Abstract. Clear cell renal cell carcinoma (ccRCC) is one of the most common types of malignant tumors and early detection contributes to a better prognosis. Finding new biomarkers for the diagnosis or treatment remains meaningful. DEF6 guanine nucleotide exchange factor (DEF6) is upregulated in ccRCC compared to normal controls, but the relationship between DEF6 expression and prognosis in ccRCC is unclear. Moreover, the potential biological functions of DEF6 in ccRCC remains unclear. In the present study, the Cancer Genome Atlas (TCGA), Gene Expression Omnibus (GEO), TISIDB and the clinical database of the Peking University First Hospital were used to analyze DEF6 expression in ccRCC. Immunohistochemistry (IHC), western blotting and reverse transcription-quantitative PCR were used to examine the DEF6 protein and mRNA expression levels in cell lines and clinical samples. Subsequently, the Kaplan-Meier method and Cox regression analyses were used to determine the impact of DEF6 expression on the overall survival of patients alongside other clinical variables in both the TCGA database and the present clinical database. The results showed that both DEF6 mRNA and protein expression levels were upregulated in ccRCC compared to normal controls. The Kaplan-Meier survival analysis showed that patients with high DEF6 expression had poor prognoses from both the TCGA database and the present clinical database. Univariate survival analysis and multivariate survival analysis revealed that DEF6 could be an independent prognostic factor for ccRCC. Additionally, bioinformatics analysis indicated that differentially expressed genes related to DEF6 expression influenced ccRCC by regulating the tumor immune microenvironment. In conclusion, overexpression of DEF6 is significantly correlated with a poor prognosis for patients with ccRCC and DEF6 may influence the biological processes involved with ccRCC by regulating the immune microenvironment.

Introduction

Renal cell carcinoma (RCC) is one of the most common malignant tumor types worldwide, representing approximately 3-4% of all human cancers (1). The major histological type of RCC is clear cell RCC (ccRCC), which accounts for >70% of all types of kidney cancers (2). The early detection of ccRCC contributes to a better prognosis (3). In the past decade, the genetic alterations behind ccRCC have been studied using bioinformatics analyses (4,5). Bioinformatics analyses have become one of the most effective tools for analyzing human diseases (6). In addition, rapid technological advances led by academic institutions have continued to broaden the application of high-throughput sequencing technology from research to the clinic (7). Furthermore, scientists have proposed numerous potential genes that are related to the prognosis for patients with ccRCC. However, few genes have been discovered that are valid targets for diagnosis or treatment (3,8). Therefore, finding further novel biomarkers of ccRCC, for the diagnosis or treatment, remains fruitful.

ccRCC is a highly immune-infiltrated tumor (9). Historically, ccRCC was one of the first malignant tumor types that responded to immunotherapy and continues to be among the most responsive (10,11). Previous studies have shown that T cells are the most abundant immune-infiltrating cells in ccRCC (12). Hence, genes that regulate the functions of immune cells, especially T cells, may be correlated with the prognosis and effectiveness of immunotherapy for patients with ccRCC.

DEF6 guanine nucleotide exchange factor [GEF (DEF6)], is a 631 amino acid Rho-family GEF (13,14). It is highly expressed in T and B cells, and regulates various immune-related processes such as the activation of CD4+ T cells.
T cells and the differentiation of T helper cells (15,16). It also regulates cell morphology in cooperation with activated Rac1, and affects cell differentiation in collaboration with integrins (17,18). More importantly, as a GEF, DEF6 can activate and affects cell differentiation in collaboration with T cells and the differentiation of T helper cells (15,16). It also

website (https://www.cancer.gov/about-nci/organiza-

clinical information were obtained from the TCGA

obtained from medical records, such as Fuhrman score and

rence was obtained from all patients. All

the Ethics Committee of Peking University First Hospital and

body mass index (BMI). The present study was supported by

proliferation, migration and invasion (19). Previous studies

that high DEF6 expression levels predicted a poor prognosis

for ccRCC. Furthermore, bioinformatics analyses revealed that DEF6 may regulate the components of the immune microenvironment to influence the processes behind ccRCC.

Materials and methods

**Extraction of clinical and gene expression data from the ccRCC databases.** Transcription profiles from high-throughput sequencing fragments per kilobase per million (HTSeq-FPKM) and corresponding clinical information were obtained from the TCGA website (https://www.cancer.gov/about-nci/organiza-
cation/ceg/research/structural-genomics/tcga), which includes 539 ccRCC samples and 72 normal samples. Nine micro-
array datasets were also downloaded from the GEO website (https://www.ncbi.nlm.nih.gov/geo/). The data from the GEO
databases were translated into log2 values for sequencing analysis. According to a previously reported method, if standardized data were not available, the raw data were downloaded. If a gene was found to have multiple probes in the same chip, the average value of all probes was taken as the expression value of the gene (24). Samples which were neither ccRCC, nor adjacent kidney tissue, were excluded from the present study. The ‘sva’ R package (version 3.6.1) was used to remove the batch effect (25). There were more than 700 specimens obtained from the following GEO datasets: GSE4282, GSE46699, GSE53757, GSE15641, GSE68417, GSE14994, GSE40435, GSE71963 and GSE76351.

**Patients and specimens.** A total of 146 paired samples from patients diagnosed with ccRCC were included in the resent study. The patients ranged in age from 20 to 83 years, including 97 males and 51 females. All patients underwent renal resec-
tion at Peking University First Hospital between June 2008 and January 2011. Clinical data of the recruited patients was obtained from medical records, such as Fuhrman score and body mass index (BMI). The present study was supported by the Ethics Committee of Peking University First Hospital and written informed consent was obtained from all patients. All procedures were performed according to the World Medical Association Declaration of Helsinki.

**Immunohistochemistry (IHC).** Firstly, the tissue samples were fixed in 10% formalin for 24 h at room temperature. Subsequently, the 4-μm paraffin-embedded tissue sections were prepared at room temperature. Immunostaining was performed using a two-step detection kit (cat. no. PV-9000; OriGene Technologies, Inc.) according to the manufacturer's protocol. The sections were deparaffinized in xylene at room temperature, rehydrated in a graded alcohol series and then boiled in citrate buffer (pH 6.0) for 30 min in an autoclave. Endogenous peroxidases were blocked by incubation in 3% H2O2 for 30 min at room temperature. The sections were washed in PBS, blocked with 10% goat serum (OriGene Technologies, Inc.) at room temperature for 1 h and incubated with anti-DEF6 (cat. no. ab247011; 1:20,000; Abcam) at 4°C overnight. The sections were washed in PBS solution three times and incubated with a reaction enhancer kit (cat. no. PV-9000; OriGene Technologies, Inc.) for 20 min at room temperature, then washed in PBS solution three times. The sections were then incubated with peroxidase-conjugated secondary antibodies (cat. no. PV-9000; 1:1,000; OriGene Technologies, Inc.) for 30 min at room temperature. All slides were counterstained with DAB solution for 3 min and 20% hematoxylin for 1 min at room temperature and dehydrated. The primary antibody diluent was used as a negative control.

**Evaluation of immunostaining staining.** Two experienced independent investigators (ZZ and CX) examined all tumor slides by examining five random fields of view and observing 100 cells per view at x400 magnification using a light micro-
scope (Olympus Corporation). The staining intensity was classified as 0 (no staining), 1 (weak), 2 (moderate) or 3 (strong); The proportion of stained tumor cells was scored as 0 (0-5%), 1 (6-25%), 2 (26-50%), 3 (51-75%), 4 (>75%). The product of these two variables was used to calculate a final score based on a previous study (26), as follows: 0 (product of 0-3); 1 (product of 4-6); 2 (product of 7-9); 3 (product of 10-12).

**Cell lines.** HK-2, 293, 786-O, 769-P, Caki-1, ACHN, A498 and OSCR2 were acquired from the American Type Culture Collection. Cells were cultured in DMEM (Invitrogen; Thermo Fisher Scientific, Inc.) or 1640 media (Invitrogen; Thermo Fisher Scientific) containing 10% FCS (Invitrogen; Thermo Fisher Scientific, Inc.) and 1% penicillin-streptomycin (Gibco; Thermo Fisher Scientific, Inc.). Cells were cultured in 10 mm culture dishes in a 5% humidified atmosphere at 37°C.

**Reverse transcription-quantitative PCR (RT-qPCR).** Total RNA from 20 paired clinical samples was obtained from 20 patients diagnosed with ccRCC by Peking University First Hospital and extracted using TRIzol reagent according to the manufacturer's protocol (Invitrogen; Thermo Fisher Scientific). Subsequently, cDNA was synthesized from 5-10 µg of total RNA using the Super Master Mix synthesis kit (Takara Bio, Inc) and the used condition of synthesizing cDNA were as follow: 15 min at 37°C, 5 sec at 85°C and 5 min at 4°C. Then, quantification of all gene transcripts was performed by RT-qPCR using the SYBR Premix ExTaq kit (Takara Bio,
Inc.) and α-tubulin was used as a normalizing control. The primer pairs used were as follows: DEF6 forward primer, 5'-TACATGCCCTACCTCAACAAGT-3' and reverse primer, 5'-TGTTCCCGTTGCTATCTGCC-3'; α-tubulin forward primer, 5'-ACCTTAACCGCCTTATTAGCCA-3' and reverse primer, 5'-CACCACGGTACAACAGGCA-3'. Each reaction was performed four times and the qPCR conditions used were as follows: 10 min at 95˚C, 40 cycles of 15 sec at 95˚C and 1 min at 60˚C. The 2-ΔΔCq method was used to calculate the relative gene expression level (27). The relative mRNA expression levels were further normalized using the following formula: Z=(x-μ)/σ, where 'x' represents the observation of the sample expression, 'μ' represents the sample mean expression, and 'σ' represents the sample SD (28).

Western blot analysis. Total proteins from cells were extracted using the NP-40 lysis buffer (cat. no. P0013F; Beyotime Institute of Biotechnology) and quantified using the BCA method. The supernatant (20 µg of protein) was denatured and separated on 10% SDS-PAGE. Samples were then transferred to 0.22-µm PVDF membranes and blocked in skimmed milk for 1 h at room temperature. After that, samples were incubated overnight at 4˚C with the antibodies against DEF6 (cat. no. ab247011; 1:1,000; Abcam) and β-actin (cat. no. sc47778; 1:1,000; Santa Cruz Biotechnology, Inc.). After incubation with peroxidase-coupled anti-rabbit IgG (cat. no. 20774; 1:1,000; Cell Signaling Technology, Inc.) at 37˚C for 2 h, bound proteins were visualized using an ECL kit (Pierce; Thermo Fisher Scientific, Inc.) and detected using a DNR Bioimaging System (DNR Bio-Imaging Systems, Ltd.). Relative protein levels were quantified using β-actin as the loading control.

Identification of differentially expressed genes (DEGs) related to DEF6 expression. According to the DEF6 median expression in ccRCC samples, the analyzed groups were divided into a low DEF6 expression group (n=269) and a high DEF6 expression group (n=270) (29). Subsequently, the ‘DESeq2’ package and ‘edgeR’ R package (version 3.6.1) were used to screen DEGs between samples. Those with an absolute log2 fold-change >1 and a false discovery rate (FDR) <0.05 in the TCGA or GEO database were considered statistically significant.

Protein-protein interaction (PPI) network analysis. STRING version 11.0 (https://string-db.org/) was used to evaluate the PPI information of all DEGs related to DEF6 expression. The PPI network included 108 nodes and 324 edges. Subsequently, the ‘DESeq2’ package and ‘edgeR’ R package (version 3.6.1) were used to screen DEGs between samples. Those with an absolute log2 fold-change >1 and a false discovery rate (FDR) <0.05 in the TCGA or GEO database were considered statistically significant.
were selected as hub genes, and interactions (combined score >0.4) were considered significant. The interaction network of these proteins was visualized using Cytoscape 3.7.1 and the molecular complex detection (MCODE) plug selected necessary modules were applied (both MCODE score and node number >4) (30).

Functional enrichment analyses of DEGs related to DEF6 expression. To further explore the mechanism of action DEF6 in regulating ccRCC, gene ontology (GO) enrichment analysis and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway analysis were performed using the ‘ggplot2’ R package (version 3.6.1; http://www.rstudio.com/). A P-value of <0.05 was considered statistically significant (31).

Moreover, gene set enrichment analysis (GSEA) was performed to identify the potential biological pathways. GSEA software (version 4.0.3; http://software.broadinstitute.org/gsea/index.jsp) was conducted on the JAVA 8.0 platform. By using the TCGA database, the high group and low group were classified according to the average mRNA expression levels of DEF6. For each analysis, gene set permutations were implemented 1,000 times. Significantly enriched gene sets were identified, which produced FDR q‑values <0.05 (32).

TISIDB analysis. The TISIDB database (http://cis.hku.hk/TISIDB) integrated high-throughput data from 988 reported immune-related anti-tumor genes. The database enabled the analysis of correlations for the expression of selected genes with the expression of lymphocytes, immunomodulators and chemokines (33). In the present study, the relationships between the expression levels of DEF6 and lymphocytes as well as immunomodulators across various types human cancers, especially ccRCC, were investigated.

Statistical analysis. In the present study, unpaired t-tests were used to compare the mRNA expression levels in ccRCC tissues and normal renal tissues from the TCGA and GEO database, using SPSS 26.0 (IBM Corp.). One-way ANOVAs were conducted to calculate the difference among multiple groups, and Bonferroni's corrections were used for the post-hoc tests. Patients with unclear values (NX, MX, GX) or missing values were excluded from subsequent analyses. Paired t-tests were performed to calculate the differences between paired samples and the Wilcoxon signed rank test was used to calculate the differences among IHC scores of clinical samples. The Pearson's correlation analysis was used to calculate the correlation between DEF6 expression and clinical parameters. The Kaplan-Meier method and Cox regression were used to compare the impact of DEF6 expression on the overall survival (OS) of patients alongside other clinical variables in both the TCGA database and the present clinical database. All P-values are based on a two-sided statistical analysis, and P<0.05 was considered to indicate a statistically significant difference.

Results

DEF6 mRNA expression levels are upregulated in ccRCC. The DEF6 mRNA expression levels were assessed in ccRCC using the GEO database, the TCGA database, clinical samples collected from the present study and cell lines. Firstly, compared with normal samples, in the ccRCC samples, the DEF6 mRNA expression levels were upregulated in the 9 GEO datasets...
(Fig. 1A). Furthermore, a forest plot, based on the standardized mean difference, showed the meta-analysis of 9 GEO datasets (Fig. 1E). The outcomes similarly indicated that DEF6 mRNA expression levels in ccRCC samples were upregulated in the 9 GEO datasets compared with the normal samples. In the TCGA database, the DEF6 mRNA expression levels were also increased in ccRCC (Fig. 1B). Furthermore, the DEF6 mRNA expression levels were elevated in 20 paired clinical samples (Fig. 1C). The relative DEF6 mRNA expression levels between the ccRCC tissues and paired normal tissues in each sample is also shown (Fig. 1F). In most samples, the mRNA expression levels of DEF6 in the ccRCC tissues was higher than that of the normal controls. In addition, the outcome of RT-qPCR in RCC cell lines showed that the DEF6 mRNA expression levels were significantly upregulated in the 786-O, ACHN, OSRC-2, and A498 cells compared to the HK-2 cells (Fig. 1D).

DEF6 protein expression levels are upregulated in ccRCC. Representative staining of tissues (Fig. 2A) and the representative scores (Fig. 2C) are shown. A higher DEF6 expression level was observed in the ccRCC tissue according to the IHC scores of 146 paired ccRCC tissues and corresponding adjacent normal renal tissues. The mean IHC scores of ccRCC tissues and adjacent normal renal tissues were 1.73 and 1.20, which showed significant difference. The frequency distribution of these scores is demonstrated in Fig. 2D. Western bolt analysis also showed that the DEF6 protein expression levels were increased in the ccRCC cell lines compared with the healthy cell lines (Fig. 1D).

Overexpression of DEF6 is an unfavorable prognostic factor for patients with ccRCC. To evaluate the prognostic value of DEF6 in ccRCC, the patient OS in both the TCGA database and the present study's clinical database were analyzed using the Kaplan-Meier log-rank test. It was found that high DEF6 expression was correlated with a poor prognosis for patients with ccRCC in both the TCGA database and the present database (Fig. 3A and B). Furthermore, the correlation between DEF6 expression and clinicopathological characteristics in patients with ccRCC was explored in the TCGA database. DEF6 expression was significantly associated with the pathologic grade (Fig. 3C), pathologic T (Fig. 3D), pathologic M (Fig. 3E) and pathologic stage (Fig. 3F).

Whether DEF6 could be an independent prognostic factor for patients with ccRCC was subsequently explored. Cox regression analysis was conducted using both the TCGA database and the present clinical database. Detailed patient information from both the TCGA database (Table SI) and the present clinical database (Table SII) is shown. As shown in Table SI, the clinical characteristics (size, pathologic T, pathologic M and pathologic N) showed a significant difference between the high- and low-DEF6 expression groups in the TCGA database. Meanwhile, as shown in Table SII, the clinical characteristics (size) showed a significant difference between the high- and low-DEF6 expression groups in the present clinical database.

In the TCGA database, univariate Cox regression analysis showed that DEF6 expression, age at diagnosis, tumor size, histologic grade, pathologic T and pathologic M stages were correlated with the OS of patients with ccRCC (Table I). Moreover, multivariate Cox regression analysis indicated that DEF6 expression, age at diagnosis and pathologic M stage were independent prognostic factors for OS (Table I).

Moreover, in the present clinical database, DEF6 expression, tumor size and pathologic T were correlated with the OS of patients with ccRCC (Table II). In addition, multivariate Cox regression analysis indicated that DEF6 expression and pathologic T stage were independent prognostic factors for OS (Table II).

Biological analysis of DEF6 in ccRCC. The analysis contained 539 ccRCC samples in the TCGA database, 101 ccRCC
samples in GSE40435 and 71 ccRCC samples in GSE53757. A total of 188 overlapping DEGs related to DEF6 expression were identified from 3 datasets, including 180 upregulated genes and 8 downregulated genes (Fig. 4A).

Moreover, GO, KEGG and GSEA analyses were conducted to explore the biological functions of these DEGs. The GO analysis showed that the DEGs were mainly enriched in the activation and regulation of the immune microenvironment (Fig. 4C). Moreover, the KEGG pathway enrichment analysis indicated that the primary functions of the identified DEGs were the regulation of the immune microenvironment. Interestingly, programmed cell death ligand 1 expression and programmed cell death 1 checkpoint pathways in cancer were also enriched in the KEGG pathway (Fig. 4D). Furthermore, GSEA analysis showed that 7 gene sets were upregulated and 4 gene sets were downregulated, with significant enrichment at both the NOM P-value<0.05 and FDR q-value<0.05 (Table III). These gene sets were mainly correlated with the immune response and tumor metabolism (Fig. 4B).

Construction of the PPI network. To identify the hub genes of these DEGs related to DEF6 expression, a PPI network was constructed using STRING 11.0 (https://string-db.org/), including 108 nodes and 324 edges. Subsequently, Cytoscape (version 3.7.1) was used to analyze the PPI network, where a low degree value is correlated with a small node size and a low co-expression value is related to a small edge size. The co-expression network showed that protein tyrosine phosphatase receptor type c (PTPRC), integrin subunit beta 2 (ITGB2), lymphocyte‑specific protein tyrosine kinase (LCK) and cytotoxic T-lymphocyte associated protein 4 (CTLA4) were hub genes with a degree >20 (Fig. 5A). The MCODE plugin was then used to explore the critical modules of target genes and two imperative modules were identified with MCODE score and node number >4 (Fig. 5B and C). Two imperative modules might act as the core regulatory network for the biological functions.

Regulation of molecules involved with the immune microenvironment by DEF6. The aforementioned biological analyses showed that DEF6 was associated with the immune microenvironment. As such, the TISIDB database was used to explore the correlations between DEF6 expression and lymphocytes and immunomodulators, using Spearman’s correlation tests.

Table I. Univariate analysis and multivariate analysis of overall survival in the TCGA database.

| Parameters                                 | Univariate analysis | Multivariate analysis |
|--------------------------------------------|---------------------|-----------------------|
|                                            | HR (95% CI)         | P-value               | HR (95% CI)         | P-value               |
| DEF6 level, high vs. low                   | 2.076 (1.502, 2.869) | <0.001                | 1.856 (1.335, 2.580) | <0.001                |
| Age at initial pathologic diagnosis, >60 vs. <60 years | 1.734 (1.248, 2.410) | <0.01                 | 1.593 (1.142, 2.222) | <0.01                 |
| Gender, male vs. female                    | 1.077 (0.776, 1.496) | 0.657                 |                       |                       |
| Size, longest dimension, ≥2 cm vs. <2 cm   | 1.526 (1.114, 2.091) | <0.01                 |                       |                       |
| Histologic grade, G1 + G2 vs. G3 + G4      | 2.332 (1.640, 3.315) | <0.001                |                       |                       |
| Pathologic T, T3 + T4 vs. T1 + T2          | 2.739 (2.003, 3.747) | <0.001                |                       |                       |
| Pathologic N, N1 vs. N0                    | 0.983 (0.722, 1.338) | 0.912                 |                       |                       |
| Pathologic M, M1 vs. M0                    | 4.289 (3.126, 5.884) | <0.001                | 2.673 (1.819, 3.928) | <0.001                |

CI, confidence interval; DEF6, DEF6 guanine nucleotide exchange factor; HR, hazard ratio.

Table II. Univariate analysis and multivariate analysis of overall survival in the clinical database of the present study.

| Parameters                                 | Univariate analysis | Multivariate analysis |
|--------------------------------------------|---------------------|-----------------------|
|                                            | HR (95% CI)         | P-value               | HR (95% CI)         | P-value               |
| DEF6 level, high vs. low                   | 7.981 (1.058, 60.211) | <0.05                 | 1.856 (1.335, 2.580) | <0.05                 |
| Age at initial pathologic diagnosis, ≥60 vs. <60 years | 1.898 (0.721, 4.966) | 0.194                 |                       |                       |
| Gender, male vs. female                    | 1.334 (0.468, 3.801) | 0.590                 |                       |                       |
| Size, longest dimension, ≥6 cm vs. <6 cm   | 6.060 (2.132, 17.219) | <0.01                 |                       |                       |
| BMI, ≥23.9 vs. <23.9                       | 0.844 (0.325, 2.187) | 0.726                 |                       |                       |
| Pathologic T, T3 + T4 vs. T1 + T2          | 10.859 (3.977, 29.647) | <0.001                | 7.101 (1.965, 25.659) | <0.01                 |
| Fuhrman score, 2+3 vs. 1                   | 4.898 (0.649, 36.992) | 0.123                 |                       |                       |

BMI, body mass index; CI, confidence interval; DEF6, DEF6 guanine nucleotide exchange factor; HR, hazard ratio.
Figs. S1A and 6A showed the correlation between DEF6 expression and tumor-infiltrating lymphocytes, and the 4 types of tumor-infiltrating lymphocytes that significant correlated with DEF6 expression, including myeloid-derived suppressor cells (MDSCs), activate CD8 T cells (Act_CD8), activate B cells (Act_B) and effector memory CD8 T cells (Tem_CD8). Moreover, Figs. S1B and 6B showed the correlation between DEF6 expression and immunoinhibitors, and the 4 significant immunoinhibitors included lymphocyte activating 3 (LAG3), programmed cell death 1 (PD-1), programmed cell death ligand 1 (PD-L1), programmed cell death ligand 2 (PD-L2), and T Cell Immunoreceptor with Ig And ITIM Domains (TIGIT) and CD96. Furthermore, Figs. S1C and 6C showed the correlation between DEF6 expression and immunostimulators, and the 4 significant immunostimulators included CD27, Lymphotoxin Alpha (LTA), Killer Cell Lectin Like Receptor K1 (KLRK1) and CD48. Figs. S1D and 6D showed the correlation between DEF6 expression and major histocompatibility complexes (MHC) molecules, and the 4 significant MHC molecules included 4 major histocompatibility complexes (HLAs), which are HLA-DOB, HLA-DPB1, HLA-DMA and HLA-DRA.

Table III. The enrichment of GSEA gene sets at both the NOM P-value <0.05 and FDR q-value <0.05.

| GS follow link to MSigDB                      | Size | ES  | NES   | NOM P-value | FDR q-value |
|---------------------------------------------|------|-----|-------|-------------|-------------|
| HALLMARK_ALLOGRAFT_REJECTION                | 199  | 0.81| 2.32  | <0.001      | 0.007       |
| HALLMARK_INTERFERON_GAMMA_RESPONSE          | 198  | 0.75| 2.2   | 0.002       | 0.016       |
| HALLMARK_INFLAMMATORY_RESPONSE              | 198  | 0.7 | 2.15  | <0.001      | 0.017       |
| HALLMARK_IL6_JAK_STAT3_SIGNALING            | 86   | 0.73| 2.14  | <0.001      | 0.014       |
| HALLMARK_INTERFERON_ALPHA_RESPONSE          | 95   | 0.74| 2.06  | 0.009       | 0.021       |
| HALLMARK_IL2_STAT5_SIGNALING                | 196  | 0.56| 2     | 0.004       | 0.032       |
| HALLMARK_COMPLEMENT                         | 199  | 0.56| 1.95  | 0.017       | 0.038       |
| HALLMARK_FATTY_ACID_METABOLISM              | 157  | -0.57| -2.07 | 0.014       | 0.037       |
| HALLMARK_OXIDATIVE_PHOSPHORYLATION          | 184  | -0.68| -2.06 | 0.017       | 0.027       |
| HALLMARK_ADIPOGENESIS                       | 193  | -0.5| -1.94 | 0.022       | 0.045       |
| HALLMARK_ANDROGEN_RESPONSE                  | 96   | -0.55| -1.94 | 0.02        | 0.037       |

ES, enrichment score; FDR, false discovery rate; NES, normalized enrichment score; GS, gene sets; NOM, normalized.

Figure 4. (A) The 188 overlapping DEGs from three datasets: TCGA, GSE40435 and GSE53757. (B) Representative GSEA results showing enrichment of the immune response and tumor metabolism in ccRCC is shown. The dot plot images of (C) GO and (D) KEGG were constructed, where larger dot sizes are correlated with higher counts and a darker red color is related to lower P-value. CAM, cell adhesion molecule; CXCR, chemokine receptor type 4; DEG, differentially expressed genes; IBD, inflammatory bowel disorder; Ig, immunoglobulin; MHC, major histocompatibility complex; PD-1, programmed cell death 1; PD-L1, programmed cell death ligand 1; Th, T helper cell.
Figure 5. (A) A protein-protein interaction network was constructed and 4 hub genes with degree >20 were identified, where a low degree is correlated with a small node size and a low co-expression value is related to a small edge size. The bluer the color, the smaller the degree and the redder the color, the higher the degree. The degree refers to the number of connections the node has with other nodes. Selected important modules of the target gene with (B) MCODE score ≥4 and (C) nodes≥4. MCODE, molecular complex detection.

Figure 6. Spearman's correlation of expression of DEF6 with expression of 4 most significant lymphocytes and immunomodulators in the TISIDB database. (A) Correlations between DEF6 and the top 4 significant lymphocytes expression levels. (B) Correlations between DEF6 and the top 4 significant immunoinhibitors expression levels. (C) Correlations between DEF6 and the top 4 significant immunostimulators expression levels. (D) Correlations between DEF6 and the top 4 significant MHC molecules expression levels. Red and blue cells indicate positive and negative correlations, respectively. The color intensity is directly proportional to the strength of the correlations. DEF6, guanine nucleotide exchange factor; MHC, major histocompatibility complex. KIRC, kidney renal clear cell carcinoma.
Hence, DEF6 may influence the immune microenvironment by regulating the aforementioned immune molecules.

**Discussion**

DEF6, also known as IBP or SWAP-70, is a GEF that regulates various processes associated with the immune microenvironment. GEFs are members of the diffuse B-cell lymphoma protein family and play a significant role in regulating the activation status of Rho-GTPases (34,35). To the best of our knowledge, Rho-GTPases are associated with oncogenic activities and contribute to the processes of malignant tumor phenotypes such as migration, invasion and metastasis (36-38). Interestingly, as an upstream activator of the Rho-GTPase family, DEF6 is involved in various cellular processes, such as the cell polarity, microtubule dynamics, membrane transport pathways and transcription factor activity (39-41). Previous studies have demonstrated that overexpression of DEF6 contributes to a poor prognosis in various cancer types, such as ovarian carcinoma and prostate cancer (42,43). However, the correlation between DEF6 expression and the prognosis in ccRCC is unclear.

A previous study showed the potential functions of DEF6 in regulating kidney podocytes (44). This study was the first to reveal the correlation between DEF6 expression and the prognosis in ccRCC. In the present study, both DEF6 mRNA and protein expression levels were explored in ccRCC. Compared with the adjacent normal tissue, the DEF6 mRNA expression level was upregulated in the TCGA database, GEO database and the present clinical samples. Moreover, 146 paired ccRCC tissues and their adjacent normal tissues were analyzed using IHC. Raised DEF6 expression levels were found in ccRCC. Univariate and multivariate Cox regression suggested that high DEF6 expression was significantly related to a poor prognosis for patients with ccRCC, which indicated that DEF6 may be an independent prognostic factor for ccRCC.

According to the DEF6 expression levels, the ccRCC samples were divided into the high expression group and low expression group. Subsequently, ‘DESeq2’ and ‘edgeR’ were used to analyze the transcription profile from the TCGA and GEO databases. A total of 188 DEGs were found for subsequent analysis, including 180 upregulated genes and 8 downregulated genes. The biological functions of these DEGs were explored using GO, KEGG and GSEA enrichment analyses.

The enrichment of biological processes indicated that DEF6 mainly regulated immune microenvironment molecules by influencing their activation, difference and adhesion. In addition, cell components showed that DEGs were enriched in various membranes such as the external side of plasma membrane and ficolin-1-rich granule membrane. These DEGs were also enriched in the composition of various complexes such as the MHC II complex and reporter complex, which suggested that DEF6 may influence the transmission of biological information. Moreover, molecular function analysis showed that DEF6 chiefly regulates cytokines and bind with various structures, such as immunoglobulins and the MHC II protein complex. Moreover, KEGG pathway analysis also indicated that DEF6 participates in various processes of the immune microenvironment. Interestingly DEF6 mainly influenced the functions of immune molecules, such as Th17 cell differentiation and natural killer cell mediated cytotoxicity, which indicated that DEF6 may regulate the development of ccRCC by influencing the immune microenvironment. In addition, GSEA analysis found 7 upregulated gene sets and 4 downregulated gene sets. Enrichment analyses of these gene sets showed that DEF6 was mainly involved in the immune response and tumor metabolism. Immune response-related signaling pathways, such as the interleukin (IL)2/STAT5 pathway and IL6/JAK/ STAT3 pathway, were upregulated. In addition, tumor metabolism-related signaling pathways, such as fatty acid metabolism and oxidative phosphorylation, were downregulated.

To explore the hub genes among DEGs, a PPI network was constructed and found 4 hub genes: PTTPRC, ITGB2, LCK and CTLA4. The detailed information of the 4 hub genes was described as follows: PTTPRC belongs to a member of the protein tyrosine phosphatase (PTP) family. PTPs are signaling molecules that regulate a variety of cellular processes, including cell growth, differentiation, mitosis and oncogenic transformation (45). PTTPRC is an essential regulator of T cell and B cell antigen receptor signaling, which also suppresses JAK kinases (46). Moreover, ITGB2 is a crucial regulator of lymphocyte trafficking, activation and residence time (47). LCK is a member of the Src family and regulates the activation of T cells (48). Recent studies have shown that LCK is expressed in various tumor types, such as breast cancer, colon cancer and lung carcinoma (49-51). In addition, CLTA-4 is a member of the immunoglobulin superfamily and mediates opposing functions in T cell activation. These functions contribute to tumor development (52). A total of 4 hub genes were found to be involved in regulating the immune microenvironment. In addition, ITGB2 and CLTA-4 were also associated with various tumor processes.

The analyses performed in the present study showed that DEF6 is associated with the immune microenvironment. To the best of our knowledge, RCC has a high level of immune infiltration and T cells are the main immune cell type enriched in ccRCC (53,54). As such, the TISIDB database was used to explore the correlation DEF6 expression with lymphocytes and immunomodulators. The results showed that DEF6 had the most significant correlation with lymphocytes (such as MDSCs, Act_CD8, Act_B and Tem_CD8), immunoinhibitors [such as LAG3, PDCDI, TIGIT and CD96], immunostimulators (such as CD27, LTA, KLRK1 and CD48), and MHC molecules (such as HLA-DOB, HLA-DPB1, HLA-DMA and HLA-DRA). CD8+ T cells have been shown to be associated with improved clinical outcomes and responses to immunotherapy (55). Moreover, TIGIT and CD96 are correlated with anti-tumor immunity (56). Combined with the outcome of the GSEA and KEGG analyses, TISIDB indicated that patients with ccRCC with high DEF6 expression may benefit more from immunotherapy.

There are several limitations to this study. Firstly, the biological functions of DEF6 in ccRCC cancer cell lines need to be verified in vitro. Moreover, the specific mechanisms of action by which DEF6 influences ccRCC remain unclear and require further study. In conclusion, the present study demonstrated for the first time that both DEF6 mRNA and protein levels were upregulated in ccRCC. Overexpression of DEF6 is
an unfavorable prognostic factor for patients with ccRCC. In addition, the mechanism of action by which DEF6 regulates ccRCC may be associated with the immune microenvironment.

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Availability of data and materials
The datasets used and/or analyzed during the current study are available from the corresponding authors on reasonable request.

Authors' contributions
JL and ZPZ designed the study. JL obtained the financial support for this study. ZPZ collected the sequencing data from public datasets and analyzed the data. LRL, TYC and CRX collected the clinical samples and performed IHC staining. Moreover, Dr Cong Huang gave us guidance on cell culture and laboratory techniques.

Ethics approval and consent to participate
For the use of the clinical materials for research purposes, the study was conducted following the Declaration of Helsinki. This study was conducted with the approval of the Peking University First Hospital Ethical Committee and informed consent was obtained from all participants.

Patient consent for publication
Not applicable.

Competing interests
The authors declare that they have no competing interests.

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