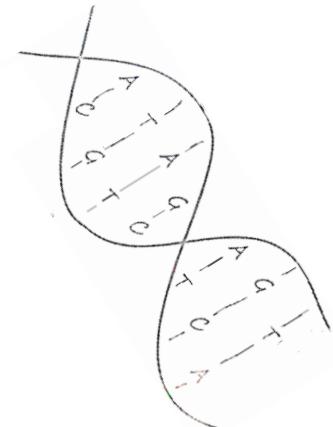




浙江大学 医学中心
ZHEJIANG UNIVERSITY MEDICAL CENTER



单细胞多组学在解析慢性肾脏病的遗传学致病机制的应用



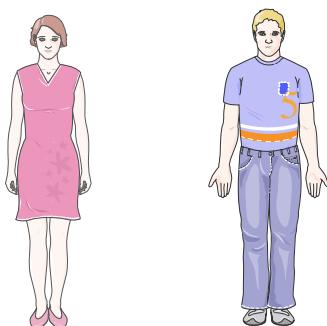
盛 欣
浙江大学医学中心

内容大纲

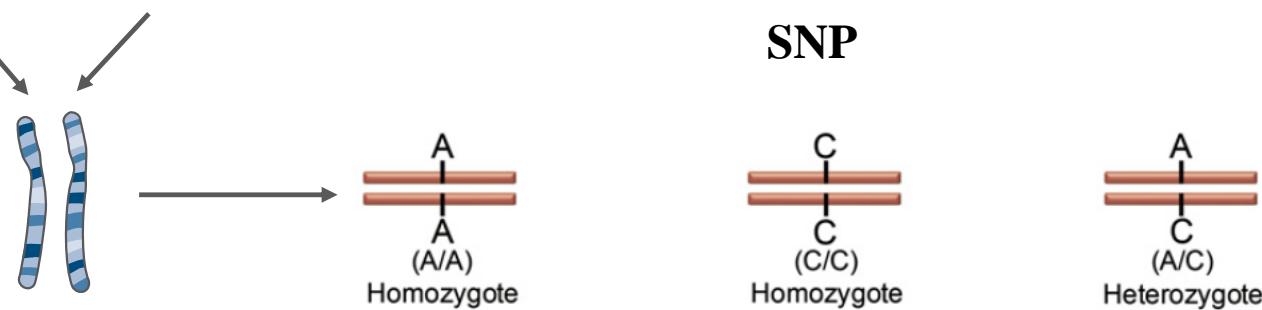
- 基本概念
- 慢性肾脏病及其并发症
- 多组学数据整合分析方法
- 肾脏具有高度细胞异质性
- 从单细胞水平解析SNP-基因-CKD机制
- 总结

单核苷酸多态 (SNP)

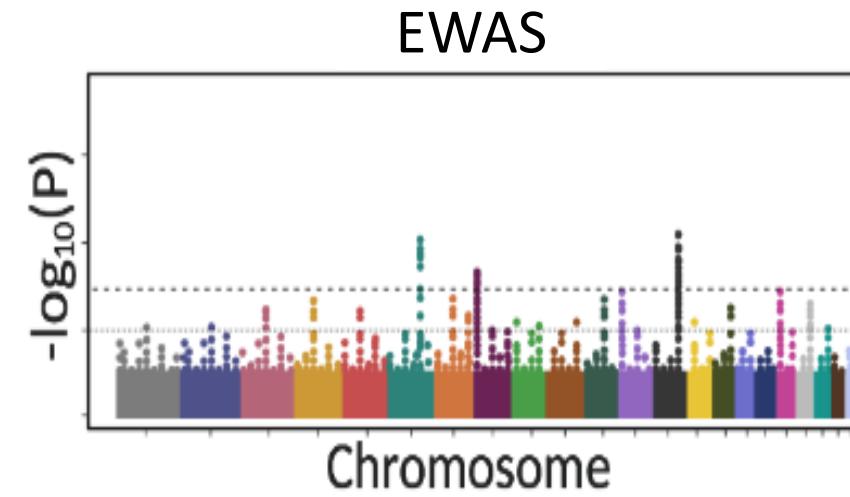
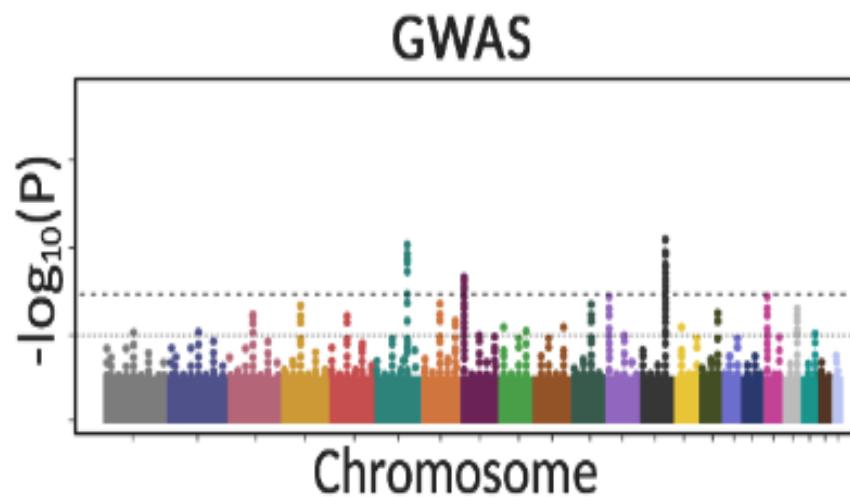
人与人的基因序列中，有99.9%以上的序列都是相同的，仅有0.1%不同。



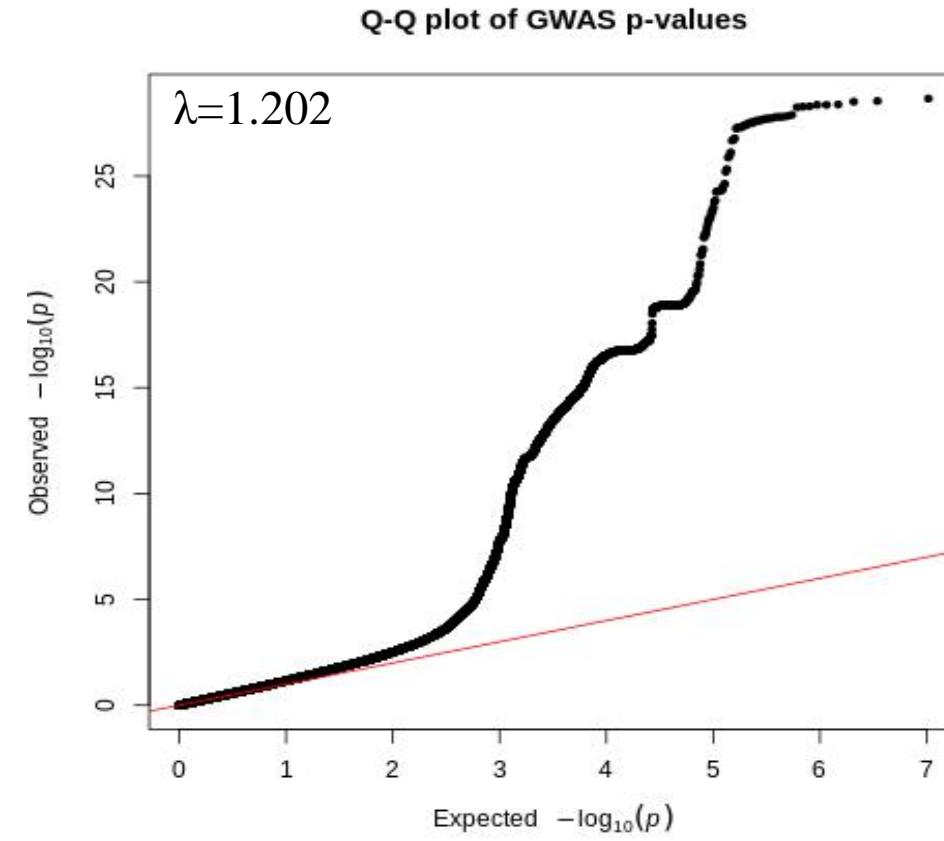
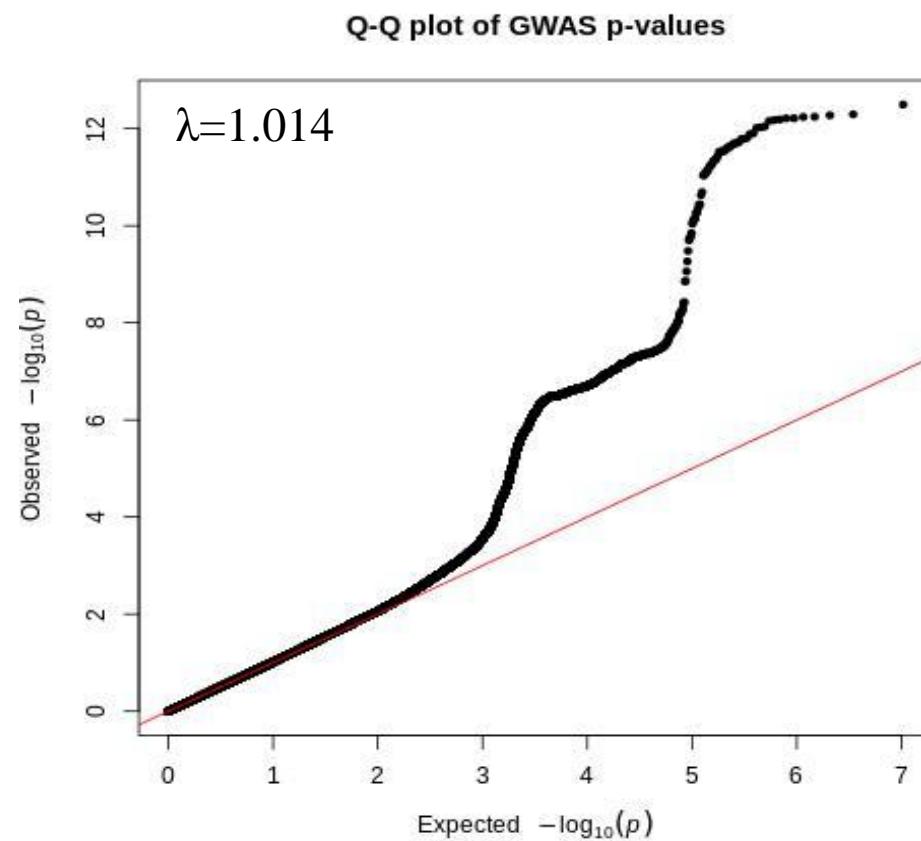
SNP: 基因组水平上由单个核苷酸的变异所引起的DNA序列多态性。
SNP是由单个碱基的转换或颠换所引起，也可由碱基的插入或缺失所致。



全X组关联分析 (XWAS)



膨胀系数：不止是线性回归



GWAS信息能有效指导新药研发

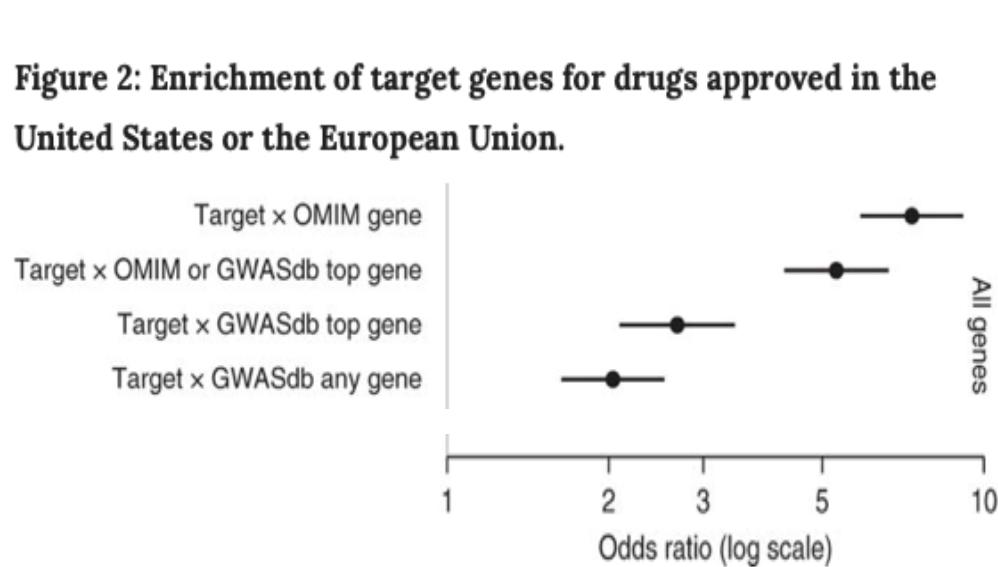
ANALYSIS

nature
genetics

The support of human genetic evidence for approved drug indications

Matthew R Nelson¹, Hannah Tipney², Jeffery L Painter¹, Judong Shen¹, Paola Nicoletti³, Yufeng Shen^{3,4}, Aris Floratos^{3,4}, Pak Chung Sham^{5,6}, Mulin Jun Li^{6,7}, Junwen Wang^{6,7}, Lon R Cardon⁸, John C Whittaker² & Philippe Sanseau²

Figure 2: Enrichment of target genes for drugs approved in the United States or the European Union.



但有GWAS还远远不够...

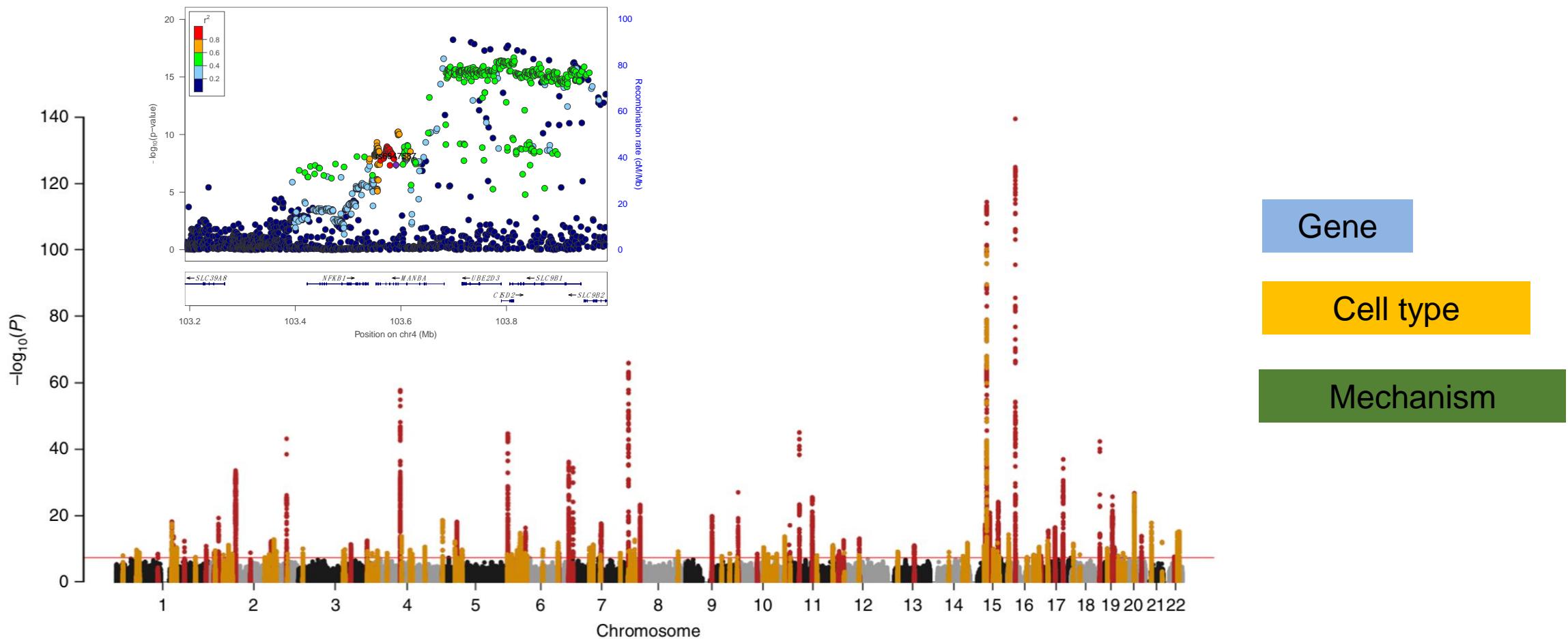
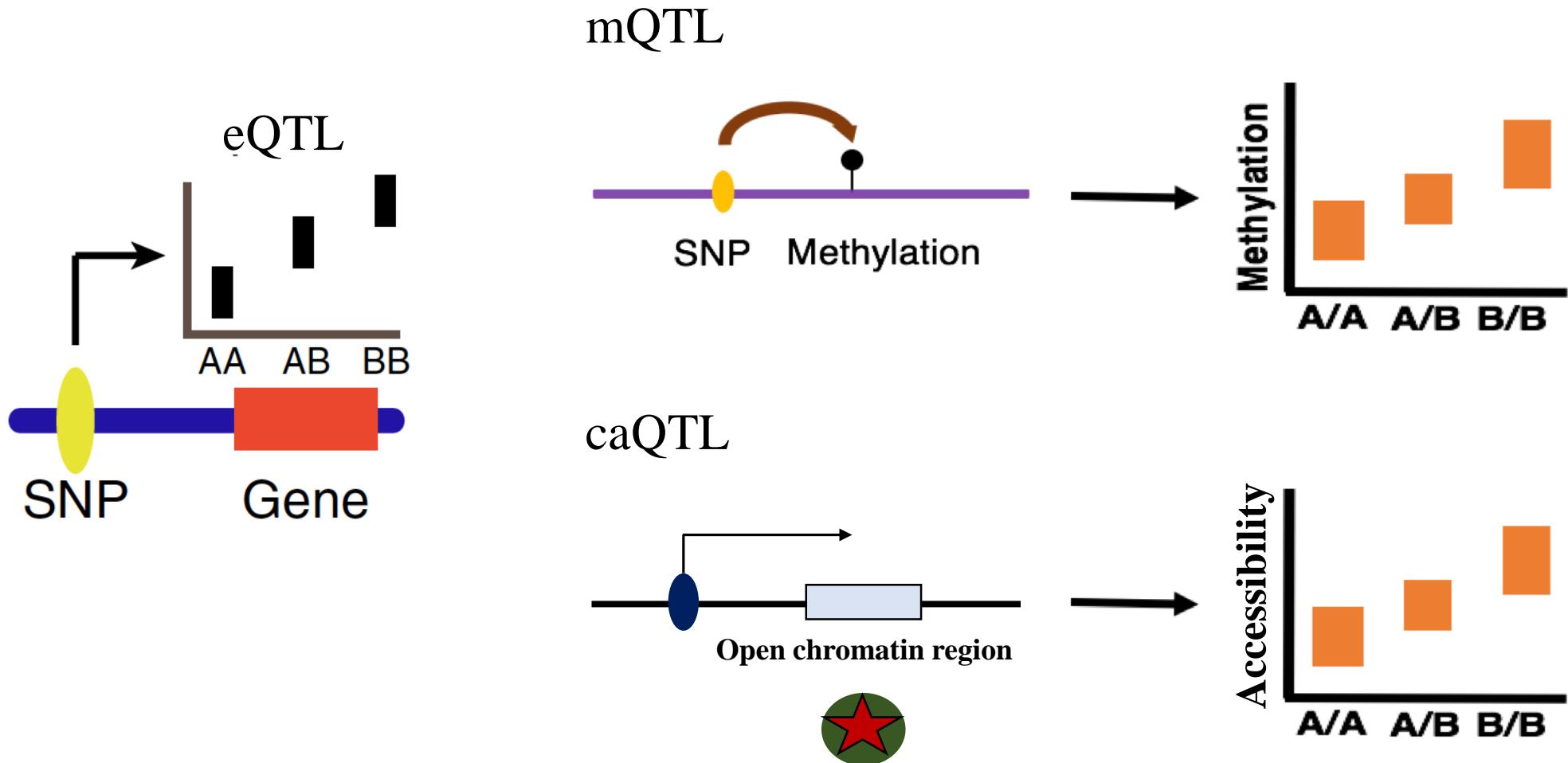


Fig. 1 Manhattan plot summarizing transethnic discovery meta-analysis of eGFR. The y axis shows the $-\log_{10} P$ -values and the x axis shows the chromosomal positions. The horizontal red line represents the thresholds of P -value = 5×10^{-8} for genome-wide significance. SNPs in red are in previously-identified kidney function loci, whereas SNPs in orange are in novel loci

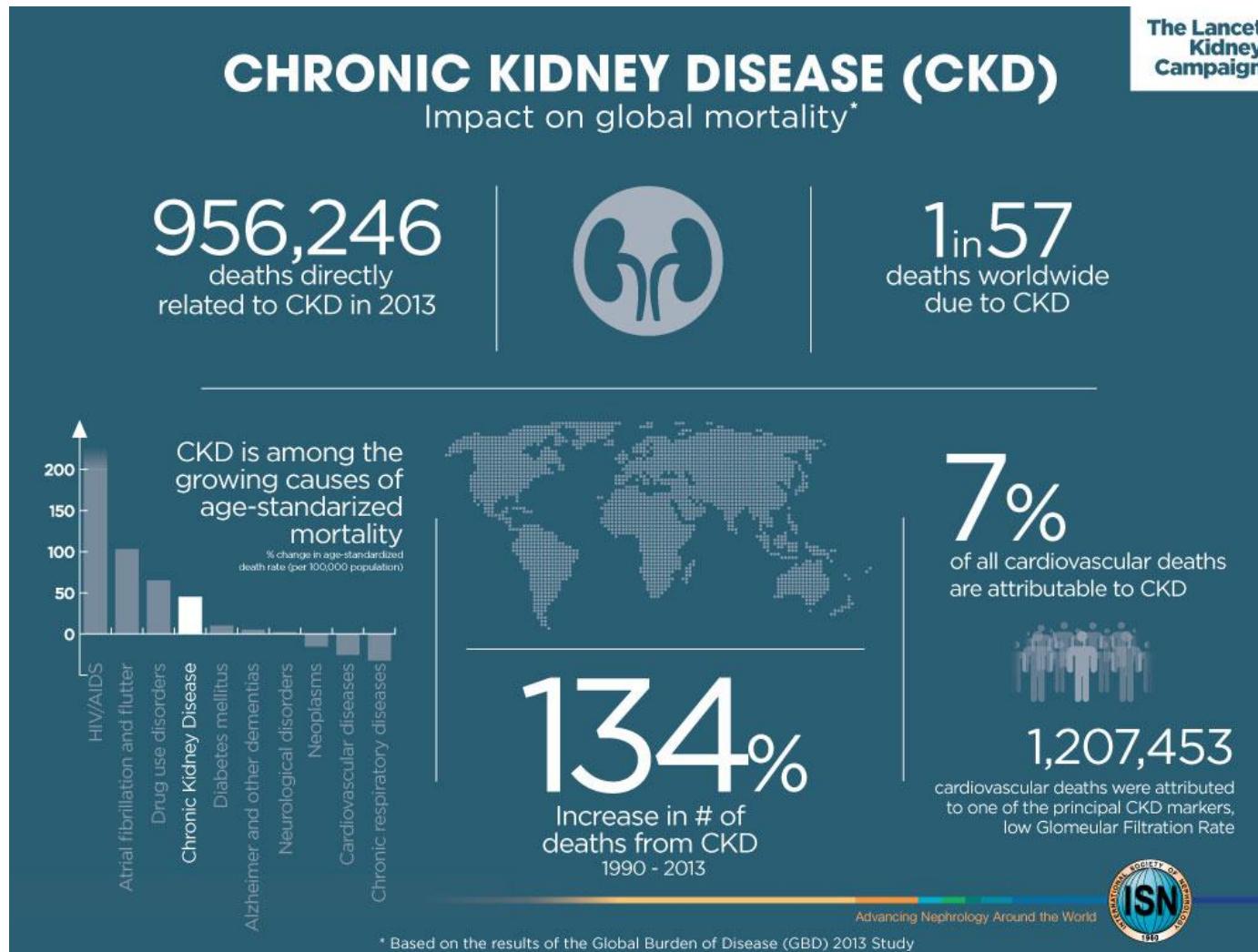
数量性状位点 (xQTL)



内容大纲

- 基本概念
- 慢性肾脏病及其并发症
- 多组学数据整合分析方法
- 肾脏具有高度细胞异质性
- 从单细胞水平解析SNP-基因-CKD机制
- 总结

慢性肾脏病 (Chronic Kidney Disease, CKD)



回顾经典：糖尿病控制及并发症研究（DCCT）

初级预防队列N=726

- 1-5年糖尿病病程
- 无视网膜病变
- 尿白蛋白肌酐比 $< 40 \text{ mg/24h}$

二级干预防队列N=715

- 1-15年糖尿病病程
- 轻到中度非增殖期视网膜病变
- 尿白蛋白肌酐比 $< 200 \text{ mg/24h}$

随机分组

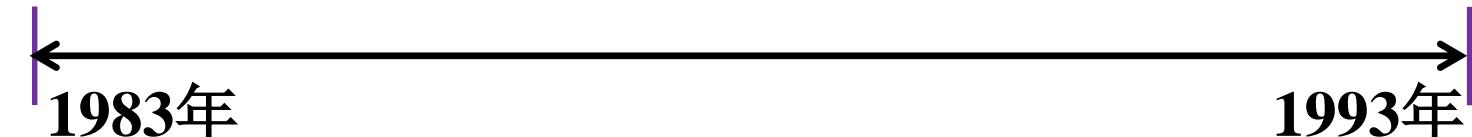


强化治疗：胰岛素泵或每天 ≥ 3 次/天胰岛素注射，N=348

标准治疗：1-2次/天胰岛素注射，N=378

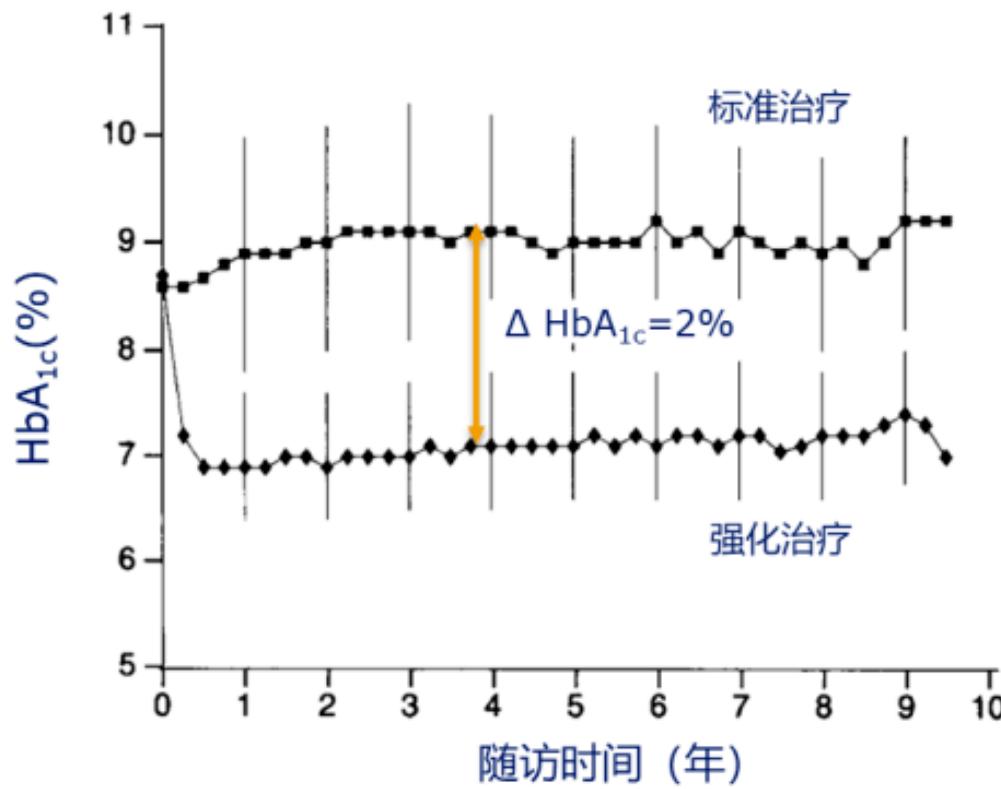
强化治疗：胰岛素泵或每天 ≥ 3 次/天胰岛素注射，N=363

标准治疗：1-2次/天胰岛素注射，N=352



1441名T1DM患者：胰岛素依赖性、年龄13-39岁、无高血压、高胆固醇血症、糖尿病严重并发症或内科疾病

回顾经典：糖尿病控制及并发症研究（DCCT）



微血管并发症	相对风险减少, % (95% 置信区间)
视网膜病变进展	63 (52-71)
视网膜黄斑变性	26 (-8-50)
严重非增殖期/增殖期视网膜病变	47 (15-67)
视网膜病变激光治疗	51 (21-70)
尿白蛋白肌酐比 (mg/24h)	
◆ ≥ 40	39 (21-52)
◆ ≥ 300	54 (19-74)
临床神经病变 (随访第5年)	60 (38-74)

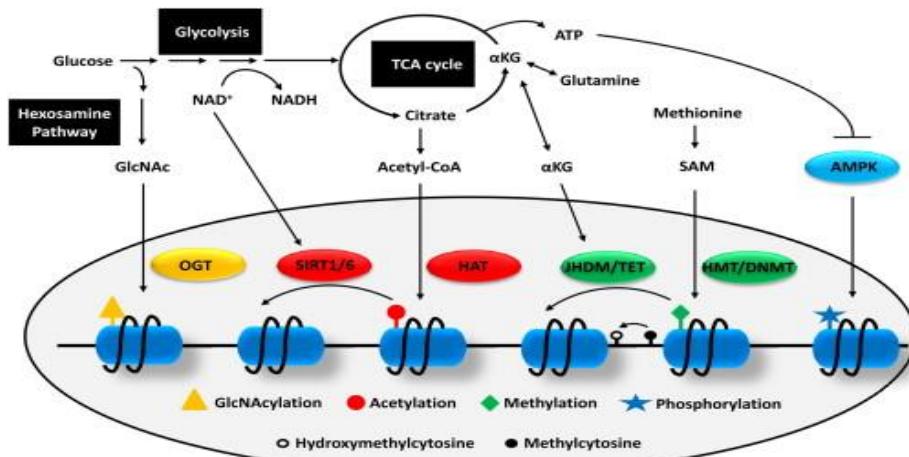
环境所留下的足迹：表观基因组修饰

- Metabolic factors play critical roles in CKD development.

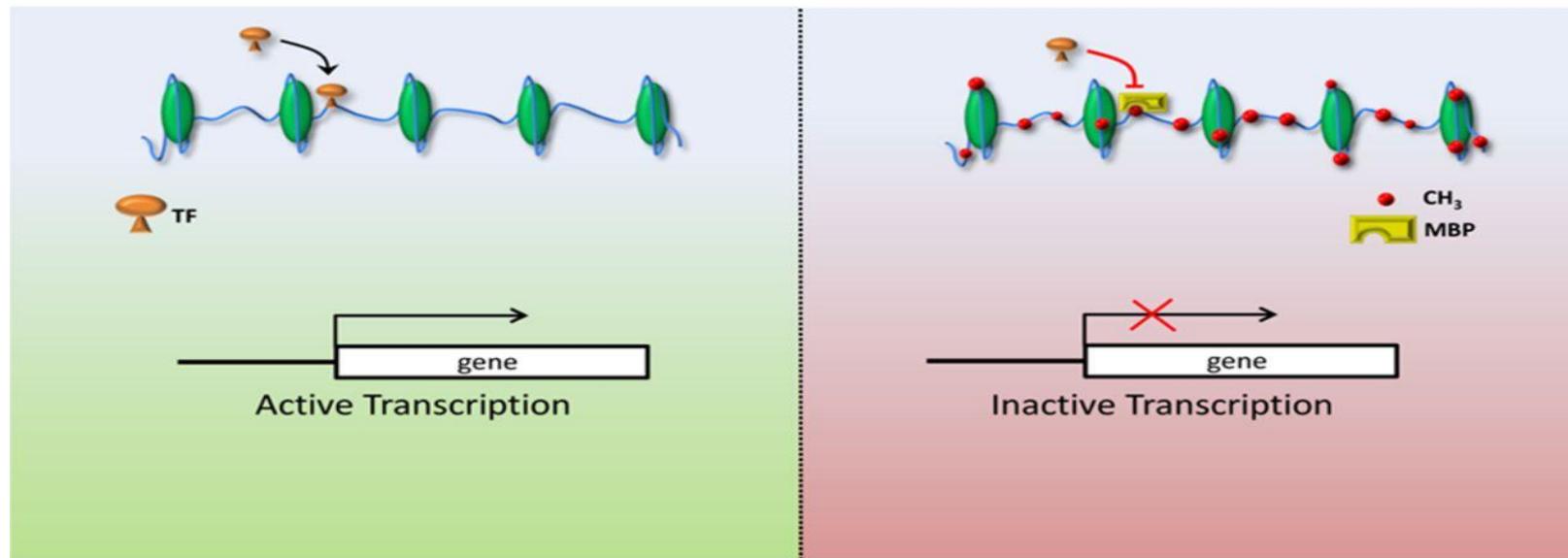
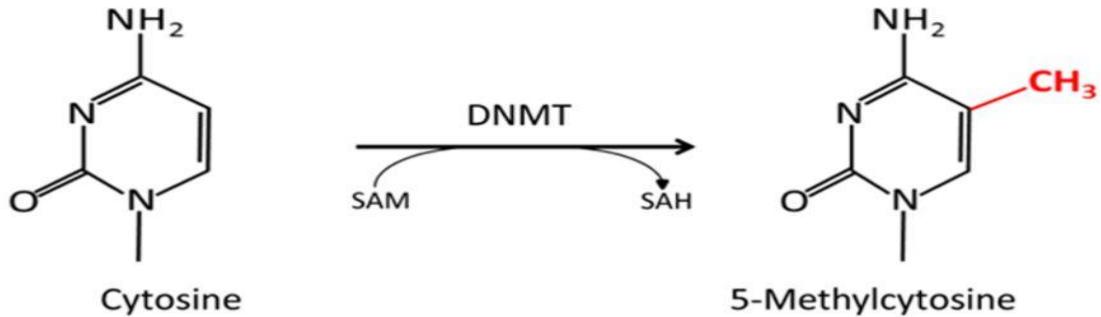


- Metabolic memory or programming:
Intrauterine nutritional deprivation or periods of hyperglycemia will increase kidney disease risk, even after several decades of good metabolic control.

- Epigenetic modifications are maintained during cell division.



环境所留下的足迹：DNA甲基化



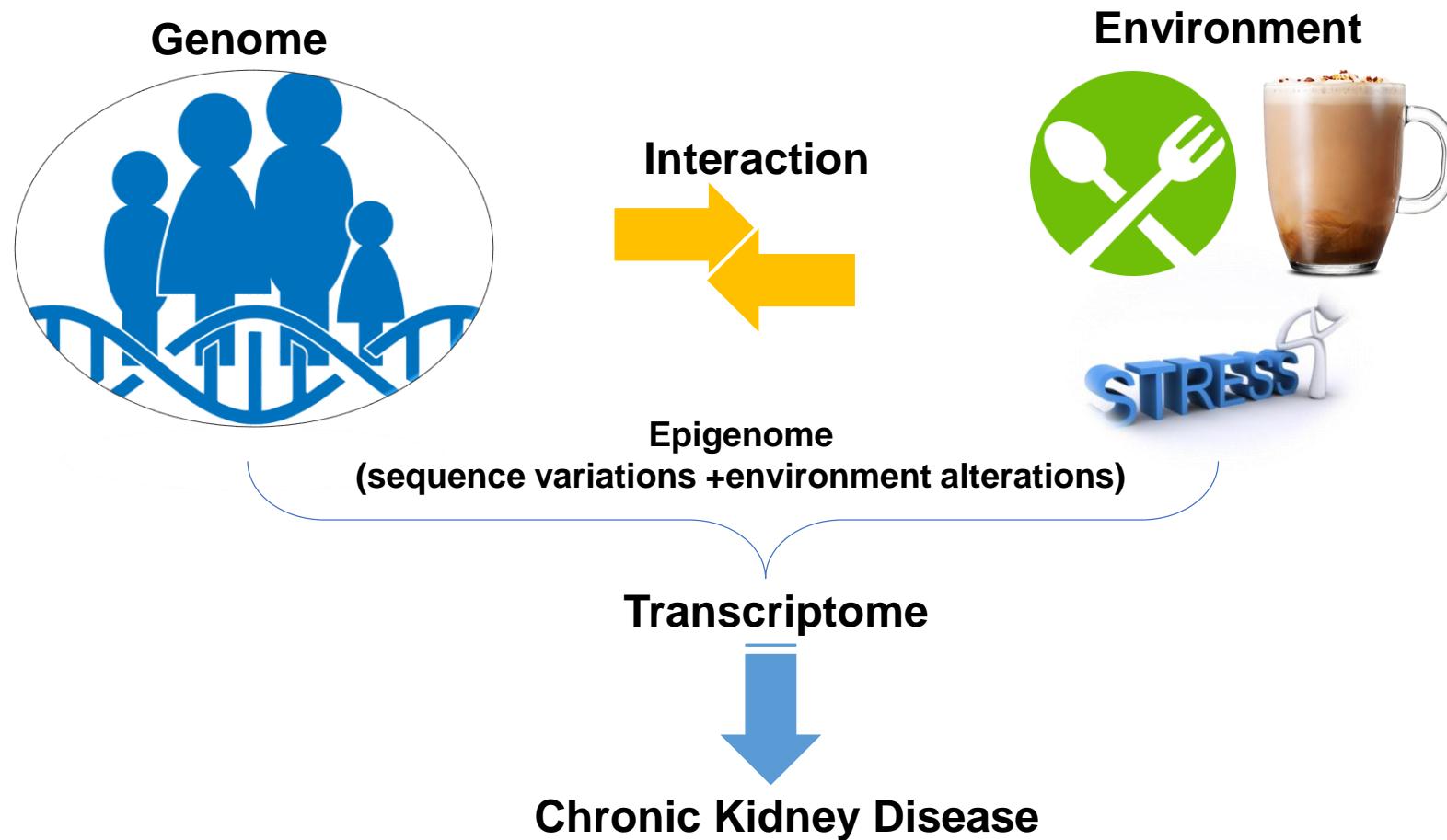
MBP: methyl CpG-binding protein

DNMT: DNA methyltransferase

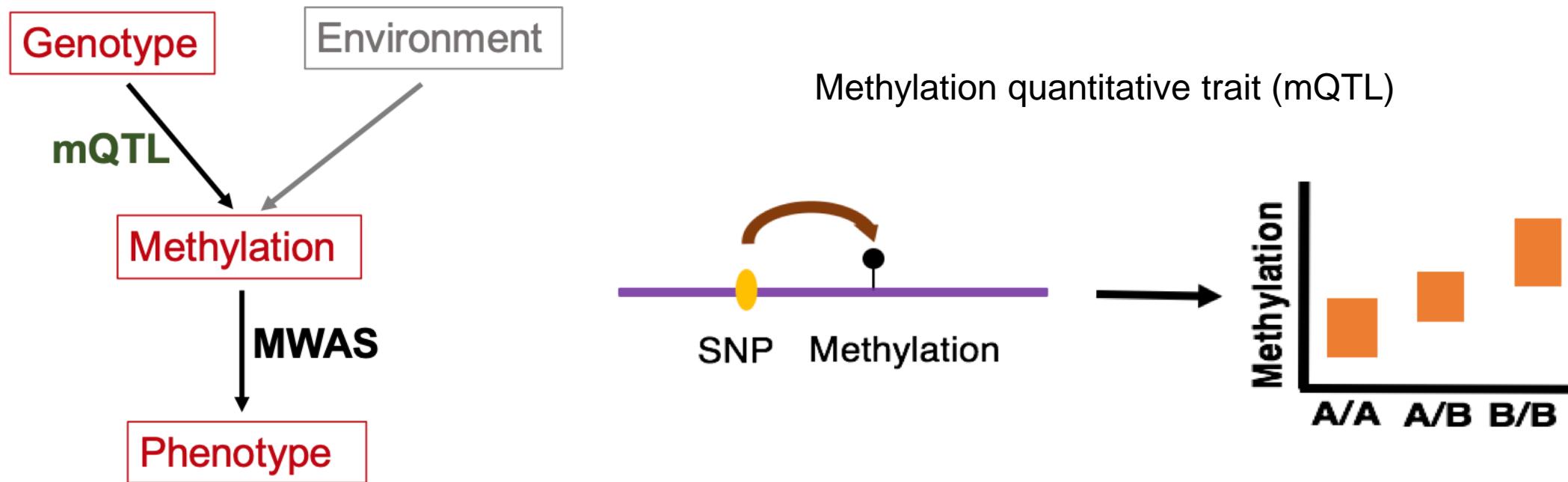
SAM: S-adenosylmethionine

SAH: S-adenosylhomocysteine

遗传-环境相互作用→糖尿病肾病 (DKD)

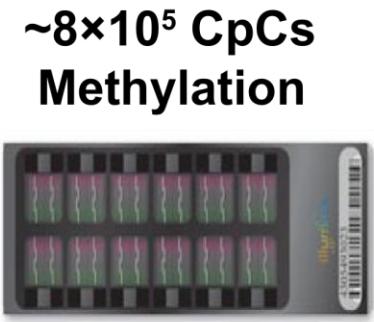
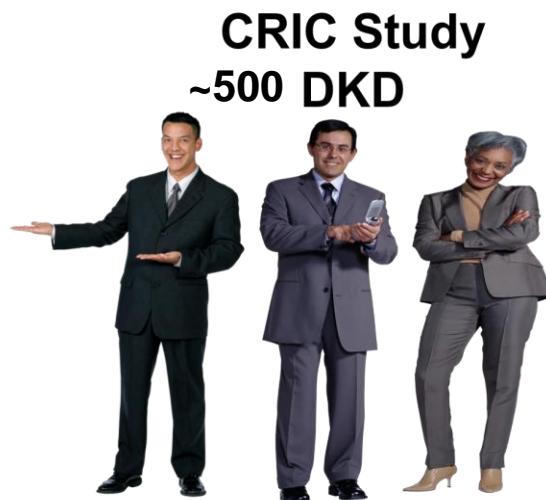


遗传-环境相互作用→糖尿病肾病 (DKD)



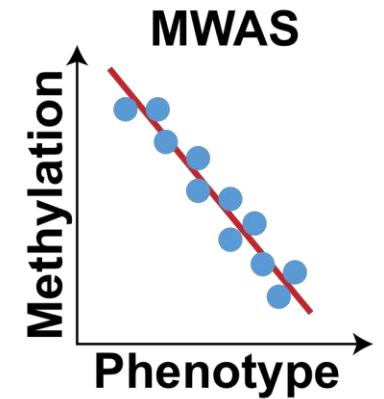
全甲基化组关联分析 (MWAS)

We defined genome-wide methylation changes associated with four DKD phenotypes: Glycemia, albuminuria, eGFR, and eGFR decline.

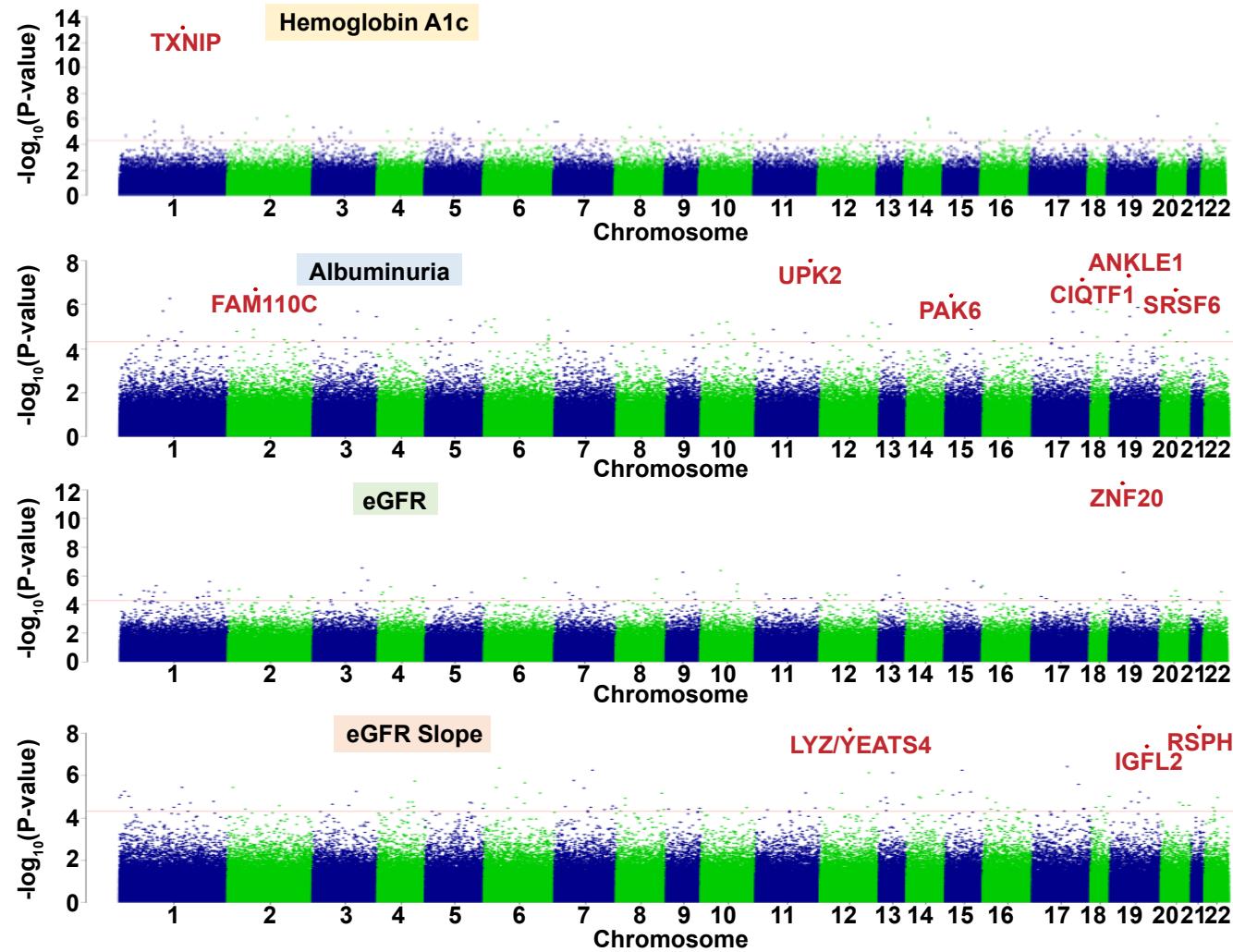


Phenotype

- Glycemia (HgbA1c)
- Albuminuria
- Kidney function (eGFR)
- Kidney function decline (eGFR slope)



与糖尿病肾病相关的甲基化位点



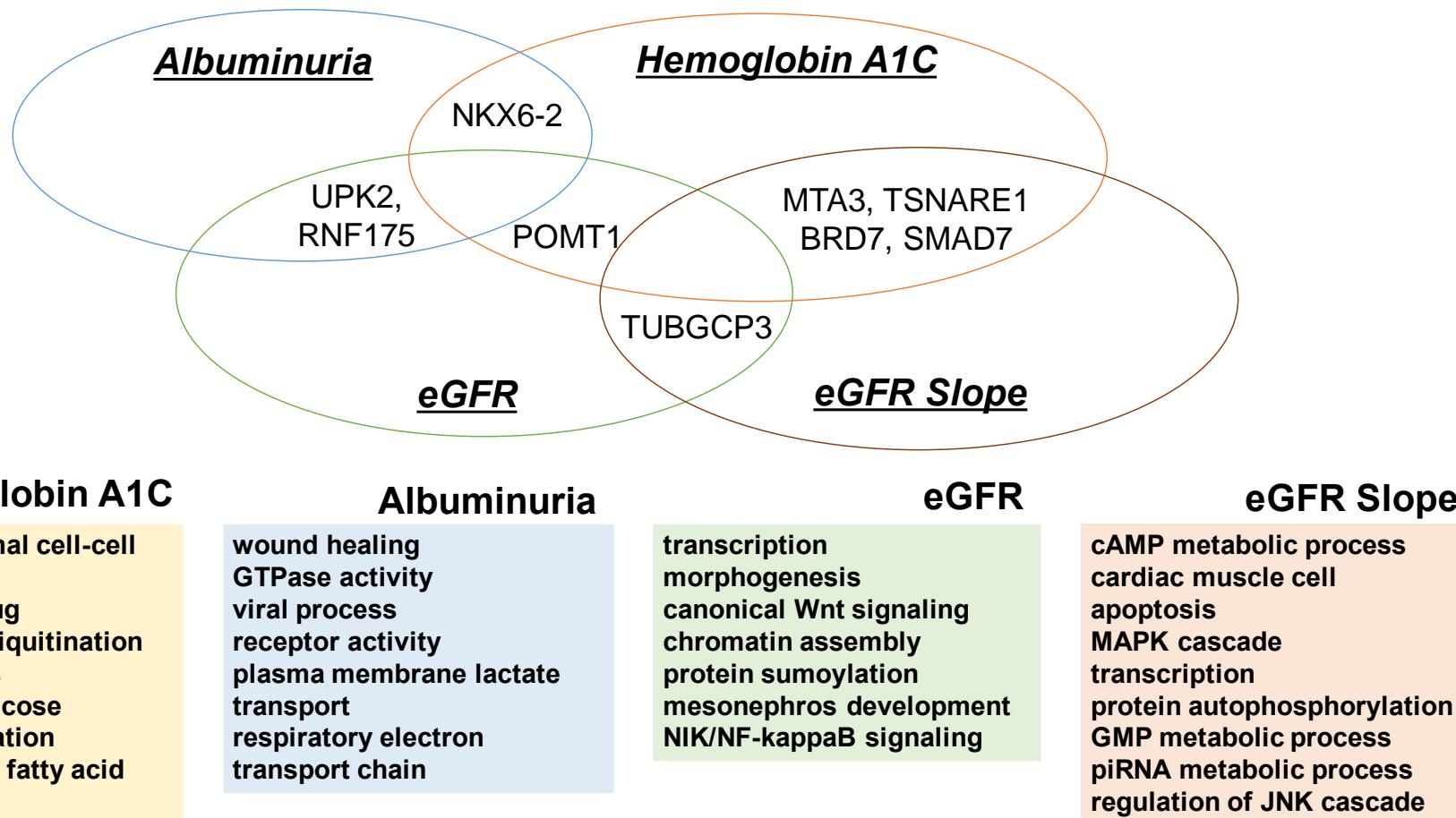
TXNIP (thioredoxin-interacting protein) plays an important role in redox homeostasis and a physiologic regulator of peripheral glucose uptake into fat and muscle in human.

Uroplakins cover urothelial apical surfaces. Mice with null mutation of Upk2 are often born with congenital kidney disease.

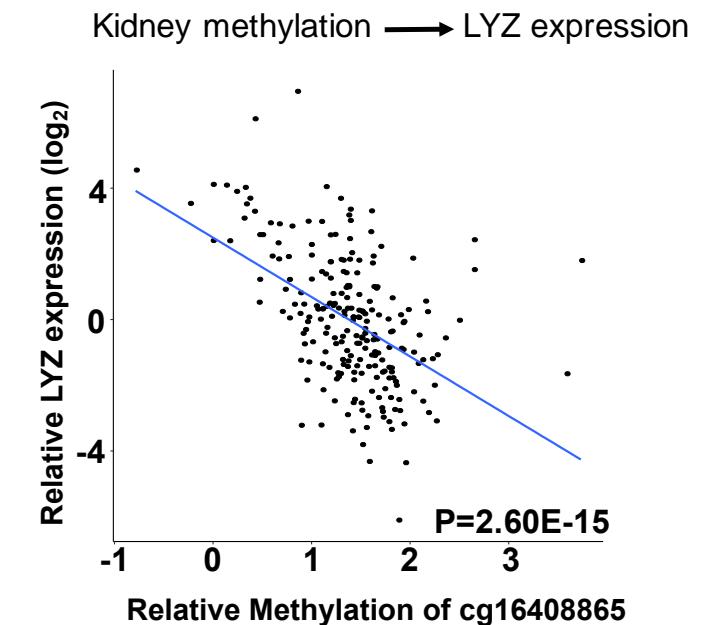
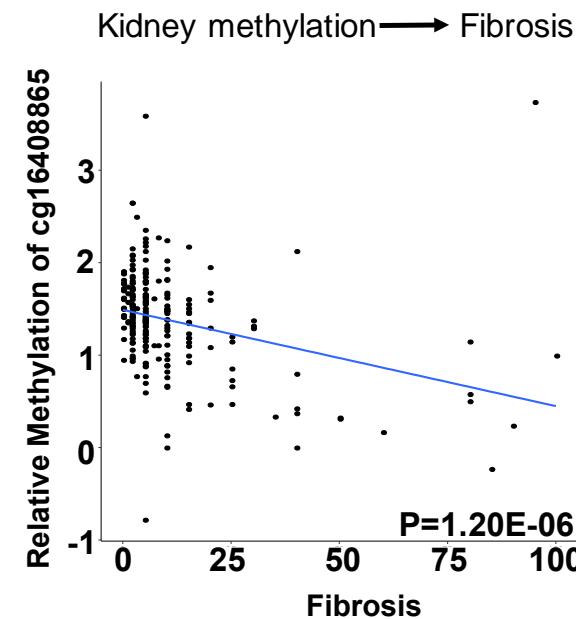
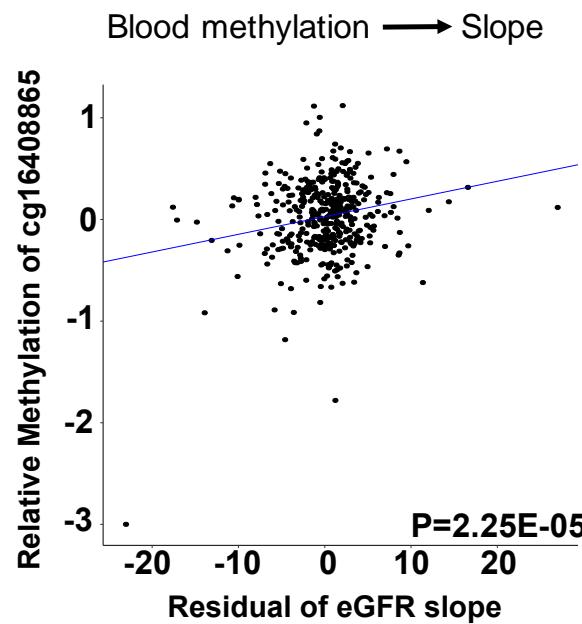
Cg17944885 (ZNF20) could be validated in multiple studies that analyzed blood or kidney samples.

高血糖与DKD的其他表型密切相关

- Methylation changes were mostly specific to the analyzed phenotypes.
- Glycemic control showed the greatest overlap with other DKD phenotypes.

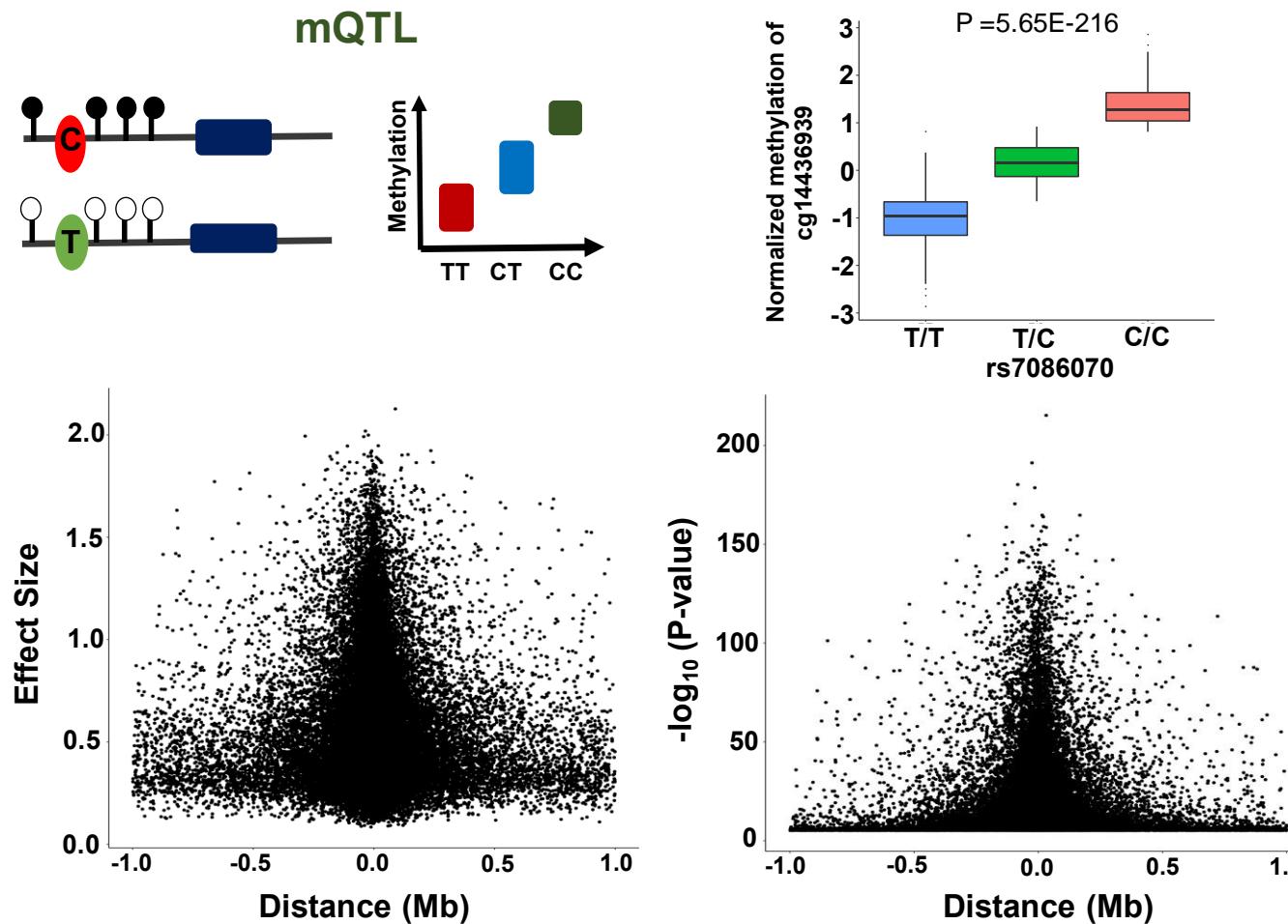


血液中与肾纤维化相关的甲基化位点

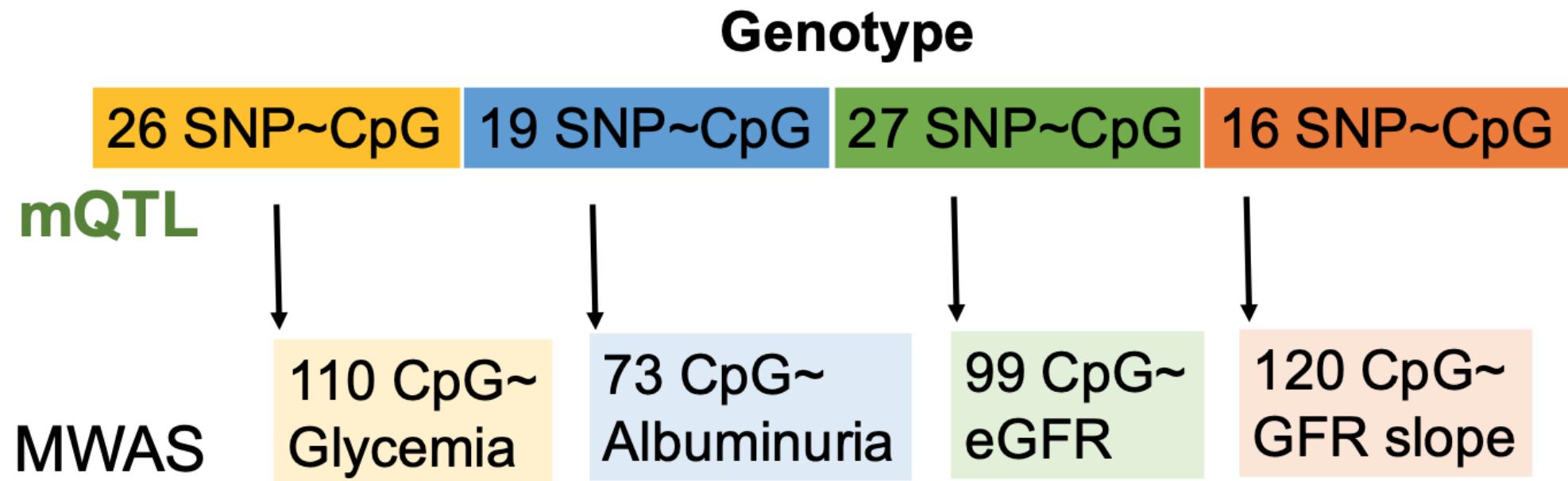


遗传因素→DNA甲基化

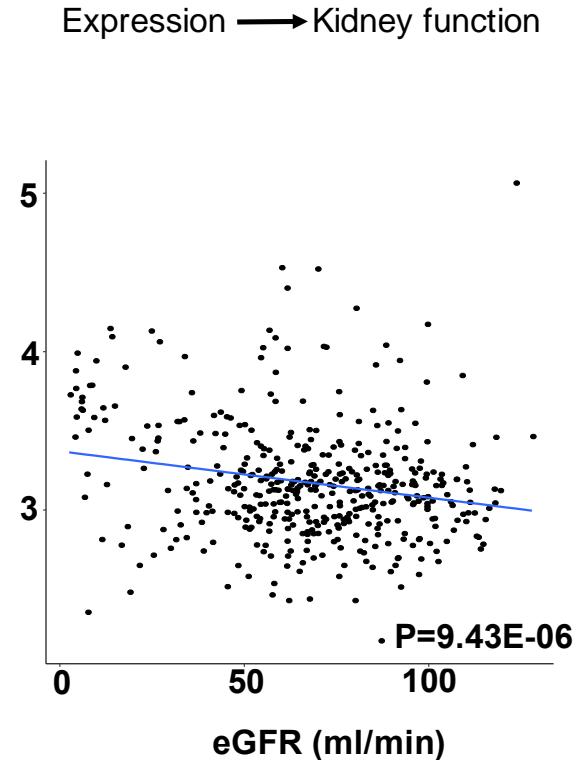
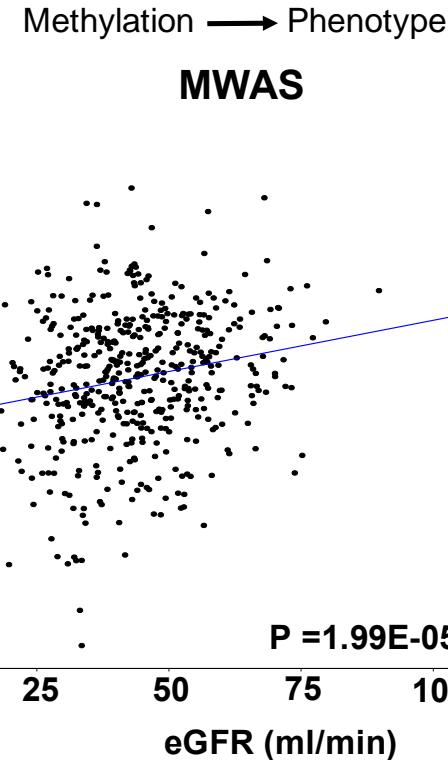
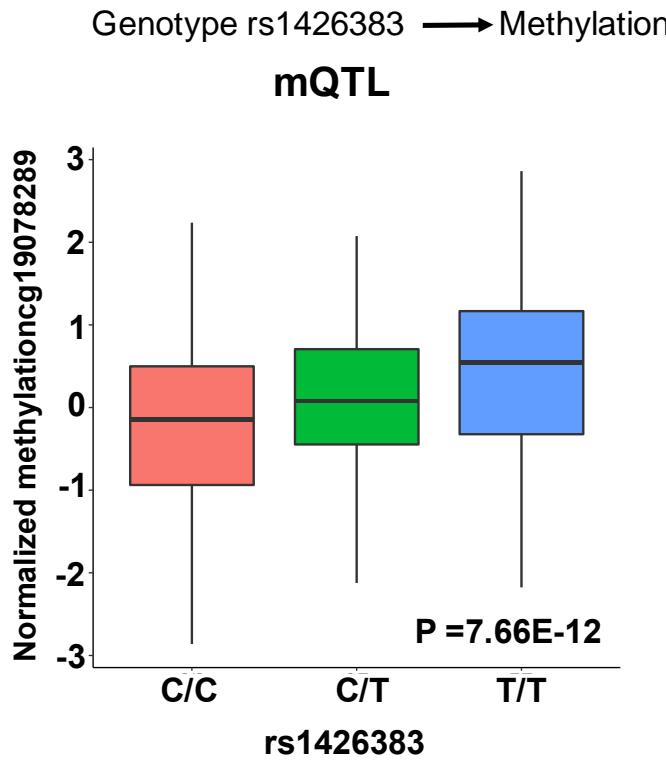
- 171,732 CpG sites were identified as significant mCpGs (CpG site that regulated by at least one SNP)



遗传因素→DNA甲基化→疾病表型



MBNL1: the muscleblind-like protein 1 locus



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贝叶斯共定位分析 (coloc)

The **posterior probability** of each possible configuration can be calculated:

H_0 : neither trait has a genetic association in the region

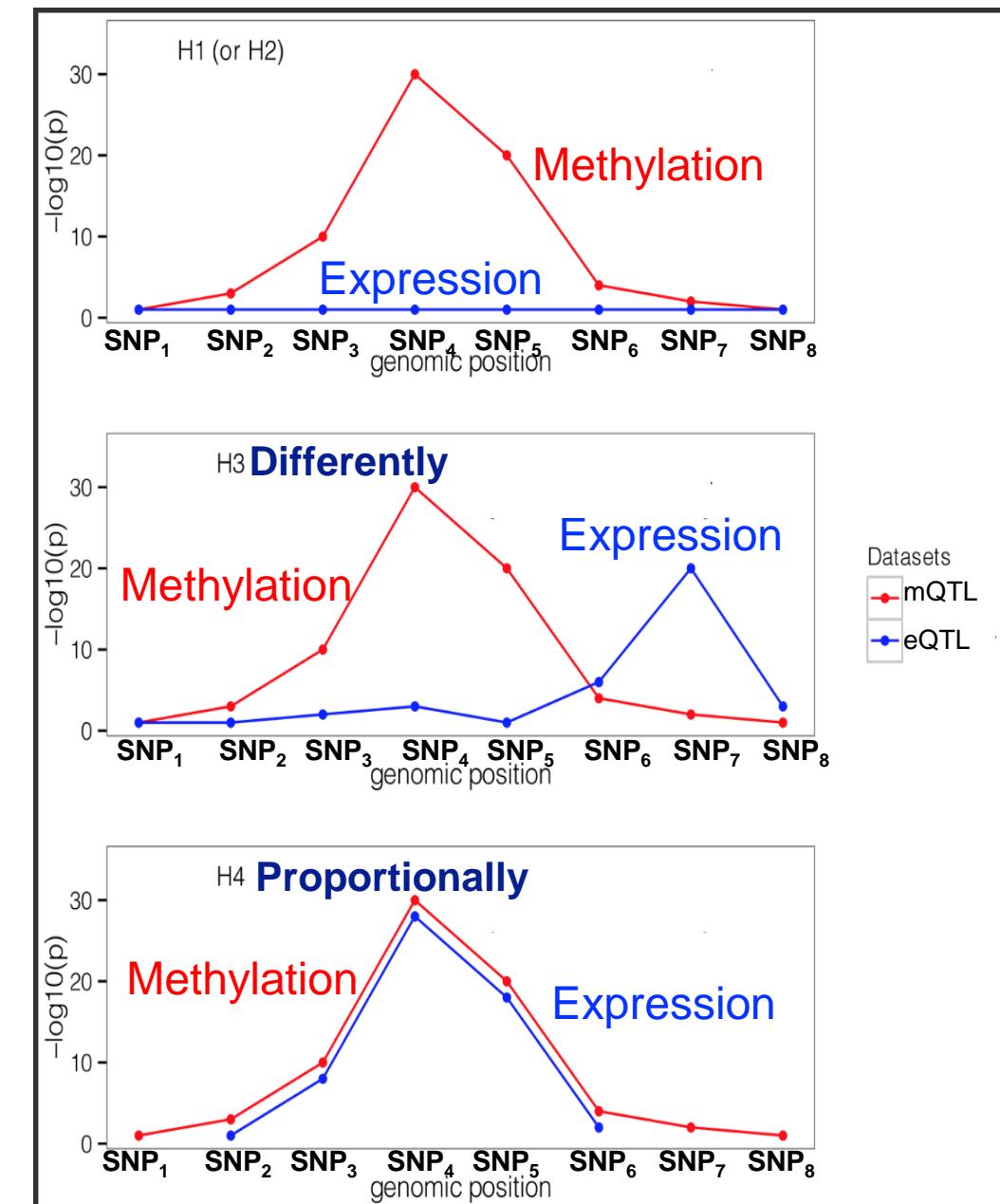
H_1 : only trait 1 has a genetic association in the region

H_2 : only trait 2 has a genetic association in the region

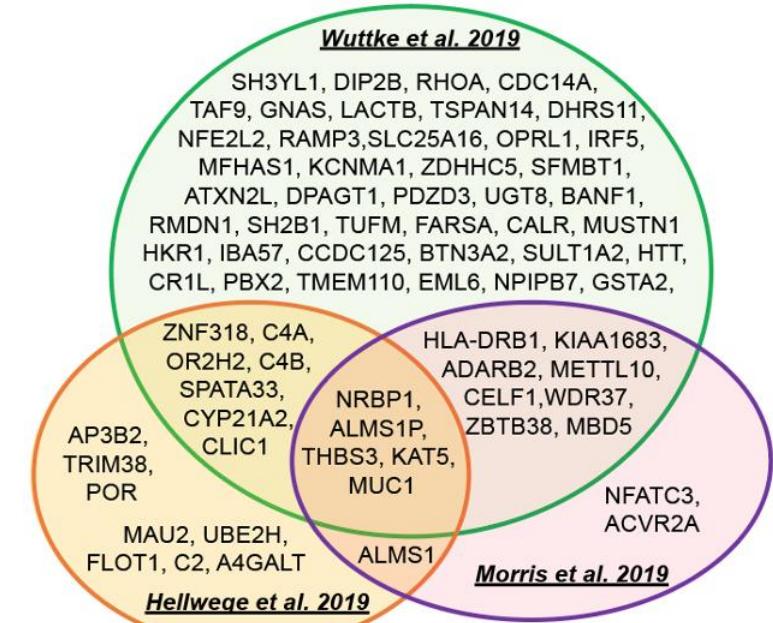
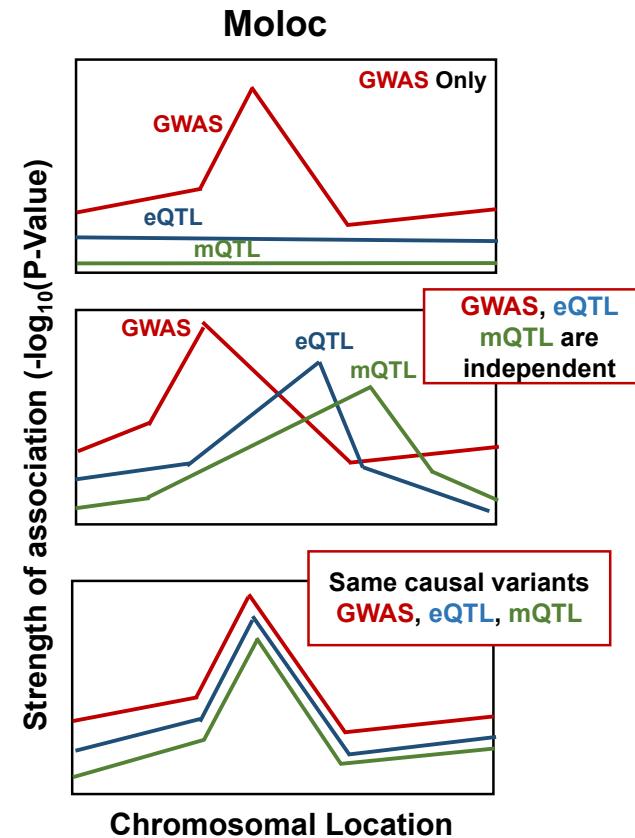
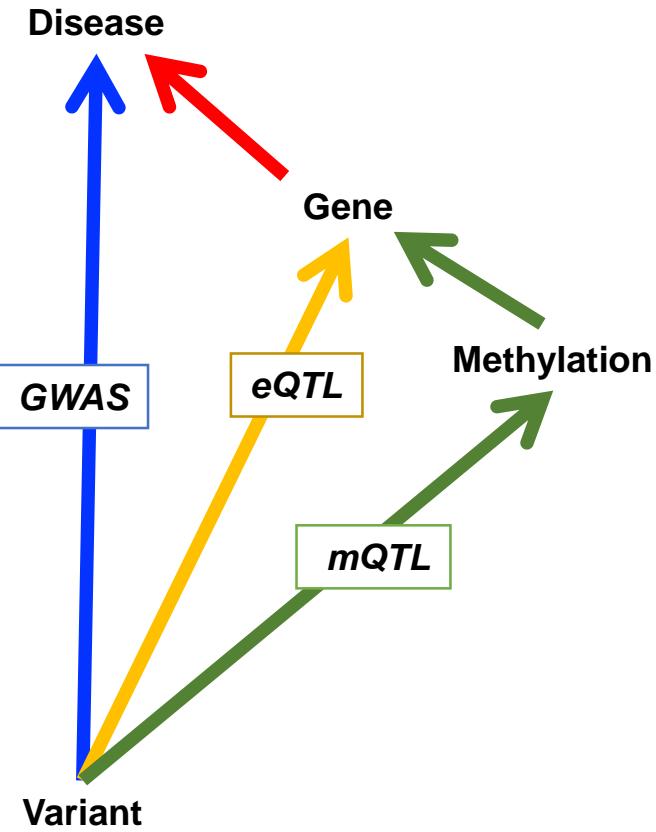
H_3 : both traits are associated, but with different causal variants

H_4 : both traits are associated and share a single causal variant

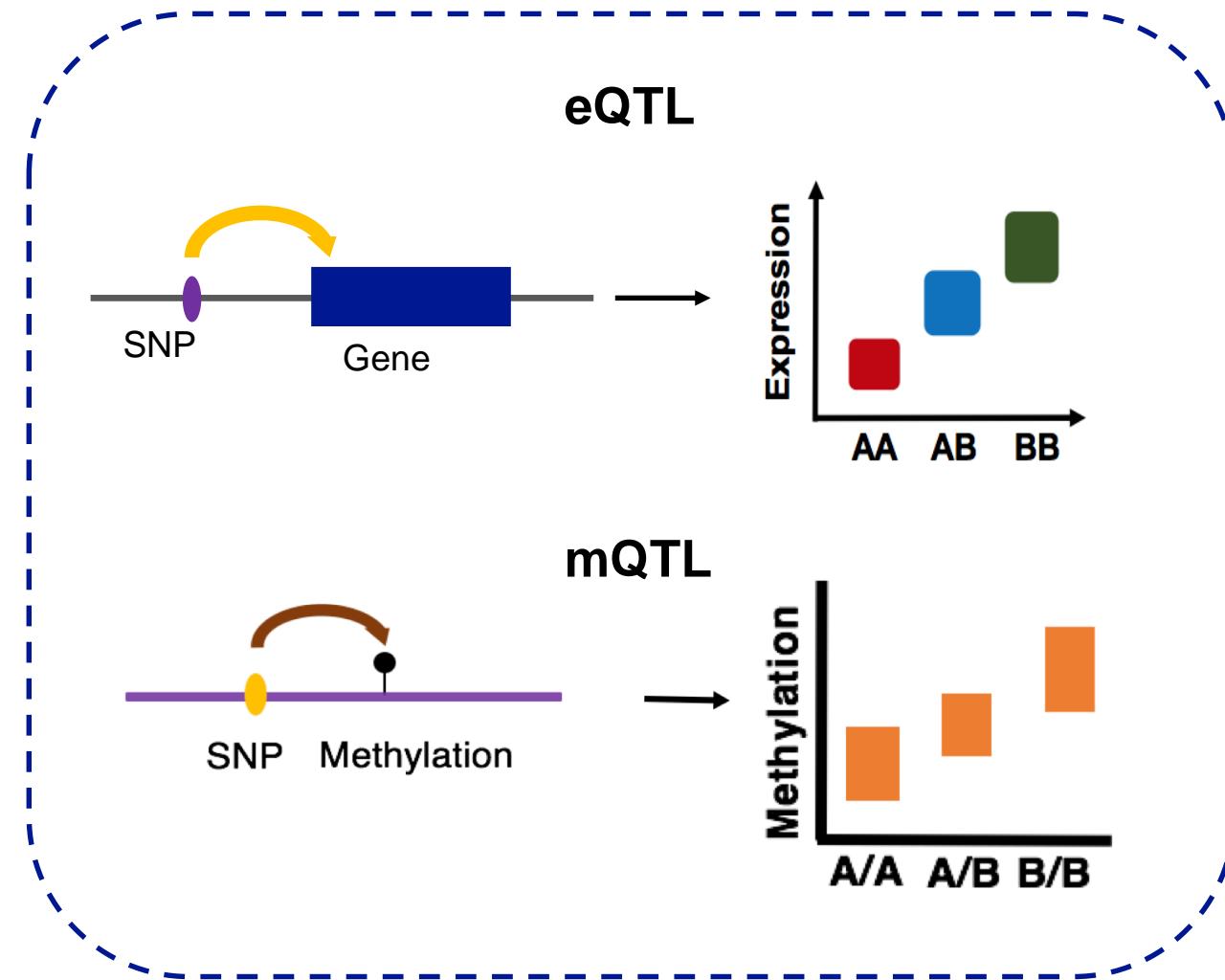
Usually use $PP4 > 0.8$ as cutoff to define co-localization.



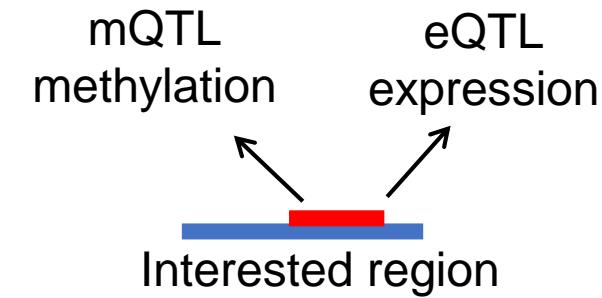
贝叶斯多性状共定位分析 (moloc)



调控因果关系推断?

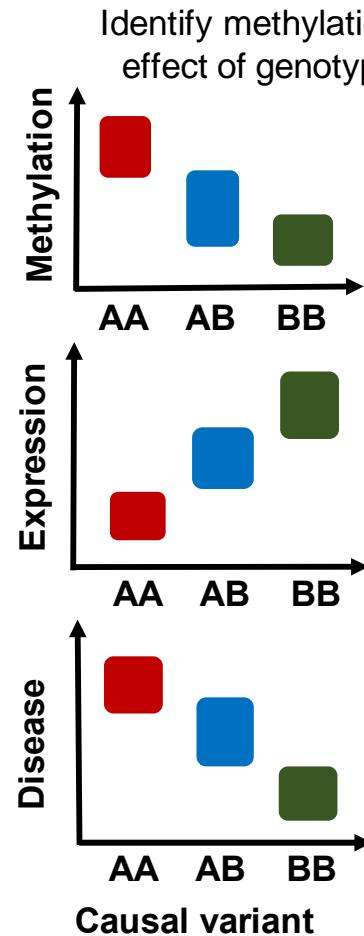
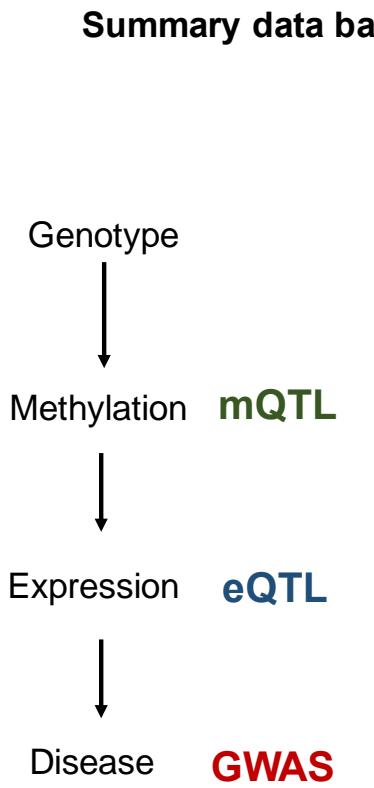


Are methylation and gene expression linked?

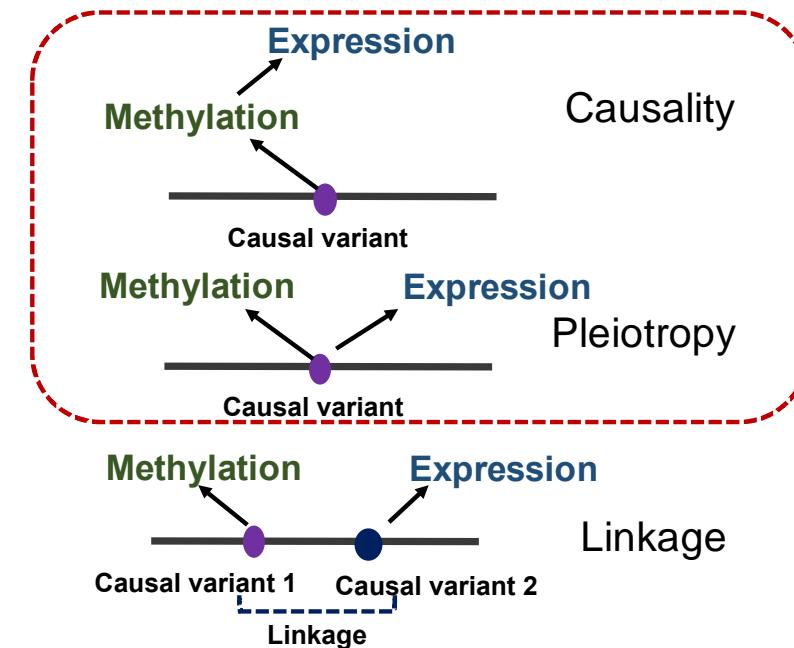


Genetic variants can influence gene expression (eQTL) and DNA methylation (mQTL).

因果推断：孟德尔随机化 (MR)



Pleiotropic association



Mendelian Randomization

Basic rules:

Here Z is an instrumental variable

$$\text{Exp}_{\text{SNP}} = \alpha_1 + \beta_1 Z + \gamma_1 U$$

If Z associated with X, and X has a causal relationship with Y (the effect of X on Y is causal)

$$\begin{aligned} \text{Trait} \quad & \text{Exp}_{\text{SNP}} \\ Y &= \alpha_2 + \beta_2 X + \gamma_2 U \\ &= \alpha_2 + \beta_2 (\alpha_1 + \beta_1 Z + \gamma_1 U) + \gamma_2 U \\ &= \alpha_2 + \beta_2 \alpha_1 + \beta_2 \beta_1 Z + (\beta_2 \gamma_1 + \gamma_2) U \\ &= \alpha_3 + \beta_3 Z + \gamma_3 U \end{aligned}$$

Z should associated with Y, if the effect of X on Y is causal.



$$\begin{aligned} \text{Beta1} &= b_{zx} \\ \text{Beta2} &= b_{xy} \\ \text{Beta3} &= b_{zy} \\ \text{Bzy} &= b_{xy} * b_{zx} \end{aligned}$$

The slope between Z and Y should be

$$\beta_3 = \beta_2 \beta_1$$

Is there slope significant? (the real relationship significantly existing?)

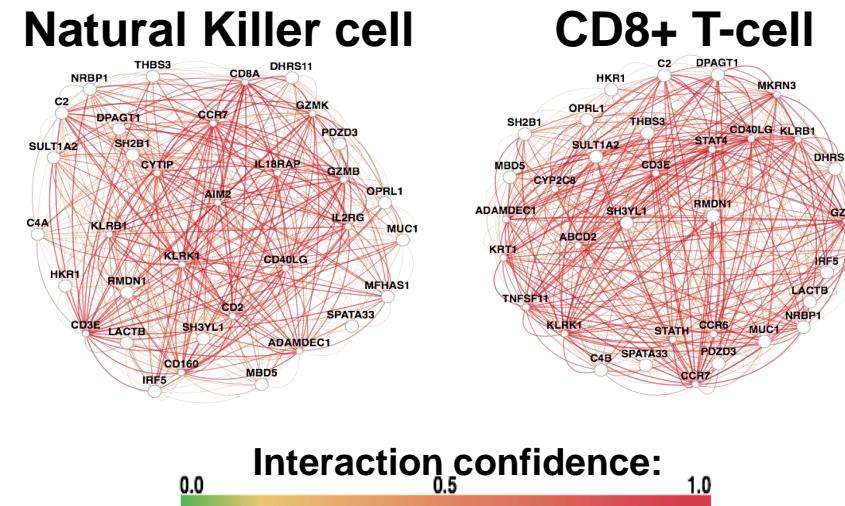
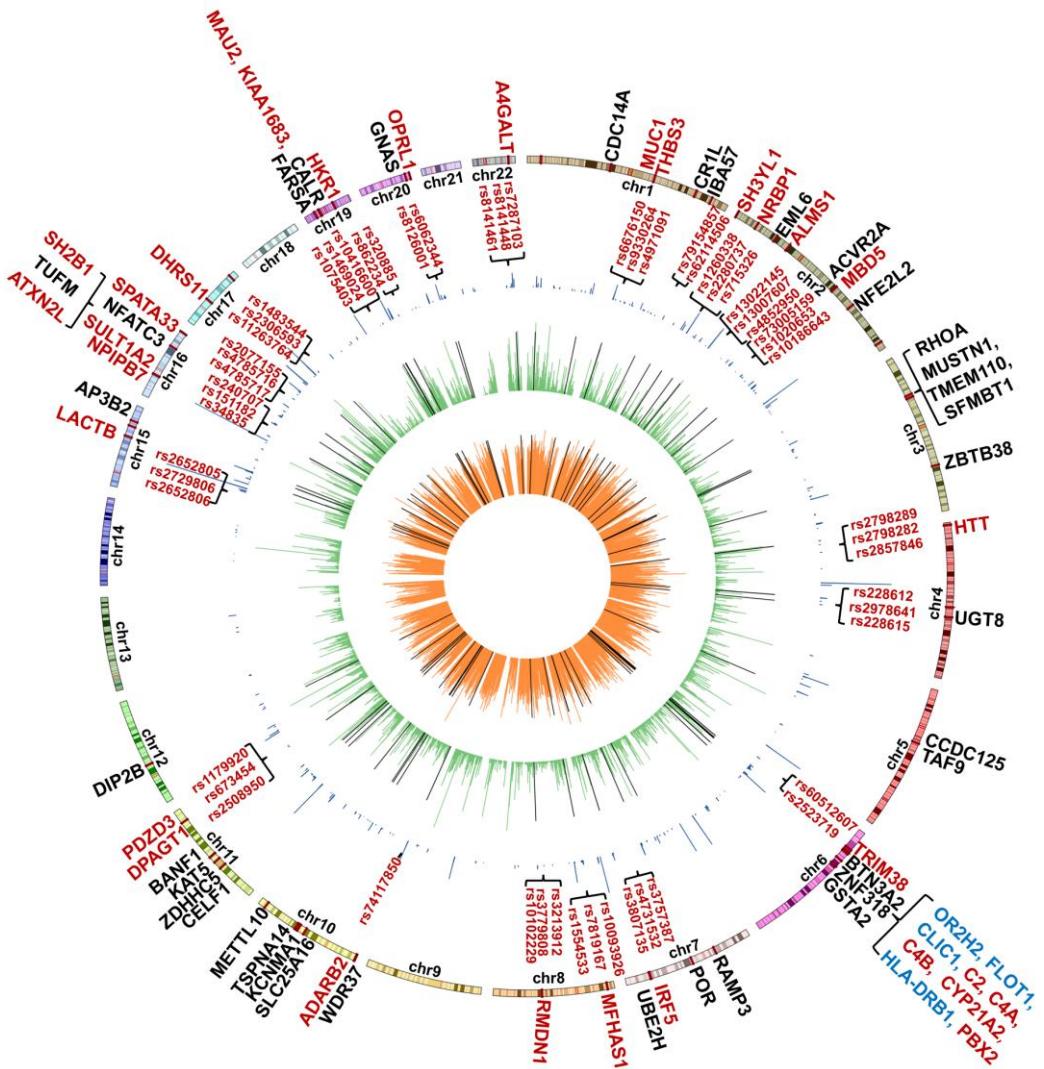


Hence, the effect of X on Y is causal

The relationship between Z and Y should be indirect. That is Z affects Y through X.

Z is the instrumental variable and Z represent the genetic variant or a combination of genetic variants. Because the fact that the alleles of a genetic variant are inherited randomly from parents to offspring, so that the relation of a genetic variant with a phenotype should not be confounded.

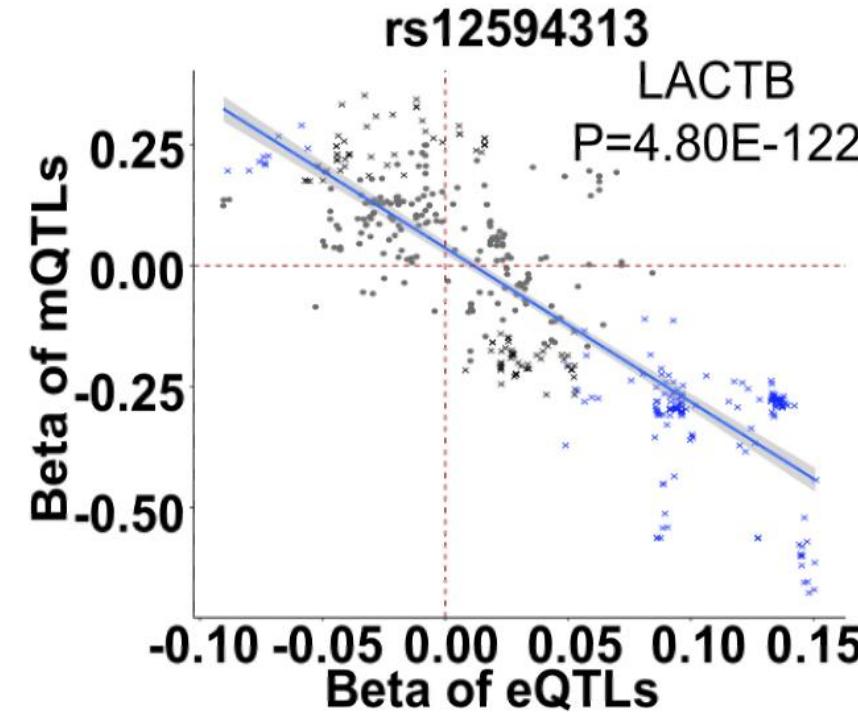
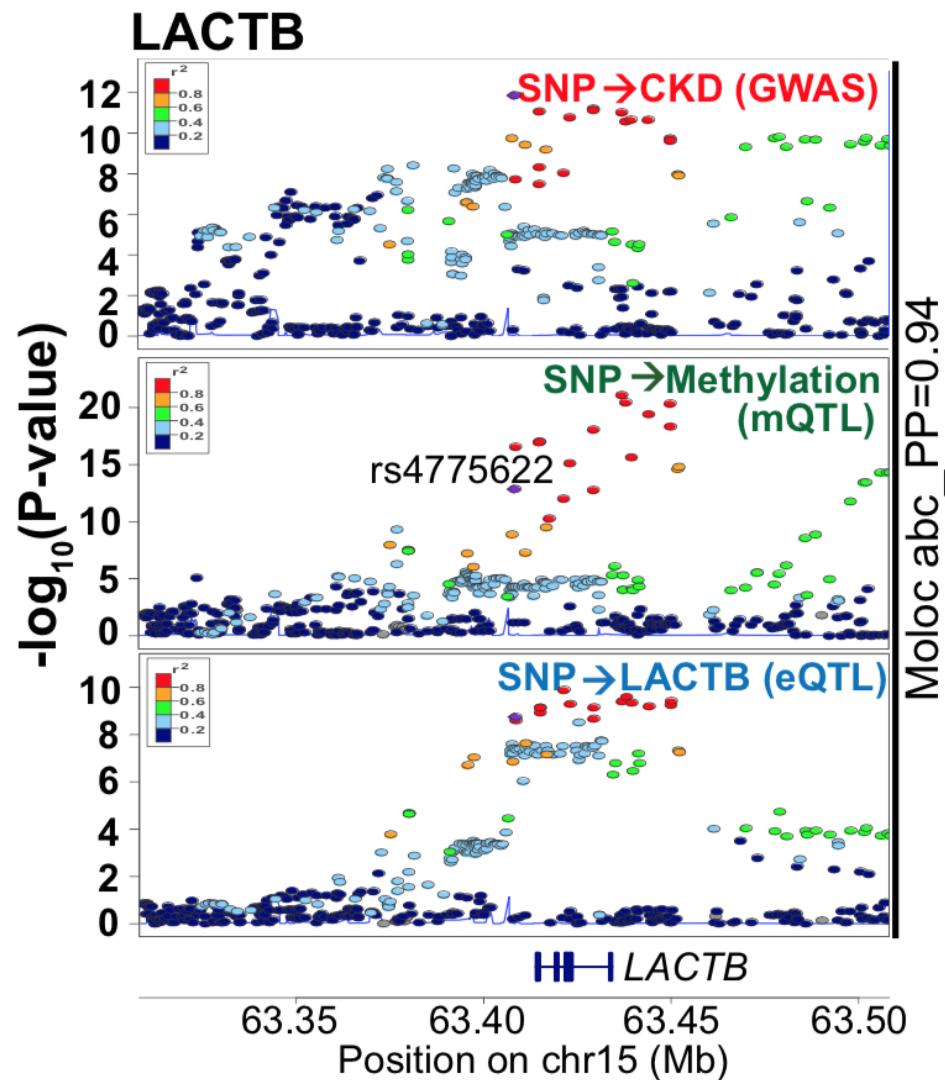
DKD致病基因鉴定



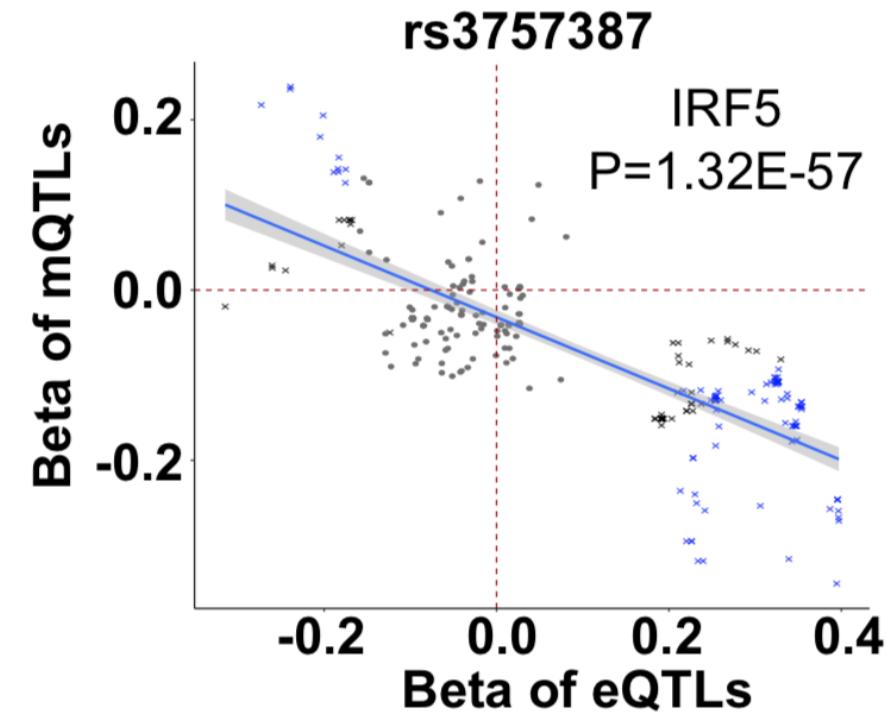
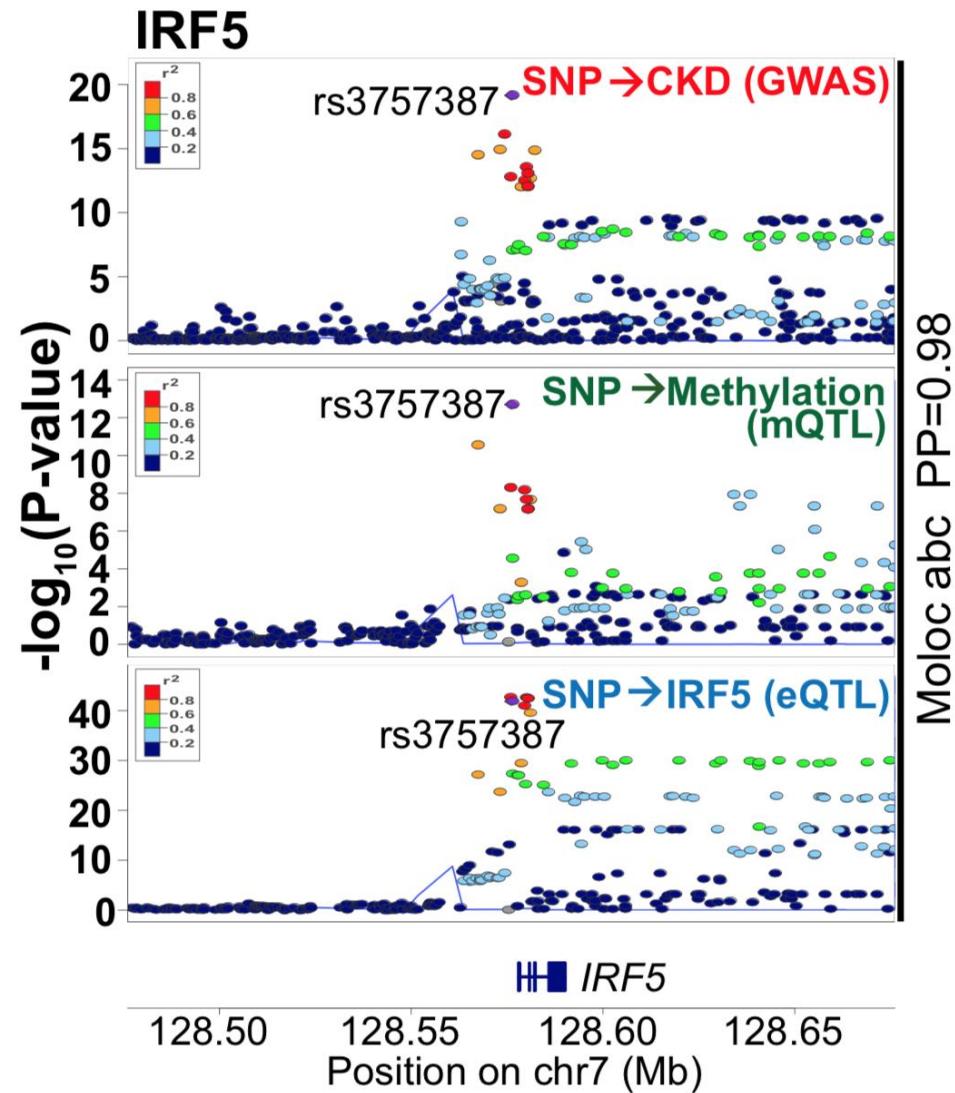
Enriched GO terms (BP):

Apoptotic cell clearance,
complement deficiency, immune
effector process etc.

LACTB: the Serine β -lactamase-like protein

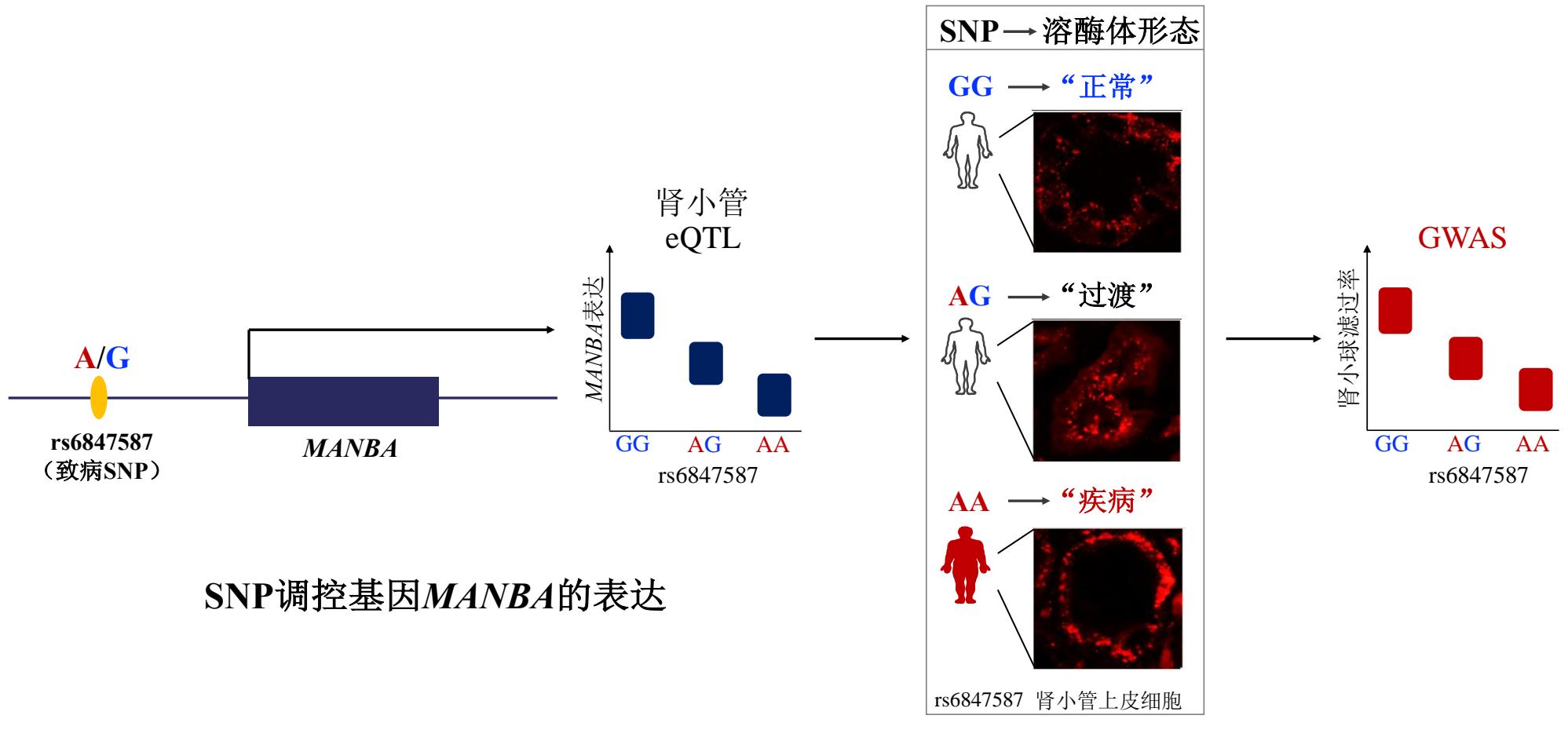


IRF5: an IFN-responsive transcription factor



SNP → 基因表达 → 疾病表型

SNP影响慢性肾脏病疾病表型（肾小球滤过率）的机制



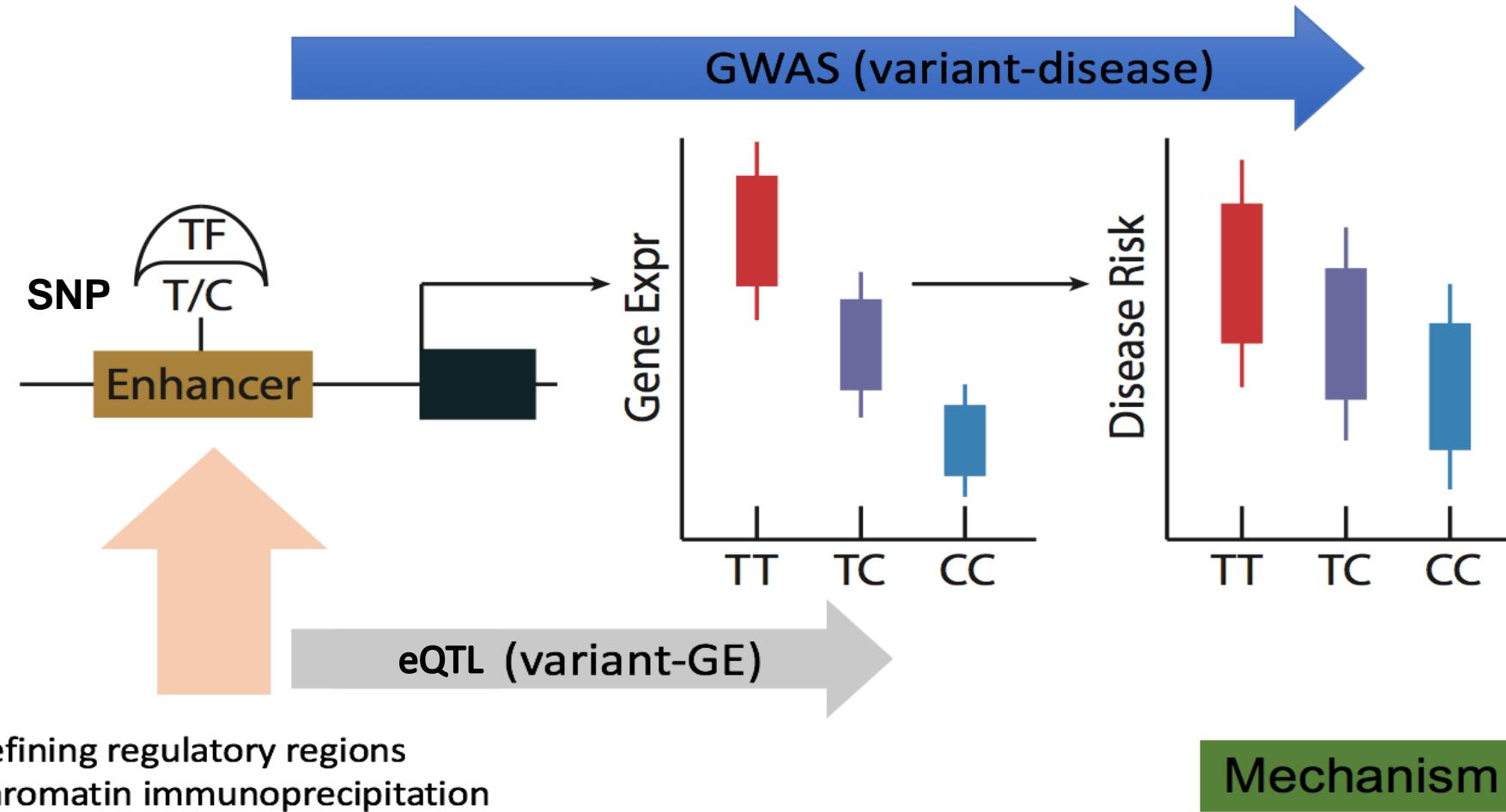
小结

- 基因组上单核苷酸位点的遗传变异（SNP）可能会引起其邻近区域的甲基化水平、染色质开放程度，以及基因表达等数量性状的变化。
- 以GWAS数据为指导，对数量性状位点（xQTL）的多组学整合分析，将有利于识别疾病的潜在致病基因，并解析疾病的发病机制。
- 贝叶斯共定位分析、孟德尔随机化方法是整合多组学分析的常用方法。

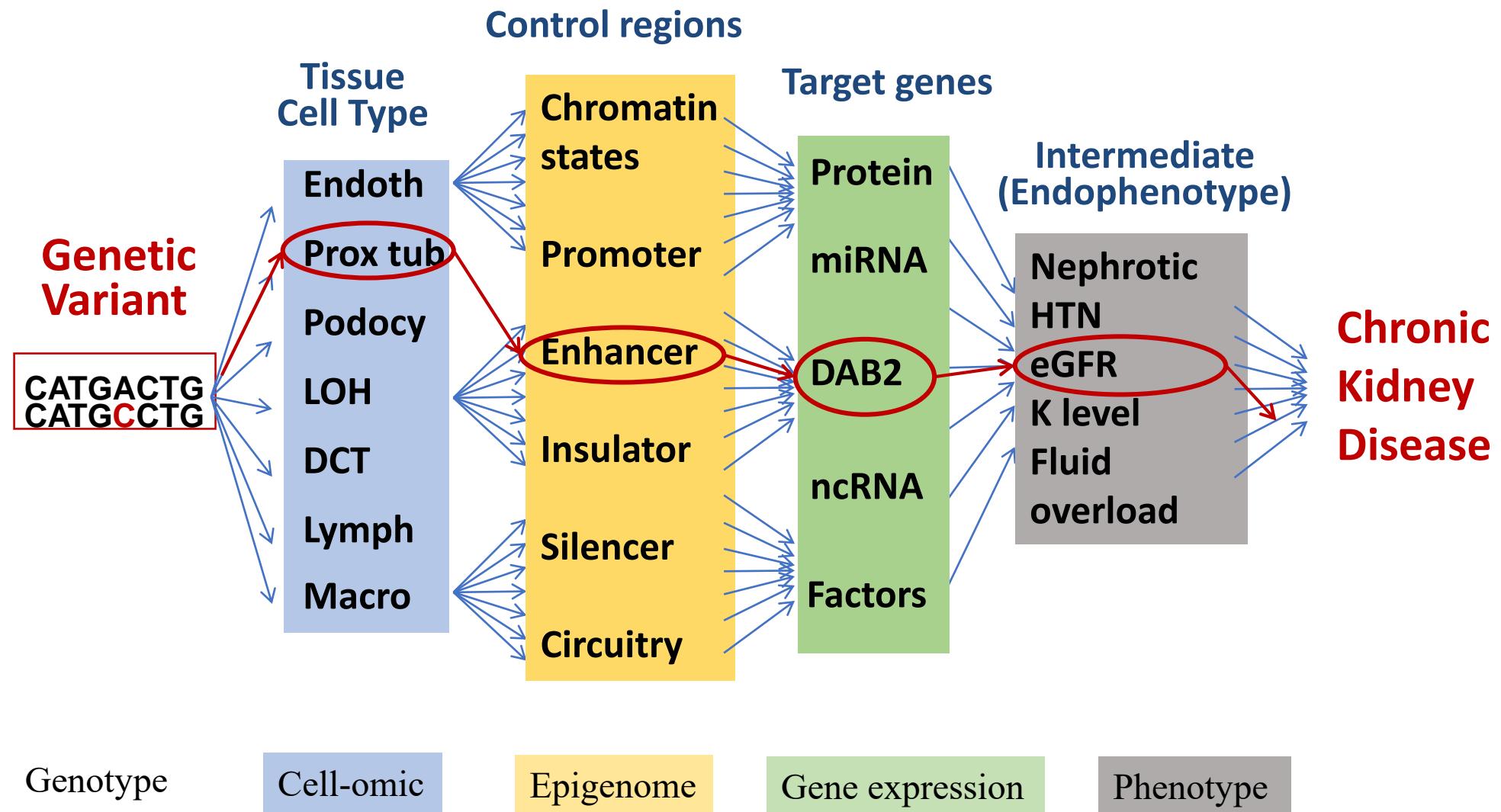
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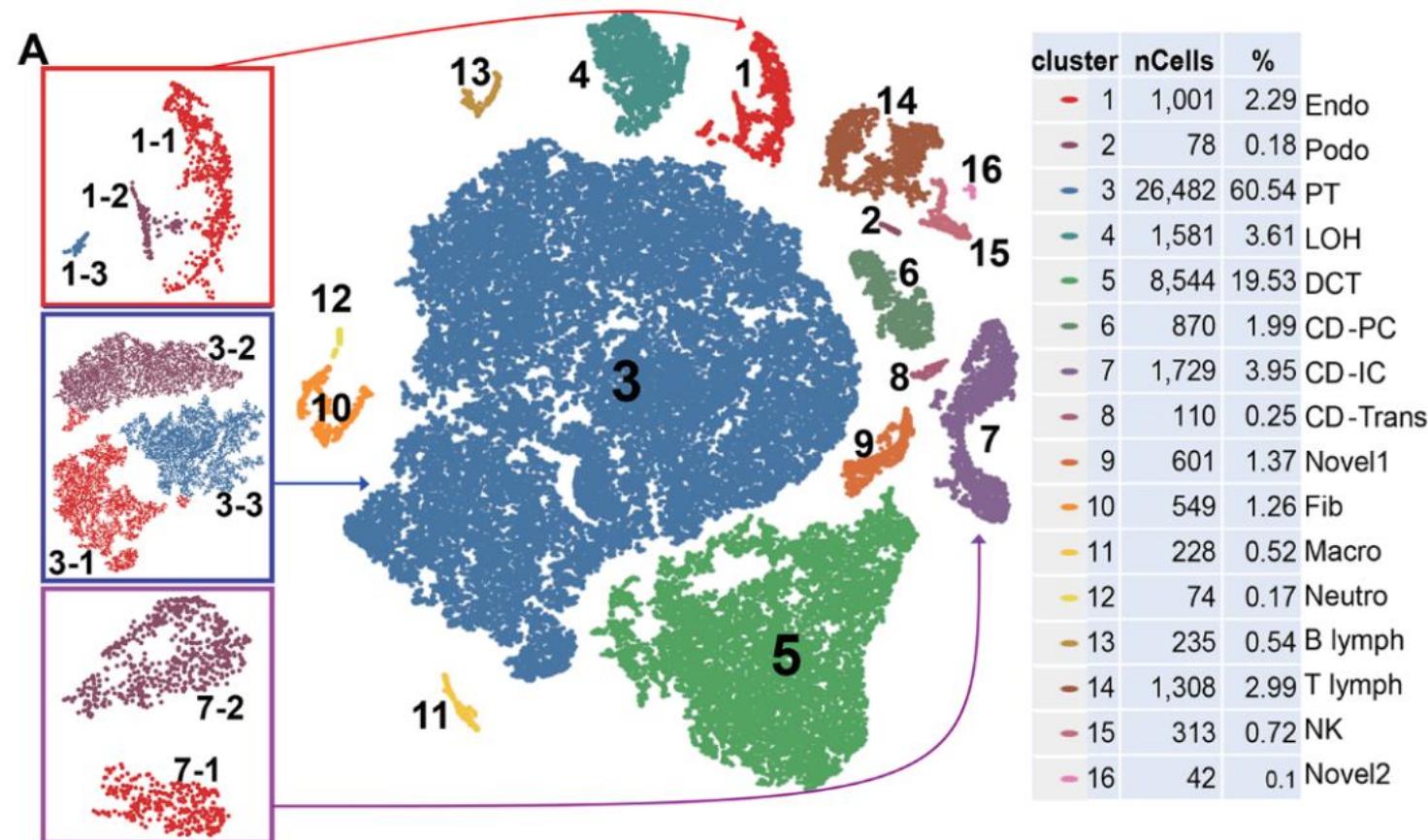
多组学整合分析：解析遗传变异的调控机制



多组学整合分析：精准定位遗传因素导致CKD的具体细胞

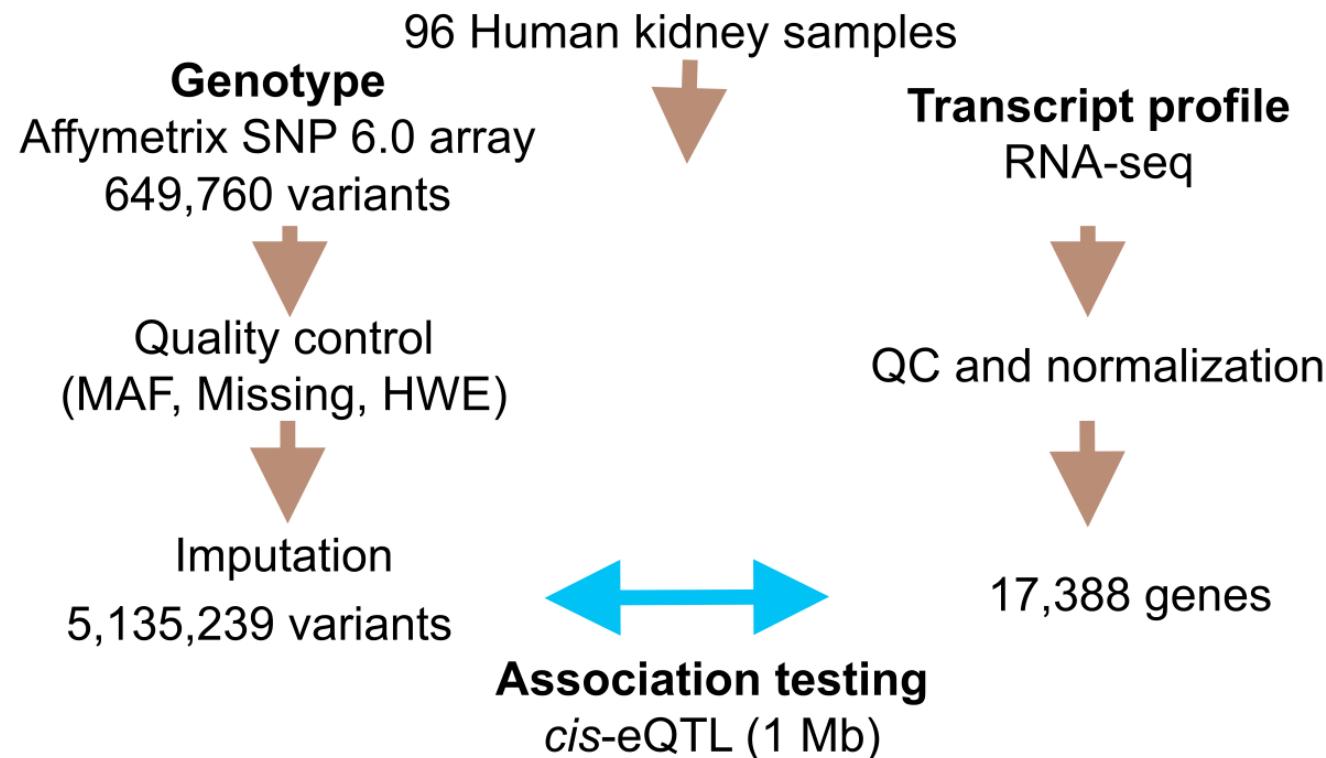


肾脏中发现新的细胞类型

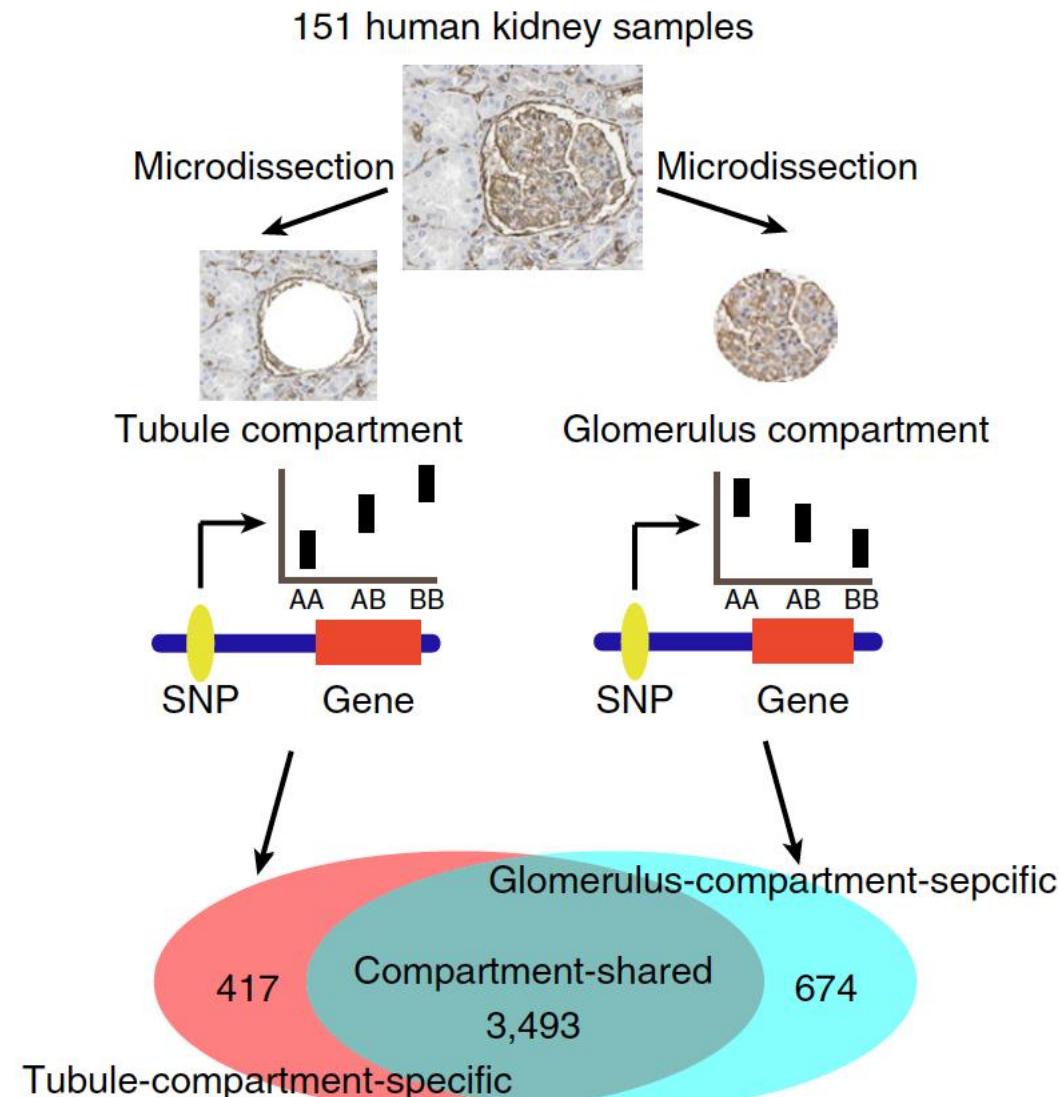


整个肾脏 (Whole Kidney)

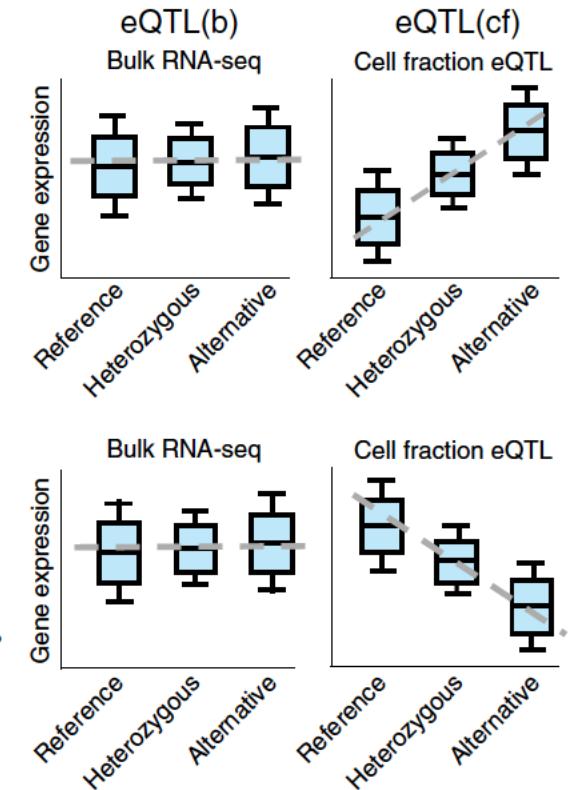
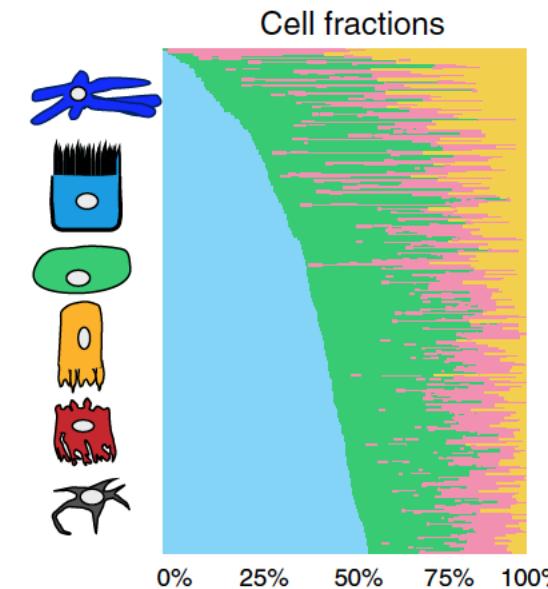
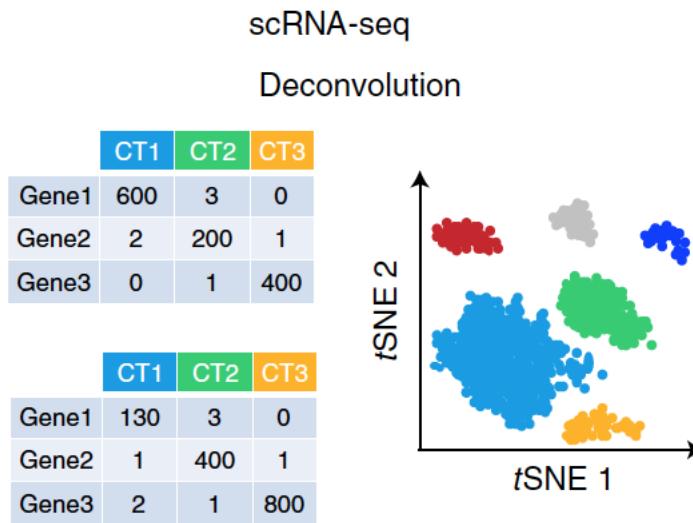
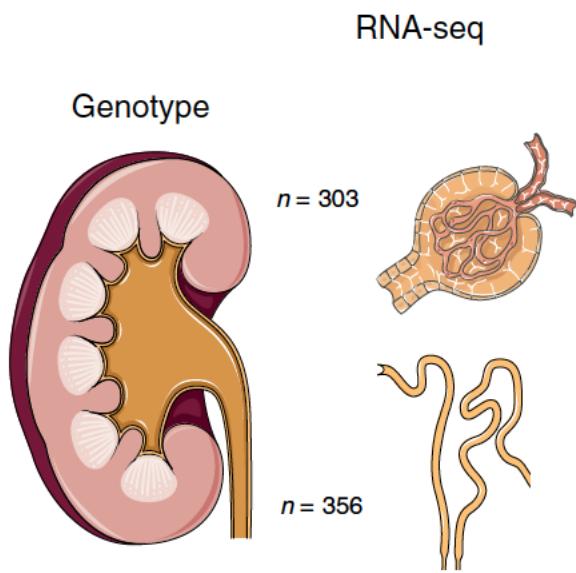
Only 9 putative causal genes were identified.



显微切割：肾小球与肾小管(Glom and Tubule)



考慮个体细胞组分的差异



eQTL的计算方式

(a) 常用计算方法:

线性回归: $\text{lm}(\text{Gene Exp} \sim \text{SNP dosage} + \text{Confounders})$

Confounders: age, gender, genetic PCs, and PEER factors (estimated hidden confounders)

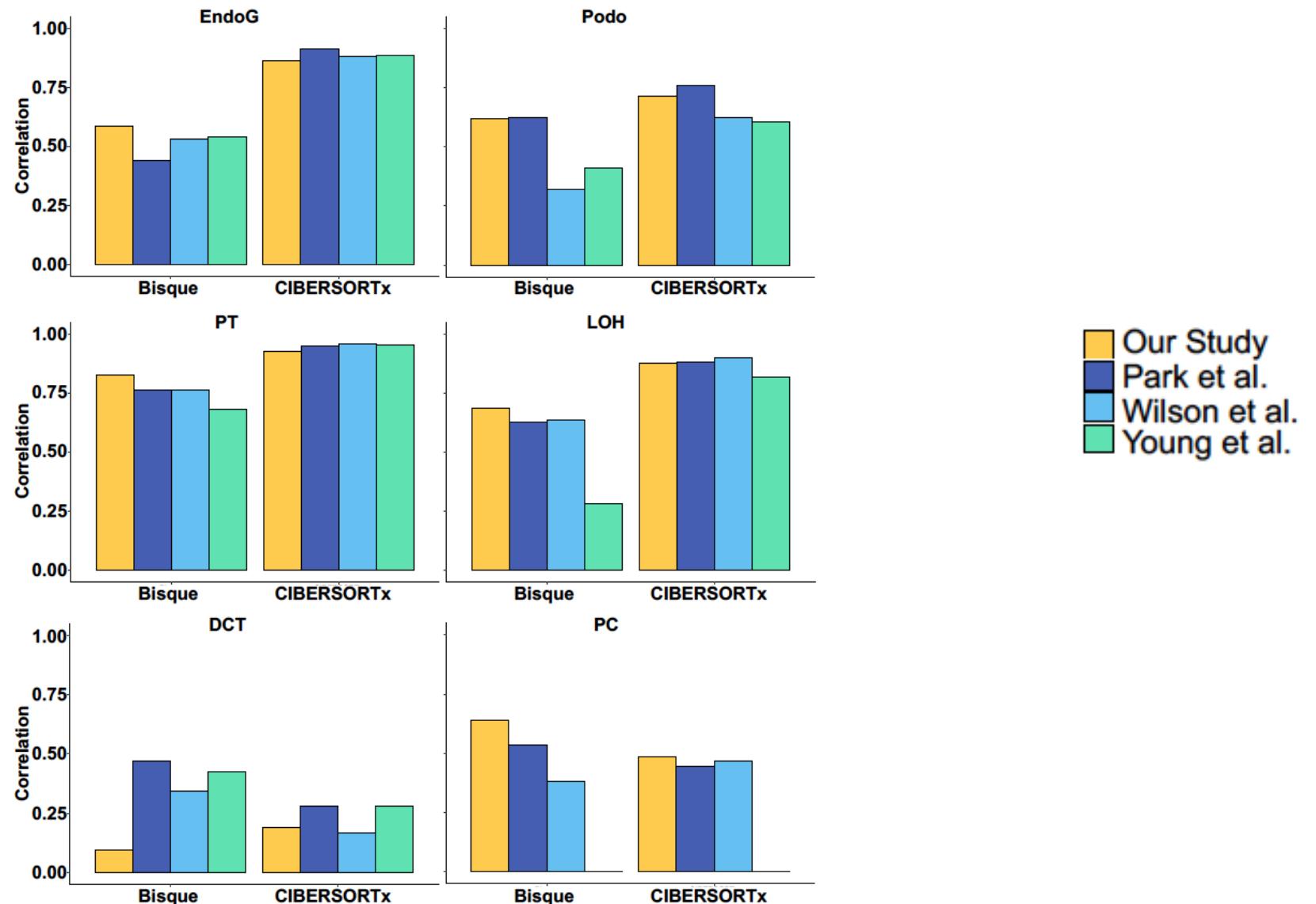
(b) Cell fraction eQTLs:

线性回归: $\text{lm}(\text{Gene Exp} \sim \text{SNP dosage} + \text{Confounders})$

Confounders: age, gender, genetic PCs, Cell fractions, and PEER factors (estimated hidden confounders)

可能存在的问题: 相似细胞之间的共线性问题, 细胞组分较少的细胞类型

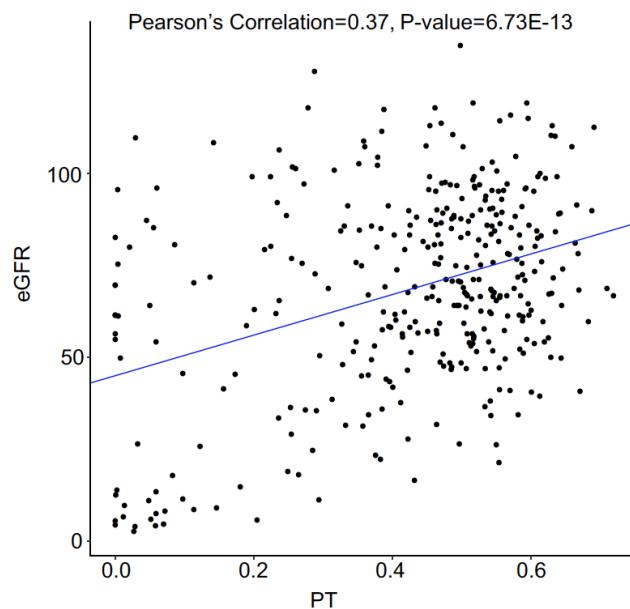
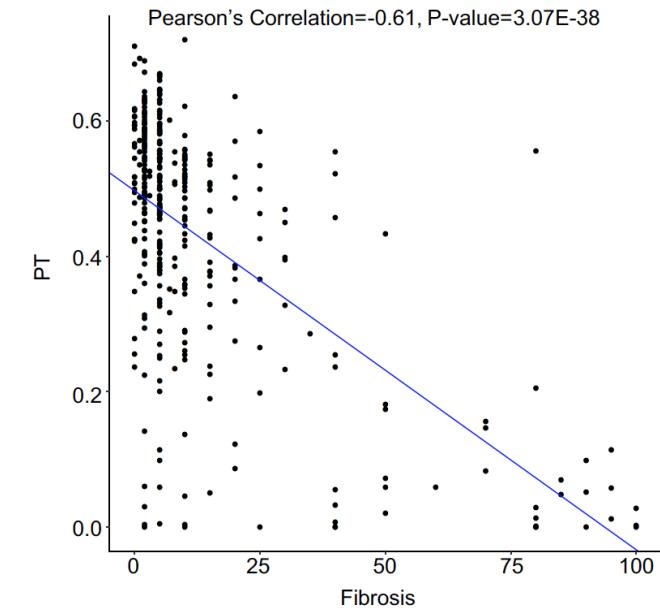
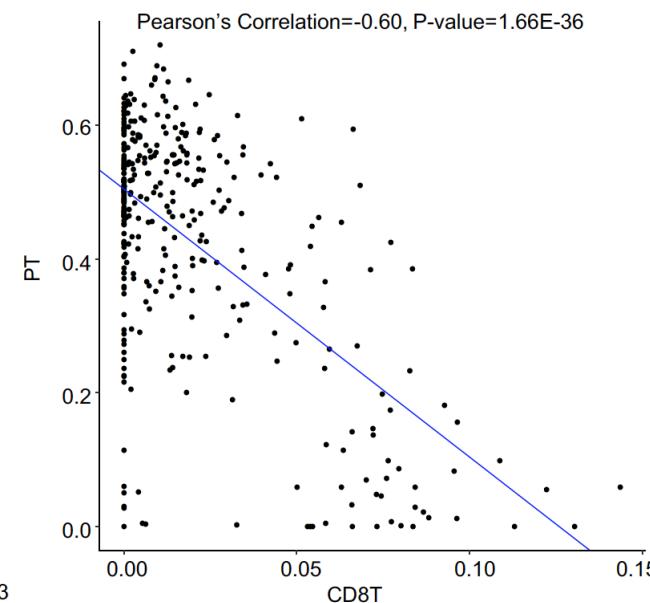
反卷积算法比较



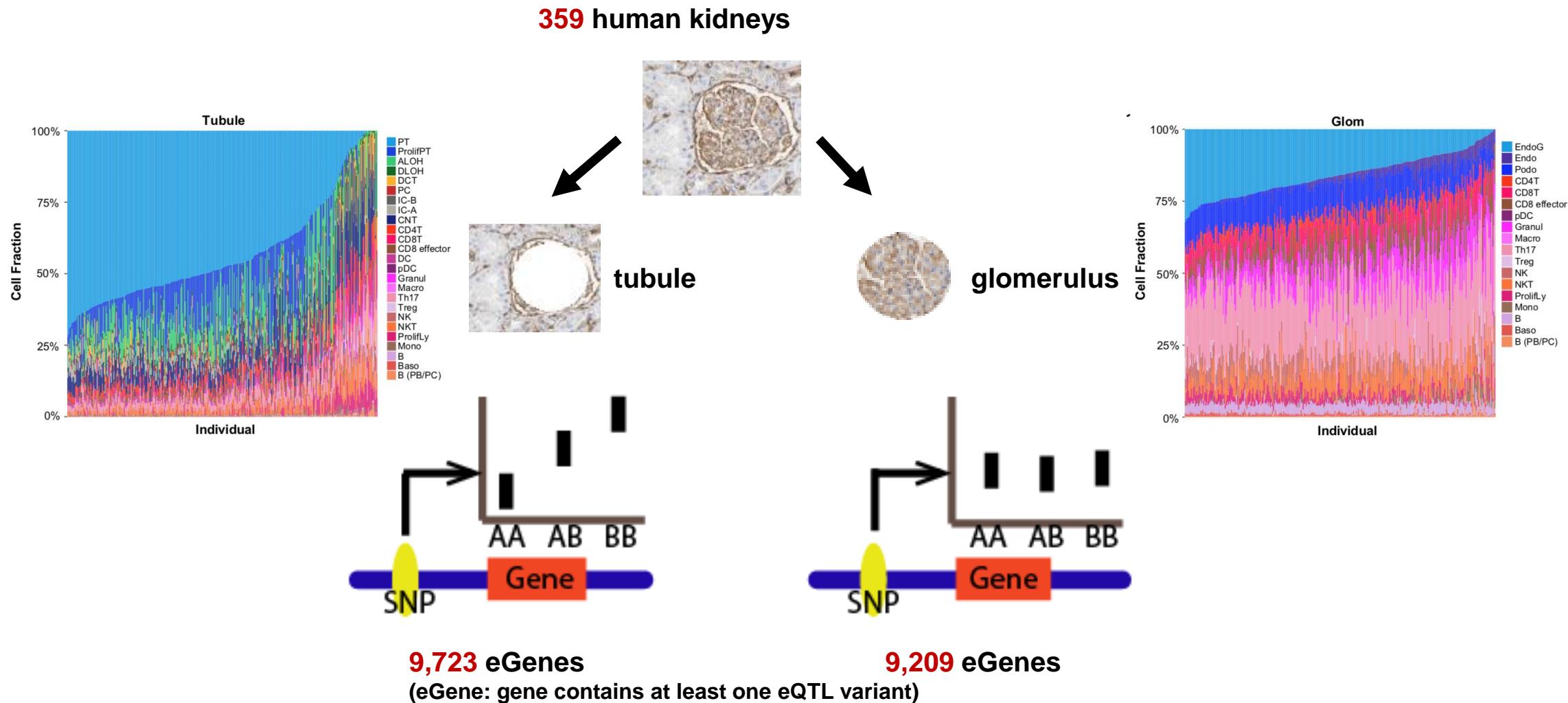
个体细胞组分随疾病状态改变

Cell Type	Pearson's Correlation	Two-sided P-value
CD8T	-0.60	1.66E-36
Macro	-0.68	1.25E-48
Treg	-0.38	2.62E-13
Mono	-0.33	1.19E-10
B	-0.41	9.57E-16

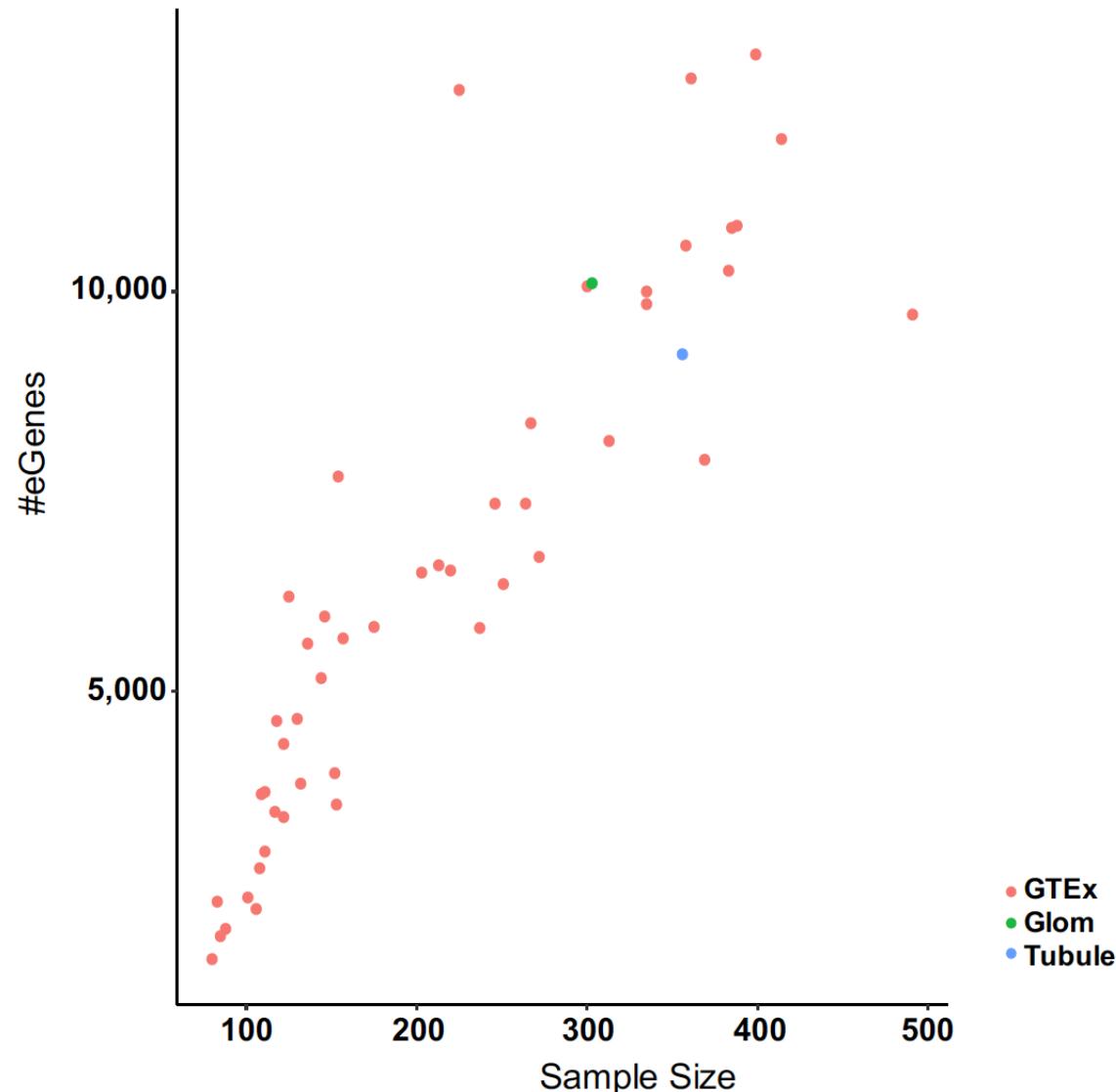
Two-sided P-value was calculated by t-test (df=354).



考慮个体细胞组分的差异

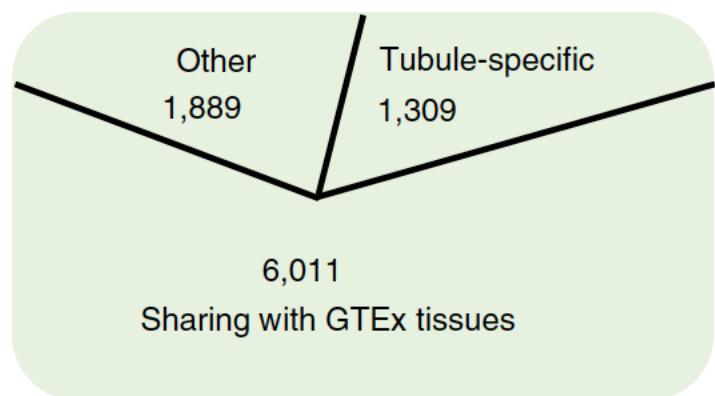
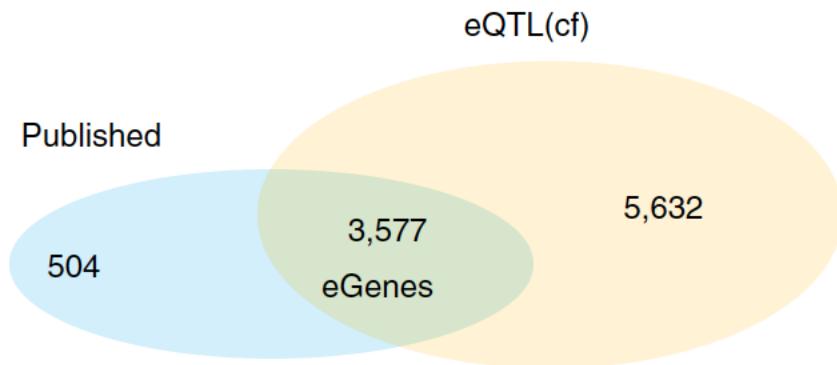


样本数量与eGene个数的线性关系

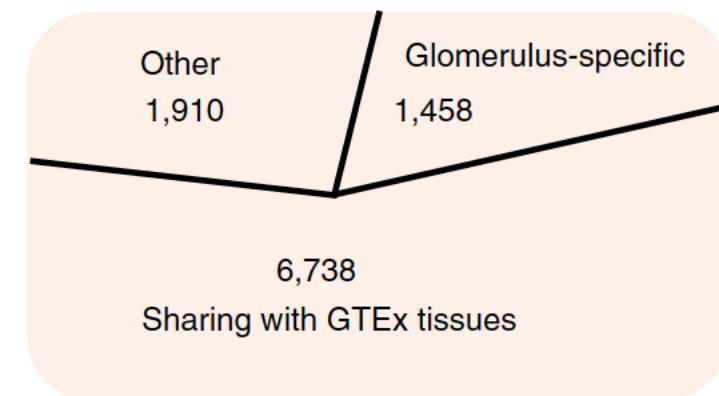
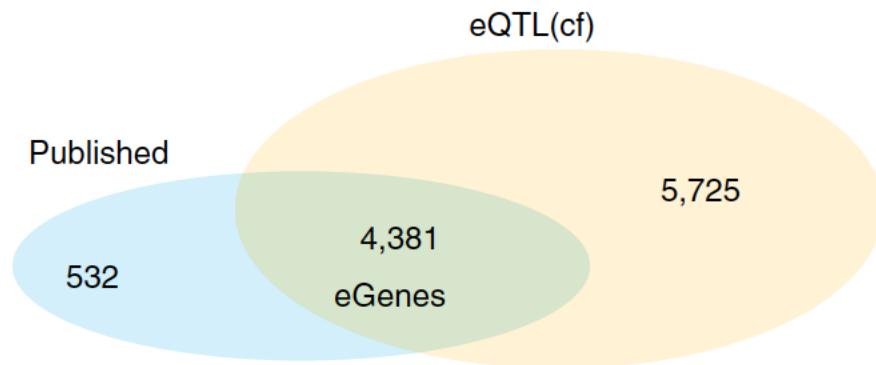


新鉴定的SNP-Gene调控

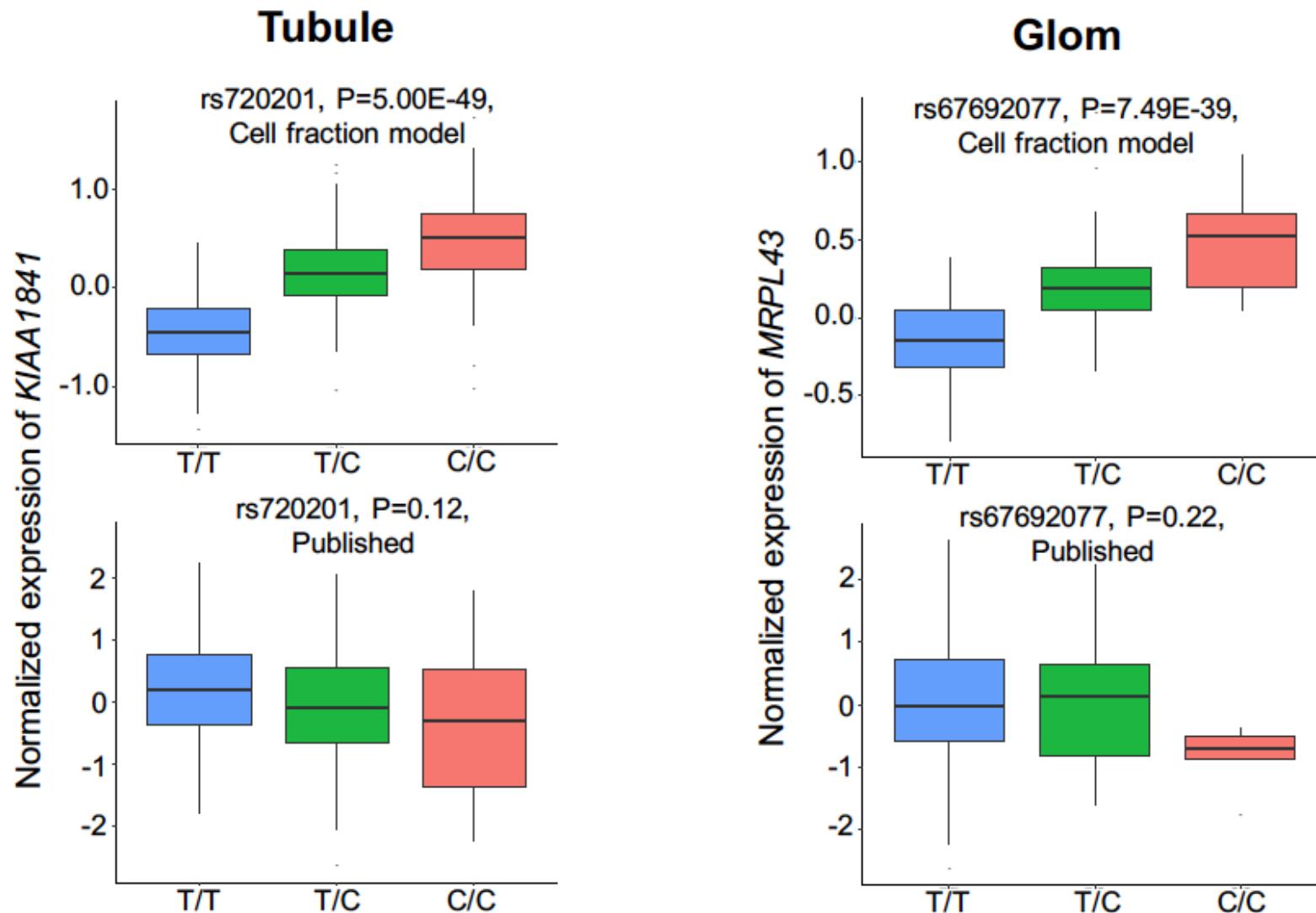
Tubule



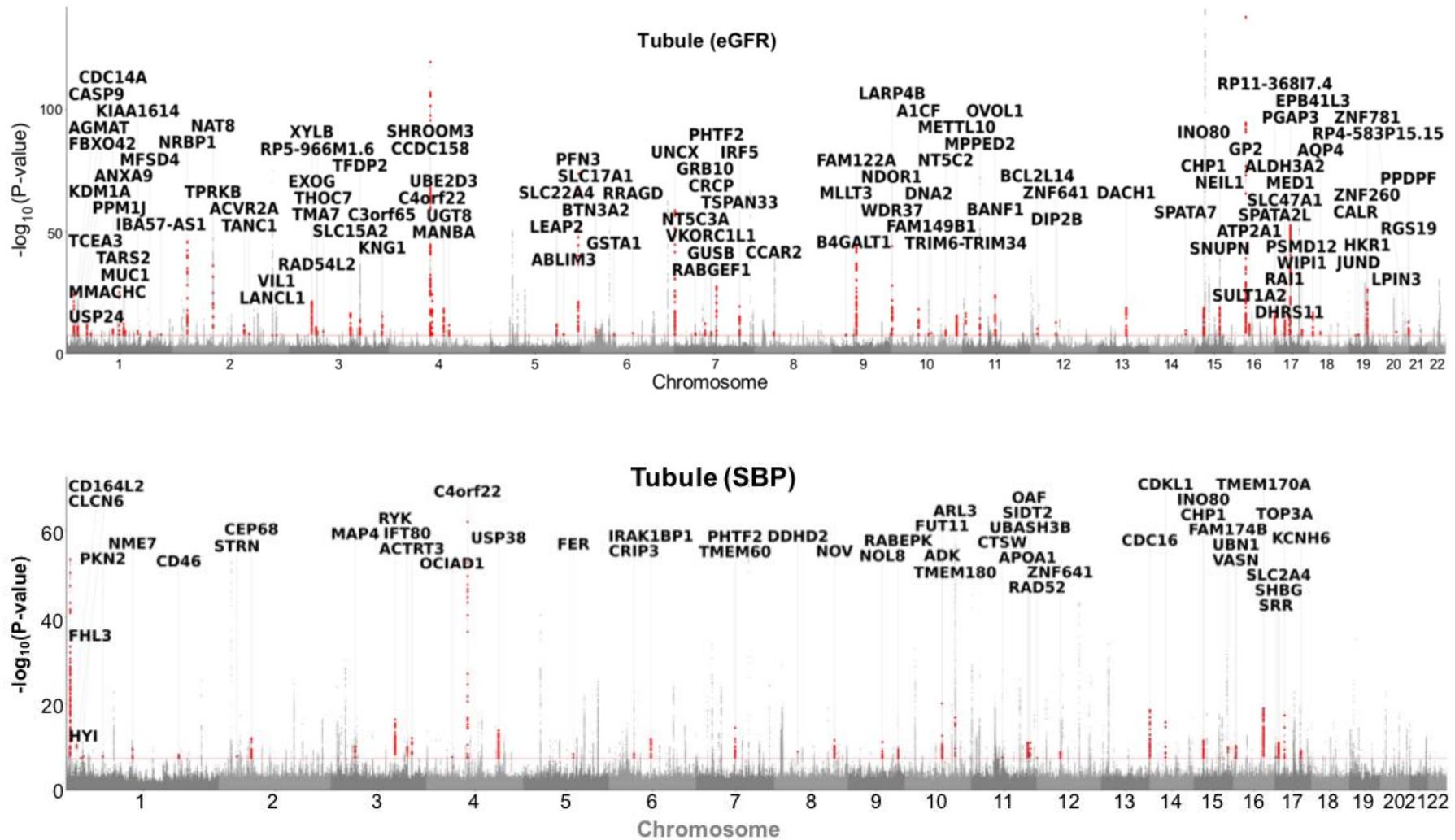
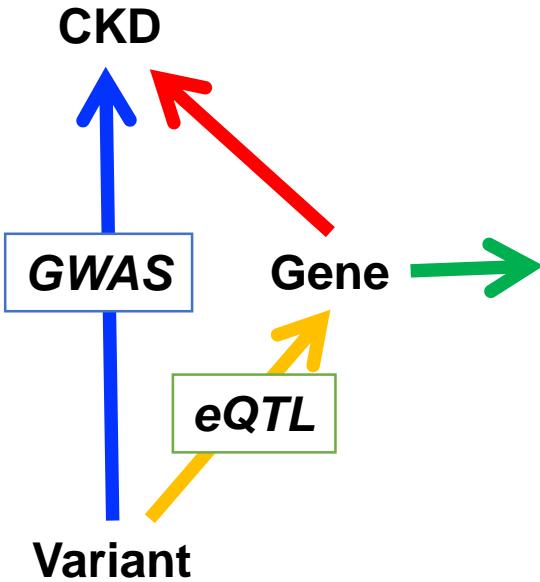
Glom



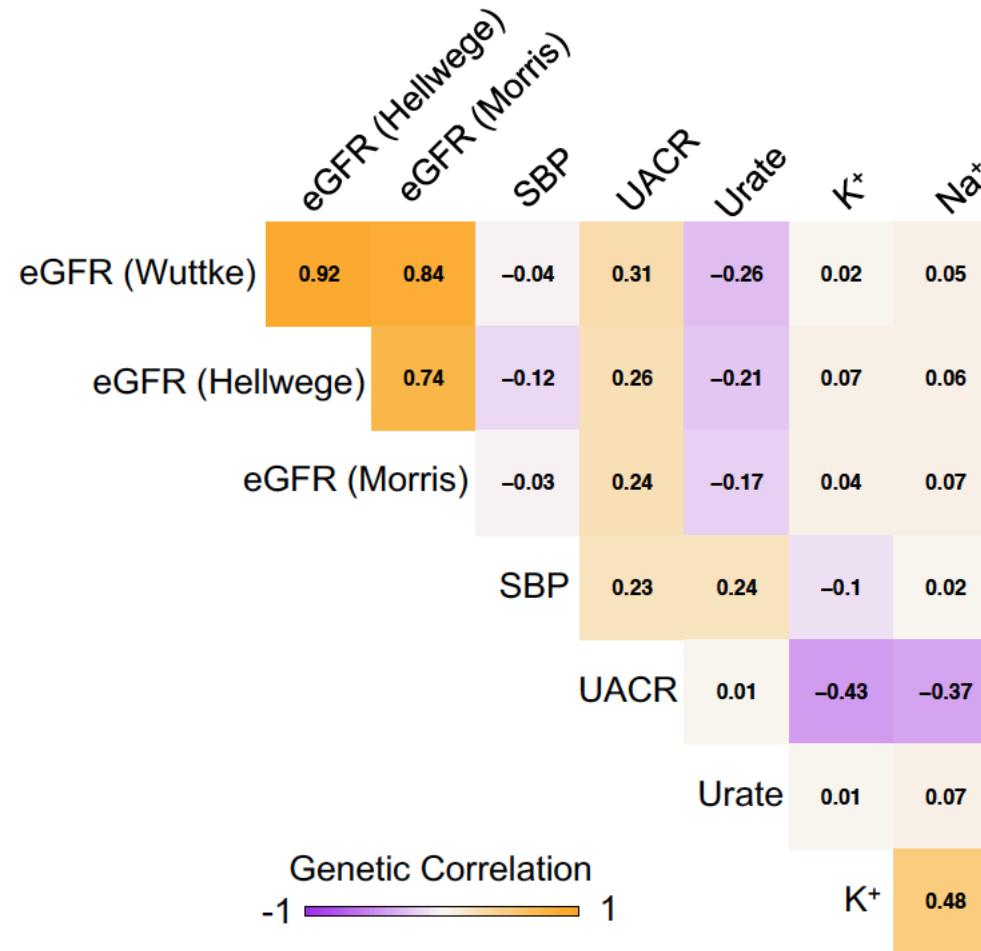
Cell fraction eQTLs



CKD和高血压的潜在致病基因



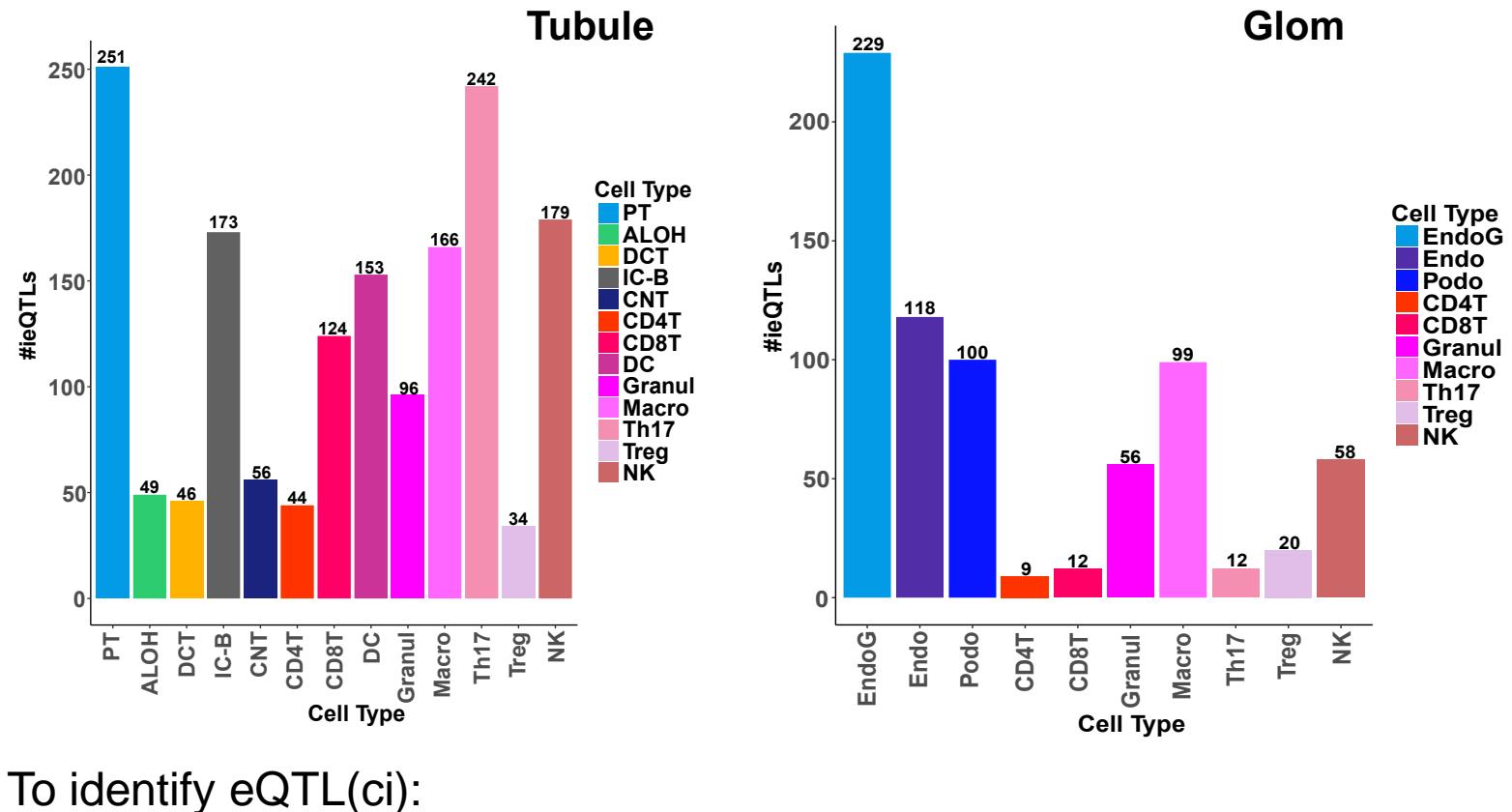
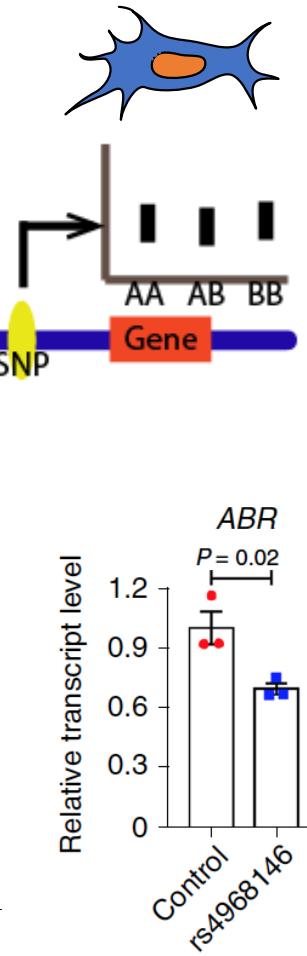
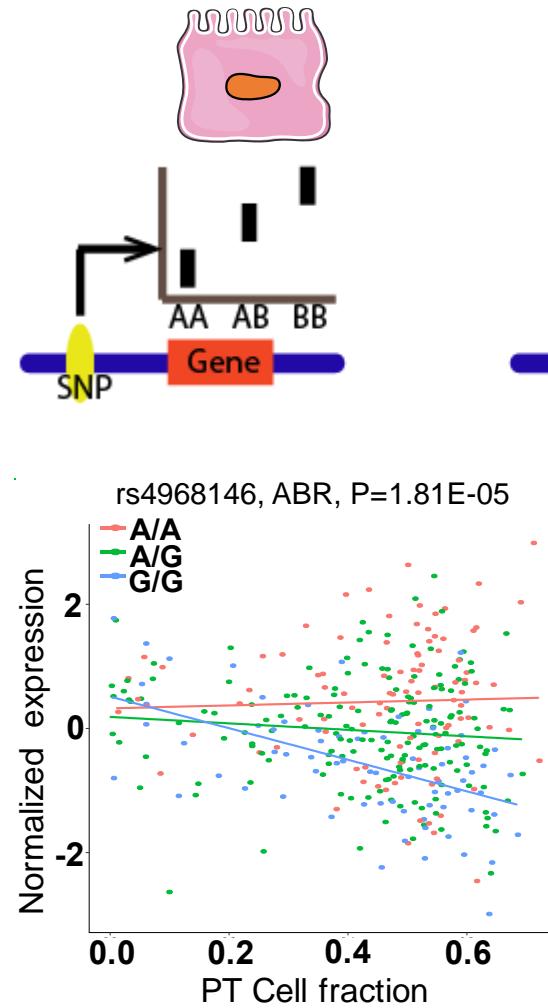
CKD疾病表型的相关性



内容大纲

- 基本概念
- 慢性肾脏病及其并发症
- 多组学数据整合分析方法
- 肾脏具有高度细胞异质性
- 从单细胞水平解析SNP-基因-CKD机制
- 总结

与细胞“互作”的：SNP-Gene调控

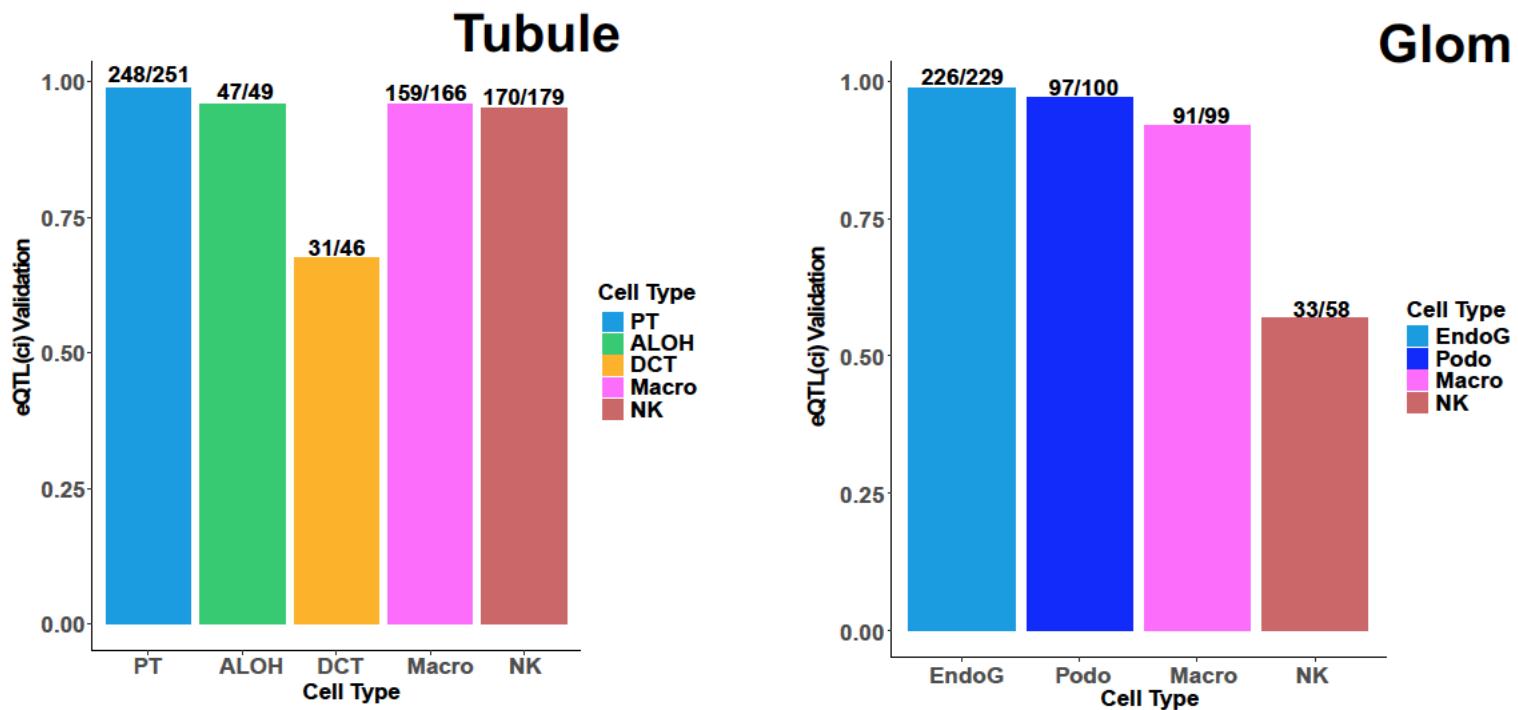


To identify eQTL(ci):

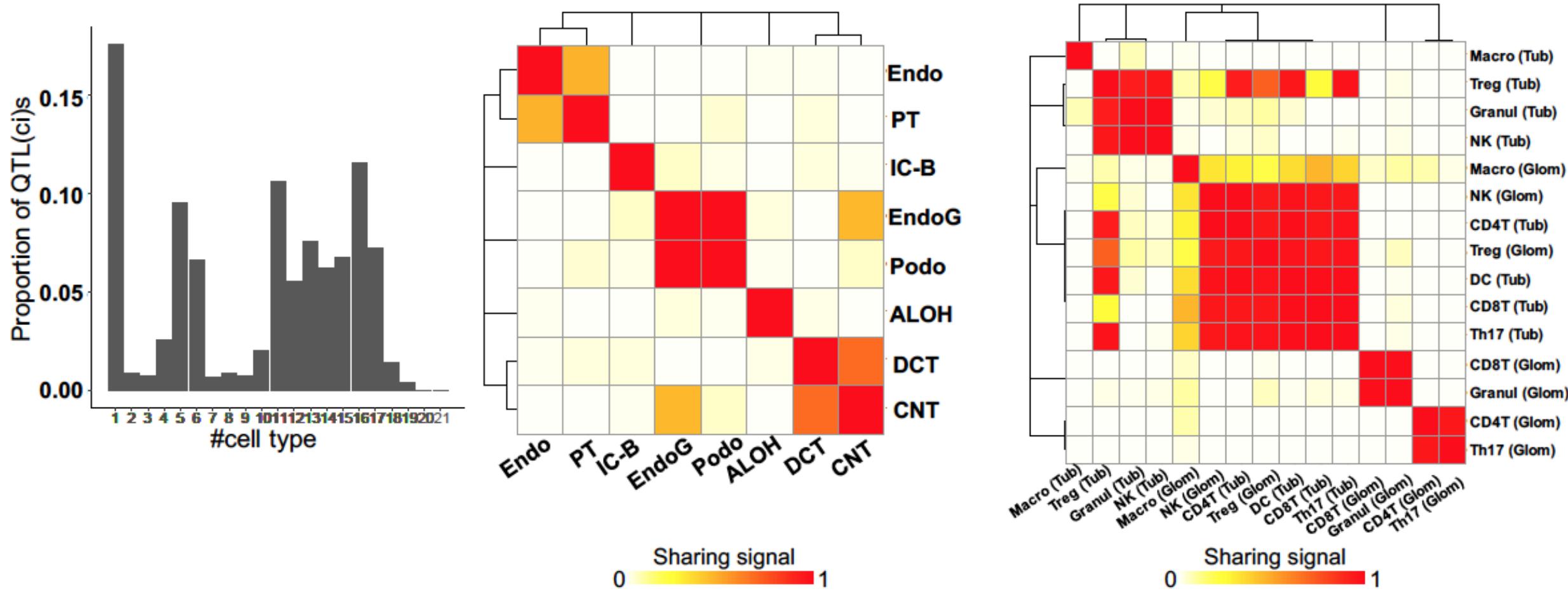
$H_0: \text{INT transformed gene expression} \sim \text{Beta2} \times \text{SNP dosage} + \text{Confounders}$ (1)

$H_1: \text{INT transformed gene expression} \sim \text{Beta1} \times \text{SNP dosage} : \text{cell fraction of interested cell type} + \text{Beta2} \times \text{SNP dosage} + \text{Confounders}$ (2)

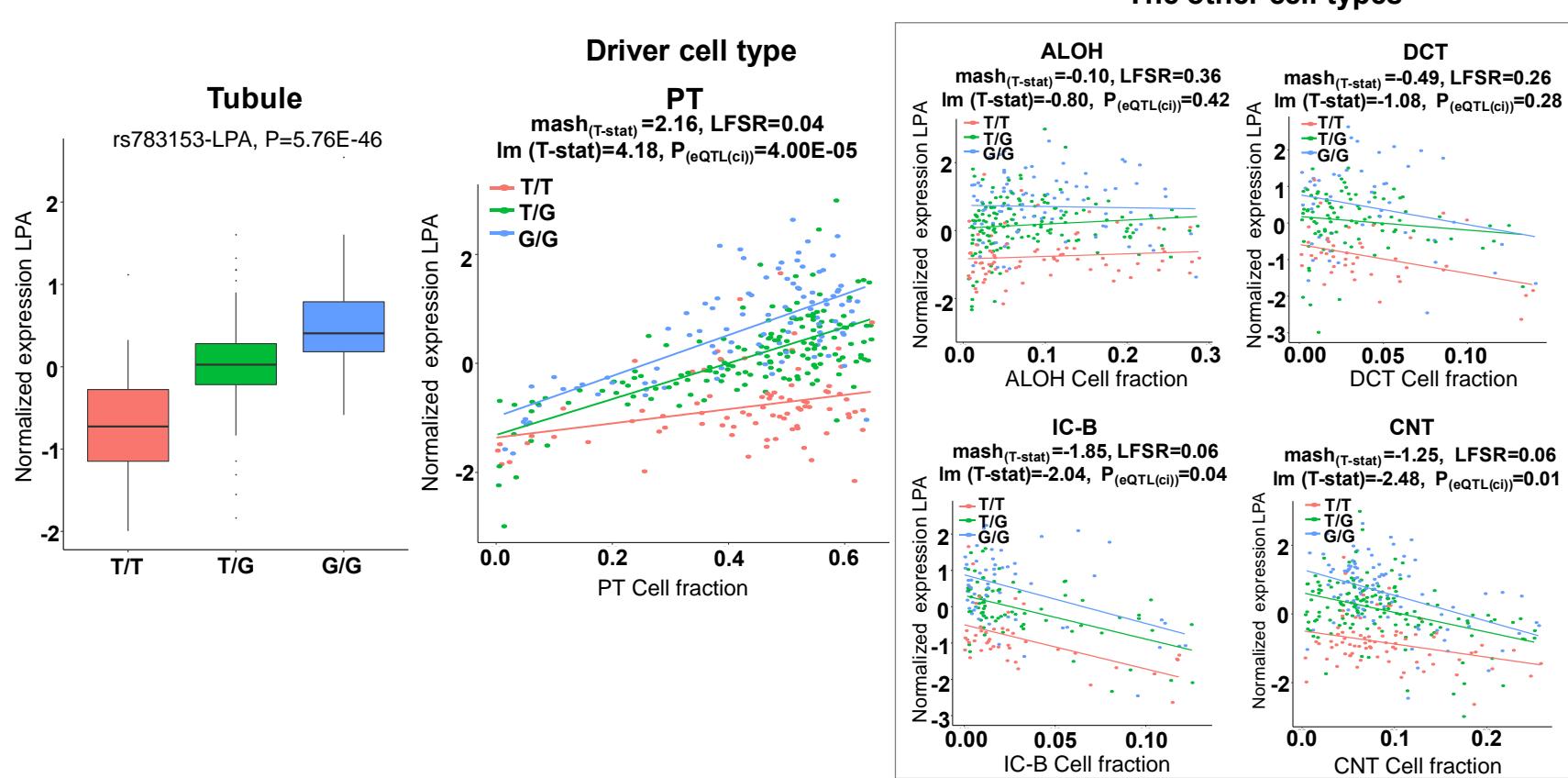
证明算法的可重复性



细胞“互作” ≠ 细胞特异性



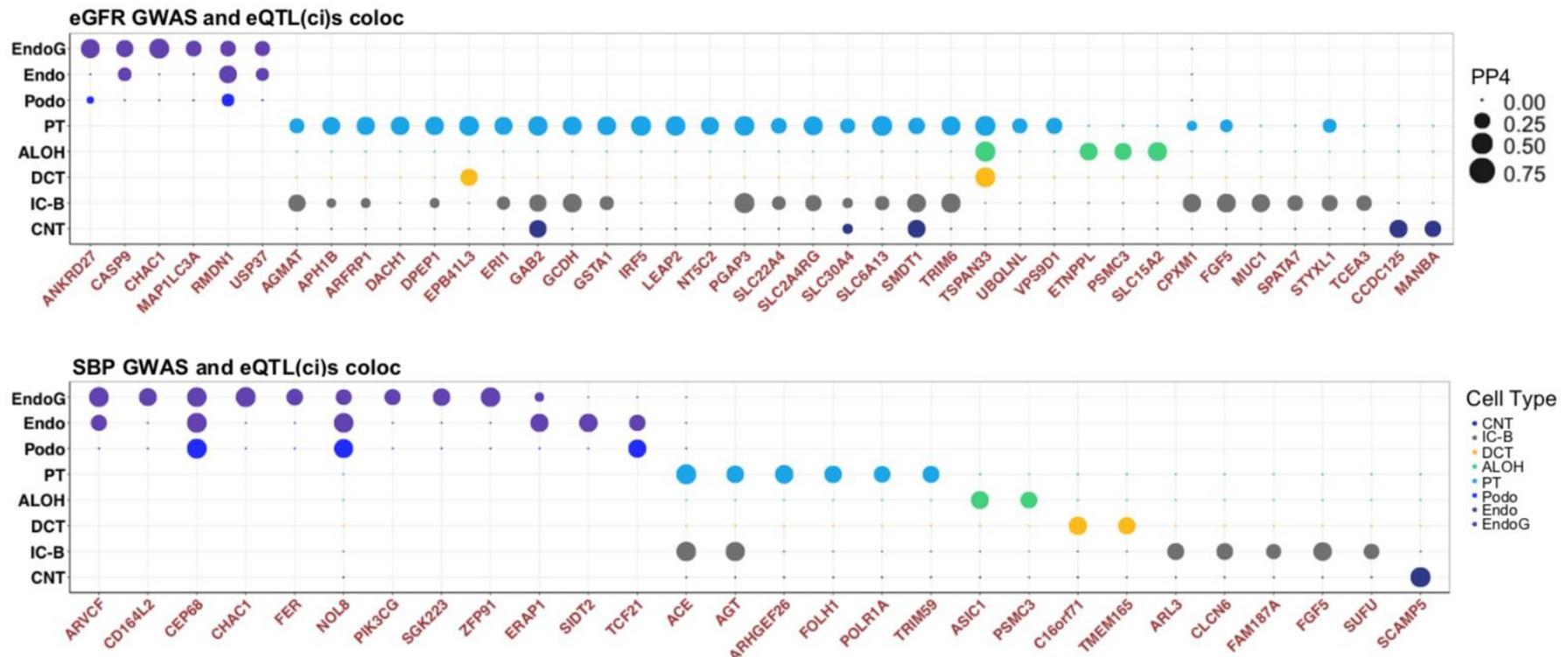
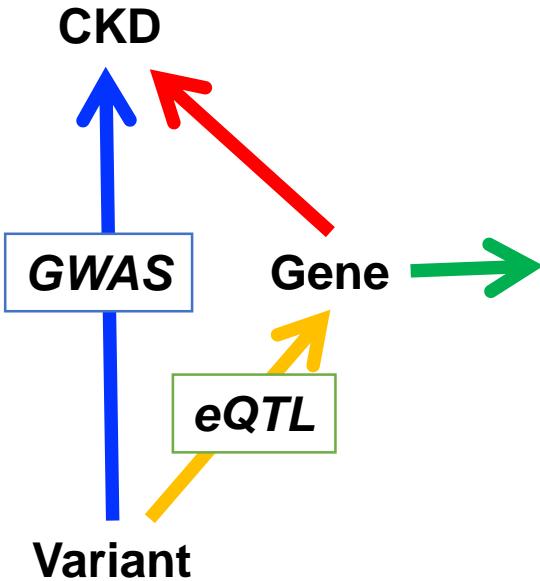
细胞特异的eQTL调控



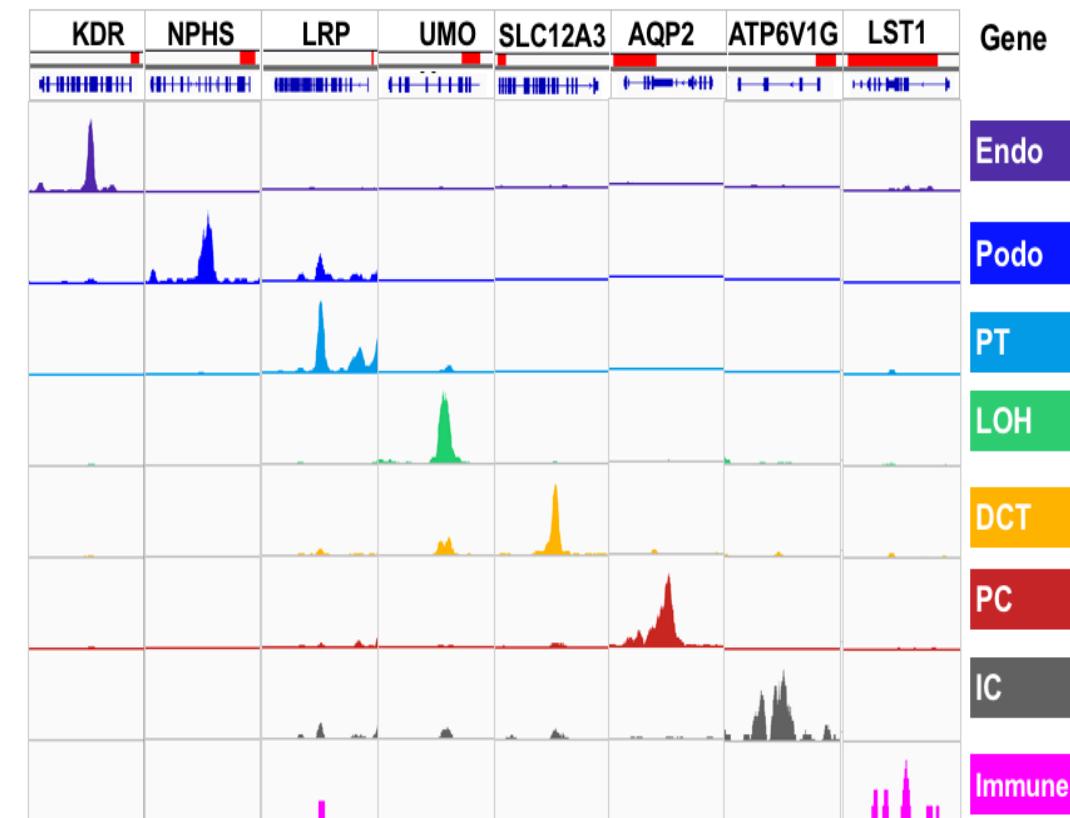
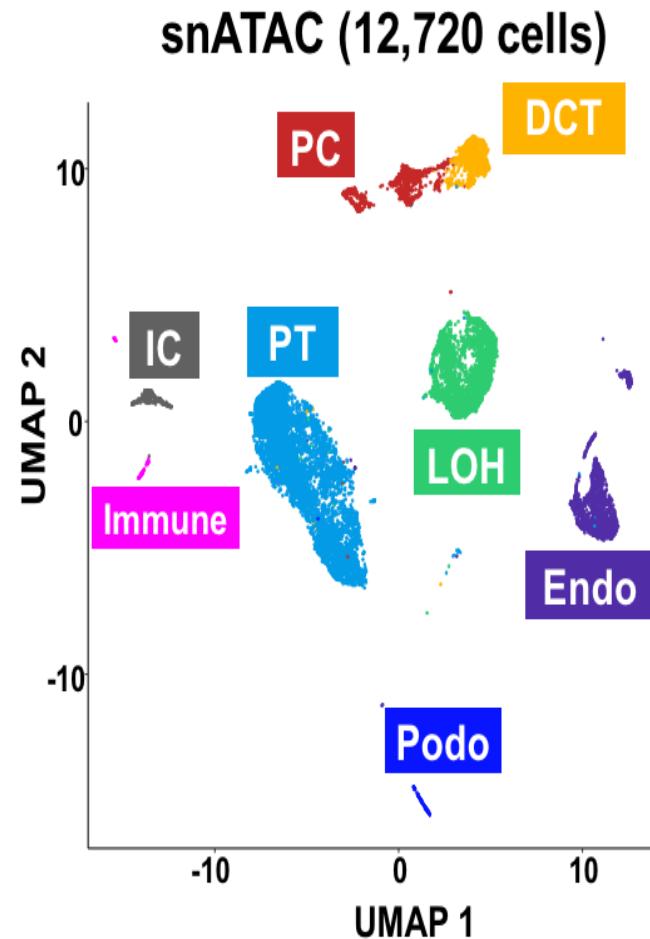
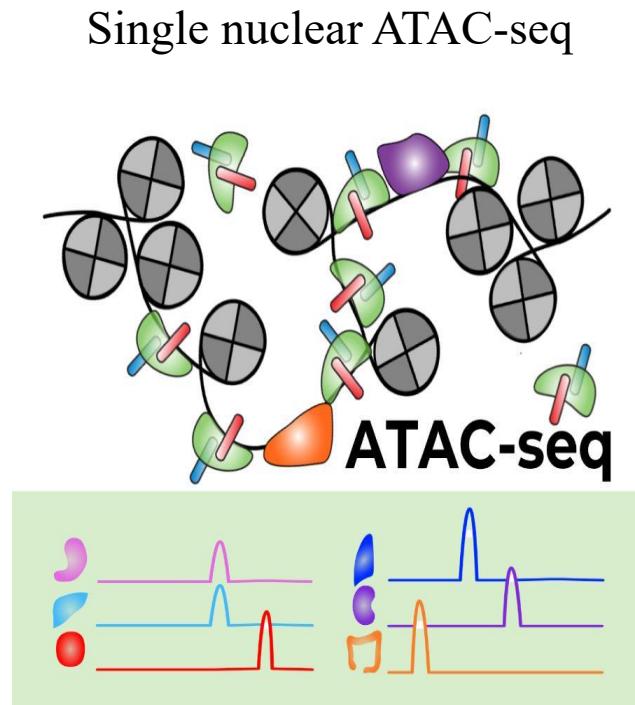
PT-specific expression in human kidney scRNA-seq data



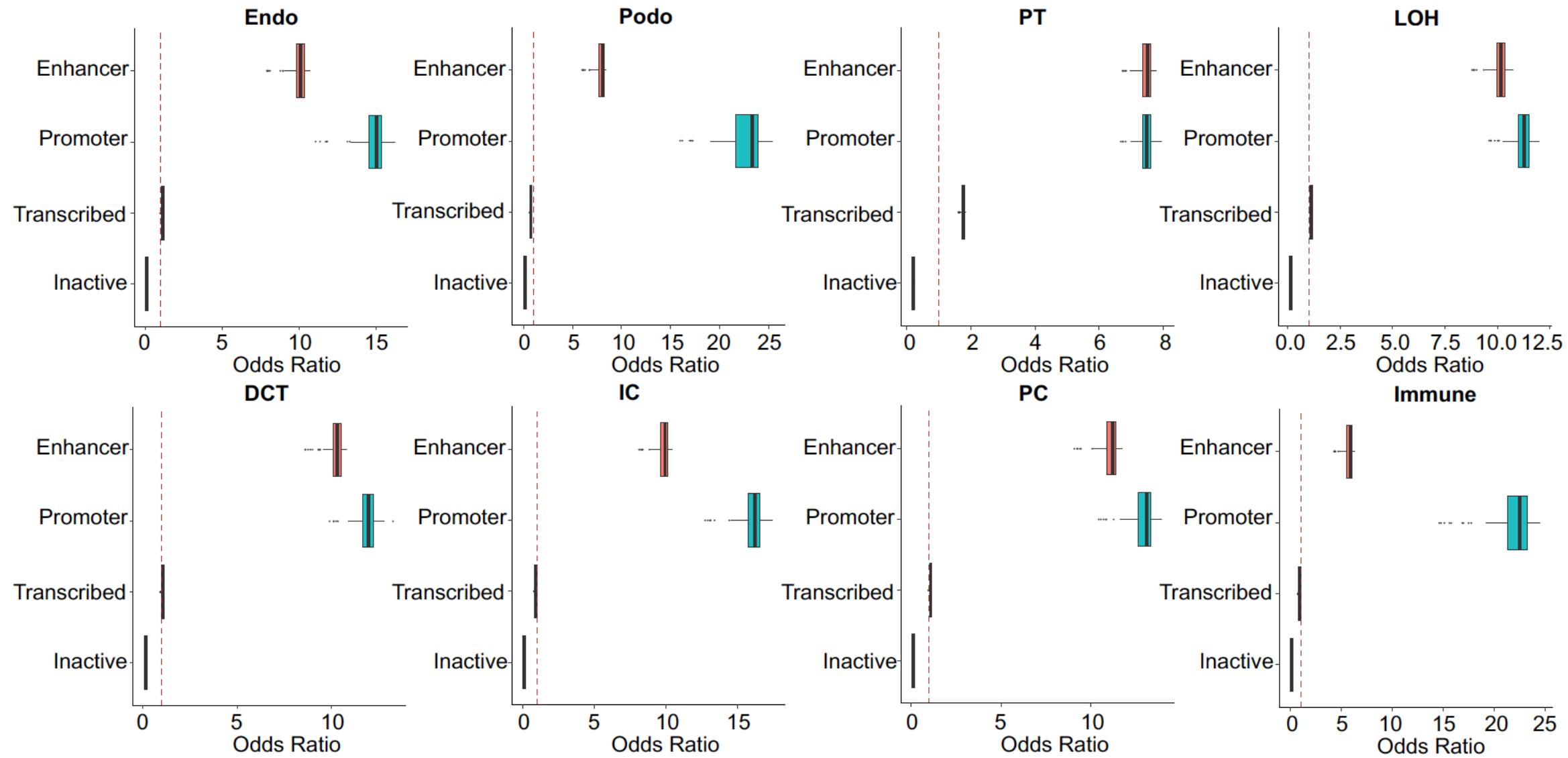
依赖特定细胞的：CKD和高血压的潜在致病基因



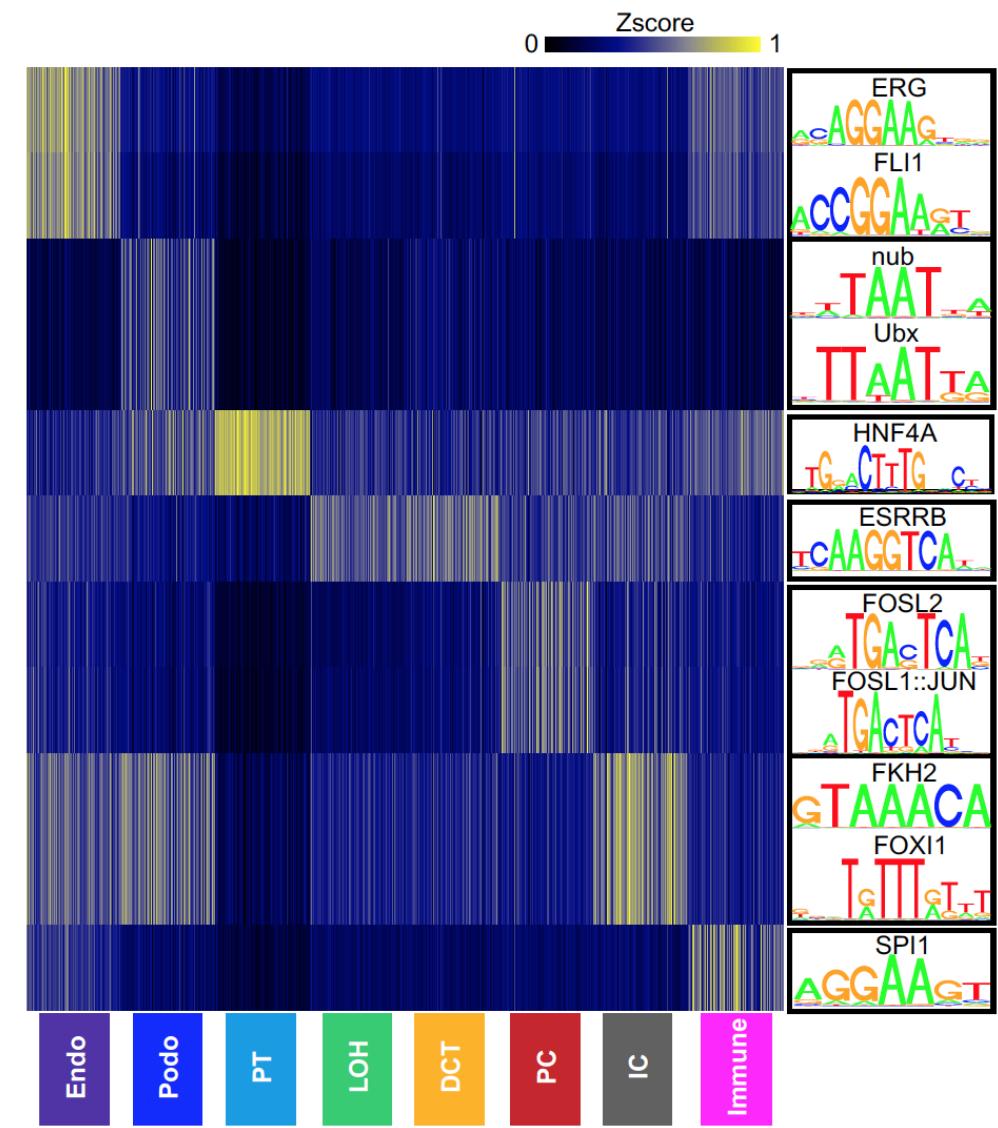
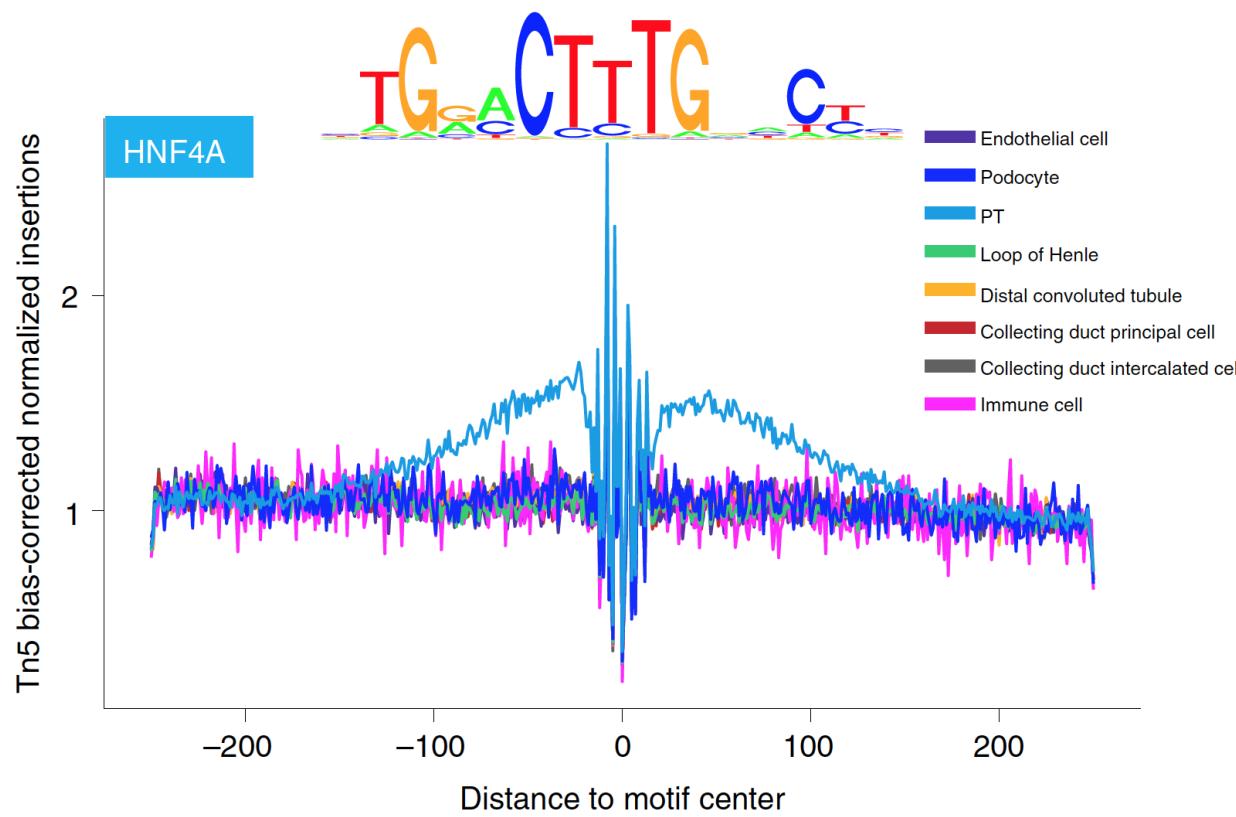
肾脏：单细胞水平染色质可及性



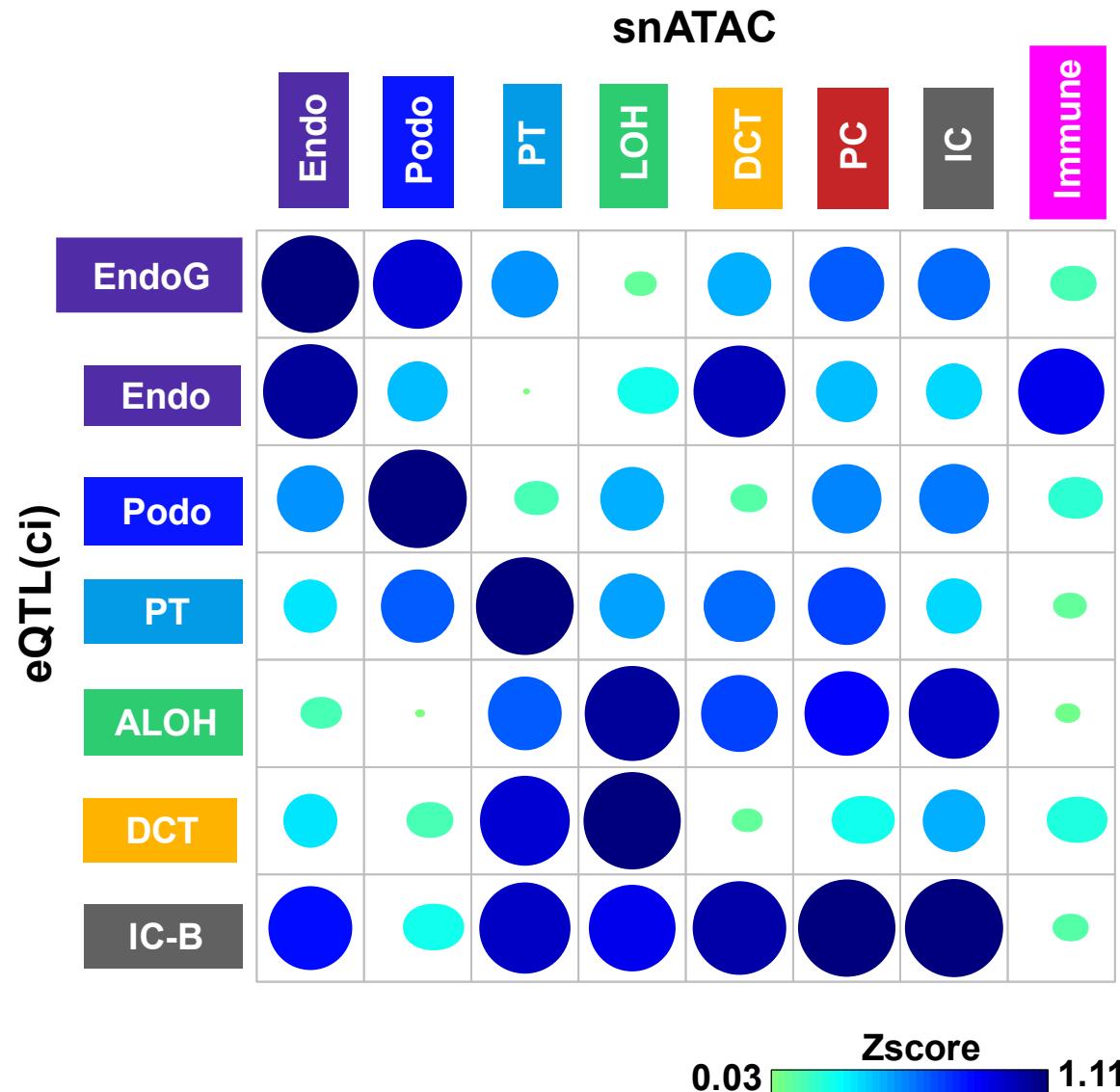
肾脏：单细胞水平开放染色质区域



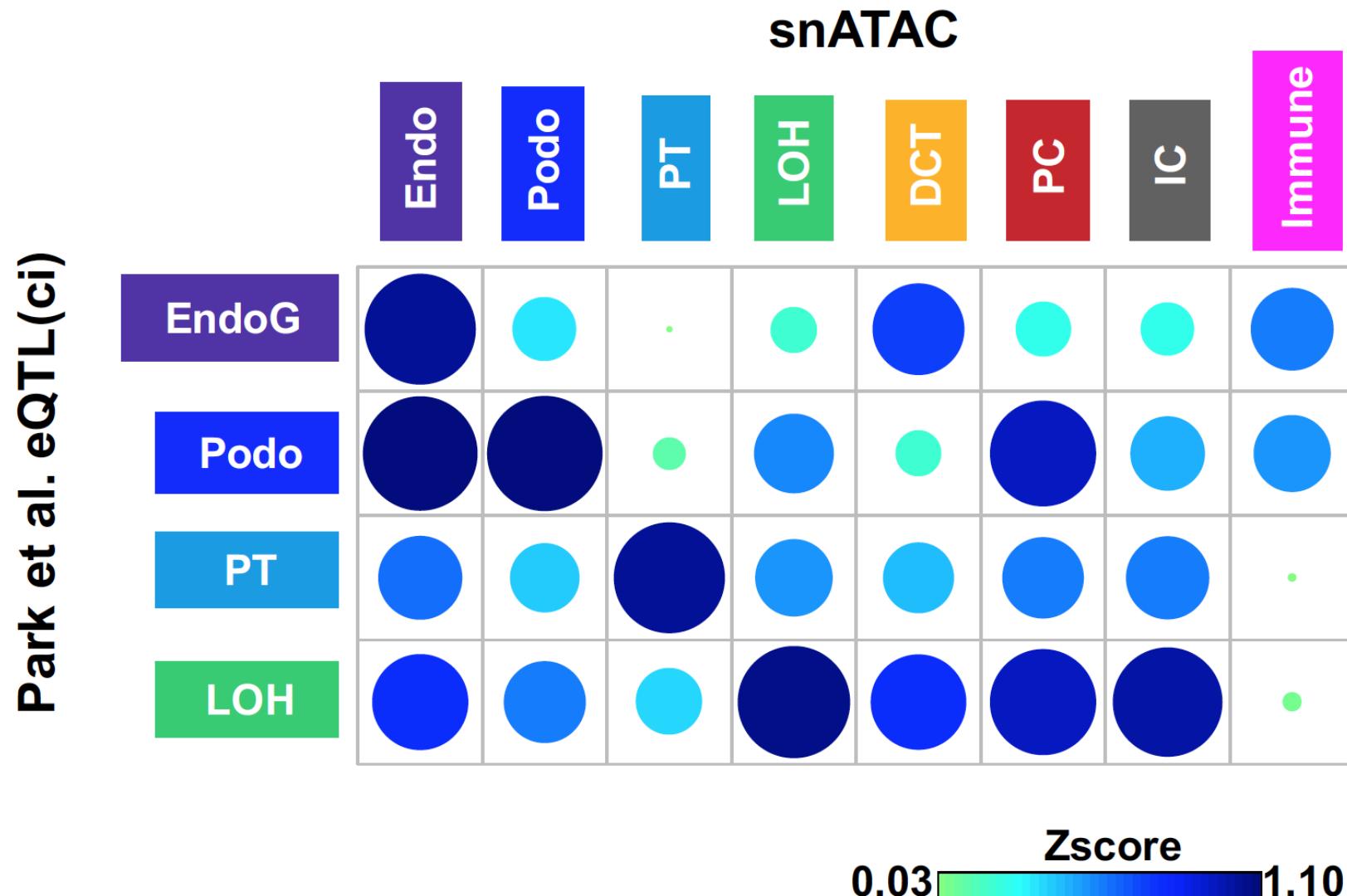
细胞特异的转录因子



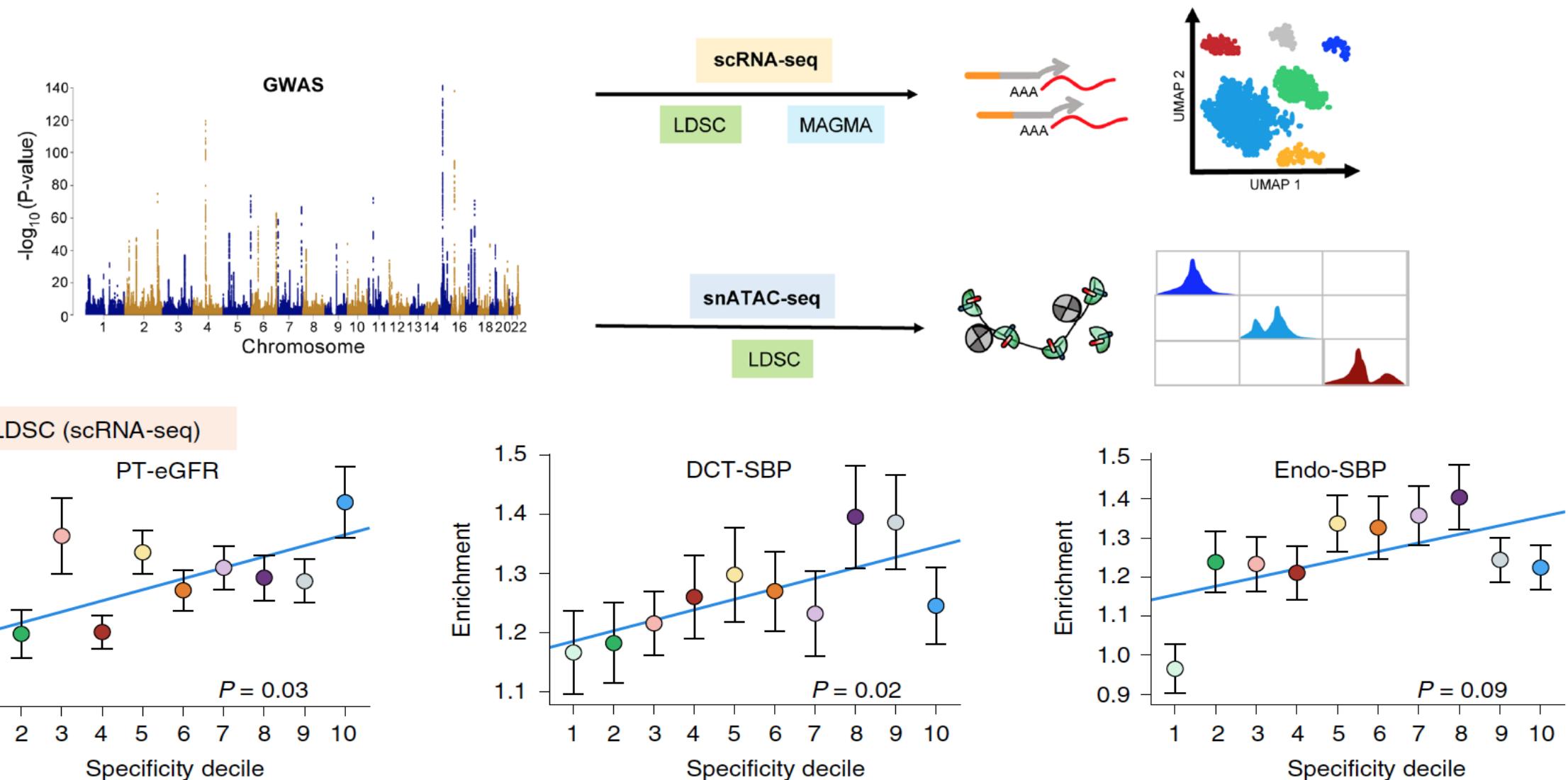
eQTL(ci) SNPs富集在相应细胞的调控区域



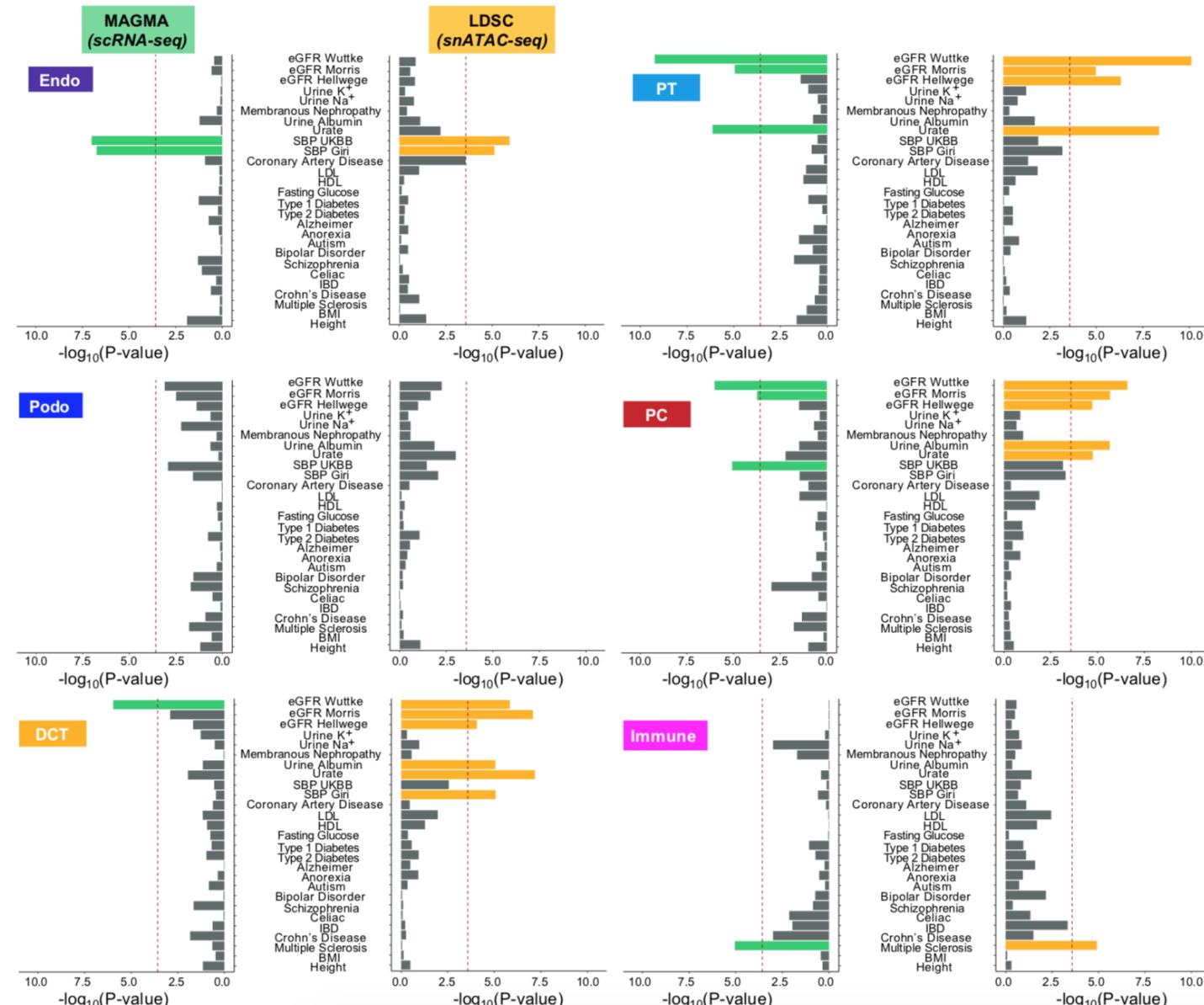
研究结论的可重复性



LDSC: 注释肾脏各类细胞所承担的生物学功能

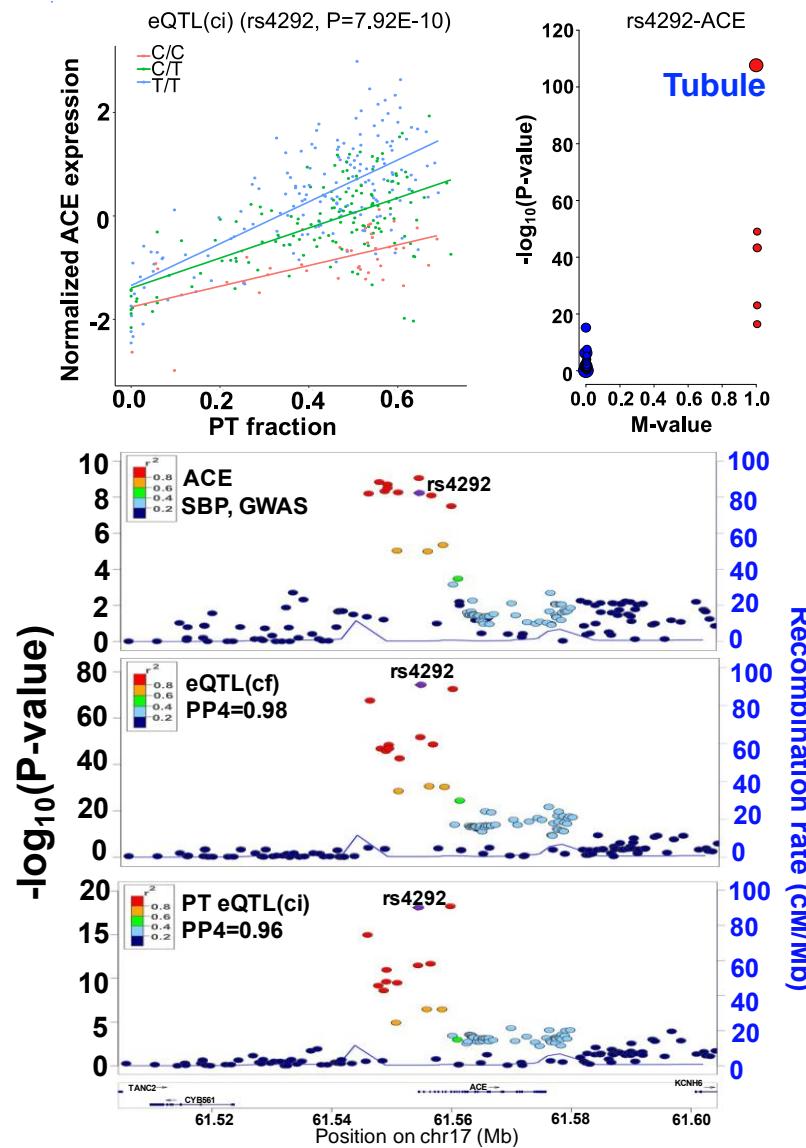


MAGMA和LDSC: 注释肾脏各类细胞所承担的生物学功能

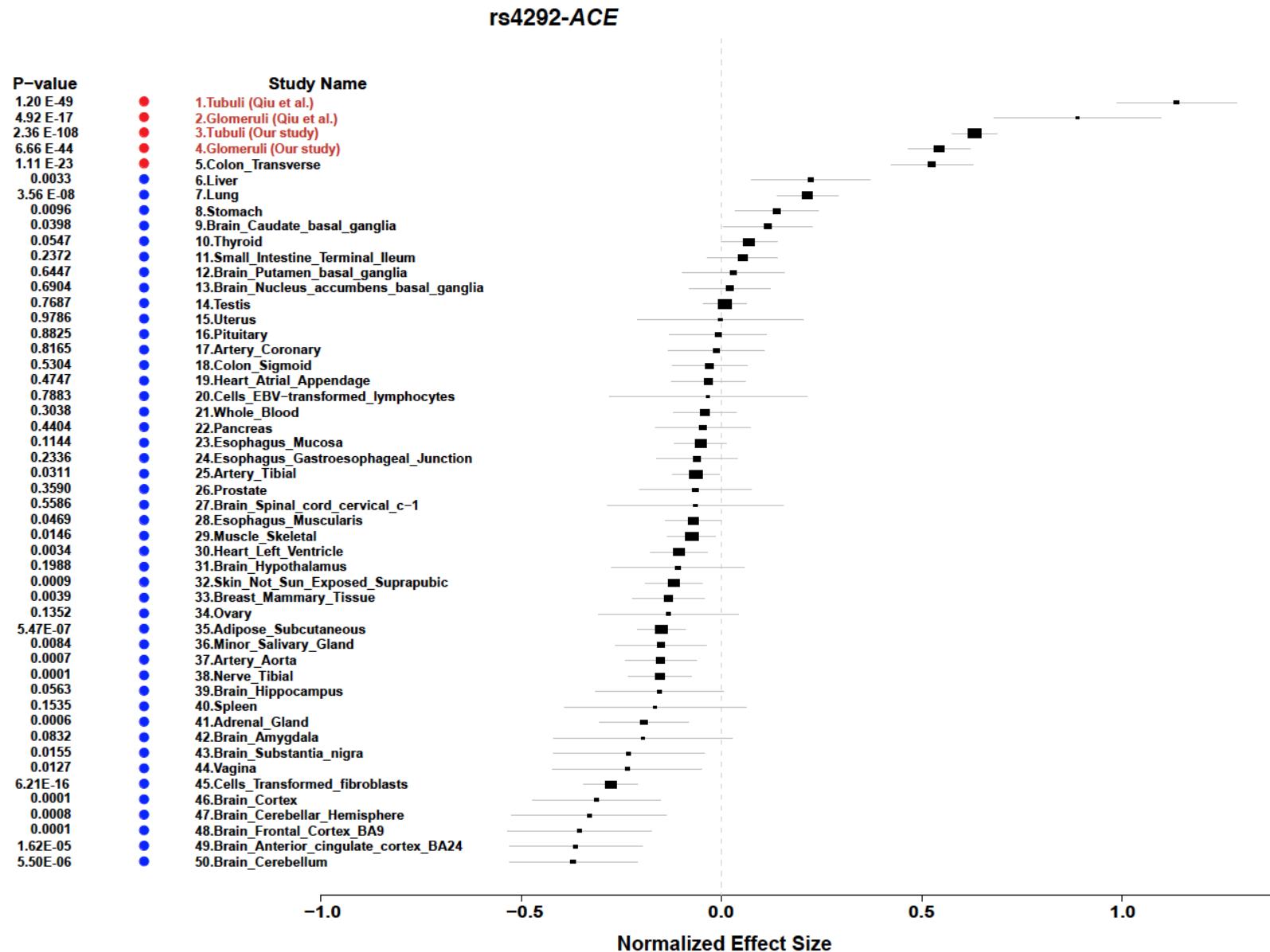


肾素-血管紧张素系统 (ACEi)

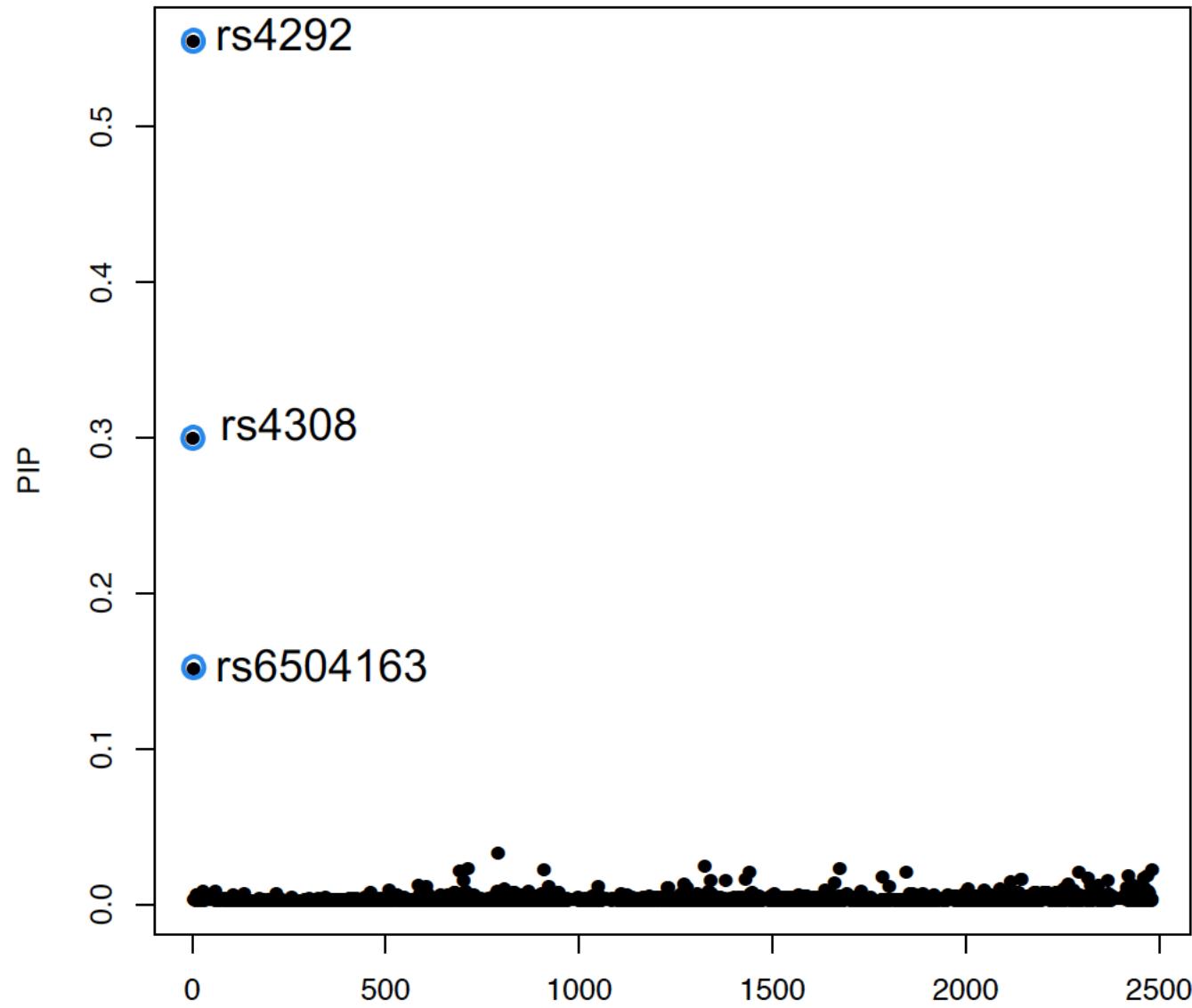
rs4292-ACE



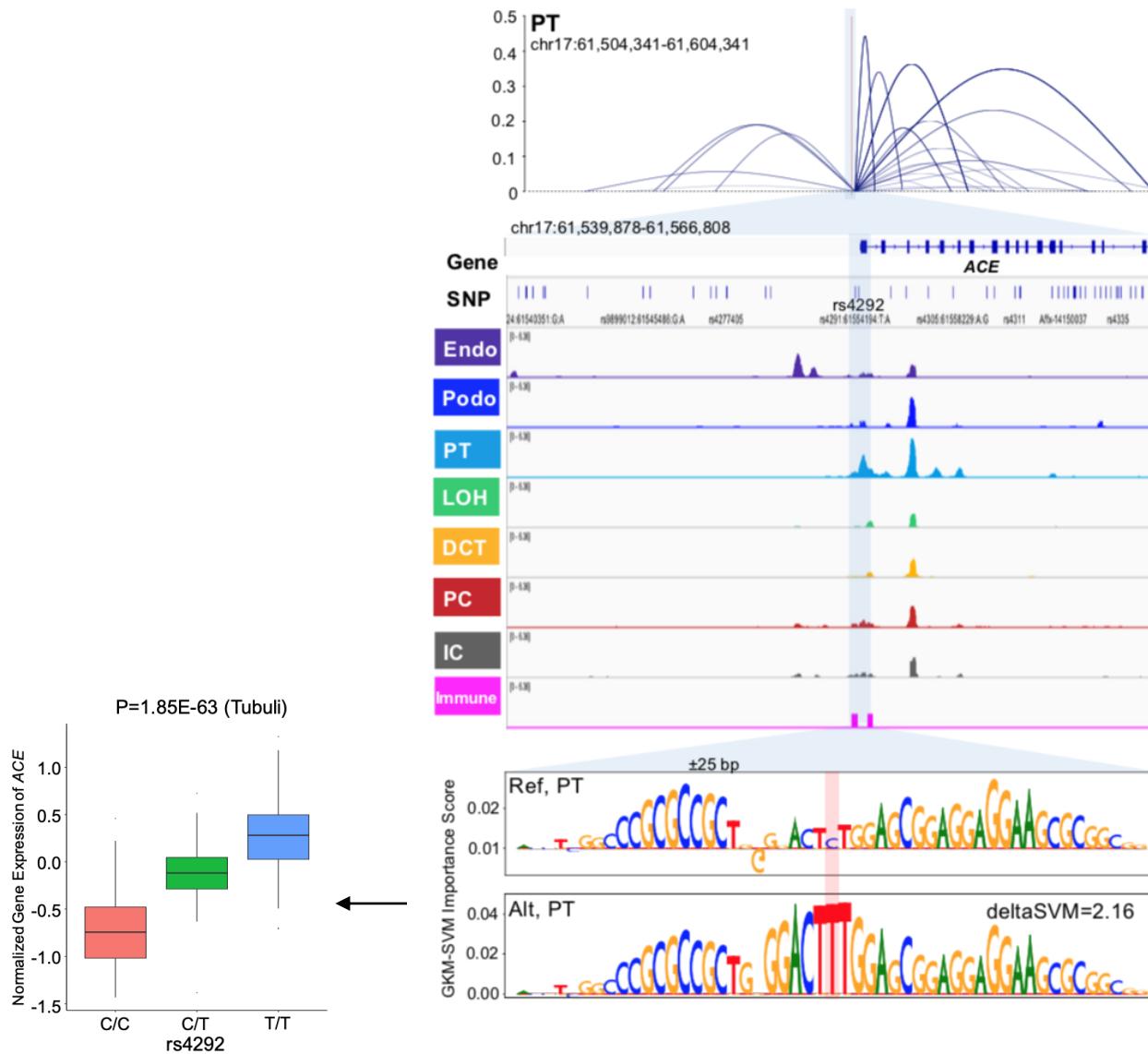
肾脏特异的调控作用：rs4292-ACE



Fine mapping: 定位致病SNP



肾素-血管紧张素系统 (ACEi)



内容大纲

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总结

- 慢性肾脏病（CKD）与高血压
- 肾脏具有高度细胞异质性
- 鉴定细胞“互作”的eQTL调控的新算法
- 组织的细胞异质性对复杂疾病的潜在致病基因鉴定十分重要