



Proteomic Dysregulations of Infertile PCOS vs. Non PCOS Females via Label-Free Quantitative Mass Spectrometry

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Article Information

Received 25 June 2025

Revised 03 July 2025

Accepted 13 July 2025

Available online 25 March 2026
(early access)

Authors' Contribution

RR and MRM conceptualized designed and supervised the study, provided supervision, and reviewed and revised the manuscript. SF and SV contributed to data acquisition, literature review, and drafting of the initial manuscript. HNK assisted in methodology and data interpretation. SAR performed the statistical analysis and interpreted the results. All authors read and approved the final version of the manuscript.

Key words

Infertility, Mass spectrometry, PCOS, Proteomics, Lipid metabolism, Clotting mechanism

ABSTRACT

Infertility presents as a complex and multifaceted challenge, characterized by inability to conceive, even after a year of engaging in regular and unprotected intercourse. The current study aims to investigate the underlying molecular mechanism of infertility related to PCOS and non-PCOS by analyzing the differential protein expression of the subgroups of infertile women compared to fertile healthy control (HC) group. Serum samples from 90 subjects were divided into three groups: infertile females with PCOS (n = 30), infertile females without PCOS (n = 30), and fertile healthy controls (n = 30). NanoLC-MS/MS analysis revealed significant differences in protein expression profiles between PCOS, non-PCOS and control groups. A total of 283 proteins were identified, with 51 significantly altered in PCOS and 62 in non-PCOS compared to controls. Thirty proteins showed similar expression patterns across subgroups, while 14 were unique to PCOS and 24 to non-PCOS. Key findings included the downregulation of CL-2 in both PCOS and non-PCOS, upregulation of CL-4 in both groups compared to controls, and unique expression patterns for CL-1, CL-3, and CL-5 of the heat map. Differentially expressed proteins included apolipoproteins, complement proteins, globins, protease inhibitors, and others. Pathway analysis indicated PCOS was associated with dysregulation of the alternative complement pathway and platelet activation, while non-PCOS showed variations in lipid metabolism and insulin-like growth factors. Comprehensive analysis revealed the significant molecular players and pathways implicated in lipid metabolism and clotting mechanisms. These findings offer valuable insights into the pathophysiology of infertility in PCOS and non-PCOS conditions. It may be used to systematically design molecular assays for understanding this disorder and improving its diagnosis and treatment in future.

INTRODUCTION

When a couple remains unable to conceive after a year of consistent unprotected intercourse, it is considered as infertility. It poses significant emotional,

psychological, and financial burden on the individuals and families. According to recent global estimates, 8-12% of couples in their reproductive age range experience infertility, with specific regional variations, such as approximately 21.9% in Pakistan (Vander-Borghet and Wyne, 2018; Ali *et al.*, 2007). The most prevalent aetiologies of infertility including, ovulatory disorders, endometrial polyps and diminished ovarian reserve, constitute significantly to female infertility. Endometriosis can lead to anatomic distortions, such as development of adhesions that can impede the fallopian tubes, thereby hindering tubal function and patency, and is therefore classified as a significant contributor of female infertility (Carson and Kallen, 2021). A considerable number of couples experience infertility where exact cause is unknown or unidentifiable. This is termed as 'unexplained

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0030-9923/2026/0001-0001 \$ 9.00/0



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infertility, as female partner exhibits normal ovulation and the male partner demonstrates normal semen analysis (Wang *et al.*, 2020).

Among ovulatory disorders, polycystic ovarian syndrome (PCOS) stands out as one of the most prevalent and clinically significant conditions, affecting approximately 70% of women with anovulation (irregular ovulation leading to the duration of cycle falling below 21 days or exceeding up to 35 days) (Carson and Kallen, 2021). In the South Asian region, particularly Pakistan, the prevalence rate is up to 52%, when compared to the white population, where it is about 20 - 25% (in the UK) (Azhar *et al.*, 2020). First described by Stein *et al.* (1935), PCOS is considered a complicated hormonal disorder (Sagvekar *et al.*, 2018) characterised by hirsutism, amenorrhea, and large polycystic ovaries (Lujan *et al.*, 2018), predominantly affecting females at child bearing age. The Rotterdam criteria is used for the diagnosis of PCOS, which require the presence of two or three of the following conditions, such as oligo-anovulation, clinical and/or biochemical hyperandrogenism and polycystic ovaries (AFC ≥ 12 and/or ovarian volume >10 ml) (Lauritsen *et al.*, 2014). PCOS is a multifactorial disease, demonstrating complex interplay between predisposition of genetic variants and environmental factors (Carmona-Ruiz *et al.*, 2015), which results in diverse representation of its features across various age and ethnic groups, thereby limiting the true representation of the syndrome (Rashid *et al.*, 2020). The major risk factors of PCOS, includes obesity, dyslipidemia, insulin resistance, hypertension, type 2 diabetes mellitus, and cardiovascular disease (CVD), further complicating its management (Group, 2003; Wilcock and Taylor, 2018; Shi *et al.*, 2021). There is no straight forward genetic test available that can be used to diagnose the condition and its underlying pathophysiological mechanisms (Rani *et al.*, 2023).

Moreover, the proteomic approaches that are non-targeted have the benefit of determining location, abundance, protein-protein interactions and posttranslational modifications (Insenser *et al.*, 2019). Any change in the expression of proteins depicts altered biological processes as proteins control the cellular functions (Khan *et al.*, 2015). Therefore, in contrast to genome-based studies, advanced proteomics techniques are fairly beneficial as they interact first-hand with the functional molecules as compared to genes or mRNA (Abdallah *et al.*, 2012). Moreover, the proteome of an organism undergoes posttranslational modifications and changes in relation to the environment cannot be predicted by the genetic coding (Abdallah *et al.*, 2012). Thus, proteomics serves to further advance the understanding of pathophysiology of the disease (Pandeswari *et al.*, 2019).

Despite the disease burden, molecular abnormalities and underlying pathophysiology of infertility caused by PCOS and non-PCOS have not yet comprehensively been explored. Currently, very limited studies have been conducted on the proteomics of PCOS in Pakistan. Hence, this study was designed to investigate the differential expression of proteins in infertile PCOS and non-PCOS groups as compared to fertile HC group by using nano LC-MS/MS mass-spectrometry. By exploring the shared differentially expressed proteins of PCOS and non-PCOS in comparison to HC, the study aims to elucidate common molecular pathways underlying infertility, providing valuable insights into the pathophysiology of PCOS-related infertility. Whereas unique differentially expressed proteins may offer clues to distinct mechanisms contributing to infertility in PCOS versus non-PCOS individuals, shedding light on the heterogeneity of infertility and potentially identifying novel diagnostic markers or therapeutic targets.

MATERIALS AND METHODS

Subject selection

A case control study was designed following the approval from Ethical Review Committee (ERC AKU 2022-7509-22501) of Aga Khan University, Karachi. Participants were recruited from the Endocrinology Clinic of the same hospital, where after obtaining an informed consent, socio-demographic data including age, weight, height, and brief medical history were recorded in order to classify fertile and infertile subjects. The subjects were categorized into three groups: healthy fertile females as healthy controls, and primary infertile females with and without PCOS as cases. The control group consisted of females matched in age, gender, and BMI with the cases, having a child aged less than 2-3 years, showing no signs of disease, and not using oral contraceptives. For the case groups, primary infertile females aged 20-40 years diagnosed with PCOS using the Rotterdam Criteria were included, regardless of ethnic background. Each group comprised 30 subjects. Exclusion criteria applied to those with secondary infertility, recent hormonal therapy, BMI < 18 or > 30 Kg/m², or diagnosed with other relevant conditions.

Sample size calculation

$$n = \frac{Z_{\alpha/2}^2 pq}{e^2}$$

N= Sample size; α = Confidence Interval = 95%; p= Prevalence= 22% (Sami and Ali, 2012); q= 1 – p; e= Error Margin= 5% = 0.05

Overall sample

$$n = \frac{(1.96)^2 (0.22)(0.78)}{(0.05)^2} = 264$$

Proportionate into fertile and infertile groups: Infertility 21.9% = 22%; $P_{\text{inf (pco)}} = 52\%$ (Amera-Tariq *et al.*, 2021); $n_{\text{inf}} = n * P_{\text{inf}} = 264 * 22\% = 58 \approx 60$; $n_{\text{fer}} = n - n_{\text{inf}} = 264 - 60 = 204$; $n_{\text{inf (pco)}} = n * P_{\text{inf (pco)}} = 60 * 52\% = 31 \approx 30$; $n_{\text{fer (nonpco)}} = n_{\text{inf}} - n_{\text{inf (pco)}} = 60 - 30 = 30$

Calculated sample size for control group came out to be 204, while for both the cases it came out be 30 each. However, due to time constraint and limitation of resources, the sample number of control group was reduced to 30 as well.

Peripheral blood collection

Blood samples (4 mL) from each study subject were collected in gel-coated vacutainers tubes, on the 2nd day of menstrual cycle. All the samples were centrifuged at 4000 rpm, and sera were transferred into the separate sterile 1.5 mL Eppendorf tubes and stored at -80°C for proteomic analysis (Gayosso-Gómez *et al.*, 2014).

Clinical and biochemical data

Radioimmunoassay was performed using an automated chemiluminescence analyzer (Abbott, USA) to evaluate the hormonal profile, encompassing follicle stimulating hormone (FSH), luteinizing hormone (LH), anti mullerian hormone (AMH), prolactin (PRL), and estradiol (E2) levels.

Mass spectrometric based proteomics

For proteomic analysis, serum samples (10 µL) from each participant were pooled based on their respective case and control groups into three separate tubes labelled as Control, PCOS, and Non-PCOS. Protein estimation of each pooled sample was measured using the Invitrogen Qubit™ Protein Assay Kit and Qubit™ 2.0 Fluorometer (Nowak *et al.*, 2021). For tryptic digestion, previously optimized protocol was followed with some modifications (Zafar *et al.*, 2021). 200 µg of protein from each sample group was mixed with 160 µL of 1M NH₄HCO₃. Following denaturation with 100 µL of 45 mM DT and alkylation with 100 µL of 100 mM iodoacetamide, 10 µg of trypsin was introduced for protein digestion. Desalting was performed using custom-made tips containing C18-resin. Peptides were eluted sequentially into Protein LoBind tubes by increasing the concentration of ACN from 30% to 70% along with 0.1% TFA. Peptide quantification was performed by Qubit™ Protein Assay Kit (Nowak *et al.*, 2021). The peptide mixture (20 µg) was applied to nano-LC-MS/MS using an Orbitrap Q-Exactive HF-X mass

spectrometer (Thermo Fisher Scientific, USA) which was connected to an EASY-LC 1000 system, as previously described in protocol (Beltran-Camacho *et al.*, 2020).

Preprocessing of proteomics data

For identification of proteins, raw files generated from Orbitrap Q-Exactive HF-X were analyzed on MaxQuant (version 2.3.2, Matrix Science, UK) in conjunction with the Andromeda search engine (Cox *et al.*, 2011). False discovery rate (FDR) was set at 1% and spectra were matched against the proteins of Homo sapiens present in UniProt/Swiss-Prot database (Zafar *et al.*, 2021). The quantitative proteomic data analysis was conducted using Perseus (v.1.6.10.50). Multiple filtration steps present in Perseus, were applied, and significant proteins were obtained using student's t test. Benjamini and Hochberg method were used for correction of p value for t-test testing with 10% FDR cutoff (Green *et al.*, 2007). Multi scatter plot, profile plot, volcano plot and heat map were obtained in order to visualize the variability of protein expression of all groups.

Functional analysis of proteomics data

The PANTHER (protein analysis through evolutionary relationships) database Version 17.0 (Enli *et al.*, 2013; Thomas *et al.*, 2022) was used to identify classes of common proteins and their proportions which was visualized using a pie chart. For pathway overrepresentation analysis, Gene Ontology database ShinyGO 0.76 (sdstate.edu) was used (Ashburner *et al.*, 2000; Consortium *et al.*, 2023). The p-value of at least 0.05 was considered as significant with FDR correction. String version 12.0 was used to elucidate protein-protein interactions and critical analysis of physical and functional association (<http://string-db.org/>) (Szkarczyk *et al.*, 2023). The KEGG (Kyoto encyclopedia of genes and genomes) pathway maps were obtained at 0.05 FDR cut off for up-regulated and down-regulated proteins in both subgroups groups in order to gain insight into the role of protein in the biological system (Ge *et al.*, 2019; Kanehisa *et al.*, 2020; Luo *et al.*, 2013).

Statistical analysis

For all the continuous variables, data were presented as mean ± standard deviation. ANOVA was performed using SPSS (IBM SPSS 21) statistical software. P-value ≤ 0.05 considered as significant.

RESULTS*Demographics and biochemical analysis*

Overall work flow of the study has been shown in Figure 1. Initially, anthropometric parameters and

biochemical profile of the subjects were compared among subgroups of infertility in comparison to healthy control fertile group. The age and BMI of these females and the hormonal profile including FSH, LH, prolactin, estradiol and AMH were also recorded in all groups as shown in Table I. The results showed that neither age nor BMI was statistically significant in comparison of infertile cases to healthy fertile control group. The hormonal profiles showed significant variation among the subgroups of infertility. In comparison of PCOS vs HC, FSH levels showed no

significant difference, whereas LH hormone levels were notably elevated, and significant reductions were observed in the mean levels of prolactin and estradiol. Conversely, when comparing the non-PCOS subgroup to the HC group, there were distinct mean reductions in the levels of LH, prolactin, and estradiol. Moreover, a comparison between the PCOS and non-PCOS subgroups revealed distinct mean elevation in LH and reductions in prolactin and estradiol levels, as delineated in Table I. AMH levels did not show any significant difference among the groups.

Table I. Bio demographical characteristics, Biochemical profile of fertile and infertile female subgroups (PCOS and non-PCOS).

Variable	HC (N=30)	PCOS (N=30)	Non-PCOS (N=30)	p value
Age	32±3.63	32.13±4.23	31.7±5.12	0.92
BMI	23.3±1.27	22.78±2.15	22.74±2.09	0.40
FSH (unit)	9.2±11.10	9.5±2.86	6.8±2.20 ^{ss}	^{ss}
LH (IU/L)	9.2±8.98	10.9±13.08 ^{***}	5.1±1.99 ^{\$\$}	^{***} ^{\$\$}
Prolactin (µg/L)	19.9±15.2	14.7±9.25 ^{##}	13.1±7.30 ^{\$\$}	^{##} ^{\$\$}
Estradiol (pmol/L)	189.5±131.29	71.7±45.04 ^{***}	77.5±64.77 ^{\$\$}	^{***} ^{\$\$}
AMH (ng/ml)	2.8±2.38	3.3±3.28	2.2±1.28	-

FSH, follicular stimulating hormone; LH, luteinizing hormone; AMH, anti mullerian hormone; HC, healthy control; PCOS, polycystic ovary syndrome; BMI, body mass index; ^{***}, showing significant difference between PCOS and HC group. ^{ss} showing significant difference between non-PCOS and HC group. ^{##} showing significant difference between PCOS and non-PCOS group.

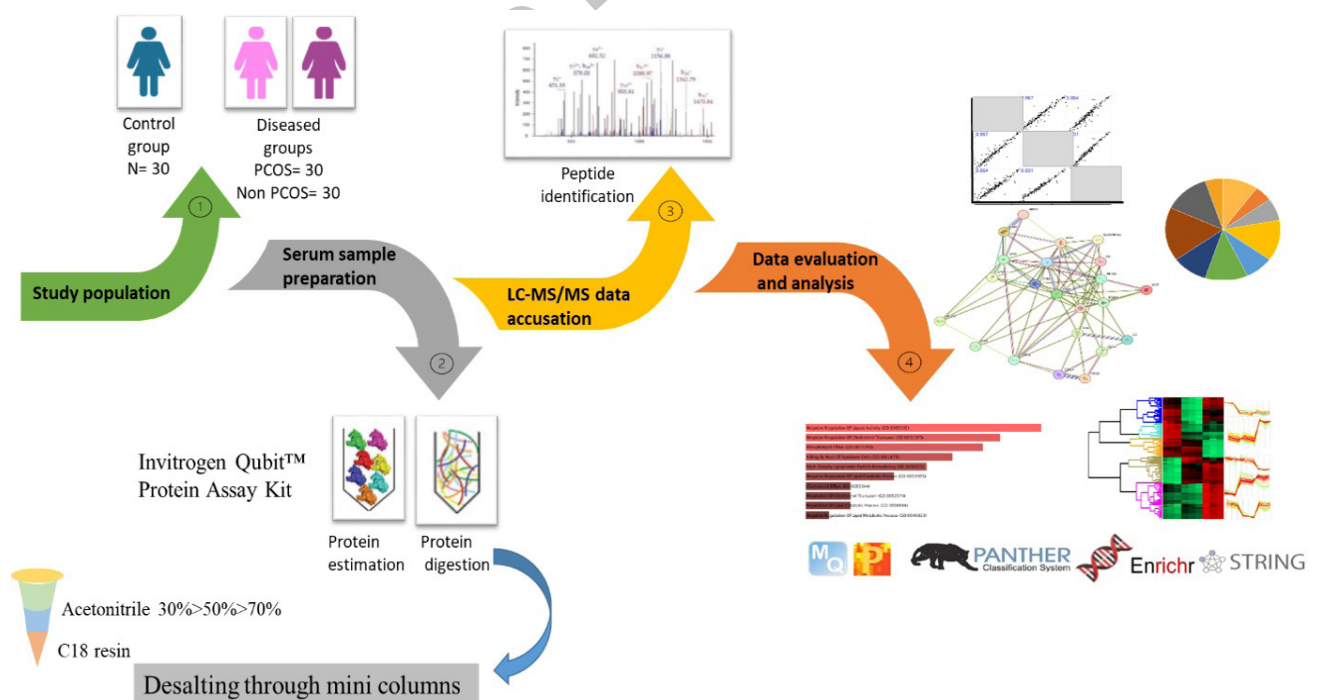


Fig. 1. Diagrammatic representation of the study.

Identification and qualitative assessment of proteins

A total number of 283 proteins were identified using MaxQuant (Fig. 2A). Complete list of identified proteins is provided in Electronic supplementary file [Supplementary Table SI](#). Multiple filtration steps were applied in Perseus (v.1.6.10.50) including, only identified by site, reversed and, removing contaminants to identify the most authentic proteins of interest (Fig. 2A). The total number of proteins in HC group was 168, whereas 176 total proteins were present in each PCOS and Non-PCOS groups. Among these 176 proteins, 10 proteins were exclusively present in infertile PCOS and non-PCOS subgroups (Electronic [Supplementary Table SII](#)), while 2 proteins were exclusive in HC. A total of 164 proteins were common in all infertility (PCOS and non-PCOS) and HC group (Fig. 2B). The variation between the infertile (PCOS and non-PCOS) and HC groups was evaluated by using multi-scatter plot analysis correlation coefficient (Fig. 2C), while profile plot showed an overall quantitative trend of proteins, with each line representing a unique protein (Fig. 2D).

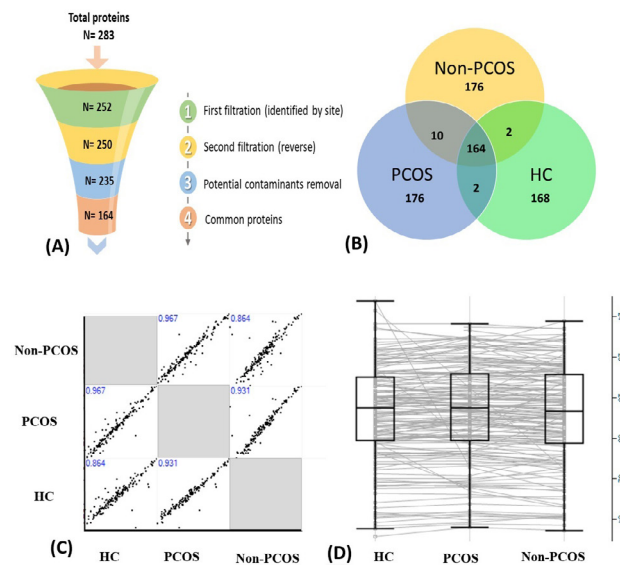


Fig. 2. Graphic representation of qualitative proteomic data analysis. (A) Filtration procedure implemented on the identified and quantified proteins. (B) Venn diagram showing proteins distribution in each group. Healthy control (HC), polycystic ovary syndrome (PCOS) and Non-PCOS. (C) Analysis of variability of proteins between the groups as shown by multi scatter plot. (D) Profile plot of proteomic data between study groups.

Quantitative proteomic data analysis

From the 283 identified proteins, 51 proteins were found to be significantly differentially expressed proteins between PCOS vs HC group. Among these, 18 proteins

exhibited up-regulation, while 33 proteins showed down-regulation. Conversely, in the non-PCOS vs HC group, we observed 62 significantly differentially expressed proteins, with 15 proteins displaying up-regulation and 47 proteins showing down-regulation. The pattern of protein expression in infertile groups was found to be consistent for about 30 proteins. Specifically, 13 proteins exhibited down-regulation and 17 showed up-regulation in both the infertility subgroups, as depicted in (Fig. 3A) and detailed in ([Supplementary Table SIII](#)). Additionally, 24 and 14 proteins were found to be exclusive identified in the non-PCOS vs HC group and PCOS vs HC group, respectively. To further understand the expression profiles, a heat map was generated based on z-score values of LFQ intensities, illustrating differential expression in both infertility subgroups (Fig. 3B). Proteins present in each cluster of heat map are listed in ([Supplementary Table SIV](#)). Clustering analysis identified five main clusters of differentially expressed proteins. Specifically, CL-1 was exclusively down-regulated in PCOS as compared to HC group, CL-2 was downregulated in both PCOS and non-PCOS group when compared to HC group, CL-3 was exclusively and significantly down-regulated in non-PCOS as compared to HC group, CL-4 displayed up-regulation in both the PCOS and non-PCOS when compared to HC group, and CL-5 was exclusively and significantly up-regulated in non-PCOS when compared to HC (Fig. 3C). The protein categories of differentially expressed proteins of PCOS vs HC group are apolipoproteins, complement component, globins, protease inhibitor, protein-binding activity modulator, serine protease and transfer/carrier proteins (Fig. 3D). The protein categories of differentially expressed proteins of non-PCOS vs HC group comprise of apolipoproteins, complement component, globins, protease inhibitor, protein-binding activity modulator, serine protease and transfer/carrier proteins and proteases (Fig. 3E). Overall both the groups revealed similar protein classes with different percentages except proteases that were only observed in non-PCOS group.

Functional analysis

To elucidate the roles of the significantly differentially expressed proteins common in PCOS vs non PCOS group, GO functional enrichment analysis was performed for upregulated and downregulated proteins respectively. Complement activation and immunoglobulin mediated immune response were the enriched biological processes by upregulated proteins while downregulated proteins were involved in negative regulation of very-low-density lipoprotein particle clearance, chylomicron remnant clearance, negative regulation of cholesterol transport and lipoprotein lipase activity, nitric oxide transport, negative

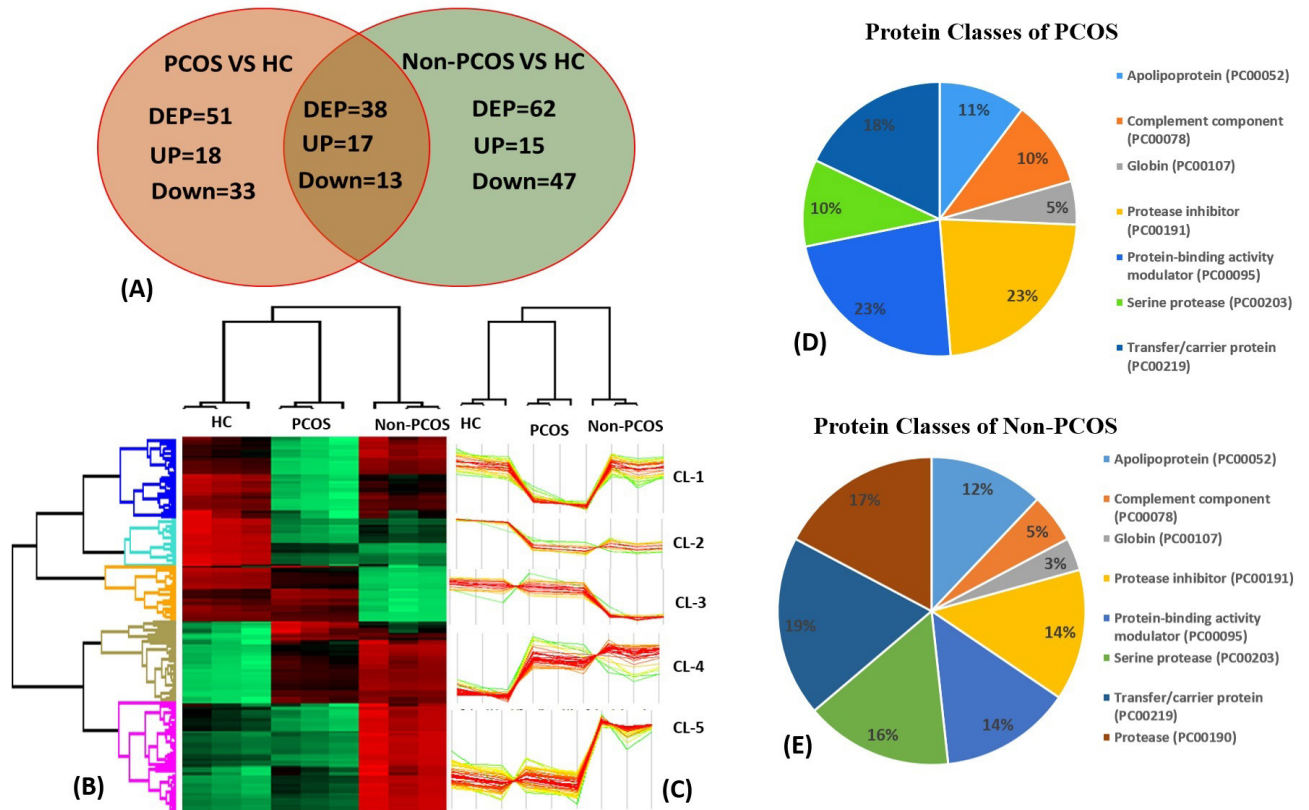


Fig. 3. Quantitative analysis of proteins in subgroups (PCOS, and non-PCOS) of infertility compared to HC. (A) Statistical data analysis revealing the differentially expressed proteins (DEPs), significantly up-regulated and down-regulated in both subgroups of infertility as well as in individual groups. (B) Heat map analysis of significant proteins in all groups. (C) Cluster analysis of significant proteins in all groups. (D) Protein classes of differentially expressed proteins of PCOS (E) Protein classes of differentially expressed proteins of non-PCOS.

GO enriched biological process by upregulated proteins (p value <0.001)	GO enriched biological process by upregulated proteins (p value <0.001)	GO enriched biological process by upregulated proteins (p value <0.001)
Complement activation	Nitric oxide transport	Regulation of high-density lipoprotein particle clearance
Immunoglobulin mediated immune response	Complement activation, classical pathway	Negative regulation of very-low-density lipoprotein particle clearance
GO enriched biological process by downregulated proteins (p value <0.001)	Humoral immune response mediated by circulating immunoglobulin	Nitric oxide transport
Negative regulation of very-low-density lipoprotein particle clearance	GO enriched biological process by downregulated proteins (p value <0.001)	Negative regulation of triglyceride catabolic process
Chylomicron remnant clearance	Negative regulation of complement activation, lectin pathway	Chylomicron remnant clearance
Negative regulation of cholesterol transport	Negative regulation of very-low-density lipoprotein particle clearance	Regulation of very-low-density lipoprotein particle remodelling
Negative regulation of lipoprotein lipase activity	Chylomicron remnant clearance	GO enriched biological process by downregulated proteins (p value <0.001)
Nitric oxide transport	Regulation of very-low-density lipoprotein particle remodelling	Complement activation
Negative regulation of triglyceride catabolic process	High-density lipoprotein particle clearance	Positive regulation of phagocytosis
High-density lipoprotein particle clearance	Regulation of phospholipid catabolic process	
Positive regulation of apoptotic cell clearance	Phospholipid efflux	
Reverse cholesterol transport	Fibrinolysis	

Fig. 4. Biological process enrichment of differentially expressed proteins common in (A) PCOS/Non PCOS group, (B) PCOS/HC group and (C) Non PCOS/HC group. Upregulated (blue) and downregulated (pink) gene ontology annotations. PCOS, polycystic ovary syndrome; HC, healthy control.

regulation of triglyceride catabolic process, high-density lipoprotein particle clearance, and reverse cholesterol transport (Fig. 4A)

Moreover, GO enriched biological functions of PCOS vs HC upregulated proteins revealed that these proteins were linked to nitric oxide transport, complement activation (classical pathway) and humoral immune response mediated by circulating immunoglobulin. The proteins that showed a decreased differential expression were involved in negative regulation of complement activation and negative regulation of very-low-density lipoprotein particle clearance, chylomicron remnant clearance, regulation of very-low-density lipoprotein particle remodelling, high-density lipoprotein particle clearance, regulation of phospholipid catabolic process, phospholipid efflux and fibrinolysis (Fig. 4B).

Conversely, GO enriched biological functions of non PCOS vs HC showed that the upregulated proteins were involved in regulation of high-density lipoprotein particle clearance, negative regulation of very-low-density lipoprotein particle clearance, nitric oxide transport, negative regulation of triglyceride catabolic process, chylomicron remnant clearance and regulation of very-

low-density lipoprotein particle remodelling while the downregulated proteins were involved in complement activation and positive regulation of phagocytosis (Fig. 4C).

Pathway over representation analysis was also conducted on both common and unique proteins associated with infertility, aiming to uncover common and unique disrupted pathways in both study groups (PCOS and non-PCOS). Shared DEPs unveiled significant dysregulation in the coagulation pathways, alongside disturbances in the classical complement system as displayed in (Fig. 5A). Protein-protein interaction (PPI) analysis conducted through STRING also demonstrated a notable association among all common proteins. Subsequently, K-means clustering was employed to classify the proteins into distinct groups. This analysis revealed that 18 out of 30 common proteins were found to be associated with the complement and coagulation pathway (red nodes). The second cluster predominantly exhibited connections to lipid metabolism (dysregulated apolipoproteins) (green nodes), while the third cluster differentiated into two proteins associated with steroid hormone binding proteins (blue nodes) (Fig. 5B).

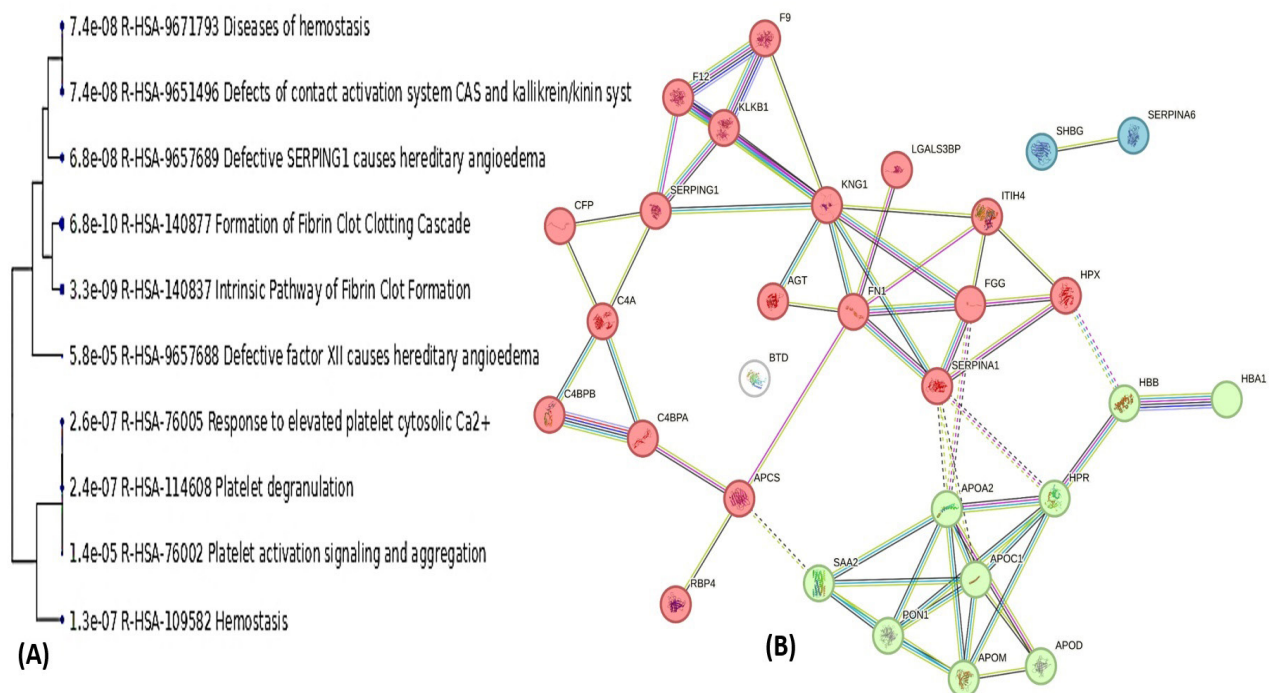


Fig. 5. Functional analysis of common proteins of infertility. Showing similar trend of gene expression in both subgroups of infertility (PCOS, and non-PCOS). (A) Pathway analysis of shared DEPs in both subgroups (PCOS, and non-PCOS) of infertility. (B) PPI network analysis of shared DEPs along with K-means clustering. Colours representation: red nodes= complement and coagulation pathway; green nodes= lipid metabolism; blue nodes= steroid hormone binding proteins.

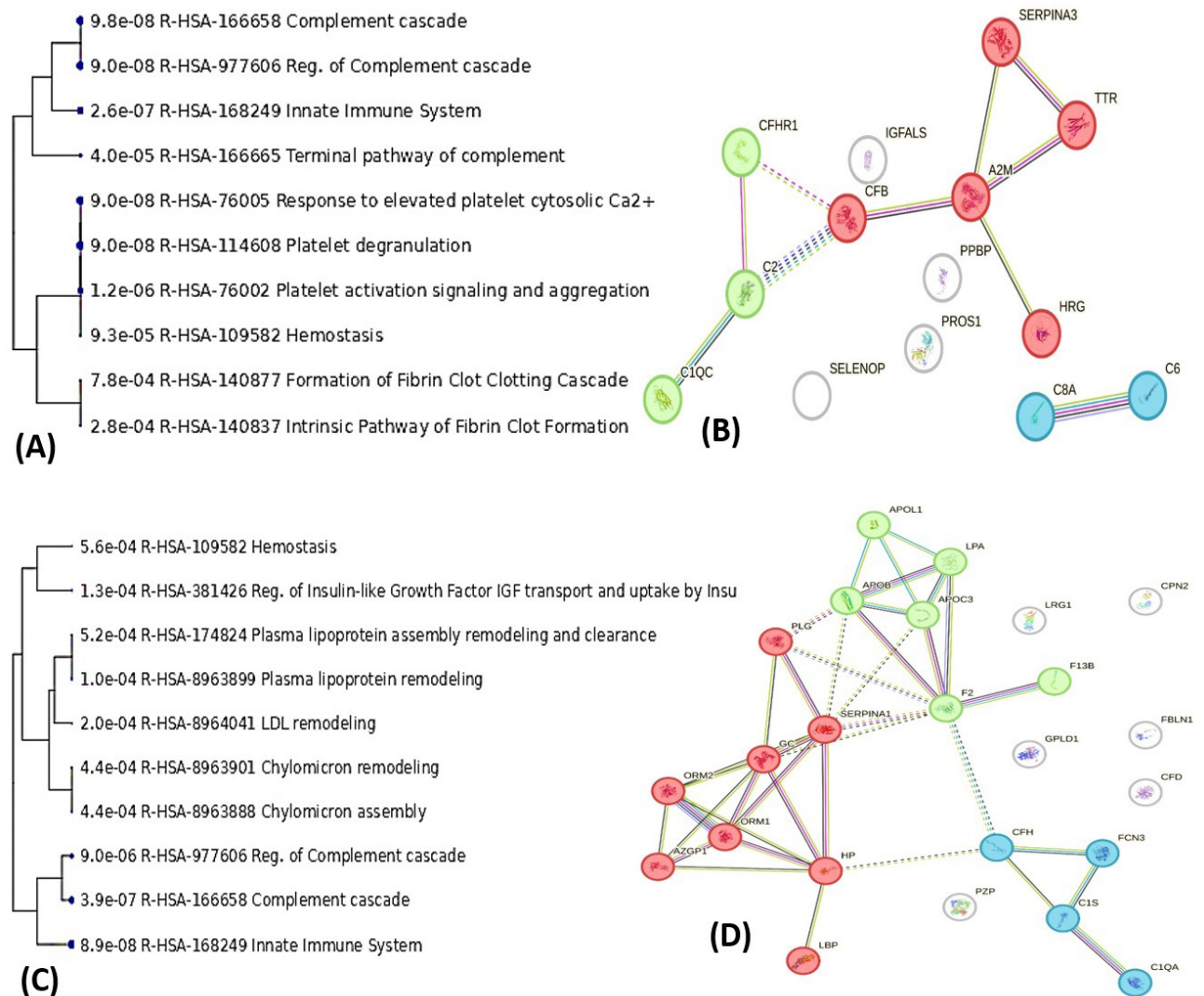


Fig. 6. Functional analysis of unique proteins of infertility. (A) Pathway analysis of exclusively DEPs of PCOS vs HC. (B) PPI network analysis of exclusively DEPs of PCOS along with K-means clustering. Colours representation: red nodes= complement activation; green nodes= cytolysis; blue nodes= blood micro particle proteins. (C) Pathway analysis of exclusively DEPs of non-PCOS vs HC. (D) PPI network analysis of exclusively DEPs along with K-means clustering. Colours representation: red nodes= acute phase response; blue nodes= immune response; green nodes= lipid metabolism proteins.

The pathway analysis of unique DEPs associated with PCOS indicated dysregulation in the whole complement cascade, and disruptions in platelet aggregation and degranulation (Fig. 6A). Additionally, the PPI network analysis highlighted strong connections of most of the DEPs as shown in (Fig. 6B), while clustering analysis categorized these proteins into functional groups, such as complement activation (red nodes), cytolysis (green nodes), and anti-inflammatory (blue nodes) proteins. The pathway analysis of unique DEPs associated with non-PCOS condition showed more pronounced disturbances

in apolipoproteins associated with the dysregulation of lipid metabolism (Fig. 6C), while PPI network analysis highlighted strong connections of the DEPs, acute phase response (red node), immune response (blue node) lipid metabolism (green node) and proteins (Fig. 6D).

The most significant pathway of both subgroups of infertility are complement and coagulation pathway, KEGG map of up-, and down-regulated proteins in both subgroups highlighted in green and red colours as shown in (Supplementary Fig. 1).

DISCUSSION

Infertility is one of the most prevalent chronic health disorders involving young adults all around the globe (Smith *et al.*, 2003). In pursuit of a comprehensive understanding of the pathology, clinical and experimental studies have been performed associated with the female reproductive health, environment and lifestyle (Bala *et al.*, 2021). Infertility associated with PCOS and non-PCOS is considered to be the complex disease rather than merely a gynaecological problem. Hence, in order to identify the molecular mechanism linked to pathophysiology of this disease, a comprehensive proteomic profile of the infertile female patients (PCOS and non-PCOS) in comparison with fertile females was performed using Nano-LC MS/MS.

Demographic results showed no significant difference in age and BMI as our samples were matched, however, the biochemical profile showed a significant difference in the levels of various hormones routinely tested for the diagnosis of the infertility. The levels of LH were significantly higher in PCOS as compared to the fertile group which is consistent with the previous findings (Yang and Chen, 2024; Mahmud *et al.*, 2022) and represents a neuroendocrine disturbance in the subjects (Pratama *et al.*, 2024). Similarly, in the non-PCOS group, both the FSH and LH levels were significantly reduced as compared to the fertile group showing hormonal disturbances. Furthermore, Estradiol and Prolactin levels were significantly decreased in both PCOS and Non-PCOS infertile groups in contrast to the fertile group. Reduced Estradiol concentration is considered to be the result of endogenous inhibitors like 5 alpha-androstane-3, 17-dione within PCOS follicular fluid, which hampers the synthesis of estradiol (Agarwal *et al.*, 1996). Contrary to our findings, prolactin has been reported to be increased in cases of PCOS and hyperprolactinemia is considered as a well-established cause of anovulatory infertility (Falaschi *et al.*, 1977; Bernard *et al.*, 2015; Melmed *et al.*, 2011). Though, some of the findings reported lower levels of prolactin in PCOS women as well, which is negatively correlated with LH and positively associated with Estradiol levels (Yang *et al.*, 2020; Glintborg *et al.*, 2014). Lastly, AMH levels did not show any significant difference between healthy, infertile PCOS and infertile non-PCOS groups though it is considered to be higher in PCOS females and serves as an established biomarker for assessing female reproductive capacity (Rudnicka *et al.*, 2021; Ran *et al.*, 2021).

Followed by biochemical analysis, proteomic analysis was performed which revealed 30 common proteins among different subgroups of infertility that showed uniform pattern of gene expression, however,

among these, 8 proteins revealed differential expression in both infertility subgroups. Among exclusively identified proteins in PCOS group, 14 proteins showed significant differential expression and non-PCOS group showed differential expression of 24 proteins.

The common proteins significantly altered in both diseased groups, were involved in complement and coagulation cascade pathways. These proteins include fibrinogen, coagulation factor 9, complement factor B, complement factor H, C2 and C4A. Fibrinogen exhibited the most statistically significant log-fold change within both infertility groups with an increase of up to 1.3 times in PCOS and up to 1.4 times in non-PCOS group. This finding is consistent with the previous findings of PCOS in younger and obese individuals (Ji *et al.*, 2015). Fibrinogen is associated with an elevated risk for cardiovascular disease which is a long term complication of PCOS (Wang *et al.*, 2017) and non-PCOS infertile women as well (Mu *et al.*, 2016; Okoth *et al.*, 2020). Moreover, it has also been suggested as a potential laboratory screening biomarker for PCOS, however, other research has questioned its status as such (Ji *et al.*, 2015; Ozgokce *et al.*, 2020).

Another protein, Coagulation factor 9, was found to be significantly up-regulated in both the infertile groups. This protein is linked to greater risk of venous thromboembolisms (Lagrange *et al.*, 2019). Moreover, blood coagulation is crucial in the interaction of embryo and endometrium (Guo *et al.*, 2021), and coagulation defects result in impaired implantation (Gerotziakas *et al.*, 2017). However, increased coagulability was documented to be less likely in these non-obese PCOS women (Moin *et al.*, 2023). Hence, we suggest that this factor could also be associated with the cause of infertility in PCOS and non-PCOS subjects. The complement system has three alternative routes which merge to a common pathway. It is one of the main component of innate immunity which is involved in inflammatory and immune response. The classical pathway is activated through C4 (Carroll, 2004) which was found to be up-regulated in both the infertile PCOS and non-PCOS groups. As opposed to classical pathway, the lectin pathway activates through C2 (Carroll, 2004) which exhibited down-regulation, as low as 0.98 times log-fold change, exclusively in the PCOS group. Complement activation plays a role in various disease processes, such as diabetes (Engström *et al.*, 2005) and cardiovascular disease (Bjerre *et al.*, 2008), both of which are linked to PCOS (Torchen, 2017; Arffman *et al.*, 2019) as well as non-PCOS related infertility (Gleason *et al.*, 2019; Tobias *et al.*, 2015). Moreover, mounting evidence suggests that PCOS is associated with an elevated risk of infection and its consequences (Kyrou *et al.*, 2020). It has also been reported that one of the classical pathways

leads to infertility by the formation of membrane attack complex (MAC) that perforates the spermatozoa leading to its destruction (Msc and Makki, 2010). Though we did not take account for antisperm antibodies which activate this pathway. Another complement protein, factor H was significantly up-regulated exclusively in non-PCOS infertile group, which is an inhibitor of complement pathway (Parente *et al.*, 2017).

Among the lipid metabolism proteins, several apolipoproteins were significantly differentially expressed. ApoA-II was found to be down-regulated in both of the infertile groups. Koike *et al.* (2021) proved the atheroprotective attribute of apoA-II in rabbits, which was substantiated by the augmentation of cholesterol efflux activity and the anti-inflammatory capabilities of HDL particles. Moreover, the risk of coronary artery disease was also found to be inversely associated with APOA-II levels in human subjects (Florea *et al.*, 2022). Hence down-regulation of ApoA-II demonstrates the increased cardiovascular risk of both infertile groups. Furthermore, Apo D and Apo M proteins were significantly up-regulated in both the infertile groups while Apo B was exclusively up-regulated in non-PCOS group, by up to 1.01 times log-fold change. Apo B and ApoM have been reported to be potential contributors to the atherosclerosis (Hahn *et al.*, 1995), diabetes and coronary artery disease (Butler *et al.*, 2023; Zannis *et al.*, 2015). Though, a cohort study revealed that the levels of Apo M were significantly reduced in the PCOS group (Butler *et al.*, 2023), which is in contrast to our findings.

Among binding proteins, sex hormone binding globulin (SHBG) was found to be down-regulated in both the infertile groups, as low as 0.97 log-fold change. Alterations in the SHBG and sex hormone levels have been linked with hypo/hyperthyroidism and in turn thyroid disorders have been linked to reduced fertility (Krassas *et al.*, 2010). Reduced SHBG levels cause an increase in androgen bioavailability, which causes anovulation and the clinical traits of PCOS to progress (Qu and Donnelly, 2020). However, evidence shows that liraglutide can reverse these symptoms (Nylander *et al.*, 2017). Another protein, S100A9 was observed to exhibit lower expression levels, as low as 0.95 log-fold change in the infertile PCOS group while higher expression, as much as 1.01 times log-fold change in the non-PCOS infertile group. This protein has not been reported in association with PCOS before, but it plays a crucial role in embryo implantation by binding with calcium ions and have been regarded as potential candidate marker for assessing endometrial receptivity (Sadigh *et al.*, 2019). The impact of this protein on adverse pregnancy outcomes is also being investigated (Verma *et al.*, 2018).

Among serine protease inhibitors, alpha1-antitrypsin (SERPINA1) was found to be up-regulated in both the infertile groups which is consistent with the previous findings in peritoneal fluid of infertile patients (Ferrero *et al.*, 2009). This protein has been suggested to potentially alleviate PCOS symptoms by inhibiting pro-inflammatory factors through direct interactions (Pan *et al.*, 2022). Moreover, SERPINA3 was down-regulated exclusively in the PCOS group as low as 0.96 times log-fold change. This protein has also not been associated with PCOS in previous findings but has been known to limit coagulation and inflammation by inhibiting Cathepsin G (Chelbi *et al.*, 2012), hence, the downregulation of this protein could likely be the reason of pronounced inflammation in PCOS patients.

In future, ELISA experiments can be carried out for further validation of these proteins. The limitations of our study included smaller sample size which could compromise the statistical significance of the study and restrict the generalizability. Moreover, it was not possible to match the lifestyle of case and control groups adding to confounders in the study.

To conclude, this study revealed significant changes in protein expression in both the infertile PCOS and non-PCOS groups utilizing the mass spectrometry technique in comparison with healthy controls. The shared and exclusive DEPs among different subgroups of infertility shed light on the diverse biological process, including lipid metabolism, complement and coagulation cascade, inflammation and reproductive functions.. The knowledge acquired through this research has the potential advantage to lay the foundation for personalized medicine strategies. Tailored treatment designed to address the precise molecular anomalies identified in each patient, may result in more effective and personalised therapeutic approaches.

DECLARATIONS

Acknowledgement

We would like to greatly acknowledge Dr. Panjwani Center for Molecular Medicine and Drug Research (ICCBS) for supporting this work by providing recurring grants for chemicals and consumables. The Villum Center for Bioanalytical Sciences at University of Southern Denmark (SDU), Odense, Denmark is acknowledged for access to advanced LC-MS/MS instrumentation. Our special thanks to department of Biological and Biomedical Sciences, Aga Khan University, Dr. Arfa Azhar for help in the recruitment of female subjects and Mussarat Ashraf for supporting in data analysis.

Grants

University Research Council Students Support Grant: 231026 and Department of Biological and Biomedical Sciences; MPhil students.

Compliance with ethical standards

All procedures performed in this study involving human participants were approved by Ethical Review Committee (ERC AKU 2022-7509-22501) of Aga Khan University, Karachi.

Consent to participate

Participants were recruited from the Endocrinology Clinic of the same hospital, after obtaining an informed consent from each participant.

Consent for publication

The author confirms that the work described has not been published before (except in the form of thesis) and it is not under consideration for publication elsewhere.

Availability of data and material

Data is available upon request.

Generative AI and AI-assisted technology statement

The authors declare that no generative AI and AI assisted technology was used in the creation of this manuscript.

Supplementary material

There is supplementary material associated with this article. Access the material online at: <https://dx.doi.org/10.17582/journal.pjz/20250625050207>

Statement of conflict of interest

The authors have declared no conflict of interest.

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Supplementary Material

Proteomic Dysregulations of Infertile PCOS vs. Non PCOS Females via Label-Free Quantitative Mass Spectrometry

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Supplementary Table I. List of all proteins identified by MaxQuant.

Majority protein IDs	Gene names	Protein names
A0A0G2JRN3	<i>SERPINA1</i>	Alpha-1 antitrypsin
P02766	<i>TTR</i>	Transthyretin
P05155	<i>SERPING1</i>	Plasma protease C1 inhibitor
E7ET33	<i>ITIH3</i>	Inter-alpha-trypsin inhibitor heavy chain H3
P08519	<i>LPA</i>	Apolipoprotein(a)
P01011	<i>SERPINA3</i>	Alpha-1-antichymotrypsin
P35542	<i>SAA2-SAA4</i>	Serum amyloid A-4 protein
P01857	<i>IGHG1</i>	Ig gamma-1 chain C region
P01880-2	<i>IGHD</i>	Ig delta chain C region
P02746	<i>CIQB</i>	Complement C1q subcomponent subunit B
P0CG04	<i>IGLC1</i>	Ig lambda-1 chain C regions
P00738	<i>HP</i>	Haptoglobin
P04278	<i>SHBG</i>	Sex hormone-binding globulin
P08603	<i>CFH</i>	Complement factor H
P00751	<i>CFB</i>	Complement factor B

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0030-9923/2026/0001-0001 \$ 9.00/0



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Majority protein IDs	Gene names	Protein names
P06681	<i>C2</i>	Complement C2
A0A0G2JMB2	<i>IGHA2</i>	Immunoglobulin heavy constant alpha 2
P08697	<i>SERPINF2</i>	Alpha-2-antiplasmin
A0A140TA49	<i>C4A</i>	Complement C4-A
A0A0G2JRQ6		Ig-like domain-containing protein
P68871	<i>HBB</i>	Hemoglobin subunit beta
P07225	<i>PROS1</i>	Vitamin K-dependent protein S
P49908	<i>SEPP1</i>	Selenoprotein P
P01860	<i>IGHG3</i>	Ig gamma-3 chain C region
P01876	<i>IGHA1</i>	Ig alpha-1 chain C region
P01859	<i>IGHG2</i>	Ig gamma-2 chain C region
P01861	<i>IGHG4</i>	Ig gamma-4 chain C region
P05156	<i>CFI</i>	Complement factor I
P68133	<i>ACTA1</i>	Actin
P69905	<i>HBA1</i>	Hemoglobin subunit alpha
P05160	<i>F13B</i>	Coagulation factor XIII B chain
P00736	<i>C1R</i>	Complement C1r subcomponent
P07360	<i>C8G</i>	Complement component C8 gamma chain
A0M8Q6	<i>IGLC7</i>	Ig lambda-7 chain C region
P04114	<i>APOB</i>	Apolipoprotein B-100
P01019	<i>AGT</i>	Angiotensin 1-4
P02775	<i>PPBP</i>	Platelet basic protein
P00747	<i>PLG</i>	Plasminogen
P04003	<i>C4BPA</i>	C4b-binding protein alpha chain
P17936-2	<i>IGFBP3</i>	Insulin-like growth factor-binding protein 3
P02656	<i>APOC3</i>	Apolipoprotein C-III
O14791	<i>APOL1</i>	Apolipoprotein L1
P23142	<i>FBLN1</i>	Fibulin-1
B1AKG0	<i>CFHR1</i>	Complement factor H-related protein 1
P10643	<i>C7</i>	Complement component C7
P09871	<i>C1S</i>	Complement C1s subcomponent
B7ZKJ8	<i>ITIH4</i>	inter-alpha-trypsin inhibitor heavy chain H4
P03952	<i>KLKB1</i>	Plasma kallikrein
P27918	<i>CFP</i>	Properdin
P01591	<i>IGJ</i>	Immunoglobulin J chain
P13671	<i>C6</i>	Complement component C6
P02679	<i>FGG</i>	Fibrinogen gamma chain
P25311	<i>AZGP1</i>	Zinc-alpha-2-glycoprotein
P05090	<i>APOD</i>	Apolipoprotein D
P00734	<i>F2</i>	Prothrombin
P02765	<i>AHSG</i>	Alpha-2-HS-glycoprotein

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Majority protein IDs	Gene names	Protein names
Q96PD5	<i>PGLYRP2</i>	N-acetylmuramoyl-L-alanine amidase
P05546	<i>SERPIND1</i>	Heparin cofactor 2
P00450	<i>CP</i>	Ceruloplasmin
P02760	<i>AMBP</i>	Alpha-1-microglobulin
P02671	<i>FGA</i>	Fibrinogen alpha chain
P02675	<i>FGB</i>	Fibrinogen beta chain
P02647	<i>APOA1</i>	Apolipoprotein A-I
P02749	<i>APOH</i>	Beta-2-glycoprotein 1
P01008	<i>SERPINC1</i>	Antithrombin-III
P02649	<i>APOE</i>	Apolipoprotein E
P51884	<i>LUM</i>	Lumican
P19827	<i>ITIH1</i>	Inter-alpha-trypsin inhibitor heavy chain H1
P01031	<i>C5</i>	Complement C5
P02787	<i>TF</i>	Serotransferrin
P15169	<i>CPN1</i>	Carboxypeptidase N catalytic chain
P04217	<i>A1BG</i>	Alpha-1B-glycoprotein
P01024	<i>C3</i>	Complement C3
P06727	<i>APOA4</i>	Apolipoprotein A-IV
P02748	<i>C9</i>	Complement component C9
P04004	<i>VTN</i>	Vitronectin
P36955	<i>SERPINF1</i>	Pigment epithelium-derived factor
P19823	<i>ITIH2</i>	Inter-alpha-trypsin inhibitor heavy chain H2
P05543	<i>SERPINA7</i>	Thyroxine-binding globulin
P43652	<i>AFM</i>	Afamin
P02774	<i>GC</i>	Vitamin D-binding protein
P10909	<i>CLU</i>	Clusterin
P0DJ18	<i>SAA1</i>	Serum amyloid A protein
P20742	<i>PZP</i>	Pregnancy zone protein
P07358	<i>C8B</i>	Complement component C8 beta chain
P01023	<i>A2M</i>	Alpha-2-macroglobulin
P27169	<i>PON1</i>	Serum paraoxonase/arylesterase 1
P08185	<i>SERPINA6</i>	Corticosteroid-binding globulin
P02751	<i>FN1</i>	Fibronectin
Q08380	<i>LGALS3BP</i>	Galectin-3-binding protein
P02654	<i>APOC1</i>	Apolipoprotein C-I
P02655	<i>APOC2</i>	Apolipoprotein C-II
P00746	<i>CFD</i>	Complement factor D
O43866	<i>CD5L</i>	CD5 antigen-like
O75636	<i>FCN3</i>	Ficolin-3
O75882	<i>ATRN</i>	Attractin
O95445	<i>APOM</i>	Apolipoprotein M

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Majority protein IDs	Gene names	Protein names
P00739	<i>HPR</i>	Haptoglobin-related protein
P00740	<i>F9</i>	Coagulation factor IX
P00748	<i>F12</i>	Coagulation factor XII
P01042	<i>KNG1</i>	Kininogen-1
P01871	<i>IGHM</i>	Ig mu chain C region
P02652	<i>APOA2</i>	Apolipoprotein A-II
P02743	<i>APCS</i>	Serum amyloid P-component
P02745	<i>C1QA</i>	Complement C1q subcomponent subunit A
P02747	<i>C1QC</i>	Complement C1q subcomponent subunit C
P02750	<i>LRG1</i>	Leucine-rich alpha-2-glycoprotein
P02753	<i>RBP4</i>	Retinol-binding protein 4
P02763	<i>ORM1</i>	Alpha-1-acid glycoprotein 1
P02790	<i>HPX</i>	Hemopexin
P04196	<i>HRG</i>	Histidine-rich glycoprotein
P06702	<i>S100A9</i>	Protein S100-A9
P07357	<i>C8A</i>	Complement component C8 alpha chain
P0DOY3	<i>IGLC3</i>	Immunoglobulin lambda constant 3
P18428	<i>LBP</i>	Lipopolysaccharide-binding protein
P19652	<i>ORM2</i>	Alpha-1-acid glycoprotein 2
P20851	<i>C4BPB</i>	C4b-binding protein beta chain
P22792	<i>CPN2</i>	Carboxypeptidase N subunit 2
P29622	<i>SERPINA4</i>	Kallistatin
P35858	<i>IGFALS</i>	Insulin-like growth factor-binding protein complex acid labile subunit
P43251	<i>BTB</i>	Biotinidase
P59665	<i>DEFA1</i>	Neutrophil defensin 1
P80108	<i>GPLD1</i>	Phosphatidylinositol-glycan-specific phospholipase D
Q14520	<i>HABP2</i>	Hyaluronan-binding protein 2
Q16610	<i>ECM1</i>	Extracellular matrix protein 1
Q96TA2	<i>YME1L1</i>	ATP-dependent zinc metalloprotease YME1L1

Supplementary Table II. List of 10 exclusive proteins found only in subgroups of infertility.

Protein IDs	Gene names	Protein names
A0A087WSY5	<i>CPB2</i>	Carboxypeptidase B2
P0DP08	<i>IGHV4-61</i>	Ig heavy chain V-II region NEWM
A0A0C4DH73	<i>IGKV1-12</i>	Ig kappa chain V-I region Wes
E5RH81	<i>CA1</i>	Carbonic anhydrase 1
G3XAK1	<i>MST1</i>	Hepatocyte growth factor-like protein
P00742	<i>F10</i>	Coagulation factor X
P02042	<i>HBD</i>	Hemoglobin subunit delta
P04430	<i>IGKV1-16</i>	Ig kappa chain V-I region BAN
P05109	<i>S100A8</i>	Protein S100-A8
P32119	<i>PRDX2</i>	Peroxiredoxin-2

Supplementary Table III. List of differentially expressed proteins in infertile PCOS and non-PCOS groups.

Protein IDs	Protein names	Gene names	Unique peptides	Log P value PCOS/ HC	FC PCOS/ HC	Log P value non-PCOS/ HC	FC non-PCOS/ HC
P43251	Biotinidase	<i>BTD</i>	3	6.626	0.956	7.869	0.911
K7EJD3	Galectin-3-binding protein	<i>LGALS3BP</i>	7	5.673	0.974	6.391	0.961
A0A0C4DGN2	Sex hormone-binding globulin	<i>SHBG</i>	3	5.402	0.978	5.689	0.974
A0A087WUD9	Plasma protease C1 inhibitor	<i>SERPING1</i>	3	4.511	0.987	5.636	0.976
H0Y4K8	Anastellin	<i>FN1</i>	63	6.054	0.972	5.436	0.98
K7EJI9	Apolipoprotein C-I	<i>APOC1</i>	5	5.715	0.977	5.126	0.984
A0A096LPE2	Serum amyloid A-4 protein	<i>SAA2</i>	4	4.971	0.984	4.404	0.989
P00748	Beta-factor XIIa part 1	<i>F12</i>	13	4.657	0.987	4.368	0.989
C9J7V5	Properdin	<i>CFP</i>	3	4.281	0.988	3.953	0.99
P02652	Apolipoprotein A-II	<i>APOA2</i>	10	4.165	0.992	4.261	0.991
C9J075	Plasma kallikrein	<i>KLKB1</i>	16	-	-	4.13	0.991
P01042-2	Bradykinin	<i>KNG1</i>	4	3.532	0.994	4.087	0.992
P20851	C4b-binding protein beta chain	<i>C4BPB</i>	4	4.915	0.985	3.385	0.994
C9JF17	Apolipoprotein D	<i>APOD</i>	6	3.588	1.007	4.068	1.009
P02790	Hemopexin	<i>HPX</i>	28	3.965	1.008	4.466	1.01
F8WF42	Serum paraoxonase	<i>PON1</i>	10	3.377	1.006	4.397	1.011
P02743	Serum amyloid P-component	<i>APCS</i>	6	4.945	1.016	5.034	1.017
A6PVY5	C4b-binding protein alpha chain	<i>C4BPA</i>	26	4.78	1.014	5.223	1.018
O95445	Apolipoprotein M	<i>APOM</i>	5	3.428	1.006	5.305	1.019
A0A0G2JPR0	Complement C4-A	<i>C4A</i>	1	3.323	1.006	5.43	1.02
B7ZKJ8	35 kDa inter-alpha-trypsin inhibitor heavy chain H4	<i>ITIH4</i>	31	3.432	1.006	5.443	1.02
A0A7P0T8D1	Angiotensin 1-4	<i>AGT</i>	12	4.958	1.016	5.487	1.021
P00740	Coagulation factor IX	<i>F9</i>	3	5.61	1.026	5.42	1.024
P00739	Haptoglobin-related protein	<i>HPR</i>	5	4.586	1.013	5.772	1.025
P02753	Plasma retinol-binding protein(1-176)	<i>RBP4</i>	7	5.518	1.022	5.774	1.026
P01009	Alpha-1-antitrypsin	<i>SERPINA1</i>	2	4.844	1.015	5.866	1.027
G3V350	Corticosteroid-binding globulin	<i>SERPINA6</i>	8	5.579	1.023	6.274	1.035
A0A2R8Y7C0	Hemoglobin subunit alpha	<i>HBA1</i>	13	7.366	1.058	8.049	1.086
A0A0J9YWK4	Hemoglobin subunit beta	<i>HBB</i>	8	7.914	1.081	8.417	1.108
C9JC84	Fibrinogen gamma chain	<i>FGG</i>	26	10.029	1.354	10.253	1.402
G3V3A0	Serpin family A member 3	<i>SERPINA3</i>	1	6.072	0.969	-	-
A0A087WT59	Transthyretin	<i>TTR</i>	10	5.751	0.976	-	-
P04196	Histidine-rich glycoprotein	<i>HRG</i>	20	5.081	0.984	-	-
A0A0G2JIE7	Complement C2	<i>C2</i>	3	4.74	0.985	-	-
P07357	Complement component C8 alpha chain	<i>C8A</i>	10	4.501	0.988	-	-
A0A0S2Z4L3	Vitamin K-dependent protein S	<i>PROS1</i>	8	4.322	0.989	-	-
P35858	Insulin-like growth factor-binding protein complex acid labile subunit	<i>IGFALS</i>	13	3.829	0.991	-	-

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Protein IDs	Protein names	Gene names	Unique peptides	Log P value PCOS/ HC	FC PCOS/ HC	Log P value non-PCOS/ HC	FC non-PCOS/ HC
B1AKG0	Complement factor H-related protein 1	<i>CFHR1</i>	3	3.609	0.992	-	-
A0A8I5KW61	Beta-thromboglobulin	<i>PPBP</i>	11	3.751	0.992	-	-
A0A0G2JH38	Complement factor B	<i>CFB</i>	35	3.733	0.993	-	-
A0A182DWH7	Selenoprotein P	<i>SEPP1</i>	3	3.384	0.993	-	-
C9JC72	Complement component C6	<i>C6</i>	20	3.311	0.994	-	-
F8W7L3	Alpha-2-macroglobulin	<i>A2M</i>	77	3.365	0.995	-	-
P02747	Complement C1q subcomponent subunit C	<i>C1QC</i>	5	5.238	1.019	-	-
O75636	Ficolin-3	<i>FCN3</i>	6	-	-	3.695	0.992
D6RBJ7	Vitamin D-binding protein	<i>GC</i>	26	-	-	3.889	0.992
A0A0D9SG88	Complement factor H	<i>CFH</i>	48	-	-	3.264	1.005
B1AH94	Apolipoprotein L1	<i>APOL1</i>	14	-	-	3.348	1.006
C9JV37	Activation peptide fragment 1	<i>F2</i>	28	-	-	3.376	1.006
C9JEV0	Zinc-alpha-2-glycoprotein	<i>AZGP1</i>	13	-	-	3.455	1.006
P02745	Complement C1q subunit A	<i>C1QA</i>	4	-	-	3.475	1.007
P02763	Alpha-1-acid glycoprotein 1	<i>ORM1</i>	9	-	-	3.745	1.007
A0A087WWY0	Apolipoprotein(a)	<i>LPA</i>	3	-	-	3.462	1.007
P18428	Lipopolysaccharide-binding protein	<i>LBP</i>	3	-	-	3.768	1.008
A0A0C4DGL8	Haptoglobin	<i>HP</i>	18	-	-	4.032	1.008
P22792	Carboxypeptidase N subunit 2	<i>CPN2</i>	9	-	-	3.78	1.008
F5GXY0	Pregnancy zone protein	<i>PZP</i>	13	-	-	3.859	1.009
A0A024R6I7	Alpha-1-antitrypsin	<i>SERPINA1</i>	2	-	-	4.168	1.009
A0A3B3IS66	Coagulation factor XIII B chain	<i>F13B</i>	7	-	-	3.775	1.009
A0A669KB70	Apolipoprotein B-100	<i>APOB</i>	181	-	-	4.272	1.01
B1AHL2	Fibulin-1	<i>FBLN1</i>	5	-	-	4.052	1.01
K7ERG9	Complement factor D	<i>CFD</i>	4	-	-	4.092	1.011
A6PVI2	Activation peptide	<i>PLG</i>	48	-	-	4.885	1.014
B0YIW2	Apolipoprotein C-III	<i>APOC3</i>	4	-	-	4.928	1.014
P80108	Phosphatidylinositol-glycan-specific phospholipase D	<i>GPLD1</i>	4	-	-	4.73	1.015
P02750	Leucine-rich alpha-2-glycoprotein	<i>LRG1</i>	11	-	-	5.024	1.016
B5MCV4	Complement C1s subcomponent	<i>C1S</i>	9	-	-	5.163	1.019
P19652	Alpha-1-acid glycoprotein 2	<i>ORM2</i>	7	-	-	5.962	1.025
Q14520	Hyaluronan-binding protein 2	<i>HABP2</i>	6	4.157	0.989	3.255	1.006
A6XND0	Insulin-like growth factor-binding protein 3	<i>IGFBP3</i>	3	4.812	0.984	3.345	1.007
A0A087WW43	Inter-alpha-trypsin inhibitor heavy chain H3	<i>ITIH3</i>	7	3.963	0.991	3.62	1.007
P06702	Protein S100-A9	<i>S100A9</i>	2	6.605	0.956	4.267	1.011
Q16610	Extracellular matrix protein 1	<i>ECM1</i>	3	3.373	0.993	4.165	1.011
E9PQD6	Amyloid protein A	<i>SAA1</i>	2	4.847	0.985	5.13	1.018
K7ER74	Apolipoprotein C-II	<i>APOC2</i>	3	4.119	0.991	6.029	1.029
A0A2R8Y793	Actin	<i>ACTA1</i>	7	5.5	0.977	6.217	1.035

HC, healthy control; PCOS, polycystic ovary syndrome; FC, fold change.

Supplementary Table IV. List of proteins present in different clusters of heat map.

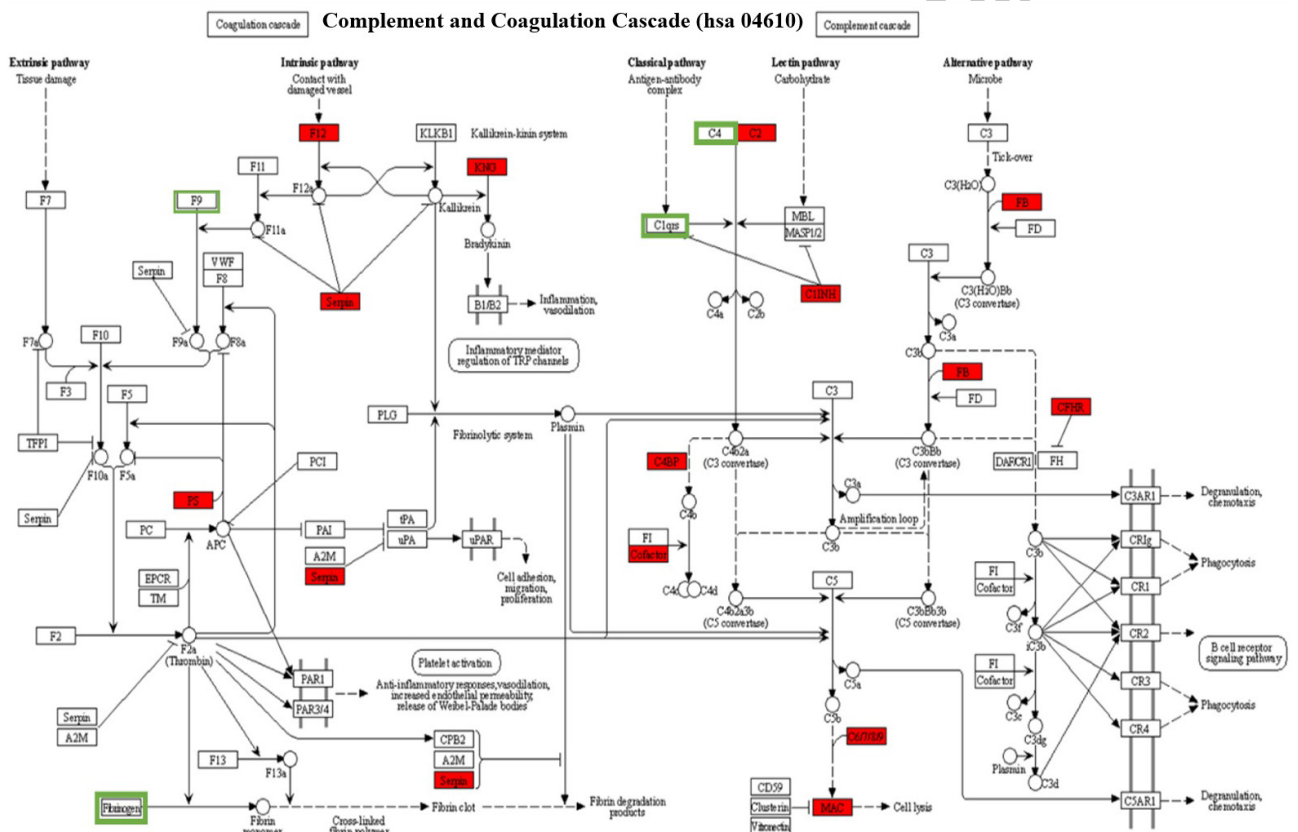
Cluster no.	Protein IDs	Gene names	Protein names
1	A0A3B3ISR2	<i>C1R</i>	Complement C1r subcomponent
1	O75882	<i>ATRNL</i>	Attractin
1	A0A182DWH7	<i>SEPP1</i>	Selenoprotein P
1	Q14520	<i>HABP2</i>	Hyaluronan-binding protein 2
1	A6XND0	<i>IGFBP3</i>	Insulin-like growth factor-binding protein 3
1	E5RG36	<i>CLU</i>	Clusterin
1	A0A0A0MSV6	<i>C1QB</i>	Complement C1q subcomponent subunit B
1	A0A087WW43	<i>ITIH3</i>	Inter-alpha-trypsin inhibitor heavy chain H3
1	P06702	<i>S100A9</i>	Protein S100-A9
1	A0A0G2JIE7	<i>C2</i>	Complement C2
1	P04196	<i>HRG</i>	Histidine-rich glycoprotein
1	A0A087WT59	<i>TTR</i>	Transthyretin
1	F8W7L3	<i>A2M</i>	Alpha-2-macroglobulin
1	P07357	<i>C8A</i>	Complement component C8 alpha chain
1	P35858	<i>IGFALS</i>	Insulin-like growth factor-binding protein complex acid labile subunit
1	P20851	<i>C4BPB</i>	C4b-binding protein beta chain
1	A0A8I5KW61	<i>PPBP</i>	Beta-thromboglobulin
1	G3V3A0	<i>SERPINA3</i>	Alpha-1-antichymotrypsin
1	A0A0S2Z4L3	<i>PROS1</i>	Vitamin K-dependent protein S
1	A0A2R8Y3M9	<i>CFI</i>	Complement factor I
1	A0A0G2JPA8	<i>SERPINF2</i>	Alpha-2-antiplasmin
2	C9J7V5	<i>CFP</i>	Properdin
2	P00748	<i>F12</i>	Beta-factor XIIa part 1
2	B1AKG0	<i>CFHR1</i>	Complement factor H-related protein 1
2	A0A0G2JH38	<i>CFB</i>	Complement factor B
2	C9JC72	<i>C6</i>	Complement component C6
2	H0Y4K8	<i>FN1</i>	Anastellin
2	K7EJ19	<i>APOC1</i>	Apolipoprotein C-I
2	A0A096LPE2	<i>SAA2-SAA4</i>	Serum amyloid A-4 protein
2	P01042-2	<i>KNG1</i>	Bradykinin
2	K7EJD3	<i>LGALS3BP</i>	Galectin-3-binding protein
2	A0A0C4DGN2	<i>SHBG</i>	Sex hormone-binding globulin
2	P02652	<i>APOA2</i>	Apolipoprotein A-II
2	P59665	<i>DEFA1</i>	Neutrophil defensin 1
3	A0A087WUD9	<i>SERPING1</i>	Plasma protease C1 inhibitor
3	P43251	<i>BTB</i>	Biotinidase
3	P29622	<i>SERPINA4</i>	Kallistatin
3	O75636	<i>FCN3</i>	Ficolin-3
3	C9J075	<i>KLKB1</i>	Plasma kallikrein
3	D6RBJ7	<i>GC</i>	Vitamin D-binding protein

Table continues on next page.....

Cluster no.	Protein IDs	Gene names	Protein names
4	B4DLR2	<i>C7</i>	Complement component C7
4	P02747	<i>C1QC</i>	Complement C1q subcomponent subunit C
4	C9JV77	<i>AHSG</i>	Alpha-2-HS-glycoprotein
4	O43866	<i>CD5L</i>	CD5 antigen-like
4	P00740	<i>F9</i>	Coagulation factor IX
4	P00739	<i>HPR</i>	Haptoglobin-related protein
4	P01009	<i>SERPINA1</i>	Alpha-1-antitrypsin
4	F8WF42	<i>PON1</i>	Serum paraoxonase/arylesterase 1
4	F5GXY0	<i>PZP</i>	Pregnancy zone protein
4	P02745	<i>C1QA</i>	Complement C1q subcomponent subunit A
4	B1AH94	<i>APOL1</i>	Apolipoprotein L1
4	P02790	<i>HPX</i>	Hemopexin
4	C9JF17	<i>APOD</i>	Apolipoprotein D
4	A0A0J9YWK4	<i>HBB</i>	Hemoglobin subunit beta
4	A6PVY5	<i>C4BPA</i>	C4b-binding protein alpha chain
4	A0A7P0T8D1	<i>AGT</i>	Angiotensin 1-4
4	A0A2R8Y7C0	<i>HBA1</i>	Hemoglobin subunit alpha
4	G3V350	<i>SERPINA6</i>	Corticosteroid-binding globulin
4	P02753	<i>RBP4</i>	Plasma retinol-binding protein(1-176)
4	C9JC84	<i>FGG</i>	Fibrinogen gamma chain
4	P02743	<i>APCS</i>	Serum amyloid P-component
5	A0A024R6I7	<i>SERPINA1</i>	Alpha-1-antiproteinase
5	A0A087WWY0	<i>LPA</i>	Apolipoprotein(a)
5	A0A087WY93	<i>SERPINA3</i>	Alpha-1-antichymotrypsin
5	A0A0C4DGL8	<i>HP</i>	Haptoglobin
5	A0A0D9SG88	<i>CFH</i>	Complement factor H
5	A0A0G2JPR0	<i>C4A</i>	Complement C4-A
5	A0A2R8Y793	<i>ACTA1</i>	Actin, alpha cardiac muscle 1
5	A0A3B3IS66	<i>F13B</i>	Coagulation factor XIII B chain
5	A0A3B3ISA6	<i>C4B</i>	C4a anaphylatoxin
5	A0A3B3ITK5	<i>C8G</i>	Complement component C8 gamma chain
5	A0A669KB70	<i>APOB</i>	Apolipoprotein B-100
5	A6PVI2	<i>PLG</i>	Activation peptide
5	B0YIW2	<i>APOC3</i>	Apolipoprotein C-III
5	B1AHL2	<i>FBLN1</i>	Fibulin-1
5	B5MCV4	<i>C1S</i>	Complement C1s subcomponent
5	B7ZKJ8	<i>ITIH4</i>	Inter-alpha-trypsin inhibitor heavy chain H4
5	C9JEV0	<i>AZGP1</i>	Zinc-alpha-2-glycoprotein
5	C9JV37	<i>F2</i>	Activation peptide fragment 1
5	E9PQD6	<i>SAA1</i>	Amyloid protein A
5	K7ER74	<i>APOC2</i>	Apolipoprotein C-II

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Cluster no.	Protein IDs	Gene names	Protein names
5	K7ERG9	<i>CFD</i>	Complement factor D
5	O95445	<i>APOM</i>	Apolipoprotein M
5	P02750	<i>LRG1</i>	Leucine-rich alpha-2-glycoprotein
5	P02763	<i>ORM1</i>	Alpha-1-acid glycoprotein 1
5	P18428	<i>LBP</i>	Lipopolysaccharide-binding protein
5	P19652	<i>ORM2</i>	Alpha-1-acid glycoprotein 2
5	P22792	<i>CPN2</i>	Carboxypeptidase N subunit 2
5	P80108	<i>GPLD1</i>	Phosphatidylinositol-glycan-specific phospholipase D
5	Q16610	<i>ECM1</i>	Extracellular matrix protein 1
5	Q96TA2	<i>YME1L1</i>	ATP-dependent zinc metalloprotease



Supplementary Fig. 1. The KEGG (Kyoto encyclopaedia of genes and genomes) pathway. Maps were obtained at 0.05 FDR cut off for up-regulated and down-regulated proteins in both subgroups groups in order to gain insight into the role of protein in the biological system.