

1 **A seven-day cycle in COVID-19 infection, hospitalization, and mortality**
2 **rates: Do weekend social interactions kill susceptible people?**

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17 **Key Words:** COVID-19, Pattern, Rhythms, Cycle, Social interaction, weekend

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1 **Abstract**

2 Seven-day cycles in numbers of COVID-19 new-cases and deaths are markedly evident in
3 most public databases (e.g. Worldometer, ECDC), but it is unclear whether they reflect
4 systematic artifacts of delays in information reporting/gathering, or have a more profound
5 basis. To address this question we located 11 databases of US states that provide date-
6 authenticated information (actual date of symptom onset and/or specimen collection, or
7 actual hospitalization or death date) that reported more than 1,000 deaths each. Numbers
8 of new cases showed a weekly cyclic pattern in 10 out of 11 states, commonly peaking on
9 weekdays, 2-6 days after the weekend, corresponding with a reported median 5-day lag
10 between infection and the manifestation of clinical symptoms. We postulate that this
11 pattern emerges from interactions with different and/or extended social-circles during
12 weekends, including increased inter-generational meetings, which in turn facilitate transfer
13 of COVID-19 from younger people to older vulnerable individuals. Furthermore, we found
14 weekly periodicity in hospitalizations in 2 out of 2 authenticated databases providing this
15 information. Actual death date, which is more difficult to attribute to individual choice, and
16 is expected to occur approximately 2-3 weeks following hospitalization, showed significant
17 7-day periodicity in 1 out of 11 states, and a trend in 2 additional states. If weekly peaks in
18 new cases can be truncated by physical/social distancing, especially during weekends, the
19 mortality of COVID-19 may be reduced, or at least hospitalization and mortality curves may
20 be flattened.

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1 **Significance Statement**

2 We believe our findings are significant, as it appears that it is difficult for people to grasp an
3 ambiguous “price” for visiting their friends and family, when they cannot be certain that
4 they are not carrying and spreading the virus. Our findings could be used as an effective
5 “tool” to demonstrate such a cost, clearly presented in terms of number of excessive
6 infections. During these days of uncertainty, we believe it is fundamental to provide
7 scientific facts that could illuminate this connection and make it tangible for the general.
8 The amplitude of the cycles we describe here is such that many thousands of infections
9 could be averted by carefully scrutinizing local policies, medical practice, and social norms.

1 **Introduction**

2 On May 25th, 2020, there were over 5.5 million reported cases of SARS-COVID-2 (COVID-
3 19) worldwide, and over 346,000 deaths¹. In many countries social distancing guidelines
4 and/or restrictions have been put in place to limit the spread of the disease. Yet, there
5 seems to be a world-wide tendency to negate some of these restrictions, at least within
6 certain subpopulations. In a few countries no formal restrictions have been imposed on
7 social interaction. Importantly, the pool of people with whom social interactions occur may
8 depend on the specific day of the week. During working days, most interactions may occur
9 within a stable pool of people, namely household members and/or co-workers, whereas
10 during weekend days, a different and more diverse pool of people may apply, including
11 parents, other relatives, and the extended social network.

12 The price of not imposing restrictions on social interactions, or partially or completely
13 defying them, is ambiguous, although it is assumed that those who do not maintain social
14 distancing are at higher risk of getting infected and spreading COVID-19. Also it is unknown
15 whether temporary or partially loosening of restrictions, specifically during a weekend or a
16 holiday (by the entire population or a subset of it), or of extending the pool of people with
17 whom we interact during the weekend, has a significant impact on the number of infected
18 people or deaths.

19 To address these questions, we scanned Our World in Data database¹ (derived from the
20 European Center for Disease Prevention and Control (ECDC) data), as well as specific
21 databases of states within the USA, for (i) daily new cases, (ii) daily numbers of
22 hospitalizations (when available), and (iii) daily deaths. We observed marked weekly

1 rhythmicity in all three indices. To test if the observed patterns are not random, and
2 whether they follow a 7-day cyclic rhythm, we studied all 12 developed North American
3 and European countries that reported more than 1,000 deaths by May 4th. Because some
4 databases ascribe events to the day the information becomes available, rather than to the
5 actual date of event, we also collected data from resources that strictly ascribe events to
6 their occurrence date. To this end we analyzed such authenticated data from 13 states
7 within the USA that had more than 1000 death and clearly indicated that the data is
8 authenticated to the date of event occurrence (of whom 11 provided data on new cases, 2
9 on hospitalization, and 11 on deaths).

10 **Results**

11 We analyzed data from March 29 to May 4th (± 3 days to create a 7-day moving average).
12 Non-authenticated databases used were Worldometer and ECDC, assessing all 12 European
13 and North-American countries having more than 1,000 deaths by May 4th. Authenticated
14 databases were all those we were able to locate in US states that ascertained date-
15 authenticated data and had more than 1,000 deaths by May 4th. Table 1 summaries all
16 results for both date-authenticated and non-date-authenticated databases.

17 **New cases and hospitalization:** We first analyzed non-authenticated data from
18 Worldometer regarding US as a whole, and found a significant 7-day periodicity in new
19 cases, having a sinus shape, peaking on Fridays and decreasing toward a nadir on Monday
20 (Figure 1A). Within the additional 11 European/North-American countries, very similar
21 patterns were observed in 8 of them (Table 1, and Supp. Fig 1 for representative countries
22 that show this periodicity).

1 We then analyzed data from US states that clearly indicated that the information provided
2 is date-authenticated. Eleven states provided such data on new cases (see a complete list in
3 Table 1), and in 10 of them a significant periodic seven day cycle was observed (See Fig. 1-3
4 and Supp. Fig. 2-4 for representative states data), commonly peaking on weekdays, 2-6
5 days after the weekend. In Massachusetts, the databases also included the percent of
6 positive tests/day, which did not show periodicity and ranged between 10%-30% along the
7 entire period studied.

8 In New York City we located date-authenticated data for new cases, hospitalization, and
9 deaths. Both new cases and hospitalizations exhibited a significant 7-day periodicity, while
10 deaths showed a statistically non-significant pattern for a 7-day periodicity (Fig 3).
11 Hospitalizations also showed a 7-day periodicity in Virginia, summing up to 2 of 2 states
12 where this authenticated information was available.

13 **Death patterns:** Based on non-authenticated databases, of the 12 European/North-
14 American countries studied, 9 presented a significant 7-day periodicity (See Fig. 1 and Supp
15 Fig 1 for examples). Within the 11 US states providing date-authenticated data on death
16 date, only 1 presented a statistically significant 7-day periodicity, 2 additional states
17 presented a noticeable but non-statistically-significant 7-day periodicity, and 8 did not
18 present any consistent periodicity.

19 **Discussion**

20 Worldmeter database showed a significant 7-day periodicity in the US, both regarding new
21 cases that peak on Fridays, and deaths that peak on Tuesdays/Wednesdays. Based on ECDC
22 non-authenticated data for North-American/European countries, a 7-day cyclic pattern

1 was evident in 8 out of the 12 countries studied, regarding new cases, and in 9 out of the 12
2 countries regarding number of deaths, with peak incidence of new cases occurring mostly
3 on Thursdays-Saturdays, and peak incidence of death occurring mostly on Wednesdays-
4 Fridays.

5 More important is data from date-authenticated US states that showed statistically
6 significant 7-day periodicity patterns. Specifically, a significant 7-day pattern was found in
7 10 of 11 states regarding new cases, 2 of 2 regarding hospitalizations, and 1 of 11 regarding
8 deaths.

9 Several characteristics of the COVID-19 pandemic and disease course are known and
10 potentially hold relevance to these observations and to our speculated interpretations.
11 First, the disease is commonly transmitted through respiratory droplets during un-
12 protected social interactions³. Second, older people account for most recognized cases of
13 the disease, and for the great majority of fatalities⁴, and asymptomatic people carry and
14 transfer the virus⁵. Third, epidemiological studies indicated that the median duration
15 between contacting a sick/carrying individual and exhibiting clinical symptoms of COVID -
16 19 infection is 4-5 days (50% up to 5 days)⁵. Also relevant to our hypothesis, a trend
17 towards increased social mixing during the weekends has been reported (including in
18 countries present in our analyses⁶), and weekend interactions are expected to involve a
19 different pool of people than weekday work & households interactions.

20 Because the restrictions imposed in many countries or adopted voluntarily include limiting
21 or eliminating workplace hours/interactions, at least for people at high-risk, we
22 hypothesize that susceptible/older people and retired people may become infected at

1 higher rates during weekend-days compared to weekdays, as a result of increased social
2 interactions with younger relatives or friends during the weekend⁶, or as a result of
3 interacting with a different and extended pool of people. Under this assumption, and in
4 accordance with the above known characteristics of COVID-19, it follows that a significant
5 portion of these vulnerable individuals will exhibit clinical signs of COVID-19 infection at
6 higher rates ~4-5 days after the weekend, on Wednesday-Friday. The pattern regarding
7 deaths is expected to exhibit greater variance due to the prolonged lag from infection to
8 death (2-3 weeks from symptoms onset to death), and a variety of medical and
9 environmental intervening variables. Both predictions are indeed evident in our results of
10 date-authenticated data. A higher infection rate during weekend days may stem from lower
11 social distancing, or higher frequency of interactions between young and old individuals
12 during the weekend. Furthermore, as most new symptomatic cases of COVID -19 occur in
13 people older than 70, and the majority of deaths are within this age group, we believe that
14 this vulnerable population is infected by unaware younger relatives or friends carrying the
15 virus. It is also likely, but need to be verified, that this vulnerable population exhibits
16 shorter and more consistent lags of time from infection to symptom onset and to death.

17 Alternative explanations for the seven-day cycle in the numbers of reported new cases and
18 reported COVID-19 mortality should be considered. A leading hypothesis would be a
19 selective delay in reporting of these events. Specifically, one may suggest that while new
20 cases and deaths are distributed evenly along all 7 days of the week, reports of these events
21 are delayed during the weekend, and erroneously ascribed to 2-4 days later, whereas
22 events occurring on weekdays are reported without delay. In the USA, however, we were
23 able to locate 11 states with more than 1000 deaths that indicate the actual date of new

1 cases, as defined by the actual day of symptoms onset, or the date a positive test was taken
2 (see Table 1). Hospitalization (2 states) and death rates (11 states) were also logged
3 accurately and authenticated. In 10 of these 11 states, the seven-day cyclic pattern is
4 clearly evident regarding new cases, and in 2 of 2 states the phenomenon is observed
5 regarding hospitalization, refuting a delayed report as an alternative explanation for a
6 weekly cycle in new cases. One may nevertheless argue that people that present symptoms
7 of COVID-19 on the weekend may delay taking a test (or reporting symptom onset) to
8 Monday/Tuesday, in contrast to those that present symptoms on weekdays, who would not
9 delay it. This scenario will yield a marked peak of new cases on Monday/Tuesday,
10 decreasing towards Friday, which is not evident. We also checked for the percent of
11 positive tests along the entire week in Massachusetts (where this information was
12 available and a 7-day cycle was significant), and found no significant cycles in this index,
13 nor a shortage in tests during weekends.

14 Weekly cycle pattern in death may stem from previous periodicity in infection dates, and is
15 unlikely to be attributed to individual choice, as most patients are anesthetized long before
16 dying. Of the 11 date-authenticated states in the US, this pattern was significant only in one
17 (Indiana), an evident but not statistically significant in additional 2 states (Massachusetts
18 and Florida). The effect sizes for death seem smaller than for new cases and hospitalization.
19 This less profound cycle would be expected given the longer delay from infection to death
20 (average estimation of 2-3 weeks), the expected greater variance in developing severe
21 symptoms leading to death, and the smaller number of events. Thus, the more pronounced
22 weekly cycles in 9 of the 12 countries based on ECDC data may indeed be attributed, at
23 least to some extent, to weekends delay in reporting deaths. However, the smaller but

1 statistically significant effects in the date-authenticated states in the USA suggest that this
2 cycle is a remnant consequence of weekend high-infection rates.

3 If our above postulations indeed account for the observed 7-day patterns, many lives may
4 be saved by greater adherence to social/physical distancing, especially during the
5 weekend. Alternatively, such adherence may at least flatten the curves of hospitalization
6 and mortality, which in some countries with restricted medical resources would also lead
7 to saving lives. In the 11 states in the US, where date-authenticated data was used, the
8 cumulative low range of daily new cases in the period assessed is about 10,000, while the
9 high range is about 15,000. If the peak days are truncated, a weekly saving of ~20,000 new
10 cases in the 11 states studied is expected. Notably, the potential of truncating weekly peaks
11 should be weighed against the expected mental and medical costs of loneliness and
12 isolation as a consequent of greater adherence to social isolation.

13 Our hypothesis should be re-tested and refined based on more complete worldwide
14 databases that accurately report events' times, if and when these will become available.

15 We hope that our analysis will trigger local health authorities to ascertain or improve the
16 accuracy of the reported dates, and act accordingly to prevent unnecessary spread of
17 COVID19 during weekends.

18 **Methods**

19 **Statistics**

20 Analyses were performed separately for each country/US state for daily numbers of new
21 cases, hospitalizations, and for deaths. In order to account for global trends in each
22 country/state, a seven-day moving average (from 3 days before to 3 days after) was

1 calculated and the detrended data (i.e. the residuals between actual data and moving
2 average) used for further analyses. Next, autocorrelations were calculated (implemented in
3 SPSS, Version 25), using lags ranging between 1 and 16 days. An autocorrelation is the
4 correlation of a signal with a delayed copy of itself as a function of a given lag. The results
5 represent the similarity between observations as a function of the time lag between them.
6 As such, a significant autocorrelation indicates the presence of a recurring pattern such
7 that values in the series can be predicted based on its preceding values. For example, if an
8 autocorrelation with a lag of 7 days is positive and significant, it means that the values of
9 the series tend to repeat themselves every seven days. We hypothesized that
10 autocorrelations with a lag of 7 days will be significantly positive. Significance and
11 confidence intervals were calculated based on a simulated Bootstrap resampling analysis.
12 For this purpose, within each dataset, data-points were randomly selected with
13 replacement from the original pool and autocorrelations were calculated 10,000 times.
14 Confidence intervals were obtained by using the 95% central values of the simulated
15 autocorrelations. P-values were calculated by counting the number of simulated
16 autocorrelations that were more extreme than the original autocorrelations, and dividing
17 by 10,000. Finally, for the purpose of comparison of the detrended time series with a fixed
18 7-day periodic sinusoidal pattern, we fitted the following model $y(t) = b_1 * (\sin(2\pi t/c + 2\pi t/c + b_2)) + b_3$ where b_1 represents the amplitude, b_2 phase shift and b_3
19 an offset while c was held constant to match the desired period of 7 days. The fit was
20 computed with a simplex search method as part of the Matlab Optimization Toolbox
21 (Matworks Inc).

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2 **Author Contributions:**

3 **Itay Ricon-Becker and Shamgar Ben-Eliyahu:** literature search, figures, study design,
4 data collection, data interpretation, writing. **Ricardo Tarrasch and Pablo Blinder:**
5 literature search, figures, study design, data collection, data analysis, data interpretation,
6 and writing.

7 **Competing interests:** None

8 **Funding:** None

9

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5 [V37byOhETU6xOCYGVkHJPasWZQ_OGeLaOORls1AZPRhTaktWDo6fIU#citation](https://ourworldindata.org/coronavirus?fbclid=IwAR0N-V37byOhETU6xOCYGVkHJPasWZQ_OGeLaOORls1AZPRhTaktWDo6fIU#citation) (2020).
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1 **Table 1: A summary of all data collected**

Authenticated (Yes/No)	Location	No. of Cases	Sig. 7-day Pattern - Cases? (Yes/No)	Peak Days ¹	No. of Other events	Sig. 7-Day Pattern? (Yes/No)	Peak Days	Data-base
Yes (only Deaths)	Florida	NA	NA	NA	Deaths: 1,343	No (p=0.069)	Tu, Th-F	State Database
Yes	Maryland	NA	NA	NA	Deaths: 1,332	No	M, W, F	State Database
Yes	NYC	129,117	Yes*** *	M-Th	Deaths: 13,736	No	NA	City Database
					Hospitalization: 34,467	Yes****	M, Tu	
Yes	Ohio	99,732	Yes*** *	M-Tu, F	Deaths: 1,273	No	NA	State Database
	Ohio [0-69]	7,108	Yes*** *	M, W, F	NA			
	Ohio [70+]	2,853	Yes*** *	M-Tu	NA			
Yes	New Jersey	98,400	Yes*** *	M-W	NA	NA	NA	State Database
Yes	Massachusetts	64,829	Yes*** *	Th-Sat	Deaths: 4,444	No (p=0.069)	M, W-Th	State Database
					Average percent of positive tests: 23.5%	No	NA	
Yes	Michigan	32,289	Yes*** *	M-W, F	NA			State Database
Yes	Connecticut	27,496	Yes*** *	M-Th	Deaths: 2,600	No	NA	State Database
Yes	Georgia	25,067	Yes*** *	M-Tu, Th	Deaths: 1,274	No	NA	State Database

Yes	Virginia	21,294	Yes*** *	M, F	Deaths: 741	No	NA	State Database
					Hospitalization: 2,506	Yes****	M	
Yes	Indiana	19,456	No	NA	Deaths: 1,214	Yes*	Tu	State Database
Yes	Colorado	14,206	Yes*** *	M, F	Deaths: 1,024	No	NA	State Database
Yes	Washington	9,541	Yes*** *	M, Th-F	Deaths: 480	No	NA	State Database
Authenticated (Yes/No)	Location	No. of Cases	Pattern significant ? (Yes/No)	Peak Days ¹	No. of Other Events	Pattern significant? (Yes/No)	Peak Days	Data-base
No	US (All states)	1,088,047	Yes****	Th-Sat	Deaths: 67,167	Yes****	Tu-W	Worldometer
No	United Kingdoms	172,056	No	NA	Deaths: 27,285	Yes****	W, F-Sat	Our-World-in-Data
No	Spain	139,060	Yes****	W-Sat	Deaths: 20,570	No (p=0.09)	W-F	Our-World-in-Data
No	Italy	124,219	Yes***	Th-Sun	Deaths: 19,748	Yes***	W, F, Sun	Our-World-in-Data
No	Germany	114,593	Yes**	Th-Sat	Deaths: 6367	Yes***	W-F	Our-World-in-Data
No	Canada	54,799	No	NA	Deaths: 3,629	Yes****	F-Sun	Our-World-in-Data
No	Belgium	42,622	No	NA	Deaths: 7,555	No (p=0.067)	NA	Our-World-in-Data
No	Netherlands	31,968	Yes****	F-Sat, M	Deaths: 4,510	Yes**	W-F, Sun	Our-World-in-Data

No	Sweden	19,271	Yes*	Th-Sat	Deaths: 2,587	Yes****	W-F	Our-World-in-Data
No	Switzerland	17,718	Yes*	F, Sun	Deaths: 1,275	Yes*	W-Th, Sat-Sun	Our-World-in-Data
No	Austria	7,841	Yes *	W-Th, Sat	Deaths: 530	No	NA	Our-World-in-Data
No	Denmark	7,477	No	NA	Deaths: 432	Yes*	Sun	Our-World-in-Data

1

2 All databases above were accessed between May 20th to May 22nd, and all data (new cases, deaths,
3 hospitalizations, and %of positive tests) were obtained for March 29th to May 4th 2020.

4 ¹ Peak days were selected based on the days that showed most new cases numbers each week
5 (compared to a seven-day running average). Only days that ranked in the three first places three
6 times or more, of the five weeks, were selected. Days are presented in abbreviated form
7 (M=Monday, Tu=Tuesday, W= Wednesday, Th=Thursday, F=Friday, Sat=Saturday, and
8 Sun=Sunday)

9 * - p-Value is less than 0.05

10 ** - p-Value of less than 0.005

11 *** - p-values is less than 0.0005

12 **** - p-Value of less than 0.00005

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14

1 **Figures:**

2 **Fig. 1: US data representative outcomes of the entire US (not date-authenticated,**
 3 **based on the Worldometer database) and two representative date-authenticated**
 4 **states (Massachusetts and Michigan)**

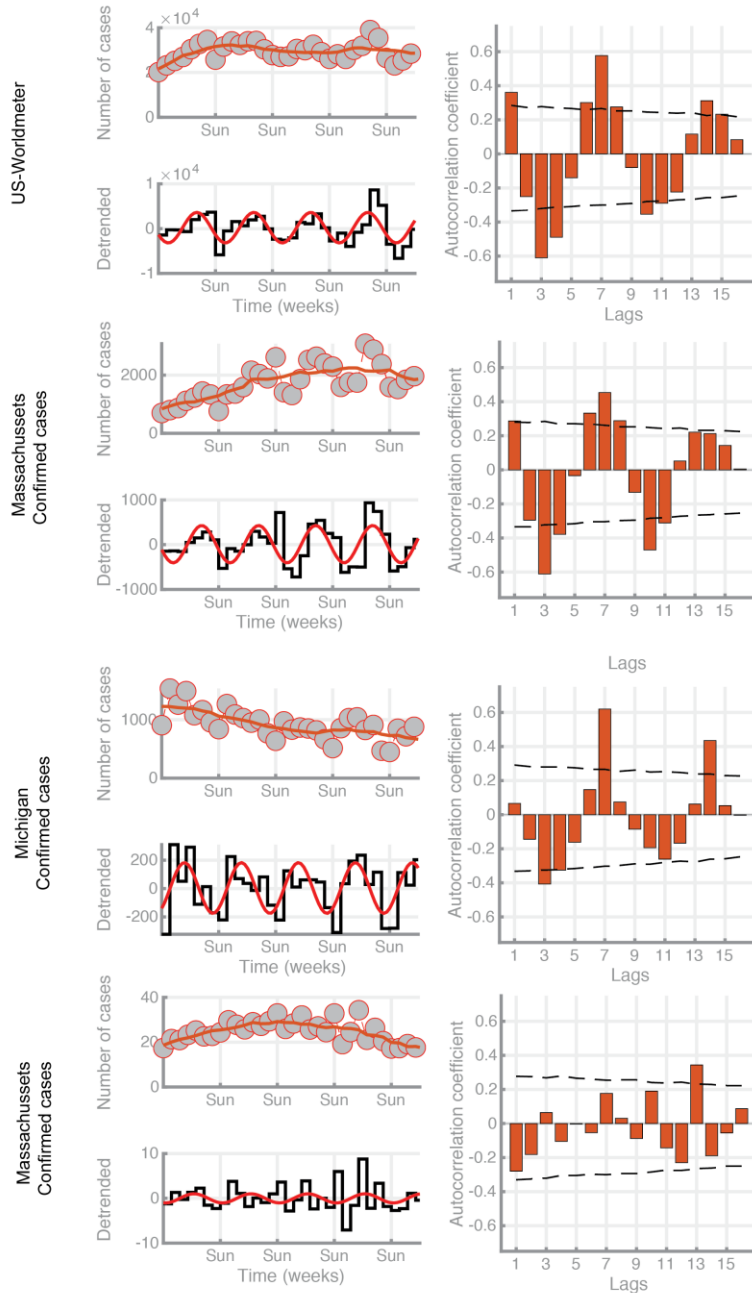
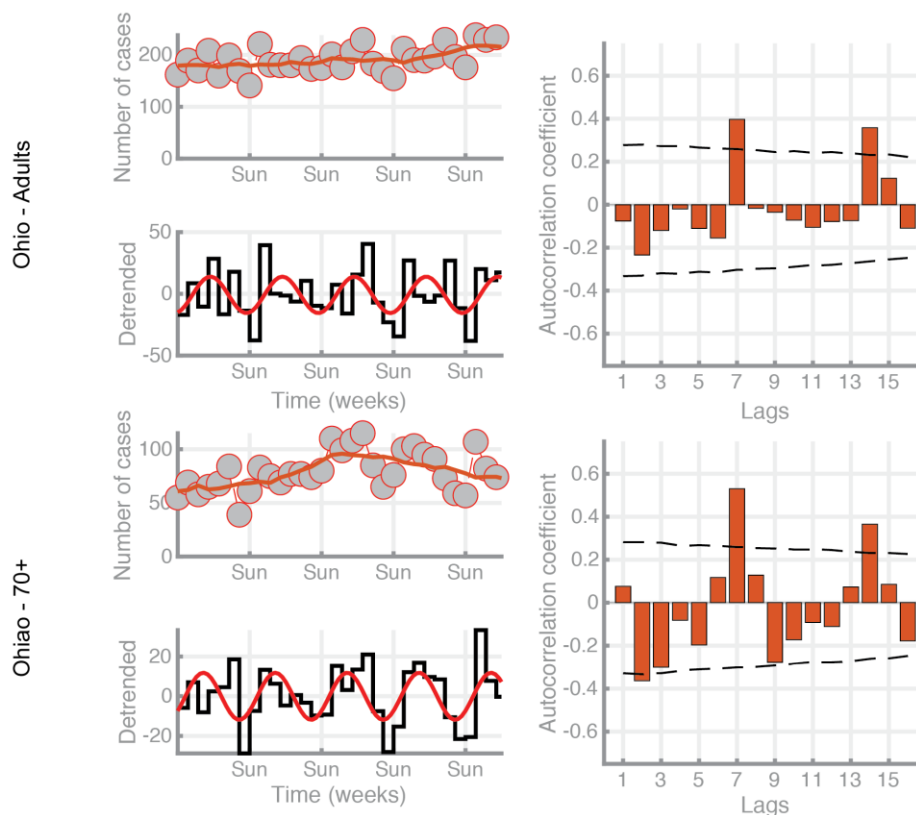


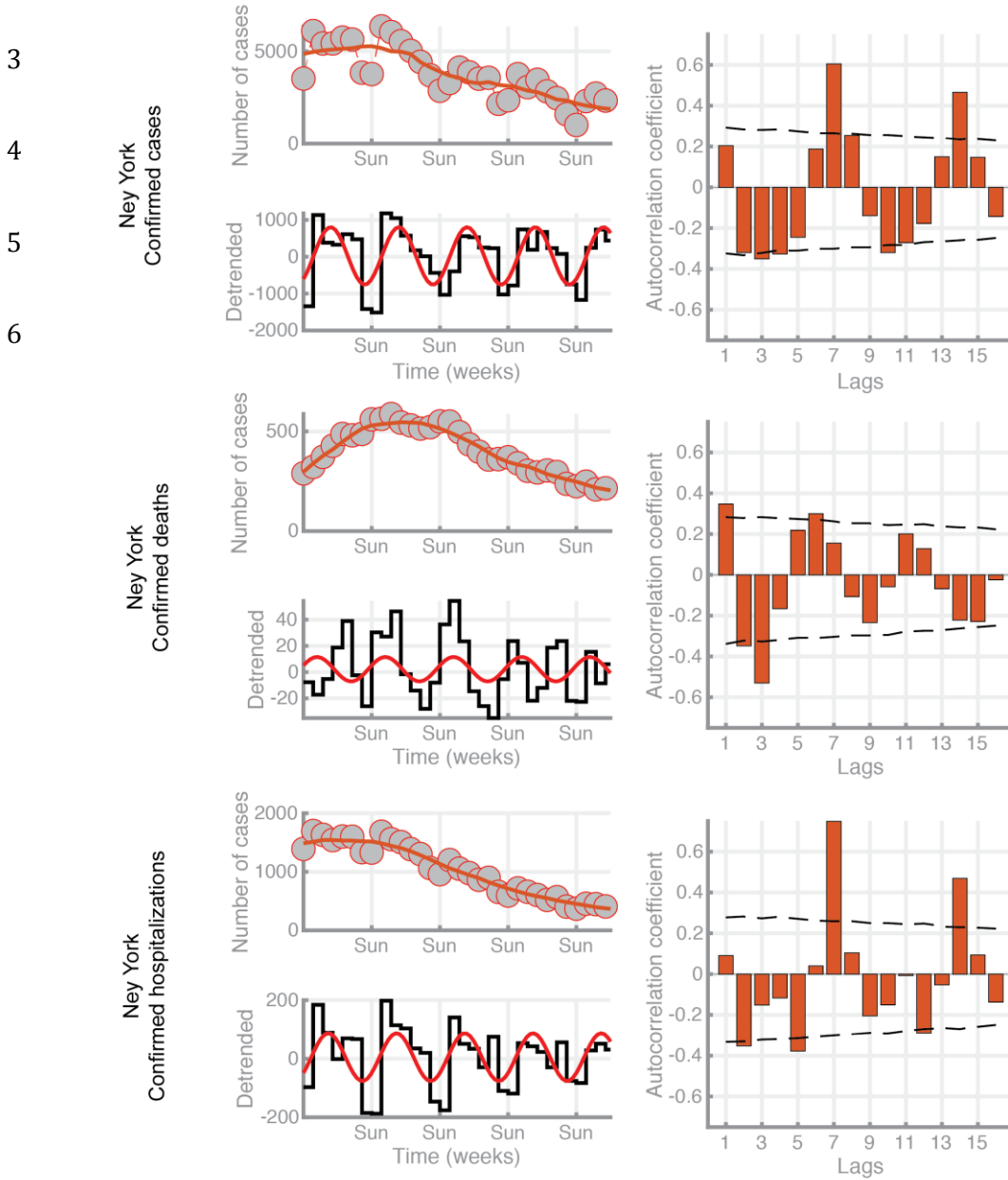
Fig. 1: The upper left pane depicts raw data (gray points) with averaged data over-imposed (red line). The lower left pane presents the detrended data (difference between the raw data and a 7-day moving average - 3 days before to 3 days after each data point, black line) with a fixed 7-day periodic sinusoidal superimposed (in red). The right pane presents autocorrelations for lags ranging between 1 and 16 days, with 95% confidence intervals (calculated based on bootstrap resampling analysis). Significant autocorrelations fall outside the 95% interval

1 **Fig. 2: Ohio New Cases by age graphs: A. Age below 70, B. Age over 70 (Date-**
2 **authenticated. Date is defined as illness onset or if onset date in unknown, the**
3 **earliest known date associated with the case).** For explanation of depicted data, see Fig.
4 1 legend.

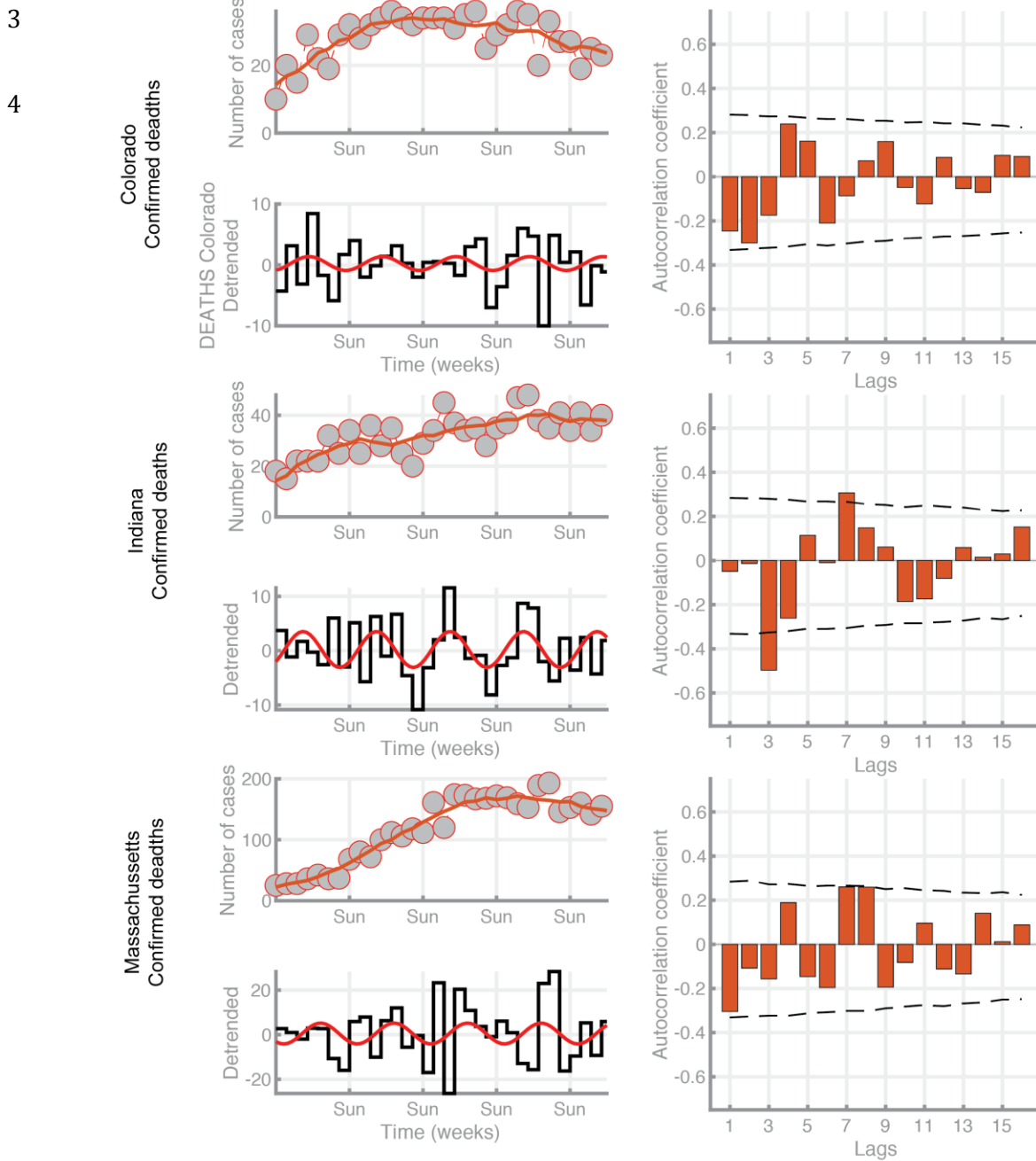


1 **Fig. 3: NYC: New Cases, Hospitalization, and Death Data (Date-authenticated).** For

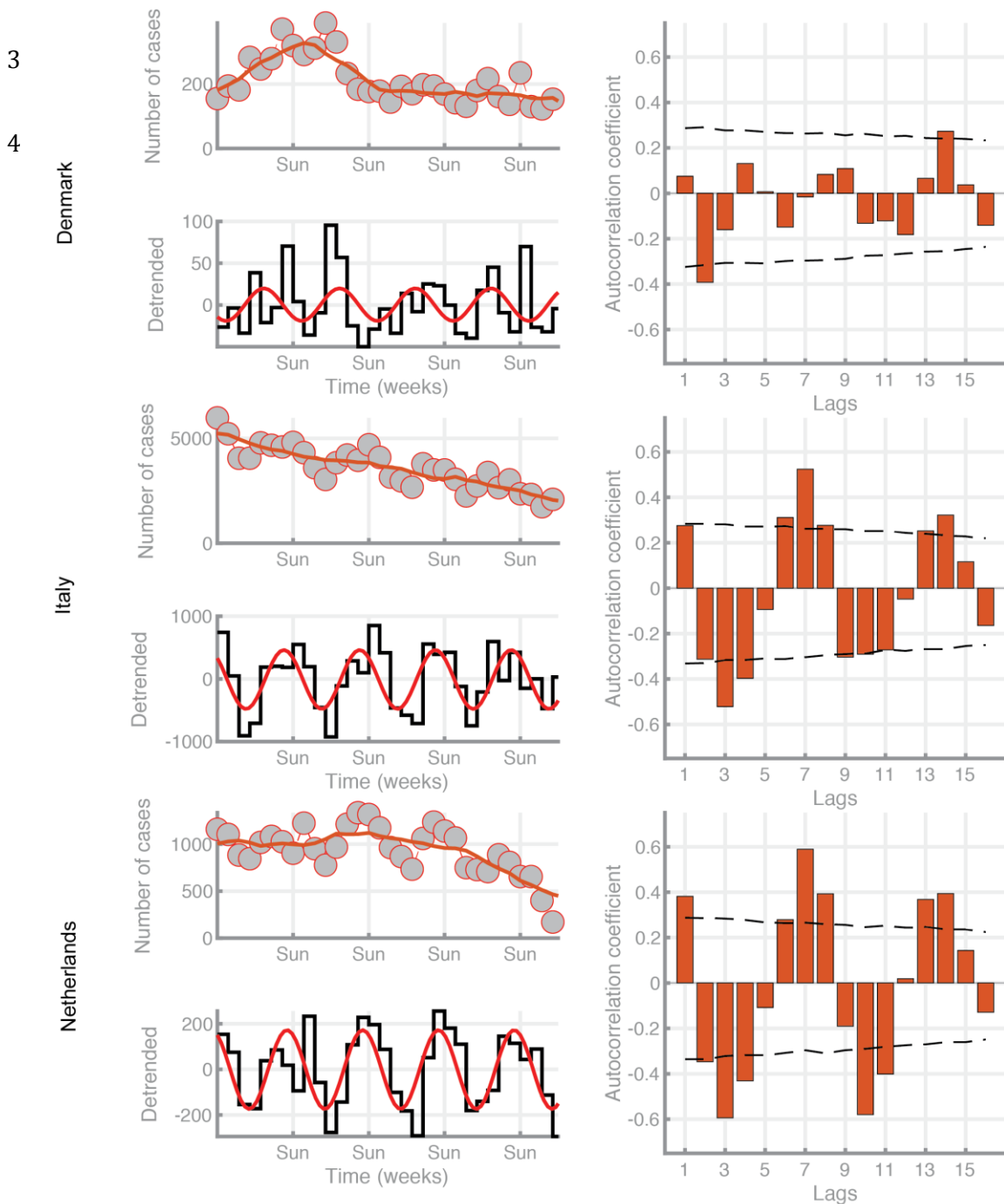
2 explanation of depicted data, see Fig. 1 legend.



1 **Fig. 4: COVID-19 Deaths in the US: representative date-authenticated states : Indiana,**
2 **Colorado, and Massachusetts.** For explanation of depicted data, see Fig. 1 legend.

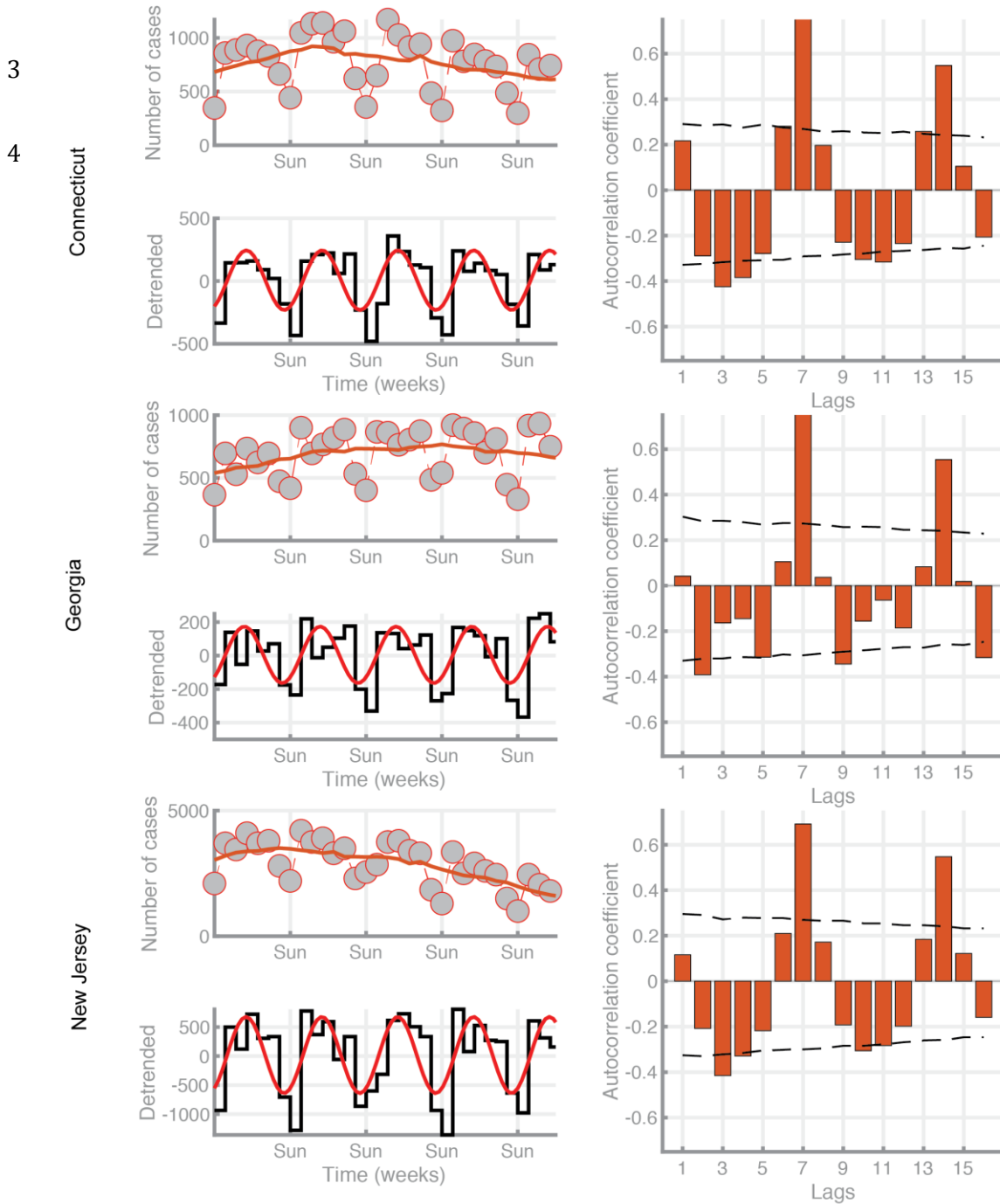


1 **Supp Fig 1: New Cases: Representative not date-authenticated countries data: Italy,**
2 **Netherlands and Denmark.** For explanation of depicted data, see Fig. 1 legend.



1 **Supp Fig 2: New Cases in representative date-authenticated states data: Connecticut.**

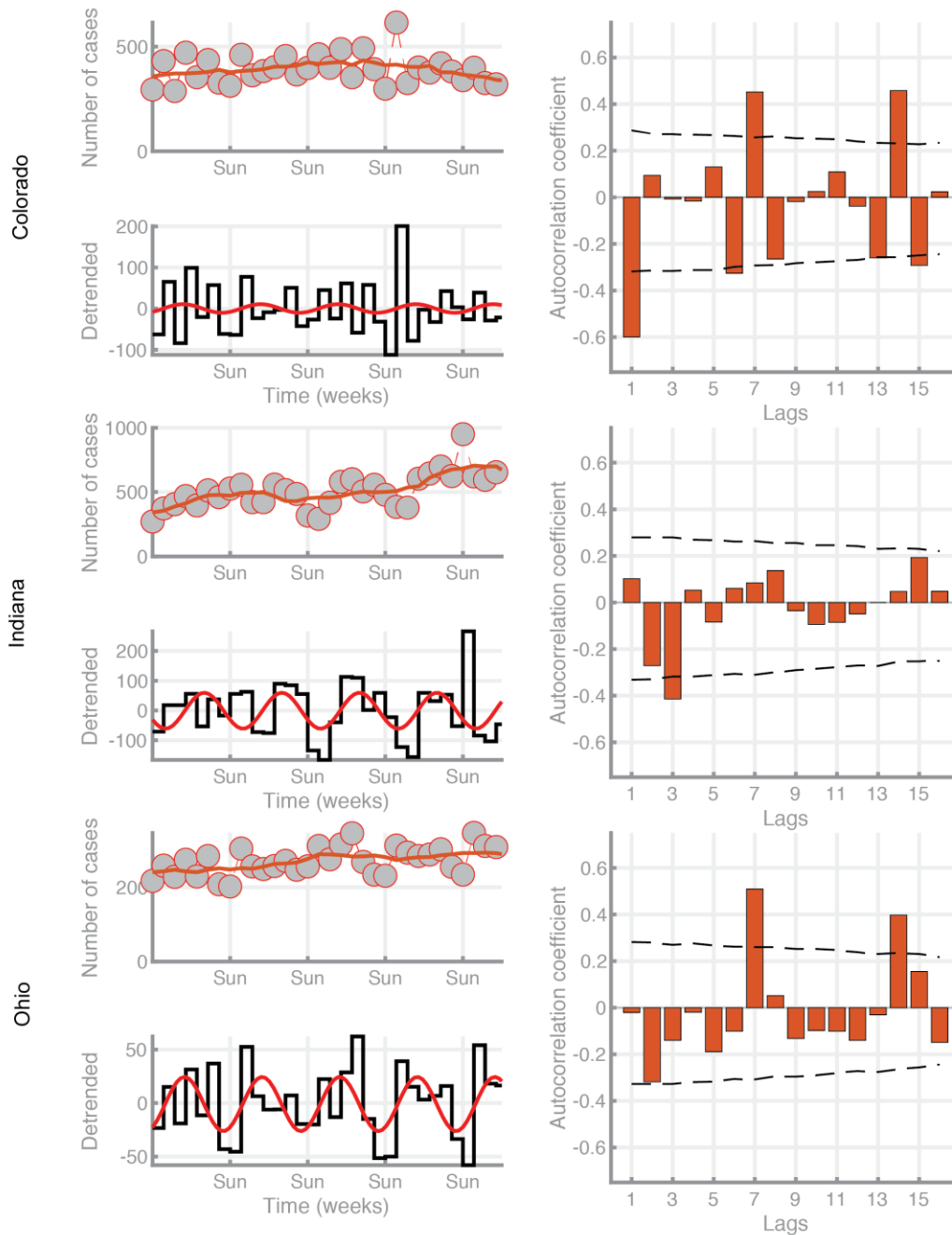
2 **Georgia, New Jersey.** For explanation of depicted data, see Fig. 1 legend.



1 **Supp Fig 3: New Cases in representative date-authenticated states data: Ohio,**
2 **Colorado, Indiana.** For explanation of depicted data, see Fig. 1 legend.

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1 **Supp Fig 4: New Cases in representative date-authenticated states data: Virginia,**
2 **Washington.** For explanation of depicted data, see Fig. 1 legend.

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