

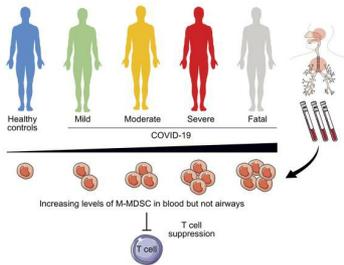
Functional monocytic myeloid-derived suppressor cells increase in blood but not airways and predict COVID-19 severity

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1 **Functional monocytic myeloid-derived suppressor cells increase in blood but**
2 **not airways and predict COVID-19 severity**

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24 **Running title:** MDSCs expand in severe COVID-19

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26 cells, Arginase-1, respiratory immunology.

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28 **Abstract**

29 The immunopathology of COVID-19 remains enigmatic, exhibiting
30 immunodysregulation and T cell lymphopenia. Monocytic myeloid-derived suppressor
31 cells (M-MDSC) are T cell suppressors that expand in inflammatory conditions, but
32 their role in acute respiratory infections remains unclear. We studied blood and airways
33 of COVID-19 patients across disease severity at multiple timepoints. M-MDSC
34 frequencies were elevated in blood but not in nasopharyngeal or endotracheal
35 aspirates of COVID-19 patients compared to controls. M-MDSCs isolated from COVID-
36 19 patients suppressed T cell proliferation and IFN γ production partly via an arginase-
37 1 (Arg-1) dependent mechanism. Furthermore, patients showed increased Arg-1 and
38 IL-6 plasma levels. COVID-19 patients had fewer T cells, and displayed downregulated
39 expression of the CD3 ζ chain. Ordinal regression showed that early M-MDSC
40 frequency predicted subsequent disease severity. In conclusion, M-MDSCs expand in
41 blood of COVID-19 patients, suppress T cells and strongly associate with disease
42 severity, suggesting a role for M-MDSCs in the dysregulated COVID-19 immune
43 response.

44 **Introduction**

45 The pathogenesis of COVID-19 caused by severe acute respiratory syndrome
46 coronavirus-2 (SARS-CoV-2) remains elusive. SARS-CoV-2 infection ranges from
47 asymptomatic disease to multi-organ failure and death (1). COVID-19 is characterized
48 by influenza-like symptoms (including fever, cough and myalgia), and in severe cases,
49 respiratory failure and acute respiratory distress syndrome, occurring in around 40%
50 of hospitalized cases (1-3). Fatal COVID-19 is caused by tissue-directed
51 immunopathology, especially in the lungs, rather than the virus itself (4, 5).
52 Furthermore, it is known that immune cells differ depending on their anatomical
53 location (6-10). Therefore, studying both systemic and respiratory immune responses
54 in COVID-19 is important to fully understand its pathogenesis and to identify factors
55 dictating disease severity.

56

57 COVID-19 is associated with substantial immune activation including elevated levels
58 of proinflammatory cytokines such as IL-6 (11). Furthermore, T cell lymphopenia
59 occurs, especially in critical cases (12), but the underlying mechanism(s) remain
60 unclear. SARS-CoV-2 specific T cells are important in combating the virus (13), and a
61 functional T cell response is critical for clearing infections in general. Myeloid-derived
62 suppressor cells (MDSCs) are myeloid immune cells with an immature phenotype and
63 potent T cell suppressive capacity (14-16). MDSCs expand in inflammatory conditions
64 including cancer, autoimmune disease and chronic viral infections like HIV and
65 hepatitis C (17). Two subpopulations of MDSCs have been identified based on
66 phenotypic and morphological features: monocytic MDSCs (M-MDSCs) and
67 polymorphonuclear MDSCs (PMN-MDSCs) with partly overlapping functions (18, 19).
68 MDSC driven mechanisms of T cell suppression include secretion of arginase 1 (Arg-

69 1) thereby catabolizing L-arginine, generation of ROS and NO, direct engagement of
70 T cell inhibitory and apoptotic receptors, and production of inhibitory cytokines such as
71 IL-10 and TGF- β (20).

72

73 Single-cell RNA sequencing, mass cytometry and flow cytometry on blood have
74 suggested that expansion of suppressive myeloid cells is a hallmark of severe COVID-
75 19 (21, 22). Furthermore, high frequency of PMN-MDSCs was recently reported to
76 correlate with disease severity in COVID-19 (23). However, functional and mechanistic
77 data on the role of M-MDSCs during COVID-19 are lacking, further confounded by the
78 lack of knowledge of the role of M-MDSCs in respiratory infections in general.

79

80 In this study, we investigated M-MDSCs in COVID-19 patients across disease severity
81 and compared to influenza patients and healthy controls. Influenza A virus infection is
82 often compared to COVID-19 due to similarities including the diverse clinical
83 presentation and the route of transmission (24, 25). We found a striking association
84 between the frequency of blood M-MDSCs and COVID-19 disease severity. However,
85 the frequency of M-MDSCs from the nasopharynx and lower airways did not correlate
86 with disease severity in COVID-19. Importantly, purified M-MDSCs were functional and
87 suppressed T cell proliferation, partly via an Arg-1 dependent mechanism. In line with
88 this, plasma Arg-1 levels were elevated in COVID-19 patients in a disease severity
89 dependent manner. Finally, we found that early frequency of blood M-MDSCs
90 predicted subsequent disease severity, suggesting both that M-MDSCs are involved
91 in the dysregulation of the immune response in COVID-19 and that they may be used
92 as a potential prognostic marker in COVID-19 patients.

93

94 **Results**

95 ***Study subject characteristics***

96 In total, 147 adults with PCR confirmed SARS-CoV-2 infection, ranging from mild to
97 fatal disease were enrolled in the study: 91 patients from hospital wards, 43 patients
98 from the intensive care unit (ICU), 3 patients from an outpatient clinic and 10 household
99 contacts, and blood as well as respiratory samples were collected longitudinally (Figure
100 1A and Supplementary figure 1). Identical samples from 44 patients with PCR
101 confirmed influenza A virus infection with mild to moderate disease, as well as from 33
102 age-matched healthy controls (HCs) were included for comparison (Figure 1B and
103 Table 1). As expected, disease severity in the COVID-19 patient cohort varied over
104 time (Figure 1C, Table 2 and Supplementary figure 1), and some patients deteriorated
105 during their hospital stay. At peak disease severity, 13% of patients were classified as
106 having mild disease, 39% as moderate, and 39% as severe (Figure 1C). Furthermore,
107 there were 12 recorded fatalities (8.1%) in the COVID-19 cohort during the observation
108 period (Figure 1C). The peak disease severity score prior to death was 6 in all but two
109 patients who had scores of 4 and 5 respectively. At the end of the study period, 110 of
110 the non-fatal, hospitalized patients had been discharged while 12 patients remained in
111 hospital all of whom were already classified as having severe disease. The distribution
112 of age varied significantly across peak disease severity groups in COVID-19 patients
113 ($p < 0.001$), as did BMI ($p < 0.001$), male gender ($p = 0.004$), and Charlson co-morbidity
114 index (CCI) ($p = 0.042$) (Table 2).

115

116 ***M-MDSC frequencies are elevated in blood from COVID-19 and influenza***
117 ***patients proportional to disease severity***

118 To investigate the dynamics of M-MDSCs during COVID-19 disease, we performed an
119 extensive analysis of samples from COVID-19 patients across disease severity and
120 compared with samples from influenza patients and HCs. PBMCs and cells from
121 nasopharyngeal aspirates (NPA) and endotracheal aspirates (ETA) were stained and
122 analysed by flow cytometry. M-MDSCs were identified as CD14⁺ cells within the
123 lineage negative (CD3⁻CD56⁻CD19⁻CD20⁻CD66⁻), HLA-DR negative population
124 (Figure 2A). In blood, the peak frequency of M-MDSCs was significantly increased in
125 both COVID-19 patients and influenza patients compared to HCs (Figure 2B). The
126 frequency of M-MDSC in NPA had a higher spread among both HC and patients
127 compared to blood (Figure 2B). Albeit in a small number of patients with mild to
128 moderate disease, influenza patients displayed a clear pattern of elevated frequencies
129 of M-MDSCs in NPA as compared to COVID-19 patients and HCs. The elevated
130 frequency of M-MDSCs in NPA in influenza patients compared to COVID-19 was also
131 evident when comparing only mild and moderate influenza and COVID-19 cases,
132 $p=0.0016$ (data not shown). In contrast to NPA, COVID-19 patients with more severe
133 disease had significantly higher peak M-MDSC frequencies in blood, while COVID-19
134 patients with mild disease had blood M-MDSC frequencies comparable to those seen
135 in HCs (Figure 2C). The frequency of M-MDSC in blood from COVID-19 patients
136 seemed to decrease over time (Figure 2D) and returned to similar frequencies as seen
137 in HCs at follow up samples taken during convalescence (33-65 days after study
138 inclusion) (Figure 2E).

139

140 Somewhat surprisingly, COVID-19 patients, in contrast to influenza patients, had low
141 frequencies of M-MDSCs in NPA, even when comparing patients with similar severity
142 of disease (Figure 2B). Since COVID-19 patients on average were included in the
143 study and sampled significantly later after onset of symptoms compared to the
144 influenza patients (18 vs 5 days, respectively), we speculated that the infection-
145 induced inflammation, including infiltration of M-MDSCs, might have transitioned into
146 the lower airways in the COVID-19 patients compared to the influenza patients. To
147 address this, we assessed whether M-MDSCs were present in ETA from the lower
148 airways in 20 intubated COVID-19 patients. However, the frequency of M-MDSCs in
149 ETA was not found to be elevated compared to NPA from the same patients (Figure
150 2F). Although the levels of blood M-MDSCs in COVID-19 patients were significantly
151 elevated compared to HCs, the M-MDSC phenotype with respect to expression levels
152 of CD62L, CCR2 and CD86 were similar (Figure 2G and data not shown). However,
153 respiratory M-MDSCs expressed significantly lower levels of CD62L, CD86 and CCR2
154 compared to blood M-MDSCs, except for the upregulation of CCR2 on ETA M-MDSCs,
155 (Figure 2G-I).

156

157 Immunomodulatory treatment could be a possible confounder when studying immune
158 cells including M-MDSCs. During the sampling period, cortisone was administered to
159 22 patients (Table 2), three of which had ongoing treatment before admission for other
160 conditions. The M-MDSC frequencies among these patients were not significantly
161 different compared to patients without cortisone treatment (Supplementary figure 2).
162 None of the patients received IL-6 inhibitors (tocilizumab) or IL-1 inhibitors during the
163 sampling period. 10 patients received chloroquine phosphate.

164

165 Levels of PMN-MDSCs were analyzed in PBMCs from COVID-19 patients and healthy
166 controls. PMN-MDSCs were identified as CD56-CD14-CD3-CD19-HLA-DR- cells
167 expressing CD66abce and lectin-like oxidized low-density lipoprotein (LDL) receptor-
168 1 (LOX-1) (26). There was a dramatic increase in frequency of PMN-MDSCs, with a
169 clear association to disease severity (Figure 2J). Similar to M-MDSCs, the levels
170 decreased over time in 6 patients where follow-up samples were collected (Figure 2K).
171 Interestingly, the frequency of CD16-expressing PMN-MDSCs was high, indicating a
172 more mature PMN-MDSC phenotype (26-28), and this subset also increased in
173 frequency with increasing disease severity (Supplementary figure 3).

174

175 Altogether, these data indicate that severe COVID-19 disease is associated with
176 elevated levels of M-MDSCs in the blood, but not in the respiratory tract, at least at the
177 time points studied.

178

179 ***M-MDSCs isolated from COVID-19 patients suppress CD4 and CD8 T cell*** 180 ***proliferation***

181 To functionally confirm the identity of M-MDSCs in COVID-19 patients, we evaluated
182 their suppressive effect on T cells. COVID-19 patient blood M-MDSCs were purified
183 and co-cultured with CFSE-labelled allogeneic PBMCs, in the presence of
184 staphylococcal enterotoxin B (SEB) for three days (Supplementary figure 4). As
185 expected, SEB induced strong T cell proliferation (Figure 3A). However, addition of M-
186 MDSCs induced a significant suppression of both CD4 and CD8 T cell proliferation in
187 a dose-dependent manner (Figure 3A-C). In line with this, SEB induced high levels of
188 IFN γ secretion that were significantly lower in co-cultures with M-MDSCs present
189 (Figure 3D). Arg-1 production is one effector mechanism by which M-MDSCs suppress

190 T cell proliferation via degradation of L-arginine that is needed for proliferation. Indeed,
191 addition of L-arginine to the M-MDSC co-cultures restored the concentration of IFN γ in
192 the cell culture supernatants (Figure 3D). Furthermore, co-cultures with M-MDSCs
193 contained high levels of Arg-1 that was undetectable in cultures without M-MDSCs
194 (Figure 3E). In co-cultures supplemented with L-arginine, Arg-1 was no longer
195 detectable, possibly due to complex formation of Arg-1 and the substrate(29) (Figure
196 3E). Importantly, the addition of recombinant L-arginine to the co-cultures decreased
197 the suppressive effect of M-MDSCs on T cells and partially restored T cell proliferation
198 (Figure 3F-G). This indicates that M-MDSCs from COVID-19 patients use Arg-1 as one
199 mechanism to suppress T cells. In conclusion, blood M-MDSC isolated from COVID-
200 19 patients are functional and can suppress T proliferation and IFN γ secretion in a
201 dose- and Arg-1 dependent manner.

202

203 Unfortunately, we were not able to assess the functional capacity of M-MDSCs isolated
204 from patients with influenza A virus infection as too few M-MDSCs could be isolated
205 from the blood volumes obtained. This should be addressed in future studies.

206

207 ***M-MDSC related cytokines are elevated in COVID-19 patients and increase***
208 ***with disease severity***

209 To further investigate the effect of elevated frequencies of blood M-MDSCs in COVID-
210 19 patients, cytokines that have been linked to M-MDSC function and activation were
211 measured in plasma and NPA at the time of study inclusion. In plasma, COVID-19
212 patients had significantly higher levels of Arg-1 than HCs, but no significant difference
213 was observed between COVID-19 patients and influenza patients (Figure 4A).
214 Interestingly, Arg-1 levels in NPA were higher than in plasma in all three groups, with

215 no significant differences between the groups (Figure 4B). Among COVID-19 patients,
216 the plasma concentration of Arg-1 was lower in patients with mild disease compared
217 to patients with moderate, severe and fatal disease ($p=0.07$, $p=0.04$ and $p=0.01$
218 respectively) (Figure 4C).

219
220 Plasma concentrations of IL-6, a potent proinflammatory cytokine important for M-
221 MDSC differentiation(14), were significantly increased in both COVID-19 patients and
222 influenza patients compared to HCs (Figure 4D), while no statistically significant
223 differences were observed in NPA (Figure 4E). Furthermore, levels of IL-6 were
224 strikingly different across the COVID-19 disease severity groups (Figure 4F). Mild
225 COVID-19 patients had significantly lower IL-6 levels than moderate patients
226 ($p=0.038$), severe patients ($p<0.001$) and patients with fatal outcome ($p=<0.01$),
227 respectively. COVID-19 patients with moderate disease also had significantly lower
228 levels than severe patients ($p=0.04$). GM-CSF, which is important for M-MDSC
229 development(14), was also measured in plasma, but was only significantly elevated in
230 influenza patients (Figure 4G).

231
232 Several factors are involved in the generation of M-MDSCs, and it has been proposed
233 that $IFN\gamma$ is involved in the licensing process of monocyte differentiation to M-MDSC
234 (30). Therefore, $IFN\gamma$ was also measured in plasma from COVID-19 patients.
235 However, in our cohort, $IFN\gamma$ was not elevated in patients compared to HCs
236 (Supplementary figure 5), and there was no association with disease severity.

237
238 Finally, concentrations of IL-10 and IL-1 β were measured in plasma of a subset of
239 COVID-19 patients with moderate to fatal disease, but only IL-10 showed an

240 association with disease severity (Figure 4H-I). In summary, cytokines involved in the
241 activation and function of M-MDSCs were elevated in plasma from COVID-19 patients,
242 and correlated with disease severity.

243

244 ***T cells are reduced in blood of COVID-19 patients and have low CD3 ζ chain***
245 ***expression***

246 Since M-MDSCs isolated from COVID-19 patients efficiently suppressed T cells in
247 vitro, the overall blood T cell frequency and function in patients was assessed in the
248 same patients using flow cytometry (Supplementary figure 6). Absolute numbers of
249 peripheral blood CD4⁺ T cells were decreased in the COVID-19 patients with
250 moderate, severe and fatal disease compared to HCs (p=0.02, p=0.001, p=0.05
251 respectively) (Figure 5A). Similarly, absolute numbers of CD8⁺ T cells were also
252 significantly decreased in moderate, severe and fatal COVID-19 disease compared to
253 HCs (p=0.004, p=0.005 and p=0.003 respectively) (Figure 5B). However, no
254 correlation between blood M-MDSC frequency and T cell count in COVID-19 patients
255 was found (Figure 5C), either at peak/bottom frequencies or at any of the longitudinal
256 time points studied in each patient (Figure 5C and data not shown).

257

258 Evidence of T cell suppression was further investigated in COVID-19 patients with
259 varying disease severity, influenza patients and HCs by quantifying the expression of
260 the CD3 ζ chain, a homodimer chain in the T cell receptor complex involved in T cell
261 proliferation and in secretion of cytokines (Figure 5D). The CD3 ζ chain is
262 downregulated in vitro in the absence of L-arginine resulting in decreased T cell
263 proliferation(31). We observed that the surface expression of the CD3 ζ chain on CD4⁺
264 and CD8⁺ T cells was significantly lower in both COVID-19 and influenza patients

265 compared to HC (Figure 5E-F) suggesting that the T cells may have an impaired
266 functional capacity. In summary, COVID-19 patients had lower T cell counts and
267 indication of impaired function of the T cells compared to HCs.

268

269 ***Early M-MDSC frequencies predict peak disease severity***

270 To evaluate the effect of blood M-MDSC frequency early during COVID-19 disease
271 subsequent disease severity, a proportional odds logistic regression was performed
272 with peak disease severity score as primary outcome. Samples were selected from
273 COVID-19 patients in whom: M-MDSC frequency was measured within two weeks
274 from onset of disease, had not already been admitted to the ICU and were not already
275 in the recovery phase (Figure 6A). This yielded a crude odds ratio (OR) of 1.43 (95%
276 CI 1.07 – 2.18), indicating that M-MDSC frequency in the first two weeks from onset of
277 symptoms could potentially be a predictor of disease severity (Figure 6B). In an initial
278 univariate ordinal regression, age but not sex had a significantly elevated OR (1.07
279 [95% CI 1.02 – 1.13] for age as compared to 2.51 [95% CI 0.76 – 8.94] for sex). A
280 model was then run including adjustment for age (Figure 6B). An overview of the
281 patients included in the analysis, the timepoints included in the model and disease
282 severity over time is shown in Figure 6C.

283

284 As shown above, M-MDSC frequency was higher in COVID-19 patients with more
285 severe disease (Figure 2C), and these patients were both predominantly male and had
286 a significantly higher age compared to COVID-19 patients with less severe disease
287 (Table 1). Therefore, the association between M-MDSC frequency and sex and age
288 was assessed, demonstrating significantly higher levels of M-MDSCs in men (Figure

289 6D) and a significant correlation between age and M-MDSC frequency ($R=0.35$,
290 $p=1.9 \times 10^{-5}$) (Figure 6E).

291

292 In summary, early M-MDSC frequencies are associated with subsequent disease
293 severity and appear to be strongly associated with age and sex.

294

295 **Discussion**

296 Understanding the immunopathogenesis of COVID-19 is critical to optimally treat
297 patients and prevent fatal outcome but also to aid in the development of specific
298 therapy and vaccines. One potential player in the immune activation to SARS-CoV-2
299 infection may be MDSCs, a subset of immune cells that in recent years have been
300 intensively studied, by us and others, mainly in relation to cancer and vaccination (14-
301 16, 18, 19, 32). Still, only limited knowledge on how M-MDSCs influence disease
302 severity during infection, including COVID-19, is available. In the present study, the
303 distribution and function of M-MDSCs during COVID-19 were investigated over time
304 and across disease severity in a comparatively large and clinically well-characterized
305 cohort.

306

307 A major finding in the current study is the association between M-MDSC frequency
308 and disease severity in COVID-19 patients. Downregulation of HLA-DR on monocytes
309 has previously been shown in severe COVID-19, possibly reflecting an increase in M-
310 MDSC frequency, and is linked to high levels of IL-6 and lymphopenia (33). We
311 observed an increased frequency of blood M-MDSC, both during COVID-19 and
312 influenza, similar to what was previously observed in HIV-1 (17, 34). While the COVID-
313 19 and the influenza patient cohorts are not completely comparable in disease severity
314 and time of sampling after symptom onset, it is still relevant to know that expansion of
315 M-MDSCs is also observed in another acute respiratory infection, and that this finding
316 is not unique for COVID-19. Furthermore, MDSCs are expanded during sepsis, and
317 are associated with poor outcome (35).

318

319 Notably, in influenza patients M-MDSC frequencies were increased in the nasopharynx
320 compared to blood, indicating that M-MDSCs are recruited to the site of infection during
321 influenza similar to what has been observed for other myeloid cells (36). In contrast,
322 the frequency of M-MDSC in the nasopharynx of COVID-19 patients was similar to
323 HCs. Although SARS-CoV-2 replication is initiated in the upper airways, it frequently
324 progresses to the lower respiratory tract (37, 38), and may result in immune cell
325 recruitment in the lower airways. However, we were unable to find higher frequencies
326 of M-MDSCs in ETA than in NPA. Nevertheless, ETA does not necessarily reflect cells
327 in the alveoli, but is rather a reflection of the trachea and bronchi. It is therefore possible
328 that M-MDSCs are present even deeper down in the lungs and/or that the time points
329 we sampled the COVID-19 patients were past the peak accumulation of M-MDSCs in
330 the airways. Alternatively, M-MDSCs could differentiate to more macrophage-like cells
331 at the site of inflammation, as seen after migration to tumor sites (39, 40), and not be
332 identified with the flow cytometry staining panel used. Migration of M-MDSC from blood
333 to the site of infection is supported by the fact that M-MDSCs in blood had upregulated
334 CD62L compared to respiratory M-MDSCs. CD62L, or L-selectin is involved in the
335 extravasation of immune cells (41). Continued studies are ongoing to further address
336 the kinetics and mechanism of myeloid cell migration in humans.

337

338 Glucocorticoid treatment may affect M-MDSCs, and is therefore important to consider
339 (42-45). In the current study, a relatively low proportion of the patients received
340 glucocorticoids, and we did not observe any differences in M-MDSC frequency
341 between patients with and without glucocorticoid treatment. It is therefore not likely that
342 this is a significant confounder in the study.

343 Characterizing MDSCs is challenging due to the lack of unique cell surface markers
344 and therefore functional analysis of the cells is critical to validate phenotypic
345 identification (15, 20). We verified the suppressive capacity of the M-MDSCs in COVID-
346 19 patients and found that HLA-DR- CD14+ cells isolated from COVID-19 patients had
347 a potent suppressive effect on T cells, demonstrating that the M-MDSCs identified by
348 flow cytometry corresponded to suppressive and functionally active cells. Identification
349 of PMN-MDSCs is even more challenging. The phenotypic identification of PMN-
350 MDSCs varies between studies, and functional analysis is thus critical also for these
351 cells (19). Additionally, PMN-MDSCs are sensitive to cryopreservation (46), and all
352 analyses require fresh cells. In the current study, this prevented us from performing
353 functional analysis of PMN-MDSCs using cryopreserved PBMCs. The focus of the
354 current study is therefore on M-MDSCs.

355

356 Since Arg-1 was important for M-MDSC activity in vitro, levels were measured in
357 plasma and NPA from patients. As expected, COVID-19 patients had increased level
358 of Arg-1 compared to HCs in plasma. There was also a connection between disease
359 severity and Arg-1 level in plasma, although there was no difference between patients
360 with moderate disease to fatal outcome. Interestingly, Arg-1 levels were in general
361 substantially higher in NPA than in plasma for all cohorts, without any association with
362 M-MDSC frequency in NPA. It is known that Arg-1 is constitutively produced in the
363 airways in bronchial epithelial cells, endothelial cells, myofibroblasts and alveolar
364 macrophages. The function, however, is unknown. Arg-1 has been suggested to be
365 involved in regulation of NO and airway responsiveness and tissue repair (29). The
366 association between M-MDSC and Arg-1 in the airways merits further examination.

367

368 Several factors are involved in the expansion of M-MDSCs, including IL-6 and GM-
369 CSF (14). Furthermore, IL-6 and IL-10 have been shown to be essential in inducing
370 emergency myelopoiesis resulting in expansion of an M-MDSC-like cell subset in
371 severe COVID-19 (47). In line with previous studies (12, 33), we demonstrated a
372 relationship between IL-6 and COVID-19 disease severity. The high levels of IL-6 could
373 contribute to the generation of M-MDSCs, especially in patients with severe disease.
374 In contrast to previous reports (48), plasma GM-CSF was not elevated in our COVID-
375 19 cohort. Instead, higher levels of GM-CSF were seen in the influenza patient cohort.
376 This could be explained by kinetics: the COVID-19 group had longer duration of
377 symptoms compared to the influenza patients and it is possible that the level of GM-
378 CSF had already decreased. The lack of IFN γ in COVID-19 patients is in line with other
379 studies showing no or minor increases in IFN γ in COVID-19 patients and a reduced
380 production of IFN γ in lymphocytes from COVID-19 patients (49-51). However, the
381 literature is not conclusive, and other studies have shown elevated levels of IFN γ in
382 COVID-19 patients (2, 52). IL-10 is also relevant in respect to MDSCs, since it can be
383 produced by MDSCs (14, 18), and plasma levels increased with rising disease severity.
384

385 A decrease in absolute numbers of both CD4⁺ and CD8⁺ T cells, in line with the data
386 presented here, has previously been demonstrated in COVID-19 patients with severe
387 disease, but the underlying mechanisms for the decrease are still unknown (53). One
388 potential such mechanism could be the downregulation of CD3 ζ chain, that results in
389 impaired T cell proliferation. CD3 ζ chain downregulation on T cells has previously been
390 observed in relation to MDSCs in several conditions including sepsis, hepatitis C
391 infection and gastric cancer (54-56). By establishing an immunosuppressive
392 environment, M-MDSCs might prevent efficient immune activation and impede the

393 development of specific adaptive responses required to clear the infection. We
394 therefore speculate that expansion of M-MDSCs contributes to the immune imbalance
395 described in COVID-19, possibly favouring disease progression.

396

397 There is an urgent need for a better understanding of the pathophysiology of COVID-
398 19, including predictors of disease severity. In the current study, we investigated
399 whether M-MDSC frequencies in blood early in disease could affect disease outcome.
400 Despite low power and risk of multicollinearity when adjusting for age, our significantly
401 elevated unadjusted crude OR suggests that elevated M-MDSC frequency in the first
402 two weeks from onset of symptoms can predict poorer disease outcome. A limitation
403 with this model is the potential confounding of disease severity at time of inclusion in
404 the model. However, by stringent selection of patients to include in the analysis, we
405 attempted to mitigate this as far as possible.

406

407 Our protein and functional M-MDSC COVID-19 data is in line with emerging RNA seq
408 data that identify a population likely to be M-MDSCs that is associated with severe
409 COVID-19 (22). Previous research has shown age and male gender are risk factors
410 for severe COVID-19 (57). Furthermore, it is known that MDSCs increase with age as
411 part of the “inflammaging” process (the inflammation seen in ageing) (58), but little is
412 known about gender differences. In line with this, we found a correlation between M-
413 MDSC frequency and both age and male gender in this cohort, possibly partly
414 explaining the increased morbidity seen in men and older patients.

415

416 In summary, we have shown that M-MDSCs are expanded in both COVID-19 and
417 influenza and are associated with disease severity in COVID-19 patients, especially in

418 men and older individuals. Although we found indirect indications of M-MDSC
419 migration, we were unable to show an increased M-MDSC frequency in respiratory
420 mucosa in the same way as in influenza, presenting an area of future research. In this
421 study we also demonstrated that M-MDSCs isolated from COVID-19 patients are
422 effective T cell suppressors in vitro and may partly explain the detrimental decrease of
423 T cells in patients with severe COVID-19. Finally, we showed that M-MDSCs may
424 predict disease outcome. The findings of this study provide yet another important piece
425 to the puzzle of fully comprehending the immunologic profile of COVID-19 and can
426 potentially contribute to future therapeutic and diagnostic advancements.

427 **Methods**

428 ***Study subjects***

429 Inclusion of COVID-19 patients was performed at Karolinska University Hospital and
430 Haga Outpatient Clinic (Haga Närakut), Stockholm, Sweden during March-May 2020.

431 Inclusion was performed at various levels of care, ranging from primary to intensive
432 care. Inclusion criteria were age >18 years and PCR-confirmed SARS-CoV-2 infection.

433 In order to recruit mild/asymptomatic cases, household contacts of COVID-19 patients
434 were screened with PCR and recruited if positive. Similarly, adult patients with PCR-
435 confirmed influenza A virus infection were recruited during January-March 2019 and
436 2020. A cohort of healthy controls (HCs) (i.e. confirmed influenza A virus and SARS-
437 CoV-2 negative by PCR) was recruited and sampled in the same way as study patients.

438
439 Degree of respiratory failure was categorized daily according to the respiratory domain
440 of the Sequential Organ Failure Assessment score (SOFA)(59). If arterial partial
441 pressure of oxygen (PaO_2) was not available, peripheral transcutaneous hemoglobin
442 saturation (SpO_2) was used instead and the modified SOFA score (mSOFA) was
443 calculated(60). Fraction of inspired oxygen (FiO_2) estimation based on O_2 flow was
444 done in accordance with the Swedish Intensive Care register definition(61). Patients
445 were subsequently categorized based on the peak respiratory SOFA or mSOFA value.
446 The 5-point respiratory SOFA score was then extended with an additional level to
447 distinguish admitted mild cases from non-admitted mild cases. Finally, fatal outcome
448 was added as a seventh level, with peak disease severity score 6 prior to death in all
449 but two patients who had scores of 4 and 5, respectively. Additionally, the resulting 7-
450 point composite peak severity score was condensed into a classification consisting of

451 mild (1-2), moderate (3-4), severe (5-6), and fatal (7) disease (Supplementary Tables
452 1 and 2).

453

454 Medical records were reviewed for clinical history, laboratory analyses, medications,
455 previous diseases and co-morbidities, and risk factors. Total burden of comorbidities
456 was assessed using the CCI (62).

457

458 ***Collection of respiratory and blood samples***

459 NPA was collected from COVID-19 and influenza patients and HCs where possible,
460 and ETA was collected from intubated COVID-19 patients in the ICU. Venous blood
461 was collected in EDTA-containing tubes from all non-ICU patients and controls. In ICU
462 patients, blood was pooled from heparin-coated blood gas syringes discarded in the
463 last 24 hours. In some ICU patients, additional venous blood samples were also
464 collected in EDTA tubes. Routine clinical chemistry analysis was performed on all
465 study subjects including HCs. Admitted patients were sampled at up to four timepoints
466 and discarded ICU patient material was collected up to ten timepoints.

467

468 ***Isolation of cells from blood and respiratory aspirates***

469 NPA and ETA samples were centrifuged at 400g/5 min/ room temperature (RT).
470 Supernatant was collected and frozen at -20°C. Cells were washed with sterile PBS
471 and mucus was removed using a 70 µm cell strainer. Blood samples were centrifuged
472 at 800g/8 min/ RT. Plasma was collected and frozen at -20°C. The cellular fraction was
473 diluted with sterile PBS and PBMCs isolated by density-gradient centrifugation at
474 900g/25 min/RT (without brake), using Ficoll-Paque Plus (GE Healthcare). Cell count
475 and viability were assessed using Trypan Blue (Sigma) exclusion with an automated

476 Countess cell counter (Invitrogen). Cells were stained fresh for flow cytometry analysis.
477 Excess PBMCs were cryopreserved in FBS (Gibco) with 10% DMSO (Sigma) and
478 stored in liquid nitrogen.

479

480 ***Flow cytometry***

481 Cells were stained using Live/Dead Blue (Invitrogen), incubated with human FcR
482 blocking reagent (Miltenyi Biotec) and stained with antibodies against the following
483 surface proteins: CD1c (AD5-8E7; Miltenyi Biotec), CD3 (SK7; BD), CD11c (B-Ly6;
484 BD), CD14 (M5E2; BD), CD16 (3GE; BioLegend), CD19 (HIB19; BioLegend), CD20
485 (L27; BD), CD45 (HI30; BD), CD56 (HCD56; BD), CD62L (SK11; BD), CD66abce
486 (TET2; Miltenyi Biotec), CD86 (2331; BD), CD123 (7G3; BD), CD141 (AD5-14H12;
487 Miltenyi Biotec), CCR2 (K036C2; BioLegend), CCR7 (150503; BD) and HLA-DR
488 (TU36; Life Technologies). If enough cells were available, a second staining was
489 performed using CD3 (SP34-2; BD), CD4 (L200; BD), CD11c (B-ly6; BD), CD14
490 (M5E2; BD), CD16 (3G8; BD), CD19 (SJ25-C1; Thermo Fisher Scientific), CD45 (HI30;
491 BD), CD56 (HCD56; BioLegend), CD66abce (TET2; Miltenyi Biotec), CD123 (7G3;
492 BD), LOX-1 (15C4; BioLegend) and HLA-DR (L243; BioLegend). All stainings were
493 performed at 4°C for 20 minutes. Cells were washed with PBS and fixed with 1-2%
494 paraformaldehyde.

495

496 The expression of CD247 (TCR ζ , CD3 ζ) on CD4 and CD8 T cells was evaluated by
497 intracellular staining. Briefly, following surface staining with Live/Dead Blue
498 (Invitrogen), CD3 (SP34-2; BD), CD4 (L200; BD) and CD8 (SK1; BD), cells were fixed
499 and permeabilized with permeabilization buffer (Thermo Fisher Scientific) and then
500 stained with anti-CD247 (6B10.2; BioLegend) at 4°C for 20 min. Samples were

501 acquired on LSRFortessa flow cytometer (BD Biosciences). Data were analysed using
502 FlowJo software 10.5.3 (TreeStar). Absolute numbers of CD4 and CD8 T cells were
503 calculated by multiplying the frequency of T cells out of total lymphocytes obtained
504 from flow cytometry data with the lymphocyte count from differential cell counts. If
505 multiple T cell frequencies were available from the same patient, the lowest T cell count
506 was used. If absolute lymphocyte count was missing, a value was linearly interpolated
507 between existing values if no more than 7 days apart.

508

509 ***M-MDSC T cell suppression assay***

510 M-MDSCs (HLA-DR⁻ CD14⁺ cells) were purified from frozen PBMCs of three COVID-
511 19 patients, following a protocol developed by Lin et al(16). HLA-DR⁺ cells were
512 depleted using anti-HLA-DR microbeads and an LD column. From the negative
513 fraction, CD14⁺ cells were positively selected using anti-CD14 microbeads. MS
514 columns and MACS separators (all Miltenyi Biotec) were used for the cell sorting.
515 Approximately 0.2 million M-MDSCs were obtained from 25-30 million PBMCs, with a
516 viability >90% and purity >85%. In parallel, cryopreserved PBMCs from a buffy coat
517 were thawed and 4 million cells were labelled with CFSE (Thermofisher). The
518 previously purified M-MDSCs were co-cultured with 0.5 million of the CFSE-labelled
519 PBMCs, at a ratio of 1:2 or 1:5, in the presence of 0.1 µg/mL SEB (Sigma-Aldrich) or
520 200µg/mL L-arginine (Sigma-Aldrich). The cells were incubated for 3 days at 37°C, in
521 RPMI 1640 (Sigma-Aldrich) medium supplemented with 10% FCS, 5 mM L-glutamine,
522 100 U/mL penicillin and streptomycin (all Invitrogen). Supernatants were collected from
523 cultures to measure secreted Arg-1 (Invitrogen) and IFN_γ (R&D Systems) by ELISA.
524 The cells were washed and surface-stained with CD3 (SK7), CD4 (OKT4), and CD8
525 (SK1) (all from BD Biosciences). Flow cytometry (LSRFortessa, BD Biosciences) was

526 performed as described above and T cell proliferation was measured, by calculating
527 the percentage of CFSE^{low} T cells.

528

529 ***Cytokine analysis***

530 Cytokine levels were measured in plasma samples, NPA supernatants and culture
531 supernatants using ELISA. IL-6, GM-CSF and IFN γ ELISAs were performed using
532 DuoSet[®] kits (R&D Systems). Arginase-1 ELISA was performed using Arginase-1
533 Human ELISA Kit (ThermoFisher). Levels of IL-10 and IL-1 β were analyzed at
534 Karolinska University Laboratory using Roche Cobas e602.

535

536 Furthermore, plasma was also analyzed using proximity extension assay (Olink
537 Proteomics, Sweden) (63), performed at Stanford University according to the
538 manufacturer's instructions. The 92-biomarker Inflammation panel was used. For this
539 paper, only data on IFN γ was included. Antibody pairs labeled with DNA
540 oligonucleotides bind the target antigen and oligonucleotides are hybridized and
541 extended by DNA polymerase. Subsequently, protein expression is measured with
542 high-throughput real-time PCR. Protein levels are presented as normalized protein
543 expression (NPX) values, an arbitrary unit in log₂ scale. Values are calculated from
544 inverted Ct values, and a high NPX value corresponds to a high protein concentration.
545 Normalization of data is performed to minimize intra- and inter-assay variation.

546

547 ***Serology***

548 Antibodies against the SARS-Cov-2 Spike (S) trimer were assessed by ELISA.
549 Recombinant proteins were received through the global health-vaccine accelerator
550 platforms (GH-VAP) funded by the Bill & Melinda Gates foundation. Briefly, 96-well

551 plates were coated with 100ng/well of S protein. Plates were incubated with a selected
552 duplicate dilution (1:50) of each plasma sample at ambient temperature for 2 hours.
553 Detection was performed with a goat anti-human IgG HRP-conjugated secondary
554 antibody (clone G18-145 from BD Biosciences) followed by incubation with TMB
555 substrate (BioLegend; cat# 421101) and stopped with a 1M solution of H₂SO₄.
556 Absorbance was read at 450nm+550nm background correction using an ELISA
557 reader. Data are reported as the average optical density (OD) value of the two
558 duplicates. An extensive analysis of respiratory and systemic antibody responses in
559 the study cohort is available (64).

560

561 **Statistics**

562 Data analysis was performed in RStudio version 1.2 (RStudio Inc., Boston, MA),
563 GraphPad Prism version 8.0 (GraphPad Software Inc., San Diego, CA) and Microsoft
564 Excel (Microsoft Corp., Redmond, WA). Routine analyses excluding cytokines and
565 flow-cytometry data were assumed to have a standard distribution and means were
566 compared using either an independent Student's t-test or a one-way analysis of
567 variance (ANOVA). Nominal patient characteristics were compared between groups
568 using a Pearson Chi-Squared test or Fisher's exact test depending on if the expected
569 count of any cell was above or below five. Cytokine and flow cytometry data were
570 presumed to have a non-standard distribution and medians were thus compared using
571 the Wilcoxon-Mann-Whitney U or Kruskal-Wallis tests depending on number of
572 cohorts. Post-hoc testing after Kruskal-Wallis was performed using Dunn's test of
573 multiple comparisons or by controlling the False Discovery Rate using Benjamini,
574 Krieger and Yekutieli's adaptive linear step-up procedure. Non-parametric
575 comparisons of dependent data were performed using Wilcoxon's Signed Ranks test.

576 For correlation analyses of continuous data, Spearman's Rho was used. Finally, a
577 proportional odds logistic regression model was constructed to evaluate predictive
578 capacity of M-MDSC frequencies on disease severity. A 95% significance level was
579 used throughout the study, and p-values abbreviated (* ≤ 0.05 , ** < 0.01 *** < 0.001).
580 Statistical tests were two-tailed throughout the study.

581
582 Missing daily severity score data was approximated by using a last-observation-
583 carried-forward (LOCF) method. For flow cytometry data, the peak M-MDSC frequency
584 and lowest T cell count were extracted. The peak value for routine laboratory analyses
585 was extracted separately for each analysis except in the case of blood differential
586 counts in which all counts were extracted for the timepoint of lowest lymphocyte count.
587 The ordinal logistic regression model was based on a subset of patients with M-MDSC
588 frequencies sampled within two weeks of onset of symptoms and before potential ICU
589 admission. For patients who developed severe disease, only timepoints prior to peak
590 disease severity were included, while in patients who developed at most mild-moderate
591 disease, timepoints occurring during recovery were excluded. In the event of multiple
592 timepoints for one patient during that period, the first timepoint was used. Unadjusted
593 OR was initially calculated individually for the predictive capacity of M-MDSC-
594 frequency, age, and sex on peak disease severity and statistically significant predictors
595 were subsequently included in a multivariate proportional odds logistic regression
596 model with the four-level peak disease severity score as primary outcome.

597

598 ***Study approval***

599 The study was approved by the Swedish Ethical Review Authority, and performed
600 according to the Declaration of Helsinki. Written informed consent was obtained from

601 all patients and controls. For sedated patients, the denoted primary contact was
602 contacted and asked about the presumed will of the patient and, if applicable, to give
603 initial oral and subsequently signed written consent. When applicable retrospective
604 written consent was obtained from patients with non-fatal outcomes.

605

606 **Contributions**

607 SF-J, SV, AL, KLo, NJ, AF and AS-S planned the study. SF-J, SV, MY, AC, IB, BÖ,
608 MJL, IS, KLe, FH performed experiments. SF-J, RF-J, BÖ, EÅ, JS, MB, NJ, AF
609 included and sampled patients and collected clinical data. JA provided relevant
610 anonymized patient clinical data. SF-J, SV, MY, RF-J, AC, IB, MJL, RL and HM
611 analyzed data. SF-J, MY, RF-J and MJL prepared figures. SF-J and AS-S wrote the
612 manuscript. All co-authors edited the manuscript.

613

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624 Washington, Seattle, WA, USA for the production of viral proteins used in the antibody
625 assay.

626 **Conflict of interest statement**

627 The authors have declared that no conflict of interest exists.

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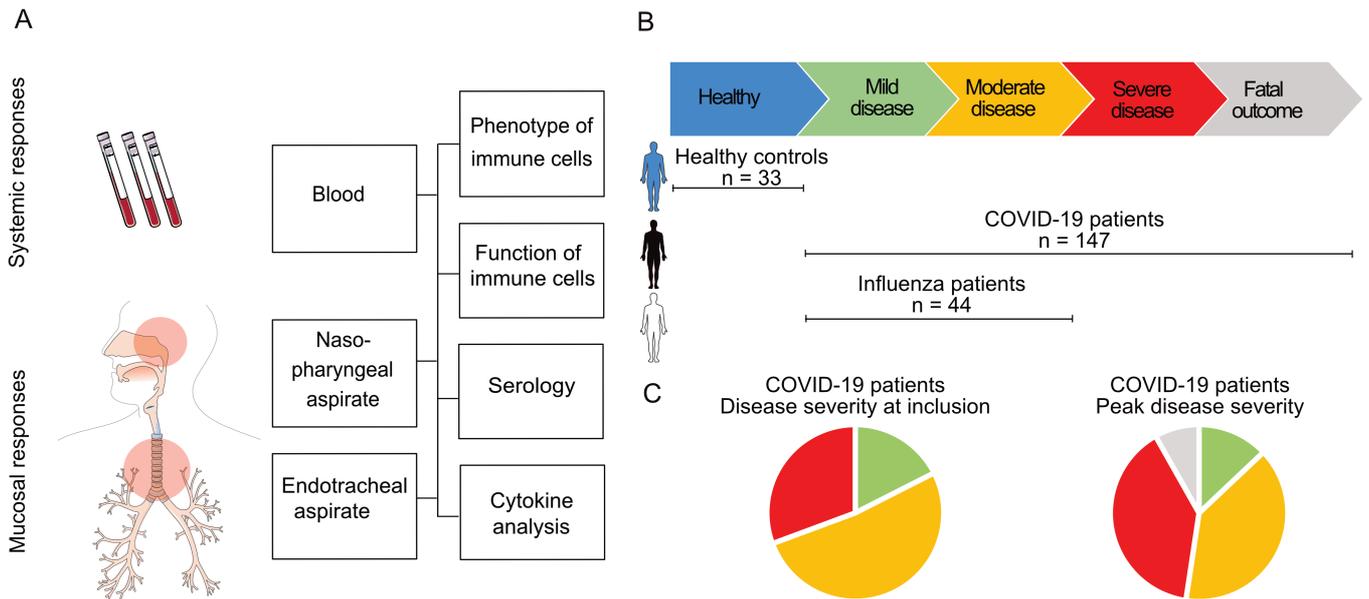


Figure 1. Study outline. (A) Blood and nasopharyngeal aspirates (NPA) were collected from COVID-19 patients, influenza patients and healthy controls. From ICU patients, endotracheal aspirates (ETA) were also collected. Cells were isolated from blood (PBMCs), NPA and ETA and were analyzed fresh using flow cytometry, and used for functional experiments. Aspirates and plasma were collected and used for serology and cytokine detection using ELISA. (B) Study subjects were included and sampled across disease severity, ranging from healthy controls (n=33) to mild or moderate influenza disease (n=44) and mild to fatal COVID-19 (n=147). (C) Pie charts show distribution of disease severity at time of inclusion and the peak disease severity of COVID-19 patients. At time of inclusion: mild 18% (n=24), moderate 52% (n=71) and severe 31% (n=42). At peak disease severity: mild 13% (n=19), moderate 39% (n=58), severe 39% (n=58) and fatal outcome 8% (n=12).

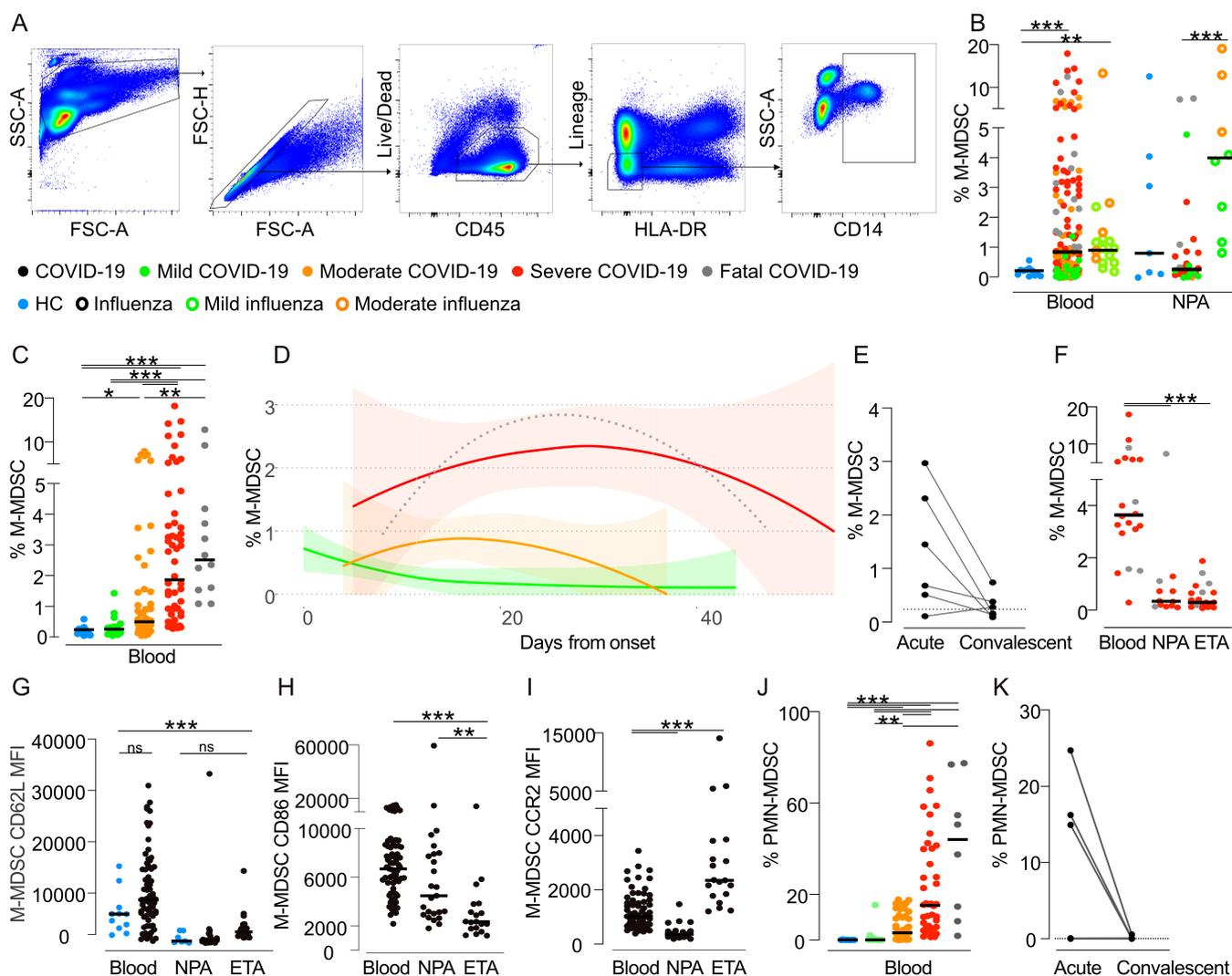


Figure 2. Frequency of respiratory and blood M-MDSC in COVID-19 patients, influenza patients and HCs. (A) Gating strategy to identify monocytic myeloid-derived suppressor cells (M-MDSCs) by flow cytometry. From live, single CD45+ leukocytes, cells expressing lineage markers (CD3, CD19, CD20, CD56, CD66abce) and HLA-DR were excluded and CD14+ M-MDSCs identified. **(B)** M-MDSCs frequency per live CD45+ cells in blood and NPA. HCs (blue): n=12 (blood), 7 (NPA). Influenza patients (open circles): n=19 (blood), 9 (NPA). COVID-19 patients (closed circles): n=140 (blood), 28 (NPA). Points color-coded by peak severity. **(C)** Peak frequency of blood M-MDSCs per live CD45+ cells across disease severity. HCs (blue): n=12. COVID-19 patients (color-coded by peak severity): mild (n=19), moderate (n=53), severe (n=56) and fatal (n=12). **(D)** Blood M-MDSC frequency over time in COVID-19 patients: mild n=17, moderate n=53, severe n=56, fatal n=12. Line shows locally estimated scatterplot smoothing (LOESS) with shaded 95% CI (fatal group CI wide, not presented). **(E)** Frequency of blood M-MDSCs in paired acute and convalescent samples from COVID-19 patients (n=6). **(F)** M-MDSC frequency in blood, NPA and ETA samples from severe (red, n=16) and fatal (grey, n=4) COVID-19 patients. **(G-I)** Surface expression on M-MDSCs from blood, NPA and ETA in HCs (blue, NPA n=7, PBMC, n=11) and COVID-19 patients (black, NPA n=25, ETA n=19, PBMC, n=69) of **(G)** CD62L **(H)** CD86 **(I)** CCR2. **(J)** Frequency of PMN-MDSCs of live CD45+ cells in blood from COVID-19 patients. HCs: n=12. COVID-19 patients: mild (n=11), moderate (n=47), severe (n=42) and fatal (n=8). **(K)** Frequency of blood PMN-MDSCs in paired acute and convalescent samples from COVID-19 patients (n=6). **(B, C, F-J)** Comparisons of M-MDSC frequencies were performed using the non-parametric Kruskal-Wallis test and subsequent Dunn's post-hoc test of multiple comparisons. In strip charts, group medians are presented as horizontal lines and individual patients as jitter points.

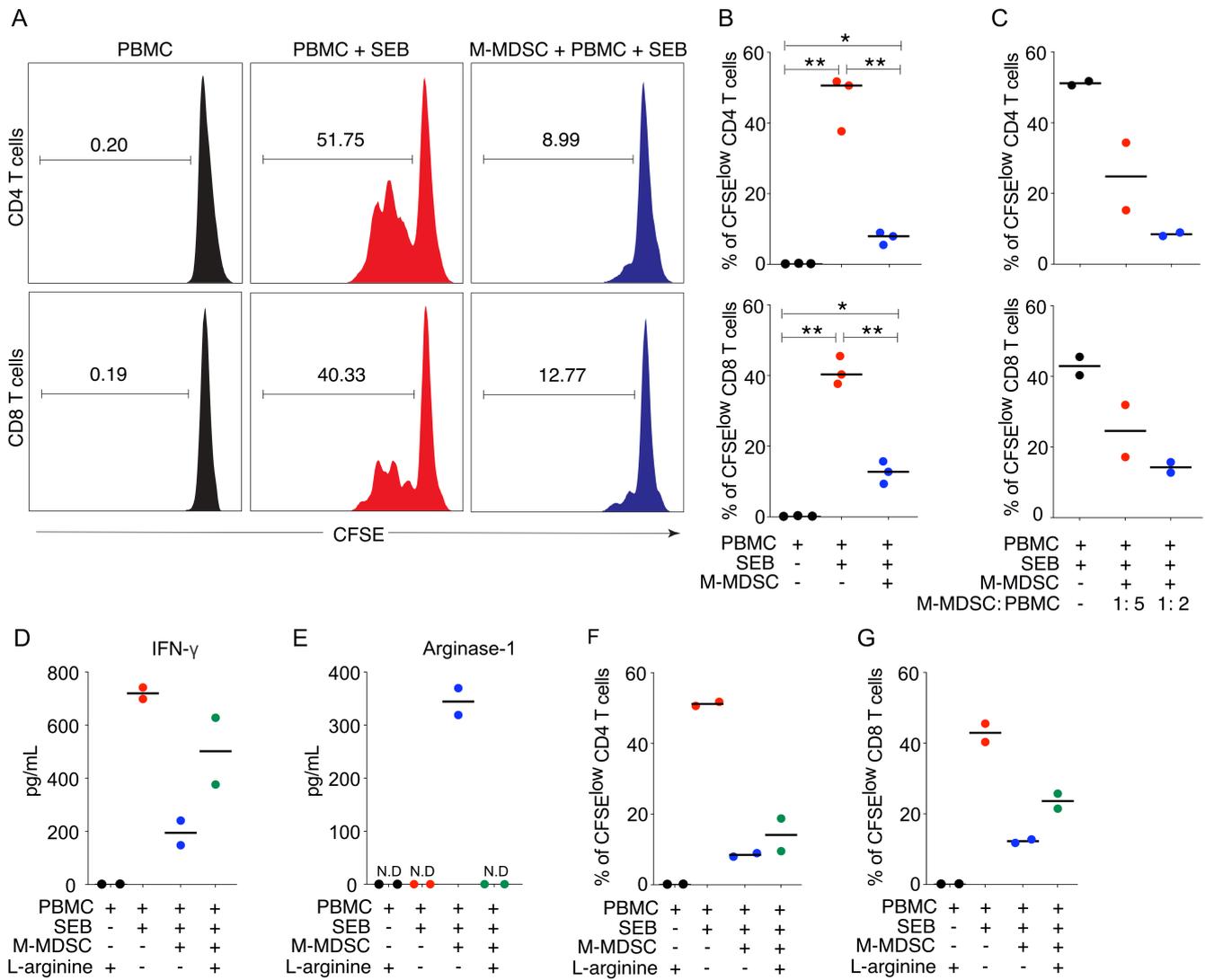


Figure 3. M-MDSCs isolated from COVID-19-patients suppress T cell proliferation partly through release of Arginase-1. (A) Blood M-MDSC isolated from COVID-19 patients were co-cultured with CFSE-labeled allogenic PBMCs in the presence of staphylococcal enterotoxin B (SEB) for 3 days, at a ratio of 1:2 (M-MDSC:PBMC). Histograms show representative CD4 and CD8 T cell proliferation as assessed by CFSE dilution and flow cytometry. Number indicate frequency of proliferating T cells. (B) Dot plots showing percentage of proliferating CD4 and CD8 cells with median (n=3). Statistical testing performed using Wilcoxon signed ranks test. (C) Isolated M-MDSC were cultured with CFSE-labelled allogenic PBMCs in the presence of SEB for 3 days. The ratio of M-MDSC: PBMC were 1:5 and 1:2. Dot plots show percentage of proliferating CD4 T and CD8 cells with median (n=2). (D) Dot plots show levels of IFN γ in supernatants from cell cultures with median (n=2). (E to G) Isolated M-MDSC were cultured with CFSE-labelled allogenic PBMCs in the presence of SEB and L-arginine for 3 days. The ratio of M-MDSC: PBMC was 1:2. (E) Dot plots show levels of Arginase-1 in supernatants from cell cultures with median (n=2). N.D.=not detectable. (F and G) Dot plots show percentage of proliferating (F) CD4 T cells and (G) CD8 T cells (n=2) with median (n=2).

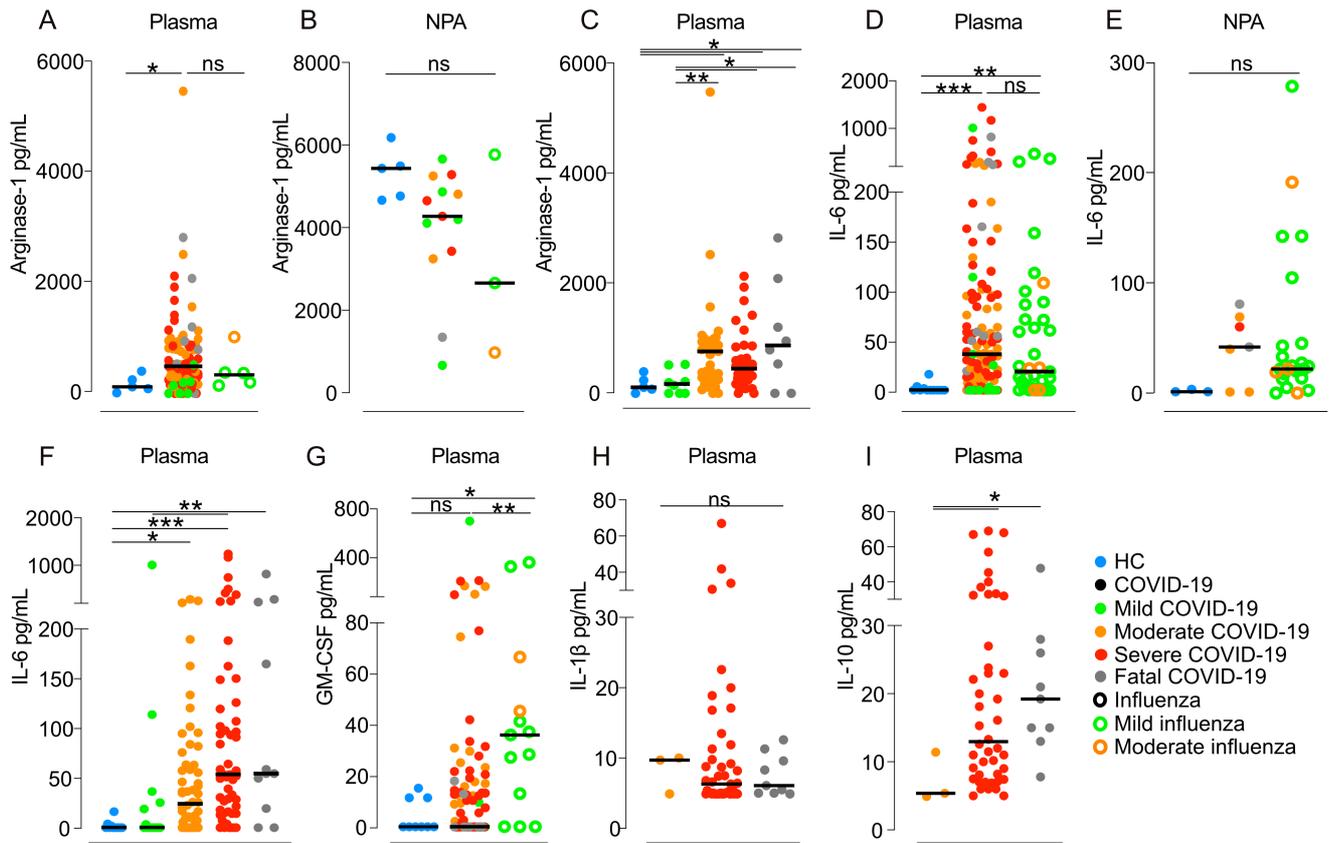


Figure 4. Levels of cytokines in blood and NPA from HCs, COVID-19 patients and influenza patients. (A-C) Arginase-1 was measured in **(A)** plasma, **(B)** NPA and **(C)** across COVID-19 disease severity in plasma. HCs (blue points): n=5 (blood), 5 (NPA). Influenza patients (open circles): n=6 (blood), 3 (NPA). COVID-19 patients (closed circles): n=93 (blood), 13 (NPA). COVID-19 patients (color-coded by peak severity): mild (n=8), moderate (n=41), severe (n=36) and fatal (n=8). **(D-F)** IL-6 as measured in **(D)** plasma, **(E)** NPA and **(F)** IL-6 across COVID-19 disease severity in plasma. HCs (blue): n=11 (blood), 3 (NPA). Influenza patients (open circles): n=37 (blood), 24 (NPA). COVID-19 patients (closed circles): n=133 (blood), 7 (NPA). COVID-19 patients: mild (n=14), moderate (n=56), severe (n=52) and fatal (n=11). **(G)** GM-CSF in plasma comparing HCs and patients. HCs (blue): n=9. Influenza patients (open circles): n=13. COVID-19 patients (closed circles, color-coded by peak severity): mild (n=12), moderate (n=38), severe (n=48) and fatal (n=8). **(H)** IL-10 and **(I)** IL-1B in plasma from patients with moderate to severe disease or fatal outcome. COVID-19 patients: moderate (n=3), severe (n=44) and fatal (n=9). **(A-I)** Medians were compared using the non-parametric Kruskal-Wallis test. Post-hoc testing was carried out while controlling the False Discovery Rate **(A-C)** or by using Dunn's test of multiple comparisons **(D-I)**. In strip charts, group medians are presented as horizontal lines and individual patients as jitter points.

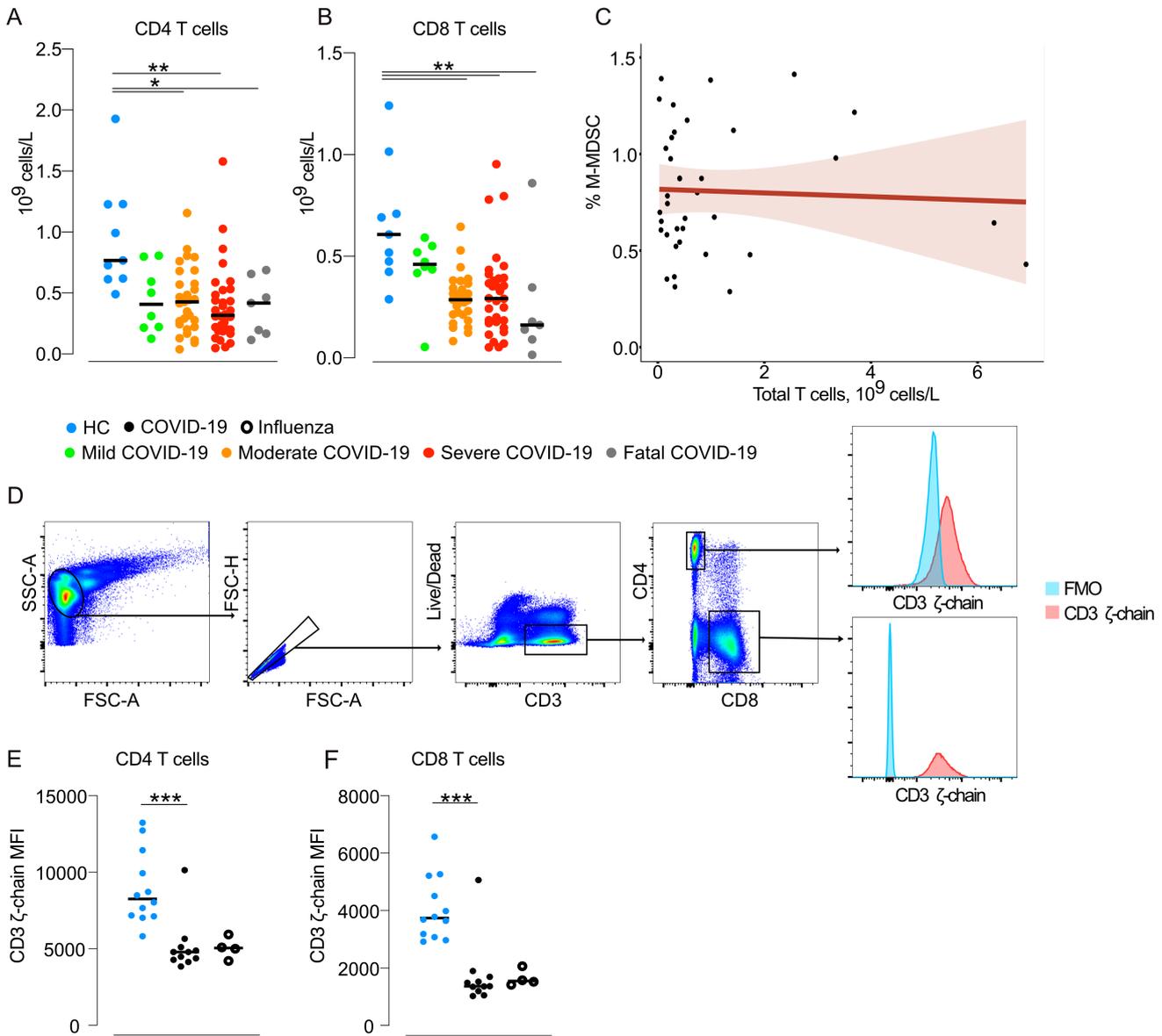


Figure 5. T cells in COVID-19 patients. (A-B) Total lowest CD4 and CD8 T cell count in blood was calculated in HCs (n=9) and COVID-19 patients across disease severity, mild (n=8), moderate (n=29), severe (n=32), fatal (n=7). (C) Spearman correlation between total CD3 $^+$ T cell count in blood and peak M-MDSC frequency. (D) Gating strategy for CD3 ζ chain analysis. Live, single CD3 $^+$ cells were identified, and separated into CD4 $^+$ or CD8 $^+$ T cells. MFI was calculated for both populations. The blue histogram represents the fluorescence-minus-one control (FMO) and the red histogram represents CD3 ζ (PE-labelled). (E-F) Thawed PBMCs from 11 COVID-19 patients, 4 influenza patients and 12 matched healthy controls were intracellularly stained for CD3 ζ chain expression. Median fluorescent intensity (MFI) was calculated for (E) CD4 $^+$ and (F) CD8 $^+$ T cells. Comparison of medians between groups was performed using Kruskal-Wallis' test and subsequent post-hoc testing by Dunn's test of multiple comparisons.

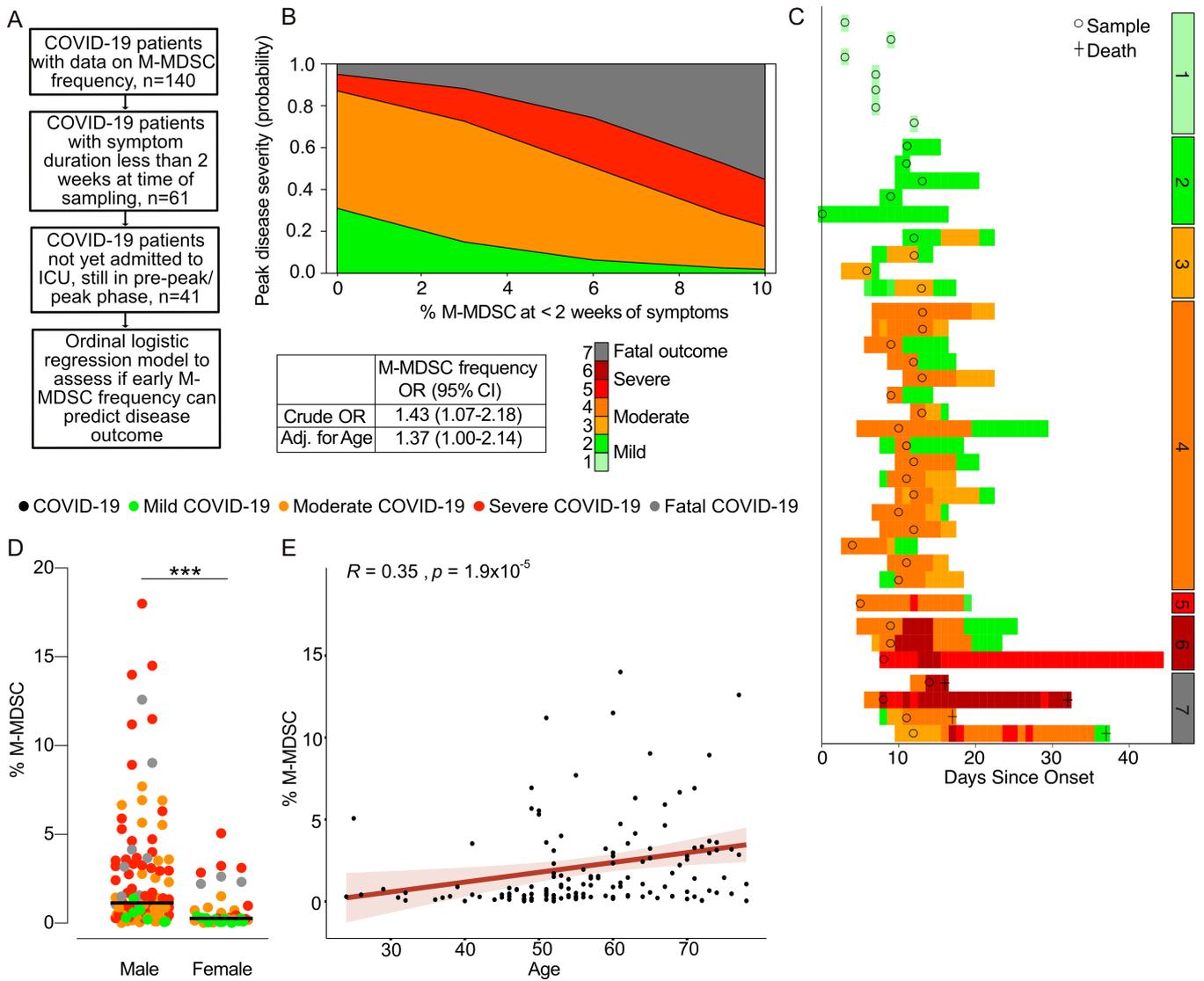


Figure 6. M-MDSC frequency predicts disease severity and is associated with male gender and age. (A) Criteria for inclusion in the ordinal logistic regression model. Only patients with up to 2 weeks of symptoms, not yet admitted to the ICU and in the pre-peak or peak phase were included in the analysis **(B)** Proportional odds logistic regression showing predictive capacity of M-MDSC-frequency on peak disease severity score. Crude and adjusted odds ratio presented. **(C)** Overview of the 41 patients included in the ordinal logistic regression model. Circles represent the samples included in the analysis and color represents daily disease severity. Patients are separated based on peak disease severity. **(D)** Peak M-MDSC frequency in men and women with COVID-19 compared using Wilcoxon-Mann-Whitney-U, circles color coded by disease severity. Men: mild (n=7), moderate (n=40), severe (n=47) and fatal (n=9). Women: mild (n=12), moderate (n=13), severe (n=9) and fatal (n=3). **(E)** Spearman correlation between age and peak M-MDSC frequency (n=140).

Table 1. Patient and control characteristics

Cohort	COVID-19	Influenza	HC	Sig. ^A
<i>n</i>	147	44	33	
Age, mean (range)	57 (24-78)	52 (18-88)	51 (24-80)	*
Male gender, <i>n</i> (%)	109 (74%)	19 (43%)	23 (70%)	
Onset to admission, days, mean (SD)	10 (5.8)	5 (3.2)	-	*
Onset to inclusion, days, mean	18 (10)	5 (2)	-	***
Co-morbidities				
CCI, mean (SD)	2 (2)	2 (2)	-	ns
BMI, mean (SD)	28.5 (4.6)	26.7 (4.9)	25.0 (3.3)	*
Hypertension, <i>n</i> (%)	55 (38%)	14 (33%)	0	ns
Diabetes, <i>n</i> (%)	34 (23%)	4 (9.1%)	0	ns
Current smoker, <i>n</i> (%)	10 (7.0%)	7 (16%)	0	ns
Laboratory analyses				
CRP (mg/L), mean (SD) ^{A,B}	206 (134)	119 (146)	2 (3)	***
WBC (x10 ⁹ /L), mean (SD) ^{A,C}	8.0 (4.3)	6.8 (3.5)	6.5 (1.5)	ns
Lymphocytes (x10 ⁹ /L), mean (SD) ^{A,C}	0.89 (0.69)	1.06 (0.47)	2.41 (1.24)	***
Neutrophils (x10 ⁹ /L), mean (SD) ^{A,C}	6.4 (3.9)	5.1 (3.5)	3.1 (0.7)	**
Monocytes (x10 ⁹ /L), mean (SD) ^{A,C}	0.44 (0.27)	0.64 (0.35)	0.54 (0.24)	**
Outcome				
Peak severity score ^D , mean (SD)	4.49 (1.65)	1.80 (0.82)	-	***
Fatal outcome, <i>n</i> (%)	12 (8.1)	0	0	

^AStatistical tests performed: One-way ANOVA; Fisher's exact test; Pearson chi-square test.

^BPeak value ^CValue at maximum lymphopenia. ^DPeak severity score on 7-grade composite scale (see Methods). Abbreviations: CCI=Charlson co-morbidity index, CRP=C-reactive protein. Normal range: CRP<3 mg/L, WBC 3.5-8.8x10⁹/L, lymphocytes 1.1-3.5 x10⁹/L, neutrophils 1.6-5.9 x10⁹/L, monocytes 0.2-0.8 x10⁹/L.

Table 2. Baseline characteristics of COVID-19 patients across disease severity.

Peak disease severity	Mild		Moderate		Severe		Fatal	Sig ^A
	1	2	3	4	5	6	7	
<i>n</i> (%)	13 (8.8)	6 (4.1)	10 (6.8)	48 (33)	19 (13)	39 (27)	12 (8.2)	
Age, mean (Range)	44 (24-72)	60 (41-72)	56 (46-78)	55 (24-76)	57 (42-74)	61 (25-77)	66 (52-78)	***
Male, <i>n</i> (%)	5 (38)	2 (33)	6 (60)	38 (79)	15 (79)	34 (87)	9 (75)	**
Onset to adm. ^B , mean (SD)	-	10.4 (2.2)	9.4 (4.1)	10.3 (4.1)	7.6 (2.9)	11.0 (8.1)	9.9 (7.6)	ns
Cortisone during sample period	0	0	0	4 (8%)	4 (21%)	12 (31%)	2 (17%)	*
Length of stay ^B , mean (SD)	-	5 (4)	6 (4)	12 (7)	25 (10)	33 (17)	-	***
Co-morbidities								
CCI, mean (SD)	1 (2)	2 (1)	1 (1)	2 (2)	2 (1)	2 (1)	3 (1)	*
BMI, mean (SD)	24.1 (4.5)	25.1 (2.2)	26.0 (3.2)	30.3 (4.2)	29.2 (5.3)	28.6 (4.7)	28.6 (2.4)	***
Hypertension, <i>n</i> (%)	1 (8.3)	0 (0)	2 (20)	20 (42)	8 (42)	15 (38)	9 (75)	**
Diabetes, <i>n</i> (%)	2 (17)	0 (0)	1 (10)	14 (29)	5 (26)	9 (23)	3 (25)	ns
Current smoker, <i>n</i> (%)	0 (0)	0 (0)	2 (20)	5 (11)	1 (5.3)	2 (5.3)	0 (0)	ns
Immunosuppressant, <i>n</i> (%)	1 (7.7)	0 (0)	0 (0)	4 (8.3)	2 (11)	5 (13)	1 (8.3)	ns
Laboratory analyses^C								
CRP ^D , mean (SD)	6 (10)	102 (54)	137 (143)	161 (87)	236 (114)	288 (106)	364 (119)	***
WBC ^E , mean (SD)	4.7 (1.6)	6.0 (1.9)	6.8 (4.9)	7.2 (2.6)	7.6 (2.7)	9.7 (5.7)	10.8 (5.3)	**
Neutrophils ^E , mean (SD)	2.4 (1.3)	4.7 (1.9)	5.4 (5.0)	5.5 (2.2)	6.2 (2.6)	8.1 (4.7)	9.7 (5.0)	***
Lymphocytes ^E , mean (SD)	1.68 (0.45)	0.76 (0.30)	0.97 (0.28)	1.06 (0.92)	0.68 (0.31)	0.72 (0.52)	0.44 (0.19)	***
Monocytes ^E , mean (SD)	0.48 (0.16)	0.34 (0.19)	0.42 (0.25)	0.49 (0.28)	0.43 (0.27)	0.44 (0.29)	0.30 (0.23)	ns
Ct-value, mean (SD)	27 (7)	24 (6)	27 (6)	26 (6)	26 (5)	24 (7)	21 (6)	ns
Seroconversion ^F , <i>n</i> (%)	9 (75)	4 (67)	9 (100)	36 (97)	17 (100)	37 (100)	10 (91)	**

^AStatistics performed: One-way ANOVA, Fisher's exact test. ^B days, ^CPeak values: CRP; Nadir values: lymphocyte count, Ct-value; WBC, neutrophil and monocyte counts at timepoint for lowest lymphocyte count. ^Dmg/L, ^E10⁹ cells/L. ^FAt any sampling timepoint. Abbreviations: CCI=Charlson co-morbidity index, CRP=C-reactive protein. Normal range: CRP <3 mg/L, WBC 3.5-8.8x10⁹/L, lymphocytes 1.1-3.5 x10⁹/L, neutrophils 1.6-5.9 x10⁹/L, monocytes 0.2-0.8 x10⁹/L.