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Overexpression of *PKMYT1* Facilitates Tumor Development and Is Correlated with Poor Prognosis in Clear Cell Renal Cell Carcinoma

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Study Design A
Data Collection B
Statistical Analysis C
Data Interpretation D
Manuscript Preparation E
Literature Search F
Funds Collection G

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Background: Protein kinase membrane-associated tyrosine/threonine (*PKMYT1*) has been found in many tumors, but its association with clear cell renal cell carcinoma (ccRCC) remains unclear.

Material/Methods: *PKMYT1* expression in ccRCC was examined in the Cancer Genome Atlas (TCGA), Gene Expression Omnibus (GEO), and Tumor Immune Estimation Resource databases. The correlation between *PKMYT1* expression and clinicopathological parameters was explored via the chi-square test. Receiver operating characteristic curves were used to estimate the diagnostic performance of *PKMYT1*. Kaplan-Meier curves, a Cox model, nomogram, time-dependent receiver operating characteristic curves, and decision curve analysis (DCA) were used to evaluate the prognostic value and clinical utility of *PKMYT1*. Genes coexpressed with *PKMYT1* in ccRCC were identified based on TCGA, the gene expression profiling interactive, and cBioPortal. Gene Set Enrichment Analysis revealed biological pathways associated with *PKMYT1* in ccRCC.

Results: Weighted gene coexpression network analysis identified *PKMYT1* as one of the genes most significantly correlated with progression of histological grade. *PKMYT1* was significantly upregulated in ccRCC compared with normal tissue ($P < 0.001$), with a trend toward differentiating between individuals with ccRCC and those who were healthy (area under the curve=0.942). High *PKMYT1* expression was correlated with unsatisfactory survival (hazard ratio=1.67, $P=0.001$), indicating that it is a risk factor for ccRCC. A nomogram incorporating *PKMYT1* level was created and showed a clinical net benefit. *PKMYT1* was strongly positively correlated with the anti-silencing function of 1B histone chaperone (*ASF1B*) gene in ccRCC.

Conclusions: *PKMYT1* is upregulated in ccRCC and its presence indicates poor prognosis, making it a potential therapeutic target for ccRCC.

MeSH Keywords: **Biological Processes • Carcinoma, Renal Cell • Database**

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Background

Kidney cancer is the 12th most common solid neoplasm, with approximately 400 000 new diagnoses and 175 000 tumor-associated deaths globally in 2018 [1]. Renal cell carcinoma (RCC) accounts for approximately 90% of all renal tumors and is derived from epithelial cells in renal tubules [2,3]. There are 3 primary histologic subtypes of RCC: clear cell RCC (ccRCC) (about 75-80% of RCC), papillary RCC (pRCC) (about 10%), and chromophobe RCC (chRCC) (about 5%) [4,5]. Rapid development in the molecular characterization of ccRCC has facilitated advances in targeted therapy, thus improving median survival of patients with advanced-stage RCC stage from less than 10 months before 2004 to 30 months by 2011 [6]. Specifically, ccRCC is usually characterized by few specific symptoms and/or laboratory abnormalities. About 20% to 30% of individuals with RCC present with advanced-stage disease at initial diagnosis. In addition, approximately 30% of patients who have localized ccRCC experience recurrence or metastasis after tumor-targeted surgery [7,8]. Thus, it is essential to further investigate reliable ccRCC-associated molecular biomarkers, thereby facilitating early diagnosis, monitoring of tumor development, and discovery of novel therapeutic targets [9].

PKMYT1 (also known as *MYT1*), a member of the WEE1 family, exerts a crucial effect on Golgi and endoplasmic reticulum assembly in mammalian cells. It is a kinase that efficiently phosphorylates cell division cycle 2 (cDC2) in both threonine-14 (Thr14) and tyrosine-15 (Thr15) in the African clawed frog, *Xenopus* [10–12]. Throughout the cell cycle, DNA-damage checkpoints can help preserve genomic stability in normal cells, a monitoring mechanism that is commonly deregulated in tumors. Because of the action of tumor-suppressor genes such as *P53*, mutations in which result in deactivation of the G1 checkpoint, most tumor cells are heavily dependent on the G2/M checkpoint, thus ensuring its genomic stability and survival advantage [13,14]. *PKMYT1* is involved in G2 arrest in oocytes and its activity is modulated through Akt phosphorylation [12]. *PKMYT1* serves as a negative modulator of the cell cycle, preventing cells from transforming from G2 to the mitosis phase through 2 pathways [15]. One is binding of *PKMYT1* localized to the cytoplasm to the cDC2/cyclin B complex, which restrains the complex from entering the nucleus [16]. In addition, *PKMYT1* suppresses cDC2 activity by phosphorylating the Thr14/Thr15 residue on cDC2 [17–20]. Considering that *PKMYT1* secures the G2/M phase transition, inhibitors that target *PKMYT1* could potentially effectively diminish the survival ability of tumor cells, which would endowed them with potential in clinical therapy [13]. Currently, an increasing number of studies have demonstrated that overexpression of *PKMYT1* facilitates proliferation, migration, invasion, and colony-forming ability, as well as epithelial-mesenchymal transition, (EMT) in multiple tumors. Specifically, a study has demonstrated that *PKMYT1* is positively associated with polo-like

kinase 1 (*PLK1*) in breast cancer. In breast cancer, the 2 genes may act synergistically to modulate tumor growth by maintaining precise control of the cell cycle and genome stability [13]. *PKMYT1* facilitates the progression of prostate cancer by targeting cyclin B1 and cyclin E1 expression [10]. *PKMYT1* also interacts with microspherule protein 1 (*MCRS1*) – an oncogene known to play a role in several tumors – to promote the malignant phenotype of gastric cancer (GC) [21]. *PKMYT1* can bind and inactivate glycogen synthase kinase 3beta and further activate beta-catenin/T-cell factor signaling, thus facilitating the development of hepatocellular carcinoma (HCC) [22]. *PKMYT1* facilitates development of ovarian cancer (OC) by negatively modulating *SIRT3* [23]. Diminished expression of *PKMYT1* is also accompanied by decreased levels of Notch1, p21, and Hes1, indicating that *PKMYT1* has potential to strengthen the activity of the Notch signal pathway in non-small cell lung cancer (NSCLC) [24]. Silencing *PKMYT1* expression eliminates irradiation-induced G2/M phase arrest and enhances the sensitivity of cancer cells to radiation [15].

Nevertheless, an understanding of the correlation between *PKMYT1* and ccRCC and the effect of *PKMYT1* on ccRCC progression remains elusive, and further investigation in this area is warranted. In the present study, multiple integrated bioinformatics approaches were used to assess the expression, prognostic effect, and potential biological function of *PKMYT1* in ccRCC.

Material and Methods

Extraction of differentially expressed gene

A flow diagram of our research approach is shown in Figure 1. The Cancer Genome Atlas (TCGA) database (<https://tcga-data.nci.nih.gov/tcga/>) was used to acquire publicly available messenger RNA (mRNA) profiles and corresponding clinicopathological data from 539 ccRCC tissues and 72 adjacent normal samples, Information recorded included: sex; race; laterality; tumor, node classification, and metastasis classification; tumor, node metastasis (TNM) stage; histologic grade; tumor status; relapse; and vital status. Differentially expressed genes (DEGs) were identified from TCGA database and GEO datasets (<http://www.ncbi.nlm.nih.gov/geo/>) (including GSE15641 [25], GSE36895 [26] and GSE40435 [27], as well as GSE105261 [28]) through the “DESeq2” package and via the “Limma” package, respectively. The thresholds were |log2-fold change (FC)| >2.0 and false discovery rate (FDR) <0.01 [29].

Weighted gene coexpression network analysis was applied for screening *PKMYT1*

Coexpression methodology with weighted gene coexpression network analysis was transformed into connection weights

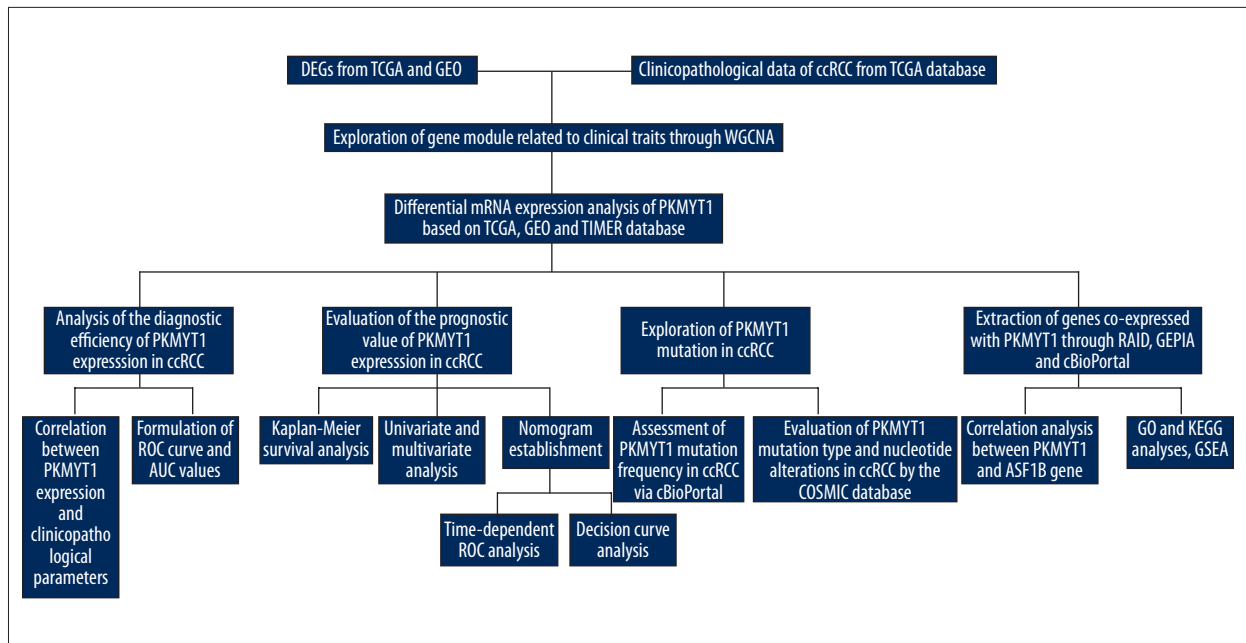


Figure 1. Flowchart of the study design.

or topology overlap measurements to assess the correlations among the identified DEGs [30,31]. The R package of weighted gene coexpression network analysis software was used to formulate a coexpression network for 2369 DEGs in 519 ccRCC tissues with specific clinicopathological data. We considered module significance (MS) as the average gene significance (GS) for all genes in a module [32]. We estimated the association between the module eigengene (ME) and clinical traits to investigate relevant modules. Commonly, the module with the greatest absolute value of MS was considered to be the one the most associated with a clinical trait, which then was selected for further analysis. The module membership (MM) was used to describe the significance of genes in the module. Thus, we identified the *PKMYT1* gene in a module via MM and GS [33]. The levels of expression of *PKMYT1* in different tumors were identified through the Tumor Immune Estimation Resource (TIMER) database (<http://timer.cistrome.org/>).

Survival analysis

Pearson's test was used to assess the correlation between *PKMYT1* expression and clinicopathological variables in ccRCC. Kaplan-Meier plots with log-rank tests were applied to evaluate the association between overall survival (OS) of patients with ccRCC and *PKMYT1* expression. A receiver operating characteristic curve (ROC) with areas under the curve (AUC) was formulated to assess the diagnostic efficiency of *PKMYT1* in ccRCC [34,35]. Cox regression analysis was used to calculate the hazard ratio (HR) with 95% confidence interval (CI) to estimate the prognostic effect of *PKMYT1* in ccRCC.

Nomogram formulation and validation

A nomogram was developed with the “rms” R package to estimate the probability of 3-, 5-, and 10-year OS in patients with ccRCC. A concordance index (C-index) and calibration plot were applied to evaluate the nomogram discrimination and the consistence between nomogram prediction and practical observation, respectively [34,35]. The time-dependent ROC curves were applied to evaluate the predictive capability of the nomogram, *PKMYT1* expression level, and additional clinicopathologic variables for 3-, 5-, and 10-year OS. DCA is a novel statistical method for evaluating the utility of a predictive model in clinical decision-making and to help select the best model for use in clinical practice [36]. DCA can estimate the clinical net benefit of every model in contrast to all or no strategies. The best model is the one with the greatest net benefit, as calculated [35,37]. We conducted DCA to compare the potential predictive benefit of the nomogram, *PKMYT1* expression, and traditional clinical parameters.

PKMYT1 mutations analysis

The cBioPortal database (<https://www.cbioportal.org/>) (last accessed May 20, 2020) was used to analyze the frequency of *PKMYT1* mutations in ccRCC. The Catalogue of Somatic Mutations in Cancer (COSMIC) (<https://cancer.sanger.ac.uk/cosmic/>) is a high-resolution database used to investigate the impact of somatic mutations on multiple human cancers, which was used to explore mutations in *PKMYT1* in ccRCC [38,39].

Gene coexpression analysis

We identified long noncoding RNAs (lncRNAs), micro-RNAs (miRNAs), mRNA, and RNA-binding protein (RBP), as well as transcription factors (TFs), that interact with *PKMYT1* using the RNA InterActome (RAID) database (RAID v2.0, www.ma-society.org/raid/) [40]. The Gene Expression Profiling Interactive Analysis (GEPIA) (<http://gepia.cancer-pku.cn/index.html>) is a database-based resource which incorporates 9736 tumors and 8587 normal tissues from TCGA and the Genotype-Tissue Expression datasets, thus analyzing RNA sequencing expression [41]. The correlation coefficient was generated via the Spearman method, which was used to estimate the degree of gene expression correlation. *PKMYT1* and the genes positively coexpressed with *PKMYT1* expression in ccRCC are shown on the x-axis and the y-axis, respectively.

Function enrichment analyses

The Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analyses were formulated by using the clusterProfiler package (version 3.14.3). Thus, the biological attributes of the genes that were positively correlated with *PKMYT1* were investigated [42]. Gene Set Enrichment Analysis (GSEA) (<https://www.broadinstitute.org/gsea/>) was used to determine the significance of the potential biological mechanisms in the high and low *PKMYT1* expression groups. Gene sets with a normalized (NOM) $P < 0.05$ and FDR < 0.25 were considered significantly enriched.

Statistical analysis

R language software (R 3.6.3 version) was used for statistical analysis. A Wilcoxon rank sum test was conducted to compare *PKMYT1* expression in ccRCC tissues and adjacent non-tumor samples using TCGA and GEO datasets. All ccRCC cases were stratified into the high- or low-expression group according to the median value of *PKMYT1* expression. The correlation between clinicopathological parameters and *PKMYT1* mRNA level was estimated via the chi-square test and the Fisher exact test. The Kaplan-Meier method was used to compare OS. Univariate and multivariate Cox analyses were formulated to determine the correlations between *PKMYT1* mRNA level and OS and with additional clinicopathological parameters. Pearson correlation analysis and Spearman correlation analysis were used to assess the relevance of *PKMYT1* and *ASF1B* gene expression values. $P < 0.05$ was deemed to be statistically significant (* $P < 0.05$; ** $P < 0.01$; *** $P < 0.001$).

Results

***PKMYT1* is a differentially expressed gene significantly associated with histological grade in ccRCC**

Based on the cutoff value of FDR < 0.05 and $|FC| > 2$, there were 2369 DEGs between ccRCC and normal kidney samples, including 1115 upregulated and 1254 downregulated DEGs. The heat map and volcano plot for mRNA profiles are shown in Figure 2A and 2B, respectively. A total of 2369 DEGs were clustered to formulate the coexpression network through weighted correlation network analysis [43]. The soft threshold power value of 3 defined the adjacency matrix, and MEs up to 0.75 were merged (Supplementary Figure 1). Ten different gene coexpression modules were identified in ccRCC after the insignificant gray module was excluded (Figure 2C). The results of an eigengene connectivity analysis of those modules are shown in Figure 2D and 2E. The yellow module, which was most significantly correlated with histological grade of ccRCC, was pivotal for predicting development of ccRCC ($r = 0.31$; $P = 1e-12$) (Figure 2F). *PKMYT1* was one of the most significant genes in the yellow module ($GS = 0.951435$, $MM = 0.935643$) (Figure 2G), which indicates that *PKMYT1* is one of the DEGs that is the most significantly correlated with histological grade in ccRCC and it potentially predicts the prognosis of ccRCC based on histological grade.

***PKMYT1* is significantly upregulated in ccRCC**

As shown in Figure 3A, *PKMYT1* expression was significantly increased in multiple solid tumors, especially in bladder urothelial carcinoma and esophageal carcinoma. Mining TCGA and multiple GEO datasets (including GSE15641, GSE36895, GSE40435, and GSE105261) further demonstrated that *PKMYT1* was markedly upregulated in ccRCC samples compared with normal kidney tissues (Figure 3B–3F).

***PKMYT1* expression is associated with clinicopathological parameters in ccRCC**

The association between *PKMYT1* expression and clinicopathological variables in 530 cases of ccRCC from the TCGA database was further analyzed through Pearson's χ^2 test. As shown in Table 1, *PKMYT1* was significantly correlated with T classification ($P < 0.0001$), N classification ($P = 0.027$), M classification ($P = 0.0058$), TNM stage ($P = 0.0021$), histologic grade ($P = 0.0002$), tumor status ($P = 0.0037$), status of relapse ($P = 0.0065$), and vital status ($P = 0.0009$). More specifically, *PKMYT1* expression was enhanced in patients whose tumors had more advanced T, N, M classification, TNM stage, and histological grade, and those in whom ccRCC had relapsed (Figure 4A–4G). There was no significant association between *PKMYT1* mRNA levels and additional clinicopathological factors, such as age, sex, and laterality of the neoplasm (Supplementary Figure 2A–2C).

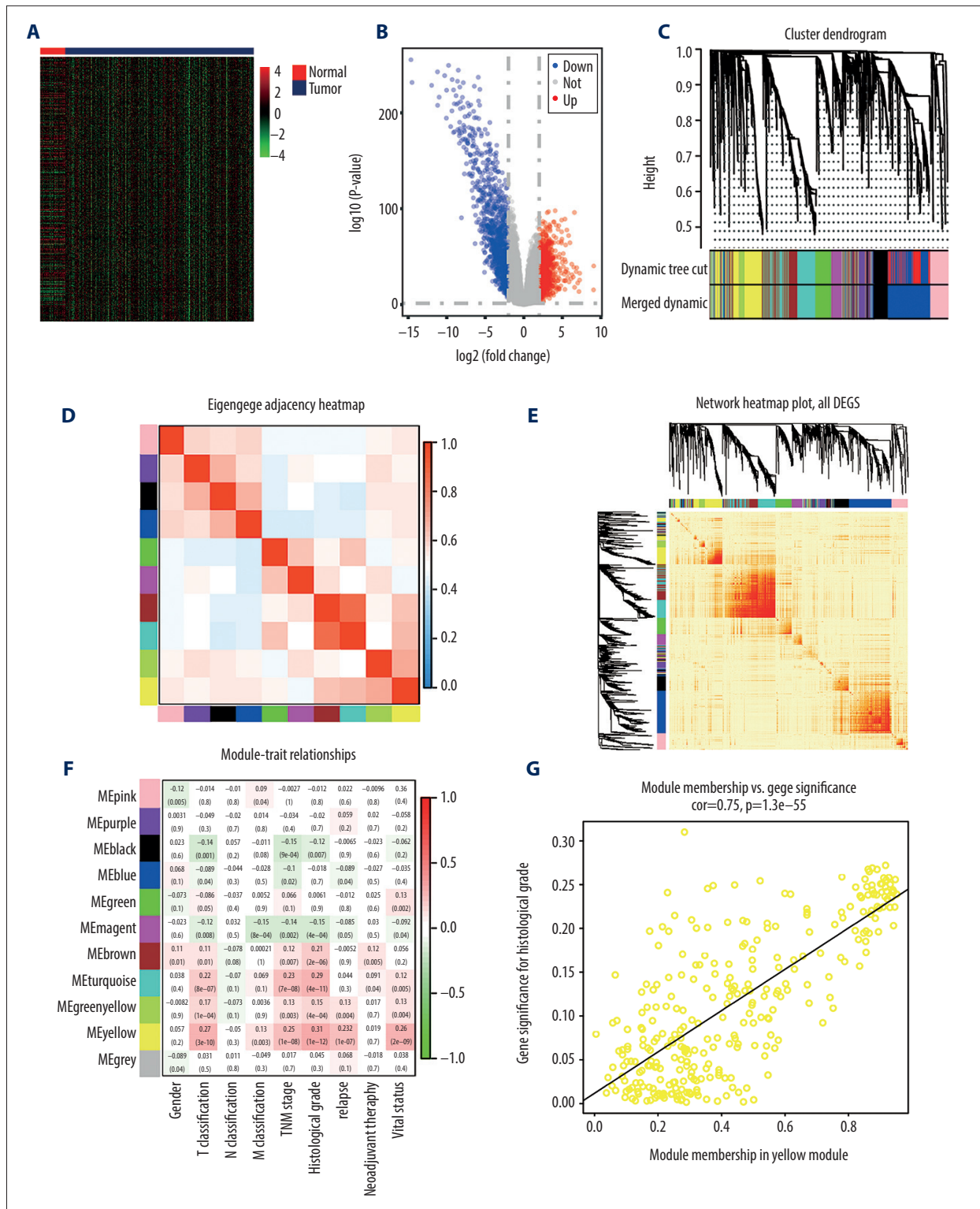


Figure 2. Bioinformatic analysis of ccRCC and normal kidney samples in the TCGA database. **(A, B)** The heatmap and volcano plot of the DEGs between 539 ccRCC cases and 72 normal samples. **(C)** Cluster dendrogram of DEGs and module screening based on gene expression profiles of 539 ccRCC samples. **(D, E)** Heatmap of the correlation coefficient expressed between modules. **(F)** Relationships between consensus module eigengenes and various clinical traits. **(G)** Analogous scatter plots for the yellow module. ccRCC – clear cell renal cell carcinoma; TCGA – The Cancer Genome Atlas; DEGs – differentially expressed genes.

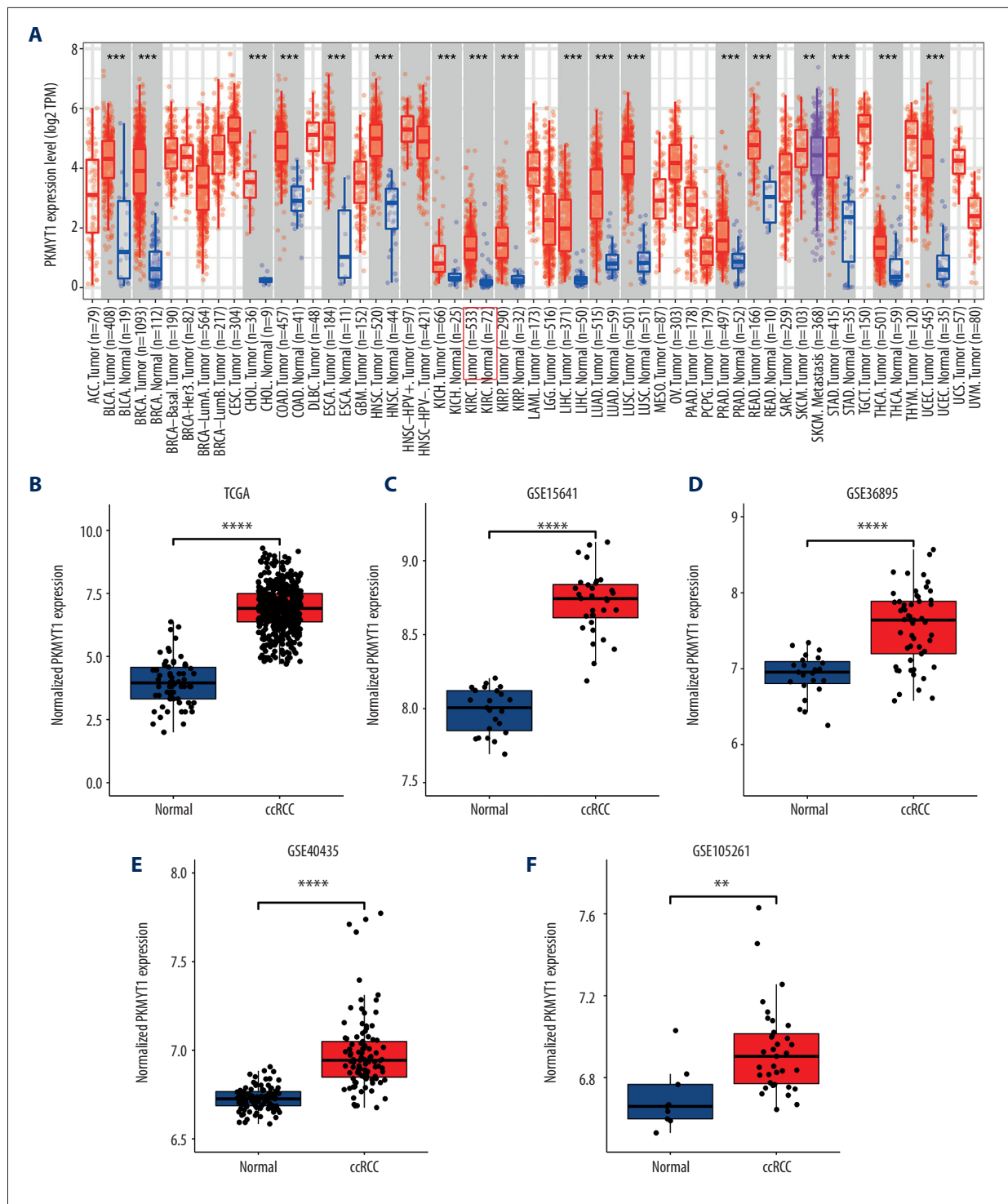


Figure 3. *PKMYT1* is prominently overexpressed in ccRCC. **(A)** *PKMYT1* mRNA expression in multiple neoplasms and adjacent normal samples from the TIMER database. *PKMYT1* mRNA level in ccRCC and normal kidney tissues from **(B)** the TCGA database, **(C)** GSE15641, **(D)** GSE36895, **(E)** GSE40435, and **(F)** GSE105261. ccRCC – clear cell renal cell carcinoma; TIMER – Tumor Immune Estimation Resource.

Table 1. Correlation between *PKMYT1* mRNA levels and clinicopathological parameters in 530 patients with ccRCC from the TCGA database.

Variable	Groups	N	Low	High	χ^2	P value
Age (years)	≤60	264	129	135	0.27136	0.6024
	>60	266	137	129		
	NA					
Sex	Male	344	164	180	2.2003	0.138
	Female	186	102	84		
	NA					
Race	White	459	226	233	0.4298*	
	Asian	8	4	4		
	Black	56	33	23		
	NA	7				
Laterality	Right	280	148	132	1.5888	0.2075
	Left	249	117	132		
	Bilateral	1				
	NA					
T classification	T1	271	149	122	22.122	P<0.0001
	T2	69	44	25		
	T3	179	72	107		
	T4	11	1	10		
	TX					
	NA					
N classification	N0	239	121	118	4.8908	0.027
	N1	16	3	13		
	NX	275				
	NA					
M classification	M0	440	231	209	7.6072	0.0058
	M1	80	28	52		
	MX	10				
	NA					
TNM stage	I	265	145	120	14.688	0.0021
	II	57	36	21		
	III	123	55	68		
	IV	82	29	53		
	NA	3				

Table 1 continued. Correlation between *PKMYT1* mRNA levels and clinicopathological parameters in 530 patients with ccRCC from the TCGA database.

Variable	Groups	N	Low	High	χ^2	P value
Histologic grade	G1	14	11	3	19.3	0.0002
	G2	227	127	100		
	G3	206	99	107		
	G4	75	23	52		
	NA	8				
Tumor status	With tumor	138	54	84	8.4434	0.0037
	Tumor-free	358	194	164		
	NA	34				
Relapse	No	373	202	171	7.3991	0.0065
	Yes	157	64	93		
	NA					
Vital status	Dead	172	68	104	10.939	0.0009
	Alive	358	198	160		
	NA					

* P value of Fisher's exact test conducted on a small sample. mRNA – messenger RNA; ccRCC – clear cell renal cell carcinoma; TCGA – The Cancer Genome Atlas.

***PKMYT1* for diagnosing and predicting prognosis of ccRCC**

An ROC curve was used to further clarify the diagnostic efficiency of *PKMYT1* expression in ccRCC cases for which information was extracted from the TCGA dataset. It showed that expression of the gene was satisfactory for differentiating between ccRCC and tissue from healthy individuals, achieving an AUC of 0.942 (95% CI: 0.906–0.978) (Figure 5A). In addition, the AUC values for the ability of *PKMYT1* to distinguish between normal renal tissue and ccRCC at TNM stages I, II, III, and IV were 0.946 (95% CI: 0.908–0.985), 0.899 (95% CI: 0.833–0.965), 0.951 (95% CI: 0.914–0.988), and 0.971 (95% CI: 0.942–0.999), respectively (Figure 5B–5E). Similarly, *PKMYT1* also demonstrated significant diagnostic value in distinguishing normal tissue from ccRCC of histological grades G1, G2, G3, and G4, with AUCs of 0.944 (95% CI: 0.897–0.991), 0.932 (95% CI: 0.890–0.973), 0.954 (95% CI: 0.919–0.989), and 0.971 (95% CI: 0.943–0.999), respectively (Figure 5F–5I). These findings demonstrate that *PKMYT1* is a potentially reliable and promising biomarker for diagnosis of even early-stage ccRCC.

To determine whether *PKMYT1* exerted a significant effect on the prognosis of ccRCC, all ccRCC cases for which information was extracted from the TCGA database were stratified into high- and low-expression groups, based on the median value

for *PKMYT1* expression. The Kaplan-Meier curve with log-rank test showed that prognosis of ccRCC was worse in those with high expression of *PKMYT1* than in those with low expression of the gene (HR=1.67, 95% CI: 1.23–2.27, P=0.001) (Figure 6A). High *PKMYT1* expression indicated shorter OS than low *PKMYT1* expression in men (HR=1.81, 95% CI: 1.23–2.67, P=0.002) and older patients (HR=1.61, 95% CI: 1.10–2.37, P=0.015) with TNM stage II ccRCC (HR=8.36, 95% CI: 1.83–38.2, P=0.006), histological grade G4 disease (HR=1.71, 95% CI: 1.00–2.91, P=0.049), and relapse (HR=1.69, 95% CI: 1.14–2.51, P=0.009) (Figure 6B–6F).

Cox regression analysis was conducted to assess the prognostic performance of *PKMYT1* mRNA level in ccRCC. Univariate analysis showed that high *PKMYT1* expression (HR=1.82, 95% CI: 1.32–2.52, P<0.001) was significantly associated with inferior OS (Table 2). The multivariate analysis also found that a high *PKMYT1* mRNA level (HR=1.79, 95% CI: 1.29–2.49, p=0.001) had the ability to predict unsatisfactory prognosis. Thus, high *PKMYT1* expression can be considered an independent risk factor for OS in ccRCC patients.

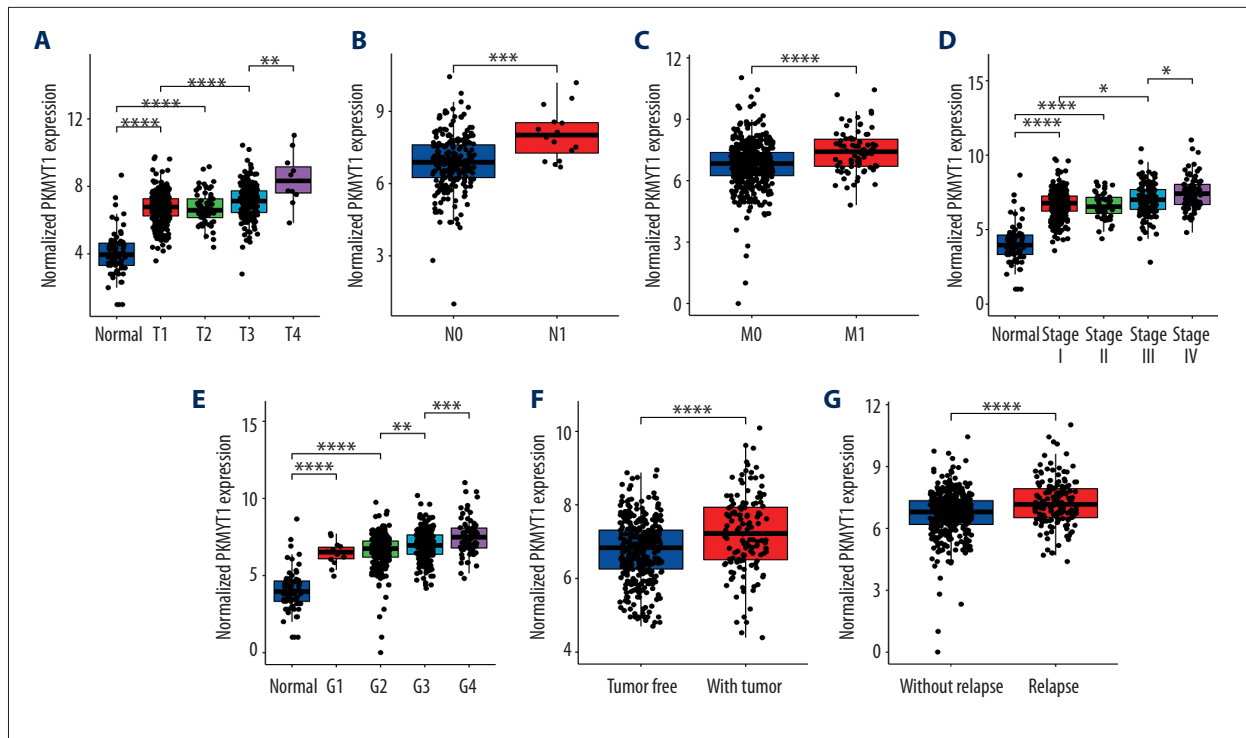


Figure 4. *PKMYT1* mRNA expression is significantly correlated with diversiform clinical variables in ccRCC cases from the TCGA database. *PKMYT1* in individuals with ccRCC stratified by (A) T classification, (B) N classification, (C) M classification, (D) TNM stage, (E) histologic grade, (F) tumor status, and (G) the status of relapse. ccRCC – clear cell renal cell carcinoma; TCGA – The Cancer Genome Atlas; TNM – tumor, node, metastasis.

A novel nomogram model was established and validated using ROC and DCA

We next formulated a nomogram that incorporated statistically significant clinical parameters identified by multivariate analysis, thus predicting 3-, 5-, and 10-year OS for patients with ccRCC. Histologic grade made the greatest contribution to prognosis, followed by tumor status, TNM stage, and *PKMYT1* expression (Figure 7A). The nomogram performed favorably for prediction, with a C-index of 0.784. The calibration plot for the probability of survival at 3, 5, and 10 years revealed great concordance between the prediction of the nomogram and actual observations (Figure 7B–7D).

We further performed ROC analysis to compare the predictive efficiency of our nomogram with that for *PKMYT1* expression, histological grade, TNM stage, and tumor status. In the ROC curves of 3-year OS in the TCGA dataset, our nomogram displayed the highest AUC value of 0.808 (95% CI: 0.779–0.867), followed by that for TNM stage (0.765, 95% CI: 0.659–0.797), tumor status (0.724, 95% CI: 0.687–0.767), histological grade (0.697, 95% CI: 0.662–0.717), and *PKMYT1* expression (0.659, 95% CI: 0.599–0.707) (Figure 8A). Similarly, the AUC for the nomogram in predicting 5-year OS achieved a value of 0.758 (95% CI: 0.693–0.801), which was greater than those for TNM stage

(0.716, 95% CI: 0.654–0.736), *PKMYT1* expression (0.699, 95% CI: 0.628–0.746), histological grade (0.669, 95% CI: 0.592–0.698), and tumor status (0.596, 95% CI: 0.518–0.622) (Figure 8B). The AUC values for the nomogram, tumor status, histological grade, TNM stage, and *PKMYT1* expression for 10-year OS were 0.739 (95% CI: 0.652–0.863), 0.678 (95% CI: 0.592–0.723), 0.661 (95% CI: 0.604–0.738), 0.653 (95% CI: 0.588–0.768), and 0.639 (95% CI: 0.572–0.743) (Figure 8C). These results underscore that our nomogram, which comprised *PKMYT1* expression level and other clinical parameters, is an optimal model for predicting long-term prognosis of ccRCC. *PKMYT1* expression had a slightly lower or similar predictive performance for predicting prognosis of ccRCC compared with that of the additional clinical indicators (including histologic grade, TNM stage, and tumor status).

In addition, we developed DCA curves to assess the clinical effect of our nomogram and other indicators, thus helping us to visually estimate their utility. As shown in Figure 8D, the DCA of our nomogram showed the greatest net benefits, followed by histological grade, *PKMYT1* expression, TNM stage, and tumor status, which suggests that *PKMYT1* expression level has relatively satisfactory clinical utility for predicting the prognosis of ccRCC.

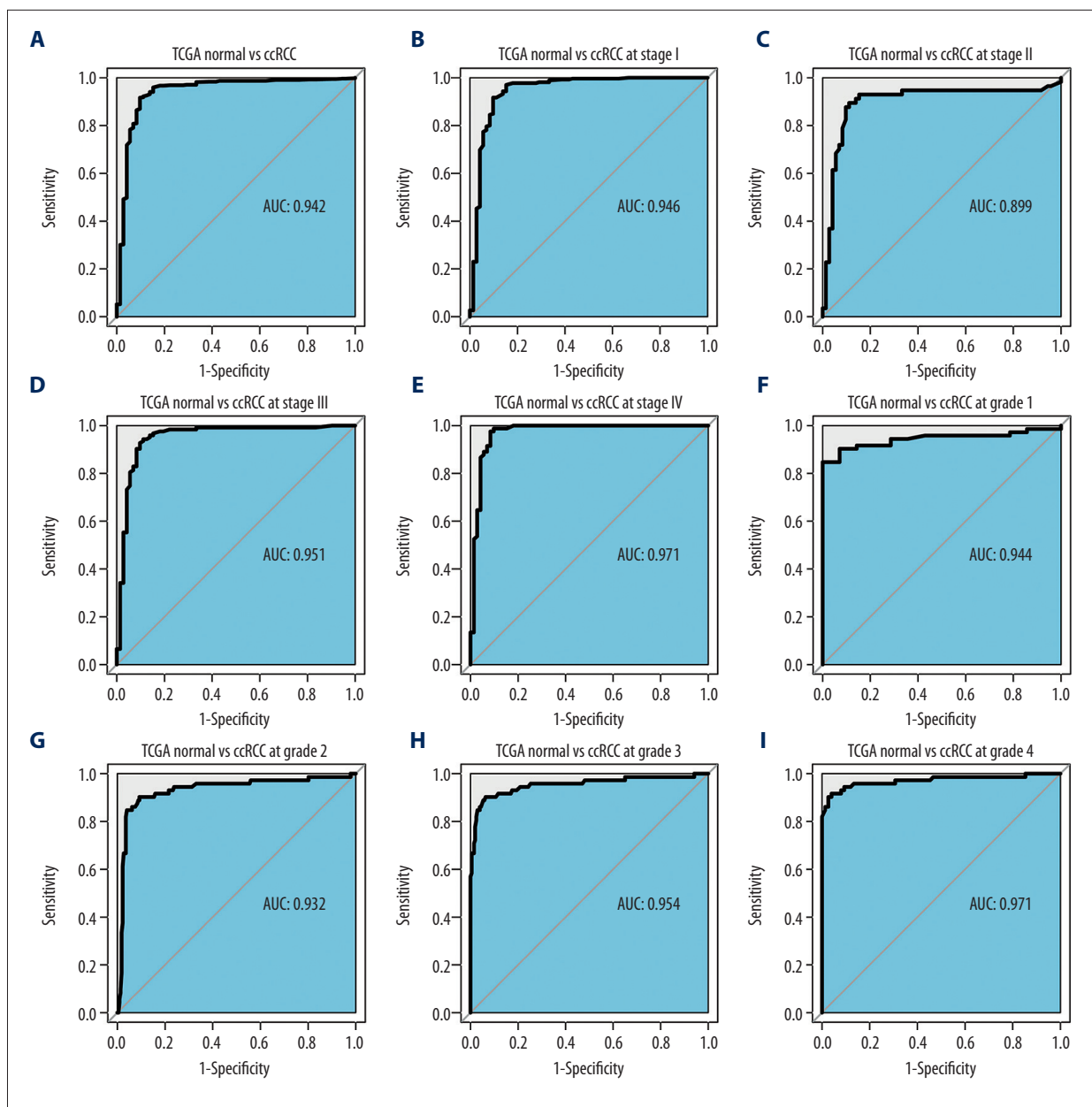


Figure 5. *PKMYT1* can effectively diagnose ccRCC. ROC curves reveal that *PKMYT1* expression can differentiate adjacent normal tissues from (A) ccRCC and ccRCC at (B) TNM I stage, (C) TNM II stage, (D) TNM III stage, (E) TNM IV stage, (F) histological stage 1, (G) histological stage 2, (H) histological stage 3, and (I) histological stage 4. ccRCC – clear cell renal cell carcinoma; ROC: receiver operating characteristic; TNM – tumor, node, metastasis.

***PKMYT1* mutations are rare in ccRCC**

The mutation frequency of *PKMYT1* in ccRCC was assessed through cBioPortal, which accounted for 0.1% (Figure 9A). Specifically, the mutant types of *PKMYT1* in ccRCC were investigated through the COSMIC database. As shown in Figure 9B, missense mutation was the most frequent type of *PKMYT1* mutation in ccRCC (27.91%), followed by synonymous mutation (11.63%), nonsense mutation (2.03%), and frameshift

mutation (0.87%). Nucleotide alterations primarily consisted of C>T, G>A, C>A and G>C mutations, the greatest proportion of which (33.57%) were C>T (Figure 9C).

***PKMYT1* positively modulates cell proliferation in ccRCC**

Using the RAID database, we investigated molecules that have potential to interact with *PKMYT1*. Specifically, *PKMYT1* displayed crosstalk with 3 lncRNAs (lncMTX2, TUG1, and

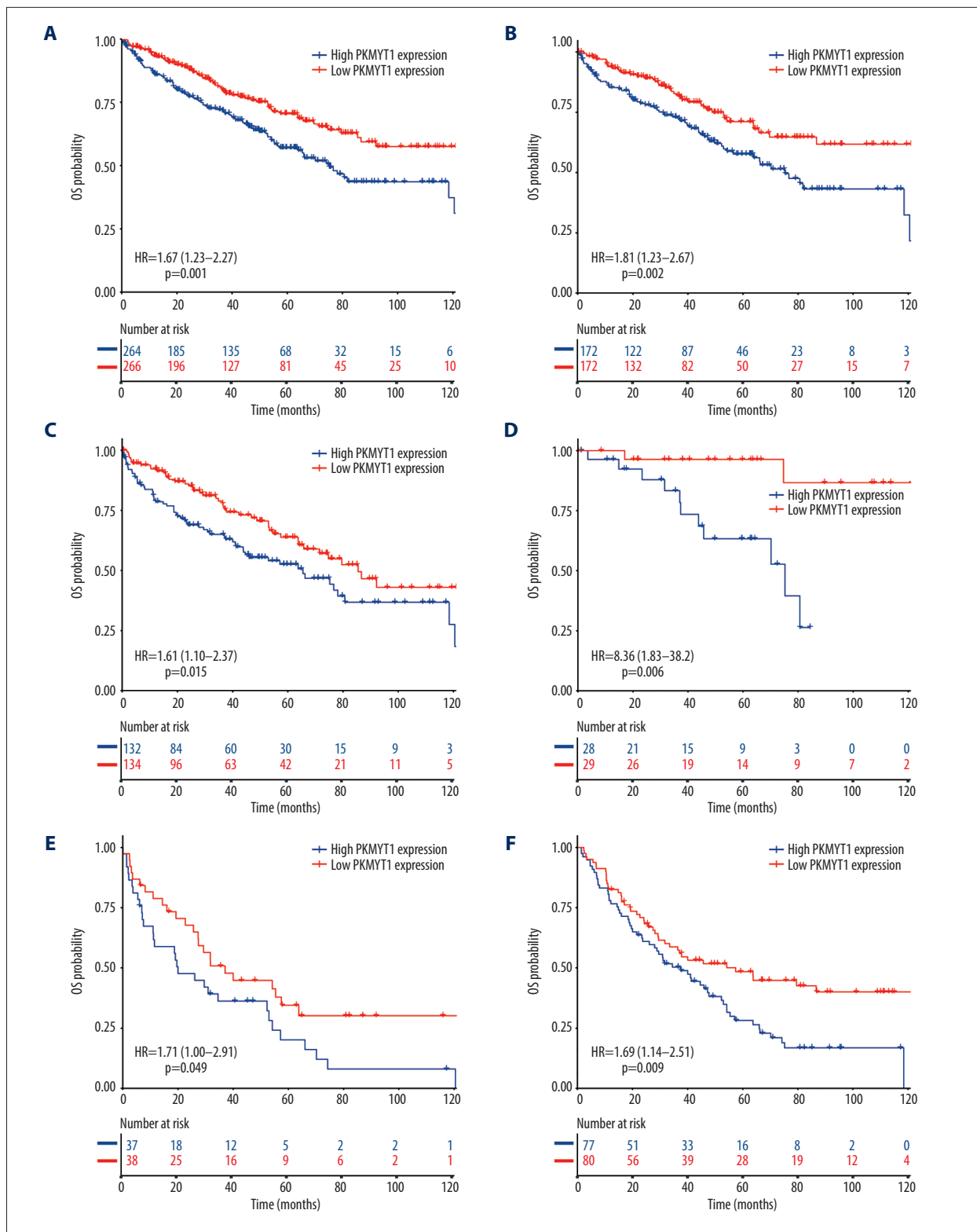


Figure 6. Overexpressed *PKMYT1* exerts negative effects on ccRCC prognosis. Survival analysis reveals OS between high *PKMYT1* and low *PKMYT1* mRNA levels in (A) all ccRCC cases from the TCGA database, (B) male patients, (C) cases with age >60 years, (D) TNM stage II, (E) cases at histological grade 4, and (F) ccRCC that has relapsed. ccRCC – clear cell renal cell carcinoma; OS – overall survival; TCGA – The Cancer Genome Atlas; TNM, tumor, node, metastasis.

Table 2. Univariate and multivariate analyses of OS in patients with ccRCC in the TCGA database.

Variable	Mean survival (Mo)	95% CI	Univariate analysis		Multivariate analysis	
			HR (95% CI)	P value	HR (95% CI)	P value
Total	44.9	42.0–47.9				
Age (years)						
≤60	46.9	42.6–51.2	Ref			
>60	43	39–47.1	1.61 (1.17–2.21)	0.004	1.25 (0.9–1.75)	0.188
Sex						
Female	45	39.6–50.3	Ref			
Male	44.9	41.4–48.5	0.96 (0.69–1.34)	0.811		
Race						
Asian	30.6	17.1–44.2	Ref			
Black	26.2	19.2–33.1	2.01 (0.26–15.7)	0.507		
White	47.3	44.1–50.4	1.89 (0.26–13.52)	0.527		
T classification						
T1+T2	47.6	44–51.3	Ref			
T3+T4	41	36–46.1	2.88 (2.09–3.97)	<0.001	0.64 (0.34–1.2)	0.164
N classification						
N0	45.2	40.8–49.5	Ref			
N1	27.2	11.7–42.6	2.81 (1.36–5.83)	0.005		
M classification						
M0	47.1	43.9–50.3	Ref			
M1	35.2	27.4–43.1	4.57 (3.3–6.32)	<0.001	1.31 (0.84–2.05)	0.236
TNM stage						
I+II	48	44.4–51.7	Ref			
III+IV	40.2	35.3–45	3.61 (2.59–5.03)	<0.001	2.54 (1.19–5.41)	0.016
Histologic grade						
G1+G2	46.6	42.4–50.9	Ref			
G3+G4	43.5	39.4–47.6	2.63 (1.83–3.76)	<0.001	1.57 (1.06–2.32)	0.024
Tumor status						
Tumor-free	46.5	43.1–50	Ref			
With tumor	40.9	35.3–46.4	5.41 (3.89–7.53)	<0.001	4.75 (2.32–9.75)	<0.001
Laterality						
Left	43.5	39.4–47.7	Ref			
Right	46.2	42–50.3	0.66 (0.48–0.91)	0.010	0.73 (0.53–1.01)	0.057
Relapse						
No	45.3	41.8–48.8	Ref			
Yes	44.2	38.7–49.8	4.17 (3.0–5.8)	<0.001	0.64 (0.31–1.32)	0.225

Table 2 continued. Univariate and multivariate analyses of OS in patients with ccRCC in the TCGA database.

Variable	Mean survival (Mo)	95% CI	Univariate analysis		Multivariate analysis	
			HR (95% CI)	P value	HR (95% CI)	P value
Neoadjuvant treatment history						
No	45.1	42.1–48.1	Ref			
Yes	39.7	24–55.3	2.19 (1.15–4.16)	0.017	1.22 (0.62–2.39)	0.561
PKMYT1 expression						
Low	47.7	43.7–51.7	Ref			
High	42.1	37.8–46.4	1.82 (1.32–2.52)	<0.001	1.79 (1.29–2.49)	0.001

OS – overall survival; ccRCC – clear cell renal cell carcinoma; TCGA – The Cancer Genome Atlas; HR – hazard ratio; 95% CI – 95% confidence interval.

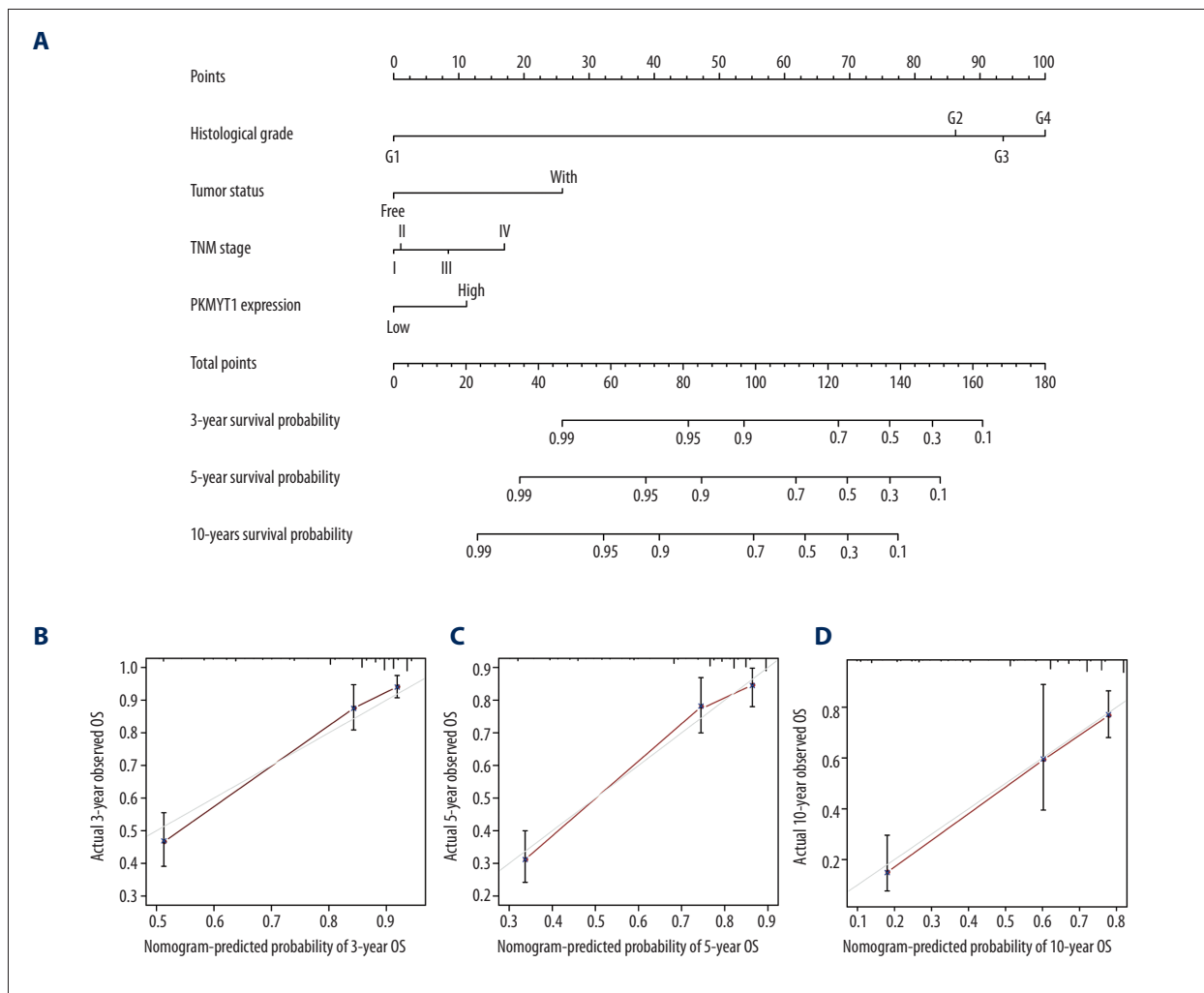


Figure 7. Nomogram model incorporating *PKMYT1* mRNA expression and significant clinical variables to predict OS of ccRCC. (A) Nomogram for predicting the 3-, 5-, and 10-year OS of patients with ccRCC from the TCGA database. The calibration curves of OS at (B) 3, (C) 5, and (D) 10 years for ccRCC. OS – overall survival; ccRCC – clear cell renal cell carcinoma; TCGA – The Cancer Genome Atlas.

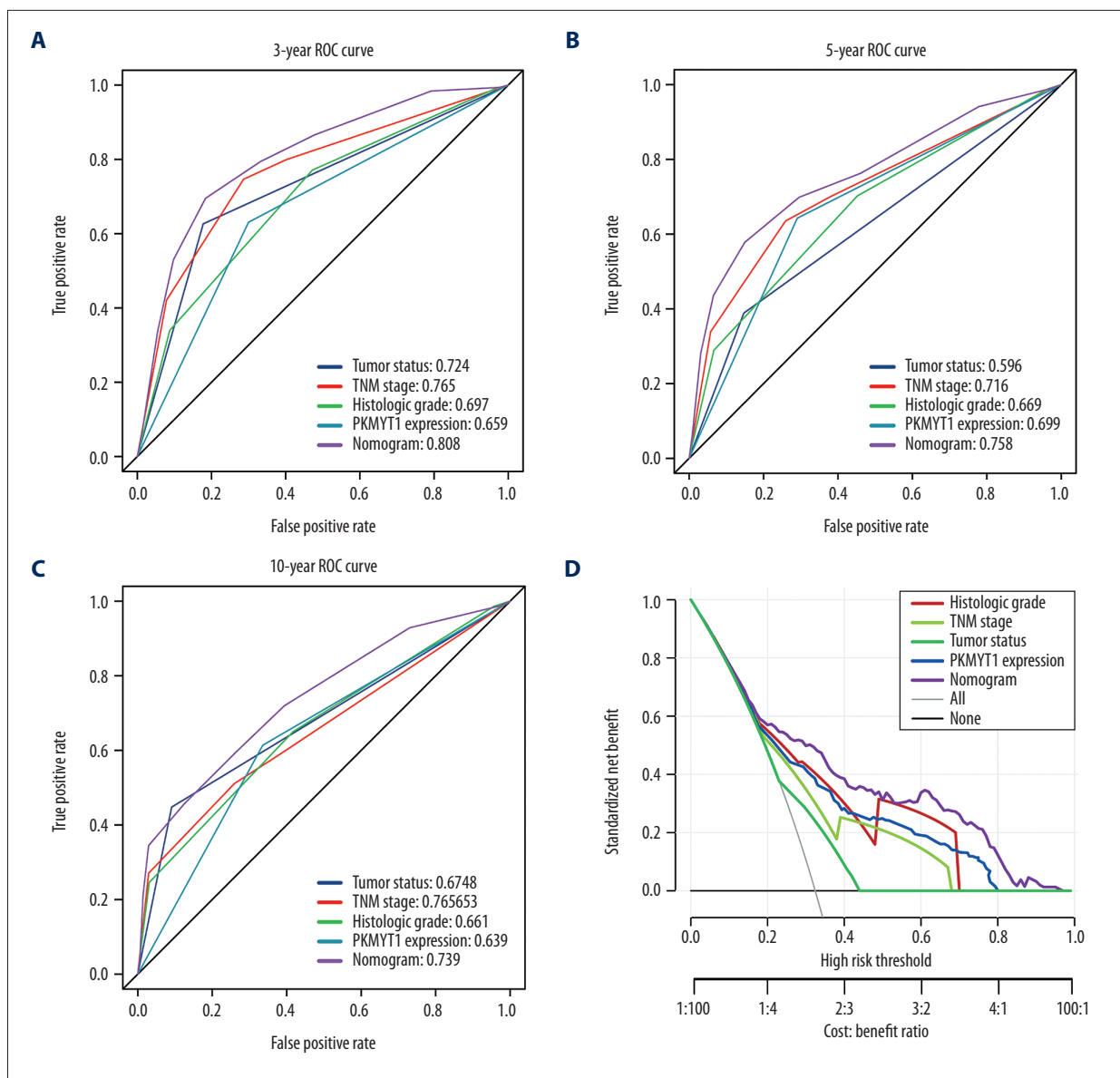


Figure 8. The predictive value and clinical utility of *PKMYT1* expression level was assessed through time-dependent ROC and DCA. ROC curves of nomogram, *PKMYT1* expression level, and additional clinical variables for (A) 3-, (B) 5- and (C) 10-year OS of ccRCC. (D) DCA curves of above parameters for prediction of OS in ccRCC. ROC – receiver operating characteristic; DCA – decision curve analysis; OS – overall survival; ccRCC – clear cell renal cell carcinoma.

linc00673), multiple miRNAs (such as has-miR-20b-5p, has-miR-371b-5p, has-miR-519d-3p, and has-miR-106b-5p), mRNAs (such as AR, CSF3R, CCAR1, KIAA1109, and SIGLEC1), RBPs (such as WDR33, GRWD1, CPSF1, METAP2, and FMR1), and TFs (such as HNF4A, SWARCC2, TAF1, REST, and FXR1) (Figure 10A). Furthermore, TCGA database (Figure 10B), the cBioPortal dataset (Figure 10C), and the GEPIA dataset (Table 3) were overlapped to select 52 common genes that are positively coexpressed with *PKMYT1* in ccRCC (Figure 10D). The functional and pathway enrichment analysis was conducted to explore the biological significance of these coexpressed genes. GO analysis

demonstrated that pathway enrichment with the top 15 genes was associated with organelle fission, nuclear division, cell cycle G2/M phase transition, and signal transduction by p53 class mediator (Figure 11A), which was roughly consistent with the findings of the KEGG analysis (Figure 11B). As revealed in the results from GSEA, high expression of *PKMYT1* was positively correlated with the cell cycle (NES=2.389, P<0.0001), p53 signaling pathway (NES=2.164, P<0.0001), DNA replication (NES=1.908, P=0.009), and base excision repair (NES=1.854, P=0.007) (Figure 11C). Therefore, *PKMYT1* plays an essential role in facilitating cell proliferation in ccRCC.

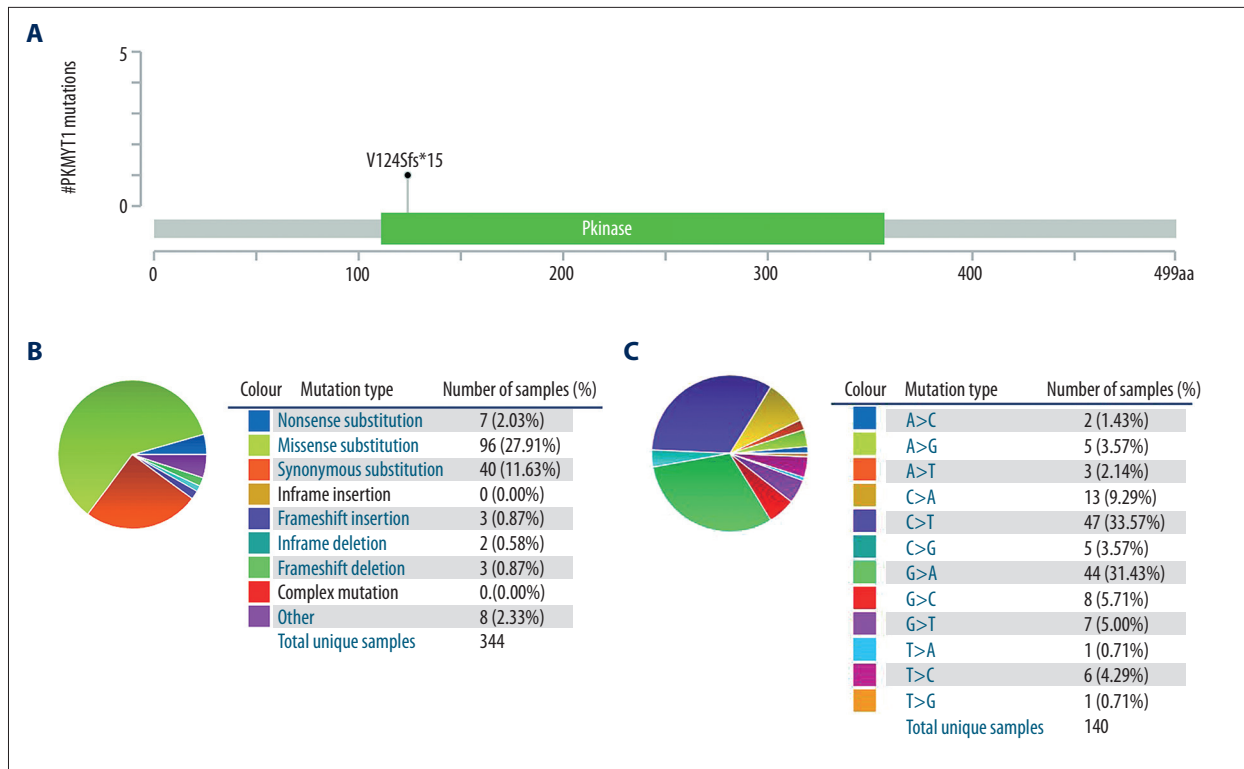


Figure 9. *PKMYT1* mutations are infrequent in ccRCC. (A) Schematic diagram describing *PKMYT1* mutations in ccRCC based on the cBioportal. (B, C) Different *PKMYT1* mutation types and the corresponding mutation proportions in ccRCC. ccRCC – clear cell renal cell carcinoma.

Expressions of *PKMYT1* and *ASF1B* in ccRCC are significantly correlated

To further explore the underlying modulation of *PKMYT1* in ccRCC, data extraction was conducted in a ccRCC cohort via cBioPortal. The regression analysis revealed that *PKMYT1* and the anti-silencing function 1B (*ASF1B*) gene displayed the greatest correlation coefficient (Spearman's correlation=0.85, $P=1.07e-145$; Pearson's correlation=0.89, $P=4.11e-174$) (Figure 12A). Similarly, data based on GEPIA also demonstrated that there was a strongly positive correlation between *PKMYT1* and *ASF1B* transcription ($R=0.65$, $P<0.0001$) (Figure 12B). A heat map in accordance with the TCGA database further confirmed the previously described positive correlation (Figure 12C). These findings suggest that *PKMYT1* may be involved in *ASF1B* signaling pathways, and may be a functional partner of *ASF1B* in ccRCC. An analysis of RNA sequencing and clinical information extracted from TCGA dataset demonstrated that levels of *ASF1B* expression were significantly enhanced in ccRCC compared with normal tissue (Figure 12D). Furthermore, based on the Human Protein Atlas (HPA) database, a high *ASF1B* level was significantly correlated with diminished OS in ccRCC ($P=0.006$) (Figure 12E).

Discussion

Globally, RCC is the 9th most common tumor in men and the 10th most common in women, resulting in over 140 000 cancer-associated deaths annually [44–46]. ccRCC is the most frequent histological subtype of RCC, with high likelihood of invasion, metastasis, and resistance to chemotherapy, which makes eliminating this neoplasm more challenging [47,48]. Thus, a growing number of studies are focusing on the pathogenesis and diagnosis of and therapy for ccRCC.

In our study, we showed the level of expression and clinical significance of *PKMYT1* in ccRCC for the first time. Initially, *PKMYT1* was found to be one of the genes significantly associated with histologic progression of ccRCC via the WGCNA method. *PKMYT1* mRNA levels were prominently increased in ccRCC compared with adjacent normal kidney samples in the TIMER, TCGA, and multiple GEO datasets. Notably, *PKMYT1* mRNA levels were positively correlated with TNM stage and histological grade of ccRCC. ROC analysis demonstrated that *PKMYT1* could distinguish healthy individuals from patients with ccRCC. In addition, enhanced *PKMYT1* levels also corresponded to unsatisfactory prognosis in patients who were male, older, had TNM stage II or histological grade G4 disease, or disease that had relapsed. *PKMYT1* expression was further

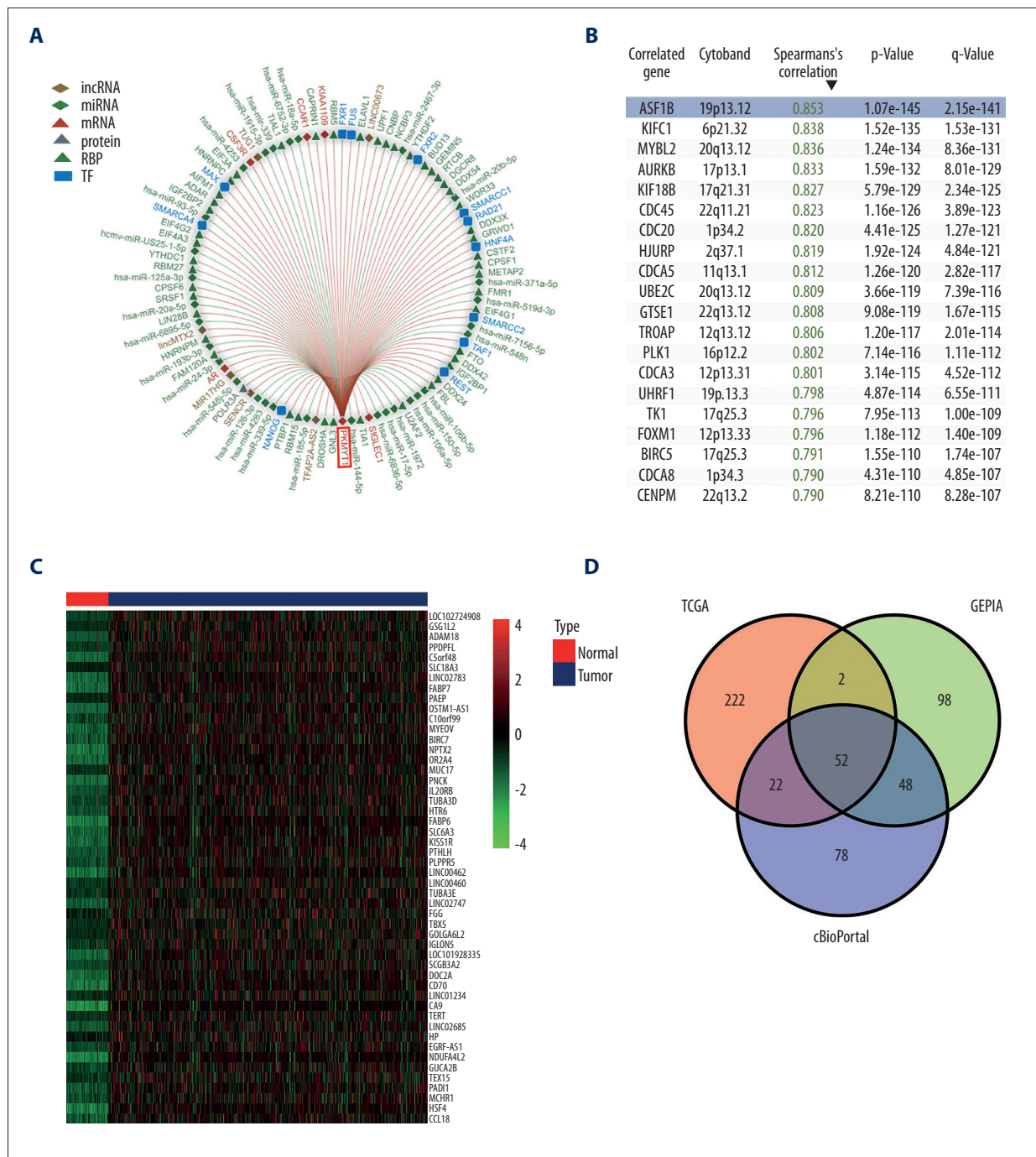


Figure 10. A series of genes coexpressed with *PKMYT1* are revealed based on multi-database mining methods. **(A)** RAID map concerning miRNA, lncRNA, RBP, and TF interacting with *PKMYT1*. **(B)** The top 20 genes positively correlated with *PKMYT1* expression in ccRCC based on the cBioPortal. **(C)** Heatmap of the top 50 DEGs that positively are coexpressed with *PKMYT1* in ccRCC through overlapping the TCGA, cBioPortal and GEPIA database. Abbreviations: RAID – RNA InterActome Database; miRNA – microRNA; lncRNA – long noncoding RNA; RBP – RNA-binding protein; TF – transcription factor; ccRCC– clear cell renal cell carcinoma; TCGA – The Cancer Genome Atlas; GEPIA – Gene Expression Profiling Interactive Analysis.

Table 3. The top 20 genes positively correlated with *PKMYT1* mRNA expression in ccRCC from the GEPIA database.

Gene symbol	Gene ID	PCC
<i>SPC24</i>	ENSG00000161888.11	0.85
<i>CENPM</i>	ENSG00000100162.14	0.77
<i>TOP2A</i>	ENSG00000131747.14	0.76
<i>RECQL4</i>	ENSG00000160957.12	0.74
<i>C16orf59</i>	ENSG00000162062.14	0.72
<i>PTTG1</i>	ENSG00000164611.12	0.72
<i>RP11-360L9.7</i>	ENSG00000253174.2	0.71
<i>POLD1</i>	ENSG00000062822.12	0.71
<i>TK1</i>	ENSG00000167900.11	0.71
<i>CDC20</i>	ENSG00000117399.13	0.69
<i>CDC45</i>	ENSG00000093009.9	0.69
<i>TROAP</i>	ENSG00000135451.12	0.69
<i>B3GNTL1</i>	ENSG00000175711.8	0.67
<i>RNASEH2A</i>	ENSG00000104889.4	0.67
<i>ATP5G1P4</i>	ENSG00000227440.1	0.66
<i>DDX39A</i>	ENSG00000123136.14	0.66
<i>BIRC5</i>	ENSG00000089685.14	0.66
<i>PAQR4</i>	ENSG00000162073.13	0.66
<i>CDCA3</i>	ENSG00000111665.11	0.66
<i>ASF1B</i>	ENSG00000105011.8	0.65

mRNA – messenger RNA; ccRCC – clear cell renal cell carcinoma; GEPIA – Gene Expression Profiling Interactive Analysis; PCC – Pearson correlation coefficient.

identified as an independent adverse factor for prognosis of ccRCC through a survival analysis that integrated a Cox analysis and a nomogram. ROC curves and DCA both demonstrated that *PKMYT1* expression levels were slightly less useful for predicting prognosis of ccRCC prognosis than traditional clinical indicators such as histological grade and TNM stage. Consistent with the present results, *PKMYT1* has been shown to be upregulated in oncogenesis and progression of multiple human tumors, including colorectal cancer [49], esophageal squamous cell carcinoma (ESCC) [11], hepatocellular carcinoma (HCC) [22], non-small-cell lung cancer (NSCLC) [24], gastric cancer (GC) [21], and breast cancer [13], as well as OC [23]. These reports have also revealed a significant correlation between increased *PKMYT1* expression and advanced clinical stage and progressive histological grade, as well as unsatisfactory prognosis of diversiform tumors [11,13,21–24,49], which coincides with the results from our study. One report revealed that mutation in the C-terminal domain of *PKMYT1* influenced the prognosis of neuroblastoma by altering its catalytic activity [50]. A previous study also demonstrated that *PKMYT1* was a potential ccRCC metastasis-related novel gene, based on expression profiles in cultures and tumor tissues [51]. Thus, high *PKMYT1*

expression potentially indicates tumor progression and it may be a promising diagnostic and prognostic biomarker for ccRCC.

In the present study, we highlighted that *PKMYT1*, as a pivotal regulator of G2/M transition, exerted a strong effect on DNA-damage recovery and maintenance of genomic stability during rapid ccRCC cell proliferation. Given the challenge of developing effective therapies for ccRCC and the significance of the G2/M checkpoint for tumor cell survival, we speculate that *PKMYT1* may be an attractive molecular target for ccRCC treatment. Another report revealed that depletion of *PKMYT1* suppressed the β -catenin/TCF pathway, which is considered a driver of the EMT in multiple tumors and that the gene can interact with primary EMT-related biomarkers, such as Twist, Snail, and Slug [22,52,53]. Similarly, downregulation of *PKMYT1* impeded Twist expression by hindering the Akt/mTOR pathway, thus suppressing migration and metastasis as well the EMT process in ESCC cells [11]. Considering that EMT is a crucial determinant for tumor metastasis and progression [52,53], targeting the β -catenin/TCF and Akt/mTOR pathway through *PKMYT1* may be a promising strategy for tumor treatment. Some *PKMYT1* inhibitors have been developed, such as the

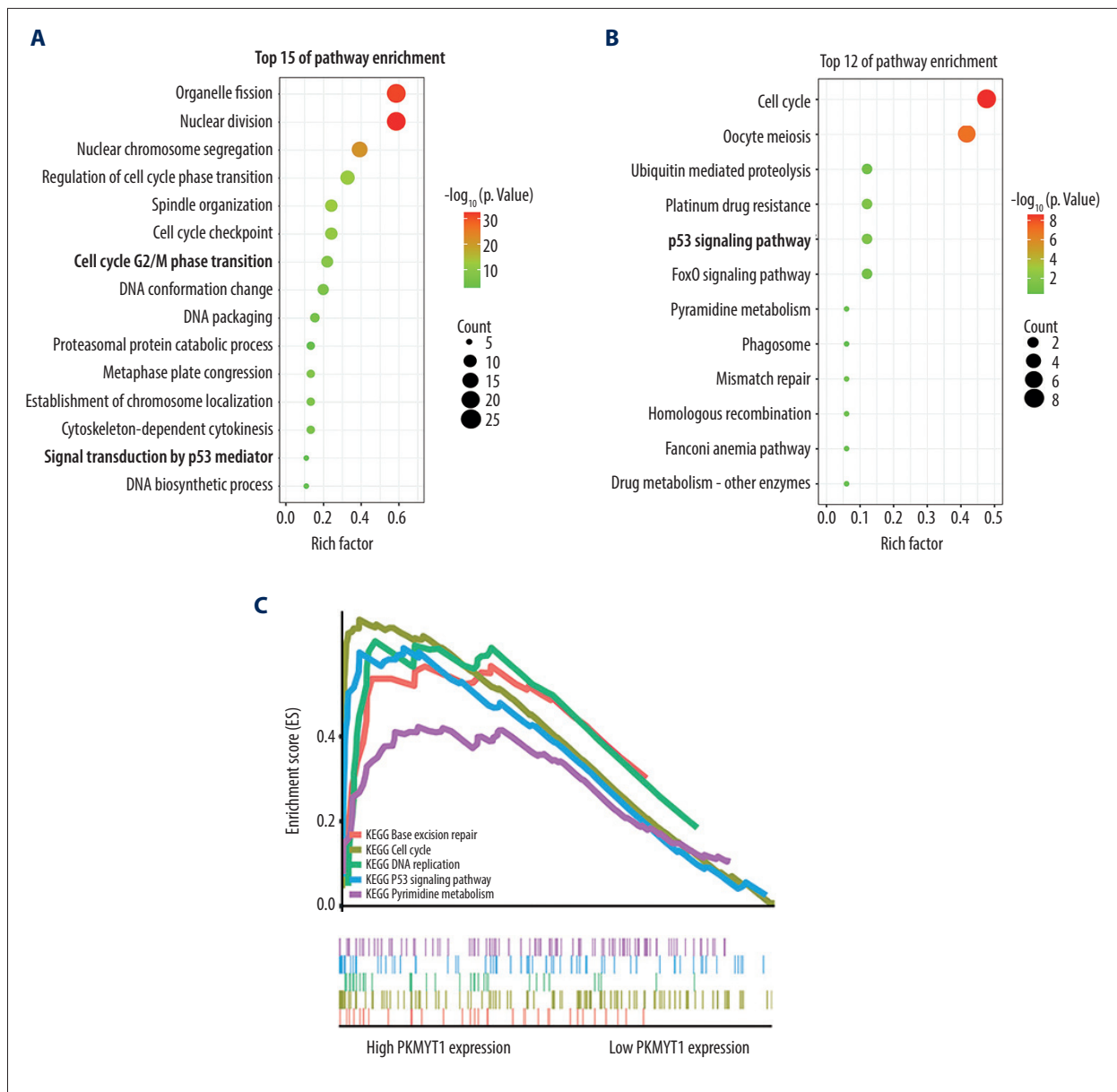


Figure 11. *PKMYT1* accelerates ccRCC progression by modulating the tumor cell cycle. The most significant signaling pathways in ccRCC found through (A) GO, (B) KEGG analysis, and (C) GSEA. Abbreviations: ccRCC – clear cell renal cell carcinoma; GO – gene ontology; KEGG – Kyoto Encyclopedia of Genes and Genomes; GSEA – gene set enrichment analysis.

tyrosine kinase inhibitors dasatinib and bosutinib. Studies of the pyridopyrimidine derivatives PD-0166285, PD-173952, and PD-173955 also are under way [11]. No potential *PKMYT1* inhibitors have yet been studied in clinical trials [14].

ASF1, a histone H3-H4 chaperone that comprises 2 ASF1 isoforms (*ASF1A* and *ASF1B*), has a critical effect on modulation of S-phase progression in multiple organisms. Depletion of both ASF1 isoforms triggers primary defects in S-phase progression. *ASF1A* and *ASF1B* are crucial to handle histones at replication forks through interaction with the MCM2-7 complex,

which is a putative helicase with the ability to unwind DNA ahead of the replication fork [54,55]. Importantly, high *ASF1B* expression correlates with faster disease development and metastasis in various tumors, including small-cell breast cancer [54], prostate cancer [56], and ccRCC [57]. Notably, in prostate cancer, downregulation of *ASF1B* suppressed expression of cyclin D1 and resulted in cell cycle arrest [56]. *ASF1B* overexpression accelerated tumor cell proliferation and migration in an Akt/P70 S6K1-dependent pathway in ccRCC [57]. Thus, the positive association of *ASF1B* and *PKMYT1* in ccRCC suggests an underlying cell cycle checkpoint mechanism related

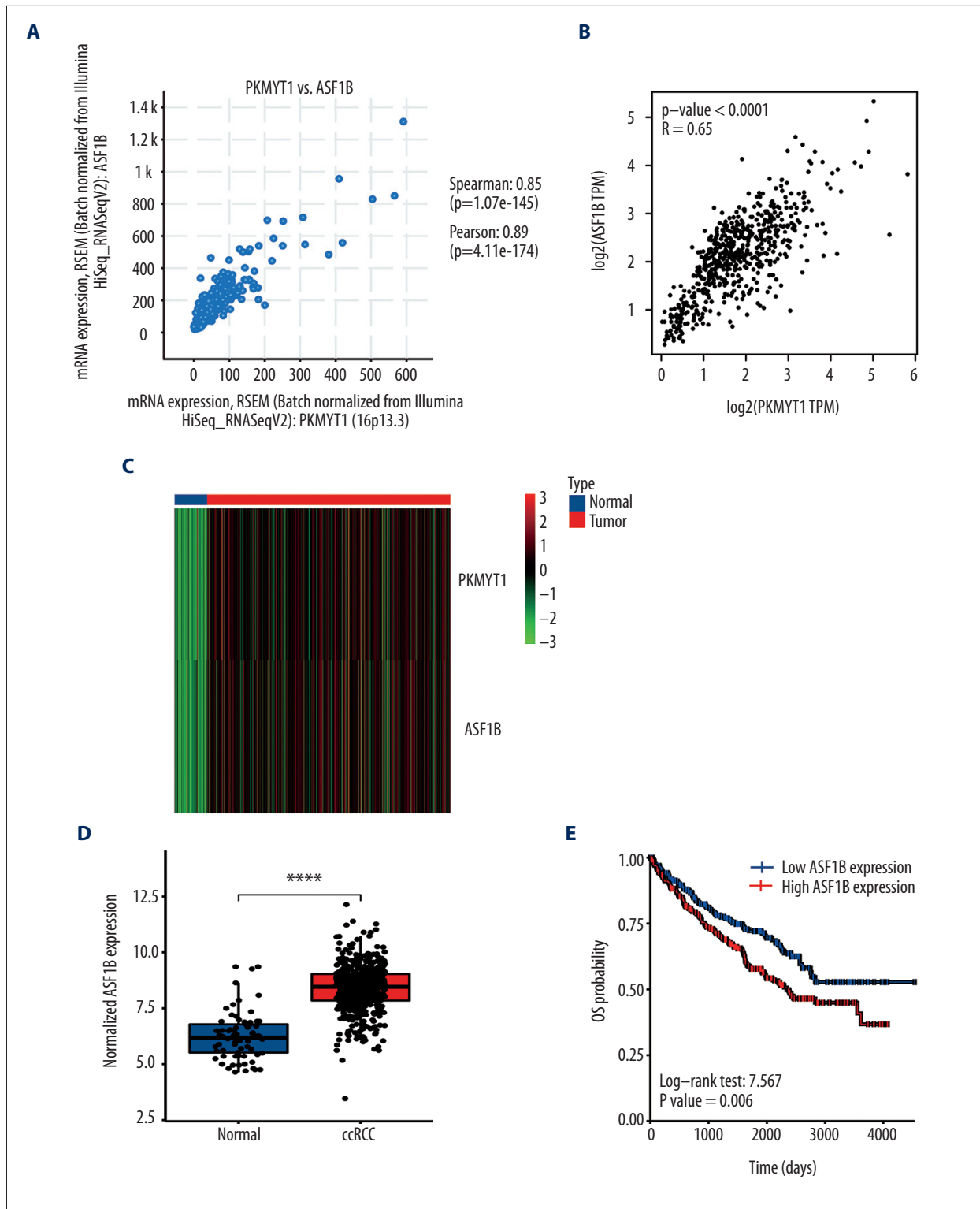


Figure 12. *PKMYT1* is significantly correlated with *ASF1B* gene expression in ccRCC. There was a positive correlation between *PKMYT1* and *ASF1B* based on (A) cBioPortal regression analysis, (B) GEPIA, and (C) TCGA dataset. (D) *ASF1B* was significantly increased in ccRCC compared with normal tissue. (E) High expression of *ASF1B* was associated with poor prognosis for ccRCC. ccRCC – clear cell renal cell carcinoma; GEPIA – Gene Expression Profiling Interactive Analysis; TCGA – The Cancer Genome Atlas.

to rapid tumor cell proliferation, based on maintenance of genomic stability. Overexpression of *PKMYT1* may result in genomic stability through modulation of G2/M in cancer cells. The *ASF1B* regulatory pathway simultaneously participates in controlling the S-phase duration for prompt cell proliferation. Co-targeting *PKMYT1* and *ASF1B* may be a promising approach to preventing ccRCC progression.

Our study deepens our comprehension of the correlation between *PKMYT1* and ccRCC, but several limitations need to be discussed. Initially, our report was performed on the basis of data extracted from multiple public biological databases. Although we identified *PKMYT1* expression at mRNA and protein levels and performed a survival analysis of ccRCC cases using information from TCGA, GEO, and HPA databases, validation of our results in further experiments is required to unravel the molecular mechanisms related to *PKMYT1* in ccRCC

development and prognosis. It is also essential to incorporate more prognostic factors into the nomogram to enhance the precision of survival analysis. Ultimately, the effects of *PKMYT1* on drug resistance associated with ccRCC require validation in future studies.

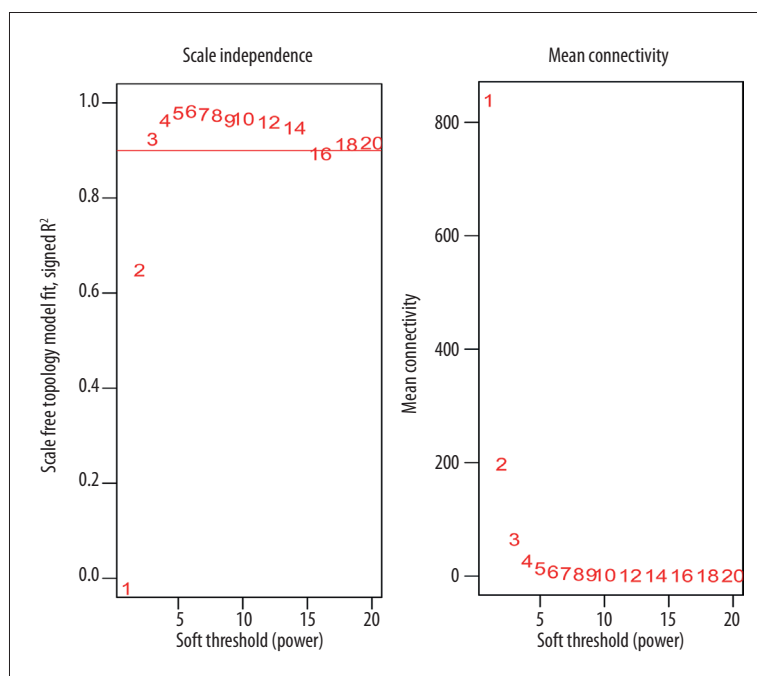
Conclusions

Our study is the first to assess the upregulation and negative prognostic role of *PKMYT1* in development of ccRCC, and our results suggest that the gene may be a promising therapeutic target for the management of ccRCC.

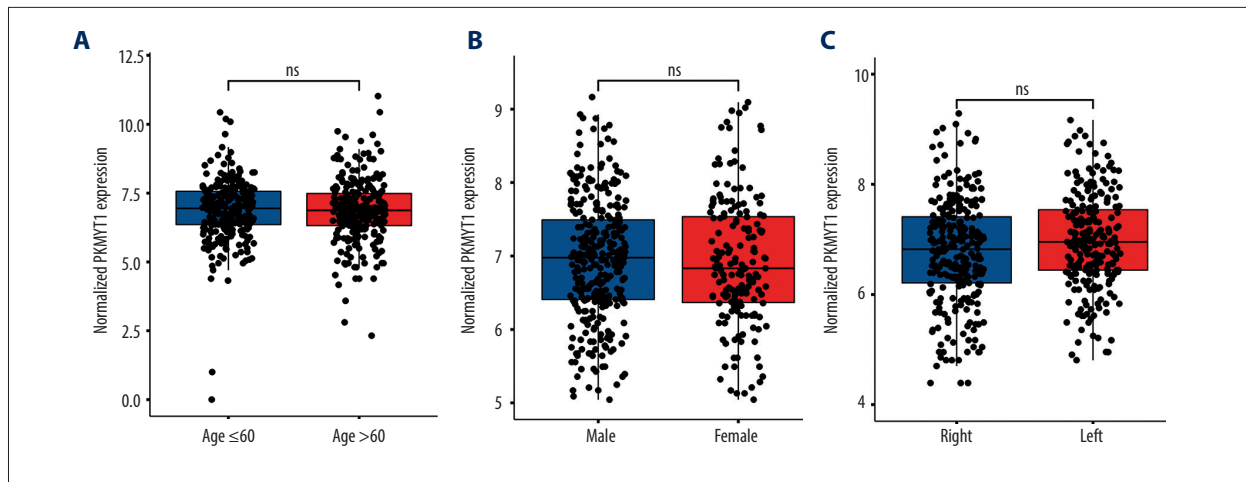
Conflicts of interest

None.

Supplementary Data



Supplementary Figure 1. Analysis of network topology for multiple soft-thresholding powers.



Supplementary Figure 2. *PKMYT1* mRNA expression in ccRCC is stratified by (A) age, (B) sex, and (C) laterality. mRNA – messenger RNA; ccRCC – clear cell renal cell carcinoma.

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