Delay of Detection Of COVID-19 Patients In Bangladesh; An Application To Cox Proportional Hazard Model

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Keywords: COVID-19, survival analysis, delay of detection

Posted Date: December 11th, 2020

DOI: <https://doi.org/10.21203/rs.3.rs-117305/v2>

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Abstract

Objective: To delineate the survival rate of the patients with coronavirus disease 2019 (COVID-19) who did the diagnostic tests lately after the development of symptoms. The aim is to determine the socio-demographic risk factors associated with the delay of the detection of COVID-19 patients.

Methods: For this cross-sectional study, 300 patients were selected who were diagnosed as COVID-19 patients in the Molecular Biology Laboratory of Chittagong Medical College, Chattogram, Bangladesh. Data were collected from May to July 2020. Clinical characteristics were obtained from over phone interviews and laboratory diagnosis by Real-time Reverse Transcriptase Polymerase Chain Reaction (rRT-PCR). Cox proportional hazard model is applied to estimate risk factors affecting the delay of detection of COVID-19 patients.

Result: Female mortality rate was 44.9% higher compared to males, graduates died 32% more than undergraduates, unmarried peoples' death rate were 56% more than married and those who were in traveling irregularly and in contact with symptomatic patients, were 86% more died than non-travelers.

Conclusion: Early diagnosis of COVID-19 can save a huge amount of lives and special attention should be emphasized on the significant explanatory variable.

Introduction

The Severe Acute Respiratory Syndrome (SARS-CoV) and the Middle East respiratory syndrome (MERS-CoV) strengthened as pathogens in the last two decades, both on humans and animals [1,2]. In December 2019, the outbreak of an emerging disease began in Wuhan, China, and spread rapidly in China and abroad. They caused over 50% mortality by respiratory tract infection, including extrapulmonary complications. SARS-CoV-2 was first identified last December by the "Chinese Center for Disease Control and Prevention" [3,4].

Current studies show that SARS-CoV-2 can transmit person-to-person through respiratory droplets in closed containment [5,6]. The World Health Organization (WHO) reported it as a health crisis on 30th January 2020 and titled the disease caused by this virus Coronavirus Disease -19 (COVID-19) on 11th February 2020. As of 31 July 2020, the number of patients has outgo 2, 54,474 globally, and more than 6002 have now expired of COVID-19.

It depicted SARS-CoV-2's genome sequence identical to SARS-CoV, which is 79.5%. Interestingly, SARS-CoV-2 was 96% analogous to a bat coronavirus [7] in the whole-genome sequencing.

Both from clinical characteristics and pathological dimensions fatal pneumonia associated with intensive care with high mortality [8]. To keep going with this economic breakdown, the members of maximum families went outside of their homes to earn money. Regular household necessities bought by them too, from markets in this pandemic situation. Besides, a so long a period of home restrictions made

them unsteady to breathe fresh air, as a result, affected by COVID-19. But initially, symptoms were mild, and especially the young generation was unwilling to maintain healthy hygiene and to follow restrictions. So, in our study, we tried to identify the hazard of late detection time of COVID-19 of those patients.

Materials And Methods

A cross-sectional study conducted from May 2020 to July 2020 in the molecular biology laboratory of Chittagong Medical College, Chattogram with 300 patients who are COVID-19 positive. Reverse Transcriptase-Polymerase Chain Reaction was performed by the QuantStudio 5 Real-Time PCR instrument (96-Well 0.2 ml Block) by Sansure Biotech Novel Coronavirus (2019-nCoV) Nucleic Acid Diagnostic Kit (PCR-Fluorescence Probing).

Patients with a SARS-CoV-2 infection confirmed by an rRT-PCR of a nasopharyngeal and oropharyngeal combined sample within 3 days of onset of symptoms included in the study. Presence of symptoms, days of illness, contact with symptomatic patient /health personnel, types of PPE used, traveling history, supplemental oxygen therapy requirement, hospital admission history, ICU support requirement, co-morbid conditions, baseline investigations, and prescription collected by the research team through interviewing over the phone.

The WHO website has provided several rRT-PCR protocols for detecting SARS-CoV-2 in different countries [9]. Here an automated system repeated the amplification process for about 45 cycles until we could detect the viral cDNA, usually by a fluorescent signal from different channels such as FAM, ROX, and CY5. We considered the SARS-CoV-2 open reading frame-ORF 1ab, and the specific conserved sequence of coding nucleocapsid protein N gene as the target regions for the conserved sequence of the double-target genes to achieve detection of sample RNA through FAM and ROX signal changes, respectively. Internal standard gene fragments (Rnase P) were detected by the CY5 signal. Here a single tube contained the specific primers, probes, dNTPs, MgCl₂, Rnasin, PCR buffer, RT enzyme, and Taq enzyme to run the entire RT-PCR reaction (24 tests/kit). We fixed positive internal control to monitor PCR inhibitors in test specimens and normal saline as a negative control. After getting the rRT-PCR results, we sorted 300 SARS-CoV-2 positive patients for our study purpose; from those we got proper history and relevant data completely.

Cox proportional hazard model (1972) [10] is a well-admired model used in survival analysis that can assess the importance of various covariates in the survival time of individuals or objects through the hazard function. It makes us capable of estimating the relationship between the hazard rate and explanatory variables without having to make assumptions on the shape of the baseline hazard function. In our study, the cox proportion hazard model was used to analyze the factor affecting COVID-19 patients' mortality because of late diagnosis by considering any extra heterogeneity present in the data. To test the equality of the hazard function, we used a log-rank test (does not make any assumptions regarding the distribution of the data set) to test the null hypothesis: all the survival curves were the same. To explore the relationship among factors having multiple-categories, we applied the Bonferroni test.

Results

This study described 300 patients with confirmed COVID-19 from May 09, 2020, to July 12, 2020, in Molecular Biology Laboratory, Chittagong Medical College, Chattogram.

Table 1: Distribution of survival status along with the Log Rank test with the selected variables

Backgroud Characteristics	Catagories	Censored	Condition of patients (Died)	Log Rank (Mantel-Cox)	df	p-value
Sex	Female	87(29.0%)	3(1.0%)	7.675	1	0.006*
	Male	183(61.0%)	27(9.0%)			
Education	Undergraduate	96(32.0%)	15(5.0%)	1.632	1	0.201
	Graduate	174(58.0%)	15(5.0%)			
Occupation	Health personnel	16(5.3%)	2(0.7%)	5.764	3	0.124
	Service holder	49(16.3%)	2(0.7%)			
	Student	57(19.0%)	3(1.0%)			
	Others	148(49.3%)	23(7.7%)			
Place of Residance	Urban	195(65.0%)	9(3.0%)	5.086	1	0.024*
	Rural	75(25.0%)	21(7.0%)			
Marital Status	Unmarried	60(20.0%)	0(0.0%)	1.025	1	0.311
	Married	210(70.0%)	30(10.0%)			
Age	0-18	20(6.7%)	1(0.3%)	1.221	3	0.748
	19-39	90(30.0%)	3(1%)			
	40-59	94(31.3%)	14(4.7%)			
	60 and above	66(22.0%)	12(4.0%)			
Family member's affected or not	Not affected	135(45.0%)	21(7.0%)	11.688	1	0.001*
	Affected	135(45.0%)	9(3.0%)			
Travelling History	Contact with symptomatic patient	114(42.7%)	3(1.1%)	17.788	2	0.000*
	Travelling regularly	90(33.7%)	9(3.4%)			

	No travelling history	36(13.5%)	15(5.6%)			
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(*) Indicates variable is the significance level at % level of significance

In the above table, censored were those who were alive during the study period, whether he/she was in isolation at home or the hospital. Total 1% female and 9% male were not alive at $p=0.006$ level of significance. 0.7% of health professionals and service holders died. Only 1% of students died in the study period. In contrast, the people related to the 'others' profession 7.7% died, which is much more.

Rural patients died 7%, where urban people were only 3% ($p = 0.024$). The death of married persons was 10%. Middle-aged adults had a 4.7% death rate, but only 0.3% of children died. Besides, 7% of death was from those group of patients whose family members unaffected and it was significantly higher ($p = 0.001$) than those who had affected family members. Surprisingly, 5.6% of patients died ($p = 0.000$) those who had no traveling history.

Table 2: Estimated Hazard Ratio obtained from Cox proportional hazard model

Variables	Categories	B	SE	df	Sig.	Exp(B)	95.0% CI for Exp(B)	
							Lower	Upper
Sex	Female	.371	.142	1	.009	1.449	1.097	1.915
	Male					1.000		
Education	Undergraduate			1		1.000		
	Graduate	.275	.140		.050	1.316	1.000	1.732
Occupation	Health personnel	.022	.264	3	.933	1.022	.610	1.714
	Service holder	.101	.165		.543	1.106	.800	1.529
	Student	.303	.157		.054	1.353	.995	1.840
	Others					1.000		
Place of Residance	Urban	.249	.137	1	.069	1.283	.981	1.678
	Rural					1.000		
Marital Status	Unmarried	.444	.239	1	.064	1.559	.975	2.492
	Married					1.000		
Age	0-18			3		1.000		
	19-39	.126	.260		.629	1.134	.681	1.890
	40-59	-.012	.162		.005	.988	.719	1.358
	60 and above	-.084	.162		.267	.920	.670	1.263
Family member's affected or not	Not affected	-.336	.122	1	.006	.715	.563	.908
	Affected					1.000		
Travelling History	Contact with symptomatic patient	.621	.192	2	.001	1.860	1.277	2.710
	Travelling regularly	.591	.199		.003	1.807	1.223	2.669
	No travelling history					1.000		

For testing the proportional hazard assumption, we checked whether the independent variables meet the proportional assumption by using a log-minus-log (LML) plot. The LML plot is a graph constructed by applying the log-log transformation of the survival function. It is a graph of the logarithm of time against the logarithm of the negative logarithm of the estimated survival function. In Figure 1 we have seen that curves of each risk factor are not crossing but they are parallel. That means all the risk factors defined properly and satisfied the proportional hazard (PH) assumptions.

From Table 2, female mortality (hazard) rate was 44.9% higher [(p = 0.009) CI (9%, 91%)] compared to the male. The factor education was likely to have in more hazardous conditions; graduate entities were likely to 1.316 times higher [CI (1.000, 1.732)] risk compared to undergraduate. Students were 1.353 times more [CI (.995, 1.840)] in hazard than 'others' category.

The residents of urban areas were in 1.283 times [CI (.981, 1.678)] high-risk zone than the rural residents. Unmarried were 1.559 times [CI (.975, 2.492)] riskier for death than married persons. 19-39 year's age group patients were 1.134 times [CI (0.681, 1.890)] more in hazardous context than 0-18 year's age group. The unaffected family members' patients were 0.715 times [CI (.563, .908)] less in risk than the affected family members' patients. Besides, the patients who were in contact with symptomatic patients were 86% more in hazard [(p = 0.001) CI (28%, 71%)] than those who had no traveling history during this pandemic.

Discussion

Delay of detection of COVID-19 plays a significant role in this survival analysis regarding the patient's status: he/she survived or not. By the cox proportional hazard model, we explored the hazard rate of several risk factors and explained the odds ratio. From this study, female (1.499 times) participants and graduates (1.316) were in more risky positions compare to other participants. The Students (1.353) were in more vulnerable positions, however, the service holders and the health and personnel were at low risk. The urban (1.283 times more affected) and the unmarried (1.559 times more affected), should be more careful as they both at high risk to befall in COVID-19. We should avoid traveling because the people who travel regularly were 1.807 times at high risk. Besides, it should restrict contact with symptomatic patients, as they were 1.860 times more vulnerable because of traveling.

This research has some limitations. First, SARS-CoV-2's information limited. Second, the information provided here based on current evidence only. Third, the sample size was small.

Conclusion

Late detection of COVID-19 can cause serious health hazards. Besides, its outcome is significantly worse than those who have comorbidities. Concomitantly it exacerbates the death of a patient. So, early earnestness to identify the SARS-CoV-2 virus and taking prompt treatment can save many lives.

Abbreviations

COVID-19: Coronavirus Disease 2019; rRT-PCR: Real-time Reverse transcription-polymerase chain reaction; SARS-CoV-2: Severe Acute Respiratory Syndrome Coronavirus 2; MERS-CoV: Middle East Respiratory Syndrome Coronavirus; ICU: Intensive Care Unit; WHO: World Health Organization; PPE: Personal Protective Equipment; LML: Log-Minus-Log; CI: Confidence Interval; SE: Standard Error; df: Degrees of freedom; Exp(B): Odds ratio.

Declarations

ETHICS APPROVAL AND CONSENT TO PARTICIPATE

Case record forms were approved by the Ethical Review Committee of the Chittagong Medical College and Hospital, Chattogram, Bangladesh, and written informed consent was obtained from each participant. No additional ethical approval is needed.

CONSENT FOR PUBLICATION

Our manuscript was approved the consent for publication from Department of Microbiology, Chittagong Medical College.

AVAILABILITY OF DATA AND MATERIAL

The data sets are available from the corresponding author on reasonable request.

CONFLICT OF INTEREST

The authors declare no conflict of interest.

FUNDING

This research received no external funding.

AUTHORS' CONTRIBUTIONS

Sujan Rudra: Database management and Statistical analysis; Shuva Das: Conceptualization, Writing - original draft preparation; Md. Ehsanul Hoque: Writing - review & editing, Supervision; Abul Kalam: Investigation and Validation; Mohammad Arifur Rahman: Methodology and Investigation; The authors read and approved the final manuscript.

ACKNOWLEDGMENTS

We thank all of the medical and ancillary hospital staff and the patients for consenting to participate. We would like to thanks the editors and referees whose constructive criticism led us to develop the presentation and maintain the quality of the paper. We would also like to thanks the Molecular Biology Laboratory of Chittagong Medical College, Chattogram, to give us almost all the opportunity for conducting the research.

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Figures

Figure 1

LML plot of the covariates of the time between investigation date and development of symptoms of patient.