

Published in final edited form as:

Obesity (Silver Spring). 2016 June; 24(6): 1257–1265. doi:10.1002/oby.21482.

Pro-Inflammatory and lipid biomarkers mediate the metabolically healthy obesity phenotype: A shotgun proteomics Study

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Abstract

Objective—The metabolically healthy obesity (MHO) phenotype is an important obesity subtype in which obesity is not accompanied by any metabolic comorbidity. However, the underlying molecular mechanisms remain elusive. In the present study, we used a shotgun proteomics approach to identify circulating biomolecules and pathways associated with MHO.

Methods—The subjects were 20 African-American women: 10 MHO cases and 10 metabolically abnormal individuals with obesity (MAO), controls. Serum proteins were detected and quantified using label-free proteomics. Differential expression of proteins between the two groups was done and the list of differentially expressed proteins was analyzed to determine enriched biological pathways.

Results—Twenty proteins were differentially expressed between MHO and controls. These proteins included: hemoglobin subunits (HBA1, p= 6.00×10^{-18}), haptoglobin-related protein (HPR, p= 1.2×10^{-15}), apolipoproteins (APOB-100, p= 1.50×10^{-40} ; APOA4, p= 1.1×10^{-14}), retinol-binding protein 4 (RBP4, p= 7.1×10^{-08}) and CRP (p= 2.0×10^{-04}). MHO was associated with lower levels of proinflammatory and higher levels of anti-inflammatory biomarkers when compared to MAO. Pathway analysis showed enrichment of lipids and inflammatory pathways, including LXR/ RXR and FXR/RXR activation, and acute phase response signaling.

Conclusions—These findings suggested that protection from dysregulated inflammatory and lipid processes were primary molecular hallmarks of MHO. The candidate biomarkers (AHSG, RBP4, and APOA4) identified in this study were potential prognostic markers for MHO.

Competing interests: The authors declare that they have no competing interests.

Raw proteomics study dataset: available at http://crggh.nih.gov/resources.cfm under Howard University Family Study (HUFS) description for download and/or collaboration

Authors' contributions: AD- conceived the study, responsible for study coordination, performed the statistical and proteomics analyses, drafted the manuscript and interpreted the data. AA- participated in the design of the study, analyzed the proteomics data, and edited and approved the manuscript. JZ- cleaned and managed the data. MZ, PD- coordinated and carried out the shot-gun proteomics in the proteomics core lab. CR- participated in the conception and design of the study, guided the research and approved the final manuscript. All Authors read and approved the final manuscript.

^{*}These authors contributed equally to this research project.

Keywords

Metabolically healthy obese; proteomics; inflammation; lipids metabolism; African Americans

Introduction

Obesity has reached epidemic proportion globally (1, 2). Annually, 300,000 excess deaths are linked to obesity and its complications, making obesity the second leading cause of premature death (3). Obesity is a major contributor to the global burden of disabilities and chronic diseases including high blood pressure, type 2 diabetes (T2D) and heart diseases (1, 2). While obesity triggers a range of complications, the most common complication is T2D, which is usually preceded by a state of insulin resistance (IR) before T2D is actually diagnosed (4). Paradoxically, it has been documented that not all individuals develop insulin resistance or T2D or other cardio-metabolic co-morbidities as they gain weight, a phenomenon referred to as metabolically healthy obesity (MHO). This paradox has led to the claim that "not all obese humans are created equal" (5). Metabolically healthy subjects with obesity (MHO) were first described about 30 years ago (6) with recent extensive description by Karelis et al. and many other investigators. (6-10). Using the presence of metabolic abnormalities as a classifying variable, individuals with obesity can be grouped into: MHO (also known as the insulin sensitive individuals with obesity) and the metabolically abnormal subjects with obesity (MAO) or the insulin resistant individuals with obesity. Reported prevalence of MHO varies between 10 and 40% of all subjects with obesity and with prevalence varying by many factors including age, ethnic background and level of physical activity (11, 12). Interestingly, interventions such as weight loss in MHO had no apparent benefit or may even be detrimental as shown by study conducted by Shin et al., and others (13, 14).

There is evidence that the MHO phenotype is not stable and that many of such individuals eventually develop metabolic abnormalities (15-17). While many MHO may convert to MAO over time, there is a substantial proportion of MHO who remained protected from obesity-related abnormalities (18). The MHO phenotype seems to span the entire life span and has been observed in individuals in their ninth decade of life, as demonstrated in the National Health and Nutrition Examination Survey (NHANES) study in which MHO was 22.1% among obese individuals 80 years and older (19). These observations suggest that the phenotype provides an opportunity to investigate the factors that distinguish MHO from MAO and thereby gain insight into biomarkers and molecular mechanisms that are involved in the development of cardio-metabolic abnormalities. Most studies of MHO phenotype have been epidemiological studies (7, 8, 18, 20-22) and only a few have focused on the molecular basis of MHO.

In the present study, we use a shotgun proteomics approach to evaluate differential protein expression between MHO and MAO individuals. By identifying differentially expressed proteins in serum samples of MHO compare with MAO, we aim to gain biological insights into the underlying biological mechanisms that mediate the MHO phenotype.

Methods

Subject selection and study design

The individuals included in this study were selected from a well phenotyped cohort of African Americans (AA) recruited in the Washington DC area to study the genetic epidemiology of complex diseases in populations of African descent, the Howard Family University Study (HUFS). Briefly, the main objective of the HUFS was to enroll and examine a randomly ascertained sample of AA families along with a set of unrelated individuals for study of the genetic and environmental bases of common complex traits including hypertension, obesity, diabetes, and associated phenotypes. In order to maximize the utility of this cohort for the study of multiple common traits, families and individuals were not ascertained based on any phenotype. All participants were recruited after an overnight fast of at least 8 hours prior to the blood draw and all collected samples were stored in a -80°C freezer pending measurement of biochemical parameters (22). This study was approved by the Howard University's institutional review board (IRB) and informed consent was obtained from each participant.

Case definition of MHO—Since this study is exploratory, we used very stringent definitions to select individuals included as MHO or MAO. A combination of three definitions of MHO was applied as follows:

Definition 1: No hypertension (BP<=130/85, no BP medication), no diabetes (glucose <=126 mg/dl), HDL-C>= 40 mg/dl for men and >=50 mg/dl for women, all conditions have to be met. This is the basic MHO definition

Definition 2: No hypertension (BP<=130/85, no BP medication), no diabetes (glucose <=100 mg/dl), HOMA<=5.1, TG/HDL <= 1.65 for men TG/HDL <= 1.32 for women, all conditions have to be met. This is the modified Wildman et al. definition (19)

Definition 3: All Definition 2 criteria + hsCRP <= 0.3 mg/dl, all conditions have to be met. This third definition takes into account inflammatory status. The CRP cut-off used was based on Karelis et al. recommendations (9).

The subjects included had to meet all three definitions, that is, these individuals (cases) would be identified as MHO by all three definitions (Figure 1, Supplementary Table S1)

Definition of controls (MAO)—MAO were selected if they were individuals with obesity who had 4 or all 5 of the following abnormalities: i.e. hypertension, no diabetes but glucose level higher than 100 mg/dl, HOMA>5.1, TG/HDL > 1.65 for men TG/HDL > 1.32 for women and hsCRP>0.3 mg/dl).

Ten MHO cases and 10 MAO controls meeting these criteria were studied.

Serum proteomics

Serum shotgun proteomics were done at the National Cancer Institute, Proteomics Core Lab. Serum samples were depleted of 14 highly abundant proteins using Agilent's Multiple

Affinity Removal System (MARS) spin cartridges which are 0.45 ml immunoaffinity cartridges packed with antibody-modified resin (Agilent Technologies, Santa Clara, CA) (list of the 14 proteins depleted, Supplementary Table S2). The enriched pool of low abundant proteins were then collected and combined from several flow-through fractions. Quality control (QC) measures were used to assess depletion efficiency by removing an aliquot of both the low and high abundant pools and subsequently resolving the proteins on a 4-12% SDS-PAGE gel. Proteins were visualized by staining with Coomassie blue.

A BCA assay was used to determine the protein concentration of the low abundant protein pools. The samples were desalted and concentrated by buffer exchange using 5000 Da molecular weight cut off (MWCO) spin concentrators followed by protein digestion using sequencing grade modified trypsin at a 1:20 trypsin-to-protein ratio. Resultant peptides were then subjected to solid phase extraction followed by fractionation using strong cation exchange (SCX) chromatography.

Each SCX peptide fraction was analyzed by microcapillary reversed-phase liquid chromatography tandem mass spectrometry (μ RPLC-MS/MS) using an Agilent 1100 capillary LC system (Agilent Technologies, Santa Clara, CA) coupled online to a linear ion trap mass spectrometer (LTQ, Thermo Electron, Waltham, MA, USA). Separations were performed using 75 μ m i.d. \times 360 μ m o.d. \times 10 cm long fused silica capillary ESI emitter columns (Polymicro Technologies, Phoenix, AZ) slurry-packed in house with 3 μ m, 300Å pore size C-18 silica (Vydac, Hysperia, CA). The mass spectrometer was operated in data-dependent MS/MS mode with each full MS scan being followed by seven MS/MS scans where the seven most intense peptide molecular ions in the MS scan were sequentially and dynamically selected for collision-induced dissociation (CID). Dynamic exclusion was employed to minimize redundant acquisition of tandem mass spectra.

Mass spectra were searched against a Uniprot human proteomic database using SEQUEST. A tryptic enzyme restriction with a maximum of two internal missed cleavage sites was used and methionine residues were considered modified in the database search. For a fully tryptic peptide to be considered legitimately identified, it had to achieve stringent charge state and proteolytic cleavage-dependent cross correlation (Xcorr) scores of 1.9 for [M+H] ¹⁺, 2.2 for [M+2H] ²⁺, 3.1 for [M+3H] ³⁺ and 4.5 for [M+4H] ⁴⁺, and a minimum delta correlation (Cn) of 0.08. SEQUEST results were further filtered and analyzed using software developed in-house. The output is a peptide count data in which columns represent peptide counts (peptide abundance) in each experimental sample and rows represent distinct proteins detected in the samples.

Data quality control and Data Analysis

A total of 16,553 proteins were identified, at least once, among the 20 subjects. The proteomics data was analyzed using DanteR/InfernoRDN R-based software packages designed to analyze quantitative proteomics data (23). Only proteins detected in both groups (n=7836 proteins) were retained for analysis. The proteomics data was reported as peptide counts (PC) which follows a Poisson distribution. To determine proteins that were differentially expressed among the two groups, analysis of variance (ANOVA) was carried out using Poisson regression. P-values were corrected for multiple testing using Benjamin

and Hochberg (FDR) and the threshold for significance was p<0.05. The list of differentially expressed proteins (DEPs) was submitted to Ingenuity pathway analysis (IPA) to determine biological functions and pathways that were enriched (IPA®, QIAGEN Redwood City, www.qiagen.com/ingenuity). The IPA biological functions analysis calculates the probability of a set of differentially expressed proteins being associated with known biological functions /canonical pathways by chance alone. The significance of the association between the DEPs and the pathways was measured in 2 ways: 1) A ratio of the number of molecules from DEP dataset that map to the pathway divided by the total number of molecules that map to the canonical pathways, 2) P-values obtained from Fisher's exact tests were used to assess the probability that observed associations between the DEPs and the canonical pathway were due to chance alone; p-values <0.05 were considered significant.

Anthropometric and clinical parameters were compared between the two groups by Student's t-test using SPSS package. Data is presented as mean± standard deviation or median and interquartile range when appropriate.

Results

Characteristics of study groups

The anthropometric and clinical characteristics of the study subjects are summarized in Table 1. The cases (MHO) and the controls (MAO) individuals were not statistically different from each other in key parameters including age, BMI, PFM and hip circumference. As expected, the 6 parameters (BP, glucose, HOMA-IR, HDL-C, TG, and hsCRP) used in the selection of the two groups are statistically different between the two groups. Notably, MHO had healthier lipid profile with considerably lower TG and higher HDL-C level compared to MAO. Inflammation, measured by adiponectin and hsCRP, was minimal in MHO compared to MAO. Notably, adiponectin, an adipokine decreased in obesity, is paradoxically higher in this group of MHO, about 1.8 times higher than in MAO group (8,074.0±3,977.4 ug/ml vs. 4,486.8±2,535.9 ug/ml).

Identification of differentially expressed proteins among MHO and MAO

At total of 56 proteins were differentially expressed (33 up-regulated and 23 down-regulated) at an unadjusted p<0.05 as shown on the volcano plot (Supplementary Figure S1 and Supplementary Table S3). After adjusting for multiple testing by the Benjamini &Hochberg method and using a p-value threshold of <0.05, 20 proteins were differentially expressed between MHO and MAO including 8 down-regulated and 12 up-regulated in MHO compared to MAO (Table 2, Supplementary Figure S2). Among the most upregulated in MHO were ApoB-100 (p=1.5×10⁻⁴⁰, FC=1.2), and Alpha-2-HS-glycoprotein, ASHG (p=3.00×10⁻²⁹, FC=1.7). Hemoglobin subunit alpha (HBA, p=6.0×10⁻¹⁸, FC=0.44) and haptoglobin-related protein (HPR, p= 1.2×10^{-15} , FC=0.6) were down-regulated in MHO compared with MAO. C-reactive protein was also confirmed to be down-regulated in MHO by the proteomic data (p= 2.0×10^{-04}). Supplementary Figure S3 is a principal components (PC) plot of the DEPs that shows separation of MHO and MAO subjects by PC1, PC2 and PC3, indicating the discriminant ability of these 20 proteins.

Enriched biological pathways in the most differentially expressed proteins in MHO compared with MAO

We used IPA to identify enriched biological pathways in the 20 DEPs. The most significant enriched canonical pathways include *LXR/RXR activation* (p=1.4×10⁻¹⁵), *FXR/RXR Activation* (p=1.9×10⁻¹⁵), *acute phase response Signaling* (2.9×10⁻¹⁴), and *complement system* (p=4.6×10⁻⁰⁸) (Figure 2, Table 3).Interestingly, a number of the enriched pathways are interconnected through DEPs that are common to more than one canonical pathway. For example, acute phase response signaling has 3 and 4 DEPs in common with complement system (C4A/C4B, CFB, SERPING1) and *LXR/RXR Activation* (AHSG, C4A/C4B, ITIH4, RBP4) respectively (Figure 3, Table 3). Figure 4 depicted DEPs mapped to the most significant enriched canonical pathways.

Discussion

Despite considerable interest in this subset of metabolically healthy individuals with obesity (MHO), the underlying molecular mechanisms remain largely unknown. Understanding the pathophysiology of MHO is important not only for this subset of individuals but also for obesity as a whole. Here, we describe the results of the first study to use high-throughput omics technology, namely shotgun label-free quantitative proteomics, to profile the proteome and obtain a comprehensive snapshot of differences in protein expression between MHO and MAO African American women.

The differentially expressed proteins between MHO and MAO underscored important inflammation-related proteins including α -2 HS glycoprotein (AHSG), C-reactive protein (CRP), Histidine-rich glycoprotein (HPR), retinol binding protein-4 (RBP4), complement factor-4A (C4A), and Inter-alpha-trypsin inhibitor heavy chain H4 (ITIH4). These molecules belong to the family of acute phase reactant proteins (APP) and are primarily produced by the liver in response to injury and infection (24) but some (e.g. RBP4) are also produced by adipocytes (25). It is well documented that inflammation is a key component of obesity and has been associated with the progression and development of obesity-related complications (26). In this study, the APPs differentially expressed can be categorized into two groups, the negative acute phase proteins (AHSG, HRG, and RBP4) all of which were over-expressed in MHO and the positive acute phase proteins (HRP, CRP, C4A, and ITIH4) which in contrast to the negative APPs were down-regulated in MHO. These changes suggest a favorable inflammatory profile associated with MHO.

Some of these molecules have also been involved in adverse metabolic events such as impairment of insulin signaling, fatty liver diseases and suppression of adiponectin production (25, 27-31). For example, high levels of AHSG, a glycoprotein produced by hepatocytes, have been linked to metabolic syndrome in certain populations (28, 30). However, the association seems to vary with gender as one report showed an association between high AHSG and metabolic syndrome in Japanese men but not in women (32). While our study showed that AHSG is upregulated in MHO, we did not see any of the other reported changes associated with elevated AHSG. For example, the MHO in this study had high adiponectin level compared to MAO (8,074.0±3,977.4 ug/ml vs. 4,486.8±2,535.9 ug/ml) and remained insulin sensitive based on their HOMA-IR values (1.8 vs. 17.3). These

results suggest at least two hypotheses – (1) AHSG is regulating its effectors differently in this MHO cohort or, 2) a threshold in AHSG concentration needs to be reached to see its effects on metabolic markers such adiponectin and insulin. In the latter scenario, AHSG may be a potential biomarker on its own or in combination with other proteins (e.g. RBP4, APOA4, APOB) to identify MHO that will convert into MAO.

RBP4, an adipokine with differential regulation in animals and humans and involved in the transport of vitamin A to peripheral tissues, has been associated with dysfunction of insulin signaling especially in animal models. However, in humans the positive association between RBP4 levels and insulin resistance has not been consistent (25, 29, 33). RBP4 could also be a promising marker to follow the changes in MHO especially now that MHO phenotype has been shown to be a transient state for many affected individuals (12).

The complement system pathway, part of the humoral immune system and involved in inflammatory processes, was over-represented among the DEPs associated with MHO. As expected, all DEPs (C4A/C4B, CFB, SERPING1) associated with the complement system were also associated with the acute phase reactant pathway showing a commonality between the over-represented pathways. All but SERPING1 were down-regulated thus agreeing with the lower inflammation profile seen in MHO. In fact, SERPING1 is an inhibitory regulator of the complement cascade and may play a role in inflammation suppression. SERPING1 has been shown to be down-regulated in patients with T2D compared to healthy controls (34).

Nine of the DEPs (AHSG, APOA4, APOB, C4A/C4B, GC, HPR, ITIH4, PON1, and RBP4) were associated with lipid metabolism. Four of these DEPs are involved in both in inflammatory and lipid processes (AHSG, C4A/C4B, ITIH4, and RBP4), underscoring the pleitropic function of some of the DEPs. Previous studies have found an association between AHSG and RBP4 and serum lipids (mainly triglycerides and HDL-C) (27, 35). ApoB100, APOA4, and PON1 are all members of the HDL/LDL family. APOA4 is secreted by enterocytes along with triglycerides and chylomicrons. For example, APOA4 has anti-oxidant and anti-inflammatory properties and is proposed to be protective against cardiovascular diseases. It also plays a role in glucose homeostasis and satiety (reduces food intake) (36). Interestingly in T2D and obesity, increased levels of APOA4 have been found to be mainly related to hypertriglyceridemia (37). In this study, APOA4 is up-regulated without hypertriglyceridemia; while the mechanisms involved need further investigation, the absence of hypertriglyceridemia seems to be protective in MHO.

A major strength of the study is the stringent and conservative approaches used to define MHO individuals (cases). Three distinct definitions of MHO were used and only individuals who were classified as MHO by all 3 definitions were selected and studied. Though this approach decreases the pool of MHO, it is more effective in reducing misclassification. In contrast to a recently published metabolomics study in MHO in which the definition of MAO used may have resulted in heterogeneous metabolic phenotypes (38), this study included only MAO with all 4 or 5 metabolic abnormalities. Secondly, the differential proteomic method combined with bioinformatics analysis (IPA) demonstrated changes in serum protein profiles in MHO individuals that were not previously reported.

While very promising, the insights provided by this study into the molecular basis of MHO should be interpreted within the following context -1) by design, this study is a discovery investigation and as such the findings need to be confirmed in larger and diversified populations; 2) the cross-sectional design provides only a snapshot of the proteome and causality should not be inferred; and 3) important confounding factors that may affect serum proteome including level of physical activity and diet were not analyzed in this study. While earlier studies didn't find any difference in dietary profile and physical activity between MHO and MAO (11, 13, 18, 26), a recent study with larger sample sizes and the use of healthy eating index (HEI) showed that MHO have a better dietary compliance to the US guidelines than MAO including intake of more whole grains, fruits and beans . However, the authors acknowledged that the effect seen may be due to reverse causality (39). Though diet and exercise are likely to affect biological functions in MHO, studies available to date have not specifically assess inflammation and other biological functions in a systematic manner. Future omics studies, including whole genome microarray, miRNA profiling and whole exome sequencing, investigating the determinants of the MHO phenotype will benefit from having data on potential lifestyle confounding variables such as physical activity and diets.

Conclusions

Our study is the first to use proteomics to investigate the mechanisms that are involved in MHO phenotype. Our findings confirm previous reports that inflammation is a key hallmark of metabolic health in individuals with obesity but also identified a network of APPs, more than what was previously reported, to be involved in the inflammatory process associated with MHO. Some of these proteins mainly RBP4, AHSG, APOA4, PON1, and APOB100 could be good candidates to follow up in longitudinal studies to determine their ability to predict which MHO individuals convert to MAO. We also found that differential regulation of lipid metabolism is also important to MHO phenotype.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Acknowledgments

Support for this study is provided by NIH grant No. 3T37TW00041-03S2 from the Office of Research on Minority Health, the National Human Genome Research Institute (NHGRI), the National Center for Research resources (NCRR) and NIDDK grant DK-54001. The contents of this publication are solely the responsibility of the authors and do not necessarily represent the official view of the National Institutes of Health. This research was supported in part by the Intramural Research Program of the Center for Research on Genomics and Global Health. The Center for Research on Genomics and Global Health is supported by the National Human Genome Research Institute, the National Institute of Diabetes and Digestive and Kidney Diseases, the Center for Information Technology, and the Office of the Director at the National Institutes of Health (Z01HG200362).

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Metabolically healthy obesity (MHO) has been extensively described in epidemiology studies and across populations. Its prevalence varies between 10- 40%. MHO has been associated with reduced inflammation as measured by adipokines such as adiponectin, CRP. However, the molecular mechanisms involved in the MHO phenotype remain elusive.

- Identification of acute phase proteins involved in the inflammatory process associated with MHO.
- The candidate biomarkers (AHSG, RBP4, and APOA4) are potential prognostic markers for MHO.
- Novel insights into the molecular determinants of the MHO phenotype can be gained by using omics tools.

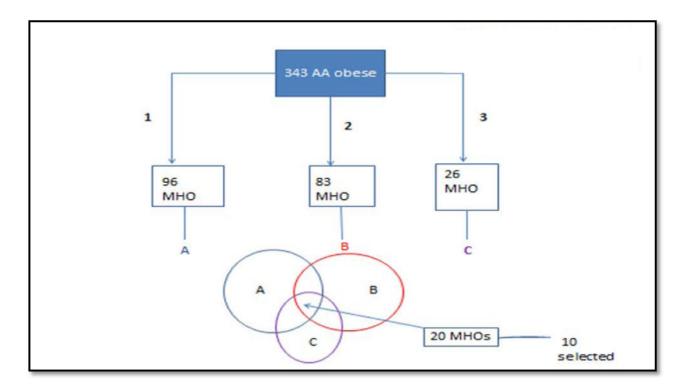


Figure 1. Flowchart for MHO cases selection

- 1, definition 1: No hypertension (BP<=130/85, no BP medication), no diabetes (glucose <=126 mg/dl), HDL-C>= 40 mg/dl for men and >=50 mg/dl for women, all conditions have to be met.
- 2, definition 2: No hypertension (BP<=130/85, no BP medication), no diabetes (glucose <=100 mg/dl), HOMA<=5.1, TG/HDL <= 1.65 for men TG/HDL <= 1.32 for women, all conditions have to be met
- 3, definition 3: definition 2 + hsCRP<= 0.3 mg/dl

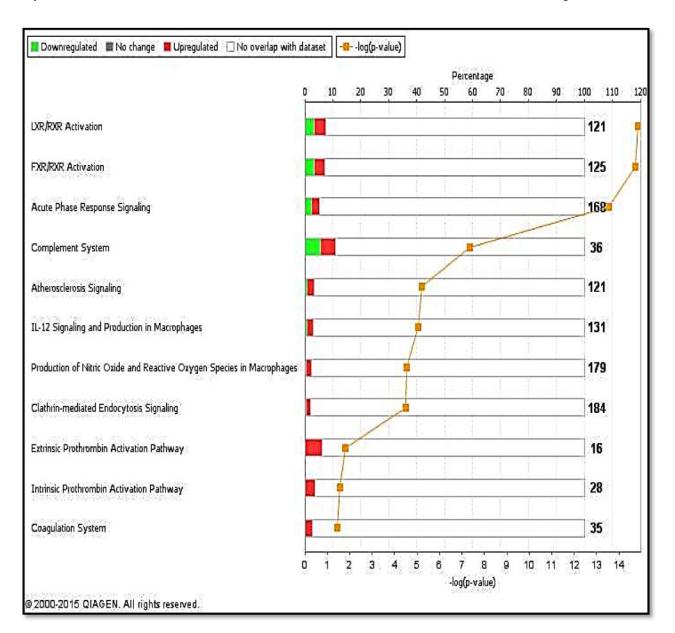


Figure 2. Canonical Pathways significantly enriched in the 20 DEPs

Bars represent total number of molecules in a pathway; upper Y-axis represents the ratio between the number of molecules in our dataset that is associated with a pathway and the total number of molecules in that pathway; Lower Y-axis represents –log of the p-value associated with each pathway. X-axis lists the top pathways associated with the DEPs.

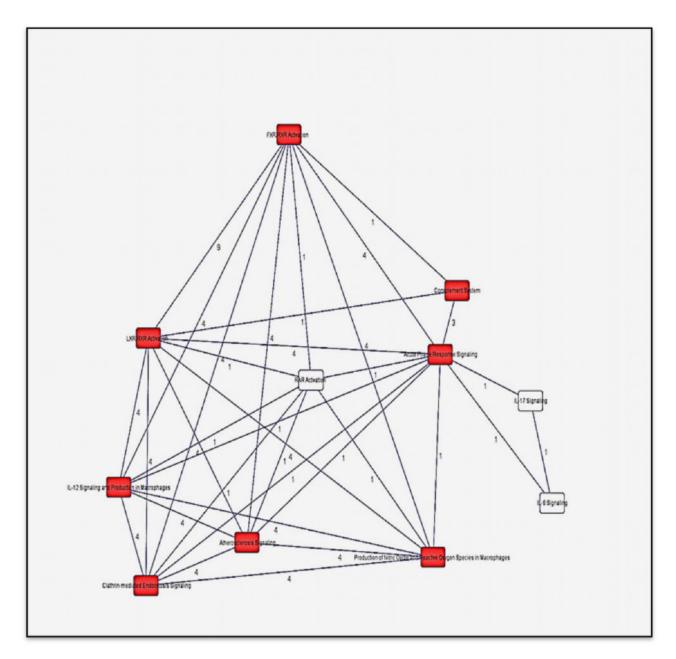


Figure 3. Overlap between the top canonical pathways enriched in the 20 DEPs
Black lines: interconnectivity between canonical pathways; red boxes: significantly enriched canonical pathways in the study; Numbers on black lines represent the number of proteins

common to two interconnected pathways.

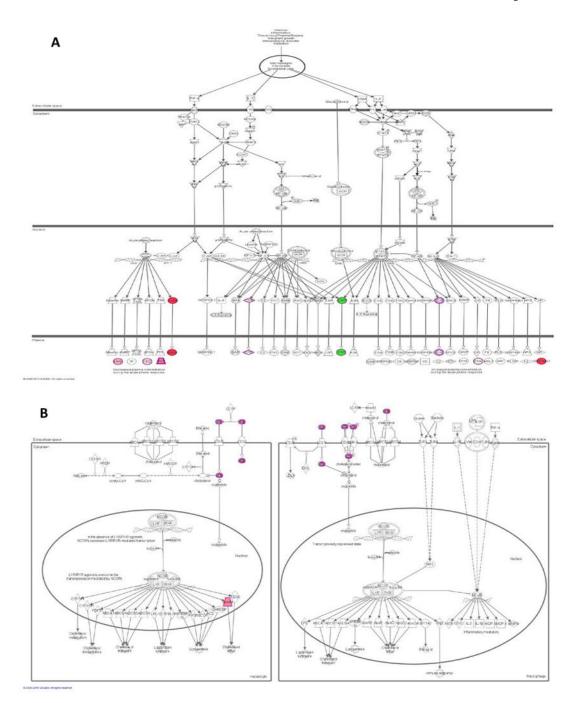


Figure 4. Two of the most significantly enriched canonical pathways in MHO

Figure 4A. Acute Phase Reactant Signaling in MHO overlapped with DEPs

Figure 4B. LXR/RXR activation in MHO overlapped with DEPs

Red: Proteins upregulated in our dataset, the intensity of the color indicates the degree of upregulation. Green: Proteins down-regulated in our dataset, the intensity of the color indicates the degree of down-regulation.

White: Proteins not specific to our dataset, but incorporated as part of the network

Table 1 Characteristics of study subjects (MAO vs. MHO)

Variables	MAO	мно	P*
Age (years)	43.3±7.8	41.3±10.1	0.6
BMI (kg/m2)	44.7±13.0	36.5±5.7	0.08
Waist (cm)	120.8±21.1	102.5±16.0	0.04
Hip (cm)	134.3±20.4	125.1±15.0	0.3
WHR	0.90 ± 0.05	0.82 ± 0.06	0.004
PFM	49.1±5.4	47.1±5.8	0.4
SBPS (mmHg)	145.8±25.1	118.2±6.3	0.007
DBPS(mmHg)	86.4±13.6	73.8±4.8	0.01
Glucose (mg/dl)	103.0±16.2	83.8±4.5	0.004
Insulin (uU/ml)	67.1±86.2	8.9±3.9	0.04
HOMA-IR	17.3±23.6	1.8±0.9	0.05
TG (mg/dl)	233.5±186.6	83.8±34.9	0.03
LDL-c (mg/dl)	103.4±35.3	130.5±51.3	0.2
HDL-c (mg/dl)	34.1±8.6	62.2±8.7	< 0.001
Adiponectin (ug/ml)	4486.8±2535.9	8074.0±3977.4	0.03
hsCRP (mg/dl)	2.2± 1.5	0.2±0.08	0.02

 $^{^{*}}$ Student's T-test, compared means between the two groups. P-value<0.05 in bold

Values in Table 1 represent mean+/- standard deviation

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List of Differentially expressed proteins among MHO and MAO African-American women determined by label-free serum proteomics Table 2

Accession #	Protein name	Gene symbol	Log effect size (log2FC)	PMHO unadjusted	Рмно вн	Average PC in MHO	Average PC in MAO
		8 Under-expr	8 Under-expressed Proteins in MHO compared to MAO	npared to MAO			
P69905	Hemoglobin subunit alpha	HBA1	-1.17	3.1×10^{-21}	6.00×10^{-18}	8.5	27.5
P00739	Haptoglobin-related protein	HPR	-0.78	7.6×10 ⁻¹⁹	1.2×10-15	19.2	42.1
P68871	Hemoglobin subunit beta	HBB	-0.80	4.6×10 ⁻¹⁵	4.5×10 ⁻¹²	13.8	30.8
P00751	Complement factor B	CFB	-0.21	1.4×10 ⁻¹⁰	1.1×10 ⁻⁰⁷	172.7	212.6
Q14624	Inter-alpha-trypsin inhibitor heavy chain H4	ITIH4	-0.16	3.0×10 ⁻⁰⁹	1.70×10 ⁻⁰⁶	242	285.1
P02741	C-reactive protein	CRP	-2.13	4.00×10 ⁻⁰⁷	2.0×10 ⁻⁰⁴	1.2	10.1
P27169	Serum paraoxonase1	PON1	-1.16	1.5×10 ⁻⁰⁶	7.0×10 ⁻⁰⁴	2.44	7.8
P0C0L4	Complement C4-A	C4A	-0.081	2.8×10^{-06}	1.0×10^{-03}	638.4	692.5
		12 Over-expr	12 Over-expressed Proteins in MHO compared to MAO	npared to MAO			
P04114	Apolipoprotein B-100	APOB	0.28	1.9×10 ⁻⁴⁴	1.5×10-40	563.1	423.7
P02765	Alpha-2-HS-glycoprotein	AHSG	0.75	7.7×10 ⁻³³	3.00×10^{-29}	79.1	37.4
P01008	Antithrombin-III	SERPINC1	0.41	3.50×10^{-26}	9.2×10 ⁻²³	163.8	108.2
P06727	Apolipoprotein A-IV	APOA4	0.40	8.5×10^{-18}	1.10×10^{-14}	113.8	76.1
P05155	Plasma protease C1 inhibitor	SERPING1	0.71	4.6×10 ⁻¹⁵	4.5×10 ⁻¹²	37	18.2
P02753	Retinol-binding protein 4	RBP4	0.46	8.1×10 ⁻¹¹	7.1×10 ⁻⁰⁸	51.5	32.5
P19823	Inter-alpha-trypsin inhibitor heavy chain H2	ттнг	0.19	2.3×10 ⁻¹⁰	1.6×10 ⁻⁰⁷	249.5	206.6
P06396	Gelsolin	GSN	0.25	3.4×10^{-10}	2.25×10^{-07}	135.1	104.3
P04196	Histidine-rich glycoprotein	HRG	0.30	8.9×10^{-10}	$5.3{\times}10^{-07}$	99.4	73.8
P19827	Inter-alpha-trypsin inhibitor heavy chain H1	ITIH1	0.15	7.5×10^{-07}	0.0004	228.9	196.6
P02774	Vitamin D-binding protein	GC	0.17	1.7×10^{-06}	0.0007	165.8	139.3
P10643	Complement component C7	C7	0.36	1.8×10-05	0.007	34.9	24.4

Cut-off P-value<0.05

Table 3

DEPs associated with the most significantly enriched canonical pathways

Name of canonical pathways	Associated DEPs	Overlap Ratio§	P-value*
LXR/RXR activation	AHSG, APOA4, APOB, C4A/C4B GC,HPR,ITIH4,PON1, RBP4	9/121 (7.4%)	1.4×10 ⁻¹⁵
FXR/RXR activation	AHSG,APOA4,APOB,C4A/C4B,GC,HPR ITIH4,PON1,RBP4	9/125 (7.2%)	1.9×10 ⁻¹⁵
Acute Phase Response Signaling	AHSG,C4A/C4B,CFB,CRP,HRG,ITIH2, ITIH4,RBP4,SERPING1	9/168 (5.4%)	2.9×10 ⁻¹⁴
Complement system	C4A/C4B,CFB,SERPING1	4/36 (11.1%)	4.6×10 ⁻⁰⁸
Atherosclerosis signaling	APOA4,APOB,PON1, RBP4	4/121 (3.3%)	6.3×10 ⁻⁰⁶

[§]Numerator of ratio represents the number of DEPs in our data set that maps to the canonical pathway; the denominator represents the total number of molecules/proteins that map to the pathway

^{*}P-value obtained by Fisher's exact test