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Research Article

Covid-19 High Attack Rate Can Lead to High Case Fatality Rate - 3

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ABSTRACT

Background: We conduct this study to identify the relationships among the Attack Rate (AR), the Mortality Rate (MR), and the Case Fatality Rate (CFR).

Material and Methods: Thirty countries and territories with > 500 Covid-19 reported cases per 10,000 population inhabitants were selected on March 10, 2021. One-sample Kolmogorov-Smirnov (K-S), one-way ANOVA, Levene, Least Significant Different (LSD), and matched paired-samples T-tests were used in this study.

Results: There was a highly significant difference concerning AR and CFR mean values. Countries with MR \ge 15 death / 10⁴ inhabitants recorded the highest level of crude mean CFR and AR values, and recorded the highest gap with leftover groups. There were independence 95% confidence intervals of mean CFR and AR values between countries with \ge 15 death / 10⁴ MR and those with MR of <10 death /10⁴. There was a significant difference between countries with MR \ge 15 death / 10⁴ inhabitants and countries with MR of <10 death / 10⁴ inhabitants through least significant difference (LSD) test for CFR (0.042 p- value) and Games Howell (GH) test for AR/10⁴ (0.000 p-value).

Conclusions: Total and mean AR and CFR are higher in high MR countries and vice versa.

Keywords: Case fatality rate; Mortality rate; Attack rate; Covid-19

INTRODUCTION

During the coronavirus disease 2019 (COVID-19) pandemic, mortality rates were heterogeneous, with some countries being hit very hard, others had a much lower death rate [1]. Many factors have been proposed as determining this heterogeneity, including demographics & social factors, comorbidities, and environmental factors (such as temperature, humidity, and air pollution the age, smoking habits, hosting of big public events, socializing habits, or the capacity of the health care system [2].

Of the many driving factors of the strong transmissibility, cluster infections play critical roles in the widespread of disease and exponentially increases the number of cases [3]. It is thought that super-spreaders of COVID-19 play a role in transmission within these clusters since they are partially contributing to the high transmission risk of SARS-CoV-2 [3].

High fatality rates were also reported within these SARS-CoV-2 cluster infections [4].

The Case Fatality Rate (CFR) sometimes called case fatality risk is used to define the probability that a case dies from the infection [5].

At country levels, CFR was seen to be different in different places and seen to be not a constant finding. Furthermore, it can decrease or increase over time [6].

Studies have reported multiple factors for variances in COVID-19 CFR among countries, these include demographics & social factors, comorbidities, and environmental factors (such as temperature, humidity, and air pollution) [2]. Although high AR was studied in small clusters and small locations its relation to MR or CFR was not identified. High mortalities during waves' peaks were attributed to system failure to cope with the increase in disease burden.

However, although the number of cases has been extensively discussed, the number of cases/population at risk i.e. AR, and its relation to MR and CFR with data at the population (country) level have not been verified.

The current paper aimed to identify the statistical correlations between AR and CFR with MR because of the COVID-19 outbreak and its difference among countries, with data at the population (country) level. From a global health perspective, there is evidence of a research knowledge gap in this field aspect.

MATERIAL AND METHODS

Materials

Thirty countries and territories were chosen. Inclusion criterion was > 500 Covid-19 reported cases per 10,000 population inhabitants. Data on covid-19 cases and deaths was selected as it was on March 10, 2021. Countries and territories were classified into three group groups: group I: countries with mortality rate \geq 15 death/10⁴ population inhabitants; and group III <10 death/10⁴ population inhabitants.

A supplementary document attached file contains original data, computed data, and references for data sources.

Definitions

The detected AR for a given country was calculated as the total number of reported cases divided by the estimated population of that country.

Crude COVID-19 CFR was calculated as the total number of COVID-19 deaths divided by the number of total COVID-19 confirmed cases by March 10, 2021 multiplied by 100.

METHODS

The following statistical data analysis approaches were used under the application of the statistical package (SPSS) *ver*. (22.0):

Descriptive data analysis: Mean value, standard deviation, standard error, (95%) confidence interval, and graphical presentation by using Bar Charts. Mean values and the two extremes values (min. and max.) were computed assuming that data followed normal distribution function.

Inferential data analysis: These were used to accept or reject the statistical hypotheses, which included the following:

- a- The One-Sample Kolmogorov-Smirnov (K-S) test. This is a goodness-of-fit test whether the observations could reasonably have come from the specified distribution.
- b- The One-Way ANOVA procedure to test the hypothesis that several means are equal. In addition to that, we applied

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after rejecting the statistical hypotheses, the Least Significant Difference (LSD) test requiring equal variances was assumed, and the Games Howell test not requiring equal variances was assumed.

- c- Levene test: was used to test homogeneity of variances for equality of variances of two and several independent groups.
- d- Matched paired-samples T-Test procedure was used to compare the means of two variables for a single group. It computes the differences between values of the two variables for each case and tests whether the average differs from zero.

RESULTS

Table 1 represents a one-sample "Kolmogorov-Smirnov" test procedure comparing the observed cumulative distribution function for studied data with a specified theoretical distribution, which proposed normal shape (i.e. bell shape), for the studied markers.

The results showed that the test's distribution was normal for the studied reading's markers since no significant levels were accounted for (*p*-value >0.05). This enabled us of applying conventional two methods of statistics: the descriptive methods of estimations (points and intervals), and the inferential statistics.

In the table 2 results showed group I countries with higher total attack rate and higher total CFR % than group III countries. Group II countries showed the lowest test coverage among the three groups. The CFR % marker recorded a high level of mean value in group I countries with a high gap concerning leftover groups, especially to group III, which accounted for the lowest level among all markers. In addition to that, first and third groups recorded an independent or non-interferer of 95% confidence interval for mean values for each other. Furthermore, group III recorded an independent (non-interferer) of 95% confidence interval for mean values concerning other groups.

Regarding the (AR/10⁴) marker, results showed that group I recorded a high level of mean value, and a high gap concerning leftover groups, especially to group III, which accounted for the lowest level for preceding markers. In addition to that, first and third groups recorded independent or non-interferer of 95% confidence interval for mean values.

 Table 1: Normal distribution function test (Goodness of fit test) for studied markers.

One-Sample Kolmogorov-Smirnov Test								
Markara	Statiation	Groups (**)						
Widikers	Statistics	I	II	Ш				
	No.	12	12	16				
	Kolmogorov-Smirnov Z	0.589	0.481	0.509				
GFR %	Asymptotic Sig. (2-tailed)	0.878	0.975	0.958				
	C.S. (*)	NS	NS	NS				
	No.	12	12	16				
104	Kolmogorov-Smirnov Z	0.591	0.848	0.739				
AR X	Asymptotic Sig. (2-tailed)	0.876	0.468	0.646				
	C.S. (*)	NS	NS	NS				
Statistical Hypothesis: Ho: Markers are followed normal distribution function								
Test distribution is Normal.								

^(*) NS: Non Sig. at *p* > 0.05.

^(*') Country mortality rate groups: group I: \geq 15 death/10⁴ population inhabitants; group II: \geq 10-15 death/10⁴ population inhabitants; and group III <10 death/10⁴ population inhabitants.

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CFR: Case Fatality Rate; AR: Attack Rate.

Concerning testing the compound statistical hypothesis, which says that studied group's concerning (CFR%, and AR/10⁴) readings were thrown from the same population, and that should be proved according to of testing equal variances were assumed, as well as equal mean values were assumed through "Levene and one-way ANOVA" tests respectively, and as illustrated in the table 3.

Concerning testing equal variances of CFR% marker, the Levene test showed that no significant differences were accounted at p > 0.05 among studied groups and a highly significant result (0.006 *p*-value) concerning AR. ANOVA test showed a significant difference at p < 0.01 among all studied groups concerning AR and CFR mean values.

The alternative statistical hypothesis says that at least two groups are not equal due to their mean values. This was tested through the LSD test for CFR% marker, and Games Howell (GH) test for AR/10⁴ marker (Table 4).

Results in the table 4 showed that no significant difference between groups I, and II regarding both studied markers, and no significant difference between groups II and III regarding the AR marker. There were significant differences among the leftover comparisons in at least at p < 0.05 for each of the studied markers. There was a significant difference between the I and III groups in both two tests.

DISCUSSION

Main findings of this study

The results of the current study suggest that the high total and mean attack rates were found in high mortality rates countries. CFRs were also higher in these countries (Table 2 and Figure 1).

It was thought that CFR is used as a measure of disease severity and is often used for predicting disease course or outcome rather than (as in our study) that CFR reflects the level of disease burden among the community.

What is already known on this topic

Commentators may consider CFR as if it's a steady and unchanging number confined to a specific disease in general and Covid-19 in particular.

Studies have reported multiple factors for variances in COVID-19
CFR among countries, these include demographics & social factors,
comorbidities, and environmental factors as early mentioned [2].

Previous literature usually attributes an increase in CFR to low estimates due to low testing. Another cause of high CFR in certain places is usually attributed to unknown or yet not yet proved causes. There is evidence suggested by recent literature that CFR is positively related to MR [7,8].

What this study adds

Our significant findings of a highly significant association between AR and CFR is supported by the following previous observations: (1) available data about South Americans and Asian countries that took the strictest measures, and they also had relatively lower COVID-19 CFR [9,10], (2) it has been noticed that in European countries that have had both large numbers of cases and deaths , the average of country-specific CFR was at 0.7%-1.3% at early times of pandemic raised to about 2%-3.31% thereafter [9], this increase in Covid-19

Table 2: Sumn	nary statistics	concern	ing studied r	narkers an	d total tests	/million am	nong different	groups of co	untries.	

Markara	Group	No	Total	Mean	Std D	Std E	95% C.I. for N		Min	Max		Total tooto/million
Warkers	Group r	NO.	TOLAT	Weall	Stu. D.	JIU. E.	L.b.	U.b.	IVIIII.	IVIAA.	TOLATIVIK	Total tests/minon
	I	12	2.082	2.271	0.598	0.173	1.891	2.651	1.351	3.261	16.377	1,069,856.523
CFR %	11	12	2.296	1.880	0.458	0.132	1.589	2.171	1.012	2.465	12.762	335,599.494
	III	16	1.024	0.872	0.464	0.116	0.624	1.119	0.157	1.827	6.763	658,743
	I	12	786.473	881.7	294.9	85.1	694.3	1069.0	510.2	1286.6	16.377	1,069,856.523
AR X 10⁴	11	12	555.616	734.1	242.5	70.0	580.0	888.2	517.5	1430.9	12.762	335,599.494
	III	16	659.951	637.7	105.3	26.3	581.6	693.8	513.3	923.3	6.763	658,743

Country mortality rate groups: group I: ≥ 15 death/10⁴ population inhabitants; group II: ≥10-15 death/10⁴ population inhabitants; and group III <10 death /10⁴ population inhabitants.

CFR: Case Fatality Rate; AR: Attack Rate; MR: Mortality Rate.

Table 3: Testing equal variances and equal mean values for studied markers concerning different groups of countries classified according to different mortality rates /10⁴ population inhabitants.

	Markor	Testing Homo Variano	geneity of ces	ANOVA- Testing Equality of Means		
	Marker	Levene Statistic	Sig. (*)	F-test	Sig. (*)	
	CFR%	0.888	0.420 (NS)	28.936	0.000 (HS)	
	AR/10⁴	5.968	0.006 (HS)	4.273	0.021 (S)	
	(*)					

^(*) HS: Highly Sig. at *p* < 0.01; S: Sig. at *p* < 0.05; NS: Non Sig. at *p* > 0.05. CFR: Case Fatality Rate; AR: Attack Rate

Table 4: Multiple Comparisons using (LSD) and (GH) tests for studied markers among studied groups of countries.

Marker	(I) Group	(J) Group	Mean Diff. (I-J)	Sig.	C.S. (*)
	CFR% 00	0.391	0.066	NS	
CFR%		III	1.399	0.000	HS
	Ш	111	1.008	0.000	HS
	I	II	147.6	0.390	NS
AR/10⁴		III	244.0	0.042	S
	Ш	111	96.4	0.424	NS
(*) HS: Highly Sig	1 at n < 0.0	1 S Sig at $n < 0$	05. Non Sic	1 at n > 0.0	5 [.] Testina

based on GH test.



Figure 1: Represent graphically plotting of bar chart regarding mean values studied marker's readings distributed in different groups of countries regarding mortality rates.

Country mortality rate groups: group I: ≥ 15 death /104 population inhabitants; group II: ≥10-15 death / 104 population inhabitants; and group III <10 death /104 population inhabitants

CFR: Case Fatality Rate; AR: Attack Rate

CFR possibly goes in parallel with an increase in the number of cases, (3) in general, the CFR of COVID-19 differs by location, and has changed during the period of the outbreak [2,4], (4) small clusters of fatal COVID-19 infections were reported previously with high CFR within these clusters (families, tourists, long-term care hospitals and facilities, etc.) [12-16]. There were also identified "vulnerable" clusters of counties in USA with high mortality incidence ratio [17]. In UK A few areas saw COVID-19 mortality more than seven times the

expected level compared with the rest of the country [18], (5) there was a positive association between population size and COVID-19 CFR [6], and (6) our findings were in concordance with recent studies which founded very high positive significant correlation between total deaths/1 million and the total number of cases / million inhabitants and a very high positive influence of the COVID-19 MR on the CFR [7,8]. Another study suggested that the crucial factor for the different death rates because of the COVID-19 outbreak is the fast implementation of public events ban [1].

We suggest that AR plays an important role in explaining variances.

The underlying cause for increased CFR with increased AR is possibly related to high viral overload as it was observed in clustering infections a phenomenon already described before that is characterized by high mortality and fatality rates.

During a disease outbreak, estimation of the (CFR) is usually used as an indication of its severity, and is used for planning and determining the intensity of a response to an outbreak as a guide to plan public health strategies [5,19,20]. As CFR value reflects the density of infection, we suggest herein, to study CFR in the context of AR rather than the CFR alone.

As far as the epidemiology concerns with the virulence of the disease in addition to the transmissibility of infectious disease [21], our finding of the significant role of AR will add an important factor explaining various virulence of epidemics in different places and times.

This study will help in the development of prevention and intervention measures to fight against this global public health crisis.

Limitations of this study

The CFR has been consistently being subjected to underestimation and overestimation [22].

Overestimate of CFR is largely due to underestimation [22] of cases especially encountered in an infection with a range of manifestations from relatively mild to severe [5].

Testing capacity, which is associated with the availability of resources and manpower, is the single most important factor that can tremendously affect the CFR [23]. The lack of availability of widespread testing leads to an ascertainment bias toward severe cases [23].

Underestimation of death accounts can lead to erroneously low CFR. This could be due to: (1) in crude estimates, some patients

encountered as non-dead are still hosted in intensive care units [2], (2) deaths caused by COVID-19 may be misattributed to other death classifications and codes [11], and (3)deaths confirmed counts are subject to time lags [4]. This means that reported cases with COVID-19 will die at a later date [5].

CONCLUSIONS

During the Covid-19 pandemic, the CFR is related to AR and MR values and measures.

Concepts regarding the severity of the disease should be directed to the ability to have high AR rather than to high CFR since CFR can change according to AR.

A possible important cause of variances CFR across the world seems to be due to be previously underestimated factor that is AR.

RECOMMENDATIONS

Increased AR is a very highly significant associated possible predictor for increased MR and CFR. Measures focused on the reduction of AR according to this study will certainly reduce MR and CFR.

Ethical approval was not required for this study, as we used publically available data, and patients were not involved.

(Appendix)

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