, 2020. Cited in footnote 3.
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- 3. "Municipalities: assessors and candidates contaminated with covid-19 in the first round?", *BFMTV* (leading 24-hour news channel):
- 4. "Covid-19: the heavy price paid by elected officials in the first round of municipal elections", *France Info* (public 24-hour news TV and radio channel):

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"Covid-19: le lourd tribut payé par les élus lors du premier tour des municipales", April, 15, 2020, https://www.francetvinfo.fr/sante/maladie/coronavirus/video-covid-19-le-lourd-tribut-paye-par-3917337.html, accessed June 25, 2020. Cited in footnote 3.
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- 5. "Mondicourt: contaminated by Covid-19 during the municipal elections, the mayor has recovered", *La voix du Nord* (large regional newspaper):
 - "Mondicourt: contaminé par le Covid-19 lors des élections municipales, le maire s'est rétabli", April, 22, 2020, https://www.lavoixdunord.fr/744294/article/2020-04-22/mondicourt-contamine-paraccessed June 25, 2020. Cited in footnote 3.
- 6. "Coronavirus: "If I had to do it all over again, I wouldn't do this vote..." "Assessors are found Covid-19 positive after the municipal elections.", 20 minutes (large free newspaper with regional editions):
 - "Coronavirus: "Si c'était à refaire, je ne ferais pas ce scrutin..." Des assesseurs positifs au Covid-19 après les municipales", March 25, 2020, https://www.20minutes.fr/politique/

2747919-20200325-coronavirus-si-refaire-ferai-scrutin-assesseurs-positifs-covid-19-apres-municipales, accessed June 25, 2020. Cited in footnote 3.

7. "Municipalities, did the end of the campaign help the spread of the coronavirus?" *La Croix* (large national newspaper):

"Municipales, la fin de campagne a-t-elle aidé à la propagation du coronavirus?" April, 1,2020, https://www.la-croix.com/France/Politique/Municipales-infections-derniere-ligne-droite-campagne, accessed June 25, 2020. Cited in footnote 3.

F.3 GDPR reference (in French)

Les informations recueillies vous concernant vont faire l'objet d'un traitement destiné à :

- UMR 8545 Paris-Jourdan Sciences économiques (Délégué à la Protection des Données : dpd.demandes@cnrs.fr). Le traitement a pour objet d'identifier l'impact des mesures de distanciation sociale prises pour le Covid-19 depuis mars 2020 sur le taux de mortalité de la population française. Il permet (sous-finalités) :
- mesurer des taux de mortalité dans des positions particulièrement exposées telles que candidat aux élections locales, élu local, commerçant ou professionnel en santé.
- d'informer les autorités publiques lorsqu'il s'agira éventuellement de faire face à une reprise de l'épidémie.

Le traitement de données concerne les candidats aux élections locales, les élus locaux, les commerçants et les professionnels en santé. Les données collectées et traitées sont :

- Données d'identification : nom prénom Sexe date de naissance commune de naissance
- Données sur leur vie professionnelle : Métier dans le secteur de la santé, Poste de direction dans une entreprise, avec numéro SIREN de l'entreprise
- Données sur les opinons politiques : Candidatures lors d'élections locales, détention d'un mandat d'élu local

Les destinataires de ces données sont les chercheurs en charge de l'étude : Laurent BACH, Arthur GUILLOUZOUIC, Clément MALGOUYRES.

Les données sont conservées pour la durée prévue du projet puis définitivement archivées.

La base légale du traitement est (article 6 du RGPD) l'exécution d'une mission de recherche publique.

Ces informations sont recueillies auprès du Ministère de l'Intérieur, de l'INSEE et de l'Agence du Numérique en Santé.

Cette recherche sur les taux de mortalité de professions exposées prévoit le recueil des données qui sont nécessaires au traitement de cette étude. Le traitement ne prévoit pas de prise de décision automatisée (article 22 du RGPD). Afin d'assurer la sécurité des données collectées, les dispositifs suivants ont été mis en place :

- Le local est sécurisé
- L'ordinateur est verrouillé par un mot de passe. Cryptage des fichiers statistiques. L'accès à l'ordinateur est réservé au seul chercheur travaillant sur ce projet sans accès possible de la part de tiers. Aucune mise en ligne sur Internet, aucune interconnexion par serveur ou autre procédé. Aucun échange de données avec qui que ce soit.

Vous pouvez accéder et obtenir copie des données vous concernant, vous opposer au traitement de ces données, les faire rectifier ou les faire effacer. Vous disposez également d'un droit à la limitation du traitement de vos données. Vous pouvez exercer ces droits en vous adressant à : Laurent Bach (laurent.bach@ipp.eu) Vous pouvez contacter également votre DPD à l'adresse suivante : CNRS – SPD Site Jean Zay – Bât Ariane 2 rue Jean Zay 54500 VANDOEUVRE LES NANCY dpd.demandes@cnrs.fr Si vous estimez que vos droits Informatique et Libertés ne sont pas respectés, vous avez la possibilité d'introduire une réclamation en ligne auprès de la CNIL ou par courrier postal.