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Review

# Key Genetic Components of Fibrosis in Diabetic Nephropathy: An Updated Systematic Review and Meta-Analysis

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**Abstract:** Renal fibrosis (RF) constitutes the common end-point of all kinds of chronic kidney disease (CKD), regardless of the initial cause of disease. The aim of the present study was to identify the key players of fibrosis in the context of diabetic nephropathy (DN). A systematic review and meta-analysis of all available genetic association studies regarding the genes that are included in signaling pathways related to RF were performed. The evaluated studies were published in English and they were included in PubMed and the GWAS Catalog. After an extensive literature review and search of the Kyoto Encyclopedia of Genes and Genomes (KEGG) database, eight signaling pathways related to RF were selected and all available genetic association studies of these genes were meta-analyzed. *ACE, AGT, EDN1, EPO, FLT4, GREM1, IL1B, IL6, IL10, IL12RB1, NOS3, TGFB1, IGF2/INS/TH cluster,* and *VEGFA* were highlighted as the key genetic components driving the fibrosis process in DN. The present systematic review and meta-analysis indicate, as key players of fibrosis in DN, sixteen genes. However, the results should be interpreted with caution because the number of studies was relatively small.

**Keywords:** renal fibrosis; diabetic nephropathy; genes; review; meta-analysis



Citation: Tziastoudi, M.; Theoharides, T.C.; Nikolaou, E.; Efthymiadi, M.; Eleftheriadis, T.; Stefanidis, I. Key Genetic Components of Fibrosis in Diabetic Nephropathy: An Updated Systematic Review and Meta-Analysis. *Int. J. Mol. Sci.* 2022, 23, 15331. https://doi.org/10.3390/ ijms232315331

Academic Editor: Carolyn M. Ecelbarger

Received: 15 October 2022 Accepted: 29 November 2022 Published: 5 December 2022

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# 1. Introduction

Diabetic nephropathy (DN) is a multifactorial condition that involves both metabolic and hemodynamic factors, but clear evidence (heritability estimates, familial clustering of the disease, and differential vulnerability according to race) also indicates that the genetic profile of individuals plays a major role [1–3]. In addition, differences in the risk of DN among ethnic populations indicate that differences in lifestyle and environment also play major roles in the selection of alleles contributing to the different risks of disease occurrence [4]. These observations suggest that some diabetic patients are programmed to develop DN. Given the continuous increase in both DM and its complications, the effort to prevent DM is still considered a global challenge.

Among the histological lesions that occur in DN [5,6], renal fibrosis (RF) is considered as a complex and irreversible process in the late stages of DN, further exacerbating the progression of the disease [6]. In RF, which takes place in response to injury and inflammation, there is excessive deposition of the extracellular matrix (ECM) and epithelial–mesenchymal transition (EMT), leading to the loss of differentiated epithelial cells and their vascular capillary bed, accumulation of myofibroblasts and inflammatory cells, and, ultimately, the formation of a scar [7]. As a result, the destruction of the normal architecture and function

of the kidney occurs [6,8,9]. RF also constitutes the end point of all kinds of chronic kidney disease (CKD), regardless of the initial cause of disease [8].

It is noteworthy that, although fibrosis is considered a detrimental condition, recent studies suggest a protective role of this process, because it helps the maintenance of crosstalk with injured proximal tubular cells supporting their regeneration [10]. In autosomal-dominant polycystic kidney disease (ADPKD), for instance, fibrosis may have some protective roles, as the fibrogenic response is generally correlated with cystic disease regression [11].

Given the undoubtedly genetic involvement in the course of DN, many genetic studies, such as linkage scans and genetic association studies (GAS), as well as meta-analyses of these studies, have been published [12–16]. A major contribution to the genetic dissection of DN has been made by genome-wide association studies (GWAS) and their meta-analyses [17–22].

In order to identify the key players of fibrosis in the context of DN, we performed a systematic review and meta-analysis of all available genetic association studies regarding genes that are involved in signaling pathways related to RF. In this study, we included all available genetic association studies regarding fibrosis-related genes that have been published within a time period of 35 months since our field synopsis of all available genetic association studies regarding DN [13] and presented an updated meta-analysis of the relevant polymorphisms.

### 2. Methods

# 2.1. Identification and Eligibility of Relevant Studies

In an effort to decipher the key genetic components of the RF process, we performed a systematic review and meta-analysis of eight signaling pathways that are closely related to RF. The eight pathways were selected after a literature review and after a search of the Kyoto Encyclopedia of Genes and Genomes (KEGG) database. The included pathways were the ACE pathway, the Relaxin pathway, the Wnt signaling pathway, the MAPK signaling pathway, the PI3KC signaling pathway, the TGFB1 signaling pathway, the NOTCH signaling pathway, and the JAK signaling pathway (Table 1). The overlap between genes in these pathways is depicted in a Venn diagram (Figure 1). The full names of the genes in each signaling pathway are shown in Supplementary Tables S1–S8. In the meta-analysis, we included GAS that examined the association of any gene included in the aforementioned pathways with DN.

**Table 1.** Pathways related to renal fibrosis in KEGG \*.

### ACE pathway

<u>ACE, ACE2, AGT, AGTR1, AGTR2, ANPEP, ATP6AP2, CMA1, CPA3, CTSA, CTSG, ENPEP, KLK1, KLK2, LNPEP, MAS1, MME, MRGPRD, NLN, PRCP, PREP, REN, THOP1</u>

### Relaxin pathway

ACTA2, <u>ADCY1</u>, ADCY2, ADCY3, ADCY4, ADCY5, ADCY6, ADCY7, ADCY8, ADCY9, AKT1, AKT2, <u>AKT3</u>, ARRB1, ARRB2, <u>ATF2</u>, ATF4, ATF6B, COL1A1, COL1A2, COL3A1, <u>COL4A1</u>, COL4A2, COL4A3, COL4A4, COL4A5, COL4A6, CREB1, CREB3, CREB3L1, CREB3L2, CREB3L3, CREB3L4, CREB5, <u>EDN1</u>, <u>EDNRB</u>, <u>EGFR</u>, FOS, GNA15, GNAI1, GNAI2, GNAI3, GNAO1, GNAS, GNB1, GNB2, <u>GNB3</u>, GNB4, GNB5, GNG70, GNG81, GNG71, GNG72, GRB2, HRAS, INSL3, INSL5, JUN, <u>KRAS</u>, MAP2K1, MAP2K2, MAP2K4, MAP2K7, MAPK10, MAPK11, MAPK11, MAPK12, MAPK13, MAPK14, MAPK3, MAPK8, MAPK9, MMP1, <u>MMP13</u>, <u>MMP2</u>, <u>MMP9</u>, NFKB1, <u>NFKBIA</u>, <u>NOS1</u>, <u>NOS2</u>, <u>NOS3</u>, NRAS, PIK3CA, PIK3CB, PIK3CD, PIK3R1, PIK3R2, PIK3R3, PLCB1, PLCB2, PLCB3, PLCB4, PRKACA, PRKACB, PRKACG, PRKCA, PRKCZ, RAF1, <u>RELA</u>, RLN1, RLN2, RLN3, RXFP1, RXFP2, RXFP3, RXFP4, SHC1, SHC2, SHC3, SHC4, <u>SMAD2</u>, SOS1, SOS2, SRC, <u>TGFB1</u>, <u>TGFBR1</u>, <u>TGFBR2</u>, <u>VEGFA</u>, <u>VEGFB</u>, <u>VEGFC</u>, VEGFD

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### Table 1. Cont.

### Wnt signaling pathway

APC, APC2, APCDD1, APCDD1L, <u>AXIN1</u>, <u>AXIN2</u>, BAMBI, BTRC, CACYBP, CAMK2A, CAMK2B, CAMK2D, CAMK2G, CBY1, CCDC88C, CCN4, CCND1, CCND2, CCND3, CER1, CHD8, CREBBP, CSNK1A1, CSNK1A1L, CSNK1E, CSNK2A1, CSNK2A2, CSNK2A3, CSNK2B, CTBP1, CTBP2, <u>CTNNB1</u>, CTNNBIP1, CTNND2, CUL1, CXXC4, <u>DAAM1</u>, DAAM2, DKK1, <u>DKK2</u>, DKK4, DVL1, DVL2, DVL3, EP300, FBXW11, FOSL1, FRAT1, FRAT2, FRZB, FZD1, FZD10, FZD2, FZD3, FZD4, FZD5, FZD6, FZD7, FZD8, FZD9, GPC4, <u>GSK3B</u>, INVS, JUN, LEF1, LGR4, LGR5, LGR6, <u>LRP5</u>, <u>LRP6</u>, <u>MAP3K7</u>, <u>MAPK10</u>, MAPK8, MAPK9, MMP7, MYC, <u>NFATC1</u>, <u>NFATC2</u>, <u>NFATC3</u>, NFATC4, NKD1, NKD2, NLK, NOTUM, PLCB1, PLCB2, PLCB3, PLCB4, PORCN, PPARD, PPP3CA, PPP3CB, PPP3CC, PPP3R1, PPP3R2, PRICKLE1, <u>PRICKLE2</u>, PRICKLE3, PRICKLE4, PRKACA, PRKACB, PRKACG, PRKCA, <u>PRKCB</u>, PRKCG, PSEN1, RAC1, RAC2, RAC3, RBX1, RHOA, RNF43, ROCK2, <u>ROR1</u>, ROR2, RSPO1, RSPO2, RSPO3, RSPO4, RUVBL1, RYK, SENP2, SERPINF1, SFRP1, SFRP2, SFRP4, SFRP5, SIAH1, SKP1, <u>SMAD3</u>, <u>SMAD4</u>, SOST, SOX17, TBL1X, TBL1XR1, TBL1Y, TCF7, TCF7L1, <u>TCF7L2</u>, TLE1, TLE2, TLE3, TLE4, TLE6, TLE7, TP53, TPTEP2-CSNK1E, VANGL1, VANGL2, WIF1, WNT1, WNT10A, WNT10B, WNT11, <u>WNT16</u>, WNT2, WNT2B, <u>WNT3</u>, WNT3A, <u>WNT4</u>, <u>WNT5A</u>, WNT5B, <u>WNT6</u>, WNT7A, WNT7B, WNT8B, WNT8B, WNT9A, WNT9B, ZNRF3

### MAPK signaling pathway

AKT1, AKT2, <u>AKT3</u>, <u>ANGPT1</u>, <u>ANGPT2</u>, ANGPT4, ARAF, AREG, ARRB1, ARRB2, <u>ATF2</u>, ATF4, BDNF, BRAF, CACNA1A, CACNA1B, CACNA1C, CACNA1D, CACNA1E, CACNA1F, CACNA1G, CACNA1H, CACNA1I, CACNA1S, CACNA2D1, CACNA2D2, CACNA2D3, CACNA2D4, CACNB1, CACNB2, CACNB3, CACNB4, CACNG1, CACNG2, CACNG3, CACNG4, CACNG5, CACNG6, CACNG7, CACNG8, CASP3, CD14, CDC25B, CDC42, CHUK, CRK, CRKL, CSF1, CSF1R, DAXX, DDIT3, DUSP1, DUSP10, DUSP16, DUSP2, DUSP3, DUSP4, DUSP5, DUSP6, DUSP7, DUSP8, DUSP9, ECSIT, EFNA1, EFNA2, EFNA3, EFNA4, EFNA5, EGF, EGFR, ELK1, ELK4, EPHA2, ERBB2, ERBB3, ERBB4, EREG, FAS, FASLG, FGF1, FGF10, FGF16, FGF17, FGF18, FGF19, FGF2, FGF20, FGF21, FGF22, FGF23, FGF3, FGF4, FGF5, FGF6, FGF7, FGF8, FGF9, FGFR1, FGFR2, FGFR3, FGFR4, FLNA, FLNB, FLNC, FLT1, FLT3, FLT3LG, FLT4, FOS, GADD45A, GADD45B, GADD45G, GNA12, GNG12, GRB2, HGF, HRAS, HSPA1A, HSPA1B, HSPA1L, HSPA2, HSPA6, HSPA8, HSPB1, <u>IGF1</u>, <u>IGF1R, IGF2</u>, IKBKB, IKBKG, <u>IL1A, IL1B, IL1R1</u>, IL1RAP, <u>INS, INSR</u>, IRAK1, IRAK4, JMJD7-PLA2G4B, JUN, JUND, <u>KDR</u>, KIT, KITLG, <u>KRAS</u>, LAMTOR3, MAP2K1, MAP2K2, MAP2K3, MAP2K4, MAP2K5, MAP2K6, MAP2K7, MAP3K1, MAP3K11, MAP3K12, MAP3K13, MAP3K14, MAP3K2, MAP3K20, MAP3K3, MAP3K4, MAP3K5, MAP3K6, MAP3K7, MAP3K8, MAP4K1, MAP4K2, MAP4K3, MAP4K4, MAPK1, <u>MAPK10</u>, MAPK11, MAPK12, MAPK13, MAPK14, MAPK3, MAPK7, MAPK8, MAPK8IP1, MAPK8IP2, MAPK8IP3, MAPK9, MAPKAPK2, MAPKAPK3, MAPKAPK5, MAPT, MAX, MECOM, MEF2C, MET, MKNK1, MKNK2, MRAS, MYC, MYD88, NF1, NFATC1, NFATC3, NFKB1, NFKB2, NGF, NGFR, NLK, NR4A1, NRAS, NTF3, NTF4, NTRK1, NTRK2, PAK1, PAK2, PDGFA, PDGFB, PDGFC, PDGFD, PDGFRA, PDGFRB, PGF, PLA2G4A, PLA2G4B, PLA2G4C, PLA2G4D, PLA2G4E, PLA2G4F, PPM1A, PPM1B, PPP3CA, PPP3CB, PPP3CC, PPP3R1, PPP3R2, PPP5C, PRKACA, PRKACB, PRKACG, PRKCA, PRKCB, PRKCG, PTPN5, PTPN7, PTPRR, RAC1, RAC2, RAC3, RAF1, RAP1A, RAP1B, RAPGEF2, RASA1, RASA2, RASGRF1, RASGRF2, RASGRP1, RASGRP2, RASGRP3, RASGRP4, RELA, RELB, RPS6KA1, RPS6KA2, RPS6KA3, RPS6KA4, RPS6KA5, RPS6KA6, RRAS, RRAS2, SOS1, SOS2, SRF, STK3, STK4, STMN1, TAB1, TAB2, TAOK1, TAOK2, TAOK3, TEK, TGFA, <u>TGFB1</u>, <u>TGFB2</u>, TGFB3, <u>TGFBR1</u>, <u>TGFBR2</u>, <u>TNF</u>, TNFRSF1A, TP53, TRADD, TRAF2, TRAF6, <u>VEGFA</u>, <u>VEGFB</u>, <u>VEGFC</u>, VEGFD

### PI3KC signaling pathway

BPNT2, CALM1, <u>CALM2</u>, <u>CALM3</u>, CALML3, CALML4, CALML5, CALML6, CDIPT, CDS1, CDS2, <u>DGKA</u>, DGKB, DGKD, DGKE, DGKG, DGKH, DGKI, DGKK, DGKQ, DGKZ, IMPA1, IMPA2, INPP1, INPP4A, INPP4B, INPP5A, INPP5B, INPP5B, INPP5D, INPP5F, INPP11, IP6K1, IP6K2, IP6K3, IPMK, IPPK, ITPK1, ITPKA, ITPKB, ITPKC, ITPR1, ITPR2, <u>ITPR3</u>, MTM1, MTMR1, MTMR14, MTMR2, MTMR3, MTMR4, MTMR6, MTMR7, MTMR8, OCRL, PI4K2A, PI4K2B, PI4KA, PI4KB, PIK3C2A, <u>PIK3C2B</u>, PIK3C2G, PIK3C3, PIK3CA, PIK3CB, PIK3CD, PIK3R1, PIK3R2, PIK3R3, PIKFYVE, PIP4K2A, PIP4K2B, PIP4K2C, PIP4P1, PIP4P2, PIP5K1A, PIP5K1B, PIP5K1C, PLCB1, PLCB2, PLCB3, PLCB4, PLCD1, PLCD3, PLCD4, PLCE1, PLCG1, PLCG2, PLCZ1, PPIP5K1, PPIP5K2, PRKCA, PRKCB, PRKCB, PTEN, SACM1L, SYNJ1, SYNJ2

### TGFB1 signaling pathway

ACVR1B, ACVR1C, ACVR2A, ACVR2B, AMH, AMHR2, BAMBI, BMP2, BMP4, BMP5, BMP6, BMP7, BMP8A, BMP8B, BMPR1A, BMPR1B, BMPR2, CDKN2B, CHRD, CREBBP, CUL1, DCN, E2F4, E2F5, EP300, FBN1, FMOD, FST, GDF5, GDF6, GDF7, GREM1, GREM2, HAMP, HJV, ID1, ID2, ID3, ID4, IFNG, INHBA, INHBB, INHBC, INHBE, LEFTY1, LEFTY2, LTBP1, MAPK1, MAPK3, MICOS10-NBL1, MYC, NBL1, NEO1, NODAL, NOG, PITX2, PPP2CA, PPP2CB, PPP2R1A, PPP2R1B, RBL1, RBX1, RGMA, RGMB, RHOA, ROCK1, RPS6KB1, RPS6KB2, SKP1, SMAD1, SMAD2, SMAD3, SMAD4, SMAD5, SMAD6, SMAD7, SMAD9, SMURF1, SMURF2, SP1, TFDP1, TGFB1, TGFB2, TGFB3, TGFBR1, TGFBR2, TGIF1, TGIF2, THBS1, THSD4, TNF, ZFYVE16, ZFYVE9

# NOTCH signaling pathway

ADAM17, APH1A, APH1B, ATXN1, ATXN1L, CIR1, CREBBP, CTBP1, CTBP2, DLL1, DLL3, DLL4, DTX1, DTX2, DTX3, DTX3L, DTX4, DVL1, DVL2, DVL3, EP300, HDAC1, HDAC2, <u>HES1</u>, HES5, HEY1, HEY2, HEYL, <u>JAG1</u>, JAG2, KAT2A, KAT2B, LFNG, MAML1, MAML2, MAML3, MFNG, NCOR2, NCSTN, NOTCH1, NOTCH2, <u>NOTCH3</u>, NOTCH4, NUMB, NUMBL, PSEN1, PSEN2, PSENEN, PTCRA, RBPJ, RBPJL, RFNG, SNW1, TLE1, TLE2, TLE3, TLE4, TLE6, TLE7

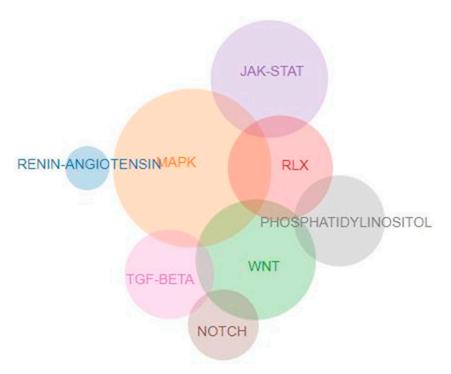
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### Table 1. Cont.

# JAK signaling pathway

AKT1, AKT2, <u>AKT3</u>, AOX1, <u>BCL2</u>, BCL2L1, CCND1, CCND2, CCND3, CDKN1A, CISH, CNTF, CNTFR, CREBBP, CRLF2, CSF2RA, CSF2RA, CSF3R, CSF3R, CSF1, CSF1, CSF2, CSF2RA, CSF2RB, CSF3, CSF3R, CSH1, CSH2, CTF1, <u>EGF</u>, EGFR, EP300, <u>EPO</u>, EPOR, FHL1, GFAP, GH1, GH2, <u>GHR</u>, GRB2, HRAS, IFNA1, IFNA10, IFNA13, IFNA14, IFNA16, IFNA17, IFNA2, IFNA21, IFNA4, IFNA5, IFNA6, IFNA7, IFNA8, IFNAR1, IFNAR2, IFNB1, IFNE, IFNG, IFNGR1, IFNGR2, IFNK, IFNL1, IFNL2, IFNL3, IFNLR1, IFNW1, <u>IL10</u>, IL10RA, IL10RB, IL11, IL11RA, <u>IL12A</u>, <u>IL12B</u>, <u>IL12RB1</u>, IL12RB2, <u>IL13</u>, IL13RA1, IL13RA2, IL15, IL15RA, IL17D, IL19, IL2, IL20, IL20RA, IL20RB, IL21, IL21R, IL22, IL22RA1, IL22RA2, IL23R, IL23R, IL24R, IL27RA, IL2RA, IL2RB, IL2RG, IL3, IL3RA, <u>IL4</u>, <u>IL4R</u>, IL5, IL5RA, <u>IL6</u>, <u>IL6R</u>, IL6ST, IL7, IL7R, IL9, IL9R, IRF9, JAK1, JAK2, JAK3, LEP, LEPR, LIF, LIFR, MCL1, MPL, MTOR, MYC, OSM, OSMR, PDGFA, <u>PDGFB</u>, <u>PDGFRA</u>, <u>PDGFRB</u>, PIAS1, PIAS2, PIAS3, PIAS4, PIK3CA, PIK3CB, PIK3CD, PIK3R1, PIK3R2, PIK3R3, PIM1, PRL, PRLR, PTPN11, PTPN2, PTPN6, RAF1, SOCS1, SOCS2, SOCS3, SOCS4, SOCS5, SOCS6, SOCS7, SOS1, SOS2, STAM, STAM2, STAT1, STAT2, STAT3, STAT4, STAT5A, STAT5B, STAT6, <u>THPO</u>, TSLP, TYK2

 $<sup>^{\</sup>ast}$  There are only published studies for underlined genes, so only these genes were meta-analyzed.



**Figure 1.** Venn diagram showing the overlap between genes in pathways.

A literature search of PubMed and the GWAS Catalog (http://www.genome.gov/gwastudies/) (accessed on 11 June 2022) was conducted, and the inclusion and exclusion criteria and data extraction were as previously described [13]. More specifically, cases were defined as diabetics with persistent micro/macroalbuminuria with or without diabetic retinopathy, whereas diseased controls were defined as diabetics with normoalbuminuria and/or normal renal function. The eligibility of the studies was assessed by two investigators (M.T. and I.S.). The reporting of the systematic review process will follow the PRISMA statement [23].

# 2.2. Data Synthesis and Analysis

The genetic association between each polymorphism and DN was assessed using the generalized odds ratio (ORG) [24,25]. The threshold for the meta-analysis was the presence of two studies per genetic polymorphism. The pooled OR was estimated using the Der Simonian and Laird random-effects model [26]. The associations are presented with ORs and their corresponding 95% confidence intervals (Cis). The between-study heterogeneity was tested with Cochran's Q statistic (considered statistically significant at p < 0.10) and its extent was assessed with the I<sup>2</sup> statistic [27,28].

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We also examined if controls confronted with Hardy–Weinberg equilibrium (HWE) predicted genotypes using Fisher's exact test for each study that provided genotype counts. We also tested for the 'small-study effect' with the Egger test [29].

### 3. Results and Discussion

### 3.1. Study Characteristics

The literature search of both PubMed and the GWAS Catalog retrieved 5058 papers after the exclusion of duplicate studies, whereas, in the meta-analysis, 180 articles were included. When an article provided data for different populations, each population was considered as a different study. Figure 2 presents a flowchart of the retrieved articles and the reasons for the exclusion of certain papers. The included studies were published between 1994 and 2021. The demographic characteristics of each study are shown in Supplementary Table S9, which was updated from our previous study [13]. Supplementary Table S9 only presents the included studies of these polymorphisms that were updated from our previous study [13].

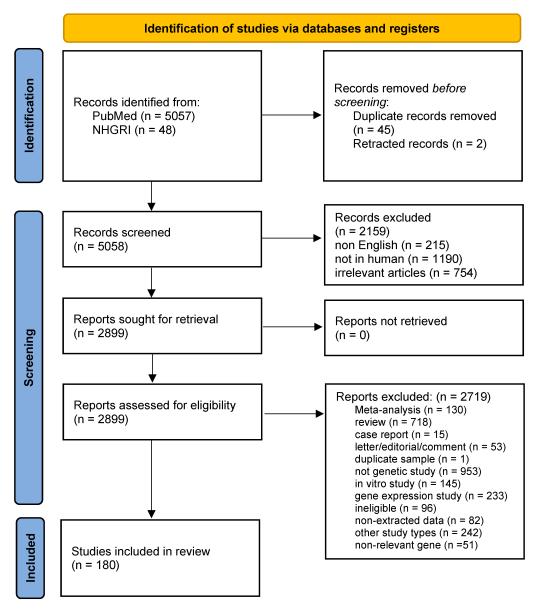


Figure 2. Flowchart of retrieved articles with specification of the reasons for exclusion.

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# 3.2. Meta-Analysis Results

Among the 884 different genes that were involved in eight pathways related to fibrosis, 134 genetic variants located in 45 different genes were meta-analyzed. In comparison with our previous study [13], the present study updated results for 24 genes. Table 2 shows the statistically significant results of the meta-analyses based on genotype counts, whereas Table 3 shows the statistically significant results of the meta-analyses based on allele counts. Tables 2 and 3 include also updated data from meta-analyses that were already published by our group in a previous study [13]. However, for the purpose of completeness, Supplementary Tables S10–S12 also present the non-significant results. Figures 3 and 4 are forest plots that show the pooled odds ratios of the significant results. Overall, sixteen genes provided significant results in all meta-analyses.

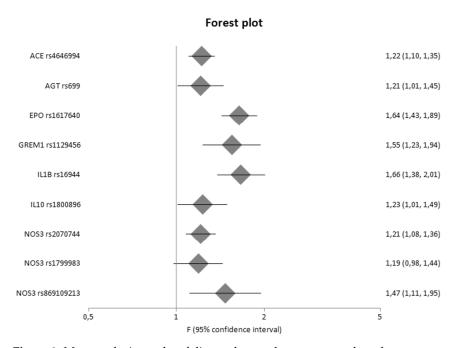


Figure 3. Meta-analysis results of diseased controls versus cases based on genotype counts.

Forest plot

# ACE rs4646994 NOS3 rs2070744 1,24 (1,02, 1,52) 1,42 (1,13, 1,77) NOS3 rs1799983 1,64 (1,21, 2,22) NOS3 rs869109213 1,72 (1,12, 2,06) 1,73 (1,46, 2,04) 1 0 0 95% confidence interval)

Figure 4. Meta-analysis results of healthy controls versus cases based on genotype counts.

**Table 2.** Results of the meta-analyses of statistically significant polymorphisms listed in alphabetical order based on genotype counts.

Gene	Variant	RS	Studies (n)	Cases/Controls (n)	RE OR <sub>G</sub>	95% LL	95% UL	I <sup>2</sup> (%)	$P_Q$	P <sub>E</sub>	Current Status		
Diseased Controls versus Cases													
ACE	I > D	rs4646994	66	11437/10984	1.22	1.10	1.35	76.34	0.00	0.70	updated		
	All in HWE	I > D	56	9383/8847	1.28	1.16	1.41	67.29	0.00	0.65			
AGT	M235T	rs699	26	5015/5253	1.21	1.01	1.45	82,45	0,00	0.84	[13]		
	All in HWE		19	3181/3655	31/3655 1.09 0.92		1.31	72.76	0.00	0.95			
EPO	G > T	rs1617640	3	1618/954	1.64	1.43	1.89	0.00	0.78	0.03	[13]		
GREM1		rs1129456 (A/T)	2	859/940	1.55	1.23	1.94	1.02	0.31	na	new		
IL1B	-511C > T	rs16944	3	774/667	1.66	1.38	2.01	0.00	0.86	0.28	[13]		
	All in HWE		3										
IL10	-1082  A > G	rs1800896	4	677/761	1.23	1.01	1.49	0	0.56	0.63	[13]		
		All in HWE	2	610/690	1.25	1.02	1.53	0	0.62	na			
NOS3	T-786C	rs2070744	9	2288/2154	1.21	1.08	1.36	0.00	0.62	0.40	updated		
	All in HWE		7	2026/1862	1.21	1.08	1.37	3.08	0.40	0.29			
NOS3	G894T	rs1799983	21	4538/3774	1.19	0.98	1.44	77.97	< 0.001	0.36	updated		
		All in HWE	19	4306/3564	1.24	1.02	1.51	77.68	0.00	0.20			
NOS3		rs869109213	2	354/444	1.47	1.11	1.95	0.00	0.95	na	updated		
		All in HWE	2							na			
Healthy Controls versus Cases													
ACE	I > D	rs4646994	30	3690/4927	1.24	1.02	1.52	83.20	0.00	0.03	[13]		
	All in HWE	I > D	29	3283/4695	1.26	1.02	1.55	82.87	0.00	0.01			
NOS3	T-786C	rs2070744	9	1583/2142	1.42	1.13	1.77	58.00	0.01	0.84	updated		

 Table 2. Cont.

	0.80	0.01	63.04	1.79	1.11	1.41	1516/2042	8		All in HWE		
updated	0.11	< 0.001	82.07	2.22	1.21	1.64	2295/2737	11	rs1799983	G894T	NOS3 G8	
	0.21	< 0.001	82.40	2.11	1.14	1.55	2247/2467	10		All in HWE		
updated	na	0.27	17.59	2.06	1.12	1.52	354/444	2	rs869109213		NOS3	
								2		All in HWE		
	0.18	0	83.64	1.96	0.86	1.30	814/1450	6	rs1800470	T869C	TGFB1	
	0.21	0.41	0	2.04	1.46	1.73	706/1103	4	All in HWE			
				ł	s versus Cases	iseased Control	Healthy Controls versus D	]				
updated	na	0.42	0.00	1.89	1.10	1.44	90/234/212	2	rs1800795	G(-174)C	IL6	
								1		All in HWE		
updated	0.46	0.53	0.00	1.43	1.17	1.29	1307/1117/1451	5	rs2070744	T-786C	NOS3	
	0.51	0.37	3.59	1.43	1.16	1.29	1240/1080/1351	4		All in HWE		
updated	0.32	0.01	70.00	1.56	1.05	1.28	1506/1255/1642	8	rs1799983	G894T	NOS3	
	0.41	0.12	40.52	1.56	1.17	1.35		7		All in HWE		
updated	na	0.23	30.45	1.63	1.04	1.30	354/444/515	2	rs869109213		NOS3	
								2		All in HWE		

RS: SNP identifier, RE  $OR_G$ : generalized random-effects odds ratio, LL: lower limit, UL: upper limit,  $I^2$ :  $I^2$  statistic,  $P_Q$ : p-value from heterogeneity testing,  $P_E$ : p-value from Egger's test.

GENE	Variant	RS	Studies (n)	Cases/ Controls (n)	RE OR	95% LL	95% UL	I <sup>2</sup> (%)	$P_Q$	PE	Current Status
Diseased Controls versus Cases											
EDN1		rs1794849	3	1176/1323	1.16	1.02	1.31	0	0.62	0.08	[13]
FLT4		rs2242221	3	1176/1323	1.14	1.01	1.29	0	0.38	0.43	[13]
IGF2/INS/TH cluster		rs1004446	3	1176/1323	1.16	1.03	1.31	0	0.49	0.22	[13]
IGF2/INS/TH cluster		rs4320932	3	1176/1323	0.84	0.73	0.96	0	0.43	0.06	[13]
VEGFA	C > A	rs2146323	3	1176/1323	0.85	0.76	0.95	0.2		0.2	[13]
Healthy Controls versus Cases											
IL12RB1		rs372889	2	1674/1719	1.243	1.130	1.367	0	0.567	-	new

**Table 3.** Results of the meta-analyses of statistically significant polymorphisms listed in alphabetical order based on allele counts.

RS: SNP identifier, RE  $OR_G$ : random effects odds ratio generalized, LL: lower limit, UL: upper limit,  $I^2$ :  $I^2$  statistic,  $P_Q$ : p-value from heterogeneity testing,  $P_E$ : p-value from Egger's test.

In meta-analyses based on genotype counts, statistically significant results were reported for angiotensin I-converting enzyme (*ACE*), angiotensinogen (*AGT*), erythropoietin (*EPO*), gremlin 1, DAN family BMP antagonist (*GREM1*), interleukin 1 beta (*IL1B*), interleukin 6 (*IL6*), interleukin 10 (*IL10*), nitric oxide synthase 3 (*NOS3*), and transforming growth factor beta 1 (*TGFB1*).

More specifically, in comparison with diseased controls versus cases, the *ACE* I/D polymorphism was significantly associated with DN with a pooled  $OR_G$  of 1.22 (95% CI 1.10–1.35), the *AGT* M235T variant showed significant results with a pooled  $OR_G$  of 1.21 (95% CI 1.01–1.45), *EPO* rs1617640 was significantly associated with DN with a pooled  $OR_G$  of 1.64 (95% CI 1.43–1.89), *GREM1* rs1129456 was associated with DN with a pooled  $OR_G$  of 1.55 (95% CI 1.23–1.94), *IL1B* -511C > T provided significant results with a pooled  $OR_G$  of 1.66 (95% CI 1.38–2.01), *IL10* -1082A > G was also associated with DN with a pooled  $OR_G$  of 1.23 (95% CI 1.01–1.49), and three variants in *NOS3* gene (rs2070744, rs1799983, rs869109213) produced significant results with pooled  $OR_G$  values of 1.21 (95% CI 1.08–1.36), 1.19 (95% CI 0.98–1.44), and 1.47 (1.11–1.95), respectively.

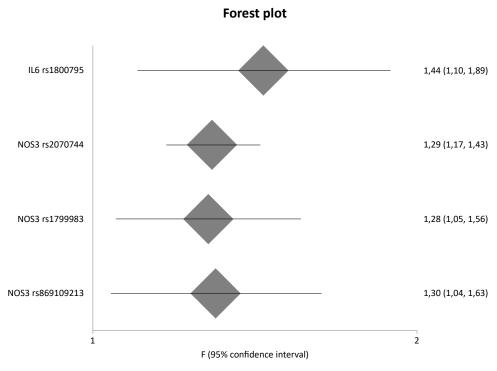
When healthy controls were compared with cases, the *ACE* I/D polymorphism was significantly associated with DN with a pooled  $OR_G$  of 1.24 (95% CI 1.02–1.52), three variants of the *NOS3* gene (rs2070744, rs1799983, and rs869109213) provided significant results with pooled  $OR_G$  values of 1.42 (95% CI 1.13–1.77), 1.64 (95% CI 1.21–2.22), 1.52 (95% CI 1.12–2.06), respectively, and TGFB1 T869C was also associated with DN with a pooled  $OR_G$  value of 1.73 (95% CI 1.46–2.04).

In a comparison of healthy controls versus diseased controls versus cases, IL6 rs1800795 was significantly associated with DN with a pooled  $OR_G$  of 1.44 (95% CI 1.10–1.89), and three variants of the NOS3 gene (rs2070744, rs1799983, rs869109213) gave significant results, with pooled  $OR_G$  values of 1.29 (95% CI 1.17–1.43), 1.28 (95% CI 1.05–1.56), and 1.30 (95% CI 1.04–1.63), respectively.

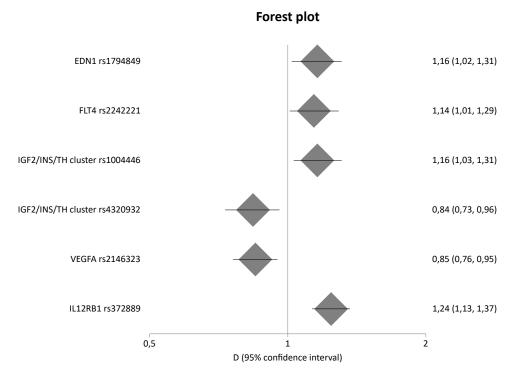
In meta-analyses based on allele counts, significant associations were reported for endothelin 1 (*EDN1*), fms-related receptor tyrosine kinase 4 (*FLT4*), insulin-like growth factor 2/insulin/tyrosine hydroxylase cluster (*IGF2/INS/TH cluster*), interleukin 12 receptor subunit beta 1 (*IL12RB1*), and vascular endothelial growth factor A (*VEGFA*).

More specifically, in a comparison of diseased controls versus cases, *EDN1* rs1794849 was significantly associated with DN, with a pooled OR of 1.16 (95% CI 1.02–1.31), *FLT4* rs2242221 produced significant results with a pooled OR of 1.14 (95% CI 1.01–1.29), two variants in the IGF2/INS/TH cluster, rs1004446 and rs4320932, were also associated with DN, with pooled OR values of 1.16 (95% CI 1.03–1.31) and 0.84 (95% CI 0.73–0.96), respectively, and *VEGFA* rs2146323 also produced significant results, with a pooled OR of 0.85 (95% CI 0.76–0.95). In a comparison of healthy controls versus cases, only one variant in

IL12RB1 (rs372889) was associated with DN, with a pooled OR value of 1.24 (95% CI 1.13-1.37). Figures 3-6 are forest plot representations of the genetic variants that are significantly associated with DN.



**Figure 5.** Meta-analysis results of healthy controls versus diseased controls versus cases based on genotype counts.



**Figure 6.** Meta-analysis results of diseased controls versus cases and healthy controls versus cases based on allele counts.

## 3.3. Discussion

To the best of our knowledge, this review constitutes the most comprehensive study of the genetic variants that are related to fibrosis in the context of DN. This study is an extension of our previous work in the field of the genetic epidemiology of DN [13], but, for the first time, we focused on fibrosis-related genes. Among the 134 genetic variants located in 45 different genes that were meta-analyzed, sixteen genes (ACE, AGT, EDN1, EPO, FLT4, GREM1, IL1B, IL6, IL10, IL12RB1, NOS3, TGFB1, IGF2/INS/TH cluster, and VEGFA) produced significant results in all meta-analyses.

The present systematic review confirmed the statistical significance of genes that are well known to have fibrotic effects, such as *ACE*, *AGT*, *EDN1*, *GREM1*, *IL1B*, *IL6*, and *TGFB1*, whereas some other genes are known for their anti-fibrotic effects, such as *EPO* and *NOS3*. The statistically significant contribution of *IL12RB1* and *TH* constitutes a novel finding, as there are no available experimental results to clarify their contribution to fibrosis.

More specifically, ACE and AGT constitute members of the renin–angiotensin system (RAS), but this finding is no surprise because RAS inhibitors also have antifibrotic effects [30]. Many lines of evidence indicate that RAS is a major regulator of renal fibrosis, as angiotensin II (AngII) promotes the release of TGF- $\beta$  and also activates the inflammatory process [31]. In addition, many publications indicate the involvement of TGFB1, which has been also characterized as the master regulator of fibrosis via the activation of both canonical and non-canonical pathways [32]. TGF- $\beta$  is also regarded as the most important inducer of endothelial-to-mesenchymal transition (EndMT) in vitro and in vivo, a form of EMT [33], suggesting that targeting the TGF- $\beta$  receptor signaling pathway could constitute a putative treatment for fibrosis [33].

Regarding the role of the innate proinflammatory cytokine IL1B, it has been reported that it induces a metabolic switch from oxidative phosphorylation to glycolysis in kidney stromal cells (SCs), promoting proximal tubule damage and fibrosis [34]. Increased expression of IL-6 and extensive and chronic activation of STAT3 were also associated with fibrosis [35]. The pathways of the anti-inflammatory mediator IL-10 have been found to be conserved in many diseases associated with fibrosis, although their underlying dissimilarity suggests that the IL-10 signaling pathway may have antifibrotic properties [36]. With regard to *IL12RB1*, to the best of our knowledge, this is a novel finding as no previous studies have found any association of *IL12RB1* with kidney fibrosis.

A previous study reported that the disruption of eNOS and ApoE genes accelerates kidney fibrosis and senescence after injury [37], indicating a protective role of the normal function of these genes. In addition, a meta-analysis that investigated the association between the (eNOS) 4b/a gene polymorphism and renal interstitial fibrosis in patients with DN demonstrated that the frequency of eNOS4bb in DN renal interstitial patients was lower than that in non-nephropathy diabetic patients and normal controls [38]. Regarding EDN1, it has been found that the endothelin receptors in renal interstitial cells do not contribute to the development of fibrosis during experimental kidney disease [39]. However, a randomized controlled trial, SONAR, regarding the selective endothelin A receptor antagonist atrasentan showed promising results, as atrasentan reduced the risk of renal events in patients with diabetes and chronic kidney disease [40]. Regarding the EPO gene, it has been found that erythropoietin attenuates renal interstitial fibrosis via the inhibition of fibrocyte accumulation [41]. Gremlin1 (Grem1), which is an antagonist of bone morphogenetic proteins, plays a key role in kidney development and renal fibrosis, and a previous study demonstrated that its levels are increased in many diseases associated with fibrosis [42]. More specifically, the grem1 levels are increased in renal fibrosis, as well as in fibrosis of the heart and lungs. FLT4, a tyrosine kinase receptor for VEGFC and VEGFD, is involved in lymphangiogenesis, which is a condition that develops during the progression of fibrosis, indicating a fibrotic effect of this factor [43].

Regarding the *IGF2/INS/TH* cluster, previous results have shown that IGF2 stimulates the differentiation into myofibroblasts that produce large amounts of collagen and other extracellular matrix proteins (ECM) [44]. Regarding *INS*, a hormone that plays a key role in

carbohydrate and lipid metabolism, it has been found that sodium–glucose cotransporter 2 (SGLT2) inhibitors suppressed kidney fibrosis in diabetic mice [45]. Regarding TH, which is involved in tyrosine-to-dopamine conversion, there are no results regarding its involvement in kidney fibrosis. Finally, VEGFA, a growth factor essential for both physiological and pathological angiogenesis, can inhibit the expression of Smad3 and miR192, thereby suppressing TGF- $\beta$ -induced EMT and improving renal fibrosis [46].

Based on the statistically significant results of the present systematic review and meta-analysis, it should be noted that mast cells (MCs) are implicated in many fibrotic conditions [47–50]. *ACE, IL1B, IL-6*, and *IL-10* are key mediators of MCs, and these genes have produced statistically significant results in the present systematic review and meta-analysis. The role of MCs in the fibrotic process is controversial, as studies in humans and in vitro data indicate a pro-fibrotic role of these cells, whereas animal studies have produced inconsistent results [51]. The reason may be the duration of the stimuli. Based on the fact that MCs are scarce in healthy human kidneys and are very rarely observed in glomeruli [52], these cells could serve as sensors of injury and could enhance the repair process. When the injury is short-lived, the MCs have an anti-fibrotic effect, but when the injury is chronic or repeated, the MCs have a pro-fibrotic effect [51]. It has been also found that an increase in the number of MCs is correlated negatively with renal function [53,54], but is correlated positively with the extent of fibrosis [54–56]. One more reason for the inconsistency is the fact that MCs produce both pro-fibrotic and anti-fibrotic mediators [49,57–60].

### 4. Conclusions

In summary, the present systematic review and meta-analysis indicate, as key players of fibrosis in DN, sixteen genes (*ACE*, *AGT*, *EDN1*, *EPO*, *FLT4*, *GREM1*, *IL1B*, *IL6*, *IL10*, *IL12RB1*, *NOS3*, *TGFB1*, *IGF2/INS/TH cluster*, and *VEGFA*). However, the results should be interpreted with caution because the number of studies in most meta-analyses is relatively small.

**Supplementary Materials:** The following supporting information can be downloaded at: https://www.mdpi.com/article/10.3390/ijms232315331/s1.

**Author Contributions:** Conceptualization, I.S.; methodology, M.T.; formal analysis, M.T.; investigation, M.T.; data curation, M.T., E.N., M.E. and T.E.; writing—original draft preparation, M.T., T.C.T. and I.S.; writing—review and editing, M.T. and T.C.T.; supervision, I.S. All authors have read and agreed to the published version of the manuscript.

Funding: This research received no external funding.

Institutional Review Board Statement: Not applicable.

Informed Consent Statement: Not applicable.

Conflicts of Interest: The authors declare no conflict of interest.

### References

- 1. Rich, S.S. Genetics of diabetes and its complications. J. Am. Soc. Nephrol. 2006, 17, 353–360. [CrossRef] [PubMed]
- 2. Cowie, C.C.; Port, F.K.; Wolfe, R.A.; Savage, P.J.; Moll, P.P.; Hawthorne, V.M. Disparities in incidence of diabetic end-stage renal disease according to race and type of diabetes. *N. Engl. J. Med.* **1989**, *321*, 1074–1079. [CrossRef] [PubMed]
- 3. Seaquist, E.R.; Goetz, F.C.; Rich, S.; Barbosa, J. Familial clustering of diabetic kidney disease. Evidence for genetic susceptibility to diabetic nephropathy. *N. Engl. J. Med.* **1989**, *320*, 1161–1165. [CrossRef] [PubMed]
- 4. Thomas, M.C.; Groop, P.H.; Tryggvason, K. Towards understanding the inherited susceptibility for nephropathy in diabetes. *Curr. Opin. Nephrol. Hypertens.* **2012**, 21, 195–202. [CrossRef]
- 5. Rutledge, J.C.; Ng, K.F.; Aung, H.H.; Wilson, D.W. Role of triglyceride-rich lipoproteins in diabetic nephropathy. *Nat. Rev. Nephrol.* **2010**, *6*, 361–370. [CrossRef]
- 6. Zhang, Y.; Jin, D.; Kang, X.; Zhou, R.; Sun, Y.; Lian, F.; Tong, X. Signaling Pathways Involved in Diabetic Renal Fibrosis. *Front. Cell Dev. Biol.* **2021**, *9*, 696542. [CrossRef]
- 7. Li, L.; Fu, H.; Liu, Y. The fibrogenic niche in kidney fibrosis: Components and mechanisms. *Nat. Rev. Nephrol.* **2022**, *18*, 545–557. [CrossRef]

Hu, L.; Ding, M.; He, W. Emerging Therapeutic Strategies for Attenuating Tubular EMT and Kidney Fibrosis by Targeting Wnt/β
-Catenin Signaling. Front. Pharmacol. 2022, 12, 830340. [CrossRef]

- 9. Edeling, M.; Ragi, G.; Huang, S.; Pavenstädt, H.; Susztak, K. *Developmental Signalling Pathways in Renal Fibrosis: The Roles of Notch, Wnt and Hedgehog*; Nature Publishing Group: Berlin/Heidelberg, Germany, 2016.
- 10. Arai, H.; Yanagita, M. Janus-Faced: Molecular Mechanisms and Versatile Nature of Renal Fibrosis. *Kidney360* **2020**, *1*, 697–704. [CrossRef]
- 11. Fragiadaki, M.; Macleod, F.M.; Ong, A.C.M. The controversial role of fibrosis in autosomal dominant polycystic kidney disease. *Int. J. Mol. Sci.* **2020**, *21*, 8936. [CrossRef]
- 12. Tziastoudi, M.; Stefanidis, I.; Stravodimos, K.; Zintzaras, E. Identification of Chromosomal Regions Linked to Diabetic Nephropathy: A Meta-Analysis of Genome-Wide Linkage Scans. *Genet. Test. Mol. Biomark.* **2019**, 23, 105–117. [CrossRef] [PubMed]
- 13. Tziastoudi, M.; Tziastoudi, M. The genetic map of diabetic nephropathy: Evidence from a systematic review and meta-analysis of genetic association studies. *Clin. Kidney J.* **2020**, *13*, 768–781. [CrossRef] [PubMed]
- 14. Tziastoudi, M.; Stefanidis, I.; Hadjigeorgiou, G.M.; Stravodimos, K.; Zintzaras, E. A systematic review and meta-analysis of genetic association studies for the role of inflammation and the immune system in diabetic nephropathy. *Clin. Kidney J.* **2017**, *10*, 293–300. [CrossRef] [PubMed]
- 15. Tziastoudi, M.; Dardiotis, E.; Pissas, G.; Filippidis, G.; Golfinopoulos, S.; Siokas, V.; Tachmitzi, S.V.; Eleftheriadis, T.; Hadjigeorgiou, G.M.; Tsironi, E.; et al. Serpin Family E Member 1 Tag Single-Nucleotide Polymorphisms in Patients with Diabetic Nephropathy: An Association Study and Meta-Analysis Using a Genetic Model-Free Approach. *Genes* 2021, 12, 1887. [CrossRef] [PubMed]
- 16. Stefanidis, I.; Tziastoudi, M.; Tsironi, E.E.; Dardiotis, E.; Tachmitzi, S.V.; Fotiadou, A.; Pissas, G.; Kytoudis, K.; Sounidaki, M.; Ampatzis, G.; et al. The contribution of genetic variants of SLC2A1 gene in T2DM and T2DM-nephropathy: Association study and meta-analysis. *Ren. Fail.* 2018, 40, 561–576. [CrossRef] [PubMed]
- 17. Maeda, S.; Osawa, N.; Hayashi, T.; Tsukada, S.; Kobayashi, M.; Kikkawa, R. Genetic variations associated with diabetic nephropathy and type II diabetes in a Japanese population. *Kidney Int. Suppl.* **2007**, 72, S43–S48. [CrossRef] [PubMed]
- 18. Germain, M.; Pezzolesi, M.G.; Sandholm, N.; McKnight, A.J.; Susztak, K.; Lajer, M.; Forsblom, C.; Marre, M.; Parving, H.H.; Rossing, P.; et al. SORBS1 gene, a new candidate for diabetic nephropathy: Results from a multi-stage genome-wide association study in patients with type 1 diabetes. *Diabetologia* **2015**, *58*, 543–548. [CrossRef]
- 19. Taira, M.; Imamura, M.; Takahashi, A.; Kamatani, Y.; Yamauchi, T.; Araki, S.I.; Tanaka, N.; Van Zuydam, N.R.; Ahlqvist, E.; Toyoda, M.; et al. A variant within the FTO confers susceptibility to diabetic nephropathy in Japanese patients with type 2 diabetes. *PLoS ONE* **2018**, *13*, e0208654. [CrossRef]
- 20. Pezzolesi, M.G.; Poznik, G.D.; Mychaleckyj, J.C.; Paterson, A.D.; Barati, M.T.; Klein, J.B.; Ng, D.P.; Placha, G.; Canani, L.H.; Bochenski, J.; et al. Genome-Wide Association Scan for Diabetic Nephropathy Susceptibility Genes in Type 1 Diabetes. *Diabetes* 2009, 58, 1403–1410. [CrossRef]
- McDonough, C.W.; Palmer, N.D.; Hicks, P.J.; Roh, B.H.; An, S.S.; Cooke, J.N.; Hester, J.M.; Wing, M.R.; Bostrom, M.A.; Rudock, M.E.; et al. A genome-wide association study for diabetic nephropathy genes in African Americans. *Kidney Int.* 2011, 79, 563–572. [CrossRef]
- 22. Hsieh, A.R.; Huang, Y.C.; Yang, Y.F.; Lin, H.J.; Lin, J.M.; Chang, Y.W.; Wu, C.M.; Liao, W.L.; Tsai, F.J. Lack of association of genetic variants for diabetic retinopathy in Taiwanese patients with diabetic nephropathy. *BMJ Open Diabetes Res. Care* **2020**, *8*, e000727. [CrossRef] [PubMed]
- 23. Page, M.J.; McKenzie, J.E.; Bossuyt, P.M.; Boutron, I.; Hoffmann, T.C.; Mulrow, C.D.; Shamseer, L.; Tetzlaff, J.M.; Akl, E.A.; Brennan, S.E.; et al. The PRISMA 2020 statement: An updated guideline for reporting systematic reviews. *BMJ* **2021**, *372*, 89. Available online: https://www.bmj.com/content/372/bmj.n71 (accessed on 14 August 2022).
- 24. Zintzaras, E. The power of generalized odds ratio in assessing association in genetic studies. *J. Appl. Stat.* **2012**, *39*, 2569–2581. [CrossRef]
- 25. Zintzaras, E. The generalized odds ratio as a measure of genetic risk effect in the analysis and meta-analysis of association studies. *Stat. Appl. Genet. Mol. Biol.* **2010**, *9*, 21. [CrossRef] [PubMed]
- 26. DerSimonian, R.; Laird, N. Meta-analysis in clinical trials. Control Clin. Trials 1986, 7, 177-188. [CrossRef]
- 27. Higgins, J.P.T.; Thompson, S.G. Quantifying heterogeneity in a meta-analysis. Stat. Med. 2002, 21, 1539–1558. [CrossRef]
- 28. Cochran, W. The combination of estimates from different experiments. Biometrics 1954, 10, 101–129. [CrossRef]
- 29. Egger, M.; Davey Smith, G.; Schneider, M.; Minder, C. Bias in meta-analysis detected by a simple, graphical test. *BMJ* **1997**, *315*, 629–634. [CrossRef]
- 30. Karihaloo, A. Anti-fibrosis therapy and diabetic nephropathy. Curr. Diabetes Rep. 2012, 12, 414–422. [CrossRef]
- 31. Mezzano, S.A.; Ruiz-Ortega, M.; Egido, J. Angiotensin II and Renal Fibrosis [Internet]. 2001. Available online: http://www.hypertensionaha.org (accessed on 11 June 2022).
- 32. Meng, X.M.; Nikolic-Paterson, D.J.; Lan, H.Y. TGF-β: The master regulator of fibrosis. *Nat. Rev. Nephrol.* **2016**, 12, 325–338. [CrossRef]
- 33. van Meeteren, L.A.; ten Dijke, P. Regulation of endothelial cell plasticity by TGF-β. *Cell Tissue Res.* **2012**, 347, 177–186. [CrossRef] [PubMed]

34. Lemos, D.R.; McMurdo, M.; Karaca, G.; Wilflingseder, J.; Leaf, I.A.; Gupta, N.; Miyoshi, T.; Susa, K.; Johnson, B.G.; Soliman, K.; et al. Interleukin-1b activates a MYC-dependent metabolic switch in kidney stromal cells necessary for progressive tubulointerstitial fibrosis. *J. Am. Soc. Nephrol.* **2018**, 29, 1690–1705. [CrossRef] [PubMed]

- 35. Ranganathan, P.; Jayakumar, C.; Ramesh, G. Proximal tubule-specific overexpression of netrin-1 suppresses acute kidney injury-induced interstitial fibrosis and glomerulosclerosis through suppression of IL-6/STAT3 signaling. *Am. J. Physiol. Ren. Physiol.* **2013**, 304, 1054–1065.
- 36. Steen, E.H.; Wang, X.; Balaji, S.; Butte, M.J.; Bollyky, P.L.; Keswani, S.G. The Role of the Anti-Inflammatory Cytokine Interleukin-10 in Tissue Fibrosis. *Adv. Wound Care* **2020**, *9*, 184–198. [CrossRef] [PubMed]
- 37. Nishimura, K.; Taguchi, K.; Kishi, S.; Brooks, C.R.; Ochi, A.; Kadoya, H.; Ikeda, Y.; Miyoshi, M.; Tamaki, M.; Abe, H.; et al. Dual disruption of eNOS and ApoE gene accelerates kidney fibrosis and senescence after injury. *Biochem. Biophys. Res. Commun.* **2021**, 556, 142–148. [CrossRef]
- 38. Sun, X.; Gan, H.; Xia, Y. A meta-analysis of the effects of endothelial nitric oxide synthase 4ba polymorphism on renal interstitial fibrosis in diabetic nephropathy. *Ann. Palliat. Med.* **2021**, *10*, 633–645. [CrossRef] [PubMed]
- 39. Neder, T.H.; Schrankl, J.; Fuchs, M.A.A.; Broeker, K.A.E.; Wagner, C. Endothelin receptors in renal interstitial cells do not contribute to the development of fibrosis during experimental kidney disease. *Pflug. Arch.* **2021**, *473*, 1667–1683. [CrossRef] [PubMed]
- 40. Heerspink, H.J.; Parving, H.H.; Andress, D.L.; Bakris, G.; Correa-Rotter, R.; Hou, F.F.; Kitzman, D.W.; Kohan, D.; Makino, H.; McMurray, J.J.; et al. Atrasentan and renal events in patients with type 2 diabetes and chronic kidney disease (SONAR): A double-blind, randomised, placebo-controlled trial. *Lancet* 2019, 393, 1937–1947.
- 41. Geng, X.C.; Hu, Z.P.; Lian, G.Y. Erythropoietin ameliorates renal interstitial fibrosis via the inhibition of fibrocyte accumulation. *Mol. Med. Rep.* **2015**, *11*, 3860–3865. [CrossRef]
- 42. Church, R.H.; Ali, I.; Tate, M.; Lavin, D.; Krishnakumar, A.; Kok, H.M.; Hombrebueno, J.R.; Dunne, P.D.; Bingham, V.; Goldschmeding, R.; et al. Gremlin1 plays a key role in kidney development and renal fibrosis. *Am. J. Physiol. Ren. Physiol.* **2017**, 312, 1141–1157. [CrossRef]
- 43. Kinashi, H.; Ito, Y.; Sun, T.; Katsuno, T.; Takei, Y. Roles of the TGF-β–VEGF-C pathway in fibrosis-related lymphangiogenesis. *Int. J. Mol. Sci.* **2018**, *19*, 2487. [CrossRef] [PubMed]
- 44. Grotendorst, G.R.; Rahmanie, H.; Duncan, M.R. Combinatorial signaling pathways determine fibroblast proliferation and myofibroblast differentiation. *FASEB J.* **2004**, *18*, 469–479. [CrossRef] [PubMed]
- 45. Li, J.; Liu, H.; Takagi, S.; Nitta, K.; Kitada, M.; Srivastava, S.P.; Takagaki, Y.; Kanasaki, K.; Koya, D. Renal protective effects of empagliflozin via inhibition of EMT and aberrant glycolysis in proximal tubules. *JCI Insight* **2020**, *5*, e129034. [CrossRef] [PubMed]
- 46. Miao, C.; Zhu, X.; Wei, X.; Long, M.; Jiang, L.; Li, C.; Jin, D.; Du, Y. Pro- and anti-fibrotic effects of vascular endothelial growth factor in chronic kidney diseases. *Ren. Fail.* 2022, 44, 881–892. [CrossRef] [PubMed]
- 47. Moretti, L.; Stalfort, J.; Barker, T.H.; Abebayehu, D. The interplay of fibroblasts, the extracellular matrix, and inflammation in scar formation. *J. Biol. Chem.* **2022**, 298, 101530. [CrossRef]
- 48. Owens, E.P.; Vesey, D.A.; Kassianos, A.J.; Healy, H.; Hoy, W.E.; Gobe, G.C. Biomarkers and the role of mast cells as facilitators of inflammation and fibrosis in chronic kidney disease. *Transl. Androl. Urol.* **2019**, *8*, S175–S183. [CrossRef]
- 49. Strattan, E.; Hildebrandt, G.C. Mast cell involvement in fibrosis in chronic graft-versus-host disease. *Int. J. Mol. Sci.* **2021**, 22, 2385. [CrossRef]
- 50. Conti, P.; Caraffa, A.; Mastrangelo, F.; Tettamanti, L.; Ronconi, G.; Frydas, I.; Kritas, S.K.; Theoharides, T.C. Critical role of inflammatory mast cell in fibrosis: Potential therapeutic effect of IL-37. *Cell Prolif.* **2018**, *51*, e12475. [CrossRef]
- 51. Bradding, P.; Pejler, G. The controversial role of mast cells in fibrosis. *Immunol. Rev.* **2018**, 282, 198–231.
- 52. El-Koraie, A.F.; Baddour, N.M.; Adam, A.G.; el Kashef, E.H.; el Nahas, A.M. Role of stem cell factor and mast cells in the progression of chronic glomerulonephritides. *Kidney Int.* **2001**, *60*, 167–172. [CrossRef]
- 53. Ehara, T.; Shigematsu, H. Contribution of mast cells to the tubulointerstitial lesions in IgA nephritis. *Kidney Int.* **1998**, *54*, 1675–1683. [CrossRef]
- 54. Kurusu, A.; Suzuki, Y.; Horikoshi, S.; Shirato, I.; Tomino, Y. Relationship between Mast Cells in the Tubulointerstitium and Prognosis of Patients with IgA Nephropathy. *Nephron* **2001**, *89*, 391–397. [CrossRef] [PubMed]
- 55. Kondo, S.; Kagami, S.; Kido, H.; Strutz, F.; Müller, G.A.; Kuroda, Y. Role of Mast Cell Tryptase in Renal Interstitial Fibrosis. *J. Am. Soc. Nephrol.* **2001**, *12*, 1668–1676. [CrossRef] [PubMed]
- 56. Roberts, I.S.D.; Brenchley, P.E.C. Mast cells: The forgotten cells of renal fibrosis. *J. Clin. Pathol.* **2000**, *53*, 858–862. [CrossRef] [PubMed]
- 57. Gieseck, R.L.; Wilson, M.S.; Wynn, T.A. Type 2 immunity in tissue repair and fibrosis. *Nat. Rev. Immunol.* **2018**, *18*, 62–76. [PubMed]
- 58. Mukai, K.; Tsai, M.; Saito, H.; Galli, S.J. Mast cells as sources of cytokines, chemokines, and growth factors. *Immunol. Rev.* **2018**, 282, 121–150.
- 59. Theoharides, T.C.; Alysandratos, K.D.; Angelidou, A.; Delivanis, D.A.; Sismanopoulos, N.; Zhang, B.; Asadi, S.; Vasiadi, M.; Weng, Z.; Miniati, A.; et al. Mast cells and inflammation. *Biochim. Et Biophys. Acta Mol. Basis Dis.* **2012**, *1822*, 21–33.
- 60. Hügle, T. Beyond allergy: The role of mast cells in fibrosis. Swiss Med. Wkly. 2014, 144, w13999. [CrossRef]