LRRK2 is a Candidate Prognostic Biomarker for Clear Cell Renal Cell Carcinoma

Chunxiu Yang  
Wuhan University Zhongnan Hospital

Jingjing Pang  
Wuhan University Zhongnan Hospital

Jian Xu  
Wuhan University Zhongnan Hospital

He Pan  
Wuhan University Zhongnan Hospital

Yueying Li  
Wuhan University Zhongnan Hospital

Huainian Zhang  
Wuhan University Zhongnan Hospital

Huan Liu  
Wuhan University Zhongnan Hospital

Shu-Yuan Xiao (✉️ syxiao@whu.edu.cn)  
Zhongnan hospital of Wuhan University  https://orcid.org/0000-0003-0484-932X

Research Article

Keywords: LRRK2, clear cell renal cell carcinoma, bioinformatic analysis, prognosis biomarker, HIF1A, EGFR, tumorigenesis, pathology

DOI: https://doi.org/10.21203/rs.3.rs-419435/v1

License: ☐ ☀️ This work is licensed under a Creative Commons Attribution 4.0 International License.  
Read Full License
Abstract

Background: Clear cell renal cell carcinoma (ccRCC), derived from renal tubular epithelial cells, is the most common malignant tumor of the kidney. The study of key genes related to the pathogenesis of ccRCC has become important for gene target therapy.

Methods: Bioinformatics analysis of The Cancer Genome Atlas (TCGA) and the NCBI Gene Expression Omnibus (GEO) database were performed to examine the expression pattern and prognostic significance of leucine-rich repeat kinase 2 (LRRK2) expression and its relationship to clinical parameters. Immunohistochemistry and Western blot were performed to verify LRRK2 expression.

Results: Bioinformatics analysis showed that LRRK2 expression was up-regulated in ccRCC, which was confirmed in ccRCC tissue immunohistochemically and by protein analysis. The level of expression was related to gender, pathological grade, stage and metastatic status of ccRCC patients. Meanwhile, Kaplan-Meier analysis showed that high expression of LRRK2 correlates to a better prognosis; protein-protein interaction network analysis showed that LRRK2 interacts with HIF1A and EGFR.

Conclusion: We found that LRRK2 may play an important role in the tumorigenesis and progression of ccRCC. Our findings provided a potential predictor and therapeutic target in ccRCC.

Background

Renal cell carcinoma (RCC) represents a highly heterogeneous group of tumor, with clear cell renal cell carcinoma (ccRCC) being the most common histologic subtype[1]. Other subtypes include papillary RCC, chromophobe RCC, clear cell papillary RCC, and several other rare types. ccRCC is believed to derive from the proximal convoluted renal tubules. Its incidence increases with age, so it is more prevalent in the elderly, more common in male patients[2]. The common clinical manifestations of ccRCC are hematuria, pain and renal mass. Histologically, ccRCC is characterized by high glycogen and lipid-rich cytoplasm. There is a genetic predisposition or hereditary factor associated with its tumorigenesis, with smoking, obesity, hypertension, chronic kidney disease and other environmental factors being contributory[3, 4]. Although molecular genetic studies have shown that mutations of several genes are associated with the pathogenesis of ccRCC, including von Hippel-Lindau (VHL), set domain-containing 2 (SETD2), BRCA1-associated protein-1 (BAP1), polybromo-1 (PBRM1), and lysine-specific histone demethylase 5C (KDM5C) [5–8], additional genes are being identified to be related to RCC from cancer genomic studies, which may have prognostic, predictive and therapeutic relevance[9]. It has been shown that loss of the short arm of chromosome 3 is closely related to ccRCC, and the central molecular signature is the inactivation of the tumor suppressor VHL gene[5, 10]. Also, myo-inositol monophosphatase 2 (IMPA2) downregulation is correlated with poor prognosis for ccRCC, and miR-25-mediated IMPA2 downregulation may be a potential therapeutic target for preventing the progression and metastatic of ccRCC[11].

Considering the high morbidity and mortality associated with ccRCC, it is essential to identify more molecular biomarkers that have diagnostic and prognostic value. In the present study, we aim to explore
the expression of differential genes in renal cell carcinoma by analyzing data from independent public
databases, and verifying the putative candidate by analyzing tumor and non-tumor control tissues. Our
results show that leucine-rich repeat kinase 2 (LRRK2) is a prognostic biomarker for ccRCC.

Materials And Methods

1. Data collection and bioinformatic analysis

All relevant data are available from the public domain databases. The RNA-seq FPKM data and
corresponding clinicopathological data of 611 samples with renal clear cell carcinoma were obtained
from The Cancer Genome Atlas databases (TCGA, https://portal.gdc.cancer.gov/). All of the 191 ccRCC
sample series matrix files were downloaded from the NCBI Gene Expression Omnibus databases (GEO,
https://www.ncbi.nlm.nih.gov/geo/), including 2 gene microarray datasets (GSE53757 and GSE71963).
The background correction and normalization were performed using the Robust Multi-array Average
(RMA) algorithm in R package “Affy”.

2. Screening for hub genes

For the two GEO datasets, the R package “limma” was applied to identify differentially expressed genes
(DEGs). The R package “EdgeR” was used to screen out DEGs based on the TCGA RNA-seq FPKM data.
All DEGs were filtered by setting for $p < 0.05$ and $|\logFC| > 1$ as cut-off criteria. The adjacency matrix was
converted to topological overlap matrix (TOM) dissimilarity matrix (1-TOM) by “tomlikeity” arithmetic, and
a scale-free topology of gene co-expression network was constructed through the R package “WGCNA”
based on genes from the GSE71963 dataset. Correlation between the module eigengenes (MEs) and
clinical trait was calculated by Pearson’s correlation analysis to identify clinic significant modules.
Subsequently, hub genes were considered those highly correlated with clinical traits as well as both the
DEGs of GEO and TCGA datasets.

Prognostic analysis was performed by using the R package “survival” based on TCGA-KIRC
clinicopathological data of 539 tumor and 72 normal samples. A univariate Cox regression analysis was
used to evaluate the association between the hub gene expression and overall survival; prognostic-related
genes were defined with $p < 0.001$ cutoff. STRING (https://string-db.org/) was used to construct the
protein-protein interaction (PPI) networks related to prognostically-related genes expressed with the
retrieval condition of Organism: Homo sapiens, and a minimum required interaction score of medium
confidence (0.400). Subsequently, a simple tabular text exported from STRING was input to the
Cytoscape v3.7.1 to obtain the top10 nodes with “betweenness” ranking method performed by
cytoHubba, which were finally defined as key hub genes.

3. Gene ontology (GO) terms and KEGG pathways
To further investigate the functional annotation and pathways of the prognostic-related genes, the R package “clusterProfiler” was adopted to carry out GO terms and KEGG pathway analyses, identified based on a threshold of adjusted $p < 0.05$.

4. Validations through other online databases

An extensive search in PubMed about the key hub genes and ccRCC was conducted to exclude genes that had been previously reported. The remaining candidate genes were validated with the Human Protein Atlas (HPA, https://www.proteinatlas.org/) by comparing the expression specificity and expression levels.

5. Gene Set Enrichment Analysis (GSEA)

Gene set enrichment analysis (GSEA) version 3.0 was used to predict the LRRK2 expression enriched pathways in ccRCC using the TCGA-KIRC datasets. ccRCC samples were divided into high and low expression level groups according to the median expression of LRRK2. The GeneChip matrix were analyzed after normalization and the number of random combinations was set to 1000. The functional annotation dataset c2.cp.kegg.v7.1.symbols.gmt was downloaded from the molecular signatures database (MsigDB, https://www.gsea-msigdb.org/gsea/msigdb) as a reference.

6. Cell lines and culture

The human ccRCC cell line Caki-1 and human embryonic kidney cell line HEK-293T were purchased from Cell Bank of the Chinese Academy of Sciences (Shanghai, China) in September 2020. The cell lines were authenticated via STR profiling and no mycoplasma contamination. The cell lines were incubated in McCoy’s 5A Medium (Gibco) and Dulbecco's Modified Eagle Medium (HyClone), respectively, supplemented with 10% fetal bovine serum (Gibco) and 1% penicillin-streptomycin (Gibco) at 37°C in 5% CO₂ humidified atmosphere.

7. Western blotting

Cells and tissues were lysed in lysis buffer supplemented with proteinase inhibitors. Protein samples were separated on SDS-PAGE and transferred to the PVDF membrane (Millipore, USA). The membranes were blocked with 5% non-fat milk for 1 hour and incubated with the primary LRRK2 antibody (1:10000, rabbit, ab133474, Abcam, USA) or GAPDH antibody (rabbit, 5174T, CST, USA) at 4°C overnight, followed by washing and incubation with HRP-conjugated goat anti-rabbit antibody (1:40000, ab205718, Abcam, USA) for 60 min at room temperature. The ECL western blotting detection kit (WBKLS0100, Millipore, USA) was used to detect the resultant bands. All experiments were performed in triplicate at least.
8. Immunohistochemical staining

Formalin-fixed paraffin embedded (FFPE) ccRCC and adjacent non-tumor tissues were obtained from our pathology archives, from September 2019 to August 2020 (detailed information in Supplementary table 1). Sections in 5-µm-thickness were prepared from each tissue block for immunohistochemistry (IHC). Briefly, the paraffin sections were dewaxed and rehydrate for antigen retrieval and following elimination of the endogenous peroxidase activity, and then were blocked in 5% goat serum uid for 1 hour at room temperature. Next, the sections were incubated with primary LRRK2 antibody (1:500, rabbit, ab133474, Abcam, USA) at 4°C overnight followed by visualization with the Dako EnVision DAB (Dako Diagnostics AG, Switzerland). For LRRK2 protein, staining localized in the cytoplasm is consider positive. Images were captured using an Olympus BX51 microscope equipped with a DP74 digital camera. Image-Pro Plus software (version 6.0) was used to assess the area and density of the stained regions, and the integrated optical density (IOD) value was obtained. The mean densitometry of the digital image (magnification, ×20) was designated as representative LRRK2 staining intensity. The signal density of the tissue areas from five randomly selected fields were counted in a blinded manner and subjected to statistical analysis.

9. Statistical analysis

Bioinformatic statistics analyses were carried out by using R v 4.0.2. Wilcox test, Wilcoxon rank sum test and logistic regression analysis were used to examine the correlation between LRRK2 expression level and clinicopathological parameters of the TCGA ccRCC samples. Kaplan-Meier survival analysis and log-rank test were performed to draw the survival curve in order to evaluate the effect of LRRK2 on overall survival, with 95% confidence interval and logarithmic rank p value. Univariate and multivariate Cox regression analyses were applied to the comparison between LRRK2 expression and other clinicopathological parameters (age, gender, grade, and TNM stage) and predicting the independent prognostic-related hazard factors. Western blotting data were analyzed using the Image J software. Experimental statistics analyses were performed using Graphpad prism 8.

Results

1. Screening of differentially expressed genes in ccRCC

1.1 Identification of LRRK2 in ccRCC

Bioinformatic approaches as described above were used to identify the target genes; the work flow of analysis is shown in Fig. 1a. First, DEGs of ccRCC were obtained from the two GEO datasets and TCGA-KIRC, with a total of 929 genes. Second, a soft threshold parameter β = 20 was selected (Fig. 1b) and a hierarchical clustering tree constructed by the correlation coefficient between genes from GSE71963, and different branches of clustering tree representing different gene modules in various colors (Fig. 1c).
Based on the weighted correlation coefficient of genes, the genes were classified according to the expression patterns, and genes with similar patterns were classified as a module. In this way, we detected genes in blue module exhibiting the most positive correlation with the tumor (Fig. 1d & e). Finally, with the intersection of the above predicted genes, 267 hub genes were included in genes of both the blue module by WGCNA and DEGs in ccRCC (Fig. 2a). To further explore the pathway functional enrichment of the identified hub genes, we performed GO terms and KEGG pathways analyses. Based on GO annotations, functions of the hub genes were related to dendritic cell migration, fatty acid metabolic process, regulation of leukocyte migration, renal system development and small molecule catabolic process. In addition, the top KEGG pathways enrichment included breast cancer, HIF-1 signaling pathway, melanoma, p53 signaling pathway and peroxisomes (Fig. 2b and 2c). For the 267 hub genes, univariate regression analysis was performed with a $p$-value cut-off of 0.001, and 54 genes were filtered out as prognostic-related genes, which were mapped to the PPI network revealing 54 nodes and 42 edges (Fig. 2d). The top10 nodes were found based on the node degree calculated by cytoHubba in Cytoscape as the key hub genes (Fig. 2e & 2f).

Subsequently, PubMed searches for these 10 key hub genes and ccRCC allow us to exclude 8 genes, namely VWF, FLT1, CCND1, PECAM1, SPARCL1, FZD1, CDH5 and INSR, as already known to be associated with ccRCC. The remaining 2 genes, LRRK2 and PDGFD were considered as candidate novel prognosis-related differentially expression genes for ccRCC. Validation of both LRRK2 and PDGFD were performed in the HPA database for expression specificity and expression levels. The LRRK2 expression is higher than PDGFD. Therefore, LRRK2 was chosen as a target gene for further study.

1.2 LRRK2 overexpression in ccRCC and its correlation with clinicopathological characteristics

As described above, analyses from the TCGA RNA-seq data revealed LRRK2 to be significantly overexpressed in ccRCC as compared to that of non-tumor controls ($p < 0.001$). An example of the differentially expressed levels is shown in Fig. 3.

A Wilcoxon signed-rank test and logistic regression analysis were used to analyze the correlation between LRRK2 expression and clinicopathological characteristics. The results suggest that the upregulation of LRRK2 expression is related to lower tumor grade, stage and TMN, and LRRK2 overexpression is more frequently seen in women (Fig. 4). Logistic regression analysis showed that LRRK2 expression is associated with gender (male vs female, OR = 0.67, 95%CI: 0.47–0.96), grade (G1 vs G4, OR = 0.16, 95% CI: 0.04–0.53), stage (I vs II, OR = 0.55, 95% CI: 0.36–0.85; I vs IV, OR = 0.44, 95% CI: 0.26–0.72), T (T1 vs T3, OR = 0.51, 95% CI: 0.35–0.75; T1 vs T4, OR = 0.07, 95% CI: 0.00-0.39), M (M0 vs M1, OR = 0.54, 95% CI: 0.32–0.88) (Table 1).
1.3 Prognostic value of LRRK2 based on TCGA-KIRC large cohorts

Kaplan-Meier survival curve from TCGA cohorts revealed that higher LRRK2 expression is associated with better prognosis in ccRCC patients (Fig. 5a). In multivariate regression analysis, the hazard ratio (HR) of LRRK2 expression (HR = 0.97, 95% CI: 0.95–0.99) is less than 1, suggesting that higher LRRK2 expression correlates with better prognosis as improved overall survival (OS). In contrast, patient age (HR = 1.03, 95% CI: 1.02–1.05), tumor grade (HR = 1.46, 95% CI: 1.16–1.85) and stage (HR = 1.98, 95% CI: 1.56–2.52) are correlated with poor prognosis, as can be expected (Fig. 5b, Table 2).
1.4 Gene Set Enrichment Analysis (GSEA) for LRRK2

GSEA was conducted between two groups, the high LRRK2 expression and low LRRK2 expression, by calculating normalized enrichment score (NES) and selecting high LRRK2 expression enriched pathways (NOM \( p\text{-val} < 0.05 \), FDR \( q\text{-val} < 0.25 \)). We identified “Prostate Cancer”, “MTOR signaling pathway”, “RIG-I-like receptor signaling pathway”, “ERBB signaling pathway”, “JAK-STAT signaling pathway” and “Apoptosis” as the potential functional enriched pathways modulated by LRRK2 (Table 3 and Fig. 6).

| MSigDB collection                  | Gene set name                  | NES   | NOM \( p\text{-val} \) | FDR \( q\text{-val} \) |
|-----------------------------------|--------------------------------|-------|-------------------------|------------------------|
| c2.cp.kegg.v7.1.symbols.gmt       | Prostate Cancer                | 2.16  | 0.006                   | 0.093                  |
|                                   | MTOR signaling pathway         | 2.12  | 0.000                   | 0.068                  |
|                                   | RIG-I-like receptor signaling pathway | 2.11  | 0.002                   | 0.047                  |
|                                   | ERBB signaling pathway         | 2.10  | 0.002                   | 0.030                  |
|                                   | JAK-STATE signaling pathway    | 2.06  | 0.002                   | 0.031                  |
|                                   | Apoptosis                      | 1.99  | 0.011                   | 0.032                  |

2. The expression of LRRK2 is highly upregulated in patients with ccRCC tissue and cell lines

To verify the findings from the above databases studies, we conducted a series of experimental validations. Overall, 30 cases of ccRCC were retrieved from the pathology archives, from 22 (73%) male and 8 (27%) female patients. The median age is 62 years (range 44–91 years). Immunohistochemically, as shown in Fig. 7a, LRRK2 is highly expressed in ccRCC, but hardly expressed in the adjacent non-tumor tissue. Based on the intensity and density of IHC staining, LRRK2 IHC score of tumor tissue is significantly higher than that of the controls (Fig. 7b). Likewise, Western blotting results showed that the expression level of LRRK2 in ccRCC tissues was significantly higher than that in adjacent non-tumor tissues (Fig. 7c).

We further validated the expression of LRRK2 in ccRCC cell line by western blotting. Compared with the normal renal epithelial cell line HEK293T, expression of LRRK2 is significantly increased in the ccRCC cell line Caki-1 (Fig. 7d).

**Protein-protein interaction (PPI) network of LRRK2**
Previous studies have shown that in addition to loss of chromosome arm 3p, there can be loss of chromosome 14q in ccRCC. The latter contains \textit{HIF1A}[12]. The PPI network prediction analysis using STRING database showed that there is a potential interaction between LRRK2 and HIF1A (score = 0.941), indicating that LRRK2 may play a vital role in the pathogenesis of ccRCC (Fig. 7e). In addition, the PPI network prediction analysis showed a potential interaction between LRRK2 and EGFR (score = 0.921), and the interaction of LRRK2 and EGFR was analyzed using the GEPIA database (http://gepia.cancer-pku.cn/), suggesting the positive correlation between them (Fig. 7f).

**Discussion**

The prognosis of renal cell carcinoma (RCC) is difficult to predict. Therefore, it is essential to identify more reliable biomarkers to guide clinical management. In this study, we found that LRRK2 is upregulated in ccRCC, and high LRRK2 expression is associated with patient outcome.

LRRK2 is a kinase encoded by LRRK2 gene. It has a complex structure with multiple domains: a RAS of complex GTPase domain (ROC), a C-terminal of ROC domain (COR), and a Ser/Thr kinase domain[13]. The HPA database shows that LRRK2 protein is mainly expressed in renal tubular epithelial cells and immune cells. The biological functions of LRRK2 include protein translation, regulation of autophagy and axonal degeneration induced by α-synuclein[13, 14]. By analyzing multiple databases for expression of individual proteins, and confirming in clinical specimens as well as cell lines, we have found that LRRK2 is over-expressed in ccRCC. We speculate that LRRK2 can be a novel prognostic biomarker of renal cell carcinoma. Analysis of the TCGA datasets has shown that high expression of LRRK2 is related to gender of the patients, tumor grade, stage, metastatic status, and prognosis of ccRCC patients.

Previously, research on LRRK2 had mainly been focused on its role in Parkinson's disease, and expression of LRRK2 in renal cell carcinoma had not been studied. The protein-protein interaction network (PPI) analysis of the STRING database has shown that LRRK2 interacts with EGFR and HIF1A. Epidermal growth factor receptor (EGFR) signaling pathway plays a critical role in the pathogenesis and progression of renal cell carcinoma[15, 16], suggested that LRRK2 may play an important role in ccRCC as well. Moreover, HIF1A is a transcription factor that regulates the expression of several hypoxia responsive genes, including vascular endothelial growth factor (VEGF)[17], platelet derived growth factor (PDGF)[18], and glucose transporters GLUT1 and GLUT4[19]. Hypoxia is a key step in the occurrence and development of renal cell carcinoma, which is mainly regulated by tumor suppressor gene VHL, and the VHL-HIF1A-VEGFA protein axis is involved in the occurrence and development of renal cell carcinoma[20]. VHL mutation and inactivation can lead to the accumulation of HIF1A transcription factors, which can trigger VEGFA transcription to promote angiogenesis and play a key role in tumorigenesis and development[20, 21]. HIF1A is detected in about 70% of ccRCC and closely related to patient survival[22]. In addition, HIF1A is necessary for the clear cell phenotype[21, 23]. Overall, we suspect that LRRK2 may play a role in the occurrence of, and influence the development of RCC by regulating HIF1A, and ultimately affect the survival of ccRCC patients. However, the exact mechanisms need to be further studied.
In conclusion, it is identified by bioinformatics analysis and confirmed in tissue specimens that LRRK2 is expressed and up-regulated in ccRCC. Our results suggest that LRRK2 may be a potential target for ccRCC treatment, but additional studies are necessary for further clarifying the mechanisms.

Abbreviations
| Abbreviation | Full Form |
|--------------|-----------|
| ccRCC        | clear cell renal cell carcinoma |
| TCGA         | The Cancer Genome Atlas |
| GEO          | Gene Expression Omnibus |
| LRRK2        | leucine-rich repeat kinase 2 |
| RCC          | renal cell carcinoma |
| VHL          | von Hippel-Lindau |
| SETD2        | set domain-containing 2 |
| BAP1         | BRCA1-associated protein-1 |
| PBRM1        | polybromo-1 |
| KDM5C        | lysine-specific histone demethylase 5C |
| IMPA2        | myo-inositol monophosphatase 2 |
| DEGs         | differentially expressed genes |
| TOM          | topological overlap matrix |
| MEs          | module eigengenes |
| PPI          | protein-protein interaction |
| HPA          | Human Protein Atlas |
| GSEA         | Gene Set Enrichment Analysis |
| MsigDB       | molecular signatures database |
| IHC          | immunohistochemistry |
| IOD          | integrated optical density |
| HR           | hazard ratio |
| OS           | overall survival |
| RMA          | Robust Multi-array Average |
| ROC          | RAS of complex GTPase domain |
| COR          | C-terminal of ROC domain |
| EGFR         | Epidermal growth factor receptor |
| VEGF         | vascular endothelial growth factor |
| PDGF         | platelet derived growth factor |
Declarations

Ethics approval and consent to participate

This study was reviewed and approved by the Medical Ethical Committee of Zhongnan hospital.

Consent for publication

Not applicable.

Availability of data and materials

The datasets generated during the current study are available in the Gene Expression Omnibus (GEO) database (https://www.ncbi.nlm.nih.gov/geo/) and the cancer genome atlas (TCGA) database (https://portal.gdc.cancer.gov/).

Competing interests

The authors have no conflicts of interest to declare.

Funding

This work was supported by Zhongnan Hospital of Wuhan University Science, Technology and Innovation Seed Fund (znpy2019092).

Author’s contribution

Chunxiu Yang and Jingjing Pang: literature design and writing (These authors contributed equally). Chunxiu Yang, Jingjing Pang and He Pan: experiment conduction, acquisition, statistical analysis and verification of data. All author: technical and material support. Shu-Yuan Xiao: design, writing, editing and final proof.

Acknowledgements

Not applicable.

References
1. Parikh M, Lara PN, Jr.: Modern Systemic Therapy for Metastatic Renal Cell Carcinoma of the Clear Cell Type. *Annual review of medicine* 2018, 69:209-221.

2. Hsieh JJ, Purdue MP, Signoretti S, Swanton C, Albiges L, Schmidinger M, Heng DY, Larkin J, Ficarra V: Renal cell carcinoma. *Nature reviews Disease primers* 2017, 3:17009.

3. Capitanio U, Bensalah K, Bex A, Boorjian SA, Bray F, Coleman J, Gore JL, Sun M, Wood C, Russo P: Epidemiology of Renal Cell Carcinoma. *European urology* 2019, 75(1):74-84.

4. Petejova N, Martinek A: Renal cell carcinoma: Review of etiology, pathophysiology and risk factors. *Biomedical papers of the Medical Faculty of the University Palacky, Olomouc, Czechoslovakia* 2016, 160(2):183-194.

5. Gossage L, Eisen T, Maher ER: VHL, the story of a tumour suppressor gene. *Nature reviews Cancer* 2015, 15(1):55-64.

6. Piva F, Santoni M, Matrana MR, Satti S, Giulietti M, Occhipinti G, Massari F, Cheng L, Lopez-Beltran A, Scarpelli M et al: BAP1, PBRM1 and SETD2 in clear-cell renal cell carcinoma: molecular diagnostics and possible targets for personalized therapies. *Expert review of molecular diagnostics* 2015, 15(9):1201-1210.

7. Sato Y, Yoshizato T, Shiraishi Y, Maekawa S, Okuno Y, Kamura T, Shimamura T, Sato-Otsubo A, Nagae G, Suzuki H et al: Integrated molecular analysis of clear-cell renal cell carcinoma. *Nature genetics* 2013, 45(8):860-867.

8. Dalgliesh GL, Furge K, Greenman C, Chen L, Bignell G, Butler A, Davies H, Edkins S, Hardy C, Latimer C et al: Systematic sequencing of renal carcinoma reveals inactivation of histone modifying genes. *Nature* 2010, 463(7279):360-363.

9. Morris MR, Latif F: The epigenetic landscape of renal cancer. *Nature reviews Nephrology* 2017, 13(1):47-60.

10. Zbar B, Brauch H, Talmadge C, Linehan M: Loss of alleles of loci on the short arm of chromosome 3 in renal cell carcinoma. *Nature* 1987, 327(6124):721-724.

11. Lin YF, Chou JL, Chang JS, Chiu IJ, Chiu HW, Lin YF: Dysregulation of the miR-25-IMPA2 axis promotes metastatic progression in clear cell renal cell carcinoma. *EBioMedicine* 2019, 45:220-230.

12. Linehan WM, Ricketts CJ: The Cancer Genome Atlas of renal cell carcinoma: findings and clinical implications. *Nature reviews Urology* 2019, 16(9):539-552.

13. Liu Z, Lenardo MJ: The role of LRRK2 in inflammatory bowel disease. *Cell research* 2012, 22(7):1092-1094.

14. Lin CH, Tsai PI, Wu RM, Chien CT: LRRK2 Parkinson's disease: from animal models to cellular mechanisms. *Reviews in the neurosciences* 2011, 22(4):411-418.

15. Liu C, Liu L, Wang K, Li XF, Ge LY, Ma RZ, Fan YD, Li LC, Liu ZF, Qiu M et al: VHL-HIF-2alpha axis-induced SMYD3 upregulation drives renal cell carcinoma progression via direct trans-activation of EGFR. *Oncogene* 2020, 39(21):4286-4298.
16. Wang S, Yu ZH, Chai KQ: Identification of EGFR as a Novel Key Gene in Clear Cell Renal Cell Carcinoma (ccRCC) through Bioinformatics Analysis and Meta-Analysis. *BioMed research international* 2019, **2019**:6480865.

17. Chen X, Liu J, He B, Li Y, Liu S, Wu B, Wang S, Zhang S, Xu X, Wang J: Vascular endothelial growth factor (VEGF) regulation by hypoxia inducible factor-1 alpha (HIF1A) starts and peaks during endometrial breakdown, not repair, in a mouse menstrual-like model. *Human reproduction* 2015, **30**(9):2160-2170.

18. Chen J, Cui X, Qian Z, Li Y, Kang K, Qu J, Li L, Gou D: Multi-omics analysis reveals regulators of the response to PDGF-BB treatment in pulmonary artery smooth muscle cells. *BMC genomics* 2016, **17**(1):781.

19. Hong J, Kim Y, Yanpallewar S, Lin PC: The Rho/Rac Guanine Nucleotide Exchange Factor Vav1 Regulates Hif-1alpha and Glut-1 Expression and Glucose Uptake in the Brain. *International journal of molecular sciences* 2020, **21**(4).

20. Wierzbicki PM, Klacz J, Kotulak-Chrzaszcz A, Wronska A, Stanislawowski M, Rybarczyk A, Ludziejewska A, Kmiec Z, Matuszewski M: Prognostic significance of VHL, HIF1A, HIF2A, VEGFA and p53 expression in patients with clear cell renal cell carcinoma treated with sunitinib as first line treatment. *International journal of oncology* 2019, **55**(2):371-390.

21. Hoefflin R, Harlander S, Schafer S, Metzger P, Kuo F, Schonenberger D, Adlesic M, Peighambari A, Seidel P, Chen CY *et al*: HIF-1alpha and HIF-2alpha differently regulate tumour development and inflammation of clear cell renal cell carcinoma in mice. *Nature communications* 2020, **11**(1):4111.

22. Fu L, Wang G, Shevchuk MM, Nanus DM, Gudas LJ: Generation of a mouse model of Von Hippel-Lindau kidney disease leading to renal cancers by expression of a constitutively active mutant of HIF1alpha. *Cancer research* 2011, **71**(21):6848-6856.

23. Gudas LJ, Fu L, Minton DR, Mongan NP, Nanus DM: The role of HIF1alpha in renal cell carcinoma tumorigenesis. *Journal of molecular medicine* 2014, **92**(8):825-836.

**Figures**
Figure 1

Modules of co-expressed genes in ccRCC a. Identification of soft threshold power (β) for scale-free network by using the "pickSoftThreshold" function; b. Cluster dendrogram of genes based on topological overlap matrix (TOM); c. Identification of modules associated clinical traits; d & e. Scatter plot of module gene significance (GS)/module membership (MM) in the blue module.
Figure 2

The screening procedures for key hub genes. a. Venn diagram among TCGA DEGs, GEO DEGs and genes from blue module by WGCNA; b & c. Gene ontology terms and functional pathways of KEGG for identified intersected genes; d. STRING analysis of prognostic-related genes; e. Top 10 nodes in STRING network ranked by “Betweenness” arithmetic; f. The rank of 10 nodes.

| Rank | Name   | Score  |
|------|--------|--------|
| 1    | VWF    | 177.619|
| 2    | FLT1   | 166.5238|
| 3    | CCND1  | 157    |
| 4    | PECAM1 | 140.8571|
| 5    | FZD1   | 120    |
| 6    | SPARCL1| 86     |
| 6    | LRRK2  | 86     |
| 8    | PDGFD  | 84     |
| 9    | CDH5   | 77.04762|
| 10   | INSR   | 44     |
Figure 3

LRRK2 differential overexpression in ccRCC tissues compared with normal control. a. LRRK2 expression level in both tumor and normal tissues based on TCGA-KIRC datasets; b. LRRK2 expression level in ccRCC tissues and paired adjacent non-tumor tissue.
Figure 4

Relationship between LRRK2 expression and clinical characteristics. a. Age (≤60 y and >60 y); b. Gender; c. Grade; d. Stage; e. Tumor size and local growth (T); f. Occurrence of distant metastases (M).
Figure 5

Prognostic analyses based on LRRK2 expression and overall survival (OS). A. Kaplan-Meier survival curve plotted based on a total of 530 patients; B. Multivariate Cox regression analysis of OS in TCGA cohorts grouped by the median of LRRK2 expression.
Figure 6

Six important enriched pathways involving LRRK2 in ccRCC according to GSEA. a. Prostate cancer pathway; b. MTOR signaling pathway; c. RIG-I-like receptor signaling pathway; d. ERBB signaling pathway; e. JAK-STAT signaling pathway; f. Apoptosis pathway.
Figure 7

Expression in kidney tissue and cell lines and protein-protein interaction of LRRK2. a. Immunohistochemical (IHC) staining of LRRK2 in paired non-tumor tissue (n=23, I. 10x and II. 20x) and clear cell renal cell carcinoma tissues (n=30, III. 10x and IV. 20x); b. IHC analysis of LRRK2 in non-tumor control and tumor tissues by image pro plus; c. LRRK2 expression in ccRCC tissues and paired adjacent non-tumor tissue by western blotting (n=3); d. LRRK2 expression in HEK293T and Caki-1 cell lines by...
western blotting (n=4); e. The protein-protein interaction network of LRRK2 by STRING database; f. Correlation between LRRK2 and EGFR in ccRCC using the GEPIA database.

**Supplementary Files**

This is a list of supplementary files associated with this preprint. Click to download.

- Supplementarytable1.docx