



The prognostic value and potential immunotherapeutic efficacy of *ACVR1* in treating gastric cancer

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Background: The discovery of biomarkers has facilitated the treatment of cancer. At present, the relationship between activin A receptor type-1 (*ACVR1*) and gastric cancer is gradually discovered. The aim of this study was to explore the expression of *ACVR1* in gastric cancer and its clinical significance, to study the relationship between *ACVR1* and tumor microenvironment (TME) for the prognosis of gastric cancer, and to further identify new targets for immunotherapy in gastric cancer.

Methods: *ACVR1* was first selected as a study gene according to several cancer and gastric cancer public datasets. Its pancancer expression was explored using the UCSC Xena database. The expression level, prognosis, and clinicopathological features of *ACVR1* in gastric cancer were analyzed using The Cancer Genome Atlas (TCGA) database. Immunohistochemistry (IHC)-based experiments were conducted to study the expression of *ACVR1* at the protein level. The IHC data were analyzed for correlations between *ACVR1* expression and various clinicopathological factors and prognosis. The correlation of this gene with the Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway, immune infiltration, immune checkpoints, drug therapy, tumor mutation burden (TMB), microsatellite instability (MSI), and mismatch repair (MMR) system was analyzed using R software.

Results: TCGA data showed that the expression of *ACVR1* was higher in gastric cancer tissues than in paracancerous tissues. Moreover, the IHC experiments indicated that *ACVR1* was upregulated in gastric cancer tissues at the protein level. Both univariate Cox and multivariate Cox results showed that the increase of *ACVR1* was closely associated with tumor stage, size, lymph node metastasis, and age. High *ACVR1* expression was linked to a poor prognosis of gastric cancer. The results also revealed that *ACVR1* was closely related to suppressive immune cells and pathways. Analyses of immune checkpoints, antitumor drug, TMB, and immune microenvironment indicated that *ACVR1* had an antitumor immune effect, promoting gastric cancer development and leading to poor immunotherapy.

Conclusions: High *ACVR1* expression can be used as an independent prognostic factor to predict the prognostic survival of patients with gastric cancer. *ACVR1* expression in gastric cancer tissues was significantly correlated with immune infiltration and may thus serve as a potential therapeutic target for gastric cancer immunotherapy.

Keywords: Gastric cancer; activin A receptor type-1 (*ACVR1*); immune microenvironment; immune checkpoints; antitumor drugs

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Introduction

Gastric cancer is the fourth leading cause of cancer death worldwide and is closely associated with infection (1). The discovery and treatment of *Helicobacter pylori* have reduced the incidence of gastric cancer. However, genetic risk factors, familial syndromes, environmental factors (2), delayed diagnosis, and lack of effective treatments (3) contribute to the prognosis of gastric cancer. Early symptoms of gastric cancer are usually not obvious; therefore, it is typically detected at an advanced stage. Despite advances in multimodal treatment such as surgery, radiotherapy, and chemotherapy, gastric cancer remains an urgent challenge in malignant tumors (4). The unique pathogenesis and active oncogenic pathways made gastric cancer hard to cure. In recent years, immunotherapy has gradually emerged, providing a new modality for gastric cancer therapy (5). A study has verified that the clinical features of gastric cancer are associated with infiltrating tumor microenvironment (TME) cells, in which the suppression of T cells causes a

reduction in the ability to mediate tumor killing and may lead to a poor prognosis (6).

TME includes immune cells, stromal cells and tumor cells. TME enhances tumor cell proliferation, migration ability, and immune escape, promoting tumor development. As the instability of the cancer cell genome predisposes increases the likelihood of drug resistance, targeting nontumor cells in the genetically relatively stable TME has significant advantages. Some patients have been shown to benefit from immunotherapy (7). Therefore, we further analyze the role of activin A receptor type-1 (*ACVR1*) in TME.

Programmed cell death protein 1 (PD-1)/programmed cell death ligand-1 (PD-L1), two of the biomarkers of gastric cancer, represent an immune checkpoint molecule involved in immunotherapy (8). PD-1 binds to PD-L1 on the surface of tumor cells and inhibits T-cell activation and proliferation, thus, facilitating immune escape (9). Other biomarkers of gastric cancer, including high microsatellite instability (MSI-H)/defective mismatch repair (dMMR), Epstein-Barr virus (EBV) positivity, and high tumor mutation burden (TMB), have also been strongly connected with immunotherapy (10,11). Immune checkpoint inhibitors for gastric cancer, such as sintilimab, pembrolizumab, and nivolumab, have been used in the clinic to benefit patients to a degree but are not sufficient to respond to individual heterogeneity. Moreover, The Cancer Genome Atlas (TCGA) data have confirmed that there is no single target that controls the progression of all gastric cancers. Therefore, it is critical to identify additional biomarkers to better characterize the relationship between gastric cancer and immunotherapy.

Transforming growth factor β (TGF- β) is a ligand family of structurally related pleiotropic cytokines (12) that control cell proliferation, differentiation, migration, and death, whose overexpression leads to extracellular matrix (ECM) deposition, epithelial–mesenchymal transition (EMT), and cancer-associated fibroblast (CAF) formation. These processes are inextricably linked to fibrotic disease and cancer development (13). *ACVR1*, which has a GS structural domain, a glycine serine–rich juxtamembrane region, and a kinase structural domain, is a type I bone morphogenetic protein (BMP) receptor that encodes TGF- β (14). *ACVR1* binds to other TGF- β receptors that together

Highlight box

Key findings

- The prognostic value and potential immunotherapeutic efficacy of activin A receptor type-1 (*ACVR1*) in treating gastric cancer.

What is known and what is new?

- Existing studies have shown that *ACVR1* is highly expressed in a variety of cancers (including gastric cancer).
- We study the relationship among the high expression of *ACVR1* in gastric cancer, immunity and prognostic value.

What is the implication, and what should change now?

- The results showed that high *ACVR1* expression may promote gastric cancer development by suppressing immunity due to the enrichment of related cancer-promoting stromal pathways, high infiltration of inhibitory immune cells, high immune escape, and low tumor mutation burden, low microsatellite instability, and *TNN* mutation rates. Moreover, the immunotherapy benefit was less in the *ACVR1* high-expression group.
- The limitation of this study was that experimental validation was insufficient, and in vivo and in vitro experiments are needed to fully explore the molecular mechanism underlying the relationship between *ACVR1* and immunity and the effect of drug therapy on gastric cancer.

mediate SMAD signaling and is involved in bone growth, retinal angiogenesis, and hepatic cell surface ferroportin production, as well as biological processes such as sweat gland cell proliferation, germ cell genesis, cardiovascular generation, neuropathic pain, and blockade of glial cell differentiation (15-22). *ACVR1* has been extensively studied in fibrodysplasia ossificans progressive (FOP) and diffuse endogenous pontine glioma (DIPG), which are two rare diseases (23). In addition, its function is being investigated in other cancer types, including multiple myeloma, breast cancer, endometrial cancer, and polycystic ovary syndrome (24-27). Overexpression of *ACVR1* promotes cell proliferation in cancer cells (28), and *ACVR1* activates the Wnt/ β -linker protein signaling pathway in breast cancer (25) and induces an imbalance in the TGF- β 1/BMP-7 pathway in hepatocellular carcinoma (HCC) cells, thereby promoting HCC cell invasion and stemness (29). Increased *ACVR1* expression has been shown to lead to increased and prolonged immune infiltration (30). However, to date, the biological functions of *ACVR1*-related to gastric cancer immunity have not been elucidated.

In this study, through bioinformatics and experimental validation, we provide evidence that *ACVR1* expression is upregulated in gastric cancer and is associated with poor prognosis for patients and that aberrant expression of *ACVR1* is not only associated with the clinical features, prognosis, TME, immune checkpoints, antitumor drug susceptibility, TMB, MSI, and DNA mismatch repair (MMR) but also participates in the promotion of malignant gastric cancer progression. We present this article in accordance with the REMARK reporting checklist (available at <https://jgo.amegroups.com/article/view/10.21037/jgo-23-984/rc>).

Methods

Data collection and process

Downloaded pancancer samples were collected from the UCSC Xena database (xenabrowser.net). RNA sequencing data involving 379 gastric cancer tissues and 34 normal tissues as well as the clinical data of 407 cases were obtained from TCGA (<https://portal.gdc.cancer.gov/>).

Immunohistochemistry (IHC) data were obtained from 212 postoperative patients diagnosed with gastric cancer in Affiliated Hospital of Nantong University from 2010 to 2011. The clinical information including survival status, survival time, date of surgery, tumor-node-metastasis

(TNM) stage, tumor stage, grading, age, and gender. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). The study was approved by the ethics board of Affiliated Hospital of Nantong University (No. 2023-L067) and informed consent was taken from all the patients. After removal of irrelevant data from the above-described set, the data were analyzed with software and visualized using the “limma” and “ggplot2” packages in R software (The Foundation of Statistical Computing), in which TCGA transcriptome data were converted to transcripts per million (TPM) format for subsequent analysis.

Immunohistochemistry (IHC) staining and scoring

The specific steps for postoperative specimens were as follows:

- (I) Treatment of coverslips and slides: slides and coverslips were soaked in an acid vat for 24 h, rinsed with tap water, rinsed at least 3 times with ddH₂O (laboratory-grade water), soaked in 95% alcohol for 24 h, and then baked in an electric thermostatic drying oven and stored for reserve.
- (II) Slicing and baking: tissues were embedded in paraffin, and a tissue slicer was used to make consecutive slices with a thickness of 5 μ m, which were affixed to APES-treated slides. The slices were baked in a 60 °C electric thermostatic oven for 6–8 h. The slices were then dried in a thermostatic oven at 60 °C.
- (III) Deparaffinization and hydration of tissue sections: the tissue sections were soaked in xylene for 15 min, placed in fresh xylene, soaked again for 15 min, and then rinsed in phosphate-buffered saline (PBS) for 5 min 3 times, anhydrous ethanol for 10 min 2 times, 95% ethanol for 5 min 2 times, 80% ethanol for 5 min, and 70% ethanol for 5 min.
- (IV) Antigen repair: first, a prepared ethylenediaminetetraacetic acid (EDTA) repair solution with a PH of 9.0 was boiled in an autoclave, and the slides were placed in the autoclave, covered with stainless steel lid, and heated with a jet at 300 W for 18 min. Running water was used to make the repair solution naturally cool down to room temperature, which was followed by rinsing with PBS for 5 min 3 times.
- (V) Blocking endogenous peroxidase: a circle was drawn with an IHC pen at the edge of the tissue,

100 μL of 3% H_2O_2 was added to each section, and incubation was conducted for 15 min at room temperature, which was followed by a rinsing with PBS for 5 min 3 times.

- (VI) Closure: slides were gently shaken dry of moisture (blotted dry of surrounding liquid), the specimen was placed in an incubation box, and the tissue was closed with PBS containing 5% bovine serum albumin (BSA) and 0.3% Triton X for 2 h at room temperature.
- (VII) Primary antibody: the *ACVR1* antibody was diluted at a ratio of 1:300, the diluted antibody was drooped into the circled tissues, incubated for 1 h at room temperature, and rinsed with PBS for 5 min \times 3 times.
- (VIII) Color development: prepared Diaminobutane (DAB) (1:20) was added dropwise onto the tissue for 5 min. The degree of color development was observed under a microscope, and when the reaction was completed, the tissue was rinsed with tap water for 5 min.
- (IX) Lining staining: hematoxylin restaining conducted the tissue for 10–20 s. Then rinsed the tissue with running water, and hydrochloric acid–ethanol differentiation was performed for 2–3 s, which was followed by a rinse with running water for 5 min.
- (X) Dehydration: 70% ethanol, 80% ethanol, 95% ethanol, and anhydrous ethanol were used to deal with the tissue in order, then xylene was used to deal with it.
- (XI) Sealing: tissue sections were sealed with neutral resin to ensure no air bubbles were produced.

For each section, two experienced pathologists randomly selected ten high-magnification fields of view under the microscope. They counted 100 cells in each field of view to calculate the ratio of the number of positive cells. Scoring was based on the percentage of positive cells, with <5% being 0, 5–25% being 1, 26–50% being 2, 51–75% being 3, and >75% being 4. Scoring was based on staining intensity, with no or light staining being 0, moderate staining being 1, and deep staining being 2. The product of the 2 scores was used as the semiquantitative immunohistochemical score.

Survival analysis

After deleting cases with missing data of downloaded TCGA samples, we calculated the median expression of *ACVR1*.

We then categorized the samples into high- and low-expression groups and examined the survival status, survival time, and the survival curve using Kaplan-Meier analysis. And the IHC data were processed to classify the expression of *ACVR1* into high- and low-expression groups based on the median value and cutoff value of the semiquantitative immunohistochemical score. The data were analyzed using the “survival” and “survminer” packages in R software (R version 4.3.0) for prognostic analysis. A P value of less than 0.05 was considered to indicate a statistically significant difference between the 2 groups.

Comprehensive analysis of clinicopathologic characteristics

The downloaded clinical data of TCGA were used to draw heatmaps of clinical traits via the “complexHeatmap” R package; meanwhile, the correlation between clinical characteristics and *ACVR1* was plotted using the “ggpubr” R package. The “ggpubr” R package was also used with IHC data to plot the correlation of clinical features with *ACVR1*. TCGA data were then analyzed using univariate Cox and multifactorial Cox to explore the relationship between clinical traits and *ACVR1* expression, while multifactorial Cox was used to examine whether *ACVR1* could be used as an independent prognostic indicator for gastric cancer. A P value less than 0.05 was considered to indicate a statistically significant difference between two groups. Based on the results of the prognostic analysis, column line graphs were constructed to assess the survival rates at 1, 3, and 5 years, after which the reliability of the results of the column line graphs was assessed using calibration curves and receiver operating characteristic (ROC) curves.

Biofunctional analysis

ACVR1 coexpressed genes were analyzed with the “ggExtra” R package, and a correlation network diagram was plotted to show the correlation between genes. Additionally, a correlation heat map was plotted using the “corrplot” R package. Gene set enrichment analysis (GSEA) of Kyoto Encyclopedia of Genes and Genomes (KEGG) was conducted for the coexpressed genes, and the results were plotted as bar charts via the “ggplot2” and “clusterProfiler” R packages, and the petal plots in “ggraph” R package were used to show the specific genes related to the pathways. Single-sample GSEA (ssGSEA) was performed using the “Hallmark” gene set to analyze the enrichment of relevant pathways in the *ACVR1* high- and low-expression

groups, and the results were visualized using the “GSVA”, “GSEABase”, “GSVA”, “GSEABase”, “GSVA”, and “GSEABase” R packages. The results were visualized using “GSVA” and “GSEABase” R packages, and pathways with a P value less than 0.05 were considered to be significantly enriched.

TME analysis

ACVR1-associated immune cells, stromal cell infiltration, and immune function were analyzed by the ESTIMATE algorithm, the CIBERSORT algorithm, the ssGSEA algorithm and Spearman correlation analysis. The results were visualized using R package “ggplot2”. Single-cell data from the Tumor Immune Single-cell Hub 2 (TISCH2) online website were used to analyze immune cell infiltration. Immunotherapy effects were predicted using the tumor immune dysfunction and exclusion (TIDE) score and the IMvigor210 immune cohort. A P value <0.05 was considered statistically significant.

Immune checkpoint and antitumor drug analysis

Differences between *ACVR1* and common immune checkpoints were determined using the “ggplot2” package. Immune checkpoint correlation heatmaps and correlation plots of 4 specific common immune checkpoints with *ACVR1* were drawn. Online data from The Cancer Imaging Archive (TCIA) were then scored to predict the outcome of anti-immune checkpoint therapy, and violin plots were created using the “ggpubr” R package. Subsequently, we analyzed the correlation between 4 common antitumor drugs and *ACVR1*. Differences in the 50% reduction in growth concentration (IC_{50}) of chemotherapeutic agents between the *ACVR1* high- and low-expression groups were estimated using the “pRRophetic” package. P values less than 0.05 were considered to be statistically significant.

TMB analysis

TMB, which reflects the number of cancer mutations, is defined as the number of somatic mutations per megabase (31). $TMB \geq 10$ mutations per megabase (mut/Mb) is considered to be a high TMB (TMB-H). TMB mutations are processed into neoantigens and presented to T cells via major histocompatibility complex (MHC) proteins. A higher TMB leads to more neoantigens, increases the chances of T-cell recognition, and is clinically associated

with better immune checkpoint blockade treatment outcomes (32).

MSI and MMR analysis

MSI can be categorized into 3 states: MSI-H, low microsatellite instability (MSI-L), and microsatellite stable (MSS), while MMR can be categorized into dMMR and proficient MMR (pMMR). In addition, dMMR is equivalent to MSI-H, while pMMR is equivalent to MSI-L or MSS. Human mismatch repair genes (MMR genes) are transcribed and translated to convey the corresponding mismatch repair proteins. A malfunction in the expression of the MMR proteins (including MLH1, MSH2, MSH6, and PMS2) can result in base mismatches during DNA replication, leading to the loss of repair, with dMMR leading to MSI (33).

Statistical analysis

All statistical analyses and graphs were completed with R version 4.3.0. The Wilcoxon rank-sum test and *t*-test were used to compare numerical differences between *ACVR1* high- and low-expression groups, while the log-rank test was conducted to assess differences in survival. Kaplan-Meier survival analysis and the log-rank test were used to evaluate survival distributions. Univariate Cox and multivariate Cox regression analyses were performed to assess the hazard ratios (HRs) and 95% confidence intervals (CIs) for different clinical characteristics and to identify independent prognostic factors. $P < 0.05$ was considered statistically significant.

Results

ACVR1 was highly expressed in multiple cancers and showed differential expression in gastric cancer

The Xena database showed that *ACVR1* expression was upregulated in cholangio carcinoma (CHOL), colon adenocarcinoma (COAD), esophageal carcinoma (ESCA), head and neck squamous cell carcinoma (HNSC), kidney renal papillary cell carcinoma (KIRP), lung adenocarcinoma (LUAD), lung squamous cell carcinoma (LUSC), pheochromocytoma and paraganglioma (PCPG), stomach adenocarcinoma (STAD), and thyroid carcinoma (THCA) (Figure 1A). In gastric cancer, TMB and MSI are closely related to the high *ACVR1* expression (Figure 1B,1C).

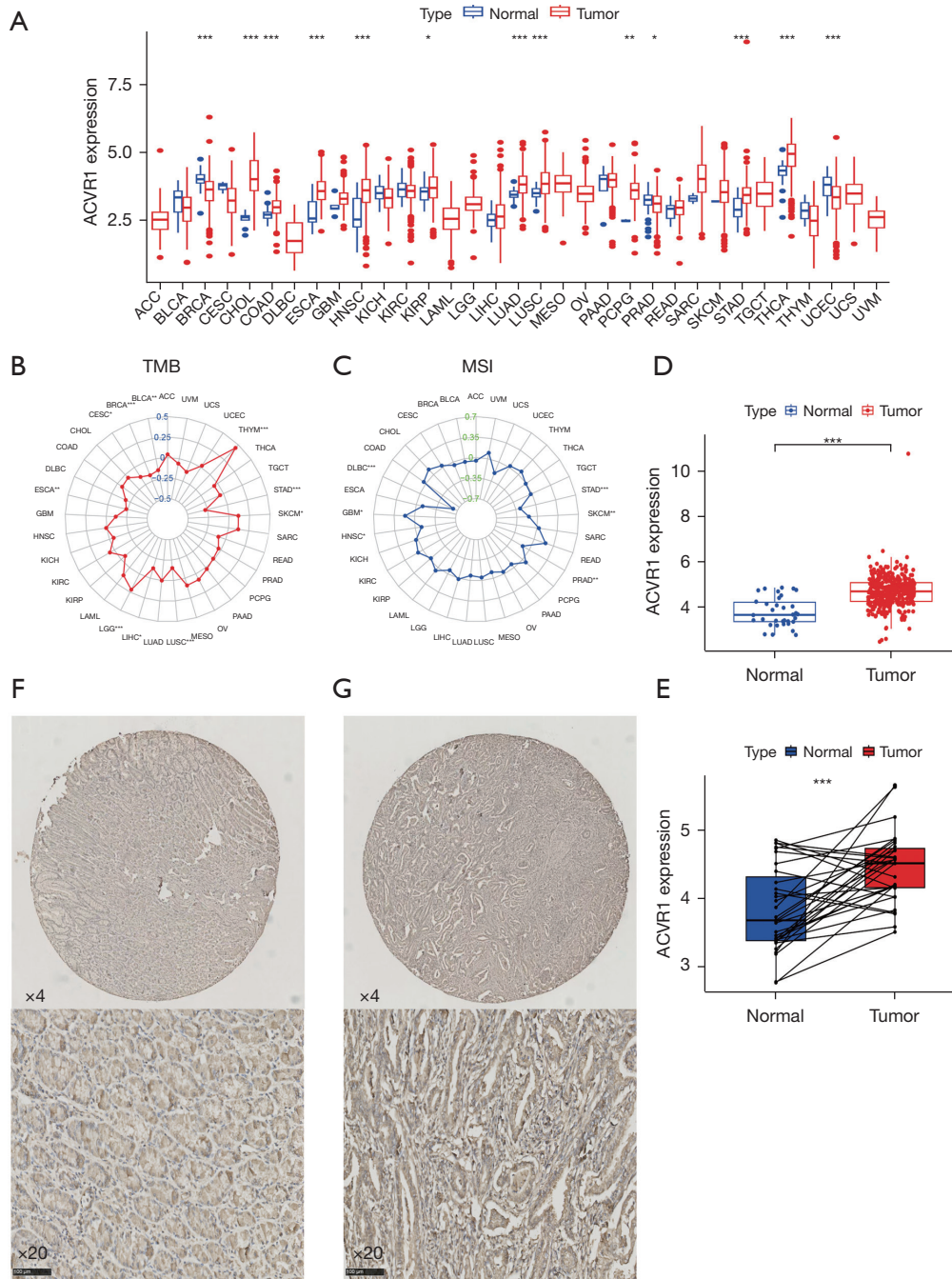


Figure 1 Differential expression of *ACVR1* in cancers. (A) *ACVR1* expression in different tumor tissues; (B) correlation of *ACVR1* with TMB in various tumor tissues; (C) correlation of *ACVR1* with MSI in different tumor tissues; (D) *ACVR1* expression in unpaired samples of normal gastric and gastric cancer tissues in TCGA database; (E) *ACVR1* expression in paired samples in normal gastric tissues and gastric cancer tissues in TCGA database; (F-G) *ACVR1* IHC taken at 4× (top) and 20× (bottom) magnification of the normal stomach with gastric cancer tissues. Scale markers are in the lower left corner of the images. *, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$. *ACVR1*, activin A receptor type-1; TMB, tumor mutation burden; MSI, microsatellite instability; TCGA, The Cancer Genome Atlas; IHC, immunohistochemistry.

The TCGA database showed that in unpaired difference analysis, the expression of *ACVR1* was higher in gastric cancer than in normal paracancerous tissues (Figure 1D). In paired differential analysis, the expression level of *ACVR1* was also high in the gastric cancer (Figure 1E). The IHC data indicated that the expression of *ACVR1* was higher in gastric cancer tissues than in normal tissues (Figure 1F,1G). Above analysis results indicate that *ACVR1* was higher in gastric cancer than in normal paracancerous tissues. It may be a procancer factor related to the poor prognosis.

Correlation of ACVR1 with survival prognosis and clinicopathologic features of gastric cancer

The results of previous studies and studies suggest that the high expression of *ACVR1* may indicate a poor prognosis in gastric cancer (34,35). In the subsequent survival curves, the difference between high and low *ACVR1* expression was noticeable, and the prognosis of the high-expression group was poorer than that of the low-expression group (Figure 2A).

Similarly, the poor prognosis of the *ACVR1* high-expression group was again verified by the results of IHC data in Affiliated Hospital of Nantong University (Figure 2B). According to the TCGA data, we found that patients in the *ACVR1* high-expression group had a shorter progression-free survival (PFS) time (Figure 2C). We again used the cutoff value of the *ACVR1* expression of TCGA data to differentiate between the high- and low-expression groups. The low-expression group has a poorer predicted prognosis (Figure 2D). The chi-squared test was used to assess the clinicopathological characteristics of TCGA data. The results showed that *ACVR1* expression was significantly correlated with tumor grade and tumor stage (Figure 3A). According to Spearman analysis, there was a significant difference in the expression of *ACVR1* in early (stage I) and advanced (stage III) gastric cancer (Figure 3B). The expression of *ACVR1* gradually increased with the increase of tumor T stage (Figure 3C). In the IHC data, the expression of *ACVR1* was increased in the advanced stage (stage III) of gastric cancer (Figure 3D). The high expression of *ACVR1* was in patients with poorly differentiated gastric cancer (Figure 3E).

Construction of column line graphs

To further investigate the relationship between *ACVR1* and overall survival (OS), we performed univariate Cox

and multivariate Cox analyses on TCGA data. Univariate Cox analysis showed that a high expression of *ACVR1*, T₂₋₄N₁₋₃M₁ stage, stage III, IV and old age (>65 years) were all associated with OS (P<0.05) (Figure 4A). Multivariate Cox analysis showed that the high expression of *ACVR1* and old age (>65 years) operated as independent prognostic factors (P<0.05) (Figure 4B). Column line plot was constructed for predicting the 1-, 3-, and 5-year survival rates based on TNM stage, pathologic grade, gender, stage, age, and *ACVR1* expression, with the results being 0.894, 0.699, and 0.642, respectively (Figure 4C). The calibration curve showed that the actual OS values were more consistent with the predicted OS consequences (Figure 4D). The area under the ROC curve for 1-, 3-, and 5-year survival rates of the forecast OS column line graphs were 0.527, 0.583, and 0.627, respectively, which, taken together, indicated a more satisfactory prediction (Figure 4E).

Coexpression analysis and enrichment analysis of ACVR1

We performed a coexpression analysis of *ACVR1*. Figure 5A shows the network diagram of the top 5 genes most positively and negatively correlated with *ACVR1*, respectively. Figure 5B shows the correlation heatmap of the top 10 genes most positively and negatively associated with *ACVR1*. Both images indicate that the extracted coexpressed genes were significantly correlated with *ACVR1*. To further clarify the underlying mechanisms, we used GSEA of the coexpressed genes of *ACVR1* in the KEGG dataset. As can be seen in Figure 5C, these coexpressed genes were significantly correlated with *ACVR1* in the areas of leukocyte transendothelial migration, neutrophil extracellular trap formation, PD-L1 expression and PD-1 checkpoint pathway in cancer, B-cell receptor signaling pathway, T helper (Th) 17 cell differentiation, Th1 and Th2 cell differentiation, NK cell-mediated cytotoxicity, T-cell receptor signaling pathway, chemokine signaling pathway, platelet activation, and Toll-like receptor signaling pathway. There was significant enrichment in these immune-related pathways, which may contribute to the development of gastric cancer via the regulation of immune-related functions or immune cells. In addition to the above pathways, these genes were also enriched in nuclear factor (NF)- κ B signaling pathway, Hedgehog signaling pathway, focal adhesion, ECM-receptor interaction, Hippo signaling pathway, mTOR signaling pathway, Wnt signaling pathway, TGF- β signaling pathway, JAK-STAT signaling pathway,

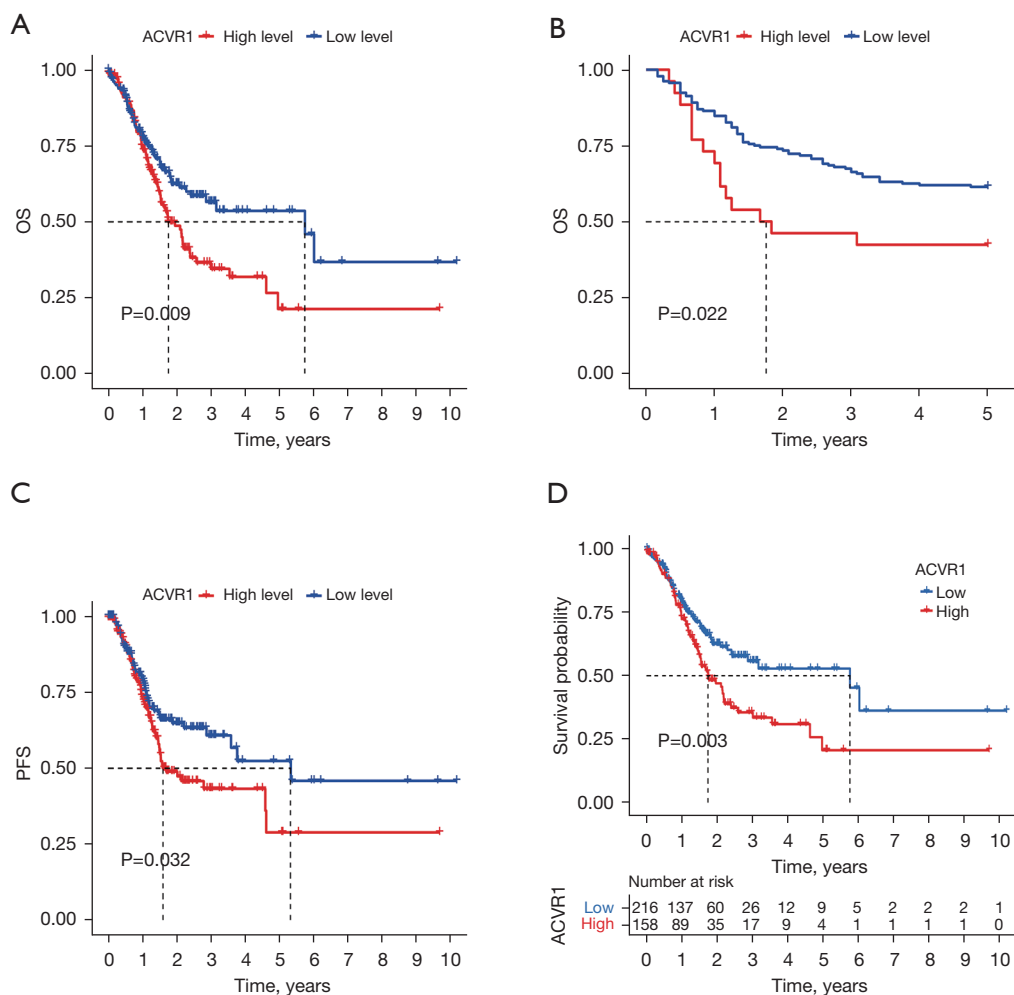


Figure 2 Prognosis of *ACVR1*. (A) OS of *ACVR1* in TCGA divided into high- and low-expression groups by median value; (B) OS of *ACVR1* in Affiliated Hospital of Nantong University cohort; (C) PFS of *ACVR1* in TCGA divided into high- and low-expression groups by cutoff value; (D) survival probability of *ACVR1* in TCGA cohort. *ACVR1*, activin A receptor type-1; OS, overall survival; TCGA, The Cancer Genome Atlas; PFS, progression-free survival.

and cytokine–cytokine receptor interaction, among others, which are involved in cell proliferation, differentiation, apoptosis, migration, and death. *ACVR1* is also enriched in metabolism-related pathways such as animal autophagy, phagosome, and lysosome, as well as hypoxia-inducible factor 1 (HIF-1) signaling pathway, which is a hypoxia-related pathway. *Figure 5D* shows all the genes enriched in the T-cell-related pathways.

Next, we performed ssGSEA using the “Hallmark” gene set, and as shown in *Figure 5E*, the high expression group was associated with allograft rejection, angiogenesis, EMT, Hedgehog transition, Hedgehog signaling, hypoxia, IL2-STAT5 signaling, IL6-JAK-STAT3 signaling,

inflammatory response, interferon α (IFN- α) response, IFN- γ response, KRAS signaling down-regulation, KRAS signaling up-regulation, mitotic spindle, Notch signaling, P53 pathway, PIK3-AKT-mTOR signaling, TGF- β signaling, TNF- α signaling via NF- κ B, ultraviolet response downregulation, ultraviolet response upregulation, Wnt- β catenin signaling, and apoptosis. These are all significantly associated with these stroma-associated pathways of oncogenic activity. Meanwhile, the low-expression group was enriched considerably in DNA repair, E2F targets, MYC target V1, G2M checkpoint, and oxidative phosphorylation, which are cell cycle event-related pathways.

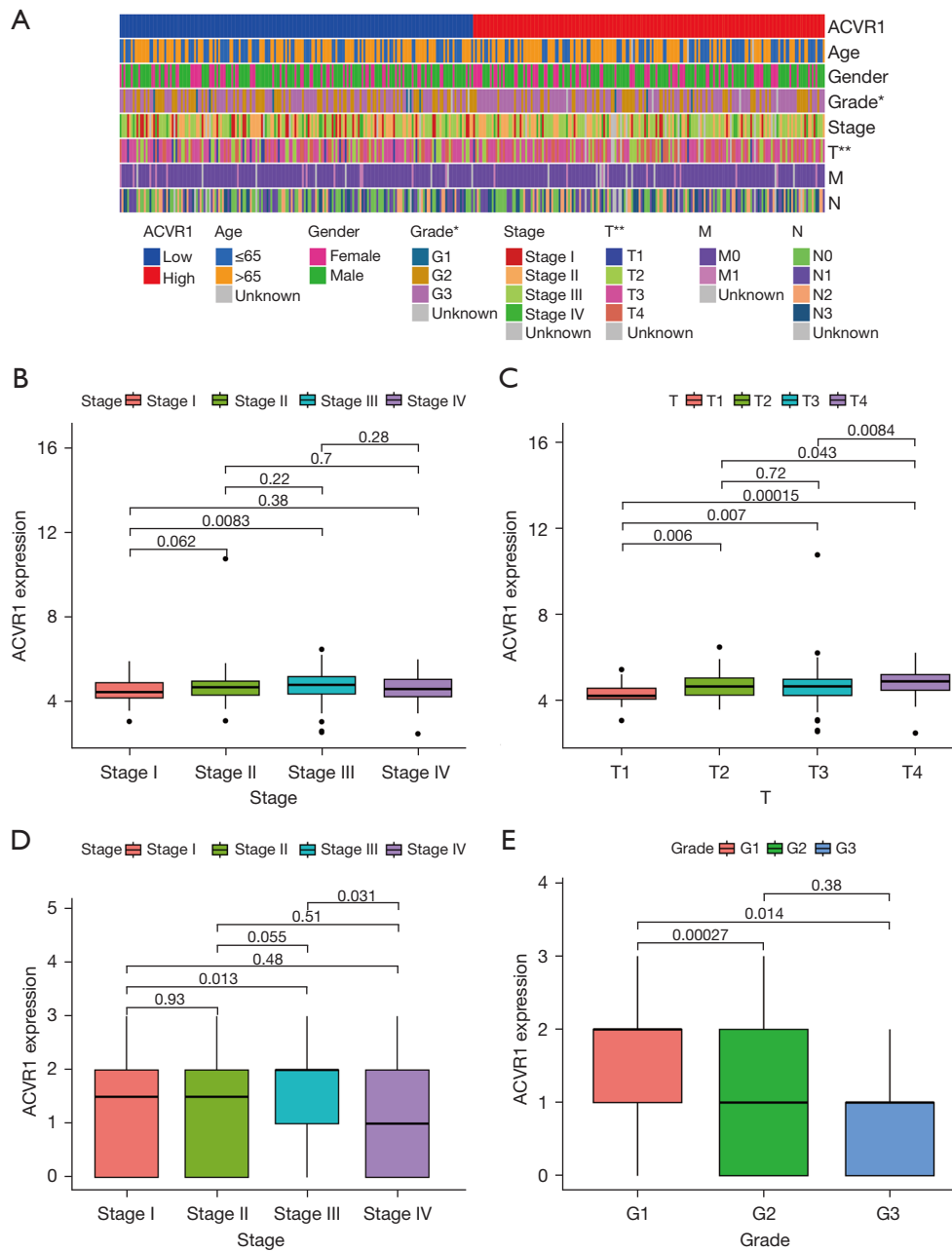


Figure 3 Clinicopathologic relevance of *ACVR1*. (A) Heatmap demonstrating the correlation between high and low *ACVR1* expression and clinicopathologic features of TCGA cohort; (B) correlation of *ACVR1* expression with tumor stage the TCGA cohort; (C) correlation between *ACVR1* expression and T stage of TCGA cohort; (D) correlation of *ACVR1* expression with tumor staging in the Affiliated Hospital of Nantong University cohort; (E) correlation of *ACVR1* expression with pathological grade in the Affiliated Hospital of Nantong University cohort. *, $P < 0.05$; **, $P < 0.01$. *ACVR1*, activin A receptor type-1; TCGA, The Cancer Genome Atlas.

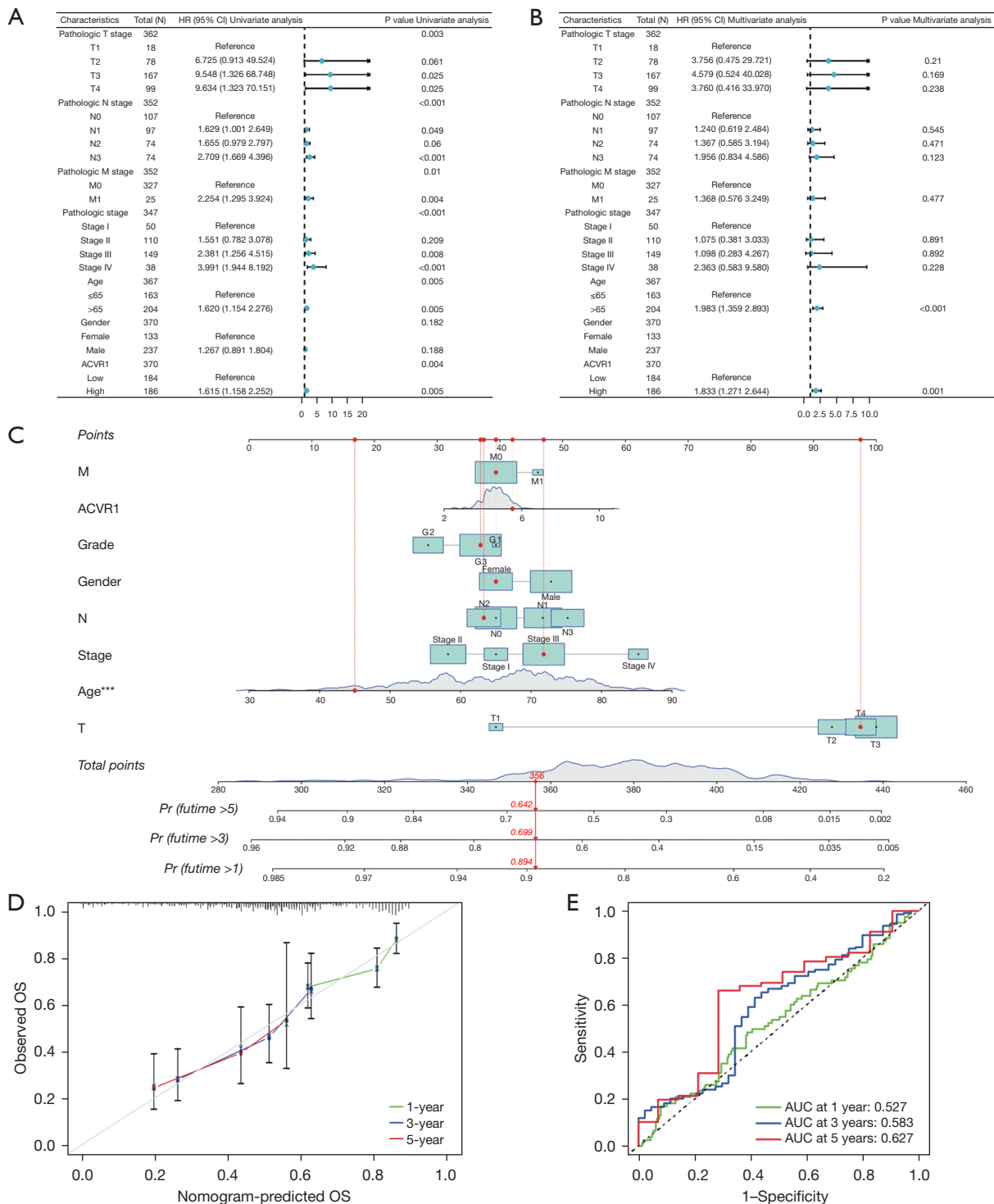


Figure 4 Establishment of column charts and assessment in TCGA cohort. (A,B) One-way Cox and multifactorial Cox regression analyses were performed; (C) construction of a prognostic column-line diagram integrating high and low *ACVR1* expression and multifactorial Cox regression analyses (including gender, TNM stage, tumor stage, age, and pathological grading) for predicting clinical outcomes in patients with gastric cancer; (D) calibration curves of prognostic column plots were used to predict 1-, 3-, and 5-year survival; (E) ROC curves for predicting 1-, 3-, and 5-year survival rates. ***, $P < 0.001$. HR, hazard ratio; CI, confidence interval; *ACVR1*, activin A receptor type-1; OS, overall survival; AUC, area under the curve; TCGA, The Cancer Genome Atlas; ROC, receiver operating characteristic.

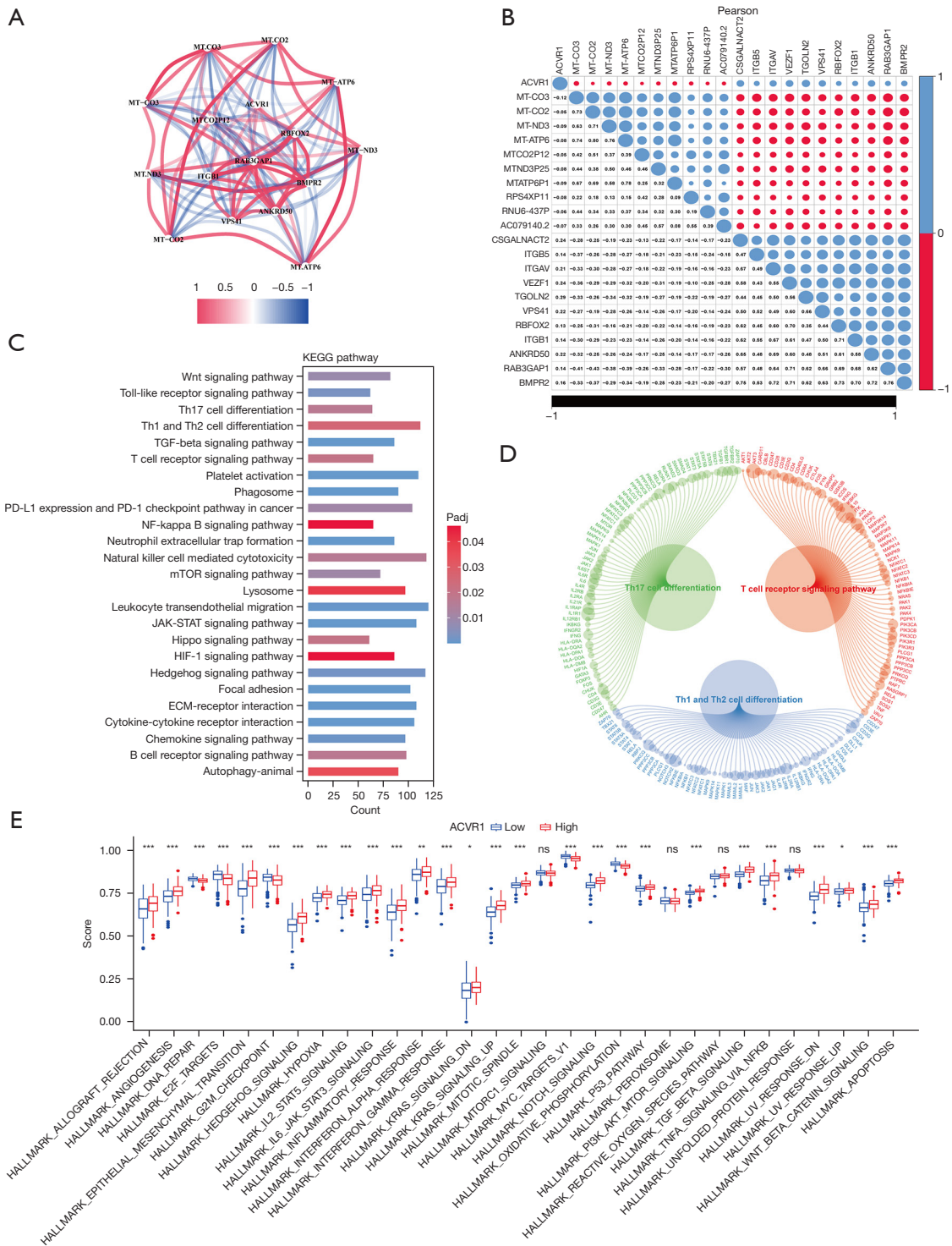


Figure 5 Coexpression and functional enrichment analysis of *ACVR1*. (A) Network diagram of the top 5 coexpressed genes positively and negatively associated with *ACVR1*; (B) heatmap of the top 10 coexpressed genes positively and negatively associated with *ACVR1*; (C) KEGG enrichment analysis of coexpressed genes; (D) all the genes related to the T-cell pathway in the pathway of coexpressed genes; (E) pathways with ssGSEA enrichment associated with high and low *ACVR1* expression. *, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$. *ACVR1*, activin A receptor type-1; KEGG, Kyoto Encyclopedia of Genes and Genomes; ssGSEA, single-sample gene set enrichment analysis.

Analysis of the correlation between ACVR1 and TME

Based on TCGA database, we used the ESTIMATE algorithm to produce scores for immune cells, stromal cells, and the availability of immune and stromal cells in the TME. The results showed that the StromalScore, ImmuneScore, and ESTIMATEScore of the high *ACVR1* expression group were higher than those of the low-expression group (Figure 6A).

We used the CIBERSORT algorithm to assess the proportion of immune cell infiltration in the *ACVR1* high- and low-expression groups. The results showed that the scores of memory B cells and plasma cells were higher in the *ACVR1* low-expression group than in the high-expression group. While the scores of resting memory CD4 T cells, M1 macrophages, M2 macrophages, and resting mast cells were higher in the *ACVR1* high-expression group than in the low-expression group (Figure 6B).

In addition, we used the ssGSEA algorithm to assess immune cell infiltration in samples of the *ACVR1* high- and low-expression groups. We found that B cells, CD8⁺ T cells, dendritic cells (DCs), interstitial DCs (iDCs), macrophages, mast cells, neutrophils, plasmacytoid DCs (pDCs), Th cells, tumor-infiltrating lymphocytes (TILs), and T regulatory cells (Tregs) had a higher degree of infiltration in the *ACVR1* high-expression group than in the low-expression group (Figure 6C). With the upregulation of *ACVR1* expression, the immune functions of antigen-presenting cell (APC) coinhibition, APC costimulation, CCR, check-point, human leukocyte antigen (HLA), parainflammation, T-cell coinhibition, T-cell costimulation, type I IFN response, AND type II IFN response were enhanced (Figure 6D). We subsequently investigated the relationship between infiltrating immune cells and *ACVR1* expression between the high- and low-expression groups using Spearman correlation analysis. As shown in Figure 6E, high *ACVR1* expression was positively correlated with resting mast cells, M2 macrophages, resting memory CD4 T cells, resting DCs, monocytes, M1 macrophages, and memory B cells. However, high *ACVR1* expression was negatively correlated with M0 macrophages, follicular Th cells, activated mast cells, memory B cells, and plasma cells.

We also performed TIDE immune escape scoring and assessed complete response (CR)/partial response (PR) and stable disease (SD)/progressive disease (PD) in patients with high *ACVR1* expression using the IMvigor210 immune cohort. We used TIDE to identify two mechanisms of tumor immune escape: induction of T-cell dysfunction in

tumors with high abundance of cytotoxic T lymphocytes (CTLs) and prevention of T-cell infiltration. It was found that immune escape was more likely to occur in the *ACVR1* high-expression group (Figure 6F). Moreover, in the IMvigor210 immunization cohort, the prognosis was worse in the high-expression group than in the low-expression group (Figure 6G). Furthermore, *ACVR1* expression was lower in the CR/PR group (Figure 6H), and in the *ACVR1* low-expression group, the ROC curve (Figure 6I) indicated that the accuracy of predicting immunotherapy in this IMvigor210 immunization cohort sample was relatively satisfactory. Taken together, these results suggest that high *ACVR1* expression may lead to increased infiltration of immunosuppressive cells over antitumor immune cells in the TME, thus promoting gastric cancer progression and resistance to immunotherapy.

Validation of ACVR1 immune infiltration at the single-cell level

We performed single-cell RNA sequencing analysis of *ACVR1* based on the gastric cancer database in the TISCH2 online website to explore the expression and distribution of *ACVR1* in TME. Two datasets, GSE134520 and GSE167297, were analyzed, revealing that *ACVR1* was significantly overexpressed in CD8 T cells, plasma cells, DCs, and mast cells (Figure 7A). Based on the results of GSE167297 data analysis, it was found that *ACVR1* was overexpressed and distributed in plasma cells, mast cells, CD8 T cells, DCs, malignant cells, gland mucous cells, and pit mucous cells (Figure 7B,7C). *ACVR1* was also overexpressed and distributed in B cells, plasma cells, mast cells, CD8 T cells, DCs, monophages and macrophages, endothelial cells, and epithelial cells (Figure 7D,7E).

Analysis of ACVR1 with immune checkpoints and antitumor drugs

Immune checkpoint blockers were associated higher survival rates compared to conventional antitumor drugs. Combining immunotherapy and chemotherapy can improve the immune status in TME and thus enhance the antitumor immune response (36). To further predict the response to immunotherapy in high *ACVR1* expression patients with gastric cancer, we analyzed the *ACVR1*-related immune checkpoints, which showed that most of the immune checkpoints were more highly expressed in

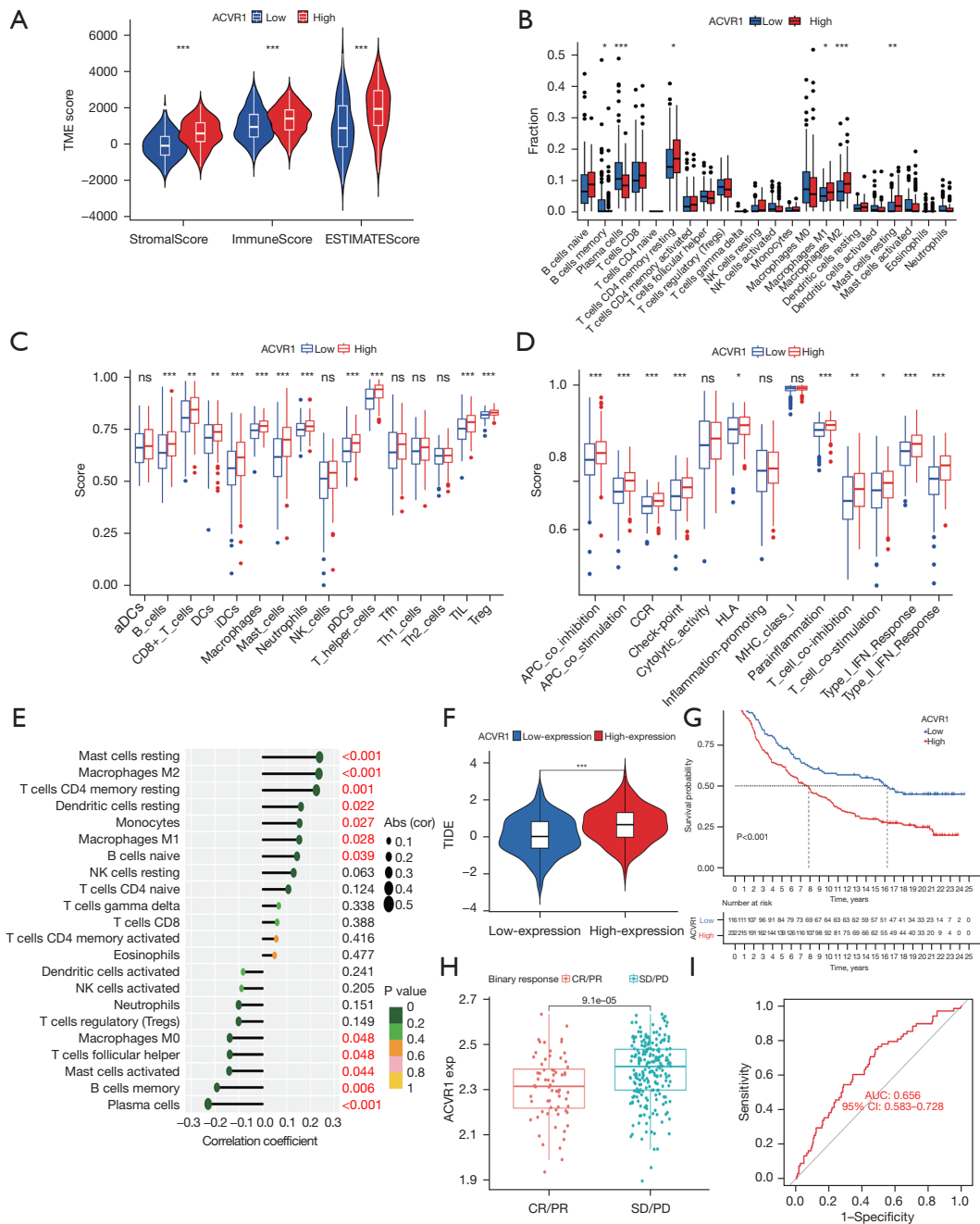


Figure 6 Correlation of *ACVR1* with immune infiltration. (A) Relationship between StromalScore, ImmuneScore and ESTIMATEScore with high and low *ACVR1* expression; (B) the CIBERSORT algorithm assessed the proportion of immune cell infiltration in the high and low *ACVR1* expression groups; (C) the ssGSEA algorithm was used to evaluate immune cell infiltration in samples from the *ACVR1* high and low expression groups; (D) the ssGSEA algorithm assesses the immune function in samples from the *ACVR1* high- and low-expression group; (E) Lollipop plot demonstrating the correlation of immune cells with *ACVR1* expression; (F) TIDE score of *ACVR1*; (G) analysis of the IMvigor210 immune cohort to assess the prognosis of patients with high *ACVR1* expression; (H) analysis of the IMvigor210 immune cohort to assess CR/PR and SD/PD in patients with high *ACVR1* expression; (I) ROC plot of IMvigor210 immune cohort in evaluating the accuracy of the cohort. *, P<0.05; **, P<0.01; ***, P<0.001. *ACVR1*, activin A receptor type-1; TME, tumor microenvironment; ssGSEA, single-sample gene set enrichment analysis; TIDE, tumor immune dysfunction and exclusion; CR, complete response; PR, partial response; SD, stable disease; PD, progressive disease; ROC, receiver operating characteristic.

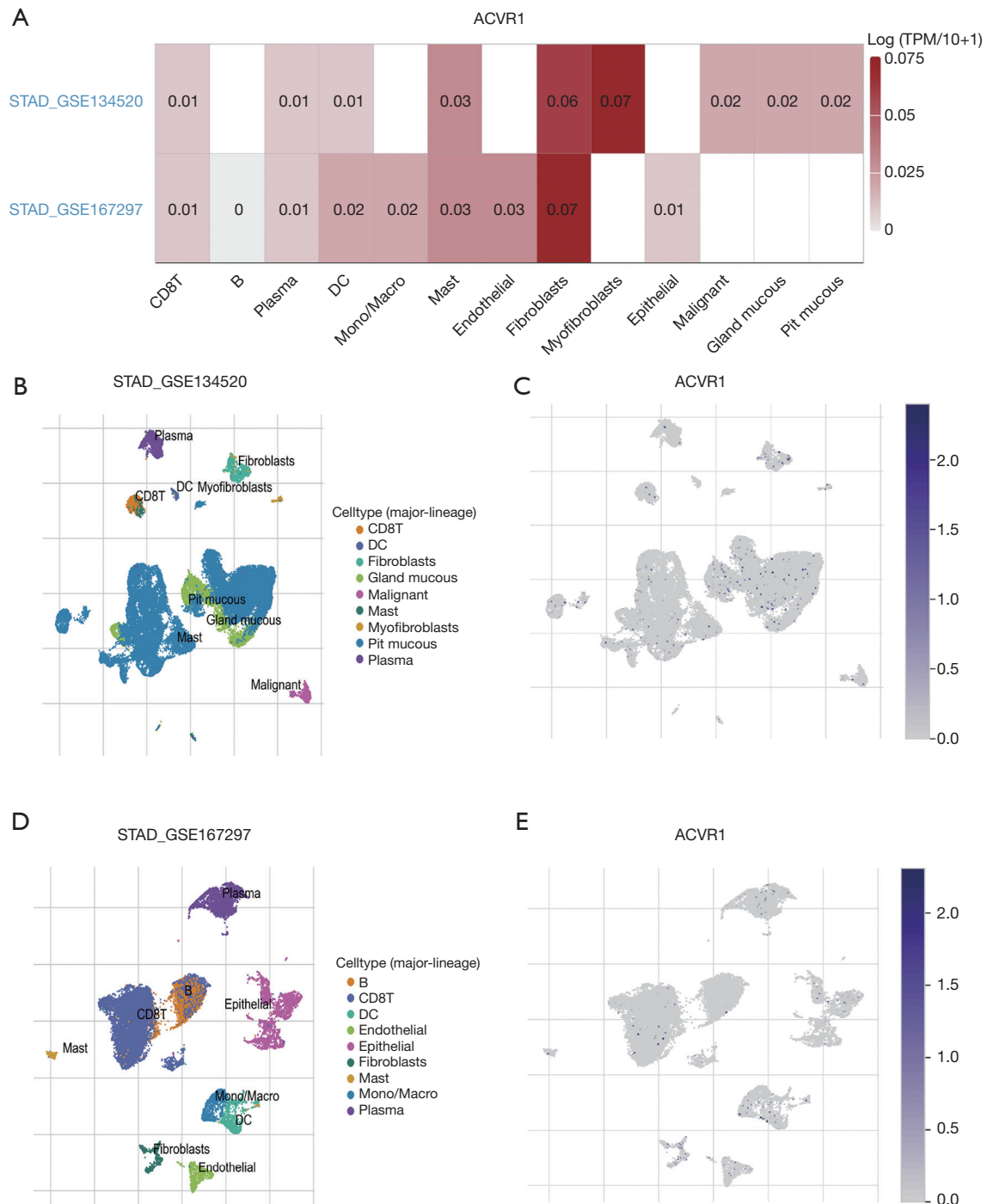


Figure 7 Relationship between *ACVR1* and immune cell infiltration in single cells. (A) Immune cell infiltration of *ACVR1* in 2 sets of gastric cancer data, GSE134520 and GSE167297; (B) distribution of immune cells in the GSE134520 gastric cancer data; (C) infiltration of immune cells associated with *ACVR1* in the GSE134520 gastric cancer data; (D) distribution map of immune cells in the GSE167297 gastric cancer data; (E) infiltration of immune cells associated with *ACVR1* in the GSE167297 gastric cancer data. *ACVR1*, activin A receptor type-1.

the *ACVR1* high-expression group (Figure 8A). According to Spearman correlation analysis, it was concluded that most of the immune checkpoints were positively regulated with *ACVR1*. Therefore, patients with high *ACVR1* expression experienced better therapeutic efficacy when treated with immune checkpoint inhibitors (Figure 8B). We analyzed PD-1, PD-L1, PD-L2, and CTLA4 among the common immune checkpoints. They were more significantly expressed in the *ACVR1*-high expression group (Figure 8C-8F). We predicted the therapeutic effects of two immune checkpoint inhibitors, PD-1 and CTLA4. The results showed that the blocking effect of these two immune checkpoints was poor in patients in the *ACVR1*-high expression group both for the negative or positive expression of PD-1 and CTLA4 (Figure 8G-8J). However, the blocking effect of the other immune checkpoint inhibitors still needs to be investigated. In addition, we also analyzed *ACVR1*-related drug treatments, and the results showed that the IC₅₀ of four common antitumor drugs, paclitaxel, erlotinib, lapatinib, and sunitinib, were significantly reduced in the *ACVR1*-high expression group (Figure 8K-8N). In conclusion, although immunotherapy alone is ineffective, combining it with other antitumor drugs may produce promising therapeutic effects. Further investigation into *ACVR1* may reveal new strategies for the treatment of gastric cancer.

Analysis of the correlation between ACVR1 and TMB

Patients with higher TMB have more prolonged survival after immune checkpoint inhibitor (PD-1/PD-L1) treatment (37). We analyzed the TMB and mutation frequency of *ACVR1*. The results showed that patients in the *ACVR1* high-expression group had lower TMB levels than the low-expression group (Figure 9A). The expression of *ACVR1* was negatively correlated with TMB (Figure 9B). Kaplan-Meier analysis showed that patients with gastric cancer with low TMB levels had a shorter survival period with high TMB (Figure 9C). According to stratified survival analysis, patients with gastric cancer in the high TMB and *ACVR1* low-expression group had the best survival prognosis, while patients with gastric cancer in the low TMB and *ACVR1* high-expression group had the worst survival prognosis (Figure 9D). Among the somatic mutations of *ACVR1*, the top 20 mutated genes in the high and low *ACVR1* expression groups were the same. However, the frequency of gene mutations in the *ACVR1* low-expression group was higher than that in the *ACVR1* high-

expression group, with the gene with the highest mutation frequency being *TTN*, followed by *TP53* and *MUC16* (Figure 9E,9F).

Correlation analysis of ACVR1 with MSI and dMMR

In the comparison of MSS, MSI-L, and MSI-H in the high and low expression of *ACVR1*, the results showed that the percentage of both MSS and MSI-L was higher in the *ACVR1* high-expression group (Figure 10A). Indeed, the differences between MSS, MSI-L, and MSI-H were significant, indicating that the MSI decreased as the expression of *ACVR1* increased (Figure 10B). We used Spearman correlation analysis and found that the expression of all 4 MMR proteins (MLH1, MSH2, MSH6, and PMS2) was increased in the *ACVR1* high-expression group (Figure 10C-10F).

Discussion

Despite a large number of studies being dedicated to eradicating gastric cancer, it remains largely uncured. Some patients receiving improved clinical outcomes and others' outcomes being unsatisfactory. Therefore, actively exploring biomarkers may help to gain insight into gastric cancer's molecular mechanisms (38). In this study, the potential biological function of *ACVR1* as a novel biomarker for gastric cancer was explored. *ACVR1* is highly expressed in various cancers, and this paper focuses on the relationship between *ACVR1* and gastric cancer. Compared with normal tissues, gastric cancer tissue showed upregulated *ACVR1* expression as demonstrated by the biosignature analysis and IHC data. The high expression of *ACVR1* predicted a poor prognosis of gastric cancer correlated with the TME.

ACVR1 was significantly associated with the clinicopathological features of tumor stage and tumor size in gastric cancer. In addition, Kaplan-Meier survival analysis (OS and PFS) of TCGA samples showed that patients with high *ACVR1* expression had a worse prognosis in gastric cancer. IHC data in Affiliated Hospital of Nantong University confirmed that patients with high expression of *ACVR1* had lower OS than low *ACVR1* expression. Univariate Cox, multivariate Cox analysis, and column line graph modeling indicated that high *ACVR1* expression could be an independent factor for poor prognosis in gastric cancer. It has been demonstrated that upregulation of *ACVR1* expression can promote angiogenesis and lymph node metastasis in gastric cancer (34). Previous *in vivo* and

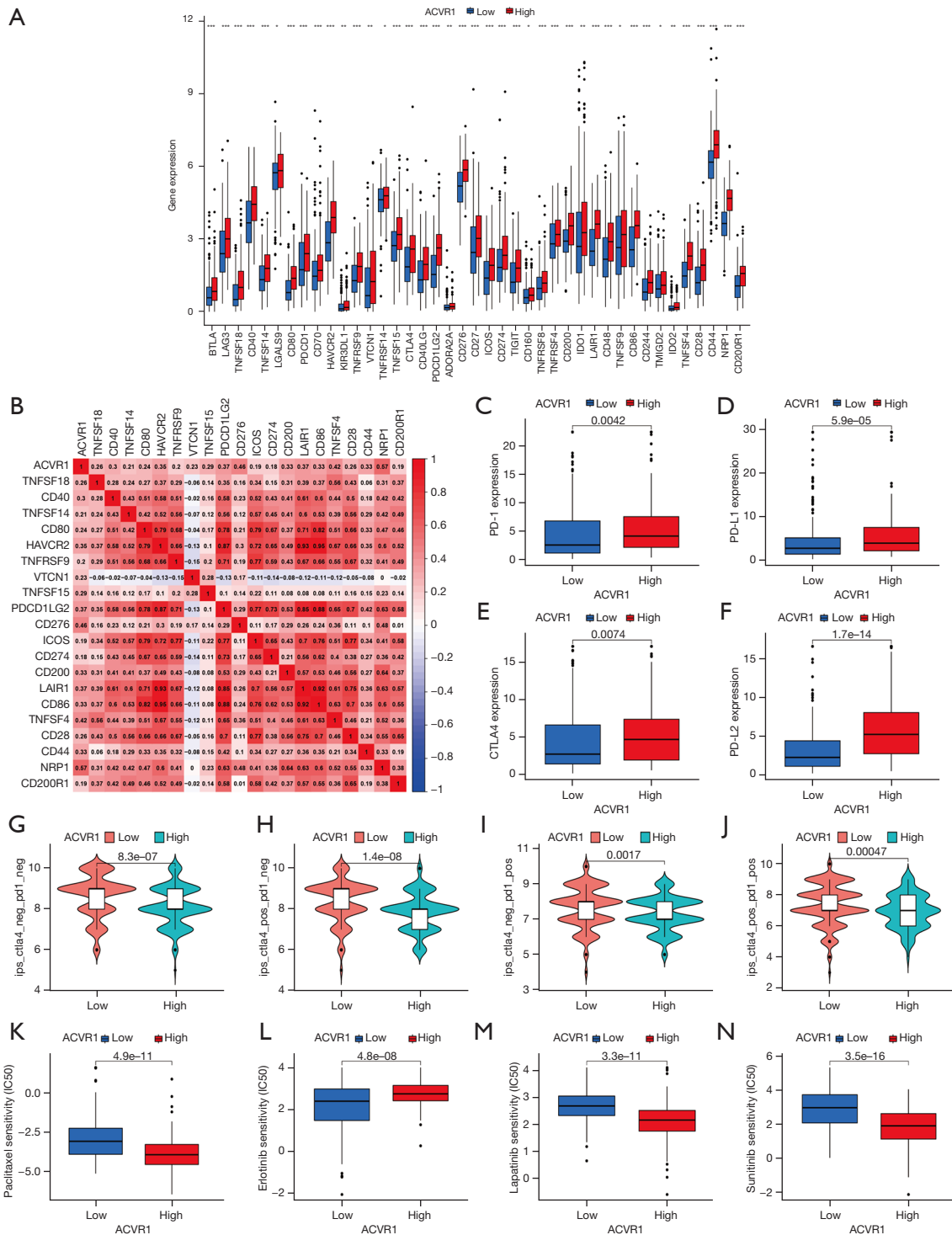


Figure 8 Analysis of *ACVR1* correlation with immune checkpoints and antitumor drugs. (A) Differential analysis between high and low *ACVR1* expression and common immune checkpoints; (B) heatmap of correlation analysis between high and low *ACVR1* expression and common immune checkpoints; (C-F) correlation analysis of *ACVR1* with PD-1, PD-L1, PD-L2, and CTLA4 immune checkpoints, respectively; (G-J) *ACVR1* and anti-PD-1, CTLA4 positive or negative immunotherapy results; (K-N) sensitivity analysis of high and low *ACVR1* expression with 4 common antitumor drugs (paclitaxel, erlotinib, lapatinib, sunitinib). *, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$. *ACVR1*, activin A receptor type-1; PD-1, programmed cell death protein 1; PD-L1, programmed cell death ligand 1; CTLA4, cytotoxic T lymphocyte-associated protein-4.

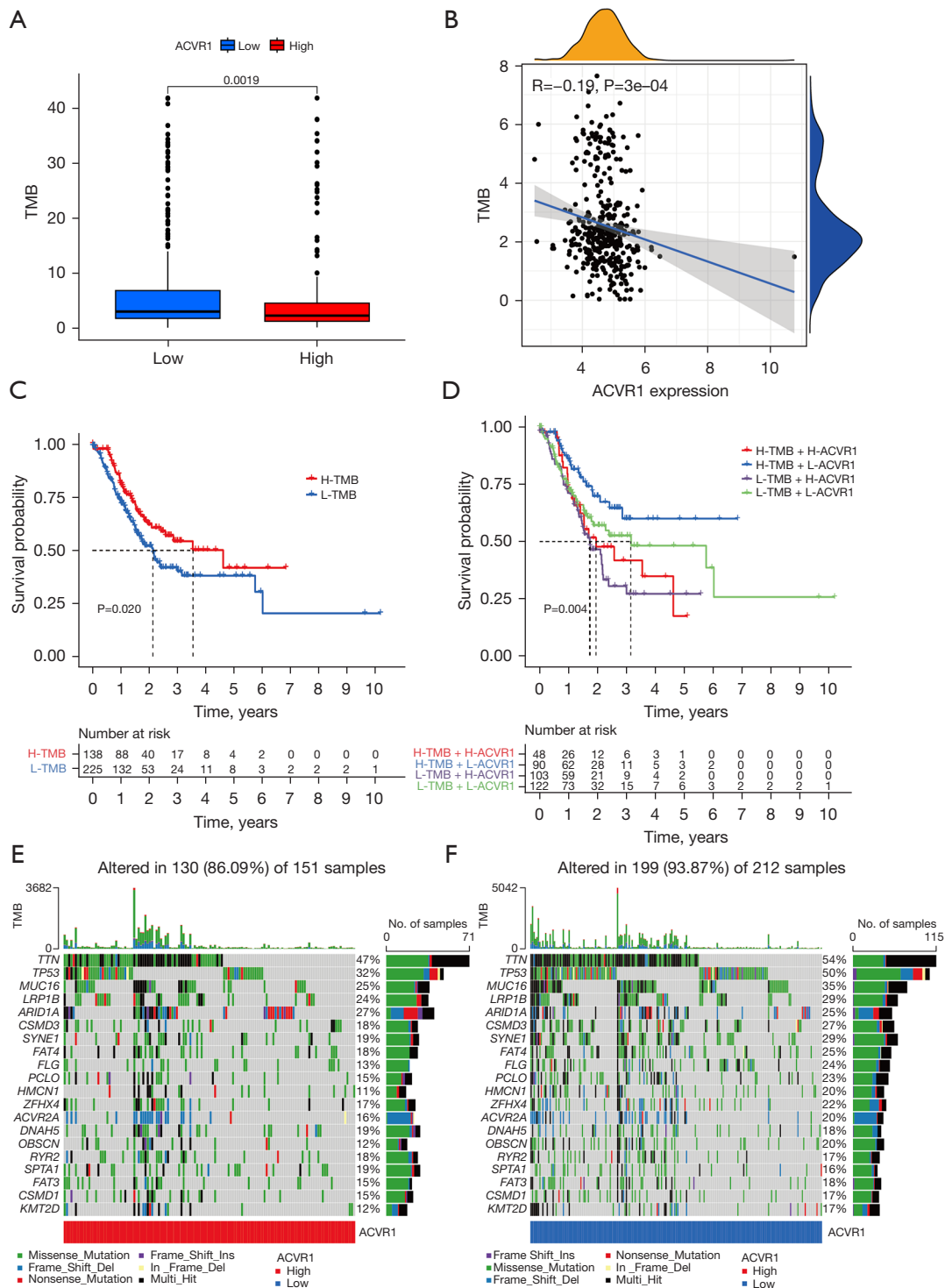


Figure 9 Correlation analysis of *ACVR1* with TMB. (A) Correlation analysis of high and low expression of *ACVR1* with TMB. (B) Negative correlation graph between *ACVR1* and TMB. (C) Correlation between high and low expression of TMB and prognosis. (D) Stratified survival curves between high and low expression of *ACVR1* and TMB and OS. (E,F) Waterfall plots of the *ACVR1* high- and low-expression groups with tumor somatic mutations. *ACVR1*, activin A receptor type-1; TMB, tumor mutation burden; OS, overall survival.

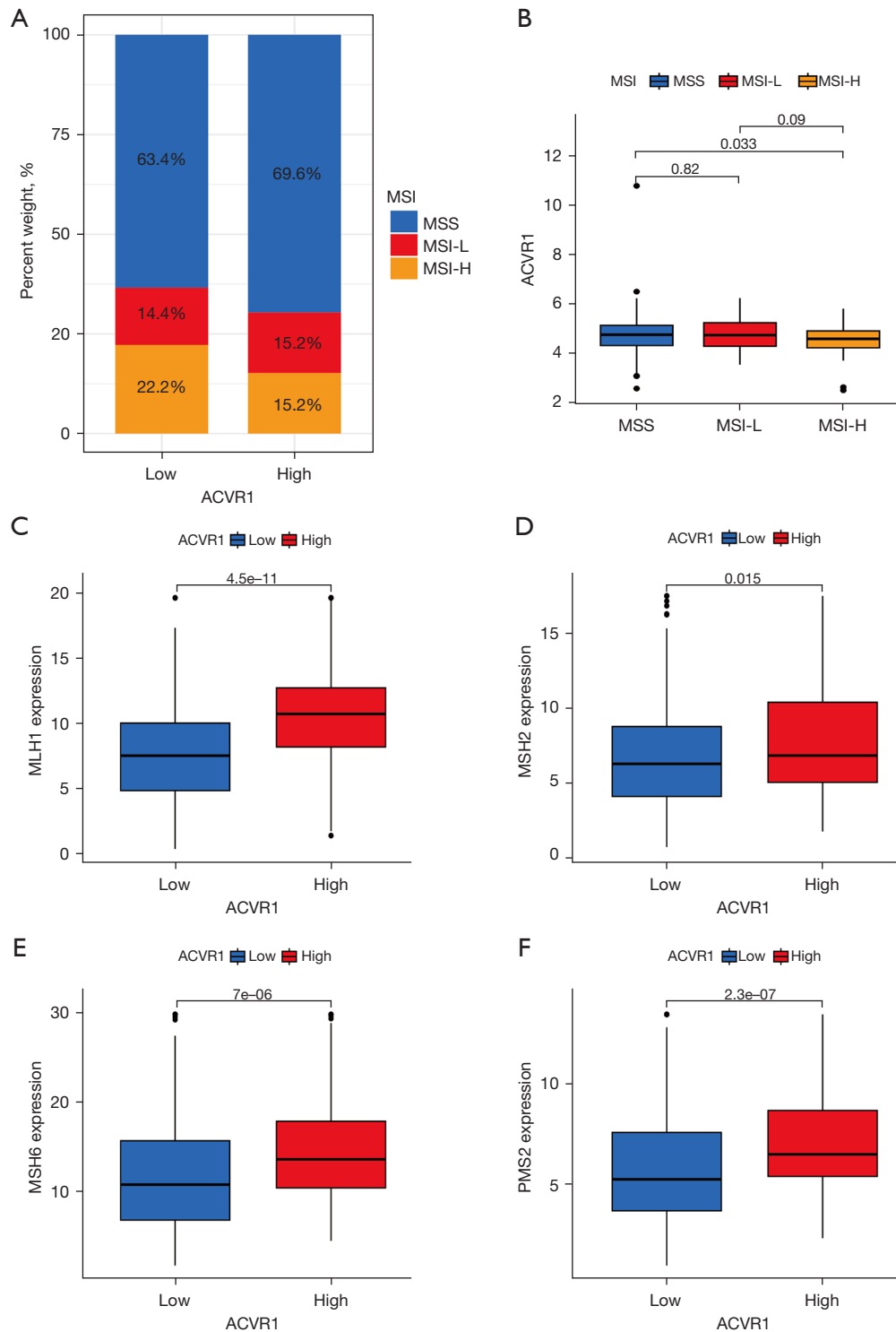


Figure 10 Correlation analysis between *ACVR1* and MSI/MMR. (A) Percentage plot of different states of MSI in the *ACVR1* high- and low-expression groups. (B) Correlation analysis between *ACVR1* expression and various forms of MSI. (C) Correlation analysis of *ACVR1* expression with MLH1. (D) Correlation analysis of *ACVR1* expression with MSH2. (E) Correlation analysis of *ACVR1* expression with MSH6. (F) Correlation analysis of *ACVR1* expression with PMS2. *ACVR1*, activin A receptor type-1; MSI, microsatellite instability; MMR, mismatch repair; MSH, muts homolog; PMS, PMS1 homolog.

in vitro experiments have confirmed the role of *ACVRI* in the proliferation, invasion, and metastasis of gastric cancer cells, which is closely related to the mechanism of action of *H. pylori* (35). In summary, a high expression of *ACVRI* may be a prognostic biomarker for gastric cancer and promote gastric carcinogenesis.

To further explore the deep mechanism of *ACVRI*, we identified the coexpressed genes of *ACVRI* and performed KEGG enrichment analysis. We found that these genes were significantly enriched in multiple immune-related pathways. The potential mechanism of *ACVRI* is that it is involved in regulating the tumor immune microenvironment. Next, the enrichment of immune-related pathways in the *ACVRI* high- and low-expression groups was analyzed. And *ACVRI* was found to be closely associated with immune cells and stroma. TME in the *ACVRI* high-expression group had a higher stroma content as shown by the ESTIMATE and biological analyses. The *ACVRI* high-expression group had a higher StromalScore and was significantly enriched in KEGG stroma-associated pathway NF- κ B signaling pathway, Hedgehog signaling pathway, focal adhesion, ECM-receptor interaction, Hippo signaling pathway, mTOR signaling pathway, Wnt signaling pathway, TGF- β signaling pathway, JAK-STAT signaling pathway, and cytokine-cytokine receptor interaction. All of them are associated with cell proliferation, differentiation, apoptosis, and migration. Among them, the TGF- β and Wnt signaling pathways promote EMT and tumor development (13). Focal adhesion is an essential step in cell migration (39). Cancer cell proliferation and metastasis depend on a hypoxic environment, and HIF-1 is an adaptive mechanism that arises in response to a hypoxic environment. Cancer cells undergo metabolic reprogramming through the HIF-1 signaling pathway to ensure that the substances required for the growth of cancer cells are provided under hypoxic conditions (40). The ssGSEA in this study showed that the stroma-associated pathway that promotes cancer was enriched in the *ACVRI* high-expression group. Therefore, high *ACVRI* expression can promote the increase of stromal components in the tumor immune microenvironment, promoting gastric cancer's invasion and metastasis.

The TME is composed of stromal components and immune cells and their pathways. In the KEGG pathway, neutrophil extracellular trap formation captures circulating cancer cells and enhances the metastatic spread of cancer, in addition to awakening dormant cancer cells and promoting cancer recurrence and metastasis (41). Leukocyte transendothelial migration, PD-L1 expression

and PD-1 checkpoint pathway in cancer, B-cell receptor signaling pathway, Th17 cell differentiation, Th1 and Th2 cell differentiation, NK cell-mediated cytotoxicity, T-cell receptor signaling pathway, chemokine signaling pathway, platelet activation, Toll-like receptor signaling pathway, and immune-related pathways immune cells play an essential role in these processes.

The relationship between immune cells and tumors is exceptionally complex, and different immune cells have exerted different functions. On the one hand, immune cells can inhibit gastric cancer. For instance, Tregs are a subpopulation of CD4⁺ T cells, which secrete immunosuppressive cytokines, granzyme A and granzyme B, to induce apoptosis of T cells, NK cells, NKT cells, APCs, and Th1 cells via cytotoxicity and phagocytosis, but can also activate macrophages, neutrophils, NK cells, and CD8 T cells to fight against tumors. CD8⁺ T cells are the main effector cells of the antitumor immune response and differentiate into cytotoxic effector T cells after activation. They induce the death of target cells through cytotoxic effects. The direct cytotoxic potential of CD4 T cells can directly kill infected and transformed cells. On the other hand, immune cells can promote gastric cancer progression. For instance, tumor-associated mast cells (TAMCs) are associated with poorer prognosis in patients with colorectal, gastric, or pancreatic ductal adenocarcinoma. In terms of tumor promotion, TAMCs achieve angiogenesis and lymphangiogenesis by secreting angiogenic components and can also suppress T-cell immunity by disrupting antitumor immunity in human gastric cancer through the expression of PD-L1. B cells can activate complement by producing IL-10, which induces protumorigenic activity. B regulatory cells inhibit the immune process by disrupting human gastric cancer's Th1–Th2 balance. Th2 cells are predominantly associated with protumorigenic activity. Meanwhile, neutrophils and macrophages can be polarized within the TME, and different polarization states exert additional functions. For example, macrophages can be polarized into 2 phenotypes, M1 or M2 macrophages (mostly the M2 type), which can inhibit T-cell and NK-cell functions by inducing the expression of TIM-3, PD-1, and CTLA-4.

On the contrary, M1-type macrophages can enhance antitumor immunity. pDCs have both tumor-promoting and tumor-suppressing effects (42), and IL-2 has been considered a key molecule in promoting T-cell proliferation and differentiation; however, a recent study has shown that IL-2 can induce CD8⁺ T cell depletion, suppressing antitumor immune responses, thus making IL-2 α double-

edged sword (43). Combining the results of the immune cell CIBERSORT algorithm, ssGSEA algorithm, and Spearman correlation analysis, we found that tumor-promoting immune cells were more densely infiltrated in the tumor immune microenvironment of the *ACVR1* high-expression group. TISCH2 online single-cell data confirmed the increased infiltration abundance of immune cells in gastric cancer with high *ACVR1* expression. Prediction of immunotherapy efficacy in high-expression *ACVR1* in the IMvigor210 immunotherapy cohort data showed poor treatment efficacy in patients with gastric cancer. The *ACVR1* high-expression group had a higher TIDE immune escape score, which predicted that the high-expression group may not benefit from immunotherapy. Cumulatively, the above analysis suggests that high *ACVR1* expression may promote the development of gastric cancer by suppressing immunity.

We examined other potential evidence concerning the association of *ACVR1* expression with immune checkpoints. Most immune checkpoints had increased expression in the *ACVR1* high-risk group and were significantly correlated with *ACVR1* expression. Among the immune checkpoint inhibitors, some PD-1 and CTLA-4 inhibitors have been validated for clinical use due to show favorable therapeutic effects (44). However, despite the upregulation of PD-1 and CTLA-4 expression in the high-risk group, predictions of therapeutic efficacy showed better immunotherapy in the low-risk group. PD-1/PD-L1 target inhibitors have proven to be therapeutically effective in a subset of patients, but a significant proportion of patients still do not benefit due to primary or secondary resistance. A study has demonstrated that both aberrant upregulation of PD-L1 expression and deletion of PD-L1 expression may lead to the ineffectiveness of PD-1/PD-L1 target inhibitors. The ineffectiveness of target inhibitors may also be the result of impaired antigen expression, presentation, recognition, activation of T cells, as well as of abnormalities in T cells themselves (45).

Abnormal T-cell immunity includes insufficient T-cell infiltration, dysfunction, and exhaustion, with the latter mostly consisting of CD8⁺ T cell depletion. Depleted CD8⁺ T cells (CD8⁺ Tex) are categorized into PD-1^{hi} and PD-1^{lo}. Immune checkpoint blockade reactivates PD-1^{lo} CD8⁺ Tex causing CD8⁺ T cells to expand, thereby promoting antitumor immunity yet leading to the apoptosis of PD-1^{hi} CD8⁺ Tex. PD-1^{hi} is accompanied by enhanced expression of coinhibitory receptors (including TIM-3, LAG-3, CD160, 2B4, TIGIT, and CTLA-4), yet cells with high expression of coinhibitory receptors cells are more severely

depleted (46). Thus, depleted T cells cannot control tumor progression at the late stages due to the presence of immune tolerance and immunosuppressive mechanisms. We further explored common chemotherapeutic and targeted agents in patients with a high expression of *ACVR1* and found that the IC₅₀ values were lower in the high-expression group, implying a more pronounced therapeutic benefit in high-expression group. Despite the suboptimal predictive results of immunotherapy, conventional therapies combined with immunotherapy may be more effective. Radiotherapy, chemotherapy, and targeted therapy may improve the immune status, eliminate suppressive immune factors, and even cause distant effects (47). In conclusion, the results suggest that high *ACVR1* expression is effective for chemotherapy and targeted therapy and can be combined with immunotherapy.

In recent years, new immunotherapy prognostic biomarkers, such as TMB, MSI, and MMR have attracted extensive research, and some have been applied in clinical practice. Therefore, we explored these biomarkers with *ACVR1*. The results showed that TMB and MSI-H were negatively correlated in the *ACVR1* high-expression group. Four markers of MMR showed upregulated expression, and pMMR was an indicator of MSI-L and MSS. Increased expression of TMB and MSI-H was favorable for immunotherapy (48), whereas low TMB and MSI-L and MSS in the *ACVR1* high-expression group suggested a poor immunotherapy effect. In particular, patients with high *ACVR1* expression and low TMB had the worst survival and prognosis. In addition, among somatic cell mutations, *TNN* has the highest mutation frequency. *TNN* has an increased degree of mutation in gastric cancer, thereby inducing the progression of gastric cancer (49). However, the *TNN* mutation frequency in our study was lower in the *ACVR1* high-risk group than in the low-risk group, indicating that the malignancy of gastric cancer was higher in the low-expression group, which implies better immunotherapy in the low-risk group and poor immunotherapy in the high-risk group (50). In conclusion, these biomarkers reconfirmed that patients with gastric cancer and high expression of *ACVR1* have poor prognosis and receive substantially reduced benefit from immunotherapy.

Conclusions

In this study, we analyzed the expression, prognosis, and immune microenvironment of *ACVR1* from TCGA database, verified the expression and prognosis of *ACVR1*

by IHC data, characterized the immune cell infiltration using single-cell data, and predicted the effect of immunotherapy using immune cohort data. The results showed that high *ACVR1* expression may promote gastric cancer development by suppressing immunity due to the enrichment of related cancer-promoting stromal pathways, high infiltration of inhibitory immune cells, high immune escape, and low TMB, MSI-L, and *TNN* mutation rates. Moreover, the immunotherapy benefit was less in the *ACVR1* high-expression group. The limitation of this study was that experimental validation was insufficient, and *in vivo* and *in vitro* experiments are needed to fully explore the molecular mechanism underlying the relationship between *ACVR1* and immunity and the effect of drug therapy on gastric cancer.

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Footnote

Reporting Checklist: The authors have completed the REMARK reporting checklist. Available at <https://jgo.amegroups.com/article/view/10.21037/jgo-23-984/rc>

Data Sharing Statement: Available at <https://jgo.amegroups.com/article/view/10.21037/jgo-23-984/dss>

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Conflicts of Interest: All authors have completed the ICMJE uniform disclosure form (available at <https://jgo.amegroups.com/article/view/10.21037/jgo-23-984/coif>). The authors have no conflicts of interest to declare.

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). The study was approved by the ethics board of Affiliated Hospital of Nantong University (No. 2023-L067) and informed consent was taken from all the patients.

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