



THE UNIVERSITY *of* EDINBURGH

Edinburgh Research Explorer

Genetically predicted high IGF-1 levels showed protective effects on COVID-19 susceptibility and hospitalization

a Mendelian randomisation study with data from 60 studies across 25 countries

Citation for published version:

Li, X, Zhou, Y, Yuan, S, Zhou, X, Wang, L, Sun, J, Yu, L, Zhu, J, Zhang, H, Yang, N, Dai, S, Song, P, Larsson, SC, Theodoratou, E, Zhu, Y & Li, X 2022, 'Genetically predicted high IGF-1 levels showed protective effects on COVID-19 susceptibility and hospitalization: a Mendelian randomisation study with data from 60 studies across 25 countries', *eLIFE*, vol. 11, e79720. <https://doi.org/10.7554/eLife.79720>

Digital Object Identifier (DOI):

[10.7554/eLife.79720](https://doi.org/10.7554/eLife.79720)

Link:

[Link to publication record in Edinburgh Research Explorer](#)

Document Version:

Peer reviewed version

Published In:

eLIFE

General rights

Copyright for the publications made accessible via the Edinburgh Research Explorer is retained by the author(s) and / or other copyright owners and it is a condition of accessing these publications that users recognise and abide by the legal requirements associated with these rights.

Take down policy

The University of Edinburgh has made every reasonable effort to ensure that Edinburgh Research Explorer content complies with UK legislation. If you believe that the public display of this file breaches copyright please contact openaccess@ed.ac.uk providing details, and we will remove access to the work immediately and investigate your claim.



1 **Title: Genetically predicted high IGF-1 levels showed protective effects on**
2 **COVID-19 susceptibility and hospitalization: A Mendelian Randomisation study**
3 **with data from 60 studies across 25 countries**

4 **Authors:** Xinxuan Li¹, Yajing Zhou¹, Shuai Yuan^{1,2}, Xuan Zhou¹, Lijuan Wang¹, Jing
5 Sun¹, Lili Yu¹, Jinghan Zhu³, Han Zhang¹, Nan Yang¹, Shuhui Dai¹, Peige Song⁴,
6 Susanna C. Larsson^{2,5}, Evropi Theodoratou^{6,7†}, Yimin Zhu^{†1}, Xue Li^{†1}

7 **Affiliations**

8 ¹ Department of Big Data in Health Science School of Public Health, Center of Clinical Big
9 Data and Analytics of The Second Affiliated Hospital, Zhejiang University School of
10 Medicine, Hangzhou, China

11 ² Unit of Cardiovascular and Nutritional Epidemiology, Institute of Environmental Medicine,
12 Karolinska Institutet, Stockholm, Sweden

13 ³ The Second School of Clinical Medicine, Southern Medical University, Guangzhou, China

14 ⁴ School of Public Health and Women's Hospital, Zhejiang University School of Medicine,
15 Hangzhou, China

16 ⁵ Unit of Medical Epidemiology, Department of Surgical Sciences, Uppsala University,
17 Uppsala, Sweden

18 ⁶ Centre for Global Health, Usher Institute, University of Edinburgh, Edinburgh, UK

19 ⁷ Cancer Research UK Edinburgh Centre, Medical Research Council Institute of Genetics
20 and Cancer, University of Edinburgh, Edinburgh, UK

21 **† Joint last authors**

22 Correspondence to:

23 Xue Li, School of Public Health and the Second Affiliated Hospital, Zhejiang University
24 School of Medicine, Hangzhou, China, email: xueli157@zju.edu.cn

25 Word count of abstract:242 / manuscript: 2575

26 Numbers of tables/ figures: 5/ 2

27 Numbers of references: 43

28

29 **Abstract**

30 **Background :** Epidemiological studies observed gender differences in COVID-19
31 outcomes, however, whether sex hormone plays a causal in COVID-19 risk remains
32 unclear. This study aimed to examine associations of sex hormone, sex hormones-binding
33 globulin (SHBG), insulin-like growth factor-1 (IGF-1) and COVID-19 risk.

34 **Methods:** Two-sample Mendelian randomization (TSMR) study was performed to explore
35 the causal associations between testosterone, estrogen, SHBG, IGF-1 and the risk of
36 COVID-19 (susceptibility, hospitalization, and severity) using GWAS summary level data
37 from the COVID-19 Host Genetics Initiative (N=1,348,701). Random-effects inverse
38 variance weighted (IVW) MR approach was used as the primary MR method and the
39 weighted median, MR-Egger, and MR-PRESSO test were conducted as sensitivity
40 analyses.

41 **Results:** Higher genetically predicted IGF-1 levels have nominally significant association
42 with reduced risk of COVID-19 susceptibility and hospitalization. For one standard
43 deviation increase in genetically predicted IGF-1 levels, the odds ratio was 0.77 (95%
44 confidence interval [CI], 0.61-0.97; $P=0.027$) for COVID-19 susceptibility, 0.62 (95%CI:
45 0.25-0.51; $P=0.018$) for COVID-19 hospitalization, and 0.85 (95%CI: 0.52-1.38, $P=0.513$)
46 for COVID-19 severity. There was no evidence that testosterone, estrogen, SHBG are
47 associated with the risk of COVID-19 susceptibility, hospitalization, and severity in either
48 overall or sex-stratified TSMR analysis.

49 **Conclusions:** Our study indicated that genetically predicted high IGF-1 levels were
50 associated with decrease the risk of COVID-19 susceptibility and hospitalization, but these
51 associations did not survive the Bonferroni correction. Further studies are needed to
52 validate the findings and explore whether IGF-1 could be a potential intervention target to
53 reduce COVID-19 risk.

54 **Funding:** We acknowledge support from NSFC (LR22H260001), CRUK (C31250/A22804),
55 SHLF (Hjärt-Lungfonden, 20210351), VR (Vetenskapsrådet, 2019-00977), and SCI
56 (Cancerfonden).

57 **Key words:** Sex hormones, IGF-1, COVID-19, Mendelian randomization.

58 **Introduction**

59 The COVID-19 pandemic has emerged as the most important health concern across the
60 globe since December 2019. A notable finding that has been noted in many affected
61 countries is a male predominance of COVID-19 related hospitalization and death.(1, 2)
62 Globally, more than 60% of deaths from COVID-19 are reported in males.(3) This
63 epidemiological pattern indicates the need for urgent public health actions, as well as for
64 further investigations on the contributing factors of sex differences in COVID-19 risk and
65 its underlying biological mechanisms.

66 Sex hormones play important roles in the immune response in which estrogen was thought
67 to be immune boosting and testosterone to be immunosuppressing.(4) Due to the higher
68 levels of testosterone in male than female, it has been hypothesized that testosterone
69 might be a promoter of SARS - CoV - 2 infection and progression in males, considering
70 the regulatory effect of androgen receptor (AR) and testosterone on the transcription of a
71 transmembrane protease serine 2, which is a critical factor enabling cellular infection by
72 coronaviruses, including SARS - CoV - 2. (2, 5, 6) Estrogen has been shown not only to
73 enhance immunological markers and response, but also to be linked to T-cell proliferation,
74 which might be involved in the immune response to the infection of SARS-CoV-2.(7) Most
75 hormone (about 60%) is tightly bound to sex hormone-binding globulin (SHBG), which is
76 an important regulator of the bioactivities of estrogens and testosterone.(8, 9) In addition,
77 sex hormone signaling could also regulate the insulin-like growth factor (IGF-1)
78 concentrations, which were also reported to be associated with acute respiratory distress
79 syndrome.(10) It is therefore hypothesized that sex hormone and its related biomarkers
80 might contribute to the sex difference of COVID-19 outcomes. A number of observational
81 studies examined the associations between sex hormones and COVID-19 risk, however,
82 the causality of these associations remains unestablished due to potential limitations of
83 observational studies (e.g., residual confounding and reverse causality) and lack of high-
84 quality data from randomized trials.(11)

85 Mendelian randomization (MR) analysis is an epidemiological approach that can
86 strengthen the casual inference by utilizing genetic variants as instrumental variables to
87 mimic biological effects of related biomarkers (12). Here, we conducted a two-sample MR
88 study to explore the causal associations testosterone, estrogen, SHBG, and IGF-1 with the
89 risk of COVID-19 (susceptibility, hospitalization, and severity) using GWAS summary level
90 data from the COVID-19 Host Genetics Initiative (COVID-19 HGI). Sex-stratified MR
91 analyses for testosterone and estradiol were further performed to explore the associations
92 in males and females separately.

93

94 **Materials and Methods**

95 **Study Design**

96 We firstly conducted a TSMR analysis to explore the causal links between testosterone,
97 estrogen, SHBG, IGF-1 and the risk of COVID-19 (susceptibility, hospitalization, and
98 severity), based on GWAS summary level data from COVID-19 Host Genetics Initiative
99 (COVID-19 HGI). We then performed sex-stratified MR analysis to further examine the
100 associations between genetically determined circulating levels of testosterone and
101 estrogen and COVID-19 outcomes in males and females separately. The design of this
102 study is explained in **Figure 1**.

103 **Genetic instruments of testosterone, estradiol, SHBG, and IGF-1**

104 Single-nucleotide polymorphisms (SNPs) associated with testosterone, estradiol, SHBG,
105 and IGF-1 levels were identified from genome-wide association analyses in up to 425,097
106 participants of European ancestry.(13, 14) Sex-stratified SNPs related to estradiol were
107 obtained from a GWAS including 147,690 males and 163,985 females in UK Biobank. (15)
108 We restricted the analysis to SNPs in linkage equilibrium which were identified in the
109 relevant GWAS at $P < 5 \times 10^{-8}$ clumped on $r^2 = 0.01$ within 10,000 kb using the 1000
110 genomes reference panel(16) to ensure sufficient statistical effectiveness. Among those
111 pairs of SNPs that had LD r^2 above the specified threshold ($r^2 = 0.01$) only the SNP with
112 the lower P value would be retained. SNPs absent from the LD reference panel were also
113 removed. To test whether there was a weak instrumental variable bias, namely genetic
114 variants selected as instrumental variables had a weak association with exposure, we
115 calculated the F statistic if it is much greater than 10 for the instrument-exposure
116 association, the possibility of weak instrumental variable bias is small. These analyses
117 were conducted using the R package “TwoSampleMR”.(17) Consequently, a total of 320,
118 316, 7 and 18 SNPs were used as instrumental variables for SHBG, testosterone, estradiol
119 and IGF-1 respectively. Given that genetic variants predicting testosterone and estradiol
120 levels differ for men and women, we selected sex-specific SNPs for testosterone (130
121 SNPs in males, 151 SNPs in females) and estradiol (10 SNPs in males and females)
122 separately for MR sensitivity analyses. Detailed information on the genetic instruments
123 were provided in the **supplementary file 1a-1d**. We used the STROBE case-control
124 checklist when writing our report.(18)

125 **Data source from COVID-19 Host Genetics Initiative**

126 We obtained the summary level data of COVID-19 susceptibility, hospitalization, and
127 severity from the Host Genetics Initiative (COVID-19-HGI) GWAS meta-analyses of data

128 across 60 studies from 25 countries (Round 5, European population) where UKB data were
129 excluded.(19) The HGI dataset included 1,348,701 participants (32,494 laboratory
130 confirmed cases of SARS-CoV-2 infection and 1,316,207 population controls) for COVID-
131 19 susceptibility, 1,557,411 participants (8316 hospitalized COVID-19 patients and
132 1,549,095 population controls) for COVID-19 hospitalization, and 1,059,456 participants
133 (4792 very severe respiratory confirmed COVID-19 cases and 1,054,664 controls) for
134 COVID-19 severity. COVID-19-HGI defined very severe respiratory confirmed COVID-19
135 cases as patients hospitalized for laboratory-confirmed SARS-CoV-2 infection who died or
136 were given respiratory support. The characteristics of the participants are shown in **Table**
137 **1**.

138 **Two-sample Mendelian randomization analyses**

139 We applied the inverse-variance weighted (IVW) method under the random-effects model
140 as the primary MR analysis. We performed sensitivity analyses, including the weighted
141 median, MR-Egger regression, leave-one-out analysis and MR Pleiotropy RESidual Sum
142 and Outlier (MR-PRESSO) methods, to examine the consistency of associations and to
143 detect and correct for potential pleiotropy. The weighted median method was performed to
144 provide unbiased causal estimates if at least 50% instrumental variables were valid.(20)
145 MR-Egger regression was used to observe and correct potential directional pleiotropy,
146 which was assessed by its intercept test.(21) MR-PRESSO method can detect SNP
147 outliers and estimate the association after removal of these outliers. The differences in
148 estimates between before and after outlier removal were examined by the embedded
149 distortion test.(22) Cochran's Q value was used to assess the heterogeneity among
150 estimates of genetic instruments and the p value for intercept in MR-Egger was used to
151 detect horizontal pleiotropy.(21) All statistical analyses were two-sided and performed in R
152 4.0.4 software using the R package Two Sample MR and MR-PRESSO.(17)

153 **Sensitivity analyses**

154 We additionally used the single-nucleotide polymorphism (SNP) rs7173595 in *CYP19A1*
155 gene, which encodes aromatase, an enzyme that converts androgens to estrogens.
156 Rs7173595 has previously been shown to be strongly associated with serum E2 levels in
157 genome-wide association studies (GWAS) of men(13, 23) and postmenopausal women
158 (24). This SNP was also associated with serum E2 in 25,502 premenopausal European
159 women (<50 years of age and not reporting a hysterectomy or that menopause has
160 occurred) in UK Biobank. The associations of serum E2 instrumented by rs7173595 in the
161 *CYP19A1* gene region with COVID-19 outcomes were estimated using the Wald ratio
162 method. We further performed a sensitivity analysis using a list of genetic instruments
163 consisting of 10 correlated SNPs ($r^2 < 0.4$) located in the *IGF-1* gene region (genomic

164 position on build GRCh37/hg19: chromosome 12:102789652-102874341) and associated
165 with IGF-1 levels at the genome-wide significance level. A matrix of linkage disequilibrium
166 among these SNPs was introduced in the MR analysis model. To control potential data
167 confounder, we selected SNPs associated with testosterone, estrogen, SHBG, and IGF-1
168 only, excluding SNPs associated with BMI which is thought to be a causal risk factor for
169 COVID-19(25) at the threshold of 5×10^{-8} in European ancestry samples by querying
170 PhenoScanner.(17) SNPs in estrogen were not exclude because their irrelevance to BMI.

171 **Results**

172 **Table 2** presents the TSMR estimates for the associations between sex hormones, SHBG,
173 IGF-1 and the risk of COVID-19 susceptibility, hospitalization and severity based on the
174 data from HGI. Higher genetically predicted IGF-1 levels have nominally significant
175 association with reduced risk of COVID-19 susceptibility and hospitalization. For one
176 standard deviation increase in genetically predicted IGF-1 levels, the odds ratio was 0.77
177 (95% confidence interval [CI], 0.61-0.97; $P=0.027$) for COVID-19 susceptibility, 0.62
178 (95%CI: 0.25-0.51; $P=0.018$) for COVID-19 hospitalization, and 0.85 (95%CI: 0.52-1.38,
179 $P=0.513$) for COVID-19 severity. Associations of IGF-1 levels with COVID-19 susceptibility
180 and hospitalization were not statistically significant after Bonferroni correction, albeit
181 showing a nominal significance at $P<0.05$. No outlying SNPs were identified by MR-
182 PRESSO analyses. Estimates from the MR-Egger and weighted mode analyses, were in
183 the same direction as those from the IVW analysis (**Figure 2, Figure 2—figure**
184 **supplement 1, Figure 2—figure supplement 2**). The MR-Egger intercept p was 0.614
185 and 0.595 for susceptibility and hospitalization, respectively, indicating the absence of
186 directional pleiotropy. The associations remained directionally consistent in the sensitivity
187 analysis based on SNPs located in the *IGF-1* gene region as instrumental variables with
188 risk of COVID-19 susceptibility (OR=0.99, 95%CI: 0.91-1.07, $P=0.777$), hospitalization
189 (OR=0.90; 95%CI: 0.74-1.10, $P=0.645$) and severity (OR=1.01; 95%CI: 0.82-1.24,
190 $P=0.415$) (**Table 3**).

191 In the analyses based on data from the genetic consortia, we found no causal associations
192 of genetically predicted testosterone with the risk of COVID-19 susceptibility (OR=0.94;
193 95%CI: 0.83-1.06, $P=0.309$), hospitalization (OR=0.82; 95%CI: 0.64-1.04, $P=0.103$), risk
194 of severity (OR=0.83; 95%CI: 0.60-1.15, $P=0.256$). Null association was also noticed
195 between SHBG and COVID-19 susceptibility (OR=0.91; 95%CI: 0.80-1.04, $P=0.182$),
196 hospitalization (OR=0.86; 95%CI: 0.66-1.11, $P=0.255$), risk of severity (OR=0.92; 95%CI:
197 0.65-1.29, $P=0.618$). Overall, no significant associations between testosterone, estrogen,
198 SHBG and COVID-19 outcomes were observed from two-sample MR analyses. Sex-
199 specific associations of genetically testosterone and estradiol levels with COVID-19 risk
200 (**Table 4**) were still nonsignificant. We noticed the P for intercept in MR-Egger regression
201 analysis was more than 0.05 for both genders, and no outlier was detected. Genetic
202 predisposition to higher serum E2 levels proxied by rs7173595 in the *CYP19A1* gene was
203 not associated with the risk of COVID-19 susceptibility (OR =0.32; 95% CI, 0.06-1.80; $P =$
204 0.195), hospitalization (OR=0.28; 95%CI: 0.01-6.46, $P=0.426$) and severity (OR=0.22;
205 95%CI: 0.00-12.73, $P=0.469$) in females; similarly, the associations remained directionally
206 consistent in males with susceptibility (OR =0.37; 95% CI, 0.08-1.67; $P = 0.195$),
207 hospitalization (OR=0.33; 95%CI: 0.02-5.11, $P=0.426$) and severity (OR=0.27; 95%CI:

208 0.01-9.26, $P=0.469$) (**Table 5**). As shown in **Table 6**, after removing SNPs associated with
209 BMI, we found similar associations of genetically predicted IGF-1 levels with the risk of
210 COVID-19 susceptibility (OR=0.76; 95%CI: 0.60-0.96, $P=0.021$), hospitalization (OR=0.61;
211 95%CI: 0.41-0.90, $P=0.014$), risk of severity (OR=0.84; 95%CI: 0.52-1.38, $P=0.497$) in
212 which we detected no moderate heterogeneity, and no indication of horizontal pleiotropy in
213 MR-Egger, and no outlier in MR-PRESSO analyses. No causal associations of genetically
214 predicted testosterone and SHBG with COVID-19 were found, but the directions were
215 consistent with results in Table 2.

216 **Discussion**

217 In this study, we assessed whether there were any causal associations between sex
218 hormone related biomarkers and the risk of COVID-19 outcomes. We found suggestive
219 evidence for associations between genetic liability to high IGF-1 levels and decreased risk
220 of COVID-19 susceptibility and hospitalization. Our findings suggest a potential role of IGF-
221 1 in COVID-19 risk and have implications for tailored treatment of COVID-19 patients.

222 Our MR findings were consistent with the multiple epidemiological studies that reported a
223 nominal association between measured IGF-1 levels and COVID-19 illness. There is one
224 observational study that demonstrated an inverse association between pre-diagnostic
225 circulating levels of IGF-1 and COVID-19 mortality risk among COVID-19 patients in UK
226 biobank.(26) Another observational study in Greece reported lower IGF-1 levels in critically
227 ill COVID-19 patients compared to their counterparts with less severe disease or without
228 COVID-19.(27) A single-cell analysis revealed that the exhaustion of CD8⁺ T cells together
229 with several cytokines including IGF-1 was associated with the pathogenesis of severe
230 SARS-CoV-2 infection.(28) Our MR analyses found a negative association between
231 genetically determined high circulating IGF-1 levels and decreased risk of COVID-19
232 susceptibility and hospitalization, indicating IGF-1 may be a protective factor of COVID-19
233 risk.

234 IGF-1 has been found to be pro-survival/anti-aging, anti-inflammatory, and antioxidant with
235 neuro- and hepatoprotective properties. A study by the Narasaraaju group demonstrated
236 that IGF-1 plays an important role in the repair of lung tissue by regulating the proliferation
237 and differentiation of alveolar epithelial cells (AECs).(29) Airway inflammation can be
238 mitigated when apoptotic cells are engulfed by pulmonary epithelial cells.(30) IGF-1 has
239 also been shown to up-regulate engulfment by professional phagocytes such as dendritic
240 cells,(31) and inhibit IL-6 production from lipopolysaccharide (LPS)-induced AECs. (32)
241 Both of these mechanisms are beneficial to the regression of local inflammation. Jakn *et*
242 *al.* showed that IGF-1 binds to insulin-like growth factor-1 receptor (IGF-1R) on airway
243 epithelial cells of non-professional phagocytic cells, which can promote the phagocytosis
244 of microparticles by airway epithelial cells.(33) Transforming growth factor β 1 (TGF- β 1)
245 derived from AECs activated alveolar macrophages (AMs) to secrete IGF-1 into the
246 alveolar fluid in response to stimulation of the airway by inflammatory signals. This AM-
247 derived IGF-1 attenuated the p38 mitogen-activated protein kinase (MAPK) inflammatory
248 signal in AECs and promoted the phagocytosis of apoptotic cells by AECs. This two-way
249 communication between AECs and AMs represents a well-tuned system for the regulation
250 of the inflammatory response in alveoli.(34) Taken together, these studies provide
251 biological evidence supporting that IGF-1 might be an important anti-inflammatory factor in

252 the alveolar microenvironment and thus may contribute to improve COVID-19 outcomes.
253 More studies are required to determine whether novel therapeutic strategy targeting on
254 IGF-1 pathway might improve COVID-19 prognosis.

255 IGF-1 level is regulated by estrogen and the functional interactions between estradiol and
256 IGF-1 signaling system involve several transcriptional and posttranscriptional mechanisms.
257 Specifically, IGF-1 can affect estrogen receptor α (ER α) action by enhancing its expression
258 and potentiating its transcriptional activity in a ligand-independent manner.(35-37) On the
259 other hand, E2 can enhance IGF-1 signaling by upregulating the expression of IGF-1,(38)
260 IGF-1 receptor,(39) and some IGF-1 binding proteins.(40) This may explain the same
261 direction from the IVW analysis of IGF-1, estradiol and COVID-19 outcomes. Estrogen is
262 found to have immune enhancing effect(7) to trigger the local immune response by
263 activating a plethora of cells such as phagocytes, dendritic cells, natural killers, and CD8⁺
264 T cells. Once these immune cells are activated, they could fight against the infection by
265 destroying the virus and thus preventing its diffusion to the lower respiratory tract or by
266 decreasing the viral load. Experimental tests have also reported that estradiol can affect
267 angiotensin-converting enzyme 2 (ACE2) and FURIN expression, with the potential of
268 mitigating SARS-CoV-2 infection.(41) However, our study did not find any supportive
269 evidence for the associations between estradiol and COVID-19, which might be due to the
270 small variance of estradiol explained by genetic instruments.

271 Our studies showed that SHBG or testosterone may not be associated with COVID-19
272 outcomes, which is consistent with the research findings of Luna Liu *et al.*(42) They also
273 observed a null causal relationship for testosterone or SHBG levels with COVID-19
274 outcomes in females and males. Meanwhile, epidemiologic data (2) indicate that while men
275 are not more predisposed to contracting COVID-19, they are more likely to develop severe
276 illness following the infection compared with women. However, our study observed null
277 causal relationship for testosterone levels with COVID-19 outcomes in both females and
278 males. According to the available evidence on the role of testosterone in COVID-19, it
279 appears that both high and low testosterone levels can be associated with poor COVID-19
280 outcomes.(43) A study demonstrated androgen deprivation therapy (ADT) exposure was
281 associated with a reduction in COVID-19 severity.(44) By contrast, the Ohio study did not
282 identify any protective effect of ADT on the severity of COVID-19 outcomes.(45) Androgen-
283 related treatments showed that transmembrane serine protease 2 (TMPRSS2) expression
284 and SARS-CoV-2 entry in human lung cells have been reduced by antiandrogens.(46-48)
285 Additionally, androgens have numerous immunosuppressive effects such as decreasing
286 proinflammatory cytokine release (e.g., IFN γ and TNF) or increasing anti-inflammatory
287 cytokine release (e.g., IL-4 and IL-10), reducing T helper 1 (Th1) and T helper 17 (Th17)
288 cell differentiation, inducing Treg differentiation and regulating B-cell development.(49-51)

289 Paradoxically, these immunosuppressive effects of testosterone might be beneficial to
290 overcome the heightened inflammatory environment that predisposes to severe COVID-
291 19. Recent research has revealed that males with COVID-19 have lower testosterone
292 levels.(52) Another study found a negative association between total testosterone levels
293 and biochemical markers of COVID-19 severity.(53) Lower testosterone concentrations
294 were associated with higher concentrations of IL-6, CRP, IL-1 receptor antagonist,
295 hepatocyte growth factor, and IFN γ -inducible protein 10.(54) Therefore additional research
296 efforts need to be made to investigate the complex relationships furtherly.

297 The major advantage of our study is the design taking the advantages of MR approach and
298 used several sensitivity analyses to test the robustness of the MR findings. The application
299 of MR analysis reduces the influence of confounding factors and reverse causality so that
300 reliable causal estimations were obtained to complement the observational findings. The
301 potential limitations of this study also need to be acknowledged. Our study may suffer from
302 weak instrument bias, especially within sensitivity analyses that restricted to smaller sets
303 of genetic instruments. In two-sample MR, this bias would tend to make estimates closer
304 to the null. Since there is no available data on recovery status for COVID-19 patients in UK
305 biobank, the current study did not take recovery as a potential competing risk into account.
306 We could not assess the sex-specific associations in IGF-1 and COVID-19 due to no data
307 by sex in HGI. Moreover, the MR was merely based on individuals of European ancestry.
308 Our findings might not be generalized to other populations. It should also be noted that the
309 study findings are based on evidence from genetic data, additional large and prospective
310 cohort studies with available IGF-1 data and information on COVID-19 susceptibility and
311 clinical outcomes are needed to validate the findings.

312 In conclusion, our study indicated that genetically predicted high IGF-1 levels were
313 associated with decrease the risk of COVID-19 susceptibility and hospitalization, but these
314 associations did not survive the Bonferroni correction of multiple testing. Further studies
315 are needed to validate the findings and explore whether IGF-1 could be a potential
316 intervention target to reduce COVID-19 risk.

317

Acknowledgments

The authors are thankful for all the participants that contributed to the UK Biobank study.

Funding Statement

The funders had no role in study design, data collection and interpretation, or the decision to submit the work for publication

Funding Information

This paper was supported by the following grants:

Natural Science Fund for Distinguished Young Scholars of Zhejiang Province (LR22H260001) to Xue Li.

CRUK Career Development Fellowship(C31250/A22804) to Evropi Theodoratou.

the Swedish Heart Lung Foundation (Hjärt-Lungfonden, 20210351), the Swedish Research Council (Vetenskapsrådet, 2019-00977), and the Swedish Cancer Society (Cancerfonden) to Susanna C. Larsson.

Data availability statement

Data analyzed in the present study are GWAS summary statistics, which have been made publicly available. GWAS summary level data of COVID-19-HGI could be downloaded from <https://www.covid19hg.org/results/>. GWAS summary level data of sex hormones and IGF-1 in UK biobank could be downloaded from GWAS catalog (http://ftp.ebi.ac.uk/pub/databases/gwas/summary_statistics/GCST90019001-GCST90020000/). All genome-wide significant SNPs have been provided in supplementary 1a to 1d. All analyses were performed using R statistical package freely available at <https://cran.r-project.org/mirrors.html>. The Two-sample MR package is available at <https://mrcieu.github.io/TwoSampleMR/>.

References

1. Grasselli G, Zangrillo A, Zanella A, Antonelli M, Cabrini L, Castelli A, et al. Baseline Characteristics and Outcomes of 1591 Patients Infected With SARS-CoV-2 Admitted to ICUs of the Lombardy Region, Italy. *Jama*. 2020;323(16):1574-81.
2. Peckham H, de Gruijter NM, Raine C, Radziszewska A, Ciurtin C, Wedderburn LR, et al. Male sex identified by global COVID-19 meta-analysis as a risk factor for death and ITU admission. *Nat Commun*. 2020;11(1):6317.
3. Richardson S, Hirsch JS, Narasimhan M, Crawford JM, McGinn T, Davidson KW, et al. Presenting Characteristics, Comorbidities, and Outcomes Among 5700 Patients Hospitalized With COVID-19 in the New York City Area. *Jama*. 2020;323(20):2052-9.
4. Strobe JD, Chau CH, Figg WD. Are sex discordant outcomes in COVID-19 related to sex hormones? *Semin Oncol*. 2020;47(5):335-40.
5. Pozzilli P, Lenzi A. Commentary: Testosterone, a key hormone in the context of COVID-19 pandemic. *Metabolism*. 2020;108:154252.
6. Cattrini C, Bersanelli M, Latocca MM, Conte B, Vallome G, Boccardo F. Sex Hormones and Hormone Therapy during COVID-19 Pandemic: Implications for Patients with Cancer. *Cancers (Basel)*. 2020;12(8).
7. Taneja V. Sex Hormones Determine Immune Response. *Front Immunol*. 2018;9:1931.
8. Raverot V, Lopez J, Grenot C, Pugeat M, Déchaud H. New approach for measurement of non-SHBG-bound testosterone in human plasma. *Anal Chim Acta*. 2010;658(1):87-90.
9. Dimou N, Mori N, Harlid S, Harbs J, Martin RM, Smith-Byrne K, et al. Circulating Levels of Testosterone, Sex Hormone Binding Globulin and Colorectal Cancer Risk: Observational and Mendelian Randomization Analyses. *Cancer Epidemiol Biomarkers Prev*. 2021.
10. Ahasic AM, Zhai R, Su L, Zhao Y, Aronis KN, Thompson BT, et al. IGF1 and IGFBP3 in acute respiratory distress syndrome. *Eur J Endocrinol*. 2012;166(1):121-9.
11. Tsang G, Insel MB, Weis JM, Morgan MA, Gough MS, Frasier LM, et al. Bioavailable estradiol concentrations are elevated and predict mortality in septic patients: a prospective cohort study. *Crit Care*. 2016;20(1):335.
12. Burgess S, Thompson SG. Mendelian randomization: methods for using genetic variants in causal estimation: CRC Press; 2015.
13. Ruth KS, Day FR, Tyrrell J, Thompson DJ, Wood AR, Mahajan A, et al. Using human genetics to understand the disease impacts of testosterone in men and women. *Nat Med*. 2020;26(2):252-8.
14. Sinnott-Armstrong N, Tanigawa Y, Amar D, Mars N, Benner C, Aguirre M, et al. Genetics of 35 blood and urine biomarkers in the UK Biobank. *Nat Genet*. 2021;53(2):185-94.
15. Schmitz D, Ek WE, Berggren E, Höglund J, Karlsson T, Johansson Å. Genome-wide Association Study of Estradiol Levels and the Causal Effect of Estradiol on Bone Mineral Density. *J Clin Endocrinol Metab*. 2021;106(11):e4471-e86.
16. Hemani G, Zheng J, Elsworth B, Wade KH, Haberland V, Baird D, et al. The MR-Base platform supports systematic causal inference across the human phenome. *Elife*. 2018;7.
17. Yavorska OO, Burgess S. MendelianRandomization: an R package for performing Mendelian randomization analyses using summarized data. *Int J Epidemiol*. 2017;46(6):1734-9.
18. von Elm E, Altman DG, Egger M, Pocock SJ, Gøtzsche PC, Vandenbroucke JP. The Strengthening the Reporting of Observational Studies in Epidemiology (STROBE) Statement: guidelines for reporting observational studies. *Int J Surg*. 2014;12(12):1495-9.
19. The COVID-19 Host Genetics Initiative, a global initiative to elucidate the role of host genetic factors in susceptibility and severity of the SARS-CoV-2 virus pandemic. *Eur J Hum Genet*. 2020;28(6):715-8.
20. Bowden J, Davey Smith G, Haycock PC, Burgess S. Consistent Estimation in Mendelian Randomization with Some Invalid Instruments Using a Weighted Median Estimator. *Genet Epidemiol*. 2016;40(4):304-14.
21. Bowden J, Davey Smith G, Burgess S. Mendelian randomization with invalid instruments: effect estimation and bias detection through Egger regression. *Int J Epidemiol*. 2015;44(2):512-25.
22. Wu F, Huang Y, Hu J, Shao Z. Mendelian randomization study of inflammatory bowel disease and bone mineral density. *BMC Med*. 2020;18(1):312.
23. Eriksson AL, Perry JRB, Coviello AD, Delgado GE, Ferrucci L, Hoffman AR, et al. Genetic Determinants of Circulating Estrogen Levels and Evidence of a Causal Effect of Estradiol on Bone Density in Men. *J Clin Endocrinol Metab*. 2018;103(3):991-1004.

24. Thompson DJ, O'Mara TA, Glubb DM, Painter JN, Cheng T, Folkerd E, et al. CYP19A1 fine-mapping and Mendelian randomization: estradiol is causal for endometrial cancer. *Endocr Relat Cancer*. 2016;23(2):77-91.
25. Freuer D, Linseisen J, Meisinger C. Impact of body composition on COVID-19 susceptibility and severity: A two-sample multivariable Mendelian randomization study. *Metabolism*. 2021;118:154732.
26. Fan X, Yin C, Wang J, Yang M, Ma H, Jin G, et al. Pre-diagnostic circulating concentrations of insulin-like growth factor-1 and risk of COVID-19 mortality: results from UK Biobank. *Eur J Epidemiol*. 2021;36(3):311-8.
27. Ilias I, Diamantopoulos A, Botoula E, Athanasiou N, Zacharis A, Tsiplis S, et al. Covid-19 and Growth Hormone/Insulin-Like Growth Factor 1: Study in Critically and Non-Critically Ill Patients. *Front Endocrinol (Lausanne)*. 2021;12:644055.
28. He L, Zhang Q, Zhang Y, Fan Y, Yuan F, Li S. Single-cell analysis reveals cell communication triggered by macrophages associated with the reduction and exhaustion of CD8(+) T cells in COVID-19. *Cell Commun Signal*. 2021;19(1):73.
29. Narasaraju TA, Chen H, Weng T, Bhaskaran M, Jin N, Chen J, et al. Expression profile of IGF system during lung injury and recovery in rats exposed to hyperoxia: a possible role of IGF-1 in alveolar epithelial cell proliferation and differentiation. *J Cell Biochem*. 2006;97(5):984-98.
30. Juncadella IJ, Kadl A, Sharma AK, Shim YM, Hochreiter-Hufford A, Borish L, et al. Apoptotic cell clearance by bronchial epithelial cells critically influences airway inflammation. *Nature*. 2013;493(7433):547-51.
31. Xuan NT, Hoang NH, Nhung VP, Duong NT, Ha NH, Hai NV. Regulation of dendritic cell function by insulin/IGF-1/PI3K/Akt signaling through klotho expression. *J Recept Signal Transduct Res*. 2017;37(3):297-303.
32. Wang H, He J, Luo Y, Mu M, Guo S, Shen L, et al. IGF-1 Promotes Endocytosis of Alveolar Epithelial Cells through PI3K Signaling. *Ann Clin Lab Sci*. 2019;49(1):3-8.
33. Han CZ, Juncadella IJ, Kinchen JM, Buckley MW, Klivanov AL, Dryden K, et al. Macrophages redirect phagocytosis by non-professional phagocytes and influence inflammation. *Nature*. 2016;539(7630):570-4.
34. Mu M, Gao P, Yang Q, He J, Wu F, Han X, et al. Alveolar Epithelial Cells Promote IGF-1 Production by Alveolar Macrophages Through TGF- β to Suppress Endogenous Inflammatory Signals. *Front Immunol*. 2020;11:1585.
35. Lange CA. Making sense of cross-talk between steroid hormone receptors and intracellular signaling pathways: who will have the last word? *Mol Endocrinol*. 2004;18(2):269-78.
36. Edwards DP, Weigel NL, Nordeen SK, Beck CA. Modulators of cellular protein phosphorylation alter the trans-activation function of human progesterone receptor and the biological activity of progesterone antagonists. *Breast Cancer Res Treat*. 1993;27(1-2):41-56.
37. Shupnik MA. Crosstalk between steroid receptors and the c-Src-receptor tyrosine kinase pathways: implications for cell proliferation. *Oncogene*. 2004;23(48):7979-89.
38. Umayahara Y, Kawamori R, Watada H, Imano E, Iwama N, Morishima T, et al. Estrogen regulation of the insulin-like growth factor I gene transcription involves an AP-1 enhancer. *J Biol Chem*. 1994;269(23):16433-42.
39. Bartucci M, Morelli C, Mauro L, Andò S, Surmacz E. Differential insulin-like growth factor I receptor signaling and function in estrogen receptor (ER)-positive MCF-7 and ER-negative MDA-MB-231 breast cancer cells. *Cancer Res*. 2001;61(18):6747-54.
40. Qin C, Singh P, Safe S. Transcriptional activation of insulin-like growth factor-binding protein-4 by 17 β -estradiol in MCF-7 cells: role of estrogen receptor-Sp1 complexes. *Endocrinology*. 1999;140(6):2501-8.
41. Glinsky GV. Tripartite Combination of Candidate Pandemic Mitigation Agents: Vitamin D, Quercetin, and Estradiol Manifest Properties of Medicinal Agents for Targeted Mitigation of the COVID-19 Pandemic Defined by Genomics-Guided Tracing of SARS-CoV-2 Targets in Human Cells. *Biomedicines*. 2020;8(5).
42. Liu L, Fan X, Guan Q, Yu C. Bioavailable testosterone level is associated with COVID-19 severity in female: A sex-stratified Mendelian randomization study. *J Infect*. 2022;85(2):174-211.
43. Ho JQ, Sepand MR, Bigdelou B, Shekarian T, Esfandyarpour R, Chauhan P, et al. The immune response to COVID-19: Does sex matter? *Immunology*. 2022;166(4):429-43.
44. Lee KM, Heberer K, Gao A, Becker DJ, Loeb S, Makarov DV, et al. A Population-Level Analysis of the Protective Effects of Androgen Deprivation Therapy Against COVID-19 Disease Incidence and Severity. *Front Med (Lausanne)*. 2022;9:774773.
45. Klein EA, Li J, Milinovich A, Schold JD, Sharifi N, Kattan MW, et al. Androgen Deprivation

- Therapy in Men with Prostate Cancer Does Not Affect Risk of Infection with SARS-CoV-2. *J Urol*. 2021;205(2):441-3.
46. Leach DA, Mohr A, Giotis ES, Cil E, Isac AM, Yates LL, et al. The antiandrogen enzalutamide downregulates TMPRSS2 and reduces cellular entry of SARS-CoV-2 in human lung cells. *Nat Commun*. 2021;12(1):4068.
47. Deng Q, Rasool RU, Russell RM, Natesan R, Asangani IA. Targeting androgen regulation of TMPRSS2 and ACE2 as a therapeutic strategy to combat COVID-19. *iScience*. 2021;24(3):102254.
48. Qiao Y, Wang XM, Mannan R, Pitchiaya S, Zhang Y, Wotring JW, et al. Targeting transcriptional regulation of SARS-CoV-2 entry factors ACE2 and TMPRSS2. *Proc Natl Acad Sci U S A*. 2020;118(1).
49. Olsen NJ, Kovacs WJ. Evidence that androgens modulate human thymic T cell output. *J Investig Med*. 2011;59(1):32-5.
50. Henze L, Schwinge D, Schramm C. The Effects of Androgens on T Cells: Clues to Female Predominance in Autoimmune Liver Diseases? *Front Immunol*. 2020;11:1567.
51. Trigunaite A, Dimo J, Jørgensen TN. Suppressive effects of androgens on the immune system. *Cell Immunol*. 2015;294(2):87-94.
52. Ma L, Xie W, Li D, Shi L, Ye G, Mao Y, et al. Evaluation of sex-related hormones and semen characteristics in reproductive-aged male COVID-19 patients. *J Med Virol*. 2021;93(1):456-62.
53. Rastrelli G, Di Stasi V, Inglese F, Beccaria M, Garuti M, Di Costanzo D, et al. Low testosterone levels predict clinical adverse outcomes in SARS-CoV-2 pneumonia patients. *Andrology*. 2021;9(1):88-98.
54. Dhindsa S, Zhang N, McPhaul MJ, Wu Z, Ghoshal AK, Erlich EC, et al. Association of Circulating Sex Hormones With Inflammation and Disease Severity in Patients With COVID-19. *JAMA Netw Open*. 2021;4(5):e2111398.

Figure legend

Figure 1. Overall Study Design

Abbreviation: IGF-1, insulin-like growth factor-1; GWAS, genome-wide association study; SNP, single nucleotide polymorphism; LD, linkage disequilibrium; IVW, inverse variance weighting; MR, mendelian randomization.

Figure2. IGF-1 and COVID-19 outcomes in mendelian randomization (MR) analyses

Abbreviation: IGF-1, insulin-like growth factor-1; SNP, single nucleotide polymorphism; IVW, inverse variance weighting; OR, odds ratio; CI, confidence interval

Table 1. Sources of data for Mendelian randomization analysis in COVID-19 HGI

Phenotype	Participants
	Meta-analysis of 35 GWAS performed in individuals of European ancestry
Susceptibility	Cases: 32,494 individuals with COVID-19 by laboratory confirmation, chart review, or self-report Controls: 1,316,207 individuals without confirmation or history of COVID-19
	Meta-analysis of 23 GWAS performed in individuals of European ancestry
Hospitalization	Cases: 8,316 hospitalized individuals with COVID-19 Controls: 1,549,095 individuals without confirmation or history of COVID-19
	Meta-analysis of 14 GWAS performed in individuals of European ancestry
Severity	Cases: 4,792 SARS-CoV-2 infected hospitalized individuals who died or required respiratory support (intubation, CPAP, BiPAP, continuous external negative pressure, high flow nasal cannula). Controls: 1,054,664 individuals without confirmation or history of COVID-19

Notes: COVID-19 outcomes are taken from the COVID-19 HGI.

Abbreviations: HGI, host genetics initiative; GWAS, genome-wide association study; UKB, UK Biobank; CPAP, continuous positive airway pressure ventilation; BiPAP, bilevel positive airway pressure ventilation.

Table 2. Sex hormones, SHBG, IGF-1 and COVID-19 outcomes in mendelian randomization (MR) analyses

Exposure	Method	Susceptibility					Hospitalization					Severity						
		SNPs	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept	SNPs	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept	SNPs	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept		
Testosterone	IVW		0.94 (0.83, 1.06)	0.309	0.006	-		0.82 (0.64, 1.04)	0.103	0.055	-		0.83 (0.60, 1.15)	0.256	0.041	-		
	MR Egger		0.93 (0.76, 1.12)	0.430	0.005	0.860		0.79 (0.55, 1.15)	0.217	0.051	0.819		0.78 (0.48, 1.27)	0.313	0.038	0.732		
	Weighted median	315	0.89 (0.71, 1.12)	0.329	-	-	303	0.81 (0.52, 1.28)	0.370	-	-	316	0.71 (0.40, 1.26)	0.246	-	-		
	Simple mode		1.13 (0.73, 1.77)	0.584	-	-		0.77 (0.27, 2.20)	0.623	-	-		0.44 (0.09, 2.18)	0.316	-	-		
	Weighted mode		0.91 (0.77, 1.08)	0.300	-	-		0.77 (0.52, 1.13)	0.180	-	-		0.65 (0.40, 1.05)	0.081	-	-		
	MR PRESSO		0.94 (1.06, 0.84)	-	-	-		0.82 (1.04, 0.65)	-	-	-		0.83 (1.15, 0.59)	-	-	-		
SHBG	IVW		0.91 (0.80, 1.04)	0.182	0.002	-		0.86 (0.66, 1.11)	0.255	0.087	-		0.92 (0.65, 1.29)	0.618	0.096	-		
	MR Egger		0.96 (0.78, 1.18)	0.708	0.002	0.494		0.83 (0.57, 1.22)	0.352	0.081	0.818		0.92 (0.56, 1.51)	0.730	0.090	0.994		
	Weighted median	319	0.90 (0.72, 1.13)	0.360	-	-	309	0.82 (0.52, 1.29)	0.391	-	-	320	0.72 (0.41, 1.27)	0.255	-	-		
	Simple mode		1.09 (0.66, 1.81)	0.735	-	-		1.18 (0.40, 3.44)	0.767	-	-		1.16 (0.25, 5.41)	0.850	-	-		
	Weighted mode		0.94 (0.78, 1.14)	0.547	-	-		0.81 (0.56, 1.18)	0.279	-	-		0.79 (0.47, 1.33)	0.376	-	-		
	MR PRESSO		0.91 (1.05, 0.80)	-	-	-		0.86 (1.11, 0.67)	-	-	-		0.91 (1.28, 0.65)	-	-	-		
Estradiol	IVW		0.54 (0.15, 1.94)	0.346	0.188	-		0.87 (0.11, 6.70)	0.895	0.769	-		0.50 (0.03, 7.64)	0.620	0.987	-		
	MR Egger		0.73 (0.04, 14.11)	0.845	0.123	0.830		0.34 (0.00, 29.54)	0.657	0.685	0.662		0.04 (0.00, 17.04)	0.345	1.000	0.401		
	Weighted median	7	0.36 (0.10, 1.35)	0.130	-	-	7	0.35 (0.03, 4.21)	0.407	-	-	7	0.30 (0.01, 7.26)	0.458	-	-		
	Simple mode		0.29 (0.03, 2.60)	0.313	-	-		0.71 (0.01, 44.94)	0.875	-	-		0.33 (0.00, 43.56)	0.673	-	-		
	Weighted mode		0.34 (0.07, 1.73)	0.241	-	-		0.38 (0.03, 4.81)	0.482	-	-		0.29 (0.01, 9.43)	0.511	-	-		
	MR PRESSO		0.54 (1.94, 0.15)	-	-	-		0.87 (3.93, 0.19)	-	-	-		0.51 (1.52, 0.17)	-	-	-		
IGF-1	IVW		0.77 (0.61, 0.97)	0.027	0.175	-		0.62 (0.25, 0.51)	0.018	0.715	-		0.85 (0.52, 1.38)	0.513	0.601	-		
	MR Egger		0.84 (0.56, 1.26)	0.408	0.145	0.614		0.72 (0.37, 1.38)	0.336	0.668	0.595		1.45 (0.67, 3.10)	0.358	0.758	0.096		
	Weighted median	16	0.76 (0.57, 1.02)	0.071	-	-	16	0.75 (0.44, 1.28)	0.294	-	-	18	0.76 (0.38, 1.53)	0.446	-	-		
	Simple mode		0.64 (0.39, 1.05)	0.097	-	-		0.66 (0.30, 1.45)	0.318	-	-		0.82 (0.27, 2.47)	0.730	-	-		
	Weighted mode		0.77 (0.58, 1.02)	0.084	-	-		0.71 (0.44, 1.17)	0.199	-	-		0.70 (0.35, 1.38)	0.319	-	-		
	MR PRESSO		0.77 (0.98, 0.61)	-	-	-		0.62 (0.88, 0.43)	-	-	-		0.85 (1.34, 0.54)	-	-	-		

Abbreviations: SNP, single nucleotide polymorphism; OR, odds ratio; CI, confidence interval; IVW, inverse variance weighting; SHBG, sex hormones-binding globulin; IGF-1, insulin-like growth factor-1.

Table 3. Sensitive analysis between serum IGF-1 levels instrumented by 10 SNPs in the IGF-1 gene region and COVID-19 outcomes.

Method	Susceptibility				Hospitalization				Severity			
	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept
IVW	0.99 (0.91, 1.07)	0.777	0.596	-	0.90 (0.74, 1.10)	0.645	0.104	-	1.01 (0.82, 1.24)	0.415	0.437	-
MR Egger	0.99 (0.93, 1.05)	0.732	0.541	0.527	0.97 (0.84, 1.11)	0.338	0.108	0.375	1.09 (0.92, 1.30)	0.953	0.372	0.590
Weighted median	1.01 (0.96, 1.06)	0.739	-	-	0.97 (0.86, 1.10)	0.620	-	-	1.05 (0.93, 1.20)	0.310	-	-
Simple mode	0.98 (0.89, 1.08)	0.685	-	-	1.12 (0.88, 1.43)	0.395	-	-	1.16 (0.88, 1.51)	0.316	-	-
Weighted mode	0.98 (0.92, 1.05)	0.596	-	-	0.94 (0.82, 1.09)	0.439	-	-	1.12 (0.92, 1.37)	0.279	-	-

Abbreviations: IGF-1, insulin-like growth factor-1; SNP, single nucleotide polymorphism; IVW, inverse variance weighting; OR, odds ratio; CI, confidence interval.

Table 4. Sex-specific associations of genetically testosterone and estradiol levels with COVID-19 risk

Exposure	Method	Susceptibility				Hospitalization				Severity			
		Male		Female		Male		Female		Male		Female	
		OR (95% CI)	<i>P</i>	OR (95% CI)	<i>P</i>	OR (95% CI)	<i>P</i>	OR (95% CI)	<i>P</i>	OR (95% CI)	<i>P</i>	OR (95% CI)	<i>P</i>
Testosterone	IVW	0.96 (0.90, 1.05)	0.463	1.06 (0.97, 1.15)	0.214	0.96 (0.83, 1.10)	0.547	1.03 (0.87, 1.22)	0.731	1.07 (0.89, 1.27)	0.479	0.88 (0.69, 1.11)	0.269
	MR Egger	0.97 (0.86, 1.09)	0.644	1.04 (0.85, 1.26)	0.713	0.88 (0.71, 1.10)	0.270	1.13 (0.76, 1.69)	0.549	0.81 (0.62, 1.08)	0.152	0.68 (0.39, 1.18)	0.169
	Weighted median	0.93 (0.83, 1.04)	0.184	1.06 (0.94, 1.19)	0.370	0.89 (0.72, 1.10)	0.277	1.08 (0.84, 1.39)	0.523	0.89 (0.67, 1.19)	0.438	0.81 (0.57, 1.14)	0.227
	<i>P</i> for intercept	1.00 (1.00, 1.00)	0.998	1.00 (0.99, 1.01)	0.854	1.00 (1.00, 1.01)	0.348	1.00 (0.99, 1.01)	0.615	1.01 (1.00, 1.02)	0.017	1.01 (0.99, 1.03)	0.314
	MR PRESSO	0.97 (0.90, 1.05)	0.464	1.06 (0.97, 1.15)	0.216	0.96 (0.83, 1.10)	0.549	1.03 (0.87, 1.22)	0.732	1.07 (0.89, 1.27)	0.478	0.88 (0.69, 1.11)	0.270
Estradiol	IVW	0.99 (0.89, 1.11)	0.923	0.95 (0.71, 1.26)	0.724	0.98 (0.81, 1.18)	0.826	1.04 (0.63, 1.73)	0.873	0.90 (0.71, 1.15)	0.403	1.39 (0.74, 7.15)	0.310
	MR Egger	1.00 (0.73, 1.36)	0.993	0.89 (0.59, 1.34)	0.598	0.93 (0.52, 1.67)	0.812	1.15 (0.56, 2.34)	0.719	0.61 (0.29, 6.15)	0.233	1.76 (0.74, 3.15)	0.234
	Weighted median	1.05 (0.92, 1.20)	0.432	0.95 (0.68, 1.32)	0.745	0.93 (0.74, 1.16)	0.508	1.32 (0.67, 2.57)	0.422	0.88 (0.65, 1.15)	0.411	1.96 (0.81, 5.15)	0.135
	<i>P</i> for intercept	1.00 (0.96, 1.04)	0.980	1.00 (0.99, 1.02)	0.669	1.01 (0.94, 1.08)	0.856	0.99 (0.96, 1.02)	0.707	1.05 (0.96, 0.15)	0.312	0.99 (0.95, 0.15)	0.441
	MR PRESSO	0.99 (0.89, 1.11)	0.925	0.95 (0.71, 1.26)	0.732	0.98 (0.81, 1.18)	0.831	1.04 (0.63, 1.73)	0.877	0.90 (0.71, 1.15)	0.425	1.39 (0.74, 2.63)	0.335

Abbreviations: OR, odds ratio; CI, confidence interval; IVW, inverse variance weighting.

Table 5. Associations of serum E2 levels instrumented by rs7173595 in the CYP19A1 gene region with COVID-19 outcomes

Sex	Phenotype	beta	se	OR (95% CI)	P effect
Female	Susceptibility	-1.14	0.88	0.32 (0.06, 1.80)	0.195
	Hospitalization	-1.27	1.60	0.28 (0.01, 6.46)	0.426
	Severity	-1.49	2.06	0.22 (0.00, 12.73)	0.469
Male	Susceptibility	-1.00	0.77	0.37 (0.08, 1.67)	0.195
	Hospitalization	-1.11	1.40	0.33 (0.02, 5.11)	0.426
	Severity	-1.31	1.80	0.27 (0.01, 9.26)	0.469

Abbreviations: E2, estradiol; OR, odds ratio; CI, confidence interval.

Table 6. Testosterone, SHBG, IGF-1 and COVID-19 outcomes in mendelian randomization (MR) analyses adjusting BMI

Exposure	Method	Susceptibility					Hospitalization					Severity				
		SNPs	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept	SNPs	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept	SNPs	OR (95% CI)	<i>P</i> effect	<i>P</i> Heterogeneity	<i>P</i> Intercept
Testosterone	IVW	306	0.95(0.83,1.07)	0.386	0.006	-	294	0.83(0.64,1.06)	0.134	0.041	-	307	0.84(0.60,1.17)	0.304	0.030	-
	MR Egger		0.93(0.77,1.13)	0.484	0.006	0.855		0.83(0.56,1.21)	0.324	0.038	0.991		0.83(0.50,1.37)	0.466	0.027	0.949
	Weighted median		0.90(0.72,1.12)	0.331	-	-		0.82(0.52,1.28)	0.375	-	-		0.71(0.42,1.21)	0.214	-	-
	Simple mode		1.13(0.70,1.82)	0.610	-	-		0.68(0.24,1.91)	0.465	-	-		0.37(0.07,1.88)	0.229	-	-
	Weighted mode		0.95(0.79,1.13)	0.540	-	-		0.81(0.56,1.17)	0.273	-	-		0.65(0.40,1.06)	0.085	-	-
	MR PRESSO		0.94(0.83,1.07)	-	-	-		0.83(0.64,1.06)	-	-	-		0.83(0.64,1.06)	-	-	-
SHBG	IVW	308	0.90(0.79,1.04)	0.160	0.002	-	198	0.84(0.64,1.10)	0.209	0.047	-	309	0.89(0.62,1.26)	0.511	0.058	-
	MR Egger		0.94(0.76,1.15)	0.538	0.001	0.663		0.81(0.54,1.21)	0.299	0.043	0.794		0.89(0.53,1.49)	0.666	0.054	0.978
	Weighted median		0.90(0.71,1.13)	0.356	-	-		0.81(0.52,1.28)	0.377	-	-		0.72(0.42,1.23)	0.230	-	-
	Simple mode		1.05(0.60,1.84)	0.860	-	-		1.25(0.42,3.78)	0.689	-	-		0.97(0.22,4.22)	0.967	-	-
	Weighted mode		0.94(0.77,1.15)	0.570	-	-		0.81(0.55,1.20)	0.295	-	-		0.72(0.43,1.22)	0.224	-	-
	MR PRESSO		0.90(0.79,1.04)	-	-	-		0.84(0.64,1.10)	-	-	-		0.89(0.62,1.26)	-	-	-
IGF-1	IVW	15	0.76(0.60,0.96)	0.021	0.172	-	15	0.61(0.41,0.90)	0.014	0.688	-	17	0.84(0.52,1.38)	0.497	0.534	-
	MR Egger		0.88(0.58,1.33)	0.554	0.168	0.390		0.77(0.39,1.50)	0.458	0.676	0.403		1.55(0.71,3.39)	0.284	0.757	
	Weighted median		0.75(0.57,0.99)	0.046	-	-		0.75(0.45,1.24)	0.260	-	-		0.75(0.38,1.48)	0.410	-	-
	Simple mode		0.65(0.38,1.11)	0.135	-	-		0.64(0.30,1.37)	0.265	-	-		0.75(0.25,2.31)	0.629	-	-
	Weighted mode		0.76(0.56,1.03)	0.096	-	-		0.71(0.44,1.15)	0.185	-	-		0.72(0.36,1.47)	0.383	-	-
	MR PRESSO		0.76(0.60,0.96)	-	-	-		0.61(0.43,0.86)	-	-	-		0.84(0.53,1.35)	-	-	-

Abbreviations: SNP, single nucleotide polymorphism; OR, odds ratio; CI, confidence interval; IVW, inverse variance weighting; SHBG, sex hormones-binding globulin; IGF-1, insulin-like growth factor-1.