

University of Groningen

Antioxidative Stress Metabolic Pathways in Moderately Active Individuals

Almuraikhy, Shamma; Anwardeen, Najeha; Doudin, Asmma; Sellami, Maha; Domling, Alexander; Agouni, Abdelali; Althani, Asmaa A.; Elrayess, Mohamed A.

Published in:
Metabolites

DOI:
[10.3390/metabo13090973](https://doi.org/10.3390/metabo13090973)

IMPORTANT NOTE: You are advised to consult the publisher's version (publisher's PDF) if you wish to cite from it. Please check the document version below.

Document Version
Publisher's PDF, also known as Version of record

Publication date:
2023

[Link to publication in University of Groningen/UMCG research database](#)

Citation for published version (APA):

Almuraikhy, S., Anwardeen, N., Doudin, A., Sellami, M., Domling, A., Agouni, A., Althani, A. A., & Elrayess, M. A. (2023). Antioxidative Stress Metabolic Pathways in Moderately Active Individuals. *Metabolites*, 13(9), Article 973. <https://doi.org/10.3390/metabo13090973>

Copyright

Other than for strictly personal use, it is not permitted to download or to forward/distribute the text or part of it without the consent of the author(s) and/or copyright holder(s), unless the work is under an open content license (like Creative Commons).

The publication may also be distributed here under the terms of Article 25fa of the Dutch Copyright Act, indicated by the "Taverne" license. More information can be found on the University of Groningen website: <https://www.rug.nl/library/open-access/self-archiving-pure/taverne-amendment>.






Take-down policy

If you believe that this document breaches copyright please contact us providing details, and we will remove access to the work immediately and investigate your claim.

Downloaded from the University of Groningen/UMCG research database (Pure): <http://www.rug.nl/research/portal>. For technical reasons the number of authors shown on this cover page is limited to 10 maximum.

Article

Antioxidative Stress Metabolic Pathways in Moderately Active Individuals

Shamma Almuraikhy ^{1,2}, Najeha Anwardeen ¹ , Asmma Doudin ¹, Maha Sellami ³ , Alexander Domling ² , Abdelali Agouni ⁴ , Asmaa A. Althani ^{1,5} and Mohamed A. Elrayess ^{1,4,*} 

¹ Biomedical Research Center, Qatar University, Doha P.O. Box 2713, Qatar

² Groningen Research Institute of Pharmacy, Drug Design, Groningen University, 9713 AV Groningen, The Netherlands

³ Physical Education Department (PE), College of Education, Qatar University, Doha P.O. Box 2713, Qatar

⁴ College of Pharmacy, QU Health, Qatar University, Doha P.O. Box 2713, Qatar

⁵ Department of Biomedical Sciences, College of Health Science, QU Health, Qatar University, Doha P.O. Box 2713, Qatar

* Correspondence: m.elrayess@qu.edu.qa; Tel.: +974-4403-6007 or +974-3350-6764

Abstract: Physical activity (PA) is known to have beneficial effects on health, primarily through its antioxidative stress properties. However, the specific metabolic pathways that underlie these effects are not fully understood. This study aimed to investigate the metabolic pathways that are involved in the protective effects of moderate PA in non-obese and healthy individuals. Data on 305 young, non-obese participants were obtained from the Qatar Biobank. The participants were classified as active or sedentary based on their self-reported PA levels. Plasma metabolomics data were collected and analyzed to identify differences in metabolic pathways between the two groups. The results showed that active participants had increased activation of antioxidative, stress-related pathways, including lysoplasmalogen, plasmalogen, phosphatidylcholine, vitamin A, and glutathione. Additionally, there were significant associations between glutathione metabolites and certain clinical traits, including bilirubin, uric acid, hemoglobin, and iron. This study provides new insights into the metabolic pathways that are involved in the protective effects of moderate PA in non-obese and healthy individuals. The findings may have implications for the development of new therapeutic strategies that target these pathways.

Keywords: physical activity; oxidative stress; metabolomics; glutathione; plasmalogen; phosphatidylcholine; vitamin A; xanthine



Citation: Almuraikhy, S.; Anwardeen, N.; Doudin, A.; Sellami, M.; Domling, A.; Agouni, A.; Althani, A.A.; Elrayess, M.A. Antioxidative Stress Metabolic Pathways in Moderately Active Individuals. *Metabolites* **2023**, *13*, 973. <https://doi.org/10.3390/metabo13090973>

Academic Editor: Elena Planells

Received: 31 July 2023

Revised: 23 August 2023

Accepted: 23 August 2023

Published: 27 August 2023



Copyright: © 2023 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (<https://creativecommons.org/licenses/by/4.0/>).

1. Introduction

Physical activity (PA) has been shown to lower body fat percentage and improve blood pressure and lipids profile, thus reducing the risk of cardiovascular diseases [1]. Recent joint guidelines of the American College of Sports Medicine and the American Heart Association have recommended 150 min of moderate activity (30 min, 5 days/week) or 60 min of vigorous PA (20 min on 3 days) for all adults. Particularly, walking regularly can improve sugar metabolism and cardiorespiratory fitness, leading to better outcomes and quality of life [2]. Regular exercise in non-obese individuals was shown to improve glucose tolerance and insulin sensitivity, as well as various lipid parameters, blood pressure, and fibrinolytic activity [3–5]. Even exercise programs that produced little or no change in maximal oxygen uptake (VO₂ max), such as long-term walking, reported improvement in insulin sensitivity [6,7].

Physical activity can induce long-term adaptations in cellular metabolism and immune cells, including decreased oxidative stress status with a concomitant rise in antioxidant factors [8]. Analyzing the functional interactions between physical exercise and the immune system is extremely complex due to the multiplicity of the underlying mediators. These

include the type, frequency, and intensity of the exercise, as well as individual variability of the immune system in response to exercise-induced stress [9].

Oxidative stress refers to the disparity between the generation of free radicals and an organism's capacity to counteract them using antioxidants [10]. Free radicals are molecules with high reactivity, possessing one or more unpaired electrons that render them unstable. These radicals are produced during routine metabolic processes, both at rest and during exercise.

The antioxidant response following physical exercise in human plasma can be evaluated using protein oxidation, lipid oxidation, and exogenous antioxidant utilization, including total antioxidant capacity (TAC). Other examples include antioxidant enzymes such as superoxide dismutase (SOD) and glutathione peroxidase (GPX), which are elevated during physical training [11]. Additionally, increased water and lipid soluble endogenous antioxidant levels, such as ascorbic acid and alpha tocopherol, can be considered [12]. Similarly, proteins that are associated with oxidative stress tend to increase with a steady-state exercise [13].

The metabolomics of human serum samples can provide a snapshot of the current metabolic state of an individual. The levels of circulating metabolites are influenced by both genetic and environmental factors, and can further depend on age, sex, menopause, and diet of the study participants. Previous investigations primarily focused on metabolomics using urine and blood, revealing significant concentration changes in 196 metabolites following aerobic or resistance exercise [13]. However, as highlighted by Schraner et al. [13], "metabolite concentrations change in different directions after exercise." Considering that earlier studies examined metabolites in diverse athlete categories and during the acute exercise phase, it became imperative to scrutinize their fluctuations in active, healthy individuals and throughout extended adaptation periods. This approach aims to offer a richer dataset to practitioners and clinicians, enhancing the depth of information available.

A better understanding of these relationships is pivotal for the planning of metabolomics studies involving human subjects and the interpretation of their results [14]. Metabolomics offers a quantitative measurement of the metabolic profiles associated with exercise in moderately active non-obese participants to identify biomarkers associated with moderate activity. Non-targeted metabolomics allows the detection of changes in response to various physiological states such as pre-/post-exercise, and enables the identification of metabolic signatures with potential translational impact for the public [15].

The emerging data highlight the involvement of antioxidative stress pathways with overall protective phenotypes. Therefore, the objective of this study is to identify the metabolic pathways underlying the protective effect of moderate PA in non-obese, apparently healthy individuals using an untargeted metabolomics approach.

2. Materials and Methods

2.1. Data Source and Study Participants

Data were collected from 305 participants from the Qatar Biobank, including questionnaires related to physical activity and laboratory results for 66 clinically-relevant metabolic traits, such as measurements of systolic and diastolic blood pressure, waist to hip ratio (WHR), body mass index (BMI), clinical chemistry, and endocrinology tests (Table 1). In addition, metabolomics data were collected on over 1000 metabolites. The study was approved by the Institutional Review Boards of the Qatar Biobank (QF-QBB-RES-ACC-00066) and Qatar University (QU-IRB 1716-E/22). All participants provided informed consent. Physically active participants were identified as those who practice any type of moderate-intensity physical activity (i.e., walking, jogging, swimming, or riding) for at least 150 min per week (moderate activity) and not more than 300 min/week according to the ACSM, WHO, and CDC guidelines and the Physical Activity Guidelines for Americans. Inclusion criteria included individuals who were young (20–30 years old), lean/overweight (BMI: 20–30 kg/m²), and healthy (no chronic diseases such as diabetes, glaucoma, macular

degeneration, blood clots, cardiovascular disease, bariatric surgery, or cancer). Accordingly, among all participants, 42% were physically active and 58% were sedentary.

Table 1. General characteristics of study participants grouped into active or sedentary.

Clinical Traits	Sedentary (177)	Moderately Active (128)	p Value
HOMA-IR (categorical)			
<1.85	87	78	0.048
>1.85	90	50	
Gender			
Male	89	70	0.486
Female	88	58	
Vital signs			
Age	26 (23–28)	26 (24–28)	0.928
BMI	24.31 (22.64–26.68)	24.67 (23.06–27.41)	0.163
Average systolic BP	105 (98–112)	107 (101–115)	0.090
Average diastolic BP	68.6 (7.75)	67.6 (7.45)	0.250
Average pulse rate	69 (63–76)	66.5 (60–73)	0.051
Blood sugar			
Fasting time (minutes)	152 (87–365)	196 (84.5–630.5)	0.138
Glucose (mmol/L)	4.8 (4.6–5.2)	4.9 (4.56–5.2)	0.525
Hb A1c	5.2 (5–5.4)	5.2 (5–5.4)	0.700
Insulin (Uu/mL)	8.2 (6–14.5)	7.5 (5.7–11.825)	0.067
HOMA-IR (continuous)	1.88 (1.325–3.165)	1.6 (1.1425–2.6025)	0.064
C-Peptide ng mL	1.84 (1.415–2.97)	1.74 (1.35–2.51)	0.087
Physical tests			
Sitting height (cm)	91.4 (86.65–134.375)	90.7 (85.5–134.275)	0.407
Weight (kg)	67 (59.2–76.5)	69.95 (61.45–77.325)	0.219
Waist size (cm)	78 (72–87)	79 (73.75–86.25)	0.404
Hip size (cm)	100.8 (6.29)	101 (6.55)	0.804
Waist to hip ratio	0.78 (0.08)	0.79 (0.07)	0.570
Handgrip (left)	28 (21–38)	33 (22–44)	0.015
Handgrip (right)	30 (22–40)	34 (26–48)	0.007
Maximum heart rate (beats/min)	128 (114–141)	120 (109.25–133)	0.002
Run time (seconds)	764 (729–764)	764 (742–764)	0.953
Blood inflammatory markers			
Hemoglobin (g/dL)	13.8 (12.4–15)	13.7 (12.5–15)	0.644
Estimated hemoglobin	17.9 (3.86)	18.6 (3.27)	0.261
Hematocrit %	41.05 (37.9–44.5)	42.1 (38.05–45.05)	0.245
Red blood cell $\times 10^6$ μ L	4.9 (4.5–5.3)	5 (4.5–5.3)	0.438
White blood cell ($\times 10^3$ μ L)	6.55 (5.3–7.725)	6.5 (5.5–7.6)	0.711
Monocyte auto %	7.2 (6–8.725)	7.1 (6.2–8.3)	0.571
Monocytes Auto ($\times 10^3$ μ L)	0.5 (0.4–0.6)	0.5 (0.4–0.6)	0.779
Neutrophil Auto ($\times 10^3$ μ L)	3.5 (2.7–4.5)	3.4 (2.75–4.4)	0.646
Neutrophil auto %	54.4 (9.31)	53.9 (9.91)	0.654
Lymphocyte Auto ($\times 10^3$ μ L)	2.2 (1.8–2.6)	2.2 (1.9–2.7)	0.550
Lymphocyte auto %	34.9 (7.69)	35.5 (8.73)	0.562
Eosinophil auto %	2.2 (1.5–3.425)	2.3 (1.35–3.75)	0.685
Eosinophils Auto ($\times 10^3$ μ L)	0.1 (0.1–0.2)	0.1 (0.1–0.2)	0.763
Basophil auto %	0.6 (0.4–0.8)	0.6 (0.4–0.7)	0.019
Basophils Auto ($\times 10^3$ μ L)	0 (0–0.1)	0 (0–0.01)	0.105
Platelet ($\times 10^3$ μ L)	238 (201.75–283.25)	235 (207–269)	0.599
Mean cell hemoglobin (pg)	28.3 (26.175–29.7)	28.2 (26.15–29.5)	0.947
Mean cell hemoglobin concentration (g/dL)	33.4 (32.6–34.025)	33.1 (32.6–33.7)	0.046
C reactive Protein (mg/L)	5 (5–5)	5 (5–5)	0.852

Table 1. Cont.

Clinical Traits	Sedentary (177)	Moderately Active (128)	p Value
Kidney profile			
Sodium (mmol/L)	140 (139–142)	141 (139–142)	0.142
Potassium (mmol/L)	4.3 (4.1–4.4)	4.3 (4.1–4.5)	0.341
Chloride (mmol/L)	101 (100–102)	101 (100–102)	0.973
Bicarbonate (mmol/L)	26 (25–27)	26 (25–28)	0.212
Urea (mmol/L)	4.2 (3.4–4.8)	4.25 (3.4–5.2)	0.221
Creatinine (mmol/L)	65 (55–77)	68 (54–79)	0.501
Calcium (mmol/L)	2.39 (0.08)	2.4 (0.07)	0.307
Calcium Corrected (mmol/L)	2.27 (0.07)	2.27 (0.06)	0.557
Phosphorus (mmol/L)	1.17 (0.17)	1.19 (0.15)	0.233
Uric Acid (μmol/L)	277 (229–335)	291.5 (238.5–344.25)	0.291
Creatine kinase	78.5 (58–116.5)	91 (66–152)	0.023
Creatine Kinase-1 (ng/mL)	1.16 (0.675–1.745)	1.13 (0.9225–1.495)	0.941
Creatine kinase-2 (u/L)	78 (57.5–122)	82 (59–142.5)	0.631
Magnesium (umol/L)	0.835 (0.05)	0.834 (0.04)	0.813
Homocysteine (umol/L)	8.1 (6.35–9.9)	8.55 (6.6–10.125)	0.383
Total Protein (g/L)	74 (71–76)	74 (72–77)	0.034
Liver profile			
Total bilirubin (umol/L)	7 (5–9)	7 (5–9)	0.963
Albumin (g/L)	46 (45–48)	47 (45–48)	0.531
Alkaline phosphatase (u/L)	65 (56–75)	63 (52–75.25)	0.241
Alanine transaminase (u/L)	17 (13–28)	18 (12–25)	0.955
Aspartate aminotransferase (u/L)	17 (15–21)	18.5 (16–22)	0.030
GGT (u/L)	14 (9–21)	13 (10.25–18.25)	0.547
GGT 2 (u/L)	20 (13–33)	15 (11–22)	0.002
Lipid profile			
HDL Cholesterol (mmol/L)	1.36 (0.33)	1.45 (0.39)	0.031
LDL cholesterol calc (mmol/L)	2.8 (2.2175–3.185)	2.83 (2.0575–3.1025)	0.560
Triglyceride (mmol/L)	0.97 (0.7–1.4)	0.9 (0.6075–1.2)	0.059
Iron profile			
Iron (μmol/L)	15.89 (11–19.97)	15 (10.85–19)	0.513
Total iron binding capacity (μmol/L)	60 (53–65)	58 (54–64.5)	0.348
Unsaturated iron binding capacity (μmol/L)	42 (35–51)	43.15 (36.125–50)	0.995
Ferritin (μg/L)	45 (12–106)	49 (17.75–109.25)	0.395
Hormones			
Free Thyroxine (pmol/L)	13.43 (12.54–14.6)	13.36 (12.48–14.45)	0.589
Free triiodothyronine (nmol/L)	4.54 (4.26–5)	4.5 (4.01–4.88)	0.312
Thyroid stimulating hormone (mU/L)	1.46 (0.95–2.26)	1.305 (0.9975–1.92)	0.223
Total Testosterone (nmol/L)	6.5 (1.36–19.51)	9.115 (1.26–21.69)	0.294
Estradiol (pmol/L)	119.5 (82–265.25)	123.5 (87–236.25)	0.579
Sex hormone binding globulin (nmol/L)	40 (26.4–64.8)	43.05 (28.15–62.62)	0.718
Vitamins			
Folate (nmol/L)	22.7 (8.04)	22.2 (7.66)	0.565
Vitamin B12 (pmol/L)	293.5 (226–375.5)	274.5 (221.25–404.75)	0.717
Dihydroxyvitamin D Total (ng/mL)	14 (11–20)	14 (11–22)	0.788

BMI body mass index, BP blood pressure, LDL low density lipoprotein, HDL high density lipoprotein, HOMA-IR homeostatic model assessment of insulin resistance, Estimated hemoglobin: hemoglobin estimated from hematocrit, Maximum heart rate (beats/min): Estimated calculated maximum heart rate, Waist size (cm): Average waist measurements, Fasting time: Fasting time in minutes. Hip and waist size: Circumference. Data are presented as mean (SD) for normal variables, median (IQR) for skewed variables, and number (percentage) for categorical parameters. Differences between the groups were tested using Student's *t*-test/Mann-Whitney U test for parametric/non-parametric variables and Fisher's exact test for nominal variables. A *p*-value significance level of 0.05 was used.

2.2. Metabolomics

Established protocols were used for untargeted metabolomics of serum samples from all participants using the Metabolon platform [15]. To measure the metabolites, we used a Waters ACQUITY ultra-performance liquid chromatography (UPLC) (Waters Corporation, Milford, MA, USA) and a Thermo Scientific Q-Exactive high resolution/accurate mass spectrometer interfaced with a heated electrospray ionization (HESI-II) source and Orbitrap mass analyzer operated at 35,000 mass resolution (Thermo Fisher Scientific, Waltham, MA, USA). A detailed description of the methodology was previously provided [15]. To identify compounds, hits were compared with existing library entries of purified standards of over 3300 purified standard compounds. Compounds were then assigned to various categories according to their sources, as previously described [16].

2.3. Statistical Analysis

The metabolomics data were log-transformed. Multivariate analysis, including unsupervised principle component analysis (PCA) and supervised orthogonal partial least square-discriminant analysis (OPLS-DA), were run using SIMCA[®] (version 16.0.1) software (Sartorius Stedim Data Analytics AB, Sweden). Two outliers were removed after PCA analysis as part of quality control prior to OPLS-DA. R version 4.0.3 (Vienna, Austria) was used to perform linear models for each metabolite (as the response variable) versus physical activity (active/sedentary) (as the explanatory variables). The model also included the following confounders: age, gender, BMI, and principle components 1 and 2 from the PCA analysis (PC1 and PC2 explain the dominant and secondary source of variability in the dataset, respectively, provided PC2 is uncorrelated with PC1. This method is applied to complex datasets to identify strong patterns and/or additional nuances present in the data). Correcting for these components accounts for their influence and focuses on the effect of the predictor variable on the dependent variable. The multiple testing correction method (False Discovery Rate (FDR)) was used to adjust the nominal p -values. $FDR < 0.1$ was considered statistically significant. Functional enrichment analysis was run on all p -value ordered metabolite lists from linear models performed in the study. This analysis was conducted based on Fisher's exact test, which was followed by the FDR multiple testing correction method. The sub-pathways were previously predefined using Metabolon, and those with less than three top hits were dropped.

3. Results

3.1. Comparing the Effect of Physical Activity on Clinical Metabolic Traits

When assessing the impact of physical activity, the active participants showed lower HOMA-IR, average pulse rate, mean cell hemoglobin concentration, GGT, and GGT 2 than their sedentary counterparts (p -value < 0.05). Additionally, the active participants exhibited higher level of handgrip (left and right), maximum heart rate, creatine kinase, AST, and HDL cholesterol (p -value < 0.05) (Table 1). When conducting these comparisons in males and females separately, data suggested lower levels of insulin, HOMA-IR, C-peptide, GGT, GGT2, and triglycerides in active compared with sedentary males, but not in females. Data also revealed that active males, but not females, had higher level of handgrip (right and left). Active females, but not males, had lower maximum heartbeat and % of basophils, but higher % of lymphocytes, HDL, and total protein compared with their sedentary counterparts (Table A1).

3.2. Multivariate Analysis of Metabolites Differentiating between Active and Sedentary Individuals

Non-targeted metabolomics was applied to determine the metabolic signatures of the 305 subjects. Two subjects deemed outliers from unsupervised principle component analysis (PCA) were removed from further analysis. OPLS-DA was performed to identify components that best differentiated between the active and sedentary cohorts. The model identified one predictive and two orthogonal components from the data, capturing 14.8% variance in the data and 52.3% variance between the study groups. The score plot

(Figure 1A) shows a clear distinction between sedentary (red) and active (green) participants, whereas the loading plot (Figure 1B) shows metabolites responsible for differentiating the two groups. These include xanthine metabolites enriched among the sedentary group, and lysoplasmalogen, plasmalogen, vitamin A, and glutathione metabolites enriched among the active participants. A variable importance in projection (VIP) plot for OPLS-DA model (Figure 1C) shows a quantitative estimation of the discriminatory power of the top 30 metabolites between active and sedentary participants.

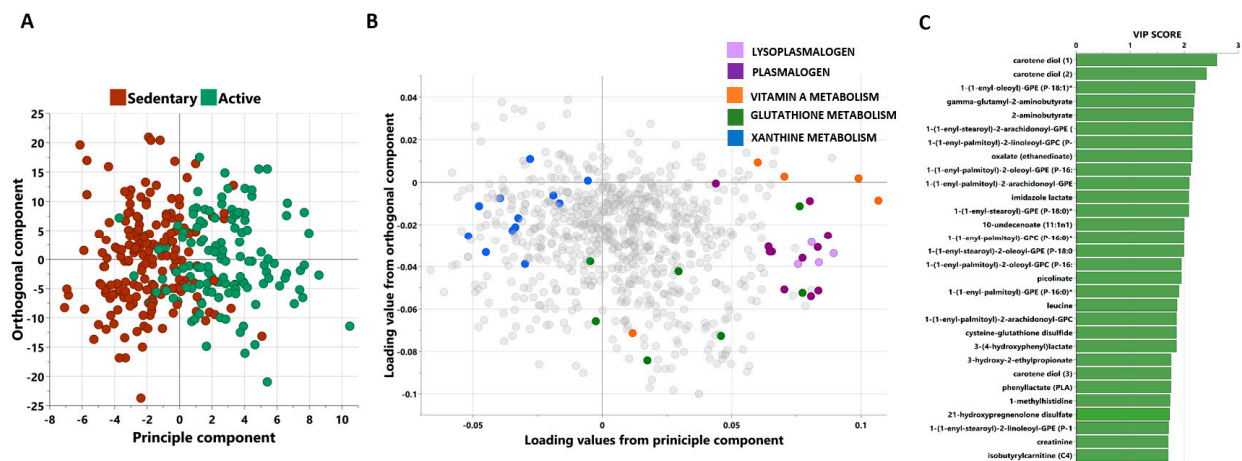


Figure 1. A score plot from the OPLS-DA model comparing active vs sedentary participants showing the class-discriminatory component 1 (x-axis) versus 1 Orthogonal component 1 (y-axis) (A), the corresponding loadings plot from the OPLS-DA model showing metabolites that belong to significantly enriched pathways (B), and a variable importance in projection (VIP) plot for the OPLS-DA model as a quantitative estimation of the discriminatory power of the top 30 metabolites between active and sedentary participants (C). Asterisks (*) indicated on IDs of some metabolites refer to compounds that have not been officially confirmed based on a standard, but their identities are known with confidence.

3.3. Univariate Analysis of Metabolites Differentiating between Moderately Active and Sedentary Individuals

A linear model was used to assess the significance of metabolite associations with physical activity status after correcting for gender, age, BMI, and principal components. Data indicated 14 metabolites associated with activity at ($FDR \leq 0.1$) (Table 2). The univariate analysis confirmed results identified through the OPLS-DA multivariate, including those belonging to vitamin A and xanthine metabolism (Figure 2).

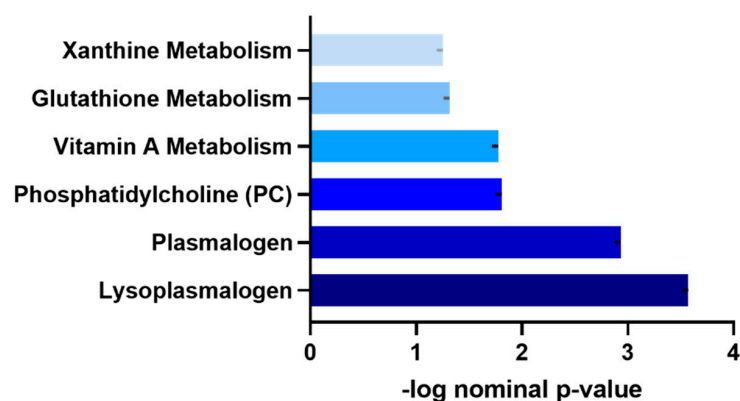


Figure 2. Bar plot of functional enrichment analysis results of active vs sedentary status based on metabolite ranks by p -value using Fisher's exact test.

Table 2. List of metabolites associated with physical activity status.

Metabolite	Sub-Pathway	Super-Pathway	Estimate	SE	p-Value	fdr
10-undecenoate (11:1n1)	Medium chain fatty acid	Lipid	0.22	0.06	7.61×10^{-5}	0.06
imidazole lactate	Histidine metabolism	Amino Acid	0.12	0.03	2.04×10^{-4}	0.06
1-palmitoyl-GPA (16:0)	Lysophospholipid	Lipid	−0.24	0.07	2.58×10^{-4}	0.06
paraxanthine	Xanthine metabolism	Xenobiotics	−0.35	0.10	3.58×10^{-4}	0.06
undecenoylcarnitine (C11:1)	Fatty acid metabolism (acyl carnitine, monounsaturated)	Lipid	0.21	0.06	3.69×10^{-4}	0.06
carotene diol (2)	Vitamin A metabolism	Cofactors and Vitamins	0.17	0.05	5.53×10^{-4}	0.08
thyroxine	Tyrosine metabolism	Amino acid	−0.11	0.03	6.31×10^{-4}	0.08
fructosyllysine	Lysine metabolism	Amino acid	−0.14	0.04	7.16×10^{-4}	0.08
carotene diol (1)	Vitamin A metabolism	Cofactors and vitamins	0.16	0.05	1.04×10^{-3}	0.09
sphingomyelin (d18:2/14:0, d18:1/14:1) *	Sphingomyelins	Lipid	−0.10	0.03	1.16×10^{-3}	0.09
fructose	Fructose, mannose and galactose metabolism	Carbohydrate	−0.09	0.03	1.23×10^{-3}	0.09
gamma-glutamyl-2-aminobutyrate	Gamma-glutamyl amino acid	Peptide	0.14	0.04	1.52×10^{-3}	0.1
N-acetylproline	Urea cycle; arginine and proline metabolism	Amino Acid	−0.13	0.04	1.77×10^{-3}	0.1
oxalate (ethanedioate)	Ascorbate and aldarate metabolism	Cofactors and Vitamins	0.13	0.04	2.12×10^{-3}	0.1

Linear regression was used to assess the significance of metabolite associations with active and sedentary physical activity status after correcting for gender, age, BMI, and Principal Components ($FDR \leq 0.1$). Asterisks (*) indicated on IDs of some metabolites refer to compounds that have not been officially confirmed based on a standard, but their identities are known with confidence.

3.4. Functional Enrichment Analysis of Metabolites Differentiating Moderately Active and Sedentary Non-Obese Individuals

Functional enrichment analysis of metabolites was performed using metabolite ranks by *p*-value from the univariate analysis. Pathway enrichment analysis revealed an over-representation of lysoplasmalogen, plasmalogen, phosphatidylcholine (PC), vitamin A, glutathione, and xanthine metabolism (Figure 2), similarly to OPLS-DA (Figure 1B).

3.5. Identifying Oxidative Stress-Related Metabolites That Significantly Changed between Sedentary and Active Participants

As glutathione metabolism was one of the enriched pathways associated with moderate physical activity, we sought further characterization of the nature of associated metabolites after correcting for age, gender, BMI, and principal components. The results showed cysteine-glutathione disulfide (Figure 3A), 2-aminobutyrate (Figure 3B), cys-gly oxidized (Figure 3C), and 2-hydroxybutyrate (Figure 3D) with nominal significant elevation in active subjects (*p*-values = 0.003, 0.008, 0.019, and 0.05, respectively), as shown in Table 3 and Figure 3.

Table 3. Investigation of glutathione metabolites associated with moderate physical activity.

Glutathione Metabolites	Estimate	SE	p-Value
cysteine-glutathione disulfide	0.229	0.078	0.003
2-aminobutyrate	0.089	0.034	0.009
cys-gly, oxidized	0.091	0.039	0.020
2-hydroxybutyrate/2-hydroxyisobutyrate	0.094	0.050	0.051
cysteinylglycine disulfide *	0.026	0.025	0.296
cysteinylglycine	−0.011	0.054	0.836
5-oxoproline	0.001	0.020	0.962

Asterisks (*) indicated on IDs of some metabolites refer to compounds that have not been officially confirmed based on a standard, but their identities are known with confidence.

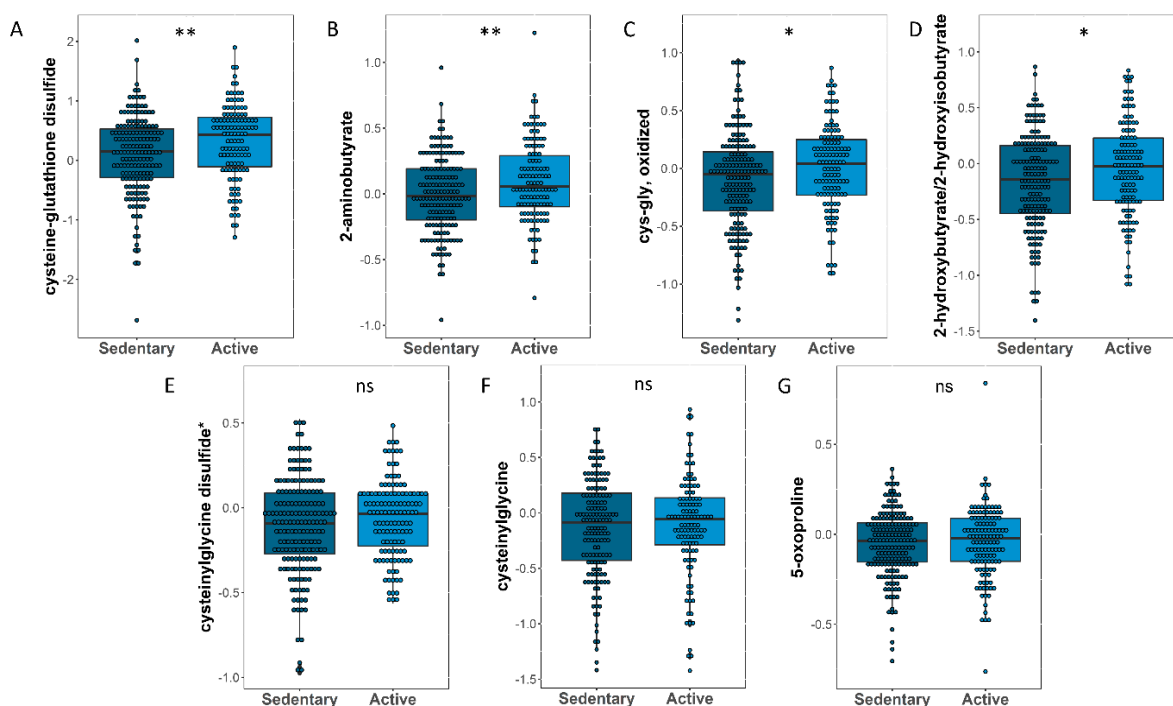


Figure 3. Box plot of metabolites related to glutathione metabolism differentiating between sedentary and active participants, including cysteine-glutathione disulfide (A), 2-aminobutyrate (B), cys-gly, oxidized (C), 2-hydroxybutyrate/2-hydroxyisobutyrate (D), cysteinylglycine disulfide (E), Cysteinylglycine (F) and 5-oxoproline (G). Asterisks (*) indicated on IDs of some metabolites refer to compounds that have not been officially confirmed based on a standard, but their identities are known with confidence. **/*/ns in the figure represents p -value <0.01 / <0.05 /not significant respectively.

3.6. Association between Antioxidant Response and Clinical Traits in Active Individuals

In order to understand the implications of elevated oxidative stress-related metabolites on the general health of active individuals, Spearman's correlation was performed between glutathione metabolites and conventional clinical traits (Figure 4). The data indicated that blood sugar-related traits showed negative correlations with glutathione metabolites, evident mainly in males but not in females. Among physical tests, handgrip was positively correlated with glutathione metabolites, but the average pulse and maximum heart rate were negatively correlated with glutathione metabolites. These associations were not significant when considering each gender separately. Among blood inflammatory markers, hemoglobin and hematocrit were positively correlated with glutathione metabolites in both genders, whereas basophils and platelets were negatively correlated. Liver function enzymes were negatively correlated with oxidative stress markers, except for cysteine-glutathione disulfide, which was positively correlated with multiple liver function enzymes in both males and females. Among the iron profile traits, iron and ferritin were positively correlated with oxidative stress markers, whereas TIBC and UIBC were negatively correlated in total, but not when considering each gender separately. Finally, among the vitamins and hormones, total testosterone was positively correlated with oxidative stress-related metabolites, evident mostly in females but not males. Conversely, estradiol was negatively correlated in general, but not when each gender was considered separately.

Studies showed that after four weeks of intensive exercise, levels of thiobarbituric acid reactive substances increased, and total antioxidant status decreased in active and sedentary individuals [26]. Hence, moderate training can be considered a strategy for promoting changes in the balance between pro- and antioxidant products [27]. Another study of sedentary non-obese, overweight, and obese volunteers revealed that a 20-min run at ~70% of VO_2 max triggers elevated levels of glutathione reductase activity in obese subjects but decreased MCP-1 in non-obese individuals, in addition to decreased EGF in all groups. These findings suggest that glutathione reductase is involved differently in the adaptive metabolic changes and redox responses induced by physical exercise in non-active subjects with different ranges of BMI. Identification of these biomarkers could help to identify people who are at higher risk of developing disease related to oxidative stress [28].

In this study, untargeted metabolomics analysis was utilized to characterize the unique serum metabolic signature of moderately active non-obese participants. The emerging data revealed significant differences in metabolite levels between sedentary and active young non-obese individuals. These differences included a clear signature of oxidative stress-related metabolites that significantly changed between sedentary and active participants.

Our results showed that plasmalogen and lysoplasmalogen metabolism was enriched among the active participants. Plasminogens are a class of membrane glycerophospholipids with unique properties. They contain a vinyl-ether-linked alkyl chain at the sn-1 position of the glycerol backbone and typically, a polyunsaturated fatty acyl chain at the sn-2 position [29,30]. Plasmalogens are critical for human health and have established roles in neuronal development, immune response, and as endogenous antioxidants that protect cells from oxidative stress. Their vinyl-ether bond could indeed be among the first targets for newly formed radicals [29,31,32]. However, the mechanistic bases of these and other biological functions of plasmalogens are not well defined. In contrast, lysoplasmalogens increase membrane fluidity to facilitate cell fusion and migration of circulating inflammatory cells through the endothelium. The lysoplasmalogen may be converted back to plasmalogen in a transacylation reaction. Alternatively, it may be degraded enzymatically by hydrolytic enzymes that include a phospholipase C, a phospholipase D, and lysoplasmalogenase. As the enzyme that cleaves the vinyl-ether bond, lysoplasmalogenase is probably an important enzyme in maintaining the balance between plasmalogen and lysoplasmalogen, thereby preserving membrane stability and function. If levels of lysoplasmalogen rise, they may disrupt and lyse cell membranes. Conversely, if lysoplasmalogen levels are too low, the transacylation reaction cannot occur. Consequently, membrane plasmalogen levels will decrease, thereby disturbing membrane structure and function by that route [33].

Our emerging data revealed that moderate exercise can trigger upregulation of both plasmalogen and lysoplasmalogen levels, highlighting the potential of these metabolites as protection mechanism against oxidative stress.

Furthermore, our results revealed the effect of exercise on increasing vitamin A in the active group. Retinoids (vitamin A) are required for maintaining many essential physiological processes in the body, including normal growth and development as well as healthy vision, immune system, reproductive functions, and skin. Additionally, vitamin A possesses an antioxidative stress function conferred by the hydrophobic chain of polyene units that can quench singlet oxygen, neutralize thiyl radicals, and stabilize peroxy radicals. Because of its structures, vitamin A can autoxidize when O_2 tension increases, thereby constituting the most effective antioxidant at low-oxygen tensions, which are typical of physiological levels found in tissues [34]. In this study, we confirmed the enrichment of vitamin A in active individuals, providing evidence of an additional antioxidant mechanism associated with moderate physical activity.

Our results also showed that PC were enriched among the active participants. PC, the most abundant of the phospholipids, have several metabolic functions in organs such as the liver and the intestine, as well as important structural and signaling functions in biological membranes [35,36]. PC liposomes have both neuroprotective and antioxidative properties through the inhibition of microglial activation [37]. PC decreased oxidative stress

in the sciatic nerve by increasing antioxidant levels (glutathione, glutathione peroxidase, and superoxide dismutase activity) [38]. The PC and phosphatidyl ethanolamine (PE) composition in skeletal muscle has been linked to insulin sensitivity. Previous studies have evaluated the relationships between skeletal muscle PC, PE, physical exercise, and insulin sensitivity. Exercise intervention for 12 weeks enhanced insulin sensitivity by 33%, skeletal muscle levels of PC by 21%, PE by 42%, and reduced the PC:PE ratio by 16%. One bicycle session, which is an example of moderate physical activity, reduced PC:PE by 5%. Furthermore, PC:PE correlated negatively with insulin sensitivity and oxidative phosphorylation [39]. Our results showed an over-representation of PC among the active participants, providing further confirmation of the underlying mechanism of physical activity in preventing oxidative stress. All the metabolites mentioned above (plasmalogen, lysoplasmalogen, vitamin A, and PC) can be taken as supplements for their known antioxidant effect. The emerging data have shown that moderate exercise could be used as a strategy to increase their levels without the need for supplement use.

Our results have also shown that glutathione metabolism is one of the enriched pathways associated with moderate physical activity. GSH is a tripeptide thiol antioxidant composed of the amino acids, glutamic acid, cysteine, and glycine [40]. It is synthesized by the sequential addition of cysteine to glutamate, followed by the addition of glycine. The sulfhydryl group (–SH) of cysteine are involved in reduction and conjugation reactions that are usually considered the most important functions of GSH. These reactions provide the means for the removal of peroxides and many xenobiotic compounds. The protective functions of glutathione are mediated through reduction, conjugation, and interaction with other non-enzymatic antioxidants. Most of the GSH in antioxidant defense in cells is utilized by three members of the glutathione peroxidase (GPx) family and by one of the peroxiredoxins (Prdx 6). These enzymes catalyze the reduction of H₂O₂ by GSH into H₂O and GSSG [41].

Since glutathione (GSH) is the most important low molecular weight antioxidant synthesized in cells, we investigated changes in glutathione metabolites in response to moderate exercise in non-obese individuals. Several metabolites related to glutathione metabolism were increased with moderate physical activity, including cysteine-glutathione disulfide, 2-aminobutyrate, cys-gly, oxidized and 2-hydroxybutyrate/2-hydroxyisobutyrate. 2-hydroxybutyrate is related to glutathione synthesis in the sense that, under conditions of oxidative stress, homocysteine metabolism is diverted from the transmethylation pathway (forming methionine) to the transsulfuration pathway (forming cystathionine), and then cysteine, to feed glutathione synthesis. 2-Oxobutyrate, produced along with cysteine from cystathionine, is subsequently reduced to 2-hydroxybutyrate by lactate dehydrogenase. Much like the production of lactate from pyruvate by the same enzyme, it has been reported that an increased [NADH]/[NAD⁺] ratio is the most important factor for the production of 2-hydroxybutyrate. This can explain the exercise-induced increase in 2-hydroxybutyrate, since a highly accelerated glycolytic flux during high-intensity exercise would result in a high rate of NAD⁺-to-NADH conversion at the level of glyceraldehyde 3-phosphate dehydrogenase [42]. Our results also showed elevated cysteine-glutathione disulfide in active individuals, confirming previous studies showing that exhaustive exercise can increase muscle glutathione disulfide content. Concentrations of the glutathione-related amino acids, glutamate, cysteine, and aspartate, were significantly increased in the same muscle after exhaustion [43].

In the present study, glutathione metabolites (Cys-gly, oxidized, 5 oxoproline, cysteinylglycine, and cysteinylglycine disulfide) were associated with higher hemoglobin and hematocrit concentrations in males and females. Recent studies have suggested that a strong effect of oxidative stress (reduced glutathione) on glutathione peroxidase and glutathione reductase levels may reduce hemoglobin concentration in diabetic patients [44]. This in line with our results, where glutathione metabolites (antioxidants) were associated with higher hemoglobin and hematocrit concentrations. When analyzing these associations in males and females separately, our data indicated that the positive associations of Cys-gly,

oxidized, and cysteinylglycine with hemoglobin were identified in males, but not in females. This is expected since differences in hemoglobin—among other measured clinical traits, including HDL and sex hormones—are sex-related [45,46].

Our data also revealed that glutathione metabolites, including Cys-gly, oxidized, cysteinylglycine, and cysteinylglycine disulfide, were positively associated with liver function enzymes, including total bilirubin in both genders. Bilirubin can also act as an antioxidant *in vitro*, but whether its redox activity is physiologically relevant is unclear. Bilirubin's unique redox activity toward $O_2^{\cdot-}$ may underlie a prominent physiologic role despite being significantly less abundant than other endogenous and exogenous antioxidants [47]. In addition to its antioxidant properties, bilirubin serves as a metabolic hormone that modulates metabolism and immune response. Like other hormones, bilirubin travels through the circulatory system and reaches its target inside cells, primarily PPAR α , to elicit gene responses [46,48].

Our data also revealed a positive correlation between uric acid and glutathione metabolites in females. Reduced GSH and uric acid are of particular interest as they are both antioxidants that are closely linked to the energy metabolism homeostasis. The magnitude of increase in plasma uric acid level following a session of intense exercise was previously suggested to be differently affected by the type of exercise performed [49]. Additionally, our data revealed a positive association between the iron profile and glutathione metabolites (Cys-gly, oxidized, cysteinylglycine, and cysteinylglycine disulfide). Previous studies have reported the accumulation of iron, ROS, and glutathione levels, as demonstrated by Wang et al., indicating their intricate relationship in cellular redox regulation and oxidative stress management [50].

Our findings are limited by the relatively small sample size and lack of assessment of the physiological mechanisms that are activated by the generation of free radicals during moderate-intensity physical exercise. Future studies will replicate these findings in larger cohorts and will assess the correlation between the identified biomarkers and the previously reported physiological mechanisms that are triggered by free radicals during exercise. Another limitation of our study is the presence of significant differences in multiple related clinical outcomes, such as lipid profile, between moderately active and inactive individuals. While these differences are integral to metabolic processes, considering them as potential confounders might affect the interpretation of our results. It should be noted that while our primary focus is on the correlation analysis across different exercise levels, there is potential value in considering the relationship between oxidative stress markers and clinical traits among sedentary individuals. Such an exploration could provide additional insights into the intricate interplay between oxidative stress and clinical characteristics, which will be considered in future studies.

5. Conclusions

This study has shed light on the metabolic pathways underlying the protective effect of moderate exercise in non-obese, apparently healthy individuals, including lysoplasmalogen, plasmalogen, PC, vitamin A, glutathione and related metabolites, and xanthine metabolism. The identification of these metabolites and metabolic pathways in response to moderate physical activity provides a better understanding of exercise physiology as a potent antioxidant mechanism with therapeutic potential. The emerging data offer valuable insights for researchers investigating the long-term effects of regular exercise on the metabolism of healthy, active adults. This is especially significant given that the majority of previous research has focused on short-term responses primarily observed in athletes.

Author Contributions: S.A., N.A., A.D. (Asmma Doudin) and M.A.E. carried out the data analysis. S.A. wrote the manuscript. M.S., A.D. (Alexander Domling), A.A., A.A.A. and M.A.E. contributed to the experimental design. M.A.E. is responsible for the integrity of the work as a whole. All authors have read and agreed to the published version of the manuscript.

Funding: This research was funded by Qatar University grant number IRCC-2022-467 (M.A.E., M.S. and A.A).

Institutional Review Board Statement: The study was approved by the Institutional Review Boards of the Qatar Biobank (QF-QBB-RES-ACC-00066) and Qatar University (QU-IRB 1716-E/22).

Informed Consent Statement: Informed consent was obtained from all subjects involved in the study.

Data Availability Statement: Data is contained within the article.

Acknowledgments: The authors would like to acknowledge the Qatar University Research Office for funding this work. Open Access funding provided by the Qatar National Library.

Conflicts of Interest: The authors declare no conflict of interest.

Appendix A

Table A1. Demographics of participants categorized by gender and physical activity.

Clinical Traits	Males			Females		
	Sedentary (89)	Moderately Active (70)		Sedentary (88)	Moderately Active (58)	
HOMA-IR (categorical)						
	<1.85	56	44	54	34	0.740
	>1.85	33	26	34	24	
Vital signs						
	Age	26 (24–28)	26 (24–28)	26 (23–28)	25 (23–28)	0.586
	BMI	25.33 (23.05–27.28)	25.5 (23.65–27.39)	23.8 (22.25–26.145)	24.235 (21.99–27.41)	0.507
	Average systolic BP	112 (106–119)	113 (106–119)	98 (93.75–104)	101 (97–106)	0.031
	Average diastolic BP	71 (65–76)	69 (62.5–75)	66 (62–70)	66.5 (61.25–71)	0.911
	Average pulse rate	66 (61–73)	64 (59–71.75)	71 (67–77.25)	70 (63.25–76)	0.165
Blood sugar						
	Fasting time (minutes)	139 (74–259)	209 (100–635)	168 (94–569.75)	174 (76.5–593.25)	0.940
	Glucose (mmol/L)	5 (4.7–5.4)	4.9 (4.7–5.2)	4.7 (4.5–5)	4.8 (4.5–5.065)	0.994
	Hb A1c	5.2 (5.1–5.4)	5.3 (5.1–5.4)	5.2 (5–5.4)	5.2 (5–5.4)	0.335
	Insulin (Uu/mL)	10 (6.4–16.775)	7 (5.47–11.87)	7.5 (6–10.15)	8 (6–11)	0.903
	HOMA-IR (continuous)	2.195 (1.39–3.735)	1.585 (1.11–2.73)	1.59 (1.215–2.28)	1.76 (1.195–2.43)	0.936
	C-Peptide ng mL	2.13 (1.57–3.39)	1.78 (1.34–2.79)	1.65 (1.39–2.42)	1.7 (1.36–2.495)	0.723
Physical tests						
	Sitting height (cm)	91.4 (88.3–96.5)	91.95 (88.27–136.12)	89.2 (84–134.65)	85.6 (82.875–133.575)	0.163
	Weight (kg)	75.7 (68.6–82.7)	76.3 (69.175–83.625)	60.6 (57.7–66.225)	61.3 (56.55–70.825)	0.722
	Waist size (cm)	87 (79–90)	83.5 (79–89)	73 (69–77)	74 (69–79)	0.319
	Hip size (cm)	99 (96–104)	101 (96–105)	101 (97–106)	103 (97–107)	0.805
	Waist to hip ratio	0.847 (0.8–0.89)	0.84 (0.795–0.865)	0.725 (0.68–0.75)	0.72 (0.70–0.77)	0.331
	Handgrip (left)	38 (34–44)	43.5 (36.25–48)	22 (18–24)	22 (18–28)	0.200
	Handgrip (right)	40 (36–46)	46.5 (39.25–50)	22.5 (20–28)	26 (20.25–28)	0.092
	Maximum heart rate (beats/min)	118.5 (108–128)	114 (106–125)	139 (129–150)	131 (120–139)	0.001
	Run time (seconds)	764 (742–764)	764 (742–764)	743 (684–764)	743 (684–764)	0.912
Blood inflammatory markers						
	Hemoglobin (g/dL)	15 (14.4–15.7)	14.9 (14.5–15.6)	12.4 (11.15–13.25)	12.45 (11.8–13)	0.516
	Estimated hemoglobin *	19.65 (18.22–20.295)	18.9 (17.65–20.54)	16.7 (3.8)	16.98 (15.36–21.02)	0.078
	Hematocrit %	44.5 (42.7–46.5)	44.7 (43–46.3)	37.9 (34.85–39.85)	37.65 (36.025–39.55)	0.356
	Red Blood Cell × 10 ⁶ µL	5.2 (5–5.6)	5.3 (5–5.5)	4.5 (4.3–4.8)	4.5 (4.225–4.9)	0.780
	White Blood Cell (× 10 ³ µL)	5.8 (4.9–7)	6.3 (5.5–7.4)	6.8 (5.75–7.95)	6.75 (5.425–7.9)	0.430
	Monocyte auto %	8.1 (6.8–9.5)	7.6 (6.4–8.8)	6.4 (5.6–7.95)	7 (5.525–7.675)	0.678
	Monocytes Auto (× 10 ³ µL)	0.5 (0.4–0.6)	0.5 (0.4–0.6)	0.4 (0.4–0.6)	0.4 (0.4–0.5)	0.661
	Neutrophil Auto (× 10 ³ µL)	3 (2.5–3.9)	3.3 (2.7–4.1)	4 (3.15–4.9)	3.4 (2.825–4.4)	0.084
	Neutrophil auto %	51.3 (46.2–58.5)	53.4 (9.64)	57.2 (8.42)	54.5 (10.26)	0.034
	Lymphocyte Auto (× 10 ³ µL)	2.2 (1.7–2.6)	2.2 (1.9–2.4)	2.2 (1.85–2.65)	2.3 (1.925–2.8)	0.366
	Lymphocyte auto %	36.4 (30.7–40.7)	34.8 (31.3–40.8)	32.5 (28.6–36.7)	35.85 (31.1–42.475)	0.016
	Eosinophil auto %	2.9 (1.7–3.6)	2.9 (1.6–4.3)	1.7 (1–2.7)	1.8 (1.125–2.85)	0.601
	Eosinophils Auto (× 10 ³ µL)	0.2 (0.1–0.2)	0.2 (0.1–0.3)	0.1 (0.1–0.2)	0.1 (0.1–0.2)	0.825
	Basophil auto %	0.6 (0.4–0.8)	0.6 (0.4–0.7)	0.7 (0.45–0.8)	0.5 (0.3–0.8)	0.023
	Basophils Auto (× 10 ³ µL)	0 (0–0.1)	0 (0–0)	0 (0–0.1)	0 (0–0.0325)	0.216
	Platelet (× 10 ³ µL)	211 (180–259.75)	220 (197–254)	268.5 (233–302.75)	254.5 (218.5–297.75)	0.308
	Mean Cell Hemoglobin (pg)	28.9 (27.2–29.9)	28.5 (26.7–29.9)	27.8 (24.8–29.45)	27.65 (25.725–29.2)	0.974
	Mean Cell Hemoglobin Concentration (g/dL)	33.8 (33.2–34.5)	33.5 (32.9–34)	33 (32.2–33.6)	32.8 (32.325–33.475)	0.546
	C Reactive Protein (mg/L)	5 (5–5)	5 (5–5)	5 (5–5)	5 (5–5)	0.990

Table A1. Cont.

Clinical Traits	Males			Females		
	Sedentary (89)	Moderately Active (70)		Sedentary (88)	Moderately Active (58)	
Kidney profile						
Sodium (mmol/L)	141 (140–142)	141 (140–142)	0.678	139 (138–141)	140 (139–141)	0.098
Potassium (mmol/L)	4.3 (4.1–4.4)	4.3 (4.2–4.5)	0.058	4.25 (4–4.425)	4.2 (4–4.4)	0.594
Chloride (mmol/L)	101 (100–102)	101 (99.25–102)	0.757	101 (100–103)	101 (100–102)	0.897
Bicarbonate (mmol/L)	27 (26–28)	27.5 (26–28)	0.106	26 (24–27)	25 (24–27)	0.775
Urea (mmol/L)	4.6 (4–5.2)	4.8 (3.95–5.4)	0.424	3.6 (2.975–4.3)	3.7 (3–4.875)	0.655
Creatinine (mmol/L)	77 (71–83)	77.5 (72–85.5)	0.238	55 (51–61)	54 (49.5–60)	0.424
Calcium (mmol/L)	2.42 (0.07)	2.41 (0.06)	0.730	2.37 (0.07)	2.39 (0.07)	0.113
Calcium Corrected (mmol/L)	2.27 (0.07)	2.27 (0.07)	0.932	2.27 (0.06)	2.27 (0.06)	0.441
Phosphorus (mmol/L)	1.15 (1.18)	1.18 (0.17)	0.319	1.18 (0.16)	1.20 (0.13)	0.447
Uric Acid (μmol/L)	333 (301–369)	338.5 (308.25–380.75)	0.714	230 (199–254.5)	238.5 (202–267)	0.459
Creatine kinase	113 (78–171.5)	121 (95.5–255)	0.036	60 (48.5–81.5)	66.5 (53.25–85.75)	0.291
Creatine Kinase_1 (ng/mL)	1.65 (1.1025–1.865)	1.385 (1.125–1.6625)	0.725	1.02 (0.6075–1.395)	0.89 (0.525–1.1425)	0.577
Creatine kinase_2 (u/L)	112 (72.75–155.5)	161 (115–255)	0.151	58 (53.5–80.5)	67 (51.25–83.5)	0.805
Magnesium (μmol/L)	0.835 (0.05)	0.83 (0.04)	0.655	0.836 (0.05)	0.82 (0.05)	0.451
Homocysteine (μmol/L)	8.95 (7.78–11.025)	9.1 (7.55–10.75)	0.689	7.5 (5.9–8.95)	7.55 (6.075–9.7)	0.546
Total Protein (g/L)	74 (71–76)	74 (71–76)	0.442	73 (70–75.25)	74.5 (72–77)	0.024
Liver profile						
Total bilirubin (μmol/L)	8 (5.85–10)	7.5 (5.7–12)	0.532	6.05 (4.075–8)	5.7 (4.2–7.4)	0.344
Albumin (g/L)	47 (46–49)	47 (46–49)	0.689	45 (44–47)	46 (45–47)	0.131
Alkaline phosphatase (u/L)	70 (60–79)	64.5 (55.25–77.5)	0.061	59.5 (51–69)	59.5 (48.75–71.75)	0.916
Alanine transaminase (u/L)	24 (18–38)	22 (17–34)	0.419	13 (9.75–17)	12.5 (11–16.75)	0.744
Aspartate aminotransferase (u/L)	19 (16–23)	21 (17–26)	0.065	16 (13.75–19)	17 (14.25–19)	0.169
GGT (u/L)	20.5 (17–28.75)	16 (12.5–23.5)	0.025	10 (8–14)	11 (9–13.25)	0.951
GGT 2 (u/L)	24 (16–38)	17 (13–23)	0.002	14 (11.75–18.5)	11 (9–17.75)	0.051
Lipid profile						
HDL Cholesterol (mmol/L)	1.20 (0.31)	1.29 (0.34)	0.100	1.53 (0.26)	1.66 (0.35)	0.018
LDL Cholesterol Calc (mmol/L)	2.96 (2.475–3.715)	2.85 (2.085–3.42)	0.127	2.635 (2.06–3)	2.8 (2.01–3.03)	0.518
Triglyceride (mmol/L)	1.12 (0.86–1.5)	0.98 (0.6225–1.2975)	0.010	0.8 (0.55–1.10)	0.8 (0.6025–1)	0.974
Iron profile						
Iron (umol/L)	18 (14.91–21.87)	16.595 (14.08–20.61)	0.286	12.95 (7.205–17.075)	12 (8.79–15)	0.995
TIBC (umol/L)	55 (51–59.5)	55 (52.25–60)	0.491	64 (60–73.5)	61 (57–70)	0.083
UIBC (umol/L)	37.25 (32.225–42)	39 (31.6–45.2)	0.222	50.7 (43–63.5)	48.9 (42–61.2)	0.382
Ferritin (μg/L)	101 (62.5–149)	105 (57.25–133.75)	0.653	13.5 (7–28.25)	19 (9–33.75)	0.200
Hormones						
Free Thyroxine (pmol/L)	13.4 (12.61–14.58)	13.36 (12.49–14.455)	0.449	13.47 (12.3075–14.7)	13.35 (12.395–14.44)	0.994
Free triiodothyronine (nmol/L)	4.7 (4.4–5.1)	4.6 (4.34–4.9475)	0.146	4.38 (4–4.85)	4.4 (3.95–4.8)	0.733
TSH (mU/L)	1.43 (0.95–2.14)	1.22 (0.9925–1.8275)	0.263	1.535 (0.915–2.46)	1.34 (1.035–1.92)	0.537
Total Testosterone (nmol/L)	19.51 (15.56–23.06)	21.34 (16.99–26.64)	0.135	1.335 (0.935–1.61)	1.19 (0.9–1.49)	0.583
Estradiol (pmol/L)	86 (69.25–113)	98 (77–121)	0.107	285 (148.75–596)	246 (165–505)	0.848
SHBG (nmol/L)	28.65 (22.15–38.2)	33 (24–43.1)	0.180	61 (45–102.7)	64 (45–91)	0.887
Vitamins						
Folate (nmol/L)	20.49 (8.0)	21.05 (16.325–26.175)	0.332	25.1 (7.74)	23.2 (17.1–28.7)	0.235
Vitamin B12 (pmol/L)	331.5 (245.25–422.75)	322 (249–429.5)	0.927	270.5 (216.5–327.5)	252 (200.5–309)	0.237
Dihydroxyvitamin D Total (ng/mL)	14 (12–20)	14 (12–21.75)	0.927	14 (10–20.5)	14.5 (10.25–23)	0.743

BMI body mass index, BP blood pressure, LDL low density lipoprotein, HDL high density lipoprotein, HOMA-IR homeostatic model assessment of insulin resistance, * hemoglobin estimated from hematocrit. Data are presented as mean (SD) for normal variables, median (IQR) for skewed variables, and number (percentage) for categorical parameters. Differences between the groups were tested using Student's *t*-test/Mann-Whitney U test for parametric/non-parametric variables and Fisher's exact test for nominal variables. A *p*-value significance level of 0.05 was used.

Table A2. Significant metabolites from the linear regression analysis that belong to enriched pathways.

Metabolite	Sub-Pathway	Super-Pathway	Estimate	SE	<i>p</i> -Value
cysteine-glutathione disulfide	Glutathione Metabolism	Amino Acid	0.23	0.08	0.003
2-aminobutyrate	Glutathione Metabolism	Amino Acid	0.09	0.03	0.009
cys-gly, oxidized	Glutathione Metabolism	Amino Acid	0.09	0.04	0.02
1-(1-enyl-oleoyl)-GPE (P-18:1) *	Lysoplasmalogen	Lipid	0.12	0.04	0.003
1-(1-enyl-stearoyl)-GPE (P-18:0) *	Lysoplasmalogen	Lipid	0.11	0.04	0.007
1-(1-enyl-palmitoyl)-GPC (P-16:0) *	Lysoplasmalogen	Lipid	0.08	0.03	0.01
1-(1-enyl-palmitoyl)-GPE (P-16:0) *	Lysoplasmalogen	Lipid	0.09	0.04	0.019
1-myristoyl-2-arachidonoyl-GPC (14:0/20:4) *	Phosphatidylcholine (PC)	Lipid	−0.16	0.06	0.011
1-palmitoyl-2-dihomo-linolenoyl-GPC (16:0/20:3n3 or 6) *	Phosphatidylcholine (PC)	Lipid	−0.11	0.04	0.013
1-myristoyl-2-palmitoyl-GPC (14:0/16:0)	Phosphatidylcholine (PC)	Lipid	−0.15	0.06	0.018
1-palmitoyl-2-palmitoleoyl-GPC (16:0/16:1) *	Phosphatidylcholine (PC)	Lipid	−0.11	0.05	0.021
1-stearoyl-2-linoleoyl-GPC (18:0/18:2) *	Phosphatidylcholine (PC)	Lipid	−0.05	0.02	0.038
1-stearoyl-2-oleoyl-GPC (18:0/18:1)	Phosphatidylcholine (PC)	Lipid	−0.07	0.03	0.048
1-(1-enyl-palmitoyl)-2-arachidonoyl-GPC (P-16:0/20:4) *	Plasmalogen	Lipid	0.08	0.03	0.009

Table A2. Cont.

Metabolite	Sub-Pathway	Super-Pathway	Estimate	SE	p-Value
1-(1-enyl-palmitoyl)-2-oleoyl-GPC (P-16:0/18:1) *	Plasmalogen	Lipid	0.08	0.03	0.013
1-(1-enyl-palmitoyl)-2-linoleoyl-GPC (P-16:0/18:2) *	Plasmalogen	Lipid	0.08	0.03	0.018
1-(1-enyl-stearoyl)-2-arachidonoyl-GPE (P-18:0/20:4) *	Plasmalogen	Lipid	0.09	0.04	0.02
1-(1-enyl-palmitoyl)-2-arachidonoyl-GPE (P-16:0/20:4) *	Plasmalogen	Lipid	0.08	0.04	0.02
1-(1-enyl-palmitoyl)-2-oleoyl-GPE (P-16:0/18:1) *	Plasmalogen	Lipid	0.07	0.03	0.026
carotene diol (2)	Vitamin A Metabolism	Cofactors and Vitamins	0.17	0.05	0.001
carotene diol (1)	Vitamin A Metabolism	Cofactors and Vitamins	0.16	0.05	0.001
carotene diol (3)	Vitamin A Metabolism	Cofactors and Vitamins	0.11	0.05	0.043
paraxanthine	Xanthine Metabolism	Xenobiotics	−0.35	0.1	<0.001
theophylline	Xanthine Metabolism	Xenobiotics	−0.25	0.09	0.009
1-methylurate	Xanthine Metabolism	Xenobiotics	−0.2	0.1	0.035
caffeine	Xanthine Metabolism	Xenobiotics	−0.29	0.14	0.044

Asterisks (*) indicated on IDs of some metabolites refer to compounds that have not been officially confirmed based on a standard, but their identities are known with confidence.

References

- Zhang, X.; Cash, R.E.; Bower, J.K.; Focht, B.C.; Paskett, E.D. Physical Activity and Risk of Cardiovascular Disease by Weight Status among U.S Adults. *PLoS ONE* **2020**, *15*, e0232893. [[CrossRef](#)] [[PubMed](#)]
- de Oliveira, L.D.S.S.C.B.; Souza, E.C.; Rodrigues, R.A.S.; Fett, C.A.; Piva, A.B. The Effects of Physical Activity on Anxiety, Depression, and Quality of Life in Elderly People Living in the Community. *Trends Psychiatry Psychother.* **2019**, *41*, 36–42. [[CrossRef](#)] [[PubMed](#)]
- Morelli, C.; Avolio, E.; Galluccio, A.; Caparello, G.; Manes, E.; Ferraro, S.; De Rose, D.; Santoro, M.; Barone, I.; Catalano, S.; et al. Impact of Vigorous-Intensity Physical Activity on Body Composition Parameters, Lipid Profile Markers, and Irisin Levels in Adolescents: A Cross-Sectional Study. *Nutrients* **2020**, *12*, 742. [[CrossRef](#)]
- Myers, J.; Kokkinos, P.; Nyelin, E. Physical Activity, Cardiorespiratory Fitness, and the Metabolic Syndrome. *Nutrients* **2019**, *11*, 1652. [[CrossRef](#)]
- Yang, D.; Yang, Y.; Li, Y.; Han, R. Physical Exercise as Therapy for Type 2 Diabetes Mellitus: From Mechanism to Orientation. *Ann. Nutr. Metab.* **2019**, *74*, 313–321. [[CrossRef](#)]
- Iaccarino, G.; Franco, D.; Sorriento, D.; Strisciuglio, T.; Barbato, E.; Morisco, C. Modulation of Insulin Sensitivity by Exercise Training: Implications for Cardiovascular Prevention. *J. Cardiovasc. Transl. Res.* **2020**, *14*, 256–270. [[CrossRef](#)]
- Sampath Kumar, A.; Maiya, A.G.; Shastry, B.A.; Vaishali, K.; Ravishankar, N.; Hazari, A.; Gundmi, S.; Jadhav, R. Exercise and Insulin Resistance in Type 2 Diabetes Mellitus: A Systematic Review and Meta-Analysis. *Ann. Phys. Rehabil. Med.* **2019**, *62*, 98–103. [[CrossRef](#)]
- Wadley, A.J.; Chen, Y.-W.; Lip, G.Y.H.; Fisher, J.P.; Aldred, S. Low Volume–High Intensity Interval Exercise Elicits Antioxidant and Anti-Inflammatory Effects in Humans. *J. Sports Sci.* **2015**, *34*, 1–9. [[CrossRef](#)] [[PubMed](#)]
- Mullins, A.L.; van Rosendal, S.P.; Briskey, D.R.; Fassett, R.G.; Wilson, G.R.; Coombes, J.S. Variability in Oxidative Stress Biomarkers Following a Maximal Exercise Test. *Biomarkers* **2013**, *18*, 446–454. [[CrossRef](#)]
- Aldred, S. Oxidative and Nitrative Changes Seen in Lipoproteins Following Exercise. *Atherosclerosis* **2007**, *192*, 1–8. [[CrossRef](#)]
- Elosua, R.; Molina, L.; Fito, M.; Arquer, A.; Sanchez-Quesada, J.L.; Covas, M.I.; Ordoñez-Llanos, J.; Marrugat, J. Response of Oxidative Stress Biomarkers to a 16-Week Aerobic Physical Activity Program, and to Acute Physical Activity, in Healthy Young Men and Women. *Atherosclerosis* **2003**, *167*, 327–334. [[CrossRef](#)]
- Powers, S.K.; Deruisseau, K.C.; Quindry, J.; Hamilton, K.L. Dietary Antioxidants and Exercise. *J. Sports Sci.* **2004**, *22*, 81–94. [[CrossRef](#)]
- Schranner, D.; Kastenmüller, G.; Schönfelder, M.; Römisch-Margl, W.; Wackerhage, H. Metabolite Concentration Changes in Humans After a Bout of Exercise: A Systematic Review of Exercise Metabolomics Studies. *Sports Med. Open* **2020**, *6*, 11. [[CrossRef](#)]
- Verri Hernandez, V.; Dordevic, N.; Hantikainen, E.M.; Sigurdsson, B.B.; Smáráson, S.V.; Garcia-Larsen, V.; Gögele, M.; Caprioli, G.; Bozzolan, I.; Pramstaller, P.P.; et al. Age, Sex, Body Mass Index, Diet and Menopause Related Metabolites in a Large Homogeneous Alpine Cohort. *Metabolites* **2022**, *12*, 205. [[CrossRef](#)]
- Al-Khelaifi, F.; Diboun, I.; Donati, F.; Botrè, F.; Alsayrafi, M.; Georgakopoulos, C.; Suhre, K.; Yousri, N.A.; Elrayess, M.A. A Pilot Study Comparing the Metabolic Profiles of Elite-Level Athletes from Different Sporting Disciplines. *Sports Med. Open* **2018**, *4*, 2. [[CrossRef](#)]
- Evans, A.M.; Bridgewater, B.R.; Liu, Q.; Mitchell, M.W.; Robinson, R.J.; Dai, H.; Stewart, S.J.; DeHaven, C.D.; Miller, L.A.D. High Resolution Mass Spectrometry Improves Data Quantity and Quality as Compared to Unit Mass Resolution Mass Spectrometry in High-Throughput Profiling Metabolomics. *J. Postgenomics Drug Biomark. Dev.* **2014**, *4*, 1. [[CrossRef](#)]
- Powers, S.K.; Duarte, J.; Kavazis, A.N.; Talbert, E.E. Reactive Oxygen Species Are Signalling Molecules for Skeletal Muscle Adaptation. *Exp. Physiol.* **2009**, *95*, 1–9. [[CrossRef](#)]
- Powers, S.K.; Jackson, M.J. Exercise-Induced Oxidative Stress: Cellular Mechanisms and Impact on Muscle Force Production. *Physiol. Rev.* **2008**, *88*, 1243–1276. [[CrossRef](#)]
- Vinña, J.; Gomez-Cabrera, M.-C.; Lloret, A.; Marquez, R.; Miñana, J.B.; Pallardó, F.V.; Sastre, J. Free Radicals in Exhaustive Physical Exercise: Mechanism of Production, and Protection by Antioxidants. *IUBMB Life* **2000**, *50*, 271–277. [[CrossRef](#)]

20. Cooper, C.E.; Vollaard, N.B.J.; Choueiri, T.; Wilson, M.T. Exercise, Free Radicals and Oxidative Stress. *Biochem. Soc. Trans.* **2002**, *30*, 280–285. [[CrossRef](#)]
21. Tauler, P.; Aguiló, A.; Gimeno, I.; Guix, P.; Tur, J.A.; Pons, A. Different Effects of Exercise Tests on the Antioxidant Enzyme Activities in Lymphocytes and Neutrophils. *J. Nutr. Biochem.* **2004**, *15*, 479–484. [[CrossRef](#)]
22. Cipryan, L. The Effect of Fitness Level on Cardiac Autonomic Regulation, IL-6, Total Antioxidant Capacity, and Muscle Damage Responses to a Single Bout of High-Intensity Interval Training. *J. Sport Health Sci.* **2018**, *7*, 363–371. [[CrossRef](#)]
23. Varamenti, E.; Tod, D.; Pullinger, S.A. Redox Homeostasis and Inflammation Responses to Training in Adolescent Athletes: A Systematic Review and Meta-Analysis. *Sports Med. Open* **2020**, *6*, 34. [[CrossRef](#)]
24. Batista, D.F.; Polegato, B.F.; da Silva, R.C.; Claro, R.T.; Azevedo, P.S.; Fernandes, A.A.; Okoshi, K.; de Paiva, S.A.R.; Minicucci, M.F.; Zornorff, L.A.M. Impact of Modality and Intensity of Early Exercise Training on Ventricular Remodeling after Myocardial Infarction. *Oxidative Med. Cell. Longev.* **2020**, *2020*, 5041791. [[CrossRef](#)]
25. Chevion, S.; Moran, D.S.; Heled, Y.; Shani, Y.; Regev, G.; Abbou, B.; Berenshtein, E.; Stadtman, E.R.; Epstein, Y. Plasma Antioxidant Status and Cell Injury after Severe Physical Exercise. *Proc. Natl. Acad. Sci. USA* **2003**, *100*, 5119–5123. [[CrossRef](#)]
26. Palazzetti, S.; Richard, M.-J.; Favier, A.; Margaritis, I. Overloaded Training Increases Exercise-Induced Oxidative Stress and Damage. *Can. J. Appl. Physiol.* **2003**, *28*, 588–604. [[CrossRef](#)]
27. Hendrix, J.; Nijs, J.; Ickmans, K.; Godderis, L.; Ghosh, M.; Polli, A. The Interplay between Oxidative Stress, Exercise, and Pain in Health and Disease: Potential Role of Autonomic Regulation and Epigenetic Mechanisms. *Antioxidants* **2020**, *9*, 1166. [[CrossRef](#)]
28. Accattato, F.; Greco, M.; Pullano, S.A.; Carè, I.; Fiorillo, A.S.; Pujia, A.; Montalcini, T.; Foti, D.P.; Brunetti, A.; Gulletta, E. Effects of Acute Physical Exercise on Oxidative Stress and Inflammatory Status in Young, Sedentary Obese Subjects. *PLoS ONE* **2017**, *12*, e0178900. [[CrossRef](#)]
29. Dorninger, F.; Werner, E.R.; Berger, J.; Watschinger, K. Regulation of Plasmalogen Metabolism and Traffic in Mammals: The Fog Begins to Lift. *Front. Cell Dev. Biol.* **2022**, *10*, 946393. [[CrossRef](#)]
30. Paul, S.; Lancaster, G.I.; Meikle, P.J. Plasmalogens: A Potential Therapeutic Target for Neurodegenerative and Cardiometabolic Disease. *Prog. Lipid Res.* **2019**, *74*, 186–195. [[CrossRef](#)]
31. Astudillo, A.M.; Balboa, M.A.; Balsinde, J. Compartmentalized Regulation of Lipid Signaling in Oxidative Stress and Inflammation: Plasmalogens, Oxidized Lipids and Ferroptosis as New Paradigms of Bioactive Lipid Research. *Prog. Lipid Res.* **2023**, *89*, 101207. [[CrossRef](#)] [[PubMed](#)]
32. Stadelmann-Ingrand, S.; Favreliere, S.; Fauconneau, B.; Mauco, G.; Tallineau, C. Plasmalogen Degradation by Oxidative Stress: Production and Disappearance of Specific Fatty Aldehydes and Fatty α -Hydroxyaldehydes. *Free Radic. Biol. Med.* **2001**, *31*, 1263–1271. [[CrossRef](#)] [[PubMed](#)]
33. Wu, L.-C.; Pfeiffer, D.R.; Calhoun, E.A.; Madaia, F.; Marcucci, G.; Liu, S.; Jurkowitz, M.S. Purification, Identification, and Cloning of Lysoplasmalogenase, the Enzyme That Catalyzes Hydrolysis of the Vinyl Ether Bond of Lysoplasmalogen. *J. Biol. Chem.* **2011**, *286*, 24916–24930. [[CrossRef](#)] [[PubMed](#)]
34. Olubukola Sinbad, O.; Folorunsho, A.A.; Olabisi, O.L.; Abimbola Ayoola, O.; Johnson Temitope, E. Vitamins as Antioxidants. *J. Food Sci. Nutr. Res.* **2019**, *2*, 214–235. [[CrossRef](#)]
35. Carlsson, E.R.; Allin, K.H.; Madsbad, S.; Fenger, M. Phosphatidylcholine and Its Relation to Apolipoproteins A-1 and B Changes after Roux-En-Y Gastric Bypass: A Cohort Study. *Lipids Health Dis.* **2019**, *18*, 169. [[CrossRef](#)]
36. Li, J.; Xin, Y.; Li, J.; Chen, H.; Li, H. Phosphatidylethanolamine N-Methyltransferase: From Functions to Diseases. *Aging Dis.* **2023**, *14*, 879. [[CrossRef](#)]
37. Hashioka, S.; Han, Y.-H.; Fujii, S.; Kato, T.; Monji, A.; Utsumi, H.; Sawada, M.; Nakanishi, H.; Kanba, S. Phosphatidylserine and Phosphatidylcholine-Containing Liposomes Inhibit Amyloid β and Interferon- γ -Induced Microglial Activation. *Free Radic. Biol. Med.* **2007**, *42*, 945–954. [[CrossRef](#)]
38. Kim, S.T.; Kyung, E.J.; Suh, J.S.; Lee, H.S.; Lee, J.H.; Chae, S.I.; Park, E.S.; Chung, Y.H.; Bae, J.; Lee, T.J.; et al. Phosphatidylcholine Attenuated Docetaxel-Induced Peripheral Neurotoxicity in Rats. *Drug Chem. Toxicol.* **2017**, *41*, 476–485. [[CrossRef](#)]
39. Lee, S.; Norheim, F.; Gulseth, H.L.; Langleite, T.M.; Aker, A.; Gundersen, T.E.; Holen, T.; Birkeland, K.I.; Drevon, C.A. Skeletal Muscle Phosphatidylcholine and Phosphatidylethanolamine Respond to Exercise and Influence Insulin Sensitivity in Men. *Sci. Rep.* **2018**, *8*, 6531. [[CrossRef](#)]
40. Irato, P.; Santovito, G. Enzymatic and Non-Enzymatic Molecules with Antioxidant Function. *Antioxidants* **2021**, *10*, 579. [[CrossRef](#)]
41. Forman, H.J.; Zhang, H.; Rinna, A. Glutathione: Overview of Its Protective Roles, Measurement, and Biosynthesis. *Mol. Asp. Med.* **2009**, *30*, 1–12. [[CrossRef](#)]
42. Pechlivanis, A.; Papaioannou, K.G.; Tsalis, G.; Saraslanidis, P.; Mougios, V.; Theodoridis, G.A. Monitoring the Response of the Human Urinary Metabolome to Brief Maximal Exercise by a Combination of RP-UPLC-MS and 1H NMR Spectroscopy. *J. Proteome Res.* **2015**, *14*, 4610–4622. [[CrossRef](#)] [[PubMed](#)]
43. Gomez-Cabrera, M.C.; Carretero, A.; Millan-Domingo, F.; Garcia-Dominguez, E.; Correias, A.G.; Olaso-Gonzalez, G.; Viña, J. Redox-Related Biomarkers in Physical Exercise. *Redox Biol.* **2021**, *42*, 101956. [[CrossRef](#)] [[PubMed](#)]
44. Waggiallah, H.; Alzohairy, M. The Effect of Oxidative Stress on Human Red Cells Glutathione Peroxidase, Glutathione Reductase Level, and Prevalence of Anemia among Diabetics. *N. Am. J. Med. Sci.* **2011**, *3*, 344–347. [[CrossRef](#)]
45. Murphy, W.G. The Sex Difference in Haemoglobin Levels in Adults—Mechanisms, Causes, and Consequences. *Blood Rev.* **2014**, *28*, 41–47. [[CrossRef](#)] [[PubMed](#)]

46. Creeden, J.F.; Gordon, D.M.; Stec, D.E.; Hinds, T.D. Bilirubin as a Metabolic Hormone: The Physiological Relevance of Low Levels. *Am. J. Physiol. -Endocrinol. Metab.* **2021**, *320*, E191–E207. [[CrossRef](#)]
47. Vasavda, C.; Kothari, R.; Malla, A.P.; Tokhunts, R.; Lin, A.; Ji, M.; Ricco, C.; Xu, R.; Saavedra, H.G.; Sbodio, J.I.; et al. Bilirubin Links Heme Metabolism to Neuroprotection by Scavenging Superoxide. *Cell Chem. Biol.* **2019**, *26*, 1450–1460.e7. [[CrossRef](#)]
48. Vitek, L.; Tiribelli, C. Bilirubin: The Yellow Hormone? *J. Hepatol.* **2021**, *75*, 1485–1490. [[CrossRef](#)]
49. Svensson, M.B.; Ekblom, B.; Cotgreave, I.A.; Norman, B.; Sjöberg, B.; Ekblom, Ö.; Sjödin, B.; Sjödin, A. Adaptive Stress Response of Glutathione and Uric Acid Metabolism in Man Following Controlled Exercise and Diet. *Acta Physiol. Scand.* **2002**, *176*, 43–56. [[CrossRef](#)]
50. Wang, Y.; Jia, Y.; Xu, Y.; Liu, X.; Wang, Z.; Liu, Y.; Li, B.; Liu, J. Exploring the Association between Glutathione Metabolism and Ferroptosis in Osteoblasts with Disuse Osteoporosis and the Key Genes Connecting Them. *Comput. Math. Methods Med.* **2022**, *2022*, 4914727. [[CrossRef](#)]

Disclaimer/Publisher’s Note: The statements, opinions and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions or products referred to in the content.