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Prognostic gene expression signature for high-grade serous ovarian cancer

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Abstract

Background: Median overall survival (OS) for women with high-grade serous ovarian cancer (HGSOC) is ~4 years, yet survival varies widely between patients. There are no well-established, gene expression signatures associated with prognosis. The aim of this study was to develop a robust prognostic signature for OS in patients with HGSOC.

Patients and methods: Expression of 513 genes, selected from a meta-analysis of 1455 tumours and other candidates, was measured using NanoString technology from formalin-fixed paraffin-embedded tumour tissue collected from 3769 women with HGSOC from multiple studies. Elastic net regularization for survival analysis was applied to develop a prognostic model for 5-year OS, trained on 2702 tumours from 15 studies and evaluated on an independent set of 1067 tumours from six studies.

Results: Expression levels of 276 genes were associated with OS (false discovery rate < 0.05) in covariate-adjusted single-gene analyses. The top five genes were *TAPI*, *ZFH4*, *CXCL9*, *FBN1* and *PTGER3* ($P < 0.001$). The best performing prognostic signature included 101 genes enriched in pathways with treatment implications. Each gain of one standard deviation in the gene expression score conferred a greater than twofold increase in risk of death [hazard ratio (HR) 2.35, 95% confidence interval (CI) 2.02–2.71; $P < 0.001$]. Median survival [HR (95% CI)] by gene expression score quintile was 9.5 (8.3 to –), 5.4 (4.6–7.0), 3.8 (3.3–4.6), 3.2 (2.9–3.7) and 2.3 (2.1–2.6) years.

Conclusion: The OTTA-SPOT (Ovarian Tumor Tissue Analysis consortium - Stratified Prognosis of Ovarian Tumours) gene expression signature may improve risk stratification in

[†]The AOCS Study members are listed in full in Appendix 1.

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DISCLOSURE

BYK served on Invitae Corporation's Advisory Board from 2017 to 2018. IAM has acted on the Advisory Boards for AstraZeneca, Clovis Oncology, Tesaro, Carrick Therapeutics and Takeda. His institution receives funding from AstraZeneca. RG is on the Advisory Boards for AstraZeneca, Tesaro, Clovis and Immunogen and does consultancy work for SOTIO. She has received support to attend conferences from AstraZeneca, Roche and Tesaro. Her institution has received research funding from Boehringer Ingelheim and Lilly/Ignyta and she is the national co-ordinating investigator for the UK for trials sponsored by AstraZeneca and Tesaro and site principal investigator for trials sponsored by AstraZeneca, Tesaro, Immunogen, Pfizer, Lilly and Clovis. PAF has received grants from Novartis, BioNtech and Cepheid as well as personal fees from Novartis, Roche, Pfizer, Celgene, Daiichi-Sankyo, TEVA, Astra Zeneca, Merck Sharp & Dohme, Myelo Therapeutics, MacroGenics, Eisai and Puma during the conduct of the study. JDB has acted on Advisory Boards for AstraZeneca and has received support from GSK to attend conferences. His institution receives funding from AstraZeneca and Aprea. UM has shares in Abcodia Ltd. Sandra Orsulic and Beth Y. Karlan have patents on predictive gene signatures in ovarian cancer (US010253368 and EU2908913). All remaining authors have declared no conflicts of interest.

clinical trials by identifying patients who are least likely to achieve 5-year survival. The identified novel genes associated with the outcome may also yield opportunities for the development of targeted therapeutic approaches.

Keywords

formalin-fixed paraffin-embedded; gene expression; high-grade serous ovarian cancer; overall survival; prognosis

INTRODUCTION

Epithelial ovarian cancer (EOC) causes ~125 000 deaths globally every year, and long-term survival rates have changed little in the past three decades.¹ Approximately 70% of women with EOC are diagnosed with advanced stage disease (stages III/IV), and fewer than 50% will survive more than 5 years.² There are five major EOC histotypes: high-grade serous, low-grade serous, endometrioid, clear cell and mucinous.³ High-grade serous ovarian cancer (HGSOC) comprises about two-thirds of cases, is responsible for most deaths and is characterized by profound genomic and clinical heterogeneity.

The most informative prognostic factors for HGSOC are International Federation of Gynecology and Obstetrics (FIGO) stage, residual disease following debulking surgery,⁴ *BRCA1* or *BRCA2* germline mutation^{5,6} and tumour-infiltrating lymphocyte scores.^{7,8} Patients with HGSOC who carry a loss-of-function germline mutation in *BRCA1* or *BRCA2* have an increased sensitivity to platinum-based chemotherapy and PARP inhibitor treatment^{9,10} and a medium-term survival advantage.⁵ However, the frequent development of drug-resistant disease⁶ limits the effectiveness of current therapies.

Gene expression data have been used to define four tumour molecular subtypes of HGSOC (C1/mesenchymal, C2/immune, C4/differentiated and C5/proliferative).^{11,12} Using transcriptome-wide data from fresh frozen tissues, The Cancer Genome Atlas (TCGA) project used 215 tumours to identify an overall survival (OS) expression signature of 193 genes that has been validated on three other HGSOC gene expression datasets.¹²

Despite these findings, gene expression biomarkers have not been implemented clinically owing to several important shortcomings. The majority of the individual markers comprising the 193 gene signature were not statistically significant across all studies, suggesting that the signature may not be robust. The sample sizes in other discovery efforts have been too small for robust statistical inference.¹² In addition, previous studies used fresh frozen samples, resulting in logistic and cost barriers to examining large clinically relevant datasets, and translation to the clinical setting.

The aim of this study was to identify a robust and clinic-ready prognostic HGSOC profile that can be applied to formalin-fixed paraffin-embedded (FFPE) tumour tissue.

PATIENTS AND METHODS

Twenty studies provided pretreatment FFPE tumour samples from 4071 women diagnosed with HGSOE (supplementary Table S1, available at *Annals of Oncology* online). All HGSOE cases with available tissue were included. During this period, patients with HGSOE were treated with chemotherapy (carboplatin and paclitaxel) after primary debulking surgery. Study protocols were approved by the respective Institutional Review Board/Ethics Approval Committee for each site (supplementary Table S1, available at *Annals of Oncology* online).

A schematic of the overall study design is shown in Figure 1. There were four main components: gene selection, gene expression assay, development of prognostic gene signature in a training set and validation of prognostic signature in an independent validation set.

Gene selection

Candidate prognostic genes were identified by carrying out an individual participant meta-analysis of six transcriptome-wide microarray studies,^{11–16} which included tumour samples from 1455 participants. Association of gene expression with OS was evaluated by Cox proportional hazards regression adjusted for molecular subtype (supplementary Table S2, available at *Annals of Oncology* online). In total, 200 genes from the meta-analysis, most achieving a permutation-based false discovery rate (FDR)¹⁷ of <0.05, and an additional 313 candidate genes based on the literature and unpublished data were selected (supplementary Tables S3 and S4, and supplementary Figure S1, available at *Annals of Oncology* online; for more details see supplementary Material, available at *Annals of Oncology* online). Five genes, *RPL19*, *ACTB*, *PGK1*, *SDHA* and *POLR1B*, were included as house-keeping genes for normalization.

Gene expression assay in tumour samples from study participants

FFPE tumour samples were processed with the NanoString nCounter technology at three different locations: Vancouver, Los Angeles and Melbourne. A control set of 48 FFPE tumour samples was run at each location and the average intraclass correlation coefficient was 0.987. Approximately 2% of the samples were run in duplicate and the average Spearman's correlation coefficient r_s was 0.995. Single-patient classification methods were used with reference samples to control for batch effects.¹⁸ The data in this publication have been deposited in NCBI's Gene Expression Omnibus¹⁹; GEO Series accession number GSE132342 (<https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE132342>). A total of 3329 samples passed quality control of which 3769 had survival data and assessable gene expression for 513 genes. Data can be found in NCBI GEO: Accession numbers GSE132342 and GPL26748.

Overall survival analysis of individual genes

Samples that contributed to the meta-analysis dataset ($n = 211$) were removed from subsequent selected analyses to enforce independence of study samples between the gene selection and final survival analysis. Time-to-event analyses were carried out for OS with

right censoring at 10 years and left truncation of prevalent cases. Associations between log-transformed normalized gene expression and survival time were tested using likelihood ratio tests with Cox proportional hazards models adjusted for age, race and stage, and stratified by study. Patients with missing race or stage information were assigned to ‘unknown’ categories. Age was modelled using a B-spline with a knot at the median age, which yielded a better fit than using knots at quartiles or categorical variables. Stage was dichotomized into early (FIGO stage I/II) and advanced (FIGO stage III/IV). Genes were scaled to have a standard deviation of one, so hazard ratios (HRs) correspond to a change of one standard deviation. A Benjamini–Hochberg FDR of <0.05 was used to identify notable associations. Because the expression of genes can be correlated, an analysis of correlated genes was performed using data from TCGA. Advanced stage ovarian cancer usually has disease spread throughout the abdomen, and therefore sensitivity analyses were performed to assess effects of the anatomical location of tumour samples included in the study by removing observations corresponding to samples known to be extraovarian ($n = 437$).

Prognostic signature development and validation

Studies were initially randomized to training set ($N = 14$) and validation set ($N = 6$). The TRI study was randomized to the validation set, but, because 107 samples were part of the meta-analysis data used for gene selection, the study was split, so those 107 samples were included in the model training dataset. Thus 2702 samples from 15 studies were used for model training and 1067 samples from 6 studies were used for validation (supplementary Table S1, available at *Annals of Oncology* online). In the training set, four modelling approaches (stepwise regression, elastic net regularized regression, boosting and random survival forests) were applied to construct competing gene expression–based biomarkers. Each was evaluated in the training data using 10-fold cross-validation for its prognostic value for OS at 2 and 5 years of follow-up using an area under the curve (AUC) measure derived from receiver operator characteristic analysis (see supplementary Material, available at *Annals of Oncology* online, for additional details). The best performing method, elastic net regularized regression, was applied to the full training set to determine the final gene signature and scoring method, which was then evaluated using the independent testing set. All models were constrained to include age and stage, where age was modelled as categorical based on quartiles of the training dataset with groups aged <53 , 53–59, 60–66, and 67. Stage was modelled as described earlier for the OS individual gene analysis.

RESULTS

Association of expression of individual genes with OS in HGSOV

In a gene-by-gene analysis of the full dataset adjusted for age, race and stage, and stratified by study, 276 of the 513 selected genes were associated with OS (FDR < 0.05). Of these, 138 were selected from the meta-analysis of six published microarray studies (supplementary Table S2, available at *Annals of Oncology* online)^{11–16} and 144 from candidate gene approaches (supplementary Tables S5 and S6, available at *Annals of Oncology* online). HRs for one standard deviation change in gene expression ranged from 0.84 to 1.19, with multiple genes exhibiting associations at very stringent significance levels (e.g. 19 genes with $P < 1 \times 10^{-8}$; supplementary Tables S5 and S6, available at *Annals of*

Oncology online). The five most significant genes were *TAP1*, *ZFHX4*, *CXCL9*, *FBN1* and *PTGER3* (Table 1). We did not find extensive evidence of high co-expression between these five genes and genes measured in TCGA project (supplementary Table S7, available at *Annals of Oncology* online). In sensitivity analyses we found that excluding samples from omentum and other extra-ovarian sites did not substantially affect the results (supplementary Tables S8 and S9, available at *Annals of Oncology* online).

Development of a novel prognostic gene signature

The four predictive modelling approaches that were evaluated in the training data using 10-fold cross-validation yielded median AUCs that ranged from 0.69 to 0.73 for 2-year OS and 0.69 to 0.74 for 5-year survival (supplementary Figure S2, available at *Annals of Oncology* online) with better prediction of 5-year OS than of 2-year OS. The elastic net approach yielded the highest median AUC for both 2- and 5-year OS and was selected for final development of the signature. Using the model on the full training dataset resulted in a prognostic signature of 101 genes in addition to age and stage (supplementary Table S10, available at *Annals of Oncology* online). Of these, 66 genes were associated with OS (FDR < 0.05) in the single gene models. There was no obvious subset of signature genes that performed as well or nearly as well as the full 101 gene signature (supplementary Figure S3, available at *Annals of Oncology* online).

Performance of the signature including age and stage was AUC 0.69 [95% confidence interval (CI) 0.65–0.73] and AUC 0.75 (95% CI 0.72–0.78) for 2- and 5-year OS, respectively (Figures 2 and 3; supplementary Figure S4, available at *Annals of Oncology* online). This was substantially better than age and stage alone with AUC 0.61 (95% CI 0.57–0.65) and AUC 0.62 (95% CI 0.59–0.67) for 2- and 5-year OS, respectively, particularly for the 5-year OS outcome with non-overlapping 95% CI. One standard deviation change in the gene expression score was associated with an HR of 2.35 [(95% CI 2.02–2.71); $P = 5.1 \times 10^{-31}$], and median survival [HR (CI)] varied substantially across quintiles of the gene expression score [9.5 (8.3 to –), 5.4 (4.6–7.0), 3.8 (3.3–4.6), 3.2 (2.9–3.7) and 2.3 (2.1–2.6) years, respectively, from smallest to largest quintile (Q1-Q5); Table 2].

For a subset of cases, there was clinical and experimental data for known prognostic factors. All samples had molecular subtype classification (Talhok et al.²⁰); residual disease was known for 1771 cases, primary chemotherapy treatment for 687, germline BRCA mutation status for 904 and nuclear CD8 tumour-infiltrating lymphocyte counts⁸ for 1111 (supplementary Table S11, available at *Annals of Oncology* online). When examined by quintile of gene expression score there were differences, as expected, for each of the known prognostic factors, including age and stage that were included in the model (Table 3). However, in sensitivity analyses, applying the signature to specific patient groups, a robustness of stratification was demonstrated, suggesting that the prognostic power of the signature is not explained by the individual factors, residual disease, treatment, BRCA status or CD8 score (Figure 3 and supplementary Figures S5–S7, available at *Annals of Oncology* online). The signature score showed modest differences by molecular subtype (supplementary Figure S8, available at *Annals of Oncology* online), and adjusting for

molecular subtype in the Cox analysis resulted in only minor changes to the HR estimates for signature quintiles (Table 2). The signature was shown to be prognostic within a homogenous group of 316 stage IIIC cases with no residual disease, within early stage cases (FIGO IA and IB) and within patients whose samples were collected from the omentum (supplementary Figures S9 and S10, available at *Annals of Oncology* online). Analysis of the signature score for paired ovarian and omental tissue from 42 of the cases showed a highly significant Pearson's correlation coefficient, $r = 0.79$ ($P = 5.4 \times 10^{-10}$; supplementary Figure S11, available at *Annals of Oncology* online).

A gene set enrichment analysis was performed for the 101 genes in the signature, as well as for genes correlated with signature genes achieving $r^2 > 0.75$ (supplementary Table S12, available at *Annals of Oncology* online). For the correlated gene analysis, the three most significant pathways involved the immune system, including the adaptive immune system and cytokine signalling. A further 10 immune pathways were significantly enriched and included interferon signalling, innate immune system and TCR signalling and antigen presentation pathways. Restricting to the signature genes only, there was also enrichment in the immune system, but the top two pathways were PI-3K (phosphoinositide 3-kinase) cascade and GPCR (G protein-coupled receptor) ligand binding. Four other pathways were related to the cell cycle and mitosis, with the remaining enriched for fibroblast growth factor receptor (FGFR) and epidermal growth factor receptor (ERBB) signalling, and one pathway related to homologous combination repair.

DISCUSSION

In a large-scale study of patients with HGSOC, we identified a 101-gene expression signature able to predict clinically relevant differences in OS. Using methods that are both economical and applicable to standard clinical sampling techniques, we showed that the signature performs substantially better than age and stage alone for prognosis of both 2- and 5-year OS. The number of patients and samples included in this study is an order of magnitude greater than previous comparable studies of gene expression and OS in patients with HGSOC.^{12,21,22} Thus, we have been able to more precisely quantify the prognostic value of gene expression.

We report definitive associations between OS and expression of 276 genes. Of the five most significant genes (*TAP1*, *ZFHX4*, *CXCL9*, *FBN1* and *PTGER3*), four have been previously reported to be associated with survival in HGSOC. The top prognostic gene, *TAP1*, is involved in the antigen-presenting pathway. Expression was reduced in metastatic HGSOC, positively associated with OS,²³ as observed here, and linked to tumour regression in response to treatment.²⁴ Further, hypomethylation of *TAP1* was associated with improved time to disease recurrence.²⁵ *CXCL9* is a chemokine that mediates the recruitment of T cells to solid tumours.²⁶ High expression of intratumoural *CXCL9* was associated with higher OS²⁷ and higher lymphocytic infiltration, which is also a robust prognostic factor in HGSOC^{8,11,28} and a feature of the immunoreactive HGSOC molecular subtype.¹¹ *CXCL9* has also been proposed as a therapeutic target due to evidence that it inhibits angiogenesis and promotes antitumour adaptive immunity.^{29–31} Strikingly, the signature was able to further refine prognostic groups within patients with high tumour-infiltrating lymphocyte

counts suggesting that *CXCL9* and *TAPI* expression may be strong indicators of immune competency in HGSOC.

FBNI is an extracellular matrix protein previously found to be a biomarker associated with early recurrence in patients with ovarian cancer who are initially sensitive to chemotherapy³² and strongly correlated with desmoplasia in HGSOC. The prostaglandin E₂ receptor *PTGER3* is expressed in ovarian tumour cells and is associated with relapse-free survival.³³ By contrast, *ZFHX4* does not have previous associations with HGSOC.

Associations between the expression of specific genes in tumour tissues and OS in patients with HGSOC may suggest new drug targets and lead to insights into biological variation in treatment response. For example, cases in the Q5 quintile with the poorest outcome had increased expression of *IGF2*, *FGFR1* and *MYC*, a possible argument for the use of *IGFR1*, *FGFR*, bromodomain (*MYC*) or a combination of PARP and CDK4/6 inhibitors (*MYC*).³⁴ More immediately, the signature may help clinicians identify patients most in need of intervention, such as patients that could potentially benefit from neoadjuvant chemotherapy (NACT). Alternatively, in clinical trials it could be used to stratify randomization by patients' risk, thereby reducing heterogeneity within subgroups and increasing heterogeneity between subgroups. The signature will be incorporated into future prospective clinical trials to determine if it can predict response to specific treatments.

Measurement of the signature required standard FFPE tissue used in routine histopathology. In addition, data preprocessing and normalization were conducted on an individual level, thus translatable to a general patient population. That is, 5-year OS prognosis of future patients can be evaluated against the patient population reported here by (i) following the same steps described here for generating the normalized gene expression data, (ii) computing an individual signature score and (iii) assigning an HR based on the score or comparing it with the reported quintiles (supplementary Material, available at *Annals of Oncology* online). NanoString gene expression is highly reproducible as seen by our quality control metrics (supplementary Material, available at *Annals of Oncology* online) and the FDA approval of the Prosigna test for breast cancer.

The question of heterogeneity by ancestry or ethnicity was beyond the scope of this study but should be pursued in future research. Another important question is whether molecular subtype can improve biomarker performance. A substantial proportion of signature genes were identified by the subtype-adjusted meta-analysis, suggesting that the strong performance of the signature is not solely attributable to differences among molecular subtypes. In addition, all of the individual genes used in the molecular subtype classification were included in development of the signature.

Although the cases received chemotherapy, the FFPE samples used in this study were chemo-naïve, as few patients had NACT during the calendar period in which these samples were collected. Because the signature appears to be prognostic in omentum samples, future studies may assess the value in NACT patients, using pretreatment omental biopsies or post-treatment tumour samples. Future work will also address if the signature can predict platinum-refractory patients.

We have developed a robust prognostic signature for HGSOc that can be used to stratify patients and identify those in need of alternative treatments. Gene set enrichment analysis applied to the signature indicates an important role for the immune system in OS and supports further investigation of immune therapy in ovarian cancer. More generally, the identification here of high-confidence prognostic genes may lead to new hypotheses for targeted treatments.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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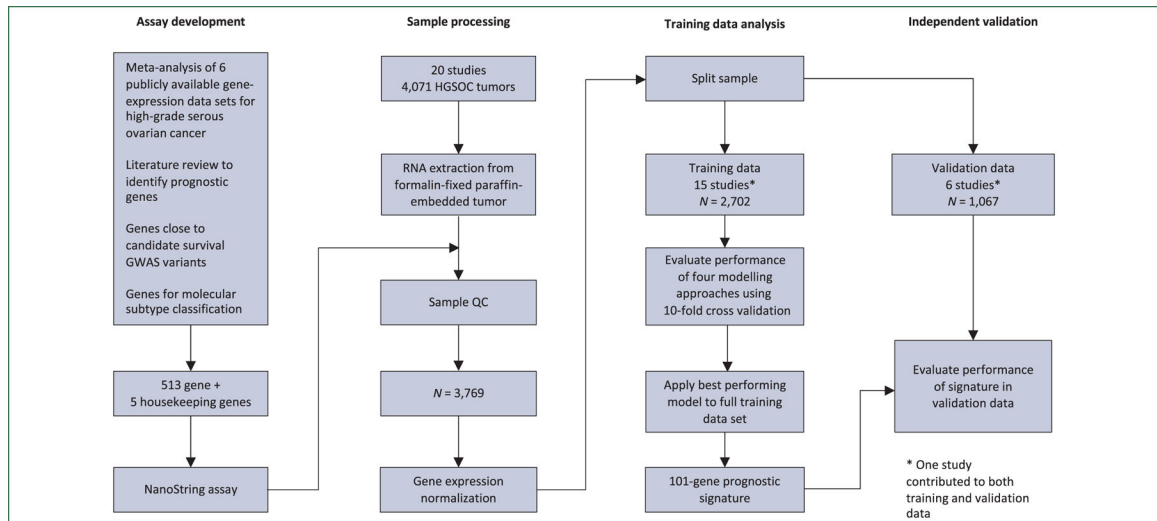


Figure 1. Schematic of study design.

*The TRI study was split across the training and validation sets due to 107 samples overlapping with the meta-analysis. GWAS, genome-wide association studies; HGSOC, high-grade serous ovarian cancer.

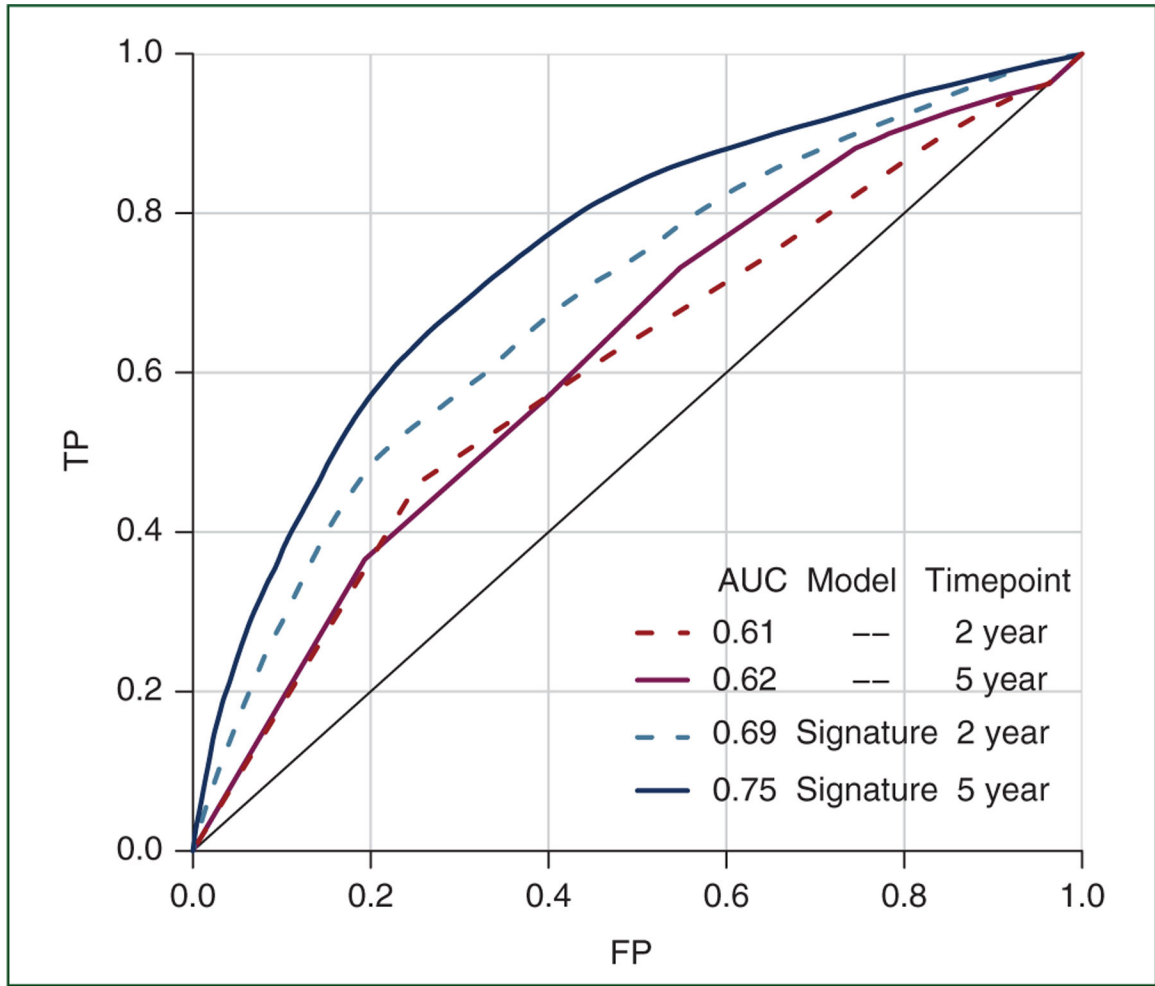


Figure 2. Receiver operator characteristic (ROC) curves for prognostic performance of the gene expression signature in independent high-grade serous ovarian cancer patients (testing data). There was no overlap between studies or patient data used to develop models (training data) and construct ROC curves and calculate area under the curve (AUC) values shown here (testing data). All models included age and stage as described in Methods section. TP denotes the true positive rate (sensitivity) and FP denotes the false positive rate (1 – specificity).

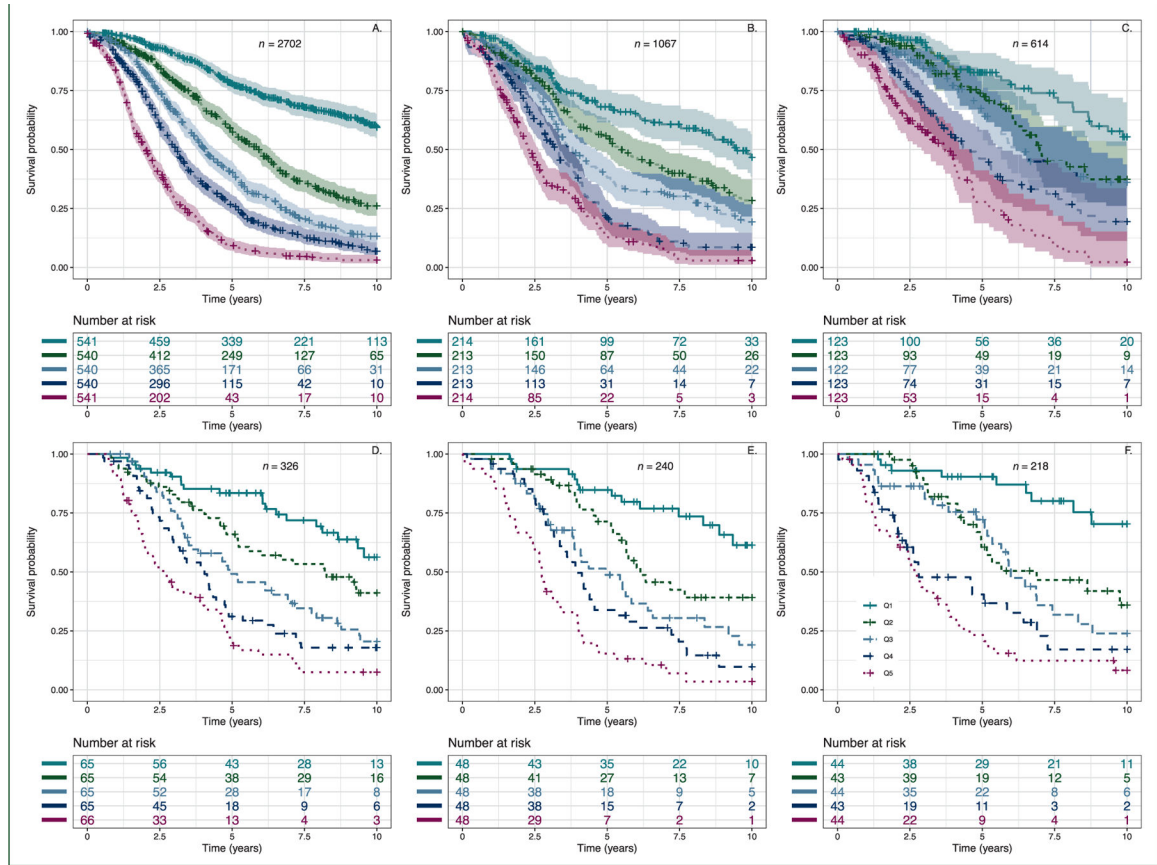


Figure 3. KaplanMeier curves of overall survival for patients (A) in the training and (B) testing sets.

Patients were assigned to quintiles (Q1–Q5) of the signature score including age and stage. Shaded areas indicate 95% confidence regions, only included for plots representing larger sample sizes. Because of limited sample size, the following plots represent all such patients in the entire dataset, training or testing: (C) no macroscopic residual disease after debulking surgery, (D) primary chemotherapy treatment 4 cycles of intravenous (IV) carboplatin area under the curve (AUC) 5 or 6 and paclitaxel 135 or 175 mg/m² every 3 weeks (actual dose known or presumed), (E) *BRCA1* or *BRCA2* germline mutation and (F) CD8 > 19.

Hazard ratios and 95% CIs for top five prognostic genes in covariate-adjusted single-gene analyses

Table 1.

Gene	HR (95% CI)	P	Selection	Correlated gene ^a	r _s
<i>TAP1</i>	0.84 (0.80–0.87)	8.3 × 10 ⁻¹⁸	Meta	PSMB9	0.89
<i>ZFX4</i>	1.19 (1.14–1.25)	1.4 × 10 ⁻¹⁵	Meta	LOC100192378	0.74
<i>CXCL9</i>	0.85 (0.82–0.88)	1.8 × 10 ⁻¹⁵	Meta and candidate	CXCR6	0.89
<i>FBN1</i>	1.18 (1.13–1.24)	4.2 × 10 ⁻¹⁴	Candidate	SPARC ^b	0.91
<i>PTGER3</i>	1.18 (1.13–1.24)	1.2 × 10 ⁻¹³	Meta	COL8A1	0.67

CI, confidence interval; HR, hazard ratio.

^aMost correlated gene according to Spearman's rank correlation coefficient, r_s, computed in The Cancer Genome Atlas (TCGA) Ovarian Serous Cystadenocarcinoma RNA-seq dataset.

^b*SPARC* was included in this project and was less significant.

Hazard ratios and 95% CIs for quintiles of the gene expression signature score in validation data

Table 2.

Quintile	N	Deaths	Median survival ^a	HR (95% CI)	Adjusted for age and stage HR (95% CI)	Adjusted for molecular subtype age and stage HR (95% CI)
Q1	214	81	9.47 (8.32 to -)	0.44 (0.33-0.58)	0.34 (0.22-0.55)	0.37 (0.23-0.59)
Q2	213	117	5.38 (4.63-6.97)	0.73 (0.57-0.93)	0.71 (0.55-0.91)	0.74 (0.58-0.96)
Q3	213	145	3.80 (3.34-4.60)			
Q4	213	158	3.23 (2.85-3.68)	1.56 (1.25-1.96)	1.56 (1.24-1.97)	1.56 (1.24-1.96)
Q5	214	179	2.27 (2.09-2.62)	2.23 (1.78-2.78)	2.11 (1.67-2.67)	2.07 (1.63-2.61)

CI, confidence interval; HR, hazard ratio.

^aMedian survival (95% CI) in years for patients in the validation set.

Table 3. Clinical data for the 3769 patients that passed quality control and the percentage of patients in each quintile of the gene expression score

	Total	Q1	Q2	Q3	Q4	Q5	P value
N	3769	754	754	753	754	754	
Median survival (years)	4.1	9.5	5.4	3.8	3.2	2.3	
% 5-year survival	41	75	57	39	25	10	
Age median	63	58	57	61	64	70	
Age range	25–89	39–78	25–86	36–82	27–89	39–86	
Age quartile Q1	894	30.8	31.3	20.0	13.4	4.5	<1 × 10 ⁻⁵⁰
Age quartile Q2	838	21.5	20.0	22.9	21.2	14.3	
Age quartile Q3	961	16.0	20.2	21.4	23.6	18.7	
Age quartile Q4	1076	13.5	10.4	16.4	21.3	38.5	
FIGO stage I/II	607	97.4	2.6	0.0	0.0	0.0	<1 × 10 ⁻⁵⁰
FIGO stage III/IV	3067	3.8	23.0	24.1	24.4	24.6	
Primary chemo ^a 1	136	16.2	22.1	23.5	19.1	19.1	0.163
Primary chemo ^a 2	190	16.3	20.0	21.6	22.1	20.0	
Primary chemo ^a 3	361	11.1	16.9	22.4	20.5	29.1	
Residual disease: No	614	32.4	22.1	17.8	15.5	12.2	<1 × 10 ⁻⁵⁰
Residual disease: Yes	1157	6.0	19.2	24.1	24.5	26.2	
Germline <i>BRCA1</i> mutation	130	23.8	31.5	26.2	11.5	6.9	2.22 × 10 ⁻⁷
Germline <i>BRCA2</i> mutation	71	28.2	26.8	18.3	18.3	8.5	
Germline no <i>BRCA1/2</i> mutation	663	19.6	16.7	18.7	20.7	24.3	
CD8 TIL score 0	192	19.8	14.6	12.5	21.4	31.8	2.46 × 10 ⁻¹⁴
CD8 TIL score 1–2	186	18.3	14.0	18.8	21.5	27.4	
CD8 TIL score 3–19	515	19.8	24.1	20.8	17.9	17.5	
CD8 TIL score >19	218	34.4	31.2	16.5	11.5	6.4	
Molecular subtype C1.MES	1105	5.4	10.4	20.7	27.4	36.0	<1 × 10 ⁻⁵⁰
Molecular subtype C2.IMM	907	23.2	28.8	21.2	16.2	10.7	
Molecular subtype C4.DIF	1144	32.6	25.5	17.9	12.8	11.2	
Molecular subtype C5.PRO	613	18.1	14.0	20.7	25.8	21.4	

	Total	Q1	Q2	Q3	Q4	Q5	P value
FIGO stage IA and IB	111	96.4	3.6	0.0	0.0	0.0	$<1 \times 10^{-50}$
FIGO stage IIIC	1979	3.1	23.7	24.6	24.1	24.6	$<1 \times 10^{-50}$
FIGO stage IIIC residual disease: No	316	6.3	31.0	24.4	20.9	17.4	6.24×10^{-45}
FIGO stage IIIC residual disease: Yes	846	2.6	21.5	25.3	24.6	26.0	

Q1 is the quintile with the best survival and Q5 the worst survival. Samples with missing data are reported in supplementary Table S11, available at *Annals of Oncology* online. P values for *BRCA1/2* mutation status were calculated for *BRCA1* or *BRCA2* mutation versus no mutation.

FIGO, International Federation of Gynecology and Obstetrics; TIL, tumour-infiltrating lymphocyte.

⁴Treatment: 1 = known to have received first-line chemotherapy treatment of 4 cycles of IV carboplatin AUC 5 or 6 and paclitaxel 135 or 175 mg/m² every 3 weeks. 2 = known to have received first-line chemotherapy treatment of 4 cycles of IV carboplatin and paclitaxel three times weekly but at doses presumed to be carboplatin AUC 5 or 6 and paclitaxel 135 or 175 mg/m². 3 = all remaining cases with chemo regimens that do not fit criteria 1 or 2 and include unknown or no chemotherapy.