

Research Paper

Development of a Novel Autophagy-related Prognostic Signature for Serous Ovarian Cancer

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Received: 2018.02.17; Accepted: 2018.07.23; Published: 2018.10.18

Abstract

Purpose: Considerable evidence suggests that autophagy plays a crucial role in the biological processes of ovarian cancer. The aim of this study was to develop a novel autophagy-related prognostic signature for serous ovarian cancer.

Methods: A univariate Cox proportional regression model was used to analyze mRNA microarray and clinical data in The Cancer Genome Atlas (TCGA) for the purpose of selecting autophagy-related prognostic genes. A multivariate Cox proportional regression model and the survival analysis were used to develop an eight-gene prognostic signature. The multivariate Cox and stratification analysis suggested that this signature was an independent prognostic factor for serous ovarian cancer patients. Bioinformatics functions were investigated by a principal components analysis and gene set enrichment analysis (GSEA). Finally, the correlation between the prognostic signature and gene mutation status was further analyzed in serous ovarian cancer, and especially with regard to the mutation status of *BRCA1* and *BRCA2* (*BRCA1/2*) genes.

Results: Distinctly different autophagy-related gene expression profiles were identified in normal ovarian tissues and serous ovarian cancer tissues. We profiled an autophagy-related gene set and identified eight genes with significant prognostic values for serous ovarian cancer. Subsequently, an autophagy-related ovarian cancer risk signature was constructed, and patients at a high-risk or low-risk for poor prognosis were identified based on their signature. High-risk patients had significantly shorter overall survival (OS) and disease-free survival (DFS) times than low-risk patients. GSEA results suggested an enhanced intensity of autophagy regulation in high-risk patients when compared with low-risk patients. When studied as an independent prognostic factor for serous ovarian cancer, the significant prognostic value of this signature could be seen in the stratified cohorts. For clinical use, we developed a nomogram that included the prognostic classifier and seven clinical risk factors. Additionally, we identified the 10 most frequently mutated genes found in serous ovarian cancer patients, and analyzed them for their differences in high-risk and low-risk patients. Among 293 patients, 62 had *BRCA1/2* gene mutations, and this result was significantly correlated with the autophagy-related prognostic signature.

Conclusions: Our findings suggest that the eight-gene autophagy-related signature could serve as an independent prognostic indicator for cases of serous ovarian cancer.

Key words: Autophagy; Serous ovarian cancer; Prognostic signature; TCGA

Introduction

As the most lethal gynecological cancer world-wide, the incidence of ovarian cancer has increased [1, 2]. Statistics indicate that 85~90% of patients with ovarian cancer have epithelial ovarian

cancer, and 75% of those patients have serous ovarian cancer [3, 4]. It is well-known that tumor stage, histological grade, and residual tumor size are significant prognostic factors for ovarian cancer

patients [5]. Moreover, recent studies suggest that the mutation status of *BRCA1/2* correlates with patient prognosis and the chemosensitivity of ovarian cancer, and especially serous ovarian cancer [6]. Based on our increased knowledge of how molecular heterogeneity affects ovarian cancer progression, it has been suggested that prognostic biomarkers be included in any clinical analysis of ovarian tumors [7].

As a mechanism for maintaining cellular homeostasis, autophagy is a process that degrades cellular components under conditions of stress or nutrient deprivation [8, 9], and therefore plays an important role in numerous biological and pathological processes, including neurodegenerative diseases, cardiomyopathy and cancer [9, 10]. However, the role of autophagy in tumorigenesis remains controversial.

Previous studies have shown that autophagy can either promote or suppress tumor cell survival at different stages of cancer development [11]. During the early stages of cancer, autophagy promotes the degradation of damaged proteins or organelles to alleviate cellular damage and chromosomal instability, and thus suppresses cancer progression [12-14]. However, once the cancer has formed, autophagy allows tumor cells to survive under stressful conditions, and thus promotes tumor progression [15-17].

The relationship between autophagy and ovarian cancer has been previously reported. For instance, autophagy was shown to promote chemoresistance and help maintain the stemness of ovarian cancer stem cells [18]. Notably, previous studies have focused on exploring the relationship between autophagy and ovarian cancer progression; however, large-scale expression data has rarely been used to investigate the role of autophagy in ovarian cancer progression and prognosis.

Because high-throughput expression data is now available, it has become feasible to use global gene expression data for analyzing the relationship between autophagy-related gene expression and the clinical outcomes of ovarian cancer patients. In this study, we detected an autophagy-related risk signature that contains eight autophagy-related genes. This signature can be used to independently and accurately predict the prognostic value for serous ovarian cancer, and is closely correlated with the mutation status of *BRCA1/2*. Broadly speaking, our study demonstrates that autophagy plays an essential role in cancer, and suggests autophagy-related genes as promising prognostic biomarkers for serous ovarian cancer.

Materials and methods

Autophagy-related gene set

We searched the HADb (Human Autophagy Database) to identify 234 genes that had been described in literature as being involved in the autophagy process [19]. In addition, 394 autophagy-related genes were obtained from the gene set (GO_autophagy, M12441) in the Molecular Signatures Database v4.0 [20].

Patient samples

Firehose was used to generate a normalized gene expression dataset and clinical information pertinent to ovarian serous cystadenocarcinoma patients from The Cancer Genome Atlas (TCGA) database. A full mutation dataset of corresponding patients was obtained from cBioPortal [21, 22]. The nine independent microarray serous ovarian cancer cohorts produced by the HGU133 Plus 2 platform were extracted from the Gene Expression Omnibus (GEO) database (accession number: GSE14407, GSE38666, GSE54388, GSE27651, GSE18520, GSE40595, GSE32062, GSE26193, and GSE51088 [23-31]). The Robust Multi-array Average method was used to normalize raw microarray datasets [32]. The ComBat method was used to remove any batch effects, and thereby eliminate discrepancies between different datasets [33]. When more than one probe matched the same gene ID, the probe with the largest expression value was used for our study. GSE26193 and GSE51088 were selected as our independent validation cohorts because they are the largest datasets containing information on overall survival among the nine different datasets. Overall, 529 samples from microarray of TCGA were used as a discovery cohort, 295 samples from the RNAseq of TCGA were used as an internal validation cohort, 79 samples from GSE26193 and 106 samples from GSE51088 were used as two independent external validation cohorts.

Procedures

The R programming language was used to perform a principle components analysis for the purpose of investigating the distinct gene expression patterns of samples.

An R package survivalROC was used to create receiver operating characteristic (ROC) curves that were used to evaluate the sensitivity and specificity of survival. Next, area under the curve (AUC) values was calculated according to the ROC curves.

GSEA was used to perform the gene set enrichment analysis [20]. A normalized enrichment

score (NES) and false discovery rate (FDR) were used to determine statistical significance.

Statistical analysis

A univariate Cox proportional hazard regression analysis was used to evaluate the association between overall survival time and gene expression values obtained from the autophagy-related gene set in the test cohort. The prognostic value of a gene was considered statistically significant when the p -value was < 0.01 . Next, a multivariate Cox proportional hazards regression analysis was performed using the pool of candidate prognostic genes. The Akaike information criterion (AIC) was used to select the most appropriate model [34]. By weighting the estimated Cox regression coefficients, the eight-gene risk signature was constructed [35]. Patients were classified into high-risk and low-risk groups according to their risk score, and using the median score as the cutoff point. The Kaplan-Meier method and log-rank test were used to estimate the overall survival (OS), disease-free survival (DFS), and progression-free survival (PFS) times of patients. Independent prognostic factors were identified using the multivariate Cox proportional hazards regression model, and hazard ratios (HR) were calculated using the Cox regression model. The prediction accuracy of this risk model was determined by a time-dependent ROC analysis. A prognostic nomogram was created for OS based on the Cox proportional hazard regression model. A concordance index (C-index) was calculated to determine the performance of the nomogram. The nomogram and calibration plot were created using the rms package of R software. The

Chi-square test was used to evaluate the correlation between *BRCA1/2* mutation status and the autophagy-related signature. R software (version 3.4.1) was used to perform the bioinformatics analysis. All tests were two-tailed and p -values < 0.05 were considered statistically significant. All data were analyzed using GraphPad Prism 7 software (GraphPad Software Inc., La Jolla, CA, USA). Differences in clinicopathologic parameters between high-risk and low-risk groups were tested by Student's t test or the X^2 test.

Results

Distinct autophagy-related phenotype expression patterns in normal ovaries and serous ovarian cancer

The general process of our study is described as Fig. 1a. A total of 117 serous ovarian cancer tissues and 52 normal ovarian tissues with mRNA expression data were obtained from 7 GEO datasets. The autophagy-related gene set included 546 genes from the Human Autophagy Database and another gene set (GO_autophagy, M12441) from the Molecular Signatures Database, v4.0. A principal components analysis comparing serous ovarian cancer tissues with normal ovarian tissues based on their autophagy-related genes showed two significantly different distribution patterns. Normal ovarian samples were distributed on the left side, while samples of serous ovarian cancer were distributed on the right side, suggesting distinctly different regulatory roles for autophagy in normal ovaries and serous ovarian cancer (Fig. 1b).

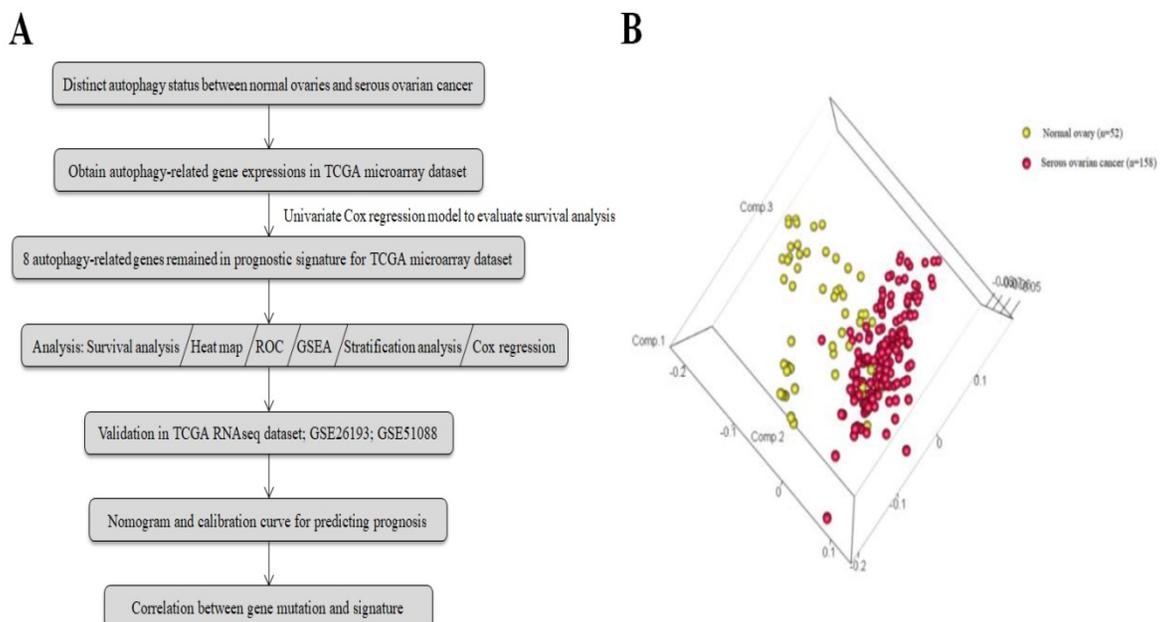


Figure 1. (a) Flowchart for the study; **(b)** Principle components analysis of autophagy-related genes in normal ovaries and serous ovarian cancer.

Identification of an autophagy-related risk signature for the prognosis of serous ovarian cancer

After reviewing the distinct autophagy-related gene expression pattern found in serous ovarian cancer when compared with the pattern found in normal ovaries, we investigated the role that autophagy might play in serous ovarian cancer, and identified an autophagy-related risk signature that might be useful for establishing the prognosis. A Cox regression analysis was performed to investigate the prognostic value of autophagy-related genes found in 529 serous ovarian cancer patients as listed in the TCGA microarray dataset. Twelve genes (*PRKG1*, *FGF7*, *WDR45L*, *NRG4*, *BLOC1S1*, *IL24*, *PEX3*, *CDKN1B*, *CD93*, *SIRT2*, *PDK4*, and *SLC22A3*) were found to be significantly associated with OS in serous ovarian cancer patients ($p < 0.01$, Table 1). The AIC was used to select the most appropriate model, which was identified as a model comprised of eight different genes (*BLOC1S1*, *IL24*, *NRG4*, *PDK4*, *PEX3*, *PRKG1*, *SIRT2*, and *WDR45L*). Among the eight genes, four genes (*BLOC1S1*, *IL24*, *NRG4*, and *WDR45L*) were identified as protective factors ($HR < 1$), while the other four genes (*PKD4*, *PEX3*, *PRKG1*, and *SIRT2*) were identified as risk factors ($HR > 1$).

Table 1. Top 12 genes significantly associated with the survival time of patients in the training dataset (N = 529)

Gene name	HR	P-value
PRKG1	1.425733	0.001275
FGF7	1.2575	0.001398
WDR45L	0.764353	0.002558
NRG4	0.858855	0.00393
BLOC1S1	0.707778	0.004328
IL24	0.618953	0.004421
PEX3	1.292997	0.00461
CDKN1B	1.280662	0.00528
CD93	1.195552	0.005346
SIRT2	1.377175	0.006193
PDK4	1.188561	0.006719
SLC22A3	1.246166	0.006826

HR: hazard ratio

We then used a multivariate Cox regression model to develop the following autophagy-related risk signature associated with the survival of serous ovarian cancer patients [36].

$$\begin{aligned} \text{Risk score} = & (-0.4175 \times \text{BLOC1S1 expression}) - (0.4852 \times \\ & \text{IL24 expression}) - (0.1413 \times \text{NRG4 expression}) + \\ & (0.1443 \times \text{PDK4 expression}) + (0.1679 \times \\ & \text{PEX3 expression}) + (0.2518 \times \text{PRKG1 expression}) + \\ & (0.4507 \times \text{SIRT2 expression}) - (0.2447 \times \\ & \text{WDR45L expression}). \end{aligned}$$

We then calculated the risk score of each patient according to their eight-gene expression signature, and used the median risk value as a cutoff point for classifying patient into a high-risk group ($n = 264$) and a low-risk group ($n = 265$), respectively. Patients in the high-risk group had a shorter overall survival time than patients in the low-risk group (median time = 2.84 years vs. 4.72 years, $p < 0.001$, Fig. 2a). In addition, we also found that patients in the high-risk group had shorter DFS times (median time = 1.15 years vs. 1.77 years, $p < 0.001$, Fig. 2b).

Serous ovarian cancer cases were divided into high-risk and low-risk groups according to the median eight-gene risk score (Fig. 2c). Protective and risky genes showed distinct expression patterns according to the risk value. High-risk patients expressed higher levels of risky genes (*PKD4*, *PEX3*, *PRKG1*, and *SIRT2*), while low-risk patients expressed higher levels of protective genes (*BLOC1S1*, *IL24*, *NRG4*, and *WDR45L*).

ROC curves for 3-year survival were used to reveal the predictive performance of the eight-gene risk signature (Fig. 2d). The 3-year AUC of our signature was 0.703, which was obviously higher than that of age (AUC = 0.626), neoplasm cancer status (AUC = 0.615), residual tumor size (AUC = 0.603), tumor stage (AUC = 0.581), and tumor grade (AUC = 0.544). These results indicated that the ability to predict survival of serous ovarian cancer patients was improved by using the eight-gene risk signature when compared to using clinical factors.

Due to such different prognostic outcomes, we sought to investigate possible differences between the high-risk and low-risk groups by using GSEA, which verified an enhanced autophagy-related signature in the high-risk group (Fig. 2e). Therefore, we propose the existence of an intense regulatory role for autophagy in high-risk serous ovarian patients.

Associations between the autophagy-related risk signature and clinicopathologic features in serous ovarian cancer patients

An analysis was performed to explore associations between various clinical parameters and our risk signature in serous ovarian cancer patients (Table 2). Results showed that the signature was significantly associated with age ($p = 0.015$), Karnofsky performance status (KPS, $p < 0.0001$), and neoplasm cancer status ($p = 0.0443$). Additionally, we found that the signature was only marginally associated with therapy outcome ($p = 0.0512$) and therapy type ($p = 0.0736$).

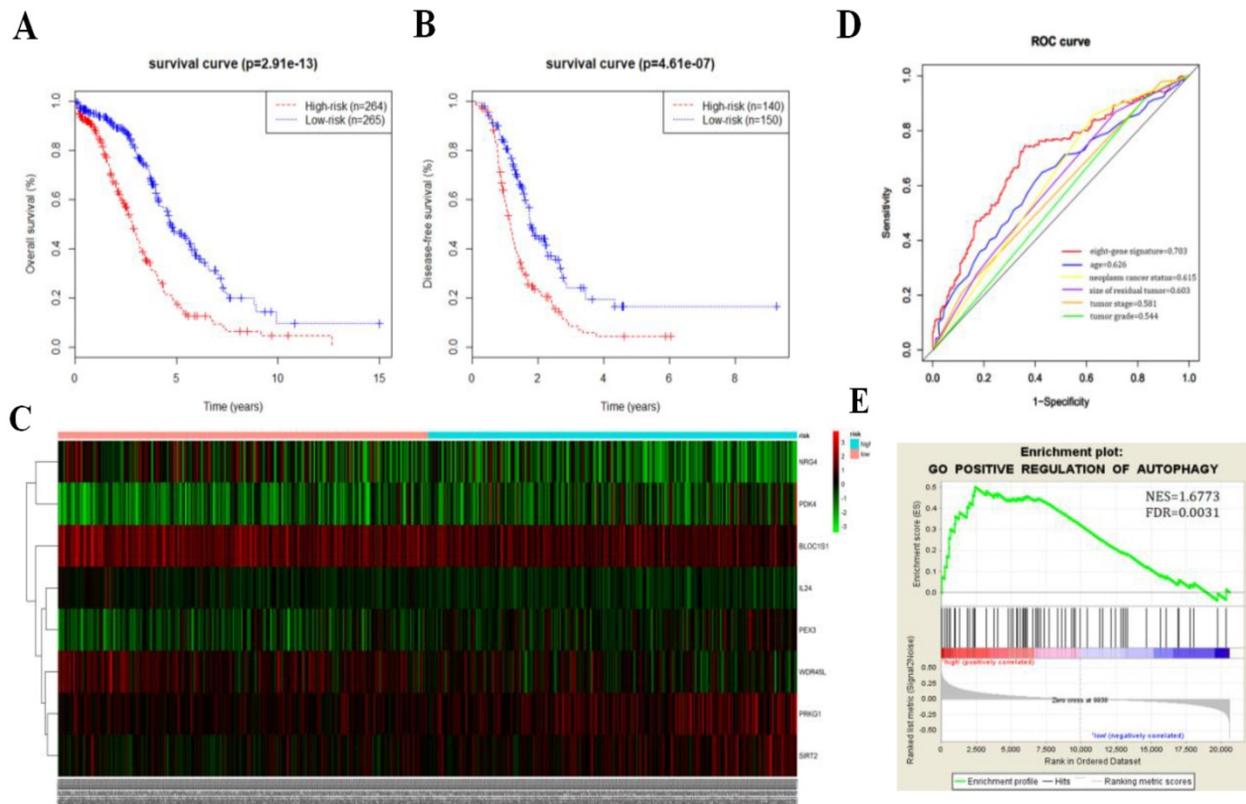


Figure 2. Kaplan-Meier curves for OS (a) and DFS (b) in the high-risk and low-risk groups when stratified by the autophagy-related signature; (c) Hierarchical clustering of high-risk and low-risk patients expressing the eight genes. Red indicates higher expression and green indicates lower expression; (d) ROC analysis at 3-year of overall survival for the eight-gene signature and classical clinicopathologic parameters in the TCGA cohort; (e) A GSEA comparing the autophagy phenotypes in high-risk and low-risk groups.

Table 2. Clinical characteristics of serous ovarian cancer patients by autophagy-related signature.

Characteristics	High-risk (n = 264)	Low-risk (n = 265)	p-value
Age			0.0015
Mean (years)	61.20	58.02	
FIGO stage			0.3899
I	10 (3.8%)	5 (1.9%)	
II	11 (4.2%)	14 (5.3%)	
III	197 (74.6%)	211 (79.6%)	
IV	44 (16.7%)	34 (12.8%)	
NA	2 (0.7%)	1 (0.4%)	
Residual tumor size			0.3594
No macroscopic	53 (20.1%)	54 (20.4%)	
1-10nm	118 (44.7%)	114 (43.0%)	
11-20nm	14 (5.3%)	20 (7.5%)	
>20nm	56 (21.2%)	44 (16.6%)	
NA	23 (8.7%)	33 (12.5%)	
Lymphatic invasion			0.8329
Yes	67 (25.4%)	65 (24.5%)	
No	36 (13.6%)	41 (15.5%)	
NA	161 (61%)	159 (60.6%)	
KPS			<0.0001
40	0 (0.0%)	2 (0.7%)	
60	6 (2.3%)	9 (3.4%)	
80	6 (2.3%)	34 (12.8%)	
100	1 (0.3%)	6 (2.3%)	
NA	251 (95.1%)	214 (80.8%)	
Venous invasion			0.4883
Yes	42 (15.9%)	43 (16.2%)	
No	38 (14.4%)	29 (10.9%)	
NA	184 (69.7%)	193 (72.9%)	
Therapy type			0.0736

Chemotherapy	216 (81.8%)	237 (89.4%)	
Immunotherapy	3 (1.1%)	2 (0.8%)	
Hormone therapy	3 (1.1%)	4 (1.5%)	
Targeted molecular therapy	8 (3.1%)	2 (0.8%)	
NA	34 (12.9%)	20 (7.5%)	
Therapy outcome			0.0512
Complete remission	132 (50.0%)	165 (62.3%)	
Partial remission	35 (13.3%)	24 (9.1%)	
Progressive disease	18 (6.8%)	18 (6.8%)	
Stable disease	19 (7.2%)	11 (4.2%)	
NA	60 (22.7%)	47 (17.7%)	
Neoplasm cancer status			0.0443
With tumor	178 (67.4%)	155 (58.5%)	
Tumor free	55 (20.8%)	80 (30.2%)	
NA	31 (11.8%)	30 (11.3%)	
New neoplasm event type			0.7736
Recurrence	128 (48.5%)	135 (51.0%)	
Progression of disease	12 (4.5%)	15 (5.7%)	
Locoregional disease	3 (1.1%)	3 (1.1%)	
Metastatic	1 (0.4%)	0 (0.0%)	
NA	120 (45.5%)	112 (42.2%)	
Radiation therapy			0.1020
Yes	4 (1.5%)	1 (0.4%)	
No	250 (94.7%)	260 (98.1%)	
NA	10 (3.8%)	4 (1.5%)	
Tumor grade			0.7955
G1	2 (0.7%)	4 (1.5%)	
G2	33 (12.5%)	33 (12.5%)	
G3	223 (84.5%)	221 (83.4%)	
G4	0 (0.0%)	1 (0.4%)	
NA	6 (2.3%)	6 (2.2%)	
Race			0.4475
White	227 (86.0%)	231 (87.2%)	
Black or African American	12 (4.6%)	11 (4.2%)	
Asian	8 (3.0%)	11 (4.2%)	
American Indian	3 (1.1%)	0 (0.0%)	
NA	14 (5.3%)	12 (4.4%)	

KPS: Karnofsky performance score

Table 3. Univariate and multivariate Cox regression analyses of the autophagy-related signature and clinical characteristics predictive of overall survival and disease-free survival.

Variable	Overall survival				Disease-free survival			
	Univariate		Multivariate		Univariate		Multivariate	
	HR	p-value	HR	p-value	HR	p-value	HR	p-value
Eight-gene signature (low-risk vs high-risk)	0.414	<0.001	0.463	<0.001	0.488	<0.001	0.599	0.007
Age (old age vs young age)	1.019	0.001	1.016	0.033	1.012	0.058		
Tumor grade (G3-G4 vs G1-G2)	1.180	0.056			1.219	0.051		
FIGO stage (stage III-IV vs stage I-II)	1.658	0.003	1.837	0.040	1.432	0.084		
Residual tumor size (no macroscopic vs visible macroscopic)	0.436	<0.001	0.578	0.037	0.478	0.002	0.487	0.016
BRCA1/2 mutation stage (no mutation vs mutation)	2.021	0.001	1.874	0.004	1.752	0.020	1.602	0.065
Lymphatic invasion (invasion vs no invasion)	1.422	0.114			1.491	0.148		
Venous invasion (no invasion vs invasion)	0.973	0.917			0.714	0.311		

HR: hazard ratio

Application of the autophagy-related signature in stratified serous ovarian cancer cohorts

In order to investigate the prognostic value of the risk signature in stratified cohorts, patients were

classified by residual tumor size, age, and lymphatic invasion status. In all cohorts, the high-risk group had shorter OS (Fig. 3) and DFS times (Fig. 4) than the low-risk group. These results suggest that classification of the autophagy-related risk signature

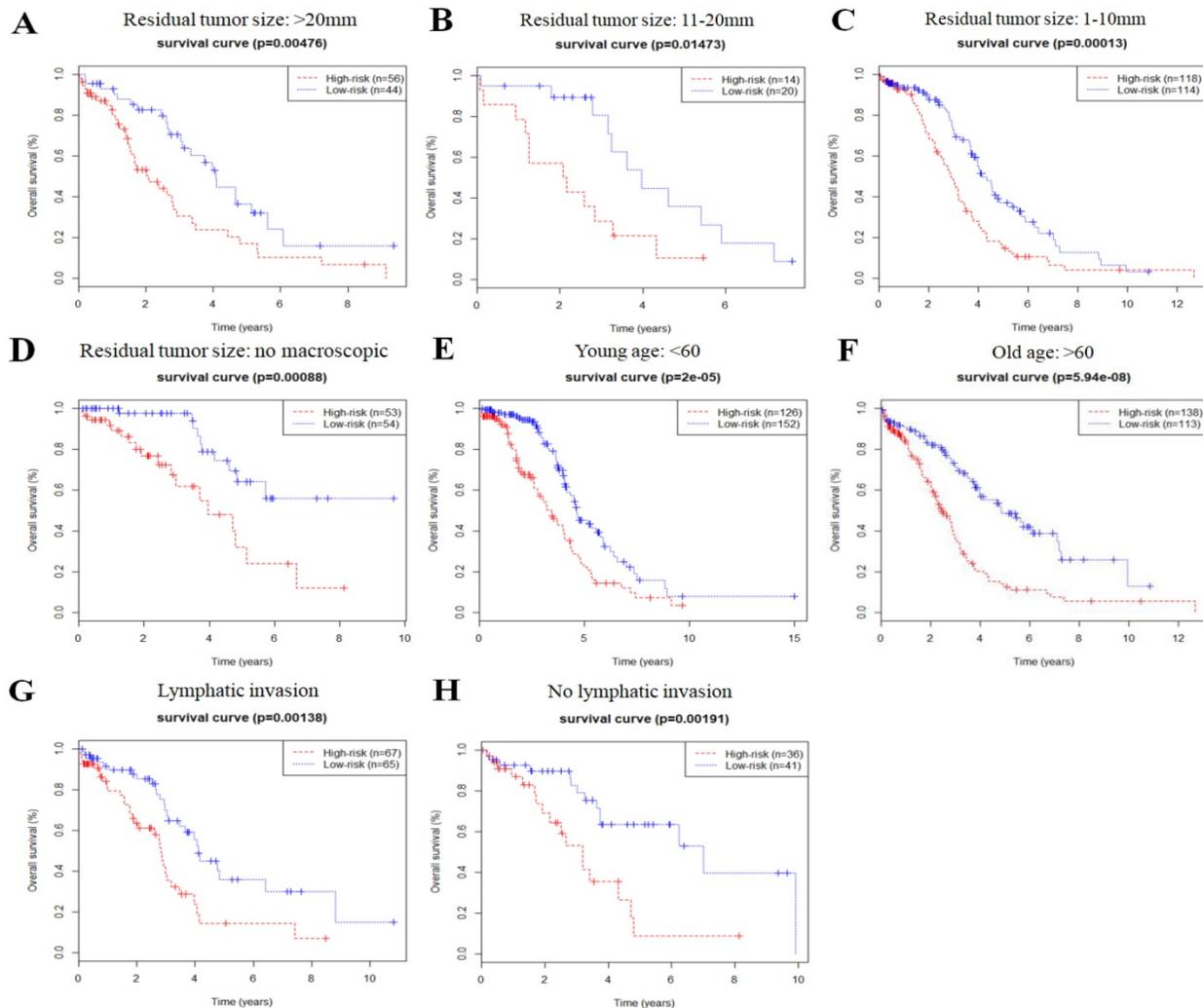


Figure 3. Overall survival of autophagy-related signature in cohorts stratified by residual tumor size (a-d), age (e-f), and lymphatic invasion status (g-h).

can be used to precisely identify patients with the poor prognosis, without the consideration of clinical parameters.

The autophagy-related signature is an independent prognostic factor for serous ovarian cancer patients

We performed univariate Cox and multivariate Cox regression analyses of data in the TCGA dataset in order to investigate whether the autophagy-related signature was an independent factor correlated with OS and DFS (Table 3). The univariate Cox analysis showed that the autophagy-related signature, age, tumor stage, residual tumor size, and *BRCA1/2* mutation status were all correlated with the overall survival of serous ovarian cancer patients; therefore, those factors were included in a multivariate Cox analysis. Moreover, the risk signature, residual tumor size, and *BRCA1/2* mutation status each showed a significant correlation with DFS. Thus our results indicated that the autophagy-related signature might

be an independent prognostic factor of OS and DFS when adjusted by those factors.

Validation of the eight-gene autophagy-related prognostic signature using three independent cohorts

To validate the eight-gene risk signature in other datasets, we calculated the risk score for each patient in the TCGA RNAseq dataset as an internal validation, in the GSE26193 and GSE51088 datasets as the independent external validations using same formula. The patients in these three datasets were then divided into a high-risk and a low-risk group based on the median risk score. As expected, the survival curves showed that in the TCGA RNAseq dataset, which including 295 patients, the overall survival time and disease-free survival time were much lower in the high-risk group (OS, median = 2.90 years vs. 4.33 years, $p < 0.001$, Fig. 5a; DFS, median = 1.22 years vs. 1.98 years, $p < 0.001$, Fig. 5b). While there was no DFS information in GSE26193, we used the

progression-free survival time to create survival curves. Similarly, in dataset GSE26193, which including 79 patients, the high-risk patients had lower overall survival time and progression-free survival time than the low-risk patients (OS, median = 2.42 years vs. 3.90 years, $p = 0.026$, Fig.5c; PFS, median = 1.66 years vs. 1.87 years, $p = 0.042$, Fig. 5d). In addition, we also found in dataset GSE51088, which including 106 patients, the overall survival time were much lower in the high-risk group (OS, median = 3.93 years vs. 5.28 years, $p = 0.034$, Fig. 5e).

We next arrayed the patients according to the risk score in the validation cohorts. The heat map displayed the expression patterns of the eight autophagy-related prognostic genes in the two

groups. As expected, in the TCGA RNAseq (Fig. 5h), GSE26193 (Fig.5f) and GSE51088 (Fig. 5g) datasets, the expression levels of the four protective genes were down-regulated in the high-risk group, while the risk factor genes were expressed at high levels. In contrast, those genes showed an opposite expression pattern in the low-risk group.

Consistent with the finding in the test cohort, the autophagy-related signature was further validated by the TCGA RNAseq, GSE26193 and GSE51088 datasets, confirming its power to independently predict the prognosis (Table 4). All these three cohorts validated that this novel autophagy-related signature could work as an independent predictor for the prognosis of serous ovarian cancer.

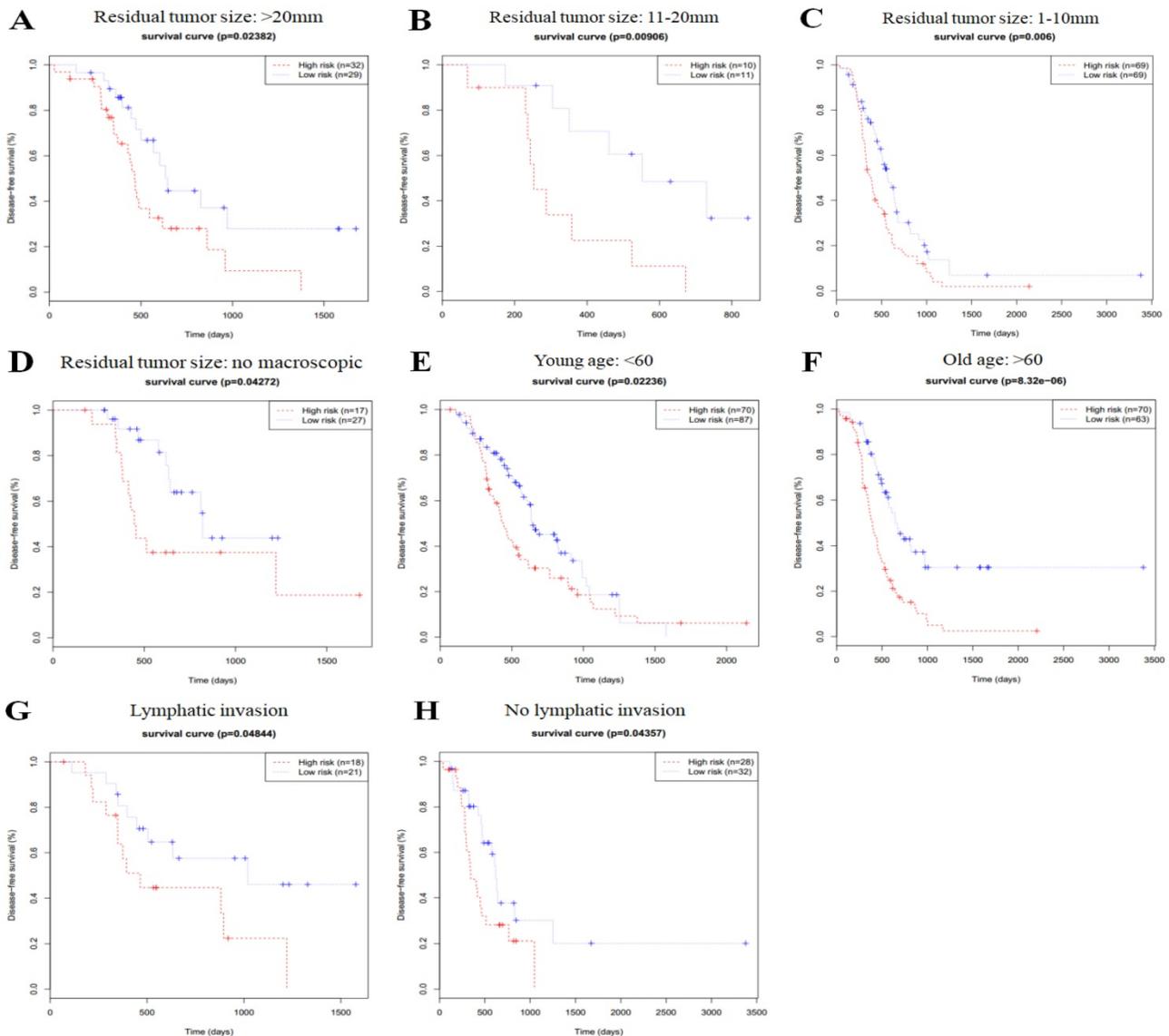


Figure 4. Disease-free survival of autophagy-related signature in cohorts stratified by residual tumor size (a-d), age (e-f), and lymphatic invasion status (g-h).

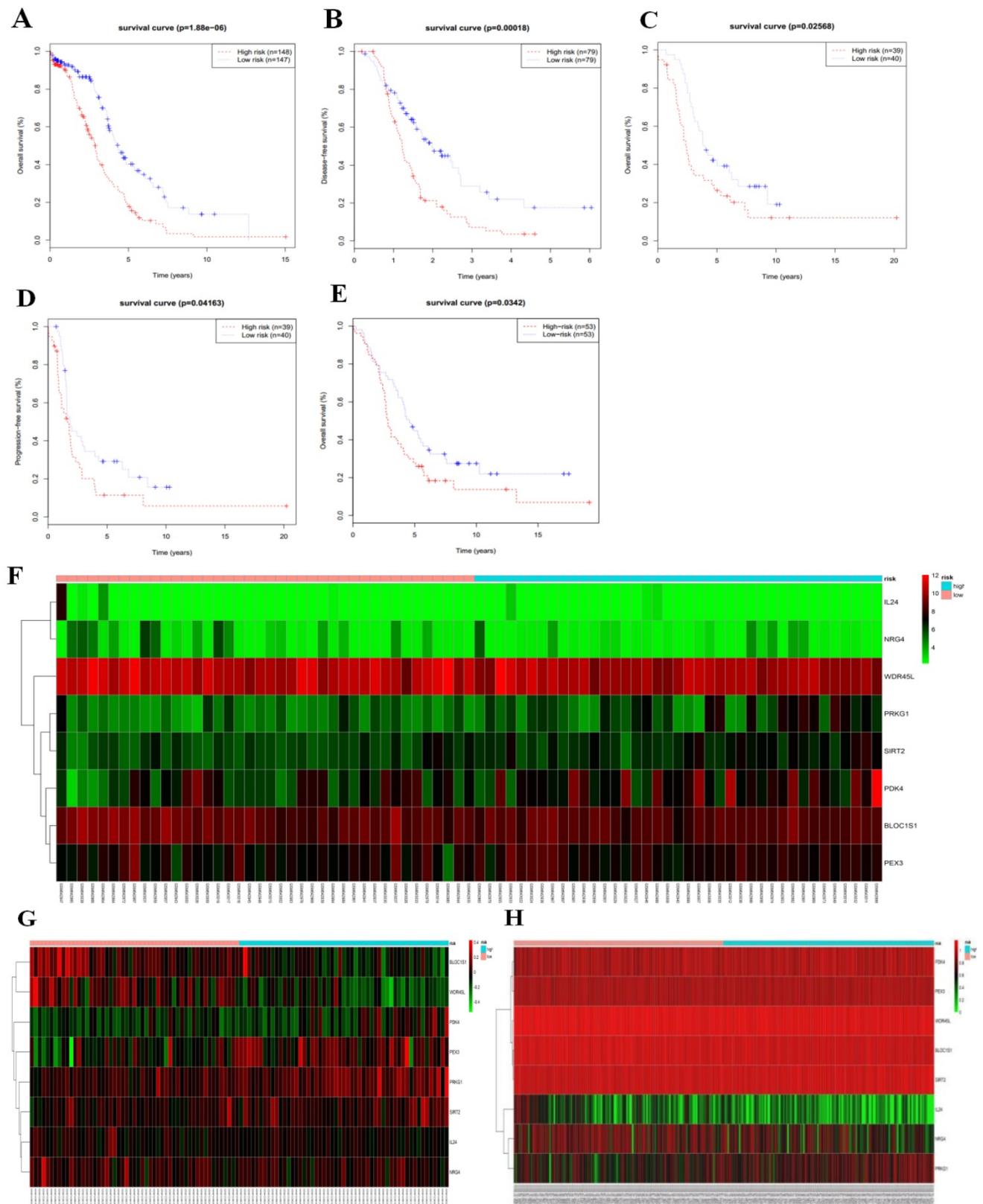


Figure 5. Based on median risk score, the eight-gene risk signature divided patients into high-risk and low-risk groups with distinct prognosis in the TCGA RNAseq cohort (a-b), GSE26193 cohort (c-d), and GSE51088 cohort (e); (f-h) Hierarchical clustering of high-risk and low-risk patients expressing the eight genes in GSE26193 (f), GSE51088 (g), and TCGA RNAseq (h). Red indicates higher expression and green indicates lower expression.

Table 4. Univariate and multivariate Cox regression analyses of the autophagy-related signature and clinical characteristics predictive of OS, DFS and PFS in each validation cohort

	Overall survival				Disease-free survival				Progression-free survival			
	Univariate		Multivariate		Univariate		Multivariate		Univariate		Multivariate	
	HR	p-value	HR	p-value	HR	p-value	HR	p-value	HR	p-value	HR	p-value
TCGA RNAseq cohort (n = 295)												
Eight-gene signature (low-risk vs high-risk)	0.477	<0.001	0.468	<0.001	0.491	<0.001	0.517	0.001				
Age (old age vs young age)	1.022	0.003	1.016	0.033	1.014	0.058						
Tumor grade (G3-G4 vs G1-G2)	1.663	0.038	1.777	0.120	1.337	0.051						
FIGO stage (stage III-IV vs stage I-II)	1.487	0.057			1.216	0.084						
Residual tumor size (visible vs no macroscopic)	1.587	0.054			1.968	0.002	1.889	0.023				
<i>BRCA1/2</i> mutation stage (mutation vs no mutation)	0.613	0.033	0.758	0.237	0.584	0.020						
Lymphatic invasion (invasion vs no invasion)	1.319	0.359			1.109	0.148						
Venous invasion (no invasion vs invasion)	0.927	0.830			0.603	0.311						
GSE26193 cohort (n = 79)												
Eight-gene signature (low-risk vs high-risk)	0.564	0.028	0.552	0.023					0.599	0.044	0.553	0.021
Tumor grade (G1-G2 vs G3-G4)	0.877	0.351							1.054	0.719		
FIGO stage (stage III-IV vs stage I-II)	1.699	0.006	1.713	0.005					1.827	0.001	1.883	0.001
GSE51088 cohort (n = 106)												
Eight-gene signature (low-risk vs high-risk)	0.624	0.035	0.556	0.021								
Age (old age vs young age)	1.023	0.012	1.008	0.148								
Tumor grade (G3-G4 vs G1-G2)	1.235	0.194										
FIGO stage (stage III-IV vs stage I-II)	1.799	0.023										
Disease status (not free vs free)	60.734	<0.001	66.097	<0.001								

HR: hazard ratio

Nomogram development for the prediction of prognostic risk

In order to provide clinicians with a quantitative approach for predicting cancer survival, we assembled a nomogram that integrated both the risk signature and various clinicopathologic risk factors. The nomogram was constructed to estimate 1-, 2-, 3-, 4-, and 5-year survival probabilities, and it showed that the signature risk score was the most essential factor among the different variables analyzed (Fig. 6a). A C-index was calculated to estimate the performance of this nomogram (Fig. 6b). The C-index of our nomogram was 0.778. A calibration plot showed optimal agreement when compared with an ideal model.

Association between the risk signature and gene mutation status, and especially *BRCA1/2* mutation status

The ten genes (*BRCA1*, *BRCA2*, *CSMD3*, *FAT3*, *HMCN1*, *MUC16*, *RYR2*, *TP53*, *TTN*, and *USH2A*) with the highest mutation rates in 293 patients were selected for further study (Fig. 7a). The association between autophagy-related signature and each of those ten genes is shown in Table 5. The data indicated that mutation of the *BRCA1* gene ($p = 0.0359$) helped to extend the survival of serous ovarian cancer patients, while mutation of the *BRCA2* ($p = 0.0691$) or *TTN* ($p = 0.0884$) gene had a marginally favorable effect on survival. The mutation status of

BRCA1/2 was recently reported to be an important favorable survival predictor in ovarian cancer [6]. Therefore, we further investigated the effect of combined *BRCA1* and *BRCA2* gene mutations. As expected, the *BRCA1/2* mutation status was significantly correlated with our risk signature ($p = 0.0062$). Therefore, the patients were arranged according to their risk scores. The genes in our autophagy-related signature showed distinct expression patterns according to the *BRCA1/2* mutation status (Fig. 7b).

Discussion

Ovarian cancer is a progressive disease that urgently needs reliable prognostic markers that can aid in its diagnosis and treatment. Computational models have recently been used to explore possible mRNA and non-coding RNA biomarkers for ovarian cancer. Furthermore, numerous studies have focused on the role that autophagy plays in tumorigenesis and the results produced by cancer therapies. While most studies of autophagy-related genes have used cell lines or animal models, our current study used high-throughput expression profiling of autophagy-related genes to investigate the progression and outcomes of serous ovarian cancer patients.

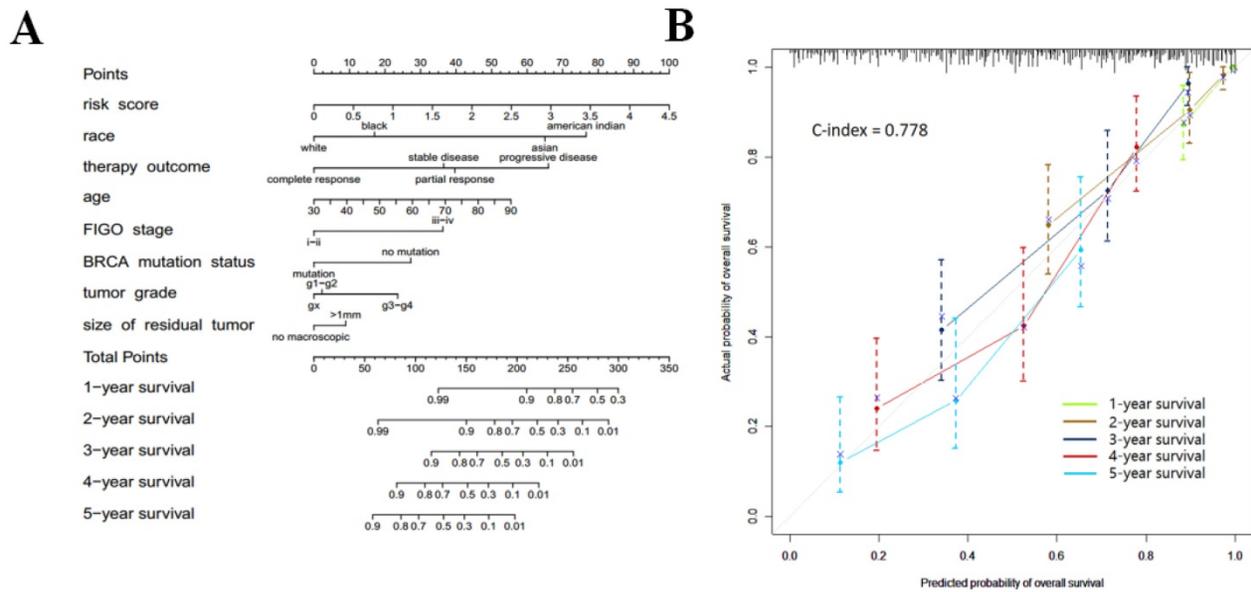


Figure 6. (a) Prognostic nomogram for serous ovarian cancer patients; (b) Calibration curves for the nomogram at different time points.

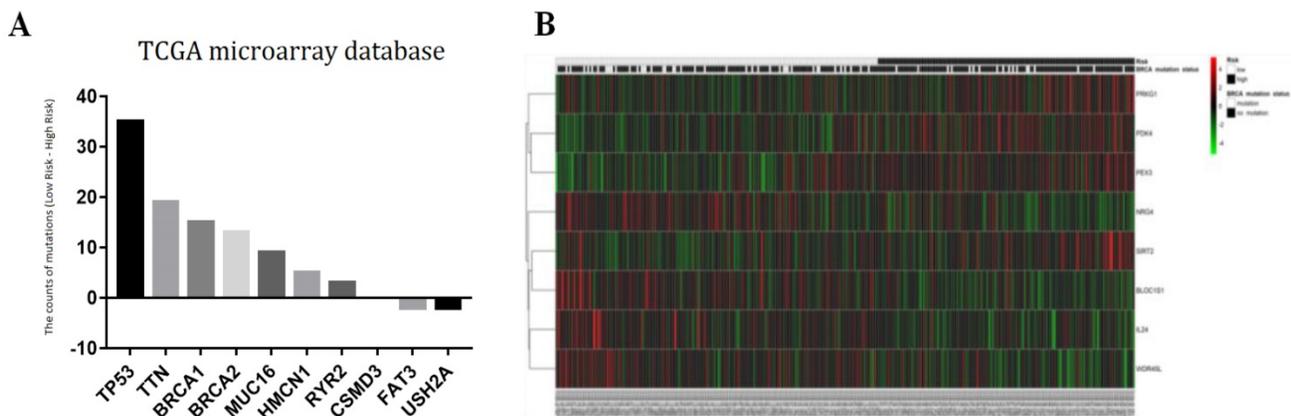


Figure 7. (a) Associations between the risk signature and *BRCA1/2* mutation status with its related genes; (b) Top ten genes with highest mutation rates in serous ovarian cancer patients of TCGA microarray database

As a novel therapeutic method, the influence of autophagy on cancer has been discussed for several years. Sui et al. suggested that autophagy might influence how patients respond to chemoradiotherapy [37]. In our study, we first examined differences in the autophagy status of normal ovarian tissues and serous ovarian cancer tissues, and then explored the role that autophagy might play in serous ovarian cancer. Although the controversial and complex role played by autophagy in cancer development has been widely discussed during the past decade, researchers have not reached a conclusion. Therefore, development of a meaningful autophagy-related risk signature should not only benefit the science of disease prognosis, but also provide clinicians with a rationale for using autophagy targeting therapies for treating serous ovarian cancer patients. In this study, we profiled an

eight-gene autophagy-related model gene signature that could predict the prognosis of serous ovarian cancer, which was further validated in the GEO database. As described in GSEA, high-risk group showed positive regulation of autophagy, which suggested that the autophagy level might be increased in high-risk group with worse prognostic outcomes. As we all known, cisplatin resistant ovarian cancer cells showed higher autophagy level compared with the cisplatin sensitivity ovarian cancer cells [38], which is in accordance with our results. Previous studies have demonstrated that patients with *BRCA1/2* gene mutations had higher chemosensitivity and better prognosis [6], while our results showed that high-risk group correlated with lower possibility of *BRCA1/2* gene mutations and higher autophagy level, which is in accordance with the conclusion that chemoresistance of ovarian cancer cells correlated

with higher autophagy level. Kwon et al performed a cDNA analysis which showed that among four high-risk genes, *PDK4* was greatly overexpressed in ovarian cancer stem cells when compared with parental ovarian cancer cells [39]. Additionally, S Sun et al demonstrated that *FAM210B* expression loss led to the regulation of *PDK4* expression, which is significantly associated with tumor metastasis and decreased survival of ovarian cancer patients [40]. Knockdown of *SIRT2* was also found to be associated with ovarian cancer cell proliferation, migration, and invasion [41]. However, with regards to other genes, most studies have focused on their role in autophagy rather than ovarian cancer. Zhao et al. reported that autophagy contributed to the resistance of osteosarcoma cells to doxorubicin [42], while *IL24* could suppress autophagy, and thus reverse the multiple drug resistance of osteosarcoma [43]. Moreover, other studies have reported that high levels of *PEX3* expression can induce pexophagy, and that ubiquitination of *PEX3* is not essential for pexophagy [44].

Table 5. Gene mutation status in serous ovarian cancer according to the autophagy-related signature.

Characteristics	High-risk (n = 130)	Low-risk (n = 163)	p-value
<i>BRCA1</i>			0.0359
mutation	9 (6.9%)	24 (14.7%)	
no mutation	121 (93.1%)	139 (85.3%)	
<i>BRCA2</i>			0.0691
mutation	9 (6.9%)	22 (13.5%)	
no mutation	121 (93.1%)	141 (86.5%)	
<i>BRCA1/2</i>			0.0062
mutation	18 (13.8%)	44 (27.0%)	
no mutation	112 (86.2%)	119 (73.0%)	
<i>CSMD3</i>			0.5995
mutation	10 (7.7%)	10 (6.1%)	
no mutation	120 (92.3%)	153 (93.9%)	
<i>FAT3</i>			0.3241
mutation	10 (7.7%)	8 (4.9%)	
no mutation	120 (92.3%)	155 (95.1%)	
<i>HMCN1</i>			0.5481
mutation	8 (6.2%)	13 (8.0%)	
no mutation	122 (93.8%)	150 (92.0%)	
<i>MUC16</i>			0.1612
mutation	7 (5.4%)	16 (9.8%)	
no mutation	123 (94.6%)	147 (90.2%)	
<i>RYR2</i>			0.7849
mutation	7 (5.4%)	10 (6.1%)	
no mutation	123 (94.6%)	153 (93.9%)	
<i>TP53</i>			0.2953
mutation	125 (96.2%)	160 (98.2%)	
no mutation	5 (3.8%)	3 (1.8%)	
<i>TTN</i>			0.0884
mutation	22 (16.9%)	41 (25.2%)	
no mutation	108 (83.1%)	122 (74.8%)	
<i>USH2A</i>			0.3178
mutation	12 (9.2%)	10 (6.1%)	
no mutation	118 (90.8%)	153 (93.9%)	

BRCA1/2 mutation status has been well studied as an important molecular marker for ovarian cancer. Ovarian cancer patients with *BRCA1/2* mutations have shorter survival times and higher degrees of chemosensitivity [6]. It is known that PARP inhibitors can inhibit the proliferation of cancer cells with *BRCA1/2* mutations. Moreover, recent studies have demonstrated that the PARP inhibitor can induce autophagy in *BRCA1* or *BRCA2* mutated breast cancer cells, and that inhibition of autophagy results in the partial inhibition of PARP inhibitor induced apoptosis [45]. These results suggest that autophagy plays an essential role in regulating *BRCA1/2* mutated cancer cell proliferation and apoptosis. Coincidentally, another study showed significant decreases in serum *LC3* and *Bcl-2* expression levels in breast cancer patients with the 5382insC *BRCA1* mutation [46]. In our current study, we found a close correlation between *BRCA1/2* mutation status and the risk signature for ovarian cancer, which suggests that autophagy might contribute to the *BRCA1/2* mutations found in serous ovarian cancer cells; however, the precise mechanism for this contribution remains unknown. Thus additional studies should be considered to further explore the role played by autophagy in *BRCA1/2* mutated serous ovarian cancer.

Further studies showed that our signature was an independent prognostic factor for serous ovarian cancer, which suggests that autophagy status might serve as an accurate prognostic indicator. In this study, patients with higher risk scores significantly showed worse prognosis. Our nomogram demonstrated that the risk score associated with the signature for each patient was the most important variable, indicating its significant meaning when predicting the prognosis of serous ovarian cancer patients. Interestingly, by building this nomogram, we found that American Indian and Asian patients showed poorer prognosis than White patients. As we all known, patients with old ages, visible residual tumor sizes, higher FIGO stages and tumor grades all showed worse prognostic outcomes, which is in accordance with our results. Additionally, patients with *BRCA1/2* gene mutations showed higher chemosensitivity and thus caused better prognosis [6]. Thus, this nomogram provided an individualized thorough estimate of survival, rather than using specific covariates. Moreover, perfect agreement was seen between actual observations and predictions in the calibration plot. Thus, the eight-gene signature could precisely predict the prognosis for serous ovarian cancer patients.

The advantage of our study is that we performed a systematic analysis of microarray data and RNAseq

data, which provided a robust statistical approach for exploring the role of autophagy in serous ovarian cancer. Although the eight-gene signature allowed for a practical independent prognosis for serous ovarian cancer, our study does have certain limitations. First, only autophagy-related genes were included in the study, and the risk signature does not represent the entire gene transcription profile associated with serous ovarian cancer. Second, our study is restricted because it is retrospective, and our results should be validated in prospective investigations.

In summary, we developed an autophagy-related gene expression model that could independently predict the overall survival of serous ovarian cancer patients. Furthermore, our results suggest that use of a targeted autophagy therapy might be a promising future strategy for treating serous ovarian cancer. Further investigations into the molecular mechanisms of autophagy will demonstrate how autophagy affects survival, and provide new suggestions for treating ovarian cancer patients. Thus, our eight-gene autophagy-related signature may predict the overall survival of serous ovarian cancer patients, and also guide the therapeutic approaches used for those patients.

Abbreviations

AIC: Akaike information criterion; AUC: area under the curve; C-index: concordance index; DFS: disease-free survival; FDR: false discovery rate; GEO: Gene Expression Omnibus; GSEA: gene set enrichment analysis; HADb: Human Autophagy Database; HR: hazard ratio; KPS: Karnofsky performance status; NES: normalized enrichment score; OS: overall survival; PFS: progression-free survival; ROC: receiver operating characteristic; TCGA: The Cancer Genome Atlas.

Acknowledgements

Supported by grant from the Shenyang Science and Technology Bureau (award number: LNCCC-A01-2015)

Competing Interests

The authors have declared that no competing interest exists.

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