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MULTI-OMIC ANALYSIS AND PLATELET FUNCTION DISTINGUISH View Article Online
DOI: 10.1039/D5FO00874C**TREATMENT RESPONSES TO HYDROXYTYROSOL IN
CARDIOVASCULAR RISK**

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Abstract

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Many phenolic compounds, such as hydroxytyrosol (HT), have been recognized for their antioxidant and cardiovascular (CV) health benefits. To address the efficacy of HT, the present study aimed to identify the relevant mechanisms associated with high-CV risk. Plasma unbiased multi-omics data were compared among two subgroups of high-CV risk patients -HT responders and non-responders-. The suppression of platelet reactivity and agonist-induced platelet activation observed after HT intervention were measured by CD61/CD62P expression and apoptotic microparticles with flow cytometry. Microbiota analysis revealed that HT treatment significantly increased and decreased the abundance of *Ruminiclostridium sp.* and *Desulfovibrio sp.*, respectively. Lipid metabolism and proteomic responses were heterogeneous within the two distinct subgroups, associated mainly with thrombotic and hemostatic signals. The metabolomic analysis further confirmed the differentially expressed metabolites within these two subgroups, highlighting improved glutathione metabolism after HT treatment. The omics datasets integration offers a more comprehensive representation of the glutathione and coagulation pathways affected by the HT treatment. Even though the translation of multi-omics technologies into the clinical landscape is slow, the current interventional study provides an overview of their utility in developing novel therapeutic opportunities to prevent CV diseases, such as suppression of platelet reactivity.

Registration number of Clinical Trial: NCT06319417

Keywords: Hydroxytyrosol, Cardiovascular, Proteomics, Metabolomics, Platelet signaling.



Abbreviations

ADP: Adenosine Diphosphate

AnV+ MPs: Annexin V+ Microparticles

β -CDs: β -Cyclodextrins

BMI: Body Mass Index

CV: Cardiovascular

CVDs: Cardiovascular Diseases

EDTA: Ethylenediaminetetraacetic Acid

EFSA: European Food Safety Authority

ELISA: Enzyme-Linked Immunosorbent Assay

EVOO: Extra Virgin Olive Oil

GP: Glycoprotein

HDL: High-density Lipoprotein Cholesterol

HPLC: High-Performance Liquid Chromatography

HT: Hydroxytyrosol

HOMA-IR: Homeostasis Model Assessment Insulin Resistance

Hs-IL-6: High Sensitive Interleukin-6

IGF: Insulin-like Growth Factor

LDL: Low-density Lipoprotein Cholesterol

MS: Mass Spectrometry

ox-LDL: Oxidized LDL

PCoA: Principal Coordinate Analysis

ROS: Reactive Oxygen Species

TG: Triglycerides

UCAM: Catholic University of Murcia



Introduction

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Cardiovascular (CV) diseases (CVD) are the leading cause of death worldwide ¹. The relationship between food intake and CV health has gained tremendous relevance ², with platelets playing a pivotal role in CVD progression. Dietary platelet-based therapeutic innovations have shown the potential to counteract excessive platelet activation ³, which is implicated in thrombus formation, stroke, and coronary syndromes ⁴. The European Commission's policy prioritizes health promotion and prevention of chronic conditions ⁵, encouraging the consumption of non-pharmaceutical compounds with health benefits beyond basic nutrition.

In particular, phenolic compounds derived from plant-based foods are promising owing to their potential role in preventing cardiometabolic diseases ^{6,7}. Hydroxytyrosol (HT) is a polyhydroxy phenol sourced in free, secoiridoid, or conjugated forms ⁸ from extra virgin olive oil (EVOO) with proven anti-inflammatory, anti-proliferative, and antimicrobial properties ^{9,10}. Its ability to scavenge reactive oxygen species (ROS) confers its potential antioxidant activity ^{8, 11}. Nonetheless, the catechol group is also responsible for the relative sensitivity of HT to oxidation and formulation challenges during food storage or processing ¹⁰, which is why few human clinical studies have elucidated the beneficial effects of HT alone. Animal and human studies typically administer extracts or a mixture of bioactive compounds but limit the direct association of HT with positive outcomes and molecular pathways ⁸. Recently, attention has been focused on establishing carrier systems that can improve the physicochemical properties of HT guest molecules ^{12, 13}. Among them, β -cyclodextrins (β -CDs) have been highlighted by the European Food Safety Authority (EFSA) as healthy food additives (E-459) ¹⁴. The encapsulation improves the stability and effectiveness of HT formulations *in vivo* ¹⁵⁻¹⁸, representing a very attractive approach for innovation in the biomedical field ¹⁷.



Several traditional plasma CV biomarkers have been used to demonstrate the health benefits of polyphenols in humans^{19, 20}. However, data on HT alone are scarce²¹. Furthermore, a comprehensive strategy using multi-omics technologies is required to investigate how its oral supplementation modulates complex biological mechanisms^{22, 23}. Hence, applying proteomics and metabolomics approaches together with diagnostic biomarkers in biomedical cohorts could potentially reveal patient-specific responses to treatment²⁴⁻²⁷, thus stratifying populations with a better outcome for a wide range of nutraceuticals^{28, 29}. The translation of multi-omics technologies into tools accessible in daily medical routines is slow. Nonetheless, these studies could lay the foundation for precision nutrition, allowing us to determine the effectiveness of nutraceuticals and other drugs, enhancing clinical outcomes.

This study investigates (i) the functional changes in *ex vivo* platelet activation, (ii) the proteome, (iii) the metabolome, and (iv) the microbiome that accompany clinical benefits in patients at high CV risk after oral HT treatment. It combines classical biomarkers, mass spectrometry (MS), 16S ribosomal RNA amplicon gene sequencing, and flow cytometry. Consequently, this work aims to demonstrate that this integrative strategy could target the therapeutic response following dietary or drug interventions, identifying dynamic pathways (proteins and metabolites) and sensitive biomarkers in CV research.



Methods

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Study design and participants

A 10-week, randomized, two-arm crossover, double-blind, controlled exploratory study was conducted between April 2022 and March 2023 at the Catholic University of Murcia (UCAM, Murcia, Spain). A total of twenty-six individuals identified as high CV-risk patients were enrolled from the *Centro de Salud de San Andrés* (Area 1, Murcia, Spain). Participants were included based on the following inclusion criteria: 20-65 years old, body mass index (BMI): 20.0-35.0 kg/m², basal cholesterol (> 200 mg/dL), and/or basal glucose (> 100 mg/dL). Exclusion criteria encompassed smoking, *diabetes mellitus*, treated hypertension, history of CV events (stroke, myocardial infarction, thromboembolism, or peripheral vascular disease), cancer, and inflammatory diseases, and the use of specific medications such as antihypertensives, antiaggregants, anticoagulants, non-steroidal anti-inflammatory drugs, antidepressants, anticholinergic or antispasmodic agents, medications affecting intestinal motility, vasodilators, lipid-lowering therapies, and fish oil supplements were also excluded. The use of other dietary supplements was assessed individually.

The study protocol was approved by the Ethics Committees of both Area 1 of the *Servicio Murciano de Salud* (2022-3-19-HCUVA) and UCAM (CE042213). Additionally, the trial was registered with Clinical Trials (ID: NCT06319417). All participants provided written informed consent in alignment with the Declaration of Helsinki and Spanish Biomedical Research Law.

Hydroxytyrosol encapsulation for caramel production

Hydroxytyrosol with a purity exceeding 99% was synthesised by Seprox Biotech, SL. (Murcia, Spain). Due to industrial confidentiality agreements, the detailed synthetic



protocol for this compound cannot be disclosed. Encapsulation followed an established protocol with modifications for this study^{30,31}. Briefly, a solution containing HiCap-100 (0.066 g/mL), maltodextrins (0.4 g/mL), and β -CDs (E-459, 0.018 mg/mL) was prepared in 500 mL of ultra-pure distilled water. The HT solution (0.605 g/mL) was added into the primary mixture utilizing a rotor-stator homogenizer (Ultra-Turrax, IKA, Germany) at 500 rpm for 5 minutes. A Mini Spray Dryer B-290 (Buchi, Germany) was used for spray-drying with controlled feed flow rates (2.5 mL/min) and air temperatures (inlet: 180 °C; outlet: 90 \pm 2 °C).

HT content and encapsulation efficiency (EE) were quantified using high-performance liquid chromatography (HPLC, Agilent 1200, Germany) with a diode array detector. A C18 Polaris column (150 x 4.6 mm, 5mm) was maintained at 35 °C. The mobile phase consisted of 0.1% formic acid (A) and methanol (B), applied under a gradient elution program (0.0 min 10% B, 5.0 min 30% B, 17.5 min 100% B, 20 min 100% B, 30.0 min 10% B, 35 min 10% B)³². Before analysis, a calibration curve was constructed using a standard HT sample (Phyproof® \geq 90%; reference substance, Phytolab) through serial 1:2 dilutions from a stock solution (25 mg in 100 mL of the initial solvent mixture). HT was identified at a retention time of 4.846 min and quantified using a calibration curve ranging from 250 to 7.8 mg/L (Supplementary Figure S1).

Drying and encapsulation efficiencies (DE and EE, respectively) were calculated using established formulas:

$$DE \text{ (Drying Efficiency)} = \frac{\text{Powder sample (g)}}{\text{Total solid content in solution (g)}} \times 100$$



$$EE \text{ (Encapsulation Efficiency)} = \frac{\text{Encapsulated compound}}{\text{Total compound added}} \times 100$$

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After this procedure, 20.20% HT powder (batch no. HT SP 211200) was used to dose the caramels. The 20% HT exhibited specific technical characteristics: bulk density of 0.27 g/cm³, tapped density of 0.43 g/cm³, pH 6.25 (10% in water), compressibility index of 58.69%, and Hausner ratio of 1.59. Caramels were produced by B-Organic (Hellín, Spain; <http://borganic.es/>) using xylitol (96.62 and 99.6%, in HT and placebo caramels, respectively) as the carrier, chosen for its low melting point and sweetening properties. Two sugar-free caramel variants (2.5 g each) were manufactured: one containing encapsulated HT (15 mg per caramel) and another serving as a placebo with empty β -CDs. Stability and HT incorporation were confirmed via HPLC batch analysis, with derivations below 9% of target concentrations.

Throughout the study period, caramel samples were examined by pulverizing four units (average mass 2.582 ± 0.061 g; RDS = 2.38%) into a fine powder using a Restch Grindomix GM 200 (Verder Scientifics GmbH&Co. KG) grinding mill for 15 s at 7000 rpm. From the resulting powder, one gram was taken to dissolve in 10 mL of a 10% methanol-water solution using a Multi Reax Vortexer (Heidolph Scientific Products GmbH) for 15 min at 2000 rpm. The dissolved samples were subsequently filtered through a 0.45 mm PTFE disc and diluted 1:10 with the same solvent mixture before undergoing HPLC analysis.

***In vitro* digestion protocol**

The *in vitro* digestion of free HT and β -CD-HT was performed following the protocol described by Minekus *et al.*³³ Briefly, simulated fluids were prepared to a final volume of 500 mL. The composition of each fluid was as follows: Simulated Salivary Fluida (SSF) contained 15.1 mL de KCl (37,3 g/L), 3.7 mL de KH₂PO₄ (68 g/L), 6.8 mL de



NaHCO₃ (84 g/L), 0.5 mL de MgCl₂(H₂O)₆ (30.5 g/L), 0.06 mL de (NH₄)₂CO₃ (48 g/L) and HCl (6 mol/L) to adjust the pH to 7.0; Simulated Gastric Fluid (SGF) contained 6.9 mL de KCl (37.3 g/L), 0.9 mL de KH₂PO₄ (68 g/L), 12.5 mL de NaHCO₃ (84 g/L), 11.8 mL de NaCl (117 g/L), 0.4 mL de MgCl₂(H₂O)₆ (30.5 g/L), 0.5 mL de (NH₄)₂CO₃ (48 g/L) and HCl (6 mol/L) to adjust the pH to 3.0; and Simulated Intestinal Fluid (SIF) contained 6.8 mL de KCl (37.3 g/L), 0.8 mL de KH₂PO₄ (68 g/L), 85 mL de NaHCO₃ (84 g/L), 38.4 mL de NaCl (117 g/L), 0.33 mL de MgCl₂(H₂O)₆ (30.5 g/L) and HCl (6 mol/L) to adjust the pH to 7.0.

The entire digestion process was carried out at 37 °C, and all samples were prepared in triplicate. The procedure was as follows: 5 mL control (water), 5 mL β-CD-HT (2 g/L of HT), 5 mL β-CD (0.018 mg/mL), and 5 mL HT (2 g/L) were added separately to a mixture containing 3.5 mL of SSF, 0.5 mL of saliva amilase (1500 U/mL SSF dissolved), 25 μL of CaCl₂ (0.3 M) and 975 μL of distilled water. The final 10 mL mixture was incubated for 2 minutes; the 10 mL bolus was mixed with 7.5 mL of SGF, 1.6 mL of porcine pepsin (25000 U/mL SGF dissolved), 0.5 μL of CaCl₂, 0.2 mL of HCl (1 M) and 695 μL of distilled water, reaching a final volume of 20 mL. This mixture was incubated at 37 °C for 2 hours. Finally, the resulting 20 mL gastric chyme was combined with 11 mL of SIF, 5 mL of a pancreatin solution (800 U/mL SFI based), 2.5 mL of fresh bile (160 mM), 40 μL of CaCl₂ (0.3 M), 0.15 mL of NaOH (1 M) and 1.31 mL of distilled water.

At the end of each digestion phase, the samples were collected in triplicate. The samples were centrifuged at 4000 g for 25 minutes at 4 °C. Then, samples were frozen at -80 °C for batch HT quantification by HPLC as described before.

Dietary intervention

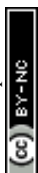


Participants were randomly assigned into two groups, with comparable clinical profiles. The intervention group consumed β -CD-HT caramels for 28 days, whereas the placebo group consumed empty β -CDs caramels. A 2-week washout period was followed by a crossover to the alternate intervention for an additional 28 days. Initially, participants receiving the active treatment switched to the placebo, and *vice versa*.

Four clinical visits were scheduled for each participant, occurring before and after both treatment phases. Blood and feces were collected and body composition was measured using Tanita BC-541 (Illinois, USA), at the four clinical visits in the morning. To monitor dietary habits during the study, participants completed a food frequency questionnaire based on the Mediterranean Diet Serving Score (MDSS). During the intervention period, the patients consumed 60 mg of encapsulated HT daily and were distributed into four caramels throughout the day, apart from meals. The HT dose (60 mg/day) was established due to the safety of the compound even at 500 mg/kg/day in mice.³⁴ Similarly, during the placebo period, the patients received 0.5 g of empty β -CDs daily, distributed into four caramels throughout the day.

Biological samples collection

Fasting peripheral venous blood samples (serum, ethylenediaminetetraacetic acid [EDTA], and sodium citrate) were extracted by a qualified nurse at the four-time point visits. Fresh blood was analyzed for biochemical, flow cytometry, and metabolomic analyses, while centrifuged plasma (15 min at 2800 g) was freeze-dried at -80 °C for proteomic and Enzyme-Linked Immunosorbent Assay (ELISA) assessments. Fecal samples were also collected from the patients in sterile duchesses and 15-80 g per sample (quantity and consistency variability), then they were frozen at -80 °C and preserved for microbiota analysis.



Biochemical determinations, ELISA, and Flow Cytometry

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Serum biochemical parameters such as lipid profile (total cholesterol, high-density lipoprotein cholesterol [HDL], low-density lipoprotein cholesterol [LDL], and triglycerides [TG]), glucose, creatinine, and insulin levels were kindly provided by the Laboratory for Clinical Analyses of the *Hospital Universitario Virgen de la Arrixaca* (Murcia, Spain). Homeostasis model assessment of insulin resistance (HOMA-IR) was also calculated. Cardiometabolic markers (leptin and adiponectin), high-sensitivity interleukin-6 (hs-IL-6), and oxidized LDL (oxLDL) were analyzed using ELISA from plasma EDTA samples. All measurements were performed using high-quality kits: hs-IL-6 antigen (1/2 dilution; Invitrogen, Cat. No. BMS213HS), oxLDL (1/100 dilution; ABclonal, Cat. No. RK09047) and adiponectin antigen (1/5000 dilution; ABclonal, Cat. No. RK00060), and leptin antigen (1/100 dilution; ABclonal, Cat. No. RK00035). The lower detection limits for hs-IL-6, oxLDL, adiponectin, and leptin were 2.5 pg/mL, 0.12 ng/mL, 30.2 pg/mL and 15.6 pg/mL, respectively.

Citrated blood (450 μ L), freshly obtained, underwent incubation with 0.02 mmol/L adenosine diphosphate (ADP) for 2 min. The expression of membrane-constitutive CD42a (GP IX) and platelet activation and aggregation markers (P-selectin and CD61, respectively) on the cell surface was evaluated using previously described methods³⁵. Additionally, Annexin V+ (AnV+) microparticles (MPs) were also examined using 0.5 and 1 μ m gating beads³⁶. Following ADP incubation, 5 μ L of blood was stained with 0.05 μ g anti-CD42a-BV395 (clone HIP1), anti-CD62P-BV786 (clone AC1.2), CD61-FITC (clone VI-PL2) (which recognizes the activated conformation of GPIIb/IIIa), and AnV-PE (all from BD Biosciences, Oxford, UK) for 15 min in AnV 5X buffer. The sample was then diluted with 1 mL AnV 1X buffer and analyzed using a FACS Celesta



flow cytometer (Becton Dickinson, Oxford, UK). Each sample was run in duplicate, with twenty thousand platelets counted per run.

Microbiota analysis

Fecal samples underwent total genomic DNA extraction using a modified repeated bead-beating technique, based on a protocol previously outlined by the University of Amsterdam (The Netherlands)³⁷. The process involved double extraction in S.T.A.R. lysis buffer (Roche, Switzerland), followed by centrifugation. The obtained supernatants were pooled and subsequently purified using the Maxwell RSC Blood DNA kit (Promega, Madison, USA)³⁷. Following this, 16S genes were amplified using a modified one-step PCR method, with the amplification products verified by agarose gel electrophoresis, as detailed in previous studies^{38, 39}. The final steps included purification and equimolar pooling of samples before sequencing, which was performed using Illumina Miseq (V3,600) technology (Illumina, Inc., San Diego, USA)^{38, 39}.

A vsearch (v2.15.2) pipeline was employed to process amplicon sequences⁴⁰. Merging of paired-end reads was performed, allowing for a maximum of 100 differences and accommodating staggered overlap. ASVs were identified from reads with an error rate lower than 1.5 using the cluster_unoise with centroids algorithm, requiring a minimum size of four. Subsequently, chimeras were eliminated using the uchime3 *de novo* method. ASV abundance in each sample was determined by mapping merged reads against ASV sequences using usearch_global with a 0.97 distance cut-off. Taxonomic assignment was conducted using R (4.2.0), and the dada2^{41, 42} assigned taxonomy functions, utilizing the Silva (v132)⁴³ reference database. A phylogenetic tree was constructed using mafft (v7.310)⁴⁴ and FastTree (2.1.11)⁴⁵.



The sample data underwent filtering based on size (greater or lower than the sample mean) and ASV count (fewer reads). Subsequently, samples with a read count significantly deviating from the mean (16380 reads) were excluded. ASV with less than four reads were eliminated, and only those present in at least 20% of the study samples were retained. This filtering process resulted in the removal of 1495 ASV due to low abundance, leaving 569 ASV for analysis. The results were then normalized using Cumulative Sum Scaling (CSS)⁴⁶. R v.3.6.2 and various R packages, including Phyloseq v.1.30.0., were employed for downstream analyses and graphical outputs. Phyloseq facilitated normalization, visualization of microbial taxonomic composition abundance, and estimation of alpha and beta diversity metrics. Statistical significance was established at $P < 0.05$. Bray-Curtis dissimilarity distance was applied for assessing beta diversity, with PERMANOVA used to determine statistically significant differences between treatments and groups. Principal Coordinate Analysis (PCoA) was utilized for data representation. In addition to PERMANOVA, ANOSIM (analysis of group similarities, similar to PERMANOVA), and PERMDISP (analysis of the homogeneity of group dispersions) were also calculated across treatments.

Proteomic analysis

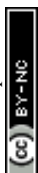
The Medical Oncology Laboratory Amsterdam UMC (The Netherlands) conducted proteomic analyses. A modified method based on published literature⁴⁷ was used for high-pH fractionation of pooled peptides from plasma EDTA samples collected at the start and end of treatment. Fractionation tips were prepared by loading ReproSil-Pur C18-AQ beads (2 mg, 5 μ m, Dr. Maisch, Germany) on two C18 membrane disks (3M Empore, Bracknell, UK) within a 200 μ L pipette tip. The tips underwent conditioning with methanol and 60% acetonitrile in aqueous NH_4FA (40 mM, pH 10), followed by NH_4FA (40 mM, pH 10). Peptide pools, dissolved in NH_4FA (40 mM, pH 10), were loaded onto



the fractionated tips and eluted in 12 steps using elution buffers with increasing acetonitrile concentrations (2, 4, 6, 8, 10, 12.5, 15, 17.5, 20, 25, 30%, and 60%) in aq. NH_4FA (40 mM, pH 10). Every sixth elution step was combined and concentrated to dryness, resulting in six fractions per pool.

The peptides underwent solubilization in 0.1% FA and were loaded onto Evotips (Evosep, Denmark) following the manufacturer's instructions. An Evosep One liquid chromatography system (Evosep, Denmark) was utilized to separate the peptides through nanoflow reversed-phase liquid chromatography using standardized gradients. The mobile phases consisted of 0.1% FA and a mixture of 0.1% FA/99.9% ACN. The analysis utilized a method designed for processing 30 samples per day (SPD), employing a C18 reverse-phase column (Bruker Daltonics) and connected to a silica emitter (Bruker Daltonics). Peptide analysis was performed on a TimsTOF HT instrument (Bruker Daltonics), with peptides introduced via nano-electrospray ion source (Captive spray source, Bruker Daltonics) operated at a spray voltage of 1400 V.

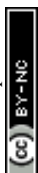
For the creation of the spectral library, peptides fractionated at high pH underwent analysis using TimsTOF HT, operating in DDA-PASEF mode. The system was configured with a 200 ms ramp time and acquired ten PASEF scans per topN acquisition cycle, yielding a 2.2s cycle time. The analysis targeted precursors within a mass range from 100-1700 m/z, an ion mobility range from 1.5-0.7 Vs cm^{-2} , and charge states from 0 (unassigned) to 5+. Precursors that either achieved this target or reached full scheduling capacity were excluded from analysis for 0.4 min. Singly charged precursors underwent filtering based on their m/z ion mobility. A quadrupole isolation width of 2 Th was applied for precursors below 700 Da, while those above 700 Da were isolated with a 3 Th width. Ion mobility dimension was calibrated among the experiments using distinctly selected ions from the Agilent ESI LC/MS Tuning Mix.



The TimsTOF HT (Bruker Daltonics) instrument, operating in the DIA-PASEF mode, examined peptides in plasma specimens. The method encompassed an ion mobility range of 1.5 to 0.7 Vs cm⁻² and an m/z range of 300 to 1200, using 10 DIA-PASEF scans with dual isolation windows per scan, yielding a cycle time of 1.1 s. The py_diAID tool⁴⁸ optimized DIA-PASEF windows for the specific sample type. Collision energy decreased from 59 eV at 1.6 Vs cm⁻² to 20 eV at 0.6 Vs cm⁻². The ion mobility dimension was calibrated among the experiments using three chosen ions from the Agilent ESI LC/MS Tuning Mix [m/z, 1/K0: (322.0481, 0)].

FragPipe v20.0 was employed to generate a spectral library from the DDA-PASEF data, utilizing the present DIA_SpecLib_Quant⁴⁹. A database was created by incorporating reverse entries and contaminants into the curated human proteome (Uniprot, March 2023, 42420 entries, including isoforms). The DDA-PASEF data was analyzed using MSFragger^{50, 51} (version 3.8) with default parameters. During the validation phase, MSBooster (version 1.1.11) was applied equally at spectra and RT levels, followed by Percolator⁵² and ProteinProphet⁵³ integrated within Philosopher⁵⁴ (version 5.0.0), for PSM validation and protein inference, respectively. The library was created using EasyPQP (version 0.1.37) and it was calibrated by RT (ciRTs based). The lowest fraction set to 0.01.7318 Vs cm⁻²; 622.0289, 0.9848 Vs cm⁻²; 922.0097, 1.1895 Vs cm⁻².

DIA data analysis was conducted using DIA-NN 1.8.1, employing a spectral library generated using FragPipe^{55, 56}. Several parameters were adjusted from their default settings: charge state was limited from 2 to 4, precursor m/z range was determined by 300 to 1200, and both mass and MS1 accuracy were configured to 14 ppm. The “Match between runs” feature was activated, while “Heuristic protein inference” was turned off. Library precursors underwent re-annotated using the reviewed human proteome (Uniprot, March 2023; 42420 entries, including isoforms). The output tables were filtered to ensure



a maximum q-value of 1% at both precursor and global protein levels. The MaxLFQ Open Access Article Online
DOI: 10.1039/D5FO00874C algorithm, implemented in R package iq, was utilized for protein quantification ⁵⁷.

Metabolomic analysis

The Laboratory of Genetic Metabolic Diseases at Amsterdam UCM (The Netherlands) conducted targeted metabolomics analysis using a previously established method ⁵⁸. Briefly, metabolic activity in a 10 μ L blood sample was halted by introducing 500 μ L of ice-cold methanol. Subsequently, a mixture of internal standards dissolved in 500 μ L of Milli-Q water was added according to the established method ⁵⁸.

The samples underwent sonication for 10 min at 40 kHz, after which 1 mL chloroform was added and thoroughly mixed. Subsequently, the samples were centrifuged (14000 g; 10 min; 4 °C), and the uppermost layer containing polar metabolites was extracted and transferred to new Eppendorf tubes. Samples were dried using a vacuum concentrator (90 °C), and then the pellets were resuspended in 100 μ L of 3:2 (Methanol:Milli-Q water). Another round of centrifugation was performed at equal conditions, collecting the resulting supernatant in vials for MS analysis. A Waters Acquity ultra-HPLC system linked to a Bruker Impact II™ Ultra-High Resolution Qq-Time-Of-Flight Mass Spectrometer was employed for metabolite analysis. The 5 μ L samples were maintained at 12 °C throughout the analysis and injected.

Merck Millipore SeQuant ZIC-cHILIC column (PEEK 100 x 2.1 mm, 3 μ m particle size) was utilized for chromatographic division. The column temperature did not vary from 30 °C. The mobile phase components were (A) acetonitrile:water in a 1:9 ratio (v/v) and (B) acetonitrile:water in a 9:1 ratio (v/v), both containing 5 mmol/L ammonium acetate. The LC gradient (0.25 mL/min flow rate) proceeded as described in the previously published



methodology [42]. Negative and positive ionization was used in MS data collection (50–1200 m/z covering). Bruker TASQ software version 2.1.22.3 was utilized for data analyses, with further normalization of the metabolites discovered.

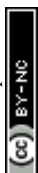
Omics integration and Statistical Analysis

The mixOmics R package (version 6.24.0) employs the DIABLO (Data Integration Analysis for Biomarker Discovery using Latent Orthogonal Components) approach for performing multivariate analysis (Stacked Partial Least-Squares Discriminant Analysis (SPLSDA) in biological datasets⁶⁰. Microbiome and flow cytometry parameters were excluded because of sample dataset limitations. Firstly, an initial model was created with two components, with the most discriminative features for each omics dataset (ten variables in each component for proteome and metabolome, and three variables for each component for biomarkers).

Continuous parameters are presented as mean \pm standard deviation and confidence interval [CI 95%]. The normality of the distribution of continuous variables was determined using the Kolmogorov–Smirnov test⁵⁹. The present analyses focused on estimating the effects of HT and the interventional period (crossover design) on outcome variables (glucose, total cholesterol, TG, LDL, HDL, oxLDL, leptin, adiponectin, and hs-IL6). The level of statistical significance was set at $P < 0.05$. To this end, differences were evaluated using a random linear regression model to control for individual variations, utilizing R v4.1.3 software and the lme4 package. For family, genus, and species, statistical significance was determined based on a P -value < 0.05 , and a $> \pm 1$ Fold Change (Log₂ Gold Change) value.

Results

Hydroxytyrosol encapsulation protects stability during caramelization



To produce caramels with a high content of pure HT, and protect the integrity of the phenolic compound from the mid-high temperatures reached during the crystallization of xylitol, HT was encapsulated in a mixture of β -CDs and maltodextrins to a fine powder by spray drying from a concentrated solution of the main components (Supplementary Figure S1). The process was adapted and satisfactorily optimized (DE=73.4 \pm 1.4%; RDS=1.9%, EE=97.5 \pm 0.4%, RDS=0.4%).

The HT composition used in this study was analysed after caramelization in order to test the chemical stability of the compound. Three HT caramel samples (mean weight 1.0039 \pm 0.0023 g; RDS=0.23%) were characterized under standard HPLC working conditions, rendering 15.002 \pm 0.149 mg HT (RDS=1%) per caramel. Furthermore, the stability of encapsulated HT in caramels was retained after the completion of the exploratory study (Supplementary Figure S2). No HT signal was detected in placebo caramel samples within the HT retention time (Supplementary Figure S2).

***In vitro* digestion stability of encapsulated hydroxytyrosol**

The *in vitro* digestion analysis revealed that although the encapsulation process safeguards HT from caramelization, less than 20% of the compound was found in the final intestinal phase, with no significant differences observed between free HT (10.3% \pm 0.7) and HT- β -CD (9.9% \pm 1.3) conditions ($P>0.05$). As expected, neither the control or the β -CD samples exhibited any detectable HT levels (Supplementary Figure S3).

The hydroxytyrosol treatment improves lipid metabolism and suppresses platelet activation

The detailed anthropometric and biochemical parameters of the entire cohort are presented in Table 1. The 26 enrolled patients were randomly allocated, and no significant

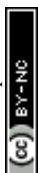


differences ($P>0.05$) were found between the two groups (starting HT or placebo). Total cholesterol and LDL levels were above 225.2 and 135.8 mg/dL, respectively, reflecting the CV risk of the population. Importantly, throughout the 2.5-month study period, there were no discernible changes in dietary habits compared to baseline conditions, as observed in the MDSS questionnaire (data not shown). No adverse effects were found after the study.

The intervention period significantly improved the lipid profile (Figure 1 and Supplementary Table S1). LDL levels were notably reduced ($P=0.025$) and better in the oxidized state ($P=0.0007$). Additionally, TG levels were significantly decreased ($P=0.002$) following HT treatment. Moreover, the hormonal balance was enhanced, as evidenced by an increase in adiponectin levels ($p=0.026$) after HT treatment (Figure 1).

Under unstimulated conditions, decreased platelet reactivity was detected after regular HT intake, as indicated by the higher CD61 expression ($P<0.0001$, Figure 2D, and Supplementary Table S1). This lower platelet aggregation was abolished by ADP stimulation (before or after HT). Moreover, platelets increased P-selectin (CD62P) expression after ADP stimulation (Figure 2D, $P<0.0001$), but this effect was also reversed by HT treatment (Figure 2D, Table S1, $P<0.0001$), suggesting lower platelet reactivity in the presence of HT. Finally, apoptotic MPs, as quantified by AnV⁺ expression in MPs, increased in response to ADP before HT treatment (Figure 2D, $P=0.023$); however, AnV⁺ expression in MPs after ADP stimulation was similar to that in unstimulated baseline conditions ($P=0.713$, Table S1 and Figure 2D).

To categorize patients according to their response to HT treatment, CV biomarkers were used to separate patients with major and minimal clinical evidence after the intervention (Supplementary Table S2). The categorization, based on the effect of HT treatment on statistically significant biochemical parameters such as LDL ($P=0.004$), oxLDL

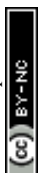


($P=0.004$, before HT), adiponectin ($P=0.026$), and TG ($P=0.046$), identified "responders" patients with major clinical improvement and "non-responders" patients with little clinical evidence after HT treatment (Table S2). This approach allowed us to categorize 15 (58%) patients as responders, and 11 (42%) as non-responders to HT treatment.

Microbiome analysis shows differential microbial genus and species after the hydroxytyrosol treatment

The relative abundances of different bacterial taxa between treatments were compared at the family, genus, and species levels, using stacked bar plots as shown in Supplementary Figure S4. At the genus level, after HT treatment patients showed a lower abundance of bacteria from the genus *Agathobacter* than those who received a placebo. A similar effect can be found in other bacterial genera such as *Faecalibacterium* and *Bifidobacterium*. However, the opposite effect was observed with the genus *Collinsella*, which showed a higher relative abundance in patients after HT treatment compared with placebo. The genus *Blautia* decreased after HT treatment, and *Subdoligranulum* or *Anaerostipes* increased after the same treatment.

Three indices were used to evaluate alpha diversity: community richness (Chao1) and uniformity of community diversity (Shannon and Simpson indices). Statistical analyses revealed no significant differences following HT treatment or placebo (Figure 3). Next, we compared the impact of each intervention period on the microbial communities. Beta diversity was evaluated by the Bray–Curtis metric, conducting a permutational multivariate analysis of variance (PERMANOVA) to determine whether there was a significant separation of samples by treatment or time point due to differences in microbiome structure. PERMDISP assesses whether the distribution or spread of the two sample groups is significantly different and is conducted in parallel with the PERMANOVA. Results from these analyses showed that although most treatment groups



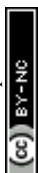
at 28 days were significantly separated from each other, HT and placebo treatments did not exhibit a significantly different community structure from the basal conditions (Supplementary Figure S4).

Subsequently, genus and species were compared at distinct time points (Figure 4). As shown in Figure 4A, the three genera displayed differential abundances following HT treatment. *Ruminiclostridium sp.* and *Holdemanella sp.* exhibited increased abundance, whereas *Desulfovibrio sp.* decreased in abundance after HT treatment (all $P < 0.05$). Comparison between HT treatment and the placebo period (Figure 4A) also revealed changes in the genus; *Catenibacterium sp.* increased after HT treatment, whereas *Holdemanella sp.* increased in abundance after the placebo period.

Notably, *Bacteroides caccae* and *Bifidobacterium kashiwanohense* exhibited increased and decreased abundances, respectively, after HT treatment (Figure 4B). Comparing the HT and placebo periods, we found that HT treatment decreased the abundance of *Bifidobacterium kashiwanohense* (Figure 4B). No changes in lower taxonomic ranks, such as family, genus, or species, were found after HT treatment in the responder and non-responder groups (data not shown).

Proteomics profiling monitors the clinical improvement after hydroxytyrosol treatment

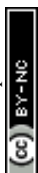
In the present study, we conducted a detailed proteomic analysis of plasma samples to gain a deeper understanding of the biological effects of HT. The data processing method described herein allowed us to investigate 673 proteins. The HT intervention led to statistically significant changes in only 19 proteins. Among these, HT treatment resulted in the overexpression of 15 proteins, whereas four proteins were under-expressed (Table 2). To identify alterations in the biological functions of differentially expressed proteins



(DEPs), biological enrichment analyses were performed using Metascape (Supplementary Figure S5). The most enriched proteins in the entire cohort (n=22) were related to three main protein clusters participating in platelet degranulation [Log(q-value) = -2.275], negative regulation of protein binding [Log(q-value) = -1.412], and regulation of Insulin-like Growth Factor (IGF) [Log(q-value) = -1.023]. Some of the identified proteins related to these pathways were IBP6 (P=0.009), IBP4 (P=0.037), and ECM1 (P=0.015) (Table 2).

To further validate the specific biological responses of the categorized patient-responder and non-responder subgroups (Supplementary Table S2), their proteomic profiles were compared at two-time points (before and after HT treatment). Among the 24 DEPs identified under basal conditions (before HT), the non-responder and responder subgroups indicated a distinct initial status in coagulation and inflammatory pathways within the two subgroups (Figure 5A). However, HT treatment produced an increase in the number of DEPs (from 24 to 139), with 113 exhibiting overexpression in the non-responder group, compared to 26 in the responder subgroup (Figure 5B). The proteins that differentiate both subgroups (before and after HT treatment) were involved in similar clusters and signaling cascades. Interestingly, HT treatment strongly affected the complement and coagulation cascades [Log(q-value) = -34.055], platelet degranulation [Log(q-value) = -20.916], processes regulated by IGF [Log(q-value) = -16.452], and proteins associated with extracellular matrix remodelling and wounding [Log(q-value) = -20.341]. Some of the identified proteins related to these pathways were FA5 (P=0.002), IL-6RB (P=0.001), VCAM-1 (P=0.002), APOB (P=0.006), and TSP4 (P=0.013) (data not shown).

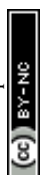
Upon further investigation, 44 proteins exhibited significant differences in the responder subgroup after HT treatment, with 13 and 31 overexpressed and underexpressed proteins,



respectively. Supplementary Figure S6 shows that changes in the proteome of patients with clinical improvement included hemostasis (Log(q-value) = -11.549), blood coagulation (Log(q-value) = -11.190), and cell-cell adhesion processes (Log(q-value) = -6.473). Additionally, regulators of the NABA ECM pathway also exhibited notable changes (Log(q-value) = -5.240). Some of the identified proteins related to these pathways were MRC-1 (P=0.028), VCAM-1 (P=0.041), and GP1BA (P=0.047) (data not shown).

Targeted metabolomics show changes in the amino acid and lipid metabolism following hydroxytyrosol treatment

Targeted analysis was used to identify the metabolites associated with amino acids, oxidative status, and lipid metabolism. The whole blood metabolome analysis after HT treatment was carried out by comparing metabolite changes between the responder and non-responder subgroups, which were defined based on their clinical improvement (Supplementary Table S2). To determine whether the HT effect was associated with differences in whole blood metabolite patterns a PLS-DA model was performed in the two categorized subgroups (Supplementary Figure S7). The responder subgroup showed a significant increase and decrease in hippuric acid and CDP-ethanolamine levels, respectively, before HT treatment (Figure 6 and Supplementary Figure S7A). In contrast, kynurine and xanthosine were more abundant in the responder subgroup after HT treatment (Figure 6 and Supplementary Figure S7B). Additionally, 2-hydroxybutyric acid and hypoxanthine levels were lower in the responder subgroup than in the non-responder subgroup (Supplementary Figure S7B). Notably, we observed an evident separation pattern within the two subgroups, as shown in Figure 6C and the score plots (Supplementary Figure S7B), indicating a more distinct metabolic signature after HT treatment.

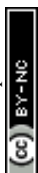


The comparison in the two categorized subgroups revealed an association with changes in glutathione, nucleotide, and amino acid metabolism (Figure 6). Especially, in the responder subgroup, HT intake induced relative increases in homoglutathione (P=0.0032), glutathione (P=0.027), and adenine (P=0.017) (Figure 6B). Conversely, uridine (P=0.010), 2-hydroxybutyric acid (P=0.019), xanthine (P=0.026), 1,3-diphosphoglyceric acid (P=0.027), and pyroglutamic acid (P=0.035) levels decreased after HT treatment (Figure 6B), suggesting the metabolic basis for the observed clinical improvement (Supplementary Figure S8).

Proteomic and metabolomic data correlated in the responder subgroup with glutathione and amino acids metabolism

Correlation analyses were performed to test possible associations between significant proteins, metabolites, and CV biomarkers (adiponectin, LDL, oxLDL, TG) in the responder subgroup (Table 3). Pearson's correlation analysis after the HT treatment identified 23 proteins with strong associations (P<0.05). Specifically, LDL levels were positively associated with APOF (r=0.82, P=0.006), APOL1 (r=0.7, P=0.037), and thrombospondin (r=0.76, P=0.018). Adiponectin was also positively correlated with angiotensinogen (r=0.87, P=0.002), serpin1 (r=0.72, P=0.03) and serpins (r=0.67, P=0.049). Proteins related to lipid metabolism (APOF and APOL1), regulation of blood coagulation (THBS1, CFHR, F13B, F13A1, HABP2, TFPI, F5, F7 and F12), inflammation (CRP, VCAM-1, CPN1, F2, and CSF1R), and proteolysis (SERPINS, MMP2, HABP2, and CPN) were correlated to these classical biomarkers of the CV status.

To further dissect the molecular dynamics after HT treatment, correlations between metabolites were determined in the responder subgroup (Table 4). In general, most metabolites and diagnostic biomarkers that correlate are related to glutathione (glutathione, glutamate, and NADH), amino acids (ornithine, pyruvate, arginine,



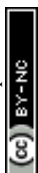
aspartame, glutamate, leucine, carnitine, and NADH), and the urea cycle (Ornithine, NADH, and glutamate) (Supplementary Figure S9). Platelet parameters strongly correlated with *de novo* triacylglycerol and cardiolipin biosynthesis (carnitine, glycerol-3P, glycerophosphoinositol, and phosphorylethanolamine).

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Multivariate analyses with the mixOmics package

MixOmics" R package was applied for omics feature selection and multiple data integration in responders vs non-responders subgroups after HT treatment. Firstly, variables with a high expression in responders were adiponectin, kynurenine, homoglutatione, creatinine, homocysteine and 1433Z_Human (Proteome). Variables with a high expression in non-responders are triglycerides, xanthine, uridine (among others), and C1R (complement system C1), A1BG (Alpha-1-B Glycoprotein). To effectively visualize the global relationships among the metabolome, proteome, and clinical biomarkers, a Circos plot (Figure 7A) was performed. Figure 7A highlights that the first components from each dataset are highly correlated to each other ($r > 0.9$). Omics integration indicates a decrease in proteins such as APOA1, APOC1, KNG1, and FA9, in the responder subgroup. These proteins are associated with the lipidomic profile and coagulation state (A1BG, C1R, and CO6). Conversely, metabolites such as homoglutatione, pyroglutamic acid, adenine, homocysteine, and kynurenine, among others, were elevated in the responder subgroup after HT treatment.

Besides, the clustered image map (cimDIABLO) based on the features selected on the first component displayed a good classification within the two phenotypic subgroups (Figure 7B).



Discussion

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One major obstacle to the translation of multi-omics technologies to routine clinical applications is its uneven maturity. As the prevalence of CV risk factors is reaching epidemic rates and healthy dietary habits are still a pending subject worldwide. This study evaluates the therapeutic landscape targeted by these integrated approaches and identifies the dynamic changes (platelet reactivity, microbiome, proteins, and metabolites) directly affected by HT in the CV context. Moreover, the analytical chromatographic assay showed that β -CDs were efficient protective carriers during the crystallization process, which has not been previously reported in the literature for interventional human diet studies. However, as shown in this study, encapsulation made no difference in terms of HT stability during *in vitro* digestion. Consistent with our findings, β -CDs do not alter HT bioavailability either ⁶¹ due to their propensity to dissolve upon contact with aqueous solutions (oral, gastric, and intestinal fluids).

The results of this clinical study demonstrated an improvement in lipid biomarkers (LDL, TG, oxLDL), adiponectin, and platelet reactivity following 60 mg/day HT. The HT dosage range used in this study aligns with the doses reported in a recent meta-analysis on olive-derived polyphenols ⁶². It is important to note that participants did not modify their dietary habits during the study period; therefore, the observed changes can be attributed solely to the nutritional intervention. Elevated LDL, TG, and ox-LDL levels play a major role in the pathogenesis of CV diseases such as atherosclerosis ⁶³⁻⁶⁵. In agreement with these data, previous human studies have shown the positive effects of HT on LDL and TG levels. However, these trials involved healthy volunteers ⁶⁶ and used a lower amount of HT combined with other compounds, which cannot grant the specific effects of this monophenol alone ⁶⁷. While reductions in ox-LDL levels have been previously reported, other lipid parameters remained unaffected in previous reports ⁶⁸.



Verhoeven et al. reported similar results attributed to a complex supplement containing monacolins in addition to HT ⁶⁹. In parallel with these findings, adiponectin levels were also increased after HT treatment in the current study. Adiponectin is the most abundant adipokine secreted by adipocytes, and its levels are typically reduced in individuals with CVD ^{70, 71}. An *in vitro* study demonstrated that HT increased adiponectin levels and speculated that this effect mitigated the inflammatory state of adipocytes ⁷². Clinical studies have also attempted to investigate the *in vivo* effects of HT; however, confounding factors, such as a caloric-deficit diet, oral medications, and a combination of other bioactive compounds mask the real effects on adiponectin levels ^{73, 74}. Conversely, the current study, characterized by a higher age range, CV risk, and elevated HT concentration intake, consistently demonstrated the ability of HT alone to enhance adiponectin levels ^{73, 74}.

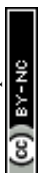
Upon platelet activation, the degranulation leads to upregulation of CD62P (P-selectin) on the platelet surface. CD61, also known as glycoprotein IIIa (GPIIIa), is a key protein involved in platelet aggregation. In the current study, platelets activate and aggregate in the presence of ADP, leading to a higher and lower mean fluorescence intensity of CD62P and CD61, respectively. However, after the HT supplementation, an increase in CD61 expression together with a decrease in cellular adhesion molecules (CD62P), and AnV expression in MPs confirmed a decrease of platelet reactivity, which is generally associated with a lower risk of thrombotic events in various clinical settings ⁷⁵. The potential effects of phytochemicals have been previously noticed in animal studies, in which HT administration led to a decrease in thromboxane 2 (TxB2) levels across a dosage range of 5-100 mg/kg/day ^{76, 77}. Some human blood studies have used HT extracts (not HT alone) or elevated HT concentrations to mitigate platelet aggregation as well ^{78, 79}. Notably, a study involving five patients with type I diabetes supplemented with an HT



concentrate extract (comprising 53% HT and 13% tyrosol) demonstrated a decrease in TxB2 levels ⁸⁰, underscoring the potential efficacy of our exploratory study in which patients were supplemented with HT alone. Overall, our study provides convincing evidence regarding the potential of HT in managing lipid levels, adipocyte inflammation, and platelet activation profiles associated with CV risk. However, the mechanisms by which HT coordinates such mechanisms have not been defined. Therefore, combining *in vivo* studies (animal and human) with -omics may be useful for elucidating complex pathways and targets *in vivo*.

One possible way to modulate the gut microbiota and CVD is through dietary interventions. Previous studies have linked HT to probiotic effects and reduction in certain bacteria associated with inflammatory diseases ⁸¹. After HT, patients showed significant changes in three gut microbial genera with increased levels of *Ruminiclostridium sp.* and *Holdemanella sp.*, both of which are associated with beneficial effects on blood pressure ⁸² and glucose tolerance ⁸³, respectively. However, these results suggest that gut microbiome alterations after HT were not related to the patient clinical benefit -responder and non-responder subgroups-, perhaps, because of the low number of patients included in the microbiome study (n=15 and n=11, respectively).

Although most studies have focused on the clinical benefits of HT, the molecular pathways involved are also of paramount importance. The current study identified 673 proteins, of which 19 showed significant differential expression after the intervention period. Notably, IBP6 and IBP4 were up-regulated, whereas ECM1 was down-regulated. Previous studies have confirmed low levels of IBP6 in patients with unstable plaques ⁸⁴, indirectly indicating the beneficial effect of HT treatment. Additionally, elevated ECM1 levels were observed during the inflammatory phase of myocardial infarction ⁸⁵, further suggesting a protective role of HT in our interventional study. While some animal studies



have determined the proteome changes in adipose tissue, liver, aorta, and heart following HT supplementation, no human studies have evaluated the proteome changes resulting from this polyhydroxylphenol ^{86, 87}.

Significant biochemical markers were used to ensure a clear distinction between individuals exhibiting distinct clinical responses to HT, enabling a more focused examination of proteomic alterations associated with CV risk. This careful selection process was aimed at facilitating a deeper investigation of the molecular mechanisms underlying HT treatment. Thus, 44 proteins were identified in the responder subgroup after HT supplementation. Of these, 13 were over-expressed and 31 were under-expressed. Most of these proteins are involved in homeostasis and coagulation. Notably, MRC-1, VCAM-1, and GP1BA levels decreased. Investigators have also found that HT can reduce VCAM-1 levels *in vitro* and animal studies ^{88, 89}, which is in agreement with current results in humans. One of the more significant findings of this study is that proteomics confirmed higher changes in coagulation and platelet-related pathways in the subgroup benefiting from HT.

Metabolomics has received increasing attention in recent years⁹⁰. When comparing the responder and non-responder subgroups, pathways involving nucleotide and amino acid metabolism were found to be altered, which has a potential role in overall metabolism ⁹¹. The metabolites showing increased levels after HT treatment were kynurenine, glutathione, adenine, homogluthathione, and creatinine. A significant body of data suggests that increased glutathione is associated with the CV prevention ⁹², whilst a increase of 2-hydroxybutyric acid can predict the development of insulin resistance ⁹³. These findings may be related to the clinical benefits of HT supplementation. This approach identified platelet functional parameters that were positively correlated with *de novo* triacylglycerol and negatively correlated with cardiolipin biosynthesis. Currently,

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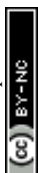


we can only speculate why this is the case, but an increase in cardiolipin levels is associated with higher platelet reactivity through platelet-activating factor receptors, and clinically with a high risk of spontaneous thrombosis ⁹⁴.

In addition, the combination of omics data offers possibilities that exceed those of the single biomarkers ^{95, 96}. Based on previous studies ^{97, 98}, we stratified our cohort into responders and non-responders. In responders, classical biomarkers were significantly associated with proteomic, metabolomics changes, reflecting beneficial effects on lipid metabolism, blood coagulation, inflammation, and proteolysis. These findings are consistent with other phenolic compounds, which exert cardioprotective effects ^{99, 100}. Integrating omics and classical biomarkers have already proven useful for patient stratification and predicting cardiometabolic risk ¹⁰¹. The multivariate analysis confirmed the two distinct groups, with responders displaying coordinated downregulation of apolipoproteins, particularly APOA1 and increase in homoglutathione and homocysteine, markers of oxidative stress ^{102, 103}. Translating this multi-omics features onto the clinical landscape indicated a more favourable cardiometabolic profile in responders. Thus, these integrative findings support the hypothesis that HT exerts synergistic effects across lipid metabolism, coagulation, and antioxidant pathways.

Finally, this study has several important limitations. First, with a small sample size, caution must be applied as the findings might not apply to the general population. Moreover, to assess long-term benefits, a follow-up study is needed. Investigating different dosages of HT in combination with other therapeutic compounds could also provide valuable insights into optimizing preventive strategies for CV diseases and other pathologies. Overall, this study raises many questions that require further investigation such as the molecular pathways or receptors for HT in platelets.

Conclusions



Mainly, we provide an overview of the utility of multi-omics in CV-nutraceutical development, which might be translated to other clinical landscapes. Our study provides convincing evidence regarding the beneficial effects of HT, showing reductions in traditional blood biomarkers (oxLDL, LDL, and TG) and increased adiponectin. Furthermore, our research offers, for the first time, insights into the antithrombotic properties of HT, such as platelet degranulation, hemostasis, and cell-cell adhesion. Integrative omics indicates the potential modulation of previously unidentified metabolic routes. From a scientific point of view, improving diagnostics and detecting patients that could benefit better from treatment, particularly in patients with multiple comorbidities such as CVD, is of paramount importance, and -omics has the potential to be used in the medical field.

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Clara Noguera-Navarro: methodology, investigation, formal analysis, data curation, validation, writing – original draft, writing–review, and editing. **Kasper T. Vinten:** methodology, investigation, metabolomics. **David Auñón-Calles:** methodology, investigation. **Carmen Carazo-Díaz:** data curation and formal analysis. **Georges E. Janssens:** methodology, investigation, metabolomics. **Silvia Montoro-García:** visualization, validation, resources, methodology, investigation, formal analysis, data curation, conceptualization, funding acquisition, project administration, writing – review and editing, supervision.

Conflicts of Interest: The authors declare that they have no known competing financial interests or personal relationships that could influence the work reported in this study.

Materials availability

This study did not generate new unique materials. Any additional information required to reanalyze the data reported in this article is available from the lead contact upon request.

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Figure legends

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Figure 1. HT and placebo treatment's effect sizes (\pm 95% confidence interval) on the plasma biomarkers. When the confidence interval does not intersect the dotted line (which reflects $r = 0$), the effect size is statistically significant.

Figure 2. Flow cytometric analysis of platelet activation in fresh blood. A) Platelet (PTL) gating strategy in FCS/SSC diagram. B) The platelet membrane antigen CD62P (P-selectin) expression before HT treatment. C) The platelet membrane antigen CD62P (P-selectin) expression after HT treatment. D) Influence of HT and ADP treatments on the membrane expression of CD61 and CD62P in total platelets (PTL) and annexin V (AnV) in microparticles (MPs). Data are presented as the mean fluorescence intensity (MFI).

Figure 3. Alpha and beta diversity of microbial communities. Boxplots depicting alpha diversity metrics of richness, A) Chao1 and evenness, B) Shannon, and C) Simpson, categorized according to treatment. Statistical differences were determined using the value of each index for each sample and employing the Kruskal-Wallis test, with a correction for multiple hypotheses applied according to Benjamín's procedure. D) Two-dimensional graphical representation of beta diversity for each microbial community based on the Bray-Curtis index as a measure of diversity through PCoA. Colours denote the treatment group.

Figure 4. Logarithmic fold changes in bacteria abundance. A) Genus and B) Species, according to ANCOM-BC analysis comparing After vs. Before HT treatment and After HT vs. After Placebo. A positive value of Fold Change indicates a higher level of abundance in the patients after HT treatment. Conversely, a negative value of Fold

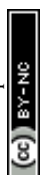


Change indicates a higher level of abundance in patients before HT treatment or the placebo group, respectively.

Figure 5. Functional enrichment analysis of the proteome comparing the responder and non-responder subgroups. A) Comparison of responder and non-responder groups before HT treatment (basal conditions); B) Comparison of responder and non-responder groups after HT treatment. A network of enriched terms coloured according to cluster identity, where nodes that share the same cluster identity are typically nearby. The most significant functions were identified based on the statistically significant differences in protein expression before and after HT treatment. \log_{10} (q-value) represents each function's significance, as shown in the table below. Process enrichment analyses were conducted using Metascape software. The pathways are listed in descending order of statistical significance.

Figure 6. Targeted metabolomics of the whole blood metabolome comparing the responder and non-responder subgroups. Volcano plot analysis of the differentially abundant metabolites. A) Before HT treatment. B) After HT treatment. The graphs depict the relative abundance of each metabolite to its statistical significance, expressed as the difference between $\log_2(\text{FC})$ and $-\log_{10}$ (q-value). C) Heatmap of metabolites after HT treatment. The heatmap scale ranged from -2 to 2 . Three patients were included in each subgroup.

Figure 7. Stacked Partial Least-Squares Discriminant Analysis (sPLS-DA) with DIABLO. A) Circos plot. The plot represents the correlations greater than 0.9 between variables of the different samples, represented on the side quadrants. B) Clustered Image Map (Euclidean distance, Complete linkage) of the multi-omics signature. Samples are represented in rows, selected features on the first component in columns.



Figures

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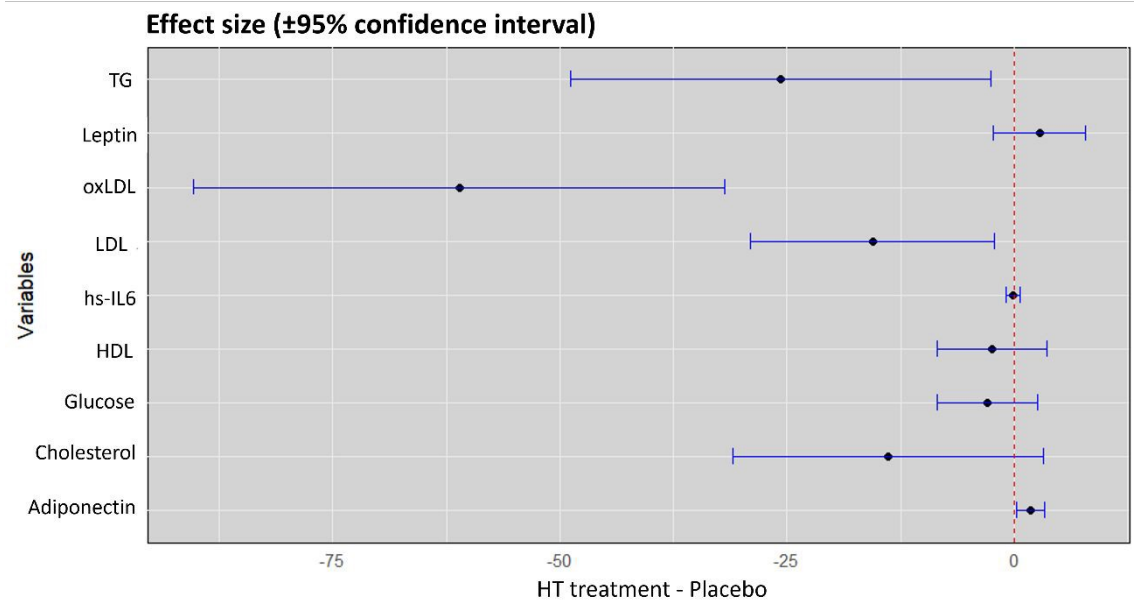


Figure 1. Noguera-Navarro et al., 2025.

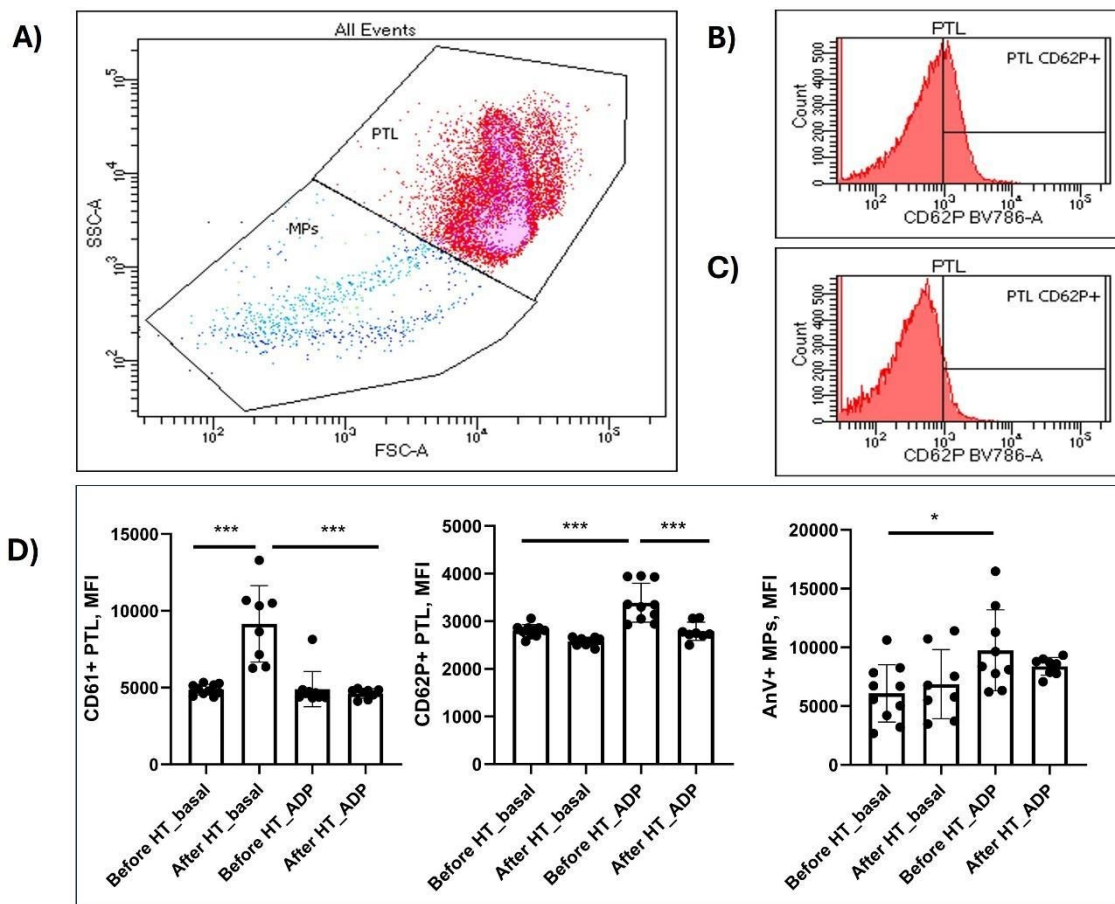


Figure 2. Noguera-Navarro et al., 2025.



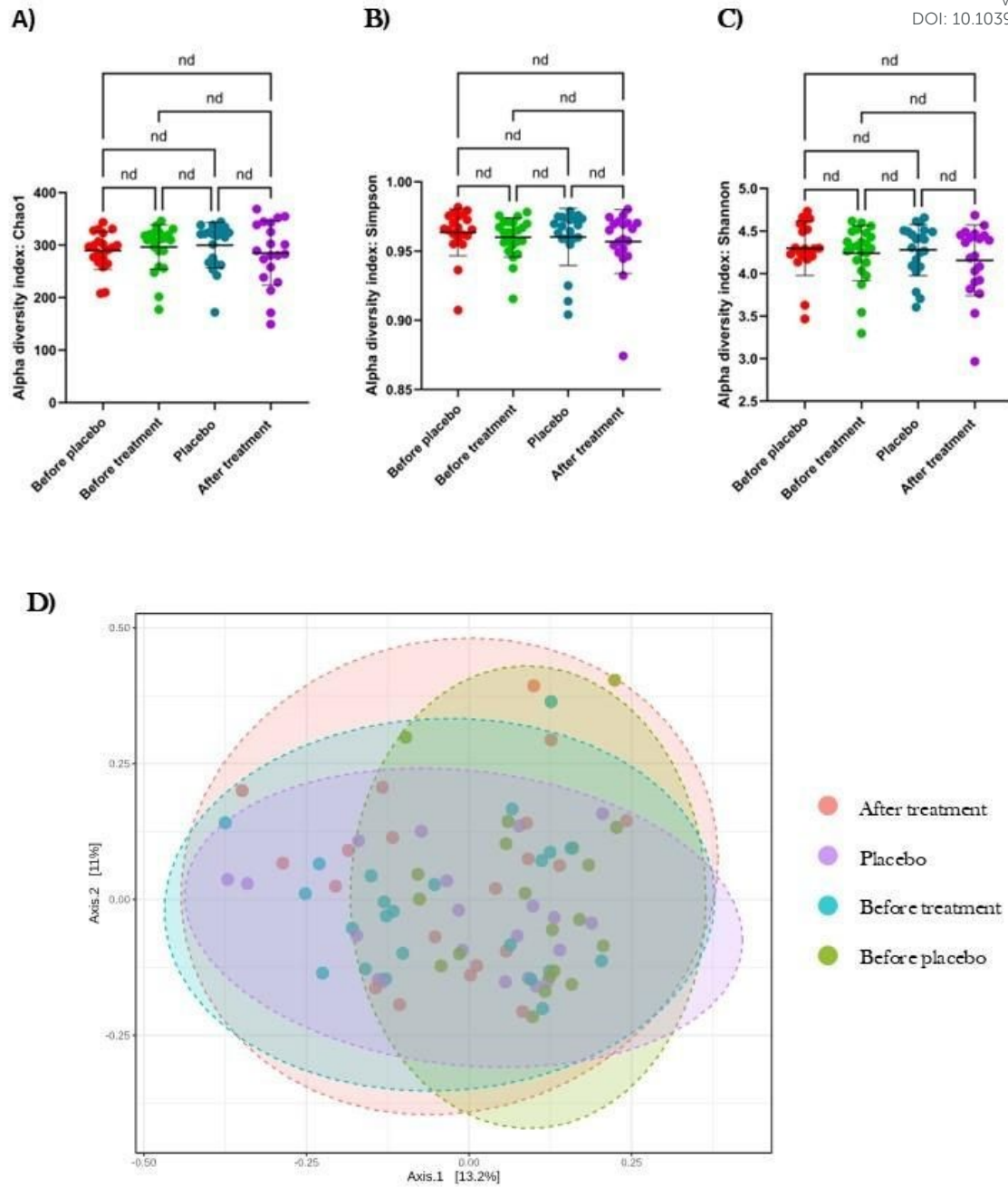


Figure 3. Noguera-Navarro et al., 2025.



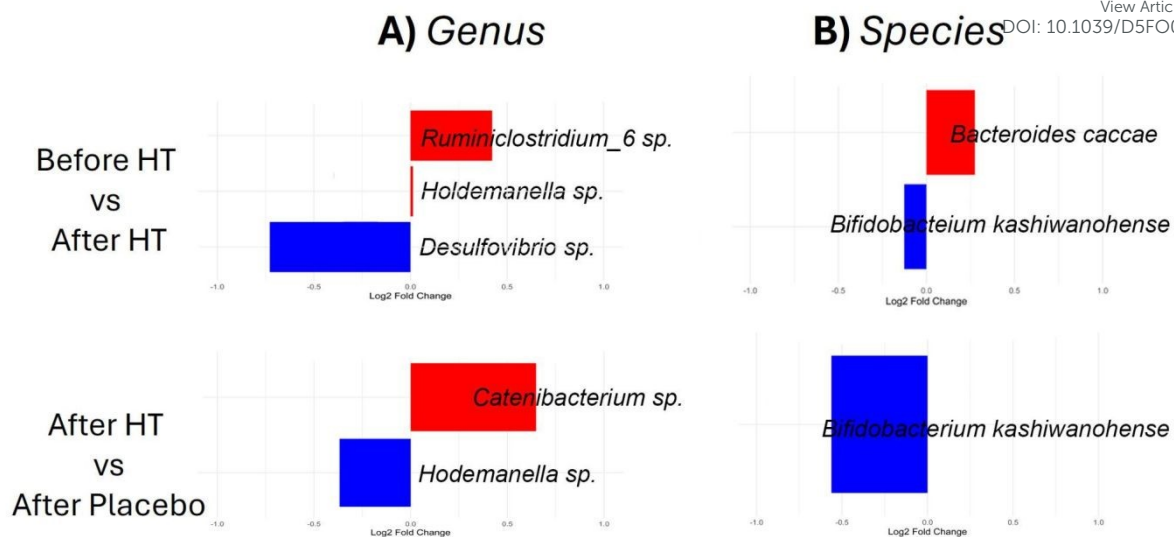


Figure 4. Noguera-Navarro et al., 2025.



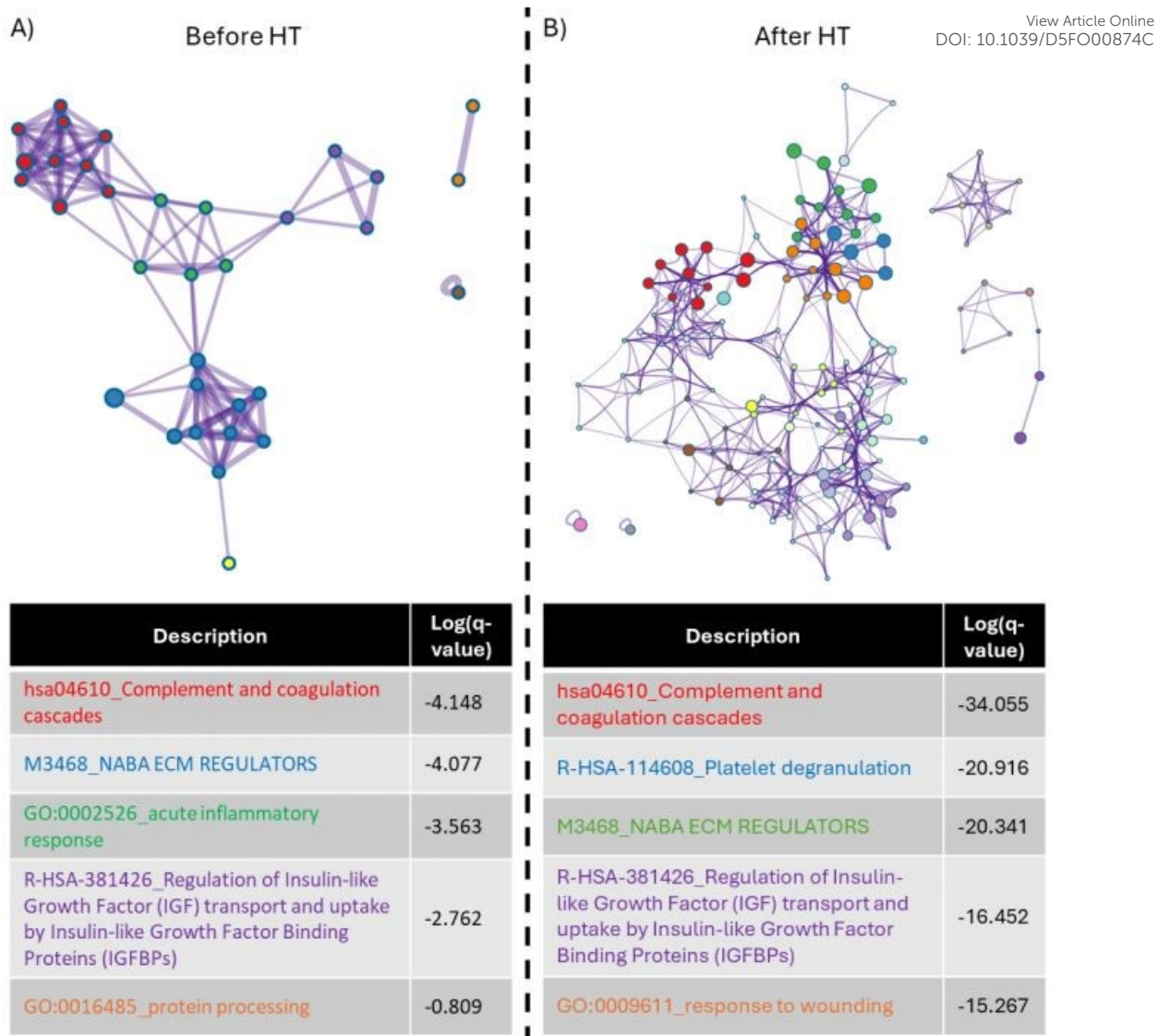
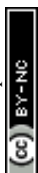


Figure 5. Noguera-Navarro et al., 2025.



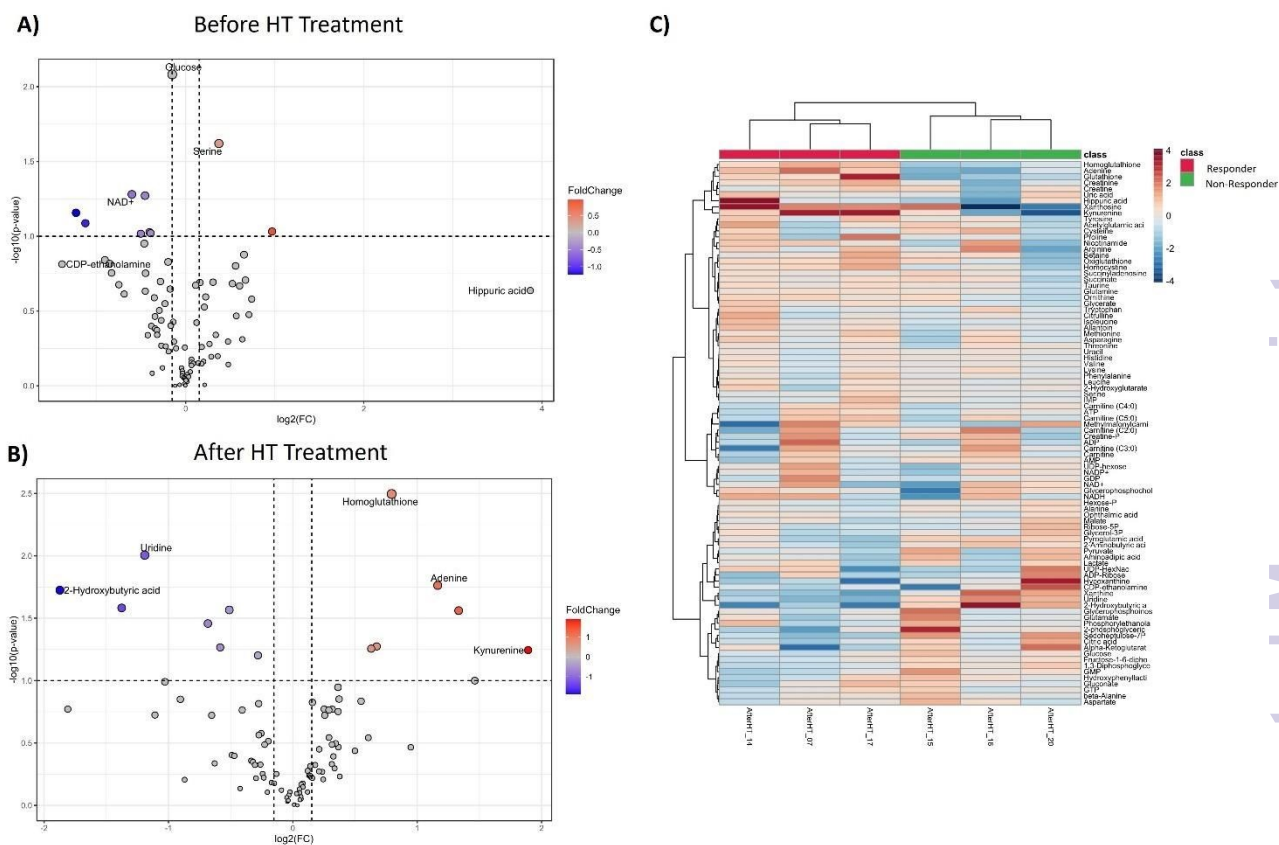


Figure 6. Noguera-Navarro et al., 2025.



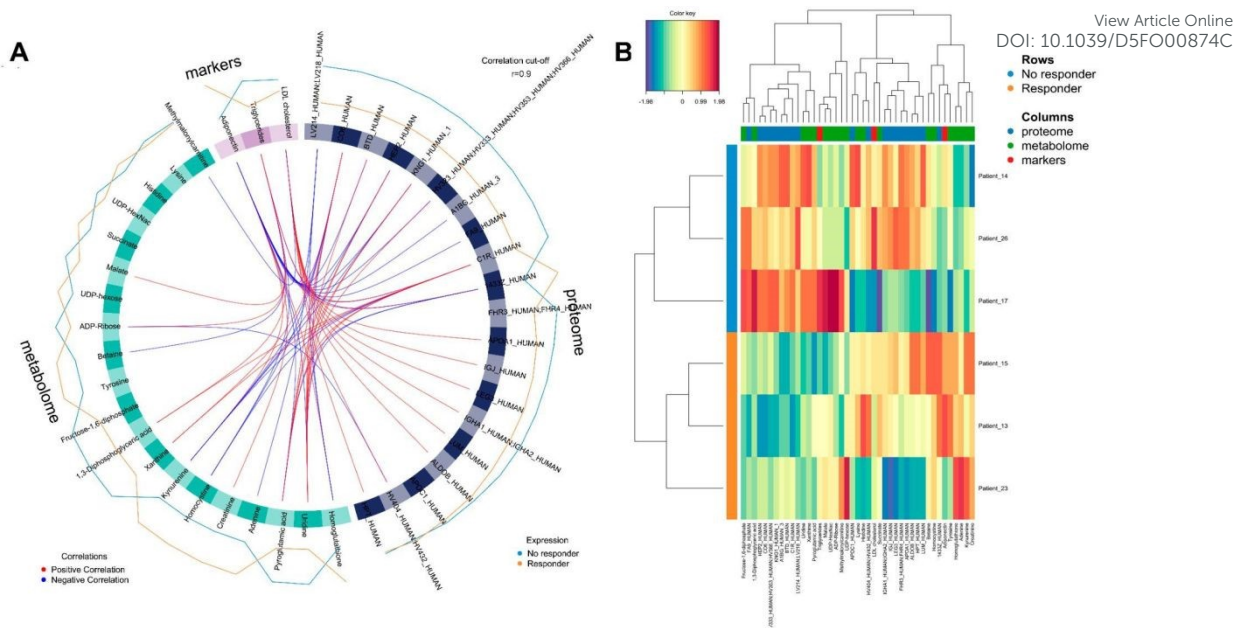


Figure 7. Noguera-Navarro et al., 2025.



Tables

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DOI: 10.1039/D5FO00874C**Table 1.** Baseline and biochemical parameters of the recruited population.

	Total (n=26)	Start HT (n=15)	Start placebo (n=11)	P-value
Baseline characteristics				
Age (years)	48.9 ±10.9	50.3 ±11	46.8 ±10.9	0.426
Male (n, %)	11 (42%)	6 (40%)	5 (45%)	0.822
BMI (Kg/m ²)	27.6 ± 5.9	27.57± 6.3	27.7 ± 5.7	0.971
Fat content, %	34.8 ±12	33.2 ±11.2	37.1 ± 13.3	0.419
Lean mass, %	26.8 ± 8.5	28.5 ±7.9	24.6 ± 9.2	0.259
Visceral fat, %	9.3 ± 4.3	9.1 ± 3.2	9.5 ± 5.6	0.813
Biochemical parameters				
Glucose (mg/dL)	86.6 ±10.2	85.6 ± 8.3	88.2 ±12.9	0.544
Total cholesterol (mg/dL)	225.2 ± 29.5	221.9 ± 30.4	230.2 ± 28.8	0.504
HDL cholesterol (mg/dL)	66.4 ± 15.2	66.3 ±15.5	66.6 ±15.7	0.962
LDL cholesterol (mg/dL)	135.8 ± 23.6	136.4 ± 24.7	135.1 ± 23.2	0.904
Triglycerides (mg/dL)	97 [81-165]	87 [79-99]	151 [114.3-177.5]	0.081
Creatine (mg/dL)	0.8 ± 0.2	0.8 ± 0.2	0.8 ± 0.1	0.520
GOT (U/L)	18.4 ± 3.6	19.1 ± 3.8	17.8 ± 3.6	0.468
GPT (U/L)	20 [13-23.3]	20 [14.5-26.5]	19.5 [11.25-14.75]	0.459
Insulin (U/mL)	9.8 [5-14.4]	9.6 [4.9-14.3]	11.2 [7.7-14.8]	0.659
HOMA-IR	2.1 [1.1-3.3]	2 [1-2.5]	2.6 [1.9-3.4]	0.403

Values are mean ± SD or median [25-75% quartiles]. T-student's and Mann-Whitney's tests were used for parametric and non-parametric parameters, respectively.



Table 2. Proteins detected by mass-spectral and differentially expressed when comparing before and after HT treatment (n=22). View Article Online
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Protein name	Difference	Pathways related	P-value
FHR1_HUMAN	-0.49	Can be associated with lipoproteins and may play a role in lipid metabolism	0.046
PROF1_HUMAN	-0.40	Binds to actin and affects the structure of the cytoskeleton	0.020
PLEK_HUMAN	-0.39	Major protein kinase C substrate of platelets	0.026
CC126_HUMAN	-0.33	Undescribed function	0.046
ACTB_HUMAN	-0.29	This gene encodes one of six different actin proteins	0.046
PRAP1_HUMAN	-0.29	Lipid-binding protein which promotes lipid absorption	0.041
IBP6_HUMAN	-0.27	Enables identical protein binding activity and insulin-like growth factor II binding activity.	0.009
TALDO_HUMAN	-0.27	Catalyzes the rate-limiting step of the non-oxidative phase in the pentose phosphate pathway	0.044
KV228_HUMAN	-0.25	V region of the variable domain of immunoglobulin light chains that participates in the antigen recognition	0.048
HV316_HUMAN	-0.24	V region of the variable domain of immunoglobulin light chains that participates in the antigen recognition	0.002
IBP4_HUMAN	-0.21	Positive regulation of insulin-like growth factor receptor signalling pathway	0.037
PTPRF_HUMAN	-0.19	Possible cell adhesion receptor	0.001
KV224_HUMAN	-0.18	V region of the variable domain of immunoglobulin light chains that participates in the antigen recognition	0.021
GGH_HUMAN	-0.15	May play an important role in the bioavailability of dietary pteroylpolyglutamates and the metabolism of pteroylpolyglutamates and antifolates	0.043
NGAL_HUMAN	-0.12	Iron-trafficking protein involved in multiple processes	0.049



TRFE_HUMAN	0.13	It is responsible for the transport of iron from sites of absorption and heme degradation to those of storage and utilization	0.049
PIP_HUMAN	0.17	Enables IgG binding activity; aspartic-type endopeptidase activity; and identical protein binding activity	0.045
ECM1_HUMAN	0.19	Stimulates the proliferation of endothelial cells and promotes angiogenesis	0.015
HV348_HUMAN	0.22	V region of the variable domain of immunoglobulin light chains that participates in the antigen recognition	0.041

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The negative values in the *Difference* section represent higher expression, while positive values represent a lower expression after HT treatment.

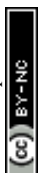


Table 3. Pearson's and Spearman's correlations between diagnostic biomarkers and differentially expressed proteins in the responder subgroup after HT treatment (n=12). View Article Online
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Biological marker	Gene	Function	R value	P-value
ADIPONECTIN	AGT	Potent regulator of blood pressure	0.87	0.002
	SERPINC1	The most important serine protease inhibitor in plasma that regulates the blood coagulation cascade	0.72	0.03
	SERPINA1; SERPINA2	Inhibitor of serine proteases. It has a moderate affinity for thrombin	0.67	0.049
TRIGLYCERIDES	THBS1	Adhesive glycoprotein that mediates cell-to-cell and cell-to-matrix interactions	-0.9	0.002
	TFPI	It possesses an antithrombotic action and the ability to be associated with lipoproteins in plasma	-0.71	0.047
LDL CHOLESTEROL	HPR	It contains HDL cholesterol	0.97	8.13E-06
	MMRN2	It acts as a negative regulator of angiogenesis	0.94	0.0002
	GPX3	It protects cells and enzymes from oxidative damage	0.89	0.001
	F10	It converts prothrombin to thrombin	0.86	0.003
	APOF	It is an important regulator of cholesterol transport	0.82	0.006
	TFPI	It possesses an antithrombotic action and the ability to associate with lipoproteins in plasma	0.79	0.01
	THBS1	Adhesive glycoprotein that mediates cell-to-cell and cell-to-matrix interactions	0.76	0.018
	APOL1	May play a role in lipid exchange and transport throughout the body	0.7	0.037
OXIDIZED LDL	SERPIND1	Thrombin inhibitor	-0.94	0.005



CPN1	It protects the body from potent vasoactive and inflammatory peptides	-0.94	0.005
F13B	It regulates the rate of transglutaminase formation by thrombin	-0.89	0.019
CFHR2	Associated with lipoproteins and may play a role in lipid metabolism	-0.89	0.019
F2	It has a role in blood homeostasis, inflammation, and wound healing	-0.89	0.019
HABP2	It activates the coagulation factor VII	-0.89	0.019
F13A1	It is activated by thrombin and calcium ions to a transglutaminase, stabilizing the fibrin clot	-0.89	0.019
MMP2	Diverse functions such as remodelling of the vasculature, angiogenesis, tissue repair, tumor invasion, inflammation, and atherosclerotic plaque rupture	-0.89	0.019
KIT	It acts as a cell-surface receptor for the cytokine KLTG/SCF	-0.89	0.019
CSF1R	It acts as a cell-surface receptor for CSF1 and IL34	-0.83	0.042



Table 4. Correlation (Pearson or Spearman) between diagnostic biological markers and significant metabolites in the responder subgroup after HT treatment (n=6). View Article Online
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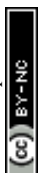
Biological marker	Metabolite	P-value
Adiponectin	Ornithine	0.043
LDL cholesterol	Aminoadipic acid	0.017
	Glycerol-3P	0.019
	Citric acid	0.025
	Pyruvate	0.027
	Alpha-Ketoglutarate	0.036
	Sedoheptulose-7P	0.037
	Uric acid	0.041
	Arginine	0.017
	Nicotinamide	0.027
Oxidized LDL	NADH	0.008
	Carnitine (C5:0)	0.04
	Phosphorylethanolamine	0.001
	Aspartate	0.035
	Glutamate	0.035
AnV+ Microparticles (n=4)	Uric acid	0.014
	Carnitine (C4:0)	0.038
	Phosphorylethanolamine	0.011
	Ribose-5P	0.04
Platelet CD61 expression, MFI (n=4)	IMP	0.004
	Glycerophosphoinositol	0.012
	Leucine	0.032
	Carnitine	0.019
	Carnitine (C3:0)	0.025
Platelet CD62P expression, MFI	Glycerol-3P	0.008
	Hypoxanthine	0.009



(n=4)	Carnitine (C2:0)	0.018
	Aspartate	0.018
	Glutathione	0.016

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Abbreviations: LDL: Low-Density Lipoproteins cholesterol, MFI: Mean Fluorescence Intensity, AnV+: Annexin V+.



This study utilizes data that are available as follows:

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1. **Data Location:** Further information and requests for resources and information should be directed to and will be fulfilled by Silvia Montoro-García (smontoro@ucam.edu).
2. **Data Access:**
 - [✓] The data are accessible at request.
3. **Restrictions:** There are no restrictions on data access.

