

1 **Iron nutrition and COVID-19 among Nigerian healthcare workers**

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81 **Abstract**

82 *Background and objectives:* The optimal iron hypothesis (OIH) posits that risk for infection is
83 lowest at a mild level of iron deficiency. The extent to which this protection results from arms
84 race dynamics in the evolution of iron acquisition and sequestration mechanisms is unclear. We
85 evaluated the OIH with regard to SARS-CoV-2, an emerging infectious agent.

86 *Methodology:* We tested 304 healthcare workers at baseline for iron deficiency (zinc
87 protoporphyrin:heme), anemia (hemoglobin), and SARS-CoV-2 (salivary PCR), and followed
88 them for ~3 months with bi-weekly SARS-CoV-2 tests. We fit logistic regression models based
89 on Akaike Information Criterion.

90 *Results:* Adequate data were available for 199 participants. Iron replete (OR: 2.87, 95% CI:
91 0.85, 9.75) and anemia (OR: 2.48; 95% CI: 0.82, 7.85) were associated with higher risk for
92 SARS-CoV-2 infection after control for covariates. Logistic regression and Cox proportional
93 hazards models of the SARS-CoV-2 outcome were similar. Anemia (OR: 1.81; 95% CI: 0.88,
94 3.71) was associated with respiratory symptoms regardless of SARS-CoV-2 infection.

95 *Conclusions and implications:* These findings provide partial support for the OIH: SARS-CoV-2
96 infection risk was elevated at the high end of the range of iron availability; however, elevated
97 risk among those with anemia was not, as expected, specific to severe iron deficiency.
98 Narrowly, for COVID-19 epidemiology, these findings accord with evidence that SARS-CoV-2's
99 ability to establish an infection is enhanced by access to iron. More broadly, these findings
100 suggest that the OIH does not hinge on a long history of evolutionary arms race dynamics in
101 access to host iron.

102

103 **Lay Summary:** Iron is necessary for both human hosts and infectious agents, including viruses.
104 We tested whether iron nutrition affected risk for infection with SARS-CoV-2, the virus that
105 causes COVID-19, and found that risk was higher in the iron replete state (the “best” iron
106 nutrition) and the anemic state.

107 **Background and objectives**

108 Iron nutrition can have multiple, complex effects on infectious disease risk. Both immune
109 cells and infectious agents require iron to support their function. Because iron that is available to
110 host cells is also available to infectious agents, it may be that risk for at least some infectious
111 diseases is decreased by iron intake that is inadequate to meet the body's overall iron needs. In
112 other words, mild iron deficiency, relative to either the iron replete state or severe iron
113 deficiency, may be optimal, at least for infectious disease risk [1–4]. We have found support for
114 this hypothesis in multiple settings in sub-Saharan Africa [1,5,6]. Here, we test whether iron
115 deficiency affects risk for infection with severe acute respiratory syndrome coronavirus 2
116 (SARS-CoV-2), the causal agent of coronavirus disease 2019 (COVID-19), among healthcare
117 workers (HCW) in Nigeria in 2021-22.

118 Testing the optimal iron hypothesis (OIH) with regard to SARS-CoV-2 risk is important to
119 understanding the evolutionary grounding of this hypothesis. Iron is a limiting resource for many
120 infectious agents. Viruses rely on host iron to efficiently replicate within infected cells, and
121 employ multiple and overlapping mechanisms to access host iron [7,8]. While evidence of
122 evolutionary arms race dynamics between infectious agent and host is most apparent for
123 bacterial pathogens—whose abilities to extract host iron, even in the face of multiple host
124 mechanisms to sequester it, provide clear evidence of the iterative nature of this arms race [9–
125 11]—it is likely that these dynamics are also at play for intracellular access to iron for many
126 viruses [7,12,13]. It is against the background of these evolutionary arms race dynamics that we
127 have posited that, as infectious agents evolve mechanisms to subvert iron sequestration and
128 extract host iron, nutritional iron deficiency may hinder infections and lower infectious disease
129 risk or severity. SARS-CoV-2, as an emerging infectious agent (or emerging infectious disease,
130 EID), may be poorly adapted to humans' iron sequestration and withholding mechanisms. If iron
131 sequestration is effective in limiting the iron available to support SARS-CoV-2 replication, there
132 is unlikely to be any hazard to the iron replete state (or conversely, any benefit to the iron

133 deficient state). Instead, those who are iron replete may have the best delivery of iron to
134 immune defense and so lowest risk for SARS-CoV-2 infection.

135 However, some evidence suggests that iron is particularly salient to risk for infection with
136 SARS-CoV-2. SARS-CoV-2 attacks hemoglobin, freeing iron from porphyrins and increasing
137 free iron [14,15]. This free iron plays a role in COVID-19's pathogenesis [14], and a positive
138 association between iron availability and disease severity has been reported among COVID-19
139 patients [16]. As with some other viruses that affect iron homeostasis, however, it remains
140 unclear the extent to which these effects enhance SARS-CoV-2's ability to establish an infection
141 or persist in the host (or whether these effects are byproducts of viral proliferation) [7]. Lower
142 risk for SARS-CoV-2 among those with blood type O [17–19] has been suggested to be
143 attributable to lower circulating iron availability [14], but this is far from definitive evidence that
144 iron is central to SARS-CoV-2's ability to establish an infection. Initial investigations of
145 hemochromatosis as a risk factor for SARS-CoV-2 infection have mixed results, but overall
146 present limited evidence that one disease-associated allele (and resultant higher plasma iron)
147 may increase risk for SARS-CoV-2 infection or a more severe course of COVID-19 [20,21].

148 Thus, overall, there is reason to expect iron deficiency increases risk for SARS-CoV-2
149 infection *and* reason to expect that it reduces risk: If SARS-CoV-2 is poorly adapted to humans'
150 many mechanisms of iron withholding and sequestration, then an iron replete host and iron
151 deficient host are unlikely to differ in the availability of iron for SARS-CoV-2, and iron deficient
152 hosts may be at higher risk for infection due to weakened immune defense. On the other hand,
153 SARS-CoV-2 may need free iron, generated by attacking host hemoglobin, to enhance its ability
154 to establish an infection and efficiently replicate in host cells, in which case an iron replete host
155 may have higher risk for infection. We assessed the impact of iron nutrition on risk for SARS-
156 CoV-2 infection. Our goals in this project were two-fold: adding nuance to our understanding of
157 the OIH and its place in the broader study of the evolutionary dynamics of humans and
158 pathogens, and expanding understanding of COVID-19 epidemiology.

159

160 **Methodology**

161 *Participants and setting*

162 We collected data in three hospitals in Lagos state and one hospital in Enugu state in
163 Nigeria. Participants were healthcare workers (HCW) in units most likely to treat COVID-19
164 patients (dedicated COVID-19 wards and medical emergency wards). We invited all HCW,
165 including both providers (physicians and nurses) and support staff, from selected wards to
166 participate, until target sample sizes (200 in Lagos, 100 in Enugu) were met. Hospital
167 employees who held administrative roles that did not involve contact with patients were not
168 included. Participation included an initial visit, at which hypothesized risk and protective factors
169 were characterized and the first COVID-19 test was performed, followed by weekly surveys for
170 symptoms and bi-weekly COVID-19 testing. (PCR positivity persists for ~2 weeks, often longer
171 [22,23], so while this testing interval did not allow us to pinpoint when participants became
172 positive, it was unlikely to miss many cases.)

173 We obtained written informed consent from all participants. The Institutional Review
174 Boards (IRBs) of Lagos State University Teaching Hospital (LASUTH), the Lagos University
175 Teaching Hospital (LUTH), and the University of Nigeria Teaching Hospital (UNTH) provided
176 ethical review and oversight. Binghamton University's IRB relied on the findings and oversight of
177 LASUTH and UNTH. All procedures were in accordance with the ethical standards of the review
178 boards and with the 1964 Helsinki Declaration and its later amendments.

179 *Survey instruments*

180 The initial survey instrument asked all participants to describe basic personal
181 characteristics (e.g., date of birth), household size and composition information, their role within
182 the hospital (physician/nurse/other), their highest degree earned, COVID-19 vaccination status,
183 past positive COVID-19 tests, and past diagnoses with diabetes or other chronic disease.
184 Follow up weekly surveys asked participants to report any of a list of infectious disease

185 symptoms, which included COVID-19 specific symptoms (e.g., loss of taste and smell) and
186 respiratory infectious diseases more generally (e.g., fever, cough); follow up surveys also asked
187 participants to report any new COVID-19 vaccination received or positive COVID-19 test from a
188 source outside the study.

189 *Anthropometry*

190 At the initial data collection, we characterized weight and height with the hospitals'
191 standard equipment.

192 *Hematology*

193 A trained phlebotomist collected venous blood at the initial data collection. Specimens
194 were transported on ice to the Hematology Laboratory at the Mainland Hospital (Lagos) or the
195 Molecular Virology Laboratory at UNTH (Enugu), where we assessed whole blood specimens
196 for hemoglobin (Hb) using a HemoCue Hb 301 hemoglobinometer; zinc protoporphyrin to heme
197 ratio (ZPP:H) using a hematofluorometer (ProtoFluor-Z, Helena Laboratories); and glycated
198 hemoglobin (HbA_{1c}) using an Infopia Clover A1c analyzer in Lagos and a SimmplexTAS
199 analyzer in Enugu.

200 *Virology*

201 Participants provided saliva specimens in sterile containers at the initial data collection
202 and biweekly for 3 months. ~1-2 ml of whole saliva was transported on ice to the Department of
203 Medical Microbiology Research Laboratory, at the College of Medicine of the University of
204 Lagos (Lagos) or the Molecular Virology Laboratory at UNTH (Enugu) and stored at -60°C until
205 analysis. Following manufacturer's instructions, we extracted viral nucleic acid from inactivated
206 specimens using a small spin column RNA extraction kit (Qiagen, Maryland, USA). We
207 amplified and reverse transcribed purified ribonucleic acid (RNA) into complementary DNA
208 using the GeneFinder COVID19 Plus RealAmp RT-PCR test kit. This kit employs qRT-PCR for
209 the qualitative identification of the SARS-CoV-2 RdRp, N, and E genes. We considered results

210 valid if internal control and cycle threshold values were within the kit manufacturer's acceptable
211 ranges.

212 *Statistical analyses*

213 We parameterized biomarker variables as follows: Diabetes, HbA1c \geq 6.5%; anemia, Hb
214 $<$ 13.0 mg/dl for males and Hb $<$ 12.0 mg/dl for females [24]; iron deficiency, ZPP:H \geq 70
215 $\mu\text{mol/mol}$ [25]. Because consensus around a ZPP:H definition for iron deficiency is lacking, with
216 published cutpoints ranging from 40 to 80 $\mu\text{mol/mol}$ [1,25–29], and in recognition of the arbitrary
217 nature of cutpoint-based definitions, we also trialed cutpoints of ZPP:H \geq 80 $\mu\text{mol/mol}$ for iron
218 deficiency and ZPP:H $<$ 40 $\mu\text{mol/mol}$ for iron replete [1]. We treated no response as missing
219 information for survey items, with the exception of previous diagnoses, weekly symptoms,
220 COVID-19 vaccination, or co-resident household members; for these variables, we assumed
221 blank responses indicated “no” or “0” (as “no” or “0” were rarely recorded responses).

222 We excluded participants from analyses if they missed more than two scheduled PCR
223 tests. We then considered any positive SARS-CoV-2 PCR test the primary outcome of interest.
224 Other outcomes of interest include symptomatic COVID-19 (a positive PCR test combined with
225 cough, sore throat, fever, shortness of breath, and/or loss of taste/smell), and symptomatic
226 respiratory infection (reported cough, sore throat, shortness of breath, fever, and/or loss of
227 taste/smell, regardless of SARS-CoV-2 PCR test results).

228 We fit logistic regression models for each outcome of interest, using the Akaike
229 Information Criterion (AIC) to select the best fit model among nested models. Predictors of
230 interest included iron deficiency or replete (by ZPP:H) and anemia (by Hb), as well as the
231 interaction between them (the OIH predicts an interaction between iron deficiency and anemia
232 such that a protective effect of iron deficiency is limited to those without anemia, or mild to
233 moderate iron deficiency). We considered COVID-19 vaccination, diabetes, overweight (body
234 mass index, BMI, \geq 25) and/or obesity (BMI \geq 30), hospital role, study site, and household size
235 (number of reported co-resident children and adults) as potential confounding variables.

236

237 **Results**

238 *Descriptive analyses*

239 304 participants initially enrolled in the study; adequate data for analysis was available
240 for 199. A total of 105 individuals were excluded from the analysis set: five individuals were
241 assigned a study ID and then declined to further participate; an additional 71 were excluded for
242 missing 3 or more PCR tests; an additional 23 were excluded for missing BMI; an additional 5
243 individuals were excluded for missing HbA_{1c}; and an additional 1 was excluded for missing
244 hospital role. The 199 participants included in analyses were not markedly different in
245 characteristics from the initial 304 participants (Table 1).

246 Both iron deficiency (16%) and anemia (33%) were common. 11% of participants tested
247 positive at least once for SARS-CoV-2 during the data collection period. Symptomatic COVID-
248 19 was uncommon among our participants, likely due to high rates of vaccination. (Most
249 vaccinated participants reported receiving the Oxford/AstraZeneca vaccine, which has better
250 efficacy for preventing severe or symptomatic disease than infection [30–32].)

251 The majority of positive test results occurred early in the evaluation period, including
252 41% of observed cases on the first test date. As such, we relied primarily on logistic regression
253 analyses to test the hypothesized relationships between iron deficiency, anemia, and SARS-
254 CoV-2 infection.

255 [TABLE 1 HERE]

256 *SARS-CoV-2 infection*

257 Comparisons of models by AIC did not support inclusion of iron deficiency variables
258 (ZPP:H \geq 70 or \geq 80 $\mu\text{mol/mol}$) in models of the SARS-CoV-2 outcome, alone or in interaction
259 with anemia. Model selection by AIC did support inclusion of iron replete (ZPP:H <40 $\mu\text{mol/mol}$
260 definition) in the final model, as well as anemia, age (in ~10-year increments of 20-29, 30-39,
261 40-49, and 50+), lean BMI (BMI < 25), household size including only adults (stratified as small

262 or <4, medium or 4-5, and large or >5), and study site (Table 2). SARS-CoV-2 infection
263 occurred more often among iron replete HCW (OR: 2.87, 95% CI: 0.85, 9.75; Table 3) and
264 those with anemia (OR: 2.48; 95% CI: 0.82, 7.85). SARS-CoV-2 infection also occurred more
265 often among older participants, was less common among those with BMI < 25, and occurred
266 more often among those in the Enugu study site.

267 We also estimated a Cox proportional hazards (CPH) regression model with the same
268 predictor variables. CPH models have the advantage of accounting for time at risk; however, the
269 preponderance of events early in the monitoring period limit our ability to rely on these models.
270 An additional 20 participants (19 negative and 1 positive) were excluded from these models due
271 to missing information on exact dates for at least one PCR test. These models showed similar
272 patterns to the logistic regression model (Supplemental Information Table S1).

273 [TABLES 2 AND 3 HERE]

274 *Respiratory infection*

275 Comparisons of models by AIC did not support inclusion of any iron nutrition variables in
276 models of the symptoms of respiratory infectious disease outcome (fever, cough, sore throat,
277 shortness of breath, and/or loss of taste or smell). Using AIC, the final model for respiratory
278 symptoms included anemia, age (in 10-year increments), sex, household size (adults and
279 children, continuous), lean BMI, and study site (Table 4). Like SARS-CoV-2 infection,
280 respiratory infectious disease symptoms were more common among HCW with anemia (OR:
281 1.81; 95% CI: 0.88, 3.71), although this pattern was more pronounced for SARS-CoV-2 infection
282 than for respiratory symptoms (Table 5). Unlike SARS-CoV-2 infection, symptoms of respiratory
283 infection declined with age (OR: 0.68, 95% CI: 0.48, 0.96). Respiratory infection symptoms also
284 increased with household size and were more common in Enugu.

285 [TABLES 4 AND 5 HERE]

286 Thus, in sum, our models suggest that iron replete was positively associated with SARS-
287 CoV-2 infection and anemia was positively associated with SARS-CoV-2 infection and

288 symptoms of respiratory infection more generally. This is partially consistent with the OIH—in
289 that it suggests SARS-CoV-2 infection risk is higher among those with more iron available—
290 although we would also have expected that association to extend to respiratory infections more
291 broadly (as captured by fever, cough, shortness of breath, and sore throat). The positive
292 associations between anemia and both SARS-CoV-2 infection and respiratory infection are also
293 consistent with our expectations of higher risk for infectious disease when nutritional strain is
294 enough to compromise immunity. However, iron status did not interact with anemia in the way
295 that we predicted (a protective effect of mild/moderate iron deficiency; i.e., a protective effect of
296 iron deficiency only among those who were not anemic), and so we cannot attribute the positive
297 association between anemia and COVID-19 or infectious symptoms to severe iron deficiency
298 alone.

299 Broadly, our results are consistent with the hypothesis that abundant host iron nutrition
300 increases risk for SARS-CoV-2 infection.

301

302 **Conclusions and implications**

303 *The optimal iron hypothesis*

304 The impact of iron nutrition on infectious disease risk results from multiple complex
305 interactions between immunity, nutrition, and infectious agents that is almost certainly simplified
306 by the optimal iron hypothesis. Iron nutrition is dynamic, and not only affects, but is affected by,
307 infectious disease processes: Infectious agents in the gastrointestinal tract can cause blood loss
308 or compete with the host for dietary iron [33–37]. Others, including SARS-CoV-2, cause
309 destruction of erythrocytes or hemoglobin, disrupting the use of iron for oxygen transportation
310 and increasing free iron in a way that is damaging to host tissues [35,36]. The immune response
311 to infection likely increases iron demands, while also severely limiting iron absorption and
312 sequestering iron away from cellular use, redistributing it to more secure compartments (e.g.,
313 bound to ferritin within macrophages), which some infectious agents have evolved to exploit

314 [9,38–43]. The complexity of these dynamics across infectious agents, environments, and hosts
315 tremendously complicates investigations of the OIH. Understanding the conditions in which iron
316 deficiency is and is not protective will be important in further understanding the evolutionary
317 dynamics of human infectious agents, and how these affect iron nutrition and disease
318 vulnerability.

319 One interesting question within investigations of the OIH regards the role of evolutionary
320 arms race dynamics. We have argued here and elsewhere [1,5,6] that protective effects of iron
321 deficiency against infection arise at least in part from these arms race evolutionary dynamics
322 between humans and our infectious agents: infectious agents' iron acquisition mechanisms
323 have an evolutionary advantage in their faster generation times, and many have overcome
324 multiple iron defense and sequestration mechanisms, and so even in the presence of complex
325 feedback pathways between iron nutrition and infection, across a broad range of infectious
326 agents, environments, and hosts, we expect restricted iron intake and absorption to reduce risk
327 for infection. However, empirical evidence to evaluate the question, “Do evolutionary arms race
328 dynamics create conditions in which iron deficiency may constitute a nutritional adaptation to
329 infectious disease?” is limited. We posit that EID can address this question, as emerging
330 infectious agents are often poorly adapted to the human host and lack a long history of
331 evolutionary arms race dynamics with humans. Here, by assessing iron nutrition and risk for an
332 EID (COVID-19, or SARS-CoV-2 infection), we begin to assess whether the OIH is indeed
333 contingent on a long evolutionary arms race for host iron.

334 Overall, our analyses suggest that the OIH is partially supported for at least one EID: the
335 iron replete state seems to increase risk for SARS-CoV-2 infection. This counters our assertion
336 that arms race dynamics contributed to selection for iron deficiency as a nutritional adaptation to
337 infectious disease. We also found that anemia seems to increase risk for SARS-CoV-2 infection,
338 but that this risk is not limited to iron deficiency anemia. This provides much more limited
339 evidence for the predicted protective effect of mild iron deficiency (relative to either the replete

340 or severe deficiency states) than we have previously documented [5,6], but does counter the
341 expectation that the OIH is relevant primarily or exclusively for infectious agents that have
342 evolved to subvert or exploit mechanisms of iron withholding. While this is only one EID, and our
343 findings are not conclusive on their own, they do suggest that a long history of arms race
344 dynamics is not necessary for nutritional adaptations that restrict iron availability to infectious
345 agents to have a protective effect against infection. Alternatively, SARS-CoV-2, through arms
346 race dynamics with other mammalian host organisms [44], may have been pre-adapted to
347 subvert human iron sequestration mechanisms, providing a poor test of the role of arms race
348 dynamics in the OIH.

349 We found elevated risk for SARS-CoV-2 infection in the most iron replete state, rather
350 than lower risk in the state of frank iron deficiency. This may be explained by differences across
351 studies in iron deficiency biomarkers and cutpoints (e.g., our cutpoint for iron replete represents
352 the low end of the range of values used to define iron deficiency); it may also arise from real
353 differences across samples, settings, or infectious agents in associations between iron nutrition
354 and infection risk. Tradeoffs in iron nutrition, between iron available to host cells and infectious
355 agents, may exist in some settings without pushing optimal iron into the range of deficiency.
356 Nonetheless, our results clearly suggest replete iron nutrition (as indicated by ZPP:H, which is
357 specific to iron nutrition but lacks consensus around cutpoints [25,27–29,45]) is associated with
358 elevated risk for SARS-CoV-2 infection. Our results may also indicate that anemia (as indicated
359 by Hb, which is affected by iron nutrition as well as multiple other factors affecting erythropoiesis
360 [46,47]) is associated with elevated risk for SARS-CoV-2 infection. This may suggest, as the
361 OIH predicts, higher risk at both ends of the range of iron availability. However, the effect of
362 anemia among our participants was not limited to iron deficiency anemia, limiting the
363 interpretability of this result with regard to the OIH.

364 In contrast to our findings for SARS-CoV-2 infection, we found little support for the OIH
365 with regard to respiratory infections in general—iron nutrition was unassociated with disease

366 risk. This was unexpected, in light of previous work. However, data in support of the OIH to date
367 come largely from children [1,5,48] and postpartum mothers [6], including those living through
368 sustained droughts [1,6]. The divergent findings in this study may point to participants'
369 underlying capacity for immune defense—which the OIH posits declines with limited iron
370 availability—as an important determinant of optimal iron, such that limited iron nutrition is more
371 likely to be protective when immune defense is compromised (e.g., due to immaturity).

372 **SARS-CoV-2 Epidemiology**

373 We observed some expected patterns in risk for SARS-CoV-2 infection: a lean BMI
374 (<25) was protective, and risk increased with age. Notably, we observed increasing risk for
375 SARS-CoV-2 infection with age among non-elderly HCW (our oldest participant was 61 years of
376 age, younger than many definitions of “elderly” employed in COVID-19 public health policy
377 [49,50]). This finding suggests that public health and prevention programs may benefit from
378 incorporating more nuance into age-based recommendations. Overall, the consistency of our
379 models with established patterns in COVID-19 risk [51,52] gives us confidence in our novel
380 finding of elevated risk among iron replete and anemic participants.

381 Elevated risk among those who are iron replete is consistent with previously published
382 suggestions that the SARS-CoV-2 virus benefits from higher iron availability, such as higher risk
383 among carriers of at least one hemochromatosis-associated allele. This suggests that SARS-
384 CoV-2’s ability to establish an infection is enhanced by its attack on hemoglobin and resultant
385 increase in free iron. Other authors have suggested iron chelation as a potential therapeutic for
386 COVID-19 [15,53]; future research may consider not only whether iron chelation alleviates the
387 adverse effects of hemoglobin destruction and excess free iron, but whether it impairs SARS-
388 CoV-2 replication in the host.

389 ***Limitations***

390 The models in Tables 3 and 5 are supported by assessment of fit using an information
391 theoretic approach. Nonetheless, caution in interpreting these results is merited, given the wide
392 confidence intervals for iron replete and anemia.

393 Most cases of SARS-CoV-2 infection in our data occurred early in the monitoring period
394 (including some at the initial visit), limiting our ability to capitalize on the longitudinal nature of
395 these data in modeling; as a result, although our approach (logistic regression for any positive
396 test during monitoring) was statistically robust, our findings are vulnerable to misattribution of
397 cause and effect.

398 Due to the complex interactions between iron nutrition and infection, previous tests of
399 the OIH have found that cross-sectional and longitudinal study designs can lead to disparate
400 conclusions [2,5], with cross-sectional assessments potentially capturing the effects of infection
401 and inflammation on risk for iron deficiency anemia, rather than effects of iron nutrition on risk
402 for infection. This is likely a concern with SARS-CoV-2 infection, as well, as it can have
403 hemolytic effects [14,16,53]. Here, we have longitudinal data, but since the majority of infections
404 happened early in the monitoring period, we have essentially collapsed the time component of
405 our data to use logistic regression models. Thus, it could be that the positive association
406 between anemia and SARS-CoV-2 infection that we observed actually reflects SARS-CoV-2
407 infection increasing risk for anemia, not the converse. However, this explanation seems unlikely,
408 given the low rate of otherwise symptomatic infections we observed; further, reverse causation
409 is not a plausible explanation for the positive association between SARS-CoV-2 infection and
410 the iron replete state that we also report.

411 We relied on a biomarker of iron deficiency, ZPP:H, that is more robust to inflammation
412 than many others (e.g., ferritin or hepcidin [54]); nonetheless, some studies have reported
413 elevated ZPP:H in the presence of inflammation [54,55]. We do not have information about
414 inflammation and so have limited ability to discriminate between nutritional iron deficiency and
415 iron withholding as determinants of ZPP:H. However, it is unlikely that elevated ZPP:H due to

416 inflammation at the outset of data collection would explain the positive association we observed
417 between iron replete (low ZPP:H) and SARS-CoV-2 infection. Similarly, we do not have
418 information about hemoglobinopathies, which may elevate ZPP:H [54]; it is thus possible that a
419 protective effect of hemoglobinopathy-associated alleles against SARS-CoV-2 infection
420 contributed to our findings.

421 Missingness was high in these data, mostly due to poor attendance for COVID-19
422 testing. Missing follow up visits are likely attributable to high workloads among participating
423 HCW during the periods of data collection in both Lagos and Enugu; if those who were excluded
424 due to missed tests were also those who were busiest with patient care, they may have been
425 particularly likely to be infected with SARS-CoV-2, resulting in under-ascertainment of cases.
426 This is unlikely, however, to have falsely produced any of the associations between SARS-CoV-
427 2 and iron deficiency or anemia that we observed.

428 PCR testing is vulnerable to false negatives, particularly in the early days of an infection
429 [56]. There is heterogeneity in how long people remain PCR positive for SARS-CoV-2 after
430 infection, with many remaining positive for a period of multiple weeks [57]. While we selected a
431 two-week testing interval to accommodate our participants' high workloads while minimizing
432 missed cases, it is possible that a case occurred in the period between tests that was not
433 captured by our sampling schedule.

434

435 *Conclusions*

436 Both replete iron nutrition and anemia may increase risk for SARS-CoV-2 infection. For
437 biological anthropologists and others interested in adaptation to infectious disease, this finding
438 suggests that the OIH is not contingent on a history of a long evolutionary battle between host
439 and infectious agent over iron. For those interested in SARS-CoV-2 virology and evolution, this
440 finding bolsters arguments that accessing host iron is a key component of the virus's ability to
441 establish an infection.

442

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446

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450

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602

Table 1. Sample characteristics

	All enrolled (n=304)		All analyzed (n=199)	
	Number	Percentage	Number	Percentage
Sex				
Female	170	56	111	56
Male	129	42	88	44
Missing data	5	2	0	0
Hospital site				
Enugu	100	33	80	40
Lagos	204	67	119	60
Hospital role				
Physician	65	21	55	28
Non-Physician	233	77	144	72
Missing data	6	2	0	0
COVID-19 vaccination status				
Yes	212	70	137	69
No	92	30	62	31
Type 2 diabetes				
Yes	37	12	22	11
No	249	82	177	89
Missing data	18	6	0	0
COVID-19 PCR test results				
Any Positive	26	9	22	11
Never Positive	202	66	177	89
Missing data	76	25	0	0
Respiratory symptoms*				
Yes	101	33	76	38
Never	203	67	123	62
Anemia				
Yes	99	32	65	33
No	191	63	134	67
Missing data	14	5	0	0
Iron deficiency (ZPP:H \geq 70 $\mu\text{mol/mol heme}$)				
Yes	58	19	32	16
No	232	76	167	84
Missing data	14	5	0	0
Iron replete (ZPP:H < 40 $\mu\text{mol/mol heme}$)				
Yes	55	18	41	21
No	235	77	158	79
Missing data	14	5	0	0
	Mean Std Dev	Range	Mean Std Dev	Range
Age in years	37.77 9.15	21-61	37.48 9.55	21-61
BMI	27.27 4.76	17.85-42.22	27.11 4.52	17.85-42.22
Household size (<18 years old)	1.44 1.65	0-10	1.48 1.63	0-10
Household size (\geq 18 years old)	2.66 2.18	1-19	2.85 2.21	1-19

*Respiratory symptoms include reported fever, cough, shortness of breath, sore throat, and/or loss of smell or taste in weekly symptom questionnaires.

Table 2. Model selection for PCR-positive SARS-CoV-2 infection

	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6	Model 7	Final Model
Anemia	Anemia	Anemia	Anemia	Anemia	Anemia	Anemia	Anemia	Anemia
Site	Site	Site	Site	Site	Site	Site	Site	Site
Age*	Age*	Age*	Age*	Age*	Age*	Age*	Age*	Age*
Household size (adults**)	Household size (adults**)	Household size (adults**)	Household size (adults**)	Household size (adults**)	Household size (adults**)	Household size (adults**)	Household size (adults**)	Household size (adults**)
Hospital role	Hospital role	Hospital role	Hospital role	Hospital role	Hospital role	Hospital role	Hospital role	
Iron replete	Iron replete	Iron replete	Iron replete	Iron replete	Iron replete	Iron replete		Iron replete
BMI < 25	BMI < 25	BMI < 25	BMI < 25	BMI < 25	BMI < 25		BMI < 25	BMI < 25
Vaccine	Vaccine	Vaccine	Vaccine					
Sex	Sex	Sex						
Household size (children***)	Household size (children***)							
Diabetes by HbA _{1C}								
AIC	129.46	127.47	124.35	123.25	121.34	121.93	121.40	120.08

*Age in 10-year increments, with 2 participants aged 60 and 61 years combined with the 50-59 group; **age \geq 18 years; ***age < 18 years

Table 3. Final model of PCR-positive SARS-CoV-2 infection							
Variable	Regression Coefficient	Coefficient 95% CI		Odds Ratio	Odds Ratio 95% CI		P-value
Intercept	-4.43	-6.04	-2.83				
Anemia	0.91	-0.21	2.03	2.48	0.82	7.85	0.11
Iron replete	1.05	-0.14	2.25	2.87	0.85	9.57	0.08
Low BMI	-1.08	-2.49	0.34	0.34	0.07	1.26	0.14
Age*	0.54	-0.02	1.10	1.71	1.00	3.08	0.06
Site (Enugu)	1.22	0.02	2.42	3.38	1.05	11.93	0.05
Household size – adults (small)	Ref						
Household size – adults (medium)	1.37	0.21	2.53	3.95	1.22	12.82	0.02
Household size – adults (large)	0.96	-0.52	2.43	2.61	0.54	10.94	0.20

*Age in 10-year increments, with 2 participants aged 60 and 61 years combined with the 50-59 group

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Table 4. Model Selection for Respiratory Infectious Disease Symptoms

	Model 1	Model 2	Model 3	Model 4	Model 5	Final Model
	Anemia	Anemia	Anemia	Anemia	Anemia	Anemia
	Site	Site	Site	Site	Site	Site
	Age*	Age*	Age*	Age*	Age*	Age*
	Household size	Household size	Household size	Household size	Household size	Household size
	Hospital role	Hospital role	Hospital role	Hospital role	Hospital role	
	Iron replete	Iron replete	Iron replete	Iron replete		
	BMI < 25	BMI < 25	BMI < 25			
	Sex	Sex		Sex	Sex	Sex
	Diabetes by HbA _{1c}					
AIC	253.8	252.1	254.5	252.1	250.5	248.5

*Age in 10-year increments, with 2 participants aged 60 and 61 years combined with the 50-59 group; **age \geq 18 years; ***age < 18 years

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Table 5. Final Model of Respiratory Infectious Disease Symptoms

Variable	Regression Coefficient	Coefficient 95% Confidence Interval		Odds Ratio	Odds Ratio 95% Confidence Interval	P-value
Intercept	-0.68	-1.41	0.05			0.07
Anemia	0.59	-0.12	1.31	1.81	0.88	3.71
Age*	-0.38	-0.73	-0.04	0.68	0.48	0.96
Male sex	-0.66	-1.33	0.02	0.52	0.26	1.01
Site (Enugu)	0.75	0.00	1.50	2.11	1.00	4.50
Total household size	0.10	-0.02	0.22	1.10	0.98	1.25

*Age in 10-year increments, with 2 participants aged 60 and 61 years combined with the 50-59 group; **age \geq 18 years; ***age $<$ 18 years

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611

612 **Iron nutrition and COVID-19 among Nigerian healthcare workers**

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688 **Supplemental Information**

Table S1. Cox proportional model of PCR-positive SARS-CoV-2 infection

Variable	Regression Coefficient	Coefficient 95% CI		Hazard Ratio	Hazard Ratio 95% CI		P-value
Anemia	0.81	-0.14	1.76	2.25	0.87	5.84	0.10
Iron replete	0.59	-0.36	1.55	1.81	0.70	4.69	0.22
BMI < 25	-0.82	-2.16	0.53	0.44	0.12	1.69	0.23
Age*	0.44	-0.07	0.95	1.55	0.94	2.58	0.09
Site (Enugu)	1.01	-0.06	2.08	2.75	0.94	8.00	0.06
Household size – adults (small)	Ref						
Household size – adults (medium)	1.08	0.10	2.07	2.95	1.10	7.91	0.03
Household size – adults (large)	0.69	-0.56	1.93	1.99	0.57	6.92	0.28

*Age in 10-year increments, with 2 participants aged 60 and 61 years combined with the 50-59 group

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