Role of Matrix Degradation, Oxidative Stress, Inflammation & Trace Elements in COVID-19 Patients: A Multivariate Study from India

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Abstract
The interrelationship between matrix degradation, oxidative stress, inflammation and trace elements can be speculated in COVID-19. The objective of the study was to evaluate the oxidative stress, inflammation and matrix degradation markers and trace elements in COVID-19 positive patients. A group of confirmed severe COVID-19 positive patients (n = 30) along with COVID-19 negative patients (n = 30) with similar symptoms were included. Both group of patients were assessed for oxidative stress markers, inflammatory cytokines, matrix metalloproteinase (MMP)s and their inhibitors along with trace elements in blood. All the data were subjected to univariate as well as multivariate analysis including PCA, PLS-DA, OPLS-DA. Diagnostic accuracy was tested by ROC curve analysis. Further relationship with Neutrophil/ lymphocyte (N/L) ratio was established if any. Increased oxidative stress, inflammation and matrix degradation is evidenced by significant rise in oxidative markers, inflammatory cytokines and MMP9/TIMP-1 ratio. Decreased Cu/Zn ratio is also observed in COVID-19 positive patients. Multivariate analysis identified SOD, Cu/Zn ratio, IL-6 and TOS, as effective discriminant among the two groups of patients. Further, accuracy was confirmed by ROC curves. Neutrophil/ lymphocyte (N/L) ratio, shows significant negative association with SOD (r = -0.75, p < 0.005) and Cu/Zn ratio (r = -0.88, p < 0.005). These data suggest the attributes of these biomarkers in disease severity. The potential use of these blood-based laboratory markers in disease prognosis seems promising and warrants further attention. Given by the symptoms and severity of the disease, it will be promising to monitor Cu/Zn ratio along with other prognostic indicators.

Keywords SARS-CoV-2 · Matrix Metalloproteinase 9 · Interleukin-6 · Trace Elements · Oxidative stress

Introduction
Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2, previously 2019-nCoV), first time detected in China, is severely contagious. The pathophysiology of COVID-19 is less studied [1, 2]. The virus invades type 2 alveolar epithelial cells, releases several cytokines and inflammatory markers. This ‘cytokine storm’ along with neutrophils, CD4 helper T cells and CD8 cytotoxic T cells cause lung inflammation [3].

Decreased bronchial wall elasticity of the bronchial wall associated with extracellular matrix (ECM) degradation, is reported earlier in COVID-19 [4]. Matrix metalloproteinases (MMP), the major ECM degrading enzymes, help in normal remodeling and repair processes of the lungs. The regulation of MMPs activity is mainly in balance with several tissue inhibitors of matrix metalloproteinases (TIMP). These proteases are over-expressed and cause lung tissue damage in several respiratory diseases. Particularly the role MMP-2, 9 and TIMPs in chronic obstructive lung disease is discussed earlier by our group [5, 6]. Similarly, there is only one study reported so far to cite MMP-9 as an early indicator of respiratory failure in COVID-19 patients [7]. Airway modeling usually involves increased MMPs action induced by pro-inflammatory cytokines. Reactive oxygen species (ROS) is shown to activate MMPs and inactivate TIMPs [5]. Cytokines such as IL-1, IL-6, and TNF-α are shown to upregulate MMPs, which can affect inflammation [8].

Zn cations are shown to inhibit SARS-coronavirus RNA polymerase activity by decreasing its replication. Higher mortality among COVID-19 patients is reported to be...
associated with low Zn level [9]. Similarly, role of copper (Cu) in COVID-19 is published elsewhere [10]. The balance between Zn and Cu is very vital as Zn neutralizes the inflammatory effect of Cu. Magnesium (Mg), another mineral having anti-inflammation, anti-oxidant property, is reported to have therapeutic beneficial effect in other lung diseases including COVID-19 [11, 12]. Neutrophil lymphocyte (N/L) ratio, a potential prognostic biomarker in COVID-19, might aid early detection of severe COVID-19 cases, thereby reducing the overall mortality [13, 14].

Although the interrelationship between matrix degradation, oxidative stress, inflammation and trace elements can be speculated in COVID-19, not much data is available till date. MMP-9/TIMP-1 ratio, a marker for airway severity is reported earlier to be associated with total oxidative stress in chronic obstructive pulmonary disease (COPD) [5], hence it would be tempting to see its role in COVID-19.

Multivariate statistical method effectively identifies the interactions between the potential biomarkers and ensures diagnostic and prognostic performances in terms of specificity, sensitivity and reliability. Principal Component Analysis (PCA), Partial Least-Squares Discriminant Analysis (PLS-DA) and Variable Importance in Projection (VIP) score is used to discriminate inflammatory biomarkers in several diseases [5, 15].

The present study is aimed to observe the changes in oxidative stress, matrix degradation, inflammation, trace elements in COVID-19 confirmed patients. The study also highlights the most useful potential marker(s) to discriminate the COVID-19 positive patients from COVID-19 negative cases using multivariate analysis followed by receiver operator characteristic (ROC) curve analysis. Further association with N/L ratio is also established, if any.

Materials and methods

Patient selection

The study screened all male patients (Jun 2020 to Sep 2020), reported at cough clinic of a tertiary care institute in India. Out of them 103 patients presented with symptoms of fever, cough, fatigue, headache or suspected to be infected with COVID-19, were randomly selected. The laboratory confirmation was done by real-time reverse transcriptase-polymerase chain reaction based positive SARS-CoV-2 RNA with throat swab samples (Genes2Me Pvt Ltd, India). Apart from patient history with fever or any respiratory illness, CT image finding for viral pneumonia, RT-PCR positivity for SARS-CoV-2 RNA, any one of the conditions such as shortness of breath, RR ≥ 30 times/min, (2) Oxygen saturation (Resting state) ≤ 93%, (3) PaO2/FiO2 ≤ 300 mmHg were the characteristics for severe patients. The details are given in flow-chart (Fig. 1). The institutional ethical permission was taken prior to the study and informed consent was taken from the patients.

Sample collection and preparation

Early morning overnight fasting venous EDTA and serum sample was collected. Complete blood cell count blood cell count was performed using an electronic cell counter (XT 2000i; Sysmex, Kobe, Japan). The plasma was used for measuring oxidative markers, cytokines and MMPs. Likewise, the serum was used to measure Cu, Zn and Mg.

Biochemical assay of plasma oxidative stress parameters

Malondialdehyde (MDA) in plasma was measured as per method suggested by Buege et al. (1978) [16]. Total oxidative stress (TOS) levels in plasma were based on the principle of ferrous ion oxidation to ferric ion [17]. Nitrate and nitrite (NOx) in plasma was measured by Griess reaction in which a deep purple azo compound was formed by Griess Reagent converting nitrate to nitrite by nitrate reductase. Similarly, plasma peroxynitrite was measured as described by VanUffelen and colleagues [18]. Uric acid under the enzymatic action of uricase is converted to allantoin and measured at 293 and 700 nm. Total bilirubin in plasma was estimated using Doumas reference method [19] (Siemens Healthcare, India). Glutathione S-transferases (GST) activity in plasma was assayed at 340 nm using substrates such as (1-chloro-2, 4-dinitrobenzene) and reduced glutathione (GSH), as described earlier [20, 21]. The activity of glutathione peroxidase (GSH-Px) was analyzed by Buege et al. (1978) and expressed in μmol/min/ml [16]. Plasma Superoxide dismutase (SOD) was measured by enzyme linked immunosorbent assay (ELISA) (ThermoFisher, USA) (Thermofisher Scientific, USA), as per the instructions by manufacturer. The analytical sensitivity of the assay is 0.044 U/ml. The intra- and inter assay sensitivities CV were < 10%.

Assessment of inflammatory cytokines

Quantification of tumour necrosis factor (TNF)-α, IL-6 and IL-10 in plasma samples was done by ELISA (Thermofisher Scientific, USA), as per the instructions by manufacturer. The assay sensitivities were as follows: IL-1 β, 0.09 pg/ml; IL-6, 0.07 pg/ml; TNF-α, <3pg/ml; IL-10, 0.09 pg/ml.
Analysis of the MMP-9, MMP-2, MMP-12, TIMP-1, TIMP-2 expression levels

ELISA commercial kits (Quantikine ELISA Kit; R&D Systems, USA) were used to measure MMP-2, MMP-12, and TIMP-1&2 levels in plasma. The assay sensitivities for MMP-9, MMP-2, MMP-12, TIMP-1 and TIMP-2 were 0.15 ng/ml, 0.28 ng/ml, <10 pg/ml, <0.08 ng/ml and 0.10 ng/ml respectively.

Assessment of trace elements (Cu, Zn) and mg

Serum copper and zinc was estimated by kit method (Randox, UK). Zinc forms a chelation with 5-Br-PAPS 2-(5-bromo-2-pyridylazo)-5-(N-propyl-N-sulfopropylamino)-phenol and copper reacts with a specific color reagent, 3,5-Di-Br-PAESA 4-(3,5-Dibromo-2-pyridylazo)-N-Ethyl-N-(3-sulfophopropyl) aniline, to form a stable, colored chelate. Mg was estimated by the end point calorimetric method using xylidyl blue.
Table 1 Baseline characteristics of the patients

| Parameters                      | COVID-19 positive patients (n = 30) | COVID-19 negative patients (n = 30) | p value |
|---------------------------------|-------------------------------------|-------------------------------------|---------|
| Age (yrs) ± SD                  | 63 ± 8.3                            | 61.3 ± 9.5                          | NS      |
| Gender                          | Male                                | Male                                | Not applicable |
| Body mass index (BMI) kg/m²     | 21.43 ± 1.62                        | 21.35 ± 1.64                        | NS      |
| Comorbidities                   |                                     |                                     |         |
| None – 11 (36%)                 | None – 16 (53%)                     | -                                   |         |
| Hypertension 8 (26%)            | Hypertension 7 (23%)                |                                     |         |
| Diabetes 9 (30%)                | Diabetes 7 (23%)                    |                                     |         |
| Osteoarthritis 2 (6%)           |                                     |                                     |         |
| Total leucocyte count (TLC) 1 x 10^9/µL, median (IQR) | 10,765 (5780)                     | 14,312 (3700)                       | p < 0.005 |
| Haemoglobin (Hb) (g/dl), median (IQR) | 12.07 (2.5)                        | 9.85 (1.12)                         | p < 0.005 |
| Absolute neutrophil count (ANC), median (IQR) | 7280 (4130)                      | 4320 (3125)                         | p < 0.005 |
| Absolute leucocyte count (ALC), median (IQR) | 1070 (630)                        | 990 (540)                           | p < 0.005 |
| Neutrophil/lymphocyte (N/L) ratio, median (IQR) | 7.1 (6.2)                         | 6.8 (4.3)                           | p < 0.005 |
| Platelet count 1 x 10^9/µL median (IQR) | 159.5 (128)                       | 210 (94)                            | p < 0.005 |
| INR, median (IQR)               | 1.11 (0.27)                         | 0.87 (0.12)                         | p < 0.005 |

Mann-Whitney U test is used to test the significance; NS- not significant

Statistical analysis

Univariate analysis:

The results were presented as a mean ± standard deviation (SD). To analyze the statistical differences between patients and control, Mann-Whitney U test was carried out using GraphPad Prism (v.6.04 GraphPad software). Spearman’s Rank Correlation was used for correlation analysis. Statistical significance was achieved when the p value was less than 0.05.

Multivariate analysis.

In order to reduce possible concentration differences between samples, normalization (by sum) and scaling by mean-centering was done. All the data sets, were subjected to multivariate statistics such as unsupervised classification modeling by PCA along with PLS-DA and orthogonal projection to latent structure with discriminant analysis (OPLS-DA), using metabolyst [22]. Separation of classes, pattern observation and outliers were identified from the PCA score plots. Further class separation was maximized using supervised classification models including PLS-DA and OPLS-DA. The goodness of fit and predictive ability in the PLS-DA and OPLS-DA models is quantitatively measured by the parameter R2, and Q2 respectively. Both the values usually vary within 0 and 1; with a perfect fit (R2) and (Q2) corresponding to 1 indicates 100% goodness of fit and predictive ability. The VIP score of a variable is calculated as a weighted sum of the squared correlations between the PLS-DA components and the original variable. A cut-off of VIP score > 1.5 was considered as the threshold for selection of the most significant variables that best correlated with the discriminative scores of PLS-DA. The cross-validation of PLS-DA model was done by permutation of the best two components and compared with 100 different permuted models. The accuracy of diagnostic model was also used to evaluate ROC curve. The sensitivity and specificity and AUC of the selected parameters were determined using this curve.

Results

The baseline characteristics of the patients and healthy volunteers are summarized in Table 1. No significant change was noticed for age and BMI of the patients. Whereas, absolute neutrophil count (ANC), absolute leucocyte count (ALC), N/L ratio, platelet count, international normalized ratio (INR) was found to be significantly changed (p < 0.005) in COVID-19 patients.

Oxidative stress markers

Figure 2 illustrates the plasma oxidative stress parameters for both groups of patients. A significant increase in oxidative parameters such as TOS (p < 0.0001), NOx (p < 0.05), MDA (p < 0.0001) and peroxynitrate (p < 0.005) was seen in COVID-19 positive patients. A significant decrease in antioxidant defence markers including GSH-PX (p < 0.05), GSH (p < 0.05), GST (p < 0.05), is also observed. No significant change for plasma uric acid and bilirubin is observed in these patients.

Plasma cytokines

The plasma cytokine levels in both groups of patients are illustrated in Fig. 2. A significant increase in IL-6, TNF-α levels was observed in COVID-19 positive patients (TNF-α; p < 0.05, IL-6; p < 0.0001). No statistical change was observed for plasma IL-10 level.
Q2 (predictive ability) were used to assess the robustness of the PLSDA model. The values of $R^2$ (0.91) and $Q^2$ (0.89) indicate that the model has good fit and can satisfactorily differentiate the COVID-19 positive from the negative patients. The effective key parameters found to discriminate the two groups of patients, were identified using their VIP values. Variables with VIP scores $>1.5$ were considered to be significant for class separation (Fig. 4). With a VIP score cut-off value of 1.5, a total of four parameters such SOD, Cu/Zn ratio, IL-6 and TOS were found to be dysregulated out of which SOD, Cu/Zn ratio were significantly down-regulated along with upregulated IL-6 and TOS levels in COVID-19 patients (Fig. 4).

Further, the generated ROC curves for these altered markers indicate these parameters have the highest accuracy in predicting COVID-19 affected patients (Fig. 5). The diagnostic potential of SOD was found to have 90% sensitivity (true positive) and 86.7% specificity (true negative) with

**Plasma MMP-9, 2, 12, TIMP-1 and 2**

The levels of MMP-9/TIMP-1, TIMP-2, MMP-2 and MMP-12 in both the groups are shown in Fig. 2. The MMP-9/TIMP-1 ratio, MMP-2 and MMP-12 levels are increased significantly in COVID-19 positive patients as compared to COVID-19 negative patients. No change is observed for TIMP-2 levels.

**Multivariate statistical analysis**

PCA obtained from both group of patients indicate unbiased clustering (Fig. 3a). A clear separation is seen between the two groups of patients by PCA scores scatter plot. Even the PLS-DA (Fig. 3b) signify enhanced class separation.

The validation of predictive capability of PLS-DA was achieved using permutation tests (Fig. 3c). The permutation tests indicate higher $R^2$ and $Q^2$ values of true model hence far better than permuted models. $R^2$ (goodness of fit) and $Q^2$ (predictive ability) were used to assess the robustness of the PLSDA model. The values of $R^2$ (0.91) and $Q^2$ (0.89) for PLS-DA and OPLS-DA ($R^2 = 0.89; Q^2 = 0.87$) indicate that the model has good fit and can satisfactorily differentiate the COVID-19 positive from the negative patients. The effective key parameters found to discriminate the two groups of patients, were identified using their VIP values. Variables with VIP scores $>1.5$ were considered to be significant for class separation (Fig. 4). With a VIP score cut-off value of 1.5, a total of four parameters such SOD, Cu/Zn ratio, IL-6 and TOS were found to be dysregulated out of which SOD, Cu/Zn ratio were significantly down-regulated along with upregulated IL-6 and TOS levels in COVID-19 patients (Fig. 4).

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Discussion

The inter-association between oxidative stress and viral infections associated with respiratory diseases is reported earlier. SARS-CoV2 like other RNA viruses seems to activate oxidative stress [23]. Several inflammatory cells such as monocytes and macrophages release a set of pro-inflammatory cytokines such as IL-1β, IL-6, TNF, IL-8, thereby, play a vital role in COVID-19 infection [24, 25]. These inflammatory cytokines might play an important role in airway remodeling by inducing the MMPs action and causing MMP over-expression and activity [5, 26]. Zinc, an

criterion <10.9. Cu/Zn ratio and TOS were having 93.3% and 96.7% of sensitivity with criterion <1.96 and >11.12 respectively along with 86.7% specificity each. Similarly, IL-6 was having 83.3% specificity and 76.7% sensitivity with criterion value of >1.87 (Fig. 5).

Further correlation with N/L ratio, indicates significant association with SOD ($r = -0.75, p < 0.005$) and Cu/Zn ($r = -0.88, p < 0.005$) (Fig. 6) unlike TOS and IL-6.

Fig. 3 Multivariate analysis; Score scatter plot (a) principal component analysis (PCA) shows discrimination among (a) COVID-19 positive and negative patients (b) Partial least square (PLS) – discriminant analysis (DA) model shows improved discrimination among both group of patients with $R^2 = 0.91$ and $Q^2 = 0.89$ (c) Permutation test statistics for the PLS-DA models indicate that the true model has much higher $R^2$ and $Q^2$ values, and thus is far better than 100 permuted models.
the importance of IL-6 and other inflammatory cytokines in COVID-19 [36].

Although the role of proteases such as MMP-2, 9, 12 and TIMP-1 are evidenced in several lung diseases [5, 37, 38], very few data are on COVID patients. Our study shows increase in MMP-9/TIMP-1 ratio, MMP-2 and 12 (Fig. 2). Recently, MMP-9 is reported as an early indicator of respiratory failure in COVID-19 patients [8]. MMP-19 is also hypothesized as possible therapeutic target of chloroquine and melatonin in COVID-19 [39]. The suitability of exploring MMPs as therapeutic targets in COVID-19 is reported recently [40]. The positive association between the MMP activity and oxidative stress in COPD is documented by our group recently [5].

Our study finds decreased Cu/Zn ratio in COVID-19 patients (Fig. 2) which is in accordance with the report published by Muhammad et al. (2021) [30]. As these elements are known to be cofactors of SOD, an antioxidant enzyme, hence decreased ratio triggers oxidative stress [41]. Several other studies also recently show decrease in Zn levels in COVID-19 patients [34, 44]. Although mechanism is not clear, surge in inflammatory cytokines seem to cause Cu deficiency [41]. The negative association between Cu/Zn ratio and N/L ratio (Fig. 6).b narrates the role of trace elements in severity of the disease.

Using univariate analysis, significant role of several indicators related to oxidative stress, inflammation, matrix degradation process, trace elements is observed among COVID-19 patients. Similarly, multivariate approach including VIP score, is able to identify best cut-off dysregulated parameters such as SOD, Cu/Zn ratio, IL-6 and TOS. ROC curves interpret the significant accuracy of the model in terms of sensitivity, specificity. The 90%, 93%, 96.7% sensitivity and 86.7% specificity in respect of parameters such as SOD, Cu/Zn ratio and TOS might enable them to be used as additional laboratory markers for COVID-19. A panel of biomarkers is reported to be preferred over single marker as far as diagnostic utility is concerned [42], which holds good in our finding. Further, significant negative association of SOD and Cu/Zn ratio, with N/L ratio, establishes their role in severity of disease.

**Conclusions**

Although, there is no in-depth explanation to identify the exact mechanism or pathway involving all these parameters simultaneously during the course of disease, but our finding clearly observes the association among oxidative stress, inflammation, matrix degradation and trace element changes in COVID-19 affected patients. Hence it will be significant to include such parameters in routine lab investigation.
Recent data reports partial improvements in clinical parameters and attenuation of cytokine storm among COVID-19 patients supplemented antioxidants orally. It can be speculated that our finding which includes oxidative and inflammatory markers can be helpful for clinicians during follow up of the COVID-19 patients. Given by the symptoms and severity of the disease, it will be promising to monitor blood Cu/Zn ratio along with other prognostic indicators. Four identified parameters such as SOD, Cu/Zn ratio, IL-6 and TOS were found to discriminate the COVID-19 positive patients from the negative ones. Although these were also suitable for diagnostic accuracy, a larger group of patient’s data can validate the finding.

The study is not without limitations. First, the study includes only severe COVID-19 patients. It would be better to investigate in mild cases and healthy controls too. Second, to understand the role of biomarkers we need to have a follow-up post treatment group, which was not possible in the study. Third, N/L ratio, a proposed ideal prognostic marker as per literature survey, was selected to compare with others. Any other relevant clinical parameter assessing the disease progression would be better indicated to assess the interrelationship. Fourth, although the results appeared

![ROC analysis](image-url)
FIG. 6 Relationship between (a) plasma superoxide dismutase (SOD) and neutrophil/lymphocyte (N/L) ratio (b) plasma Cu/Zn ratio and neutrophil/lymphocyte (N/L) ratio. Spearman’s rank correlation coefficient is denoted by $r$. A significant negative correlation by Spearman’s rank correlation analysis was observed before and after therapy ($p < 0.005$).

to be quite encouraging, the sample size was too less ($n = 30$) to draw any conclusion regarding the mechanism of the disease. It would be worthwhile to study in a larger group of patients involving a multicentric approach. Similarly, it will be ideal to replicate our findings in an independent dataset. Last, it was not possible to rule out any outliers due to the existing co-morbidities such as hypertension, diabetes mellitus and osteoarthritis among these patients.

Abbreviations
SARS-CoV-2  severe acute respiratory syndrome coronavirus 2.
ECM  extracellular matrix, MMP:Matrix metalloproteinases.
TIMP  tissue inhibitors of matrix metalloproteinases.
ROS  Reactive oxygen species.
COPD  chronic obstructive pulmonary disease.
PCA  Principal Component Analysis.
PLS-DA  Partial Least-Squares Discriminant Analysis.
VIP  Variable Importance in Projection.
MDA  malondialdehyde.
TOS  total oxidative stress.
GST  Glutathione S-transferases.
GSH-Px  Glutathione peroxidase.
SOD  Superoxide dismutase.
TNF  tumour necrosis factor.
OPLS-DA orthogonal projection to latent structure with discriminant analysis.
ANC  absolute neutrophil count.
ALC  absolute leukocyte count.
INR  international normalized ratio.
ROC  Receiver Operator Characteristic.

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Declarations

Conflict of Interest The authors declare that they have no competing interests.

Compliance with ethical standards The study protocol was approved by the Institutional Ethics Committee. Informed consent was obtained from all subjects prior to enrolment.

References
1. Yüce M, Filiztekin E, Özkaya KG. COVID-19 diagnosis —A review of current methods. Biosens Bioelectron. 2021;172:112752.
2. Parasher A. COVID-19: Current understanding of its pathophysiology, clinical presentation and treatment. Postgrad Med J. 2020;97:postgradmedj–2020.
3. Rahman S, Montero MTV, Rowe K, Kunik F. Epidemiology, pathogenesis, clinical presentations, diagnosis and treatment of COVID-19: A review of current evidence. Expert Rev Clin Pharmacol. 2021;1:21.
4. Leng L, Cao R, Ma J, Mou D, Zhu Y, Li W, et al. Pathological features of COVID-19-associated lung injury: a preliminary proteomics report based on clinical samples. Signal Transduction and Targeted Therapy [Internet]. 2020 [cited 2021 Jan 14];5. Available from: https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7557250/.
5. Singh B, Ghosh N, Saha D, Sarkar S, Bhattacharyya P, Chaudhury K. Effect of doxycycline in chronic obstructive pulmonary disease - An exploratory study. Pulm Pharmacol Ther. 2019;58:101831.
6. Singh B, Jana SK, Ghosh N, Das SK, Joshi M, Bhattacharyya P, et al. Metabolomic profiling of doxycycline treatment in chronic obstructive pulmonary disease. J Pharm Biomed Anal. 2017;132:103–8.
7. Ueland T, Holter J, Holten A, Müller K, Lind A, Bekken G, et al. Distinct and early increase in circulating MMP-9 in COVID-19 patients with respiratory failure. The Journal of Infection [Internet]. 2020 [cited 2021 Jan 14];81:e41–3. Available from: https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7320854/.

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8. Siwik DA, Colucci WS. Regulation of Matrix Metalloproteinases by Cytokines and Reactive Oxygen/Nitrogen Species in the Myocardium. Heart Failure Reviews [Internet]. 2004 [cited 2019 Jun 12];9:43–51. Available from: https://link.springer.com/article/https://doi.org/10.1023/B%3AHREV0000011393.40674.13.

9. Anuk AT, Polat N, Akdas S, Erol SA, Tanacan A, Biriken D, et al. The Relation Between Trace Element Status (Zinc, Copper, Magnesium) and Clinical Outcomes in COVID-19 Infection During Pregnancy. Biological Trace Element Research [Internet]. 2020 [cited 2021 Jan 14];1:1–10. Available from: https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7685187/.

10. Hackler J, Heller RA, Sun Q, Schwarzer M, Diegmann J, Bachmann M, et al. Relation of Serum Copper Status to Survival in COVID-19. Nutrients [Internet]. 2021 [cited 2022 Jan 5];13:1898. Available from: https://pubmed.ncbi.nlm.nih.gov/34072977/.

11. Tang C-F, Ding H, Jiao R-Q, Wu X-X, Kong L-D. Possibility of magnesium supplementation for supportive treatment in patients with COVID-19. European Journal of Pharmacology [Internet]. 2020 [cited 2021 Jan 15];886:173546. Available from: https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7486870/.

12. Powell C, Dwan K, Milan SJ, Beasley R, Hughes R, Knopp-Sihota JA, et al. Inhaled magnesium sulfate in the treatment of acute asthma. Cochrane Database of Systematic Reviews. 2012;12:CD003898.

13. Yang A-P, Liu J, Tao W, Li H. The diagnostic and predictive role of NLR, d-NLR and PLR in COVID-19 patients. Int Immunopharmacol. 2020;84:106504.

14. Li X, Liu C, Mao Z, Xiao M, Wang L, Qi S, et al. Predictive values of neutrophil-to-lymphocyte ratio on disease severity and mortality in COVID-19 patients: a systematic review and meta-analysis. Critical Care. 2020;24.

15. Patro ARK, Mohanty S, Prusty BK, Singh DK, Gaikwad S, Sashwat T, et al. Cytokine Signature Associated with Disease Severity in Dengue. Viruses. 2019;11:34. [16] JA, Buege SD, Aust. [30] Microsomal lipid peroxidation, Methods in Enzymology. (1978) 302–310. https://doi.org/10.1016/s0076-6879(78)52032-6.

16. Buege JA, Aust SD. [30] Microsomal lipid peroxidation. Methods in Enzymology [Internet]. 1978;302–10.

17. Jansen E, Ruskovska T. Comparative Analysis of Serum (Anti)oxidative Status Parameters in Healthy Persons. International Journal of Molecular Sciences [Internet]. 2013 [cited 2022 Jul 4];14:6106–15.

18. VanLIFELEN EB, Van der ZEE J, de KASTER MB, VanSTEFVENICK J, ELFERINK GRJ. Intracellular but not extracellular catalysis of nitroxyl anion into nitric oxide leads to stimulation of human neutrophil migration. Biochemical Journal. 1998;330:719–22.

19. Meidaninikjeh S, Sabouni N, Marzouni HZ, Bengar S, Khalili A, Jafari R. Monocyte and macrophages in COVID-19: Friends and foes. Life Sciences. 2021;269:119910.

20. Merad M, Martin JC. Pathological inflammation in patients with COVID-19: a key role for monocytes and macrophages. Nature Reviews Immunology [Internet]. 2020;20:1–8. Available from: https://www.nature.com/articles/s41577-020-0331-4

21. Merad M, Martin JC. Pathological inflammation in patients with COVID-19: a key role for monocytes and macrophages. Nature Reviews Immunology [Internet]. 2020;20:1–8. Available from: https://www.nature.com/articles/s41577-020-0331-4

22. Solun B, Shoenfeld Y. Inhibition of metalloproteinases in therapy for severe lung injury due to COVID-19. Medicine in Drug Discovery [Internet]. 2020 [cited 2021 Apr 17];7:100052. Available from: https://www.sciencedirect.com/science/article/pii/S2590086220300397.

23. Chong J, Soufan O, Li C, Caraus I, Li S, Bourque G, et al. Metalloproteinase-9 as a host protein target of chloroquine and melatonin for immunoregulation in COVID-19: A network-based
meta-analysis. Life Sciences [Internet]. 2020;257:118096. Available from: https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7361122/

40. Hardy E, Fernandez-Patron C. Targeting MMP-Regulation of Inflammation to Increase Metabolic Tolerance to COVID-19 Pathologies: A Hypothesis. Biomolecules. 2021;11:390.

41. Ebrahimi M, Norouzi P, Aazami H, Moosavi-Movahedi AA. Review on oxidative stress relation on COVID-19: Biomolecular and bioanalytical approach. International Journal of Biological Macromolecules [Internet]. 2021 [cited 2021 Oct 28];189:802–18. Available from: https://www.ncbi.nlm.nih.gov/pmc/articles/PMC8372478.

42. Dutta M, Singh B, Joshi M, Das D, Subramani E, Maan M, et al. Metabolomics reveals perturbations in endometrium and serum of minimal and mild endometriosis. Scientific Reports. 2018;8.

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