



1 *Type of the Paper (Communication)*

## 2 **HLA-B\*44 and C\*01 prevalence correlates with Covid19 spreading** 3 **across Italy**

4 **Pierpaolo Correale<sup>\*1</sup>, Luciano Mutti<sup>\*2</sup>, Francesca Pentimalli<sup>3</sup>, Giovanni Baglio<sup>4</sup>, Rita Emilena**  
5 **Saladino<sup>5</sup>, Pierpaolo Sileri<sup>4</sup>, Antonio Giordano<sup>2,6, #</sup>**

6 1 Unit of Medical Oncology, Oncology Department, Grand Metropolitan Hospital 'Bianchi Melacrino  
7 Morelli' Reggio Calabria I-89124, Italy; [correalep@yahoo.it](mailto:correalep@yahoo.it)

8 2 Sbarro Institute for Cancer Research and Molecular Medicine, Center for Biotechnology, College of Science  
9 and Technology, Temple University, Philadelphia, Pennsylvania, PA 19122, USA; [luciano.mutti@hotmail.it](mailto:luciano.mutti@hotmail.it)

10 3 Cell Biology and Biotherapy Unit, Istituto Nazionale Tumori-IRCCS-Fondazione G. Pascale, I-80131,  
11 Napoli, Italy; [f.pentimalli@istitutotumori.na.it](mailto:f.pentimalli@istitutotumori.na.it)

12 4 Ministry of Health, Italy; [g.baglio@sanita.it](mailto:g.baglio@sanita.it); [piersileri@yahoo.com](mailto:piersileri@yahoo.com)

13 5 Tissue Typing Unit Grand Metropolitan Hospital 'Bianchi Melacrino Morelli' Reggio Calabria I-89124, Italy;  
14 [ritaemilena.saladino@gmail.com](mailto:ritaemilena.saladino@gmail.com)

15 6 Department of Medical Biotechnologies, University of Siena, Italy; [antonio.giordano@unisi.it](mailto:antonio.giordano@unisi.it)

16

17 \* Dr. Pierpaolo Correale and Prof. Luciano Mutti have to be equally considered as first author.

18

19 # Correspondence: Prof. Antonio Giordano, Sbarro Institute for Cancer Research and Molecular Medicine,  
20 Center for Biotechnology, College of Science and Technology, Temple University, BioLife Science Bldg.  
21 Suite 333, 1900 North 12th Street, Philadelphia, PA 19122, USA Tel:001 215-204 9520 Fax: 001- 215-204 9522  
22 Email: [giordano@temple.edu](mailto:giordano@temple.edu)

23 Received: date; Accepted: date; Published: date

24

25 **Abstract:** Covid-19 spreading is showing huge, unexplained, differences between northern and  
26 southern Italy. We hypothesized that the regional prevalence of specific class I HLA alleles, which  
27 shape the anti-viral immune-response, might partly underlie these differences.

28 Through an ecological approach, wWe analyzed whether a set of HLA alleles (A, B, C), known to be  
29 involved in the immune response against infections, correlates with Covid-19 incidence. Covid-19  
30 data were provided by the National Civil Protection Department, whereas HLA allele prevalence was  
31 retrieved through the Italian Bone-Marrow Donors Registry. Among all the alleles HLA-A\*25, B\*08,  
32 B\*44, B\*15:01, B\*51, C\*01, and C\*03 showed a positive log-linear correlation with Covid-19 incidence  
33 rate fixed at the National outbreak peak on April 9<sup>th</sup> 2020 (Pearson's coefficients between 0.50 and 0.70,  
34 p-value<0.0001), whereas HLA-B\*14, B\*18, and B\*49 showed an inverse log-linear correlation  
35 (Pearson's coefficients between -0.47 and -0.59, p-value<0.0001).

36 When alleles were examined simultaneously using a multiple regression model to control for  
37 confounding factors, HLA-B\*44 and C\*01 still resulted positively and independently associated with  
38 Covid-19: a growth rate of 16% (95%CI: 0.1%-35%) per 1% point increase in B\*44 prevalence; and of 19%  
39 (95%CI: 1%-41%) per 1% point increase in C\*01 prevalence.

40 Our epidemiologic analysis, despite the limits of the ecological approach, is strongly suggestive of  
41 unravels a permissive role of HLA-C\*01 and B\*44 towards SARS-CoV-2 infection, which warrants  
42 further investigation in case-control studies. This study opens a new potential avenue for the  
43 identification of sub-populations at risk, which could provide Health Services a tool to define more  
44 targeted clinical management strategies and priorities in vaccination campaigns.

45 **Keywords:** SARS-Cov2; coronavirus; Covid-19; HLA class I; viral infection susceptibility.  
46

---

## 47 **1. Introduction**

48 Covid-19 has been declared a pandemic by the WHO [1]. Italy showed an explosive and apparently  
49 unrestrainable evolution throughout the country rapidly achieving one of the highest infection and  
50 mortality rates worldwide since the first case diagnosed in the province of Lodi in Lombardy, on  
51 February, 21<sup>st</sup>, 2020 [2]. Italian authorities are strictly monitoring the outbreak and report a large  
52 gradient of frequency that decreases from the northern to the southern and the Islands, across the  
53 twenty regions of the country [3]. So far this gradient has not been significantly modified even  
54 though the epidemic had the possibility to spread all along the peninsula due to massive migratory  
55 fluxes of individuals escaping from the high risk regions to return in their native landscapes [4] and  
56 a delayed restrictive response by the different regional authorities until March 2020. The incidence of  
57 Covid-19 cases reported by the national authorities also indicates relevant differences in the  
58 infection spreading within single provincial areas composing some of the most affected Italian  
59 regions (Figure 1a,b). Many socio-political as well as environmental hypotheses have been proposed  
60 to explain inter- and intra-regional differences and the reasons for such an aggressive spreading  
61 throughout northern Italy but so far no clear demonstration has been provided.

62 Emerging data are showing that both T cell and humoral response to Covid-19 infection may equally  
63 contribute to virus clearance and protective memory [2,5], however in a minority of infected patients  
64 an inappropriate immune-response may lead to virus spread from the oropharyngeal district to lung  
65 and other tissues including kidney and the central nervous system (CNS) [2,6,7]. Moreover an  
66 exaggerated cell mediated response to the virus in the alveolar tissue may be responsible of the  
67 dreaded cytokines storm and interstitial pneumonitis leading to a fatal acute respiratory distress  
68 syndrome (ARDS) [2,8–10].

69 Considering the crucial role played by Class I/II Human leukocyte antigen (HLA) molecules in  
70 triggering the anti-viral immune-response, it has been hypothesized that different HLA alleles may  
71 define an individual susceptibility to Covid-19 infection and spreading as reported for other viruses  
72 as well as the two different corona-viruses responsible for Severe Acute Respiratory Syndrome  
73 (SARS) and Middle East Respiratory Syndrome (MERS) [5]. In this context a number of studies is  
74 searching for selected HLAs with very efficient ability to present viral-antigen derived epitope  
75 peptides to cytotoxic T cells. The identification of highly immunogenic peptide epitopes recognized  
76 by specific T Cell Receptors (TCRs) might in fact, provide potential candidates for vaccine  
77 development [11–13]. By triggering and sustaining of the human host immune-defences to the virus,  
78 specific class I HLA alleles may also be involved in the occurrence of other symptoms, morbidity or  
79 lethality. Indeed, a study in a small cohort of Covid-19 Chinese patients suggested that specific HLA  
80 alleles might correlate with disease occurrence [14]. So, we set out to perform an exploratory  
81 epidemiological analysis, [through an ecological approach](#), aimed to investigate whether known

82 differences existing in HLA-A, B, and C allele distribution among the Italian population could be  
83 correlated to Covid-19 incidence and spreading throughout the peninsula.

## 84 2. Results

### 85 *Correlation between HLA-A, B, and C allele frequency and Covid-19 incidence in Italian Provinces*

86 We extrapolated HLA-A, B, and C allele frequency within the different Italian regions and relative  
87 intraregional provinces from the Italian Bone Marrow Donors Registry (IBMDR) report published on  
88 February, 2010 [15]. This is the largest published national database of– bone marrow healthy donors  
89 and includes a cohort of 370,000 individuals. The regional distribution of the HLA, A, B, C alleles  
90 was also evaluated in a more recent IBMDR high-definition-analysis database including a further  
91 120,926-individual sample size typed with high resolution method, which did not show significant  
92 changes compared to the previous report [16]. These data extrapolated from healthy donors  
93 represent a reliable surrogate of the real HLA-allele frequency scenario existing within the Italian  
94 population inhabiting different geographic areas of the country.

95 We examined allele prevalence across the Italian regions and detected a higher frequency of  
96 HLA-A\*25, B\*08, B\*44, B\*15:01, B\*51, and C\*01, and C\*03 alleles in the northern regions compared  
97 with that recorded in the southern regions. A reverse situation was instead observed for HLA-B\*14,  
98 B\*18, and B\*49 alleles, which frequency is higher in the southern regions (Figure 1c, d and  
99 supplementary data). All of the other HLA A, B, and C alleles in the database did not show  
100 substantial inter and intraregional differences (data from IBMDR database) [15].

101 We subsequently analyzed all the selected alleles examined and classified (HLA-A\*25, B\*08, B\*44,  
102 B\*15:01, B\*51, B\*14, B\*18, B\*49, C\*01, and C\*03) allele frequency for each province of the twenty  
103 Regions and then-compared them with Covid-19 incidence, as reported by the Italian Department of  
104 Civil Protection. Our results showed the existing correlation between HLA allele frequency and  
105 Covid-19 incidence rate fixed at the National outbreak peak on April 9<sup>th</sup> 2020, suggesting that the  
106 shape of the relationship is non-linear: as HLA prevalence increases, Covid-19 incidence varies  
107 according to an exponential trend (Figure 2a). In particular, HLA-A\*25, B\*08, B\*44, B\*15:01, B\*51,  
108 C\*01, and C\*03 alleles showed a positive log-linear correlation with Covid-19 incidence rate. On the  
109 other hand, HLA-B\*14, B\*18, and B\*49 alleles, whose frequency is higher in the southern regions,  
110 showed an inverse log-linear correlation (Figure 2a).

111 Pearson's coefficient (r) was calculated as a measure of the correlation between the logarithm of the  
112 Covid-19 incidence and the prevalence of different HLA alleles, in accordance with the exponential  
113 model. The complete correlation matrix with Pearson's coefficients and corresponding p-value for  
114 each couple of variables are shown in Table\_1. Such HLA alleles that were positively correlated to  
115 Covid-19 infection revealed Pearson's coefficient values between 0.50 and 0.70 ( $p < 0.0001$ ), and those  
116 inversely correlated between -0.47 and -0.59 ( $p < 0.0001$ ).

117 In order to control for mutual confounding (also including in the model the regions as confounders),  
118 the above mentioned HLA-alleles were examined simultaneously by performing a multivariable

119 regression analysis whose results showed that only HLA-B\*44 and C\*01 alleles maintained a positive  
120 and independent association with Covid-19 incidence (Table 2).

121 The exponential of the regression coefficient (growth rate) allowed us to quantify an increase of 16%  
122 (95%CI: 0.1%-35%) in Covid-19 incidence per 1 percentage point increase in B\*44 prevalence (Table  
123 2). Considering that the range of B\*44 prevalence varies in Italy from 4% to 12%, the risk of  
124 developing Covid-19 for the highest level of prevalence can be estimated 3 times greater than that  
125 for the lowest one (Table 2).

126 Similarly, for C\*01 the growth rate was 19% (95%CI: 1%-41%): considering a range of prevalence  
127 among provinces from 1% to 9%, the risk of developing Covid-19 for the highest level of prevalence  
128 is 4 times higher.

129 In order to provide further proof of evidence concerning the correlation of permissive HLA allele  
130 prevalence and Covid-19 incidence, we focused on two regions (Emilia Romagna and Marche)  
131 where the prevalence of B\*44 and C\*01 alleles is unevenly distributed among the different provinces  
132 (Figure [32b](#)).

133 Remarkably, in these regions the identified correlation seems to account for the intra-regional  
134 differences that are currently unexplained, such as the low incidence of Covid-19 in the province of  
135 Ferrara, compared with the other Emilian provinces, which are highly affected by the virus.  
136 Similarly, Pesaro-Urbino, which is the most affected province in the Marche Region, is also one with  
137 the highest prevalence of HLA-B\*44 (Figure 1e, f). In the latter case, the exponential regression curve  
138 explains almost all the variance of the data ( $r^2 = 0.9172$ ) and the prevalence of B\*44 can almost predict  
139 exactly the incidence of Covid-19 (Figure [32b](#)).

140

### 141 3. Discussion

142 Our epidemiologic analysis, through a geographical ecological approach, identified putative  
143 permissive class I alleles that are potentially unable to trigger an efficient immune-response unable  
144 to counteract SARS-Cov-2 infection.

145 In particular, we selected from the widest national genetic study that reports HLA data (in terms of  
146 allele prevalence) from almost 500,000 bone marrow donors representing the population from the  
147 whole national territory, those that presented stable inter- and intra-regional differences in  
148 prevalence to examine whether they could underlie the geographic differences in Covid-19  
149 incidence. By univariate analysis we found that HLA-A\*25, B\*08, B\*44, B\*15:01, B\*51, and C\*01, and  
150 C\*03 alleles showed a positive correlation with Covid-19 incidence rate, whereas found HLA A\*25,  
151 B\*08, B\*44 and C\*01 allele prevalence in the healthy population as geographically correlated with the  
152 incidence of Covid-19 in Italy, while the prevalence of HLA-B\*14, B\*18, and B\*49 showed an  
153 opposite trend.

154 Then, we tested the association between Covid-19 incidence and HLA alleles independently of each  
155 other using a multivariable regression analysis. Importantly, as an alternative approach to stratified

156 analysis, the Italian regions were included as covariates in the model to control for the confounding  
157 effect of the geographical context, and at the same time to verify the association of interest regardless  
158 of the North-South gradient.

159 Interestingly, ~~specific HLA assets in the Italian epidemiologic scenario seems to explain, at least~~  
160 ~~partly, the dramatic differences in the infection rate recorded in nearby provinces within the same~~  
161 ~~regions. Overall, the inter and intra regional infection rate clearly seems to mirror the frequency of~~  
162 ~~HLA A\*25, B\*08, B\*44, B\*15:01, B\*51, and C\*01, and C\*03 alleles in the healthy inhabitants. In our~~  
163 ~~analysis, we also identified other HLA alleles showing significant differences in frequency among~~  
164 ~~the twenty Italian regions but none of the selected alleles could be independently associated to~~  
165 Covid-19 incidence if not associated to the permissive HLA B\*44 and C\*01.

166 Our results are not surprising considering that class I HLA molecules have the specific task of  
167 binding and presenting antigen-derived epitope peptides to the TCR of epitope peptide specific T  
168 cells. With this mechanisms class I HLA molecules are critically involved in both CTL replication  
169 and ability to recogniz~~e~~ and destroy virus infected target cells. These 9-10 mere epitope peptides  
170 derive from the intracellular processing of protein antigens operated by the proteasome system prior  
171 being complexed with HLA molecules and translated on the membrane to be exposed to the TCR of  
172 the immune-effectors [17–20]. Peptide ability to bind HLA molecules is allele specific and is  
173 restricted by specific amino-acidic consensus motifs that allows their anchorage to different HLA  
174 molecules [21,22]. In this context, two individuals carrying the same antigen but different HLA  
175 profile may give rise to a completely different T cell mediated immune-response since they may  
176 have completely different amount of HLA specific antigen derived epitopes. This hypothesis has  
177 been confirmed in several studies concerning a number of different viruses as well as tumour  
178 antigens and autoimmune models [11,23–29].

179 Our model suggests that healthy individuals carrying HLA-B\*44 and/or C\*01 and in less extent  
180 HLA-A\*25, HLA-B\*08 alleles may be more susceptible to SARS-CoV-2 infection; indeed, they could  
181 be unable to present a sufficient amount of immune-dominant virus derived epitope peptides and  
182 consequently, they would be unable to mount a fast and efficient anti-viral immune response. It can  
183 be hypothesized that, in these patients, the virus may freely spread from the oropharyngeal mucosae  
184 starting a more efficient replication. Consistently, both HLA-B\*44 and C\*01 alleles, that we identified  
185 as possibly permissive to SARS-CoV-2 infection in Italy, have also been associated to known  
186 inflammatory autoimmune diseases [30–34], a fact that highlights their ability to trigger  
187 non-proficient and often inappropriate immunological reactions. The latter finding deserve to be  
188 explored in direct experimental approaches aimed to investigate whether the expression of these  
189 HLA allele also correlates with more aggressive disease outcomes and development of interstitial  
190 pneumonitis. Interestingly, inheritance of HLA-B\*44 was shown to underlie susceptibility to  
191 recurrent sinopulmonary infection [35] A further consideration stems from the knowledge that the  
192 HLA-C\*01 allele, which resulted as the most permissive to SARS-CoV-2 infection in our study, also  
193 represents the specific ligand of Killer cell Immunoglobulin like receptors (KIRs), KIR2DL2 and

194 KIR2DL3 [36–38]. These receptors are able to inhibit the activity of Natural killer cells, which  
195 represent the first line of host defence to the infection before the occurrence of a more specific T cell  
196 response [39]. An hypothesis that deserves further and more accurate investigation.

197 In the present study we could not assess whether a correlation existed among HLA alleles and  
198 Covid-19-associated morbidity (Hospitalization) and lethality rates owing to the absence of reliable  
199 data and the presence of significant confounding factors, including delay in hospital records  
200 transmission, saturation of emergency rooms and presence of co-morbidities. However, our data  
201 concerning the number of Covid-19 cases might be biased towards more severe outcomes because  
202 the initial national screening was mostly limited to symptomatic or hospitalized individuals,  
203 therefore implying in our correlation analyses an association with severity of the disease.

204 Although Despite the limits of our this type of ecological, epidemiologic approach has intrinsic  
205 limits, it also has the advantage of considering a large number of cases which are readily available  
206 through public-access data sets. Indeed, geographical ecological studies are often the first to identify  
207 risk factors for a variety of diseases, which are then verified through subsequent studies [40].

208 Our observational study identifies HLA-C\*01 and B\*44 alleles as potential genetic determinants for  
209 the identification of individuals at risk, which warrants to be further investigated in case-control  
210 studies. To this purpose we are currently investigating the expression of different HLA alleles in  
211 pauci-symptomatic patients affected by Covid-19 and those with severe interstitial pneumonitis. We  
212 suppose that Class I and II HLA genotyping in Covid-19 patients could be easily achieved and  
213 cost-effective and could provide the bases for the identification of individuals with high risk of  
214 interstitial pneumonitis and cytokines storm who should be immunised first as the vaccine will  
215 become available.

216 Overall our results, by identifying the potential relevance of HLA-C\*01 and B\*44 alleles in  
217 developing Covid-19 open new avenues of investigation not only to understand the diffusion and  
218 the physio-pathogenesis of the disease, but also to inform future vaccination campaign priorities and  
219 clinical management strategies while promoting the research of other potential permissive alleles  
220 and high risk population worldwide.

## 221 4. Materials and Methods

### 222 4.1 Data source and population sample

223 HLA is a highly polymorphic genetic system. The frequency of specific HLA alleles varies significantly  
224 among the different populations inhabiting the twenty geographic regions that compose the Italian  
225 Republic. We retrieved the HLA allele frequency data, which were recorded within the different  
226 regions and each relative intra-regional provinces, from the database of the IBMDR. We referred to an  
227 IBMDR database Analysis, published on February, 1<sup>st</sup>, 2010 [15], containing data that were collected in  
228 a twenty year interval from a cohort of 370,000 volunteer donors with known provincial and regional  
229 birthplace origin. These allele frequencies, expressed in % as calculated through the Arlequin  
230 Software, were also compared to those reported in an updated version of the IBMDR database,  
231 including data collected on a further 120,926 volunteer donor cohort [16]. Samples from the more

232 recent cohort were typed with high resolution method [16]; frequency data were grouped in regions  
233 and the same allele nomenclature version was used (according to the Immuno Polymorphism  
234 Database(IPD)-International ImMunoGeneTics project (IMGT/HLA) Database Release 3.32, April  
235 2018; <http://www.ebi.ac.uk/ipd/imgt/hla/stats.html>). The same allele frequency distribution across  
236 Italy was confirmed.

237 Data concerning the total number of individuals infected by SARS-CoV-2 per province (updated to  
238 April 9<sup>th</sup>, 2020) were provided by the Italian Department of Civil Protection, the institution under the  
239 Presidency of the Council of Ministers that manages the emergency at national level. Data were  
240 provided as aggregate numbers, in an anonymized manner.

241

#### 242 *4.2 Statistical analysis*

243 The relationship between SARS-CoV-2 infection and the frequency of HLA alleles was explored as  
244 part of an ecological approach aimed at assessing the degree of correlation between the incidence of  
245 Covid-19 and the prevalence of HLA alleles, both measured on a geographical basis (taking each  
246 Italian province as the unit of observation).

247 For each allele, values were preliminarily plotted in a scatter diagram and the curve with the best fit  
248 (corresponding to exponential curve) was selected using the least squares method, which is the most  
249 widely used procedure for developing estimates of the model parameters. The estimated regression  
250 equations are indicated at the top of each graph (Figures 2, 3).

251 ~~Pearson's coefficient was used as a measure of the correlation between the logarithm of Covid-19~~  
252 ~~incidence and the prevalence of different HLA alleles, taking each Italian province as the unit of~~  
253 ~~observation. Data were expressed as percentages of total population.~~

254 Consistent with the exponential model, Pearson's coefficient (r) was calculated as a measure of the  
255 correlation between the logarithm of the Covid-19 incidence and the prevalence of different HLA  
256 alleles. For each value of r, the corresponding p-value was also considered, in order to assess the  
257 statistical significance of the correlation (with respect to the null hypothesis of no log-linear  
258 correlation).

259 ~~Multivariable regression analysis was also performed in order to assess the association between~~  
260 ~~SARS-CoV-2 infection and each HLA allele independently from the others, also including regions in~~  
261 ~~the model as confounders.~~

262 Finally, the association between the logarithm of the Covid-19 incidence (considered as dependent  
263 variable) and HLA alleles independently of each other was tested using a multivariable regression  
264 analysis. Furthermore, the Italian regions were included as covariates in the model (which is an  
265 alternative approach to stratified analysis), in order to control for confounding of geographical  
266 context, and at the same time to verify the association of interest regardless of the North-South  
267 gradient.

268 All statistical analyses were conducted using STATA software 11.0 version (StataCorp LLC, Texas,  
269 USA). Microsoft Excel was used to draw maps.

270

271 **Supplementary Materials:** Supplementary materials can be found at [www.mdpi.com/xxx/s1](http://www.mdpi.com/xxx/s1).

272 **Author Contributions:** Conceptualization, P.C., L.M., A.G.; methodology, F.P., G.B., R.E.S.; formal analysis,  
273 P.C. F.P. G.B. R.E.S. P.S.; data curation, F.P. G.B. R.E.S.; writing—original draft preparation, P.C., L.M.;  
274 writing—review and editing, P.C., L.M., F.P., G.B., R.E.S.; supervision, P.S., A.G. All authors have read and  
275 agreed to the submitted version of the manuscript.

276 **Acknowledgments:** We are grateful to Dr Nicoletta Sacchi, Director of the Italian Bone Marrow Donors  
277 Registry (IBMDR), for providing access to the database and to the Tissue Typing Unit of Grand Metropolitan  
278 Hospital in Reggio Calabria in the persons of its director Dr Giuseppa Romeo responsible of Calabrian Regional  
279 Bone Marrow Donors' Registry and Dr Marina Francone for the precious contribution provided.

280 **Conflicts of Interest:** The authors declare no conflict of interest.

## 281 References

- 282 1. WHO Virtual press conference on COVID-19. Available online:  
283 [https://www.who.int/docs/default-source/coronaviruse/transcripts/who-audio-emergencies-](https://www.who.int/docs/default-source/coronaviruse/transcripts/who-audio-emergencies-coronavirus-press-conference-full-and-final-11mar2020.pdf?sfvrsn=cb432bb3_2)  
284 [coronavirus-press-conference-full-and-final-11mar2020.pdf?sfvrsn=cb432bb3\\_2](https://www.who.int/docs/default-source/coronaviruse/transcripts/who-audio-emergencies-coronavirus-press-conference-full-and-final-11mar2020.pdf?sfvrsn=cb432bb3_2).
- 285 2. Mutti, L.; Pentimalli, F.; Baglio, G.; Maiorano, P.; Saladino, R.E.; Correale, P.; Giordano, A.  
286 Coronavirus Disease (Covid-19): What Are We Learning in a Country With High Mortality  
287 Rate? *Front. Immunol.* **2020**, *11*, 1208, doi:10.3389/fimmu.2020.01208.
- 288 3. Livingston, E.; Bucher, K. Coronavirus Disease 2019 (COVID-19) in Italy. *JAMA* **2020**,  
289 doi:10.1001/jama.2020.4344.
- 290 4. Leaked coronavirus plan to quarantine 16m sparks chaos in Italy | World news | The  
291 Guardian Available online:  
292 [https://www.theguardian.com/world/2020/mar/08/leaked-coronavirus-plan-to-quarantine-1](https://www.theguardian.com/world/2020/mar/08/leaked-coronavirus-plan-to-quarantine-16m-sparks-chaos-in-italy)  
293 [6m-sparks-chaos-in-italy](https://www.theguardian.com/world/2020/mar/08/leaked-coronavirus-plan-to-quarantine-16m-sparks-chaos-in-italy).
- 294 5. Li, X.; Geng, M.; Peng, Y.; Meng, L.; Lu, S. Molecular immune pathogenesis and diagnosis of  
295 COVID-19. *J. Pharm. Anal.* **2020**, doi:<https://doi.org/10.1016/j.jpha.2020.03.001>.
- 296 6. Cheng, Y.; Luo, R.; Wang, K.; Zhang, M.; Wang, Z.; Dong, L.; Li, J.; Yao, Y.; Ge, S.; Xu, G.  
297 Kidney disease is associated with in-hospital death of patients with COVID-19. *Kidney Int.*  
298 **2020**, doi:10.1016/j.kint.2020.03.005.
- 299 7. Moriguchi, T.; Harii, N.; Goto, J.; Harada, D.; Sugawara, H.; Takamino, J.; Ueno, M.; Sakata,  
300 H.; Kondo, K.; Myose, N.; et al. A first Case of Meningitis/Encephalitis associated with  
301 SARS-Coronavirus-2. *Int. J. Infect. Dis.* **2020**, doi:10.1016/j.ijid.2020.03.062.
- 302 8. Wu, F.; Zhao, S.; Yu, B.; Chen, Y.-M.; Wang, W.; Song, Z.-G.; Hu, Y.; Tao, Z.-W.; Tian, J.-H.;  
303 Pei, Y.-Y.; et al. A new coronavirus associated with human respiratory disease in China.  
304 *Nature* **2020**, *579*, 265–269, doi:10.1038/s41586-020-2008-3.
- 305 9. Huang, C.; Wang, Y.; Li, X.; Ren, L.; Zhao, J.; Hu, Y.; Zhang, L.; Fan, G.; Xu, J.; Gu, X.; et al.  
306 Clinical features of patients infected with 2019 novel coronavirus in Wuhan, China. *Lancet*  
307 **2020**, *395*, 497–506, doi:10.1016/S0140-6736(20)30183-5.
- 308 10. Xu, Z.; Shi, L.; Wang, Y.; Zhang, J.; Huang, L.; Zhang, C.; Liu, S.; Zhao, P.; Liu, H.; Zhu, L.; et  
309 al. Pathological findings of COVID-19 associated with acute respiratory distress syndrome.  
310 *Lancet. Respir. Med.* **2020**, doi:10.1016/S2213-2600(20)30076-X.
- 311 11. Nguyen, A.; David, J.K.; Maden, S.K.; Wood, M.A.; Weeder, B.R.; Nellore, A.; Thompson, R.F.  
312 Human leukocyte antigen susceptibility map for SARS-CoV-2. *J. Virol.* **2020**,

- 313 doi:10.1128/jvi.00510-20.
- 314 12. Hyun-Jung Lee, C.; Koohy, H. In silico identification of vaccine targets for 2019-nCoV.  
315 *F1000Research* **2020**, *9*, 145, doi:10.12688/f1000research.22507.1.
- 316 13. Lurie, N.; Saville, M.; Hatchett, R.; Halton, J. Developing Covid-19 Vaccines at Pandemic  
317 Speed. *N. Engl. J. Med.* **2020**, doi:10.1056/nejmp2005630.
- 318 14. Wang, W.; Wei, Z.; Zhang, J.; He, J.; Zhu, F. Distribution of HLA allele frequencies in 82  
319 Chinese individuals with coronavirus disease-2019. *HLA* **2020**, doi:10.1111/tan.13941.
- 320 15. Amoroso, A.; Ferrero, N.; Rendine, S.; Sacchi, N. *Le caratteristiche HLA della popolazione Italiana:  
321 analisi di 370.000 volontari iscritti all'IBMDR; Analysis 1.; Sindacato nazionale dirigenti sanitari  
322 SSN e ARPA (SDS-SNABI), 2010;*
- 323 16. Sacchi, N.; Castagnetta, M.; Miotti, V.; Garbarino, L.; Gallina, A. High-resolution analysis of  
324 the HLA-A, -B, -C and -DRB1 alleles and national and regional haplotype frequencies based  
325 on 120 926 volunteers from the Italian Bone Marrow Donor Registry. *HLA* **2019**, *94*, 285–295,  
326 doi:10.1111/tan.13613.
- 327 17. Wölfel, T.; Klehmann, E.; Müller, C.; Schütt, K.H.; Meyer zum Büschenfelde, K.H.; Knuth, A.  
328 Lysis of human melanoma cells by autologous cytolytic T cell clones. Identification of human  
329 histocompatibility leukocyte antigen A2 as a restriction element for three different antigens. *J.  
330 Exp. Med.* **1989**, *170*, 797–810, doi:10.1084/jem.170.3.797.
- 331 18. Crowley, N.J.; Darrow, T.L.; Quinn-Allen, M.A.; Seigler, H.F. MHC-restricted recognition of  
332 autologous melanoma by tumor-specific cytotoxic T cells. Evidence for restriction by a  
333 dominant HLA-A allele. *J. Immunol.* **1991**, *146*, 1692–9.
- 334 19. Hunt, D.F.; Henderson, R.A.; Shabanowitz, J.; Sakaguchi, K.; Michel, H.; Sevilir, N.; Cox, A.L.;  
335 Appella, E.; Engelhard, V.H. Characterization of peptides bound to the class I MHC molecule  
336 HLA-A2.1 by mass spectrometry. *Science* (80-. ). **1992**, *255*, 1261–1263,  
337 doi:10.1126/science.1546328.
- 338 20. McDonnell, A.M.; Robinson, B.W.S.; Currie, A.J. Tumor antigen cross-presentation and the  
339 dendritic cell: where it all begins? *Clin. Dev. Immunol.* **2010**, *2010*, 539519,  
340 doi:10.1155/2010/539519.
- 341 21. Falk, K.; Rötzschke, O.; Stevanović, S.; Jung, G.; Rammensee, H.G. Allele-specific motifs  
342 revealed by sequencing of self-peptides eluted from MHC molecules. *Nature* **1991**, *351*,  
343 290–296, doi:10.1038/351290a0.
- 344 22. Gross, G.; Margalit, A. Targeting tumor-associated antigens to the MHC class I presentation  
345 pathway. *Endocr. Metab. Immune Disord. Drug Targets* **2007**, *7*, 99–109,  
346 doi:10.2174/187153007780832064.
- 347 23. Simmonds, M.J.; Gough, S.C.L. Genetic insights into disease mechanisms of autoimmunity.  
348 *Br. Med. Bull.* **2004**, *71*, 93–113, doi:10.1093/bmb/ldh032.
- 349 24. Li, S.; Jiao, H.; Yu, X.; Strong, A.J.; Shao, Y.; Sun, Y.; Altfield, M.; Lu, Y. Human leukocyte  
350 antigen class I and class II allele frequencies and HIV-1 infection associations in a Chinese  
351 cohort. *J. Acquir. Immune Defic. Syndr.* **2007**, *44*, 121–31,  
352 doi:10.1097/01.qai.0000248355.40877.2a.
- 353 25. Vejbaesya, S.; Thongpradit, R.; Kalayanaroj, S.; Luangtrakool, K.; Luangtrakool, P.; Gibbons,  
354 R. V.; Srinak, D.; Ngammthaworn, S.; Apisawes, K.; Yoon, I.-K.; et al. HLA Class I Supertype  
355 Associations With Clinical Outcome of Secondary Dengue Virus Infections in Ethnic Thais. *J.*

- 356 *Infect. Dis.* **2015**, *212*, 939–47, doi:10.1093/infdis/jiv127.
- 357 26. Hudson, L.E.; Allen, R.L. Leukocyte Ig-like receptors - A Model for MHC class I disease  
358 associations. *Front. Immunol.* **2016**, *7*, doi:10.3389/fimmu.2016.00281.
- 359 27. Rallón, N.; Restrepo, C.; Vicario, J.L.; Del Romero, J.; Rodríguez, C.; García-Samaniego, J.;  
360 García, M.; Cabello, A.; Górgolas, M.; Benito, J.M. Human leucocyte antigen  
361 (HLA)-DQB1\*03:02 and HLA-A\*02:01 have opposite patterns in their effects on susceptibility  
362 to HIV infection. *HIV Med.* **2017**, *18*, 587–594, doi:10.1111/hiv.12494.
- 363 28. Falfán-Valencia, R.; Narayanankutty, A.; Reséndiz-Hernández, J.M.; Pérez-Rubio, G.;  
364 Ramírez-Venegas, A.; Nava-Quiroz, K.J.; Bautista-Félix, N.E.; Vargas-Alarcón, G.;  
365 Castillejos-López, M.D.J.; Hernández, A. An Increased Frequency in HLA Class I Alleles and  
366 Haplotypes Suggests Genetic Susceptibility to Influenza A (H1N1) 2009 Pandemic: A  
367 Case-Control Study. *J. Immunol. Res.* **2018**, *2018*, 3174868, doi:10.1155/2018/3174868.
- 368 29. Correale, P.; Saladino, R.E.; Nardone, V.; Giannicola, R.; Agostino, R.; Pirtoli, L.; Caraglia, M.;  
369 Botta, C.; Tagliaferri, P. Could PD-1/PDL1 immune checkpoints be linked to HLA signature?  
370 *Immunotherapy* **2019**, *11*, 1523–1526.
- 371 30. Sanders, P.A.; Thomson, W.; Dyer, P.A.; Grennan, D.M. Haplotypes bearing HLA-A, -B, and  
372 -DR: Bf and C4 genes in rheumatoid arthritis families. *Tissue Antigens* **1989**, *33*, 21–9,  
373 doi:10.1111/j.1399-0039.1989.tb01673.x.
- 374 31. Orchard, T.R.; Thiyagaraja, S.; Welsh, K.I.; Wordsworth, B.P.; Gaston, J.S.H.; Jewell, D.P.  
375 Clinical phenotype is related to HLA genotype in the peripheral arthropathies of  
376 inflammatory bowel disease. *Gastroenterology* **2000**, *118*, 274–278,  
377 doi:10.1016/S0016-5085(00)70209-5.
- 378 32. Grams, S.E.; Moonsamy, P. V.; Mano, C.; Oksenberg, J.R.; Begovich, A.B. Two new HLA-B  
379 alleles, B\*4422 and B\*4704, identified in a study of families with autoimmunity. *Tissue*  
380 *Antigens* **2002**, *59*, 338–340, doi:10.1034/j.1399-0039.2002.590417.x.
- 381 33. Ueta, M.; Kannabiran, C.; Wakamatsu, T.H.; Kim, M.K.; Yoon, K.-C.; Seo, K.Y.; Joo, C.-K.;  
382 Sangwan, V.; Rathi, V.; Basu, S.; et al. Trans-ethnic study confirmed independent associations  
383 of HLA-A\*02:06 and HLA-B\*44:03 with cold medicine-related Stevens-Johnson syndrome  
384 with severe ocular surface complications. *Sci. Rep.* **2014**, *4*, 5981, doi:10.1038/srep05981.
- 385 34. Jung, E.S.; Cheon, J.H.; Lee, J.H.; Park, S.J.; Jang, H.W.; Chung, S.H.; Park, M.H.; Kim, T.-G.;  
386 Oh, H.-B.; Yang, S.-K.; et al. HLA-C\*01 is a Risk Factor for Crohn's Disease. *Inflamm. Bowel*  
387 *Dis.* **2016**, *22*, 796–806, doi:10.1097/MIB.0000000000000693.
- 388 35. Johnston, D.T.; Mehaffey, G.; Thomas, J.; Young, K.R.; Wiener, H.; Li, J.; Go, R.C.P.;  
389 Schroeder, H.W. Increased frequency of HLA-B44 in recurrent sinopulmonary infections  
390 (RESPI). *Clin. Immunol.* **2006**, *119*, 346–50, doi:10.1016/j.clim.2006.02.001.
- 391 36. Fadda, L.; Körner, C.; Kumar, S.; van Teijlingen, N.H.; Piechocka-Trocha, A.; Carrington, M.;  
392 Altfeld, M. HLA-Cw\*0102-restricted HIV-1 p24 epitope variants can modulate the binding of  
393 the inhibitory KIR2DL2 receptor and primary NK cell function. *PLoS Pathog.* **2012**, *8*,  
394 e1002805, doi:10.1371/journal.ppat.1002805.
- 395 37. Mori, M.; Wichukchinda, N.; Miyahara, R.; Rojanawiwat, A.; Pathipvanich, P.; Miura, T.;  
396 Yasunami, M.; Ariyoshi, K.; Sawanpanyalert, P. Impact of HLA allele-KIR pairs on disease  
397 outcome in HIV-infected Thai population. *J. Acquir. Immune Defic. Syndr.* **2018**, *78*, 356–361,  
398 doi:10.1097/QAI.0000000000001676.

- 399 38. Pende, D.; Falco, M.; Vitale, M.; Cantoni, C.; Vitale, C.; Munari, E.; Bertaina, A.; Moretta, F.;  
400 Del Zotto, G.; Pietra, G.; et al. Killer Ig-Like Receptors (KIRs): Their Role in NK Cell  
401 Modulation and Developments Leading to Their Clinical Exploitation. *Front. Immunol.* **2019**,  
402 *10*, 1179, doi:10.3389/fimmu.2019.01179.
- 403 39. Vitale, M.; Cantoni, C.; Della Chiesa, M.; Ferlazzo, G.; Carlomagno, S.; Pende, D.; Falco, M.;  
404 Pessino, A.; Muccio, L.; De Maria, A.; et al. An Historical Overview: The Discovery of How  
405 NK Cells Can Kill Enemies, Recruit Defense Troops, and More. *Front. Immunol.* **2019**, *10*, 1415,  
406 doi:10.3389/fimmu.2019.01415.
- 407 40. Grant, W.B. The role of geographical ecological studies in identifying diseases linked to UVB  
408 exposure and/or vitamin D. *Dermatoendocrinol.* 2016, *8*.
- 409

410

411

412



© 2020 by the authors. Submitted for possible open access publication under the terms and conditions of the Creative Commons Attribution (CC BY) license (<http://creativecommons.org/licenses/by/4.0/>).

413

414

415 **Figure legends.**

416 **Figure 1 | Covid-19 incidence, HLA-B\*44 and C\*01 prevalence in Italian Provinces** | **A.** The  
417 graphical map shows the twenty Italian regions each constituted by various provinces. **B.** The  
418 graphical map shows quintiles of Covid-19 incidence across Italian provinces. Incidence data were  
419 calculated as the number of laboratory-confirmed Covid-19 cases up to 04/09/2020 divided by the  
420 number of residents, according to the official National data (supplementary data). **C.** and **D.** The  
421 graphical maps show B\*44 and C\*01 prevalence (%) in Italian Provinces. **E.** and **F.** The graphical  
422 maps show Covid-19 incidence and B\*44 prevalence (%) in the provinces of Emilia Romagna and  
423 Marche. Geographical maps were built through Microsoft Excel. All Covid-19 incidence and HLA  
424 prevalence values are reported as Supplementary data.

425

426 **Figure 2 | Correlation between Covid-19 incidence rate and HLA prevalence** | The graphs show  
427 the correlation between Covid-19 incidence and the prevalence of HLA-A\*25, B\*08, B\*44, B\*15:01,  
428 B\*51, B\*14, B\*18, B\*49, C\*01, and C\*03, expressed as percentages, for all the available Italian  
429 provinces. For each correlation the R-squared value is provided at the top ~~right~~ of the graph along  
430 with the estimated regression equations. The R and p values are reported in table 1.

431

432 **Figure 3 | Correlation between Covid-19 incidence rate and HLA-B\*44 prevalence in Emilia**  
433 **Romagna and Marche provinces** | The graphs show the correlation between Covid-19 incidence  
434 and the prevalence of HLA-B\*44 prevalence, both expressed as percentages, for all the provinces of  
435 Emilia Romagna (top panel) and all the available provinces of Marche (bottom panel). For each  
436 correlation the R-squared value is provided at the top ~~right~~ of the graph along with the estimated  
437 regression equations. For Emilia Romagna: R=0,6813 and p value=0.0628; for Marche R=0,9577 and p  
438 value=0,0423.

439