# PeerJ

Proteomic analysis of human cervical adenocarcinoma mucus to identify potential protein biomarkers

Zhifang Ma<sup>1,\*</sup>, Jie Chen<sup>2,\*</sup>, Ting Luan<sup>2</sup>, Chengzhuo Chu<sup>1</sup>, Wangfei Wu<sup>1</sup>, Yichao Zhu<sup>3,4</sup> and Yun Gu<sup>1</sup>

<sup>1</sup> Department of Pathology, Women's Hospital of Nanjing Medical University, Nanjing Maternity and Child Health Care Hospital, Nanjing, China

- <sup>2</sup> Department of Gynecology, Women's Hospital of Nanjing Medical University, Nanjing Maternity and Child Health Care Hospital, Nanjing, China
- <sup>3</sup> Department of Physiology, Nanjing Medical University, Nanjing, China
- <sup>4</sup> State Key Laboratory of Reproductive Medicine, Nanjing Medical University, Nanjing, China

\* These authors contributed equally to this work.

# ABSTRACT

**Background:** Cervical cancer is the most common gynecological cancer, encompassing cervical squamous cell carcinoma, adenocarcinoma, and other epithelial tumors. There are many diagnostic methods to detect cervical cancers but no precision screening tool for cervical adenocarcinoma at present.

**Material and methods:** The cervical mucus from three normal cervices (Ctrl), three endocervical adenocarcinoma (EA), and three cervical adenocarcinoma in situ (AIS) was collected for proteomic analysis. The proteins were screened using liquid chromatography-mass spectrometry analysis (LC-MS). The biological function of the differently expressed proteins were predicted by Gene Ontology (GO). **Results:** A total of 711 proteins were identified, including 237 differently expressed proteins identified in EA/Ctrl comparison, 256 differently expressed proteins identified in AIS/Ctrl comparison, and 242 differently expressed proteins identified in AIS/EA comparison (up-regulate  $\geq 1.5$  or down-regulate  $\leq 0.67$ ). Functional annotation was performed using GO analysis on 1,056 differently expressed proteins to identify those that may impact cervical cancer, such as heme protein myeloperoxidase, which is involved in the immune process, and APOA1, which is associated with lipid metabolism.

**Conclusion:** We used proteomic analysis to screen out differently expressed proteins from normal cervical mucus and cervical adenocarcinoma mucus samples. These differently expressed proteins may be potential biomarkers for the diagnosis and treatment of cervical adenocarcinoma but require additional study.

**Subjects** Biochemistry, Bioinformatics, Oncology, Women's Health **Keywords** Cervical adenocarcinoma, Mucus, Proteomics, Biomarkers

## **INTRODUCTION**

Cervical cancer is the third most commonly diagnosed gynecological malignancy and the fourth leading cause of cancer-related death in women worldwide resulting in approximately 530,000 deaths annually (*Siegel et al., 2014*; *Bray et al., 2018*). Advances in

Submitted 21 February 2020 Accepted 21 June 2020 Published 28 July 2020

Corresponding authors Yichao Zhu, zhuyichao@njmu.edu.cn Yun Gu, guyun1979@njmu.edu.cn

Academic editor Vladimir Uversky

Additional Information and Declarations can be found on page 13

DOI 10.7717/peerj.9527

Copyright 2020 Ma et al.

Distributed under Creative Commons CC-BY 4.0

#### **OPEN ACCESS**

the treatment of cervical cancer, such as radiotherapy, chemotherapy, surgery, immunotherapy, and targeted therapy have not improved the survival rate for cervical cancer (*Dai et al., 2016; Menderes et al., 2016*). Cervical adenocarcinoma follows squamous cell carcinoma as the most common subtype of cervical carcinoma and is increasing in incidence and prevalence in younger populations (*Seoud, Tjalma & Ronsse, 2011*). Novel biomarkers for the diagnosis and treatment of cervical adenocarcinoma are needed.

A number of studies have looked at the prevention and treatment of cervical cancer by investigating relevant cells, tissues, and subtypes (Farzanehpour et al., 2019; Li, Hong & Wijayakulathilaka, 2019; Siegler et al., 2019). Cervical cancer is commonly diagnosed through cytological screening known as a Pap test in conjunction with the detection of high-risk human papillomaviruses (hr-HPVs). However, these tests are expensive and depend on good infrastructure and well-trained personnel (Lynge et al., 2014). Proteomics analysis is a powerful tool for monitoring the change of protein levels to discover new biomarkers in many cancers, such as colorectal cancer, pancreatic cancer, and neuroendocrine cervical cancer (Lin et al., 2014; Pan, Brentnall & Chen, 2015; Chauvin & *Boisvert*, 2018). Unique membrane proteins have been identified using proteomics analysis of the cervical cancer cell lines (Pappa et al., 2018). Lee et al. (2011) studied the proteome of cervical mucus plugs and suggested its role for maintaining pregnancy and parturition. However, their approach was not suitable for comparative studies of mucins among different groups. The constitutive protein composition of cervical mucus in fertile women and changes in the cervical mucus proteome were identified throughout the menstrual cycle but interactions between immunoglobulin, defense-binding protein, and other cervical mucus proteins were not studied (Grande et al., 2015).

We sought to discover novel biomarkers using proteomics analysis of nine uterine cavity mucus samples to produce a better prognosis for cervical cancer and to provide a reliable reference for the diagnosis and treatment of patients with cervical adenocarcinoma.

# **MATERIALS AND METHODS**

### **Patient materials**

Cervical mucus samples were obtained from three patients with normal cervices (Ctrl), three patients with endocervical adenocarcinoma (EA), and three patients with adenocarcinoma in situ of the cervix (AIS) who had a total hysterectomy at the Nanjing Maternity and Child Health Care Hospital Affiliated to Nanjing Medical University. The three normal cervical mucus samples acted as the control. All samples were stored at -80 °C and an initial histopathological diagnosis was conducted. The cancerous cervical mucus and normal cervical mucus were used for proteomic analyses. This study was approved by the ethical committee of Women's Hospital of Nanjing Medical University, Nanjing Maternity and Child Health Care Hospital ((2019) KY-040). All participants voluntarily agreed to participate in this study and signed an informed consent.

### **Protein extraction**

Protein extraction was performed on the nine cervical mucus samples. Each sample was mixed and homogenized with a protein lysate (7 M Urea, 2 M Thiourea, 4% SDS, 40 mM Tris-HCl, pH 8.5, 1 mM Phenylmethanesulfonyl fluoride (PMSF), 2 mM Ethylene Diamine Tetraacetic Acid (EDTA)) on ice for 5 min. Dithiothreitol (DTT) (Solarbio, 428F0422) was added to a final concentration of 10 mM, followed by an ice bath ultrasound for 15 min and centrifuging at 13,000g at 4 °C for 20 min. After ultracentrifugation, supernatants of each samples were collected to a new centrifugal tube. Cold acetone was added into the centrifugal tube at four times the volume and was left to stand overnight at -20 °C. The protein precipitation was collected and left to air dry. A total of 8 M urea/100 mM tetraethyl-ammonium bromide (TEAB) (BCBC6216; Sigma-Aldrich, St. Louis, MO, USA) (pH 8.0) solution was added to re-dissolve the protein and then DTT was added to the final concentration of 10 mM and the solution was immersed in water at 56 °C for 30 min for the reduction reaction. Iodoacetamide (IAM, Aladdin, J1513091) was added to the final concentration of 55 mM and placed at room temperature in the dark for 30 min to produce an alkylation reaction. The Bradford method was used to measure protein concentrations.

### iTRAQ labeling and peptide fractionation

The extracted proteins were digested by trypsin to obtain the corresponding peptides, which were desalted by the Durashell C18 column (5 ms, 100 A, 4.6 × 250 mm) (Agela, Tianjin, China) and vacuum dried. The peptides were dissolved with 0.5 M TEAB and labeled using the itraq-8 standard kit (SCIEX, Shanghai, China) according to the manufacturer's instructions. The samples were labeled and mixed and the mixed peptides were then graded and separated using the Ultimate 3000 HPLC system (DINOEX; Thermo, Waltham, MA, USA). The separation of peptides was achieved by increasing acetonitrile (CAN) concentration under alkaline conditions. The flow rate was 1 ml/min with one tube collected per minute. A total of 42 secondary fractions were collected and combined into 12 fractions, which were desalinated and vacuum-dried on the Strata-X column. All analyses were completed by Wuhan Genecreate Biological Engineering Co. LTD. (Wuhan, China).

### Protein identification and quantification

Proteins were identified by the Proteinpilot<sup>TM</sup> V4.5 search engine. At least one unique peptide section of each protein line and the unused score was believed to be no more than 1.3 (just above 95%). The reliability of each identified peptide and its protein quantification was considered for protein quantification.

### Liquid chromatography-mass spectrometry analysis

Mass spectrometry data were collected using the TripleTOF 5600+ liquid/mass coupling system (SCIEX, Shanghai, China). Polypeptide samples were dissolved in 2% acetonitrile/0.1% formic acid and analyzed using the TripleTOF 5600+ mass spectrometer. The peptide solution was added to the C18 capture column (5 ms, 100 ms  $\times$  20 mm) and

eluted on the C18 analysis column (3 ms, 75 ms × 150 mm) for 90 min with a flow rate of 300 nL/min. The two mobile phases were buffer A (2% acetonitrile, 0.1% formic acid, 98% H<sub>2</sub>O) and buffer B (98% acetonitrile, 0.1% formic acid, 2% H<sub>2</sub>O). The primary mass spectrometry for information dependent acquisition (IDA) was scanned at 250 ms ion accumulation time and a secondary mass spectrometry of 30 precursor ions was collected at 50 ms ion accumulation time. The MS1 spectrum was collected in the range of 350–1,500 m/z, and the MS2 spectrum was collected in the range of 100–1,500 m/z. The dynamic exclusion time for precursor ions was set at 15 s.

### **Functional annotation**

Gene ontology (GO) functional annotation analysis was performed for all identified proteins and the GO functions of the cellular component, biological process, and molecular function corresponding to all proteins were analyzed. Detailed information can be found at http://www.geneontology.org. COG analysis and KEGG analysis were used to further determine the function of the proteins.

### **Statistics**

The mean value of the ratio of repeated samples was normalized using the median as the difference multiple by the samples to be compared. The minimum *P* value of Student's *t* test of the single sample of parewise comparisons between repeated samples was used as the significance difference of the samples. Differential proteins were screened according to fold change and *P* value. The proteins were considered to be statistically different when the difference of multiples was  $\geq 1.5$  (i.e., up-regulate  $\geq 1.5$  and down-regulate  $\leq 0.67$ ) with a *P* value < 0.05 after the significance statistical test.

# RESULTS

## Identification of proteins in cervical mucinous

A total of nine samples were divided into three groups to identify differently abundant proteins. Ctrl was comprised of samples from three healthy individuals, EA was comprised of samples from three invasive cervical adenocarcinoma individuals, and AIS had samples from three cervical adenocarcinoma in situ individuals. All samples were obtained from individuals with an average age of 41 years. The clinicopathological characteristics of these patients were summarized in Table 1. A diagnosis of endocervical adenocarcinoma was suggested by the cervical sample based on cytomorphological features. The postoperative pathological diagnosis showed no lymph node metastasis and immunohistochemical results directed the diagnosis and subsequent treatment. The corresponding pathological images were obtained for the nine samples and were stained by HE (Fig. 1A; Fig. S1).

The unique peptides only needed to be detected in one protein to identify the protein in the sample. The presence of the peptide uniquely determined the presence of the corresponding protein. The biordinate distribution of the unique peptides representing the unique proteins was identified (Fig. 1B). The abscissa is the number of unique peptides contained in the protein, and the left ordinate is the number of proteins corresponding to

Table 1 Patient information of samples used in the comprehensive proteomic experiments.										
Sample	Age (years old)	FIGO stage	Lesion size (mm)	Depth of invasion	Immunohistochemical diagnosis	Chief complaint	ТСТ	HPV		
Ctrl-1	49	Normal				Routine physical	_	-		
Ctrl-2	44	Normal				Routine physical	_	_		
Ctrl-3	40	Normal				Routine physical	_	_		
AIS-1	36	0	2.3		Ki67(70%+), P16(+), ER(-)	Cervical lesions	ASC-H	+		
AIS-2	32	0	2		Ki67(90%+), P16(+), ER(-)	Vaginal bleeding after intercourse	+	+		
AIS-3	43	0	<7		Ki67(50%+), P16(+), ER(+)	Cervical lesions	_	+		
EA-1	54	II A1	12	>1/3 & <2/3	Ki67(60%+), P16(+), ER(-)	Vaginal bleeding after menopause	NK	NK		
EA-2	37	I B2	≤20	>1/3 & <2/3	Ki67(2%+), P16(++), ER(+)	Menstrual disorder and abnormal leukorrhea	ASC-US	+		
EA-3	34	II A1	30	>1/3 & <2/3	Ki67(80%+), P16(-), ER(-)	Vaginal bleeding after intercourse	HSIL	+		

Note:

Ctrl indicates the normal cervix; AIS indicates adenocarcinoms in situ of the cervical; EA indicates endocervical adenocarcinoma; TCT, Thinprep cytology test; HPV, human papillomavirus; –, negative; +, positive; NK, not known; ASC-H, atypical squamous cells: cannot exclude high-grade squamous intraepithelial lesion; ASC-US, atypical squamous cells of undetermined significance; HSIL, high-grade squamous intraepithelial lesions.

the abscissa. The right ordinate is the cumulative protein proportion corresponding to the abscissa. The results showed that there are 757 protein with at least 2 unique peptides, accounting for 78.12% of the total protein. The length distribution of the peptide was obtained by mass spectrometer (Fig. 1C). The average length of the peptides identified was 12.55, which was a reasonable peptide length. A protein with more peptide support makes the protein more reliable, therefore, protein coverage can indirectly reflect the overall accuracy of the identification results. Pie charts were used to represent the percentage of different proteins within the identification coverage (Fig. 1D). The identified range of proteins was 37.56% and proteins with coverage  $\geq 20\%$  accounted for 39.73% of the total protein. The average protein coverage was 21.90%.

#### The analysis of differential protein expression profile

Of the identified proteins, 237, 256, and 242 proteins were significantly differently expressed in the comparison group EA/Ctrl, AIS/Ctrl, and AIS/EA using iTRAQ quantitative analysis with the filtered threshold of up-regulation  $\geq 1.5$  or down-regulation  $\leq 0.67$ , respectively. The number of significant protein differences was shown between two samples (Fig. 2A). The results of the relative quantitative analysis of the protein showed that the protein ratio distribution followed a normal distribution pattern. The distribution of the differential multiples of all quantifiable proteins was shown (Fig. 2B), where the *x*-coordinate represented the value of the differential multiples after logarithmic transformation in base 2. The part greater than 0 indicated the up-regulated expression, while the part less than 0 indicated the down-regulated expression. The hierarchical clustering heat map of the intersection of 98 differently significant proteins in the comparison groups visually aggregated samples/proteins with higher similarity (Fig. 2C). The data source is the log2 logarithmic value of the protein abundance ratio between the two comparison groups. Quantitative information of some significantly different proteins is shown in Table 2. Some of the differently expressed proteins were

# **Peer**J



Figure 1 Identification of proteins related to cervical mucus. (A–C), pathological section analysis of normal cervical mucus (HE,  $\times$ 200); (D–F), pathological section analysis of cervical adenocarcinoma mucus in situ (HE,  $\times$ 200); (G–I), pathological section analysis of endocervical adenocarcinoma mucus (HE,  $\times$ 200). (J) All peptide length profiles. (K) The distribution of unique peptide number of identified proteins. (L) Pie chart of protein identification coverage distribution. Full-size  $\Box$  DOI: 10.7717/peerj.9527/fig-1

significantly expressed in the groups EA/Ctrl, AIS/Ctrl, and AIS/EA, including myeloperoxidase and APOA1, which are involved in the development of cancer (*Cine et al., 2014; Castelao et al., 2015; Zadeh Fakhar et al., 2019*).

## Functional and pathway analysis of identified proteins

It is important to analyze the function of genes as more genomes are sequenced and to predict the function of unknown genes to guide further experiments. We provided the commonly used GO annotation and KEGG annotation results, as well as the COG annotation results to comprehensively reflect the function of proteins obtained from different databases to reveal the biological significance of proteins in various life activities.



**Figure 2 Distribution and stratification of differential proteins.** (A) Distribution of differential proteins between two groups. (B) Protein abundance distribution map. The figure shows the distribution of the differential multiples of all quantifiable proteins, where the *x*-coordinate represents the value of the differential multiples after logarithmic transformation in base 2. The part greater than 0 indicates the up-regulated expression, while the part less than 0 indicates the down-regulated expression. (C) Hierarchical clustering heat map of differential protein layers. Rows represent protein clustering and the column represents the sample pair clustering situation. As the protein ratio changes from small to large, the heat map color shows a corresponding green-black-red change.

Full-size DOI: 10.7717/peerj.9527/fig-2

A total of 1,118 proteins were identified and functionally annotated (Fig. 3A). The annotation information for some proteins varied among the databases due to the limitations of the background annotation library. GO, COG, and KEGG annotated 1,056 proteins, 569 proteins, and 751 proteins, respectively. GO is a community-based bioinformatics resource that supplies information about gene product function using ontologies to represent biological knowledge (*Gene Ontology Consortium, 2015*).

	Accession	Gene	Protein.names	P-value		
Up-regulated						
EA:Ctrl	P04114 APOB Apolipoprotein		Apolipoprotein	0.0402		
	P02751	FINC	Fibronectin	0.0069		
	P13646	K1C13	Keratin, type I cytoskeletal 13	0.001		
	P02549	SPTA1	Spectrin alpha chain, erythrocytic 1	0.0003		
	P04040	CATA	Catalase	0.0034		
	P19013	K2C4	Keratin, type II cytoskeletal 4	0.0006		
	P05164	PERM	Myeloperoxidase	0.0007		
	P11277	SPTB1	Spectrin beta chain, erythrocytic	0.0024		
	P13796	PLSL	Plastin-2	0.0082		
	P01833	PIGR	Polymeric immunoglobulin receptor	0.0003		
AIS:EA	P67775	PP2AA	Serine/threonine-protein phosphatase 2A catalytic subunit alpha isoform	0.000000115		
	P69892	HBG2	Hemoglobin subunit gamma-2	0.00000523		
	Q13228	SBP1	Methanethiol oxidase	0.0001		
	P02656	APOC3	Apolipoprotein C-III	0.0001		
	P07476	INVO	Involucrin	0.0003		
	P01019	ANGT	Angiotensinogen	0.0004		
	P68371	TBB4B	Tubulin beta-4B chain	0.0005		
	P07195	LDHB	L-lactate dehydrogenase B chain	0.0006		
	P00352	AL1A1	Retinal dehydrogenase 1	0.0008		
	P02647	APOA1	Apolipoprotein A–I	0.0009		
AIS:Ctrl	Q8NFI4	F10A5	Putative protein FAM10A5	0.000000278		
	P63208	SKP1	S-phase kinase-associated protein 1	0.00000389		
	P69892	HBG2	Hemoglobin subunit gamma-2	0.0001		
	Q6XQN6	PNCB	Nicotinate phosphoribosyltransferase	0.0001		
	P00918	CAH2	Carbonic anhydrase 2	0.0002		
	P00915	CAH1	Carbonic anhydrase 1	0.0005		
	P27169	PON1	Serum paraoxonase/arylesterase 1	0.0005		
	P02549	SPTA1	Spectrin alpha chain, erythrocytic 1	0.0006		
	P30043	BLVRB	Flavin reductase (NADPH)	0.0006		
	P04040	CATA	Catalase	0.0006		
Down-re	gulated					
EA:Ctrl	P0C0L5	CO4B	Complement C4-B	0.0064		
	P01009	A1AT	Alpha-1-antitrypsin	0.0000254		
	P08603	CFAH	Complement factor H	0.0014		
	A8K2U0	A2ML1	Alpha-2-macroglobulin-like protein 1	0.0025		
	P02647	APOA1	Apolipoprotein A–I	0.0027		
	P00738	HPT	Haptoglobin	0.0078		
	P02675	FIBB	Fibrinogen beta chain	0.0108		
	Q09666	AHNK	Neuroblast differentiation-associated protein AHNAK	0.0002		
	P04083	LPC1	Calpactin II	0.0003		
	P02671	FIBA	Fibrinogen alpha chain	0.0034		

Table 2 Quantitative information on the top 10 proteins with significant differences among different components.

Table 2 (continued)						
	Accession	Gene	Protein.names	P-value		
AIS:EA	P0DOX2	IGA2	Immunoglobulin alpha-2 heavy chain	0.0000343		
	P04792	HSP27	Heat Shock 27 kDa Protein 1	0.0000452		
	P05164	PERM	Myeloperoxidase	0.0001		
	P02751	FINC	Fibronectin	0.0001		
	P13647	K2C5	Keratin, type II cytoskeletal 5	0.0002		
	P06702	MRP14	Migration inhibitory factor-related protein 14	0.0003		
	P27797	CALR	Calreticulin	0.0003		
	P25815	S100P	Protein S100-P	0.0004		
	P07737	PROF1	Profilin-1	0.0007		
	P15104	GLNA	Glutamine synthetase	0.0007		
AIS:Ctrl	P14618	KPYM	Pyruvate kinase PKM	0.0000178		
	P35321	SPR1A	Cornifin-A	0.0001		
	P59665	MRS	Myeloid-Related Sequence	0.0002		
	P05164	PERM	Myeloperoxidase	0.0003		
	P02750	A2GL	Leucine-rich alpha-2-glycoprotein	0.0003		
	P02748	CO9	Complement component C9	0.0004		
	P31151	S10A7	Protein S100-A7	0.0005		
	Q9UBG3	Sep-53	Squamous epithelial heat shock protein 53	0.0006		
	P80188	NGAL	Neutrophil gelatinase-associated lipocalin	0.0006		
	O60218	AK1BA	Aldo-keto reductase family 1 member B10	0.0007		

GO was used to annotate 1,056 proteins and statistical analysis was performed based on biological, cellular, and molecular processes (Figs. 3B–3D). We conducted independent functional annotation analysis on the up-regulated and down-regulated differently expressed proteins to better analyze the functions of differently expressed proteins. A total of 237 differently expressed proteins (98 significantly up-regulated and 139 significantly down-regulated proteins) in EA/Ctrl were used as examples to compare and analyze their GO annotation results. Functional notes were analyzed for groups AIS/Ctrl and AIS/EA, and the results are shown in a bar chart in Figs. 4A–4C.

# DISCUSSION

Cervical cancer is an aggressive gynecological cancer of the uterine cervix. It is the fourth most common cancer among women worldwide, with an estimated 570,000 cases and 311,000 deaths in 2018 (*Bray et al., 2018; Takayanagi et al., 2019*). Although the incidence of cervical squamous cell carcinoma has decreased, the incidence of cervical adenocarcinoma remains high (*Vizcaino et al., 1998; Smith et al., 2000; Fan et al., 2014; Saida et al., 2019; Ouyang et al., 2020*). Cervical adenocarcinoma currently accounts for up to 25% of all cases of cervical cancer in many western countries (*Parkin & Bray, 2006*). It is important to find novel biomarkers relevant to cervical adenocarcinoma to improve the current treatment strategies and prognosis of this disease.

# Peer\_





We identified differently expressed proteins from subjects with cervical adenocarcinoma and those with normal cervices using the iTRAQ proteomics approach. A total of 1,118 proteins were identified in three separate trials with 711 common proteins identified. We found that the number of proteins with significant differences in the comparison groups (EA/Ctrl, AIS/Ctrl and AIS/EA) was 237, 256 and 242, respectively. We performed functional annotations, including GO, COG, and pathway analysis, on significantly differently expressed proteins. Expression level clustering analysis and



**Figure 4 GO function annotation statistics results of differential proteins.** The yellow column shows that (A) EA/Ctrl (B) AIS/EA (C) AIS/Ctrl upregulation of the functional classification proportion of differential proteins under three standard bodies. The green column is the functional classification proportion under the three ontologies of down-regulated differential proteins.

Full-size 🖾 DOI: 10.7717/peerj.9527/fig-4

functional enrichment analysis was performed on all significantly differently expressed proteins to determine significant proteins.

There are many studies on the proteomics of cervical cancer but few studies on the proteomics analysis of cervical adenocarcinoma (*Kacerovsky & Tosner, 2009*; *Serafín-Higuera et al., 2016*; *Zhang et al., 2016*). We performed proteomics analysis on samples of cervical adenocarcinoma mucus and normal cervical mucus a nd used functional annotation to screen out differently expressed proteins relevant to immune, metabolic, cell adhesion, and other cellular processes, such as myeloperoxidase and APOA1. Differently expressed proteins may be used as markers for the early diagnosis of cervical adenocarcinoma, however, our study had some limitations, including its small sample size and lack of other techniques to verify the accuracy of our analysis.

Myeloperoxidase, also known as MPO or PERM, is an important member of the heme peroxidase-cyclooxygenase superfamily. It is mainly expressed in neutrophils and monocytes (*Ndrepepa, 2019*). The enzyme can be released through phagocytosis in phagocytic bodies to catalyze the synthesis of hypochlorous acid (HOCl) from hydrogen peroxide (H<sub>2</sub>O<sub>2</sub>) and chloride ions (Cl<sup>-</sup>) (*Arnhold & Flemmig, 2010*; *Gomez-Mejiba et al., 2010*; *Rossmann et al., 2011*). HOCl is a potent microbicidal agent that damages DNA, proteins, and lipids. Myeloperoxidase is associated with many cancer types, including lung, ovarian, colorectal, and prostate cancers (*Arslan, Pinarbasi & Silig, 2011*; *Li et al., 2011*; *Castillo-Tong et al., 2014*; *Zou et al., 2018*) through the –463G/A promoter polymorphism. It is reported that the –463G/A MPO gene polymorphism is not associated with cervical intraepithelial neoplasia and susceptibility to cervical cancer and the genotype GG of MPO with a higher transcriptional activity is a protective against cervical cancer (*Mustea et al., 2007*; *Castelao et al., 2015*; *Natter et al., 2016*). Additional studies are required to determine the role of myeloperoxidase in cervical tumorigenesis.

APOA1 belongs to the apolipoprotein A1/A4/E family and participates in lipid metabolism, including the intracellular reuse of fatty acids; it is an important carrier and cofactor (*Zhang & Yang, 2018*). APOA1 is the major component of high density lipoprotein. *Rhee, Byrne & Sung (2017)* reported that the ratio of HDL cholesterol to APOA1 may be a risk marker for cancer mortality. APOA1 is dysfunctional in cervical squamous cell carcinoma and is identified as a biomarker (*Guo et al., 2015*). We suggest that APOA1 may be a novel candidate marker for cervical adenocarcinoma and further study is needed to determine its functional mechanisms.

We collected nine cervical mucus samples and summarized their clinical information. Menopause is a life phenomenon with characteristics of estrogen secretion depletion and the cessation of menstruation. The majority of women enter menopause between the ages of 49 and 52 (*Takahashi & Johnson, 2015*) and some studies show no statistically significant difference in cervical cancer rates of premenopausal and postmenopausal women (*Holt et al., 2017; Zhang et al., 2018*). A total of 8 of the 9 individuals used in this study were premenopausal and the postmenopausal individual was within the age range of normal menopause. All samples met the guidelines for normal menstruation and had a uterus of normal size and position; the cervix was enlarged or inflamed in all nine cases.

Our sample size was small due to variable sample collection time, long cycles, and limited scientific research funds. We intend to verify our results in the future and will investigate the mechanism of biomarkers used in the treatment of cervical adenocarcinoma.

# CONCLUSION

We performed a non-targeted proteomics study to profile differently expressed proteins in cervical adenocarcinoma. The proteins studied may serve as potential biomarkers for cervical cancer research and treatment.

# **ADDITIONAL INFORMATION AND DECLARATIONS**

## Funding

This study was supported by the National Natural Science Foundation of China (81802105, 81702561); Jiangsu Provincial Key Research and Development Program (BE2017619); and by the Nanjing Technological Development Program (201605049). The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

# **Grant Disclosures**

The following grant information was disclosed by the authors: National Natural Science Foundation of China: 81802105 and 81702561. Jiangsu Provincial Key Research and Development Program: BE2017619. Nanjing Technological Development Program: 201605049.

# **Competing Interests**

The authors declare that they have no competing interests.

## **Author Contributions**

- Zhifang Ma conceived and designed the experiments, performed the experiments, analyzed the data, prepared figures and/or tables, and approved the final draft.
- Jie Chen conceived and designed the experiments, performed the experiments, analyzed the data, prepared figures and/or tables, and approved the final draft.
- Ting Luan performed the experiments, prepared figures and/or tables, and approved the final draft.
- Chengzhuo Chu performed the experiments, prepared figures and/or tables, and approved the final draft.
- Wangfei Wu performed the experiments, prepared figures and/or tables, and approved the final draft.
- Yichao Zhu conceived and designed the experiments, authored or reviewed drafts of the paper, and approved the final draft.
- Yun Gu conceived and designed the experiments, authored or reviewed drafts of the paper, and approved the final draft.

### **Human Ethics**

The following information was supplied relating to ethical approvals (i.e., approving body and any reference numbers):

Women's Hospital of Nanjing Medical University, Nanjing Maternity and Child Health Care Hospital, approval to carry out the study within its facilities (2019) KY-040.

### **Data Availability**

The following information was supplied regarding data availability:

The raw data are available in Figshare: Gu, Yun (2020): WIFF.zip. figshare. Dataset. DOI 10.6084/m9.figshare.11864337.v1.

#### **Supplemental Information**

Supplemental information for this article can be found online at http://dx.doi.org/10.7717/ peerj.9527#supplemental-information.

# REFERENCES

- Arnhold J, Flemmig J. 2010. Human myeloperoxidase in innate and acquired immunity. *Archives of Biochemistry and Biophysics* 500(1):92–106 DOI 10.1016/j.abb.2010.04.008.
- Arslan S, Pinarbasi H, Silig Y. 2011. Myeloperoxidase G-463A polymorphism and risk of lung and prostate cancer in a Turkish population. *Molecular Medicine Reports* 4(1):87–92.
- Bray F, Ferlay J, Soerjomataram I, Siegel RL, Torre LA, Jemal A. 2018. Global cancer statistics 2018: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. CA: A Cancer Journal for Clinicians 68(6):394–424 DOI 10.3322/caac.21492.
- Castelao C, Da Silva AP, Matos A, Inacio A, Bicho M, Medeiros R, Bicho MC. 2015. Association of myeloperoxidase polymorphism (G463A) with cervix cancer. *Molecular and Cellular Biochemistry* **404(1-2)**:1–4 DOI 10.1007/s11010-015-2359-5.
- Castillo-Tong DC, Pils D, Heinze G, Braicu I, Sehouli J, Reinthaller A, Schuster E, Wolf A, Watrowski R, Maki RA, Zeillinger R, Reynolds WF. 2014. Association of myeloperoxidase with ovarian cancer. *Tumour Biology* 35(1):141–148 DOI 10.1007/s13277-013-1017-3.
- Chauvin A, Boisvert FM. 2018. Proteomics analysis of colorectal cancer cells. *Methods in Molecular Biology* 1765:155–166.
- **Cine N, Baykal AT, Sunnetci D, Canturk Z, Serhatli M, Savli H. 2014.** Identification of ApoA1, HPX and POTEE genes by omic analysis in breast cancer. *Oncology Reports* **32(3)**:1078–1086 DOI 10.3892/or.2014.3277.
- Dai S, Lu Y, Long Y, Lai Y, Du P, Ding N, Yao D. 2016. Prognostic value of microRNAs in cervical carcinoma: a systematic review and meta-analysis. *Oncotarget* 7(23):35369–35378 DOI 10.18632/oncotarget.9294.
- Fan DM, Tian XY, Wang RF, Yu JJ. 2014. The prognosis significance of TGF-beta1 and ER protein in cervical adenocarcinoma patients with stage Ib~IIa. *Tumour Biology* 35(11):11237–11242 DOI 10.1007/s13277-014-2110-y.
- Farzanehpour M, Mozhgani SH, Jalilvand S, Faghihloo E, Akhavan S, Salimi V, Azad TM. 2019. Serum and tissue miRNAs: potential biomarkers for the diagnosis of cervical cancer. *Virology Journal* 16(1):116 DOI 10.1186/s12985-019-1220-y.
- Gene Ontology Consortium. 2015. Gene Ontology Consortium: going forward. *Nucleic Acids Research* 43(Database issue):D1049–D1056.

- Gomez-Mejiba SE, Zhai Z, Gimenez MS, Ashby MT, Chilakapati J, Kitchin K, Mason RP, Ramirez DC. 2010. Myeloperoxidase-induced genomic DNA-centered radicals. *Journal of Biological Chemistry* 285(26):20062–20071 DOI 10.1074/jbc.M109.086579.
- Grande G, Milardi D, Vincenzoni F, Pompa G, Biscione A, Astorri AL, Fruscella E, Luca ADe, Messana I, Castagnola M, Marana R. 2015. Proteomic characterization of the qualitative and quantitative differences in cervical mucus composition during the menstrual cycle. *Molecular Biosystems* 11(6):1717–1725 DOI 10.1039/C5MB00071H.
- Guo X, Hao Y, Kamilijiang M, Hasimu A, Yuan J, Wu G, Reyimu H, Kadeer N, Abudula A. 2015. Potential predictive plasma biomarkers for cervical cancer by 2D-DIGE proteomics and ingenuity pathway analysis. *Tumour Biology* 36(3):1711–1720 DOI 10.1007/s13277-014-2772-5.
- Holt HK, Zhang L, Zhao FH, Hu SY, Zhao XL, Zhang X, Pan QJ, Zhang WH, Smith JS, Qiao YL. 2017. Evaluation of multiple primary and combination screening strategies in postmenopausal women for detection of cervical cancer in China. *International Journal of Cancer* 140(3):544–554 DOI 10.1002/ijc.30468.
- Kacerovsky M, Tosner J. 2009. Proteomics and biomarkers for detection of cervical cancer. *Ceska Gynekol* 74(5):335–338.
- Lee DC, Hassan SS, Romero R, Tarca AL, Bhatti G, Gervasi MT, Caruso JA, Stemmer PM, Kim CJ, Hansen LK, Becher N, Uldbjerg N. 2011. Protein profiling underscores immunological functions of uterine cervical mucus plug in human pregnancy. *Journal of Proteomics* 74(6):817–828 DOI 10.1016/j.jprot.2011.02.025.
- Li H, Hong J, Wijayakulathilaka W. 2019. Long non-coding RNA SNHG4 promotes cervical cancer progression through regulating c-Met via targeting miR-148a-3p. *Cell Cycle* **18(23)**:3313–3324 DOI 10.1080/15384101.2019.1674071.
- Li Y, Qin Y, Wang ML, Zhu HF, Huang XE. 2011. The myeloperoxidase –463 G>A polymorphism influences risk of colorectal cancer in southern China: a case-control study. *Asian Pacific Journal of Cancer Prevention* 12(7):1789–1793.
- Lin LH, Chang SJ, Hu RY, Lin MW, Lin ST, Huang SH, Lyu PC, Chou HC, Lai ZY, Chuang YJ, Chan HL. 2014. Biomarker discovery for neuroendocrine cervical cancer. *Electrophoresis* 35(14):2039–2045 DOI 10.1002/elps.201400014.
- Lynge E, Rygaard C, Baillet MV, Dugue PA, Sander BB, Bonde J, Rebolj M. 2014. Cervical cancer screening at crossroads. *APMIS* 122(8):667–673 DOI 10.1111/apm.12279.
- Menderes G, Black J, Schwab CL, Santin AD. 2016. Immunotherapy and targeted therapy for cervical cancer: an update. *Expert Review of Anticancer Therapy* 16(1):83–98 DOI 10.1586/14737140.2016.1121108.
- Mustea A, Heinze G, Sehouli J, Koensgen D, Wolf A, Gutu L, Sofroni D, Pirvulescu C, Braicu EI, Schuster E, Zeillinger R, Tong D. 2007. The -463G/A polymorphism in myeloperoxidase gene and cervical cancer. *Anticancer Research* 27(3B):1531–1535.
- Natter C, Polterauer S, Pils S, Castillo-Tong DC, Zeilinger R, Heinze G, Hefler L, Grimm C. 2016. Association of -463G/A MPO gene polymorphism and risk of cervical intraepithelial neoplasia. Archives of Gynecology 293(4):865–869 DOI 10.1007/s00404-015-3869-9.
- Ndrepepa G. 2019. Myeloperoxidase: a bridge linking inflammation and oxidative stress with cardiovascular disease. *Clinica Chimica Acta* **493**:36–51 DOI 10.1016/j.cca.2019.02.022.
- Ouyang P, Cai J, Gui L, Liu S, Wu NY, Wang J. 2020. Comparison of survival outcomes of neoadjuvant therapy and direct surgery in IB2/IIA2 cervical adenocarcinoma: a retrospective study. *Archives of Gynecology and Obstetrics* **301(5)**:1247–1255 DOI 10.1007/s00404-020-05505-6.

- Pan S, Brentnall TA, Chen R. 2015. Proteomics analysis of bodily fluids in pancreatic cancer. Proteomics 15(15):2705–2715 DOI 10.1002/pmic.201400476.
- Pappa KI, Christou P, Xholi A, Mermelekas G, Kontostathi G, Lygirou V, Makridakis M, Zoidakis J, Anagnou NP. 2018. Membrane proteomics of cervical cancer cell lines reveal insights on the process of cervical carcinogenesis. *International Journal of Oncology* 53(5):2111–2122.
- Parkin DM, Bray F. 2006. Chapter 2: the burden of HPV-related cancers. *Vaccine* 24(Suppl. 3):11–25.
- Rhee EJ, Byrne CD, Sung KC. 2017. The HDL cholesterol/apolipoprotein A–I ratio: an indicator of cardiovascular disease. *Current Opinion in Endocrinology & Diabetes and Obesity* 24(2):148–153 DOI 10.1097/MED.000000000000315.
- Rossmann C, Rauh A, Hammer A, Windischhofer W, Zirkl S, Sattler W, Malle E. 2011. Hypochlorite-modified high-density lipoprotein promotes induction of HO-1 in endothelial cells via activation of p42/44 MAPK and zinc finger transcription factor Egr-1. *Archives of Biochemistry and Biophysics* **509**(1):16–25 DOI 10.1016/j.abb.2011.02.016.
- Saida T, Sakata A, Tanaka YO, Ochi H, Ishiguro T, Sakai M, Takahashi H, Satoh T, Minami M. 2019. Clinical and MRI characteristics of uterine cervical adenocarcinoma: its variants and mimics. *Korean Journal of Radiology* 20(3):364–377 DOI 10.3348/kjr.2018.0458.
- Seoud M, Tjalma WA, Ronsse V. 2011. Cervical adenocarcinoma: moving towards better prevention. *Vaccine* 29(49):9148–9158 DOI 10.1016/j.vaccine.2011.09.115.
- Serafín-Higuera I, Garibay-Cerdenares OL, Illades-Aguiar B, Flores-Alfaro E, Jiménez-López MA, Sierra-Martínez P, Alarcón-Romero LC. 2016. Differential proteins among normal cervix cells and cervical cancer cells with HPV-16 infection, through mass spectrometry-based Proteomics (2D-DIGE) in women from Southern Mexico. *Proteome Science* 14(1):10 DOI 10.1186/s12953-016-0099-4.
- Siegel R, Ma J, Zou Z, Jemal A. 2014. Cancer statistics, 2014. CA: A Cancer Journal for Clinicians 64(1):9–29 DOI 10.3322/caac.21208.
- Siegler E, Reichman Y, Kugelman N, Mackuli L, Lavie O, Ostrovsky L, Shaked-Mishan P, Segev Y. 2019. Low-risk human papillomavirus types in cervical intraepithelial neoplasia 2–3 and in invasive cervical cancer patients. *Journal of Lower Genital Tract Disease* 23(4):248–252 DOI 10.1097/LGT.00000000000486.
- **Smith HO, Tiffany MF, Qualls CR, Key CR. 2000.** The rising incidence of adenocarcinoma relative to squamous cell carcinoma of the uterine cervix in the United States: a 24-year population-based study. *Gynecologic Oncology* **78**(2):97–105 DOI 10.1006/gyno.2000.5826.
- Takahashi TA, Johnson KM. 2015. Menopause. *Medical Clinics of North America* 99(3):521–534 DOI 10.1016/j.mcna.2015.01.006.
- Takayanagi A, Florence TJ, Hariri OR, Armstrong A, Yazdian P, Sumida A, Quadri SA, Cohen J, Tehrani OS. 2019. Brain metastases from cervical cancer reduce longevity independent of overall tumor burden. *Surgical Neurology International* 10:176 DOI 10.25259/SNI\_37\_2019.
- Vizcaino AP, Moreno V, Bosch FX, Munoz N, Barros-Dios XM, Parkin DM. 1998. International trends in the incidence of cervical cancer: I. Adenocarcinoma and adenosquamous cell carcinomas. *International Journal of Cancer* 75(4):536–545 DOI 10.1002/(SICI)1097-0215(19980209)75:4<536::AID-IJC8>3.0.CO;2-U.
- Zadeh Fakhar HB, Zali H, Rezaie-Tavirani M, Darkhaneh RF, Babaabasi B. 2019. Proteome profiling of low grade serous ovarian cancer. *Journal of Ovarian Research* 12(1):64 DOI 10.1186/s13048-019-0535-z.

- Zhang C, Ding W, Liu Y, Hu Z, Zhu D, Wang X, Yu L, Wang L, Shen H, Zhang W, Ren C, Li K, Weng D, Deng W, Ma D, Wang H. 2016. Proteomics-based identification of VDAC1 as a tumor promoter in cervical carcinoma. *Oncotarget* 7(32):52317–52328 DOI 10.18632/oncotarget.10562.
- Zhang X, Guan XJ, Li JQ, Zhang LH, Yang JH. 2018. Clinical analysis of cervical intraepithelial lesion in postmenopausal women. *Zhonghua Fu Chan Ke Za Zhi* 53(10):705–710.
- Zhang Y, Yang X. 2018. Prognostic significance of pretreatment apolipoprotein A–I as a noninvasive biomarker in cancer survivors: a meta-analysis. *Disease Markers* 2018:1–9 DOI 10.1155/2018/1034037.
- Zou S, Pan X, Hua C, Wu M, He B, Chen Z. 2018. Myeloperoxidase –463 G/A polymorphism is associated with lung cancer risk: a meta-analysis with 7420 cases and 9132 controls. *Journal of Cancer Research and Therapeutics* 14(Supplement):S282–S287 DOI 10.4103/0973-1482.202888.