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Identification of Aggravation-Predicting Gene Polymorphisms in Coronavirus Disease 2019 Patients Using a Candidate Gene Approach Associated With Multiple Phase Pathogenesis: A Study in a Japanese City of 1 Million People

IMPORTANCE: The pathology caused by the coronavirus disease 2019 is mediated by host-mediated lung inflammation, driving severity, and mortality. Polymorphisms in genes encoding host inflammation and immune-related molecules may be associated with the development of serious pathologies, and identifying such gene polymorphisms may lead to the identification of therapeutic targets.

OBJECTIVES: We attempted to identify aggravation-predicting gene polymorphisms.

DESIGN: We use a candidate gene approach associated with multiple phase pathogenesis in coronavirus disease 2019 patients among a cohort in Hiroshima, a city with a population of 1 million, in Japan. DNA samples from the study populations were genotyped for 34 functional polymorphisms from 14 distinct candidate genes, which encode proteins related to viral cell entry, regulation of viral replication, innate immune modulators, regulatory cytokines, and effector cytokines.

SETTING AND PARTICIPANTS: Three core hospitals providing different services for patients with coronavirus disease 2019 under administrative control. A total of 230 patients with coronavirus disease 2019 were recruited from March 1, 2020, to March 31, 2021.

MAIN RESULTS AND MEASUREMENTS: Among the 14 genes, we found rs1131454 in *OAS1* and rs1143627 in *IL1B* genes as independent genetic factors associated with disease severity (adjusted odds ratio = 7.1 and 4.6 in the dominant model, respectively). Furthermore, we investigated the effect of multiple phase pathogenesis of coronavirus disease 2019 with unbiased multifactor dimensionality reduction analysis and identified a four-gene model with rs1131454 (*OAS1*), rs1143627 (*IL1B*), rs2074192 (*ACE2*), and rs11003125 (*MBL*). By combining these polygenic factors with polyclinical factors, including age, sex, higher body mass index, and the presence of diabetes and hypertension, we proposed a composite risk model with a high area under the curve, sensitivity, and probability (0.917, 96.4%, and 74.3%, respectively) in the receiver operating characteristic curve analysis.

CONCLUSIONS AND RELEVANCE: We successfully identified significant genetic factors in *OAS1* and *IL1B* genes using a candidate gene approach study as valuable information for further mechanistic investigation and predictive model building.

KEY WORDS: candidate gene approach; coronavirus disease 2019; gene polymorphisms; severe acute respiratory syndrome coronavirus 2; severity

The coronavirus disease 2019 (COVID-19) pandemic, caused by the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2), has resulted in a global and local health crisis. The transmission capacity of SARS-CoV-2 exceeds expectations, and the outbreak puts our medical systems at a risk of collapse.

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Patients with COVID-19 show a broad spectrum of disease progression. The identification of risk factors for severe COVID-19 is desperately needed to distinguish patients who are at increased risk and to treat them more aggressively with limited social resources. Clinical studies on COVID-19 have revealed the clinical risk factors for severe diseases, such as age, sex, obesity, and comorbidities (1–6). The pathogenesis of severe COVID-19 and the associated respiratory failure has been extensively investigated, revealing the mechanism of viral entry, life cycle, and the host immune system that sometimes responds excessively, resulting in unfavorable outcomes, such as cytokine storms and acute respiratory distress syndrome (7–10). Our understanding of the pathogenesis of COVID-19 suggests that the clinical severity could be determined by the host response against SARS-CoV-2 in multiple pathogenetic phases.

Gene polymorphism could be one of the determinants of individual responses against viruses, including SARS-CoV-2. The candidate gene approach is a useful technique for identifying genetic risk factors for complex disorders, such as severe COVID-19 (11). Candidates having a plausible physiologic role in a major pathway implicated in the pathogenesis of COVID-19 should be tested for the effects of the genetic variants of potentially contributing genes.

Here, we used the candidate gene approach study to delineate the host genetic factors contributing to severe COVID-19 with respiratory failure. Furthermore, we investigated the interaction of multiple phase pathogenesis with unbiased and model-free statistical methods, multifactor dimensionality reduction (MDR) analysis, and estimated the genetic factors as predictors of severe COVID-19.

MATERIALS AND METHODS

Study Participants

We recruited a total of 230 patients with COVID-19, which was confirmed by SARS-CoV-2 viral RNA polymerase chain reaction (PCR) test using nasopharyngeal swabs or other relevant biological fluids. Patients were recruited from three hospitals: Funairi Citizens Hospital, Hiroshima Prefectural Hospital, and Hiroshima University Hospital in Hiroshima, Japan, from March 1, 2020, to March 31, 2021 (initial phase of the pandemic). Priority vaccination for the elderly began in Hiroshima Prefecture on April 12, 2021.

Therefore, the cohort in this study consisted of unvaccinated patients. These three hospitals provided different services for patients with COVID-19 under administrative control. The judgment to admission was pretty depend on the pandemic situation. At the time of initial emerging phase, the judgment was strictly made even for asymptomatic patients to isolate them. Funairi Citizens Hospital treated mainly mild cases (participants, $n = 162/526$), and Hiroshima Prefectural Hospital treated mild to moderate II and some severe cases ($n = 46/354$). If the patients' condition became severe, Hiroshima University Hospital treated them in the ICU ($n = 22/26$). This study was approved by the Ethical Committee for Human Genome Research of Hiroshima University (Hi-258). Written informed consent was obtained from all the participants.

Disease severity was classified according to the National Institutes of Health guideline, Version October 22 (Rocky Mountain Laboratories, National Institute of Allergy and Infectious Diseases), based on the use of oxygen supplementation or mechanical ventilation (12). Severity was graded according to the maximum respiratory support received at any point during hospitalization (mild, no oxygen supplementation; moderate I, supplemental oxygen therapy only; moderate II, noninvasive ventilatory support [nasal high flow]; and severe, invasive ventilatory support including extracorporeal membrane oxygenation). Other data collection was performed by physicians from local medical records using an electronic data form. The data were gathered from the Hiroshima University Hospital for analyses. Comorbidities, such as diabetes mellitus (DM), hypertension (HT), chronic respiratory disease, chronic renal disease, cancer, and steroid use, were defined upon receiving the prescription. Symptoms, laboratory data, were collected from the date of the initial assessment after admission (**Supplemental Table 1**, <http://links.lww.com/CCX/A839>). Missing data were happened at random on the observed variables only among nonsevere patients (body mass index [BMI], $n = 17$; comorbidities, $n = 2$; and smoking history, $n = 11$, respectively).

Candidate Gene Polymorphism

DNA samples from the study populations were genotyped for 34 functional polymorphisms from 14 distinct candidate genes. These gene polymorphisms in genes encoding viral cell entry, regulation of viral replication, innate immune modulators, regulatory cytokines, and effector

cytokines were selected based on the best evidence from studies published under the key words “sepsis,” “respiratory infection,” “pneumonia,” “inflammation,” and “SARS.” If there were more than four candidate polymorphisms for a gene, the selection was based on the number of publications in single nucleotide polymorphism (SNP) database of National Center for Biotechnology information (13–38). Given the possibility of genetic diversity between the Japanese and East Asian populations (39), we performed the analysis even through the selected SNPs had low minor allele frequencies in database. The descriptions of the polymorphisms and minor allele frequencies for each marker are shown in **Supplemental Table 2** (<http://links.lww.com/CCX/A840>).

Genotyping

Blood samples of the study population were collected by EDTA containing tubes with blinded unique identification number. Peripheral blood mononuclear cells (PBMCs) were isolated from the blood draw. Genomic DNA was extracted from PBMCs or whole blood using a QIAcube (QIAGEN, Hilden, Germany). Then, most SNP genotyping was carried out using TaqMan SNP genotyping assays (Thermo Fisher Scientific, MA) according to the manufacturer’s protocol. In brief, two allele-specific TaqMan probes containing distinct fluorescent dyes and a PCR primer pair were used to detect specific SNP targets. Quantitative PCR, as a readout, was performed with Roter-GeneQ (QIAGEN). We describe the zygosity of A>G SNP as AA or GG, and AG in the case of homozygosity and heterozygosity, respectively.

Zygosity for the deletion (D)–insertion (I) polymorphism of the *ACE* gene was determined using an assay to distinguish D and I alleles based on PCR and visualization by electrophoresis (190bp and 490bp amplicon, respectively) (40). Samples classified as the deletion-homozygotic (DD) genotype were subjected to PCR again to eliminate DD mistyping by detecting a 319bp amplicon implied I allele (13).

Statistical Analysis

All statistical analyses were performed as complete-case analysis using the JMP Pro statistical software package 15.0.0 (SAS Institute, Cary, NC) and the Java software MDR v.3.0.2 (investigated by Jason Moore, available at www.sourceforge.net). The Pearson chi-square test was used to test for

differences in distribution between nonsevere and severe patients. Logistic regression analysis was used to estimate the associations between genotypes and risk of severe disease by computing the odds ratios (ORs) and 95% CIs from multivariate logistic regression analyses (adjusting for age > 65 yr, sex, BMI > 25, presence of diabetes, and hypertension, $n = 211$). The impact of polymorphism was tested in additive, dominant, and recessive models.

We employed the multifactor dimensionality reduction (MDR) analysis, a nonparametric and model-free method developed by Ritchie et al (41), and detected and characterized high-dimensional gene-gene interactions in studies with relatively small sample sizes ($n = 230$). MDR analysis identified nonhypothetical genotype combinations (or possible interactions) associated with disease severity. In brief, the dataset was divided into a model building set and a model testing set to evaluate the accuracy rate. The model with maximum cross-validation (CV) consistency (CVC) and balanced testing accuracy (TA) was selected as the best model. In this study, the MDR analysis was conducted using the dichotomous groupings of the polymorphisms selected by the candidate gene approach, and we used a 10-fold CV to calculate the mean CVC and balanced TA. The best models with the maximum TA and CVC were identified. The MDR results were considered statistically significant at the level of 0.05. The predictive models are created by a logistic regression model based on the forward-backward selection according to the mentioned clinical and genetic candidate factors ($n = 211$). Then, the accuracy of the severity risk model was assessed using the area under the curve (AUC).

RESULTS

Clinical Characteristics of the Study Population

Twenty-eight of the 230 patients (12.2%) with COVID-19 infection were categorized as having severe infection, which required mechanical ventilation during hospitalization (severe group). Ninety-eight (70 nonsevere and 28 severe) and six patients were received corticosteroid therapy and anti-interleukin (IL)-6 antibody therapy for COVID-19 infection, respectively (Supplemental Table 1, <http://links.lww.com/CCX/A839>). There was no patient who was received anti-IL-1 antibody or anti-tumor necrosis factor alpha antibody therapy. Seven patients in the severe group (25.0%) died due to COVID-19 infection, but no mortality was observed in those with mild

to moderate COVID-19 infection (nonsevere group). Consistent with previous reports, the severe group was older (age mean \pm SD, 64.2 ± 13.3 vs 48.5 ± 16.7), more likely to be male (82.1% vs 55.0%), and had higher BMI (27.6 ± 4.8 vs 23.7 ± 3.9) and comorbidities (82.1% vs 49.0%) compared with the nonsevere group (Table 1). Diabetes and hypertension were more frequent in the severe group than in the nonsevere group (50.0% vs 11.4%, and 71.4% vs 18.3%, respectively). There was no difference in the periods in which patients were admitted to the hospital after the start of symptoms between the severe

and nonsevere groups. In our cohort, history of smoking was not associated with the severity of infection; however, the current smoking status seems to be a beneficial factor associated with the severity, contrary to the expectation that smoking exacerbated COVID-19 infection. We observed that younger patients who were associated with a lower risk of severe disease were more frequently current smokers, resulting in a conflicting association. Based on these observations, we set five clinical covariates (age ≥ 65 yr, sex, BMI ≥ 25 , diabetes, and hypertension) for covariate adjustment of the genetic analyses.

TABLE 1.
Characteristics of the Study Population

Demographics/Comorbidities	Nonsevere (<i>n</i> = 202)				Nonsevere vs Severe <i>p</i> ^b
	Mild ^a (<i>n</i> = 163)	Moderate I ^a (<i>n</i> = 36)	Moderate II ^a (<i>n</i> = 3)	Severe ^a (<i>n</i> = 28)	
Severity at admission (mild/moderate I/II/severe), <i>n</i>	163/0/0/0	15/21/0/0	3/0/0/0	4/12/3/9	
Number of death (mortality), <i>n</i> (%)	0	0	0	7 (25.0)	
Age, mean \pm SD	53.1 \pm 16.3	59.7 \pm 12.7	67.7 \pm 13.3	64.2 \pm 13.3	< 0.0001
Age < 65/ \geq 65 ^c , <i>n</i>	147/16	25/11	2/1	12/16	< 0.0001
Gender (male/female) ^c , <i>n</i>	85/78	24/12	2/1	23/5	0.004
BMI, mean \pm SD	23.4 \pm 3.5	25.2 \pm 5.3	23.5 \pm 1.0	27.6 \pm 4.9	< 0.0001
BMI < 25/ \geq 25 ^{c,d} , <i>n</i>	104/42	22/14	2/1	10/18	0.0006
Days from initial symptoms to admission, mean \pm SD	4.4 \pm 4.2	5.9 \pm 4.1	3.0 \pm 1.8	4.5 \pm 3.1	0.776
Current smoking, <i>n</i> (%)	43 (26.4)	3 (8.3)	0	0	0.002
History of smoking ^d , <i>n</i> (%)	84 (51.5)	19 (52.8)	1 (33.3)	12 (42.9)	0.536
Diabetes ^{c,d} , <i>n</i> (%)	16 (9.8)	7 (19.4)	0	14 (50.0)	< 0.0001
Hypertension ^{c,d} , <i>n</i> (%)	23 (14.1)	12 (33.3)	2 (66.7)	20 (71.4)	< 0.0001
Chronic respiratory disease ^d , <i>n</i> (%)	9 (5.5)	0	0	1 (3.6)	0.813
Chronic renal disease ^d , <i>n</i> (%)	0	1 (2.8)	0	2 (7.1)	0.026
Cancer ^d , <i>n</i> (%)	3 (1.8)	0	0	0	0.373
Steroid user ^d , <i>n</i> (%)	0	1	0	1 (3.6)	0.192
Cardiovascular infarction ^d , <i>n</i> (%)	4 (2.5)	0	0	4 (14)	0.008
Cardiac failure ^d , <i>n</i> (%)	1 (0.6)	0	0	1 (3.6)	0.192
Brain infarction ^d , <i>n</i> (%)	1 (0.6)	0	0	2 (7.1)	0.026
Organ transplantation ^d , <i>n</i> (%)	1 (0.6)	1	0	0	0.468

BMI = body mass index.

^aDefinition of severity classified according to National Institutes of Health guideline version October 22: mild, no oxygen supplementation; moderate I, supplemental oxygen therapy only; moderate II, noninvasive ventilatory support; and severe, invasive ventilatory support.

^bUnivariate statistic (Wilcoxon signed rank test/ χ^2 test).

^cSet up as adjusting covariates for genetic analysis.

^dMissing data: BMI (*n* = 17), comorbidities (*n* = 2), and history of smoking (*n* = 11).

Candidate Gene Approach Identified Independent Risk Genotypes for Severe COVID-19

The SNP descriptions are listed in Supplemental Table 2 (<http://links.lww.com/CCX/A840>). Two of the final set of 34 SNPs in the 14 genes investigated were found to be significantly associated with the severe group. The SNPs rs1131454 (*OAS1* gene, GG genotype) and rs1143627 (*IL1B* gene, GG genotype) had an increased risk of severe disease at approximately seven-fold and four-fold (adjusted OR [aOR], 7.1; 95% CI, 1.8–27.6 and aOR, 4.6; 95% CI, 1.3–17.0), respectively, in the dominant model (Table 2). The additive effect, which included a baseline risk (absence of the risk allele), a medium risk (heterozygosity for the risk allele), and a high risk (homozygosity for the risk allele), was not observed in these two SNPs for severity of COVID-19 infection. Additionally, the two SNP statuses were not associated with the five clinical covariates for adjustments.

Collectively, we identified independent genetic risk factors, homozygotes with the minor allele of rs1131454 in the *OAS1* gene and rs1143627 in *IL1B*, as predictors of severe COVID-19 infection.

Multiple Phase Model in MDR Analysis

We further investigated the gene-gene interactions and/or multiple phase effects of gene polymorphisms with MDR analyses. We obtained no consistent model in the two- or three-way gene interaction analyses. However, the four-way gene interaction analysis revealed a

significant association between rs1131454 (*OAS1*), rs1143627 (*IL1B*), rs2074192 (*ACE2*), and rs11003125 (*MBL*) (CVC, 0.9; TA, 0.74; $p < 0.0001$; Table 3).

Predictive Model for Severe COVID-19 Infection

We evaluated whether the use of polygenetic risk factors improved the prediction of severe disease once the individual was infected with COVID-19. We performed a receiver operating characteristic (ROC) curve analysis with the risk score calculated by a multivariable model using forward-backward stepwise binary logistic regression analysis. ROC curve analyses revealed that a higher AUC was found for the combination of five clinical factors and two gene factors compared among those with only two genetic factors that were identified as independent risk, four genetic factors by MDR analysis, and clinical factors (AUC = 0.915 vs 0.640, 0.709, and 0.885, respectively; Fig. 1). The genetic factors identified in the MDR analysis also slightly improved the AUC in the ROC analysis combined with clinical factors (AUC = 0.917). The final model (clinical and MDR genes) equation was: Predictive score (P) = $1/(1 + e^{-x})$

$$x = -1.12 + 0.767 \times (-1 \text{ if yr} \leq 64 \text{ 1 if yr} \geq 65) + 0.577 \times (-1 \text{ if female 1 if male}) + 0.549 \times (-1 \text{ if BMI} < 25 \text{ 1 if BMI} \geq 25) + 0.757 \times (-1 \text{ if DM negative 1 if DM positive}) + 0.998 \times (-1 \text{ if HT negative 1 if HT positive}) + 0.882 \times (-1 \text{ if rs1131454 others 1 if rs1131454 GG}) + 0.480 \times (-1 \text{ if rs1143627 others 1 if rs1143627 GG}) + 0.574 \times (-1 \text{ if rs11003125 CC 1 if rs11003125 others}).$$

The composite risk score including polygenetic and polyclinical factors showed a high sensitivity and

TABLE 2.
Genetic Risk Factors for Severe Coronavirus Disease 2019

Gene	Additive Model		Recessive Model		Dominant Model	
	Adjusted OR ^a	<i>p</i>	Adjusted OR	<i>p</i>	Adjusted OR	<i>p</i>
<i>OAS-1</i> (rs1131454)						
AA (<i>n</i> = 82)	Reference		7.1 (1.8–27.6)	0.0048	2.1 (0.63–6.8)	0.225
AG (<i>n</i> = 96)	1.1 (0.3–4/3)	0.913				
GG (<i>n</i> = 33)	7.4 (1.6–34.6)	0.0112				
<i>IL1B</i> (rs1143627)						
AA (<i>n</i> = 61)	Reference		4.6 (1.3–17.0)	0.0207	2.23 (0.57–8.64)	0.244
AG (<i>n</i> = 103)	1.5 (0.3–6.4)	0.608				
GG (<i>n</i> = 47)	6.1 (1.1–33.3)	0.0372				

OR = odds ratio.

We describe the zygosity of A>G SNP as AA or GG, and AG in the case of homozygosity and heterozygosity, respectively.

^aAdjusted with age > 65, gender, body mass index > 25, diabetes, and hypertension.

TABLE 3.
Interaction Gene Models for Severe Coronavirus Disease 2019 by Multifactor Dimensional Reduction Assay

No. of Gene	Best Interaction Model			Testing Accuracy	Cross-Validation Consistency	p
1	OAS1 rs1131454			0.61	9/10	0.0018
2	OAS1 rs1131454	Chemokine ligand 5 rs2280788		0.64	4/10	0.0102
3	OAS1 rs1131454	IL1B rs1143627	ACEI Alu repeat zygosity for the insertion-deletion polymorphism	0.68	3/10	0.0007
4	OAS1 rs1131454	IL1B rs1143627	ACE2 rs2074192 MBL rs11003125	0.73	8/10	< 0.0001

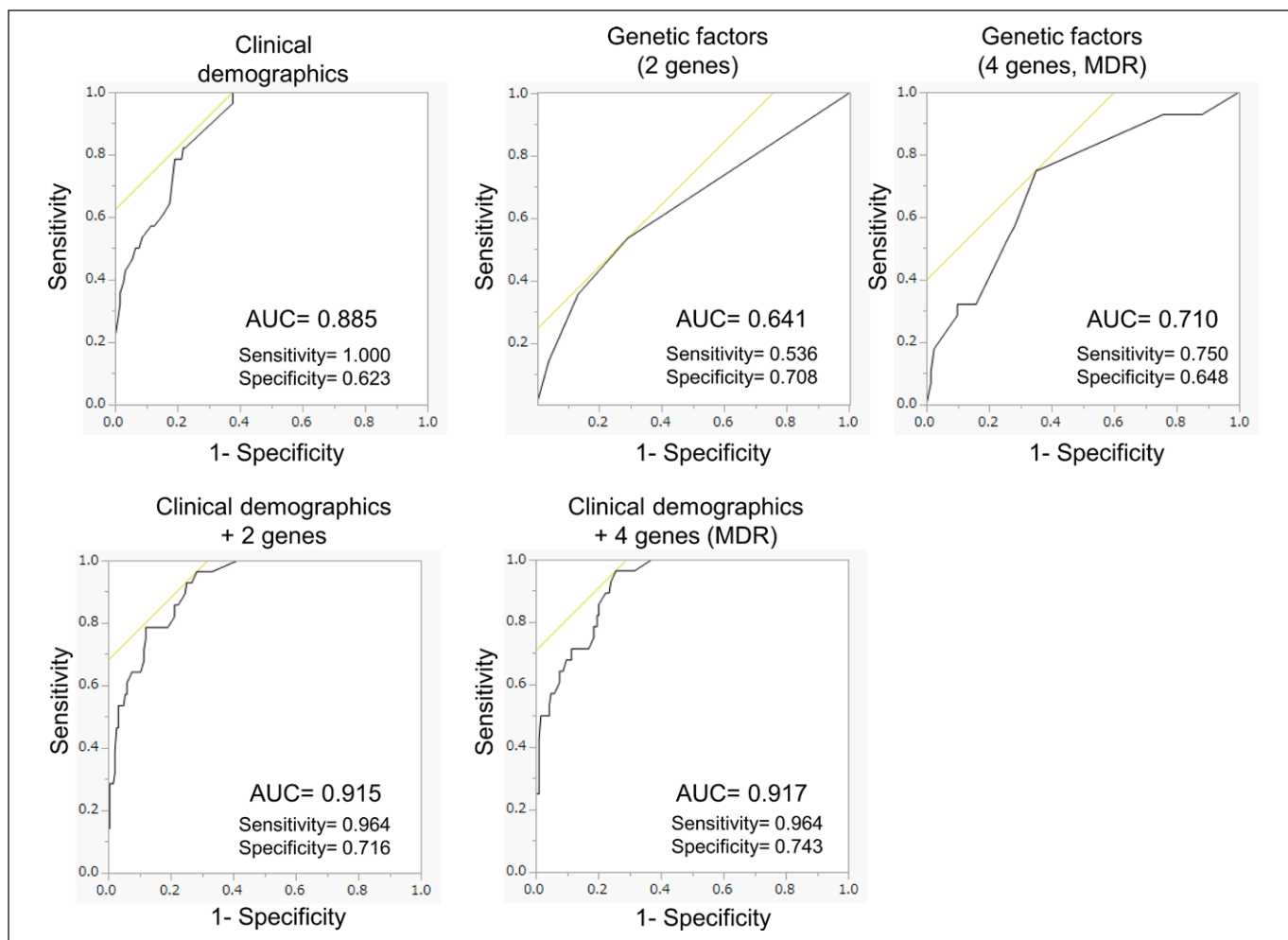


Figure 1. Receiver operating characteristic (ROC) analysis of predictive models for severe coronavirus disease 2019 infection. ROC curve analysis were performed with indicated clinical or genetic covariates to calculate the predictive value, area under the curve (AUC), sensitivity, and specificity. Clinical demographics and two genes indicated age greater than 65, gender, body mass index greater than 25, the presence of diabetes and hypertension, and rs1131454 (*OAS1*) and rs1143627 (*IL1B*), respectively. Four genes additionally indicated rs2074192 (*ACE2*) and rs11003125 (*MBL*) identified in multifactor dimensionality reduction (MDR) analysis.

specificity (96.4% and 74.3%, respectively) to effectively predict the population that may be associated with severe disease after COVID-19 infection.

DISCUSSION

The pathogenesis of COVID-19 infection is complex and involves a series of reactions in the host's body. Gene polymorphisms have been reported to be associated with these reactions against viral infections, including previous coronavirus infections such as SARS and Middle East respiratory syndrome, through multiple mechanisms (19, 22–24). Several genome-wide association studies (GWASs) have been conducted in Europe to identify unbiased targets for risk diagnosis and the therapeutic design for COVID-19 infection. The first report on GWAS of severe COVID-19 detected two genetic susceptibility loci, at Chr3p21.31 and Chr9q34.2, using a meta-analysis of two case/control panels of a thousand people from Italy and Spain (42). The gene cluster around the Chr3p21.31 locus, reported as a Neanderthal genetic variant, has been reproducibly associated with severe COVID-19 by at least three GWASs, indicating a common genetic mechanism underlying severe COVID-19 (42–44). Although genetic risk analysis is helpful to identify the natural risk of the population and even of individuals as well, it is sometimes difficult to conduct a study with a sufficient number of participants to identify unbiased targets, such as GWAS. The gene candidate strategy focuses on selected target genes and can reduce the required number of participants based on the information provided in previous reports. Using the candidate gene approach, we identified independent genetic factors, rs1131454 (*OAS1* gene) and rs1143627 (*IL1B* gene), which are associated with severe disease in the Japanese population. Both pathways identified in our cohort, *OAS1* and *IL-1b*, have convincing points of action, such as regulating viral replication and promoting an excessive immune response, respectively. The *OAS1* pathway, which activates RNase L to cleave the RNA virus genome, has also been identified as a candidate genetic mechanism in a European cohort by a GWAS using different SNPs (rs4767027 as a protein quantitative trait loci) (45). We investigated SNPs (rs2660 and rs10774671), which have high linkage disequilibrium with rs4767027, and both SNPs showed no significant impact on severe disease in our cohort. One of the previous reports revealed

that the Neanderthal risk haplotype, which includes the *OAS1* rs4767027, is almost absent in East Asian populations, such as in the Japanese, owing to its geographical distribution (44). However, the finding that a pathway involves disease severity with the same gene through different polymorphisms in different races could be interpreted to enhance the biological significance of the *OAS1* pathway in COVID-19 infection. It is likely that a genetic predisposition in the *OAS1* pathway may contribute to the dysregulation of the innate antiviral immune system in COVID-19 patients. Similarly, *IL-1* signaling has been reported as the most important pathway for the severity of COVID-19 patients in another GWAS (46). Together with these findings, further investigation of the mechanisms underlying the pathogenesis of COVID-19 through the pathways, *OAS1* and *IL-1b*, would be potential opportunities to develop new prevention and treatment strategies.

Previous investigators have developed risk scoring systems to identify severe or fatal disease outcomes based on clinical findings after infection, such as laboratory tests or CT findings with demographic characteristics, to guide clinical decisions (47–50). A representative study in the United Kingdom has worked on nearly 20 million primary care records to identify the factors associated with COVID-19–related death by creating a new health analytics platform and building a predictive model with a high *C*-statistic (0.93) (1). Owing to these valuable efforts (1–6), several demographics, such as age, sex, obesity, and comorbidities, are well-established risk factors for severe infection. Our independent genetic predictors for severe COVID-19 infection improved the predictive model with a combination of demographic characteristics (Fig. 1). The prediction of the aggravation in COVID-19 patients by clinical demographics was superior in sensitivity but not in specificity. On the other hand, that by genetic factors was superior in specificity but not in sensitivity. The predictive model by combining both factors and complemented each other resulted in a high in AUC. A model created through MDR analysis that originally investigated gene-gene interaction by unbiased computational model building can be interpreted as a model with multiple phase effects in different pathogenesis stages in COVID-19 infection. The model using four genes listed by MDR analysis provided a slight but further improvement for the prediction of severe COVID-19 infection before the individuals were infected. Under the current COVID-19 pandemic, a

better predictive model can provide valuable information for selecting potentially risky individuals to set up focused prophylaxis such as prior vaccination or develop strategies for the management of hospitalization or priority treatment in the limited social resources. However, the clinical significance of the slight improvement in AUC by adding genetic factors from two to four would be a subject for future research.

The limitation of this study is the sample size, which could provide only limited power. We believe this illustrates that using the proper methodology, such as the candidate gene approach, the genetic effects can be revealed even with a modest sample size. However, we cannot exclude the possibility that our findings may be due to chance and should be interpreted with caution. Further studies with large participants are required to validate our findings.

CONCLUSIONS

In conclusion, a candidate gene approach to investigate gene polymorphism for severe COVID-19 infection successfully identified independent genetic factors, OAS1 and IL-1B. This is valuable information for further mechanistic investigations and predictive model building.

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