	Severity group 1	Severity group	Severity group	Severity group 3
		2a	2b	
Age (mean ± S.D)	50 ± 19	63 ± 13*	$63 \pm 14*$	$64 \pm 14*$
Gender (male/female,	22 / 19	29 / 12	20 / 5	22 / 5
n)				
Diabetes mellitus (%)	24	29	44	41
Hypertension (%)	29	54	52	56
Current smoking (%)	10	22	20	22
Drugs used:				
Favipiravir (%)	34	51	32	52
Nafamostat (%)**	12	20	24	52
Remdesivir (%)**	5	41	64	33
Steroids (%)**	12	54	76	85

Supplemental Table S1. Characteristics of the subjects enrolled in this study.

Age is represented as the median \pm S.D. The differences in age among the three severity classes of COVID-19 were assessed using an independent Kruskal-Wallis test, followed by the Games Howell test for post hoc analysis. * P < 0.01 vs. severity 1. The differences in the sex distribution, frequency of complications, and rate of use of specific drugs among the three patient groups classified according to the severity of COVID-19 were assessed by the Chi square test. ** P < 0.01.

Models	Number of samples (%)					
	Training set		Validation set			
	Low severity	High severity	Low severity	High severity		
	group	group	group	group		
Day 4-7 S1 vs. S2a, 2b, 3	23 (35)	42 (65)	9 (33)	18 (67)		
Day 4-7 S1, 2a vs. S2b, 3	52 (83)	12 (18)	22 (81)	5 (19)		
Day 5-8 S1 vs. S2a, 2b, 3	25 (30)	58 (70)	10 (29)	24 (71)		
Day 5-8 S1, 2a vs. S2b, 3	62 (77)	19 (23)	27 (77)	8 (23)		
Day 6-9 S1 vs. S2a, 2b, 3	27 (29)	67 (71)	11 (28)	28 (72)		
Day 6-9 S1, 2a vs. S2b, 3	65 (69)	29 (31)	27 (69)	12 (31)		
Day 7-10 S1 vs. S2a, 2b, 3	28 (26)	78 (74)	12 (27)	33 (73)		
Day 7-10 S1, 2a vs. S2b, 3	70 (66)	37 (35)	29 (66)	15 (34)		
Day 8-11 S1 vs. S2a, 2b, 3	27 (20)	105 (80)	12 (27)	33 (73)		
Day 8-11 S1, 2a vs. S2b, 3	71 (57)	54 (43)	30 (58)	22 (42)		
Day 9-12 S1 vs. S2a, 2b, 3	26 (20)	105 (80)	12 (21)	45 (79)		
Day 9-12 S1, 2a vs. S2b, 3	67 (51)	64 (49)	29 (52)	27 (48)		

Supplemental Table S2. Summary of the training and validation datasets.

Best values for models		Hyperparame	eters		
		Max length	Min child	Colsample	Subsample
		2.5.5.0	weight	bytree	
Possible values		3, 5, 7, 9,	0, 0.2, 0.4, 0.6,	0, 0.3, 0.5, 0.7,	0, 0.3, 0.5, 0.7,
Dov 4.7 S1 vc	aliniaal	10, 11	0.8, 1	0.9, 1.0	0.9, 1.0
Day $4-7$ S1 vs. S2a 2b 3	chinical	10	0	0.5	0.7
52 u , 20, 5	clinical + antibody	10	0.4	0.7	0.9
Day 4-7 S1, 2a vs.	clinical	10	0	1.0	1.0
520, 5	clinical + antibody	10	0.4	0.3	0.5
Day 5-8 S1 vs.	clinical	10	0.4	0.3	0.5
32a, 20, 3	clinical + antibody	10	0.6	0.3	1.0
Day 5-8 S1, 2a vs.	clinical	10	0.4	0.5	0.7
520, 5	clinical + antibody	10	0.4	0.9	0.9
Day 6-9 S1 vs.	clinical	10	0.4	0.5	0.5
32a, 20, 3	clinical + antibody	10	0	0.9	0.3
Day 6-9 S1, 2a vs.	clinical	10	0	0.7	0.7
320, 5	clinical + antibody	10	0.8	1.0	0.3
Day 7-10 S1 vs.	clinical	10	0.8	0.3	0.3
S2a, 2b, 3	clinical + antibody	10	0.4	0.9	1.0
Day 7-10 S1, 2a	clinical	10	0.2	0.7	0.5
vs. 520, 5	clinical + antibody	10	0.2	0.3	0.5
Day 8-11 S1 vs. S2a, 2b, 3	clinical	10	0.8	0.7	1.0
	clinical + antibody	10	0	1.0	0.7
Day 8-11 S1, 2a vs.	clinical	10	0	0.5	0.5

Supplemental Table S3. Drawn optimal hyperparameter values for each model.

S2b, 3	clinical antibody	+	10	0	0.5	0.7
Day 9-12 S1 vs.	clinical		10	0.8	0.3	0.5
S2a, 20, 5	clinical antibody	+	10	0.2	0.3	0.5
Day 9-12 S1, 2a	clinical		10	0.2	0.7	0.5
vs. 520, 5	clinical antibody	+	10	0.8	0.3	0.7

"clinical" indicates that variables of age, gender, diabetes mellitus, hypertension, current smoking, CRP, and D-Dimer were input into the models, and "clinical + antibody" indicates that the clinical information described above plus the antibody titers were input into the models.

Supplemental Table S4. The accuracy of the model to distinguish severity group 2a or over from severity group 1 in the validation set.

day	true severity	estimated S1 (n)	estimate S2a, 2b, 3 (n)	error rate	accuracy
day	S1 (n)	4	3	0.43	0.80
1-6	S2a, 2b, 3 (n)	1	12	0.08	0.80
day	S1 (n)	14	3	0.18	0.01
7-12	S2a, 2b, 3 (n)	4	55	0.07	0.91

A. Clinical data alone

B. Clinical dat	a + antiboo	ly data
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day	true severity	estimated S1	estimate S2a,	error rate	accuracy		
		(n)	2b, 3 (n)	(%)			
day	S1 (n)	5	2	0.29	0.95		
1-6	S2a, 2b, 3 (n)	1	12	0.08	0.83		
day	S1 (n)	10	7	0.41	0.92		
7-12	S2a, 2b, 3 (n)	6	53	0.10	0.85		

Supplemental Table S5. The accuracy of the model to distinguish severity groups 2b and 3 from severity groups 1 and 2a in the validation set.

day	true severity	estimated S1, 2a (n)	estimate S2b, 3 (n)	error rate (%)	accuracy
day	S1, 2a (n)	15	1	0.06	0.95
1-6	S2b, 3 (n)	2	2	0.50	0.85
day	S1, 2a (n)	37	7	0.16	0.76
7-12	S2b, 3 (n)	11	21	0.34	0.70

A. Clinical data alone

B. Clinical data + antibody data

day	true severity	estimated S1 (n)	estimate S2a, 2b, 3 (n)	error rate (%)	accuracy		
day	S1, 2a (n)	15	1	0.06	0.00		
1-6	S2b, 3 (n)	1	3	0.25	0.90		
day	S1, 2a (n)	35	9	0.20	0.70		
7-12	S2b, 3 (n)	7	25	0.22	0.79		

Supplemental Table S6. The accuracy of the model to distinguish severity group 2a or over from severity group 1 in the validation set and the cases of breakthrough infections.

day	true severity	estimated S1	estimate S2a,	error rate	accuracy
		(n)	2b, 3 (n)	(%)	
day	S1 (n)	7	2	0.22	0.81
4-7	S2a, 2b, 3 (n)	3	15	0.17	0.81
day	S1 (n)	4	6	0.60	0.76
5-8	S2a, 2b, 3 (n)	2	22	0.08	0.76
day	S1 (n)	9	2	0.18	0.02
6-9	S2a, 2b, 3 (n)	1	27	0.04	0.92
day	S1 (n)	6	6	0.50	0.84
7-10	S2a, 2b, 3 (n)	1	32	0.03	0.84
day	S1 (n)	3	8	0.73	0.91
8-11	S2a, 2b, 3 (n)	2	39	0.05	0.81
day	S1 (n)	5	6	0.55	0.84
9-12	S2a, 2b, 3 (n)	3	42	0.07	0.84

A. The validation set (Clinical data + antibody against N antigen)

B. The cases of breakthrough infections (Clinical data + antibodies against N antigen)

day	true severity	estimated S1 (n)	estimate S2a, 2b, 3 (n)	error rate (%)	accuracy	
day	S1 (n)	4	1	0.20	0.70	
4-7	S2a, 2b, 3 (n)	1	3	0.25	0.78	
day	S1 (n)	3	1	0.25	0.00	
5-8	S2a, 2b, 3 (n)	0	6	0.00	0.90	
day	S1 (n)	1	2	0.67	0.75	
6-9	S2a, 2b, 3 (n)	0	6	0.00	0.75	
day	S1 (n)	1	2	0.67	0.75	
7-10	S2a, 2b, 3 (n)	0	5	0.00	0.75	
day	S1 (n)	0	4	1.00	0.64	
8-11	S2a, 2b, 3 (n)	0	7	0.00	0.64	
day	S1 (n)	0	4	1.00	0.60	
9-12	S2a, 2b, 3 (n)	0	6	0.00	0.00	

Supplemental Table S7. The accuracy of the model to distinguish severity groups 2b and 3 from severity groups 1 and 2a in the validation set and the cases of breakthrough infections.

day	true severity	estimated S1	estimate S2a,	error rate	accuracy
		(n)	2b, 3 (n)	(%)	
day	S1, 2a (n)	21	1	0.05	0.80
4-7	S2b, 3 (n)	2	3	0.40	0.89
day	S1, 2a (n)	24	3	0.11	0.80
5-8	S2b, 3 (n)	4	4	0.50	0.80
day	S1, 2a (n)	25	2	0.07	0.97
6-9	S2b, 3 (n)	3	9	0.25	0.87
day	S1, 2a (n)	26	3	0.10	0.75
7-10	S2b, 3 (n)	8	7	0.53	0.75
day	S1, 2a (n)	22	8	0.27	0.75
8-11	S2b, 3 (n)	5	17	0.23	0.75
day	S1, 2a (n)	24	5	0.17	0.77
9-12	S2b, 3 (n)	8	19	0.30	0.77

A. The validation set (Clinical data + antibody against N antigen)

B. The cases of breakthrough infections (Clinical data + antibodies against N antigen)

day	true severity	estimated S1 (n)	estimate S2a, 2b, 3 (n)	error rate (%)	accuracy
day	S1, 2a (n)	6	0	0.00	0.67
4-7	S2b, 3 (n)	3	0	1.00	
day	S1, 2a (n)	5	0	0.00	0.60
5-8	S2b, 3 (n)	4	1	0.80	
day	S1, 2a (n)	4	0	0.00	0.56
6-9	S2b, 3 (n)	4	1	0.80	
day	S1, 2a (n)	2	1	0.33	0.25
7-10	S2b, 3 (n)	5	0	1.00	
day	S1, 2a (n)	3	2	0.40	0.36
8-11	S2b, 3 (n)	5	1	0.83	
day	S1, 2a (n)	3	2	0.40	0.40
9-12	S2b, 3 (n)	4	1	0.80	



Severity 1 Severity 2a Severity 2b Severity 3

Supplemental Figure S1. Approximate curves for the antibody ratio kinetics in COVID-19 patients classified by the disease maximum severity.

Local polynomial regression curves were fitted to indicate the antibody ratios until day 12 after symptom onset in COVID-19 patients classified according to the disease severity. (A) IgM(S1/N), (B) IgM(RBD/N), (C) IgM(RBD/S1), (D) IgG(S1/N), (E) IgG(RBD/N), (F) IgG(RBD/S1), (G) IgA(S1/N), (H) IgA(RBD/N), (I) IgA(S1/RBD).



Severity 1 Severity 2a Severity 2b Severity 3

Supplemental Figure S2. The difference in antibody ratios among COVID-19 patients classified by the disease maximum severity.

We compared the various antibody ratios among patients with COVID-19 classified by the disease severity, as described in the *Material and Methods* section, on day 4-5, day 6-7, day 8-9, day 10-11, day 12 after symptom onset. *P < 0.05, **P < 0.01. The horizontal bar represents the median, the box bar represents the lower and upper quartiles, and the fine bar represents the minimum and maximum.



Clinical data + antibody data

Supplemental Figure S3. The workflow to predict severity groups of 2a or over, which represents one of tree estimators in the optimum model, on day 4-7.

Supplemental Figure S4. The feature importance in the model constructed using a machine learning technique to distinguish severity groups of 2a or over and severity group 1.

Clinical data

Clinical data + antibody data

Supplemental Figure S5. The workflow to predict severity group 2b or 3, which represents one of the tree estimators in the optimum model on days 4-7.

Supplemental Figure S6. The feature importance in the model constructed by a machine learning technique to distinguish severity groups 2b and 3 from severity groups 1 and 2a.

Supplemental Figure S7. ROC analyses of the analysis models constructed using a machine learning technique for predicting the maximum severity of COVID-19 when we sub-grouped the data on day 1-6 and day 7-12.

The ROCs of the analysis models constructed using a machine learning technique for predicting the COVID-19 severity, using the data obtained on day 1–6 (A, C), and day 7–12 (B, D), are shown. The models were constructed to distinguish severity groups 2a or over from severity group 1 (A, B) or distinguish severity groups 2b and 3 from severity groups 1 and 2a (C, D). The yellow curves represent the ROCs of the model constructed using clinical parameters and the green curves represent those of the model constructed using both clinical and antibody data.

Time course after the onset (day)

Supplemental Figure S8. The feature importance in the model constructed using a machine learning technique to distinguish severity groups of 2a or over and severity group 1, when we sub-grouped the data on day 1-6 and day7-12.

Time course after the onset (day)

Supplemental Figure S9. The feature importance in the model constructed by a machine learning technique to distinguish severity groups 2b and 3 from severity groups 1 and 2a, when we sub-grouped the data on day 1-6 and day7-12.

Supplemental Figure S10. ROC analyses of the analysis models constructed using a machine learning technique with only antibodies against N antigen for predicting the severity of COVID-19.

The ROCs of the analysis models constructed using a machine learning technique for predicting the COVID-19 severity, using the data obtained on day 4–7 (A, G), day 5–8 (B, H), day 6–9 (C, I), day 7–10 (D, J), day 8–11 (E, K), and day 9–12 (F, L), are shown. The models were constructed to distinguish severity groups 2a or over from severity group 1 (A–F) or distinguish severity groups 2b and 3 from severity groups 1 and 2a (G–L). The yellow curves represent the ROCs of the model constructed using clinical parameters and the green curves represent those of the model constructed using both clinical and antibody data (IgM(N), IgG(N), and IgA(N)).

Supplemental Figure S11. The workflow to predict severity groups of 2a or over, which represents one of tree estimators in the optimum model, on day 6-9 (A) and that to predict severity group 2b or 3, which represents one of the tree estimators in the optimum model on days 4-7 (B).

Supplemental Figure S12. The feature importance in the model constructed using a machine learning technique with the clinical data with the antibody data of antibodies against N antigen to distinguish severity groups of 2a or over and severity group 1 (A) and to distinguish severity groups 2b and 3 from severity groups 1 and 2a (B).