

Host genetic markers associated with severe COVID-19: A systematic review

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Fecha de recepción: febrero de 2021.

Fecha de aceptación: junio de 2021.

ABSTRACT

Background: Severity of COVID-19 has been linked to several factors. As any other polygenic-multifactorial phenotype, genotype is not determinant in this prediction but may add actionable information. There is no consensus yet as to which genetic markers are useful, but several studies have been published that postulate different hypotheses acknowledging the relevance of including host genetics among the variables that predict the risk for severe forms of the disease. **Objective:** The objective of this study is to perform a systematic review that summarizes the projects, studies and postulated markers in order to establish if their application in clinical practice is currently feasible. **Materials and methods:** A comprehensive search was conducted in Pubmed. The inclusion criterion was studies of patients with COVID-19 who had germinal genetic markers of interest sequenced. The selected studies had to include at least a group of patients with the severe form of the disease. **Results:** 7 studies that met the criteria were included, which involved 6347 individuals. Markers for 19 genes have been postulated as relevant. **Conclusion:** The performed analysis indicates that multiple markers may be correlated with worse evolution of COVID-19; however, great heterogeneity has been found among the studies, which still precludes their translation into clinical practice.

KEYWORDS

Genetic predisposition, COVID-19, Computational biology, Genomics, Genome-Wide association study

Marcadores genéticos del huésped asociados con COVID-19 grave: una revisión sistemática

RESUMEN

Antecedentes: La severidad de COVID-19 depende de múltiples factores. Del mismo modo que en cualquier fenotipo multigénico-multifactorial, la genética no es determinante en esta predicción pero sí puede brindar información accionable. No hay un consenso aún sobre cuáles son los marcadores genéticos de utilidad pero sí hay varios estudios que postulan diferentes hipótesis, reconociendo la importancia de incluir la genética entre las variables que predicen riesgo de cuadros graves. **Objetivo:** El objetivo del estudio es realizar una revisión sistemática que resuma los proyectos/estudios realizados y los marcadores postulados con el fin de establecer si actualmente es posible su uso en la práctica clínica. **Materiales y métodos:** Se realizó una búsqueda exhaustiva en Pubmed. El criterio de inclusión fue estudios de pacientes COVID-19 con secuenciación de marcadores genéticos germinales de interés. Los estudios seleccionados debían incluir un grupo de pacientes que desarrollaron formas graves de la enfermedad. **Resultados:** 7 estudios cumplieron los criterios, los cuales involucran a 4604 individuos. Se postularon como relevantes marcadores en 19 genes. **Conclusión:** El análisis realizado evidencia múltiples marcadores que podrían estar correlacionados con peor evolución de COVID-19; sin embargo se detectó gran heterogeneidad en los resultados lo cual no permite aún la traslación a la clínica.

PALABRAS CLAVE

Predisposición genética, COVID-19, Biología computacional, Genómica, Estudio de asociación del genoma completo

INTRODUCTION

In December 2019, a large number of individuals developed pneumonia in the city of Wuhan, which attracted the interest of China, and the whole world [1]. After the identification of a coronavirus as the source of this outbreak, and the realization that it had the ability to provoke a severe acute respiratory syndrome [2], the Coronavirus Study Group taxonomically recognized it as being related to SARS-CoV, so they named it SARS-CoV-2 [3]. On February 11th 2020, World Health Organization (WHO) defined the name for the disease caused by this virus as COVID-19 [4], and on March 11th they characterized it as a pandemic [5]. On January 10th 2020, the first whole sequence of SARS-CoV-2 was published, and by April 7th 2020 more than 500 sequences had been deposited in GenBank [6].

Even though in most cases COVID-19 is associated with mild symptoms, or SARS-CoV-2 infected individuals may even be asymptomatic, the mortality risk for severe forms of the disease is high. In patients with mild and moderate symptoms, currently

available treatments include oxygen administration, antiretrovirals, immunomodulators and antithrombotics. New treatments for COVID-19 are constantly being tested, but no consensus or a definite solution for severe forms of the disease has yet been found.

The pandemic has affected, by June 8th 2021, 173,609,772 people worldwide [7] and has caused 3,742,653 deaths [7]. According to WHO, by June 8th, there are 102 vaccines in clinical development, 185 in pre-clinical development and 2,092,863,229 people have been vaccinated [7]. At least 60% of the population is currently considered to need to be vaccinated in order for the region to achieve herd immunity, but this percentage is under revision and depends on several variables [8].

Although many vaccines are being developed, herd immunity will not be achieved in many countries (especially those with middle and low income that have had greater difficulties in acquiring vaccines) in the short term, so a greater comprehension of the factors that determine the risk for more severe forms of the disease is key to adopt prevention and treatment strategies, not only at population level but also con-

sidering personalized/precision medicine paradigms. Even more, new viral strains may emerge in the future, which makes currently available vaccines useless. Many host characteristics have been postulated, and some of them considered proven, as representing risk factors for worse evolution of COVID-19, including age [9,10], gender [9,10] and the presence of certain comorbidities [11], mainly diabetes [9,12,13,14], cardiovascular disease [9,14], hypertension [9] and obesity [14,15,16]. However, the list of risk factors is not yet considered complete [10].

The identification and analysis of the genetic sequence of SARS-CoV-2 have been a central breakthrough for its classification and the development of vaccines in record time. On the other hand, host genetics may play an important role in the prediction of the disease progression, and therefore represent a valuable addition to the list of individual risk factors.

The interest in identifying such markers has been high in the last months of the pandemic. Genomics, bioinformatics and artificial intelligence (particularly the use of Machine learning techniques to infer models based on large volumes of data [17]) have been some of the core scientific disciplines involved in these findings. This applies both to the study of the viral genome and the human one as well. The Host Genetics Initiative [18] has had a central role in these last studies, since it functions as a public data repository for host genetic markers involved with the response to COVID-19. In a similar fashion, GISAID [19,20] has become one of the main databases for viral sequences.

From a molecular point of view, ACE2 -a protein mainly expressed in AT2 alveolar cells- has been found to act as a cellular binding site for the viral spike protein [21,22,23,24]. ACE2 had previously been confirmed as a binding site for other previously known coronaviruses, SARS-CoV and NL63 [25,24]. Another relevant finding was that the product of the human STMP3 gene is used by the virus to perform a cleavage that allows the fusion of the membranes and the following entry of the virus to the cell [23].

In the same cellular types that express ACE2, other key genes for the entry of the virus may be found at high levels: ITGB6, CAV2 [24], as well as genes that allow the newly formed viral particles to leave the cell: CHMP3, CHMP5, CHMP1A and VPS37B [24]. ACE2 levels in lung tissue vary among healthy individuals, in part due to ethnic factors. In people of East Asian origin, higher expression of ACE2 has been seen,

which has partially been explained by the differences in allelic frequency in genetic variants in eQTL sites (loci involved in differences in the gene expression) in this population in comparison with others [24,25,26].

Regarding immunity, Human leukocyte antigen (HLA) has a known relevant role in the susceptibility to several viral infections [27] and the severity associated with the disorders these infections may provoke [28]. Even though HLA depends on the genetic characteristics of the individual, the gold standard for HLA testing is currently not genetic sequencing, and correlations between HLA subtypes and their corresponding genotypes is not known in many cases. The same is true for blood ABO groups, which have been associated with severity of COVID-19 by some authors. Variants in the gene IFITM3 have been reported as potentially related to a more severe evolution in patients infected with influenza H749 or H1N1 [29,30]. Patients who harbour certain deleterious variants in the Myxovirus resistance A (MxA) gene appear to have a worse evolution as well, since this gene codes for an antiviral protein stimulated by α and β interferon [31].

Several studies have been performed in order to identify an association between host genetic markers and worse clinical evolution of the SARS-CoV-2 infection, mainly Genome-Wide Association Studies (GWAS).

In this work we had the objective to perform a systematic review in order to evaluate the clinical utility of applying germinal host genetic markers that have been postulated to have predictive value for severe forms of COVID-19.

MATERIALS AND METHODS

This study is a systematic review with qualitative methodology.

SEARCH STRATEGY

A comprehensive search in Pubmed was conducted, using the following expression based on MESH terms: "Genetic Predisposition to Disease"[Mesh] and COVID-19[Mesh].

The search was not restricted by a temporal variable, since the problem is naturally restricted in time, or by the population included.

The search was complemented by articles referenced by the papers originally identified in the search.

ELIGIBILITY CRITERIA

Observational studies which compare severe with non-severe cases on which any kind of genetic test would have been performed to acquire data on the patients' genotype were included.

TABLE 1 SHOWS INCLUSION CRITERIA.

Inclusion criteria
Observational studies which associate genetic markers with severe COVID-19 by: sequencing just one or several single nucleotide polymorphisms (SNP), a full gene, a group of genes of interest, or the individual's whole exome or genome or genotyping through microarray for a group of polymorphic markers (mainly SNP) or using already available molecular information of evaluated COVID-19 patients.
Authors report statistical significance of the results with a P-Value < .05 and odds ratio proving the effect of the variant on susceptibility to severe COVID-19 or bringing the data to allow odds ratio calculation.

EXCLUSION CRITERIA

TABLE 2 SHOWS EXCLUSION CRITERIA.

Exclusion criteria
Studies that did not make use of COVID-19 patients' genotypes obtained by sequencing, but inferred them from hypothesis, bioinformatic analysis only or speculations from population allele frequencies by country or ethnic group.
Studies that did not evaluate genetic predisposition to severe forms of COVID-19 but merely to being infected by SARS-CoV-2.
Studies that only tested for gene expression or other molecular data from the patients but not their germinal DNA.
Studies in which controls are taken from biobanks but are not tested for COVID-19.
Series and report cases
Non peer-reviewed articles

In order to evaluate the eligibility and inclusion/exclusion criteria for the articles that resulted from the search, titles, abstracts and part of the discussion were reviewed.

COLLECTED DATA

The main type of data extracted from each study was the genetic markers deemed to be associated with worse evolution of COVID-19. Other relevant data were extracted as well, such as ethnic group, number of individuals evaluated in each study and technology used for sequencing or genotyping (see Table 3). The number of papers in which each marker was deemed relevant was evaluated, with the ponderation of the statistical significance of the correlation among the complete set of markers and the severity of COVID-19, using Odds ratios (OR), 95% confidence intervals (CI) and the corresponding p-value. ORs calculations were checked using R language [32] -epitools package [33]- as well as CI and p-value, taking into account the number of severe and non severe cases, with or without each variant of interest. (see Table 4).

BIASES

One of the possible biases in the present analysis is ethnic origin, since the studies mostly include individuals of European ancestry, a phenomenon which may be seen in most studies of multigenic and multifactorial phenotypes [34]. The analysis includes any studies that fulfill inclusion criteria regardless of the ethnic background of the patients evaluated, and if a marker has been postulated by more than one study, a separate analysis by ethnic group would be performed, in order to solve such bias.

EVALUATION OF THE METHODOLOGICAL QUALITY OF THE STUDIES

The Newcastle-Ottawa Scale (NOS) was applied so as to evaluate the methodological quality of the studies included in the present systematic review [35].

RESULTS

The applied search strategy allowed us to identify 207 articles of interest, published until May 2021.

After applying the inclusion and exclusion criteria, 6 papers were considered relevant for the analysis. The review of the references cited by these articles revealed 1 additional paper that was subsequently added to the review (see the flow diagram in Figure 1).

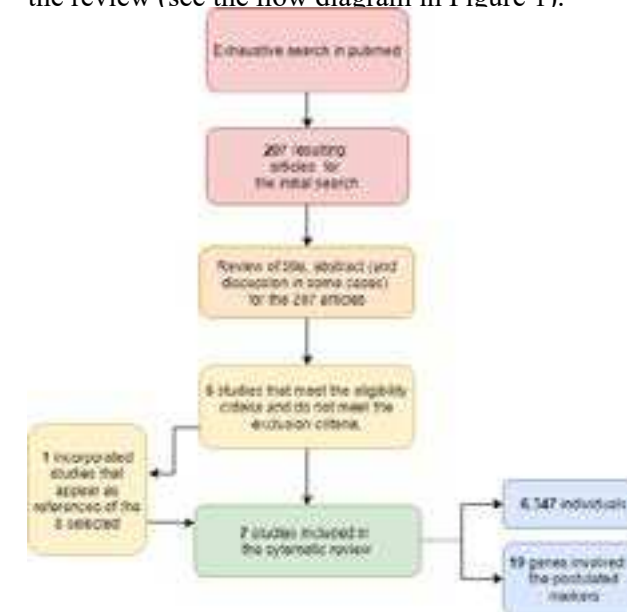


FIGURE 1. FLOW DIAGRAM FOR THE SYSTEMATIC REVIEW

The quality of the included papers was evaluated using the NOS [35]. Table 5 presents the articles included in the review, in a descending order by their score.

TABLE 5. NOS FOR EACH INCLUDED PAPER

Primer Autor	Escala Newcastle-Ottawa (NOS)
David Ellinghaus [36]	8
Eleni Gavrilaki [37]	8
Sushma Verma [38]	7
JuanGómez [39]	7
Jianchang Hu [40]	6
Jihad G. Youssef [41]	5
Yonghong Zhang [42]	5

Out of the 7 included papers, the one with the highest number of patients included 3815 individuals, whereas the one with the lowest number evaluated just 13, with a median number of 81 for the whole set of papers. The total number of included patients was 6.347. Several polymorphic markers and genes were identified as possibly associated with severe forms of COVID-19. However, great heterogeneity was seen among the studies. Table 6 compares the postulated markers from each study.

An additional set of papers was identified that did not fully satisfy inclusion criteria, mainly due to the comparison of severe cases with controls taken from biobanks that had not been tested for the disease (nor was there information available regarding severity of the disease, had they been infected). However, some of them have identified potential markers of severity and should therefore be regarded as relevant [43,44,45,46].

Publicación	Título	Países	Revisión	Sexo	Edad	Control	Genes	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs	SNPs
David Ellinghaus [36]	Genome-wide association study of severe COVID-19 with transcriptome-wide analysis	USA, UK, Spain, Germany, France, Italy, Netherlands, Belgium, Sweden, Denmark, Norway, Finland, Iceland, Ireland, Portugal, Greece, Austria, Czech Republic, Slovakia, Hungary, Poland, Slovenia, Croatia, Serbia, Bosnia and Herzegovina, Montenegro, Albania, Kosovo, Macedonia, Bulgaria, Romania, Moldova, Ukraine, Georgia, Armenia, Azerbaijan, Kazakhstan, Kyrgyzstan, Uzbekistan, Turkmenistan, Tajikistan, China, India, Pakistan, Bangladesh, Nepal, Sri Lanka, Myanmar, Thailand, Cambodia, Laos, Vietnam, Philippines, Malaysia, Singapore, Indonesia, Brunei, Timor-Leste, East Timor, Papua New Guinea, Vanuatu, Solomon Islands, Fiji, Tonga, Samoa, Tokelau, Niue, Cook Islands, Tonga, Samoa, Tokelau, Niue, Cook Islands	2021	2021	11,400	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200	1,200
Eleni Gavrilaki [37]	Genetic architecture of severe COVID-19 in Long Island region	USA	2021	2021	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000
Sushma Verma [38]	Genetic architecture of severe COVID-19 in Long Island region	USA	2021	2021	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000
JuanGómez [39]	Genetic architecture of severe COVID-19 in Long Island region	USA	2021	2021	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000
Jianchang Hu [40]	Genetic architecture of severe COVID-19 in Long Island region	USA	2021	2021	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000
Jihad G. Youssef [41]	Genetic architecture of severe COVID-19 in Long Island region	USA	2021	2021	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000
Yonghong Zhang [42]	Genetic architecture of severe COVID-19 in Long Island region	USA	2021	2021	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000	1,000

TABLE 3. DATA EXTRACTED FROM EACH PAPER INCLUDED IN THE REVIEW

- ta-Analysis, and Meta-Regression Analysis. *Clin Infect Dis*. 2020 11 19;71(16):2199-206.
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