Identification and verification of vascular cell adhesion protein 1 as an immune-related hub gene associated with the tubulointerstitial injury in diabetic kidney disease

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ABSTRACT

Diabetic kidney disease (DKD) is the leading cause of chronic kidney disease (CKD) and end-stage renal disease (ESRD), but the pathogenesis is not completely understood. Tubulointerstitial injury plays critical roles in the development and progression of DKD. The present study aimed to investigate the profile of tubulointerstitial immune cell infiltration and reveal the underlying mechanisms between tubular cell injury and interstitial inflammation in DKD using bioinformatics strategies. First, xCell analysis identified immune cells displaying significant changes in the DKD tubulointerstitium, including upregulated CD4\(^+\) T cells, Th2 cells, CD8\(^+\) T cells, M1 macrophages, activated dendritic cells (DCs) and conventional DCs, as well as downregulated Tregs. Second, pyroptosis was identified as the main form of cell death compared with other forms of programmed cell death. Vascular cell adhesion protein 1 (VCAM1) was identified as the top ranked hub gene. The correlation analysis showed that VCAM1 was significantly positively correlated with pyroptosis and infiltrated immune cells in the tubulointerstitium. Upregulation of VCAM1 in the DKD tubulointerstitium was further verified in European Renal cDNA Bank cohort and was observed to negatively correlate with the glomerular filtration rate (GFR). Our in vitro study validated increased VCAM1 expression in HK-2 cells under diabetic conditions, and pyroptosis inhibition by disulfiram decreased VCAM1 expression, inflammatory cytokine release and fibrosis. In conclusion, our study identified upregulated VCAM1 expression in renal tubular cells, which might interact with infiltrated immune cells, thus promoting fibrosis. The FDA-approved drug disulfiram might improve fibrosis in DKD by targeting tubular pyroptosis and VCAM1 expression.

Introduction

Diabetic kidney disease (DKD), one of the common complications of diabetes, is the main cause of chronic kidney disease (CKD) and end-stage renal disease (ESRD) in many developed and developing countries [1,2]. According to a report from the China Kidney Disease Network (CK-NET), DKD accounts for 26.70% of all cases of CKD and imposes a large medical and economic burden [2]. The pathogenesis of DKD is complex and involves a multitude of different pathways. Clarifying the pathological characteristics and pathogenesis of DKD will help improve clinical management and identify new therapeutic targets.

Although glomerular damage is the main pathological feature of DKD, increasing evidence has recently revealed that tubulointerstitial pathology [3,4], such as tubular atrophy, interstitial fibrosis and interstitial infiltrating immune cells, plays a critical role in the development and progression of DKD [5–7]. Bioinformatics analysis represents an effective method to process large amounts of data within an extremely short time and provides valuable information about disease. However, bioinformatics studies investigating tubulointerstitial gene expression and immune cell infiltration in DKD are relatively rare.

The purpose of the study was to describe the characteristics of DKD tubulointerstitial immune
cell infiltration and to identify some key immune and inflammatory genes to provide novel insights into the pathogenesis and therapy of DKD. First, a tubulointerstitial microarray dataset of DKD was downloaded from the Gene Expression Omnibus (GEO) database. Using xCell [8], a web-based tool that performs a cell type enrichment analysis of gene expression data for 64 immune and stromal cell types, we first investigated the differences in immune cell infiltration between kidney tissues from individuals with DKD and normal controls. Second, a list of immune-related differentially expressed genes (DEGs) was screened, among which the top ranked gene, vascular cell adhesion protein 1 (VCAM1), was identified as an immune hub gene in DKD. Then, gene sets associated with various forms of cell death were obtained from GeneCards, and pyroptosis was evaluated as the main form of cell death using gene set variation analyses (GSVA) and gene set enrichment analysis (GSEA). Furthermore, a correlation analysis showed that VCAM1 was positively correlated with tubulointerstitial immune cell infiltration and pyroptosis. The upregulated expression of VCAM1 in the DKD tubulointerstitium was further verified in the European Renal cDNA Bank (ERCB) cohort and was observed to negatively correlate with renal function in patients with DKD. An in vitro study validated increased VCAM1 expression in HK-2 cells cultured under diabetic conditions, and pyroptosis inhibition by the FDA-approved drug disulfiram decreased VCAM1 expression, inflammatory cytokine release and fibrosis. Therefore, tubular pyroptosis and upregulated VCAM1 expression might be targets for immunotherapy of DKD.

**Methods**

**Microarray data**

The human gene expression dataset GSE30529 was downloaded from the National Center for Biotechnology Information (NCBI) Gene Expression Omnibus (GEO, http://www.ncbi.nlm.nih.gov/geo/) database. GSE30529 (GPL571 [HG-U133A_2] Affymetrix Human Genome U133A 2.0 Array) consisted of 10 DKD tubule samples and 12 control samples.

**Data processing**

The raw data in the GSE30529 dataset were downloaded and processed using the Limma package and Oligo package, respectively [9,10]. Data processing included background correction, normalization, and expression calculation. When multiple probes mapped to one gene symbol, the average expression level of the probes was calculated and regarded as the gene expression level of that gene. The probes that did not map to genes were removed.

**Immune and stromal cell analyses**

We applied a novel gene signature-based method, xCell [8], to estimate the cell type enrichment score and determine the profile of infiltrating immune cells in the tubulointerstitium of patients with DKD. xCell is a method for cell-type enrichment analysis using single-sample gene set enrichment analysis (ssGSEA) that calculates the enrichment scores for 64 cell types, including 34 types of immune cells, 30 types of stromal cells and other cells. It outperforms other extensive in silico analyses (including CIBERSORT) by encompassing more types of immune cells and employing a spillover compensation technique to reduce dependencies between closely related cell types. The 34 types of immune cells were categorized into nine groups, including CD4+ T cell subpopulations, CD8+ T cell subpopulations, gamma delta T cells (γδ cells), NK cells, NKT cells, B cell subpopulations, monocyte/macrophage subpopulations, dendritic cell (DC) subpopulations, and granulocyte subpopulations.

**Gene set enrichment analyses (GSEA) and gene set variation analyses (GSVA)**

Using the hallmark gene sets as the reference gene set, we performed gene set variation analyses (GSVA) between DKD and control tissues using the GSVA Bioconductor package [11]. The thresholds were set to enrichment score change > 1.0, p value < 0.05 and t value > 2. Both upregulated and downregulated pathways were identified. GSEA were performed with the GSEA desktop.
Construction of gene sets representing various forms of cell death and enrichment analysis

Gene sets representing various forms of cell death were constructed by searching the keywords ‘pyroptosis’, ‘necrosis’, ‘necroptosis’, ‘apoptosis’, ‘ferroptosis’ and ‘autophagy’ in the GeneCards database (https://www.genecards.org/). Nonprotein-coding genes were removed. The gene sets are listed in Table S2. Then, gene set GSEA were performed using the GSEA desktop application to evaluate the cell death form in the tubulointerstitium. We performed Pearson’s correlation analyses to explore the relationship between pivotal cell death pathways and infiltrating immune cells in the tubulointerstitium.

Analysis of DEGs and screening of immune hub genes

DEGs between patients with DKD and healthy controls were identified using the ‘Limma’ package [13] in R software. An adjusted P value<0.05 and log fold change (FC) ≥ 1 were set as the thresholds. A list of 1793 immune-related genes was downloaded from ImmPort Shared Data (https://www.immport.org/shared/genelists), and shared genes from the two datasets (DEGs and immune-relevant gene list) were determined by constructing a Venn diagram to identify immune-related hub genes. An analysis of the functional interactions between proteins might provide insights into the mechanisms of DKD. In the present study, the protein–protein interaction (PPI) networks of ninety-four shared genes were generated using the STRING database (https://string-db.org/) and visualized with Cytoscape software (version 3.7.1, http://www.cytoscape.org/). An interaction score greater than 0.4 (medium confidence) was considered statistically significant. CytoHubba, a Cytoscape plugin, was utilized to explore PPI network hub genes [14]. Among them, VCAM1 was the top ranked gene. Enrichment analyses of the thirty immune hub genes identified above were performed using the online tool Metascape [15]. Furthermore, dominant modules in the PPI network were identified using the MCODE plugin (version 1.4.2, http://apps.cytoscape.org/apps/MCODE) in Cytoscape software.

VCAM1 verification and correlation analysis with renal function

The differential expression of VCAM1 in the tubulointerstitium of patients with DKD was validated in the ERCB cohort (31 healthy controls and 17 patients with DKD). Then, we used the Nephroseq v5 online database (http://v5.nephroseq.org), an integrated data-mining platform for gene expression datasets of kidney diseases, to validate the correlation between VCAM1 expression and clinical traits of patients with DKD using Pearson’s correlation analysis in ERCB. A p value of <0.05 was considered statistically significant.

Immunofluorescence staining for VCAM1

Paraffin kidney sections from a patient clinicopathologically diagnosed with DKD were stained with VCAM1. This was approved by the Committee on Research Ethics of Peking University First Hospital (NO. 20171280). Nonspecific binding was blocked with 3% BSA following fixation and heat treatment for antigen retrieval. Kidney sections were then incubated with a primary antibody against VCAM1 (Abcam, ab134047, 1:200) overnight at 4°C. The Cy3-labeled secondary antibody was obtained from Jackson ImmunoResearch. Tumor-adjacent kidney tissue from nephrectomy samples was used as a normal control. Representative images were captured using a Leica DFC 7000 T camera via Leica Application Suite V4.7.1 software.

Cell culture

Human kidney 2 (HK-2) cells, a human kidney proximal tubular epithelial cell line, were cultured in DMEM (11885084, Gibco, USA) supplemented with 10% fetal bovine serum (10099141 C, Gibco, USA) and 1% penicillin/streptomycin (10378016, Gibco, USA) at 37°C with 5% CO₂. High glucose (30 mM D-glucose, G8644, Sigma) [16] and TNF-α (40 ng/ml, 300–01A, PeproTech) [17] were used to induce HK-2 cell injury in vitro. Disulfiram
were - GraphPad - Statistical (Tiangen, KR116). RNAsimple - RNA Cybernetics, sis - bated dilution VCAM1 - night fat-free denned PAGE) denatured Assay tration RIPA - Total - determined RIPA - next, denatured proteins were separated by sodium dodecyl sulfate-polyacrylamide gel electrophoresis (SDS-PAGE) and then electrically transferred to polyvinylidene difluoride membranes (Millipore, IPVH00010). The membranes were blocked for 60 minutes with 5% fat-free milk dissolved in Tris-buffered saline containing 0.1% Tween 20 (TBST). The blots were incubated with the following relevant primary antibodies overnight at 4°C: GSDMD (Abcam, ab209845, 1:1000), VCAM1 (Abcam, ab134047, 1:2000), and tubulin (ZSbio, TA-10, 1:5000). An incubation with a 1:1000 dilution of the HRP-conjugated secondary antibody was carried out for 1 hour at room temperature. After five washes with TBST, the membranes were incubated with the chemiluminescence substrate (Millipore, WBKLS0100) for 5 minutes, and images were captured using an Image Quant LAS 4000 Mini system (GE Healthcare). The semiquantitative analysis was conducted using ImageJ software (Media Cybernetics, Silver Spring, MD).

RNA isolation and RT–PCR analysis

Total RNA was extracted from HK-2 cells using an RNAsimple Total RNA Kit (Tiangen, DP419) according to the manufacturer’s instructions and reverse transcribed into cDNAs using a FastKing RT Kit (Tiangen, KR116). Quantitative PCR was performed using a StepOne Real-Time PCR System (Applied Biosystems, USA) with two-step methods. The sequences of the primers used are shown in Table S4. Comparative gene expression was calculated using the 2\(^{-ΔΔCt}\) method as described previously.

Statistical analysis

GraphPad Prism 6.0 software was used for statistical analyses, and data are presented as the means ± SEM. A two-tailed unpaired t test was applied for comparisons between two groups. Differences at the P < 0.05 level were considered statistically significant.

Results

Accumulating evidence has recently revealed that tubulointerstitial pathology plays a critical role in the development and progression of DKD. The purpose of the study was to describe the characteristics of DKD tubulointerstitial immune cell infiltration and to identify some key immune and inflammatory genes to provide novel insights into the pathogenesis and therapy of DKD. In this study, we investigated the profile of tubulointerstitial immune cell infiltration and identified pyroptosis as the main form of programmed cell death in DKD using bioinformatics analysis. VCAM1 was identified as the top ranked immune-related hub gene and was positively correlated with pyroptosis and infiltrated immune cells. Furthermore, VCAM1 expression was validated to be elevated in renal tubular cells cultured under diabetic conditions. The FDA-approved drug dulisifiram inhibited renal tubular cell pyroptosis and decreased VCAM1 expression, inflammatory cytokine levels and fibrosis in vitro.

1. Bioinformatics analysis workflows

The main steps of the workflow are shown in Figure 1. We first evaluated immune cell infiltration in tubulointerstitial tissues from patients with DKD. Then, pyroptosis, the main form of cell death in the tubulointerstitium, was discovered using GSEA and GSVA. Next, we screened a list of genes closely related to immune cell infiltration and identified hub genes. A correlation analysis was performed between infiltrated immune cells, pyroptosis and the top hub gene VCAM1. In addition, clinical validation and immunostaining validation of VCAM1 were performed. Finally, we further conducted an in vitro study to validate the results of the bioinformatics analysis.

2. Immune and stromal cell deconvolution analyses

We used xCell, which generates cell type enrichment scores using bulk gene expression data, to determine the cell types potentially involved in tubulointerstitial injury in diabetic kidney.
disease. The enrichment scores of 64 cell types, including 34 types of immune cells and 30 types of stromal and other cells, were obtained for each sample (Figure 2(a)). The heatmap of 34 immune cell enrichment scores was illustrated to identify the immune landscape of the DKD tubulointerstitium (Figure 2(b)). Compared with normal controls, the majority of T cell enrichment scores were relatively higher in patients with DKD, except for CD4+ Tcm, Tregs, Th1 cells, CD8+ naïve T cells, and NKT cells. The enrichment scores of aDCs and cDCs were higher in patients with DKD than in controls, while Pro-B cell enrichment scores were lower in patients with DKD. The enrichment scores of most monocyte/macrophage and granulocyte subsets were not significantly different between DKD and normal tissues, except for M1 macrophages and neutrophils and eosinophils (Table 1).

### 3. GSVA and GSEA of hallmark gene sets

We conducted GSVA and GSEA of hallmark gene sets in the GSE30529 dataset to explore the biological processes in the DKD tubulointerstitium. The ‘interferon alpha response’, ‘interferon gamma response’ and ‘complement’ signaling pathways were significantly activated in the pathogenesis of tubulointerstitial injury in patients with DKD (Figure 3(a)), and a summary of the GSVA results is displayed in Table S1. Then, we performed GSEA of the three gene sets and found that they were positively enriched in the dataset. The normalized enrichment scores (NESs) were 1.89, 1.89 and 1.70, with normalized p values (NOM p values) of 0.000, 0.002 and 0.019, respectively (Figure 3(b-d)). All of the significantly enriched pathways identified using GSEA are shown in Table 2. These data suggest that the tubulointerstitial immune response plays important roles in the pathogenesis of DKD.

### 4. GSEA and GSVA of gene sets associated with various forms of cell death

Cell death includes pyroptosis, necrosis, necroptosis, apoptosis, ferroptosis and autophagy, which have different pathophysiological mechanisms...
and signaling pathways. Various forms of programmed cell death have been reported to be associated with inflammatory and immune responses. We constructed gene sets of several types of cell death from GeneCards (Table S2) and then performed both GSEA and GSVA to explore the involvement of cell death in tubulointerstitial inflammation in patients with DKD. GSEA showed that only the pyroptosis gene sets were significantly enriched in patients with DKD compared to normal controls (NES = 1.65, FDR q value = 0.023, FWER = 0.013) (Figure 3(e)). Meanwhile, the GSVA score for pyroptosis was the highest
Table 1. Immune cell enrichment score between DKD patients and normal controls.

| CD4 T-cell subpopulations            | DKD (n = 10) Mean±SEM | Controls (n = 12) Mean±SEM | P       |
|--------------------------------------|------------------------|-----------------------------|---------|
| CD4+ naïve T cells                   | 0.2182 ± 0.04533       | 0.02152 ± 0.01897           | 0.0004  |
| CD4+ memory T cells                  | 0.03083 ± 0.01899      | 0.0 ± 0.0                   | 0.089   |
| CD4+ Tem cells                       | 0.3547 ± 0.06656       | 0.03815 ± 0.01823           | < 0.0001|
| CD4+ Tcm                             | 0.08278 ± 0.02926      | 0.2400 ± 0.04904            | 0.0166  |
| CD4+ T cells                         | 0.2146 ± 0.07198       | 0.1041 ± 0.03488            | 0.1605  |
| Tregs                                | 0.1340 ± 0.05184       | 0.3015 ± 0.04379            | 0.0218  |
| Th1 cells                            | 0.2988 ± 0.07070       | 0.2918 ± 0.05794            | 0.9388  |
| Th2 cells                            | 0.1490 ± 0.02815       | 0.00575 ± 0.003196          | < 0.0001|

| CD8 T-cell subpopulations            |                        |                             |         |
|--------------------------------------|------------------------|-----------------------------|---------|
| CD8+ naïve T cells                   | 0.1662 ± 0.04715       | 0.04317 ± 0.02055           | 0.0195  |
| CD8+ Tcm                             | 0.1898 ± 0.05078       | 0.4839 ± 0.04030            | 0.0002  |
| CD8+ Tem                             | 0.1845 ± 0.03968       | 0.00305 ± 0.00305           | < 0.0001|
| CD8+ Tem                             | 0.1787 ± 0.05322       | 0.02509 ± 0.01853           | 0.0083  |
| Tyró cells                           | 0.1690 ± 0.02973       | 0.02208 ± 0.009720          | < 0.0001|
| NKT cells                            | 0.6282 ± 0.17111       | 1.212 ± 0.1347              | 0.0132  |
| NK cells                             | 0.05611 ± 0.01876      | 0.0 ± 0.0                   | 0.0036  |

| B-cell subpopulations                |                        |                             |         |
|--------------------------------------|------------------------|-----------------------------|---------|
| Pro B cells                          | 0.01628 ± 0.008225     | 0.08178 ± 0.02194           | 0.0174  |
| B cells                              | 0.01107 ± 0.007845     | 0.004783 ± 0.004783         | 0.4861  |
| Naive B cells                        | 0                      | 0                           | na      |
| Memory B cells                       | 0.00178 ± 0.00178      | 0.0 ± 0.0                   | 0.284   |
| Class-switched memory B cells        | 0.09659 ± 0.03978      | 0.07359 ± 0.01983           | 0.5918  |
| Plasma cells                         | 0.04556 ± 0.01902      | 0.0240 ± 0.01357            | 0.3563  |

| Monocyte/macrophage subpopulations   |                        |                             |         |
|--------------------------------------|------------------------|-----------------------------|---------|
| Monocytes                            | 0.0131 ± 0.007538      | 0.0 ± 0.0                   | 0.0701  |
| Macrophages                          | 0.05801 ± 0.02097      | 0.003867 ± 0.002350         | 0.0107  |
| Macrophages M1                       | 0.02475 ± 0.008952     | 0.0006083 ± 0.0006083       | 0.0078  |
| Macrophages M2                       | 0.1425 ± 0.03050       | 0.1866 ± 0.03174            | 0.3341  |

| DC subpopulations                    |                        |                             |         |
|--------------------------------------|------------------------|-----------------------------|---------|
| DCs                                  | 0.1409 ± 0.03783       | 0.1027 ± 0.01948            | 0.3565  |
| Activated DCs                        | 0.1304 ± 0.02466       | 0.02792 ± 0.009929          | 0.0005  |
| Conventional DCs                     | 0.1838 ± 0.03091       | 0.0 ± 0.0                   | < 0.0001|
| Plasmacytoid DCs                     | 0.02108 ± 0.01302      | 0.003875 ± 0.003875         | 0.1862  |
| Immature DCs                         | 0.1495 ± 0.02799       | 0.1745 ± 0.01547            | 0.423   |

| Granulocyte subpopulations           |                        |                             |         |
|--------------------------------------|------------------------|-----------------------------|---------|
| Neutrophils                          | 0.06308 ± 0.01922      | 0.01414 ± 0.01003           | 0.028   |
| Eosinophils                          | 0.5600 ± 0.1245        | 0.9532 ± 0.1106             | 0.0281  |
| Mast cells                           | 0.05088 ± 0.009363     | 0.04652 ± 0.008655          | 0.7361  |
| Basophils                            | 0.5578 ± 0.1299        | 0.6611 ± 0.09912            | 0.5276  |

NOTE: Bold indicates statistical significance.
Abbreviations: DC, dendritic cell; Tem, effector memory T cell; Tyró cell, gamma delta T cell.

among all of the various types of cell death analyzed (Table S3). Based on this result, pyroptosis might play a role in the pathogenesis of tubulointerstitial injury in patients with DKD.

5. Correlation between pyroptosis enrichment scores and immune cell infiltration

GSVA enrichment scores for pyroptosis were correlated with the proportion of immune cell infiltration by performing Pearson’s correlation analysis to evaluate the correlation between pyroptosis and infiltration of immune cells. We observed a significant correlation between 15 types of immune cells and pyroptosis (Figure 4), suggesting a close interaction between immune cells and pyroptosis.

6. Discovery of core genes

A differential expression analysis was performed to further explore the relationship between renal tubular cells and interstitial immune cells. Four hundred eight DEGs were identified between the DKD group and the control group, among which 65 were downregulated and 343 upregulated (Figure 5). An immune-related gene list was
downloaded to identify immune-related DEGs, and ninety-four hub immune genes were acquired from two datasets (DEGs and the immune-related gene list) via a Venn diagram (Figure 6(a)). The PPI analysis of these ninety-four hub immune genes revealed 97 nodes and 694 interactions (Figure S1). The top 30 nodes calculated by the MCC algorithm were identified as hub genes (Figure 6(b) and Table 3). Then, an enrichment analysis of hub genes was performed. ‘Allograft
Table 2. Significantly enriched hallmark gene sets using GSEA.

| GS follow link to MSigDB | GS DETAILS                                    | NES | NOM p-val | FDR q-val | LEADING EDGE |
|--------------------------|-----------------------------------------------|-----|-----------|-----------|--------------|
| 1                        | HALLMARK INTERFERON ALPHA RESPONSE            | 1.89| 0.000     | 0.062     | tags = 71%, list = 13%, signal = 81% |
| 2                        | HALLMARK INTERFERON GAMMA RESPONSE            | 1.89| 0.002     | 0.033     | tags = 61%, list = 14%, signal = 70% |
| 3                        | HALLMARK ANGIogenesis                         | 1.81| 0.006     | 0.048     | tags = 37%, list = 8%, signal = 40%  |
| 4                        | HALLMARK E2F TARGETS                          | 1.71| 0.006     | 0.105     | tags = 48%, list = 21%, signal = 60%  |
| 5                        | HALLMARK COMPLEMENT                           | 1.70| 0.019     | 0.099     | tags = 35%, list = 13%, signal = 39%  |
| 6                        | HALLMARK EPITHELIA MESENCHYMAL TRANSITION     | 1.66| 0.046     | 0.103     | tags = 43%, list = 14%, signal = 49%  |
| 7                        | HALLMARK G2M CHECKPOINT                       | 1.61| 0.028     | 0.140     | tags = 42%, list = 19%, signal = 51%  |
| 8                        | HALLMARK ALLOGRAFT REJECTION                  | 1.55| 0.115     | 0.196     | tags = 43%, list = 14%, signal = 49%  |
| 9                        | HALLMARK APOPTOSIS                            | 1.54| 0.020     | 0.184     | tags = 31%, list = 14%, signal = 35%  |
| 10                       | HALLMARK MITOTIC SPINDLE                     | 1.52| 0.037     | 0.187     | tags = 30%, list = 16%, signal = 35%  |
| 11                       | HALLMARK UV RESPONSE DN                      | 1.48| 0.044     | 0.225     | tags = 46%, list = 22%, signal = 58%  |

7. Correlation between VCAM1 expression and infiltrating immune cells

Among the hub genes identified, the top ranked gene, VCAM1, was identified as an immune hub gene in DKD. VCAM1 expression was significantly increased in patients with DKD (Figure 7(a)). The correlation analysis revealed that VCAM1 expression was positively correlated with CD4+ T cells (r = 0.6877, P = 0.0004), CD4+ T memory cells (r = 0.7102, P = 0.0002), Th2 cells (r = 0.7358, P < 0.0001), CD8+ Tcm (r = 0.6026, P = 0.0030), Tγδ cells (r = 0.5864, P = 0.0041), NK cells (r = 0.5692, P = 0.0057), macrophages (r = 0.5068, P = 0.0161), M1 macrophages (r = 0.4441, P = 0.0384), aDCs (r = 0.4814, P = 0.0233) and cDCs (r = 0.6562, P = 0.0009) but negatively correlated with naïve CD8+ T cells (r = −0.4909, P = 0.0204) and NKT cells (r = −0.5548, P = 0.0074) (Figure 7(b)).

8. Correlation between VCAM1 expression and pyroptosis

Pearson’s correlation analysis was performed to explore whether the upregulated VCAM1 expression in the tubulointerstitium of patients with DKD is related to pyroptosis. VCAM1 expression was closely related to pyroptosis (r = 0.6118, P = 0.0025) (Figure 7(c)), indicating that pyroptosis might induce VCAM1 expression.

9. Verification, clinical relevance and immunofluorescence validation of VCAM1 expression

We performed a relevant analysis of patients with DKD from ERCB to validate the change in VCAM1 expression in the DKD tubulointerstitium. Consistently, VCAM1 mRNA levels were significantly higher in patients with DKD (Figure 8(a)). VCAM1 expression was negatively correlated with the glomerular filtration rate (GFR) (r = −0.5920, P = 0.0123) (Figure 8(b)) and positively correlated with serum creatinine levels (r = 0.5189, P = 0.0328) (Figure 8(c)). Furthermore, we performed immunofluorescence staining for VCAM1 in the kidney tissue of a patient with DKD and found that renal tubular VCAM1 expression was significantly upregulated compared with that in normal kidneys (Figure 8(d)).
10. DSL inhibited HG-induced HK-2 cell pyroptosis, VCAM1 expression and inflammation

We exposed cultured HK-2 cells to high glucose (HG) and TNF-α for 48 h to further determine the relationship between tubular cell pyroptosis, VCAM1 expression and the tubulointerstitial immune response in individuals with DKD. Since the core of cellular pyroptosis is membrane pore formation induced by cleaved GSDMD, we detected the levels of the GSDMD N-terminal domain (GSDMD-NT) and VCAM1 using western
blotting. Both HG and TNF-α induced a significant increase in the GSDMD-NT level, indicating the occurrence of GSDMD-mediated pyroptosis (Figure 9(a)). VCAM1 expression was increased statistically significantly in the HG and TNF-α groups (Figure 9(b)). Disulfiram (DSL) is an effective inhibitor of GSDMD pore formation [18]. Pretreatment with DSL alleviated the increase in GSDMD-NT levels, indicating that cell pyroptosis was suppressed. Meanwhile, VCAM1 expression was decreased in the DSL-pretreated HG and TNF-α groups (Figure 9(a-b)). Next, the inflammatory cytokine profile was determined using qPCR. The results showed that the upregulated levels of the MCP-1, IL-1β, and IL-18 mRNAs induced by HG were significantly decreased after DSL pretreatment. In the TNF-α group, the level of IL-1β decreased significantly in the DSL-pretreated group, and the levels of MCP-1 and IL-18 did not change significantly. No significant changes in the levels of the IL-6 and IL-8 mRNAs were found in any group. Importantly, TGF-β1 mRNA levels were also decreased in the DSL-pretreated groups. Thus, pyroptotic tubular cells increased VCAM1 expression under diabetic conditions and that a treatment abrogating pyroptosis inhibited VCAM1 expression, inflammation and fibrosis in HK-2 cells.

**Discussion**

Currently available clinical approaches to treating DKD remain limited and mainly involve strict control of hyperglycemia, proteinuria and blockade of the renin-angiotensin system. Considering the high incidence of DKD-related ESRD, novel and satisfactory strategies are urgently needed. Accumulating evidence indicates a paramount role for immunity and inflammation in the pathogenesis of DKD. To date, several studies have screened genes related to glomerular immune cell infiltration in individuals with DKD [20]. Relevant tubulointerstitial studies are rare. In the present study, we analyzed immune cell infiltration and relevant cellular processes in the tubulointerstitium using bioinformatics analysis. Pyroptosis was identified as the main form of cell death among other forms of programmed cell death using GSEA and GSVA. VCAM1 was identified as the top immune-related hub gene and was correlated with renal function in patients with DKD. An *in vitro* study validated increased VCAM1 expression in HK-2 cells cultured under diabetic conditions, and pyroptosis inhibition by disulfiram decreased VCAM1 expression, inflammatory cytokine release and fibrosis. Our study identified a close relationship between renal proximal
tubular pyroptosis, VCAM1 expression and the interstitial immune response in patients with DKD. This study provided new insights into the tubulointerstitial pathogenesis of DKD and helped identify new therapeutic targets.

xCell was used to conduct a comprehensive evaluation of immune cell infiltration in individuals with DKD. The results revealed that both innate immunity and adaptive immunity were involved in tubulointerstitial injury in patients with DKD. Macrophages and DCs constitute the mononuclear phagocyte (MNP) system [21], which is particularly complex in the kidney [22]. We found that macrophages, especially M1 macrophages, were significantly enriched in the tubulointerstitium. Interstitial macrophage M1 infiltration was reported to be correlated with interstitial fibrosis, tubular atrophy, renal function and proteinuria in both animal models and human kidneys [23–25]. M1 macrophages produce cytokines and proteins (e.g., Nitric oxide, platelet derived growth factor, IL-1, and TNF-α), causing damage to vascular endothelial cells and the proliferation of fibroblasts and mesangial cells, thus

Figure 6. Identification of hub genes and enrichment analysis. (a) Venn diagram of DEGs and immune-related gene list. (b) Top thirty hub immune genes identified via PPI network. (c) Hallmark, KEGG and GO enrichment analysis of the top thirty hub immune genes. (d) Two most significant clusters of DEGs were identified by MCODE plugin.
Table 3. Top 30 in network string_interactions.tsv ranked by MCC method.

| Rank | Name     | Score  | Rank | Name     | Score  |
|------|----------|--------|------|----------|--------|
| 1    | VCAM1    | 5.41E+09 | 15   | HLA-B    | 9.63E+08 |
| 2    | CXCL8    | 1.46E+09 | 17   | HLA-F    | 9.63E+08 |
| 3    | CCL2     | 1.46E+09 | 18   | B2M      | 9.63E+08 |
| 4    | CCL3     | 1.46E+09 | 19   | HLA-DRA  | 9.62E+08 |
| 5    | CCL9     | 1.45E+09 | 20   | HLA-DPA1 | 9.62E+08 |
| 6    | CXCL1    | 1.45E+09 | 21   | HLA-DQA1 | 9.62E+08 |
| 7    | CXCL12   | 1.45E+09 | 22   | HLA-DPB1 | 9.62E+08 |
| 8    | CX3CR1   | 1.45E+09 | 23   | IRF1     | 9.60E+08 |
| 9    | CCL20    | 1.44E+09 | 24   | GBP2     | 9.59E+08 |
| 10   | CCL19    | 1.44E+09 | 25   | PTPRC    | 9.47E+08 |
| 11   | TLR1     | 9.30E+08 | 26   | PSMB8    | 8.38E+08 |
| 12   | TLR7     | 9.80E+08 | 27   | KNG1     | 4.80E+08 |
| 13   | C3       | 9.64E+08 | 28   | TRIM22   | 4.79E+08 |
| 14   | HLA-E    | 9.63E+08 | 29   | CXCL6    | 4.79E+08 |
| 15   | HLA-C    | 9.63E+08 | 30   | EGF      | 1.21E+07 |

aggravating interstitial fibrosis [26]. DCs were reported to be involved in tubulointerstitial inflammation in various progressive kidney diseases, such as lupus nephritis [27, 28] and crescentic glomerulonephritis [29, 30]. Nevertheless, the role of DCs in many prevalent kidney diseases, such as DKD and hypertensive nephropathy, remains poorly understood. Only one animal study showed that CD11+ dendritic cells infiltrated the glomeruli in NOD mice and correlated with albuminuria [31], and no dendritic cell infiltration in the interstitium has been reported until now. The vast majority of kidney DCs are conventional DCs (cDCs) expressing CD11b and C-X3-C motif receptor 1 [32, 33]. We observed increased infiltration of DCs, especially activated DCs and conventional DCs, indicating a role for these cells in the tubulointerstitial pathogenesis of DKD. Previous studies have reported that DCs were identified morphologically within the tubulointerstitium and might be involved in the pathogenesis of interstitial inflammation in drug-induced acute interstitial nephritis [34] and light chain-associated tubulointerstitial nephritis [35]. The specific mechanism requires additional experimental evidence. Helper (CD4+) T cells, cytotoxic (CD8+) cells and B cells are important components of the adaptive immune system. In the present study, we observed a significant enrichment of CD4+ and CD8+ T cells in the interstitium, suggesting a role of adaptive immunity in DKD. An animal study also detected CD4+ and CD8+ T cell infiltration in the kidney interstitium in streptozotocin-treated rats [36]. In people with diabetes, the number of CD4+ T cells and CD20 cells positively correlates with proteinuria [37]. However, not all CD4+ T cells promote DKD. In our study, Tregs were downregulated in DKD and almost negatively correlated with pyroptosis (r = −0.4191, P = 0.0522). Tregs appear to play a protective role in DKD.

Renal tubular epithelial cells are critical for both tubular and glomerular function, and injured renal tubular cells are an important source of proinflammatory cytokines and chemokines [38]. Multiple studies have shown that regulated cell death (RCD), such as pyroptosis, necroptosis, apoptosis, and ferroptosis, of tubular cells contributes to the pathophysiology of kidney diseases [39–42]. Various forms of cell death were evaluated using GSEA and GSVA in the dataset to explore the relationship between tubular injury and interstitial immune cell infiltration. The results revealed that pyroptosis was the main form of cell death compared with necroptosis, apoptosis, ferroptosis and autophagy. The accumulation of reactive oxygen species (ROS) in individuals with DKD caused by hyperglycemia is an important activator of the NLRP3 (NOD-like receptor family, pyrin domain containing 3) inflammasome. Then, NLRP3 activates gasdermin D (GSDMD) and rapidly causes cell membrane rupture and the release of cell contents, leading to an inflammatory response [43]. Pyroptosis is involved in many chronic progressive diseases, such as Alzheimer’s disease [44], Parkinson’s disease [45], atherosclerosis [46] and rheumatoid arthritis [47]. The correlation analysis indicated that pyroptosis was positively correlated with increased interstitial infiltration of CD4+ T cells, CD4+ T memory cells, Th2 cells, CD8+ T cells, CD8+ Tcm cells, CD8+ Tem cells, Tγδ cells, NK cells, M1 macrophages, activated dendritic cells, conventional dendritic cells and neutrophils. Thus, our study showed that tubular pyroptosis-related interstitial inflammation might be an important contributor to DKD. This finding was confirmed in studies of patients with diabetes and db/db mice showing that tubular epithelial cell pyroptosis was accompanied by increases in the levels of the NLRP3 inflammasome, IL-1β, and
TGF-β. Treatments targeting renal tubular epithelial cell pyroptosis may be critical to interfere with the progression of DKD.

The highest MCC score for VCAM1 obtained from cytoHubba suggested that it is an important hub gene in DKD tubulointerstitial injury. VCAM1 has an important role in leukocyte adhesion to the endothelium. According to an animal study, MRL/lpl (a murine model of lupus nephritis) kidneys display increased VCAM1 expression in the endothelium, cortical tubules and glomeruli [48]. Patients with type 2 diabetes present significantly elevated serum levels of VCAM1, and VCAM1 levels correlate well with the extent of albuminuria [49]. Previous studies on VCAM1 in DKD mainly focused on glomerular endothelial cells [50]. However, researchers have not completely elucidated whether renal tubular expression of VCAM1 is elevated in patients with DKD and whether it is involved in the interstitial pathogenesis of DKD. In the present study, the upregulated VCAM1 mRNA levels in the DKD tubulointerstitium in the public GEO dataset and markedly increased tubular VCAM1 expression in HK-2 cells confirmed the upregulation of tubular VCAM1 expression in DKD. As VCAM1 is an adhesion molecule for lymphocytes and monocytes, the positive correlation between VCAM1 expression and interstitial immune cell infiltration in our study indicated that elevated tubular

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**Figure 7. Correlation between VCAM1 and immune cells and pyroptosis.** (a) VCAM1 expression was significantly increased in patients with DKD (n = 10) compared with healthy controls (n = 12) (**P < 0.001). (b) Correlation of VCAM1 with the infiltrated immune cells. (c) Correlation of VCAM1 with pyroptosis.
VCAM1 expression might allow the protein to interact with mononuclear cells, thus promoting interstitial fibrosis. Consistent with our results, an increase in tubular VCAM1 expression in episodes of acute renal allograft rejection is associated with T cell and monocyte infiltration [50]. In addition, we found a positive correlation between VCAM1 expression and pyroptosis, indicating that pyroptotic tubular cells might release VCAM1. The in vitro study validated increased VCAM1 expression in HK-2 cells cultured under diabetic conditions, and pyroptosis inhibition by disulfiram decreased VCAM1 expression, inflammatory cytokine release and fibrosis. Our study highlighted renal tubular VCAM1 as an attractive target for DKD therapy (Figure 10). However, the molecular mechanisms underlying the interaction between tubular VCAM1 and immune cell infiltration in the interstitium are essentially unknown, and further experimental studies are urgently needed [51].

The current work investigated the profile of tubulointerstitial immune cell infiltration and identified pyroptosis as the main form of programmed cell death in patients with DKD using bioinformatics analysis. The analytical methods were reliable and novel. The correlation between immune cells, pyroptosis, and VCAM1 provides new information for future research. Furthermore, our in vitro study suggested that the FDA-approved drug disulfiram might alleviate interstitial inflammation and fibrosis by inhibiting tubular pyroptosis and VCAM1 expression in individuals with DKD. However, this study has several limitations. First, the sample size applied in the analysis was small.
Second, the results were not confirmed by conducting in-depth experiments. Thus, more combined samples will be needed, and the results will be verified by experiments in DKD animal models and DKD cohorts.

**Conclusion**

In conclusion, using bioinformatics analysis, we investigated the profile of tubulointerstitial immune cell infiltration and evaluated the role of pyroptosis in the tubulointerstitium of DKD. VCAM1 was identified as the hub gene and was positively correlated with pyroptosis and infiltrated immune cells. In addition, VCAM1 expression was validated to be elevated in renal tubular cells. *In vitro* study revealed that the FDA-approved drug disulfiram inhibited renal tubular cell pyroptosis and decreased VCAM1 expression, inflammatory cytokine levels and fibrosis. Further
experimental analyses of pyroptosis and VCAM1 in the tubulointerstitium of patients with DKD might identify targets for immunotherapy.

Availability of Data and Materials
The datasets used during the study are available from the corresponding author upon reasonable request.

Author contributions
Y.J. analyzed the data, interpreted the results and drafted the article. Y.J. and Z.X. conceived, designed and organized the study. H.X. and L.T. analyzed the data. Y.J. and Q.Y. conducted the in vitro experiments and VCAM1 immunostaining. Z. X. interpreted the results and revised the manuscript. All the authors read and approved the final manuscript.

Funding
This research was supported by grants from China Postdoctoral Science Foundation (No.2018M640808) and ‘San-ming’ Project of Medicine in Shenzhen (No.SZSM201812097).

References
[1] Yang C, Gao B, Zhao X, et al. Executive summary for China Kidney Disease Network (CK-NET) 2016 Annual Data Report. Kidney Int. 2020;98 (6):1419–1423.
[2] Johansen KL, Chertow GM, Foley RN, et al.: US Renal Data System 2020 Annual Data Report: epidemiology of Kidney Disease in the United States. Am J Kidney Dis. 2021;77(1):A7–A8.
[3] Zhu X, Xiong X, Yuan S, et al. Validation of the interstitial fibrosis and tubular atrophy on the new pathological classification in patients with diabetic nephropathy: a single-center study in China. J Diabetes Complications. 2016;30(3):537–541.
[4] An Y, Xu F, Le W, et al. Renal histologic changes and the outcome in patients with diabetic nephropathy. Nephrol Dial Transplant. 2015;30(2):257–266.
[5] Gilbert RE. Proximal Tubulopathy: prime Mover and Key Therapeutic Target in Diabetic Kidney Disease. Diabetes. 2017;66(4):791–800.
[6] Hickey FB, Martin F. Role of the Immune System in Diabetic Kidney Disease. Curr Diab Rep. 2018;18 (4):20.
[7] Tang SCW, Yiu WH. Innate immunity in diabetic kidney disease. Nat Rev Nephrol. 2020;16(4):206–222.
[8] Aran D, Hu Z, Butte AJ. xCell: digitally portraying the tissue cellular heterogeneity landscape. Genome Biol. 2017;18(1):220.
[9] Irizarry RA, Hobbs B, Collin F, et al. Exploration, normalization, and summaries of high density oligonucleotide array probe level data. Biostatistics. 2003;4 (2):249–264.
[10] Limma SGK. linear models for microarray data. In: Gentleman R, VJ C, Huber W, et al, editors.
bioinformatics and computational biology solutions using R and bioconductor. Vols. 397–420. New York NY: Springer; 2005. p. 2005.

[11] Hanzelmann S, Castelo R, Guinney J. GSVA: gene set variation analysis for microarray and RNA-seq data. BMC Bioinformatics. 2013;14(1):7.

[12] Subramanian A, Tamayo P, Mootha VK, et al. Gene set enrichment analysis: a knowledge-based approach for interpreting genome-wide expression profiles. Proc Natl Acad Sci U S A. 2005;102(43):15545–15550.

[13] Ritchie ME, Phipson B, Wu D, et al. limma powers differential expression analyses for RNA-sequencing and microarray studies. Nucleic Acids Res. 2015;43(7):e47.

[14] Chin CH, Chen SH, Wu HH, et al. cytoHubba: identifying hub objects and sub-networks from complex interactome. BMC Syst Biol. 2014;8(4):S11.

[15] Zhou Y, Zhou B, Pache L, et al. Metascape provides a biologist-oriented resource for the analysis of systems-level datasets. Nat Commun. 2019;10(1):1523.

[16] Zhan M, Usman IM, Sun L, et al. Disruption of renal tubular mitochondrial quality control by Myo-inositol oxygenase in diabetic kidney disease. J Am Soc Nephrol. 2015;26(6):1304–1321.

[17] Satou R, Miyata K, Katsurada A, et al. Tumor necrosis factor-[alpha] suppresses angiotsingen expression through formation of a p50/p50 homodimer in human renal proximal tubular cells. Am J Physiol Cell Physiol. 2010;299(4):C750–9.

[18] Hu JJ, Liu X, Xia S, et al. FDA-approved disulfiram inhibits pyroptosis by blocking gasdermin D pore formation. Nat Immunol. 2020;21(7):736–745.

[19] Zhang Y, Zhang R, Han X. Disulfiram inhibits inflammation and fibrosis in a rat unilateral ureteral obstruction model by inhibiting gasdermin D cleavage and pyroptosis. Inflamm Res. 2021;70(5):543–552.

[20] Yu K, Li D, Xu F, et al. IDO1 as a new immune biomarker for diabetic nephropathy and its correlation with immune cell infiltration. Int Immunopharmacol. 2021;94:107446.

[21] van Furth R, Cohn ZA. The origin and kinetics of mononuclear phagocytes. J Exp Med. 1968;128(3):415–435.

[22] Nelson PJ, Rees AJ, Griffin MD, et al. The renal mononuclear phagocytic system. J Am Soc Nephrol. 2012;23(2):194–203.

[23] Chow F, Ozols E, Nikolic-Paterson DJ, et al. Macrophages in mouse type 2 diabetic nephropathy: correlation with diabetic state and progressive renal injury. Kidney Int. 2004;65(1):116–128.

[24] Chow FY, Nikolic-Paterson DJ, Ma FY, et al. Monocyte chemotactrant protein-1-induced tissue inflammation is critical for the development of renal injury but not type 2 diabetes in obese db/db mice. Diabetologia. 2007;50(2):471–480.

[25] Klessens CQF, Zandbergen M, Wolterbeek R, et al. Macrophages in diabetic nephropathy in patients with type 2 diabetes. Nephrol Dial Transplant. 2017;32(8):1322–1329.

[26] Rousselle A, Kettritz R, Schreiber A. Monocytes Promote Crescent Formation in Anti-Myeloperoxidase Antibody-Induced Glomerulonephritis. Am J Pathol. 1998-1915;187(9):2017.

[27] Castellano G, Trouw LA, Fiore N, et al. Infiltrating dendritic cells contribute to local synthesis of C1q in murine and human lupus nephritis. Mol Immunol. 2010;47(11–12):2129–2137.

[28] Tucci M, Quatraro C, Lombardi L, et al. Glomerular accumulation of plasmacytoid dendritic cells in active lupus nephritis: role of interleukin-18. Arthritis Rheum. 2008;58(1):251–262.

[29] Evers BD, Engel DR, Bohner AM, et al. CD103+ Kidney Dendritic Cells Protect against Crescentic GN by Maintaining IL-10-Producing Regulatory T Cells. J Am Soc Nephrol. 2016;27(11):3368–3382.

[30] Brahler S, Zinselmeyer BH, Raju S, et al. Opposing Roles of Dendritic Cell Subsets in Experimental GN. J Am Soc Nephrol. 2018;29(1):138–154.

[31] Xiao X, Ma B, Dong B, et al. Cellular and humoral immune responses in the early stages of diabetic nephropathy in NOD mice. J Autoimmun. 2009;32(2):85–93.

[32] Soos TJ, Sims TN, Barisoni L, et al. CX3CR1+ interstitial dendritic cells form a contiguous network throughout the entire kidney. Kidney Int. 2006;70(3):591–596.

[33] Kruger T. Identification and functional characterization of dendritic cells in the healthy murine kidney and in experimental glomerulonephritis. J Am Soc Nephrol. 2004;15(3):613–621.

[34] Cheng M, Gu X, Herrera GA. Dendritic cells in renal biopsies of patients with acute tubulointerstitial nephritis. Hum Pathol. 2016;54:113–120.

[35] Cheng M, Gu X, Turbat-Herrera EA, et al. Tubular Injury and Dendritic Cell Activation Are Integral Components of Light Chain-Associated Acute Tubulointerstitial Nephritis. Arch Pathol Lab Med. 2019;143(10):1212–1224.

[36] Mensah-Brown EP, Obieneche EN, Galadari S, et al. Streptozotocin-induced diabetic nephropathy in rats: the role of inflammatory cytokines. Cytokine. 2005;31(3):180–190.

[37] Moon JV, Jeong KH, Lee TW, et al. Aberrant recruitment and activation of T cells in diabetic nephropathy. Am J Nephrol. 2012;35(2):164–174.

[38] Liu BC, Tang TT, Lv LL, et al. Renal tubule injury: a driving force toward chronic kidney disease. Kidney Int. 2018;93(3):568–579.

[39] Linkermann A. Nonapoptotic cell death in acute kidney injury and transplantation. Kidney Int. 2016;89(1):46–57.
[40] Linkermann A, Skouta R, Himmerkus N, et al. Synchronized renal tubular cell death involves ferroptosis. Proc Natl Acad Sci U S A. 2014;111(47):16836–16841.

[41] Martin-Sanchez D, Ruiz-Andres O, Poveda J, et al. Ferroptosis, but Not Necroptosis, Is Important in Nephrotoxic Folic Acid-Induced AKI. J Am Soc Nephrol. 2017;28(1):218–229.

[42] Mulay SR, Linkermann A, Anders HJ. Necroinflammation in Kidney Disease. J Am Soc Nephrol. 2016;27(1):27–39.

[43] Cookson BT, Brennan MA. Pro-inflammatory programmed cell death. Trends Microbiol. 2001;9(3):113–114.

[44] Zheng Z, Li G. Mechanisms and Therapeutic Regulation of Pyroptosis in Inflammatory Diseases and Cancer. Int J Mol Sci. 2020;21(4):1456.

[45] Wang S, Yuan YH, Chen NH, et al. The mechanisms of NLRP3 inflammasome/pyroptosis activation and their role in Parkinson’s disease. Int Immunopharmacol. 2019;67:458–464.

[46] Hoseini Z, Sepahvand F, Rashidi B, et al. NLRP3 inflammasome: its regulation and involvement in atherosclerosis. J Cell Physiol. 2018;233(3):2116–2132.

[47] Vande Walle L, Van Opdenbosch N, Jacques P, et al. Negative regulation of the NLRP3 inflammasome by A20 protects against arthritis. Nature. 2014;512(7512):69–73.

[48] Wuthrich RP. Vascular cell adhesion molecule-1 (VCAM-1) expression in murine lupus nephritis. Kidney Int. 1992;42(4):903–914.

[49] Rubio-Guerra AF, Vargas-Robles H, Lozano Nuevo JJ, et al. Correlation between circulating adhesion molecule levels and albuminuria in type-2 diabetic hypertensive patients. Kidney Blood Press Res. 2009;32(2):106–109.

[50] Sun L, Sun C, Zhou S, et al. Tamsulosin attenuates high glucose- induced injury in glomerular endothelial cells. Bioengineered. 2021;12(1):5184–5194.

[51] Lin Y, Kirby JA, Browell DA, et al. Renal allograft rejection: expression and function of VCAM-1 on tubular epithelial cells. Clin Exp Immunol. 1993;92(1):145–151.