Research Paper

ISG20 serves as a potential biomarker and drives tumor progression in clear cell renal cell carcinoma

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ABSTRACT

Clear cell renal cell carcinoma (ccRCC) is one of the most common malignancies and lacks reliable biomarkers for diagnosis and prognosis, which results in high incidence and mortality rates of ccRCC. In this study, ISG20, HJURP, and FOXM1 were identified as hub genes via weighted gene co-expression network analysis (WGCNA) and Cox regression analysis. Samples validation showed that only ISG20 was up-regulated in ccRCC. Therefore, ISG20 was selected for further study. High ISG20 expression was associated with poor overall survival and disease-free survival. Furthermore, the expression of ISG20 could effectively differentiate ccRCC from normal tissues and was positively correlated to clinical stages. Functional experiments proved that knockdown of ISG20 expression could obviously inhibit cell growth, migration, and invasion in ccRCC cells. To find the potential mechanisms of ISG20, gene set enrichment analysis (GSEA) was performed and revealed that high expression of ISG20 was significantly involved in metastasis and cell cycle pathways. In addition, we found that ISG20 could regulate the expression of MMP9 and CCND1. In conclusion, these findings suggested that ISG20 promoted cell proliferation and metastasis via regulating MMP9/CCND1 expression and might serve as a potential biomarker and therapeutic target in ccRCC.

INTRODUCTION

Renal cell carcinoma (RCC) is one of the common lethal tumors in the urologic system, which is characterized by high incidence and high mortality rates. RCC accounts for 80-90% of all renal tumors [1, 2]. Recent cancer statistics estimated that there will be 73,820 new cases of RCC and 14,770 people will die of RCC in the USA in 2019 [3]. Clear cell renal cell carcinoma (ccRCC) is the most common subtype of RCC, which accounts for approximately 80% of RCC [4–6]. Because of resistance to radiotherapy and traditional chemotherapy, surgery becomes the most effective treatment for localized RCC patients [7]. For advanced RCC or metastatic RCC (mRCC), molecular targeted therapy has become a new first-line treatment

[8–11] and promoted the median survival time for approximately 3 years [12]. Unfortunately, many patients are still insensitive to targeted therapy. Therefore, it is necessary to find new biomarkers and therapeutic targets in ccRCC.

Nowadays, microarray [13] and high throughput sequencing [14, 15] techniques are frequently applied to identify generally genetic alterations. Integrated bioinformatics analyses are further used to find potential molecular mechanisms of tumorigenesis and progression. In this study, weighted gene co-expression network analysis (WGCNA) and cox regression analysis were utilized to screen hub genes in ccRCC.

Previous studies revealed that the interferon stimulated genes (ISGs) produce proteins acted as antiviral effectors in many virus infectious diseases [16-18]. Interferon stimulated exonuclease Gene 20 (ISG20), also named as estrogen-regulated transcript 45 protein, is an RNA exonuclease which induced by interferons (IFN types I and II) or double-stranded RNA [19-21]. ISG20 is able to cleave single-stranded RNA or DNA and is significantly associated with host antiviral innate immune defense [22, 23]. Furthermore, previous studies indicated that ISG20 played a vital role in tumorigenesis and progression of neoplasms. ISG20 with exonuclease activity could promote angiogenesis in vitro [24]. Lin et al. also demonstrated that ISG20 enhanced angiogenesis and supported progression of hepatocellular carcinoma (HCC) regulated by thyroid hormone [25]. To the best of our knowledge, the function of ISG20 has not been reported in ccRCC. In this study, we focused on the biological function and molecular mechanism of ISG20 via integrated bioinformatics analysis and functional experiments of ccRCC in vitro.

RESULTS

Identification of Differentially Expressed Genes (DEGs) in ccRCC

Gene expression data and clinical data were extracted from the GSE66272 dataset. According to the cut-off criteria, a total of 1025 genes were identified as DEGs (Figure 1A). In addition, the top 50 genes were exhibited in a heat map (Figure 1B).

Functional enrichment analysis of DEGs

The "clusterProfiler" package was used to perform GO and KEGG enrichment analysis in R. As shown in Figure 1C, biological process analysis indicated that the DEGs were significantly associated with organic anion transport, small molecule catabolic process, and leukocyte cell-cell adhesion. Cellular component results revealed that the DEGs mainly located in the apical plasma membrane, apical part of cell and basolateral plasma membrane. In the molecular function group, the DEGs were obviously enriched in secondary active transmembrane transporter activity, active transmembrane transporter activity, and anion transmembrane transporter activity. Moreover, the KEGG pathway enrichment analysis was performed to further uncover the potential biological functions of DEGs. As shown in Figure 1D, KEGG analysis exhibited that the DEGs were significantly correlated to the PPAR signaling pathway, cell adhesion molecules and multiple metabolic pathways (fructose/mannose metabolism, glycolysis/gluconeogenesis, tyrosine metabolism, cholesterol metabolism and arachidonic acid metabolism).

Screening hub genes via WGCNA and Cox regression analysis

The "WGCNA" package was utilized to screen hub modules significantly related to clinical characters. In our study, the power of $\beta = 5$ was selected as the soft threshold to ensure a scale-free network (Figure 2B). As shown in Figure 2C, seven modules (turquoise, yellow, blue, red, brown, green and grey module) were identified based on the gene expression pattern. Furthermore, the correlations between modules were all less than 0.8 (Figure 2D, 2E) and no modules need to be merged. According to the correlations between modules and clinical traits, the brown module (T stage: r = 0.47, p = 0.02; M stage: r = 0.36, p = 0.07; Grade: r = 0.67, p = 2e-04) was identified as a hub module for further analysis (Figure 2F). In this study, we selected T stage, M stage, and Grade as cutoff parameters to screen key genes (Figure 2G, 2I). Nine genes (TOP2A, NUF2, KIF4A, HJURP, FOXM1, CDCA8, CDCA5, CDC45, and ISG20) with significant clinical correlation were identified as key genes in the brown module (Table 1). Then, univariate and LASSO cox regression analysis of OS and DFS were performed to further screen hub genes. Univariate cox regression analysis indicated that nine key genes regarded as prognostic factors in both OS and DFS (Table 2). In LASSO cox regression, four genes (HJURP, FOXM1, CDC45, and ISG20) were identified in OS analysis (Figure 3A, 3B) and seven genes (TOP2A, KIF4A, HJURP, FOXM1, CDCA8, CDC45 and ISG20) were identified in DFS analysis (Figure 3C, 3D). As shown in Figure 3E, ISG20, HJURP and FOXM1 were identified as hub genes according to the results of cox regression analysis.

Hub genes validation

RT-PCR assay was performed to verify the mRNA expression level of hub genes in ccRCC tissues. We selected ISG20 for further study as only ISG20 was elevated in ccRCC tissues (Figure 4A). Then, we

further validated the expression level of ISG20 in other databases. In the TCGA KIRC and GSE40453 datasets, the expression of ISG20 in the ccRCC group was significantly higher than that in the control group (Figure 4B–4D). In the Oncomine database, ISG20 was also obviously up-regulated in ccRCC tissues (Figure 4E). Moreover, we found that ISG20 was elevated both in mRNA and protein levels of ccRCC cell lines (Figure 4F, 4G). And as shown in Figure 4H–4I, the protein expression level of ISG20 was also up-regulated in ccRCC tissues. Therefore, we believe that ISG20 is indeed up-regulated in ccRCC.

The expression level of ISG20 was significantly associated with clinicopathological features

As shown in Figure 5A–5E, the expression of ISG20 was observably positively correlated with multiple clinical stages (TNM stage, Grade stage, T stage, N stage and M stage). The survival analysis (Figure 5F–5G) showed that high expression of ISG20 predicted poor OS (HR = 1.64, p = 0.001) and DFS (HR = 1.70, p = 0.003). In addition, ROC curve analysis also exhibited that ISG20 had good diagnostic value in ccRCC. The expression level of ISG20 could effectively differentiate



Figure 1. Identification of DEGs and functional annotation. (A) The DEGs were exhibited as a volcano plot. Red dot: up-regulated genes; Blue dot: down-regulated genes; Black dot: unchanged genes. (B) Heatmap of the top 50 genes. (C) GO enrichment analysis of DEGs. Green columns: biological process group; Orange columns: cellular component group; Blue columns: molecular function group. (D) KEGG pathway enrichment analysis of DEGs. Top 20 enriched pathways were exhibited. DEGs: differentially expressed genes; GO: Gene Ontology; KEGG: Kyoto Encyclopedia of Genes and Genomes.



Figure 2. WGCNA of DEGs. (A) Sample clustering of GSE66272 to detect outliers. (B) Analysis of scale-free fit parameter and mean connectivity for various soft-thresholding power. (C) Dendrogram of DEGs clustered based on a dissimilarity measure (1-TOM). The DEGs were divided into seven modules (turquoise, yellow, blue, red, brown, green and grey module). (D, E) Clustering of module eigengene and eigengene adjacency heatmap to calculate the correlation between each module. No modules need to be merged. (F) The brown module was selected as the most important module according to the correlation between MEs and clinical traits. (G–I) Nine genes (TOP2A, NUF2, KIF4A, HJURP, FOXM1, CDCA8, CDCA5, CDC45, and ISG20) were identified as key genes in the brown module due to significantly associated with T stage, M stage, and Grade. WGCNA: weighted gene co-expression network analysis; DEGs: differentially expressed genes; TOM: topological overlap matrix; MEs: module eigengenes; MM: module membership; GS: gene significance.

Table 1. Key	genes identified	by	WGCNA.
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Gene —	Cor.geneModuleMembership	Cor.geneTraitSignificance			
	ME brown	T stage	M stage	Grade	
TOP2A	0.920	0.487	0.423	0.706	
NUF2	0.864	0.532	0.514	0.689	
KIF4A	0.893	0.640	0.429	0.662	
HJURP	0.879	0.559	0.411	0.688	
FOXM1	0.917	0.639	0.434	0.701	
CDCA8	0.884	0.462	0.518	0.743	
CDCA5	0.925	0.556	0.436	0.699	
CDC45	0.802	0.525	0.506	0.682	
ISG20	0.904	0.483	0.461	0.715	

WGCNA: weighted gene co-expression network analysis; ME: module eigengene.

Table 2. Univariate cox regression analysis of key genes.

Variable	Univariate analysis (OS)			Univariate analysis (DFS)		
	HR	95%CI	P-value	HR	95%CI	P-value
TOP2A	1.774	1.305-2.413	0.000	1.906	1.321-2.749	0.001
NUF2	1.840	1.353-2.503	0.000	1.673	1.169-2.394	0.005
KIF4A	1.984	1.456-2.702	0.000	1.942	1.350-2.794	0.000
HJURP	2.594	1.879-3.579	0.000	2.138	1.487-3.075	0.000
FOXM1	2.118	1.550-2.894	0.000	2.026	1.407-2.917	0.000
CDCA8	2.052	1.500-2.806	0.000	1.954	1.359-2.810	0.000
CDCA5	2.027	1.482-2.771	0.000	1.896	1.320-2.725	0.001
CDC45	1.894	1.395-2.573	0.000	1.530	1.071-2.185	0.019
ISG20	1.652	1.219-2.239	0.001	1.703	1.188-2.441	0.004
Age	1.750	1.288-2.378	0.000	1.363	0.957-1.942	0.086
Gender	0.942	0.692-1.282	0.704	1.427	0.960-2.120	0.079
T stage	1.901	1.615-2.237	0.000	2.506	2.033-3.089	0.000
N stage	3.847	2.083-7.107	0.000	5.731	2.874-11.428	0.000
M stage	4.295	3.149-5.856	0.000	8.628	5.944-12.524	0.000
TNM stage	1.875	1.645-2.138	0.000	2.668	2.238-3.180	0.000
Grade	2.250	1.842-2.748	0.000	3.028	2.348-3.905	0.000

OS: overall survival; DFS: disease-free survival; HR: hazard ratio; CI: confidence interval; TNM: Tumor Node Metastasis.

ccRCC from normal renal tissues (Figure 5H, 5I). Thus, ISG20 may become a potential diagnostic and prognostic biomarker in ccRCC.

ISG20 promoted cell proliferation, invasion, and migration in ccRCC

To verify the effect of ISG20 on the biological function of ccRCC cell, the siRNA targeting ISG20 (si-ISG20) was used to knockdown the expression of ISG20 in 786O, A498, and OSRC-2 cell lines. The RT-PCR and western blotting results indicated that the si-ISG20 could effectively inhibit the expression of ISG20 in ccRCC cell (Figure 6A, 6B). As shown in Figure 6C– 6H, down-regulation of ISG20 significantly inhibited cell proliferation. Furthermore, down-regulation of ISG20 also dramatically reduced the capacity of invasion and migration in ccRCC cells (Figure 7). These findings suggested that ISG20 drove tumor progression and played the role of the oncogene.

ISG20 facilitated proliferation and metastasis of ccRCC via mediating the expression of CCND1/MMP9

GSEA results indicated that the high expression of the ISG20 group mainly enriched in metastasis pathways and cell cycle pathways (Figure 8A, 8B). We also found



Figure 3. Identification of hub genes via LASSO cox regression analysis. (A) Partial likelihood deviance of OS for the LASSO coefficient profiles. (B) LASSO coefficient profiles of the nine key genes for OS. (C) Partial likelihood deviance of DFS for the LASSO coefficient profiles. (D) LASSO coefficient profiles of the nine key genes for DFS. (E) ISG20, HJURP, and FOXM1 were identified as hub genes according to the LASSO cox regression analysis of OS and DFS. LASSO: least absolute shrinkage and selection operator; OS: overall survival; DFS: disease-free survival.

that the silencing of ISG20 could significantly downregulate the expression of MMP9 and CCND1 in ccRCC (Figure 8C). In addition, both MMP9 and CCND1 were both up-regulated in ccRCC (Figure 8D). Therefore, we believed that ISG20 might promote the progression of ccRCC through up-regulating CCND1 and MMP9 (Figure 8E).

DISCUSSION

ccRCC is one of the most common neoplasms characterized by high metastasis and recurrence rates. At the time of diagnosis, with about 30% of patients

exist metastatic lesions due to lack of effective biomarkers and obvious clinical manifestations [8]. Molecular targeted therapy prolongs the survival time of some patients, but there are still many patients who are insensitive to targeted therapy [26, 27]. At immunotherapy, specifically immune present. checkpoint inhibitors, has become a new promising strategy for ccRCC [28-30]. However, intrinsic resistance or acquired resistance of immunotherapy is still observed [31]. Therefore, it is very necessary to further study the molecular mechanisms of progression and discover new therapeutic targets of ccRCC.





In this present study, we performed differential analysis and functional annotation to screen important genes in ccRCC. WGCNA and Cox regression analysis were used to reduce dimension and identify hub genes. ISG20, HJURP, and FOXM1 were selected as candidate biomarkers for further analysis and study.

GO enrichment analysis indicated that the biological functions of DEGs were mainly correlated to transmembrane transporter activity and cell-cell adhesion. Numerous studies have shown that transmembrane transport plays an important role in various physiological responses and functions [32–34]. For tumor cells, abnormal transmembrane transport may be a vital factor in maintaining their survival and malignancy [35, 36]. It has been reported that mono-carboxylate transporter 1 (MCT1) and monocarboxylate transporter 4 (MCT4) promotes proliferation and metastasis of renal cancer cells [37]. In hepatocellular carcinoma, previous study proved that YAP1/TAZ activated the mTORC1 pathway via up- regulating amino acid transporters (SLC38A1 and SLC7A5)





to promote cell growth [38]. Cell adhesion is another important biological behavior which obviously associated with cell proliferation, migration, and invasion [39], especially in cancers. Labernadie et al. uncovered that cancer-associated fibroblasts (CAFs) enhanced tumor invasion via heterophilic adhesion between CAFs and tumor cells [40]. In the glioblastoma, acetyl-coenzyme A (acetyl-CoA) acted as an activator for the Ca²⁺-NFAT signaling pathway to drive cell adhesion and migration [41]. In our present study, KEGG analysis also showed that the DEGs were enriched in cell adhesion molecules (CAMs) pathway. Furthermore, KEGG enrichment analysis revealed that the DEGs were obviously associated with multiple





metabolic pathways. ccRCC is known for metabolic disorders. Recent studies indicated that abnormal metabolism played an important role in the occurrence and progression of ccRCC. Bianchi et al. proved that aerobic glycolysis was a grade-dependent feature and fatty acid oxidation was needed in different grade tumors [42]. Multi-omics characterization also revealed that abnormal glycolysis and pentose phosphate pathway promoted tumor growth [43]. Lucarelli et al. indicated that aerobic glycolysis reprogramming is pivotal in the initial phases of tumorigenesis in ccRCC [44]. Therefore, a better understanding of the molecular mechanisms underlying this metabolic alteration will be crucial for the identification of novel potential therapeutic targets and biomarkers.

It was worth mentioning that the DEGs were significantly associated with the peroxisome

proliferators-activated receptors (PPAR) signaling pathway. A large number of studies have demonstrated that the occurrence and development of tumors are closely related to PPAR signaling pathways. It has been reported that PPAR-delta maintains cell survival in an energy-poor environment for chronic lymphocytic leukemia [45]. Zuo et al. also verified that up-regulated PPAR- δ/β could promote the susceptibility to colon cancer in villin-PPAR mice models [46]. Thus, in-depth study of the mechanisms of the PPAR pathway will help us in conquering cancer.

In this study, we selected ISG20 for further study through the validation of ccRCC tissues. ISG20, an interferon regulated gene, codes for a 20-kDa protein with 181amino acid [47]. Both type I (IFN- α/β) and type II (IFN- γ) IFNs can induce the expression of ISG20 [17]. As a 3'-5' exonuclease, ISG20 cleaves both DNA and RNA to



Figure 7. ISG20 improved the migration and invasion abilities of ccRCC cells in vitro. (A–C) Transwell assays analysis of the effect of ISG20 knockdown on cell migration and invasion. (D–F) Wound healing assays analysis of the effect of ISG20 knockdown on the migration of ccRCC cells. ccRCC: clear cell renal cell carcinoma. Data are represented as mean ± SD. ***, P < 0.001; **, P < 0.01; *, P < 0.05.



Figure 8. ISG20 positively regulated the expression of MMP9/CCND1 in ccRCC. (A–B) GSEA revealed that the high ISG20 expression group was associated with metastasis pathways and cell cycle pathways. (C) The expression of MMP9 and CCND1 were down-regulated in ccRCC cells with knockdown of ISG20. (D) The expression of MMP9 and CCND1 were elevated in the TCGA KIRC dataset and ccRCC tissues. (E) The schematic representation of elevated the expression of ISG20 could drive tumor progression via up-regulating the expression of MMP9/CCND1. ccRCC: clear cell renal cell carcinoma; TCGA KIRC: The Cancer Genome Atlas Kidney Clear Cell Carcinoma. Data are represented as mean \pm SD. ***, P < 0.001; **, P < 0.01; *, P < 0.05.

protect against virus and bacteria [48]. Qu et al. found that ISG20 reduced influenza A virus (IAV) replication by interacting with nucleoprotein [49]. Apart from fighting against pathogens, ISG20 also plays an important role in other diseases [50], especially in tumors. It has been reported that ISG20 is up-regulated in cervical cancer [51]. In this study, we also found that the expression of ISG20 was elevated in online databases (TCGA and Oncomine) and ccRCC samples. In human glioma, ISG20 was positively correlated to immune checkpoints (PD-1 and PD-L1) and suppressed the adaptive immune response. In addition, high expression of ISG20 predicted poor overall survival in glioma [52]. Similarly, patients with high expression of ISG20 had shorter overall survival time in our present study. We found that ISG20 was positively associated with the clinical stage (TNM stage and Grade stage) in ccRCC. Furthermore, ROC curve analysis indicated that ISG20 could effectively differentiate ccRCC samples from normal samples. These results suggested that ISG20 might serve as a potential biomarker in ccRCC.

To further reveal the biological function and molecular mechanisms of ISG20, GSEA was performed and indicated that high expression of ISG20 mainly enriched in metastasis and cell cycle pathways. Functional experiments also verified that knockdown of ISG20 could obviously inhibit proliferation, migration, and invasion of ccRCC cells. In addition, ISG20 could positively regulate the expression of MMP9 and CCND1 in ccRCC cells in our study. MMP9 (matrix metalloproteinase 9), a member of the matrix metalloproteinase (MMP) family, is involved in the breakdown of extracellular matrix [53]. Type IV and V collagens are the major substrates for MMP9. A large number of studies have proved that MMP9 plays an important role in tumorigenesis, proliferation, apoptosis, invasion, and angiogenesis as well [54-57]. Moreover, it has been reported that the expression and activation of MMP9 are regulated by multiple signal pathways, such as STAT3 pathway [58, 59], JNK pathway [60, 61] and PI3K/AKT pathway [62]. A number of studies also uncovered that MMP9 could promote the migration and invasion of ccRCC cells [63, 64]. Gong et al. verified that P2RX6 facilitated the invasion and metastasis of RCC cells through the Ca²⁺-p-ERK1/2-MMP9 signal pathway [65]. CCND1 (cyclin D1), a member of the highly conserved cyclin family, acts as a regulator of CDK kinases for mediating cell cycle [66, 67]. Mutations, amplification and overexpression of CCND1 are observed frequently in a variety of tumors and these may contribute to tumorigenesis [68-70]. Several studies indicated that non-coding RNA was involved in tumor progression via regulating the expression of CCND1. Ai et al. revealed that LINC01355 inhibited cell proliferation through suppressing the transcription of CCND1 [71]. In cervical cancer, a research group demonstrated that miR-2861 suppressed tumor cell growth and invasion by targeting EGFR/AKT2/CCND1 pathway [72]. However, the specific molecular mechanisms between ISG20 and MMP9/CCND1 are still unclear and need further study.

In summary, ISG20, HJURP, and FOXM1 were identified as hub genes via WGCNA and Cox regression analysis in this study. Clinical samples analysis showed that only ISG20 had an obvious difference between ccRCC tissues and normal renal tissues. Survival analysis and ROC curve analysis indicated that ISG20 had good diagnostic and prognostic value, which could become a candidate biomarker in ccRCC. In addition, high expression of ISG20 promoted the proliferation, migration, and invasion of ccRCC cells via regulating MMP9/CCND1 expression. To the best of our knowledge, this is the first study on the biological functions of ISG20 in ccRCC. These findings may provide a new therapeutic target for ccRCC. However, the specific mechanisms of ISG20 in ccRCC still need further research.

MATERIALS AND METHODS

Data download and study design

The gene expression data and clinical data were downloaded from the GSE66272 dataset and The Cancer Genome Atlas (TCGA) database (<u>https://www.cancer.gov/tcga</u>). The microarray data of GSE66272 included 52 matched ccRCC tissues and adjacent normal tissues. Furthermore, the study design was exhibited in a flow diagram (Figure 9).

Identification of differentially expressed genes (DEGs)

The "limma" package [73] was used to screen DEGs according to the cutoff criterion of adjusted P-value (adj. P) < 0.05 and |log Fold Change| (|log FC|) > 2.0. A heat map was drawn to exhibit the expression difference between ccRCC samples and normal samples of the top 50 DEGs.

Functional and pathway enrichment analysis

Gene Ontology (GO) analysis and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway analysis were performed to further study the potential biological function of DEGs via the "clusterProfiler" package [74] in R. P value < 0.05 was selected as the cutoff point.

Weighted gene co-expression network analysis (WGCNA)

The "WGCNA" package [75] was used to construct a coexpression network in R. Firstly, GSE66272 was evaluated via sample clustering to detect outliers. No samples were removed according to the height cutoff point = 60 in this study (Figure 1A). Then, the Pearson's correlation matrices were performed between each of the gene pairs. A weighted adjacency matrix was constructed using a power function $a_{mn} = |c_{mn}|^{\beta}$ (c_{mn} = Pearson's correlation between gene m and gene n; $a_{mn} = adjacency$ between gene m and gene n). The adjacencies achieved scale-free topology based on the soft threshold power β . Next, the adjacencies were transformed into a topological overlap matrix (TOM). Average linkage hierarchical clustering was performed to divide DEGs into different modules according to the correlations between each gene. Genes with high absolute correlation were clustered into the same module. Finally, the correlation between module eigengenes (MEs) and clinical traits was calculated to identify the clinically significant modules. Furthermore, we calculated the correlation between genes and clinical traits (cor.geneTraitSignificance) and the correlation between genes and MEs (cor.geneModuleMembership) as



Figure 9. Flow diagram of the study. Data collection and analysis were exhibited in the flow diagram. DEGs: differentially expressed genes; WGCNA: weighted gene co-expression network analysis; LASSO: least absolute shrinkage and selection operator; GSEA: gene set enrichment analysis; ROC: receiver operator characteristic.

well. In this study, we selected |cor.geneTraitSignificance| > 0.4 and |cor.geneModuleMembership| > 0.8 of T stage, M stage and Grade as the cutoff criterion to screen clinical key genes.

Cox regression analysis

Firstly, the "survival" package (https://CRAN.Rproject.org/package=survival) was applied to perform univariate cox regression analysis for both overall survival (OS) and disease-free survival (DFS) in R. Genes with p value < 0.05 were identified as useful genes. Then, least absolute shrinkage and selection operator (LASSO) cox regression analysis was performed to further screen hub genes via "glmnet" [76, 77] package in R.

Hub genes validation in TCGA, GEO, and Oncomine databases

The mRNA expression data of hub genes were downloaded from The Cancer Genome Atlas (TCGA), Gene Expression Omnibus (GEO) and Oncomine (<u>https://www.oncomine.org</u>) databases including Beroukhim Renal, Gumz Renal, Jones Renal, and Yusenko Renal dataset. GraphPad Prism software was used to verify the difference between ccRCC samples and normal samples. P value < 0.05 was considered statistically significant.

Survival analysis and receiver operator characteristic (ROC) curve analysis

Survival data were obtained from the TCGA database. The ccRCC samples were divided into high expression group and low expression group according to the median mRNA expression of hub genes. Kaplan-Meier survival curves of overall survival (OS) and disease-free survival (DFS) were performed by GraphPad Prism. Meanwhile, the ROC curves were also drawn by GraphPad Prism to evaluate the diagnostic value of hub genes. P value < 0.05 was considered statistically significant.

Gene set enrichment analysis (GSEA)

All ccRCC samples were divided into high expression group and low expression group according to the median mRNA expression of hub genes. Then, GSEA software [78, 79] was used to find potential molecular mechanisms of hub genes. Nominal P < 0.05 and false discovery rate (FDR) < 0.25 were selected as cutoff criteria.

Human ccRCC specimens

A total of 37 pairs of ccRCC tissues and adjacent normal renal tissues (5 cm away from the margin of

the tumor tissues) were collected from Wuhan Union Hospital between 2017 and 2018. All patients were provided with informed consent. Moreover, this study was approved by the Human Research Ethics Committee of Huazhong University of Science and Technology (HUST).

Cell culture

The 786O, Caki-1, A498, OSRC-2, and HK-2 cell lines were purchased from the American Type Culture Collection (ATCC) in this study. All cells were cultured in Dulbecco's modified Eagle's medium (Gibco, USA) with 10% fetal bovine serum (FBS) and 1% penicillinstreptomycin (Gibco, USA). Moreover, all cells were maintained in 37 °C and 5% CO₂ culture environment.

RNA extraction and quantitative real-time PCR (qRT-PCR)

Total RNA of cells and tissues was extracted with the TRIzol reagent (Invitrogen, CA). Then. the concentration and purity of RNA were detected via using Nanodrop 2000c spectrophotometer (Thermo Scientific, USA). Reverse transcription kit (Takara, China) was used to synthesize cDNA for further analysis. The qRT-PCR assay was performed to amplify cDNA with SYBR Green Master Mix (Vazyme, China). Primer sequences were listed as follows: GAPDH 5'- GCACCGTCAAGGCTGAGAAC-3': Forward: GAPDH Reverse: 5'- TGGTGAAGACGCCAGTGGA-3'; ISG20 Forward: 5'- CTCGTTGCAGCCTCGTGAA-3': ISG20 Reverse: 5'- CGGGTTCTGTAATCGGT GATCTC-3'; HJURP Forward: 5'- CACAAAGC CATCAAGCATCATC-3'; HJURP Reverse: 5'- TCA GAGCAGGGTATGAAGTTCT-3'; FOXM1 Forward: 5'-CGTCGGCCACTGATTCTCAAA-3': FOXM1 Reverse: 5'- GGCAGGGGATCTCTTAGGTTC-3'.

Western blotting (WB)

Western blotting was performed as previously described [80]. In brief, the protein of cells and tissues were extracted with the radio-immunoprecipitation assay (RIPA) lysis buffer (Beyotime, China) containing 1mM Phenylmethylsulfonyl fluoride (Beyotime, China) and 1X protease inhibitor cocktail (MCE, USA). The concentration of protein was detected using the BCA assay kit (Beyotime, China). 60ug of protein was subjected to 12% SDS-PAGE and transferred to PVDF membranes (Millipore, USA). Then, the membranes were blocked in 5% nonfat dried skimmed milk for 2h at room temperature. After that, the PVDF membranes were incubated with primary antibodies containing ISG20 (1:100, Abclonal, China, A14744), GAPDH (1:50000, Abclonal, China, AC002), β -actin (1:50000, Abclonal,

China, AC026), MMP9 (1:1000, Abclonal, China, A2095), CCND1 (1:10000, Abcam, ab134175) and NDRG2 (1:5000, Abcam, ab169775) overnight at 4 °C. Finally, the membranes were incubated with corresponding secondary antibodies (1:3000, Proteintech, China, SA00001-1 and SA00001-2) for 1.5h at room temperature and visualized with the ChemiDoc-XRS+ system (Bio-Rad, USA).

Immunohistochemical (IHC) assay

IHC assay was performed as previously described [81]. In short, the ccRCC tissues and adjacent renal tissues were fixed using the following step: formalin fixation, dehydration, and paraffin embedding. After that, the tissue sections were incubated with a primary antibody against ISG20 (1:50, Abclonal, China, A14744) overnight at 4 °C. Then, the tissue sections were washed with PBS two times and incubated with a second antibody (1:400, Proteintech, China, SA00001-2) for 1.5h at room temperature.

Transient transfection assay

The siRNA oligonucleotide sequences targeting ISG20 (si-ISG20) and the negative control (si-NC) siRNA were designed and synthesized from TranSheepBio (Shanghai, China). The si-ISG20 and si-NC with a final concentration of 75 nM were transfected into the ccRCC cells of 70% confluence with Lipofectamine® 3000 (Invitrogen, USA) according to the manufacturer's protocol. Cells were collected 48h after si-RNA transfection as subsequent assays. The si-ISG20 sequences were listed as below: si-ISG20#1: 5'-GAGAUCACCGAUUACAGAATT-3'; si-ISG20#2: 5'-GAUCCUGCAGCUCCUGAAATT-3'.

Cell viability assays

In this study, cell counting kit-8 (CCK-8) assay and colony formation assay were performed to evaluate the effect of ISG20 on cell proliferation.

Cell counting kit-8 assay: 1000 cells were planted in 96-well plates per well with 100ul of medium. Then, CCK-8 solution (MCE, USA) was added to 96-well plates with 10ul per well. After incubation for 2.5h at 37 °C, the optical density (OD) value of each well was measured at 450 nm with a spectrophotometer (Bio-Rad, USA). The OD value was assessed at after 0, 24, 48, 72 and 96h upon treatments, respectively.

Colony formation assay: 1000 cells were seeded in 6well plates per well with 2ml complete medium and cultured for 2 weeks. Then, the cells were fixed with methanol and stained with 0.1% crystal violet.

Transwell assay

Firstly, cells were incubated in FBS-free medium for 24h. Secondly, a total of 10000 cells were seeded into the upper chamber with 200 μ l of FBS-free medium for the migration assay and 20000 cells were seeded into the upper chamber which was pre-coated with Matrigel (Bio-Rad, USA) for the invasion assay. In addition, 600ul medium with 10%FBS was added into the bottom chamber. Thirdly, the invasive cells were fixed with methanol and stained with 0.1% crystal violet after incubated for 24h. Finally, five fields were randomly selected to count the cells.

Wound healing assay

Cells were seeded into the 6-well plates with equal numbers. Then, cells were wounded by a 10ul pipette tip when its confluence reached 80-90%. Images of the wound were obtained at 0 and 24h.

Statistical analysis

All statistical analyses were performed using GraphPad Prism 7.0 (GraphPad software, Inc., La Jolla, CA, USA) and each experiment was conducted in triplicate. All data were represented as mean \pm SD. Student's t-test and Pearson's χ^2 test were used to analyze the data in this study. The significance value was determined when p < 0.05.

Abbreviations

ccRCC: clear cell renal cell carcinoma; RCC: renal cell carcinoma; mRCC: metastatic RCC; WGCNA: weighted gene co-expression network analysis; ISGs: interferon genes; interferon stimulated ISG20: stimulated exonuclease Gene 20; HCC: hepatocellular carcinoma; TCGA: The Cancer Genome Atlas; DEGs: differentially expressed genes; GO: Gene Ontology; KEGG: Kyoto Encyclopedia of Genes and Genomes: TOM: topological overlap matrix; MEs: module eigengenes; OS: overall survival; DFS: disease-free survival; LASSO: least absolute shrinkage and selection operator; GEO: Gene Expression Omnibus; ROC: receiver operator characteristic; GSEA: gene set enrichment analysis; FDR: false discovery rate; HUST: Huazhong University of Science and Technology; ATCC: American Type Culture Collection; FBS: fetal bovine serum; qRT-PCR: quantitative real-time PCR; WB: western blotting; RIPA: radio-immunoprecipitation assay; IHC: immunohistochemical; CCK-8: cell counting kit-8; OD: optical density; MCT1: monocarboxylate transporter 1; MCT4: monocarboxylate transporter 4; CAMs: cell adhesion molecules; PPAR: peroxisome proliferators-activated receptors; IAV: influenza A virus; MMP: matrix

metalloproteinase; MMP9: matrix metalloproteinase 9; CCND1: cyclin D1; HR: hazard ratio; CI: confidence interval; TNM: Tumor Node Metastasis; MM: module membership; GS: gene significance; CGA KIRC: The Cancer Genome Atlas Kidney Clear Cell Carcinoma.

AUTHOR CONTRIBUTIONS

XZ and KC designed this study. TX, SG, HR, and JL performed data collection and analysis. TX, YL, ZS, and QC performed the majority of the experiments. TX and HR wrote the manuscript and contributed to preparing and making figures and tables. LB, DL and KW collected the clinical samples and managed the clinical data. JT and JS reviewed the relevant literature. HY, HL, XZ, and KC provided conceptual advice and critically reviewed the manuscript. All authors read and approved the final manuscript.

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CONFLICTS OF INTEREST

The authors declare that they have no conflicts of interest.

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REFERENCES

 Ljungberg B, Bensalah K, Canfield S, Dabestani S, Hofmann F, Hora M, Kuczyk MA, Lam T, Marconi L, Merseburger AS, Mulders P, Powles T, Staehler M, et al. EAU guidelines on renal cell carcinoma: 2014 update. Eur Urol. 2015; 67:913–24. <u>https://doi.org/10.1016/j.eururo.2015.01.005</u> PMID:<u>25616710</u>

- Campbell S, Uzzo RG, Allaf ME, Bass EB, Cadeddu JA, Chang A, Clark PE, Davis BJ, Derweesh IH, Giambarresi L, Gervais DA, Hu SL, Lane BR, et al. Renal Mass and Localized Renal Cancer: AUA Guideline. J Urol. 2017; 198:520–29. <u>https://doi.org/10.1016/j.juro.2017.04.100</u> PMID:28479239
- Siegel RL, Miller KD, Jemal A. Cancer statistics, 2019. CA Cancer J Clin. 2019; 69:7–34. <u>https://doi.org/10.3322/caac.21551</u> PMID:<u>30620402</u>
- Moch H, Cubilla AL, Humphrey PA, Reuter VE, Ulbright TM. The 2016 WHO Classification of Tumours of the Urinary System and Male Genital Organs-Part A: Renal, Penile, and Testicular Tumours. Eur Urol. 2016; 70:93– 105.

https://doi.org/10.1016/j.eururo.2016.02.029 PMID:26935559

- Linehan WM. Genetic basis of kidney cancer: role of genomics for the development of disease-based therapeutics. Genome Res. 2012; 22:2089–100. <u>https://doi.org/10.1101/gr.131110.111</u> PMID:<u>23038766</u>
- Yan BC, Mackinnon AC, Al-Ahmadie HA. Recent developments in the pathology of renal tumors: morphology and molecular characteristics of select entities. Arch Pathol Lab Med. 2009; 133:1026–32. <u>https://doi.org/10.1043/1543-2165-133.7.1026</u> PMID:<u>19642729</u>
- Capitanio U, Montorsi F. Renal cancer. Lancet. 2016; 387:894–906. <u>https://doi.org/10.1016/S0140-6736(15)00046-X</u> PMID:<u>26318520</u>
- Gong J, Maia MC, Dizman N, Govindarajan A, Pal SK. Metastasis in renal cell carcinoma: biology and implications for therapy. Asian J Urol. 2016; 3:286–92. <u>https://doi.org/10.1016/j.ajur.2016.08.006</u> PMID:<u>29264197</u>
- 9. Rini BI, Halabi S, Rosenberg JE, Stadler WM, Vaena DA, Archer L, Atkins JN, Picus J, Czaykowski P, Dutcher J, Small EJ. Phase III trial of bevacizumab plus interferon alfa versus interferon alfa monotherapy in patients with metastatic renal cell carcinoma: final results of CALGB 90206. J Clin Oncol. 2010; 28:2137–43. <u>https://doi.org/10.1200/JCO.2009.26.5561</u> PMID:<u>20368558</u>
- Sternberg CN, Davis ID, Mardiak J, Szczylik C, Lee E, Wagstaff J, Barrios CH, Salman P, Gladkov OA, Kavina A, Zarbá JJ, Chen M, McCann L, et al. Pazopanib in locally advanced or metastatic renal cell carcinoma: results of a randomized phase III trial. J Clin Oncol. 2010; 28:1061–68.

https://doi.org/10.1200/JCO.2009.23.9764 PMID:20100962

- Rini BI, Escudier B, Tomczak P, Kaprin A, Szczylik C, Hutson TE, Michaelson MD, Gorbunova VA, Gore ME, Rusakov IG, Negrier S, Ou YC, Castellano D, et al. Comparative effectiveness of axitinib versus sorafenib in advanced renal cell carcinoma (AXIS): a randomised phase 3 trial. Lancet. 2011; 378:1931–39. <u>https://doi.org/10.1016/S0140-6736(11)61613-9</u> PMID:22056247
- Pal SK, Choueiri TK, Karam JA, Heng DY. Metastatic renal cell carcinoma: contending with a sea change in therapy. Urol Oncol. 2015; 33:507–08. <u>https://doi.org/10.1016/j.urolonc.2015.10.005</u> PMID:<u>26584741</u>
- Sudo S, Konno Y, Odagiri T, Kato T, Hosaka M, Takeda M, Watari H, Kaneuchi M, Sakuragi N. Exploration of biomarkers for lymph node metastasis in patients with endometrial cancer using exon-expression microarray. J Clin Oncol. 2011; 29:5100. https://doi.org/10.1200/jco.2011.29.15 suppl.5100
- 14. Nilsson RH, Anslan S, Bahram M, Wurzbacher C, Baldrian P, Tedersoo L. Mycobiome diversity: highthroughput sequencing and identification of fungi. Nat Rev Microbiol. 2019; 17:95–109. <u>https://doi.org/10.1038/s41579-018-0116-y</u> PMID:<u>30442909</u>
- Grosselin K, Durand A, Marsolier J, Poitou A, Marangoni E, Nemati F, Dahmani A, Lameiras S, Reyal F, Frenoy O, Pousse Y, Reichen M, Woolfe A, et al. High-throughput single-cell ChIP-seq identifies heterogeneity of chromatin states in breast cancer. Nat Genet. 2019; 51:1060–66. https://doi.org/10.1038/s41588-019-0424-9

PMID:31152164

- Horio T, Murai M, Inoue T, Hamasaki T, Tanaka T, Ohgi T. Crystal structure of human ISG20, an interferoninduced antiviral ribonuclease. FEBS Lett. 2004; 577:111–16. <u>https://doi.org/10.1016/i.febslet.2004.09.074</u> PMID:15527770
- Zheng Z, Wang L, Pan J. Interferon-stimulated gene 20kDa protein (ISG20) in infection and disease: review and outlook. Intractable Rare Dis Res. 2017; 6:35–40. <u>https://doi.org/10.5582/irdr.2017.01004</u>
 PMID:<u>28357179</u>
- Liu Y, Nie H, Mao R, Mitra B, Cai D, Yan R, Guo JT, Block TM, Mechti N, Guo H. Interferon-inducible ribonuclease ISG20 inhibits hepatitis B virus replication through directly binding to the epsilon stem-loop structure of viral RNA. PLoS Pathog. 2017; 13:e1006296.

https://doi.org/10.1371/journal.ppat.1006296 PMID:28399146

- 19. Espert L, Rey C, Gonzalez L, Degols G, Chelbi-Alix MK, Mechti N, Gongora C. The exonuclease ISG20 is directly induced by synthetic dsRNA via NF-kappaB and IRF1 activation. Oncogene. 2004; 23:4636-40. https://doi.org/10.1038/sj.onc.1207586 PMID:15064705
- 20. Espert L, Degols G, Gongora C, Blondel D, Williams BR, Silverman RH, Mechti N. ISG20, a new interferoninduced RNase specific for single-stranded RNA, defines an alternative antiviral pathway against RNA genomic viruses. J Biol Chem. 2003; 278:16151–58. https://doi.org/10.1074/jbc.M209628200 PMID:12594219
- 21. Espert L, Degols G, Lin YL, Vincent T, Benkirane M, Mechti N. Interferon-induced exonuclease ISG20 exhibits an antiviral activity against human immunodeficiency virus type 1. J Gen Virol. 2005; 86:2221-29. https://doi.org/10.1099/vir.0.81074-0 PMID:16033969
- 22. Degols G, Eldin P, Mechti N. ISG20, an actor of the innate immune response. Biochimie. 2007; 89:831-35. https://doi.org/10.1016/j.biochi.2007.03.006 PMID:17445960
- 23. Nguyen LH, Espert L, Mechti N, Wilson DM 3rd. The interferonhuman and estrogen-regulated ISG20/HEM45 gene product degrades single-stranded RNA and DNA in vitro. Biochemistry. 2001; 40:7174–79. https://doi.org/10.1021/bi010141t PMID:11401564
- 24. Taylor KL, Leaman DW, Grane R, Mechti N, Borden EC, Lindner DJ. Identification of interferon-beta-stimulated genes that inhibit angiogenesis in vitro. J Interferon Cytokine Res. 2008; 28:733-40. https://doi.org/10.1089/jir.2008.0030 PMID:18937547
- 25. Lin SL, Wu SM, Chung IH, Lin YH, Chen CY, Chi HC, Lin TK, Yeh CT, Lin KH. Stimulation of Interferon-Stimulated Gene 20 by Thyroid Hormone Enhances Angiogenesis in Liver Cancer. Neoplasia. 2018; 20:57-68. https://doi.org/10.1016/j.neo.2017.10.007 PMID:29195126
- 26. Qu L, Ding J, Chen C, Wu ZJ, Liu B, Gao Y, Chen W, Liu F, Sun W, Li XF, Wang X, Wang Y, Xu ZY, et al. Exosome-Transmitted IncARSR Promotes Sunitinib Resistance in Renal Cancer by Acting as a Competing Endogenous RNA. Cancer Cell. 2016; 29:653-68. https://doi.org/10.1016/j.ccell.2016.03.004 PMID:27117758
- 27. Elgendy M, Fusco JP, Segura V, Lozano MD, Minucci S, Echeveste JI, Gurpide A, Andueza M, Melero I, Sanmamed MF, Ruiz MR, Calvo A, Pascual JI, et al. Identification of mutations associated with acquired

resistance to sunitinib in renal cell cancer. Int J Cancer. 2019: 145:1991-2001.

https://doi.org/10.1002/ijc.32256 PMID:30848481

- 28. McDermott DF, Cheng SC, Signoretti S, Margolin KA, Clark JI, Sosman JA, Dutcher JP, Logan TF, Curti BD, Ernstoff MS, Appleman L, Wong MK, Khushalani NI, et al. The high-dose aldesleukin "select" trial: a trial to prospectively validate predictive models of response to treatment in patients with metastatic renal cell carcinoma. Clin Cancer Res. 2015; 21:561-68. https://doi.org/10.1158/1078-0432.CCR-14-1520 PMID:25424850
- 29. Hammers H. Immunotherapy in kidney cancer: the past, present, and future. Curr Opin Urol. 2016; 26:543-47. https://doi.org/10.1097/MOU.00000000000338 PMID:27533501
- 30. Escudier B. Emerging immunotherapies for renal cell carcinoma. Ann Oncol. 2012 (Suppl 8); 23:viii35-40. https://doi.org/10.1093/annonc/mds261 PMID:22918926
- 31. Kelderman S, Schumacher TN, Haanen JB. Acquired and intrinsic resistance in cancer immunotherapy. Mol Oncol. 2014; 8:1132-39. https://doi.org/10.1016/j.molonc.2014.07.011 PMID:25106088
- 32. Halestrap AP. The SLC16 gene family structure, role and regulation in health and disease. Mol Aspects Med. 2013; 34:337-49. https://doi.org/10.1016/j.mam.2012.05.003 PMID:23506875
- 33. Macintyre AN, Gerriets VA, Nichols AG, Michalek RD, Rudolph MC, Deoliveira D, Anderson SM, Abel ED, Chen BJ, Hale LP, Rathmell JC. The glucose transporter Glut1 is selectively essential for CD4 T cell activation and effector function. Cell Metab. 2014; 20:61-72. https://doi.org/10.1016/j.cmet.2014.05.004 PMID:24930970
- 34. Gale PA, Pérez-Tomás R, Quesada R. Anion transporters and biological systems. Acc Chem Res. 2013; 46:2801-13. https://doi.org/10.1021/ar400019p PMID:23551251
- 35. Schulte ML, Fu A, Zhao P, Li J, Geng L, Smith ST, Kondo J, Coffey RJ, Johnson MO, Rathmell JC, Sharick JT, Skala MC, Smith JA, et al. Pharmacological blockade of ASCT2-dependent glutamine transport leads to antitumor efficacy in preclinical models. Nat Med. 2018; 24:194-202.

https://doi.org/10.1038/nm.4464 PMID:29334372

36. Payen VL, Mina E, Van Hée VF, Porporato PE, Sonveaux P. Monocarboxylate transporters in cancer. Mol Metab. 2019. [Epub ahead of print].

https://doi.org/10.1016/j.molmet.2019.07.006 PMID:<u>31395464</u>

- 37. Guo C, Huang T, Wang QH, Li H, Khanal A, Kang EH, Zhang W, Niu HT, Dong Z, Cao YW. Monocarboxylate transporter 1 and monocarboxylate transporter 4 in cancer-endothelial co-culturing microenvironments promote proliferation, migration, and invasion of renal cancer cells. Cancer Cell Int. 2019; 19:170. <u>https://doi.org/10.1186/s12935-019-0889-8</u> PMID:31297034
- 38. Park YY, Sohn BH, Johnson RL, Kang MH, Kim SB, Shim JJ, Mangala LS, Kim JH, Yoo JE, Rodriguez-Aguayo C, Pradeep S, Hwang JE, Jang HJ, et al. Yes-associated protein 1 and transcriptional coactivator with PDZ-binding motif activate the mammalian target of rapamycin complex 1 pathway by regulating amino acid transporters in hepatocellular carcinoma. Hepatology. 2016; 63:159–72.

https://doi.org/10.1002/hep.28223 PMID:26389641

- Bachmann M, Kukkurainen S, Hytönen VP, Wehrle-Haller B. Cell Adhesion by Integrins. Physiol Rev. 2019; 99:1655–99. <u>https://doi.org/10.1152/physrev.00036.2018</u> PMID:<u>31313981</u>
- Labernadie A, Kato T, Brugués A, Serra-Picamal X, Derzsi S, Arwert E, Weston A, González-Tarragó V, Elosegui-Artola A, Albertazzi L, Alcaraz J, Roca-Cusachs P, Sahai E, Trepat X. A mechanically active heterotypic E-cadherin/N-cadherin adhesion enables fibroblasts to drive cancer cell invasion. Nat Cell Biol. 2017; 19:224–37.

https://doi.org/10.1038/ncb3478 PMID:28218910

- Lee JV, Berry CT, Kim K, Sen P, Kim T, Carrer A, Trefely S, Zhao S, Fernandez S, Barney LE, Schwartz AD, Peyton SR, Snyder NW, et al. Acetyl-CoA promotes glioblastoma cell adhesion and migration through Ca²⁺-NFAT signaling. Genes Dev. 2018; 32:497–511. <u>https://doi.org/10.1101/gad.311027.117</u> PMID:<u>29674394</u>
- 42. Bianchi C, Meregalli C, Bombelli S, Di Stefano V, Salerno F, Torsello B, De Marco S, Bovo G, Cifola I, Mangano E, Battaglia C, Strada G, Lucarelli G, et al. The glucose and lipid metabolism reprogramming is gradedependent in clear cell renal cell carcinoma primary cultures and is targetable to modulate cell viability and proliferation. Oncotarget. 2017; 8:113502–15. <u>https://doi.org/10.18632/oncotarget.23056</u> PMID:<u>29371925</u>
- Lucarelli G, Rutigliano M, Sallustio F, Ribatti D, Giglio A, Lepore Signorile M, Grossi V, Sanese P, Napoli A, Maiorano E, Bianchi C, Perego RA, Ferro M, et al. Integrated multi-omics characterization reveals a distinctive metabolic signature and the role of

NDUFA4L2 in promoting angiogenesis, chemoresistance, and mitochondrial dysfunction in clear cell renal cell carcinoma. Aging (Albany NY). 2018; 10:3957–85. https://doi.org/10.18632/aging.101685 PMID:30538212

- 44. Lucarelli G, Loizzo D, Franzin R, Battaglia S, Ferro M, Cantiello F, Castellano G, Bettocchi C, Ditonno P, Battaglia M. Metabolomic insights into pathophysiological mechanisms and biomarker discovery in clear cell renal cell carcinoma. Expert Rev Mol Diagn. 2019; 19:397–407. <u>https://doi.org/10.1080/14737159.2019.1607729</u> PMID:30983433
- 45. Li YJ, Sun L, Shi Y, Wang G, Wang X, Dunn SE, Iorio C, Screaton RA, Spaner DE. PPAR-delta promotes survival of chronic lymphocytic leukemia cells in energetically unfavorable conditions. Leukemia. 2017; 31:1905–14. <u>https://doi.org/10.1038/leu.2016.395</u> PMID:<u>28050012</u>
- 46. Zuo X, Xu M, Yu J, Wu Y, Moussalli MJ, Manyam GC, Lee SI, Liang S, Gagea M, Morris JS, Broaddus RR, Shureiqi I. Potentiation of colon cancer susceptibility in mice by colonic epithelial PPAR-δ/β overexpression. J Natl Cancer Inst. 2014; 106:dju052. <u>https://doi.org/10.1093/jnci/dju052</u> PMID:24681603
- 47. Gongora C, David G, Pintard L, Tissot C, Hua TD, Dejean A, Mechti N. Molecular cloning of a new interferon-induced PML nuclear body-associated protein. J Biol Chem. 1997; 272:19457–63. https://doi.org/10.1074/jbc.272.31.19457 PMID:9235947
- Weiss CM, Trobaugh DW, Sun C, Lucas TM, Diamond MS, Ryman KD, Klimstra WB. The Interferon-Induced Exonuclease ISG20 Exerts Antiviral Activity through Upregulation of Type I Interferon Response Proteins. MSphere. 2018; 3:e00209–18. <u>https://doi.org/10.1128/mSphere.00209-18</u> PMID:30232164
- 49. Qu H, Li J, Yang L, Sun L, Liu W, He H, Influenza A. Influenza A Virus-induced expression of ISG20 inhibits viral replication by interacting with nucleoprotein. Virus Genes. 2016; 52:759–67. <u>https://doi.org/10.1007/s11262-016-1366-2</u> PMID:27342813
- 50. Zheng Z, Wang L, Pan J. Estradiol and proinflammatory cytokines stimulate ISG20 expression in synovial fibroblasts of patients with osteoarthritis. Intractable Rare Dis Res. 2017; 6:269–73. <u>https://doi.org/10.5582/irdr.2017.01062</u> PMID:<u>29259855</u>
- 51. Rajkumar T, Sabitha K, Vijayalakshmi N, Shirley S, Bose MV, Gopal G, Selvaluxmy G. Identification and

validation of genes involved in cervical tumourigenesis. BMC Cancer. 2011; 11:80. https://doi.org/10.1186/1471-2407-11-80 PMID:21338529

- 52. Gao M, Lin Y, Liu X, Li Y, Zhang C, Wang Z, Wang Z, Wang Y, Guo Z. ISG20 promotes local tumor immunity and contributes to poor survival in human glioma. Oncolmmunology. 2018; 8:e1534038. <u>https://doi.org/10.1080/2162402X.2018.1534038</u> PMID:<u>30713788</u>
- Bauvois B. New facets of matrix metalloproteinases MMP-2 and MMP-9 as cell surface transducers: outside-in signaling and relationship to tumor progression. Biochim Biophys Acta. 2012; 1825:29–36. <u>https://doi.org/10.1016/j.bbcan.2011.10.001</u> PMID:<u>22020293</u>
- Wessely A, Waltera A, Reichert TE, Stöckl S, Grässel S, Bauer RJ. Induction of ALP and MMP9 activity facilitates invasive behavior in heterogeneous human BMSC and HNSCC 3D spheroids. FASEB J. 2019; 33:11884–93.

https://doi.org/10.1096/fj.201900925R PMID:<u>31366234</u>

55. Sokołowska J, Urbańska K. Immunohistochemical assessment of metalloproteinases MMP2 and MMP9 expression in canine various subtypes of lymphomas in relation with proliferative and apoptotic markers. Pol J Vet Sci. 2019; 22:203–11. <u>https://doi.org/10.24425/pjvs.2019.127087</u>

PMID:<u>31269334</u>

- 56. Nazir SU, Kumar R, Singh A, Khan A, Tanwar P, Tripathi R, Mehrotra R, Hussain S. Breast cancer invasion and progression by MMP-9 through Ets-1 transcription factor. Gene. 2019; 711:143952. https://doi.org/10.1016/j.gene.2019.143952 PMID:<u>31265880</u>
- 57. Huang G, Jiang M, Wu X, Xiang X, Zhang L, Yu J. KGFR and MMP9 expression are correlates with cancer growth in cervical carcinoma. Minerva Med. 2019. [Epub ahead of print]. <u>https://doi.org/10.23736/S0026-4806.19.06243-8</u> PMID:<u>31317692</u>
- Schröer N, Pahne J, Walch B, Wickenhauser C, Smola S. Molecular pathobiology of human cervical high-grade lesions: paracrine STAT3 activation in tumor-instructed myeloid cells drives local MMP-9 expression. Cancer Res. 2011; 71:87–97. <u>https://doi.org/10.1158/0008-5472.CAN-10-2193</u> PMID:<u>21199798</u>
- Liang Y, Yang N, Pan G, Jin B, Wang S, Ji W. Elevated IL-33 promotes expression of MMP2 and MMP9 via activating STAT3 in alveolar macrophages during LPS-

induced acute lung injury. Cell Mol Biol Lett. 2018; 23:52.

https://doi.org/10.1186/s11658-018-0117-x PMID:<u>30410547</u>

- Chen YJ, Chang LS. Simvastatin induces NFκB/p65 down-regulation and JNK1/c-Jun/ATF-2 activation, leading to matrix metalloproteinase-9 (MMP-9) but not MMP-2 down-regulation in human leukemia cells. Biochem Pharmacol. 2014; 92:530–43. <u>https://doi.org/10.1016/j.bcp.2014.09.026</u> PMID:<u>25316568</u>
- Wang H, Liu Y, Shi J, Cheng Z. ORMDL3 knockdown in the lungs alleviates airway inflammation and airway remodeling in asthmatic mice via JNK1/2-MMP-9 pathway. Biochem Biophys Res Commun. 2019; 516:739–46. https://doi.org/10.1016/j.bbrc.2019.06.122

PMID:31255288

- 62. Luo S, Li H, Mo Z, Lei J, Zhu L, Huang Y, Fu R, Li C, Huang Y, Liu K, Chen W, Zhang L. Connectivity map identifies luteolin as a treatment option of ischemic stroke by inhibiting MMP9 and activation of the PI3K/Akt signaling pathway. Exp Mol Med. 2019; 51:37. https://doi.org/10.1038/s12276-019-0229-z PMID:30911000
- 63. Chen S, Liu W, Wang K, Fan Y, Chen J, Ma J, Wang X, He D, Zeng J, Li L. Tetrandrine inhibits migration and invasion of human renal cell carcinoma by regulating Akt/NF-κB/MMP-9 signaling. PLoS One. 2017; 12:e0173725. https://doi.org/10.1371/journal.pone.0173725 PMID:28288190
- 64. Wang P, Zhang LD, Sun MC, Gu WD, Geng HZ. Overexpression of mir-124 inhibits MMP-9 expression and decreases invasion of renal cell carcinoma cells. Eur Rev Med Pharmacol Sci. 2018; 22:6308–14. <u>https://doi.org/10.26355/eurrev_201810_16041</u> PMID:<u>30338828</u>
- 65. Gong D, Zhang J, Chen Y, Xu Y, Ma J, Hu G, Huang Y, Zheng J, Zhai W, Xue W. The m⁶A-suppressed P2RX6 activation promotes renal cancer cells migration and invasion through ATP-induced Ca²⁺ influx modulating ERK1/2 phosphorylation and MMP9 signaling pathway. J Exp Clin Cancer Res. 2019; 38:233. <u>https://doi.org/10.1186/s13046-019-1223-y</u> PMID:<u>31159832</u>
- 66. Musgrove EA, Caldon CE, Barraclough J, Stone A, Sutherland RL. Cyclin D as a therapeutic target in cancer. Nat Rev Cancer. 2011; 11:558–72. <u>https://doi.org/10.1038/nrc3090</u> PMID:21734724
- 67. Lim S, Kaldis P. Cdks, cyclins and CKIs: roles beyond cell cycle regulation. Development. 2013; 140:3079–93.

https://doi.org/10.1242/dev.091744 PMID:23861057

 Eisfeld AK, Kohlschmidt J, Schwind S, Nicolet D, Blachly JS, Orwick S, Shah C, Bainazar M, Kroll KW, Walker CJ, Carroll AJ, Powell BL, Stone RM, et al. Mutations in the CCND1 and CCND2 genes are frequent events in adult patients with t(8;21)(q22;q22) acute myeloid leukemia. Leukemia. 2017; 31:1278–85. https://doi.org/10.1038/leu.2016.332 PMID:27843138

 laccarino I, Afify L, Aukema SM, Reddemann K, Schütt P, Flür M, Klapper W. t(11;14)-positive mantle cell lymphomas lacking cyclin D1 (CCND1) immunostaining because of a CCND1 mutation or exclusive expression of the CCND1b isoform. Haematologica. 2018; 103:e432–35. https://doi.org/10.3324/haematol.2018.192435 PMID:29773591

- 70. Xu J, Lin DI. Oncogenic c-terminal cyclin D1 (CCND1) mutations are enriched in endometrioid endometrial adenocarcinomas. PLoS One. 2018; 13:e0199688. <u>https://doi.org/10.1371/journal.pone.0199688</u> PMID:<u>29969496</u>
- 71. Ai B, Kong X, Wang X, Zhang K, Yang X, Zhai J, Gao R, Qi Y, Wang J, Wang Z, Fang Y. LINC01355 suppresses breast cancer growth through FOXO3-mediated transcriptional repression of CCND1. Cell Death Dis. 2019; 10:502. https://doi.org/10.1038/s41419-019-1741-8 PMID:31243265
- 72. Xu J, Wan X, Chen X, Fang Y, Cheng X, Xie X, Lu W. miR-2861 acts as a tumor suppressor via targeting EGFR/AKT2/CCND1 pathway in cervical cancer induced by human papillomavirus virus 16 E6. Sci Rep. 2016; 6:28968.

https://doi.org/10.1038/srep28968 PMID:27364926

- 73. Ritchie ME, Phipson B, Wu D, Hu Y, Law CW, Shi W, Smyth GK. limma powers differential expression analyses for RNA-sequencing and microarray studies. Nucleic Acids Res. 2015; 43:e47. <u>https://doi.org/10.1093/nar/gkv007</u> PMID:<u>25605792</u>
- 74. Yu G, Wang LG, Han Y, He QY. clusterProfiler: an R package for comparing biological themes among gene clusters. OMICS. 2012; 16:284–87.

https://doi.org/10.1089/omi.2011.0118 PMID:22455463

75. Langfelder P, Horvath S. WGCNA: an R package for weighted correlation network analysis. BMC Bioinformatics. 2008; 9:559. <u>https://doi.org/10.1186/1471-2105-9-559</u> PMID:19114008

76. Friedman J, Hastie T, Tibshirani R. Regularization Paths for Generalized Linear Models via Coordinate Descent. J Stat Softw. 2010; 33:1–22. <u>https://doi.org/10.18637/jss.v033.i01</u> PMID:20808728

 Simon N, Friedman J, Hastie T, Tibshirani R. Regularization Paths for Cox's Proportional Hazards Model via Coordinate Descent. J Stat Softw. 2011; 39:1–13.

https://doi.org/10.18637/jss.v039.i05 PMID:27065756

78. Subramanian A, Tamayo P, Mootha VK, Mukherjee S, Ebert BL, Gillette MA, Paulovich A, Pomeroy SL, Golub TR, Lander ES, Mesirov JP. Gene set enrichment analysis: a knowledge-based approach for interpreting genome-wide expression profiles. Proc Natl Acad Sci USA. 2005; 102:15545–50. <u>https://doi.org/10.1073/pnas.0506580102</u>

PMID:<u>16199517</u>

- 79. Mootha VK, Lindgren CM, Eriksson KF, Subramanian A, Sihag S, Lehar J, Puigserver P, Carlsson E, Ridderstråle M, Laurila E, Houstis N, Daly MJ, Patterson N, et al. PGC-1alpha-responsive genes involved in oxidative phosphorylation are coordinately downregulated in human diabetes. Nat Genet. 2003; 34:267–73. https://doi.org/10.1038/ng1180 PMID:12808457
- 80. Wang K, Xu T, Ruan H, Xiao H, Liu J, Song Z, Cao Q, Bao L, Liu D, Wang C, Cheng G, Liang H, Chen Z, et al. LXRα promotes cell metastasis by regulating the NLRP3 inflammasome in renal cell carcinoma. Cell Death Dis. 2019; 10:159. https://doi.org/10.1038/s41419-019-1345-3

PMID:30770793

 Liu Y, Cheng G, Song Z, Xu T, Ruan H, Cao Q, Wang K, Bao L, Liu J, Zhou L, Liu D, Yang H, Chen K, Zhang X. RAC2 acts as a prognostic biomarker and promotes the progression of clear cell renal cell carcinoma. Int J Oncol. 2019; 55:645–56.

https://doi.org/10.3892/ijo.2019.4849 PMID:31364727