



# Evaluating prognostic value and relevant gene signatures of tumor microenvironment characterization in esophageal carcinoma

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**Background:** Tumor microenvironment (TME) cells are an important part of tumor tissues. There is increasing evidence that the TME plays a vital role in tumor prognosis, and is associated with patient survival in various kinds of malignances. To date, very little research has been conducted on how to effectively use TME to better evaluate the prognosis of patients with esophageal carcinoma (EC). The concept of a “TME score” was introduced to better distinguish the prognosis of patients.

**Methods:** We employed bioinformatic methods to investigate the TME infiltration patterns of 160 patients with EC from the Cancer Genome Atlas (TCGA) cohort. TME clusters were identified using k-means clustering methods with 1,000 resampling times. The significance of the survival difference among patients belonging to different TME clusters was assessed by the log-rank test and Kaplan-Meier survival curves. Correlations between immune cell types and survival were calculated by a Cox regression, and the Pearson correlation coefficient (PCC) was used to measure the relationship among different immune cell types. We classified patient into 2 subtypes based on the optimal breakpoint of TME score determined by R package maxstat.

**Results:** Two TME phenotypes were defined based on the immune cell type fractions, and patients with a high TME score phenotype had a better prognosis than those with a low TME score phenotype. Kaplan-Meier analysis for differentially expressed micro ribonucleic acids (RNAs) and messenger RNAs also showed that different TME score subtypes were significantly associated with the prognosis of EC. Just as tumor mutational burden can predict the efficacy of immunotherapy, the TME score can predict the efficacy of immune checkpoint inhibitors (ICIs). The genomic alterations of 2 TME score subtypes of EC further revealed that genomic instability is prevalent in TMEs, and patients with a low TME score subtype have a more unstable chromosome status than those with a high subtype.

**Conclusions:** Thus, TME score is an emerging prognostic biomarker for predicting the efficacy of ICIs.

**Keywords:** Esophageal carcinoma (EC); tumor microenvironment (TME); TME score; prognosis

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## Introduction

Esophageal carcinoma (EC) is one of the most aggressive cancers. It was ranked 9th and 6th for incidence and mortality, respectively (1). The standard therapies for EC are surgery, chemotherapy, and radiotherapy. At present, 3 therapeutic agents that target human epidermal growth factor receptor 2 (HER2), vascular endothelial growth factor 2 (VEGF-2), and programmed cell death protein 1 (PD-1) have been approved by the Food and Drug Administration for use in the treatment of EC (2). Despite the great improvement in therapeutics, the 5-year survival rate remains poor at about 20% (3). Previous genomic studies have explored the molecular mechanism underlying EC, but have mainly focused on the genetic landscape (4,5). The tumor microenvironment (TME), which contains diverse immune cell populations, is thought to play a crucial role in the development of different cancers, including breast cancer (6), gastric cancer, (7) and esophageal squamous cell carcinoma (ESCC) (8), which suggest that immune cell infiltration may be associated with the clinical outcome of cancers. During the development of tumors, oncogenic mutations allow cells to proliferate independently, and cancer cells interact with the TME to favor its expansion and evade immune control (9).

A greater understanding of the tumor-immune microenvironment may provide more advanced prognostic biomarkers for the immunotherapy and chemotherapy of EC patients and reveal additional novel targets. Immunotherapy is a well-known treatment that works by modulating the TME. In addition, research suggests that the success of chemotherapy and radiotherapy relies on the induction of a durable anti-cancer immune response, and that targeted therapies also have immuno-stimulatory effects (10-12). It may be that the TME could change as a result of therapies, such that there is a shift from a pre-existing immune response to a therapy-induced immune response (10). Thus, information about the prediction of a treatment response and prognosis could be gained from the TME. A number of studies have highlighted the role of the TME in prognosis and treatment. For example, a high density of T lymphocytes in TME is associated with a favorable prognosis, while a high density of M2 macrophages is associated with a poor prognosis (10,13,14). The TME context determined at the time of diagnosis could also reflect the immune response and chemotherapy benefits (15,16). PD-L1 blockade is used in the microsatellite instability (MSI)/deficient mismatch repair (dMMR) treatment of colorectal cancers (17). Thus, it is necessary to integrate existing medical and

targetable biomarkers, including the MSI biomarker, for the comprehensive evaluation of EC.

Recently, several bioinformatic approaches have been developed to analyze the TME and have been successfully applied to some tumor types (7,18-22). However, little research has been conducted on the use of TME in evaluating the therapeutic efficacy of treatments for patients with EC. In relation to estimations of the TME, immunohistochemistry (IHC) is the only technique to yield quantitative information on the TME. In this study, we analyzed the immune infiltrate of EC by using a proposed computational algorithm, and developed an approach to quantify the TME infiltration pattern (i.e., the TME score). Our results showed that the TME score has potential clinical application and value in evaluating the prognosis of patients with EC. We further showed significant correlations between some micro ribonucleic acids (miRNAs) and differentially expressed genes (DEGs) in TME patterns with survival.

## Purpose

This study designed a TME signature to evaluate the comprehensive TME of EC, and a TME score classifier to effectively predict the prognosis of EC patients. The results of this study may have important implications for identifying subgroups of EC patients with a low or high risk of poor survival. We present the following article in accordance with the REMARK reporting checklist (available at <https://dx.doi.org/10.21037/jgo-21-371>).

## Methods

### Data collection

Molecular data from The Cancer Genome Atlas (TCGA) for EC (ESCA) type, including RNA-sequencing data, single nucleotide variation (SNV) data, copy number variations (CNV) data, miRNA microarray data, methylation microarray data, and corresponding clinical data with survival information, were downloaded (<https://portal.gdc.cancer.gov/projects/TCGA-ESCA>). Samples without survival information were excluded from further analysis. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

### Data pre-processing

Genes with a low expression level were removed by

supplying RNA-sequence raw read count data to the filterByExpr function of edgeR. The read counts of the remaining genes were then normalized by Voom in the Limma package to facilitate the deconvolution of immune cell types.

### *Estimation of immune cell type fractions in the TME*

CIBERSORT is an algorithm for enumerating the immune cell composition of complex tissues from bulk RNA-sequencing data. To infer the fraction of 22 immune cell types in tumor samples, a leukocyte gene signature matrix containing 547 genes (LM22) was used. We uploaded the pre-treated RNA-sequencing data (n=160) to the CIBERSORT web portal (<http://cibersort.stanford.edu/>), and ran the algorithm using the LM22 gene signature and 1,000 iterations.

### *Consensus clustering for TME-immune cells*

TME clusters were identified using k-means clustering methods in the ConsensusClusterPlus R package with 1,000 resampling times to ensure the stability of classification. The optimal number of clusters was determined using the elbow method. The significance of the survival difference among patients belonging to different TME clusters was assessed by the log-rank test and Kaplan-Meier survival curves. Correlations between immune cell types and survival were calculated by a Cox regression, and the Pearson correlation coefficient (PCC) was used to measure the relationship among different immune cell types.

### *DEG analysis*

DEGs between 2 TME clusters were identified by the Limma package in R with the significance criteria of a P value <0.001 and a  $|\log_2FC| > 1$ . To explore the biological implications of these DEGs, functional enrichment was carried out using the R package cluster Profiler.

### *Generation of TME gene signatures*

The signature genes most closely related to the TME clusters were selected by reducing redundant DEGs using a random-forest algorithm. These selected signature genes were then separated into 2 categories according to the correlation between the Cox coefficients and survival. After calculating the coefficient, we defined the TME score,

which is similar to the previously reported Gene expression Grade Index (GGI), to summarize the total expression level of survival-related genes. The TME score is calculated according to the following formula:

$$\text{TME score} = \sum \text{voom}(X) - \sum \text{voom}(Y) \quad [1]$$

where X is the expression level of genes whose Cox efficient is positive, and Y is the expression level of genes whose Cox efficient is negative.

### *Exploration of the predictive value of the TME score for prognosis*

To explore the relationship between the TME score and the survival of patients, we classified patient into 2 groups based on the optimal breakpoint of TME score determined by R package maxstat. The survival difference between the high and low TME score groups of patients was assessed using the log-rank test and Kaplan-Meier survival curve.

### *Identification of somatic mutations*

SNV data, comprising 160 samples, were used to obtain the mutation spectrum. The maftools R package was used.

### *Identification of mutational signatures*

SNV data, including 160 samples, were used to obtain the mutational spectra and identify mutational signatures. The maftools and SomaticSignatures R packages were used.

### *Inference of the copy number variation (CNV)*

The CNV regions among the 160 samples were detected by the Genomic Identification of Significant Targets in Cancer (GISTIC) module of GenePattern. The confidence interval was set as 0.95, and the other parameters were default values.

### *Identification of prognosis-related differentially expressed mRNA and miRNA*

As stated above, the patients were divided into high and low TME score groups. The differentially expressed messenger RNA (mRNA) and miRNA of these 2 groups were identified using the Limma R package. Next, a gene annotation enrichment analysis was performed. For the analysis of mRNA, the significance criteria were a  $P < 5e-2$  and a  $|\log_2FC| > 0.5$ . For the analysis of miRNA,

the significance criteria were a  $P < 0.05$  and a  $|\log_2FC| > 0.25$ . For the analysis of methylated sites, the significance criteria were a  $P < 0.05$  and a  $|\log_2FC| > 0.5$ . Finally, the differentially expressed mRNA and miRNA were further analyzed to explore their correlation with survival.

### Statistical analysis

The data analyses were performed with R software (version 3.5.1) and R Bioconductor packages. The normality of the variables was examined using the Shapiro-Wilk normality test. Comparisons between the 2 groups were performed using unpaired Student *t*-tests if the variables were normally distributed, or otherwise, Wilcoxon rank-sum tests. The survival curves for the subgroups were generated using the Kaplan-Meier method, and the statistical significance of the differences was determined using the log-rank test. A  $P < 0.05$  was considered statistically significant.

### Data availability

All the data used in this study are available for download from The Cancer Genome Atlas (TCGA) database (<https://portal.gdc.cancer.gov/projects/TCGA-ESCA>).

## Results

### TME score subtypes as biomarkers for predicting the prognosis of EC

All 160 EC samples with overall survival (OS) information were used for the prognostic model construction. A univariate Cox regression analysis was performed to estimate the prognostic relationship between the TME score and OS in EC patients. Through the LASSO Cox regression model with minimized lambda, the TME score subtypes of 22 immune cell signatures were selected to build the TME score model in the training cohort. The associations between the TME score subtypes of 22 immune cell signatures and OS are presented in *Figure 1A*. Additionally, the TME cell network depicted tumor-immune cell interactions and their effects on the OS of patients with EC using Cytoscape (see *Figure 1B* and *Table S1*). The TME cell network revealed that eosinophils (23), M0 macrophages, M2 macrophages, natural killer cells, and memory cluster of differentiation (CD)4+ T cells (24) were significantly negatively correlated with prognosis in EC, while activated dendritic cells (DCs), CD8+ T cells, and naive CD4+ T cells were significant

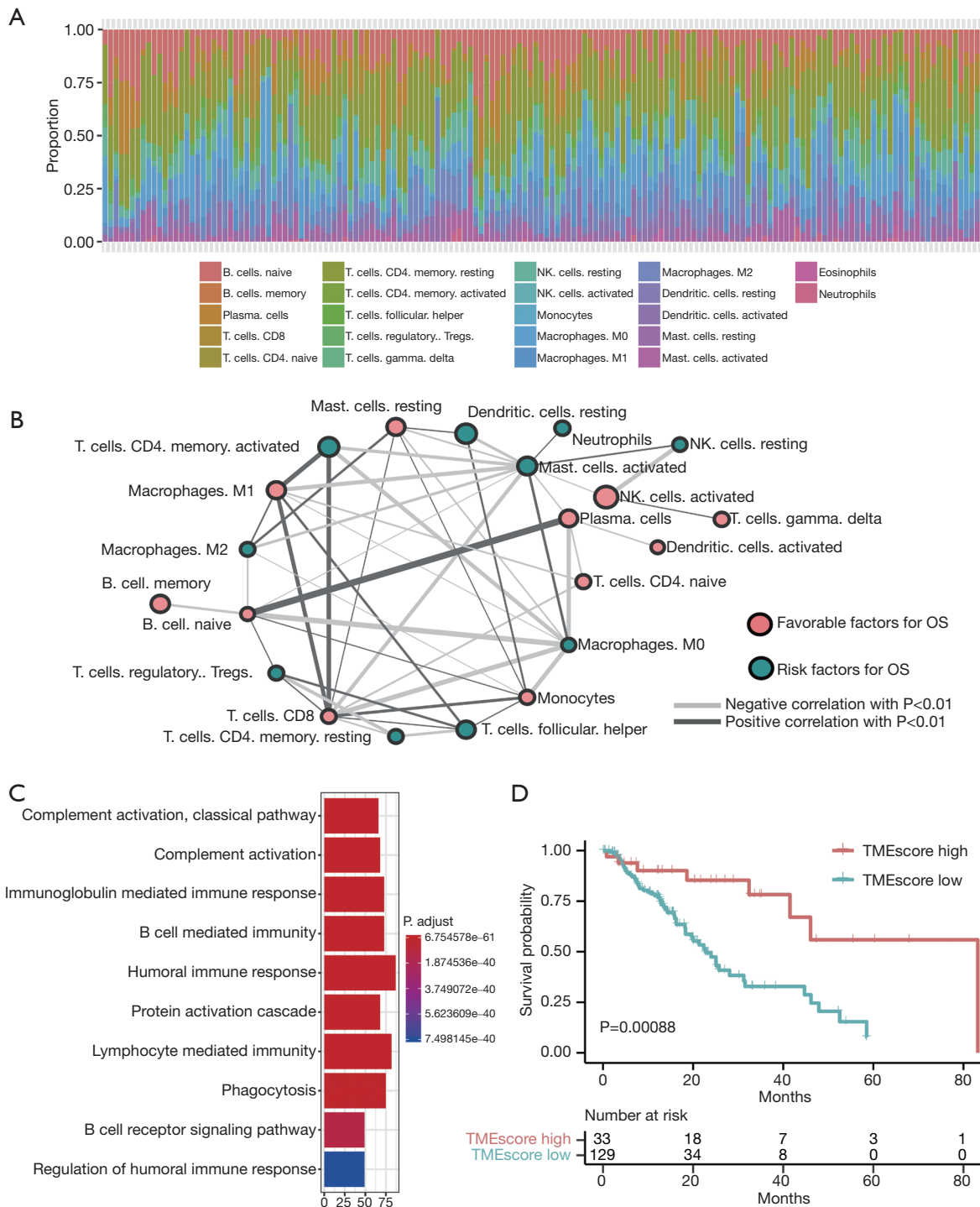
positively correlated with prognosis in EC. Next, to achieve optimal clustering stability, the unsupervised clustering method (K-means) and elbow method were applied to identify the optimal K value. As a result, 2 clusters were determined ( $K=2$ ; see *Figure S1A, S1B*). 2 distinct patterns of TME cell infiltration were determined ( $K=2$ ) by unsupervised clustering (see *Figure S1C*), and the proportions of immune cells were slightly different between TME cluster 1 and TME cluster 2 (see *Figure S1D*).

To further understand the biological and clinical differences of the TME phenotypes, the DEGs between TME cluster 1 and TME cluster 2 were determined using the Limma R package. As a result, a total of 687 DEGs were included in the TME classification. An unsupervised clustering analysis was also performed to group patients into genomic subtypes (see *Figure S2A, S2B*), and the clustering results were significantly consistent with those of the TME phenotype groups (Pearson's Chi-squared test,  $P < 0.001$ ). We then applied a random-forest algorithm to reduce redundant DEGs, and obtained 119 signature genes (see *Tables S2, S3*). The biological processes of gene ontology (GO) terms were mainly enriched in immune-related processes, such as the humoral immune response mediated by circulating immunoglobulin, the immunoglobulin-mediated immune response, and B cell-mediated immunity (see *Figure 1C*). Based on the Cox regression model, the Cox coefficient for each cluster signature gene was obtained and used to calculate TME scores. As a result, the samples were classified into high or low TME scores. A further survival analysis stratified by TME score demonstrated that patients with a high TME score phenotype had a better prognosis than those with a low TME score phenotype (see *Figure 1D*).

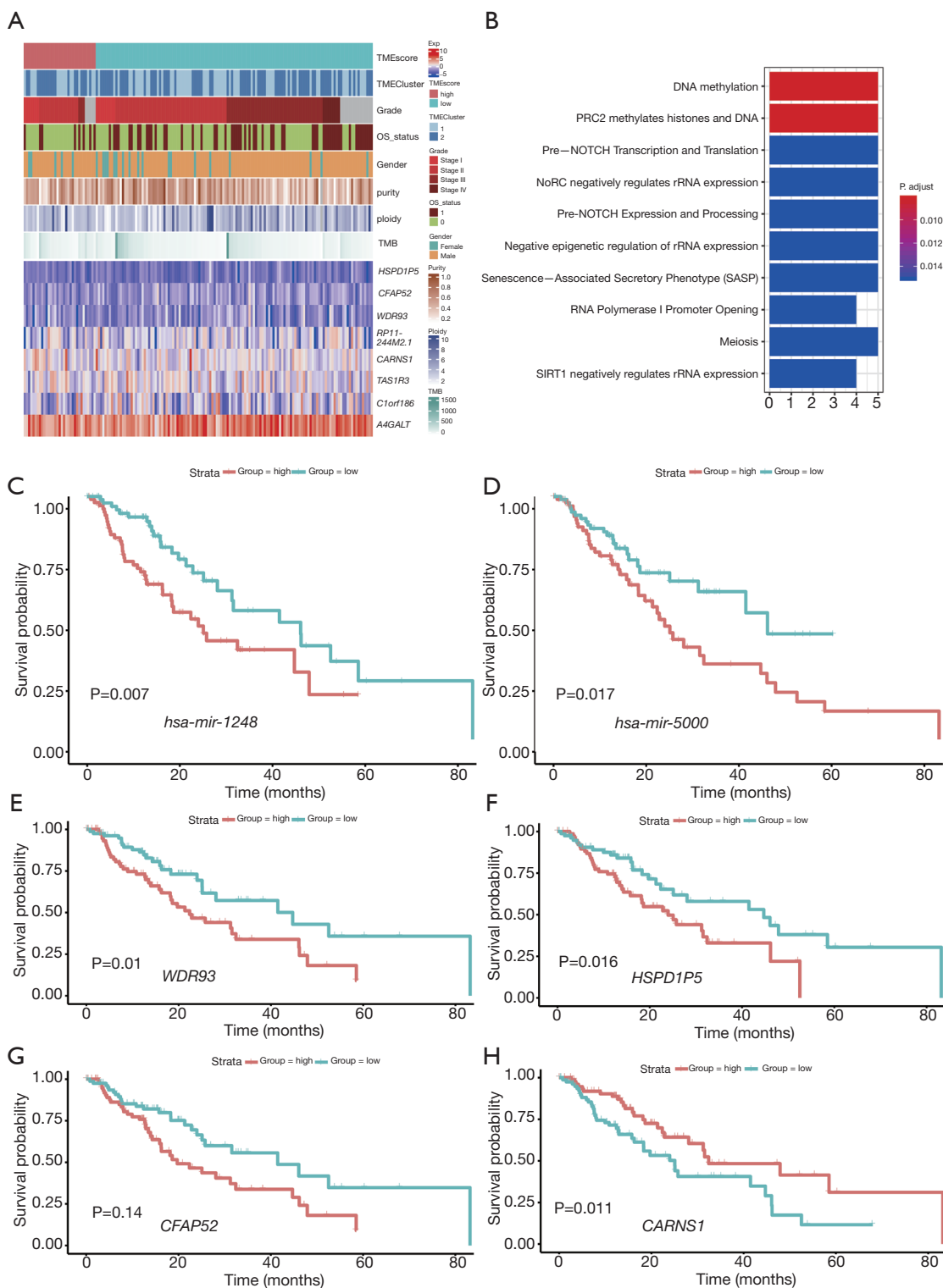
In terms of clinical characteristics, we performed unsupervised hierarchical clustering on 202 survival-related genes, and the significant prognostic effect of the immune subtypes in 160 patients with EC were classified as either high or low TME score subtypes (see *Figure 2A* and *Table S4*). The survival difference was independent of grade, gender, purity, ploidy, tumor mutational burden (TMB), and OS status. The DEG analysis revealed 15 differentially expressed miRNAs and 202 differentially expressed mRNAs between the high and low TME score subtypes with a threshold adjusted  $P < 0.05$  and a  $|\log_2FC| > 0.25$  for miRNA, and an adjusted  $P < 0.05$  and a  $|\log_2FC| > 0.5$  for mRNA, respectively.

The biological processes of GO terms revealed that differentially expressed mRNAs were mainly enriched





**Figure 1** Landscape of the TME in EC and the characteristics of TME subtypes. (A) A hierarchical heatmap of different immune cells. (B) Tumor-immune cell interaction network. The lines connecting the cells represent cellular interactions. The thickness of the line represents the strength of the correlation. The positive correlation is indicated in dark gray, and the negative in light gray. The size of each cell represents their survival effect. Favorable factors for overall survival are indicated in red, and risk factors are indicated in green. (C) The GO enrichment analysis of the 119 signature genes. (D) Kaplan-Meier curves for high (n=33) and low (n=129) TME score patient groups. Log-rank test, P<0.001. TME, tumor microenvironment; EC, esophageal carcinoma.



**Figure 2** Identification of novel genomic prognostic biomarkers. (A) Comprehensive genome landscape of ESCA. (B) The expression values of differentially expressed mRNA in TME clusters. Higher expression levels of *bas-mir-1248* (C), *bas-mir-5000* (D), *WDR93* (E), *CFAP52* (F), and *HSPD1P5* (G) were associated with shorter OS, while a higher expression level of *CARNS1* (H) was associated with longer OS. TME, tumor microenvironment; OS, overall survival.

in deoxyribonucleic acid (DNA) methylation, polycomb repressive complex 2 (PRC2) methylates histones, and DNA (see *Figure 2B*). However, we did not find any differentially expressed methylated site. Based on the Cox regression model, 2 differentially expressed miRNAs and 8 differentially expressed mRNAs were identified that had a statistically significant effect on the prognosis of EC. In relation to these, the 2 miRNAs were hsa-mir-1248 and hsa-mir-5000, and the top 4 genes were *WDR93*, *CARNS1*, *CFAP52*, and *HSPD1P5*. The OS of patients with a high expression level of has-mir-1248 or has-mir-5000 was significantly shorter than that of patients with a lower expression level (see *Figure 2C,2D*). Similarly, a high expression level of *WDR93*, *CFAP52*, and *HSPD1P5* was a negative prognostic factor of EC (see *Figure 2E-2G*), while a high expression level of *CARNS1* was a good prognostic factor in EC (see *Figure 2H*).

#### **Genomic alterations of high and low TME score subtypes of EC**

The landscape of the somatic mutations of EC are presented in *Figure 3A*. Missense mutation mainly caused by SNV was the major type of mutation, and C>T was the major type of base substitution (5). The most frequently mutated genes were TP53 in the high and low TME score subtypes. There were 28 genes with significant differences in mutation frequencies between the high and low TME score subtypes, among which the top 10 differential genes were *TP53*, *TTN*, *CSMD3*, *SYNE1*, *FLG*, *RIMS2*, *KMT2D*, *MUC16*, *MUC4*, *RYR2*, *LRP1B*, *DNAH5*, and *PCLO* (see *Figure 3A, B, C*). *TTN-AS1*, which is a potential diagnostic and prognostic biomarker (25), is highly expressed in ESCC tissues, and promotes ESCC cell proliferation and metastasis (26). *PCLO*, which is frequently mutated and amplified in ESCC, represents a novel prognostic biomarker and therapeutic target for patients with ESCC (27). These results further confirmed that these high-frequency mutated genes play a key role in the prognosis of EC tumors.

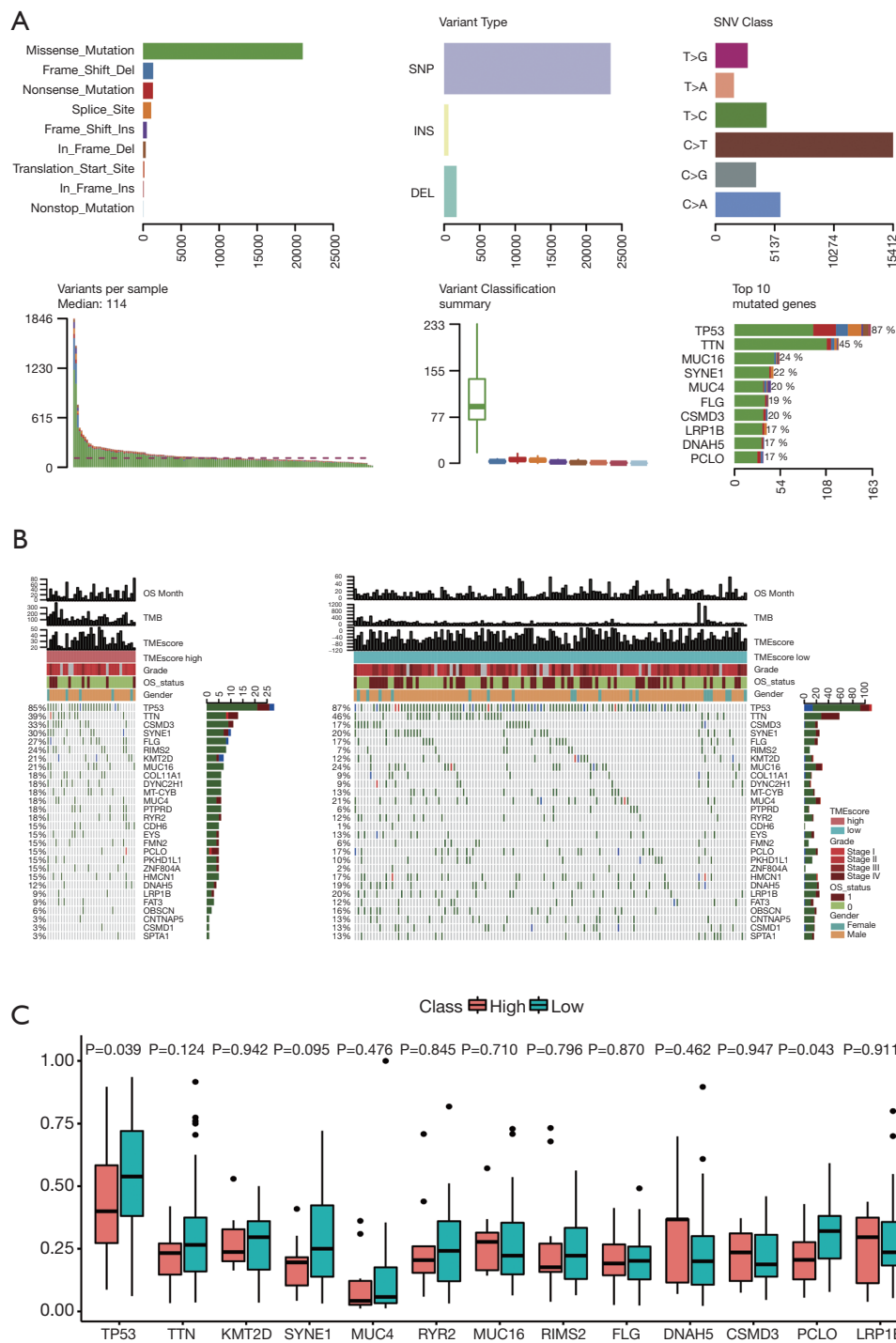
The mutational signature analysis showed that the high TME score subtype was associated with Signature 1, Signature 13, Signature 17, and Signature 29 (see *Figure 4A*). Conversely, the a TME score subtype was associated with Signature 1, Signature 3, and Signature 17 (see *Figure 4B*). Signature 1 was associated with the spontaneous deamination of 5-methylcytosine, Signature 3 was associated with DNA-double-strand breaks repair, and Signature 29 was associated with exposure to tobacco mutagens.

We also investigated genomic instability. The CNVs analysis by GISTIC showed that the significantly amplified regions included 7p, 8q, and 20q. Further, deletions, including at 3p, 5q, and 18q, frequently occurred in the low TME score subtype (see *Figure S3A*), whereas deletions, including at 9p, 4p, 5q, 18q, and 21q, frequently occurred in the high TME score subtype (see *Figure S3B*). The minimal common region analysis showed that amplifications of 8q24.21 and 11q13.3, and deletions of 2q22.1, 9p21.3, and 16q23.1 frequently occurred in the high TME score subtype, while amplifications of 11q13.3, and deletions of 7q31.1, 9p21.3, 16q23.1, and 4q22.1 frequently occurred in the low TME score subtype (see *Figure S3C* and *Table S5*). Thus, the results indicated a more instable status of chromosomes in patients with the low TME score subtype. Based on the CNV results, the tumor purity and ploidy analysis by ABSOLUTE showed that the estimated tumor purity ranged from 0.19 to 1.00, and tumor ploidy ranged from 1.75 to 10.12 (see *Figure 4C,4D* and *Table S6*). Thus, genomic disorder appears to be a common phenomenon in the process of tumorigenesis. There were no significant differences between high and low TME score subtypes in terms of tumor purity and ploidy (Wilcoxon-test, P=0.62 and P=0.55, respectively).

#### **The therapeutic benefits of the TME score**

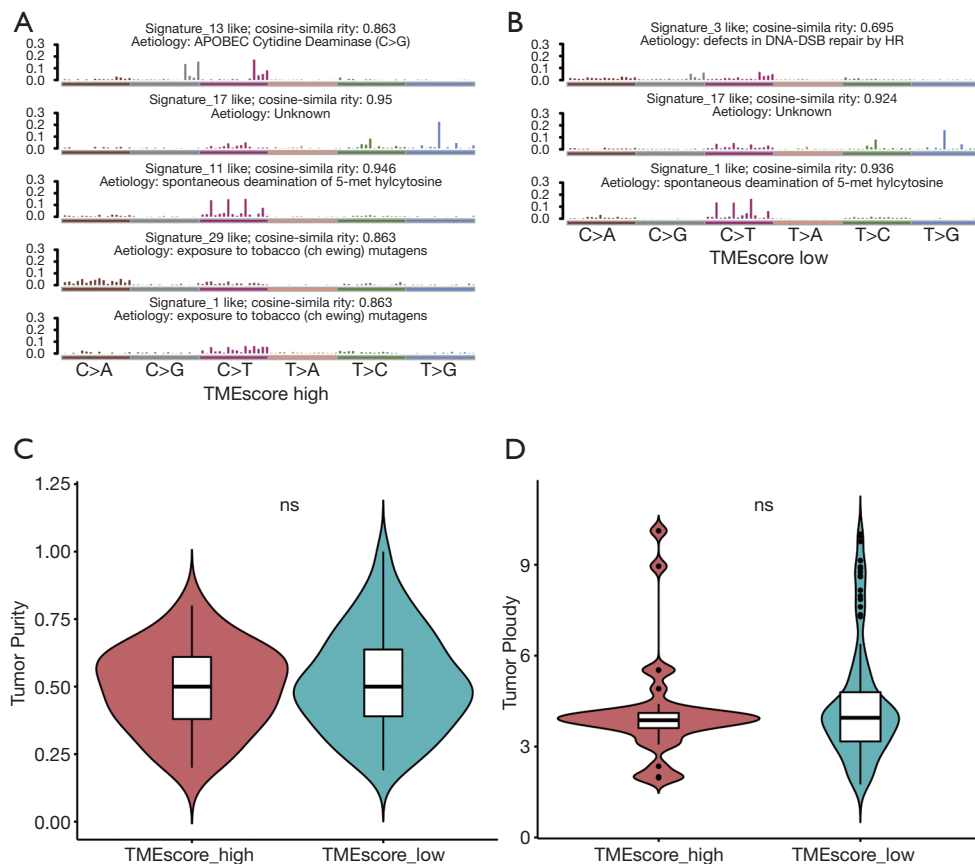
The TME score is a significant biomarker that can add prognostic information beyond that provided by standard clinicopathological properties (6,28). The predictive value of the TME score in EC patients who received adjuvant chemotherapy was examined in a TCGA-ESCA data set. The EC cohort included the mRNA expression levels of patients (n=42) before and after treatment with adjuvant chemotherapy. However, there was no statistically significant differences between patients before and after treatment with adjuvant chemotherapy (P=1.0 for TCGA-ESCA; see *Figure S4A*). In addition, MSI/dMMR was more frequently detected in well-differentiated tumors and younger patients ( $\leq 60$  years) (29). We also tried to determine whether the TME score and MSI were correlated. The relationship between the TME score and MSI was identified in EC patients who were examined in the Sensor score and MANTIS (n=157) cohorts. The TME score was not correlated with MSI/dMMR (see *Figure S4B,S4C*).

The inhibition of the immunological checkpoints that block T-cell inhibitory molecules programmed cell death protein-1 (PD-1) and PD-L1, including pembrolizumab



**Figure 3** TME characteristics and the cancer somatic genome in EC. (A) The landscape of somatic mutations of EC. (B) The oncoPrint was constructed by those with low TME scores on the right (blue) and those with high TME scores on the left (red). Individual patients are represented in each column. The top bar plot indicates OS, TMB, and TME score per patient. The right bar plot shows the mutation frequency of each gene in the separate TME score groups. Clinical stage, gender, and OS status are shown as patient annotations. (B) Mutation frequency of the genes in the 2 TME score groups. (C) Mutation frequency of the genes in the 2 TME score groups. TME, tumor microenvironment; EC, esophageal carcinoma; OS, overall survival.





**Figure 4** Mutational signatures associated with TME score subtypes. (A) Mutational signatures associated with the high TME score subtype on the left and the low TME score subtype on the right. (B) The tumor purity and ploidy in the 2 TME score subtypes. (C) Tumor purity of TMEscore-high and TMEscore-low groups. (D) Ploidy of tumor samples in the TMEscore-high and TMEscore-low groups. The 'ns' mean 'no significant difference'. TME, tumor microenvironment.

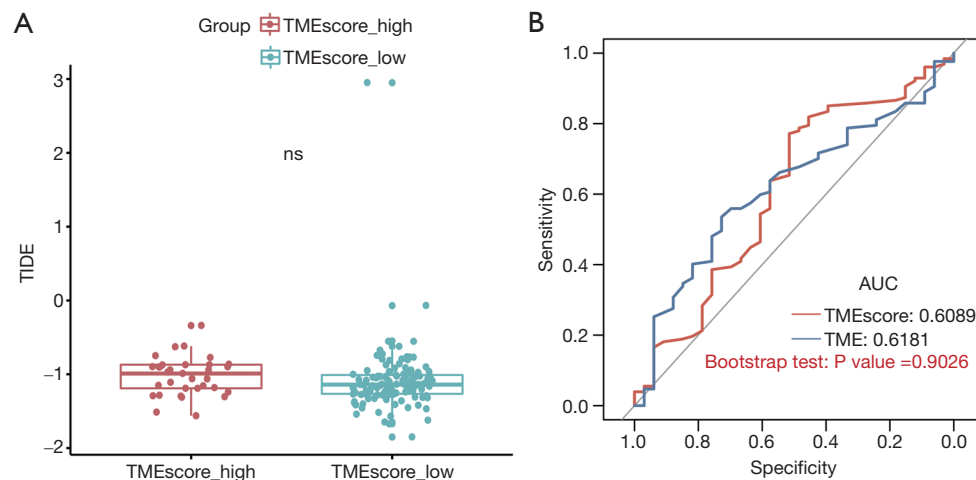
and nivolumab, has emerged as an anti-cancer treatment with unprecedented and synergistic survival benefits (30). Previous research has implied that the Tumor Immune Dysfunction and Exclusion (TIDE) (<http://tide.dfci.harvard.edu/>) score model was significantly associated with the efficacy of immunotherapy (31). Further, the TIDE model was used to evaluate the clinical efficacy of immunotherapy for high and low TME score subtypes. By Kaplan-Meier analysis, the high TIDE score subtype had much shorter progression free survival than low TIDE score subtype in EC (Figure S4D). However, there was not significant difference in survival prediction efficiency between the low TIDE score subtype and the high TIDE score subtype (Wilcoxon-test,  $P=0.054$ ) (Figure 5A).

The TMB, which may predict clinical responses to immune checkpoint inhibitors (ICIs) (32), was evaluated

by a receiver operating characteristic (ROC) analysis. The TMB is an emerging biomarker for predicting the efficacy of ICIs. The predictive efficiency of the TME score was not significantly different to that of the TMB based on ROC curve analysis (in which the AUC of the TME score was 0.6089 and the AUC of the TMB was 0.6181;  $P>0.051$ ; see Figure 5B). Thus, both TME score and TMB can be used to predict the survival of patients with EC, and the results were consistent.

## Discussion

Accurate prognostic assessment is critical in selecting the appropriate treatment. Oncologists routinely use the tumor, node, metastasis (TNM) staging system in prognosis. However, as clinical outcomes vary among patients in the



**Figure 5** TME score is a prognostic biomarker that predicts the effect of immunotherapy. (A) The distribution of TIDE in the 2 TME score subtypes. (B) The ROC curves of TME score and the TMB. TME, tumor microenvironment; ROC, receiver operating characteristic; TMB, tumor mutational burden. The ‘ns’ mean ‘no significant difference’.

same TNM stages, a more accurate prognostic biomarker is needed (15,33,34). Recently, research has revealed that the TME is a promising predictive and prognostic biomarker. IHC has been widely used in various types of cancers, including EC, to characterize the tumor-immune infiltrates within the TME. However, IHC has a number of limitations; for example, it cannot assay cell types simultaneously. In the present study, we applied a bioinformatics method called CIBERSORT that integrates the expression signatures of immune cells to rapidly profile the immune infiltration patterns in EC. Notably, we found a significantly negative correlation between eosinophils and EC survival. Tumor infiltrating eosinophils have been reported in a variety of solid tumors, including EC; however, the function of eosinophils may depend on the cellular composition of the local TME and the presence of a costimulatory factor (35). Eosinophils are regulated by numerous molecules in which interleukin-5 (IL-5) is the most specific and critical cytokine that is responsible for eosinophils biology (36-38). IL-5 has been reported to be positively regulated by miR-1248 (39). MiR-1248 expression in the low TME score group was higher than that of the high TME score group, and high miR-1248 expression was associated with poor survival. These results further support the findings of this study.

We also constructed a classifier, referred to as the TME score, and examined its use as a prognostic tool. We found a significant correlation between the TME score and patients’ survival, which suggests that the TME score

is a prognostic biomarker for EC. However, due to the scarcity of RNA-sequencing data sets containing clinical information, validations using other datasets could not be conducted in the current study. Thus, we explored the biological underpinnings of the TME signature to support our findings. Previous studies have suggested that *TP53* mutations lead to poor clinical outcomes in a number of cancers (40), and *PCLO*, which is a frequently mutated and amplified gene in ESCC, contributes to ESCC aggressiveness by stabilizing *EGFR* and promoting *EGFR* dependent signaling (27). Our in-depth analysis also revealed a more instable chromosome status in the low TME score group, which may also contribute to the association between low TME scores and poor clinical prognosis.

There are increasing evidences that the TME has substantial promise as a predictive biomarker for the ICI response (41-43). Jiang *et al.* developed a computational method, called the TIDE tool, to predict the immune checkpoint blockade response (44). However, the lack of treatment information in the study suggests that TIDE tool may not be applicable. Therefore, it is necessary to accept immunotherapy-related RNA data sets to evaluate the value of TME score in predicting the benefit of immunotherapy.

## Conclusions

We designed a TME signature to evaluate the comprehensive TME. Our TME score classifier effectively predicted

the prognosis of EC patients, which may have important implications for identifying subgroups of EC patients who are at low or high risk of poor survival.

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## Footnote

*Reporting Checklist:* The authors have completed the REMARK reporting checklist (available at <https://dx.doi.org/10.21037/jgo-21-371>).

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*Ethical Statement:* The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). Institutional ethical approval and informed consent were waived.

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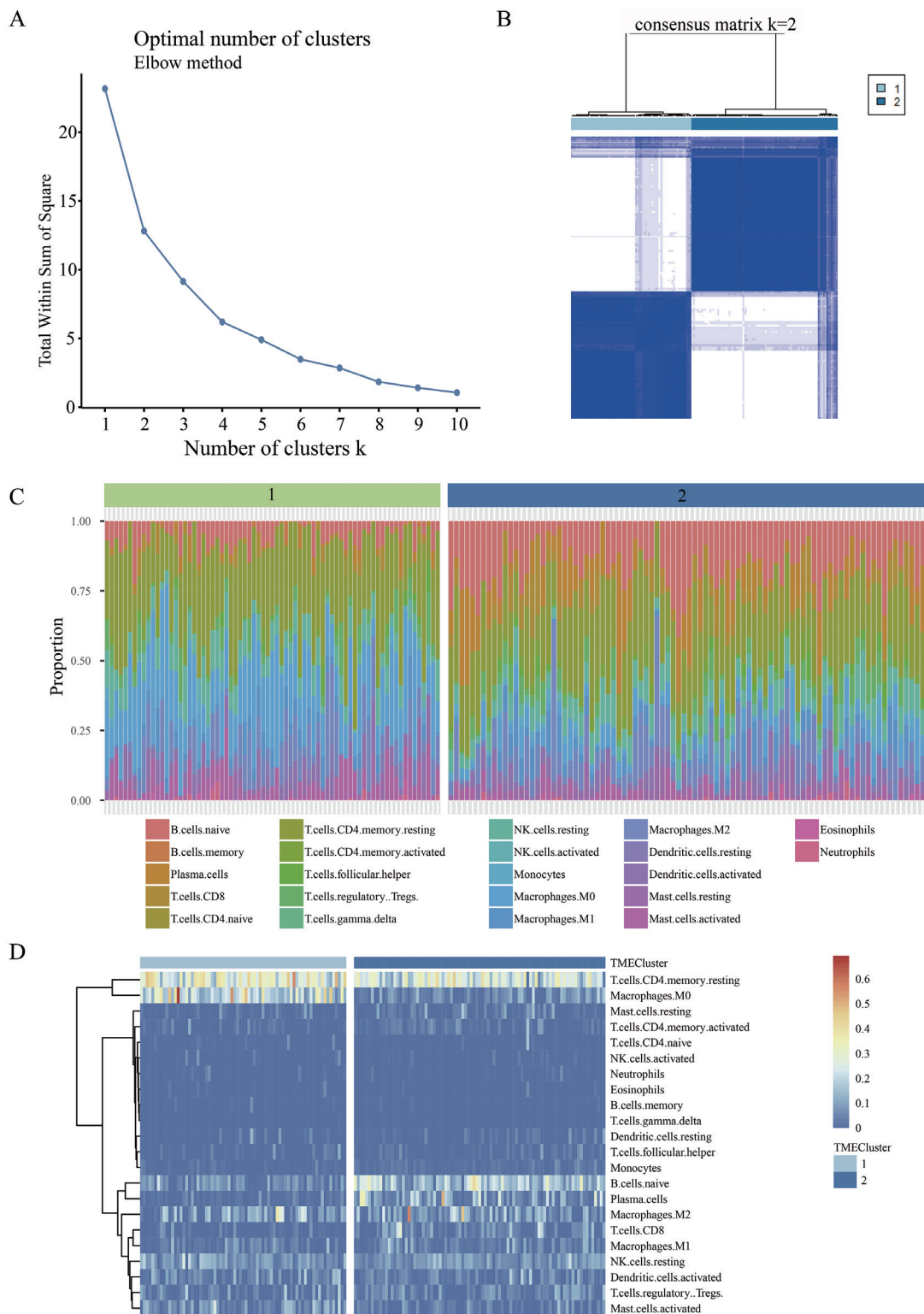
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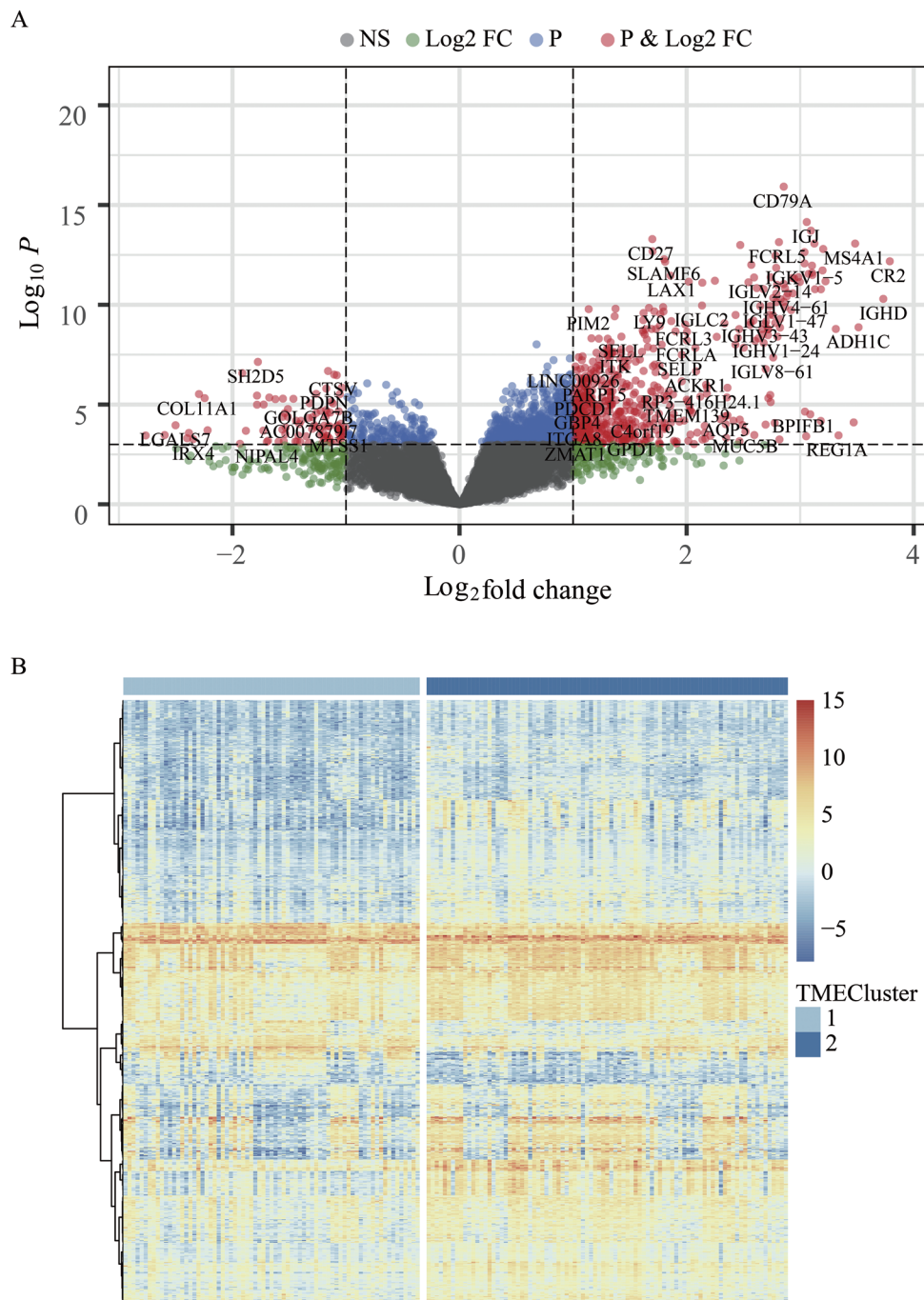
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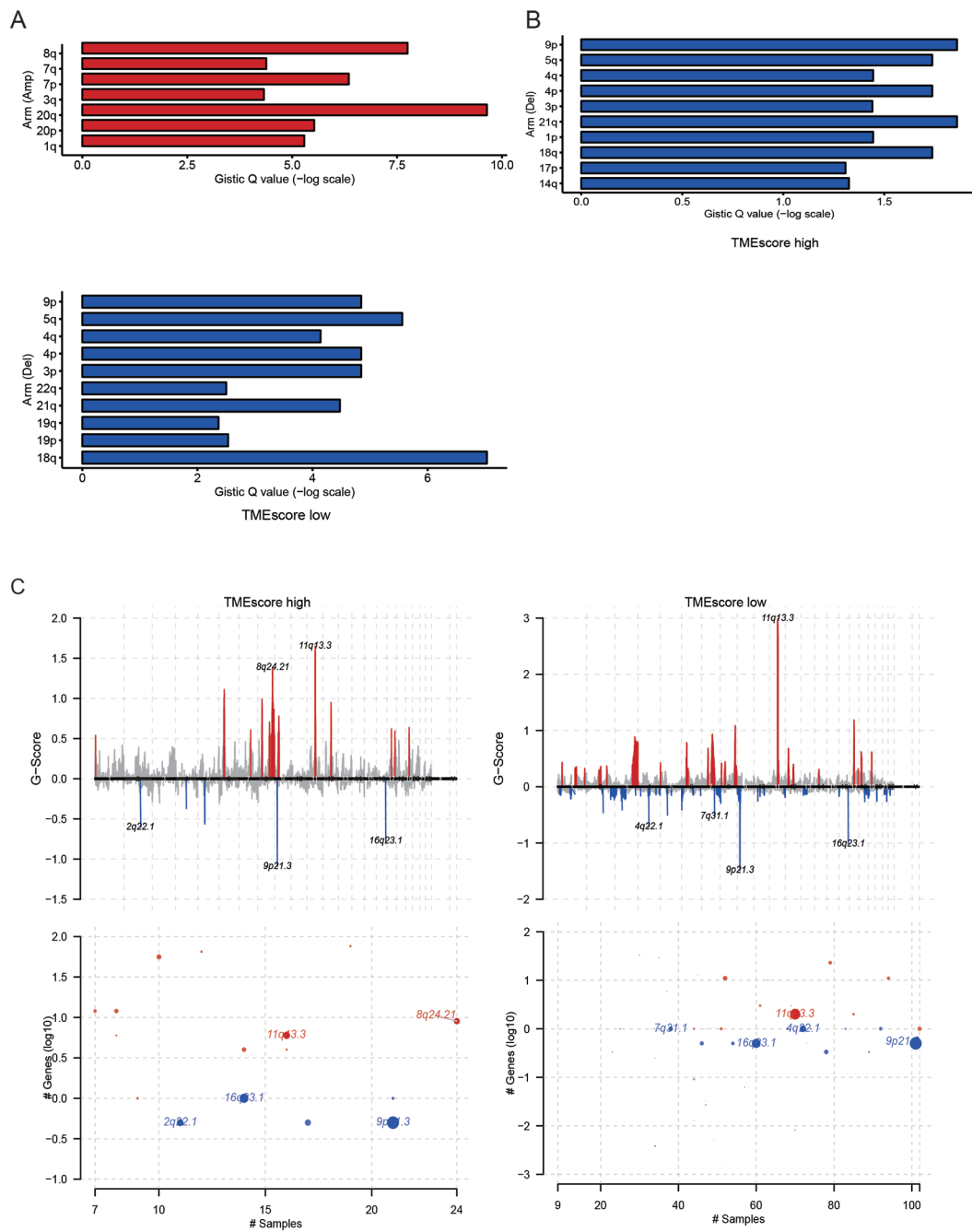
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**Figure S1** TME clusters of tumor-immune cells. (A) Estimation of the optimal k value using the elbow method. (B) Consensus matrixes for K=2, displaying the clustering stability using 1,000 iterations of hierarchical clustering. (C) The proportion of tumor-immune cells in the TME cluster. (D) The distribution of tumor-immune cells in the TME cluster. TME, tumor microenvironment.

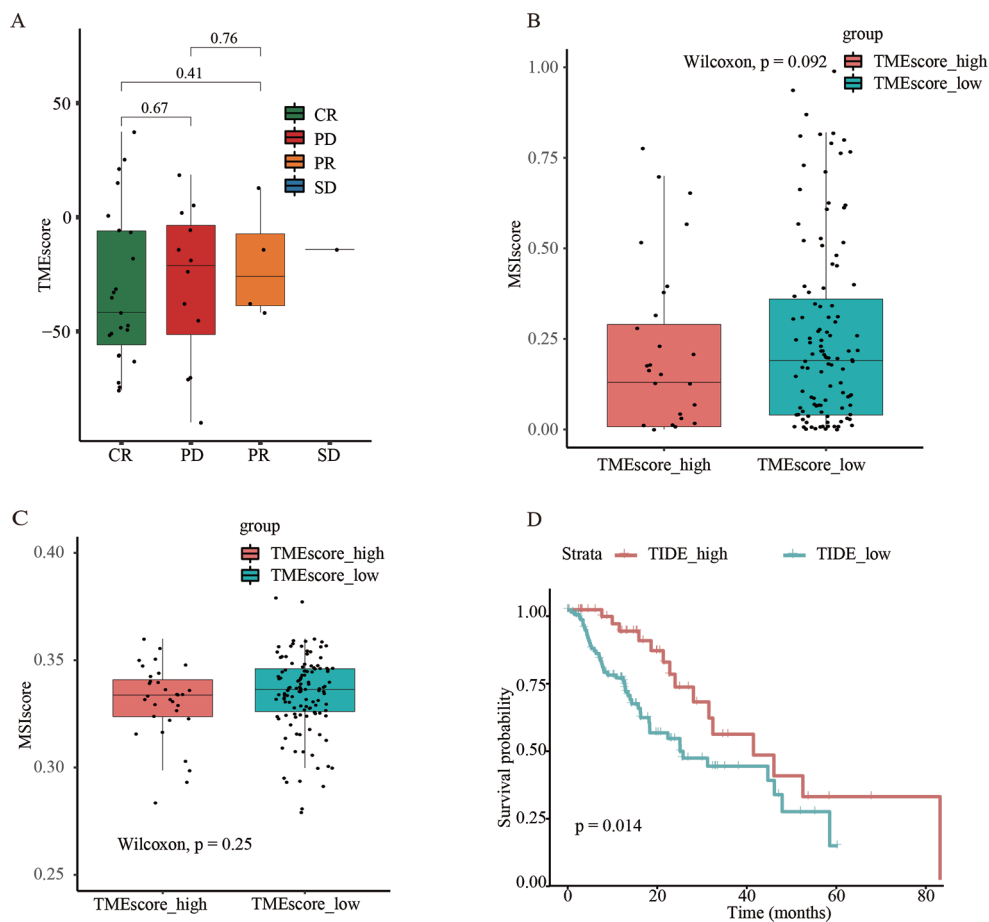


**Figure S2** DEGs in the TME cluster. (A) Volcano plot of DEGs in the TME cluster. (B) The expression values of DEGs in the TME cluster. DEG, differentially expressed gene; TME, tumor microenvironment.



**Figure S3** CNV analysis and Kaplan-Meier analysis. (A) Amplification and deletion of chromosome arm in both TME score-low. (B) Deletion of chromosome arm in both TME score-high. (C) Distribution of localized copy number amplification and deletion regions (red and blue represent amplification and deletion, respectively). CNV, copy number variation; TME, tumor microenvironment.





**Figure S4** The therapeutic benefits of the TME score. (A) The predictive value of the TME score in EC patients who received adjuvant chemotherapy. (B and C) The relationship between the TME score and MSI (Sensor score and MANTIS cohorts). (D) Kaplan-Meier curves of TIDE. TME, tumor microenvironment; EC, esophageal carcinoma; MSI, microsatellite instability; TIDE, the Tumor Immune Dysfunction and Exclusion; CR, complete response; PR, partial response; SD, stable disease; PD, progressive disease.

**Table S1** Association between immune cells and OS

Cell type	Risk	Log rank P value
Eosinophils	Negative	0.033
NK.cells.activated	Positive	0.106
Dendritic.cells.resting	Negative	0.179
T.cells.CD4.memory.activated	Negative	0.202
Mast.cells.activated	Negative	0.254
B.cells.memory	Positive	0.292
T.cells.follicular.helper	Negative	0.297
Macrophages.M1	Positive	0.306
Mast.cells.resting	Positive	0.315
Plasma.cells	Positive	0.320
Neutrophils	Negative	0.565
T.cells.gamma.delta	Positive	0.579
Monocytes	Positive	0.608
T.cells.CD4.naive	Positive	0.619
T.cells.regulatory.Tregs.	Negative	0.623
T.cells.CD8	Positive	0.629
Macrophages.M2	Negative	0.648
T.cells.CD4.memory.resting	Negative	0.655
NK.cells.resting	Negative	0.690
Macrophages.M0	Negative	0.719
B.cells.naive	Positive	0.775
Dendritic.cells.activated	Positive	0.833

OS, overall survival.

**Table S2** Signature genes positively associated with the TME

Genes
<i>IGHD, SSTR1, PKDCC, ACSL5, HOXB-AS3, PPP1R16B, DPCR1, SLC44A4, GNA14, DSC3, MECOM, PRR26, C16orf74, CHDH, DUOX2, HOMER3, SLC1A1, CTSE, ALDOB, TRHDE, CD5, SNAI2, MYRF, SHROOM3, GSDMB, BCL2L14, CLDN2, USH1C, ERN2, HABP2, CA13, WDR66, DMBT1, TMPRSS2, GPD1, AC007386.4, TJP3, SULT1B1, ZG16B, PPP1R1B, FAM221A, RP11-324O2.3, AKR7L, ADH6, GATA6, KCNK5, FUT4, TESC, VILL, S1PR5, IRF8, PIWIL4, DNAJC22, PRR15L, FA2H, C9orf152</i>

TME, tumor microenvironment.

**Table S3** Signature genes negatively associated with the TME

Genes
<i>IGJ, CAPN5, CCNI2, HID1, ATP10B, VNN2, PLEKHA6, ARHGEF38, AGT, MYO1A, KLRB1, RP11-1220K2.2, HOXB6, CES3, LGALS4, HSH2D, C4orf19, MUC5B, ARTN, BTNL8, ICA1, TMC5, IQGAP2, VSIG2, SULT1C2, SLC37A1, CDX2, TRIM31, PLS1, METTL7B, SH3BGR2, SAMD5, SMIM24, MYO15B, ADAM28, GTF2IRD2P1, KIAA1244, NCMAP, PPARG, RP3-395M20.8, SLC4A4, DDAH1, HNF4G, TOX3, AC006042.6, CTSS, PIGR, IL17RB, RAB17, OLFM4, ONECUT2, XK, SHH, SLC3A1, GATM, SPNS3, NOSTRIN, DLX2, HNF1A-AS1, GPR35, CYP3A5, PIP5K1B, RP11-739N20.2</i>

TME, tumor microenvironment.

**Table S4** Landscape of the TME in EC

	Age	Sex	Grade	OS (months)	OS status	Score	Group	TME cluster	Purity	Ploidy	TMB
TCGA-JY-A93E-01	61	Male	Stage III	25.2	0	54.21742431	High	2	0.46	3.75	147
TCGA-L5-A8NT-01	69	Male	Stage II	27.1	0	53.3347741	High	2	0.39	3.45	87
TCGA-L5-A4OU-01	81	Male	Stage II	28.98	0	48.7687703	High	2	0.42	4.14	146
TCGA-2H-A9GR-01	80	Male	Stage II	32.42	1	45.99521387	High	2	0.2	3.87	380
TCGA-2H-A9GK-01	43	Male	Stage III	7.62	1	45.73813682	High	2	0.29	3.62	153
TCGA-JY-A938-01	75	Male	Stage II	34.82	0	43.65594148	High	2	0.7	3.85	115
TCGA-JY-A6FD-01	51	Female	Stage II	67.97	0	43.03214985	High	2	0.65	3.93	114
TCGA-IG-A4P3-01	48	Male	Stage II	18.63	1	40.02176844	High	2	0.34	4.04	227
TCGA-Z6-A8JD-01	53	Male	Stage II	3.42	0	36.12807077	High	2	0.25	4.11	182
TCGA-L5-A4OH-01	71	Male	Stage I	32.59	0	29.05047095	High	2	0.62	3.08	174
TCGA-JY-A939-01	77	Male	Stage II	21.68	0	25.80574654	High	2	0.34	4.05	52
TCGA-KH-A6WC-01	82	Male	Stage I	6.27	0	25.41678036	High	2	0.61	2	26
TCGA-L5-A8NW-01	55	Male		46.06	1	23.96660129	High	2	0.4	4.4	130
TCGA-L5-A8NE-01	77	Male	Stage II	55.45	0	22.37891614	High	2	0.28	4.91	214
TCGA-LN-A49S-01	59	Male	Stage II	13.14	0	21.97945873	High	2	0.48	8.95	126
TCGA-V5-AASX-01	74	Male		8.97	0	20.61834862	High	2	0.61	3.84	267
TCGA-L5-A8NU-01	84	Male	Stage II	83.18	1	18.42952882	High	2	0.38	3.76	52
TCGA-L5-A4OG-01	79	Female	Stage I	4.7	0	17.35606076	High	2	0.5	3.85	148
TCGA-L5-A43C-01	81	Male		3.15	0	47.81625028	High	1	0.62	3.61	121
TCGA-VR-A8EY-01	44	Female	Stage II	33.67	0	45.20710463	High	1	0.74	2.35	91
TCGA-JY-A6FB-01	77	Male	Stage I	60.35	0	42.01339107	High	1	0.54	3.94	75
TCGA-IG-A5B8-01	72	Male	Stage I	0.79	1	41.79390784	High	1	0.38	3.91	231
TCGA-LN-A49U-01	62	Male	Stage II	15.34	0	37.46944181	High	1	0.57	4.17	115
TCGA-S8-A6BW-01	51	Male	Stage I	20.37	0	36.18138671	High	1	0.56	3.29	85
TCGA-L5-A88Z-01	70	Female	Stage II	7.39	0	35.72041456	High	1	0.8	1.97	59
TCGA-JY-A6FG-01	50	Male	Stage III	41.49	1	32.61960224	High	1	0.57	4.11	186
TCGA-L5-A891-01	51	Male		3.75	0	31.09107672	High	1	0.45	5.52	144
TCGA-JY-A6FH-01	53	Male	Stage II	47.34	0	30.47473299	High	1	0.22	5.53	108
TCGA-JY-A93F-01	58	Female	Stage I	24.01	0	27.08134281	High	1	0.55	4.09	84
TCGA-IG-A3QL-01	54	Male	Stage II	35.18	0	25.3023902	High	1	0.64	3.07	77
TCGA-LN-A8HZ-01	56	Male	Stage II	12.32	0	21.31145906	High	1	0.5	10.12	91
TCGA-V5-A7RC-06	55	Male		3.42	1	18.62300766	High	1	0.47	3.67	75
TCGA-IG-A97I-01	58	Male	Stage II	12.16	0	17.30201828	High	1	0.65	2	99
TCGA-IG-A625-01	60	Male	Stage III	12.81	1	16.51455926	Low	2	0.66	4.39	45

**Table S4** (continued)

**Table S4** (continued)

	Age	Sex	Grade	OS (months)	OS status	Score	Group	TME cluster	Purity	Ploidy	TMB
TCGA-JY-A93D-01	51	Male	Stage III	31.54	1	15.18941367	Low	2	0.25	2.25	95
TCGA-Z6-AAPN-01	57	Male	Stage II	2.66	0	14.52304483	Low	2	0.41	3.81	330
TCGA-V5-AASV-01	67	Male	Stage II	15.34	0	12.97941016	Low	2	0.75	2.26	89
TCGA-Z6-A9VB-01	53	Male	Stage III	1.31	0	12.9078017	Low	2	0.34	7.61	81
TCGA-IG-A3YB-01	61	Male	Stage III	2.63	0	12.72295909	Low	2	0.61	4.02	65
TCGA-V5-A7RB-01	59	Male		5.29	1	12.17730994	Low	2	0.44	4.17	243
TCGA-L5-A8NV-01	75	Male	Stage II	52.53	1	11.44863349	Low	2	0.38	3.14	100
TCGA-JY-A93C-01	47	Male	Stage III	23.16	0	11.09493906	Low	2	0.55	3.16	80
TCGA-L5-A4OX-01	60	Male	Stage II	7.42	1	7.248113325	Low	2	0.65	3.87	115
TCGA-2H-A9GJ-01	57	Male	Stage I	58.51	1	6.561082013	Low	2	0.7	2.1	97
TCGA-L5-A88T-01	86	Male	Stage II	22.8	0	4.715614504	Low	2	0.45	4.01	77
TCGA-L5-A43E-01	74	Male	Stage I	30.22	0	4.53116118	Low	2	0.51	3.36	135
TCGA-L5-A893-01	71	Female	Stage I	3.02	0	3.793991204	Low	2	0.37	4.2	120
TCGA-2H-A9GL-01	74	Male	Stage III	5.91	1	3.728631352	Low	2	0.4	4.8	186
TCGA-L7-A6VZ-01	62	Male	Stage III	10.35	0	0.837002711	Low	2	0.53	8.16	98
TCGA-XP-A8T6-01	54	Male	Stage II	25.07	1	0.530405828	Low	2	0.58	3.8	71
TCGA-R6-A8W5-01	60	Male	Stage IV	15.77	1	-5.454195281	Low	2	0.2	4.15	93
TCGA-IC-A6RF-01	69	Female	Stage I	15.67	0	-5.605140944	Low	2	0.55	2.14	95
TCGA-2H-A9GO-01	58	Male	Stage IV	16.23	1	-6.066961882	Low	2	0.43	8.63	118
TCGA-IG-A8O2-01	62	Male	Stage III	4.66	1	-8.354187901	Low	2	0.61	3.2	109
TCGA-VR-AA7I-01	70	Male	Stage III	15.9	1	-10.56812262	Low	2	0.41	3.21	90
TCGA-IC-A6RE-01	59	Male	Stage II	7.69	0	-11.77930855	Low	2	0.2	4.17	462
TCGA-JY-A6FE-01	49	Male	Stage III	3.68	1	-12.74473462	Low	2	0.38	5.28	51
TCGA-L5-A4OE-01	81	Male	Stage III	23.98	1	-13.04247581	Low	2	0.49	4.6	261
TCGA-L7-A56G-01	65	Male		10.84	1	-14.06502577	Low	2	0.55	3.38	63
TCGA-L5-A4OI-01	79	Male	Stage III	19.97	0	-16.44914795	Low	2	0.66	4.33	1263
TCGA-L5-A8NG-01	77	Male	Stage III	35.94	0	-17.24047186	Low	2	0.3	7.86	118
TCGA-LN-A7HY-01	50	Male	Stage III	12.02	0	-17.91795058	Low	2	0.35	7.35	142
TCGA-L5-A8NH-01	54	Male	Stage IV	12.91	1	-23.41078436	Low	2	0.27	3.96	112
TCGA-VR-AA4D-01	53	Male	Stage II	46.16	1	-23.69132545	Low	2	0.62	3.41	77
TCGA-LN-A9FP-01	60	Female	Stage II	12.02	0	-31.30560784	Low	2	1	4.02	415
TCGA-LN-A49W-01	73	Male	Stage III	13.24	0	-32.73325872	Low	2	0.43	3.88	73
TCGA-Q9-A6FW-01	61	Male	Stage III	7.82	0	-32.85359156	Low	2	0.5	4.21	185
TCGA-LN-A8I1-01	67	Female	Stage II	13.17	0	-33.6048471	Low	2	0.63	3.46	87

**Table S4** (continued)



**Table S4** (continued)

	Age	Sex	Grade	OS (months)	OS status	Score	Group	TME cluster	Purity	Ploidy	TMB
TCGA-M9-A5M8-01	58	Male	Stage II	33.08	0	-34.38105001	Low	2	0.93	2.79	65
TCGA-VR-A8EP-01	51	Male	Stage III	27.07	0	-37.48722031	Low	2	0.42	3.08	86
TCGA-V5-AASW-01	72	Male		9.26	0	-38.88859241	Low	2	0.31	6.12	117
TCGA-2H-A9GI-01	68	Male	Stage III	14.29	1	-39.75879743	Low	2	0.55	2.66	164
TCGA-L5-A4OJ-01	70	Female	Stage I	20.99	0	-39.98917851	Low	2	0.86	1.84	218
TCGA-JY-A6FA-01	51	Male	Stage II	44.71	1	-41.72588264	Low	2	0.72	2.09	106
TCGA-L5-A4OS-01	86	Female	Stage II	58.54	0	-44.40856809	Low	2	0.19	2.7	82
TCGA-LN-A4A8-01	52	Male	Stage II	15.51	0	-49.16551343	Low	2	0.71	2.03	99
TCGA-IG-A7DP-01	50	Female	Stage III	14.85	0	-50.76160756	Low	2	NA	NA	18
TCGA-2H-A9GH-01	44	Male	Stage II	31.24	1	-51.05706433	Low	2	1	8.76	106
TCGA-L5-A4OT-01	77	Male	Stage IV	4.89	1	-53.32740443	Low	2	0.49	3.92	205
TCGA-VR-A8Q7-01	60	Male	Stage III	52.23	0	-54.60812908	Low	2	0.29	2.89	72
TCGA-L5-A88V-01	60	Male	Stage III	2.6	0	-55.41109354	Low	2	0.72	3.28	144
TCGA-L5-A4OP-01	67	Female	Stage I	7.16	0	-55.62208884	Low	2	0.69	1.81	100
TCGA-L5-A8NS-01	76	Male	Stage II	13.4	0	-55.69961448	Low	2	0.39	3.52	339
TCGA-L5-A4OO-01	75	Male	Stage III	3.32	0	-56.64870948	Low	2	0.33	4.28	88
TCGA-R6-A6KZ-01	42	Male		5.06	1	-56.70320347	Low	2	0.35	5.13	135
TCGA-VR-A8EZ-01	47	Male	Stage III	18.17	1	-59.50710675	Low	2	0.41	6.18	201
TCGA-R6-A6XQ-01	58	Male		6.34	1	-60.02680856	Low	2	0.8	9.95	65
TCGA-LN-A7HZ-01	49	Male	Stage II	13.17	0	-60.29923058	Low	2	0.5	10.01	60
TCGA-IG-A51D-01	63	Male	Stage II	17.02	0	-60.4165586	Low	2	0.5	4.27	134
TCGA-L5-A88S-01	84	Male	Stage I	15.47	0	-60.78792761	Low	2	0.82	3.9	75
TCGA-L5-A8NM-01	84	Female	Stage II	7.75	1	-61.38506461	Low	2	0.39	2.05	1085
TCGA-LN-A7HV-01	58	Male	Stage II	10.51	0	-63.02904193	Low	2	0.47	3.47	93
TCGA-R6-A8W8-01	72	Male		2.89	1	-64.20673312	Low	2	0.63	2.75	110
TCGA-L5-A8NN-01	81	Male	Stage III	5.49	0	-69.31018669	Low	2	0.31	5.21	98
TCGA-LN-A4A3-01	61	Male	Stage III	12.75	0	-70.93140118	Low	2	0.31	3.97	64
TCGA-LN-A4MQ-01	46	Male	Stage III	12.32	0	-74.28300264	Low	2	0.65	8.93	73
TCGA-L5-A8NI-01	79	Male	Stage III	13.47	1	-76.40098706	Low	2	0.34	5.4	167
TCGA-L5-A8NJ-01	77	Male	Stage III	16.46	0	-81.51012839	Low	2	0.75	2.83	194
TCGA-LN-A5U5-01	57	Male	Stage IV	4.47	1	-84.47230082	Low	2	0.53	4.18	45
TCGA-IG-A5S3-01	69	Female	Stage II	23.39	0	-87.4784893	Low	2	0.73	9.77	62
TCGA-VR-A8EX-01	63	Male	Stage IV	28.09	1	-87.52962306	Low	2	0.4	3.52	115
TCGA-LN-A49Y-01	77	Male	Stage II	12.45	0	-89.85395773	Low	2	0.88	3.84	199

**Table S4** (continued)

**Table S4** (continued)

	Age	Sex	Grade	OS (months)	OS status	Score	Group	TME cluster	Purity	Ploidy	TMB
TCGA-R6-A6DQ-01	74	Female		7.59	1	-114.0412025	Low	2	0.46	3.25	69
TCGA-2H-A9GQ-01	80	Male	Stage III	4.2	1	13.06857923	Low	1	0.75	2.84	231
TCGA-L5-A8NF-01	57	Male	Stage IV	2.66	1	12.69563845	Low	1	0.46	2.77	110
TCGA-LN-A49P-01	71	Male	Stage II	12.32	0	12.53733958	Low	1	0.49	4.39	63
TCGA-2H-A9GM-01	53	Male	Stage II	13.93	1	9.873298261	Low	1	0.51	3.96	91
TCGA-VR-AA4G-01	51	Female	Stage III	18.04	0	8.866454168	Low	1	0.33	3.94	62
TCGA-LN-A49O-01	47	Male	Stage II	13.4	0	5.367156069	Low	1	0.51	3.16	54
TCGA-L5-A43J-01	90	Male	Stage II	4.3	1	3.731963763	Low	1	0.57	3.52	512
TCGA-S8-A6BV-01	76	Male	Stage III	20.01	0	2.090819664	Low	1	0.45	2.99	112
TCGA-IG-A4QS-01	71	Male	Stage III	3.88	1	-1.83201075	Low	1	0.53	5.8	198
TCGA-L5-A4OW-01	56	Female	Stage II	7.13	1	-2.787619378	Low	1	0.52	4.12	240
TCGA-R6-A8WG-01	60	Male		12.68	1	-3.098706595	Low	1	0.82	3.59	93
TCGA-LN-A49M-01	62	Male	Stage II	12.65	0	-6.422834234	Low	1	0.76	1.75	192
TCGA-LN-A4A4-01	36	Male	Stage III	12.58	0	-6.813949388	Low	1	0.62	2.53	82
TCGA-L5-A88Y-01	76	Male		0.36	0	-8.955868983	Low	1	0.33	4.64	197
TCGA-Z6-A8JE-01	57	Male	Stage III	2.1	0	-10.84671593	Low	1	0.62	9.14	180
TCGA-RE-A7BO-01	72	Female	Stage II	7	1	-12.49851705	Low	1	0.5	3	185
TCGA-R6-A8WC-01	56	Male		2.3	0	-13.70848969	Low	1	0.84	2.59	109
TCGA-L5-A88W-01	67	Male	Stage II	25.1	1	-14.06053101	Low	1	0.69	2	124
TCGA-V5-A7RC-01	55	Male		3.42	1	-18.72974578	Low	1	0.57	3.74	75
TCGA-VR-A8EO-01	49	Male	Stage II	25.79	0	-23.7775511	Low	1	0.78	3.87	44
TCGA-L5-A4ON-01	65	Male	Stage II	18.33	1	-26.34589442	Low	1	0.3	5.53	135
TCGA-LN-A8I0-01	52	Male	Stage II	13.37	0	-29.53674198	Low	1	0.35	4.59	43
TCGA-R6-A6Y0-01	54	Male		53.91	0	-32.32474332	Low	1	0.2	4.78	151
TCGA-LN-A5U7-01	46	Male	Stage II	25.23	0	-35.06070703	Low	1	0.73	3.15	102
TCGA-V5-A7RE-01	45	Male	Stage I	16.43	0	-37.76349412	Low	1	0.28	3.59	139
TCGA-VR-A8ET-01	64	Male	Stage II	1.54	1	-39.55431646	Low	1	0.52	4.1	22
TCGA-IG-A3YA-01	53	Male	Stage III	20.76	0	-42.55182807	Low	1	0.27	4.11	80
TCGA-LN-A9FO-01	42	Male	Stage II	0.13	0	-44.22696249	Low	1	0.57	3.37	92
TCGA-IG-A3YC-01	62	Male	Stage III	20.11	0	-45.12874276	Low	1	0.22	3.16	48
TCGA-IG-A97H-01	36	Male	Stage II	14.49	0	-46.98362277	Low	1	0.25	6.39	87
TCGA-LN-A9FQ-01	62	Male	Stage II	12.84	0	-47.01459067	Low	1	0.53	7.97	86
TCGA-LN-A4A9-01	58	Male	Stage II	11.53	1	-47.05091624	Low	1	0.68	2.39	182
TCGA-LN-A5U6-01	54	Male	Stage II	12.32	0	-47.24856243	Low	1	0.63	8.61	92

**Table S4** (continued)

Table S4 (continued)

	Age	Sex	Grade	OS (months)	OS status	Score	Group	TME cluster	Purity	Ploidy	TMB
TCGA-XP-A8T8-01	49	Male	Stage II	14.36	0	-48.16359601	Low	1	0.8	3.87	87
TCGA-IG-A3I8-01	51	Female	Stage II	33.25	0	-48.2079175	Low	1	0.6	3.79	95
TCGA-VR-A8EU-01	51	Male	Stage IV	18.3	1	-48.50146976	Low	1	0.4	7.3	82
TCGA-R6-A6L4-01	27	Male		16.29	1	-49.71571801	Low	1	0.44	5.03	118
TCGA-VR-A8EQ-01	73	Male	Stage III	22.8	1	-51.52771905	Low	1	0.47	5.08	146
TCGA-L5-A8NR-01	81	Female	Stage III	8.71	0	-52.01169279	Low	1	0.45	5.74	270
TCGA-2H-A9GF-01	67	Male	Stage III	25.76	1	-53.23604225	Low	1	0.32	7.29	228
TCGA-LN-A49X-01	44	Male	Stage II	12.61	0	-54.12930235	Low	1	0.4	4.3	60
TCGA-L5-A4OM-01	54	Female	Stage I	47.9	1	-55.03814574	Low	1	0.48	4.35	61
TCGA-LN-A7HW-01	59	Male	Stage II	11.99	0	-57.65183778	Low	1	0.56	3.2	86
TCGA-R6-A6XG-01	64	Male		38.37	0	-58.51737922	Low	1	0.47	5.12	161
TCGA-IG-A6QS-01	54	Male	Stage II	9.95	1	-59.31469209	Low	1	0.7	1.95	60
TCGA-L5-A8NK-01	84	Female	Stage II	13.53	0	-59.54210181	Low	1	0.27	3.7	172
TCGA-LN-A9FR-01	70	Male	Stage II	12.25	0	-60.50119561	Low	1	0.65	8.88	64
TCGA-L5-A8NQ-01	71	Male	Stage II	21.35	1	-61.21314301	Low	1	0.51	4.25	216
TCGA-LN-A4A5-01	49	Male	Stage II	22.37	1	-67.35119522	Low	1	0.44	3.83	63
TCGA-IG-A50L-01	58	Male	Stage III	0.53	0	-67.3520416	Low	1	0.5	4.27	102
TCGA-R6-A6DN-01	58	Male		7.98	1	-68.73193408	Low	1	0.45	5.6	80
TCGA-LN-A4A1-01	60	Male	Stage II	12.58	0	-70.14206226	Low	1	0.64	4.15	104
TCGA-LN-A7HX-01	72	Male	Stage II	12.22	0	-72.31835444	Low	1	0.58	3.88	88
TCGA-VR-A8ER-01	54	Male	Stage III	12.42	1	-75.7553627	Low	1	0.88	3.05	52
TCGA-L5-A8NL-01	56	Male	Stage III	13.21	0	-78.04101838	Low	1	0.22	4.33	84
TCGA-2H-A9GN-01	70	Male	Stage III	8.94	1	-80.40115402	Low	1	0.23	4.63	97
TCGA-VR-A8EW-01	57	Male	Stage III	8.11	1	-81.96905515	Low	1	0.39	5.24	101

TME, tumor microenvironment; EC, esophageal carcinoma.

**Table S5** The recurrent occurring CNV regions

Group	Cytoband	q value
TMEscore high	11q13.3 (amp)	3.65E-08
	8q24.21 (amp)	1.16E-06
	6p21.1 (amp)	8.24E-05
	9p21.3 (del)	1.54E-19
	16q23.1 (del)	1.49E-11
	2q22.1 (del)	1.75E-07
	5q11.2 (del)	1.50E-06
TMEscore low	11q13.3 (amp)	9.69E-74
	17q12 (amp)	4.45E-19
	8q24.21 (amp)	1.54E-16
	7q21.2 (amp)	5.56E-13
	3q26.2 (amp)	5.17E-12
	6p21.1 (amp)	6.06E-10
	7p11.2 (amp)	3.46E-08
	12p12.1 (amp)	4.30E-08
	18q11.2 (amp)	5.82E-07
	19q12 (amp)	6.65E-07
	9p21.3 (del)	6.86E-95
	16q23.1 (del)	3.06E-54
	4q22.1 (del)	1.09E-31
	5q12.1 (del)	4.10E-18
	7q31.1 (del)	1.55E-19
	2q22.1 (del)	1.06E-15
	6p25.3 (del)	6.02E-13
	3p14.2 (del)	6.02E-13
	8p23.2 (del)	4.74E-07
	20p12.1 (del)	1.47E-05
	21q22.12 (del)	1.18E-05
	18q21.2 (del)	1.01E-05
	10p11.21 (del)	1.98E-05
	3p14.3 (del)	2.23E-05
	7q36.3 (del)	1.98E-05
7q36.3 (del)	1.98E-05	
9p23 (del)	2.80E-05	

CNV, copy number variation.

**Table S6** Purity and ploidy of patients with EC

Sample	Purity	Ploidy
TCGA-LN-A49M-01	0.76	1.75
TCGA-L5-A4OP-01	0.69	1.81
TCGA-L5-A4OJ-01	0.86	1.84
TCGA-IG-A6QS-01	0.7	1.95
TCGA-L5-A88Z-01	0.8	1.97
TCGA-KH-A6WC-01	0.61	2
TCGA-IG-A97I-01	0.65	2
TCGA-L5-A88W-01	0.69	2
TCGA-LN-A4A8-01	0.71	2.03
TCGA-L5-A8NM-01	0.39	2.05
TCGA-JY-A6FA-01	0.72	2.09
TCGA-2H-A9GJ-01	0.7	2.1
TCGA-IC-A6RF-01	0.55	2.14
TCGA-JY-A93D-01	0.25	2.25
TCGA-V5-AASV-01	0.75	2.26
TCGA-VR-A8EY-01	0.74	2.35
TCGA-LN-A4A9-01	0.68	2.39
TCGA-LN-A4A4-01	0.62	2.53
TCGA-R6-A8WC-01	0.84	2.59
TCGA-2H-A9GI-01	0.55	2.66
TCGA-L5-A4OS-01	0.19	2.7
TCGA-R6-A8W8-01	0.63	2.75
TCGA-L5-A8NF-01	0.46	2.77
TCGA-M9-A5M8-01	0.93	2.79
TCGA-L5-A8NJ-01	0.75	2.83
TCGA-2H-A9GQ-01	0.75	2.84
TCGA-VR-A8Q7-01	0.29	2.89
TCGA-S8-A6BV-01	0.45	2.99
TCGA-RE-A7BO-01	0.5	3
TCGA-VR-A8ER-01	0.88	3.05
TCGA-IG-A3QL-01	0.64	3.07
TCGA-VR-A8EP-01	0.42	3.08
TCGA-L5-A4OH-01	0.62	3.08
TCGA-L5-A8NV-01	0.38	3.14

**Table S6** (continued)

**Table S6** (continued)

Sample	Purity	Ploidy
TCGA-LN-A5U7-01	0.73	3.15
TCGA-IG-A3YC-01	0.22	3.16
TCGA-LN-A49O-01	0.51	3.16
TCGA-JY-A93C-01	0.55	3.16
TCGA-LN-A7HW-01	0.56	3.2
TCGA-IG-A8O2-01	0.61	3.2
TCGA-VR-AA7I-01	0.41	3.21
TCGA-R6-A6DQ-01	0.46	3.25
TCGA-L5-A88V-01	0.72	3.28
TCGA-S8-A6BW-01	0.56	3.29
TCGA-L5-A43E-01	0.51	3.36
TCGA-LN-A9FO-01	0.57	3.37
TCGA-L7-A56G-01	0.55	3.38
TCGA-VR-AA4D-01	0.62	3.41
TCGA-L5-A8NT-01	0.39	3.45
TCGA-LN-A8I1-01	0.63	3.46
TCGA-LN-A7HV-01	0.47	3.47
TCGA-L5-A8NS-01	0.39	3.52
TCGA-VR-A8EX-01	0.4	3.52
TCGA-L5-A43J-01	0.57	3.52
TCGA-V5-A7RE-01	0.28	3.59
TCGA-R6-A8WG-01	0.82	3.59
TCGA-L5-A43C-01	0.62	3.61
TCGA-2H-A9GK-01	0.29	3.62
TCGA-V5-A7RC-06	0.47	3.67
TCGA-L5-A8NK-01	0.27	3.7
TCGA-V5-A7RC-01	0.57	3.74
TCGA-JY-A93E-01	0.46	3.75
TCGA-L5-A8NU-01	0.38	3.76
TCGA-IG-A3I8-01	0.6	3.79
TCGA-XP-A8T6-01	0.58	3.8
TCGA-Z6-AAPN-01	0.41	3.81
TCGA-LN-A4A5-01	0.44	3.83
TCGA-V5-AASX-01	0.61	3.84
TCGA-LN-A49Y-01	0.88	3.84

**Table S6** (continued)

**Table S6** (continued)

Sample	Purity	Ploidy
TCGA-L5-A4OG-01	0.5	3.85
TCGA-JY-A938-01	0.7	3.85
TCGA-2H-A9GR-01	0.2	3.87
TCGA-L5-A4OX-01	0.65	3.87
TCGA-VR-A8EO-01	0.78	3.87
TCGA-XP-A8T8-01	0.8	3.87
TCGA-LN-A49W-01	0.43	3.88
TCGA-LN-A7HX-01	0.58	3.88
TCGA-L5-A88S-01	0.82	3.9
TCGA-IG-A5B8-01	0.38	3.91
TCGA-L5-A4OT-01	0.49	3.92
TCGA-JY-A6FD-01	0.65	3.93
TCGA-VR-AA4G-01	0.33	3.94
TCGA-JY-A6FB-01	0.54	3.94
TCGA-L5-A8NH-01	0.27	3.96
TCGA-2H-A9GM-01	0.51	3.96
TCGA-LN-A4A3-01	0.31	3.97
TCGA-L5-A88T-01	0.45	4.01
TCGA-IG-A3YB-01	0.61	4.02
TCGA-LN-A9FP-01	1	4.02
TCGA-IG-A4P3-01	0.34	4.04
TCGA-JY-A939-01	0.34	4.05
TCGA-JY-A93F-01	0.55	4.09
TCGA-VR-A8ET-01	0.52	4.1
TCGA-Z6-A8JD-01	0.25	4.11
TCGA-IG-A3YA-01	0.27	4.11
TCGA-JY-A6FG-01	0.57	4.11
TCGA-L5-A4OW-01	0.52	4.12
TCGA-L5-A4OU-01	0.42	4.14
TCGA-R6-A8W5-01	0.2	4.15
TCGA-LN-A4A1-01	0.64	4.15
TCGA-IC-A6RE-01	0.2	4.17
TCGA-V5-A7RB-01	0.44	4.17
TCGA-LN-A49U-01	0.57	4.17
TCGA-LN-A5U5-01	0.53	4.18

**Table S6** (continued)



**Table S6** (continued)

Sample	Purity	Ploidy
TCGA-L5-A893-01	0.37	4.2
TCGA-Q9-A6FW-01	0.5	4.21
TCGA-L5-A8NQ-01	0.51	4.25
TCGA-IG-A51D-01	0.5	4.27
TCGA-IG-A50L-01	0.5	4.27
TCGA-L5-A4OO-01	0.33	4.28
TCGA-LN-A49X-01	0.4	4.3
TCGA-L5-A8NL-01	0.22	4.33
TCGA-L5-A4OI-01	0.66	4.33
TCGA-L5-A4OM-01	0.48	4.35
TCGA-LN-A49P-01	0.49	4.39
TCGA-IG-A625-01	0.66	4.39
TCGA-L5-A8NW-01	0.4	4.4
TCGA-LN-A8I0-01	0.35	4.59
TCGA-L5-A4OE-01	0.49	4.6
TCGA-2H-A9GN-01	0.23	4.63
TCGA-L5-A88Y-01	0.33	4.64
TCGA-R6-A6Y0-01	0.2	4.78
TCGA-2H-A9GL-01	0.4	4.8
TCGA-L5-A8NE-01	0.28	4.91
TCGA-R6-A6L4-01	0.44	5.03
TCGA-VR-A8EQ-01	0.47	5.08
TCGA-R6-A6XG-01	0.47	5.12
TCGA-R6-A6KZ-01	0.35	5.13
TCGA-L5-A8NN-01	0.31	5.21
TCGA-VR-A8EW-01	0.39	5.24
TCGA-JY-A6FE-01	0.38	5.28
TCGA-L5-A8NI-01	0.34	5.4
TCGA-L5-A891-01	0.45	5.52
TCGA-JY-A6FH-01	0.22	5.53
TCGA-L5-A4ON-01	0.3	5.53
TCGA-R6-A6DN-01	0.45	5.6
TCGA-L5-A8NR-01	0.45	5.74
TCGA-IG-A4QS-01	0.53	5.8
TCGA-V5-AASW-01	0.31	6.12

**Table S6** (continued)

**Table S6** (continued)

Sample	Purity	Ploidy
TCGA-VR-A8EZ-01	0.41	6.18
TCGA-IG-A97H-01	0.25	6.39
TCGA-2H-A9GF-01	0.32	7.29
TCGA-VR-A8EU-01	0.4	7.3
TCGA-LN-A7HY-01	0.35	7.35
TCGA-Z6-A9VB-01	0.34	7.61
TCGA-L5-A8NG-01	0.3	7.86
TCGA-LN-A9FQ-01	0.53	7.97
TCGA-L7-A6VZ-01	0.53	8.16
TCGA-LN-A5U6-01	0.63	8.61
TCGA-2H-A9GO-01	0.43	8.63
TCGA-2H-A9GH-01	1	8.76
TCGA-LN-A9FR-01	0.65	8.88
TCGA-LN-A4MQ-01	0.65	8.93
TCGA-LN-A49S-01	0.48	8.95
TCGA-Z6-A8JE-01	0.62	9.14
TCGA-IG-A5S3-01	0.73	9.77
TCGA-R6-A6XQ-01	0.8	9.95
TCGA-LN-A7HZ-01	0.5	10.01
TCGA-LN-A8HZ-01	0.5	10.12
TCGA-IG-A7DP-01	NA	NA

EC, esophageal carcinoma.