(Check for updates

OPEN ACCESS

EDITED BY József Tímár, Semmelweis University, Hungary

*CORRESPONDENCE Youcheng Zhang, zhangyouchengphd@163.com

RECEIVED 20 October 2022 ACCEPTED 09 January 2023 PUBLISHED 19 January 2023

CITATION

Deng H, Li T, Wei F, Han W, Xu X and Zhang Y (2023), High expression of TMEM200A is associated with a poor prognosis and immune infiltration in gastric cancer. *Pathol. Oncol. Res.* 29:1610893. doi: 10.3389/pore.2023.1610893

COPYRIGHT

© 2023 Deng, Li, Wei, Han, Xu and Zhang. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

High expression of TMEM200A is associated with a poor prognosis and immune infiltration in gastric cancer

Hongyang Deng, Tengfei Li, Fengxian Wei, Wei Han, Xiaodong Xu and Youcheng Zhang*

Department of General Surgery, Hepatic-Biliary-Pancreatic Institute, Lanzhou University Second Hospital, Lanzhou, China

Background: Gastric cancer (GC) is one of the global malignant tumors with high incidence and poor prognosis. Exploring new GC molecular markers is important to improve GC prognosis. Transmembrane protein 200A (TMEM200A) is a member of the family of transmembrane proteins (TMEM). This study is the first to investigate the potential function of TMEM200A and its relationship with immune infiltration in GC.

Methods: The differential expression of TMEM200A was determined through the Cancer Genome Atlas (TCGA) and Gene Expression Omnibus (GEO) databases. The receiver operating characteristic (ROC) curve was drawn to assess the diagnostic value of TMEM200A for GC. The relationship between TMEM200A and the clinical characteristics of patients with GC was investigated using the Wilcoxon test and the Kruskal-Wallis test. The effect of TMEM200A on overall survival (OS) was identified using the Kaplan-Meier method, the Logrank test, the univariate/multivariate Cox regression analysis, and the nomogram prediction model. The co-expressed genes and gene set enrichment analysis (GSEA) were used to explore the potential biological functions of TMEM200A. We used the Tumor Immune Estimation Resource (TIMER) database and the ssGSEA algorithm to estimate the relationship between TMEM200A and immune cell infiltration. Furthermore, we investigated the correlation of TMEM200A with immune checkpoint/immune cell surface markers using the TCGA-STAD data set. Finally, we identified prognosis-related methylation sites in TMEM200A using MethSurv.

Results: TMEM200A was highly expressed in GC tissues. TMEM200A had a good diagnostic value for GC. High expression of TMEM200A may shorten the OS of GC patients and may be an independent risk factor for OS in GC patients. TMEM200A participates in the construction of a predictive model with a good predictive effect on the survival rate of GC patients at 1, 3, and 5 years. Co-expressed genes and GSEA indicated that TMEM200A may be an adhesion molecule closely associated with tumor invasion and metastasis. In addition, TMEM200A may be significantly associated with immune cell infiltration and immune checkpoint expression. We also found that TMEM200A contains three methylation sites associated with a poor prognosis.

Conclusion: Upregulated TMEM200A may be a promising prognostic marker for GC and is closely associated with the tumor microenvironment (TME).

KEYWORDS

bioinformatics, gastric cancer, prognostic, immune infiltration, TMEM200A

Introduction

Gastric cancer (GC) is common cancer worldwide, with more than one million new cases each year (1, 2). It has the sixth highest number of new cancers diagnoses in the world by 2020 (2). Furthermore, patients with GC are often diagnosed at an advanced stage and therefore do not have a high 5-year survival rate, making it the third leading cause of cancer-related death after lung and liver cancer (2). GC is most prevalent in East Asia, North East Asia, and North Africa (2). Surgery and platinum chemotherapy combined with fluorouracil are the mainstays of GC treatment, but the median overall survival (OS) after chemotherapy in patients with unresectable or metastatic GC is only less than 1 year (3). Despite the benefit seen with trastuzumab in combination with chemotherapy targeting the HER2 gene over chemotherapy alone (4), HER2 gene overexpression occurs in only 10%-20% of GC patients (5). Therefore, the development of new GC biomarkers or specific therapeutic targets may improve the prognosis of patients with GC.

The transmembrane protein (TMEM) family genes encode a large class of proteins that span the lipid bilayer of the cell membrane. Due to their transmembrane properties, TMEM proteins play a role in signal transduction, ion transport, and cell adhesion (6). Although some studies (7) suggested that genes in the TMEM family are essential for tumor metastasis. Different families of TMEM gene play cancer promoting or cancer suppressing roles in different cancers (8–10). In general, current knowledge of the biological functions and mechanisms of action of TMEM genes is limited (11). A deeper understanding of the function of TMEM genes may be a promising direction for the development of new therapeutic targets for tumors.

Transmembrane protein 200A (TMEM200A) (Ensembl ID: ENSG00000164484) is also a member of the TMEM gene family, located on chromosome 6q23.1, with a total length of 77.536 Kb and comprising 8 exons. Recent studies have shown that TMEM200A can regulate adipose morphology by affecting adipogenesis (12). Tan et al. (13) found that the expression of TMEM200A in familial pancreatic cancer was lower than in sporadic pancreatic cancer. Furthermore, 6q23.1, where TMEM200A is located, was identified to be associated with cleft palate and cleft lip (14). Bioinformatics analysis based on the Cancer Genome Atlas (TCGA) and Gene Expression Omnibus (GEO) databases showed that (15) TMEM200A could be used to construct a risk score for the prognosis of acute myeloid leukemia (AML) and could effectively predict the OS of patients with AML. Zhang et al. (16) analyzed somatic mutations and RNA prognostic markers in GC and initially found that TMEM200A may have a prognostic value in GC. However, the role of TMEM200A in GC is still unknown.

This study compared the differential expression of TMEM200A between GC tissues and adjacent/normal gastric mucosal tissues based on TCGA and GEO databases. Subsequently, Kaplan-Meier method, univariate/multivariate Cox regression analysis confirmed the effect of TMEM200A on the prognosis of patients with GC and used the expression of TMEM200A and clinical characteristics of patients with GC to construct a nomogram model to predict the prognosis. In addition, co-expression of genes and GSEA explored the possible biological functions and signaling pathways involved in TMEM200A. In addition, we calculated the relationship between TMEM200A and 28 immune cell infiltrates using the ssGSEA algorithm. We also explored the correlation between TMEM200A expression and immune checkpoint expression to further understand the impact of TMEM200A on the tumor microenvironment (TME). Finally, we analyzed the effect of TMEM200A on the sensitivity of common chemotherapeutic drugs and DNA methylation sites.

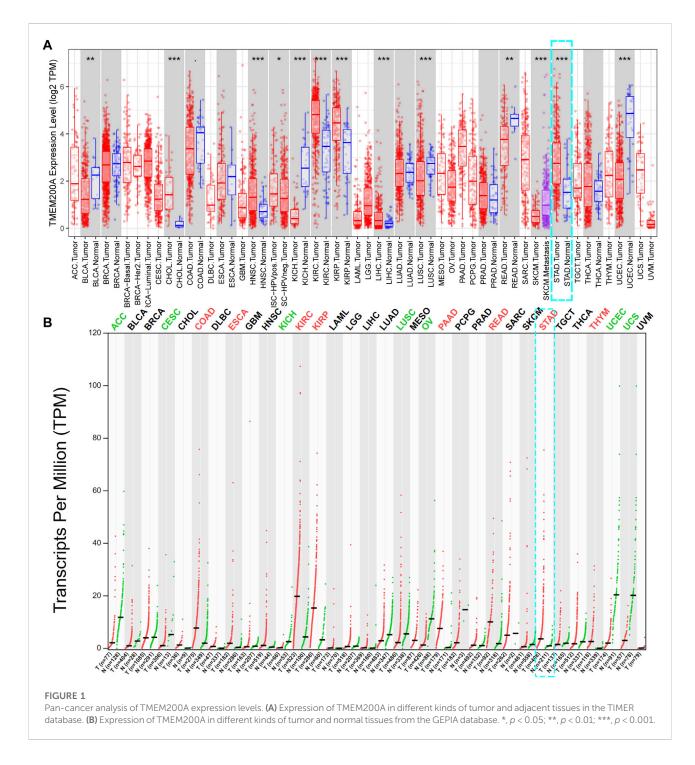
Materials and methods

Expression analysis of TMEM200A in pancancer

The Tumor Immune Estimation Resource (TIMER) database (https://cistrome.shinyapps.io/timer/) is based on the TCGA database (https://portal.gdc.cancer.gov/) of 10,897 cancer patients from 32 types of cancer and can be used to calculate differences in gene expression in cancer and adjacent tissues, correlations between genes and levels of immune cell infiltration (17). The GEPIA database (http://gepia.cancer-pku. cn/detail.php) contains gene expression data for 9736 tumor tissues from the TCGA database and 8587 normal tissues from the GTEx database (18). We used these two databases to assess the differential expression of TMEM200A in different kinds of tumor.

Expression analysis of TMEM200A in GC

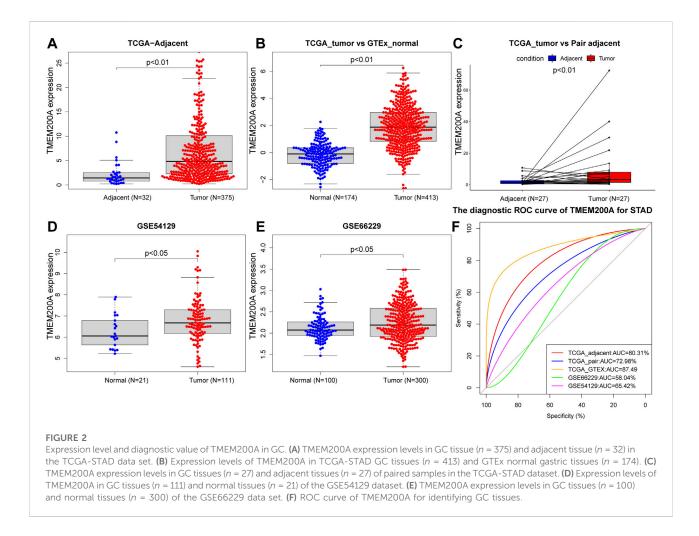
We downloaded RNA sequencing (RNA-Seq) data in TPM format for GC samples and adjacent samples from the TCGA database. The combined RNA-Seq data from the TCGA GC and GTEx databases of normal gastric tissue were downloaded from



the UCSC database (https://xenabrowser.net/datapages/) (19), and the TCGA-STAD data set relating the clinical profile and survival information. In addition, we acquired GEO database to obtain GSE54129 and GSE66229 (20) as expression validation datasets. The receiver operating characteristic (ROC) curve was drawn to evaluate the diagnostic ability of the TMEM200A to identify GC tissue.

Association of TMEM200A expression with clinical characteristics and prognosis of patients with GC

Excluding samples with partially missing data, we used RNA-Seq data for TMEM200A from 350 patients with GC in the TCGA-STAD dataset and the corresponding clinical information to assess



the prognostic impact of TMEM200A. The median expression value of TMEM200A was used to divide the GC samples into high and low expression groups. We used Kaplan-Meier curves to visualize the effect of TMEM200A on the OS of patients with GC and calculated the *p*-value using the Log-Rank test. To evaluate the independence of the prognostic impact of TMEM200A, we performed a univariate/multivariate Cox regression analysis of patient clinical information such as age, sex, and stage with TMEM200A expression. Furthermore, we evaluated the efficacy of TMEM200A in predicting the prognosis of patients with gastric cancer using the nomogram prediction model and assessed the effect of the model by the C-index and the calibration curve. GSE15459 (21) contains RNA-Seq and survival information from 192 GC patients as an external validation cohort.

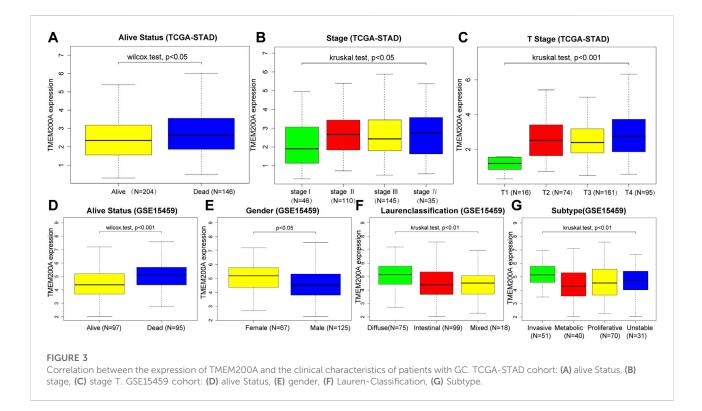
Analysis of TMEM200A co-expressed genes in GC

The LinkedOmics database (http://www.linkedomics.org/ login.php/) collects clinical characteristics, RNA-Seq,

methylation and other multi-omics data from 32 TCGA cancer cohorts and 10 CPTAC cancer cohorts (22). We used it to calculate the co-expressed genes of TMEM200A, using volcano plot to represent all co-expressed genes and heat map to represent the top 50 genes positively and negatively correlated with TMEM200A. The Pearson correlation test was used to calculate the correlation coefficient, and genes with p < 0. 05 and correlation coefficients greater than or equal to 0. 3 were selected for Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analysis by using the clusterProfiler (23) package R.

Gene set enrichment analysis

MSigDB (http://www.gsea-msigdb.org/gsea/index.jsp) collects various annotated gene sets, and gene set enrichment analysis (GSEA) is an algorithm to calculate the variability of these gene sets between two different phenotypes (24). We made the log(TPM+1) normalized TCGA-STAD dataset as input RNA-Seq data and used TMEM200A as the phenotypic label



and permutation tests 1000 times and considered statistically significant at p < 0.05 and FDR <0.05. The reference gene sets were c5.go.bp.v2002.1.Hs.symbols.gmt, c5.go.cc.v2002.1.Hs. symbols.gmt, c5.go.mf.v2002.1.Hs.symbols.gmt, and c2.cp.kegg. v2022.1.Hs. symbols.gmt. All results were sorted by NES value from largest to smallest.

Correlation analysis between TMEM200A and immune cell infiltration

The TIMER database was used to estimate the relationship between TMEM200A and the abundance of six immune cell infiltrates. Subsequently, we further calculated the relationship between TMEM200A and 28 subtypes of immune cell infiltration by the ssGSEA algorithm of the GSVA (25) R package, and the results were presented in box plots. Finally, we explored the correlation of TMEM200A expression with immune checkpoint and immune cell surface marker expression in GC tissues using the TCGA-STAD dataset with adjacent excluded samples, and the results were visualised by radar plots.

Drug sensitivity

We calculated the semi-inhibitory concentrations (IC50) of chemotherapeutic agents for each sample in the

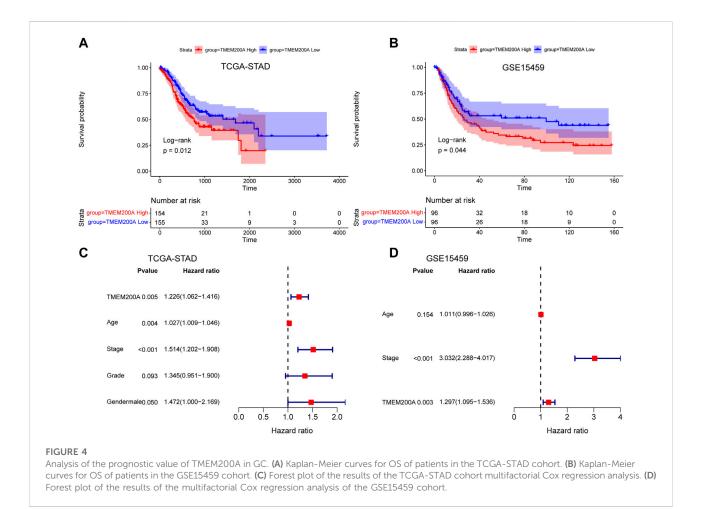
TCGA-STAD dataset using the pRRophetic R package. The patients were divided into high and low TMEM200A expression groups by median TMEM200A expression values and the difference in IC50 of these chemotherapeutic agents between the two groups was analyzed by the Wilcoxon test.

Methylation analysis of TMEM200A

DNA methylation is a process of epigenetic modification of genes in which methyl groups are mounted on DNA cytosines. Aberrant DNA methylation plays a key role in cancer development. MethSurv (https:// biit.cs.ut.ee/methsurv/) (26) is a database for DNA methylation survival analysis based on 7,358 patients of 25 different cancers in TCGA. We used MethSurv to analyze the methylation sites of TMEM200A and the impact of these methylation sites on the survival of patients with GC.

Statistical analysis

The R software (version 4.1.2) and its packages were applied to test and visualize. p-values less than 0.05 were considered statistically significant.



Results

TMEM200A is aberrantly expressed in a variety of cancers

TIMER database analysis showed that TMEM200A was significantly more expressed in cholangiocarcinoma (CHOL), head and neck squamous carcinoma (HNSC), kidney renal clear cell carcinoma (KIRC), kidney renal papillary cell carcinoma (KIRP), liver hepatocellular carcinoma (LIHC), stomach adenocarcinoma (STAD) and less expressed in bladder urothelial carcinoma (BLCA), kidney chromophobe (KICH), (LUSC), lung squamous cell carcinoma rectum adenocarcinoma (READ), skin cutaneous melanoma (SKCM) and uterine corpus endometrial carcinoma (UCEC) compared to paraneoplastic tissues (Figure 1A). The results of GEPIA database analysis suggested that TMEM200A was highly expressed in colon adenocarcinoma (CODA), esophageal carcinoma (ESCA), KIRC, KIRP, pancreatic adenocarcinoma (PADD), READ, STAD and thymoma (THYM) and low in adrenocortical carcinoma (ACC), cervical squamous cell

carcinoma and endocervical adenocarcinoma (CESC), KICH, LUSC, ovarian serous cystadenocarcinoma (OV), UCEC, uterine carcinosarcoma (UCS) compared to normal tissues (Figure 1B). Interestingly, we found that TMEM200A was significantly upregulated in GC tissues compared to both adjacent and normal tissues.

TMEM200A is significantly highly expressed in GC and has diagnostic value

After comparison, we further confirmed that the expression of TMEM200A in 375 GC samples was significantly higher than that in 32 adjacent tissues in the TCGA-STAD dataset (p < 0.01) (Figure 2A). Moreover, the expression of TMEM200A in 413 GC tissues was also significantly higher than that in 174 normal tissues (p < 0.01) (Figure 2B). Subsequently, the TCGA dataset of 27 paired GC and adjacent tissues (p < 0.01) (Figure 2C) and the GSE54129 (p < 0.05) (Figure 2D) and GEE66229 (p < 0.01) (Figure 2E) datasets also found that TMEM200A was significantly elevated in GC. ROC analysis revealed that

Characteristics	TCGA-STAD	GSE15459 (N = 192) N (%)	
	(N = 350)		
	N (%)		
Alive status			
Alive	204 (58.29)	97 (50.52)	
Dead	146 (41.71)	95 (49.48)	
Gender			
Female	124 (35.43)	67 (34.9)	
Male	226 (64.57)	125 (65.1)	
Age			
≥55	289 (83.29)	152 (79.17)	
<55	58 (16.71)	40 (20.83)	
Unkown	3	0	
Stage			
I	46 (13.14)	31 (16.15)	
II	110 (31.43)	29 (15.1)	
III	145 (41.43)	72 (37.5)	
IV	35 (10)	60 (31.25)	
Unkown	14 (4)	0	
Subtype	(-)		
Invasive	_	51 (26.56)	
Metabolic	_	40 (20.83)	
Proliferative	_	70 (36.46)	
Unstable	_	31 (16.15)	
Laurenclassification		51 (10.15)	
Diffuse		75 (39.06)	
Intestinal		99 (51.56)	
Mixed		18 (9.38)	
Grade		16 (9.36)	
Gl	0 (2 57)		
G1 G2	9 (2.57)	—	
	125 (35.71)	—	
G3	207 (59.14)	—	
Unkown	9 (2.57)	—	
M stage	212 (00.1.4)		
M0	312 (89.14)	—	
M1	23 (6.57)	—	
Unkown	15 (4.29)	—	
N stage			
N0	103 (29.43)	—	
N1	93 (26.57)	—	
N2	72 (20.57)	—	
N3	71 (20.29)	—	
Unkown	11 (3.14)	—	
T stage			
T1	16 (4.57)	_	
T2	74 (21.14)	—	
T3	161 (46)		

TABLE 1 Clinicopathologic characteristics of GC path	ients in TCGA-
STAD and GSE15459 datasets.	

TABLE 1 (Continued) Clinicopathologic characteristics of GC patients in
TCGA-STAD and GSE15459 datasets.

haracteristics	TCGA-STAD	GSE15459	
	(N = 350)	(N = 192)	
	N (%)	N (%)	
T4	95 (27.14)	_	
Unkown	4 (1.14)	_	

TMEM200A had a strong value in the diagnosis of GC. The AUC of TMEM200A in TCGA-STAD, TCGA-STAD paired samples, TCGA-GTEx dataset, GSE54129 and GSE66229 dataset were 0.803, 0.730, 0.875, 0.580, and 0.654 (Figure 2F). These results indicated that TMEM200A was abnormally upregulated in GC.

Correlation between TMEM200A and the clinical characteristics of GC patients

Table 1 summarized the clinical characteristics of all GC patients in the TCGA-STAD and GSE15459 data sets. As shown in Figures 3A-G, highly expressed TMEM200A is distributed in dead patients (TCGA-STAD, alive vs. dead, p < 0.05) (GSE15459, alive vs. dead, p < 0.001) and female patients (GSE15459, female vs. male, p < 0.05). Furthermore, the expression of TMEM200A tends to increase with increasing clinical stage (TCGA-STAD, stage I vs. stage II vs. stage III vs. stage IV, p < 0.05) and T stage (TCGA-STAD, T1 vs. T2 vs. T3 vs. T4, p < 0.001). The distribution of TMEM200A expression varies between Lauren-classification and subtypes, with the highest expression in the diffuse and invasive types of GC. These results suggested that the high expression of TMEM200A appears to accelerate the progression of GC and contribute to its more aggressive nature.

Upregulated TMEM200A has GC prognostic value

In the TCGA-STAD cohort (n = 309), the calculation of the Kaplan-Meier method showed that patients with GC with high expression of TMEM200A had significantly lower OS than those with low expression of TMEM200A (p = 0.012) (Figure 4A). Univariable/multifactor Cox regression analysis illustrates that TMEM200A was an independent risk factor for OS in GC patients (HR = 1.226, 95%CI = 1.062–1.416, p = 0.005) (Figure 4C) (Table 2). Based on the results of multifactor Cox analysis, we selected factors with p less than 0.05 to construct a nomogram prediction model for predicting the survival rate of

	Univariate cox regression		Multivariate cox regression	
Factors	HR (95%CI)	<i>p</i> -value	HR (95%CI)	<i>p</i> -value
TMEM200A	1.244 (1.082–1.431)	0.002	1.226 (1.062–1.416)	0.005
Age	1.021 (1.004-1.040)	0.017	1.027 (1.009-1.046)	0.004
Stage	1.474 (1.191–1.825)	< 0.001	1.514 (1.202–1.908)	< 0.001
Grade	1.377 (0.974-1.946)	0.070	1.345 (0.951900)	0.093
M Stage	1.675 (0.877-3.200)	0.118	_	_
N Stage	1.308 (1.118-1.530)	< 0.001	_	_
T Stage	1.299 (1.040-1.622)	0.021	_	_
Gender	1.464 (0.998–2.147)	0.051	1.472 (0.999–2.169)	>0.05

TABLE 2 Results of a univariable/multivariate Cox regression analysis based on the TCGA-STAD dataset.

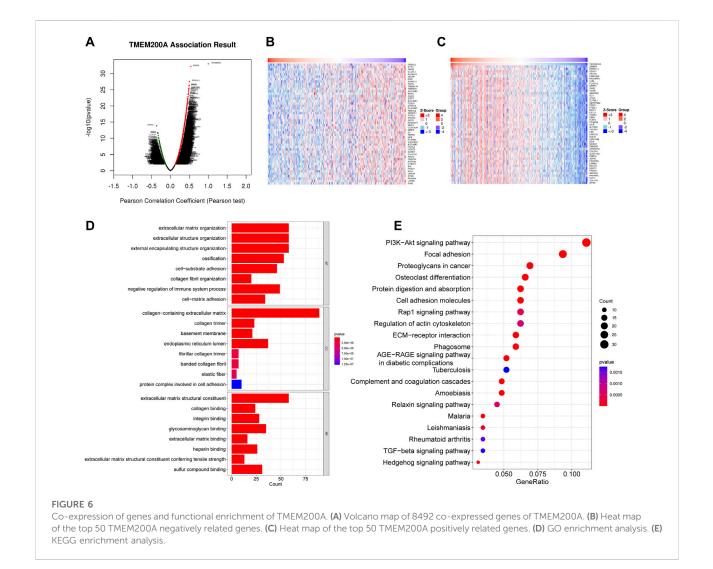
Α В 0 10 20 30 40 50 60 70 80 90 100 70 80 20 50 60 Points Points Age Age 45 70 20 95 40 45 50 55 60 65 70 75 80 Stage Stage TMEM200A TMEM200A 4 8 0 0.5 1 1.5 2 2.5 3 3.5 4 4.5 5 5.5 6 6.5 7 3 5 6 Total Points Total Points 100 120 140 160 80 0 20 40 60 80 40 60 120 160 200 1-year survival 1-year survival 0.9 0.8 0.7 0.6 0.50.40.3 0.9 0.8 0.7 0.6 0.5 0.4 3-year survival 3-year survival 0.8 0.7 0.6 0.5 0.4 0.3 0.2 0.1 0.9 0.9 0.7 0.6 0.5 0.4 0.3 0.2 0.1 0.8 5-year survival 5-year survival 0.8 0.7 0.6 0.50.40.30.2 0.1 0.9 0.7 0.6 0.5 0.4 0.3 0.2 0.1 D 0.8 С Е G F H n=306 de 0.8 0.2 0.4 0.6 0.8 0.4 0.2 0.4 0.6 0.6 0.8 n=192 d=9d Gray: ideal n+192 d+66 Gray: ideal n=192 d=5 Gray, idea 8+1000 B+1000

FIGURE 5

Nomogram plots of combining TMEM200A and clinical features evaluated survival in GC. (A) Nomogram for predicting OS in TCGA-STAD dataset. (B) Nomogram for predicting OS in the GSE15459 dataset. (C–E) The calibration curve of the nomogram for one, three, and 5 years of OS prediction in TCGA-STAD dataset. (F–H) The calibration curve of the nomogram for one, three, and 5 years of OS prediction in GSE15459 dataset.

	Univariate cox regression		Multivariate cox regression	
Factors	HR (95%CI)	<i>p</i> -value	HR (95%CI)	<i>p</i> -value
TMEM200A	1.197 (1.010–1.420)	0.038	1.297 (1.095–1.536)	0.003
Age	0.999 (0.984-1.016)	0.969	1.011 (0.996-1.026)	0.154
Stage	2.790 (2.140-3.635)	< 0.001	3.031 (2.288-4.017)	< 0.001
Gender	1.402 (0.908–2.165)	0.127	_	_

TABLE 3 Results of a univariable/multivariate Cox regression analysis based on the GSE15459 dataset.



GC patients at 1, 3 and 5 years. The C-index of the model is 0.635 (95% CI = 0.608-0.662) (Figure 5A), indicating good predictive performance. The calibration curves show a good fit between the probability of predicting OS at 1, 3 and 5 years in GC patients and the actual probability of occurrence on the column graphs (Figures 5C–E).

Furthermore, we further validated the effect of TMEM200A on the survival of patients with GC by analyzing the GSE15459 dataset (n = 192). Similarly, GC patients with high TMEM200A expression have lower OS than those with low TMEM200A expression (p = 0.044) (Figure 4B). Meanwhile, TMEM200A is an independent risk factor for the prognosis of patients with GC

10.3389/pore.2023.1610893

(HR = 1.297, 95%CI = 1.095–1.536, p = 0.003) (Figure 4D) (Table 3). We also used TMEM200A in conjunction with age and stage to build the nomogram prediction model (Figure 5B), which also had excellent prediction performance (C-index = 0.766, 95%CI = 0.745–0.787) (Figures 5F–H). In conclusion, our study showed that upregulation of TMEM200A expression promoted GC progression and shortened the OS of GC patients.

Co-expression gene analysis of TMEM200A

LinkedOmics identified a total of 8492 TMEM200A related genes (Figure 6A) of which 3,074 were negatively related (Figure 6B) and 5,418 were positively related (Figure 6C). The total number of TMEM200A positively correlated genes with correlation coefficients greater than or equal to 0.3 was 705. In general, co-expressed genes have similar biological functions. The results of GO analysis suggested that the co-expressed genes of TMEM200A might be part of the cytoskeleton and were associated with ECM and adhesion (Figure 6D). KEGG pathway enrichment analysis revealed that co-expressed genes of TMEM200A could be involved in the P13K-Akt signaling pathway, focal adhesion, cell adhesion molecules, Rap1 signaling pathway, regulation of actin cytoskeleton, ECM-receptor interaction, phagosome, relaxin signaling pathway, TGF-beta signaling pathway, and Hedgehog signaling pathway (Figure 6E). These results suggested that TMEM200A may promote the progression of GC through these pathways.

GSEA

GSEA unraveled that extracellular matrix assembly, FC receptor signaling pathway, phagocytic vesicle, extracellular matrix structural constituent, fibronectin binding, transforming growth factor beta binding, B cell receptor signaling pathway, cell adhesion molecules CAMs, ECM receptor interaction, focal adhesion, T cell receptor signaling pathway, TGF-beta signaling pathway and ubiquitin-mediated proteolysis (Figures 7A, B), indicating that elevated TMEM200A could participate in the occurrence and progression of GC through these pathway. This is similar to the results of the coexpression gene enrichment analysis of TMEM200A, suggesting that TMEM200A may be an important adhesion molecule in extracellular information traffic and may be relevant to tumor immunity.

TMEM200A is associated with immune cell infiltration in GC

We found a positive correlation between TMEM200A and CD4⁺ T cells (r = 0.284, p = 3.18e-08), macrophages (r = 0.337, p = 2.88e-11), neutrophils (r = 0.179, p = 5.36e-04) and DC (r = 0.186e-04) and DC (r = 0.188e-04) and DC (r = 0.

0.278, p = 5.37e-08) based on TIMER (Figure 8A). Furthermore, 23 of the 28 immune cell subtypes could be regulated by TMEM200A expression in GC tissues (Figure 8B). In addition, we found that TMEM200A is positively correlated with 17 common surface markers of the immune cell (Table 4). Interestingly, TMEM200A was positively correlated with the expression of 12 immune checkpoints in GC (Figure 8C), such as PDCD1, TIGIT, LAG3, CD40, etc. The above results suggested that TMEM200A probably modulated the tumor immune microenvironment (TIME) of GC, promoted the infiltration of immune cells and upregulates the expression of various immune checkpoints, which helped GC cells to immune escape and thus promoted the progression of GC.

Drug sensitivity analysis

After estimated the IC50 of 138 drugs in TCGA-STAD patients, we discovered that GC patients with high TMEM200A expression might gradually tolerate for AKT.inhibitor.VIII, Afatinib, Gefitinib, Lapatinib, and Metformin, while All-trans Retinoic Acid, Cytarabine, Nilotinib, and Crizotinib may have better therapeutic effects in these patients (Figure 9).

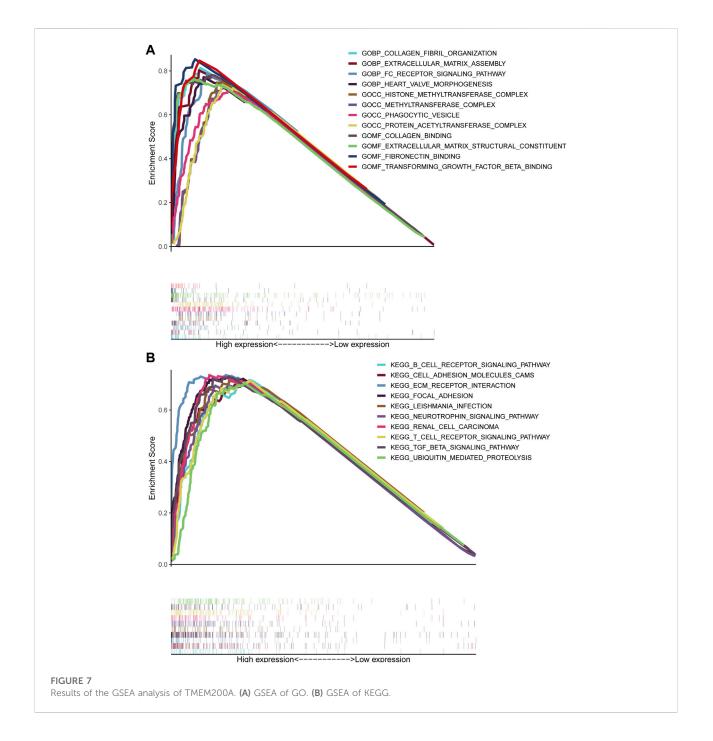
DNA methylation sites of TMEM200A

Six TMEM200A methylation sites were identified by using MethSury tool (Figure 10A), all of which had high methylation levels. Hypermethylation at three of these methylation sites was positively associated with a poor prognosis in patients with GC (Figures 10B–D). These results suggested that methylated of TMEM200A may be a potential method to inhibit the progression of GC.

Discussion

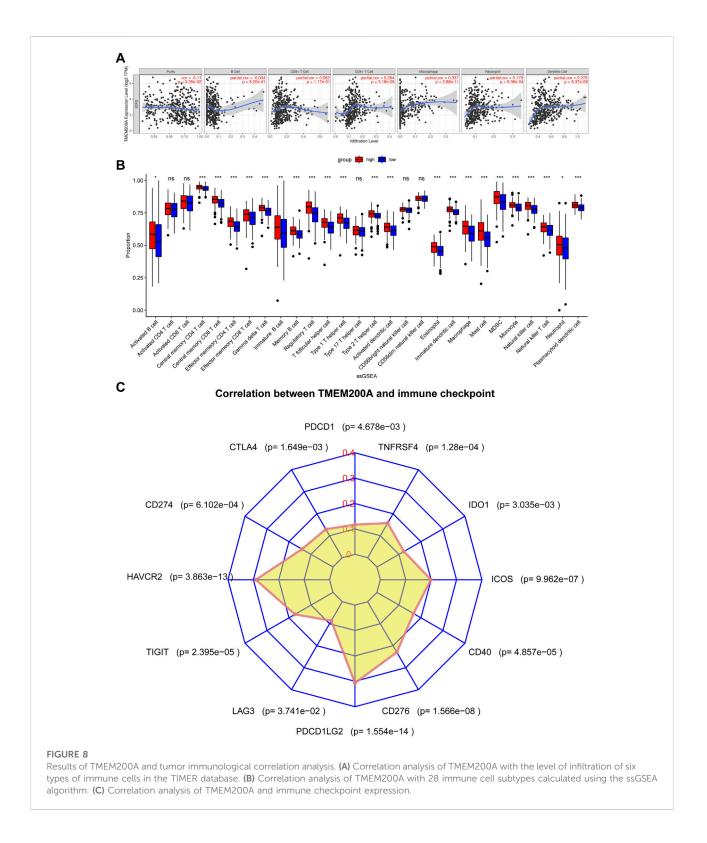
Several members of the TMEM family gene have been shown to be associated with the progression of digestive cancers. For example, interference with TMEM45B expression inhibited the proliferation migration and invasive ability of GC cells (27). TMEM106C is highly expressed in LIHC and suppresses LIHC proliferation and metastasis after knockdown (28). High TMEM180 expression have been shown to predict low survival in COAD (29). Therefore, the intrinsic link between the members of the TMEM family gene and the tumors of digestive system deserves further exploration.

To the best of our knowledge, the expression and biological function of TMEM200A in GC and the relationship with TIME have not been reported. In this study, we found that TMEM200A was upregulated in GC tissues and shortened the OS of GC

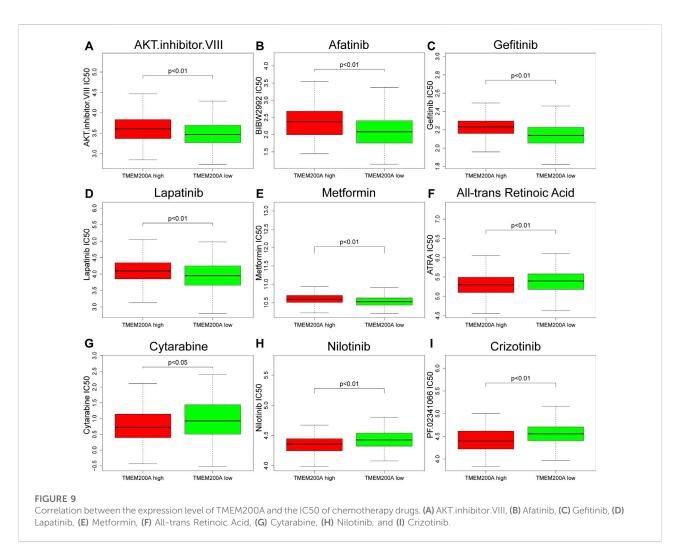


patients by calculating multiple data sets. Meanwhile, we also believe that the expression of TMEM200A increased with increasing staging. More importantly, univariable/multifactor cox regression analysis based on two independent data sets showed that TMEM200A was an independent risk factor for OS in patients with GC. Based on the results of multifactor Cox regression analysis, we constructed a nomogram prediction model using TMEM200A in combination with the clinical characteristics of GC patients, which had excellent predictive power for 1-year, 3year and 5-year survival rates of GC patients. It is reasonable to assume that high expression TMEM200A promotes the progression of GC and is a valid prognostic indicator of GC. Therefore, it is necessary to further clarify the pathways through which it can exert its cancer-promoting effects.

Enrichment analysis of TMEM200A co-expressed genes revealed that TMEM200A could act as a cell membrane adhesion molecule possibly involved in the P13K-AKT pathway, focal adhesion, cell adhesion molecules,

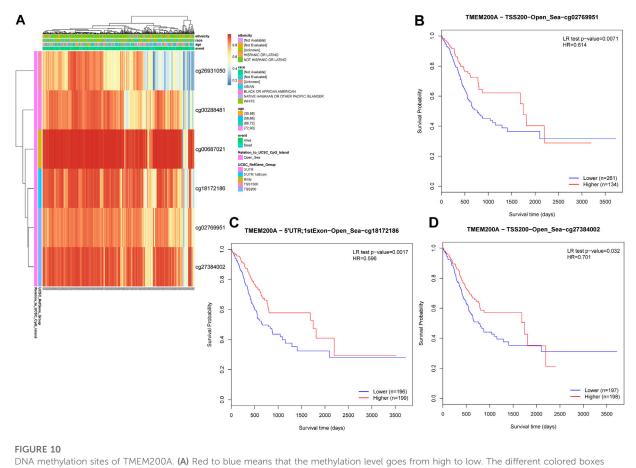


Rap1 signaling pathway, regulation of actin cytoskeleton, ECMreceptor interaction, phagosome, relaxin signaling pathway, TGF-beta signaling pathway, and Hedgehog signaling pathway. It has been shown (30) that the P13-AKT pathway can promote GC cells proliferation and invasion. Rap1, which has a high sequence similarity to the Ras oncoprotein, is an important Ras regulator and has been shown to be highly associated with cell adhesion and integrin function (31).



Relaxin signaling promotes the proliferation, migration, invasion, and adhesion of prostate cancer cells (32) and is associated with ECM remodeling (33). The TGF-beta pathway is involved in the construction of an immunosuppressive microenvironment that assists in the immune escape of tumor cells (34). Organoid models of GC confirmed (35) that the Hedgehog signaling pathway mediates PD-L1 expression in GC cells via mTOR. Focal adhesion is part of the integrin adhesion complex and plays an important role in cell migration, cell cycle and cell proliferation (36). Similarly, cell migration is regulated by the cytoskeleton (37). In addition, it has been shown (38) that ECM is a key mediator of the metastatic spread of cancer cells. Overall, these findings demonstrated that TMEM200A may act as an adhesion molecule and be an accomplice in increased aggressiveness and metastasis of GC cells. These results were also corroborated by the GSEA results. Interestingly, we observed an additional enrichment of GSEA for the transforming growth factor beta binding, the B cell receptor signaling pathway, and the T cell receptor signaling pathway, suggesting that TMEM200A may be closely associated with the TIME.

The TIME, which includes tumor cells, immune cells and cytokines (39), has long been implicated in the progression, metastasis and recurrence of cancer (40). Our study found a positive correlation between TMEM200A and CD4⁺ T cell infiltration, macrophages, neutrophils, and DCs. CD4⁺ T cells are known to function through cytokines that help immune cells such as CD8⁺ T. However, CD4⁺CD25⁺ Treg cells, a subpopulation of CD4+ T cells, suppress effector T cells and thereby impair anti-tumor immunity (41, 42). Tumor-associated macrophages (TAM) are key regulators of GC progression (43). Especially, M2 TAMs promote GC angiogenesis and participate in the construction of TIME (43). Neutrophils are the main line of defense against exogenous agents, but their release of neutrophil extracellular traps (NETs) is one of the key regulatory players in promoting tumor recurrence and metastasis (44). In addition, we found that TMEM200A high expression samples were infiltrated by 23 more immune cell



represent ethnicity, race, age, event, relation to UCSC CpG Island, and UCSC RefGene Group. (B) The Kaplan-Meier curve of TMEM200A-TSS200-Open_Sea-cg02769951 in patients with GC. (C) The Kaplan-Meier curve of TMEM200A-5'UTR; 1stExon-Open_Sea-cg18172186 in patients with GC. (D) The Kaplan-Meier curve of TMEM200A-TSS200-Open_Sea-cg27384002 in patients with GC.

subtypes, which was also confirmed by correlation analysis of immune cell surface markers with TMEM200A. Collectively, these results suggested that TMEM200A is closely related to immune cells and likely induces immune cell infiltration that affects the prognosis of patients with GC.

Immune checkpoints are involved in creating an immunosuppressive microenvironment that promotes tumor immune escape. Our study revealed that TMEM200A is positively correlated with the expression of 12 common immune checkpoints, with HAVCR2, TIGIT, PDCD1LG2, CD276, and ICOS being the most significant. HAVCR2 (TIM-3) and TIGIT are co-inhibitory receptors expressed on T cells, Foxp3⁺ Treg cells, macrophages and DC cells that can cause T cell failure (45). PDCD1LG2 (PD-L2), the second ligand of PD-1, inhibits antitumor immunity by suppressing T-cell activation (46). It has been shown (47), that CD276 expression on the surface of tumor stem cells can promote their evasion of immune responses and epithelial mesenchymal transformation. T cells expressing ICOS can inhibit the proliferation of CD8⁺ T cells

(48). These results demonstrated that although TMEM200A may induce immune cell infiltration in GC tissues. On the other hand, TMEM200A may inhibit antitumor immunity by up-regulating immune checkpoints.

Chemotherapy is a postoperative adjuvant treatment for patients undergoing surgery for GC and a dividend treatment for patients with advanced disease (5), but the ensuing resistance of tumor cells is also a cause of metastasis and recurrence (49). This study found that GC patients with high TMEM200A expression may be sensitive to All-trans Retinoic Acid, Cytarabine, Nilotinib, and Crizotinib, but may be resistant to AKT.inhibitor.VIII, Afatinib, Gefitinib, Lapatinib, and Metformin. These findings may help reduce drug resistance and improve the prognosis of GC patients. DNA methylation is considered to be one of the causes of tumorigenesis (50). As DNA methylation is reversible, regulation of DNA methylation has been suggested to be a promising direction for cancer therapy. Finally, we identified three TMEM200A methylation sites that were negatively associated with the prognosis of GC

Cell type	Marker	Cor	<i>p</i> -value
CD8 ⁺ T cell	CD8A	0.174	7.28E-04
	CD8B	0.164	1.45E-03
CD4 ⁺ T cell	CD4	0.37	1.24E-13
	CD40LG	0.276	5.66E-08
	CXCR4	0.287	1.62E-08
T cell (general)	CD2	0.254	6.17E-07
	CD3E	0.199	1.09E-04
	CD3D	0.184	3.44E-04
	CD6	0.25	9.63E-07
	SH2D1A	0.249	1.09E-06
	TRAT1	0.225	1.07E-05
	CD3G	0.24	2.66E-06
B cell	CD19	0.123	1.76E-02
	CD79A	0.136	8.51E-03
Monocyte	CD86	0.371	1.12E-13
	CSF1R	0.392	3.19E-15
TAM	CD68	0.057	2.71E-01
	CCL2	0.287	1.51E-08
	IL10	0.287	1.47E-08
M1 macrophage	IRF5	0.183	3.79E-04
1 0	PTGS2	0.196	1.37E-04
M2 macrophage	CD163	0.385	1.02E-14
1 0	VSIG4	0.347	4.82E-12
	MS4A4A	0.378	3.57E-14
Neutrophil	\$100A12	0.028	5.91E-01
I I	CEACAM3	0.132	1.06E-02
	CCR7	0.208	4.74E-05
	FPR1	0.31	9.07E-10
	SIGLEC5	0.251	8.12E-07
	CSF3R	0.266	1.78E-07
	FCAR	0.198	1.12E-04
	FCGR3B	0.136	8.19E-03
Nature killer cell	KIR2DL1	0.078	1.31E-01
	KIR2DL3	0.118	2.25E-02
	KIR2DL4	-0.017	7.38E-01
	KIR3DL1	0.101	4.97E-02
	KIR3DL2	0.063	2.24E-01
	KIR3DL3	-0.083	1.11E-01
	XCL1	0.122	1.85E-02
	XCL2	0.139	7.19E-03
	NCR1	0.055	2.84E-01
DC	ITGAX	0.375	5.38E-14
-	HLA-DPA1	0.18	4.58E-04
	HLA-DRA	0.2	9.68E-05
	HLA-DQB1	0.105	4.23E-02
	HLA-DQB1 HLA-DPB1	0.105	4.23E-02 3.64E-03
	CCL13	0.225	1.09E-05
	HSD11B1	0.223	2.23E-16
	1001101		
		(Continued in next column)	

TABLE 4 Results of correlation analysis of TMEM200A and immune cell surface markers.

TABLE 4 (*Continued*) Results of correlation analysis of TMEM200A and immune cell surface markers.

Cell type	Marker	Cor	<i>p</i> -value
Th1	TBX21	0.167	1.18E-03
	TNF	0.063	2.20E-01
	STAT1	0.196	1.29E-04
	STAT4	0.269	1.25E-07
Th2	IL13	0.131	1.11E-02
	GATA3	0.126	1.44E-02
	STAT5A	0.325	1.11E-10
	STAT6	0.281	3.14E-08
Tfh	VSIG4	0.347	4.82E-12
Th17	STAT3	0.302	2.32E-09
Treg	TGFB1	0.324	1.24E-10
	STAT5B	0.455	1.57E-20
	CCR8	0.371	1.17E-13
	FOXP3	0.292	8.44E-09
T cell exhaustion	TIGIT	0.216	2.40E-05
	GZMB	0.113	2.90E-02
	TOX	0.302	2.33E-09
	HAVCR2	0.363	3.86E-13
	LAG3	0.108	3.74E-02
	CTLA4	0.162	1.65E-03
	PDCD1	0.146	4.68E-03

The meaning of the bold values is p < 0.05.

patients. Increasing the degree of methylation of these sites may provide new ideas for GC therapy.

Conclusion

In conclusion, we found that TMEM200A was upregulated in GC and that high TMEM200A expression predicted poor prognosis. TMEM200A may play a role as an adhesion molecule regulating the GC immune microenvironment and promoting the invasive metastasis of GC cells. In addition, TMEM200A may upregulate immune checkpoints to help GC cells evade anti-tumor immune responses. Our study elucidated the potential role of TMEM200A in GC for the first time. We believed that TMEM200A deserves further study.

Data availability statement

Publicly available datasets were analyzed in this study. This data can be found here: All data generated or analysed during this study are available in the TCGA (https://portal.gdc.cancer.gov) and GEO (https://www.ncbi.nlm.nih.gov/geo/). The names of the repositories and accession number be found in the article/Materials and Methods, TCGA-STAD GSE54129, GSE66229, and GSE15459.

Author contributions

HD analyzed the data and wrote the manuscript. TL refined the language and logic. FW, WH, and XX revised the manuscript. YZ is the corresponding author and provided ideas and study design. All the authors contributed directly, substantially, and intellectually to this manuscript and approved the submitted version.

Funding

This work was supported by National Natural Science Foundation of China (82060800), the Science and Technique

References

1. Bray F, Ferlay J, Soerjomataram I, Siegel RL, Torre LA, Jemal A. Global cancer statistics 2018: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. *CA Cancer J Clin* (2018) 68(6):394–424. doi:10.3322/ caac.21492

 Sung H, Ferlay J, Siegel RL, Laversanne M, Soerjomataram I, Jemal A, et al. Global cancer statistics 2020: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. CA Cancer J Clin (2021) 71(3):209–49. doi:10.3322/caac.21660

3. Smyth EC, Nilsson M, Grabsch HI, van Grieken NC, Lordick F. Gastric cancer. *Lancet* (2020) 396(10251):635–48. doi:10.1016/S0140-6736(20)31288-5

4. Bang YJ, Van Cutsem E, Feyereislova A, Chung HC, Shen L, Sawaki A, et al. Trastuzumab in combination with chemotherapy versus chemotherapy alone for treatment of HER2-positive advanced gastric or gastro-oesophageal junction cancer (ToGA): A phase 3, open-label, randomised controlled trial. *Lancet* (2010) 376(9742):687–97. doi:10.1016/S0140-6736(10)61121-X

5. Lordick F, Carneiro F, Cascinu S, Fleitas T, Haustermans K, Piessen G, et al. Gastric cancer: ESMO clinical practice guideline for diagnosis, treatment and follow-up. Ann Oncol (2022) 33(10):1005–20. doi:10.1016/j.annonc.2022.07.004

6. Zhao Y, Zhang K, Pan H, Wang Y, Zhou X, Xiang Y, et al. Genetic analysis of six transmembrane protein family genes in Parkinson's disease in a large Chinese cohort. *Front Aging Neurosci* (2022) 14:889057. doi:10.3389/fnagi.2022.889057

7. Marx S, Dal Maso T, Chen JW, Bury M, Wouters J, Michiels C, et al. Transmembrane (TMEM) protein family members: Poorly characterized even if essential for the metastatic process. *Semin Cancer Biol* (2020) 60:96–106. doi:10. 1016/j.semcancer.2019.08.018

8. Zhuang J, Huang Y, Zheng W, Yang S, Zhu G, Wang J, et al. TMEM100 expression suppresses metastasis and enhances sensitivity to chemotherapy in gastric cancer. *Biol Chem* (2020) 401(2):285–96. doi:10.1515/ hsz-2019-0161

9. Schmit K, Michiels C. TMEM proteins in cancer: A review. Front Pharmacol (2018) 9:1345. doi:10.3389/fphar.2018.01345

10. Zhang S, Dai H, Li W, Wang R, Wu H, Shen M, et al. TMEM116 is required for lung cancer cell motility and metastasis through PDK1 signaling pathway. *Cell Death Dis* (2021) 12(12):1086. doi:10.1038/s41419-021-04369-1

11. Koteluk O, Bielicka A, Lemańska Ż, Jóźwiak K, Klawiter W, Mackiewicz A, et al. The landscape of transmembrane protein family members in head and neck cancers: Their biological role and diagnostic utility. *Cancers (Basel)* (2021) 13(19): 4737. doi:10.3390/cancers13194737

12. Lundbäck V, Kulyté A, Arner P, Strawbridge RJ, Dahlman I. Genome-Wide association study of diabetogenic adipose morphology in the GENetics of adipocyte lipolysis (GENiAL) cohort. *[J] Cell* (2020) 9(5):1085. doi:10.3390/cells9051085

13. Tan M, Schaffalitzky De Muckadell OB, Jøergensen MT. Gene expression network analysis of precursor lesions in familial pancreatic cancer. J Pancreat Cancer (2020) 6(1):73–84. doi:10.1089/pancan.2020.0007

14. Vieira AR, de Carvalho FM, Johnson L, DeVos L, Swailes AL, Weber ML, et al. Fine mapping of 6q23.1 identifies TULP4 as contributing to clefts. *Cleft Palate Craniofac J* (2015) 52(2):128–34. doi:10.1597/13-023

15. Nie L, Zhang Y, You Y, Lin C, Li Q, Deng W, et al. The signature based on seven genomic instability-related genes could predict the prognosis of acute myeloid

Innovation Project of Lanzhou University Second Hospital (CY2022-MS11), the Doctoral Supervisors Research Fund of Lanzhou University Second Hospital (bdkyjj-04), and Gansu Provincial Science Foundation for Young Scholars of China (21JR1RA161, 21JR1RA149, and 21JR11RA114).

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

leukemia patients. Hematology (2022) 27(1):840-8. doi:10.1080/16078454.2022. 2107970

16. Zhang X, Zheng P, Li Z, Gao S, Liu G. The somatic mutation landscape and RNA prognostic markers in stomach adenocarcinoma. *Onco Targets Ther* (2020) 13:7735–46. doi:10.2147/OTT.S263733

17. Li T, Fan J, Wang B, Traugh N, Chen Q, Liu JS, et al. Timer: A web server for comprehensive analysis of tumor-infiltrating immune cells. *Cancer Res* (2017) 77(21):e108–10. doi:10.1158/0008-5472.CAN-17-0307

18. Tang Z, Li C, Kang B, Gao G, Li C, Zhang Z. Gepia: A web server for cancer and normal gene expression profiling and interactive analyses. *Nucleic Acids Res* (2017) 45(W1):W98–W102. doi:10.1093/nar/gkx247

19. Goldman MJ, Craft B, Hastie M, Repečka K, McDade F, Kamath A, et al. Visualizing and interpreting cancer genomics data via the Xena platform. *Nat Biotechnol* (2020) 38(6):675–8. doi:10.1038/s41587-020-0546-8

20. Oh SC, Sohn BH, Cheong JH, Kim SB, Lee JE, Park KC, et al. Clinical and genomic landscape of gastric cancer with a mesenchymal phenotype. *Nat Commun* (2018) 9(1):1777. doi:10.1038/s41467-018-04179-8

21. Ooi CH, Ivanova T, Wu J, Lee M, Tan IB, Tao J, et al. Oncogenic pathway combinations predict clinical prognosis in gastric cancer. *Plos Genet* (2009) 5(10): e1000676. doi:10.1371/journal.pgen.1000676

22. Vasaikar SV, Straub P, Wang J, Zhang B. LinkedOmics: Analyzing multiomics data within and across 32 cancer types. *Nucleic Acids Res* (2018) 46(D1): D956–d963. doi:10.1093/nar/gkx1090

23. Yu G, Wang LG, Han Y, He QY. clusterProfiler: an R package for comparing biological themes among gene clusters. *Omics* (2012) 16(5):284–7. doi:10.1089/omi. 2011.0118

24. Subramanian A, Tamayo P, Mootha VK, Mukherjee S, Ebert BL, Gillette MA, et al. Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. *Proc Natl Acad Sci U S A* (2005) 102(43): 15545–50. doi:10.1073/pnas.0506580102

25. Hänzelmann S, Castelo R, Guinney J. Gsva: Gene set variation analysis for microarray and RNA-seq data. *BMC Bioinformatics* (2013) 14:7. doi:10.1186/1471-2105-14-7

26. Modhukur V, Iljasenko T, Metsalu T, Lokk K, Laisk-Podar T, Vilo J. MethSurv: A web tool to perform multivariable survival analysis using DNA methylation data. *Epigenomics* (2018) 10(3):277–88. doi:10.2217/epi-2017-0118

27. Shen K, Yu W, Yu Y, Liu X, Cui X. Knockdown of TMEM45B inhibits cell proliferation and invasion in gastric cancer. *Biomed Pharmacother* (2018) 104: 576–81. doi:10.1016/j.biopha.2018.05.016

28. Duan J, Qian Y, Fu X, Chen M, Liu K, Liu H, et al. TMEM106C contributes to the malignant characteristics and poor prognosis of hepatocellular carcinoma. *Aging (Albany NY)* (2021) 13(4):5585–606. doi:10.18632/aging.202487

29. Shiraishi T, Ikeda K, Tsukada Y, Nishizawa Y, Sasaki T, Ito M, et al. High expression of TMEM180, a novel tumour marker, is associated with poor survival in stage III colorectal cancer. *BMC Cancer* (2021) 21(1):302. doi:10.1186/s12885-021-08046-6

30. Chen ZF, Wang J, Yu Y, Wei W. MicroRNA-936 promotes proliferation and invasion of gastric cancer cells by down-regulating FGF2 expression and activating

P13K/Akt signaling pathway. Eur Rev Med Pharmacol Sci (2020) 24(12):6707-15. doi:10.26355/eurrev_202006_21658

31. Shah S, Brock EJ, Ji K, Mattingly RR. Ras and Rap1: A tale of two GTPases. Semin Cancer Biol (2019) 54:29–39. doi:10.1016/j.semcancer.2018.03.005

32. Feng S, Agoulnik IU, Li Z, Han HD, Lopez-Berestein G, Sood A, et al. Relaxin/ RXFP1 signaling in prostate cancer progression. *Ann N Y Acad Sci* (2009) 1160: 379–80. doi:10.1111/j.1749-6632.2008.03793.x

33. Ng HH, Shen M, Samuel CS, Schlossmann J, Bennett RG. Relaxin and extracellular matrix remodeling: Mechanisms and signaling pathways. *Mol Cel Endocrinol* (2019) 487:59–65. doi:10.1016/j.mce.2019.01.015

34. Peng D, Fu M, Wang M, Wei Y, Wei X. Targeting TGF- β signal transduction for fibrosis and cancer therapy. *Mol Cancer* (2022) 21(1):104. doi:10.1186/s12943-022-01569-x

35. Koh V, Chakrabarti J, Torvund M, Steele N, Hawkins JA, Ito Y, et al. Hedgehog transcriptional effector GLI mediates mTOR-Induced PD-L1 expression in gastric cancer organoids. *Cancer Lett* (2021) 518:59–71. doi:10.1016/j.canlet. 2021.06.007

36. Mishra YG, Manavathi B. Focal adhesion dynamics in cellular function and disease. *Cell Signal* (2021) 85:110046. doi:10.1016/j.cellsig.2021.110046

37. Seetharaman S, Etienne-Manneville S. Cytoskeletal crosstalk in cell migration. *Trends Cel Biol* (2020) 30(9):720–35. doi:10.1016/j.tcb.2020.06.004

38. Gilkes DM, Semenza GL, Wirtz D. Hypoxia and the extracellular matrix: Drivers of tumour metastasis. *Nat Rev Cancer* (2014) 14(6):430–9. doi:10.1038/nrc3726

39. Lv B, Wang Y, Ma D, Cheng W, Liu J, Yong T, et al. Immunotherapy: Reshape the tumor immune microenvironment. *Front Immunol* (2022) 13:844142. doi:10. 3389/fimmu.2022.844142

40. Fu T, Dai LJ, Wu SY, Xiao Y, Ma D, Jiang YZ, et al. Spatial architecture of the immune microenvironment orchestrates tumor immunity and therapeutic response. *J Hematol Oncol* (2021) 14(1):98. doi:10.1186/s13045-021-01103-4

41. Granito A, Muratori L, Lalanne C, Quarneti C, Ferri S, Guidi M, et al. Hepatocellular carcinoma in viral and autoimmune liver diseases: Role of CD4⁺ CD25⁺ Foxp3⁺ regulatory T cells in the immune microenvironment [J]. *World J Gastroenterol* (2021) 27(22):2994–3009. doi:10.3748/wjg.v27.i22.2994

42. Shen LS, Wang J, Shen DF, Yuan XL, Dong P, Li MX, et al. CD4(⁺)CD25(⁺) CD127(low/-) regulatory T cells express Foxp3 and suppress effector T cell proliferation and contribute to gastric cancers progression [J]. *Clin Immunol* (2009) 131(1):109–18. doi:10.1016/j.clim.2008.11.010

43. Gambardella V, Castillo J, Tarazona N, Gimeno-Valiente F, Martínez-Ciarpaglini C, Cabeza-Segura M, et al. The role of tumor-associated macrophages in gastric cancer development and their potential as a therapeutic target. *Cancer Treat Rev* (2020) 86:102015. doi:10.1016/j.ctrv.2020.102015

44. Demkow U. Neutrophil extracellular traps (NETs) in cancer invasion, evasion and metastasis. *Cancers (Basel)* (2021) 13(17):4495. doi:10.3390/cancers13174495

45. Das M, Zhu C, Kuchroo VK. Tim-3 and its role in regulating anti-tumor immunity. *Immunol Rev* (2017) 276(1):97–111. doi:10.1111/imr.12520

46. Latchman Y, Wood CR, Chernova T, Chaudhary D, Borde M, Chernova I, et al. PD-L2 is a second ligand for PD-1 and inhibits T cell activation. *Nat Immunol* (2001) 2(3):261–8. doi:10.1038/85330

47. Wang C, Li Y, Jia L, Kim JK, Li J, Deng P, et al. CD276 expression enables squamous cell carcinoma stem cells to evade immune surveillance. *Cell Stem Cell* (2021) 28(9):1597–613.e7. e7. doi:10.1016/j.stem.2021.04.011

48. Nagase H, Takeoka T, Urakawa S, Morimoto-Okazawa A, Kawashima A, Iwahori K, et al. ICOS(⁺) Foxp3(⁺) TILs in gastric cancer are prognostic markers and effector regulatory T cells associated with *Helicobacter pylori* [J]. *Int J Cancer* (2017) 140(3):686–95. doi:10.1002/ijc.30475

49. Zhang X, Xie K, Zhou H, Wu Y, Li C, Liu Y, et al. Role of non-coding RNAs and RNA modifiers in cancer therapy resistance. *Mol Cancer* (2020) 19(1):47. doi:10.1186/s12943-020-01171-z

50. Das PM, Singal R. DNA methylation and cancer. J Clin Oncol (2004) 22(22): 4632-42. doi:10.1200/JCO.2004.07.151

Glossary

GC gastric cancer OS overall survival TMEM transmembrane protein TMEM200A transmembrane protein 200A TCGA the cancer genome atlas GEO gene expression omnibus ROC receiver operating characteristic curve TME tumor microenvironment TIMER tumor immune estimation resource RNA-Seq RNA sequencing CHOL cholangiocarcinoma HNSC head and neck squamous carcinoma KIRC kidney renal clear cell carcinoma KIRP kidney renal papillary cell carcinoma LIHC liver hepatocellular carcinoma STAD stomach adenocarcinoma BLCA less expressed in bladder urothelial carcinoma KICH kidney chromophobe LUSC lung squamous cell carcinoma

READ rectum adenocarcinoma SKCM skin cutaneous melanoma UCEC uterine corpus endometrial carcinoma CODA colon adenocarcinoma ESCA esophageal carcinoma PADD pancreatic adenocarcinoma THYM thymoma ACC adrenocortical carcinoma CESC cervical squamous cell carcinoma and endocervical adenocarcinoma OV ovarian serous cystadenocarcinoma UCS uterine carcinosarcoma GO gene ontology KEGG kyoto encyclopedia of genes and genomes GSEA gene set enrichment analysis IC50 semi-inhibitory concentrations TIME tumor immune microenvironment TAMs tumor-associated macrophages NETs neutrophil extracellular traps