



**QUEEN'S
UNIVERSITY
BELFAST**



Multiomic approaches to chronic disease risk

Dr Gareth McKay
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16th June 2023

**SHAPING A
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Outline of my talk

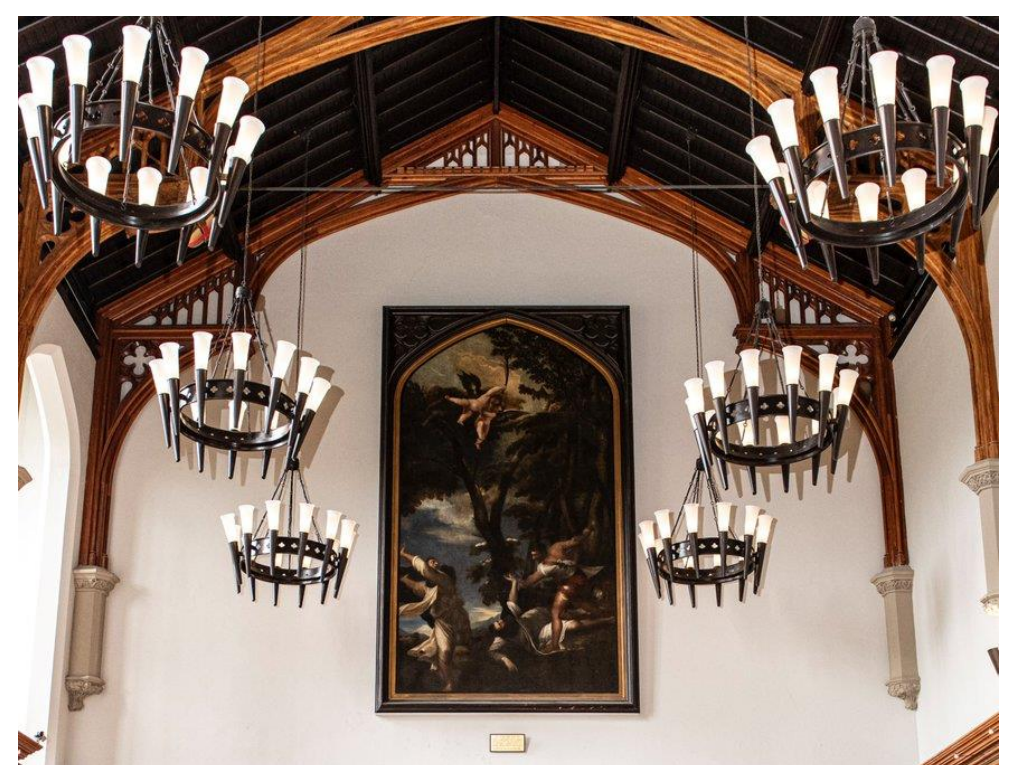
- A bit about me
- Northern Ireland
- Queen's University Belfast
- My career track
- Chronic Kidney Disease (CKD)
- Diabetic Kidney Disease (DKD) Genetics
- DKD Epigenetics
- DKD miRNA analysis
- Oculomics
- Kidney Transplantation



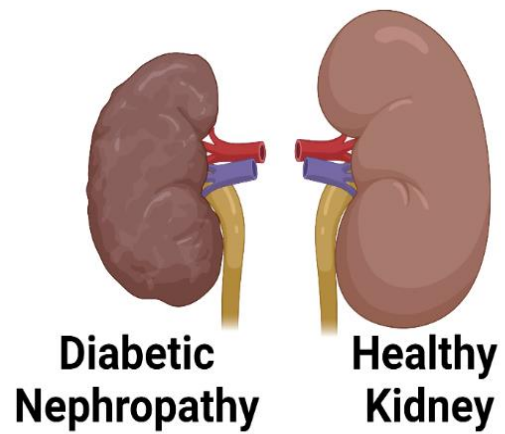
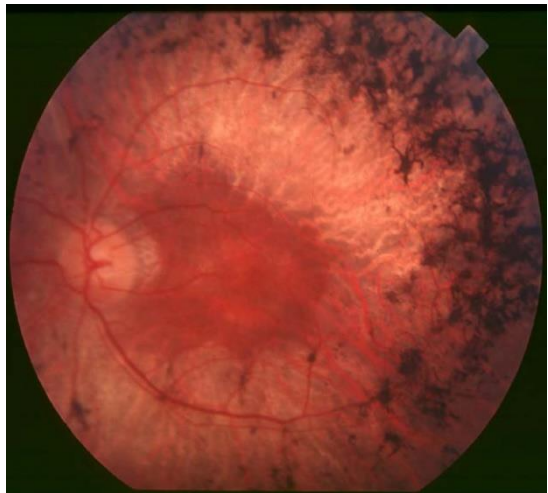




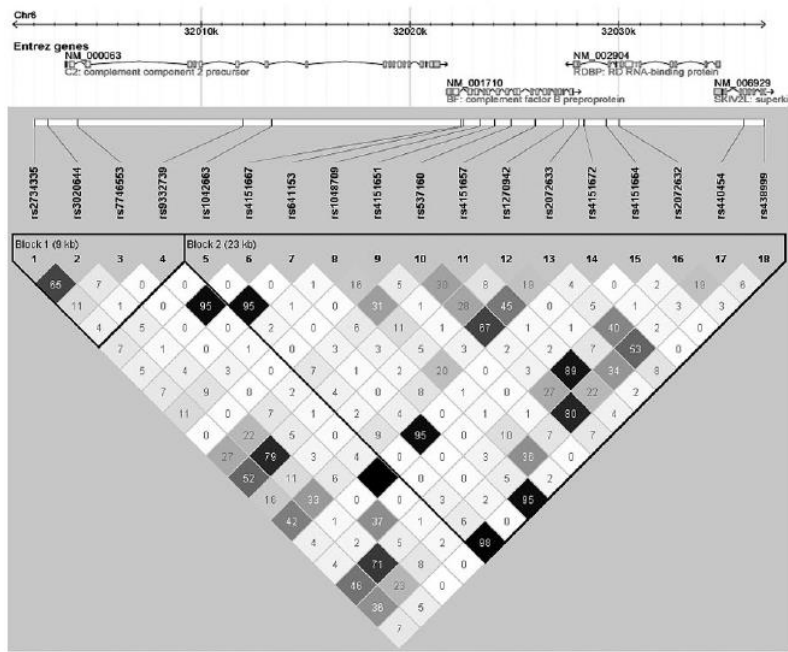








<https://ec.europa.eu/info/funding-tenders/opportunities/portal/screen/work-as-an-expert>



Further Assessment of the Complement Component 2 and Factor B Region Associated with Age-Related Macular Degeneration

Gareth J. McKay,¹ Giuliana Silvestri,¹ Christopher C. Patterson,² Ruth E. Hogg,¹ Usba Chakravarthy,¹ and Anne E. Hughes²
Investigative Ophthalmology & Visual Science, February 2009, Vol. 50, No. 2



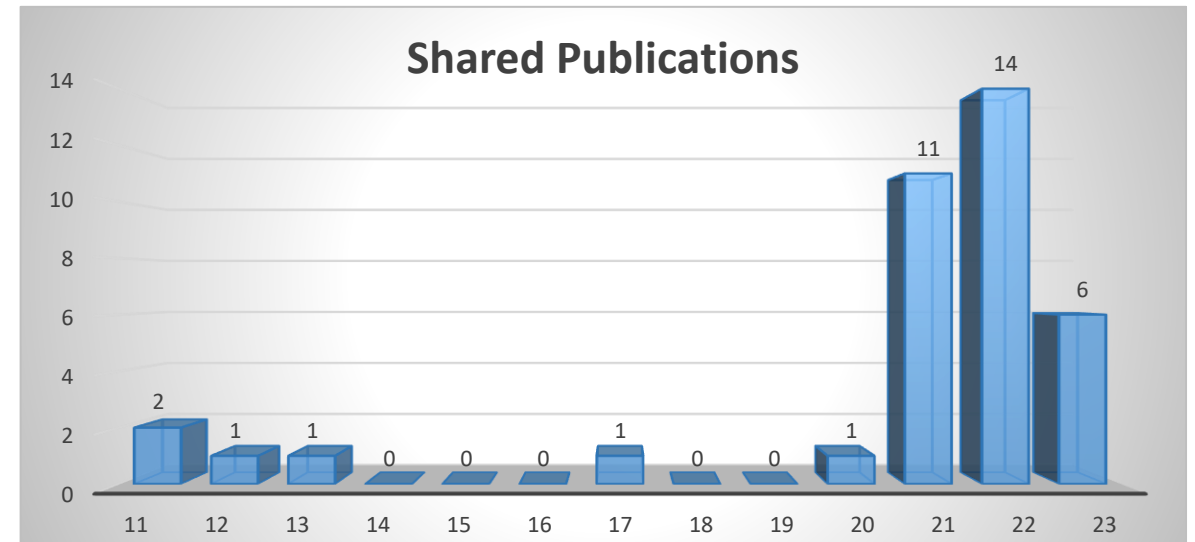
From: Ammarin Thakkinstian <raatk@mahidol.ac.th>
 Sent: 08 June 2010 01:57
 To: Gareth McKay <g.j.mckay@qub.ac.uk>
 Subject: C2-C3 & AMD

Tuesday, June 08, 2010

Dear Prof McKay,

We are currently performing a meta-analysis of C2-C3 polymorphisms and age-related macular degeneration. We have identified your study that published in

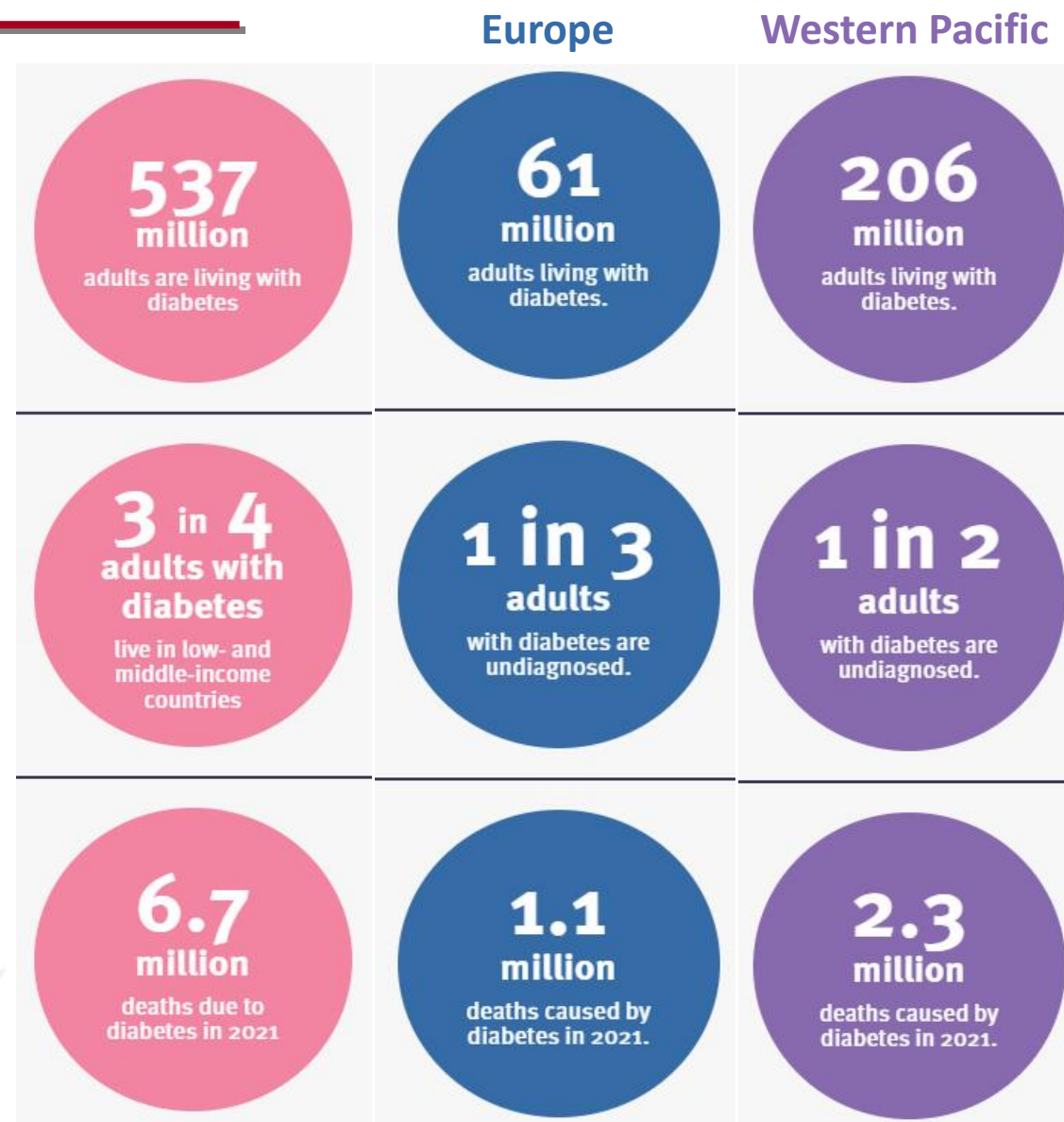
McKay GJ, Silvestri G, Patterson CC, Hogg RE, Chakravarthy U, Hughes AE: **Further assessment of the complement component 2 and factor B region associated with age-related macular degeneration.** *Invest Ophthalmol Vis Sci* 2009, **50**:533-539.



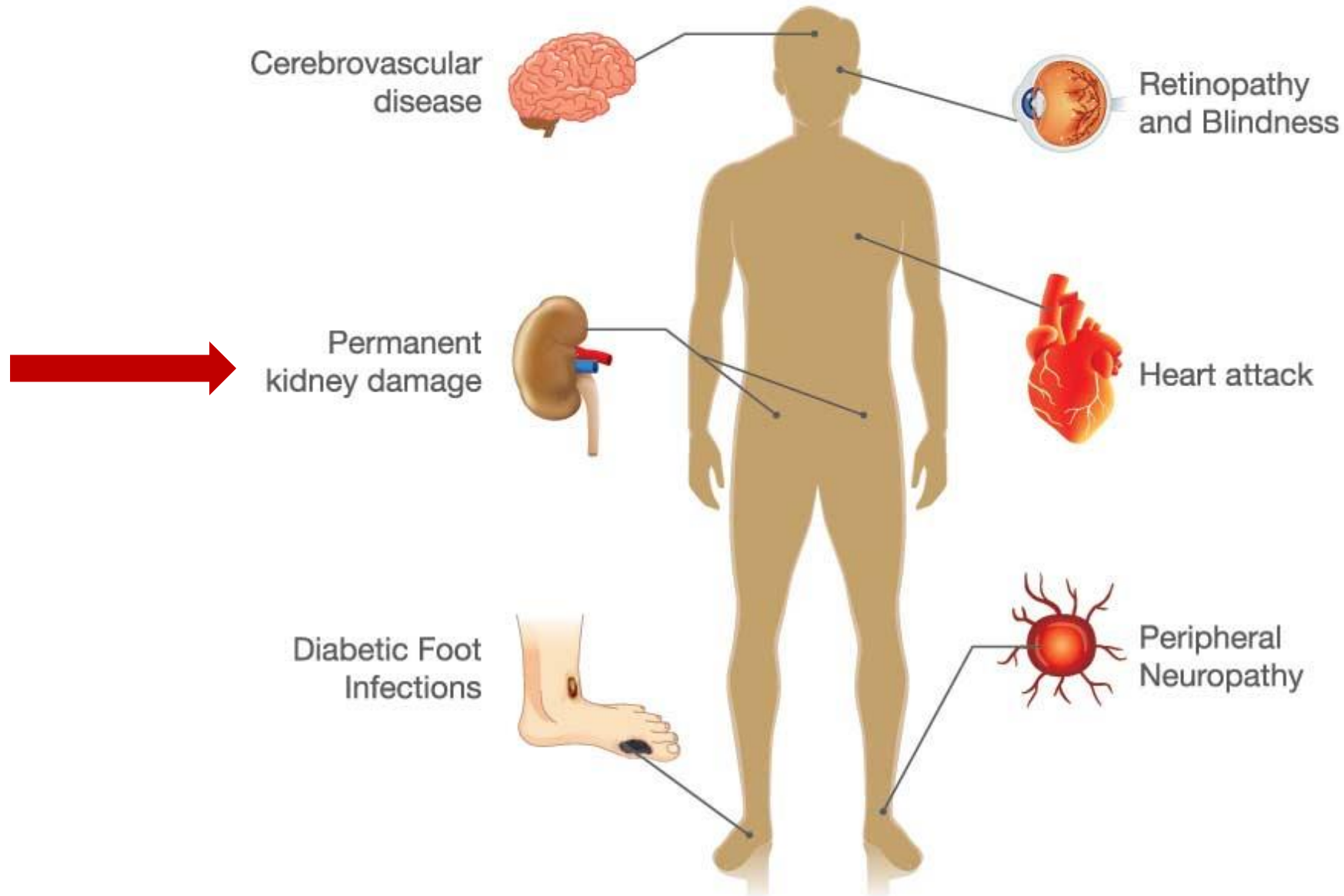
Diabetes and it's complications



Diabetes around the world in 2021

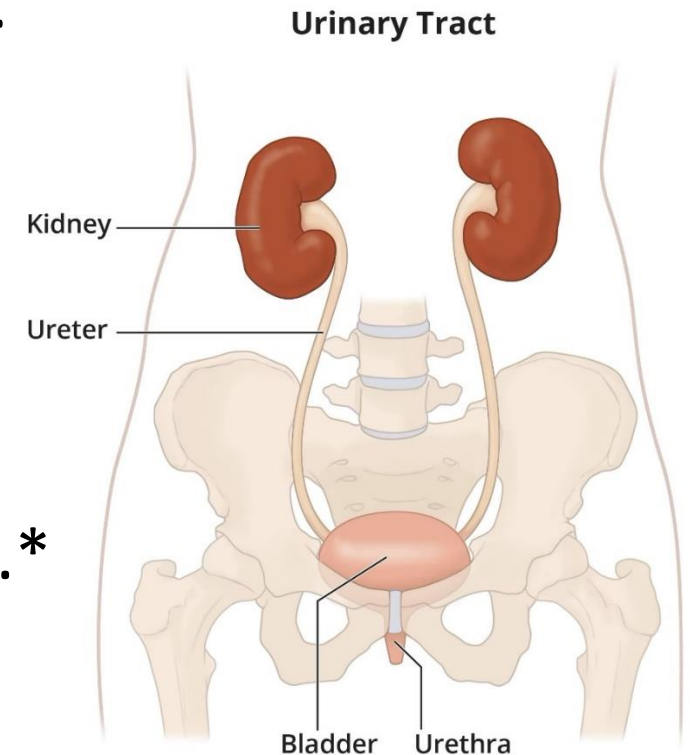
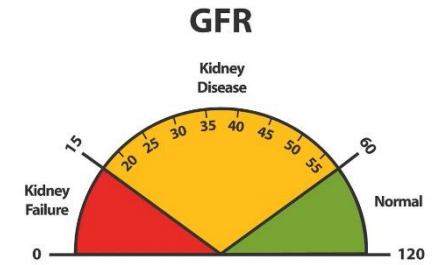


Complications of diabetes mellitus



Chronic Kidney Disease (CKD)

- CKD means your kidneys are damaged and can't filter blood the way they should.
- The disease is called “chronic” because the damage to your kidneys happens slowly over a long period of time.
- This damage can cause waste to build up in your body.
- CKD can also cause other health problems.
- Estimated to be the 5th leading cause of death by 2040.*



Criteria for CKD

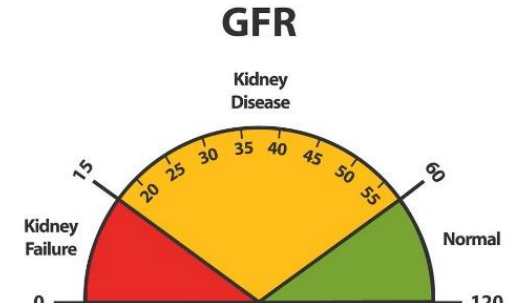
➤ Abnormalities of kidney structure or function, present for >3 months, with implications for health

➤ Either of the following must be present for >3 months:

➤ GFR <60 mL/min/1.73 m²

➤ ACR >30 mg/g

➤ Markers of kidney damage (one or more*)



Prognosis of CKD by GFR and Albuminuria Categories

				Albuminuria categories Description and range		
				A1	A2	A3
				Normal to mildly increased	Moderately increased	Severely increased
				<30 mg/g <3 mg/mmol	30-299 mg/g 3-29 mg/mmol	≥300 mg/g ≥30 mg/mmol
GFR categories (ml/min/1.73 m ²) Description and range	G1	Normal or high	≥90			
	G2	Mildly decreased	60-90			
	G3a	Mildly to moderately decreased	45-59			
	G3b	Moderately to severely decreased	30-44			
	G4	Severely decreased	15-29			
	G5	Kidney failure	<15			

Green: low risk (if no other markers of kidney disease, no CKD); Yellow: moderately increased risk; Orange: high risk; Red, very high risk.
KDIGO 2012

*Markers of kidney damage can include nephrotic syndrome, nephritic syndrome, tubular syndromes, urinary tract symptoms, asymptomatic urinalysis abnormalities, asymptomatic radiologic abnormalities, hypertension due to kidney disease.

Main CKD Risk Factors

Non-Modifiable

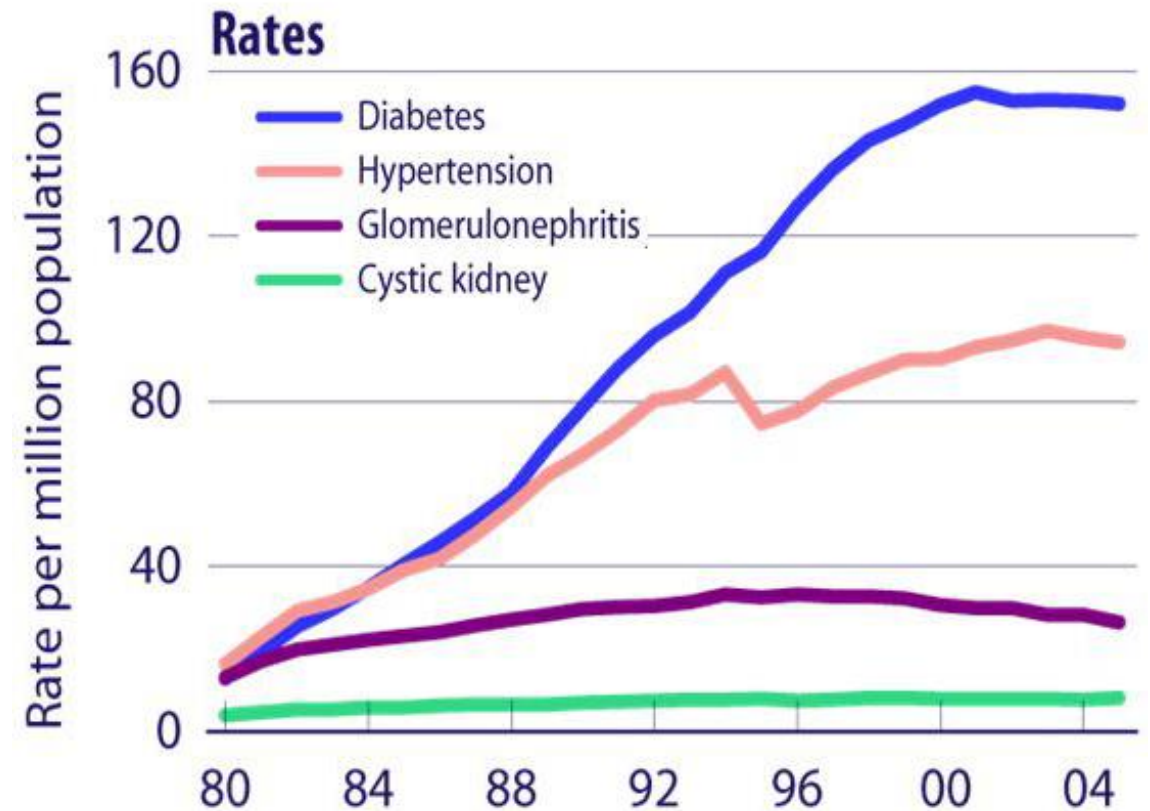
- Family history of kidney disease, diabetes, or hypertension
- Age 60 or older (GFR declines normally with age)
- Belong to a population group with a high rate of diabetes or hypertension, such as African Americans, Hispanic Americans, Asian, Pacific Islanders, and American Indians

Main CKD Risk Factors

Modifiable

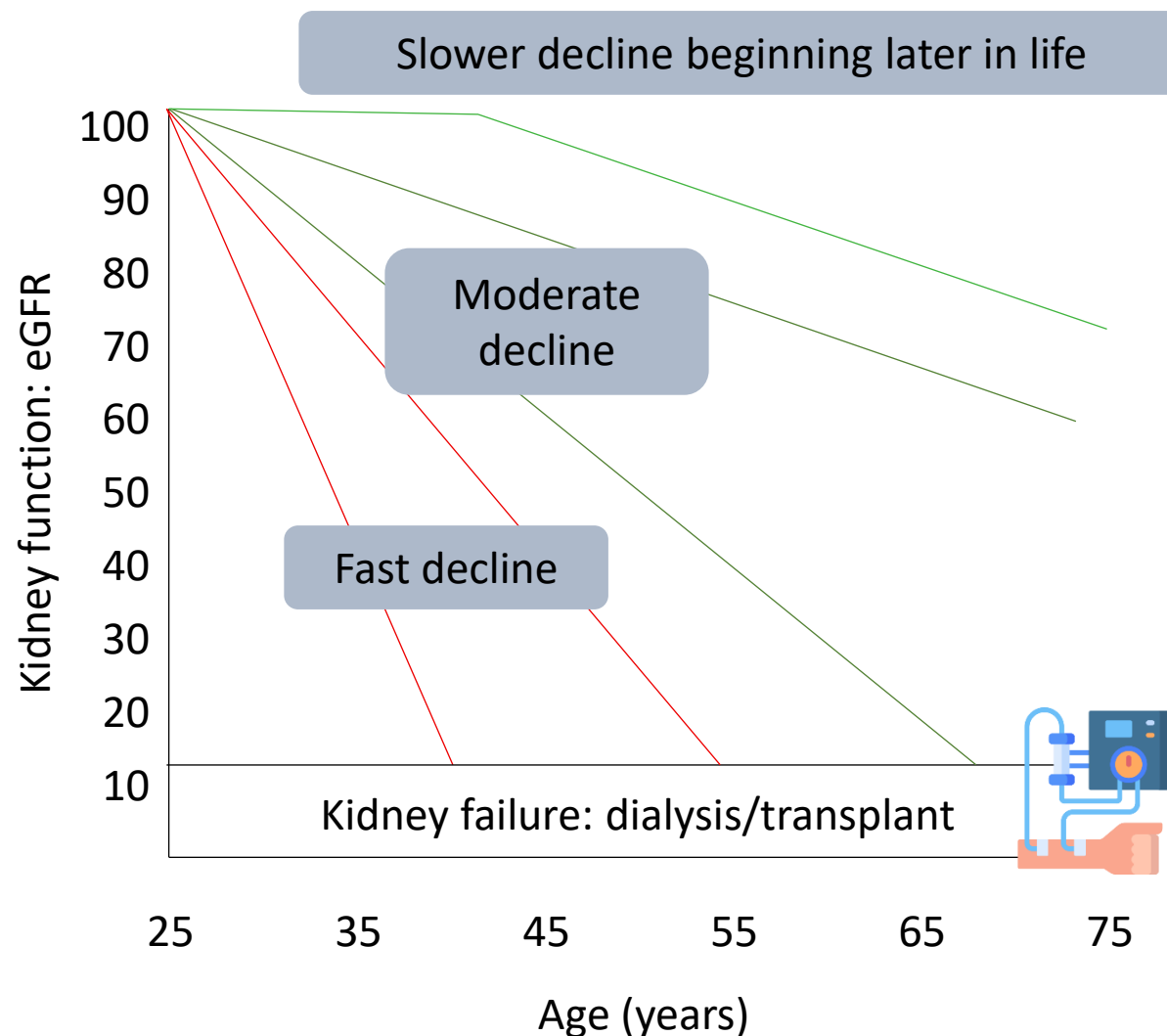
- Diabetes
- Hypertension
- History of AKI
- Frequent NSAID use

Diabetes and hypertension are leading causes of kidney failure



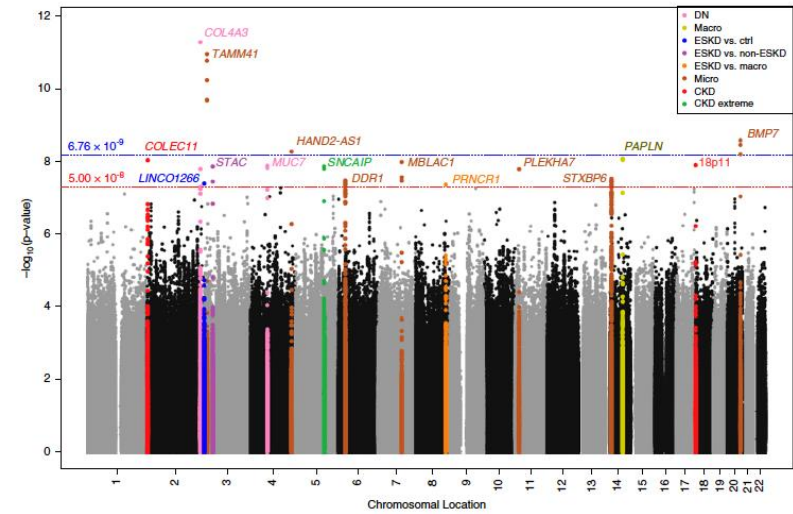
Incident ESRD rates, by primary diagnosis, adjusted for age, sex, & race.

CKD Treatment Rationale



- Renal function declines over time
- Eventual outcome: dialysis or kidney transplant
- Rate of decline varies by person
- Aim: slow the rate of decline = delay dialysis/transplant

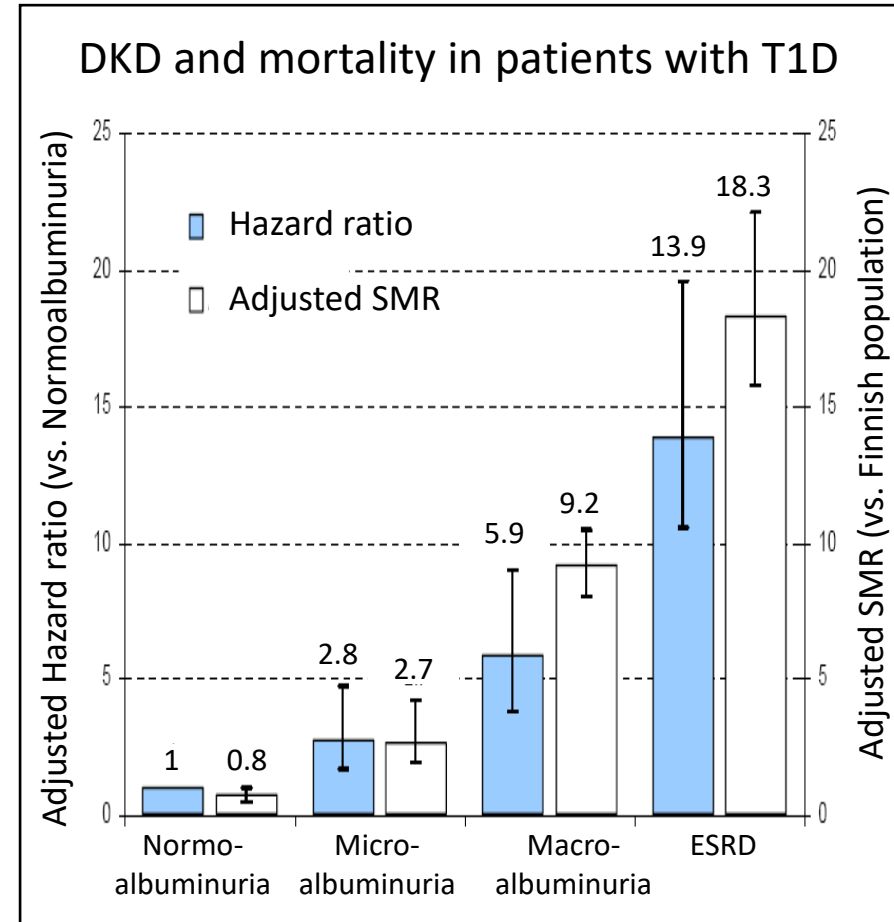
Genetics and Diabetic Kidney Disease





Genetic Risk in Diabetic Kidney Disease (DKD)

- Diabetes is the leading risk factor for kidney disease
- Type 1 diabetes affects >9M people globally, with approximately 40% developing diabetic kidney disease.
- All excess mortality in type 1 diabetes (T1D) associated with DKD¹
- DKD is inherited in T1D^{2,3,4}
 - Sibling risk $\lambda_s = 2.1$ ³



¹ Groop et al. Diabetes 2009

² Seaquist et al. NEJM 1989; ³ Quinn et al. Diabetologia 1996; ⁴ Harjutsalo et al Diabetes 2004

Familial aggregation of DKD

The New England Journal of Medicine

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Volume 320

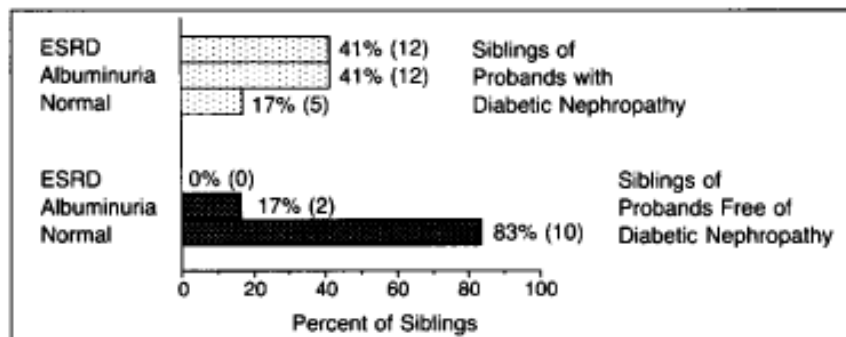
MAY 4, 1989

Number 18

FAMILIAL CLUSTERING OF DIABETIC KIDNEY DISEASE

Evidence for Genetic Susceptibility to Diabetic Nephropathy

ELIZABETH R. SEAQUIST, M.D., FREDERICK C. GOETZ, M.D., STEPHEN RICH, PH.D., AND JOSÉ BARBOSA, M.D.



2.1-2.3X increased risk of DKD
in T1D siblings of probands with DKD

GWAS in DKD

ORIGINAL ARTICLE

Genome-Wide Association Scan for Diabetic Nephropathy Susceptibility Genes in Type 1 Diabetes

Marcus G. Pezolesi,¹ G. David Poznik,¹ Josyf C. Mychaleckyj,² Andrew D. Paterson,^{3,4} Michelle T. Barati,⁵ Jon B. Klein,² Daniel P.K. Ng,^{1,6} Grzegorz Placha,^{1,7} Luis H. Canani,^{1,8} Jacek Bochenski,¹ Daryl Waggott,⁹ Michael L. Merchant,² Bozena Krolewski,¹ Lucia Mirea,^{4,9} Krzysztof Wanic,¹ Pisut Katavetin,¹ Masahiko Kure,¹ Pawel Wolkow,^{1,10} Jonathon S. Dunn,¹ Adam Smiles,¹ William H. Walker,¹ Andrew P. Boright,¹¹ Shelley B. Bull,^{4,9} the DCCT/EDIC Research Group,^{*} Alessandro Doria,¹ John J. Rogus,¹ Stephen S. Rich,² James H. Warram,¹ and Andrzej S. Krolewski¹

- *Diabetes* 2009
- US GoKinD
- Discovery 1.7K
- No GWS findings

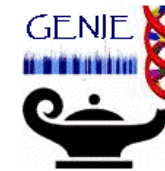
OPEN ACCESS Freely available online

PLOS GENETICS

New Susceptibility Loci Associated with Kidney Disease in Type 1 Diabetes

Niina Sandholm^{1,2,3,9}, Rany M. Salem^{4,5,6,9}, Amy Jayne McKnight^{7,9}, Eoin P. Brennan^{8,9,9}, Carol Forsblom^{1,2}, Tamara Isakova¹⁰, Gareth J. McKay⁷, Winfred W. Williams^{6,11}, Denise M. Sadlier^{8,9},

- 2012
- GENIE
- Discovery 6K
- 2 GWS findings for ESRD, none for DN



Chromosome 2q31.1 Associates with ESRD in Women with Type 1 Diabetes

Niina Sandholm,^{*††} Amy Jayne McKnight,[§] Rany M. Salem,^{||†**} Eoin P. Brennan,^{†† ††} Carol Forsblom,^{*†} Valma Harjutsalo,^{*†§§} Ville-Petteri Mäkinen,^{*† |||} Gareth J. McKay,[§] Denise M. Sadlier,^{††††} Winfred W. Williams,^{**†††} Finian Martin,^{††††} Nicolae Mircea Panduru,^{***} Lise Tarnow,^{†††††} Jaakko Tuomilehto,^{§§ §§§|||††††} Karl Tryggvason,^{****} Gianpaolo Zerbini,^{††††} Mary E. Comeau,^{††††} Carl D. Langefeld,^{††††} for the FIND Consortium; Catherine Godson,^{††††} Joel N. Hirschhorn,^{||†**} Alexander P. Maxwell,^{§ §§§§} Jose C. Florez,^{||**†††} and Per-Henrik Groop,^{*† |||||} on behalf of the FinnDiane Study Group and the GENIE Consortium.

- *JASN* 2013
- FinnDiane
- Discovery 3.6K
- 1 GWS finding for ESRD in women

T2D DKD GWAS

RESEARCH ARTICLE

Genome-Wide Association and Trans-ethnic Meta-Analysis for Advanced Diabetic Kidney Disease: Family Investigation of Nephropathy and Diabetes (FIND)

Sudha K. Iyengar¹*, John R. Sedor^{2,3}*, Barry I. Freedman⁴*, W. H. Linda Kao⁵†

- *PLoS Genet* 2015
- FIND
- Discovery 6.2K
- Multiethnic
- One GWS finding at *SCAF8-CNKSR3*

1414

Diabetes Volume 67, July 2018



A Genome-Wide Association Study of Diabetic Kidney Disease in Subjects With Type 2 Diabetes

Natalie R. van Zuydam,^{1,2} Emma Ahlqvist,³ Niina Sandholm,^{4,5,6} Harshal Deshmukh,⁷ N. William Rayner,^{1,2,8}

- *Diabetes* 2018
- SUMMIT
- Discovery 5.7K
- No reproducible findings
- Combined T2D+T1D

Diabetic Nephropathy Collaborative Research Initiative



FinnDiane
InterDiane
UK-ROI
Scotland
France-Belgium
Steno (Denmark)
Sweden
Italy

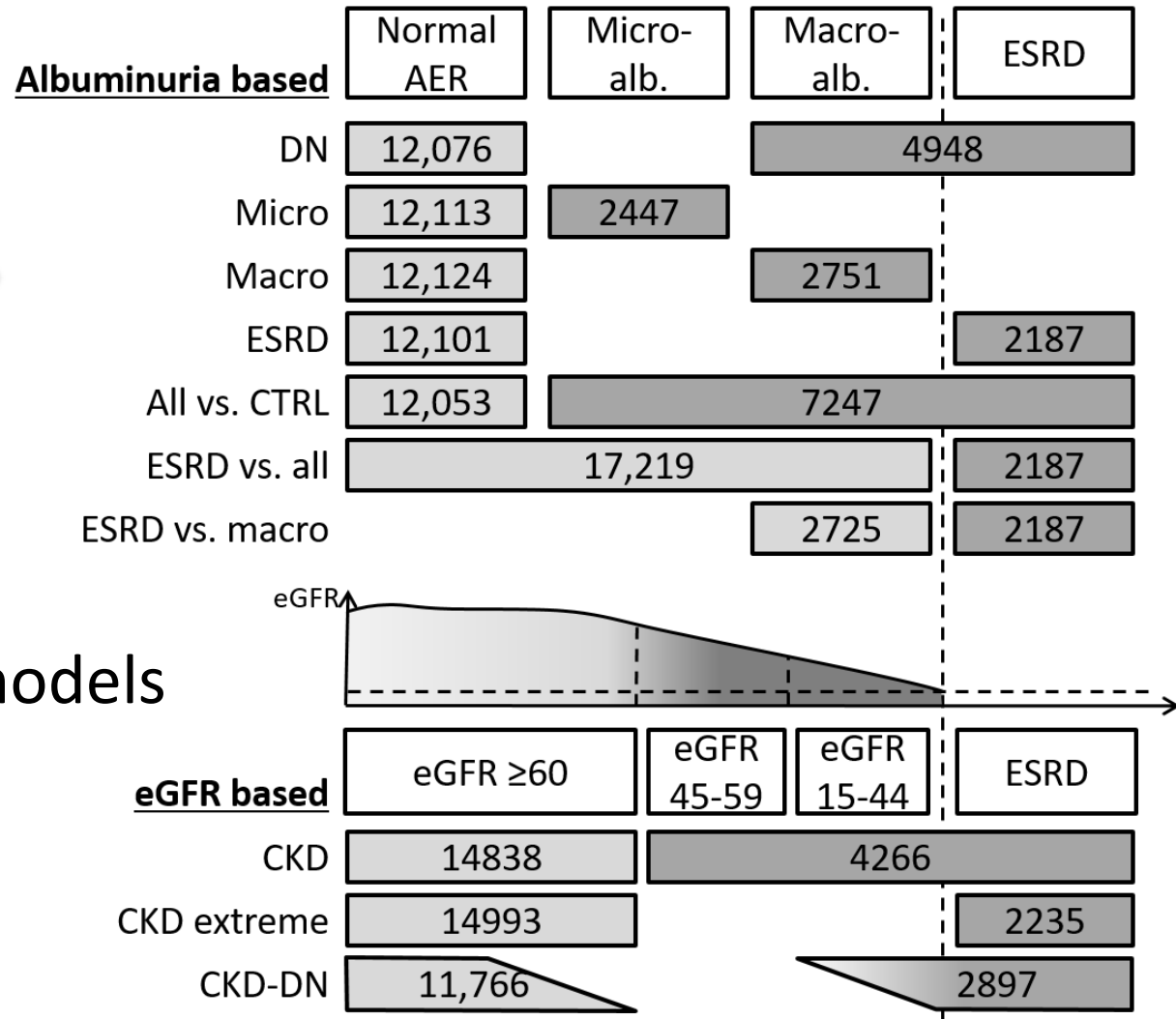


17 cohorts

19,406 participants

Phenotype definitions and contrasts

- Genome-wide significance:
 - $P < 5 \times 10^{-8}$
- 10 phenotypes + 2 covariate models
- 7.4 effective tests
- Experiment-wide significance:
 - $P < 6.76 \times 10^{-9}$



Genome-Wide Association Study of Diabetic Kidney Disease Highlights Biology Involved in Glomerular Basement Membrane Collagen

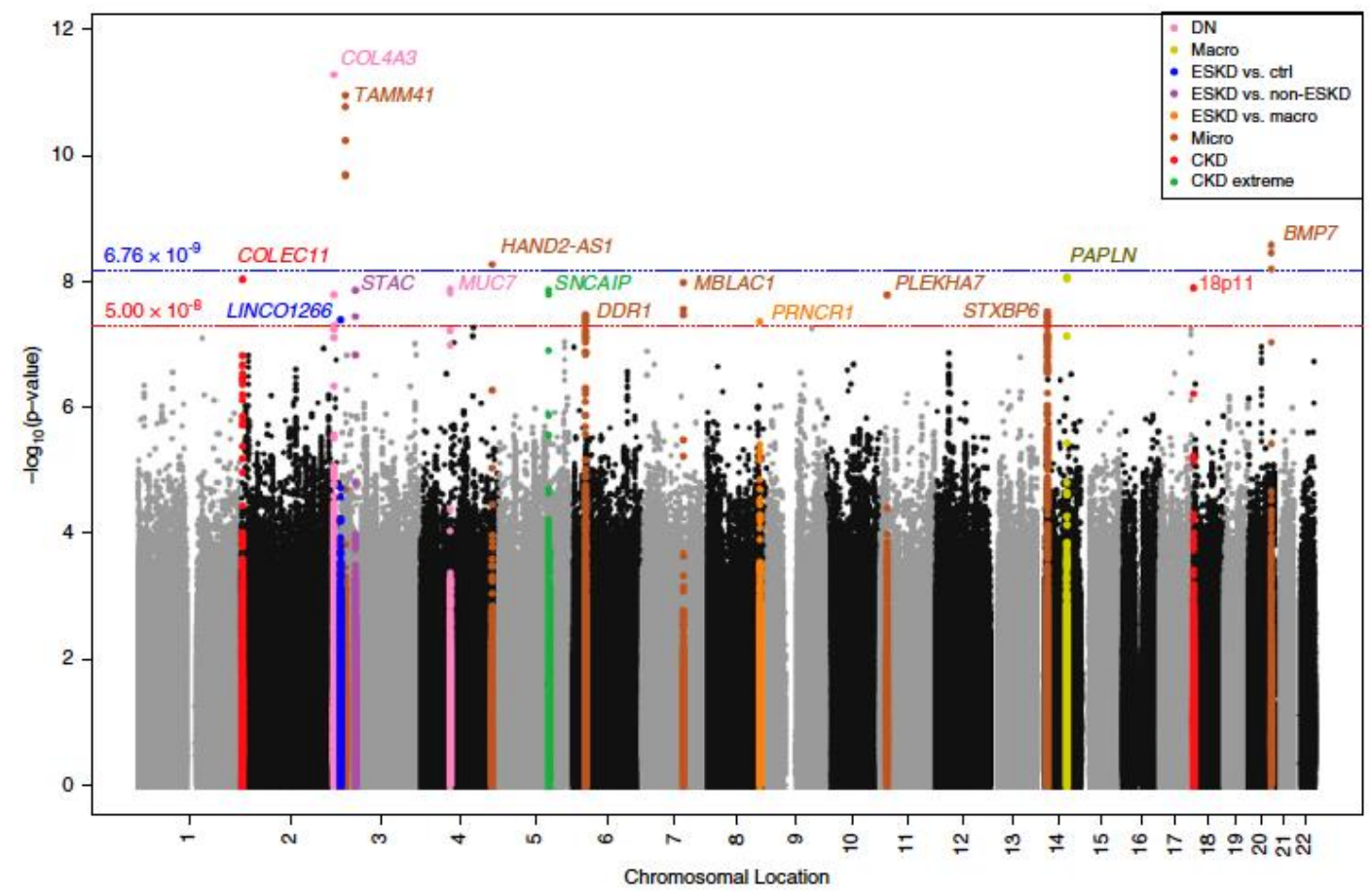
Rany M. Salem¹, Jennifer N. Todd^{2,3,4}, Niina Sandholm^{5,6,7}, Joanne B. Cole^{2,3,4}, Wei-Min Chen⁸, Darrell Andrews⁹, Marcus G. Pezzolesi¹⁰, Paul M. McKeigue¹¹, Linda T. Hiraki¹², Chengxiang Qiu¹³, Viji Nair¹⁴, Chen Di Liao¹², Jing Jing Cao¹², Erkkka Valo^{5,6,7}, Suna Onengut-Gumuscu⁸, Adam M. Smiles¹⁵, Stuart J. McGurnaghan¹⁶, Jani K. Haukka^{5,6,7}, Valma Harjutsalo^{5,6,7,17}, Eoin P. Brennan⁹, Natalie van Zuydam^{18,19}, Emma Ahlqvist²⁰, Ross Doyle⁹, Tarunveer S. Ahluwalia²¹, Maria Lajer²¹, Maria F. Hughes⁹, Jihwan Park¹³, Jan Skupien¹⁵, Athina Spiliopoulou¹¹, Andrew Liu²², Rajasree Menon^{14,23}, Carine M. Boustany-Kari²⁴, Hyun M. Kang^{23,25}, Robert G. Nelson²⁶, Ronald Klein²⁷, Barbara E. Klein²⁷, Kristine E. Lee²⁷, Xiaoyu Gao²⁸, Michael Mauer²⁹, Silvia Maestroni³⁰, Maria Luiza Caramori²⁹, Ian H. de Boer³¹, Rachel G. Miller³², Jingchuan Guo³², Andrew P. Boright¹², David Tregouet^{33,34}, Beata Gyorgy^{33,34}, Janet K. Snell-Bergeon³⁵, David M. Maahs³⁶, Shelley B. Bull³⁷, Angelo J. Canty³⁸, Colin N.A. Palmer³⁹, Lars Stechemesser⁴⁰, Bernhard Paulweber⁴⁰, Raimund Weitgasser^{40,41}, Jelizaveta Sokolovska⁴², Vita Rovite⁴³, Valdis Pirags^{42,44}, Edita Prakapiene⁴⁵, Lina Radzeviciene⁴⁶, Rasa Verkauskiene⁴⁶, Nicolae Mircea Panduru^{6,47}, Leif C. Groop^{20,48}, Mark I. McCarthy^{18,19,49,50}, Harvest F. Gu^{51,52}, Anna Möllsten⁵³, Henrik Falhammar^{54,55}, Kerstin Brismar^{54,55}, Finian Martin⁹, Peter Rossing^{21,56}, Tina Costacou³², Gianpaolo Zerbini³⁰, Michel Marre^{57,58,59,60}, Samy Hadjadj^{61,62,63}, Amy J. McKnight⁶⁴, Carol Forsblom^{5,6,7}, Gareth McKay⁶⁴, Catherine Godson⁹, A. Peter Maxwell⁶⁴, Matthias Kretzler^{14,23}, Katalin Susztak¹³, Helen M. Colhoun¹⁶, Andrzej Krolewski¹⁵, Andrew D. Paterson¹², Per-Henrik Groop^{5,6,7,65}, Stephen S. Rich⁸, Joel N. Hirschhorn^{2,3}, Jose C. Florez^{3,4,66,67} and SUMMIT Consortium, DCCT/EDIC Research Group, GENIE Consortium

JASN 30: 2000–2016, 2019



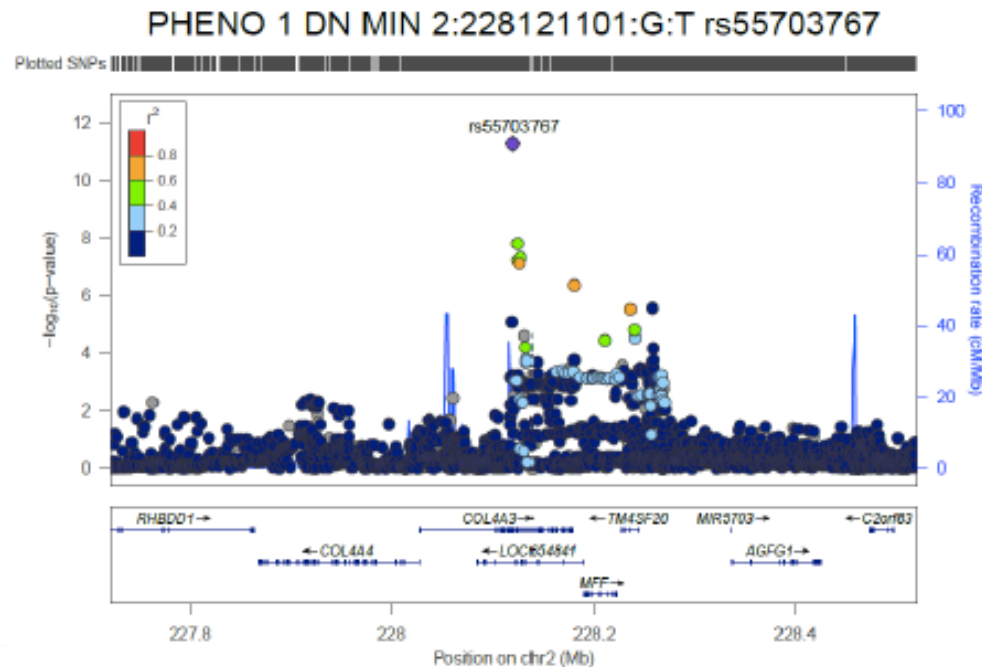
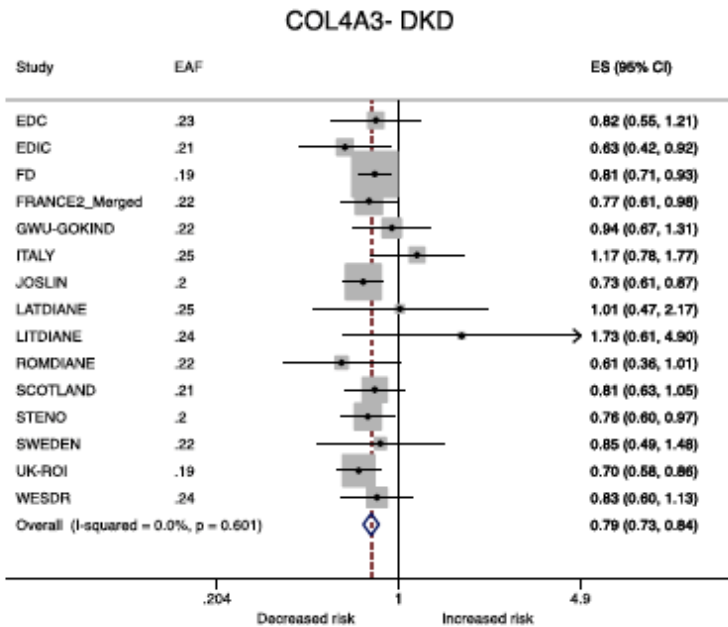
16-Jun-23

Results



Jc

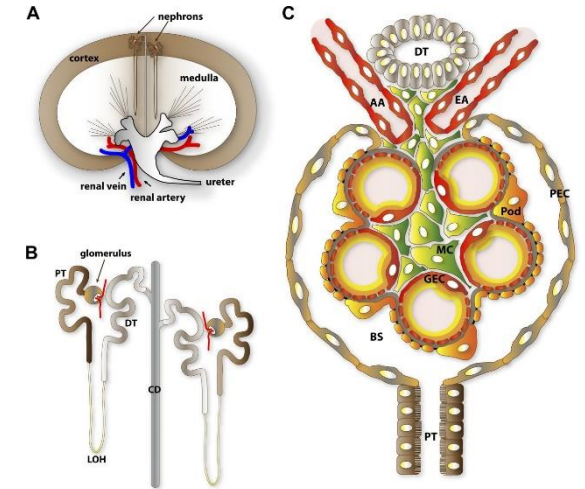
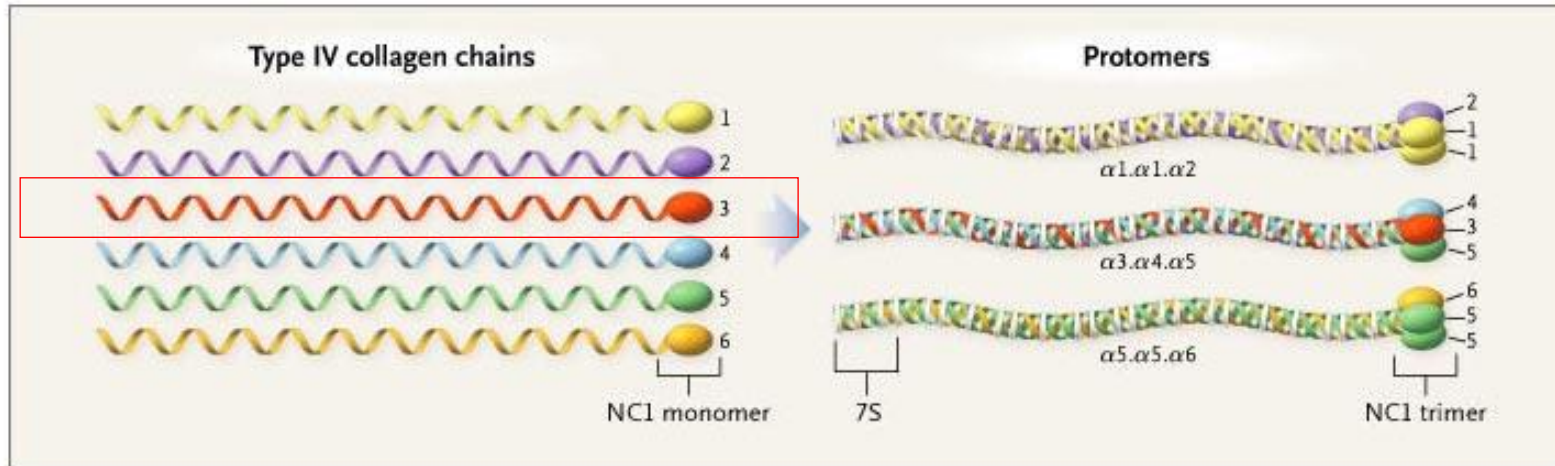
Associations point to causal genes: Asp326Tyr in *COL4A3* protects against DKD



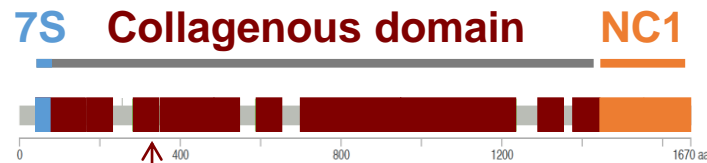
OR 0.79, $P = 5 \times 10^{-12}$
EAF 20.6% in Europeans
(5% in Africa, 11% in Asia)



COL4A3 encodes the α -3 subunit of type IV collagen, a major structural component of the GBM



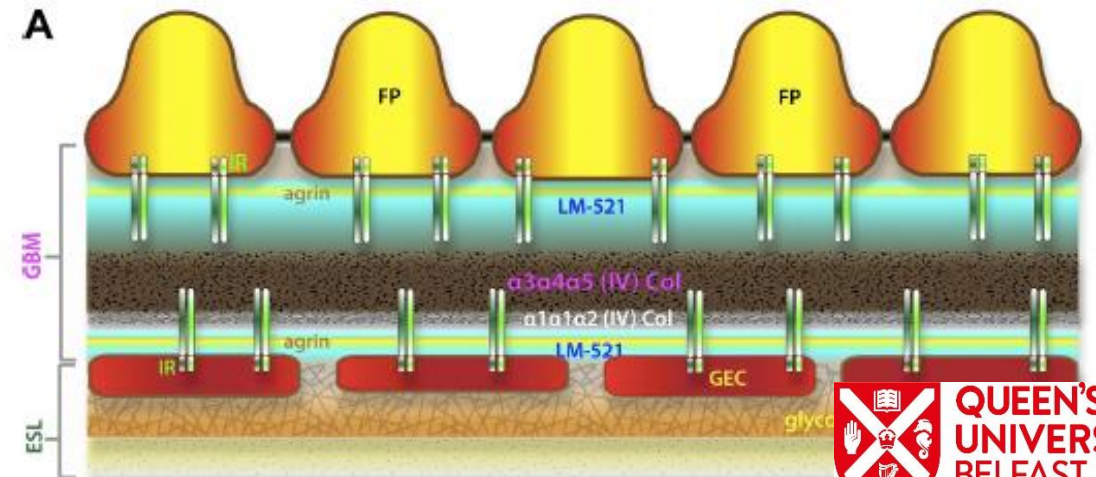
Hudson BG et al. *N Engl J Med* 2003;348:2543-2556



rs55703767
Asp326Tyr

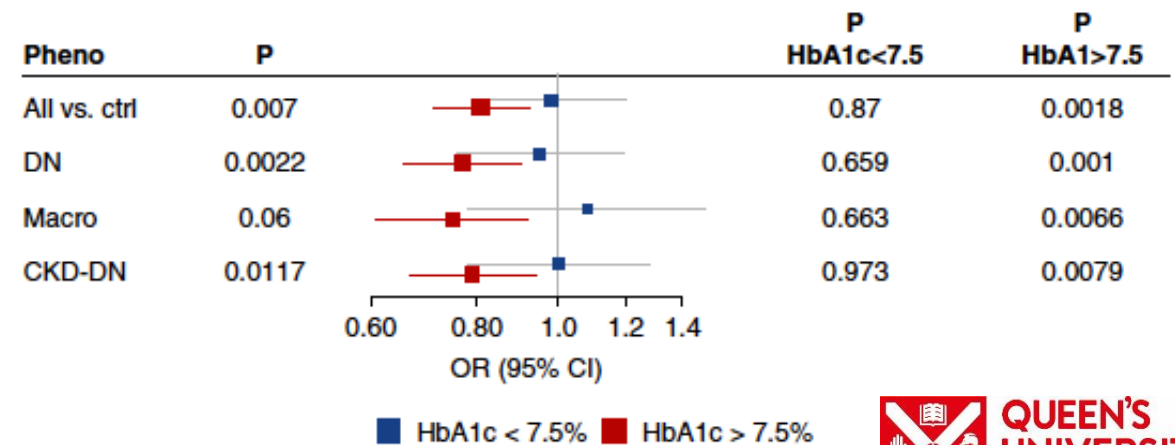
16-Jun-23

Rizaldy Scott & Susan Quaggin
J Cell Biol 2015;209:199-210

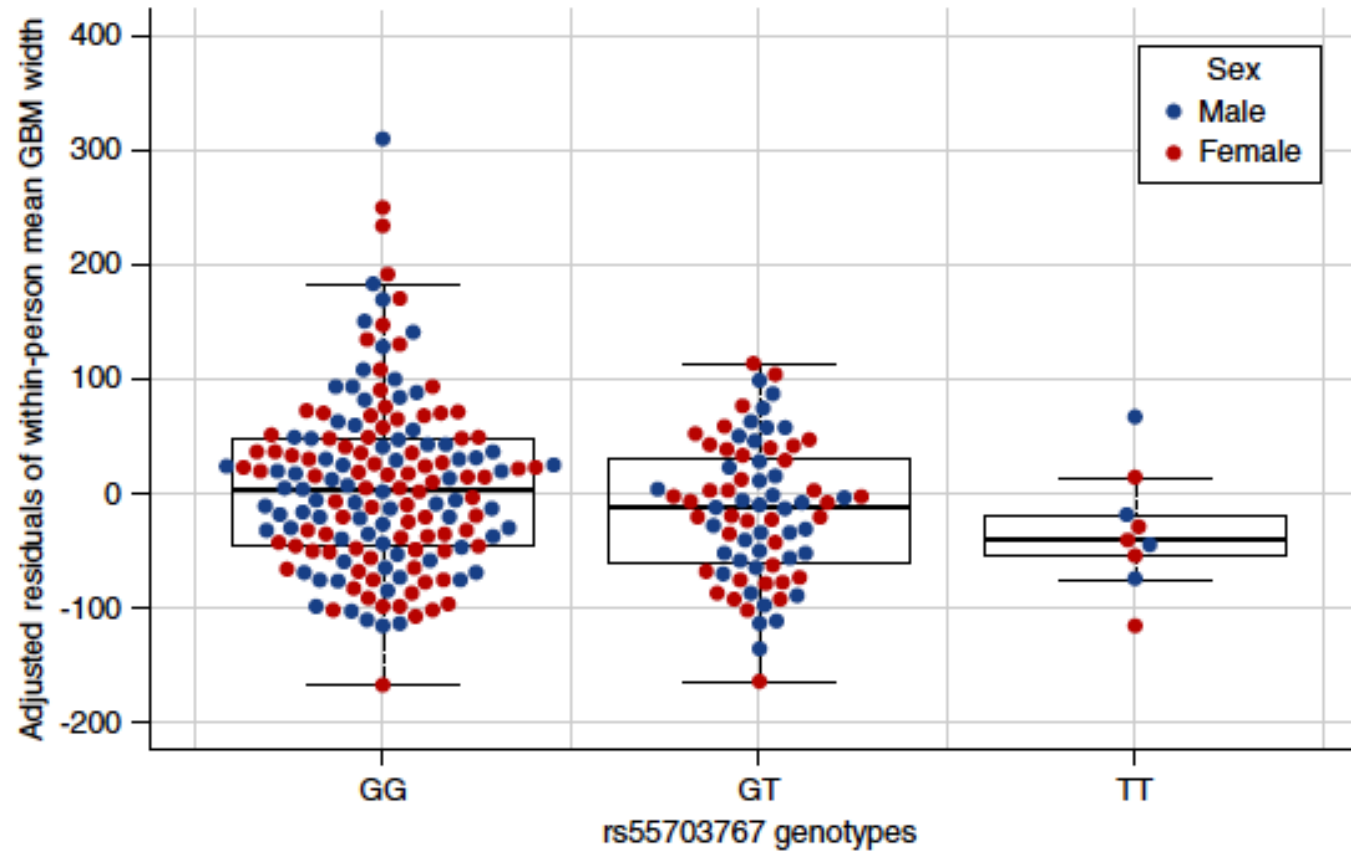


Interaction with hyperglycemia

- Stratification by HbA_{1c} in FinnDiane
- **Time-weighted mean HbA_{1c}**: 1-129 measurements (mean 19, IQR 9-32)
- Stratification at HbA_{1c} ≥7.5% or <7.5% (58 mmol/mol)
- **COL4A3**: Association when HbA_{1c} ≥7.5%, no association when HbA_{1c} <7.5% (protection in hyperglycemia)
- Protection stronger in conventional arm of DCCT/EDIC (higher HbA_{1c})
- → **Diabetic** nephropathy



Ultrastructural phenotype



- RASS study
- $N=253$
- Thicker GBM seen in DKD
- Protective T allele associated with thinner GBM (22.8 nm per allele, $P=0.006$)



With SUMMIT T2D: COL4A3 remains the top association signal

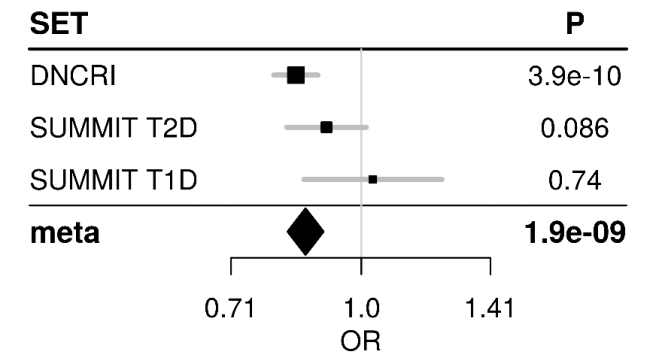
Table 1: GWAS meta-analysis result summary: loci with $p < 5 \times 10^{-8}$

Phenotype	CHR:POS	SNP	EA	NEA	EAF	OR (95% CI)	P-value	Dir	N (studies)	Genes
Novel locus										
CKD+DKD	5:166978230	rs72831309	A	G	0.039	2.08 (1.62 - 2.67)	9.8×10^{-9}	+++	8,570 (7)	<u>TENM2</u> *
Previous loci										
CKD	2:3745215	rs12615970	A	G	0.867	1.31 (1.20 - 1.44)	9.4×10^{-9}	+??	18,488 (13)	<u>ALLC</u> , <u>COLEC11</u>
All vs. Ctrl	2:228121101	rs55703767	T	G	0.207	0.86 (0.82 - 0.90)	1.9×10^{-9}	-+-	26,898 (24)	<u>COL4A3</u> *
CKD+DKD	2:228121101	rs55703767	T	G	0.210	0.81 (0.75 - 0.88)	4.7×10^{-8}	-+-	17,611 (17)	<u>COL4A3</u> *
Severe DKD	2:228121101	rs55703767	T	G	0.208	0.82 (0.77 - 0.87)	3.6×10^{-11}	---	21,898 (23)	<u>COL4A3</u> *
ESRD	3:926345	rs115061173	A	T	0.014	9.40 (4.22 - 20.93)	4.1×10^{-8}	+??	4,827 (3)	<u>LINC01266</u> , <u>CNTN6</u> *
Micro	3:11910635	rs142823282	A	G	0.983	0.15 (0.08 - 0.27)	8.3×10^{-10}	-??	6,076 (2)	<u>TAMM41</u>
ESRD vs. All	3:36566312	rs116216059	A	C	0.016	8.73 (4.13 - 18.45)	1.4×10^{-8}	+??	3,667 (2)	<u>STAC</u> - <u>DCLK3</u>
Severe DKD	4:71358776	rs191449639	A	T	0.005	32.42 (9.77 - 107.59)	1.3×10^{-8}	+??	7,768 (2)	<u>MUC7</u> , <u>AMTN</u>
Micro	7:99728546	rs77273076	T	C	0.008	9.16 (4.29 - 19.56)	1.1×10^{-8}	+??	7,500 (2)	<u>MBLAC1</u> - <u>ZNF3</u>
ESRD vs. macro	8:128100029	rs551191707	CA	C	0.122	1.69 (1.40 - 2.04)	4.4×10^{-8}	+??	3,634 (7)	<u>PRNCRI</u>
Micro	11:16937846	rs183937294	T	G	0.993	0.06 (0.02 - 0.16)	1.7×10^{-8}	-??	6,076 (2)	<u>PLEKHA7</u> *
CKD	18:1811108	rs185299109	T	C	0.007	20.75 (7.30 - 59.00)	1.3×10^{-8}	+??	7,223 (2)	<u>LINC00470</u>

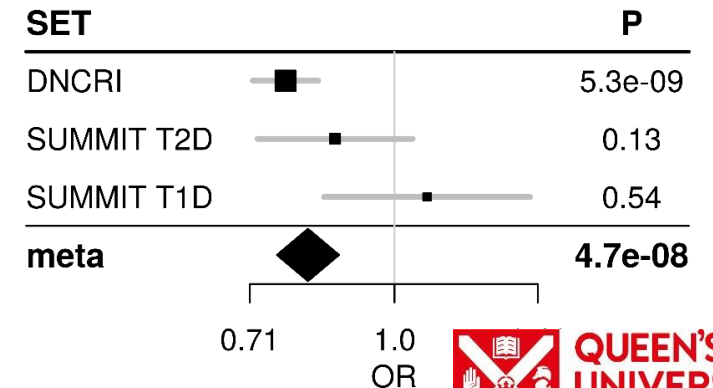
EA: Effect allele. NEA: Non-effect allele. EAF: Effect allele frequency. Dir: Direction of association in DNCRI (T1D), SUMMIT T2D, and in SUMMIT T1D, respectively. N (studies): Number of contributing individuals and (studies). Genes: closest gene(s). * indicates gene prioritized by PoPS. Genes underlying the lead SNP are underlined.

$n=26,785$

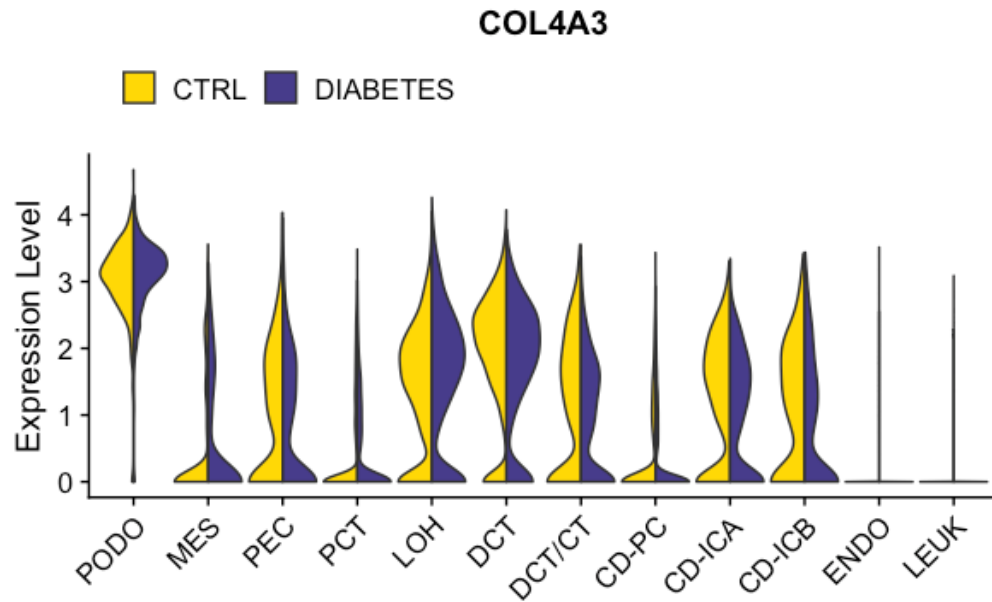
COL4A3 allvcntrl



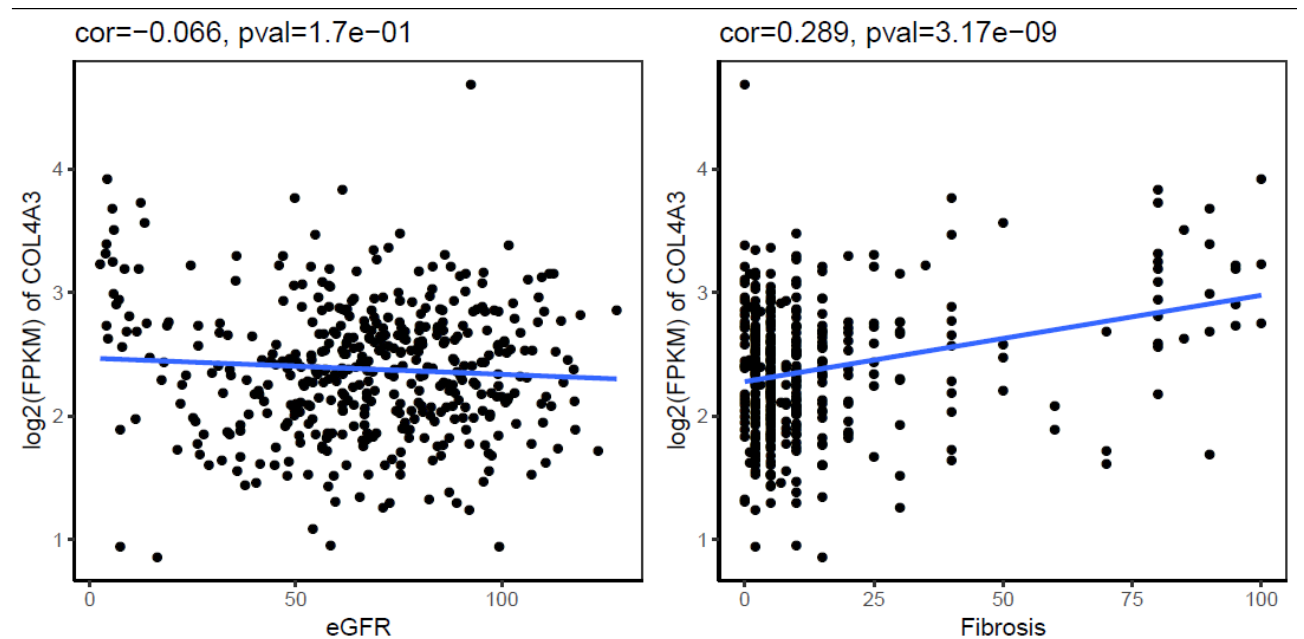
COL4A3 ckddn



COL4A3 is preferentially expressed in podocytes, and correlates with fibrosis



Human Diabetic Kidney data set (23,980 nuclei)
Wilson *et al.* *PNAS* 2019
<http://humphreyslab.com/SingleCell/displaycharts.php>



433 tubular and 335 glomerular nephrectomy samples with DKD and hypertensive CKD
Sandholm, Cole *et al.* (*Diabetologia* 2022)

GWAS Summary for DKD

- In the largest genetic study of T1DKD to date, we discovered 16 novel associations with DKD
- We found a **protective** missense variant in *COL4A3*
 - The effect was **consistent** across cohorts
 - The association appeared to be driven by **proteinuria**
 - The association was dependent on **glycemia**
 - The variant was associated with a **structural phenotype**
- Expression of *COL4A3* is correlated with fibrosis in human tubular samples
- There is no association of the index variant with levels of *COL4A3* expression
- Other signals lie in genes with biological plausibility (*COLLEC11*, *DDR1*)

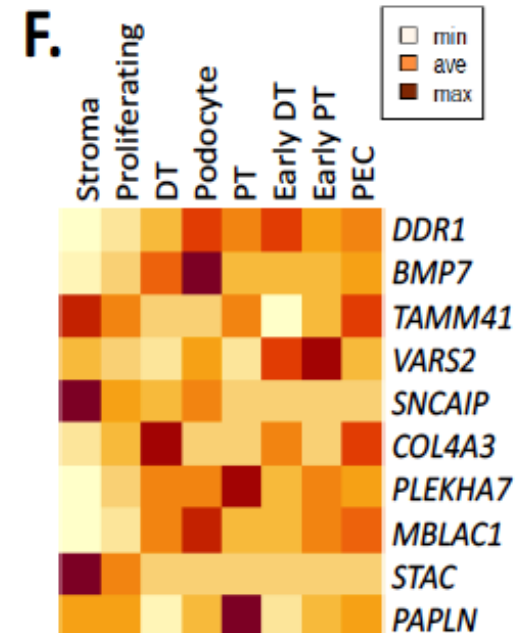
Current directions for DKD genetics studies – GENIE 3

- Combine T1DKD and T2DKD (>150K)

Cohort	Ethnicity	Investigator	N cases	N controls	N total
FIND	Multiethnic	Sudha Iyengar	4,295	2,189	6,484
ANDIS	European	Leif Groop	5,250	15,750	21,000
DIREVA	European	Leif Groop	1,375	4,125	5,500
SDR	European	Leif Groop	1,500	4,500	6,000
Singapore	Asian	Xueling Sim	1,665	1,778	3,443
BioVU	Multiethnic	Adriana Hung	2,264	8,199	10,463
MVP	Multiethnic	Adriana Hung	24,960	27,408	52,368
GoDARTS	European	Colin Palmer	1,728	2,576	4,304
Chennai	South Asia	V. Mohan	1,703	3,423	5,126
ARIC	Multiethnic	Rany Salem	406	1,584	1,990
CHS	Multiethnic	Rany Salem	181	533	714
MESA	Multiethnic	Rany Salem	614	891	1,505
FHS	European	Rany Salem	177	189	366
UKBiobank	Multiethnic	Rany Salem	5,507	28,939	34,346
HCHS-SOL	Latino	Rany Salem	653	1,890	2,543
JHS	African	Rany Salem	142	339	481
Total:			52,420	104,213	156,633

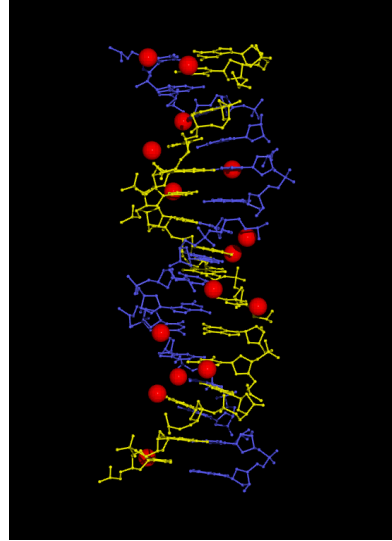
- Functional follow-up of findings

- Humanized mouse – J. Miner
- Collagen biomechanics – R. Lennon
- Organoids – M. Kretzler, J. Harder, P.-H. Groop
- scRNAseq – K. Susztak



V. Nair,
M. Kretzler,
J. Harder
(unpublished)

Epigenetics and Diabetic Kidney Disease



What is 'Epi' genetics?

- 'Epi': On top of
- Definition:
 - changes in organisms caused by modification of gene expression rather than alteration of the genetic code itself.

Genes are inherited - they only influence individual development if they are expressed.
Gene *expression* depends on a range of factors including those in the environment



Genes determine
How cells function



Epigenetic influences affect
how and when genes are
expressed

Epigenetics



The nature-nurture debate



The historical debate over the relative influence of genes and environment on characteristics



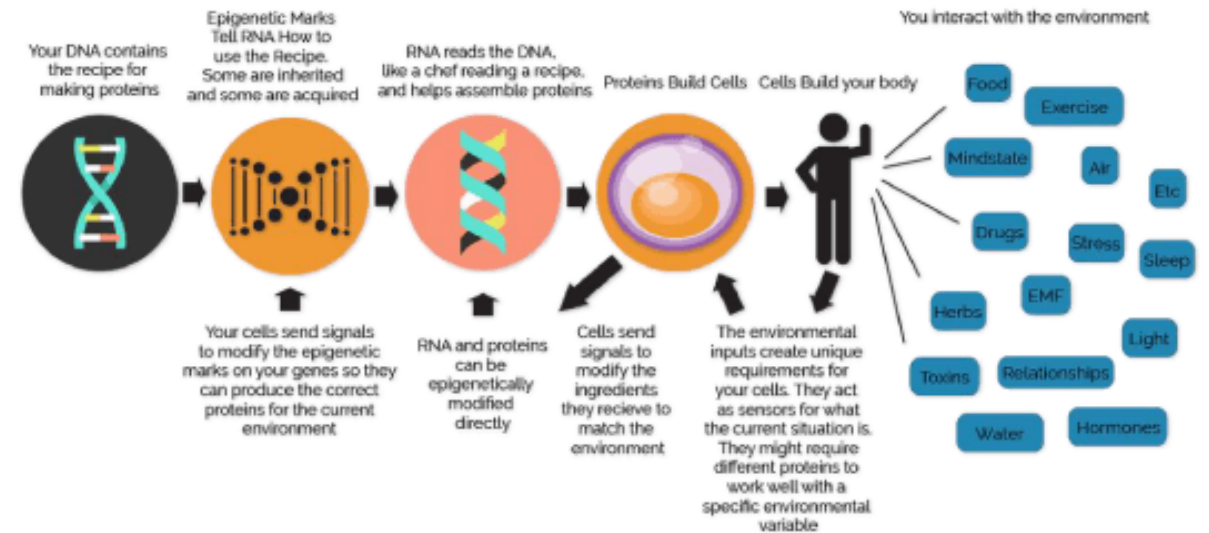
Nature = genetic influence



Nurture = environmental influence

What Is Epigenetics?

Epigenetics is the feedback loop between our genes and the environment. Our cells are constantly updating our genes on what they need to thrive. By inputting the correct environmental variables for your genes, food included, we can create health, longevity, and leave beneficial epigenetic marks for the next generation. Epigenetic coaching is all about optimizing the inputs to your system.



Epigenetics is the science of how genes and environment interact within an individual

Epigenetic mechanisms

Expression of a DNA depends on two main factors acting on the DNA tail:

Methylation

- Adding a methyl group prevents the genes being read and effectively ‘turns off’ the genes.

Acetylation

- Adding an acetyl group makes a DNA strand accessible and effectively turns it on.

Diet, exercise, pollution, sun exposure, smoking/toxins and age are the main factors effecting epigenetics



Epigenome-wide meta-analysis identifies DNA methylation biomarkers associated with diabetic kidney disease

nature communications



Article

<https://doi.org/10.1038/s41467-022-34963-6>

Epigenome-wide meta-analysis identifies DNA methylation biomarkers associated with diabetic kidney disease

Received: 27 April 2022

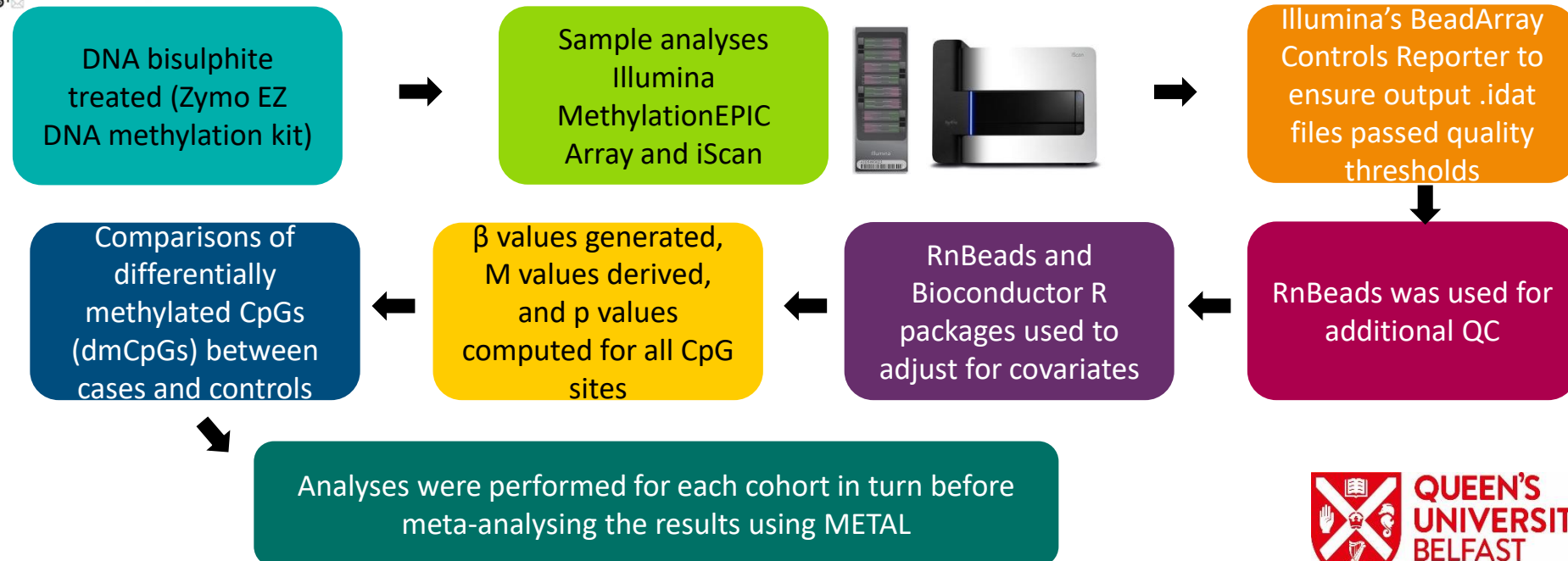
Accepted: 14 November 2022

Published online: 22 December 2022

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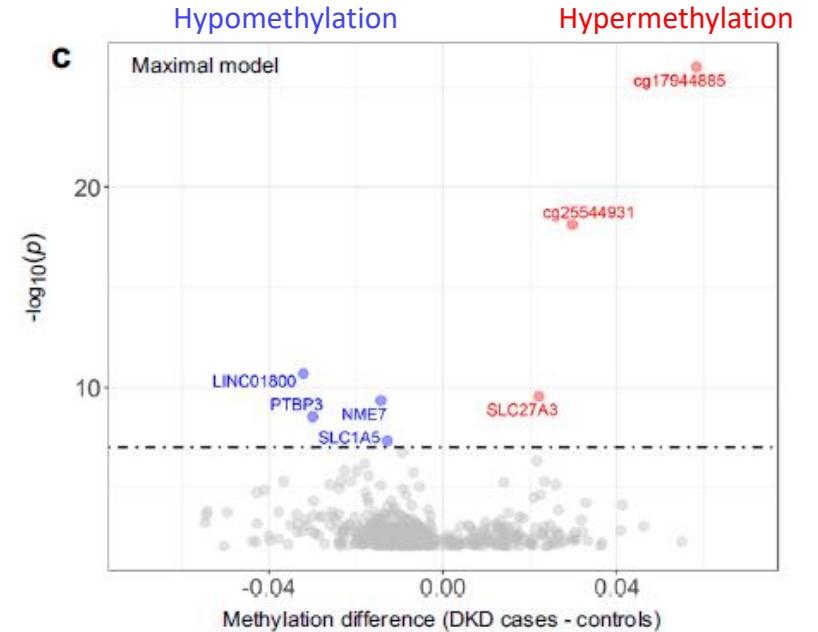
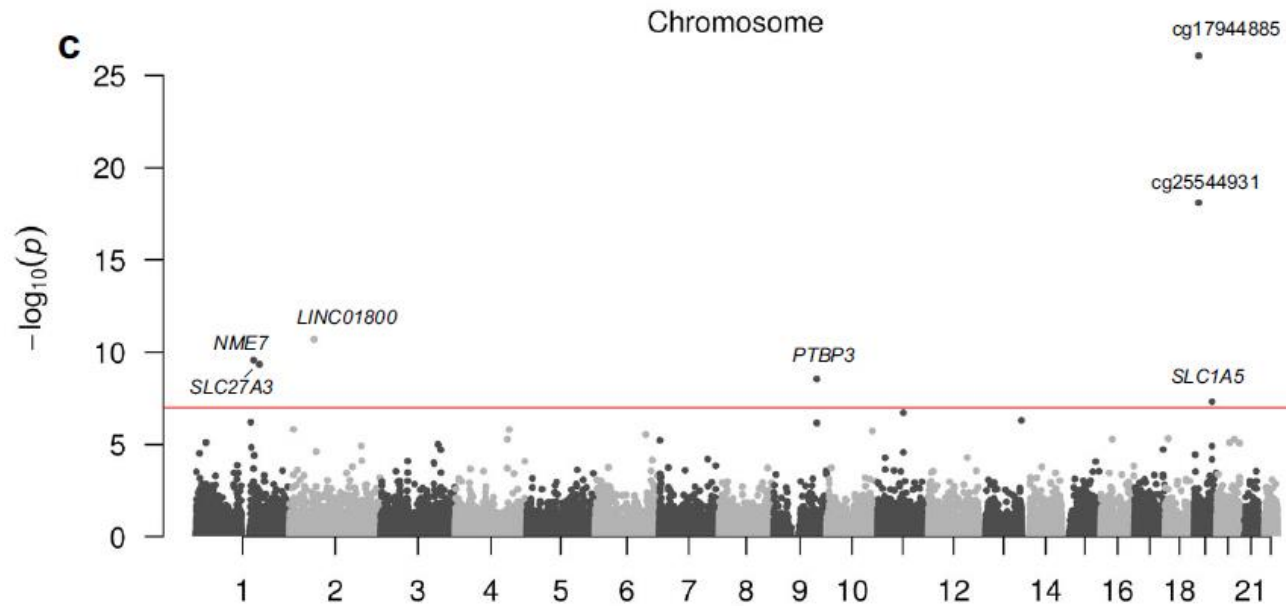
Laura J. Smyth^{1,20}, Emma H. Dahlström^{2,3,4,20}, Anna Syreeni^{2,3,4}, Katie Kerr¹, Jill Kilner¹, Ross Doyle⁵, Eoin Brennan⁵, Viji Nair⁶, Damian Fermin⁷, Robert G. Nelson⁸, Helen C. Looker⁸, Christopher Wooster¹, Darrell Andrews⁵, Kerry Anderson¹, Gareth J. McKay¹, Joanne B. Cole^{9,10}, Rany M. Salem¹¹, Peter J. Conlon¹², Matthias Kretzler¹³, Joel N. Hirschhorn^{9,14,15}, Denise Sadlier¹⁶, Catherine Godson⁵, Jose C. Florez^{9,10,17}, GENIE consortium¹, Carol Forsblom^{2,3,4}, Alexander P. Maxwell^{1,18}, Per-Henrik Groop^{2,3,4,19}, Niina Sandholm^{2,3,4} & Amy Jayne McKnight¹ ✉

- Evidence suggests epigenetic alterations, such as DNA methylation, in DKD.
- Blood-derived genome-wide DNA methylation assessed for association with DKD in 1304 well characterised individuals from T1D cohorts from United Kingdom/Ireland & Finland.
- DKD cases had persistent macroalbuminuria (AER > 300mg/ml) in urine; controls had normal range AER despite a long duration of T1D (≥15 yrs).



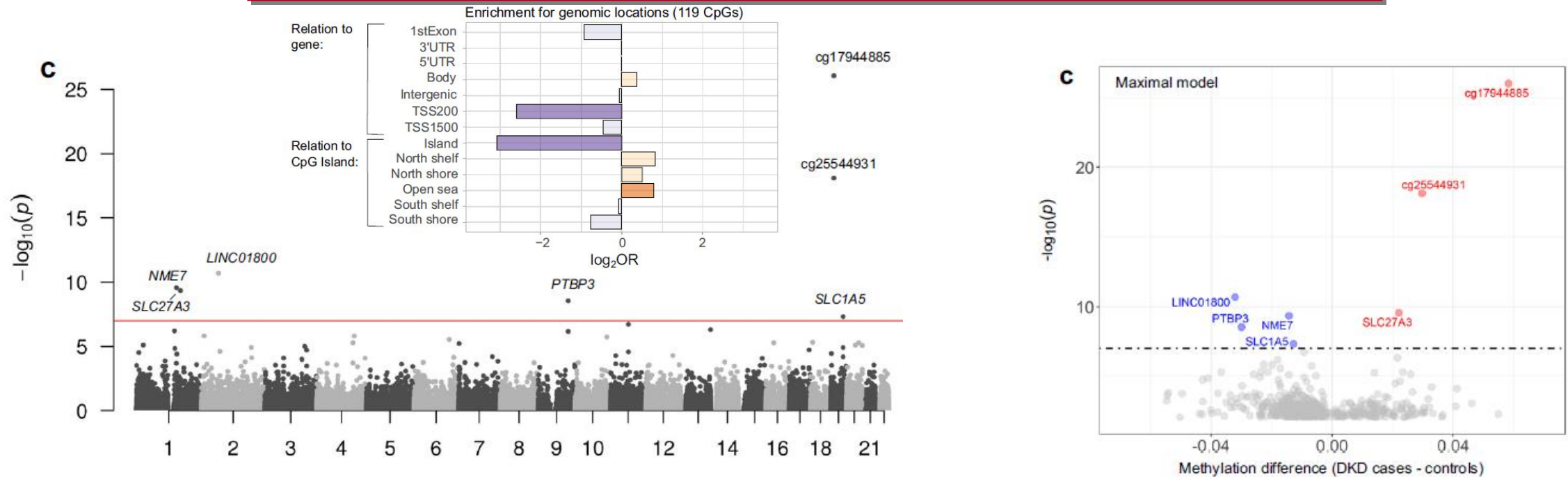
16-Jun-23

Epigenetic associations with diabetic kidney disease



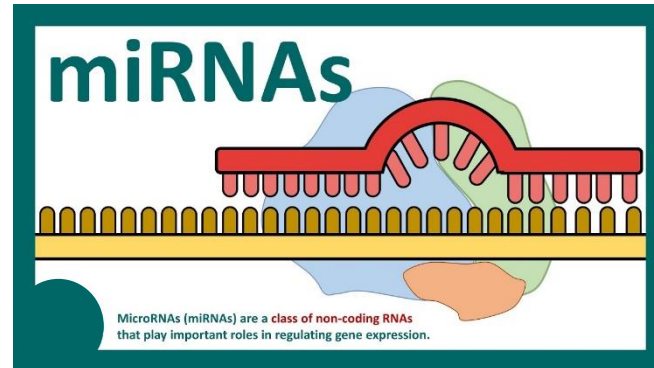
- Meta-analysis identified 32 differentially methylated CpGs with DKD in T1D, 18 of which were located within genes differentially expressed in kidneys or correlated with pathological traits in DKD.
- Follow-up data was available for 397 DKD cases to evaluate progression to kidney failure.
- Methylation at 21 of the 32 CpGs were shown to predict development of kidney failure, potentially identifying individuals at greater risk for DKD in T1D.

Epigenetic associations with diabetic kidney disease



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- Methylation at 21 of the 32 CpGs were shown to predict development of kidney failure, potentially identifying individuals at greater risk for DKD in T1D.
- Enrichment of DKD-associated CpGs in TSS.

miRNA's and Diabetic Kidney Disease



Differential urinary exosomal microRNA expression in DKD

Differential Expression of Urinary Exosomal MicroRNAs miR-21-5p and miR-30b-5p in Individuals with Diabetic Kidney Disease

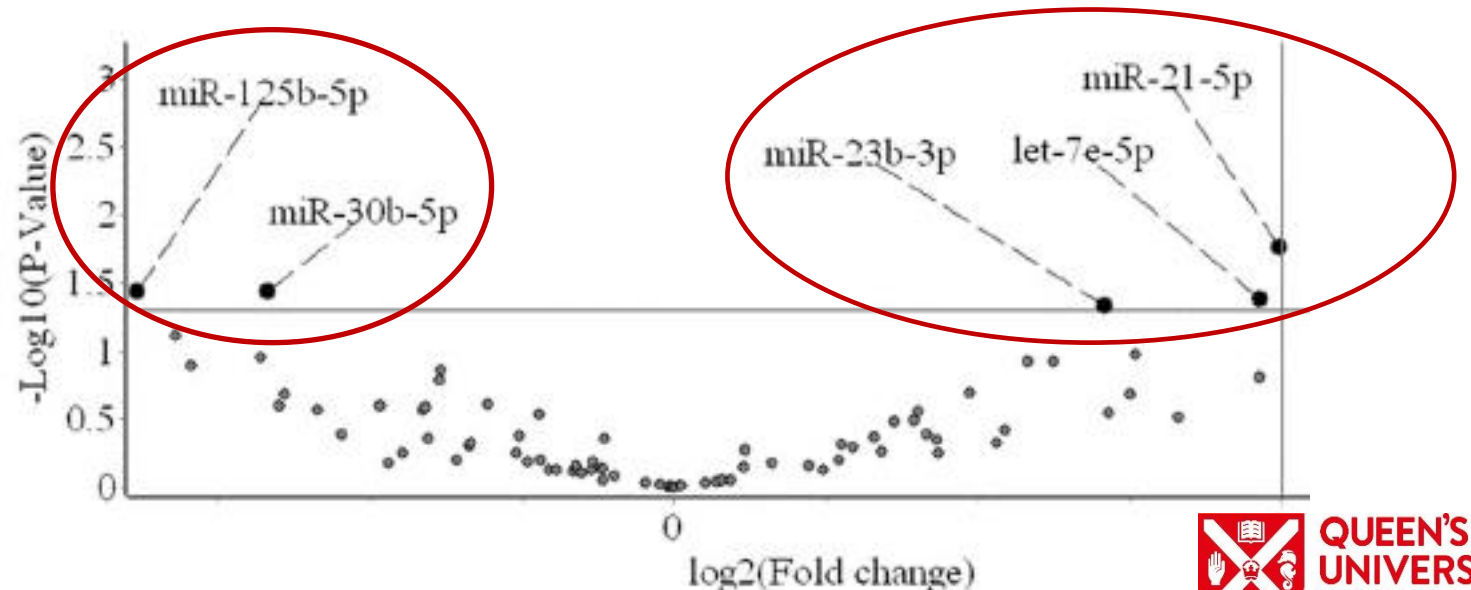
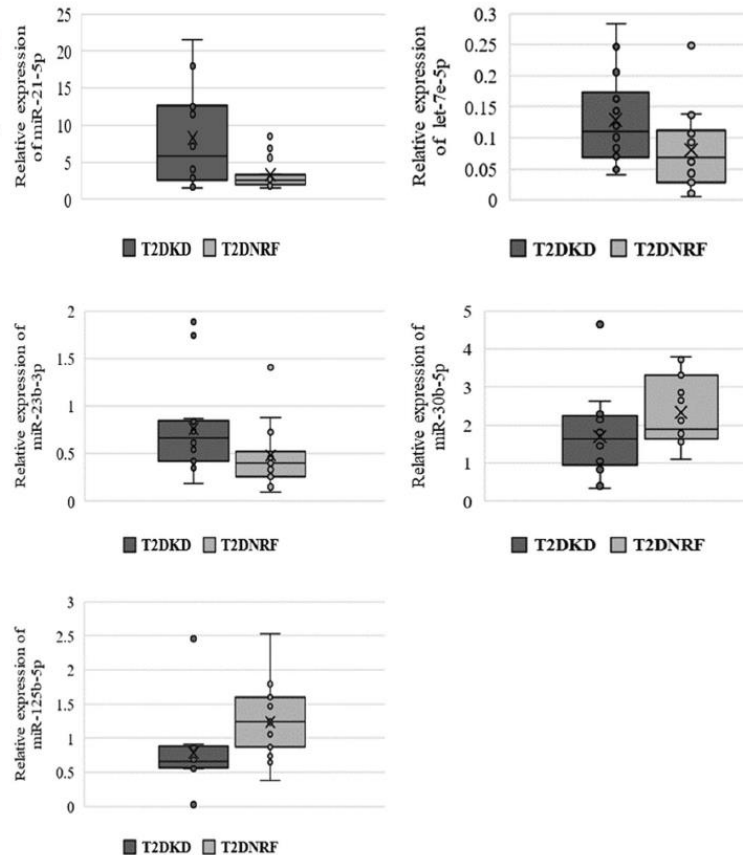
Jinnan Zang¹, Alexander P. Maxwell¹, David A. Simpson^{1,2} & Gareth J. McKay¹
SCIENTIFIC REPORTS | (2019) 9:10900 | <https://doi.org/10.1038/s41598-019-47504-x>

**SCIENTIFIC
REPORTS**
nature research

- Current biomarkers to identify DKD lack sensitivity to detect early kidney damage.
- miRNAs - short, non-coding, regulatory RNA molecules commonly found in urinary exosomes and differentially expressed as renal function declines.
- **Study 1** - We evaluated 87 urinary exosomal miRNA expression using **miRCURY PCR panel** (Qiagen) in a discovery cohort with T2DKD (n = 14) and age & sex-matched controls with T2D & normal renal function (T2DNRF; n = 15).

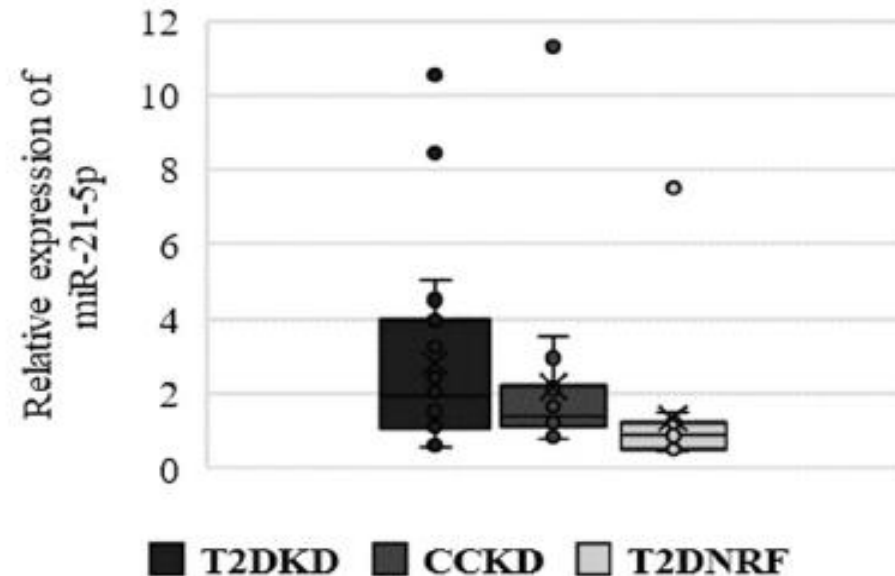
Differential urinary exosomal microRNA expression in DKD

- In discovery cohort, miR-21-5p, let-7e-5p & miR-23b-3p significantly upregulated in T2DKD.
- Conversely, miR-30b-5p & miR-125b-5p significantly lower in T2DKD.



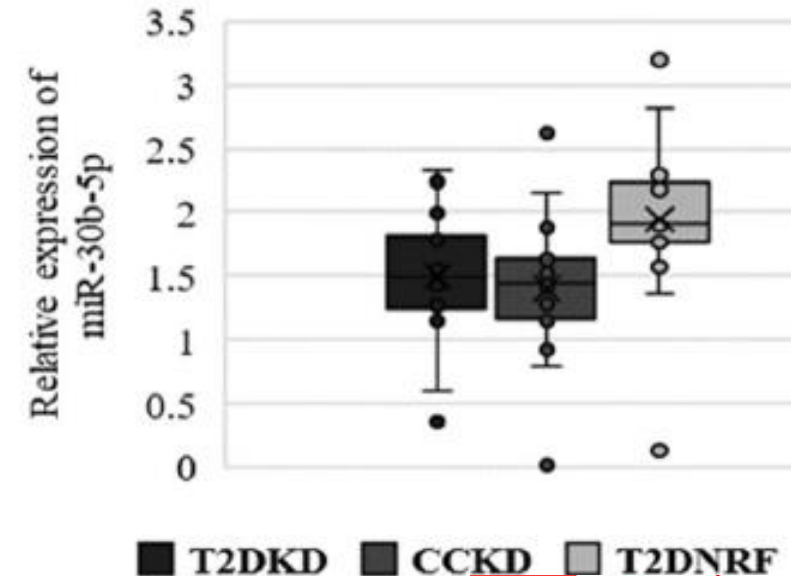
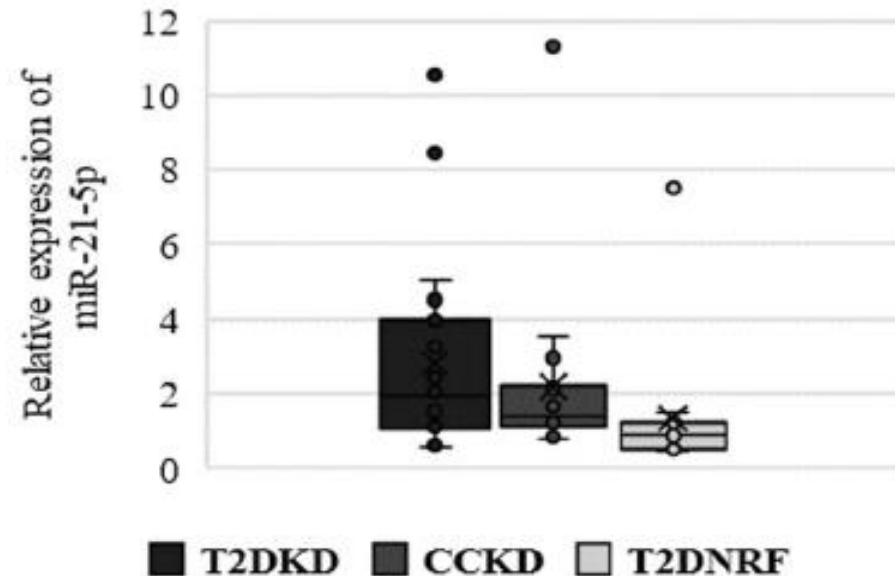
Differential urinary exosomal microRNA expression in DKD

- Independent miRNA validation performed in 2nd cohort with T2DKD (n = 22) & 2 control groups: T2DNRF (n = 15) & CKD controls without diabetes (CCKD; n = 18).
- Independent validation confirmed 2 miRs - up-regulation of miR-21-5p in T2DKD (2.13-fold, p = 0.006) & CCKD (1.73-fold, p = 0.024).

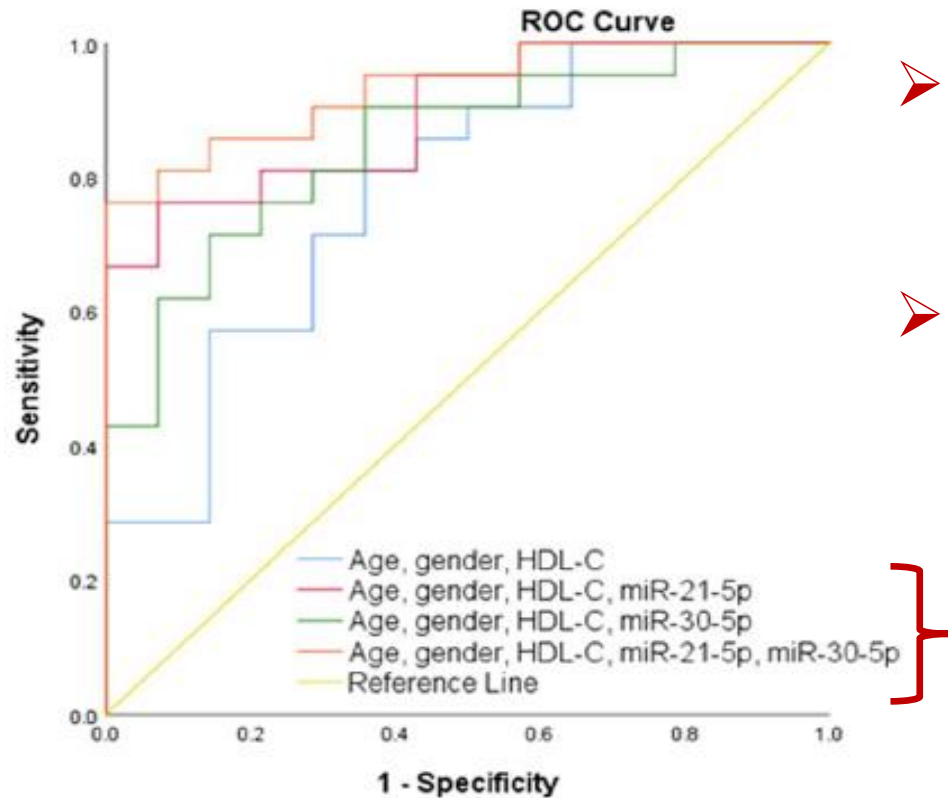


Differential urinary exosomal microRNA expression in DKD

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- Independent validation confirmed 2 miRs - up-regulation of miR-21-5p in T2DKD (2.13-fold, p = 0.006) & CCKD (1.73-fold, p = 0.024).
- In contrast, miR-30b-5p was downregulated in T2DKD (0.82-fold, p = 0.006) & CCKD (0.66-fold, p < 0.002).

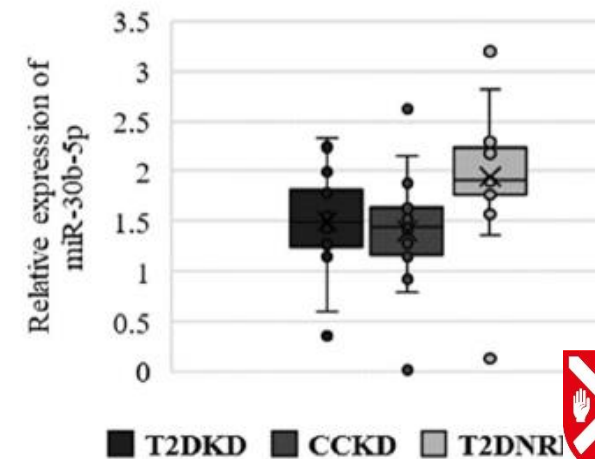
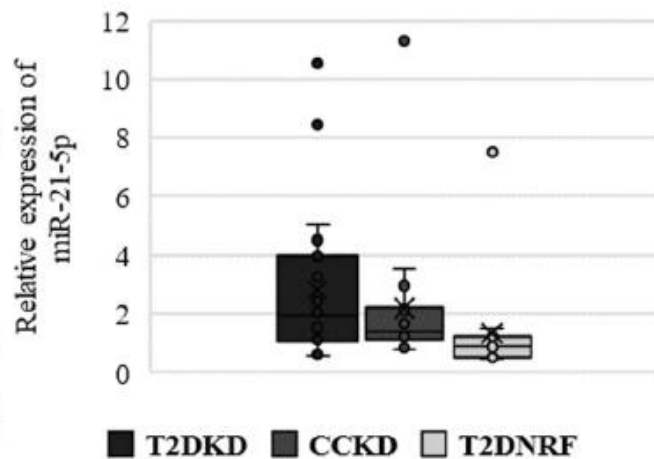


Differential urinary exosomal microRNA expression in DKD



- Identified differential expression of miR-21-5p & miR-30b-5p in individuals with either DKD or poor renal function.
- These miRNAs may represent potential biomarkers associated with the pathogenesis of renal dysfunction.

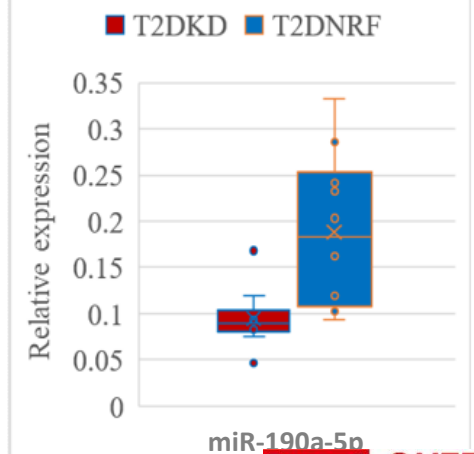
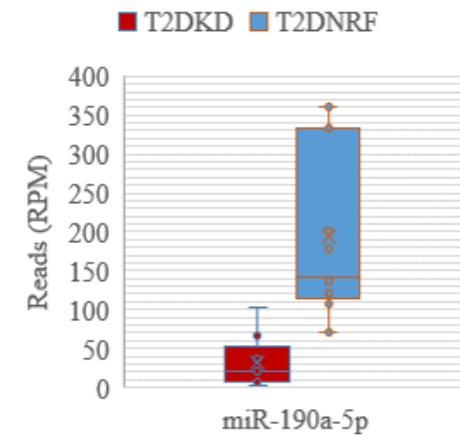
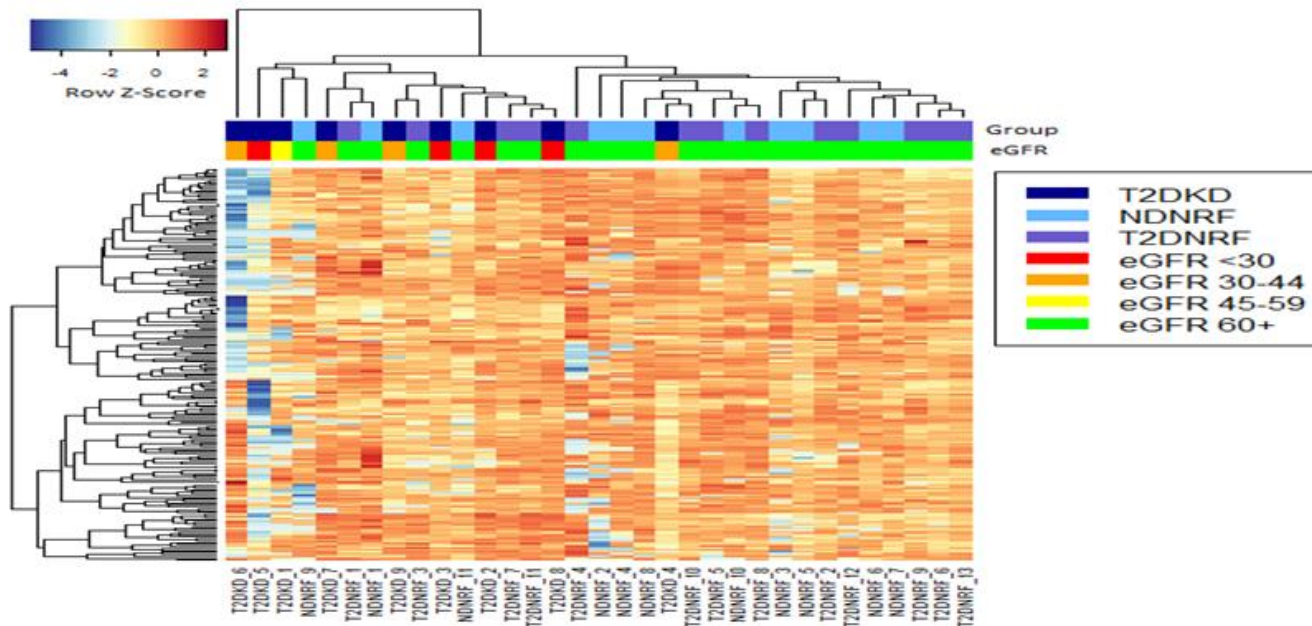
Covariates	AUC	95% CI	p value
Discovery Cohort			
Age, gender, HDL	0.560	0.338-0.782	0.593
Age, gender, HDL, miR-21-5p	0.830	0.673-0.986	0.004
Age, gender, HDL, miR-30b-5p	0.714	0.517-0.911	0.058
Age, gender, HDL, miR-21-5p, miR-30b-5p	0.813	0.652-0.974	0.006
Validation Cohort			
Age, gender, HDL	0.779	0.620-0.938	0.006
Age, gender, HDL, miR-21-5p	0.895	0.793-0.996	< 0.001
Age, gender, HDL, miR-30b-5p	0.850	0.724-0.977	0.001
Age, gender, HDL, miR-21-5p, miR-30b-5p	0.932	0.853-1.000	< 0.001



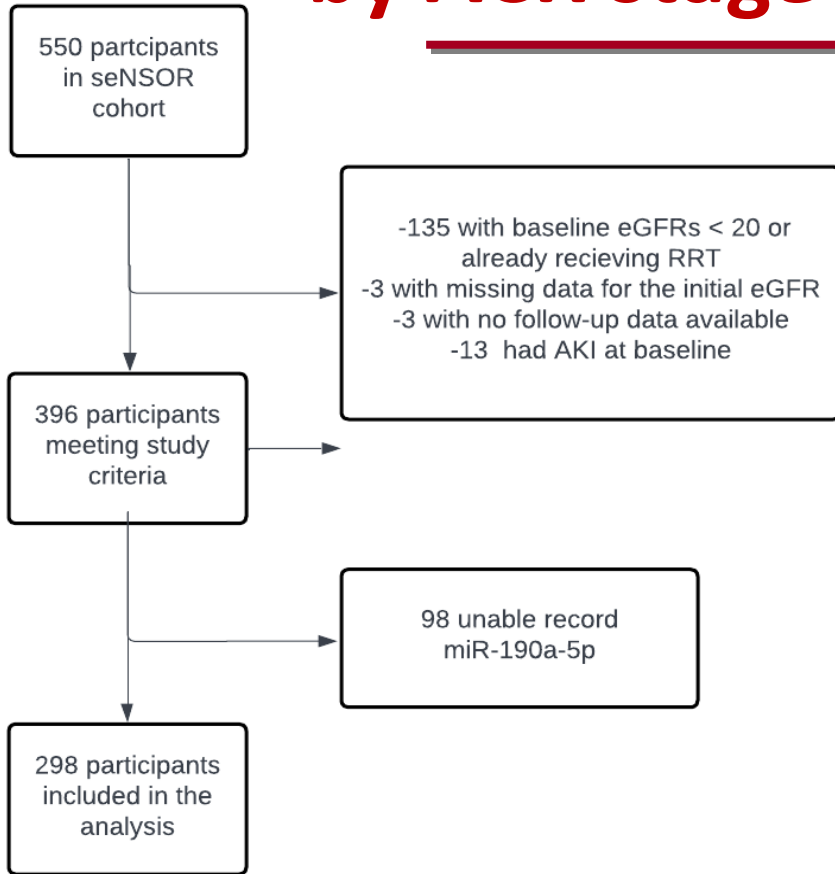
Reduced tubular miR-190a-5p a novel biomarker for stratification of patients with DKD

Study 2

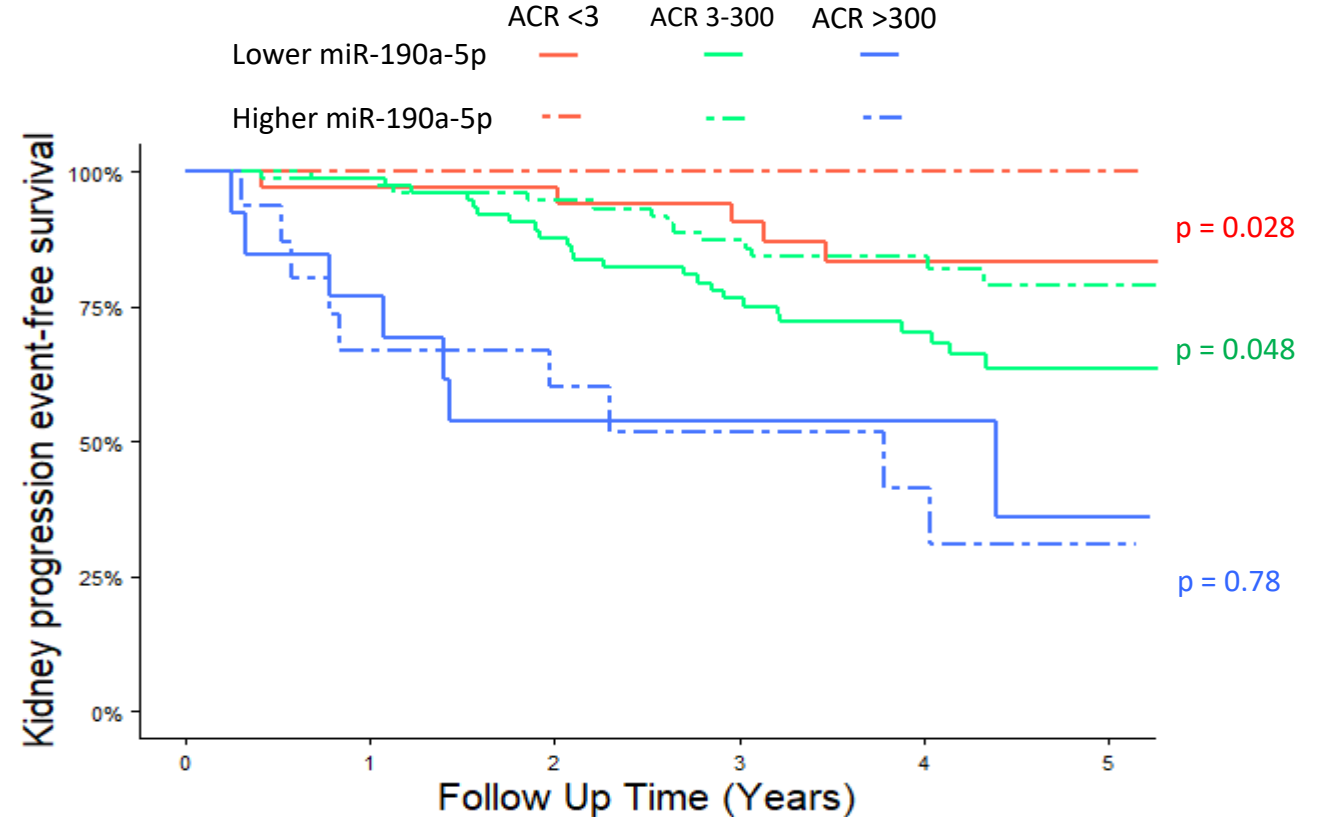
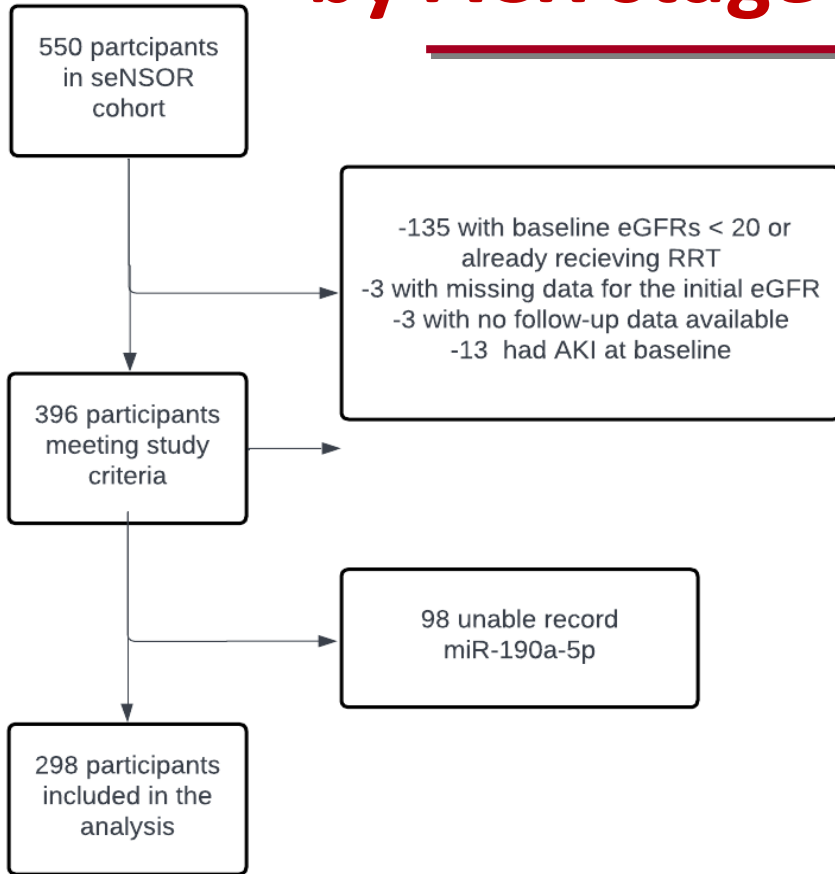
- Used miRNA NGS RNA - Seq to measure differential expression of plasma miRNAs in T2D patients with (n=9) or without (n=13) kidney disease and non-diabetic normal renal function (n=11) to identify novel miR biomarkers of kidney dysfunction.
- miR-190a-3p significant lower expression in T2DKD in discovery and validation cohorts



Validation of differential expression of mir-190a-5p in DKD by ACR stage in the prospective seNSOR cohort

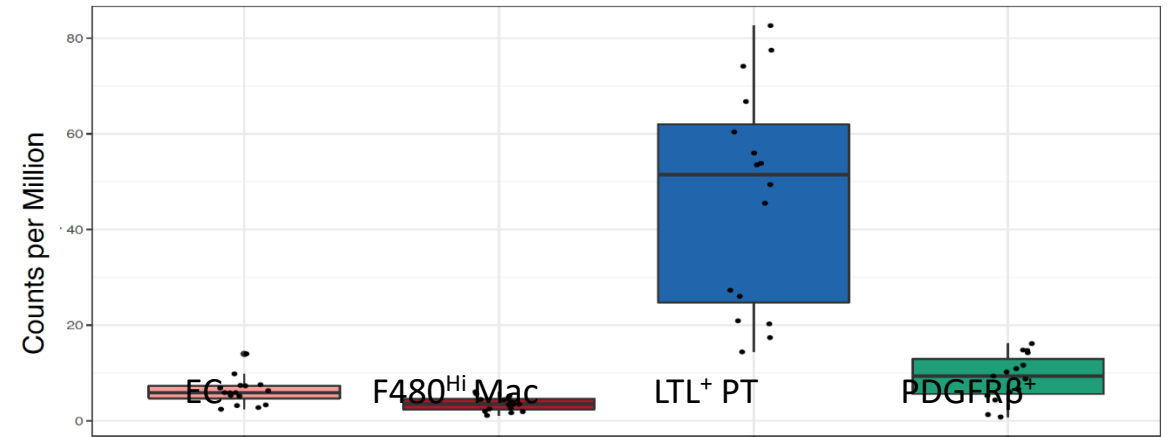
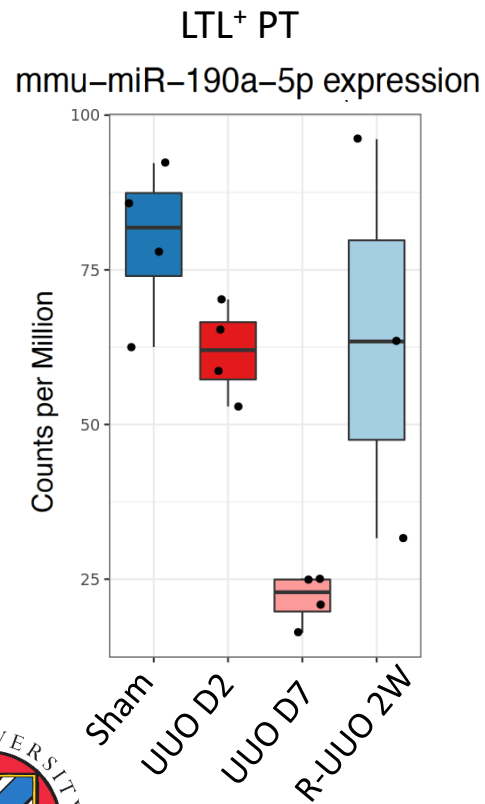
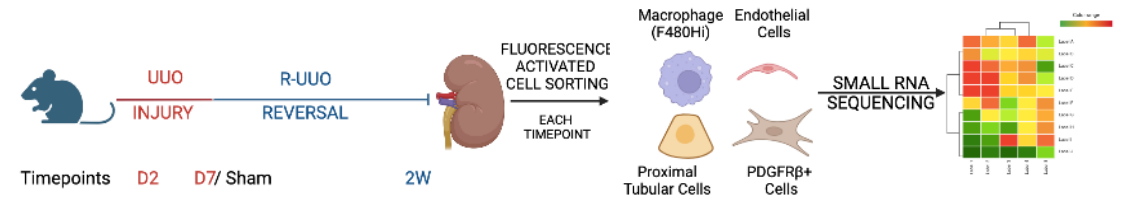


Validation of differential expression of miR-190a-5p in DKD by ACR stage in the prospective seNSOR cohort



- Positive correlation between miR-190a-5p and eGFR ($\rho = 0.12$, $p=0.04$) & inversely with age ($\rho = -0.12$, $p=0.04$).
- MiR-190a-5p levels below the median predict CKD progression in those with minimal and moderate albuminuria (ACR < 300mg/mmol respectively) but not in those with severe albuminuria (ACR > 300mg/mmol).

Murine model miR-190a-5p expression



- miR-190a expression in renal cell types in the reversible unilateral ureter obstruction mouse model shows enrichment in proximal tubule cells
- miR-190a expression falls significantly following injury before increasing again during the repair phase.
- Low serum miR-190a may predict declining renal function in patients with low or moderate proteinuria, independent of existing risk factors.

Oculomics in Diabetic Kidney Disease



What and why measure the eye?

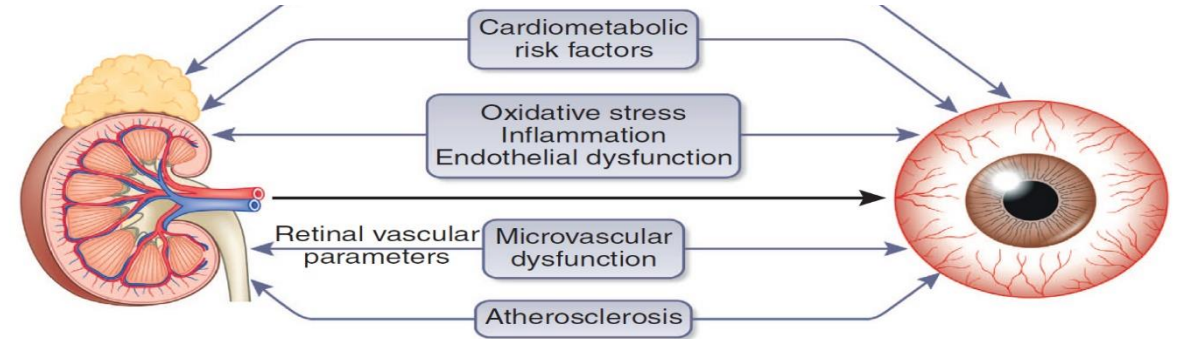
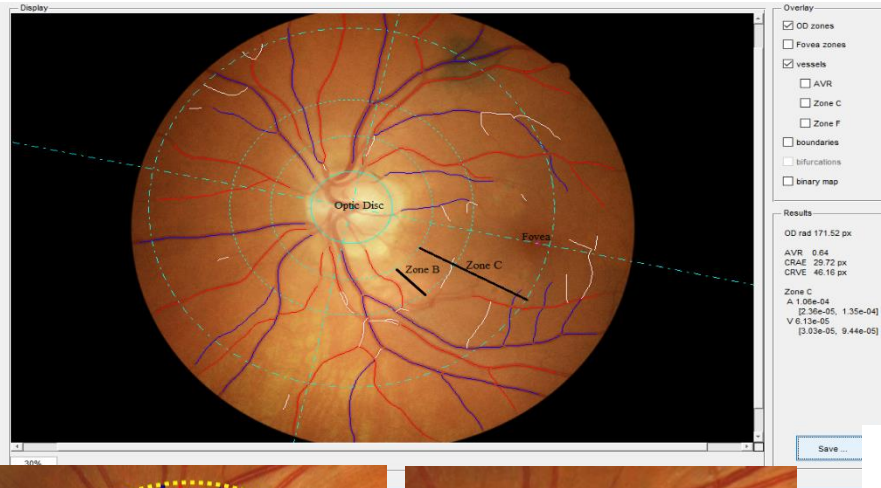


Figure 1 | Common pathogenetic mechanisms underlying renal and retinal diseases.
Wong et al., *Kidney International*, 2013

- Leading causes of visual impairment associated with CKD (Zhu et. al. 2020) - Cataract, AMD, Glaucoma and any retinopathy
- Previous reported associations between renal function & retinal parameters.
- Associations may reflect systemic vascular effects and /or renovascular damage.
- Retina - a non-invasive, opportunistic microvascular imaging.
- Significant advances in retinal imaging technology and analysis applications.
- Renal biopsy and vascular imaging procedures are invasive

Northern Ireland Cohort of Longitudinal Ageing (NICOLA)

- Stratified random sample of ~8500 men/women aged 50+ in Northern Ireland
- Computer Assisted Personal Interview (CAPI) plus self-completed questionnaires- social, behavioural, economic and environmental aspects of ageing including diet, mental health, physical activity
- Longitudinal : Repeated measures every 2-4 years
- Health Assessment –
 - Cardiovascular, cognitive and respiratory health
 - Visual function including retinal imaging
 - Anthropometry
 - Physical function
 - Biological samples

O'Neill et al. *BMC Nephrology* (2020) 21:382
<https://doi.org/10.1186/s12882-020-02031-0>

BMC Nephrology

RESEARCH ARTICLE

Open Access

Association of retinal venular tortuosity with impaired renal function in the Northern Ireland Cohort for the Longitudinal Study of Ageing



R. A. O'Neill, A. P. Maxwell, F. Kee, I. Young, B. McGuinness, R. E. Hogg and McKay GJ*

CKD status	SeCr	Adjusted (Min)			Adjusted (Full)		
Retinal parameter	OR	95% CI	P Value	OR	95% CI	P Value	
Arteriolar Calibre (PX)	1.23	0.25, 5.96	0.80	1.85	0.30, 11.50	0.51	
Venular Calibre (PX)	0.97	0.18, 5.43	0.98	0.59	0.08, 4.24	0.60	
Arteriolar Fractal dimension	1.11	0.92, 1.34	0.28	1.16	0.94, 1.42	0.16	
Venular Fractal dimension	0.84	0.70, 1.00	0.05	0.86	0.70, 1.06	0.17	
Arteriolar Tortuosity	1.05	0.88, 1.25	0.60	1.03	0.85, 1.24	0.77	
Venular Tortuosity	1.34	1.13, 1.58	<0.01	1.29	1.08, 1.54	<0.01	

Min adj: age, sex; Full: age, sex, diabetes, smoking, education, BMI, antihypertensive medication, MABP, triglycerides, HDL & LDL.

Associations of retinal thinning and poorer renal function

Paterson *et al. BMC Nephrology* (2020) 21:37
<https://doi.org/10.1186/s12882-019-1679-1>

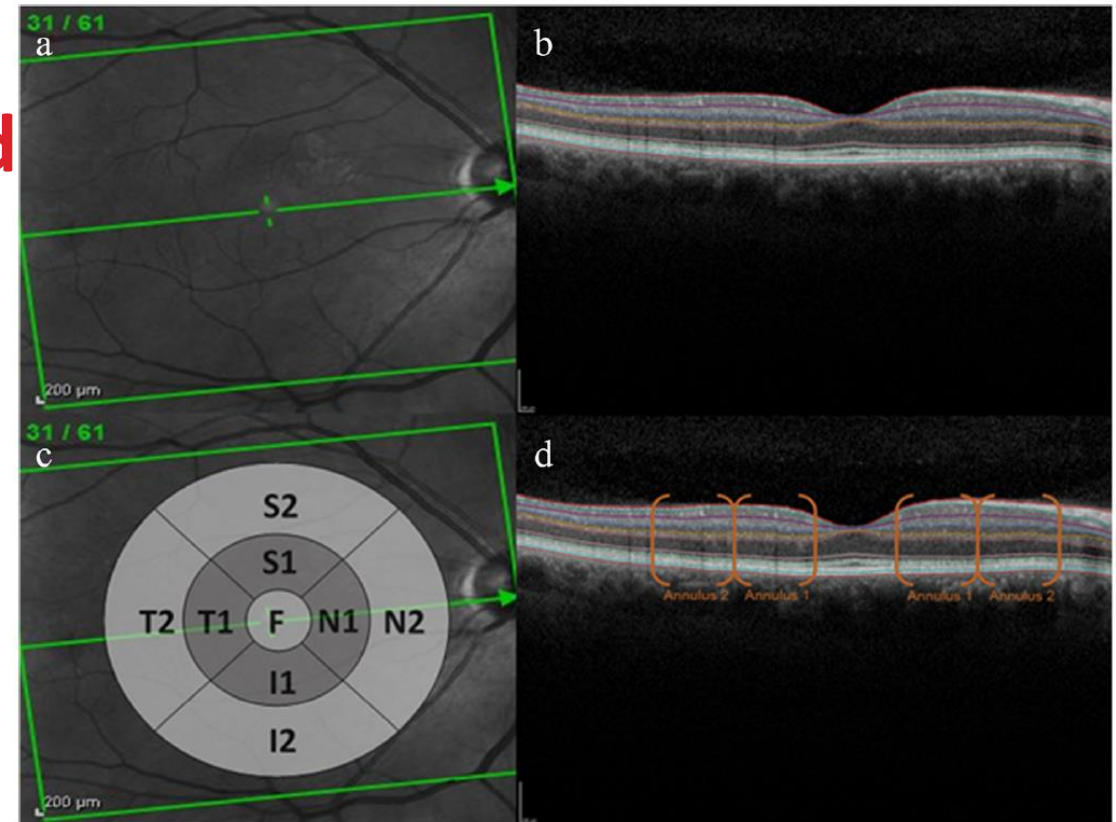
BMC Nephrology

RESEARCH ARTICLE

Open Access

Association of reduced inner retinal thicknesses with chronic kidney disease

Euan N. Paterson¹, Meera L. Ravindran¹, Kayleigh Griffiths¹, Claire A. Le Velly¹, Chris C. Cardwell¹, Rachel V. McCarter¹, Patrick Nicol¹, Jay K. Chhablani², Mohammed Abdul Rasheed³, Kiran Kumar Vupparaboina³, Thomas J. MacGillivray⁴, Mark Harbinson⁵, Alexander P. Maxwell¹, Ruth E. Hogg¹ and Gareth J. McKay^{1*} 



- Participants (n = 241) - mean age: 65 yrs; mean eGFR: 67 ml/min/1.73m². 39% had diabetes.
- Reduced retinal thickness, in particular a thinner inner retinal layer and microvascular complexity associated with CKD stage 4–5 independent of other important risk factors (age, MABP, diabetes, LDL, BMI & sex).
- Associations were limited to layers of the retina supplied by the retinal microvasculature.
- No associations with early stage CKD, but distinct association with CKD st 4–5.

Sparser retinal microvascular network and reduced renal function

Paterson et al. *BMC Nephrology* (2021) 22:72
<https://doi.org/10.1186/s12882-021-02273-6>

BMC Nephrology

RESEARCH ARTICLE

Open Access

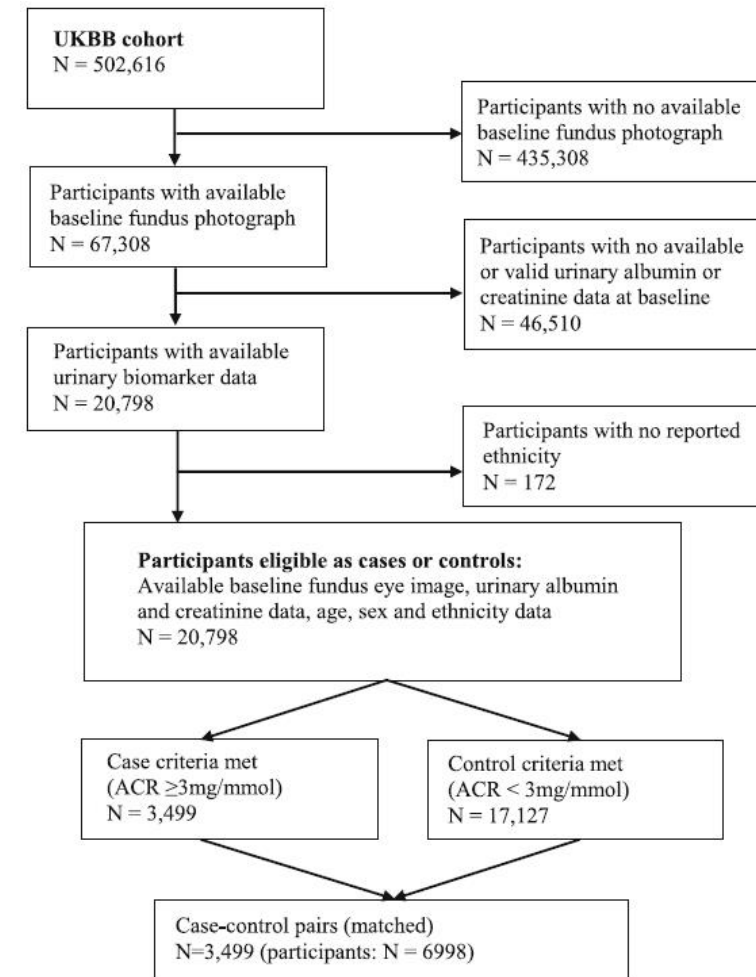


Investigation of associations between retinal microvascular parameters and albuminuria in UK Biobank: a cross-sectional case-control study

Euan N. Paterson¹, Chris Cardwell¹, Thomas J. MacGillivray², Emanuele Trucco³, Alexander S. Doney⁴, Paul Foster⁵, Alexander P. Maxwell¹, Gareth J. McKay^{1*} and on behalf of The UK Biobank Eye and Vision Consortium

Table 5 Associations between retinal microvascular parameters (Z scores) vs CKD status based on ACR > 3 mg/mmol and/or eGFR < 60 ml/min/1.73m² (SCr & SCys)

	Model 1 OR (95% CI)	p	Model 2 OR (95% CI)	p	Model 3 OR (95% CI)	p
CRAE	1.01 (0.90, 1.13)	0.88	0.97 (0.86, 1.09)	0.59	0.98 (0.87, 1.11)	0.79
CRVE	1.03 (0.93, 1.14)	0.59	0.99 (0.89, 1.10)	0.90	1.01 (0.90, 1.12)	0.92
AVR	0.92 (0.81, 1.05)	0.21	0.93 (0.80, 1.06)	0.27	0.93 (0.80, 1.07)	0.29
FDa	1.22 (1.07, 1.39)	0.003	1.20 (1.05, 1.37)	0.01	1.22 (1.06, 1.39)	0.01
FDv	1.27 (1.08, 1.49)	0.004	1.28 (1.08, 1.52)	0.01	1.26 (1.06, 1.50)	0.01
Torta	0.99 (0.90, 1.09)	0.85	1.00 (0.90, 1.10)	0.92	0.98 (0.89, 1.08)	0.69
Tortv	0.97 (0.88, 1.07)	0.52	0.98 (0.88, 1.08)	0.65	0.98 (0.88, 1.08)	0.63



➤ Case-control study of healthy versus unhealthy ACR suggests reduced retinal microvascular FD, i.e. sparser retinal microvascular networks, associate with albuminuria & lower eGFR.

Not all associations are positive!

OPEN Retinal microvascular parameters are not associated with reduced renal function in a study of individuals with type 2 diabetes

Received: 21 November 2017
Accepted: 22 February 2018
Published online: 02 March 2018

Gareth J. McKay¹, Euan N. Paterson¹, Alexander P. Maxwell¹, Christopher C. Cardwell¹, Ruixuan Wang², Stephen Hogg², Thomas J. MacGillivray³, Emanuele Trucco² & Alexander S. Doney¹

Baseline Variables	Sample n = 1068	Progressors n = 335	Non-progressors n = 570	p
Age, yrs (SD)	63.0 (7.6)	62.5 (7.7)	63.1 (7.8)	0.21
Gender, female (%)	521 (49)	168 (50)	281 (49)	0.81
eGFR, ml/min/1.73 m ² (SD)	94.0 (17.2)	98.6 (21.3)	91.3 (14.3)	<0.001
SBP, mmHg (SD)	138 (13)	139 (14)	137 (13)	0.08
DBP, mmHg (SD)	77 (8)	76 (9)	77(8)	0.63
HbA _{1c} , % (SD); mmol/mol	7.41 (1.38); 57.5	7.51 (1.36); 58.6	7.40 (1.41); 57.4	0.25
Diabetic retinopathy present, n (%)	244 (23)	82 (25)	118 (21)	0.19
Mean follow-up period, yrs (SD)	3.01 (0.35)	3.02 (0.35)	3.02 (0.34)	0.98

Retinal microvascular parameter (per unit increase)	Unadjusted β eGFR (95% CI)	p	Adjusted β eGFR (95% CI)	p
Calibre				
Central retinal arteriolar equivalent	-0.47 (-0.87, -0.07)	0.02	-0.38 (-0.80, 0.05)	0.08
Central retinal venular equivalent	-0.30 (-0.60, 0.00)	0.05	-0.27 (-0.58, 0.05)	0.10
Arteriovenous ratio	-3.32 (-21.81, 15.16)	0.72	-0.52 (-19.64, 18.60)	0.96
Fractal dimension				
Arteriolar	-18.41 (-36.92, 0.10)	0.05	-17.64 (-36.71, 1.44)	0.07
Venular	-3.74 (-22.79, 15.31)	0.70	-3.46 (-23.36, 16.43)	0.73
No. of First branches in zone C				
Arteriolar	-0.67 (-1.63, 0.30)	0.17	-0.50 (-1.50, 0.49)	0.32
Venular	0.66 (-0.43, 1.75)	0.24	0.82 (-0.31, 1.95)	0.15
Tortuosity				
^a Arteriolar	-0.01 (-2.66, 2.65)	1.00	-0.01 (-2.75, 2.73)	0.99
^a Venular	-3.20 (-6.73, 0.32)	0.08	-2.22 (-5.86, 1.43)	0.23

clinical investigation

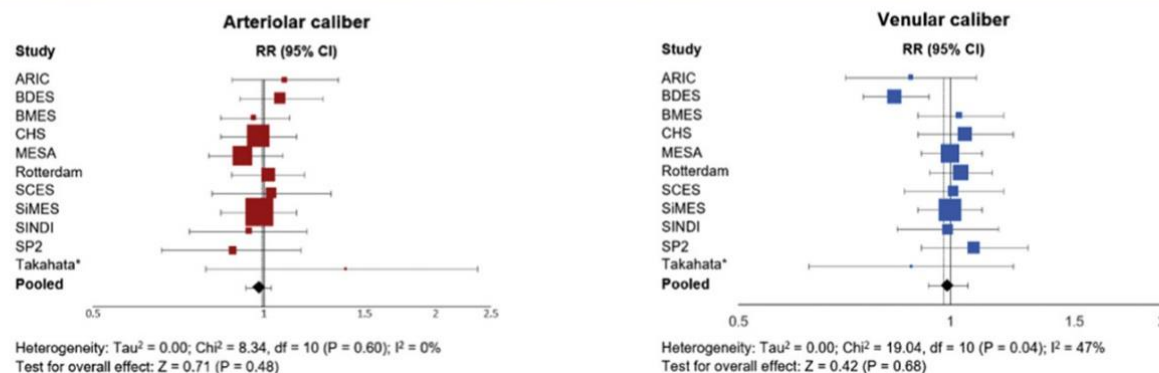
www.kidney-international.org

A systematic review and participant-level meta-analysis found little association of retinal microvascular caliber with reduced kidney function

Weng Kit Lye^{1,12}, Euan Paterson^{2,12}, Christopher C. Patterson², Alexander P. Maxwell², Riswana Banu Binte Mohammed Abdul¹, E. Shyong Tai¹, Ching Yu Cheng¹, Takamasa Kayama³, Hidetoshi Yamashita⁴, Mark Sarnak⁵, Michael Shlipak⁶, Kunihiro Matsushita⁷, Unal Mutlu^{8,9}, Mohammad A. Ikram⁸, Caroline Klaver^{8,9}, Annette Kifley¹⁰, Paul Mitchell¹⁰, Chelsea Myers¹¹, Barbara E. Klein¹¹, Ronald Klein¹¹, Tien Y. Wong¹, Charumathi Sabanayagam^{1,13} and Gareth J. McKay^{2,13}

A systematic review and participant-level meta-analysis found little association of retinal microvascular caliber with reduced kidney function.

Relative Risk (RR) for estimated glomerular filtration rate <60ml/min/1.73m² per 20 μ m increase in caliber. 44,803 individuals across 11 cohorts

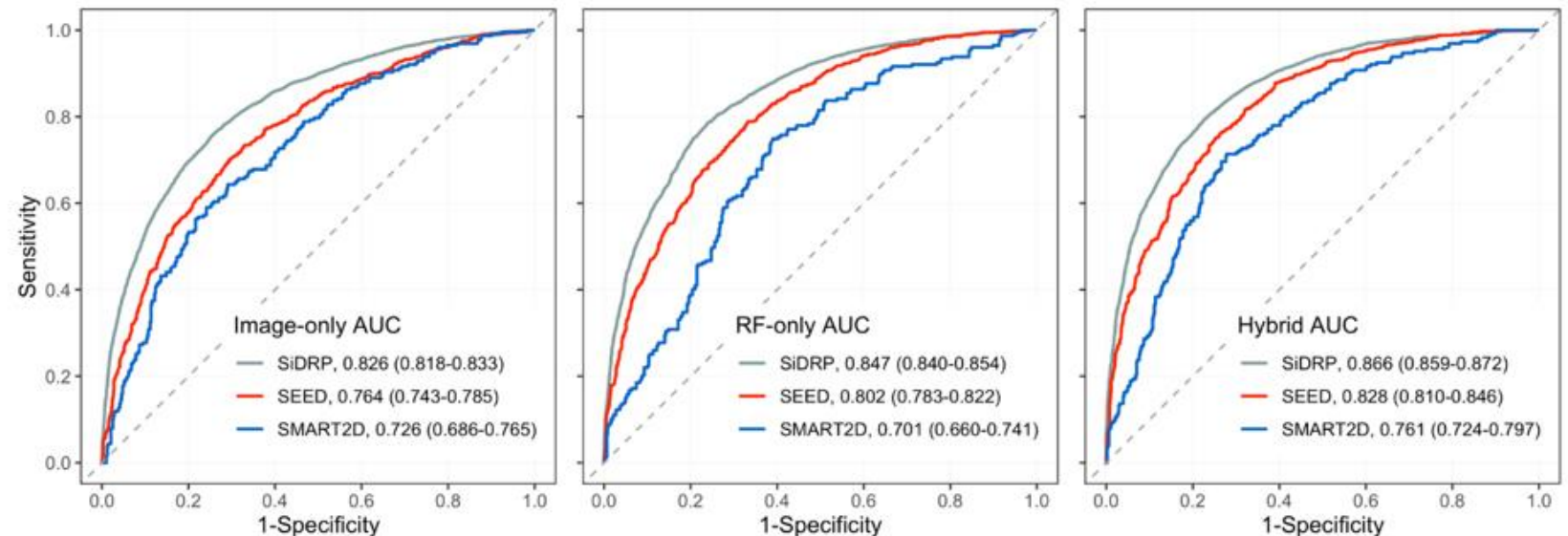


Pooled analyses adjusted for age, center, gender, ethnicity (if multi-ethnic cohort), education, current smoking, diabetes, hypertension, BMI, total cholesterol, and fellow vessel central retinal arteriolar equivalent.

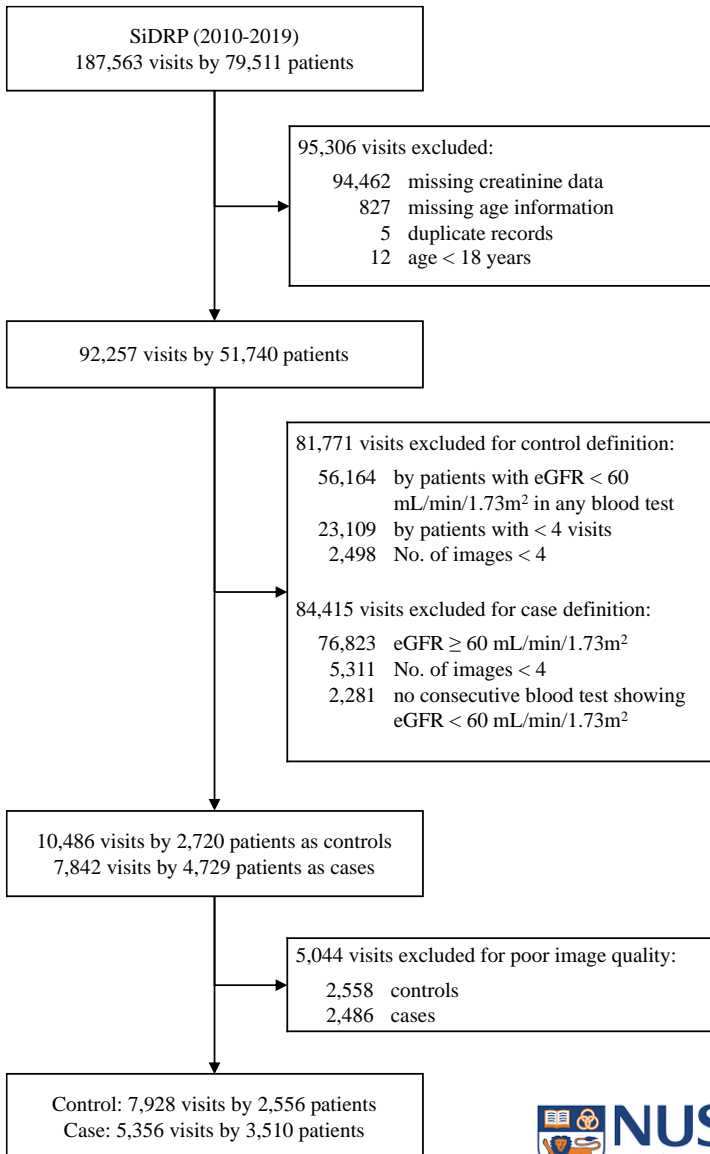
CONCLUSION:
No evidence of association between retinal microvascular caliber and CKD stages 3-5 was found in pooled analyses of 11 population-based cohorts.

Deep learning algorithms to detect DKD from retinal photographs in multi-ethnic populations with diabetes

- The CNN DLA was trained on 26568 retinal images from 6066 SiDRP participants.
- The DLA models were based on ResNet18 architecture with pre-training using a large-scale diabetic retinopathy dataset (Kaggle) to improve generalizability.
- Prediction compared with the ground truth label and revised via back-propagation.
- 5-fold cross-validation to evaluate model performance.



Models showed reasonable performance, fairing well in internal validation (AUC image-only = 0.826), with moderate performance in external validation (AUC image-only = 0.764 in SEED; 0.726 in SMART2D). In particular, the image-only model performed comparably well in all datasets compared to the RF-only model.



Greater retinal microvascular complexity & dementia risk in diabetes

Retinal vascular measures from diabetes retinal screening photographs and risk of incident dementia in type 2 diabetes: A GoDARTS study

Alexander S. F. Doney^{1*}, Aditya Nar¹, Yu Huang¹, Emanuele Trucco², Tom MacGillivray³, Peter Connelly⁴, Graham P. Leese¹, Gareth J. McKay⁵ and on behalf of the INSPIRED consortium

 **frontiers** | Frontiers in Digital Health

TYPE Original Research
PUBLISHED 31 August 2022
DOI 10.3389/fdgth.2022.945276

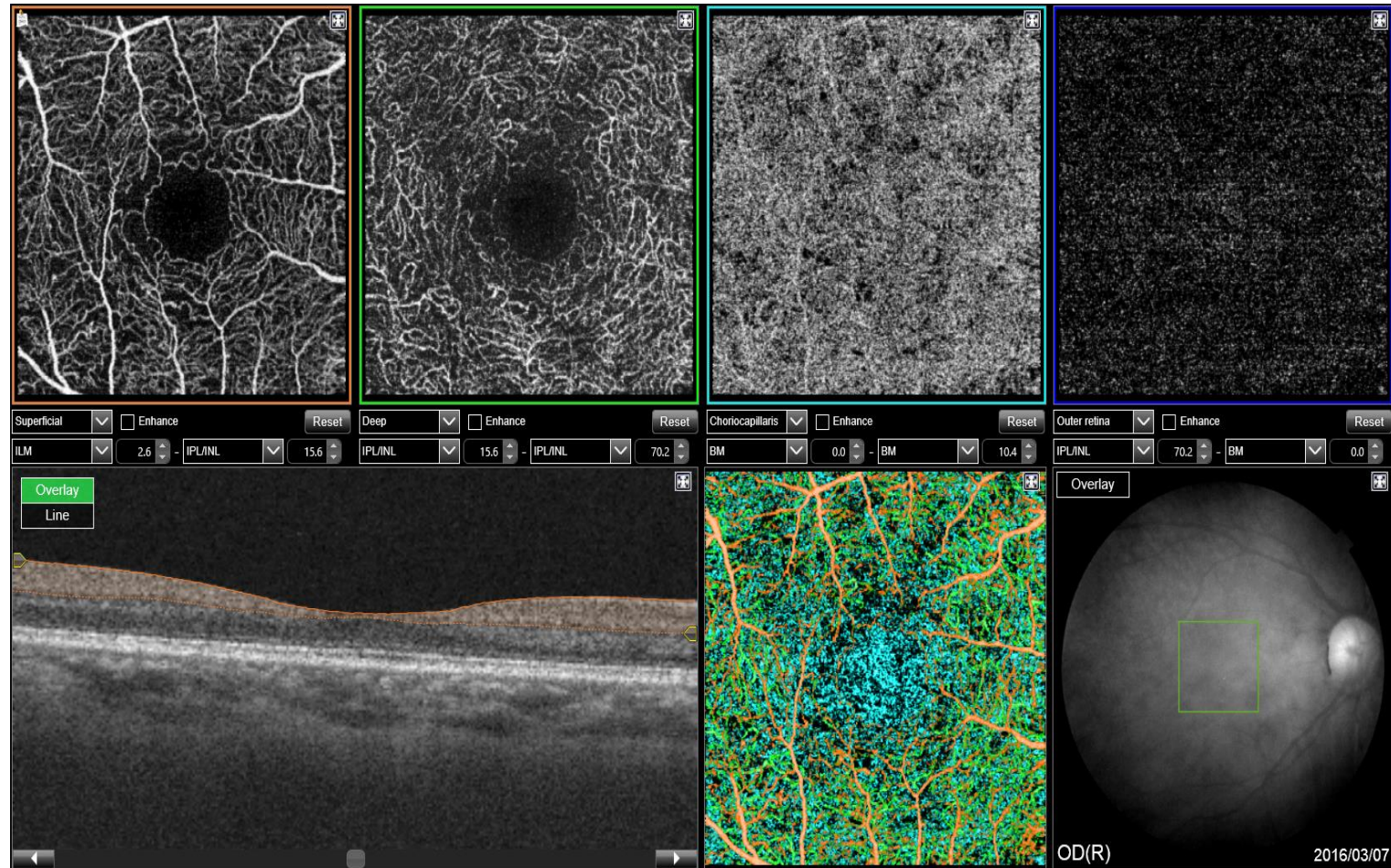
- Cox's proportional hazards with entry time as date of image acquisition and exit time as first available evidence of dementia diagnosis in EMR.
- Censoring was end of available follow-up EMR data or death without EMR evidence of dementia.
- In addition to ACD, AD and VD also considered separately.

TABLE 3 AUC changes with inclusion of RVMs.

	Model with RVMs		Model without RVMs		<i>p</i>
	AUC	95% CI	AUC	95%CI	
ACD	0.7896	0.7731–0.8060	0.7855	0.7689–0.8022	0.022
VD	0.7831	0.7566–0.8095	0.7769	0.7498–0.8041	0.090
AD	0.7759	0.7532–0.7986	0.7685	0.7459–0.7910	0.094

- Increased retinal FDa associated with greater risk of ACD (HR 1.17; 1.08–1.26) & AD (HR 1.33; 1.16–1.52).
- RVMs help predict future dementia incidence independent of other risk factors.
- Differences in retinal microvascular parameters are easily measured and help predict dementia susceptibility in patients with diabetes which may inform dementia prevention strategies.

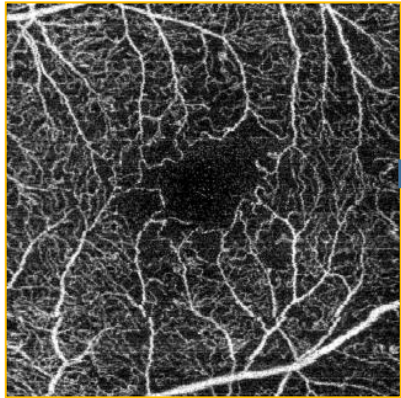
Optical coherence tomography angiography



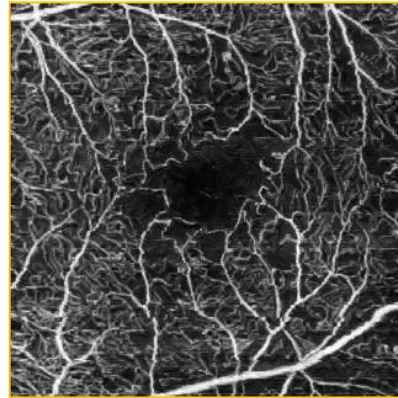
Courtesy of Dr Carol Cheung, The Chinese University of Hong Kong

Automated image processing

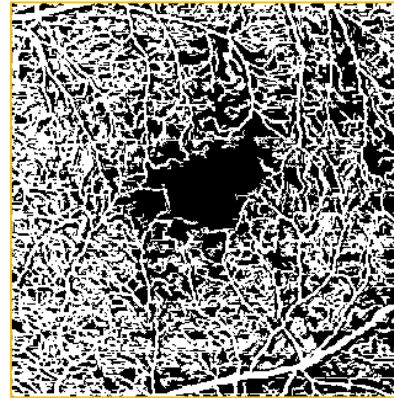
OCT-angiogram



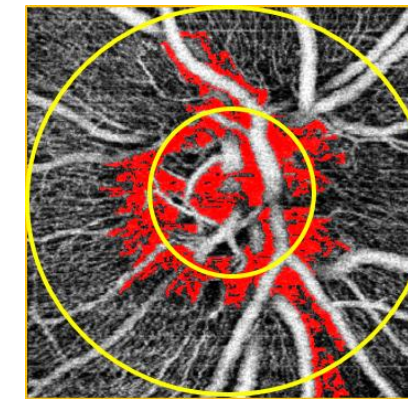
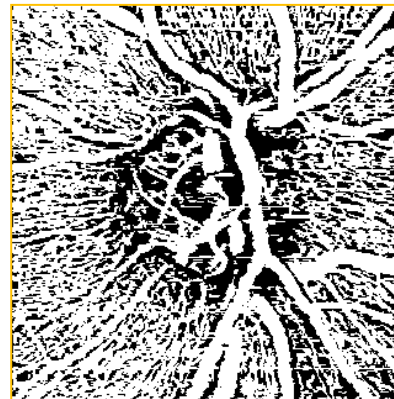
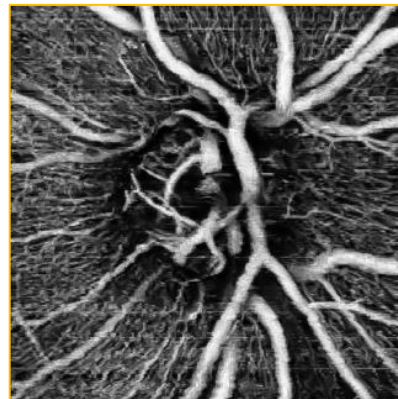
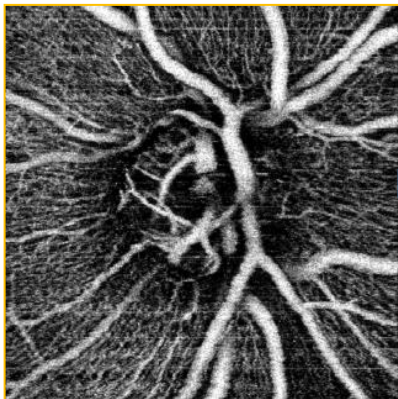
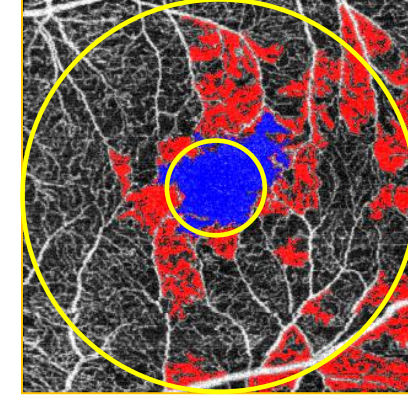
Denoising



Binarization

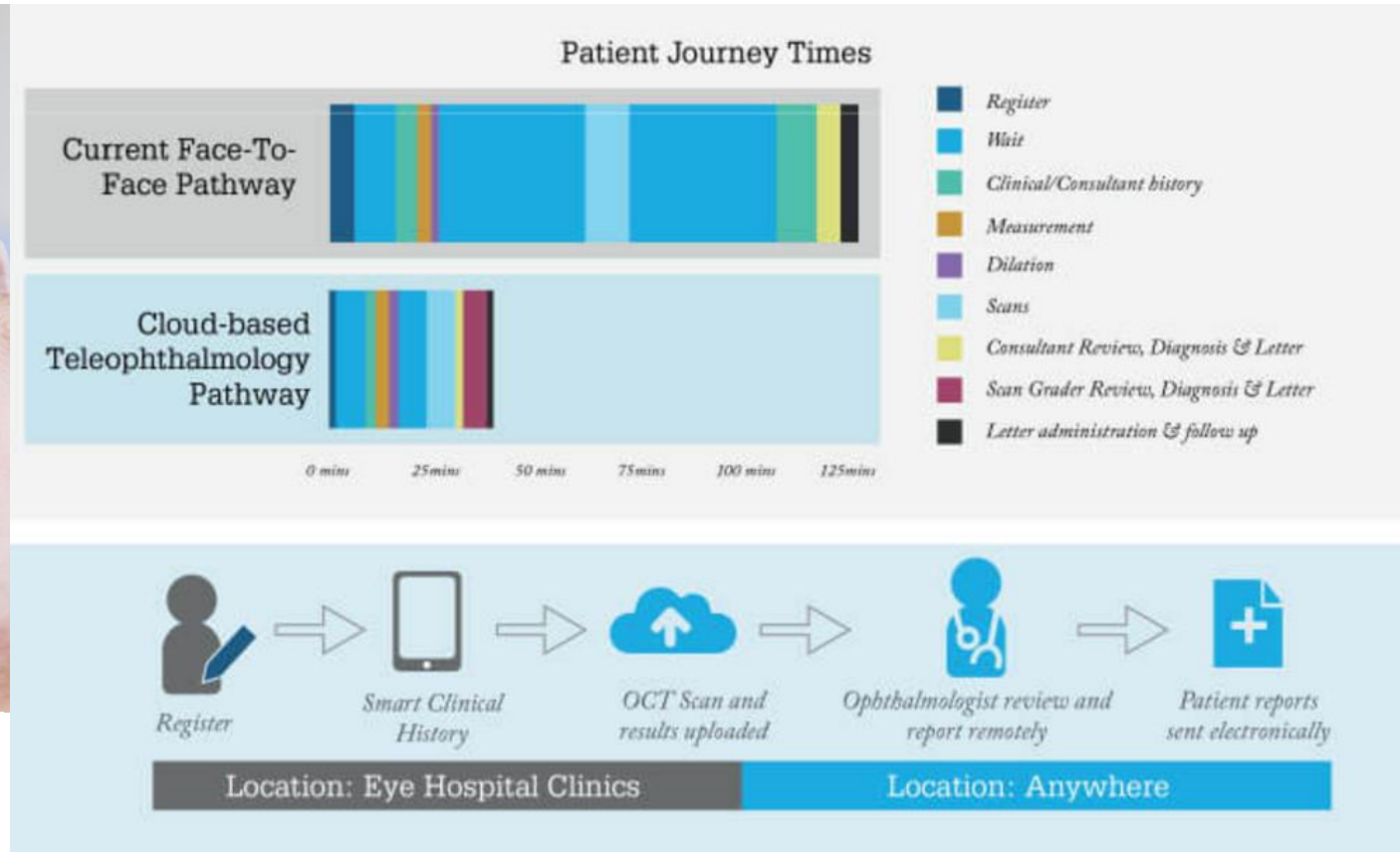


Quantification



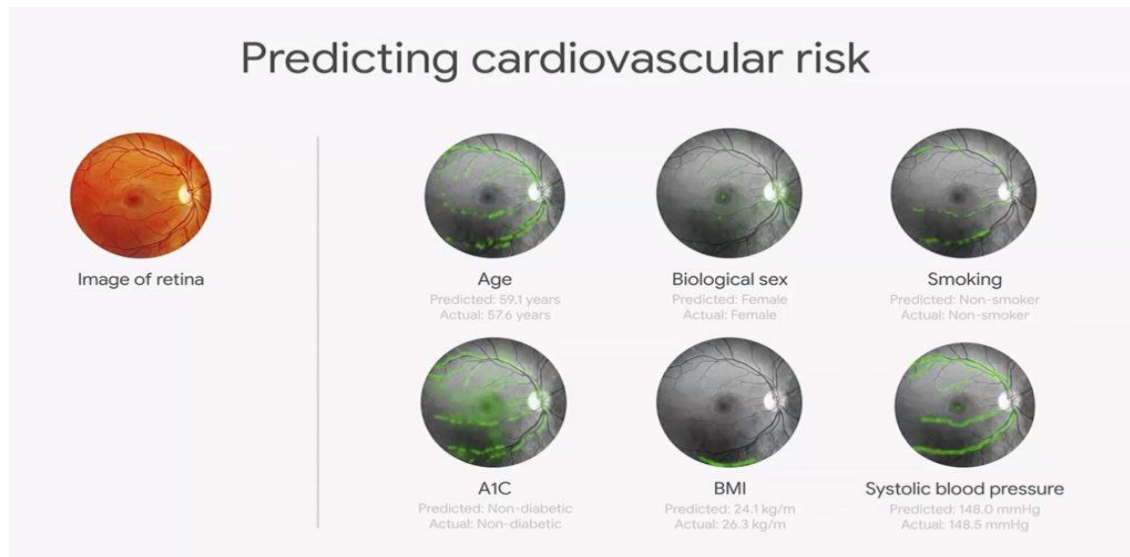
Courtesy of Dr Carol Cheung, The Chinese University of Hong Kong

Impact – Cloud-based technology



Optomed

Impact - The future is AI?



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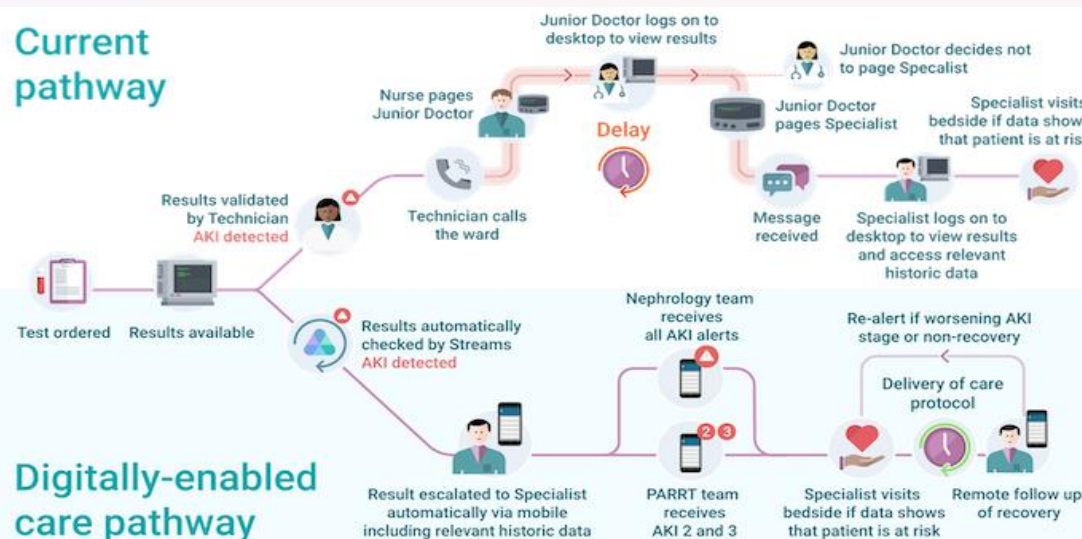
<https://doi.org/10.1038/s41551-018-0195-0>

Prediction of cardiovascular risk factors from retinal fundus photographs via deep learning

Ryan Poplin^{1,4}, Avinash V. Varadarajan^{1,4}, Katy Blumer¹, Yun Liu¹, Michael V. McConnell^{2,3}, Greg S. Corrado¹, Lily Peng^{1,4*} and Dale R. Webster^{1,4}

Traditionally, medical discoveries are made by observing associations, making hypotheses from them and then designing and running experiments to test the hypotheses. However, with medical images, observing and quantifying associations can often be difficult because of the wide variety of features, patterns, colours, values and shapes that are present in real data. Here, we show that deep learning can extract new knowledge from retinal fundus images. Using deep-learning models trained on data from 284,335 patients and validated on two independent datasets of 12,026 and 999 patients, we predicted cardiovascular risk factors not previously thought to be present or quantifiable in retinal images, such as age (mean absolute error within 3.26 years), gender (area under the receiver operating characteristic curve (AUC)=0.97), smoking status (AUC=0.71), systolic blood pressure (mean absolute error within 11.23 mmHg) and major adverse cardiac events (AUC=0.70). We also show that the trained deep-learning models used anatomical features, such as the optic disc or blood vessels, to generate each prediction.

Current pathway



Direction of travel



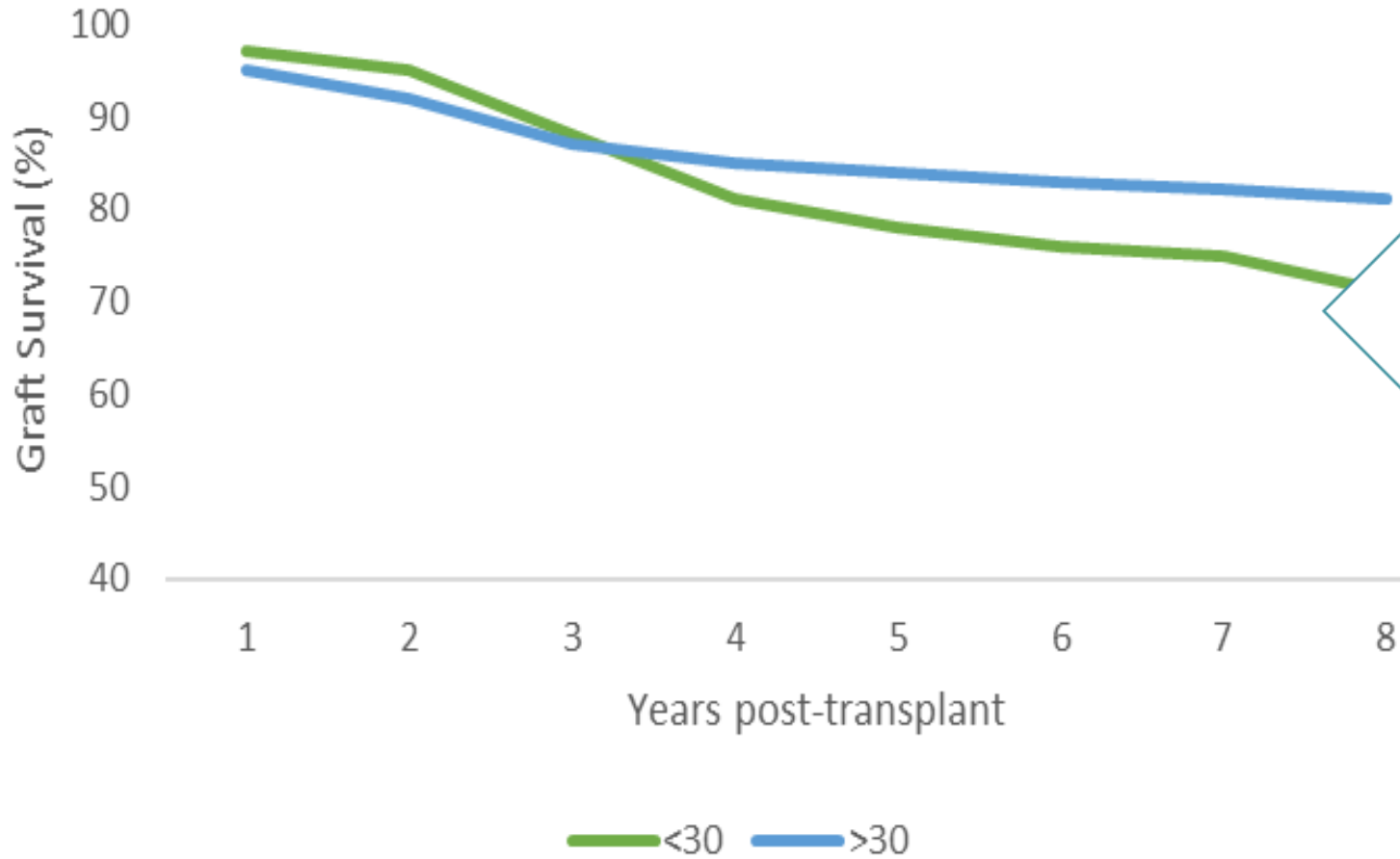
➤ Oculomics

➤ DKD – MultiOmics Health

Expanding multiomics approaches –

Why do kidney transplants fail so early in young people?

Graft Survival- <30 vs >30



Why does this matter?



Individual



Cost



Societal



3200 renal transplants annually in the UK & Ireland

25% of our local transplants in individuals <30

The science and the art of post transplant care



TOO HOT



TOO COLD



JUST RIGHT

In clinical practice determination of 'optimal' immunosuppression is limited by little objective parameters to guide decision making

Monitoring transplant function

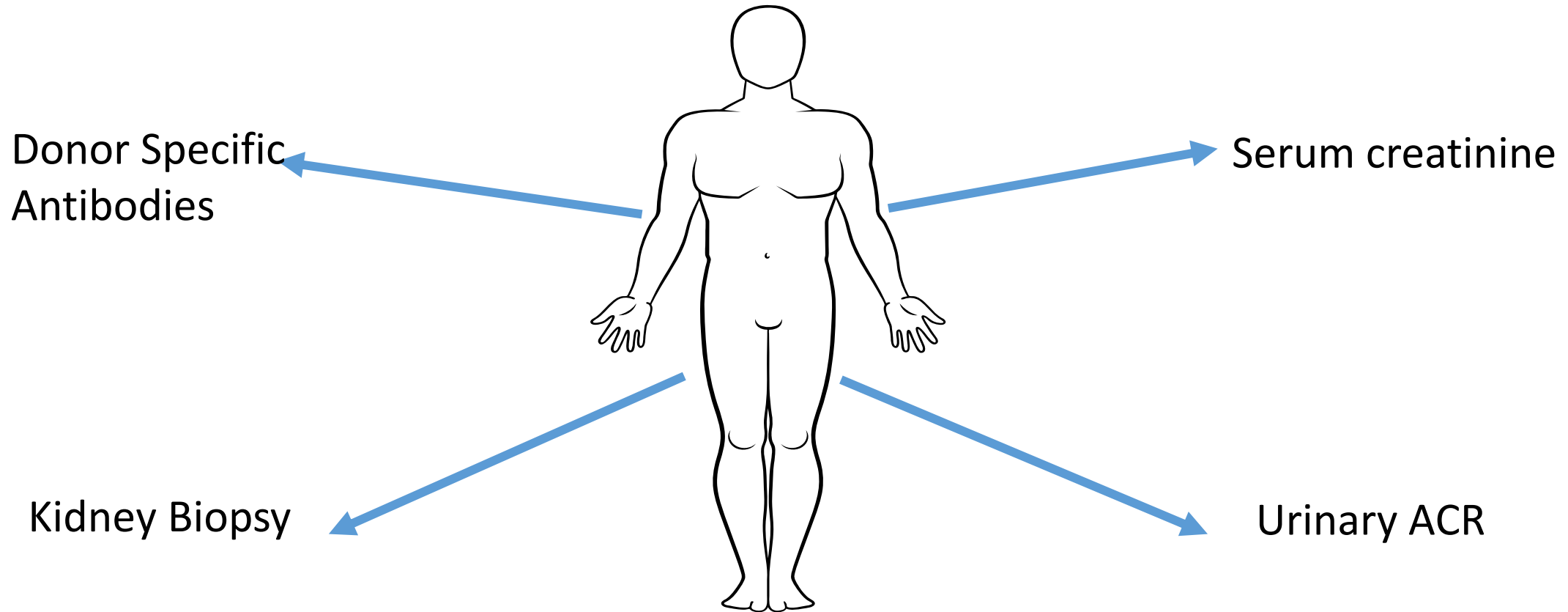
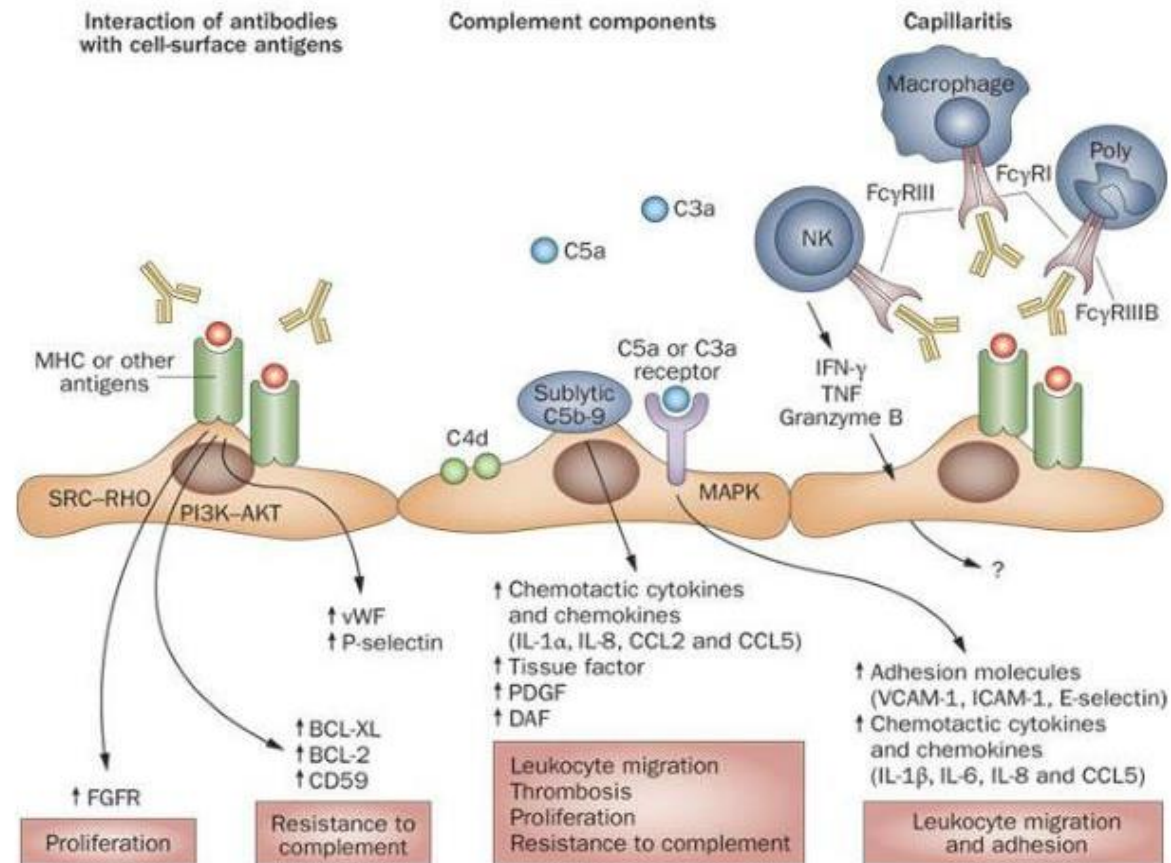


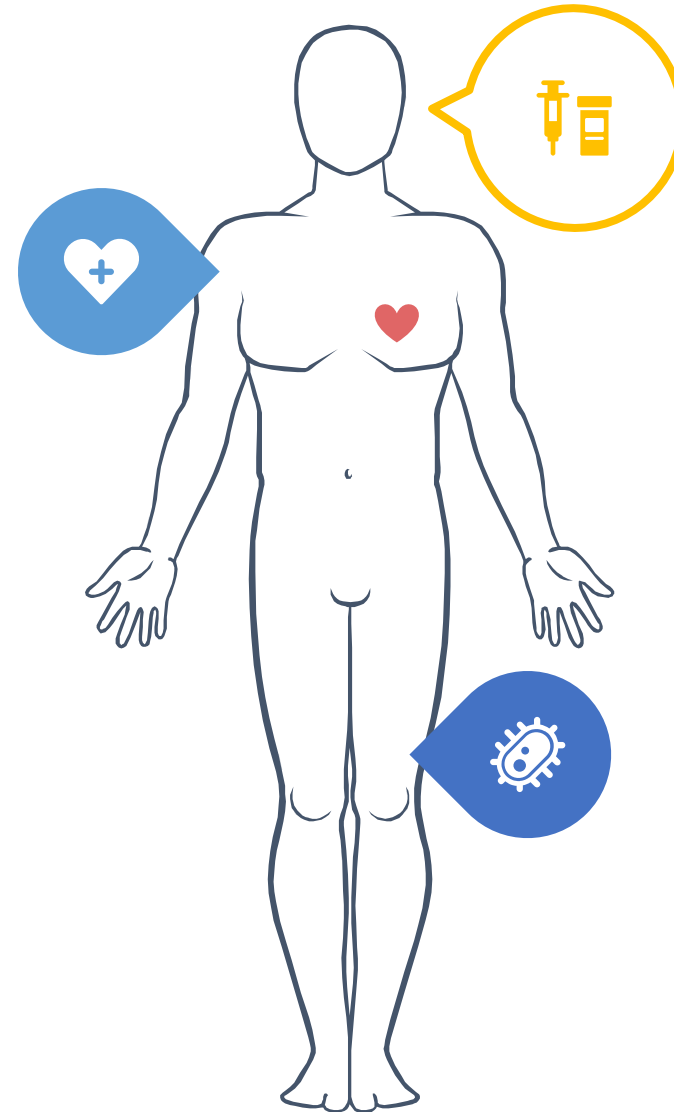
Figure 1 Mechanisms of donor-specific antibody-mediated endothelial injury in renal allografts



Farkash, E. A. & Colvin, R. B. (2012) Diagnostic challenges in chronic antibody-mediated rejection
Nat. Rev. Nephrol. doi:10.1038/nrneph.2012.61

Why do kidney transplants fail so early in young people?

- better understanding of underlying pathophysiology
- identify early molecular markers of immunological mediated graft injury



Relatively large
homogenous renal
transplant population

Database- excellent
phenotypical
knowledge of our
patients

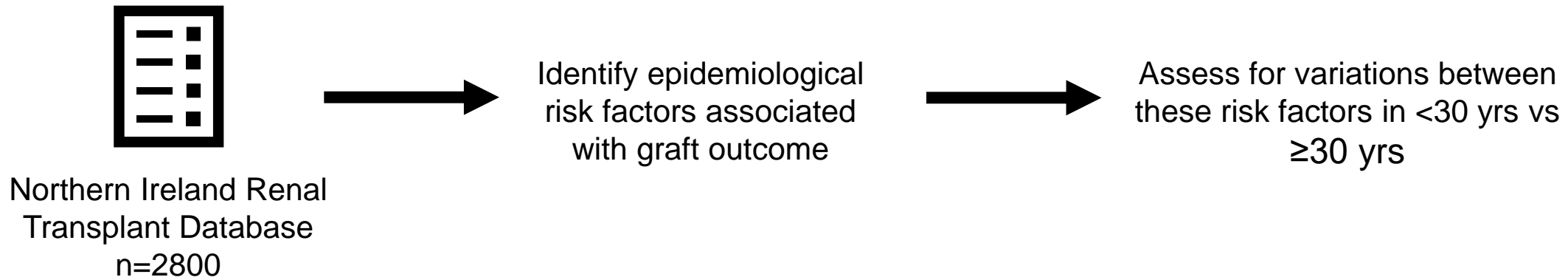


Access to historical samples
stored as part of H&I testing

Aim

To explore long-term kidney transplant outcomes between younger recipients (< 30 yrs) vs older recipients (\geq 30 yrs) for association with:

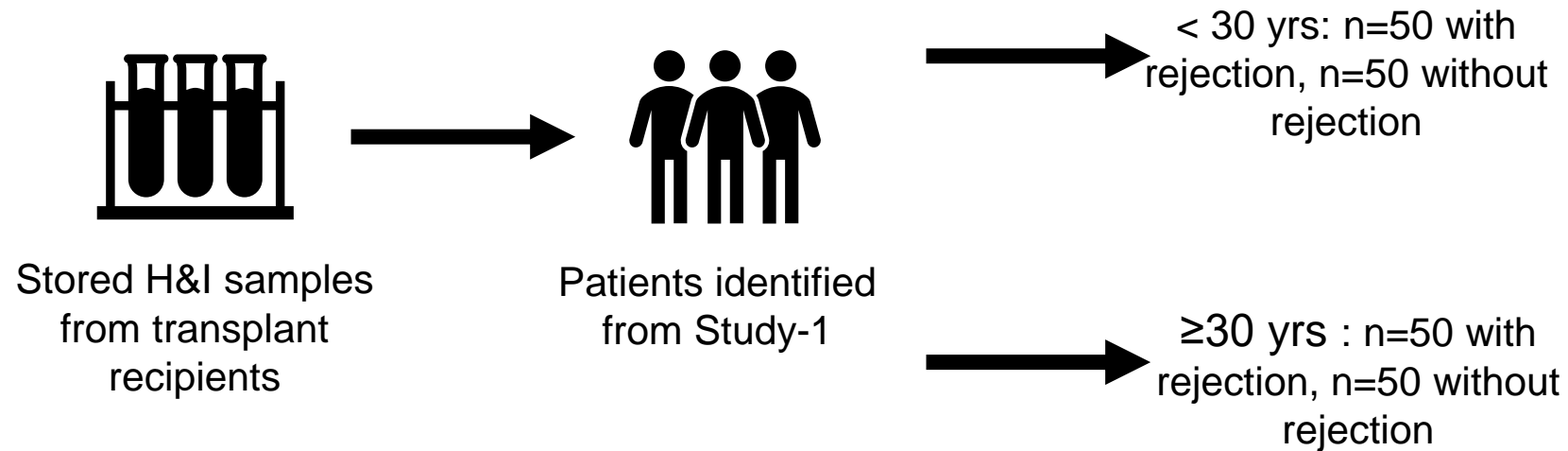
- Differences in epidemiological risk factors.



Aim

To explore whether variable long-term kidney transplant outcomes between younger recipients (< 30 yrs) vs older recipients (\geq 30 yrs) is associated with:

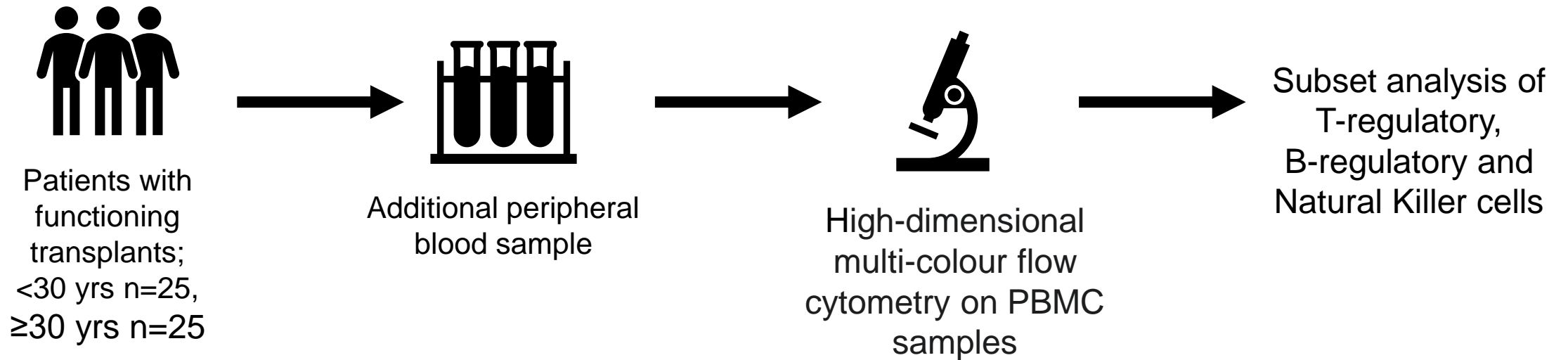
- Proteomic/ Metabolomic/ WGS profiles.



Aim

To explore whether variable long-term kidney transplant outcomes between younger recipients (< 30 yrs) vs older recipients (≥ 30 yrs) is associated with:

➤ Cellular modelling.



Concluding Thoughts

- Comorbid chronic disease prevalence for many conditions continue to increase with improved healthcare provision and ageing populations
- Genetics studies facilitate identification of risk variants & disease pathways
- Epigenetic studies help determine the influence of environmental & lifestyle factors on disease outcomes
- miRNA's influence gene expression and may help predict disease outcomes
- Oculomics enables non-invasive evaluation of microvascular health
- Proteomic & metabolomic approaches can identify potential drug targets.
- Multiomic approaches & data integration represent novel opportunities to elucidate disease risks, mechanisms, progression and therapeutic targets.



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Thank you for your attention