

Age-related changes in plasma biomarkers and their association with mortality in COVID-19

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GRAPHICAL ABSTRACT Overview of the study findings. TNF-R1: tumour necrosis factor receptor 1; TREM-1: triggering receptor expressed on myeloid cells 1. Figure partially created with BioRender.com.

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for ∼89% of these deaths [\[4\]](#page-14-0). The biological substrate behind this steep increase is currently unknown. Ageing-related comorbidities mediate only a fraction of the increased mortality [[2](#page-14-0), [5\]](#page-14-0).

COVID-19 is associated with increased concentrations of inflammation markers and cytokines, coagulation disturbances, and endotheliopathy [[6](#page-14-0), [7\]](#page-14-0). Multiple randomised controlled trials (RCTs) that sought to ameliorate these host response disturbances have reported improved clinical outcomes [[8](#page-14-0)]. However, in the largest RCTs, the two current cornerstones of treatment for hospitalised COVID-19 patients, namely dexamethasone and tocilizumab, showed little if any beneficial effect in the oldest patients [\[9](#page-14-0)–[11\]](#page-14-0). In general, ageing is associated with changes in innate and adaptive immunity; in this context, "inflammageing" refers to a state of sustained low-grade inflammation, while "immunosenescence" refers to a gradual decline in the ability to generate an effective immune response to new antigens [\[12](#page-14-0)]. Although the impact of ageing on the immune system has been implicated in the pathogenesis of COVID-19 in elderly patients [[13,](#page-14-0) [14](#page-15-0)], studies analysing the influence of older age on host response disturbances in COVID-19 are scarce. Previous studies focused on plasma cytokine levels in a limited number of patients across various disease severities, showing an association of ageing with an increased inflammatory response [\[15](#page-15-0)–[18\]](#page-15-0). The influence of ageing on endothelial and coagulation responses in COVID-19 has yet to be evaluated. Exploring age-driven differences in these key host response domains and their association with ageing-driven mortality in COVID-19 may lay a foundation for identifying new immunomodulating targets.

The primary objective of this study was to gain insight into the association between age and aberrations in key pathophysiological pathways implicated in COVID-19 in patients admitted to a general hospital ward (i.e. noncritically ill). To this end, we determined 43 biomarkers reflective of alterations in the endothelial and coagulation response, inflammation and organ damage, and cytokine and chemokine release. Our secondary objective was to associate the host response aberrations detected in older patients with the increased 30-day mortality rate in this age group.

Methods

Study design and population

Data were derived from the ELDER-BIOME study [\(ClinicalTrials.gov](https://clinicaltrials.gov/): NCT02928367) and the Amsterdam University Medical Center (UMC) COVID-19 Biobank Study (AUMC 2020_065) (see [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) for details). Both studies were approved by the Amsterdam UMC Ethics Committee. ELDER-BIOME is a prospective observational study in Amsterdam UMC (both locations: Academic Medical Center (AMC) and Vrije Universiteit Medical Center (VUMC), Amsterdam, The Netherlands) and Flevo Hospital (Almere, The Netherlands) [[7\]](#page-14-0). The Amsterdam UMC COVID-19 Biobank Study is conducted in both AMC and VUMC, according to which leftover blood from clinical care is processed for plasma storage. Written informed consent was obtained from all patients or their legal representatives. Within the Amsterdam UMC Biobank Study, an additional option of deferred consent using an opt-out method was implemented for incapacitated patients. We included patients above 18 years of age with COVID-19-related symptoms and a positive severe acute respiratory syndrome coronavirus 2 $(SARS-CoV-2)$ PCR or suspected COVID-19 with a CO-RADS computed tomography score ≥ 4 reflecting a high (4) or very high (5) suspicion for pulmonary involvement of COVID-19 [\[19](#page-15-0)]. Inclusion was done between February 2020 and September 2021. Patients were sampled within 48 h of general ward admission (or, in a separate cohort, intensive care unit (ICU) admission). Patients were divided according to four "waves" (for details, see [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) and [\[20](#page-15-0)]). Biomarker data were validated in an external cohort (for details, see [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) and [\[21](#page-15-0)]). [Figure 1](#page-3-0) depicts an overview of the COVID-19 cohorts. In addition, mortality data were also obtained from patients admitted to a general ward because of community-acquired pneumonia caused by pathogens other than SARS-CoV-2 (for details, see [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) and [\[7\]](#page-14-0)).

Assays

43 host response biomarkers were measured in EDTA anticoagulated plasma using Luminex (R&D, Minneapolis, MN, USA) and the Bio-Plex 200 System (Bio-Rad, Hercules, CA, USA) (for the selection rationale, see [supplementary methods\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). Biomarkers were stratified into four pathophysiological domains (for details, see [supplementary table S1](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)). Reference values were obtained from 10 heathy and 19 age- and sex-matched noninfectious outpatient clinic controls ([ClinicalTrials.gov:](https://clinicaltrials.gov/) NCT02928367).

Statistics

Patients were stratified according to age decades: $\langle 50, \rangle 50 - \langle 60, \rangle 60 - \langle 70, \rangle 70$ years. This method was chosen to facilitate visualisation and clinical interpretation. Differences in 30-day survival were visualised by Kaplan–Meier curves. Biomarker data were log-transformed. Overall differences between age decades among biomarkers within one host response domain were visualised by a principal component

FIGURE 1 Overview of the three COVID-19 cohorts in which plasma biomarkers were measured. The external validation cohort was derived from a publicly available dataset of nonintubated COVID-19 patients in whom plasma proteins were measured by the Olink proximity extension assay [[21\]](#page-15-0). ICU: intensive care unit. Figure partially created with BioRender.com.

analysis (PCA), as described previously [\[7\]](#page-14-0). Differences in the principal component (PC) scores between age groups were analysed by ANOVA. The correlation of PC scores with ageing on a continuous scale was determined by Spearman's correlation test.

Differences in individual biomarker levels between age decades were quantified using Hedges' g effect size and visualised using heatmaps [\[22](#page-15-0)]. Additionally, in a regression analysis, age was modelled as a continuous variable. The association's strength was analysed using Spearman's correlation. In a separate analysis, the association of individual biomarkers with ageing was also explored upon admission to the ICU.

We investigated if age-driven alterations in biomarker concentrations were associated with the age-driven increase in 30-day mortality rates by mediation analysis [[5](#page-14-0), [23](#page-15-0)]. A biomarker or PC score needed to be significantly associated with both ageing and 30-day mortality to enter the mediation analysis (for assumptions and details, see [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)) [[24\]](#page-15-0). Biomarkers associated with both ageing and mortality were also externally validated in an independent cohort entailing 196 hospitalised nonintubated COVID-19 patients (for cohort description, see [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)) (figure 1) [[21\]](#page-15-0). Lastly, we performed a cluster analysis on patients aged ≥ 70 years to assess the uniformity of their host response, using Ward's method [\[25](#page-15-0)]. Furthermore, we investigated whether the mediating effects of biomarkers were influenced by the patient's host response phenotype.

The regression and mediation models all consisted of an unadjusted and an adjusted approach. Adjusted models evaluating the association of ageing with biomarker concentrations contained covariates associated with either ageing or COVID-19 care and included: demographics (inclusion hospital, sex and inclusion wave), age-related comorbidities (hypertension, diabetes, malignancies, immunosuppression, and chronic cardiac, neurological, respiratory and kidney disease), age- and biomarker-related chronic medication prior to admission (antiplatelet and anticoagulant drugs), and COVID-19-related immunomodulating treatments before sampling (corticosteroids, anti-interleukin (IL)-6 and imatinib). When evaluating the association of ageing with mortality, we used the same covariates plus covariates that may impact mortality in the adjusted model: immunomodulating treatments on admission but after sampling and the use of antibiotics and remdesivir (for details and assumption testing, see [supplementary methods\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials).

Details on missingness and handling of missingness are described in the [supplementary methods](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials), [supplementary Excel file](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) (sheet 1), and [supplementary tables S2 and S3](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). p-values of all analyses were multiple testing corrected using the Benjamini–Hochberg (BH) procedure. A BH-adjusted p-value <0.05 was considered statistically significant.

Results

Patients, presentation and outcome

464 patients with COVID-19 admitted to a general hospital ward were included ([table 1](#page-4-0)). Of these, 89 patients (19.2%) were aged ≤ 50 years at hospital admission, 111 (23.9%) $\geq 50 - \leq 60$ years, 135 (29.1%) ≥ 60 - <70 years and 129 (27.8%) ≥ 70 years. The distribution of sex did not differ between age decades. Most patients were enrolled during the second and third COVID-19 waves in the Netherlands, in which the SARS-CoV-2 alpha variant occurred and became dominant. The age distribution was similar between

Data are presented as median (interquartile range) or n (%), unless otherwise stated. BMI: body mass index; qSOFA: quick Sequential Organ Failure Assessment; MEWS: Modified Early Warning Score; CURB: confusion, urea >7 mmol·L⁻¹, respiratory rate ≥30 breaths·min⁻¹, blood pressure
<90 mmHg (systolic) or ≤60 mmHg (diastolic). [#]: prior to admission; [¶]: Charlson history of an organ transplant, immune deficiency or chronic use of immunosuppressants (see [supplementary table S7](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) for details); [§]: 4C Mortality, a validated COVID-19 score [\[26](#page-15-0)], was calculated without the age and obesity component; ^f: CURB score was calculated without the age component.

> waves [\(supplementary table S4](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)). The proportion of patients with comorbidities increased with age. While the duration of symptoms prior to admission did not differ between age groups, older patients presented with higher disease severity scores [\[26](#page-15-0)]. Routine laboratory values were comparable between age groups except for an age-dependent decrease in lymphocyte counts and an age-dependent increase in the neutrophil/lymphocyte ratio and creatinine. COVID-19-related treatments, primarily supplemental oxygen and dexamethasone, were comparable between age groups [\(table 2\)](#page-5-0). Mortality rates increased with age in a nonlinear fashion [\(table 2](#page-5-0) and [figure 2\)](#page-6-0). For example, an increase in age from 60 to 70 years was

Data are presented as median (interquartile range) or n (%), unless otherwise stated. SARS-CoV-2: severe acute respiratory syndrome coronavirus 2; ICU: intensive care unit. [#]: prednisolone and hydrocortisone; ": numbers do not add up to 100% as some patients suffered from both pulmonary embolism and deep venous thrombosis; ⁺: ICU admission after sampling; ^s: for any cause within 28 days of the initial admission; ^f: for 98.2% of patients, worsening of COVID-19 was reported as a causal element for mortality by the treating physician.

> associated with an increased 30-day mortality (OR 1.15, 95% CI 1.12–1.18) [\(figure 2b\)](#page-6-0), which was independent of demographics, age-related comorbidities and COVID-related treatments (adjusted OR 1.14, 95% CI 1.11–1.17) ([figure 2c](#page-6-0)). In pneumonia patients with pathogens other than SARS-CoV-2 [\(supplementary table S5](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)), a similar increase in age (from 60 to 70 years) was not associated with an increase in 30-day mortality odds (OR 1.01, 95% CI 0.996–1.025). Moreover, older patients with pneumonia caused by pathogens other than SARS-CoV-2 showed significantly lower mortality rates compared with those with SARS-CoV-2: $\geq 60 - \leq 70$ years ($p \leq 0.01$) and ≥ 70 years ($p \leq 0.001$) [\(supplementary figure S1\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials).

Association of ageing with aberrations in distinct host response domains

We determined 43 host response biomarkers in plasma obtained within 48 h after admission and stratified these in four pathophysiological domains. First, we performed PCA to visualise overall differences between age decades among biomarkers within each pathophysiological domain ([figure 3\)](#page-7-0). All domains showed significant differences between age groups. The strongest association with ageing was observed for PC1 of the systemic inflammation and organ damage domain ($p = -0.41$), which was mainly driven by increased plasma concentrations of soluble tumour necrosis factor receptor 1 (sTNF-R1), soluble triggering receptor expressed on myeloid cells 1 (sTREM-1) and soluble receptor for advanced glycation end-products (sRAGE) in older patients ([figure 3b\)](#page-7-0). The second strongest association was that of ageing with PC2 of the endothelial and coagulation domain ($p=0.30$), which was driven mainly by increased

FIGURE 2 Mortality analysis of COVID-19 patients admitted to the general ward stratified by age decade. a) Kaplan–Meier plot of patients stratified by age decade. b) Risk of 30-day mortality with age modelled as a continuous variable. Given the nonlinear relationship between age and mortality, a restricted cubic spline function with three inner knots at default quantile locations was used. To calculate the odds ratio, the reference was set to 60 years of age. c) The same method as b), but now the 30-day mortality odds ratio is adjusted for demographics (inclusion hospital, sex and inclusion wave), age-related comorbidities (hypertension, diabetes, malignancies, immunosuppression, and chronic cardiac, neurological, respiratory and kidney disease), age-related chronic medication (antiplatelet and anticoagulant drugs), and COVID-19-related treatments both before and after sampling (corticosteroids including dexamethasone, anti-interleukin-6, imatinib, remdesivir and antibiotics).

plasma angiopoietin (ANG)-2/1 ratios, soluble thrombomodulin (sThrombomodulin) and soluble E-selectin (sE-selectin) in older patients ([figure 3a\)](#page-7-0). The cytokine domain was characterised by two groups of cytokines: one in which the levels increased with ageing ([figure 3c](#page-7-0), PC2), primarily driven by granulocyte–macrophage colony-stimulating factor (GM-CSF) and the anti-inflammatory cytokines IL-1RA and IL-10, and one in which the levels were relatively similar between age groups ([figure 3c;](#page-7-0) PC1). A mixed pattern was observed for the chemokine domain in which some chemokine levels increased with ageing (e.g. C-X-C motif chemokine ligand 10 (CXCL10)), while others decreased with ageing (e.g. C-C motif chemokine ligand 5 (CCL5)) [\(figure 3d](#page-7-0); PC2). The complete contribution of each biomarker to a PC score is depicted in [supplementary table S6](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). As a sensitivity analysis, we further stratified patients \geq 70 years into patients \geq 70– <80 years (n=86) and \geq 80 years (n=43), which did not modify the age-dependent trends to a significant extent [\(supplementary figure S2\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials).

Association of ageing with individual host response biomarkers

We next compared individual host response biomarkers across age decades. Of the 43 determined biomarkers, 36 were significantly different between patients and controls [\(supplementary Excel file](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials), sheet 2). [Figure 4a](#page-9-0) shows biomarker concentrations in patients \geq 70 years compared with the other age groups, expressed as Hedges' g, a commonly used effect size measure [[22\]](#page-15-0). This analysis confirmed that most biomarkers reflective of endothelial and coagulation activation, as well as those indicating systemic inflammation and organ damage, were higher in patients \geq 70 years, while chemokines and cytokines demonstrated a mixed pattern. A sensitivity analysis in which we further stratified patients \geq 70 years into patients ≥70– <80 and ≥80 years yielded similar results [\(supplementary figure S2](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)). For almost all biomarkers that had a significant association with ageing in the unadjusted model, significance was maintained after adjusting for demographics, comorbidities and COVID-19-related immunomodulating treatments before sampling (see Methods for details). On a continuous scale, sTNF-R1, sTREM-1, sThrombomodulin and tenascin-C demonstrated the strongest (positive) correlations with ageing [\(figure](#page-9-0) [4b\)](#page-9-0). The strength and direction of the associations of each biomarker with ageing showed strong similarities across inclusion waves ([supplementary figure S3\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials).

FIGURE 3 Principal component analysis (PCA) of host response domain differences between age groups: a) endothelial and coagulation response, b) systemic inflammation and organ damage, c) cytokines and d) chemokines. Principal components (PCs) 1 and 2 are plotted per domain. For each domain, the x- and y-axes are labelled with the percentage of the total variance within that domain that is explained by PC1 and PC2. respectively. The complete contribution of each biomarker to a PC score is depicted in [supplementary table S6.](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) The ellipse indicates the central 10% of each age group, colour coded as indicated in the key at the bottom of the figure. The arrows indicate the direction (arrow orientation) and strength (arrow length) of the correlation between each biomarker and the PCs. Next to each PCA plot are box plots with 1.5 interquartile range whiskers of PC1 and PC2. Herein, upper p-values were obtained by ANOVA between age groups: p-values with accompanying p-values were generated using a Spearman's correlation with ageing on a continuous scale. Note that a negative association of a PC with ageing may still reflect a positive association with biomarker concentration, as reflected by the direction of the arrows. Post-hoc testing was done with a Tukey test. *: p<0.05; **: p<0.01; ***: p<0.001. ANG: angiopoietin; sTie-2: soluble Tie-2; sE-selectin: soluble E-selectin; sThrombomodulin: soluble thrombomodulin; sVCAM-1: soluble vascular cellular adhesion molecule 1; PAI-1: plasminogen activator inhibitor 1; sCD31: soluble cluster of differentiation 31; sRAGE: soluble receptor for advanced glycation end-products; sTNF-R1: soluble tumour necrosis factor receptor 1; sTREM-1: soluble triggering receptor expressed on myeloid cells 1; SP-D: surfactant protein D; CD40L: CD40 ligand; PD-L1: programmed death ligand 1; CCL: C-C motif chemokine ligand; CXCL: C-X-C motif chemokine ligand; IL: interleukin; TNF: tumour necrosis factor; GM-CSF: granulocyte–macrophage colony-stimulating factor; IFN: interferon.

> In a separate analysis, we evaluated the age distribution of the same biomarkers in a cohort of critically ill COVID-19 patients sampled on ICU admission (n=157) ([supplementary tables S8 and S9](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials), and [supplementary Excel file,](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) sheet 3). Differences between patients ≥ 70 years and younger patients were largely reproduced in this cohort of critically ill patients, especially compared with the youngest age group (<50 years), although, due to the relatively low sample size, statistical significance was not always reached [\(supplementary figure S4a\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). Akin to the patients admitted to a general ward, sTNF-R1, sTREM-1, sThrombomodulin, tenascin-C and soluble vascular cellular adhesion molecule 1 (sVCAM-1) demonstrated a significant positive correlation with ageing ([supplementary figure S4b](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials)).

Association of age-driven host response alterations with mortality

We used mediation analysis to evaluate whether age-driven alterations in biomarker concentrations were associated with the age-driven increase in 30-day mortality rates [\[23](#page-15-0)]. First, we identified biomarkers associated with both ageing and 30-day mortality before and after adjustment for possible confounding and ageing-related factors ([figure 5a](#page-10-0) and b). Then, we estimated which proportion of the age effect on mortality was associated with age-related alterations in PC scores of pathophysiological domains and individual host response biomarkers ([figure 5c](#page-10-0)). In the unadjusted model ([figure 5d](#page-10-0)), PC scores of the endothelial and coagulation (PC2), systemic inflammation and organ damage (PC1), and chemokine (PC2) and cytokine (PC2) domains explained a significant part of the age effect on mortality. With regard to individual biomarkers, ageing-related differences in sThrombomodulin, sVCAM-1, sTNF-R1, sTREM-1 and sRAGE were most associated with the age-driven increase in 30-day mortality. Notably, in an external cohort, sThrombomodulin, sVCAM-1, sTNF-R1 and tenascin-C showed the strongest association with ageing and 30-day mortality ([supplementary figure S5\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials); sTREM-1 was not measured [\[21](#page-15-0)]. In the adjusted model [\(figure 5e](#page-10-0)), only ageing-driven changes in sVCAM-1, sRAGE, CXCL10, CCL2, GM-CSF and IL-10 were associated with the ageing-driven increase in 30-day mortality. Collectively, these results demonstrate that a proportion of the association between age and mortality is explained by specific host response differences.

Given that the host response in older patients may not be uniform, we performed a cluster analysis in patients \geq 70 years. We identified three clusters ([supplementary figure S6\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). Cluster 3 had the highest age and 30-day mortality, with high endothelial cell activation, inflammation and organ damage markers but low cytokine concentrations ([supplementary tables S10 and S11,](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) and [supplementary Excel file,](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) sheet 4). Cluster 1 had high coagulation markers and cytokine concentrations [\(supplementary figure S6\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). Cluster 2 had a mixed host response. The top three mediating biomarkers in the primary analysis (sThrombomodulin, sTNF-R1 and sTREM-1) were important for assigning patients to Cluster 3 [\(supplementary table S12\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials). Furthermore, the majority of biomarkers and all PC scores that were significant mediators in the primary analysis remained predictors of mortality in patients \geq 70 years, which was, with a few exceptions, independent of the assigned cluster [\(supplementary table S13\)](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials).

Discussion

We report the association of ageing with 43 biomarkers reflective of host response disturbances in four key pathophysiological domains relevant for the pathogenesis of COVID-19 and relate these immune deviations to 30-day mortality. While our results confirm previously reported increased plasma concentrations of IL-6, IL-10, IL-15, TNF-α and CXCL8 in older COVID-19 patients [[15](#page-15-0)–[18](#page-15-0)], we

FIGURE 4 Association of host response biomarkers with ageing. a) Heatmap depicting the magnitude of biomarker differences (Hedges' g) between patients >70 years compared with other age groups. p-values were obtained from a linear (if linear) or cubic spline regression analysis (if nonlinear) in which age was modelled as a continuous variable. The adjusted model included demographics, age-related comorbidities, age and biomarker-related chronic medication, and COVID-19-related immunomodulating treatments before sampling (see Methods for details). Red indicates higher levels in patients \geqslant 70 years and blue indicates lower levels in patients \geqslant 70 years. $\overset{\ast}{\texttt{.}}$ biomarkers with a nonlinear relationship with ageing on a continuous scale. b) Volcano plot depicting the strength of the correlation between a biomarker and ageing. Red dots represent a significant positive correlation, blue dots represent a significant negative correlation and grey dots represent a nonsignificant correlation. Both the adjusted and unadjusted p-values are multiple testing corrected using the Benjamini–Hochberg (BH) procedure for testing 43 biomarkers. *: p<0.05; **: p<0.01; ***: p<0.001. ANG: angiopoietin; sTie-2: soluble Tie-2; sE-selectin: soluble E-selectin; sThrombomodulin: soluble thrombomodulin; sVCAM-1: soluble vascular cellular adhesion molecule 1; PAI-1: plasminogen activator inhibitor 1; sCD31: soluble cluster of differentiation 31; sRAGE: soluble receptor for advanced glycation end-products; sTNF-R1: soluble tumour necrosis factor receptor 1; sTREM-1: soluble triggering receptor expressed on myeloid cells 1; SP-D: surfactant protein D; CD40L: CD40 ligand; PD-L1: programmed death ligand 1; CCL: C-C motif chemokine ligand; CXCL: C-X-C motif chemokine ligand; IL: interleukin; TNF: tumour necrosis factor; GM-CSF: granulocyte–macrophage colony-stimulating factor; IFN: interferon.

FIGURE 5 Mediation analysis for ageing-associated mortality and host response biomarkers upon admission to the general ward. a) Quadrant plot. The x-axis depicts the increase in the 30-day mortality odds ratio per 25% increase of the biomarker derived from an unadjusted logistic regression with the log-transformed biomarker as the explanatory variable and 30-day mortality as the response variable. The y-axis shows the percentage change in the biomarker concentration per 1-year increase in age which was derived from an unadjusted linear regression analysis with the log-transformed biomarker as the response variable. Biomarkers in both the top-right and bottom-left corners are most likely associated with an age-dependent increase in 30-day mortality. The significance of the coefficient was multiple testing corrected using the Benjamini–Hochberg procedure for testing 43 biomarkers. b) Quadrant plot. The same method as in a), but both coefficients are now adjusted for demographics (inclusion hospital, sex and inclusion wave), age-related comorbidities (hypertension, diabetes, malignancies, immunosuppression, and chronic cardiac, neurological, respiratory and kidney disease), age- and biomarker-related chronic medication (antiplatelet and anticoagulant drugs), and COVID-19-related treatments both before and after sampling (corticosteroids including dexamethasone, anti-interleukin (IL)-6, imatinib, remdesivir and antibiotics). c) Diagram of mediation analysis. The adjusted model contained the same covariates as in b). d) Unadjusted and e) adjusted mediation analysis results. Only biomarkers and principal components (PCs) significantly associated with ageing and 30-day mortality were analysed. Confidence intervals were obtained from 1000 times bootstrapping. The higher the proportion of mediation, the stronger the association of the age-dependent differences in that biomarker and the age-dependent increase in 30-day mortality. The PCs and their contributing biomarkers are depicted in [figure 3](#page-7-0). The complete contribution of each biomarker to a PC score is depicted in [supplementary table S6.](http://erj.ersjournals.com/lookup/doi/10.1183/13993003.00011-2023.figures-only#fig-data-supplementary-materials) Endocoag: endothelial and coagulation; ANG: angiopoietin; sTie-2: soluble Tie-2; sE-selectin: soluble E-selectin; sThrombomodulin: soluble thrombomodulin; sVCAM-1: soluble vascular cellular adhesion molecule 1; PAI-1: plasminogen activator inhibitor 1; sCD31: soluble cluster of differentiation 31; sRAGE: soluble receptor for advanced glycation end-products; sTNF-R1: soluble tumour necrosis factor receptor 1; sTREM-1: soluble triggering receptor expressed on myeloid cells 1; SP-D: surfactant protein D; CD40L: CD40 ligand; PD-L1: programmed death ligand 1; CCL: C-C motif chemokine ligand; CXCL: C-X-C motif chemokine ligand; IL: interleukin; TNF: tumour necrosis factor; GM-CSF: granulocyte–macrophage colony-stimulating factor; IFN: interferon.

> additionally show that many cytokines are hardly affected by age. More importantly, we provide novel evidence that ageing is associated with greater activation of the endothelial cell and coagulation system and elevation of inflammation and organ damage markers in COVID-19. Mediation analysis indicated that specific host response alterations explained part of the age effect on mortality. Our study is the most comprehensive analysis of the association between ageing, the host response and mortality in COVID-19.

> Older COVID-19 patients showed strongly increased plasma concentrations of several biomarkers in the endothelial cell and coagulation activation and inflammation and organ damage domains. Likely, alterations in these pathophysiological areas are at least in part intertwined. These (and all other) analyses were done in an unadjusted model and a model adjusted for variables associated with ageing. While these two types of analyses yielded largely similar results, we consider the unadjusted analyses a better reflection of the clinical setting of older patients, who (as an example) inherently have more comorbidities and comedication than young patients. Ageing results in endothelial senescence, characterised by vascular inflammation and endothelial dysfunction; age-associated increases in endothelial cell NF-κB activity have been implicated herein [\[27](#page-15-0), [28\]](#page-15-0). Endotheliopathy is a key feature of COVID-19, and earlier studies reported elevated plasma levels of a variety of biomarkers indicating endothelial cell activation and dysfunction [[29](#page-15-0)]. The current results suggest that COVID-19 is associated with more profound endotheliopathy in older patients. Indeed, patients \geqslant 70 years had higher plasma levels of biomarkers reflective of endothelial cell activation (sThrombomodulin and sVCAM-1), a more disturbed glycocalyx function (syndecan) and a more disrupted barrier function (ANG-2). Loss of glycocalyx is an important feature of endothelial dysfunction in COVID-19 [[29, 30\]](#page-15-0). The glycocalyx consists of proteoglycans and glycosaminoglycans that inhibit immune activation and provide an anticoagulant surface [[30\]](#page-15-0). Hence, a more disturbed glycocalyx function in older COVID-19 patients may contribute to systemic hyperinflammation and coagulation activation. Higher ANG-2 and ANG-2/1 ratios in older COVID-19 patients point in the same direction: ANG-2 promotes endothelial inflammation and vascular leakage, high plasma ANG-2 has been associated with a worse outcomes in COVID-19 [\[31](#page-15-0)], and high ANG-2/1 ratios strongly correlated with mortality in critically ill patients with acute lung injury [\[32](#page-15-0)]. Of note, the association between older age and increased endotheliopathy seems specific for COVID-19, since we recently reported lower endothelial cell activation and dysfunction in old compared with young patients with sepsis [[33\]](#page-15-0).

> Ageing was an independent risk factor for mortality in this study group of noncritically ill COVID-19 patients, confirming previous investigations that analysed populations with more diverse baseline disease severities [\[1, 2](#page-14-0), [5](#page-14-0)]. More importantly, we showed that the age-dependent increase in mortality at day 30 was associated with certain age-associated host response changes. sTNF-R1 has been identified as a cornerstone biomarker for the unfavourable "hyperinflammation" subtype of acute respiratory distress syndrome [[34\]](#page-15-0). sTNF-R1 levels may be a surrogate marker for the activation of cell-associated TNF-R1,

which can trigger vascular leakage and neutrophilic inflammation in the lung [\[35](#page-15-0)–[37\]](#page-15-0). Blockade of TNF-R1 in healthy individuals led to a decrease in neutrophil transmigration and endothelial injury upon inhalation of endotoxin [[35\]](#page-15-0), and TNF-R1-deficient mice were strongly protected against the formation of pulmonary oedema in an acute lung injury model [\[37](#page-15-0)]. Collectively, these data suggest that targeting the TNF-R1 pathway may improve the outcome especially in older COVID-19 patients. TREM-1 is a receptor on innate immune cells that amplifies the immune response triggered by Toll-like and NOD receptors [[38\]](#page-15-0). Increased concentrations of sTREM-1 are reflective of TREM-1 pathway activity [[39\]](#page-15-0), and are associated with increased cytokine levels and disease severity in patients with COVID-19 [[40\]](#page-15-0). In our study, ageing was associated with enhanced sTREM-1 levels, which was a main determinant in the ageing-driven increase in 30-day mortality. Of interest, in a recently completed study in COVID-19 patients requiring ventilator support, infusion of an inhibitor of the TREM-1 pathway improved clinical outcomes, including 28-day mortality (data from a press release by the sponsor [[41\]](#page-15-0)).

Plasma IL-6 was associated with both ageing and COVID-19 mortality. However, in the mediation analysis, the age-dependent increase in 30-day mortality was not associated with increased IL-6 concentrations, suggesting that IL-6 blockade may not benefit older COVID-19 patients. In agreement, the beneficial effect of tocilizumab in the overall population of COVID-19 patients enrolled in a large RCT was not observed in the oldest patients [[11\]](#page-14-0). Similarly, dexamethasone, the other cornerstone immunomodulatory treatment in COVID-19, did not improve outcome in older patients [[9](#page-14-0), [10](#page-14-0)]. Elevated inflammation in older patients may result in steroid resistance [\[42](#page-15-0)], which could contribute to a lower response to a standard dose of dexamethasone. Collectively, these findings suggest that age-associated changes in the host response during COVID-19 influence the effect of immune modulation in this disease and that age should be taken into account for patient selection.

Our study has strengths and limitations. We provide comprehensive insight into aberrations in host response pathways considered important for the pathogenesis of COVID-19 in a large population. Our primary analysis only included noncritically ill patients, yet we provided preliminary evidence that ageing is associated with similar host response changes in COVID-19 patients admitted to the ICU. Our study was purely observational and no definitive conclusion can be drawn on causality. Nonetheless, by using mediation analysis, we provide novel insights into which host response pathways contribute to ageing-associated mortality. While admission plasma biomarkers are used to guide immunomodulating trials [[43\]](#page-16-0), they do not cover the full spectrum of the host response and we may have missed host response alterations that were not (yet) visible on a protein level. Our study was conducted before the broad availability of vaccines and the dominance of the omicron variant. Yet, in populations with a high vaccination coverage, a significant portion of hospitalised COVID-19 patients remain unvaccinated [[44\]](#page-16-0). While the omicron variant still disproportionally affects older patients [[45\]](#page-16-0), robustness of the ageing-associated alterations remains to be validated.

This study documents that ageing is associated with substantial and broad alterations in the host response to COVID-19 across several pathophysiological domains. Increased concentrations of several biomarkers (in particular, sTNF-R1 and sTREM-1) were associated with the ageing-driven increase in 30-day mortality. These results may lay a foundation for new immunomodulatory targets in older patients.

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Data sharing: Data can be shared upon reasonable request after approval of a proposal with a signed data access agreement and always in collaboration with the study group.

Author contributions: E.H.A. Michels, B. Appelman, H. Peters-Sengers and T. van der Poll contributed to the concept and design of the study. E.H.A. Michels, B. Appelman, J. de Brabander, R.B.E. van Amstel, C.C.A. van Linge, O. Chouchane, A.R. Schuurman, T.D.Y. Reijnders, T.A.L. Sulzer, A.M. Klarenbeek, R.A. Douma and the Amsterdam UMC COVID-19 Biobank Study Group were responsible for the collection of the data and samples and the sample preparation. E.H.A. Michels performed the data analysis. H. Peters-Sengers verified the analytical methods. E.H.A. Michels drafted the first manuscript in consultation with B. Appelman, H. Peters-Sengers and T. van der Poll. T. van der Poll supervised the project and provided the funding. All authors provided intellectual input and both revised and approved the final version of the manuscript.

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