

Effectiveness of mask in COVID-19 pandemic in Japan

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Conflict of Interest:

No author has any conflict of interest, financial or otherwise, to declare in relation to this study.

ICMJE Statement

Contributors JK was responsible for the coordination of the study and responsible for the data setting. YO developed the model and TS illustrated the results. All authors contributed to the writing of the final manuscript.

Abstract

Background: The government of Japan had stopped to recommend to wear mask on March 13, 2023.

Object: The object of this study was to estimate effectiveness of mask to prevent for COVID-19 infection through comparison among before and after the stopping date.

Method: The effective reproduction number($R(t)$) was regressed on dummy variable for after the stopped date as well as vaccine coverage, vaccine coverage with some delay, temperature, humidity, mobility, share of the mutated strains, counter measures including Go To Travel Campaign and an Olympic Games. The study period was February, 2020 through April 14, 2023, as of May 7, 2023.

Results: We selected the specification with 90 days lag of waning. In this specification, the second to the fourth vaccination coverage, the first to the third emergency of status, Goto Travel Campaign and Olympic games and the share of other minor mutated strains were significantly negative. Conversely, share of the alpha, delta, and BA2, BA5, and XBB.1.5 sublineage of omicron variant strain,

The estimated coefficient of after the stopped date for recommendation to wear mask was negative but insignificant.

Discussion: The obtained estimated results showed that mask may not have effectiveness to reduce infection of COVID-19. However, it might suggest that the most of Japanese have continued to wear mask without the government did not recommend after the stopped date.

Keywords: SARS-CoV-2, effective reproduction number, mask, , vaccine coverage, waning in vaccine effectiveness

1. Introduction

Since emerging SARS-CoV-2, the Japanese government had recommended to wear mask to prevent for infection. It had ceased on March 12, 2023. Japan was one of the latest country in which cancelled recommendation to wear mask. At the same time, in Japan, people was recommended to wear mask by the government, however, wearing mask has never been legally enforced. Even without enforcement, the most Japanese had wearied mask during pandemic, and many Japanese have kept to wear mask even after March 12, 2023. Therefore,

Wearing mask may be effective to reduce infectivity of SARS-CoV-2 by many observational study [1-12] . Moreover, one experimental study found partially effectiveness[13]. Conversely, one randomized clinical trial found no effective[14], though placebo for mask . Therefore, definitive conclusions about effectiveness of wearing mask might not be reached, yet.

The object of this study was to examine how mask reduced infectivity of SARS-CoV-2 in Japan. So as to do it, we regressed the effective reproduction number, $R(t)$, on mask policy, which was one in the period before the stopped date and zero after the stopped date, as well as other situation such as vaccine coverage, the mutated strain, counter measures or climate condition.

2. Methods

This study examined the numbers of patients including asymptomatic patients reported by the Ministry of Health, Labour and Welfare (MHLW) from February 1, 2020 to April 14, 2023 published [3] as of May 7, 2023.

Estimation procedure for effective reproduction number was the same as previous study [4]. In short, let $f(t)$ was the empirical distribution of incubation period and $g(t)$ was the empirical distribution of the period from onset to be reported in public. Both were indicated in the previous study [4]. Then, $\sum_{s=1} g(s)x(t+s)$ should be estimated the number of patients whose onset date was t where $x(t)$ was newly confirmed symptomatic patients on date t . $x(t)$ was assumed to be constant proportion of total of newly confirmed symptomatic patients including asymptomatic cases, which were reported in public. Now, $\sum_{k=1} \sum_{s=1} f(k)g(s)x(t+s+k)$ should be estimated the number of patients who were infected on date t . On the other hand, let $h(t)$ was distribution of infectivity power defined in the previous study [4]. Then,

$\sum_{k=1} \sum_{s=1} h(k)g(s)x(t+s+k)$ was sum of infectivity on date t . Therefore,

$R(t) = \sum_{k=1} \sum_{s=1} f(k)g(s)x(t+s+k) / \sum_{k=1} \sum_{s=1} h(k)g(s)x(t+s+k)$. Note that proportion of asymptomatic cases does not affect $R(t)$ if it was constant.

The dummy variable for mask policy was one for before March 13, 2023, the stopped date, and was zero for after the stopped date.

Data indicating the shares of mutated variants among all cases were published by the Tokyo Metropolitan Government. Unfortunately, detailed information about mutated strains has not been published for the entirety of Japan. We used four measures for the mutant strain shares in Tokyo, Japan: alpha, delta, BA.1, BA.2, BA.5 and XBB.1.5. sublineage of omicron variant strains [5].

We use average temperature and relative humidity data for Tokyo during the day as climate data because national average data are not available. We obtained data from the Japan Meteorological Agency (<https://www.data.jma.go.jp/gmd/risk/obsdl/index.php>).

Temperatures were measured in degrees Celsius.

Additionally, we identified several remarkable countermeasures in Japan: four state-of-emergency declarations, a travel campaign, and school closure and voluntary event cancellation (SCVEC). The latter, SCVEC, extended from February 27 through March in 2020: this countermeasure required school closure and cancellation of voluntary events, and even cancellation of private meetings. The first state of emergency was declared on April 7, 2020. It ceased at the end of May. It required school closures, shutting down of some businesses, and voluntary restriction against going out. To subsidize travel and shopping at tourist destinations, the “Go To Travel Campaign (GTTC)” started on July 22, 2020. It was halted at the end of December 2020.

The second state of emergency was declared on January 7, 2021 for the 11 most-affected

prefectures. This countermeasure required restaurant closure at 8:00 p.m., with voluntary restrictions against going out, but it did not require school closure. It continued until March 21, 2021. The third state of emergency was declared on April 25, 2021 for four prefectures: Tokyo, Osaka, Hyogo, and Kyoto. Later, the application areas were extended gradually. They never covered the entirety of Japan.

To clarify associations among $R(t)$ and mask policy in addition to current and the past vaccine coverage, the mutant strains, climate, mobility, the Olympic Games, and countermeasures, we used ordinary least squares regression to regress the daily $R(t)$ on daily current vaccine coverage and daily past vaccine coverage as well as dummy variables for the Games, weekly shares of variant strains, daily climate, mobility, and dummy variables for countermeasures.

Because mobility data provided by Apple Inc. or Google Inc. had been ceased to provide to public in March 13, 2022 and October 15, 2022, we could not use mobility data. Thus we discard mobility to estimate for the effective reproduction number in this study even before date of ceased to release, though the previous studies used these information [4].

Information about vaccine coverage was provided Prime Minister's office [6]. We examined the current vaccine coverage of the second to fifth dose without delay. As of the end of year 2022, fifth dose of vaccination had been started. However, its coverage was not high at the end of study period as shown in Figure 1. If a vaccine perfectly protects the

recipient from infection, then the estimated coefficient of vaccine coverage would be -0.01 if one assumes an average of $R(t)$ with no vaccination in the study period. That would indicate that vaccine coverage increased by one percentage point could be expected to reduce $R(t)$ by 0.01 . If the estimated coefficient of vaccine coverage were larger than -0.01 , then it might reflect imperfect personal prevention. Conversely, if the estimated coefficients of vaccine coverage were smaller than -0.01 , then herd immunity might be supposed to have contributed to prevention of infection among non-recipients.

Waning of vaccine effectiveness was measured by the estimated coefficient of vaccine coverage in the past. Particularly, we examined every 30 days prior until 150 days prior. Note that the waning of the fifth vaccination may not be identified with longer lag because it had started recently. We expected the estimated coefficient to be positive if waning was occurring. If its estimated coefficient was positive but smaller than the absolute value of the estimated coefficient of current vaccine coverage, then waning was presumed to be partially occurring. Vaccination was presumed to be effective even if a part of effectiveness was waning. If the estimated coefficient of vaccine coverage in the past was positive and almost equal to the absolute value of the estimated coefficient of current vaccine coverage, then waning was presumed to be complete. We might not expect vaccine effectiveness until that time. Conversely, if the estimated coefficient of vaccine coverage in the past was positive and larger than the absolute value of the estimated coefficient of current vaccine coverage, then

the vaccine might raise infectivity eventually. We supposed waning of vaccine effectiveness in the second and third vaccination because the fourth vaccination had just started in the study period. We also estimate it without any vaccine coverage in the past which implies to be no waning of vaccine effectiveness. We selected length of lag in vaccine coverage in the past through adjusted coefficient of determinant which was a measure of goodness of fit when the number of explanatory variables were not the same.

We specified major variant strains as alpha and delta, and BA.1, BA.2, BA.5 and XBB.1.5 sublineage of omicron variant strain and other minor sublineages of omicron variant strain of omicron such as BF.7, BA.2.75 or BQ1.1. The other minor sublineages of omicron variant strain was represented by dummy variables for November, 2021 and after period. Because the original Wuhan strain had disappeared until October, 2021, we added this dummy variables for November, 2021. Its estimated coefficients might represent average infectivity among the other minor sublineages of omicron variant strain comparison with the original Wuhan strain. On the other hand, constant term represents infectivity of the original Wuhan strain.

We expected the sign of the explanatory variables as follows: mask was expected to reduced infectivity. Also, vaccine coverage in any time reduced infectivity, however, its lag was supposed to be raise infectivity due to waning. The mutated strains were supposed to raise infectivity. Counter measure as the emergency status or SCVEC were supposed to

decline the infectivity. Conversely, Olympic Games and/or GTTC which enhanced to move persons might raise infectivity. We adopted 5% as the significance level.

3. Results

Figure 1 depicts vaccine coverage second and third dose with a 14-day delay. It showed the alpha variant strain emerged in March, 2021 and reached peak on May, 2021. The delta variant strain emerged April, 2021 and reached peak on November, 2021. The micron BA.1, 2, and 5 emerged in December, 2021, February and May, 2021, respectively. The former two variant strains reached peak on February and May, 2022. The latest proportion of XBB.1.5. in Figure 1 was around 40%.

The complete rate for the second vaccination had started in increasing since April, 2021 and it surpassed 80% in November, 2021 but it changed to be almost flat. The third vaccination had started in December, 2021, and it reached 65% in July, 2022.

Figure 2 depicts $R(t)$ during the study period. The highest was around 9 in December, 2021 when the Delta variant strain dominated. The second highest was around in March, 2020 before the first emergency state declared. In other period, it fluctuated around one and less than three at highest.

Table 1 presents estimation results. We selected the specification with 90 days lag of waning of the second and third dose vaccination. In this specification, vaccine coverage except for the fifth dose and those with lag, and proportion of alpha and delta, and BA.2 and BA.5 sublineage of omicron variant strain, the first to third emergency status were significant with the expected sign. Conversely, SCVEC and the fourth emergency status were significant but with unexpected sign. Goto Travel Campaign and Olympic games were significantly negative and thus these events reduced infectivity. The estimated coefficient of mask was negative, but not significant. XBB.1.5 sublineage of omicron variant strain was insignificant. Other minor sublineages of omicron variant strain was significantly negative and thus these minor sublineages had lower infectivity than the original Wuhan strain. This result might be consistent with these sublineage could not be major sublineage.

Discussion

The obtained estimated results showed that wearing mask might not reduce infectivity. However, we have to remind that the period without mask was very short, just a one month. Therefore, estimation result about effectiveness of mask might change over time. Moreover, after the stopped date, all Japanese did not necessarily cease to wear mask on the stopped date. Quite a few Japanese might continue to wear mask after the stopped date. Unfortunately, proportion of wearing mask day by day was not available. If majority

of Japanese had continued to wear mask after the stopped date, our estimation procedure cannot find out some effectiveness of mask.

XBB.1.5 may not have higher infectivity than BA.1, BA.2, BA.5 or the other minor sublineage of omicron variant strains. Note that basic reproduction number, R , for the mutated strain was probably impossible because vaccination was prevailed, and other counter measures was activated, and we cannot exclude these effect completely.

Conversely, $R(t)$ can always be calculated as described above because it depends these several factors affected infectivity as well as R . Therefore, this study focused $R(t)$ instead of R and then examined how several factors affected it. Therefore, the obtained result in this study might be more reliable than the previous study. Actually, in Figure 2, the last peak caused by XBB.1.5. was not much higher than other peak such as the high peak around the end of 2021 caused by BA.1. or peaks in 2020 caused by the original strain.

We found that waning of the second to fourth dose vaccine with 90 days prior was the most appropriate specification. This duration may be comparable with earlier studies of waning [6,7], which reached their conclusions based on antibody titer or test negative design. Readers must be reminded that waning estimated for the present study might include behavioral changes among the vaccinated persons to adoption of more risky behavior that is prone to exacerbating infectivity. Such behaviors and the vaccine itself affect waning results, but they are not separately discernible based on results of this study. Weakening of

immunoreaction and behavioral change are separate factors, but their mutual effects might be the most important for management of public health.

Limitations

First, we assumed implicitly that epidemiological characteristics including incubation period or delay in reports were the same among the original strain, alpha, delta, and sublineages of omicron variant strains. However, results of one study indicated that the delta variant strain has a shorter incubation period than either original strain [15].

Secondly, readers must be reminded when interpreting the obtained results that they do not indicate causality. Even though it was not significant, the obtained results of this study demonstrated that a negative association exists between mask and infectivity. That finding does not necessarily mean that mask wearing reduced infectivity. The lower infectivity might have caused or might have even simply coincided with mask wearing.

Conclusion

The obtained estimated results showed that mask may not have effectiveness to reduce infection of COVID-19. Because the length of the period after the stopped date was too short, data accumulation should be necessarily to consider this problem.

The present study is based on the authors' opinions: it does not reflect any stance or policy of their professionally affiliated bodies.

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Ethical considerations

All information used for this study was from official data published on the internet. There is therefore no ethical issue related to this study.

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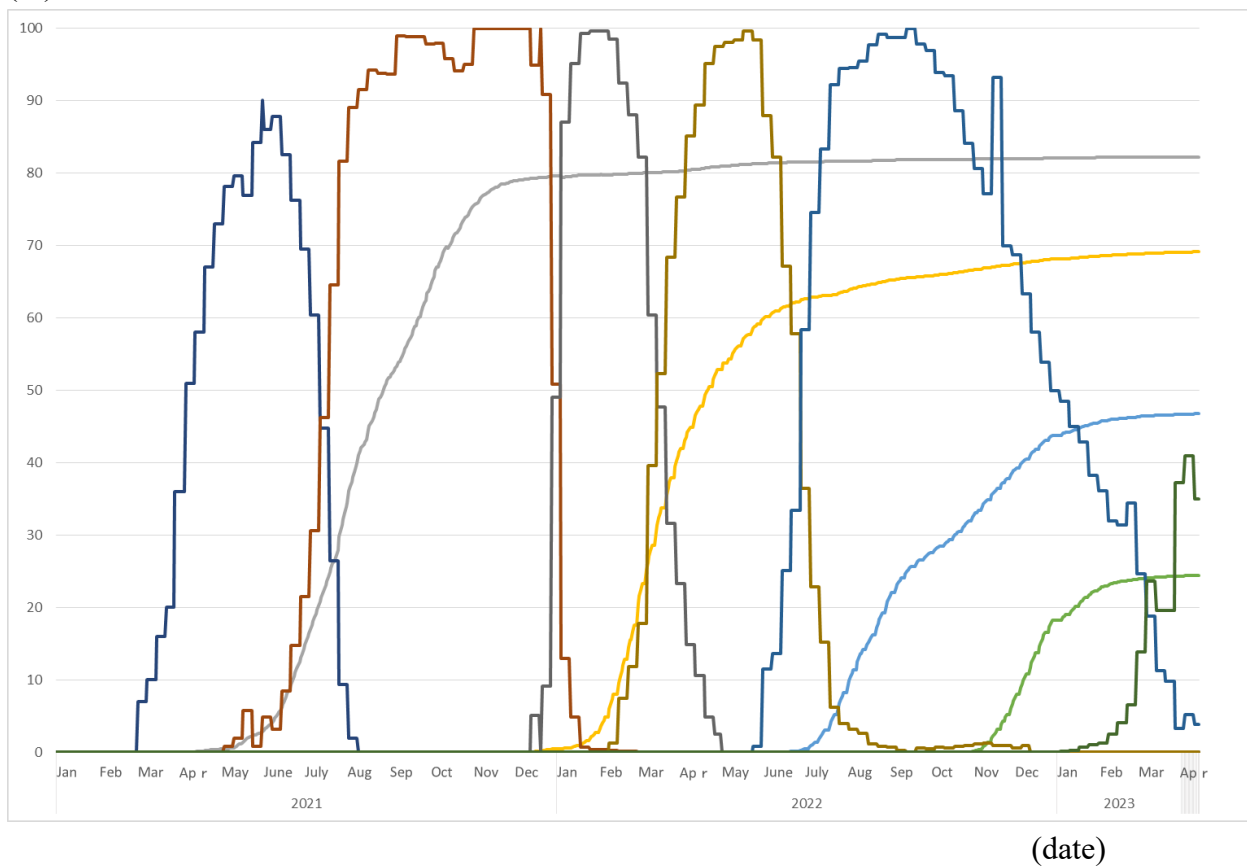
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Figure 1: Vaccine coverage and shares of alpha and delta variant strains in 2021 until April 14, 2023.

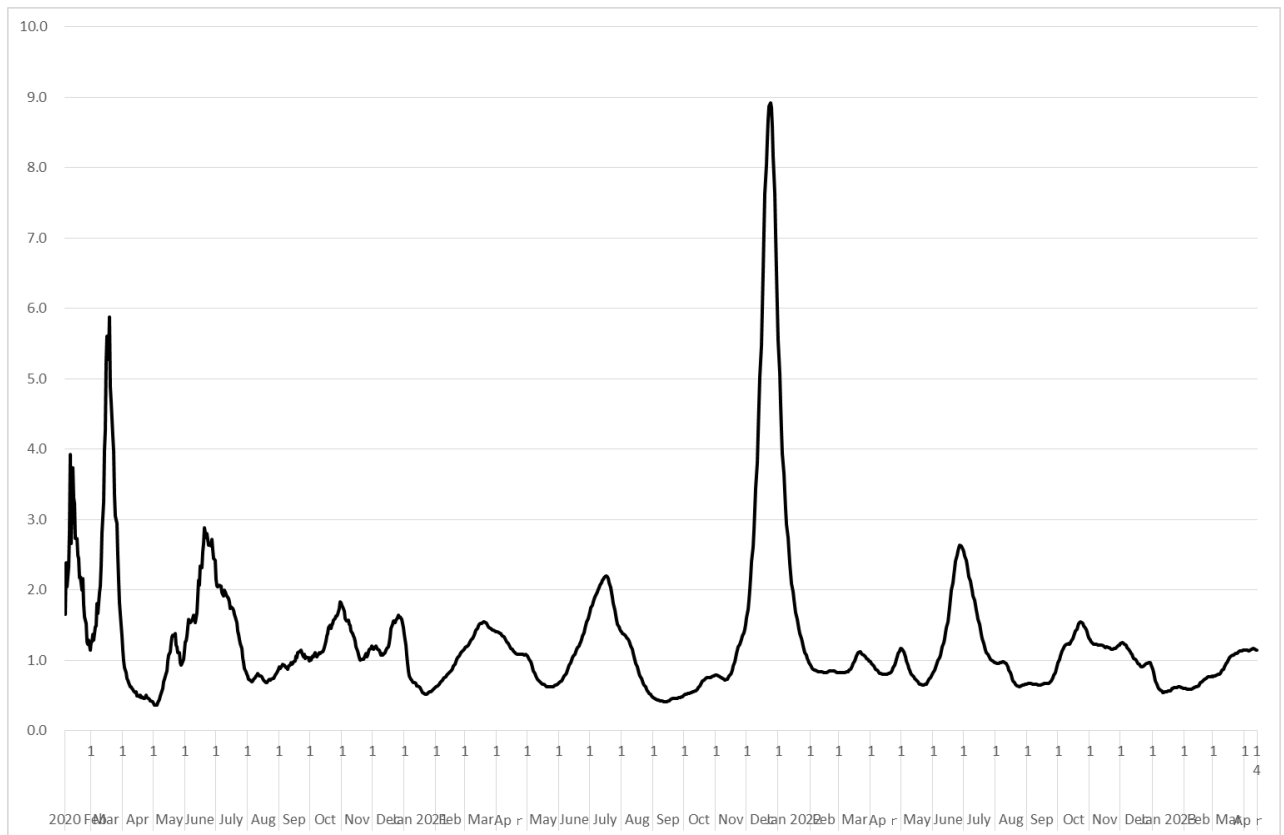
(%)



Note: Five lines with chevron shape and dark green increasing line started since September, 2022 indicates proportion of variant strains. The most left dark blue line with chevron shape indicates alpha variant strain, and orange line was delta variant strain. Following three lines were BA.1, BA.2 and BA.5 sublineage of omicron variant strain. Dark green increasing line started since January, 2023 indicates XBB.1.5 sublineage of omicron variant strain. Monotonically increasing four line except for dark green increasing line started since September, 2022 indicates vaccine coverage rate. Gray increasing line indicates vaccine coverage of second dose, yellow increasing line was the third dose, and blue increasing line was the fourth dose. Light green increasing line started since November, 2022 indicates vaccine coverage of the fifth dose.

Figure 2: Effective reproduction number from February, 2020 through April 14, 2023

$R(t)$



(date)

Note: The line represents the effective reproduction number in Japan from February, 2020 through April 4, 2023, as of the Mat 7, 2023. Calculation procedures are explained in the main text.

Table 1: Estimation results of $R(t)$ on wearing mask, vaccine coverage, prevalence of the variant strains, and Olympic Games with the climate condition, and countermeasures.

Lag of waning (days)	Without waning		30		60	
	Estimated coefficient	<i>p</i> -value	Estimated coefficient	<i>p</i> -value	Estimated coefficient	<i>p</i> -value
Mask	-0.2464	0.461	-0.38074	0.237	-0.40667	0.161
Temperature	-0.02076	0	-0.01131	0.043	-0.00123	0.806
Humidity	-0.00215	0.247	-0.00119	0.504	0.000951	0.552
SCVEC	0.812375	0	0.892072	0	0.984042	0
1 st State of emergency	-1.10841	0	-1.09469	0	-1.08184	0
GTTC	-0.6385	0	-0.64972	0	-0.66562	0
2 nd State of emergency	-1.10132	0	-1.01657	0	-0.90239	0
3 rd State of emergency	-0.80946	0	-1.23766	0	-1.49257	0
4 th State of emergency	-0.16575	0.343	0.653678	0.001	2.108386	0

Olympic	0.405953	0.084	0.979412	0	0.584988	0.004
Games						
Vaccine						
coverage of						
the second	-0.02602	0	-0.10221	0	-0.20939	0
dose(%)						
Vaccine						
coverage of						
the second			0.103125	0	0.22143	0
dose with lag						
(%)						
Vaccine						
coverage of						
the third	0.017037	0.17	-0.07128	0	-0.06695	0
dose(%)						
Vaccine						
coverage of						
the third dose			0.061777	0	0.039942	0
with lag (%)						

Vaccine

coverage of

-0.034

0

-0.13781

0

-0.07304

0

the fourth

dose(%)

Vaccine

coverage of

the fourth

0.115788

0

0.084662

0

dose with lag

(%)

Vaccine

coverage of

-0.01382

0.507

0.002085

0.94

-0.00378

0.841

the fifth

dose(%)

Vaccine

coverage of

-0.00405

0.87

-0.02129

0.224

the fifth dose

with lag (%)

Share of

alpha variant -0.00058 0.8 0.006706 0.007 0.01208 0

strain (%)

Share of delta

variant strain 0.00705 0.075 0.000544 0.917 0.039499 0

(%)

Share of

BA.1

sublineage of -0.01055 0.004 -0.01577 0.001 -0.00484 0.175

omicron

variant strain

(%)

Share of

BA.2

sublineage of -0.02371 0.001 -0.01369 0.072 0.012392 0.086

omicron

variant strain

(%)

Share of

BA.5

sublineage of

-0.01411 0.065 -0.00324 0.721 0.013995 0.083

omicron

variant strain

(%)

Share of

XBB.1.5subli

neage of

0.002404 0.838 -0.00138 0.919 0.010105 0.437

omicron

variant strain

(%)

Other minor

sublineage of

2.558108 0 1.650689 0 -0.16053 0.483

omicron

variant strain

Constant 2.569967 0 2.464329 0 2.156395 0

Adjusted R^2 0.3461 0.4082 0.521

Notes: The dependent variable was $R(t)$, effective reproduction number; GTTC stands for “Go To Travel Campaign”; SCVEC denotes school closure and voluntary event cancellation. The sample period was February 1, 2021 through April 14 2023, as of May 7, 2023. Number of observations was 1168. Yellow markers indicate significant except for constant term.

(cont.)

Lag of		90	120		150	
waning(days)						
Explanatory variable	Estimated coefficient	<i>p</i> -value	Estimated coefficient	<i>p</i> -value	Estimated coefficient	<i>p</i> -value
Mask	-0.40099	0.268	0.322583	0.35	-0.46104	0.129
Temperature	-0.00403	0.408	-0.01033	0.042	-0.0102	0.049
Humidity	-5.2E-05	0.973	-0.00169	0.302	-0.00217	0.194
SCVEC	0.953756	0	0.887228	0	0.907865	0
1 st State of emergency	-1.08693	0	-1.10164	0	-1.07804	0
GTTC	-0.66193	0	-0.65809	0	-0.63388	0

2 nd State of emergency	-0.92837	0	-1.0043	0	-0.98503	0
3 rd State of emergency	-1.12537	0	-0.85535	0	-0.79955	0
4 th State of emergency	1.280828	0	-0.56675	0	-1.58278	0
Olympic Games	-0.53779	0.008	-0.31224	0.139	-0.43737	0.046
Vaccine coverage of the second dose(%)	-0.19167	0	-0.107	0	-0.11821	0
Vaccine coverage of the second dose with lag (%)	0.25483	0	0.157661	0	0.176541	0
Vaccine coverage of	-0.18161	0	-0.10208	0	-0.0786	0

the third

dose(%)

Vaccine

coverage of

0.134595	0	0.23112	0	-0.17779	0
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the third dose

with lag (%)

Vaccine

coverage of

-0.07236	0	-0.23592	0	0.271068	0
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the fourth

dose(%)

Vaccine

coverage of

0.095956	0	0.07737	0.279	0.049666	0.038
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dose with lag

(%)

Vaccine

coverage of

-0.02405	0.255	-0.0761	0.365	-0.14118	0.003
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the fifth

dose(%)

Vaccine						
coverage of						
the fifth dose	-0.0215	0.623	-0.00148	0.973	-0.06968	0.646
with lag (%)						
Share of						
alpha variant	0.006132	0.002	0.000535	0.79	-0.00024	0.908
strain (%)						
Share of delta						
variant strain	0.0647	0	0.052727	0	0.071927	0
(%)						
Share of						
BA.1						
sublineage of						
omicron	-0.00345	0.307	-0.01015	0.002	-0.00718	0.062
variant strain						
(%)						
Share of						
BA.2	0.061337	0	0.014336	0.056	-0.01823	0.008
sublineage of						

omicron

variant strain

(%)

Share of

0.023212

BA.5

sublineage of

0.004

-0.04647

0

0.012107

0.15

omicron

variant strain

(%)

Share of

XBB.1.5

sublineage of

0.007506

0.68

-0.04552

0.009

0.008566

0.534

omicron

variant strain

(%)

Other minor

sublineage

-2.89005

0

-0.86875

0.001

0.668418

0.002

of omicron

variant

strain

Constant	2.274261	0	1.787472	0	2.579917	0
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Adjusted R^2	0.5442		0.4951		0.46440.4708	
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