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United States**

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Drivers of Covid-19 Vaccinations: Vaccine Administration and Delivery Efficiency in the United States

Abstract

This paper adds some formal research to the success of ongoing efforts to combat the COVID-19 pandemic by examining the drivers of the administration and delivery efficiency of coronavirus vaccines. For this purpose, we use data from the 50 US states and place the formal analysis in the context of socio-economic drivers of vaccinations. Results show that state economic prosperity and rural population aid vaccine administration and delivery efficiency. Delivery efficiency improves in states with more nursing homes per capita, in states with more COVID-19 deaths, and with more health workers. The subset of health workers, including physicians and nurses, did not significantly impact administration or efficiency. On the other hand, vaccination efficiency was lower in states with a centralized public health agency. States with a larger share of the elderly population and those with Democrats as governors were no different from others with regard to vaccinations. Robustness checks are performed using vaccination from a more recent period. Finally, a state's legacy of corrupt activity, across two different time dimensions, is broadly consistent with the greasing effects of corruption. While the study uses data from a single nation that is among the first to start vaccinating its population, the findings have relevance for other nations, especially in the Global South, that are starting vaccinations or lagging behind in delivering vaccines.

JEL-Codes: H500, H750, I100, I180, K420.

Keywords: Covid-19, coronavirus, vaccine, efficiency, rural, deaths, health workers, corruption, networking, United States.

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1. Introduction

The recent rollouts of coronavirus vaccines in many nations (and some others in development), coupled with the increasing cases/deaths from the disease, has added public and political pressures to the speedy and widespread dissemination of the vaccines. Even the business sector has recently become concerned with sluggish vaccine rollout due to adverse implications for the speed of economic recovery.¹ This heightened interest has resulted in greater scrutiny of the rollout process, with some initial reports of inefficiency and mismanagement emerging.²

Some nations have been better than others in vaccinating their populations; for example, Israel's dexterity in vaccinating its population has received positive reviews.³ While the institutional, economic, and geo-political factors vary substantially across nations and they might explain key reasons behind vaccination differences,⁴ it is more surprising, even troubling, when subnational governments within a union exhibit substantial differences in vaccine rollouts. In the US some states have much better than others at vaccinating their populations, while others have lagged, sometimes due to no apparent reasons.⁵ Some have touted a lack of transparency behind a failure to understand differences in vaccinations across jurisdictions.⁶ However, before any corrective policy measures can be undertaken, a good understanding of the factors driving vaccinations across US states is essential. It is towards that goal this paper is focused.

Although new information about vaccination efficiency and coverage is literally emerging every day, it seems useful, both from an academic and a policy perspective, to analyze the available information in terms of the identification of bottlenecks in vaccinations. What are the significant factors driving the delivery of vaccines? Are the determinants of vaccine delivery different from the drivers of vaccine delivery efficiency?

In this respect, we examine both the dissemination or diffusion of vaccines (vaccines actually administered) and the efficiency in their delivery (what percent of vaccines that a state received were actually administered) during the early stages of the vaccine rollout in the US. Given the gravity of the underlying health and economic consequences surrounding the pandemic, the speed and spread of vaccinations are of paramount importance both to the public and the policymakers. An identification of significant factors facilitating vaccinations and promoting the efficiency of vaccinations would be useful information not only for jurisdictions where

¹ <https://nypost.com/2021/01/22/wall-street-bigs-press-for-speedier-covid-vaccine-rollout-in-ny/>

² <https://www.msn.com/en-us/news/world/angry-eu-looks-to-restrict-vaccine-exports-as-supply-shortage-hits-slow-rollout/ar-BB1d6Upt>; <https://www.nytimes.com/2021/01/05/opinion/letters/covid-vaccinations.html>

³ <https://www.cbc.ca/news/world/israel-covid-vaccinations-1.5859396>

⁴ For example, using cross-country data, Goel et al. (2020) show how corruption impacts dimensions of supply chain differently.

⁵ See, for examples, https://www.stltoday.com/news/local/govt-and-politics/west-virginia-touts-covid-19-vaccination-success-story-as-national-rollout-sputters/article_b68f9b1a-1feb-5d88-a460-52f2927a0034.html; <https://philadelphia.cbslocal.com/video/5234110-gov-tom-wolf-blaming-federal-government-for-slow-covid-19-vaccine-rollout-in-pennsylvania/>

⁶ <https://news.sky.com/story/covid-19-success-of-vaccine-rollout-will-remain-a-mystery-so-long-as-government-keeps-key-data-under-wraps-12186846>

vaccinations are underway but also in areas that are looking to start vaccinating their populations.

This paper adds initial formal research to the success of ongoing efforts to combat the COVID-19 pandemic by examining the drivers of the administration and delivery efficiency of coronavirus vaccines in the early stages of the vaccination campaign in the US. For this purpose, we use data from the 50 US states and place the formal analysis in the context of socio-economic drivers of vaccinations. Results show that state economic prosperity and rural population aid both vaccine administration and delivery efficiency. Furthermore, more populous states have a weaker record for both outcome measures, at least in the early stages of the vaccine rollout campaign. Delivery efficiency improves in states with more nursing homes per capita, in states with more COVID-19 deaths, and with more health workers. On the other hand, vaccination efficiency was lower in states with a centralized public health agency. States with a larger share of the elderly population and those with Democrats as governors were no different from others with regard to vaccinations. Finally, a state's legacy of corrupt activity is broadly consistent with the greasing effects of corruption in improving vaccine administration and delivery efficiency. Some policy implications based on the evolving data are discussed in the concluding remarks.

The structure of the rest of the paper includes the model in the next section, followed by data and estimation, results, and conclusions.

2. Model

To study the drivers of vaccinations and vaccination efficiency, we start a simple linear model, that accounts for cross-state economic, political social, and medical differences. With the unit of observation at the state level, the estimated equations, with alternative dependent variables, take the following general forms

$$\text{VACadmin} = f(\text{Healthcare market}_j, Z, \text{Corruption}) \quad \dots(1)$$

$$\text{VACeff} = g(\text{Healthcare market}_j, Z, \text{Corruption}) \quad \dots(2)$$

$$j = \text{PHYSICIANS, NURSES, HEALTHworkers, CENTRALIZED, SlhHOSP, nHOMES}$$

$$Z = \text{RGDPpc, RURAL, POP, 65plus, cDEATHS, Governor}$$

The two dependent variables capture qualitatively different aspects of the vaccine delivery process: (i) VACadmin captures the dissemination of vaccinations; and (ii) VACeff captures the related efficiency – the percentage of vaccines that a state received that were actually administered. VACadmin would be related to supply shortages, administrative bottlenecks and supply chain issues (see Goel (2009), Goel et al. (2020)), while VACeff would be primarily limited by internal bottlenecks. Viewed alternatively, our consideration of the two dimensions of vaccinations can be seen as addressing external versus internal bottlenecks, respectively.

The correlation between VACadmin and VACeff in our sample is 0.84 for the January 12, 2021 date considered below and 0.64 for the latter, February 2, 2021, date.

RGDPpc captures economic prosperity in a state, with more prosperous states having a better infrastructure that would aid the administration and efficiency of vaccinations. The effect of rural populations could go either way. On the one hand, a greater rural population would face challenges in accessing medical facilities (reducing vaccinations); on the other hand, informal social and communications networks might be stronger in rural areas that might aid in the dissemination of information in accessing vaccinations. Greater elderly populations (65plus) would facilitate vaccinations as many states have prioritized the vaccinations of elderly residents. Further, the size of a state (in terms of the state population (POP)), could impact vaccinations. One could envision scenarios where smaller states might have coordination or networking advantages at least in the initial vaccination stages. In our sample, a little over a quarter of the population resided in rural areas, and about 17 percent of the population in a state on average was above the age of 65 (Table 1).

Coronavirus deaths (cDEATHS) would also add a sense of urgency to vaccinations in a state, *ceteris paribus*. Conversely, some states with high death rates might direct some resources to curb mortality rates than on preventing infections. As of January 14, 2021, the average COVID-19 related deaths per 100,000 state residents were about 112.

All these aspects, however, are subject to the political climate in a state. Depending upon the political inclinations of the chief executive of a state, the response to combating COVID-19 might be different (British Medical Journal (2020), Chabner (2020)). Accordingly, we include a dummy variable, Governor, to denote states with a governor belonging to the Democratic party (see Potrafke (2018) for a broader discussion of the role of the political variables). In our sample, 46 percent of states had a Democratic governor in 2021.

Later in the paper (Section 4.4), we consider the possible influence of a state's legacy of corruption, using two alternative measures: (a) a five-year average of federal corruption convictions by state (per capita), CorruptSR; and (b) a corresponding ten-year average of corruption convictions, CorruptLR.⁷ Corrupt activity is variable across states in the United States, with convictions showing some lumpiness in specific years. The averaging over time smooths some of the large variations and the longer-term considered would also better account for changes in political regimes in cases certain regimes are more closely associated with corrupt activity.

Corruption can proxy for institutional weakness and can potentially have a “greasing” or “sanding” effect on vaccinations, with possible variations on the administration and efficiency of vaccinations (see Goel et al. (2021)). A legacy of corruption would grease vaccination drives when some intrusive regulations may be bypassed to deliver vaccinations (by health providers or

⁷ Corruption convictions have been widely used in the literature to capture state-level corruption in the United States (see, for examples, Goel and Nelson (2011)).

supply chain participants), yet states with high corruption activity might impose greater transaction cost, slowing down or sanding vaccination drives.

Finally, and perhaps most pertinent to the issue at hand, the role of the medical establishment in a state would crucially influence the delivery and efficiency of vaccinations. In this regard, we consider a number of dimensions: (i) the number of physicians per capita (PHYSICIANS); (ii) registered nurses per capita (NURSES); (iii) more broadly, the number of health workers per capita in a state (HEALTHworkers); (iv) the number of nursing homes per capita (nHOMES); (v) a measure, CENTRALIZED, capturing whether a state's public health system was centralized; and (vi) accounting for fiscal decentralization in health spending by including SIhHOSP to capture state and local government spending on health and hospitals. On average, there were about 6 nursing homes per 100,000 state population, and forty percent of states had a centralized public health agency.

We would expect more medical personnel to aid vaccinations, *ceteris paribus*, although this would partly depend upon the prevalence of other diseases in the state, and the contractual obligations in different states. More nursing homes, at least in the initial vaccine rollout stages where nursing home residents are prioritized, could be expected to result in greater vaccinations in a state, *ceteris paribus*. State and local spending on health signifies government participation in health at the local level. On the one hand, this signal greater state capacity to undertake vaccinations; on the other hand, greater health spending might capture commitments elsewhere (institutional inertia) that might slow or inhibit the rapid transition to vaccinations. Finally, the degree of centralization of the health agency captures decision-making, coordination, and red-tape aspects.

We turn next to a discussion of the data employed and the estimation strategy used.

3. Data and estimation

3.1 Data

Data on COVID-19 vaccine doses distributed to the states and the number of doses administered are taken from the US Center for Disease Control (CDC) COVID Tracker website. These data are updated daily by the CDC and two different dates were selected for analysis. State-level tracking began on December 14, 2020, about the time initial shipments began to rollout to the states. While states were given some guidance from the CDC it was up to the individual states to develop their own distribution plan and prioritization guidelines for who would receive the vaccine. Further, in states with decentralized health care systems, vaccine roll out decisions and implementation plans were often left to local officials. Even in mid-January 2021, state-local distribution plans were still fluid in many jurisdictions, with timelines and priority group designations varying widely across the states. Based on this, we chose – admittedly somewhat arbitrarily - state-level outcomes as of January 12, 2021, as the first date of analysis. Even by

this date, most states were still vaccinating the Phase 1a priority group generally consisting of health care workers and long-term care residents and staff.⁸

The upper-left chart of Figure 1 provides a visual display of the differences among the states in vaccine distribution as of the January 12, 2021, data. The top ten states are shown at the top, depicted in navy blue while the ten states with the lowest number of vaccines administered are shown dark green. As of that date, the mean number of vaccinations for 100 thousand population [VACadmin] was 3,008, with a wide range from 1,355 (Arkansas) to 5,766 (West Virginia). Similarly, the upper-right chart of the figure shows state differences in their record of vaccination efficiency. The mean state vaccination rate [VACeff] stood at 37%, with a low of 15% (Arkansas) to a high of 74% (North Dakota).

To gain additional insights into the state records as they gained more experience with vaccine rollouts the date of February 2, 2021, was also separately analyzed. By this date, most states had advanced to the Phase 1b priority group, with those that followed the CDC Advisory Committee on Immunization Practices (ACIP) recommendations now vaccinating persons 75 years and older.⁹ A few states had even progressed to Priority group 1c. All states included some seniors in the eligibility group, although the minimum age varied from 65 years old to 75 depending upon the state.¹⁰ The bottom two charts show the differences among the states with respect to the two outcome measures considered in the analysis. The mean state vaccination rate by that date rose to 65%, while the average number of vaccinations for 100 thousand population [VACadmin] stood at 9,934, also showing a wide range from 7,087 (Idaho) to 16,874 (Alaska).

Data on state-level COVID-19 related deaths (confirmed or probable) [cDEATHS] are drawn from The COVID Tracking Project and are based on the cumulative outcomes as of January 14, 2021. Mean death rates per 100 thousand population stood at 112.5, a figure that masks wide variation among the states where the death toll was only 22 in Hawaii while New Jersey had a high of 228.

State capacity to conduct the vaccine rollout is measured by per capita state and local government spending on health and hospitals (current expenditure only) for fiscal year 2018

⁸ Four states added seniors to the Phase 1a eligibility groups, including Georgia and Florida (65 and older), Tennessee (75+), and West Virginia (80+). By January 11, 2021, 13 other states and the District of Columbia either had or were transitioning to Phase 2a, and made some older senior age groups eligible for the vaccine. <https://www.kff.org/coronavirus-covid-19/issue-brief/the-covid-19-vaccination-line-an-update-on-state-prioritization-plans/>

⁹ <https://www.cdc.gov/vaccines/hcp/acip-recs/vacc-specific/covid-19.html>

¹⁰ <https://www.kff.org/other/state-indicator/state-covid-19-vaccine-priority-populations/?currentTimeframe=0&print=true&sortModel=%7B%22colId%22:%22Location%22,%22sort%22:%22asc%22%7D#> In analysis not reported below to conserve space, a variable representing the minimum age group eligible for vaccinations at each point in time was also included in each model. The inclusion of this variable did not materially affect any of the conclusions drawn below regarding the other variables in the model. These results are available upon request.

[SlhHOSP].¹¹ Along another related dimension, the classification of the organizational structure of state and local public health agencies with respect to their functional and administrative relationships is based on an analysis undertaken by the Association of State and Territorial Health Officials in 2011. In that analysis states, with no local health units were classified as centralized [CENTRALIZED = 1] as were states where the local units are led by state employees. States where local public health units were led by local employees were classified as decentralized [CENTRALIZED = 0]. Based on their classification criteria, 32 states had decentralized public health systems in 2011 while the other 18 were centralized.¹²

Further details of all of the other variables used in the analysis, descriptive statistics, and data sources can be found in Table 1.

3.2 Estimation

For the vaccinations administered [VACadmin] outcome measure model estimation is carried out using Ordinary Least Squares with robust standard errors. Regarding the vaccination delivery efficiency outcome measure [VACeff], its value lies within the [0, 1] interval when expressed as a fraction. An OLS regression in this case suffers from similar problems to linear probability models involving binary response dependent variables. In particular, predicted probabilities can lie outside the [0, 1] interval and linear relationships with the independent variables in the model are assumed. To address this, we follow Papke and Wooldridge (1996) and Wooldridge (2010) when using VACeff dependent variables and fit a fractional response model using the quasi-likelihood estimator, under the assumption that the distribution of the conditional mean of the response variable follows a logistic functional form.

4. Results

We alternatively discuss results with our two dependent variables: VACadmin and VACeff, respectively.

4.1 Drivers of dissemination of vaccines (VACadmin)

Table 2 reports results with VACadmin as the dependent variable and the very early stages of the vaccine rollout. In the data as of January 12, 2021, a little over 3,000 folks per 100,000 state population were vaccinated on average. The related results are reported in Table 2 with four different modeling variations.

¹¹ Ideally, a more narrowly based measure would have been preferred as the Census Bureau lumps together all forms of spending on public health in the data it reports, including activities such as animal control and inspections of food handling establishments that are not directly relevant in the current context.

¹² In a preliminary analysis, decentralization of state public health care was alternatively measured as the local government share of total state-local government spending on health and hospitals. The results (not reported to conserve space) are qualitatively similar to what is reported below using the centralization [CENTRALIZED] classification variable.

The results show that more prosperous and states with greater rural populations had greater vaccine administrations. The resulting coefficients are statistically significant in all the models estimated. In terms of relative magnitudes, the elasticity of VACadmin with respect to RGDPpc turns out to be 0.8, while that with respect to RURAL is about a fourth of that at 0.2.¹³ From a policy perspective, of course, changing the composition of rural population is very slow, time-consuming, and often politically challenging. Furthermore, smaller states (POP) seem to be more adept at vaccinating their populations, consistent with the notion of relatively better coordination, networking, and communication in such states.

When medical personnel are considered in Models 1.2 and 1.4, we see that the coefficients on PHYSICIANS and NURSES are statistically insignificant, whereas that on the broader measure, HEALTHworkers, is positive and significant (at the 10% level). This finding alludes to the importance of the importance of other health workers - e.g., appointment schedulers, nursing assistants, etc. These resources are especially important in the context of COVID vaccinations with inherent short shelf life, scarcity, and refrigeration/transportation requirements. In addition, states with more nursing homes were no different from others in vaccine administration.

Turning to structural issues with respect to governmental involvement in the health system of states, states with a centralized health systems were no different from others with regard to vaccinations. Surprisingly, state and local governments spending greater amounts on health and hospitals tended to have lower vaccine administrations.¹⁴ This finding can be viewed in the context of greater government size (involvement) leading to some decision-making lethargy or to a lack of coordination between federal and local governments - especially relevant in the context of COVID-19 vaccines (e.g., transportation, logistics, execution, etc.).

With regard to the other controls, states with Democrats as governors, with more elderly populations, relatively greater COVID-19 deaths, more nursing homes, and with more health workers were no different from others when it came to administering vaccines.¹⁵

Later, in Section 4.3, we shall examine how these results change as the initial vaccination period has elapsed, along with a change in government leadership at the federal level. We turn next to examining the determinants of vaccine efficiency.

4.2 Drivers of vaccine administration efficiency (VACeff)

The results with VACeff as the dependent variable using January 12, 2021 data are reported in Table 3. The average percent of vaccines that were delivered over what was received in each state was about 37 percent (Table 1). Vaccine efficiency is especially relevant in the case of the

¹³ Elasticities are calculated based on sample means.

¹⁴ It is also important to acknowledge, however, that the measure of health spending used here encapsulated spending on public health in other areas not directly relevant to the administration of the vaccine to the population (see Footnote 8).

¹⁵ The insignificance of the Governor may be seen in the context of the structural inability or political unwillingness of states' chief executives to appreciably steer vaccination policy in certain directions.

COVID-19 vaccines, not only to save lives but also to conserve and prevent a scarce resource from going waste (because unused vaccines that been on the shelf for a while lose their efficacy).

The results for vaccine efficiency provide some interesting contrasts with those with vaccine administration. More prosperous states again had some comparative advantages in administering efficiency; however, the coefficient on RGDPpc was statistically significant in two of the four models estimated. On the other hand, vaccination efficiency was no different in more rural states and the administration advantages of smaller states were largely absent when it came to efficiency - the coefficient on POP was negative and marginally significant in Model 2.4.

More significant differences in administration and efficiency emerge with respect to the impact of COVID-19-related deaths. States with a greater number of deaths had greater vaccination efficiency. It may be the case that states with more deaths also had more hospitalization, enabling better and faster alternative uses of unused vaccines or it may have to do with the politics of speedy vaccinations in states facing adverse press reports with more deaths.

Again, as in Table 2, vaccination efficiency was greater in states with more health workers, but not just with more nurses and physicians. On the other hand, both structural centralization of state health systems (CENTRALIZED) and fiscal decentralization tended to undermine efficiency.

Efficiency was, however, higher in states with more nursing homes, in the early phases of the vaccine rollout. Whether this is still the case, when the initial focus on nursing home vaccinations becomes less relevant over time, will be examined in the following section. Thus, some of the negative efficiency effects of a state's direct involvement are countered by nursing homes and overall health employment.

As in Table 2 with regard to vaccine administration, states with elderly populations and with Democratic governors were no different from others with regard to vaccination efficiency.

4.3 Robustness checks using alternative timing of vaccine rollouts

As the information on vaccine rollouts is coming out with regularity, we conducted a robustness check by measuring the two dependent variables at an alternative date - February 2, 2021. Besides checking for robustness at an alternative time period where states had greater experience in confronting the challenges faced by the vaccine rollout, the second date also accounts for a change in the government ideology with a change in the presidency and the balance of power at the federal level and in states across the Union.

The corresponding results, using variants of Tables 2 and 3, are reported in Table 4. Overall, we find that the results for the latter period are mixed, with some support for earlier findings and some remarkable differences.

First, more prosperous continued to see vaccine administration gains, but now such states had no efficiency advantages. Quantitatively, the elasticity of VACadmin with respect to RGDPpc is almost half in the latter period compared to the former period (specifically, 0.4 in Model 1.1A,

compared to 0.8 in Model 1.1). Qualitatively similar was the case for more rural states (although the coefficient on RURAL is also marginally significant in Model 2.3A).

Second, less populous states showed greater administration and efficiency than larger states. While the results for vaccine administration support earlier findings, those with efficiency now are stronger than the earlier period. On the other hand, the effects of COVID-19-related deaths on efficiency are not evident in the latter period.

Third, and perhaps most striking, are the results with respect to the medical variables. All three, SlhHOSP, CENTRALIZED, and nHOMES, had no impact on vaccine efficiency, while the negative sign on SlhHOSP supported earlier results with regard to vaccine administration. Further, the negative impact of nursing homes on vaccine administrations is likely due to nearly all states broadening the priority for vaccinations and most nursing home populations already being vaccinated in the United States.¹⁶

Finally, as in Tables 2 and 3, the influences of Democratic governors and elderly populations continue to be statistically insignificant.¹⁷

Whereas the data on vaccination success will emerge over time, this study provides intermediate information when there is still time to fine-tune the vaccination process, with positive implications for avoiding unnecessary loss of life.

4.4 Impact of a legacy of corruption

The corruption risks associated with all phases of the COVID-19 vaccine rollout, from manufacture to allocation and distribution, have been well recognized by international bodies (United Nations Office on Drugs and Crime (2020)).¹⁸ In the present context, state corruption can potentially have an impact on vaccinations, with the effects being either positive or negative, depending upon whether enabling or retarding effects of corrupt activity prevail. Since corruption is hard to detect, especially concurrently, we consider two measures of state-level corruption lagged over two different time periods: (a) CorruptSR: a five-year average of per capita federal corruption convictions in a state; and (b) CorruptLR: a corresponding 10-year average (see Table 1 for details).

Key questions addressed in this respect are:

- What are the effects of corruption legacy on vaccination administration and efficiency?
- Are the effects of short-run and long-run corruption alike?

¹⁶ <https://www.cdc.gov/coronavirus/2019-ncov/vaccines/index.html>.

¹⁷ Some wealthier elderly might be engaging in vaccine tourism and that might have something to do with the insignificance of the resulting coefficient (<https://www.theguardian.com/us-news/2021/jan/31/us-vaccine-tourism-state-borders-covid-19-shots>).

¹⁸ As a recent example in the US regarding vaccine distribution, see “Coming to a black market near you: Covid-19 vaccine.” <https://www.nbcnews.com/news/us-news/coming-black-market-near-you-covid-19-vaccine-n1253504>.

These results, inserting the two corruption variables alternatively into Models from Table 2-4, are reported in Table 5. Since most of the other findings are in broad agreement with earlier results, we will focus on the corruption variables.

Overall, the impact of corruption is consistent with the greasing theory - states with a greater legacy of corruption seem relatively better adept at vaccination administration and efficiency. This might have to do with weak institutions in such states, where strict vaccination mandates (e.g., who to vaccinate on priority, how to ship vaccines and at what temperatures, policing, etc.) better enable administration and efficiency of vaccines. Furthermore, the presence of corruption might increase vaccinations when clinics are open longer/special hours, or allow the jumping of queues via the payment of bribes. It could also be the case that more corrupt states are better able to inflate vaccination success data to show better performance.

We do, however, find differences in the impact of corruption, especially those of *CorruptSR*, across vaccines administration and vaccine efficiency - with *VACeff* positively and significantly impacted by greater corruption legacy (albeit at the 10% level in Model 3.3a). Viewed differently, in the latter vaccination period (data from February 2, 2021), short term corruption legacy significantly and positively impacts vaccination administration, but not administration.

Relatively speaking the effects of short-term corruption are stronger than longer-term corruption - effects of corruption tend to dissipate over time. Thus, the answer to the second question posed above is a no. Quantitatively, comparing Models 3.1a and 3.1b, the elasticity of *VACadmin* with respect to *CorruptSR* is somewhat smaller than that with respect to *CorruptLR* (0.08 versus 0.10, respectively (both based on vaccination data from January 12, 2021)). It would be interesting to see how these findings change as data on concurrent corruption emerges.

The concluding section follows.

5. Concluding remarks

The scale and gravity of the coronavirus have caused alarm and health anxieties worldwide. The concerns resulted in the push by various governments and the private sector to rapidly develop a successful vaccine. As a result, a number of vaccines have been approved, with many in various stages of development. Now the focus has partly shifted to rapid and equitable public dissemination of the scarce vaccines. Given the disparities in the administration of vaccines (see Figure 1) and reports of wastage, an understanding of the factors driving the spread of efficiency of these vaccines seems in order.

This paper formally contributes to the examination of the success of efforts to combat the COVID-19 pandemic by studying the drivers of the administration and delivery efficiency of coronavirus vaccines. For this purpose, we use recent data from the 50 US states and place the analysis in the context of socio-economic drivers of vaccinations.

Besides adding insights into the effective use of public resources, an understanding of the challenges faced by a large-scale vaccination rollout would also help the timeline towards achieving herd immunity in the fight against the current COVID-19 pandemic. Our research offers some important insights from the perspective of a developed economy.

Results show that state economic prosperity and rural population aid vaccine administration and delivery efficiency. More prosperous states have greater resources and better infrastructure to vaccinate their populations, *ceteris paribus*. Our estimates show that a ten percent increase in per capita state GDP would increase vaccinations per capita by about nine percent. It remains to be seen, however, whether improvements over time in the supply chain and with the approval of additional vaccine candidates would preserve or dissipate the advantages of wealthier states.

Delivery efficiency improves in states with more nursing homes per capita, in states with more COVID-19 deaths, and with more health workers. However, the advantage of states with more nursing homes should dissipate after the initial goal of prioritizing nursing home residents is largely achieved. The subset of health workers, including physicians and nurses, did not significantly impact administration or efficiency. A part of this insignificance might be attributed to the fact these health professionals are bound by the CDC and state guidelines pertaining to vaccinations. On the other hand, vaccination efficiency was lower in states with a centralized public health agency. This might have to do with coordination issues, especially when people in line for vaccinations fail to show up and centralized guidelines about how to avoid vaccines from being wasted might not be as effective as local insights into how to get the vaccinations fast to those in the second or waiting tier. Finally, states with a larger share of the elderly population and those with Democrats as governors were no different from others with regard to vaccinations. Globally, Japan and Italy stand out as nations with substantial elderly populations.

Robustness checks are performed using vaccination from a more recent period. They show some differences in the impacts of drivers of both vaccinations and related efficiency. Finally, a state's legacy of corrupt activity, across two different time dimensions, is broadly consistent with the greasing effects of corruption. The prevalence of corruption is even more variable across global nations, (see <https://www.transparency.org/en/>), compared to states within the U.S.

From a policy angle, some of the significant drivers of vaccinations identified in this research, e.g., state population, rural population and even state prosperity change slowly over time and are not amenable to short-term policy manipulations. With respect to the others, such as health sector employment and spending, while more amenable to policy changes, any recommendations would need to be tempered by the fact that they would be based on fast-changing underlying (especially vaccinations) data. On the other hand, political compulsions in states with high COVID cases might force policymakers to change strategies to combat the pandemic.

It is nevertheless our hope that this work provides some initial analytical insights into an important area. While the study uses data from a single nation (United States) that is among the first to start vaccinating its population, the findings have relevance for other nations, especially in the Global South, that are starting vaccinations or lagging behind in delivering vaccines. For

example, the positive vaccination effects of economic prosperity or economic development suggest that poorer nations will likely face extra challenges in vaccinating their populations. This may be related to under-development of institutions, infrastructure, vaccine development and the ability to administer the vaccines, etc. Future work with the benefit of additional data, both within the US context and elsewhere, can be expected to further enlighten us.

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Table 1
Variable definitions, summary statistics and data sources

Variable	Mean (standard deviation)	Source
Vaccine administration: COVID-19 vaccinations administered, per 100 thousand state population. [VACadmin]	as of Jan. 12, 2021 3007.6 (1014.0)	[1]
	as of Feb. 2, 2021 9934.6 (1818.6)	
Vaccine administration efficiency: Percentage of distributed COVID-19 vaccinations that were administered. [VACeff]	as of Jan 12, 2021 36.9 (11.9)	[1]
	as of Feb. 2, 2021 64.5 (7.3)	
Real GDP per capita – 2019 (chained 2012 dollars, in thousands). [RGDPpc]	54.5 (10.4)	[2]
State population – 2019 (per 100 thousand, in natural logs). [POP]	3.69 (1.03)	[2]
Percentage of population classified as rural, 2010. [RURAL]	26.4 (14.6)	[3]
Percentage of population over 65 years old, 2019. [65plus]	17.0 (1.9)	[4]
Democratic governor, 2020. (1=yes, 0=no). [Governor]	0.46 (0.50)	[5]
COVID-19 total deaths per 100 thousand population (Jan. 14, 2021). [cDEATHS]	112.5 (49.7)	[6]
Number of active RN licenses per 100 thousand population, 2021 [NURSES]	1723.8 (394.8)	[7]
Professionally active physicians per 100 thousand population, 2020 [PHYSICIANS]	299.4 (77.9)	[8]
State and local government spending on health and hospitals, fiscal year 2018, \$ per capita [SlhHOSP]	809.1 (451.9)	[9]
Centralized state public health agency, 2011. (1 = yes; 0 = no). [CENTRALIZED]	0.4 (0.49)	[10]
Nursing homes per 100 thousand population, 2016 [nHOMES]	5.8 (2.8)	[11]
Workers in health care occupations 100 thousand population, 2019 [HEALTHworkers]	4670.3 (701.7)	[12]
Corruption, short-term: Average corruption convictions per 100 thousand population, average over 2015-19 [CorruptSR]	0.259 (0.24)	[13]
Corruption, long-term: Average corruption convictions per 100 thousand population, average over 2010-19 [CorruptLR]	0.291 (0.21)	[13]

Notes: All data are by state.

Sources:

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Table 2
Drivers of COVID-19 Vaccinations Administration (Jan. 12, 2021)

Outcome measure: Vaccines Administered per 100k Population [VACadmin]

Model →	1.1	1.2	1.3	1.4
Real GDP per capita [RGDPpc]	42.61** (3.2)	33.38** (2.1)	44.23** (3.8)	25.92* (1.7)
State Population [POP]	-353.49** (2.6)	-335.57** (2.0)	-323.05** (2.3)	-377.63** (3.1)
Rural population [RURAL]	22.58** (2.2)	20.83* (1.9)	25.14** (2.2)	12.07 (1.1)
Population over 65 [65plus]	42.32 (0.5)	-2.50 (0.0)	41.86 (0.5)	-7.91 (0.1)
Democratic Governor [Governor]	-161.35 (0.7)	-174.21 (0.7)	-138.94 (0.6)	-247.51 (1.0)
COVID-19 deaths [cDEATHS]	1.87 (0.8)	1.11 (0.4)	0.26 (0.1)	-0.21 (0.1)
Registered nurses [NURSES]		0.25 (0.5)		
Physicians [PHYSICIANS]		1.45 (0.6)		
State spending on health and hospitals [SlhHOSP]			-0.78** (5.3)	
Centralized public health [CENTRALIZED]			-411.62 (1.3)	
Nursing homes [nHOMES]			34.73 (0.8)	
Health care workers [HEALTHworkers]				0.40* (1.8)
Observations	50	49	50	50
F-statistic	3.98**	4.19**	7.67**	5.17**
R-square	0.39	0.41	0.58	0.43

Notes: See Table 1 for variable details. All models include a constant term and are estimated via ordinary least squares with robust standard errors. The numbers in parentheses are (absolute value) t-statistics. Vaccine administered data are for January 12, 2021.

* denotes statistical significance at the 10% level, and ** denotes significance at the 5% level (or better).

Table 3
Drivers of COVID-19 Vaccination Delivery Efficiency (Jan. 12, 2021)

Outcome measure: Percent of Distributed Vaccines Administered [VACeff]

Model →	2.1	2.2	2.3	2.4
Real GDP per capita [RGDPpc]	0.016** (2.1)	0.014 (1.6)	0.014** (2.2)	0.008 (0.9)
State Population [POP]	-0.110 (1.5)	-0.119 (1.4)	-0.116 (1.5)	-0.125* (1.8)
Rural population [RURAL]	0.009 (1.5)	0.009 (1.5)	0.005 (0.9)	0.003 (0.5)
Population over 65 [65plus]	0.040 (0.9)	0.030 (0.6)	0.050 (1.2)	0.012 (0.3)
Democratic Governor [Governor]	-0.058 (0.5)	-0.055 (0.5)	-0.037 (0.3)	-0.107 (0.9)
COVID-19 deaths [cDEATHS]	0.003** (2.6)	0.003** (2.2)	0.002** (2.0)	0.002* (1.7)
Registered nurses [NURSES]		-0.000 (0.1)		
Physicians [PHYSICIANS]		0.001 (0.5)		
State spending on health and hospitals [SlhHOSP]			-0.0003** (3.4)	
Centralized public health [CENTRALIZED]			-0.347** (2.5)	
Nursing homes [nHOMES]			0.051** (2.4)	
Health care workers [HEALTHworkers]				0.0002** (2.2)
Observations	50	49	50	50
Wald χ^2	14.45**	17.11**	57.86**	21.36**
Pseudo R-square	0.01	0.01	0.02	0.01

Notes: See Table 1 for variable details. All models include a constant term in these fractional response regressions assuming a logit distribution. Dependent variable converted into [0,1] fraction prior to estimation. The numbers in parentheses are (absolute value) z-statistics. Vaccine data are for January 12, 2021.

* denotes statistical significance at the 10% level, and ** denotes significance at the 5% level (or better).

Table 4
Drivers of COVID-19 Vaccination Administration and Delivery Efficiency
(Feb. 2, 2021)

Model →	1.1A	1.3A	2.1A	2.3A
Outcome measure →	Vaccines Administered per 100k Population [VACadmin]		Percent of Vaccines Administered [VACeff]	
Real GDP per capita [RGDPpc]	70.28** (2.5)	85.98** (3.0)	0.007 (1.4)	0.009 (1.4)
State Population [POP]	-630.76** (2.5)	-702.00** (2.2)	-0.090* (1.7)	-0.109** (2.0)
Rural population [RURAL]	37.39** (2.1)	68.37** (3.1)	0.006 (1.3)	0.009* (1.9)
Population over 65 [65plus]	0.19 (0.0)	-1.36 (0.0)	-0.007 (0.2)	-0.002 (0.1)
Democratic Governor [Governor]	263.96 (0.6)	172.15 (0.4)	0.058 (0.8)	0.042 (0.5)
COVID-19 deaths [cDEATHS]	-1.65 (0.4)	2.42 (0.7)	0.001 (0.9)	0.001 (1.4)
State spending on health and hospitals [SlhHOSP]		-0.886** (2.7)		-0.0001 (1.0)
Centralized public health [CENTRALIZED]		-307.02 (0.5)		-0.110 (1.1)
Nursing homes [nHOMES]		-249.86** (2.4)		-0.028 (1.5)
Observations	50	50	50	50
F-statistic/ Wald χ^2	3.89**	3.35**	15.81**	14.61*
R-square/ Pseudo R-square	0.35	0.47	0.003	0.004

Notes: See Table 1 for variable details. All models include a constant term. Models 1.1A and 1.3A are estimated via ordinary least squares. Models 2.1A and 2.3A are fractional response regressions assuming a logit distribution. Dependent variable converted into [0,1] fraction prior to estimation. The numbers in parentheses in all models are robust (absolute value) t-statistics/z-statistics. Vaccine data are for February 2, 2021. * denotes statistical significance at the 10% level, and ** denotes significance at the 5% level (or better).

Table 5								
Drivers of COVID-19 Vaccination Administration and Delivery Efficiency: Impact of legacy of corruption								
Model →	3.1a	3.2a	3.1b	3.2b	3.3a	3.4a	3.3b	3.4b
Outcome measure →	Vaccines Administered per 100k Population [VACadmin]				Percent of Vaccines Administered [VACeff]			
Period →	Jan. 12, 2021		Feb. 2, 2021		Jan. 12, 2021		Feb. 2, 2021	
Real GDP per capita [RGDPpc]	44.33** (3.4)	45.08** (3.4)	72.05** (2.6)	73.90** (2.6)	0.02** (2.2)	0.02** (2.2)	0.01 (1.5)	0.01 (1.4)
State Population [POP]	-349.56** (2.5)	-340.52** (2.5)	-626.73** (2.5)	-611.78** (2.4)	-0.11 (1.4)	-0.11 (1.4)	-0.09* (1.8)	-0.09* (1.7)
Rural population [RURAL]	19.81* (1.8)	18.77* (1.7)	34.55* (1.9)	31.80* (1.8)	0.01 (1.3)	0.01 (1.3)	0.01 (1.1)	0.01 (1.1)
Population over 65 [65plus]	37.46 (0.5)	47.87 (0.6)	-4.81 (0.0)	8.31 (0.1)	0.04 (0.9)	0.04 (0.9)	-0.01 (0.3)	-0.01 (0.2)
Democratic Governor [Governor]	-96.17 (0.4)	-112.13 (0.5)	330.96 (0.7)	336.01 (0.7)	-0.04 (0.3)	-0.05 (0.4)	0.08 (1.0)	0.07 (0.9)
COVID-19 deaths [cDEATHS]	0.77 (0.3)	0.49 (0.2)	-2.79 (0.6)	-3.67 (0.8)	0.003** (2.1)	0.003** (2.1)	0.00 (0.4)	0.00 (0.5)
Corruption - 5-year avg. [CorruptSR]	885.05* (1.8)		909.77 (1.0)		0.30* (1.8)		0.30** (2.3)	
Corruption - 10-year avg. [CorruptLR]		1048.64* (2.0)		1535.34 (1.3)		0.26 (1.0)		0.23 (1.1)
Observations	50	50	50	50	50	50	50	50
F-statistic/ Wald χ^2	4.53**	4.09**	3.38**	3.32**	28.73**	19.20**	22.53**	15.58**
R-sq./ Pseudo R-sq.	0.44	0.44	0.36	0.37	0.01	0.01	0.004	0.004
Notes: See Table 1 for variable details. All models include a constant term. Models 1.1A and 1.3A are estimated via ordinary least squares. Models 2.1A and 2.3A are fractional response regressions assuming a logit distribution. Dependent variable converted into [0,1] fraction prior to estimation. The numbers in parentheses in all models are robust (absolute value) t-statistics/z-statistics. Vaccine data are for February 2, 2021. * denotes statistical significance at the 10% level, and ** denotes significance at the 5% level (or better).								

Figure 1
Vaccines Administered and Vaccines Efficiency: Top Ten and Bottom Ten US States

