

Analysis And Application Of Cohort Data

Cardiovascular Disease Precision Medicine Driven By Population Genomics

2025/10/14

International Training Course
on Biological Data Integration,
Sharing and Application

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INNOVATION TOGETHER

Wang lab

A data-driven
precision medicine lab

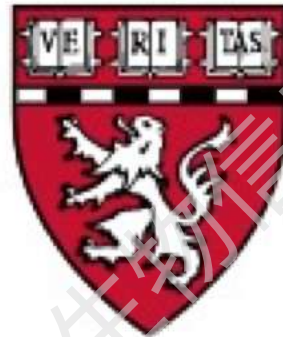
Personal Profile



Jilin University
2005 – 2009
BSc
Animal Science



CAS-PICB
2009 – 2015
PhD
Computational
Biology



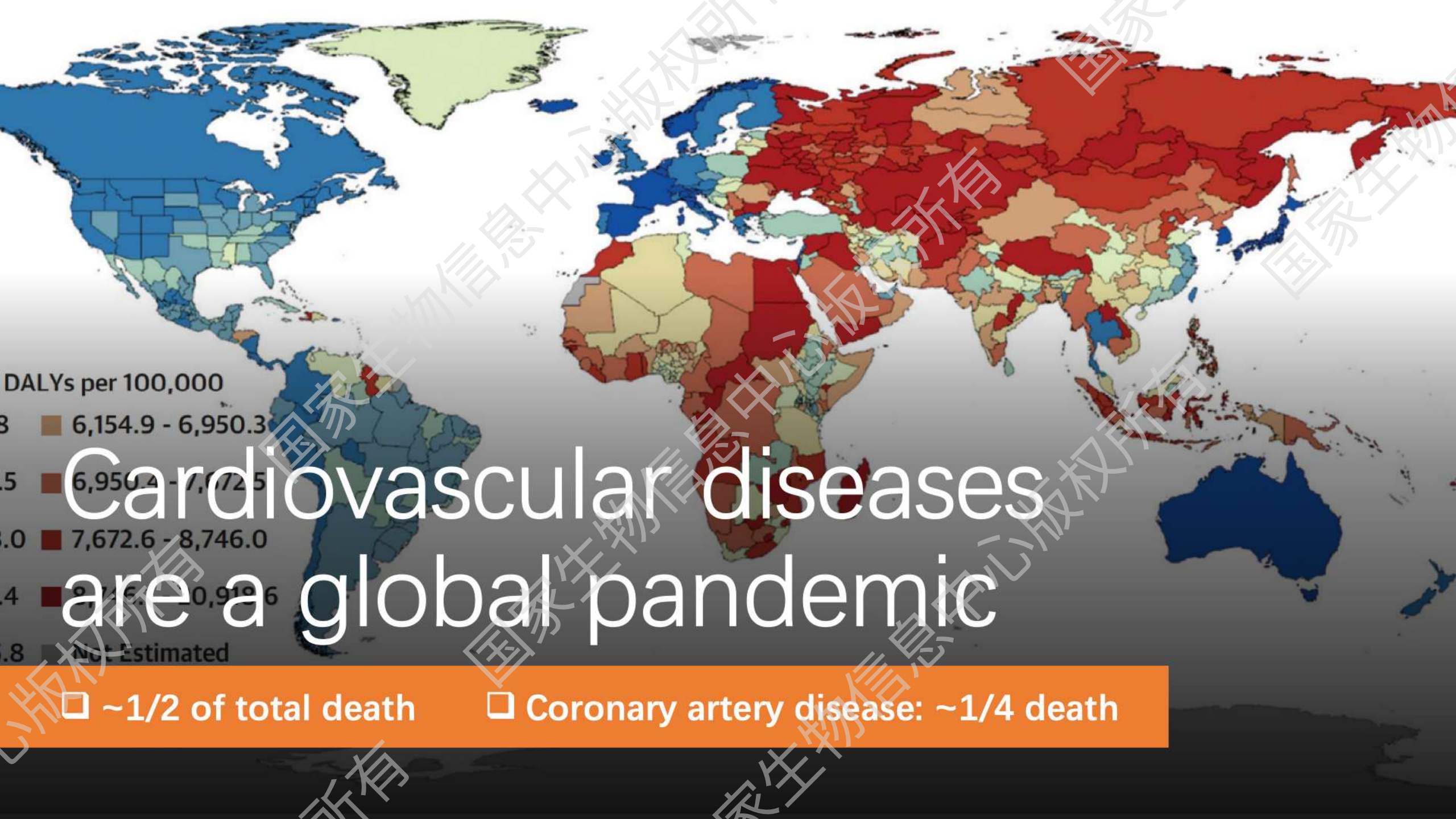
Harvard Medical
School
2015 – 2018
Postdoc
Medical Genetics



Broad Institute of MIT and
Harvard
Massachusetts General
Hospital
2018.01 – 2021.10
Computational Biologist



CAS-CNCB
2021.10 –
Principal Investigator
PhD Supervisor



DALYs per 100,000

3 6,154.9 - 6,950.3

5 6,950.4 - 7,672.5

0 7,672.6 - 8,746.0

4 8,746.1 - 10,912.6

8 Not Estimated

Cardiovascular diseases are a global pandemic

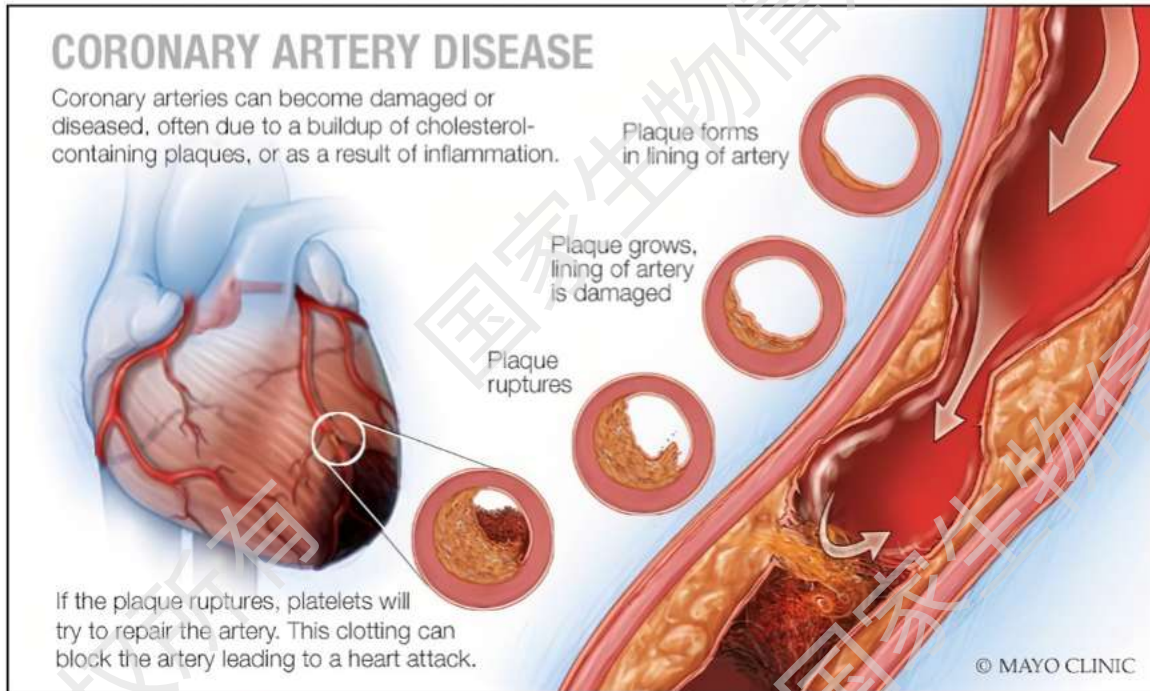
□ ~1/2 of total death

□ Coronary artery disease: ~1/4 death

Coronary Artery disease and atrial fibrillation

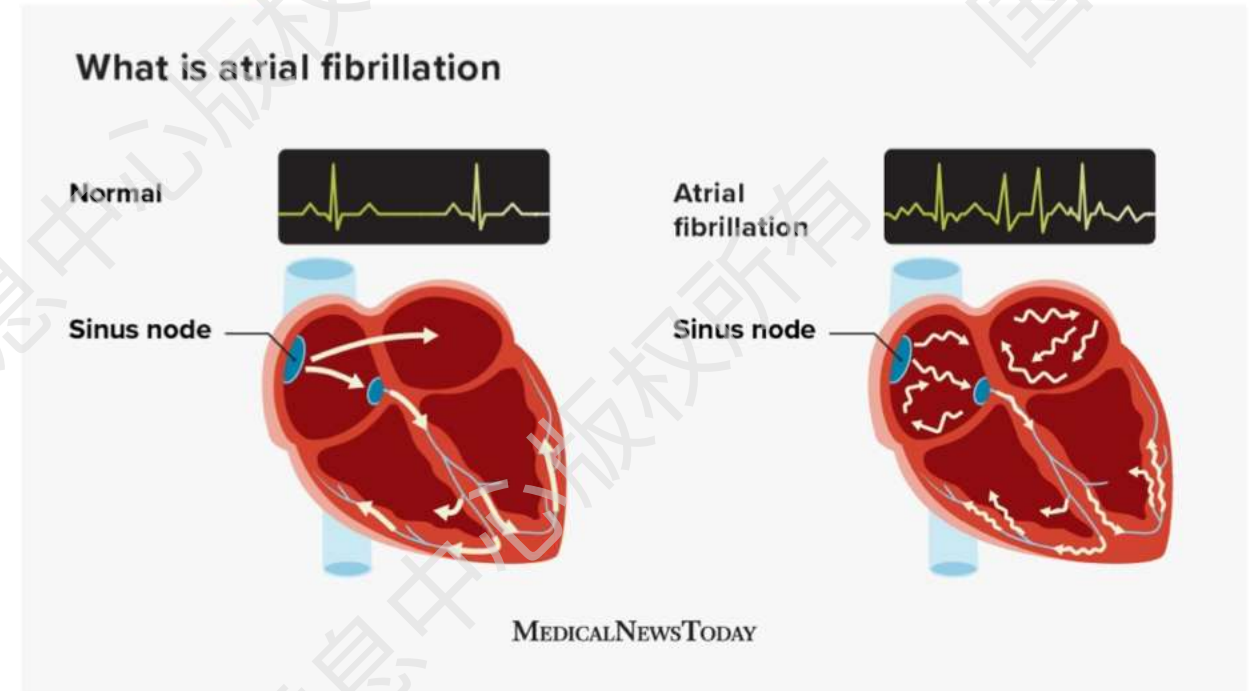


- ❖ **CAD: Atherosclerosis in the coronary arteries reduces blood flow to the heart**



Myocardial infarction chest pain and tightness

- ❖ **AF: Rapid, disordered electrical activity in the atria causes an irregular heartbeat**



Stroke incidence in AF patients: 17.5%

Stroke incidence in general population: 0.35%

Genes and Phenotypes



Aaron at 7

Me at 18

You reap what you sow: Genes have a strong determinative effect on phenotypes

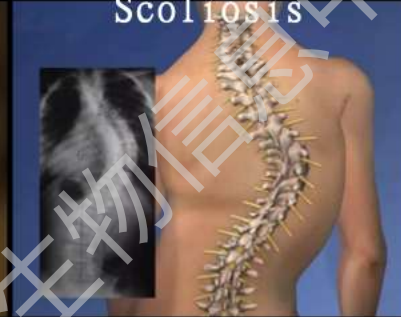
Genes and Human Disease and Health



ALS



Congenital
Scoliosis



Gaucher
Disease



Laron Syndrome



Epidermolysis
Bullosa



IgG4-Related
Disease



Marfan Syndrome



ALS



Wilson
Disease



Albinism



Hereditary
Angioedema



Congenital Myasthenic
Syndromes



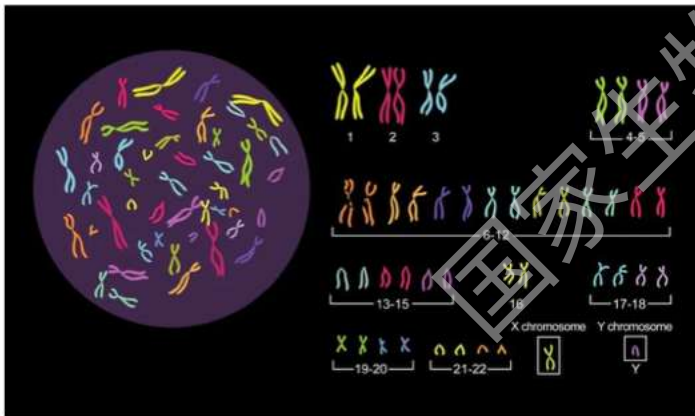
Hereditary Spastic
Paraplegia



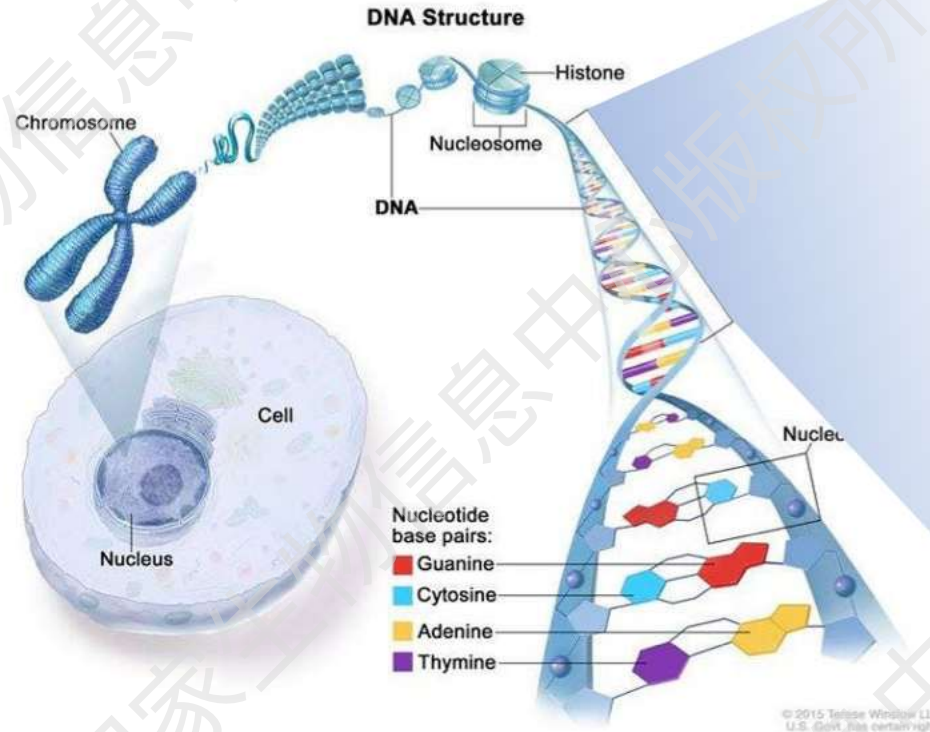
CMT



A gene is a functional sequence within the genome



Human Chromosome Map



Four bases of genome:
A Adenine
G Guanine
C Cytosine
T Thymine

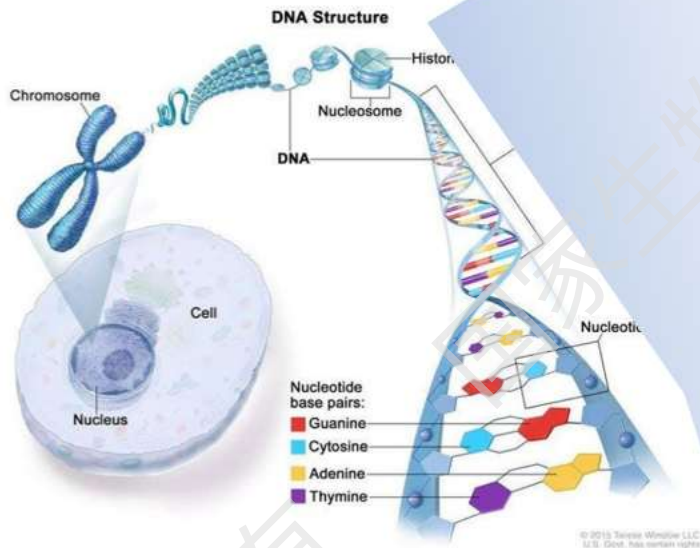
- The human DNA molecule is approximately 2.2 meters long
- The human genome contains 3.2 billion BP

```

1  accccccac  cccaggcct  aatgggccag  gggcagggg  ttgagaggt
51  ggggagatgg  gctctgagac  tataaagcca  gggggggccc  agcagccctc
101 AGCCCTCCAG GACAGGCTGC ATCAGAAGAG GCCATCAGC AGgtctgttc
151 caagggcctt  tgcgtcaggt  gggctcagga  ttccagggtg  gctggacccc
201 agggcccagc  tctgcagcag  ggaggacgtg  gctgggctcg  tgaagcatgt
251 ggggggtgag  ccagggggcc  caaggcaggg  cacttgccct  tcagcctgcc
301 tcagccctgc  ctgtctccca  gATCACTGTC CTCTGCGCAT GGCCCTGTGG
351 ATGCGCCTCC TGCCCTGCTG GCGCTGCTG GCCCTCTGG GACCTGACCC
401 AGCCGCAGCC TTTGTGAACC AACACCTGTG CGGCTCACAC CTGGTGGAAG
451 CTCTCTACCT AGTGTGCGGG GAACGAGGCT TCTCTACAC ACCCAAGACC
501 CGCCGGGAGG CAGAGGACCT GCAGGgtgag ccaactgccc attgtgtccc
551 ctggccgccc ccagccacccc cctgctcctg gcctcccac ccagcatggg
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701 cccagtcaga  atctcagcct  gaggacgggt  ttggcttcgg  cagccccag
751 atacatcaga  ggggtggcac  gctcctccct  ccactgcccc  ctcaaacaaa
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1251 ctctgtgtgc  cctctgcctc  gccgtgttgc  cggaacctgc  totgcgcggc
1301 acgtcctggc  agTGGGGCAG GTGGAGCTGG GCGGGGGCCC TGGTGCAGGC
1351 AGCCTGCAGC CCTTGGCCCT GGAGGGGTCC CTGCAGAAGC GTGGCATTGT
1401 GGAACAATGC TGTACCAGCA TCTGCTCCCT CTACCAGCTG GAGAACTACT
1451 GCAACTAGAC GCAGCCGCAC GGCAGCCCCA CACCCGCCGC CTCCTGCACC
1501 GAGAGAGATG GAATAAAGCC CTTGAACCAg cctgtgtgtg ccgtctgtgt
1551 gtcttggggg ccctggggcca agcccactt  cccggcactg ttgtgagccc
1601 tctccacqtc tctccacqtc ctctgqqtg
    
```

Human Insulin Gene

A gene is a functional sequence within the genome



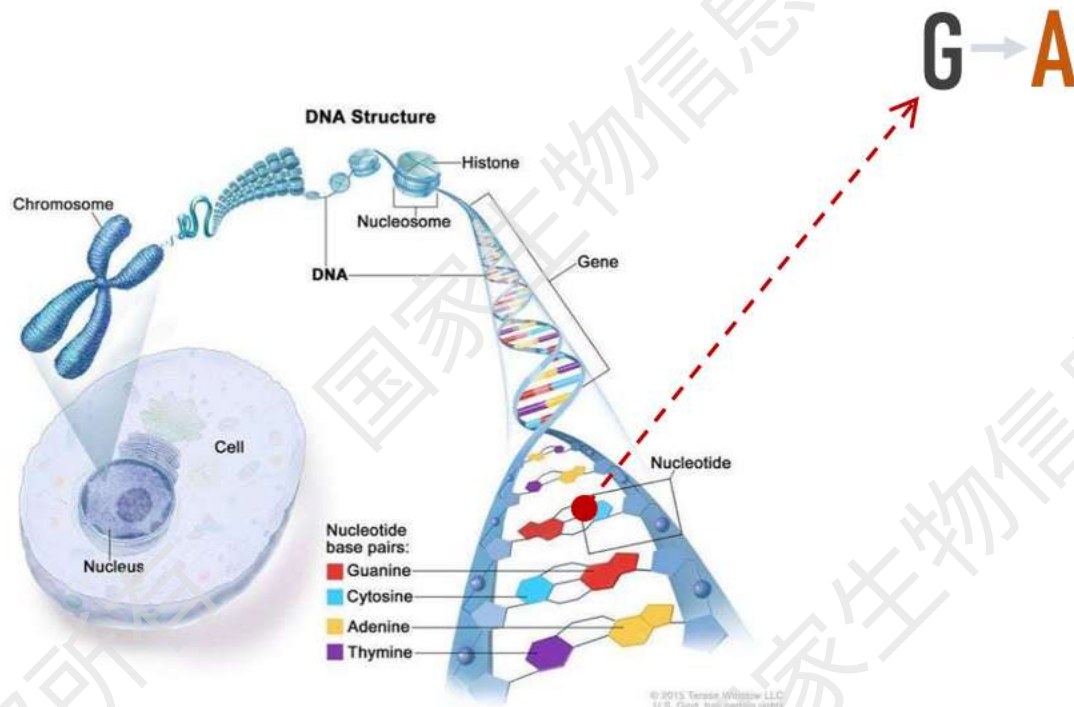
**The human genome contains
3.2 billion BP**

**If you printed out the
~3 billion letters of the
human genome in size 12
font, it would stretch
from Houston to Boston!**

Note: Only 20 letters of DNA sequence shown. To imagine what ~3 billion letters of the human genome sequence would look like, multiple this by 153 million!



How is our genome sequenced?

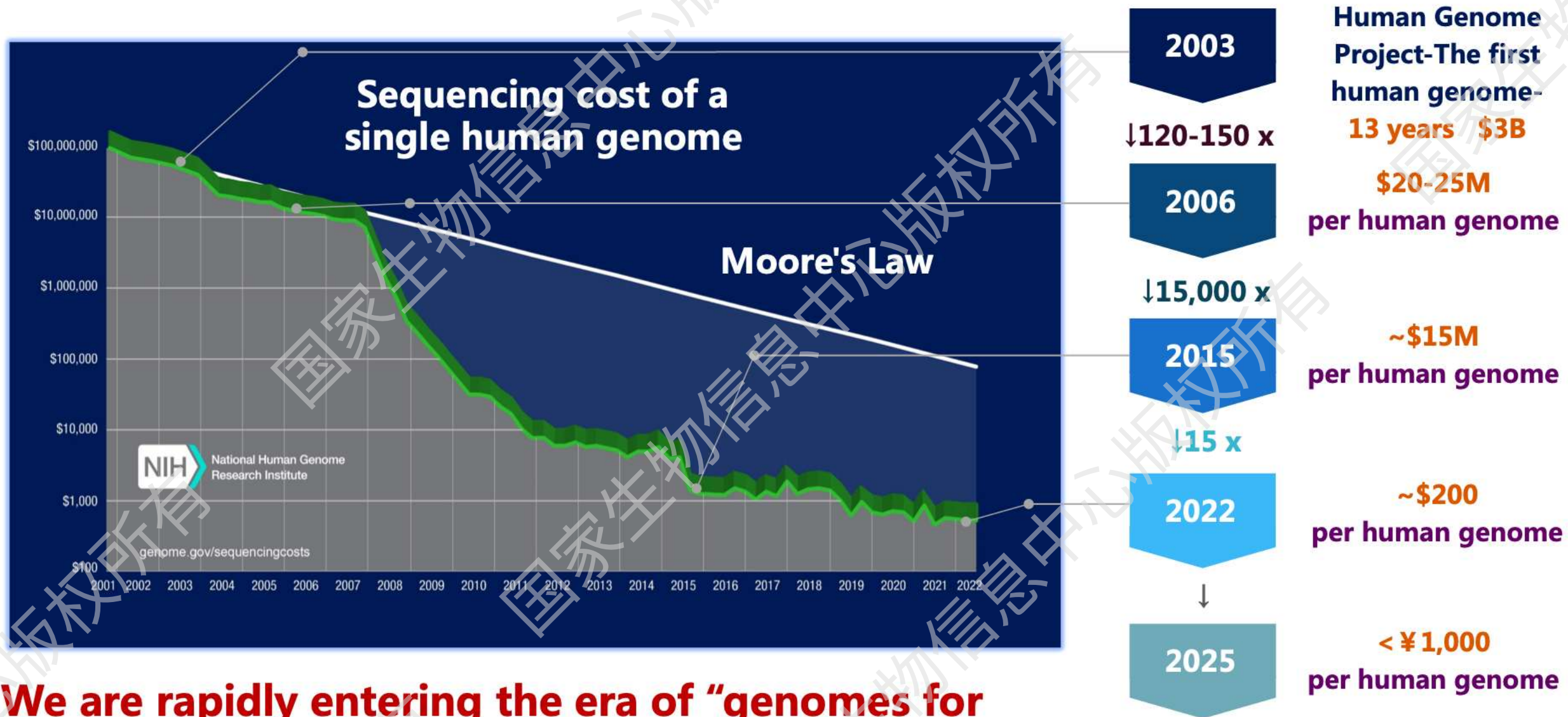


H1: GCTCATCGG--CTCGA
H2: GCTCATCGA--CTCTA



DNA Sequencer

Sequencing costs have decreased faster than Moore's Law



We are rapidly entering the era of "genomes for everyone"

Based on 30x sequencing depth Credit: 谢晓亮院士

Genetic differences are the basis of disease and phenotypic variation

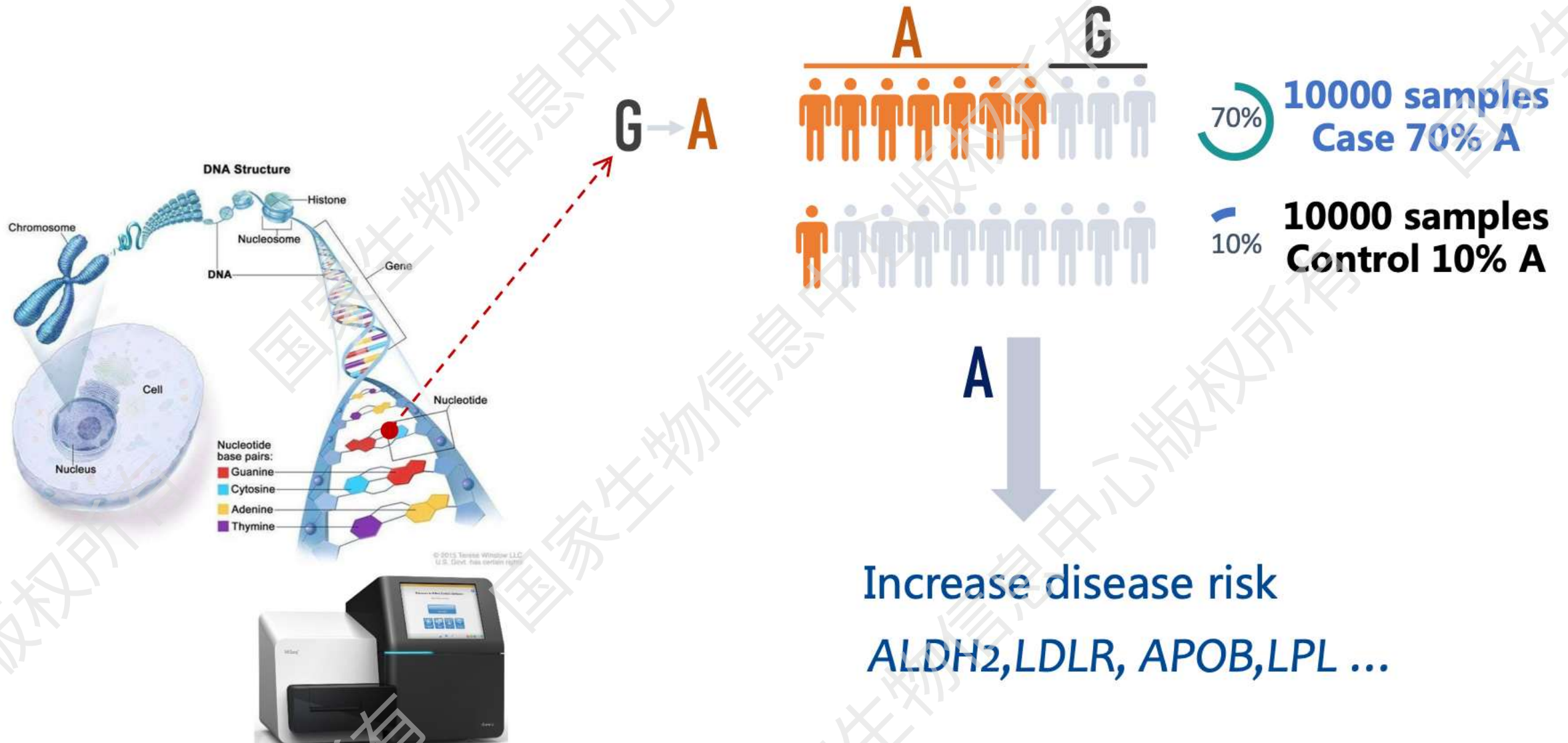


<https://www.genome.gov/dna-day/15-ways/human-genomic-variation>

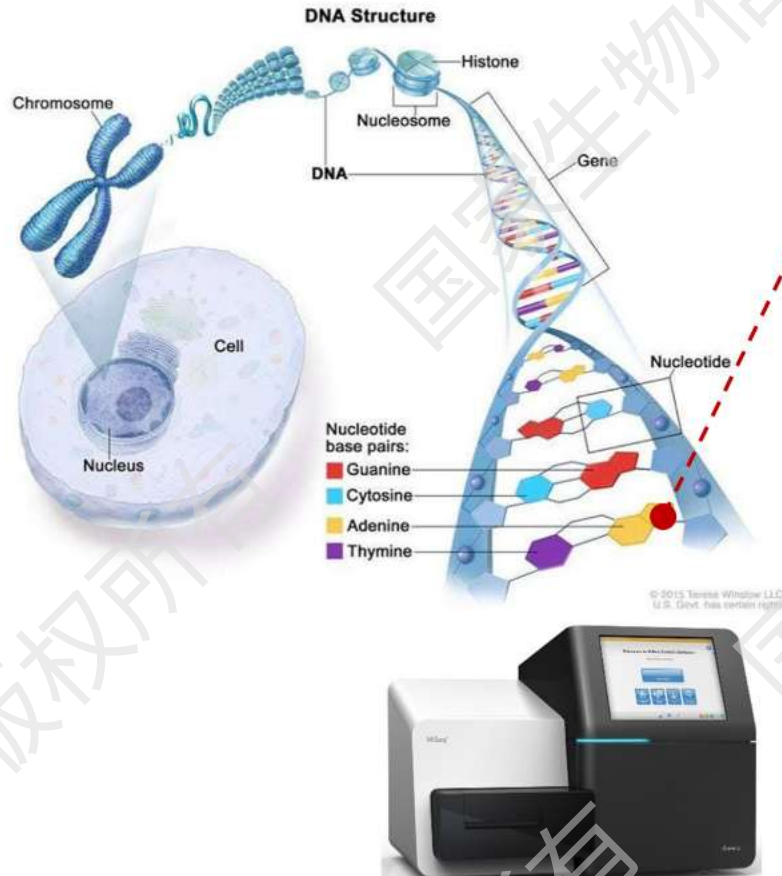
The genome sequences of individuals differ by approximately 0.1%



Identifying gene–disease/phenotype associations



Identifying gene–disease/phenotype associations



A → C



50% **10000 samples**
Case 50% C



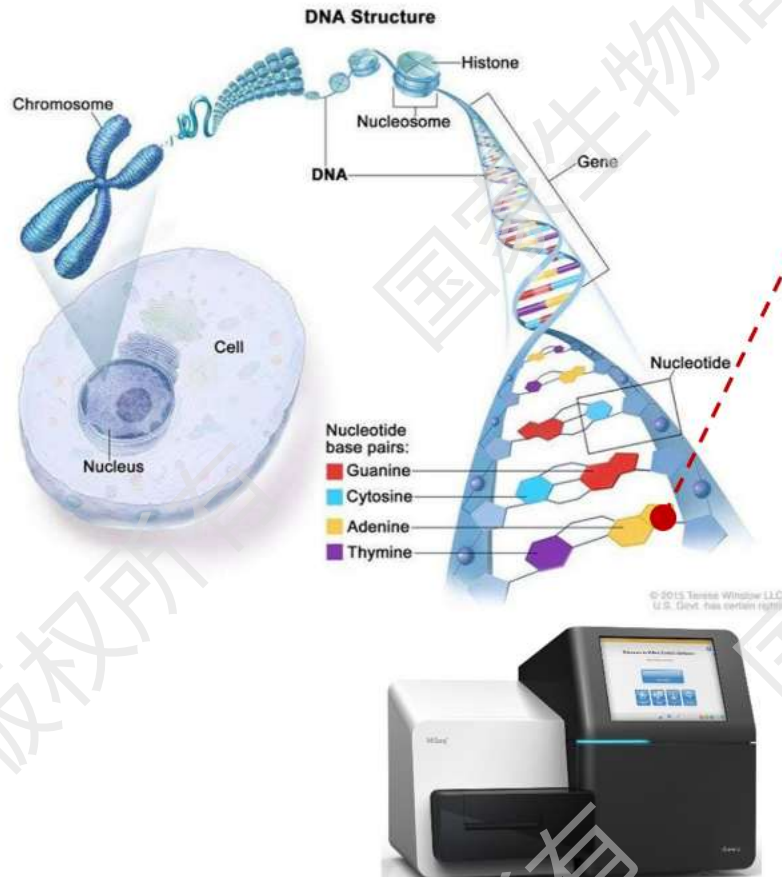
90% **10000 samples**
Control 90% C



Reduce disease risk

PCSK9, ASGR1, NPC1L1 ...

Identifying gene–disease/phenotype associations



A → C



50% 10000 samples
Case 50% C



90% 10000 samples
Control 90% C

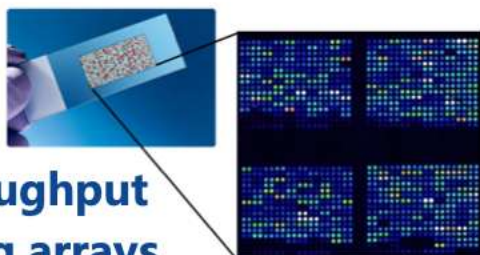
C

➡ **Not all mutations
are harmful**

Reduce disease risk

PCSK9, ASGR1, NPC1L1 ...

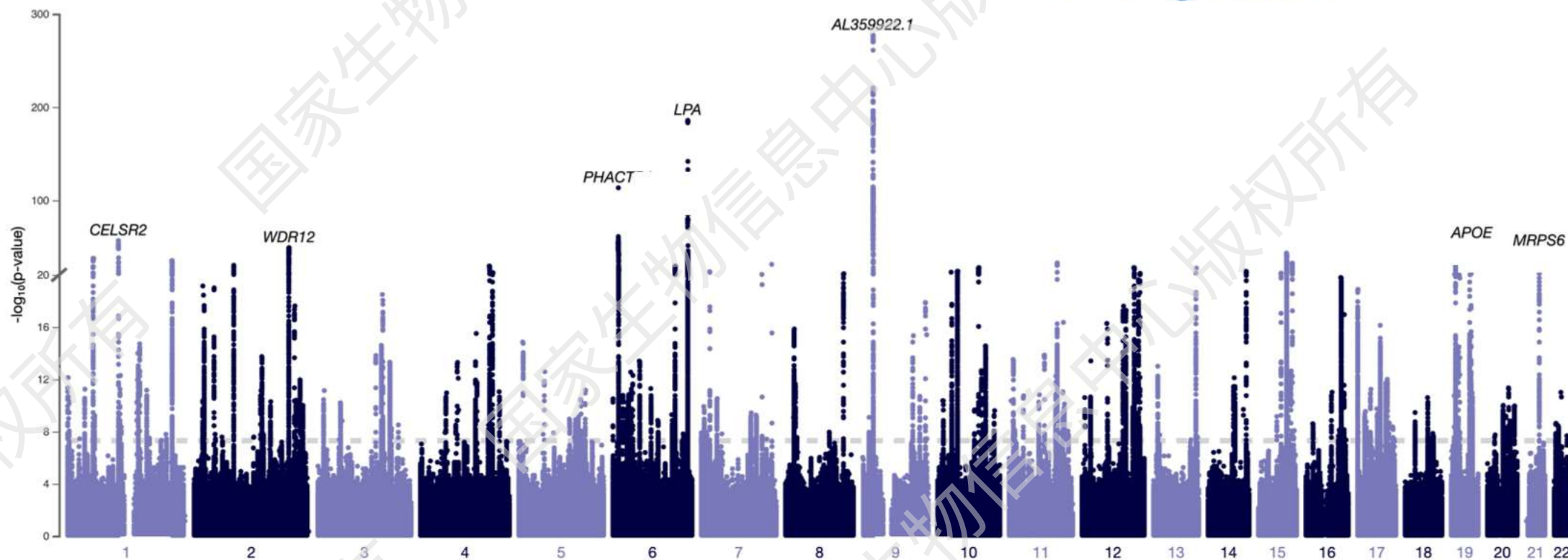
Coronary Artery Disease Genetic Map



High-throughput
genotyping arrays



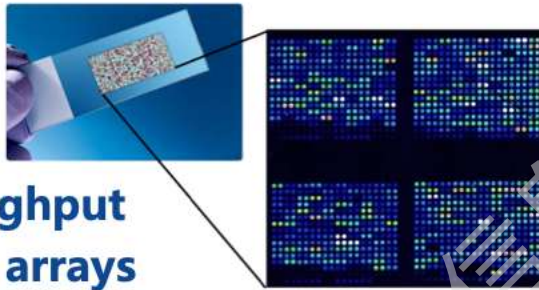
**180,000 CAD cases+ 1.17M
controls GWAS
~ 300 genes**



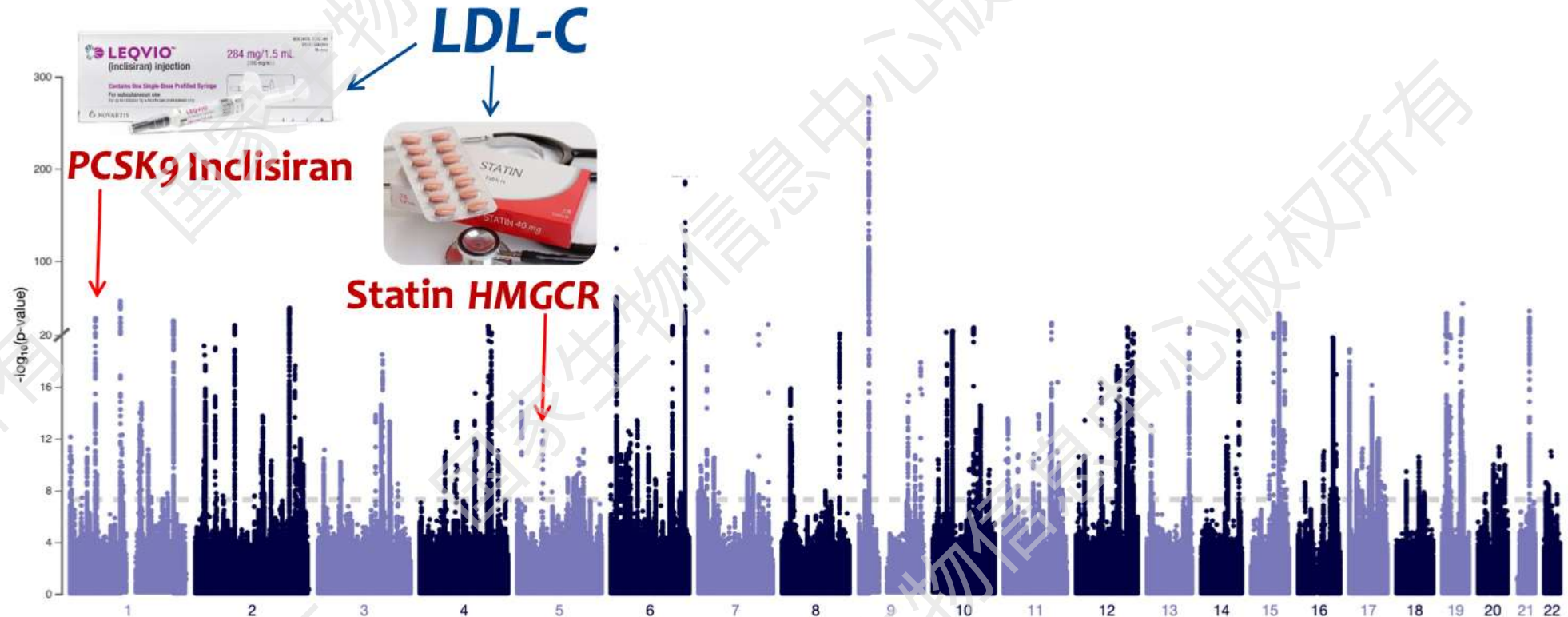
CAD Genetic Map and Medications



High-throughput
genotyping arrays



180,000 CAD cases+ 1.17M
controls GWAS
~ 300 genes



Gene-targeted drugs are revolutionizing CAD treatment and prevention



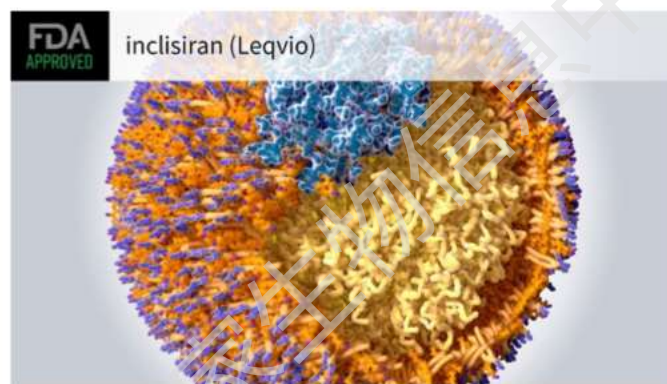
Every day
Statins



Every six months
PCSK9 siRNA

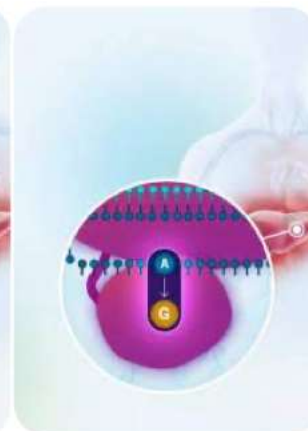


Once in a lifetime
Gene editing



Novartis Leqvio (inclisiran)
Long-term lipid-lowering

NEJM. 2020
medpagetoday.com



The unmet medical needs



**Can we
treat
every CAD
patient
effectively?**

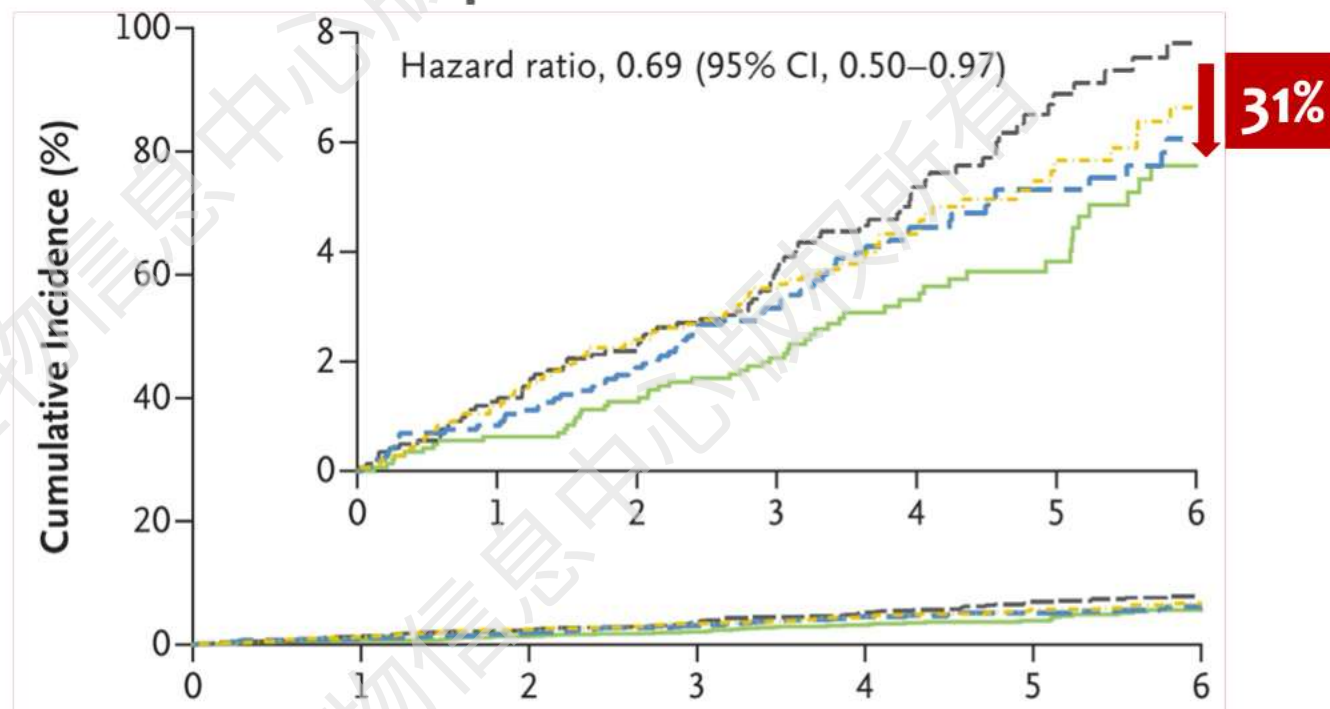
The unmet medical needs



Can we
treat
every patient
effectively?

- Lipid-lowering
- Blood-pressure-lowering ↓ 30~55%
- Antiplatelet

Composite cardiovascular event



5,713 individuals | 4.6 years of follow-up

MEDICAL NEEDS

Identify new genetic drug targets for CAD



Risk Factors for Cardiovascular Disease



Genetic



Monogenic



Somatic mutation



Polygenic



Telomere length



Family history

Environment



Smoking



Drinking



Diet



Aging



Exercise



Sleep

CAD



Biobank Japan

32k cases
146k controls

ALDH2, rs671 (p.Glu504Lys)
Freq 24%, OR = 1.22, P < 6.4⁻⁷⁰

Koyama et al., *Nat Genet*, 2020

50% ~ 60%

Heritability

AF



UK Biobank

30k cases
416k controls

CDH2, Rare LoF Variants
OR = 10.23, P = 5.3⁻³

Zöller B, et al., *Circ Genomic Precis Med*, 2024

60 ~ 62%

Heritability

Challenges in Developing New Drugs for Cardiovascular Disease



Good cholesterol
associated with
reduced
CAD risk

Lilly ~\$800m

>7 years
Phase III
Abandoned

Evacetrapib

CSL ~\$1B

~6 years
Phase III
Failed

CSL-112

MSD ~\$500m

~4 years
Not marketed

Anacetrapib

Pfizer ~\$800m

>6 years
Phase III
Failed

Torcetrapib



Bad cholesterol
associated with
increased
CAD risk

Pfizer ~\$1B

~11 years
Success

Atorvastatin

MSD ~\$600m

~6 years
Success

Ezetimibe

AMGEN >\$1B

~7 years
Success

Evolocumab

NOVARTIS ~\$2B

>7 years
Success

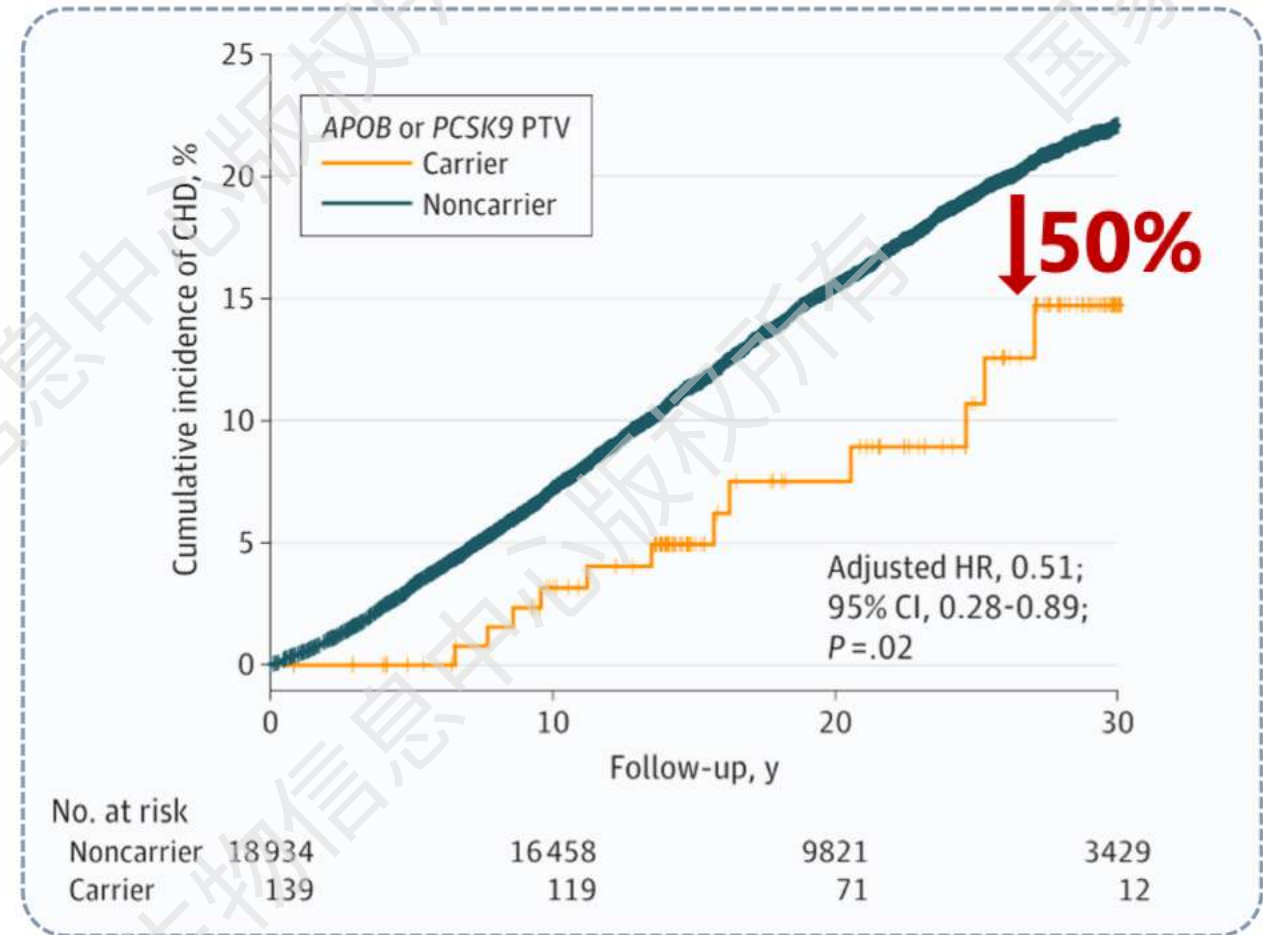
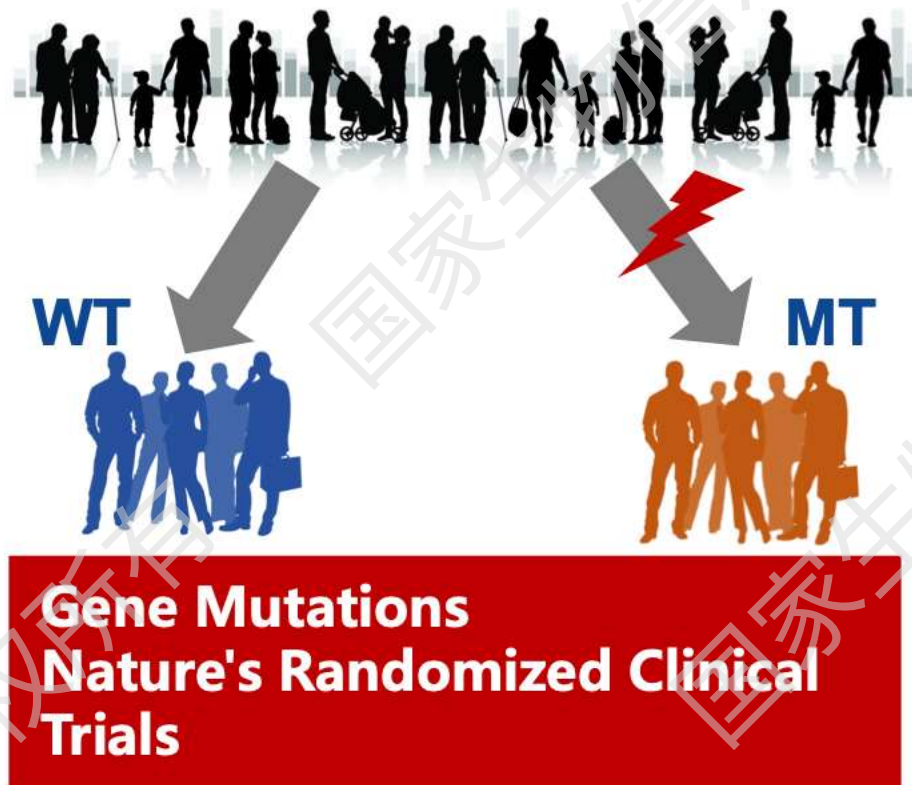
Inclisiran

How to increase clinical trial success rates, shorten development time, and reduce R&D costs?

Identifying gene–disease/phenotype associations



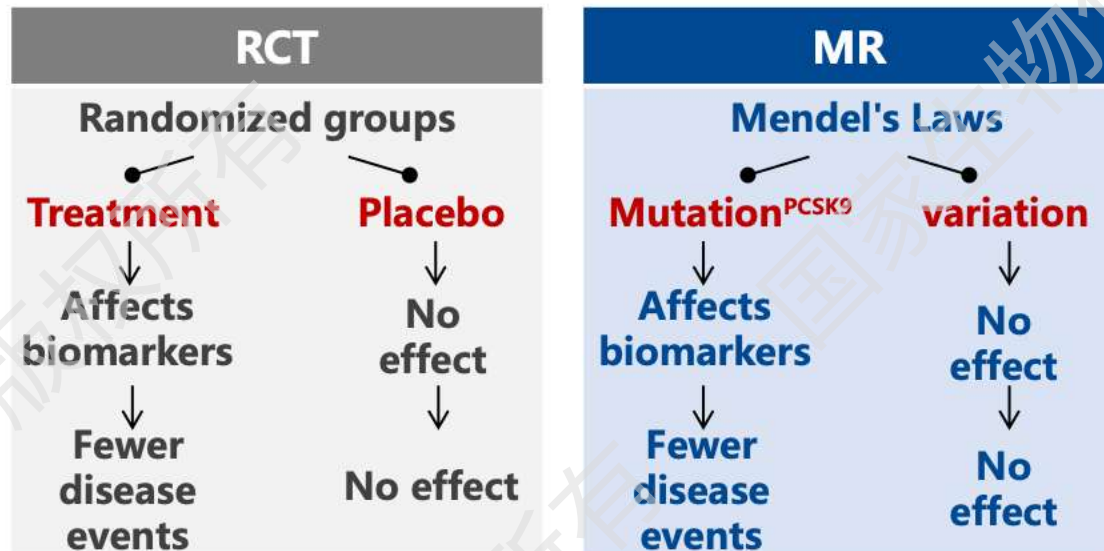
Loss-of-function mutations in PCSK9 Reduce the risk of CAD by 50%



New Opportunities in Drug Development from Population-Scale Genomic Data



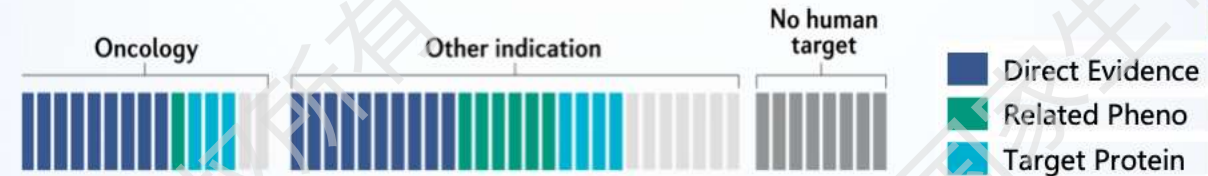
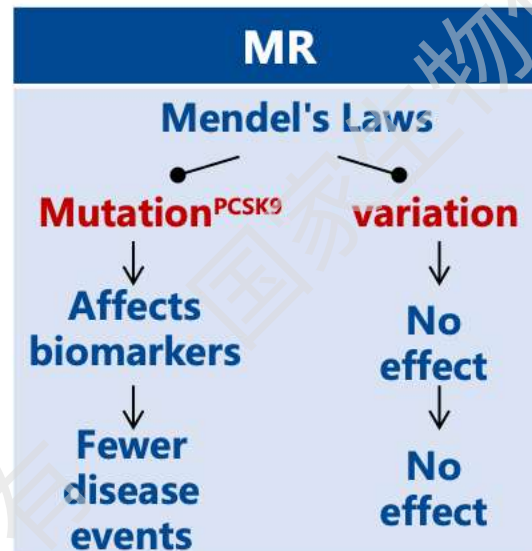
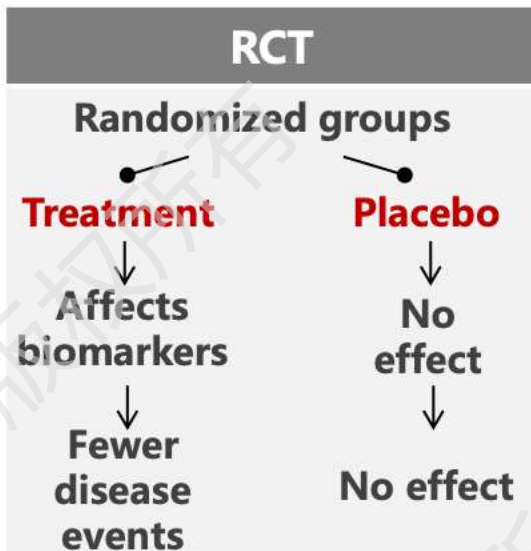
- ❑ Genetics influences the fundamental mechanisms of disease
- ❑ Represents lifelong exposure to risk factors
- ❑ Not affected by confounding factors
- ❑ Helps clarify causal relationships
- ❑ Helps assess safety issues



New Opportunities in Drug Development from Population-Scale Genomic Data



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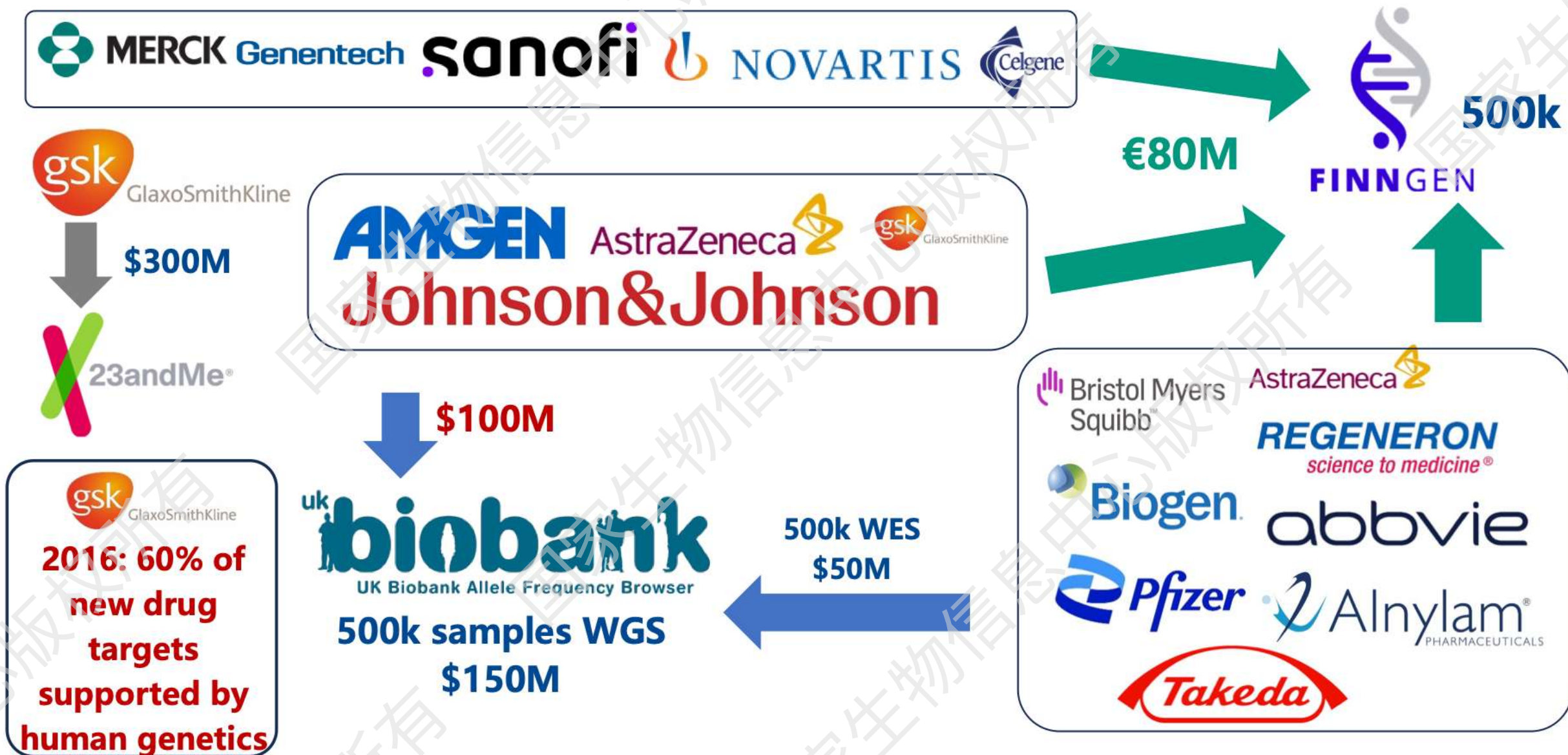
2021 FDA approvals: 50 drugs, ~2/3 supported by genetic evidence

Ochoa et al., Nat Rev Drug Discov. 2022;

Genetic evidence increases drug development success 2–3 fold

Nelson MR, Tipney H, Painter JL, et al. Nat Genet. 2015;
Minikel EV, Painter JL, Dong CC, Nelson MR. Nature. 2024

Top pharma companies invest in population cohorts

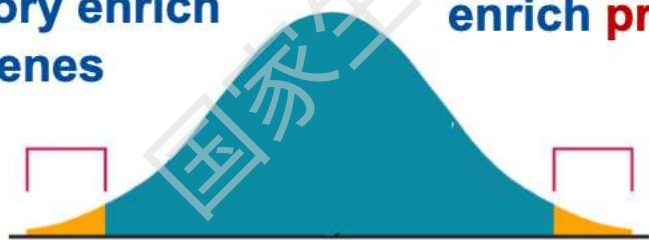




Large families + Extreme phenotype + Large population cohort

Early onset and
family history enrich
risk genes

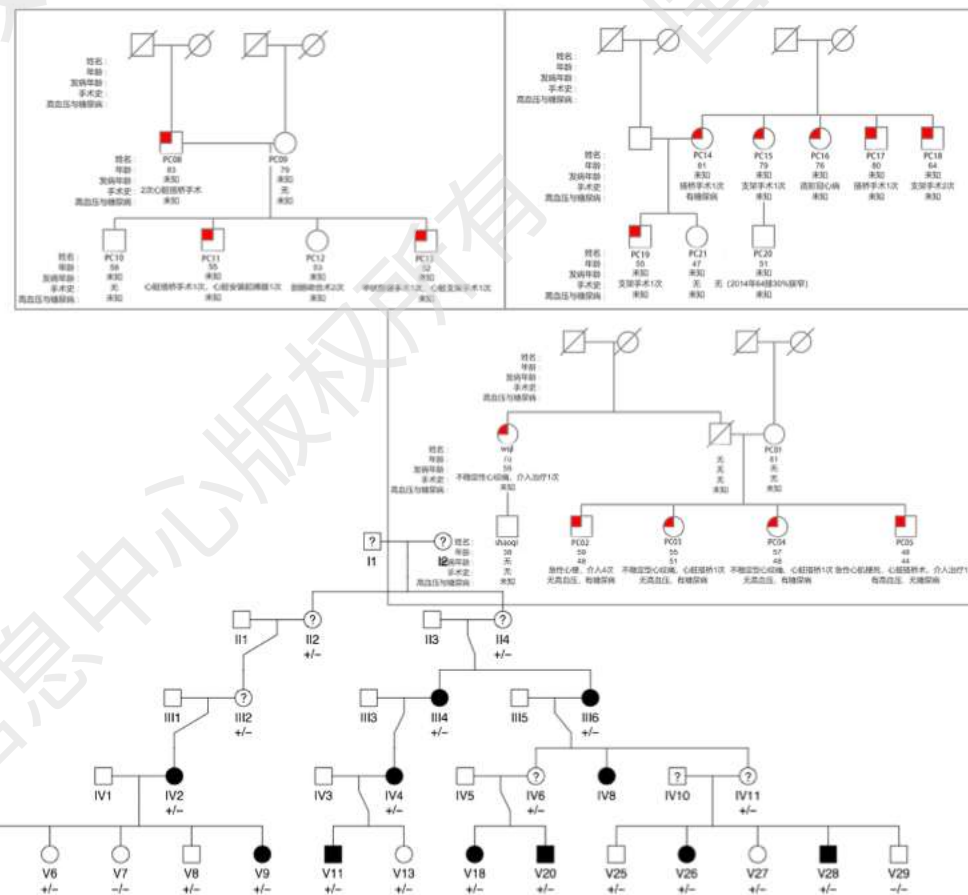
Old health cohort
enrich **protective genes**

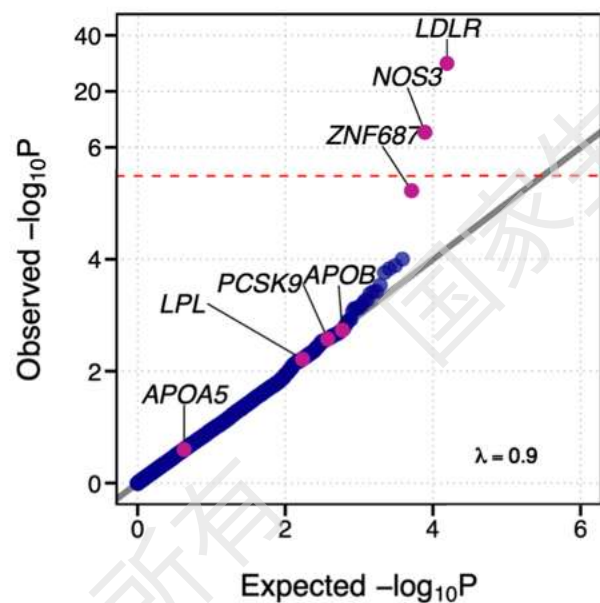


Extreme phenotype cohort

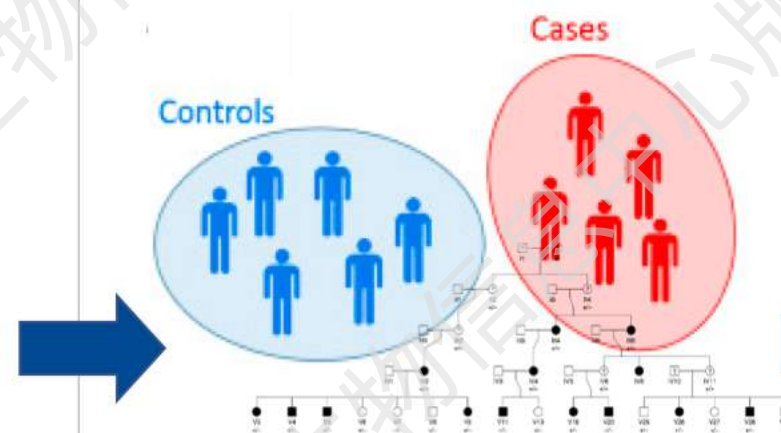
- **Early onset CAD (1k WGS)**
 - **Male ≤ 45 years, female ≤ 45 years, with a family history**
 - Low conventional risk, eg. blood pressure and blood lipids
 - No smoking, no drinking, and with a good lifestyle
- **Old health cohort (1k WGS)**
 - **>75 years, good vascular status, no obvious signs of plaque**

large families





Phase 1
41K CAD cases
217K controls
Circ Genomic Precis Med. 2022



Phase 2
50K CAD cases
510K controls



Phase 3
850K cohort



BROAD
INSTITUTE



MASSACHUSETTS
GENERAL HOSPITAL

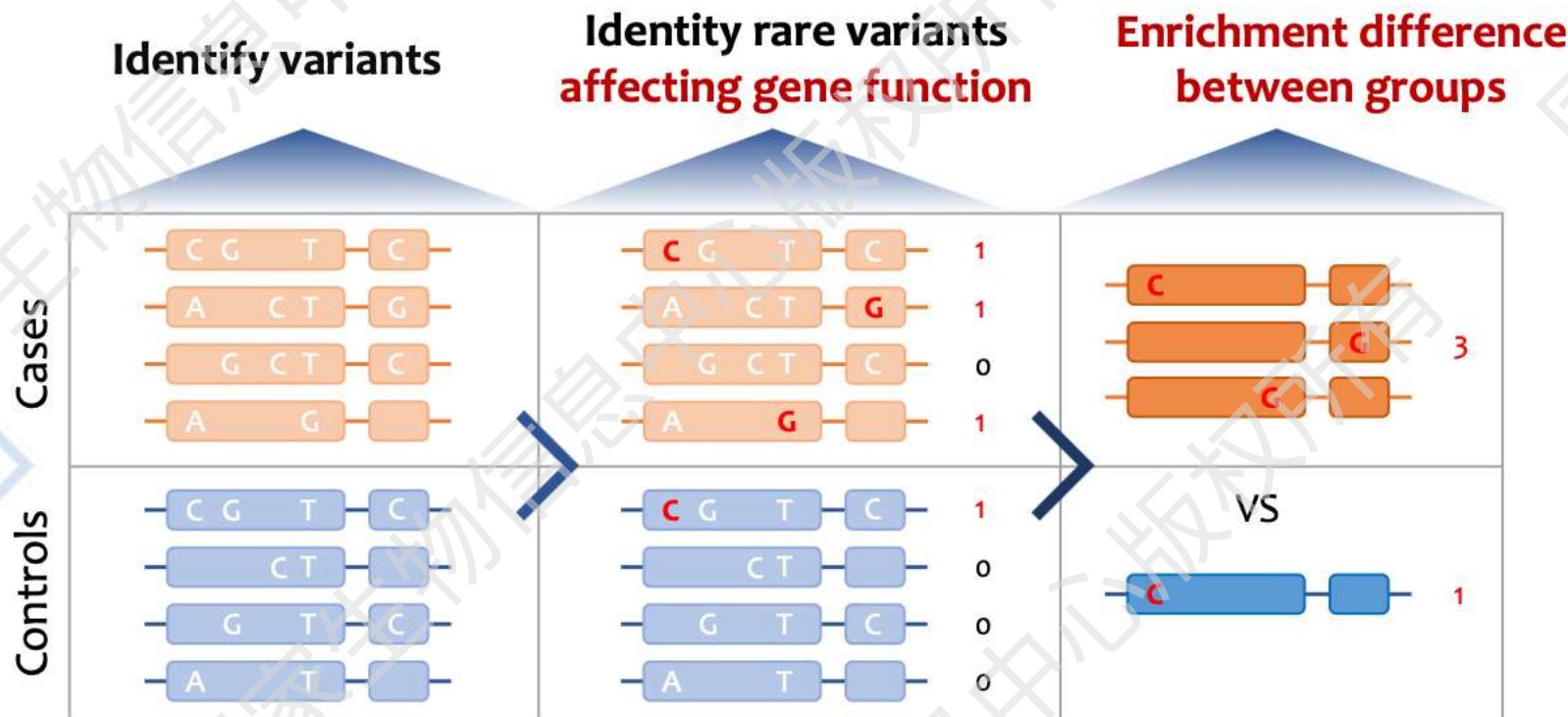
Gene coding variants enrichment analysis



41K
CAD cases



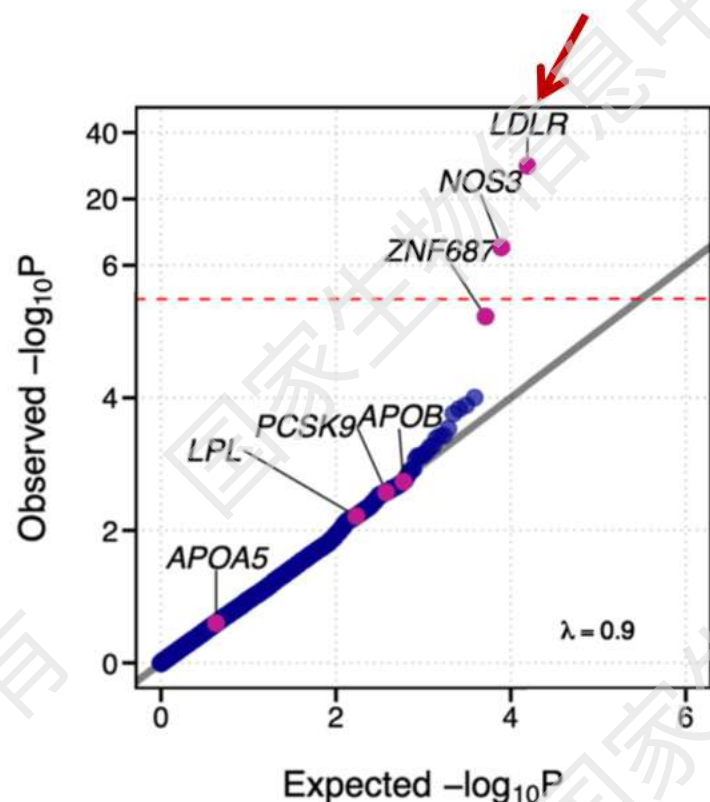
220K
controls



MAF < 1/10000

Circ Genomic Precis Med. 2022

Gene coding variants enrichment analysis



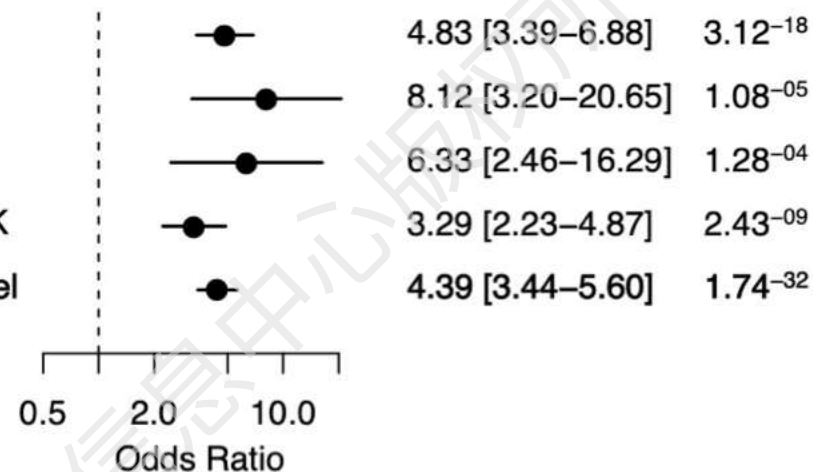
LDLR

- Odds ratio 4.39
- $P = 1.74 \times 10^{-32}$
- 0.91% cases vs 0.34% controls

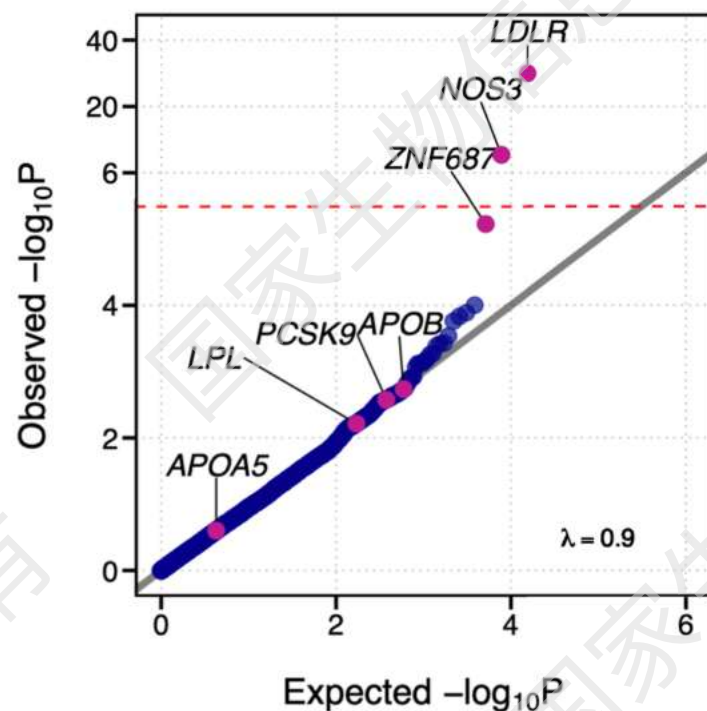
DATA

MI Gen ExSeq
MI Gen WGSseq
UK Biobank 13K
UK Biobank 200K
Fixed effect model

Odds ratio [95CI] P Value



Gene coding variants enrichment analysis



Nitric oxide synthase 3 (NOS3)

- Odds ratio 2.42
- $P = 5.5 \times 10^{-9}$

DATA

MIGen ExSeq

MIGen WGSeq

UK Biobank 200K

UK Biobank 13K

Fixed effect model

Odds ratio [95CI] P Value

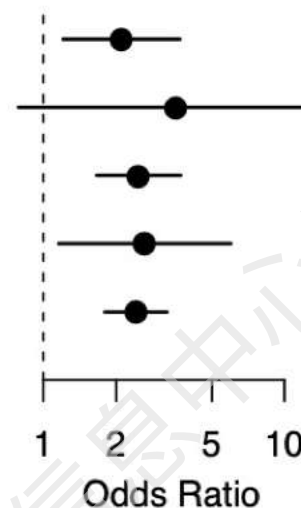
2.11 [1.20–3.68] 8.96×10^{-3}

3.55 [0.79–15.97] 9.90×10^{-2}

2.48 [1.66–3.71] 8.99×10^{-6}

2.63 [1.15–5.99] 2.16×10^{-2}

2.42 [1.80–3.26] 5.50×10^{-9}



20k protein-coding genes: $0.05/20000 = 2.5 \times 10^{-6}$

Molecular phenotypes dissect disease mechanisms

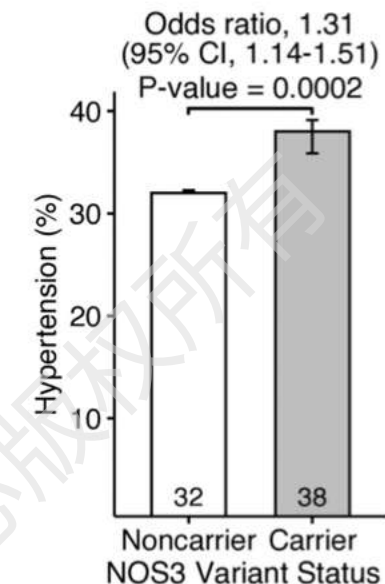


Association between NOS3 rare variants status with lipid and blood pressure biomarkers

TRAIT (unit)	Beta	P-value
LDL cholesterol* (mg/dL)	+ 0.8	0.47
HDL cholesterol* (mg/dL)	- 0.2	0.64
Total cholesterol* (mg/dL)	- 0.4	0.76
Triglycerides* (mg/dL)	- 0.2	0.94
Systolic blood pressure† (mmHg)	+ 3.3	5.0 ⁻⁶
Diastolic blood pressure† (mmHg)	+ 1.3	7.0 ⁻⁴

UKB lipid/blood pressure data of 200,000 individuals

Carriers of NOS3 rare variants
1.31-fold increased risk of hypertension



Pathway biomarkers indicate NOS3 is lipid-independent
Genomic-based patients subtyping and treatment

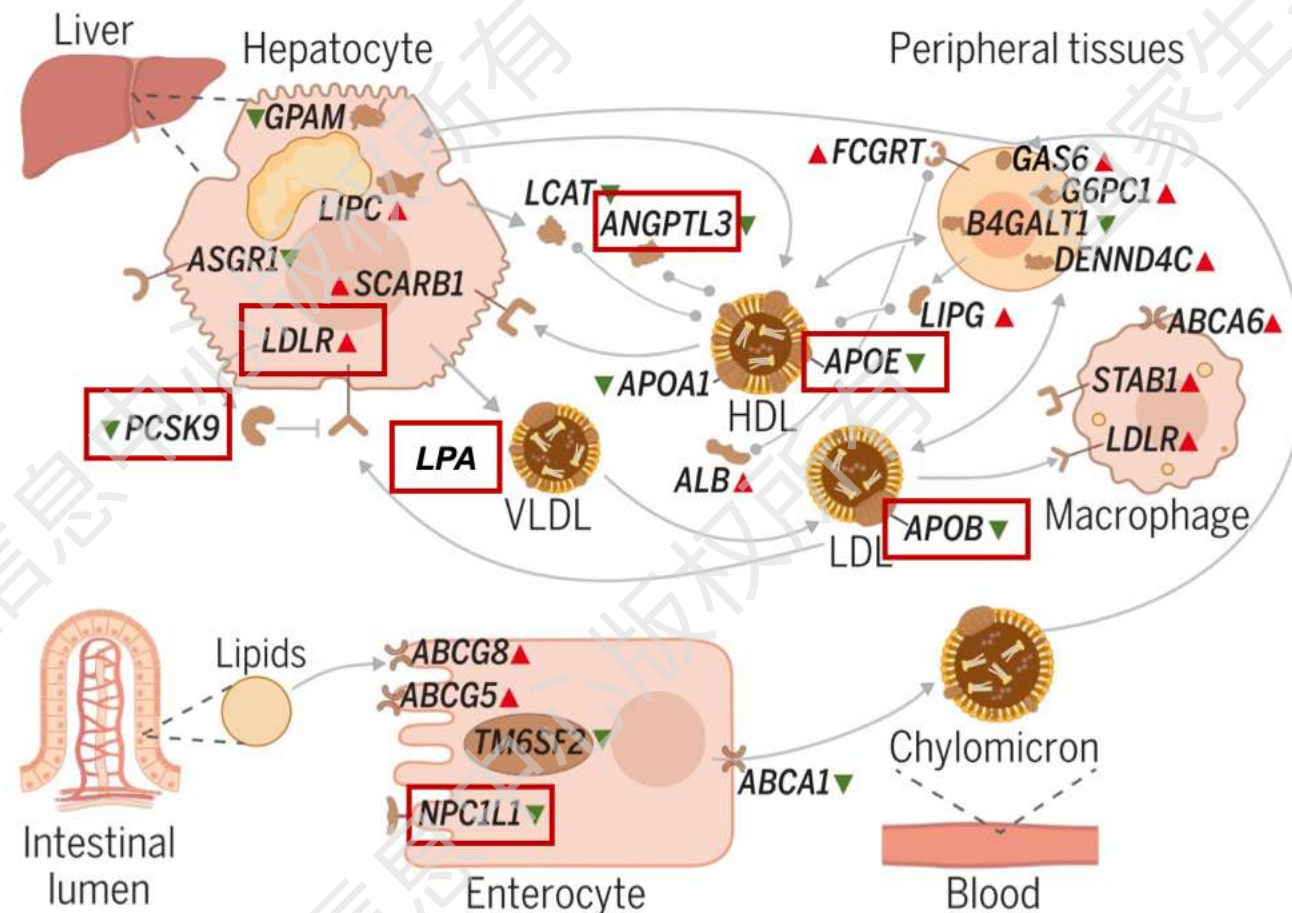
Preliminary results of the phase 2 study



Phase 2
50K CAD cases
510K controls

Rank	Gene	P Value
1	LPA	2.3^{-48}
2	LDLR	2.3^{-27}
3	APOE*	2.3^{-24}
4	PCSK9*	2.3^{-14}
6	LPL*	3.3^{-8}
13	ANGPTL4	6.5^{-7}
20	APOB*	3.3^{-6}
	NOS3	1.8^{-6}
	NPC1L1*	3.3^{-3}

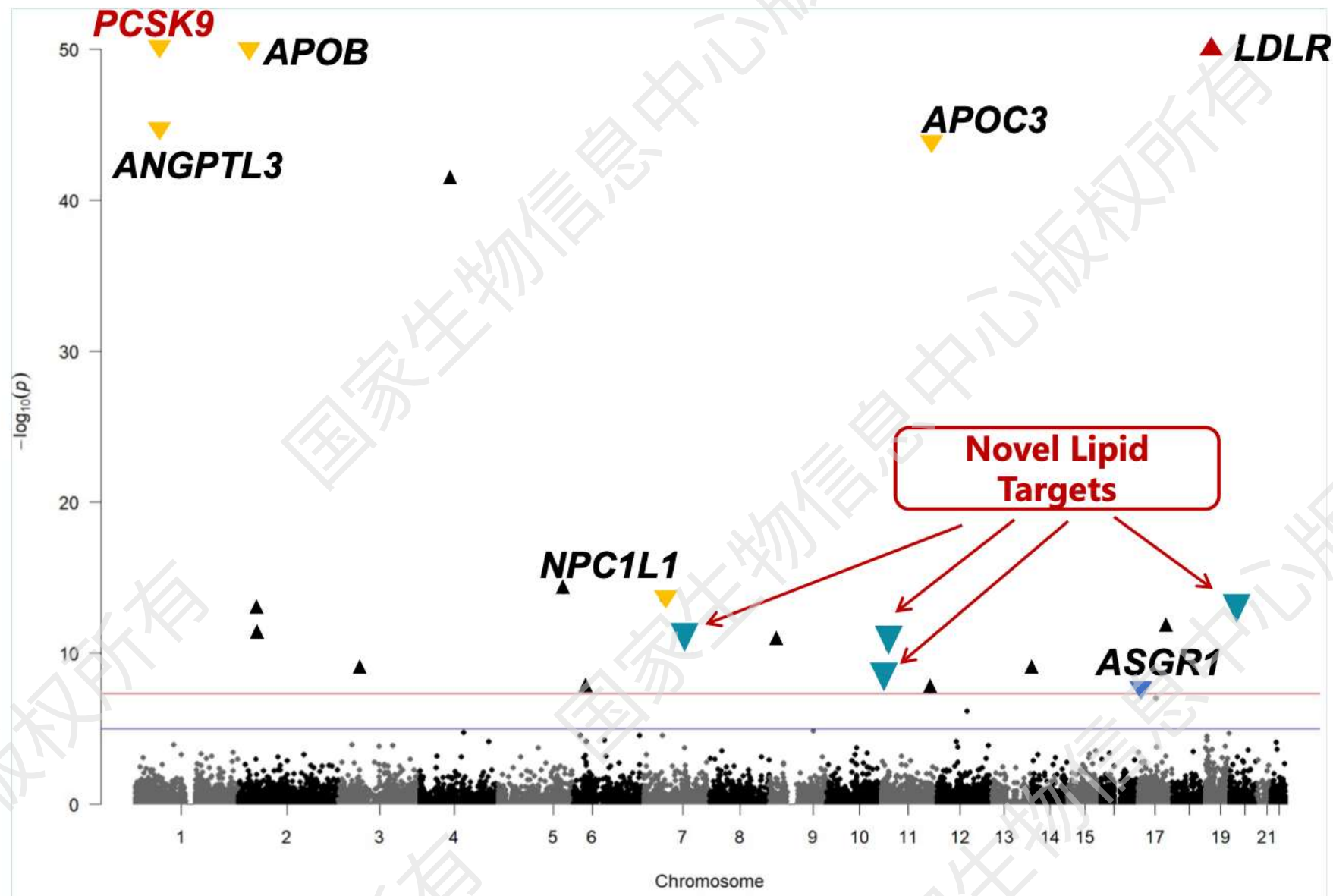
* Genetic variants reduce disease risk
Unpublished



Human lipid pathway

Fiziev P, McRae J, Dron J, et al. *Science*, 2023.

Genetic Map of Low-Density Lipoprotein Cholesterol in 500,000 Individuals



biobank^{uk}

UK Biobank
500k LDL Burden Test
mask1 : lofHC

- ▲ beta > 0, Elevated lipid levels
- ▼ beta < 0, Reduced lipid levels
- Well-established genes
- FDA approved
- Clinical trials
- Novel lipid targets

Unpublished



廖双巧

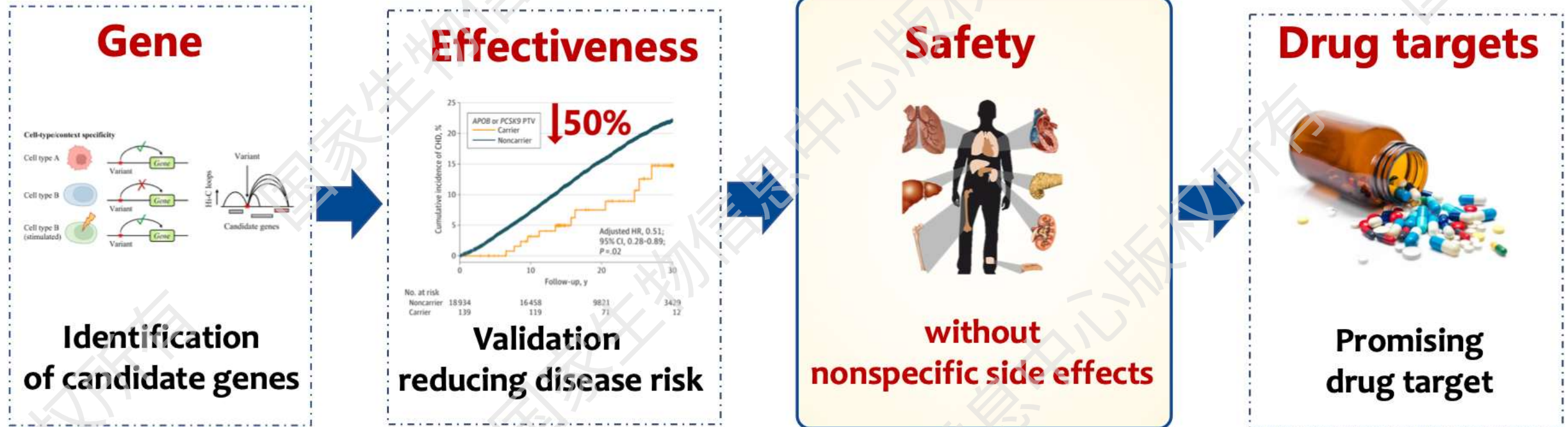
[Rare] protective genetic variants guide the successful development of new drugs



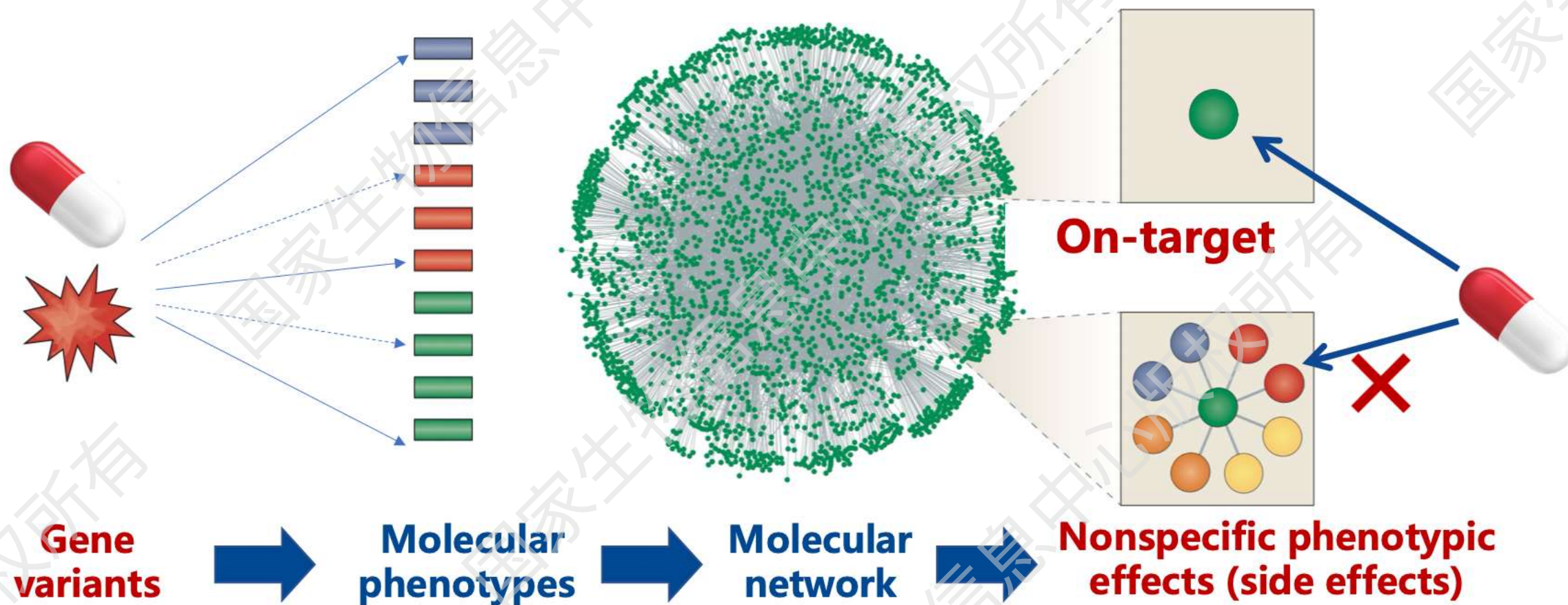
Gene	PCSK9	APOC3	NPC1L1	ANGPTL3
Mutation frequency	1/50	1/150	1/650	1/300
Performance	80% ↓risk	40% ↓risk	53% ↓risk	34% ↓risk
FDA approved Drugs	McAb: Alirocumab, Evolocumab ASO: Inclisiran	ARO-APOC3 (RNAi)	Ezetimibe	EvinaCumab

NEJM. 2015, 2017, 2018, 2021

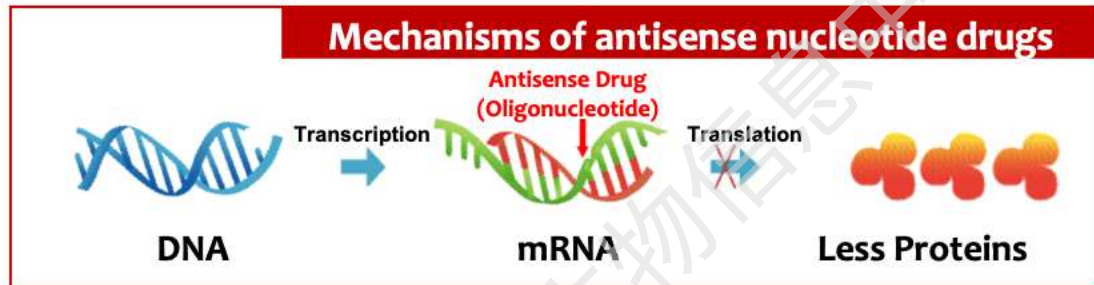
From candidate genes to promising drug targets



Drug safety assessment by molecular phenotypes

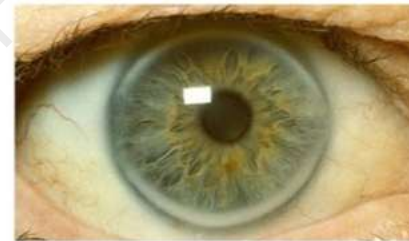


Gene variants mimic drug effects



- **Volanesorsen: FDA-approved treatment for familial chylomicronemia syndrome**
 - Antisense oligonucleotide drugs
 - The target is **APOC3** gene, which reduces triglyceride
 - **Phase 3 study:**
45% with acute thrombocytopenia

Familial Chylomicronemia Syndrome triglyceride abnormality



Drug safety assessment by molecular phenotypes



**UK Biobank whole exome
sequencing of 42K individuals**

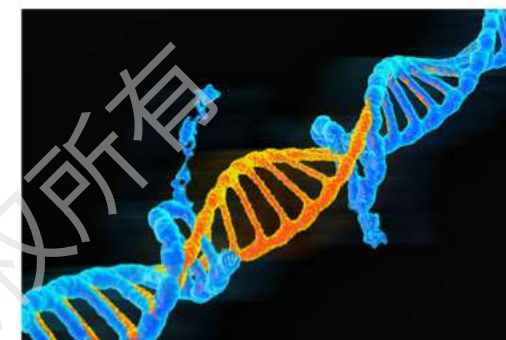


**235 APOC3 carriers
Loss-of-function variant**



Gene mutation mimics drug intervention

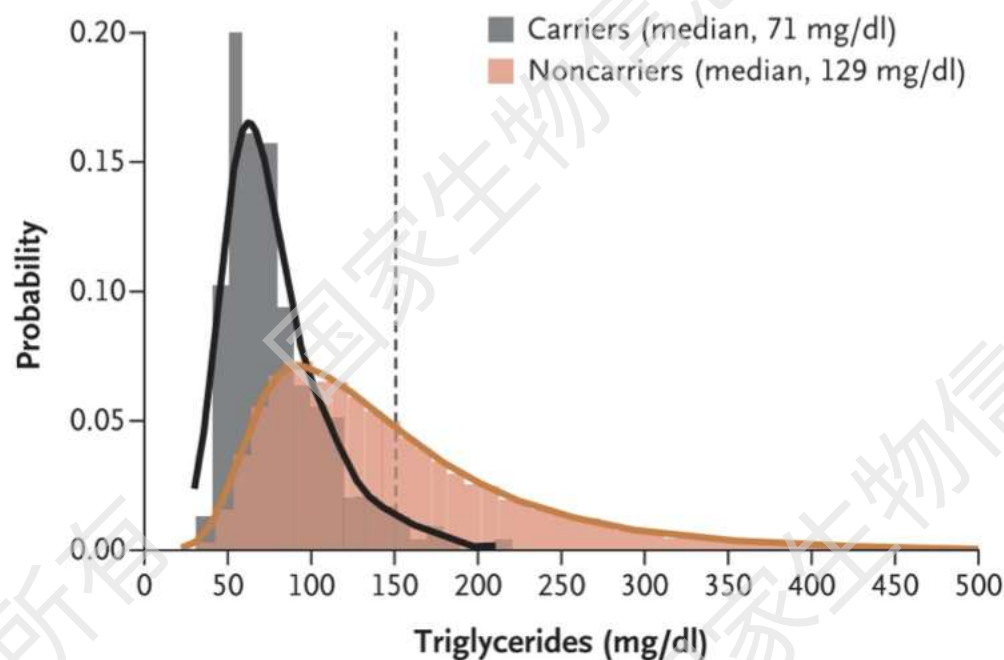
➤ **Gene mutation:
clinical trial by nature**



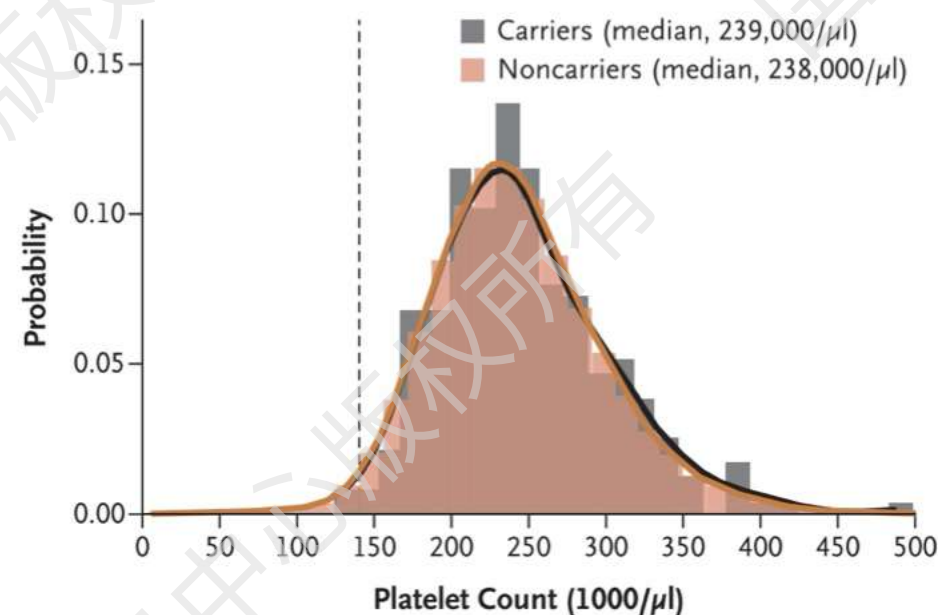
**UK Biobank
10 years follow-up**

New England Journal of Medicine, 2019

APOC3 genetic mutations lower triglyceride levels

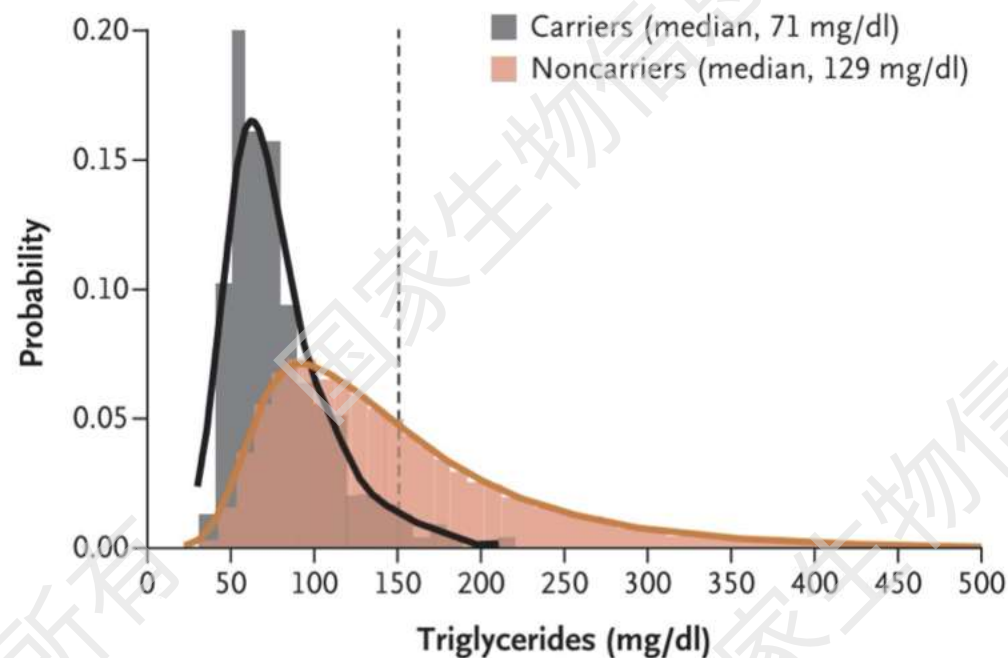


APOC3 genetic mutations and platelet count levels

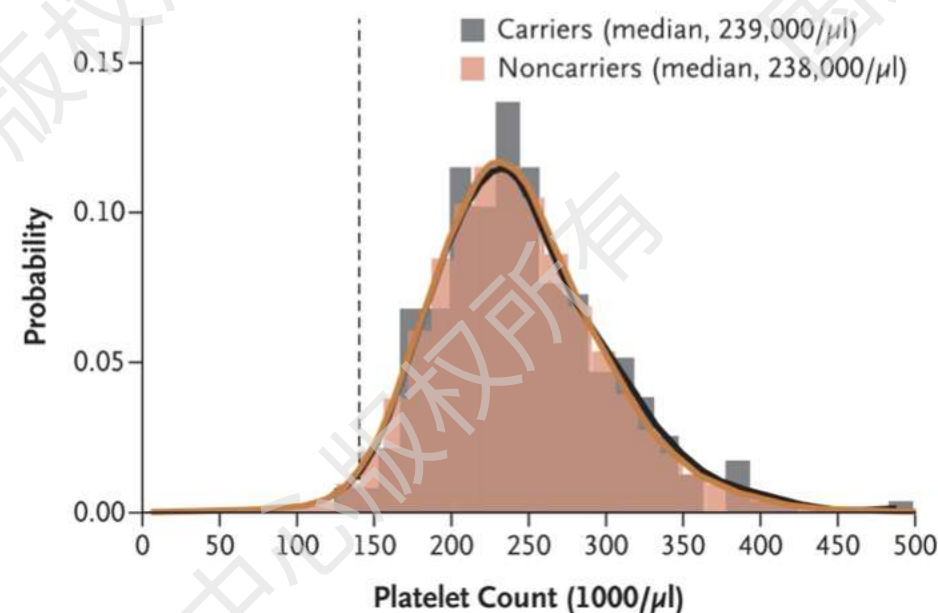


New England Journal of Medicine, 2019

APOC3 genetic mutations lower triglyceride levels



APOC3 genetic mutations and platelet count levels



The side effects of Volanesorsen is from the drug delivery system rather than the target gene

Target gene is safe!

Drug safety assessment by molecular phenotypes



The NEW ENGLAND JOURNAL of MEDICINE

1. Witztum JL, Gaudet D, Freedman SD, et al. Volanesorsen and triglyceride levels in familial chylomicronemia syndrome. *N Engl J Med* 2019;381:531-42.
2. The TG and HDL Working Group of the Exome Sequencing Project, National Heart, Lung, and Blood Institute. Loss-of-function mutations in *APOC3*, triglycerides, and coronary disease. *N Engl J Med* 2014;371:22-31.

DOI: 10.1056/NEJMc1912350

effect and early intervention to mitigate and maintain patients on treatment. We are currently investigating a galactosamine-conjugated *APOC3* oligonucleotide that, in a phase 1-2a study conducted over the course of 3 months, achieved similar levels of triglyceride reduction at lower doses, with no reported changes in platelet counts.³

Reply from drug developers:

Phase 1-2a, modified delivery system with the same target had no side effects

The same approach is applicable to other genetic drug development

The NEW ENGLAND JOURNAL of MEDICINE

ORIGINAL ARTICLE

Olezarsen for Hypertriglyceridemia in Patients at High Cardiovascular Risk

**Phase 2b :
Olezarsen -- Targeting
APOC3 messenger RNA
(ASO), found no safety
concerns**

Prohaska et al, NEJM 2024.



**For a longer,
healthier life**



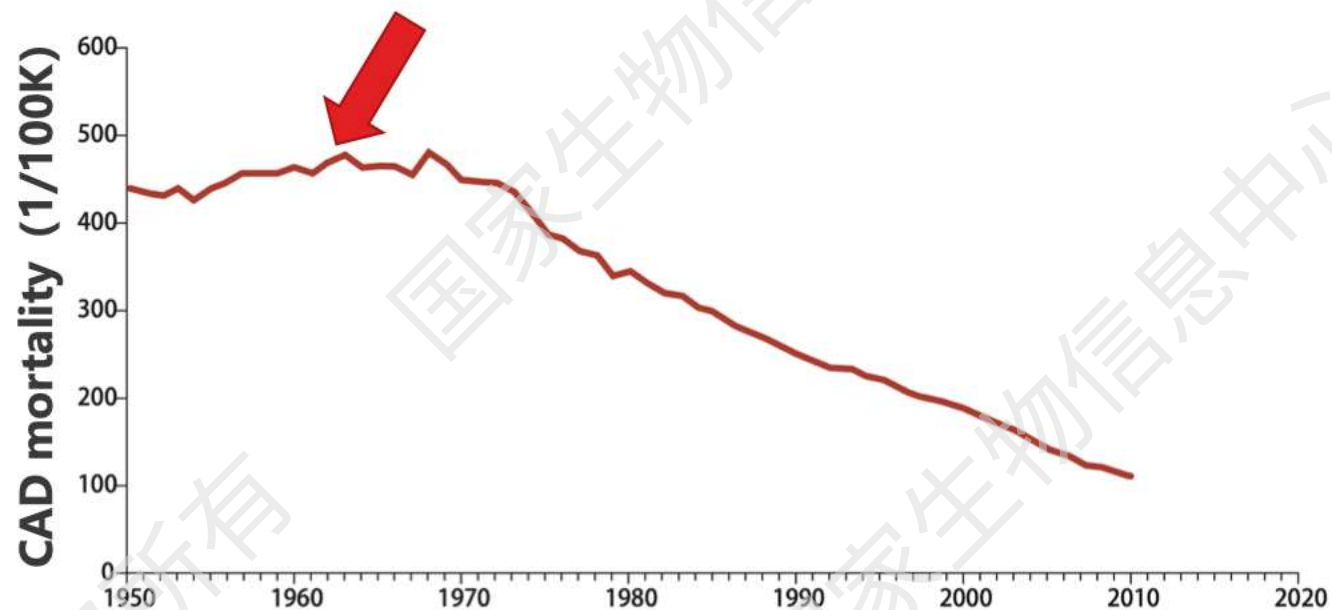
American Heart Association®

Healthy for Good™



**80% of
Cardiovascular diseases are
preventable**

Discovery of CAD risk factors in 1960s



CAD mortality in the United States

Nabel, E. G. & Braunwald, E. N. *Engl. J. Med.* (2012).
Mensah, G. A. et al. *Circulation Research* (2017).



The NEW ENGLAND
JOURNAL of MEDICINE

200th NEJM ANNIVERSARY ARTICLE

The discovery of CAD risk factors and implementation of preventive medicine have led to remarkable achievements in disease prevention!

Preventive Medicine Based on Probabilistic Models



AMERICAN COLLEGE of CARDIOLOGY ASCVD Risk Estimator Plus

Estimate Risk | Therapy Impact | Advice

Current Age ^{*} Sex ^{*} ☐ Male ☐ Female Race ^{*} ☐ White ☐ African American ☐ Other

Systolic Blood Pressure (mm Hg) ^{*} Diastolic Blood Pressure (mm Hg)

Total Cholesterol (mg/dL) ^{*} HDL Cholesterol (mg/dL) ^{*}

History of Diabetes? ^{*} ☐ Yes ☐ No Smoker? ^{*} ☐ Current ☐ Former ☐ Never

On Hypertension Treatment? ^{*} ☐ Yes ☐ No On a Statin? ^{*} ☐ Yes ☐ No On Aspirin Therapy? ^{*} ☐ Yes ☐ No

$$f(x) = (3.514) \log(\text{age}) + (0.672) \log(\text{Total} - C) + (0.481) \log(\text{LDL} - C) + (-0.269) \log(\text{HDL} - C) + (0.478) \log(\text{treatedSBP}) + (0.446) \log(\text{untreatedSBP}) + (0.482) \text{smoking} + 9.046(\text{DM}) + (0.125) \log(\text{UACR}) + (-0.008)(\text{eGFR}) + (-2.118)(\log(\text{age}) * \text{DM})$$

<https://tools.acc.org/ascvd-risk-estimator-plus/#!/calculate/estimate/>

ACC/AHA Clinical Guidelines :
PCE 10-Year ASCVD risk > 7.5%
Recommended Drug Interventions

10-year risk for ASCVD

Low-risk (<5%)

Borderline risk (5% to 7.4%)

* Intermediate risk (7.5% to 19.9%)

* High risk (≥20%)

Preiss, D. & Kristensen, S. L. Can. J. Cardiol. (2015).
2013 ACC/AHA Guideline on the Assessment of Cardiovascular Risk.

Huangdi Neijing: "Superior doctors prevent illness before it occurs."

ACC/AHA PCE: American College of Cardiology/American Heart Association Pooled Cohort Equations

New concept of disease precision prevention



Disease treatment
↓
Health management

Early identification of individuals at high risk of disease, especially before the onset of disease, through targeted lifestyle interventions and cholesterol-lowering medications can reduce lifetime disease risk!

- **How can we effectively identify high-risk individuals?**

Monogenic Genetic Diagnosis and Disease Risk Stratification



**Olympic Volleyball Star
Flo Hyman**

Marfan syndrome
FBN1 mutation + Sudden Cardiac
Death



2013 Angelina Jolie
BRCA1 + Family history

Mother died from ovarian
cancer at age 56

Genetics in Medicine (2022) 24, 1407–1414



Genetics
in
Medicine

www.journals.elsevier.com/genetics-in-medicine

ACMG STATEMENT

ACMG SF v3.1 list for **reporting of secondary findings** in clinical exome and genome sequencing: A policy statement of the American College of Medical Genetics and Genomics (ACMG)

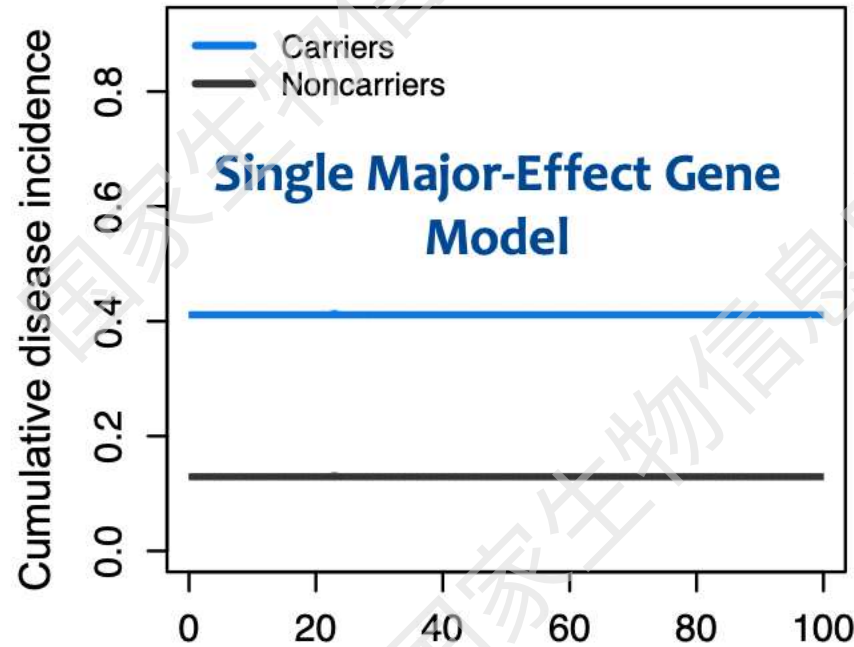
Miller DT, et al. *Genet Med.* 2023;25(8)



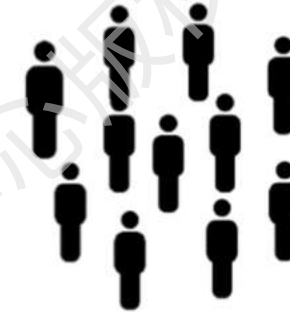
➤ 81 clinically actionable gene list

Phenotype	Gene
...	...
Familial hypercholesterolemia	LDLR
	APOB
	PCSK9
Marfan syndrome, Loeys-Dietz syndromes, and familial thoracic aortic aneurysms and dissections	FBNI1
	TGFBR1
	TGFBR2
	SMAD3
	ACTA2
	MYH11
...	...

Monogenic Genetic Diagnosis and Disease Risk Stratification

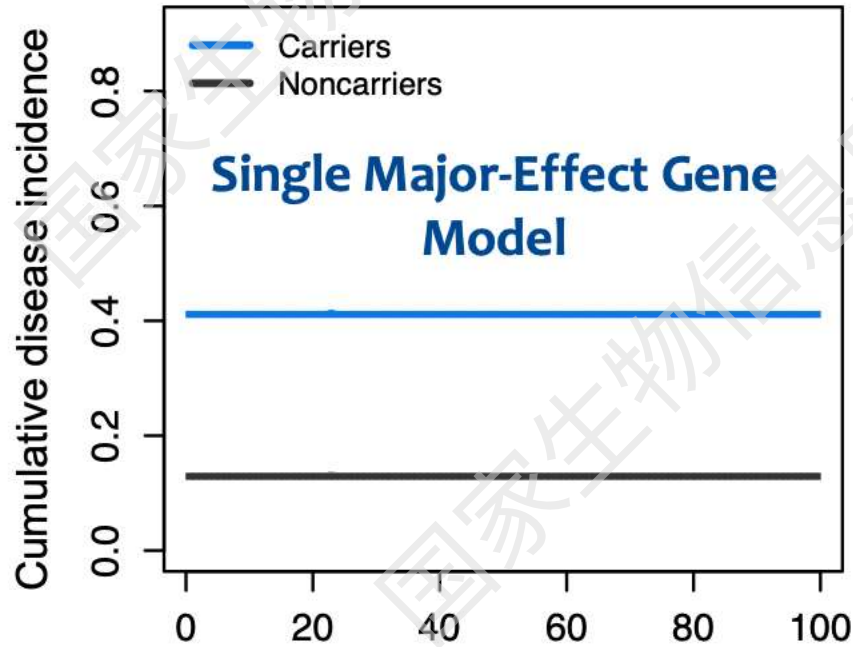


**High risk
Carriers**



**Average risk
Noncarriers**

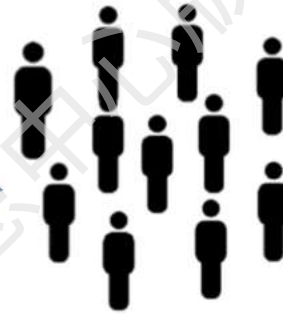
Monogenic Genetic Diagnosis and Disease Risk Stratification



High risk Carriers

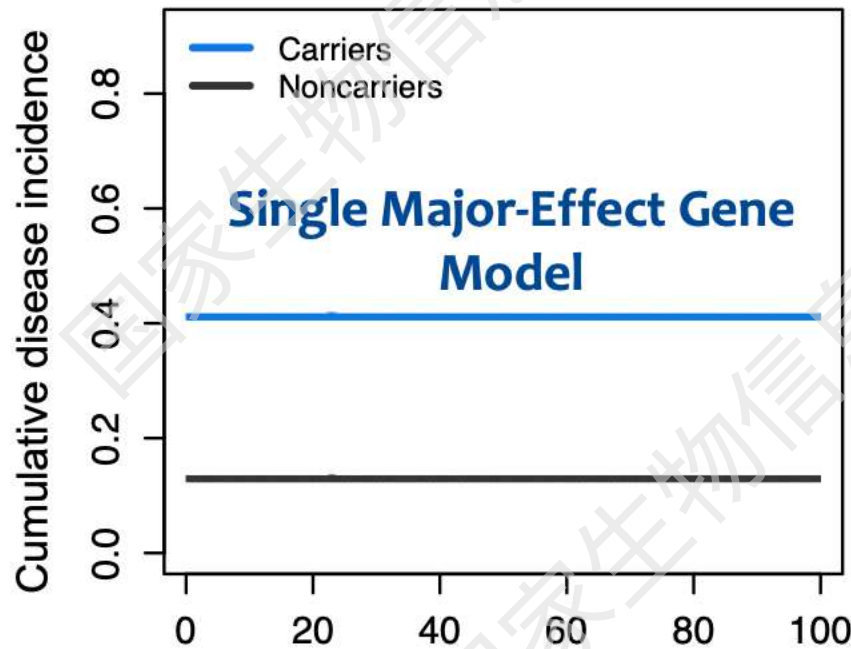


- Population Explained
- Incomplete Penetrance



Average risk Noncarriers

➤ Monogenic diagnosis has limited utility in complex diseases — CAD



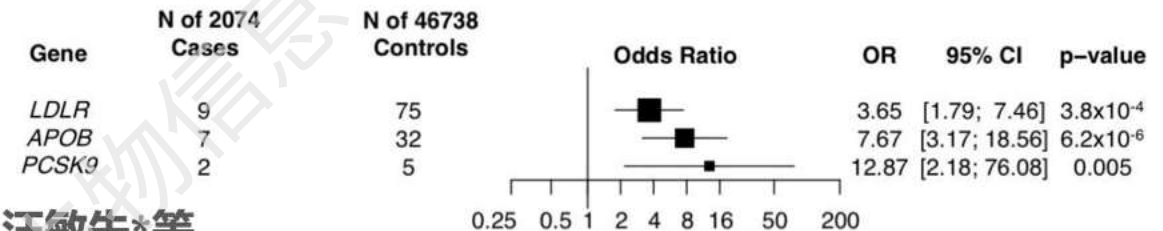
LDLR | 41k CAD cases and 220k controls

- Odds ratio 4.39
- $P = 1.74 \times 10^{-32}$
- **0.91% cases vs 0.34% controls**

DATA		Odds ratio [95CI]	P Value
MIGen ExSeq		4.83 [3.39–6.88]	3.12×10^{-18}
MIGen WGSseq		8.12 [3.20–20.65]	1.08×10^{-5}
UK Biobank 13K		6.33 [2.46–16.29]	1.28×10^{-4}
UK Biobank 200K		3.29 [2.23–4.87]	2.43×10^{-9}
Fixed effect model		4.39 [3.44–5.60]	1.74×10^{-32}



汪敏先*等,
Circ Genomic Precis Med. 2022



汪敏先*等

Nature Communications. 2020

➤ Monogenic diagnosis has limited utility in complex diseases — AF



Gene	Case Carrier	Control Carrier	Percentage	Odds Ratio	95%CI	P-value	Total
<i>TTN</i>	571	4149	1.1344%	1.66	1.52-1.81	1.09E-26	4720
<i>RPL3L</i>	343	2722	0.7366%	1.57	1.39-1.77	1.36E-12	3065
<i>PKP2</i>	75	485	0.1346%	1.76	1.45-2.14	2.17E-07	560
<i>C10orf71</i>	45	305	0.0841%	1.96	1.41-2.74	1.95E-04	350
<i>CTNNA3</i>	41	193	0.0562%	2.51	1.74-3.62	4.71E-06	234
<i>KDM5B</i>	34	196	0.0553%	2.58	1.75-3.80	1.08E-05	230
<i>MYBPC3</i>	30	175	0.0493%	2.75	1.81-4.17	1.40E-05	205
<i>RBM20</i>	25	147	0.0413%	2.18	1.39-3.44	1.66E-03	172
<i>PPA2</i>	15	115	0.0312%	2	1.03-3.88	3.96E-02	130
<i>PDLIM3</i>	15	95	0.0264%	2.22	1.09-4.52	2.80E-02	110
<i>DES</i>	16	85	0.0243%	2.99	1.70-5.26	5.55E-04	101
<i>FLNC</i>	14	79	0.0224%	2.71	1.25-5.87	1.13E-02	93
<i>PRKAG2</i>	11	63	0.0178%	2.72	1.13-6.54	2.51E-02	74
<i>ACTN2</i>	12	62	0.0178%	2.67	1.37-5.19	7.48E-03	74
<i>GATA5</i>	9	40	0.0118%	2.32	1.29-4.17	1.00E-02	49
<i>ACTC1</i>	6	33	0.0094%	3.6	1.04-12.49	4.36E-02	39
<i>GATA6</i>	6	18	0.0058%	4.8	1.81-12.74	4.40E-03	24
<i>TGFB3</i>	5	11	0.0038%	7.13	2.13-23.87	3.32E-03	16
<i>WWTR1</i>	3	11	0.0034%	3.5	1.19-10.27	9.65E-03	14
<i>NKX2-5</i>	4	8	0.0029%	7.5	1.91-29.44	6.01E-03	12
<i>CDH2</i>	3	6	0.0022%	10.23	2.26-46.22	5.31E-03	9

Polygenic Risk Score (PRS)



MIT
Technology
Review

Genetic Fortune-Telling



“polygenic risk score” (PRS)

Next-gen individual health
prediction



Method innovations

25k

100k

1M

5M



- \$38million for 5 Years,
- Consortium Members



- \$75million for 5 Years,
- RRS for 10 conditions,
- return results to 25,000 participants.

Wisdom

Women Informed to
Screen Depending on
Measures of Risk (Wisdom
Study) (WISDOM)

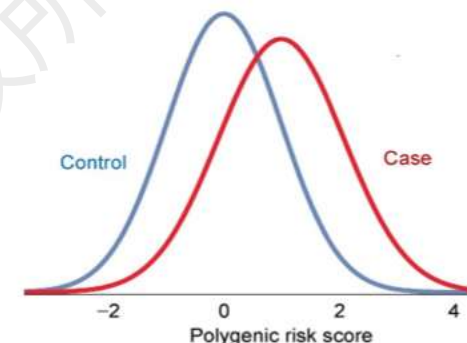


- [ClinicalTrials.gov Identifier: NCT02620852](https://clinicaltrials.gov/ct2/show/study/NCT02620852),



Accelerating detection
of disease (ADD)
challenge program

- improve risk prediction, early detection and early intervention,
- By 2023 to 2024, return PRS for 5 million participants.



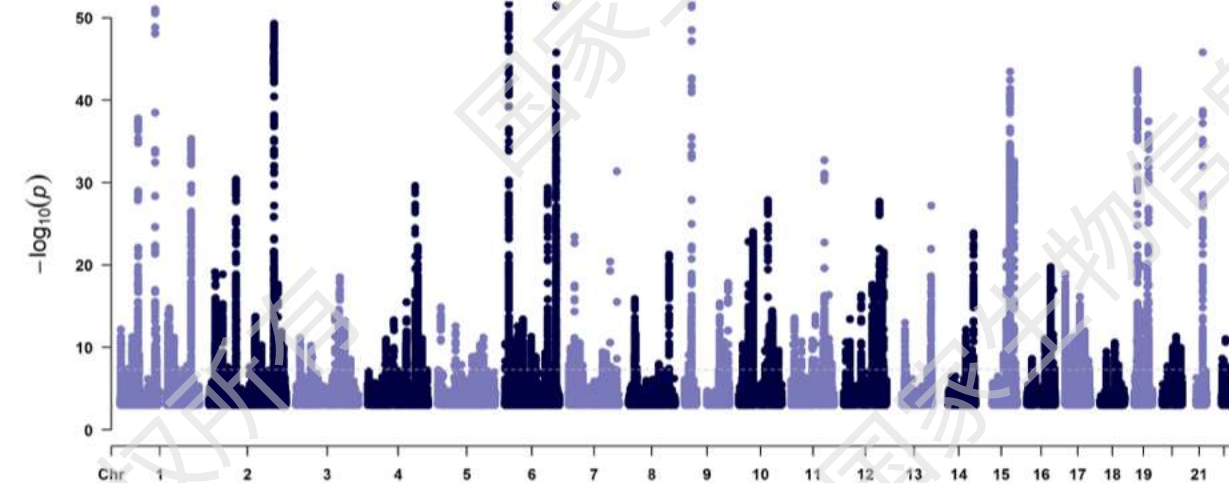
$$PRS_i = \sum_j^M \hat{\beta}_j \times dosage_{ij}$$

Early warning across common complex diseases , “Measure once, benefit for life”

Genetic Maps for CAD and AF



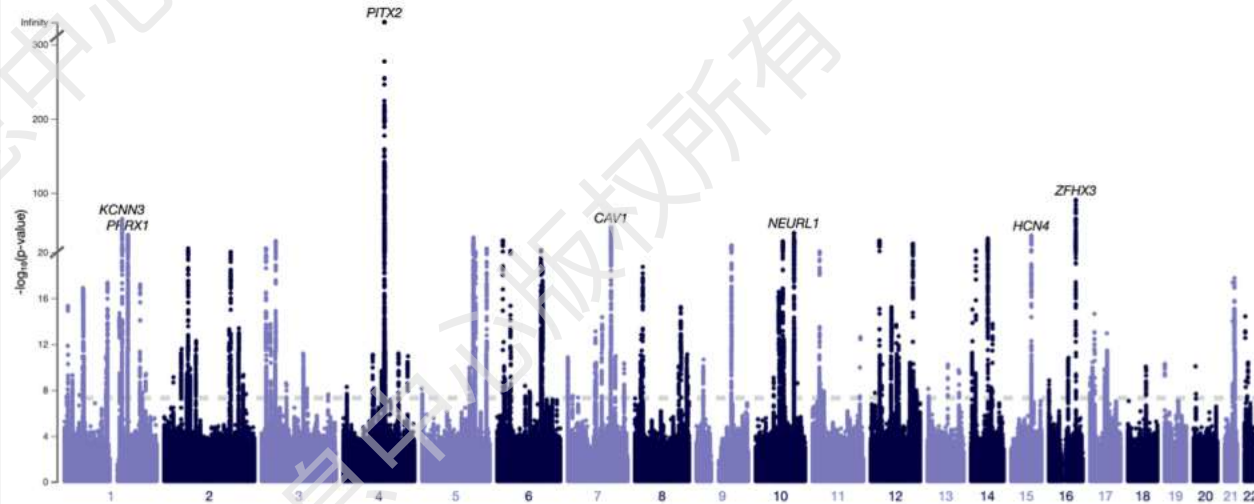
180k CAD cases + 1,170k controls
300 independent genetic loci
identified by GWAS



Aragam KG, et al., *Nat Genet.* 2022



85k AF cases + 540k controls
>300 independent genetic loci
identified by GWAS



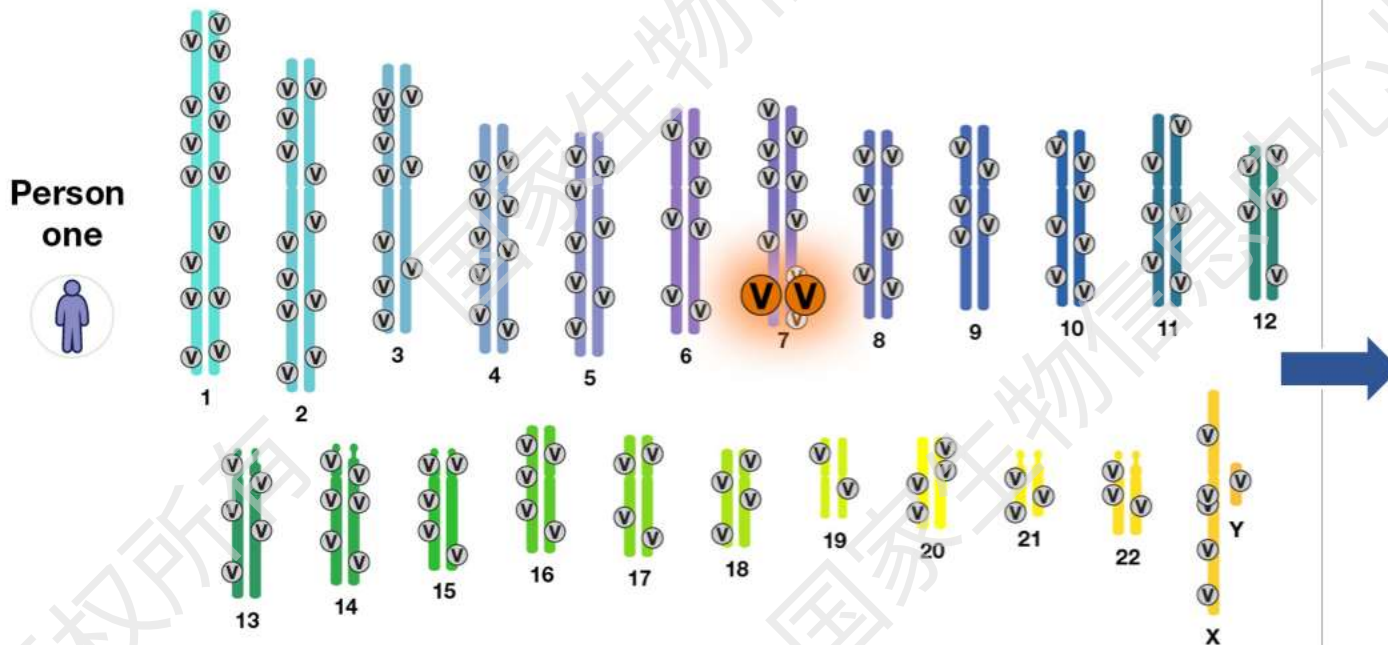
Anurag V., et al., *Science.* 2024

Polygenic Risk Score (PRS)



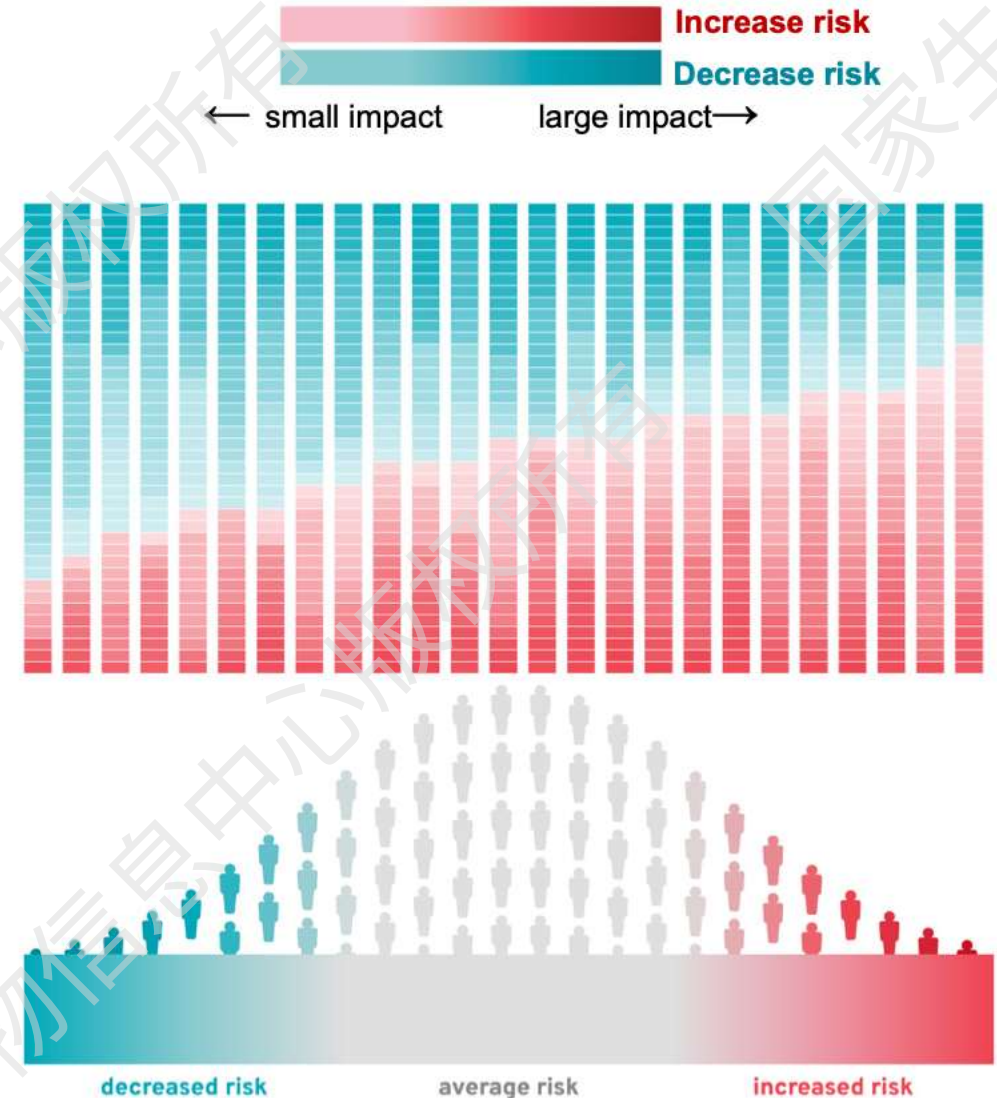
CAD GWAS

180K CAD cases + 1.2M controls
identified 300 associated variants

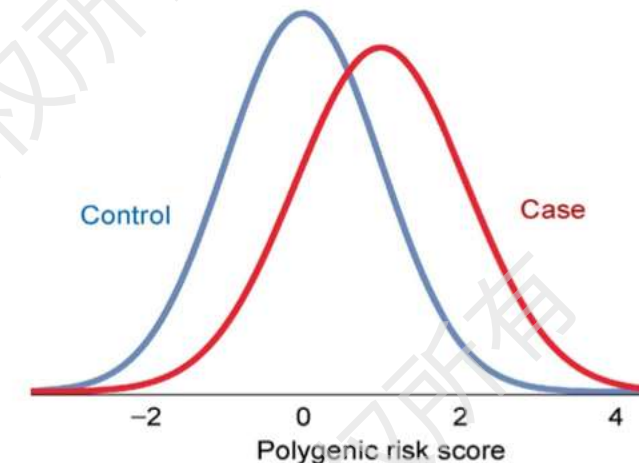
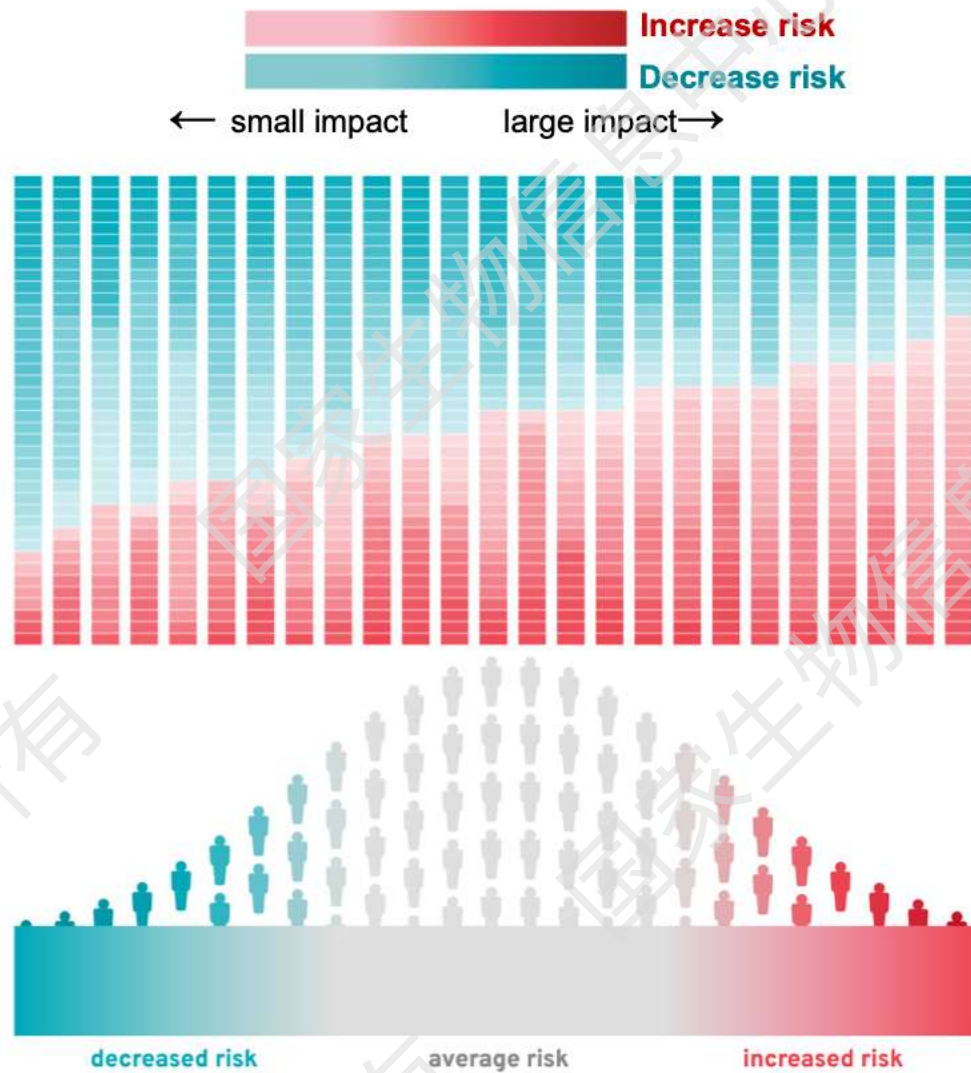


Letter 'V' for disease-associated mutations.

Rare large-effect mutations and common variants with small effects.



Polygenic Risk Score (PRS)



$$PRS_i = \sum_j^M \hat{\beta}_j \times \text{dosage}_{ij}$$

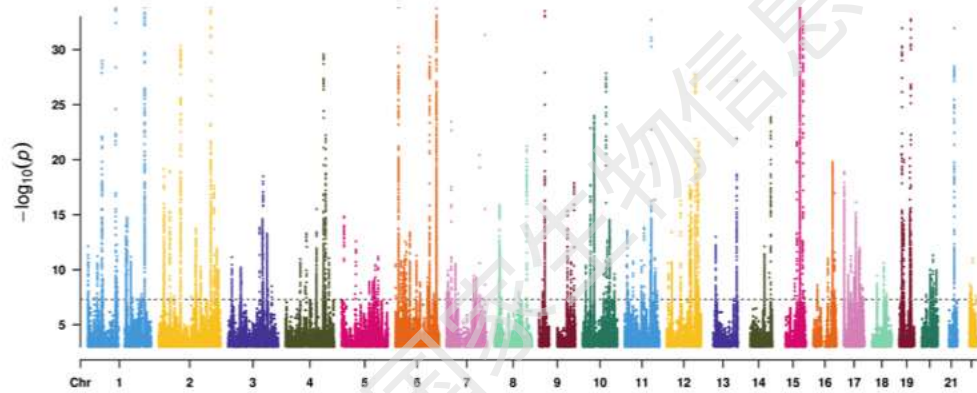
Effect size

#Effective alleles

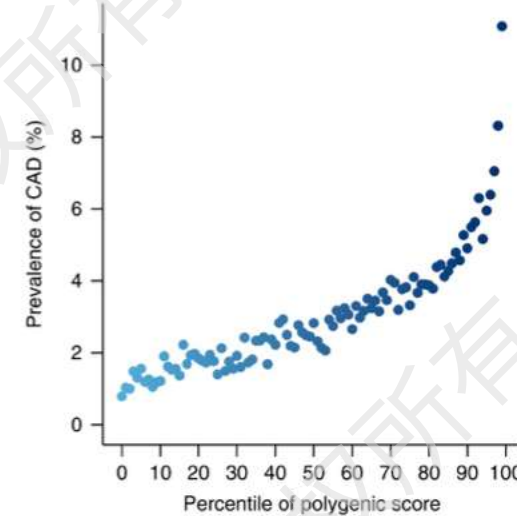
Polygenic Risk Score (PRS)



CAD GWAS



CAD risk



$$PRS_i = \sum_j^M \hat{\beta}_j \times \text{dosage}_{ij}$$

Effect size

No. of Effective alleles

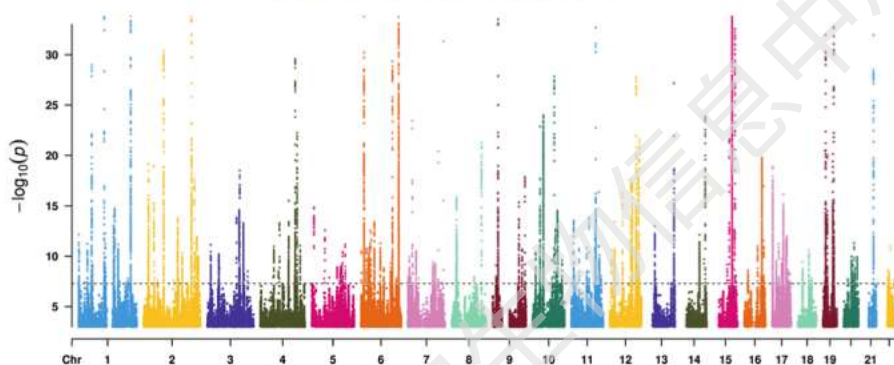
**Traditional methods
Predict CAD risk of by CAD GWAS weight**

Khera A V., et al. *Nat Genet.* 2018

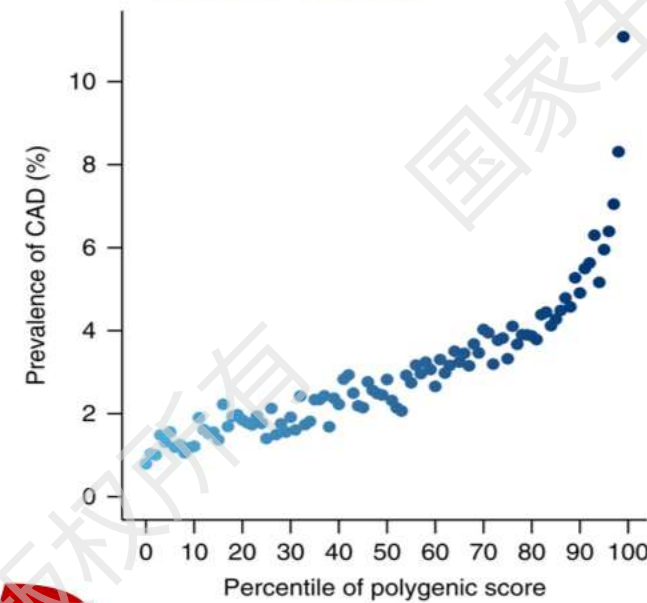
Polygenic risk model for CAD -- GPS 2.0



CAD GWAS



CAD risk



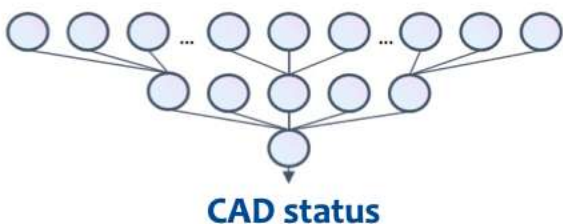
Risk factors and comorbidities



Polygenic risk model for CAD -- GPS 2.0



Two-layer neural network



GWAS Summary Statistics
(CARDIoGRAM, GLGC, MEGASTROKE, MVP, BBJ, FinnGen, Genes & Health)

Multi-Ancestry

East Asian
 PRS_{CAD}

European
 PRS_{CAD}

South Asian
 PRS_{CAD}



Multi-Traits

GPSmult

Integrate genetic risk
from multi-ethnic and comorbidity GWASs

➤ **CardiogramPlusC4D**

>260k CAD cases

>1170k control

➤ **CAD comorbidities and risk factors**

> 1.3M individuals

Polygenic risk model for CAD -- GPS 2.0



UK Biobank 300 K



GPS Low Risk

GPS High Risk



6%

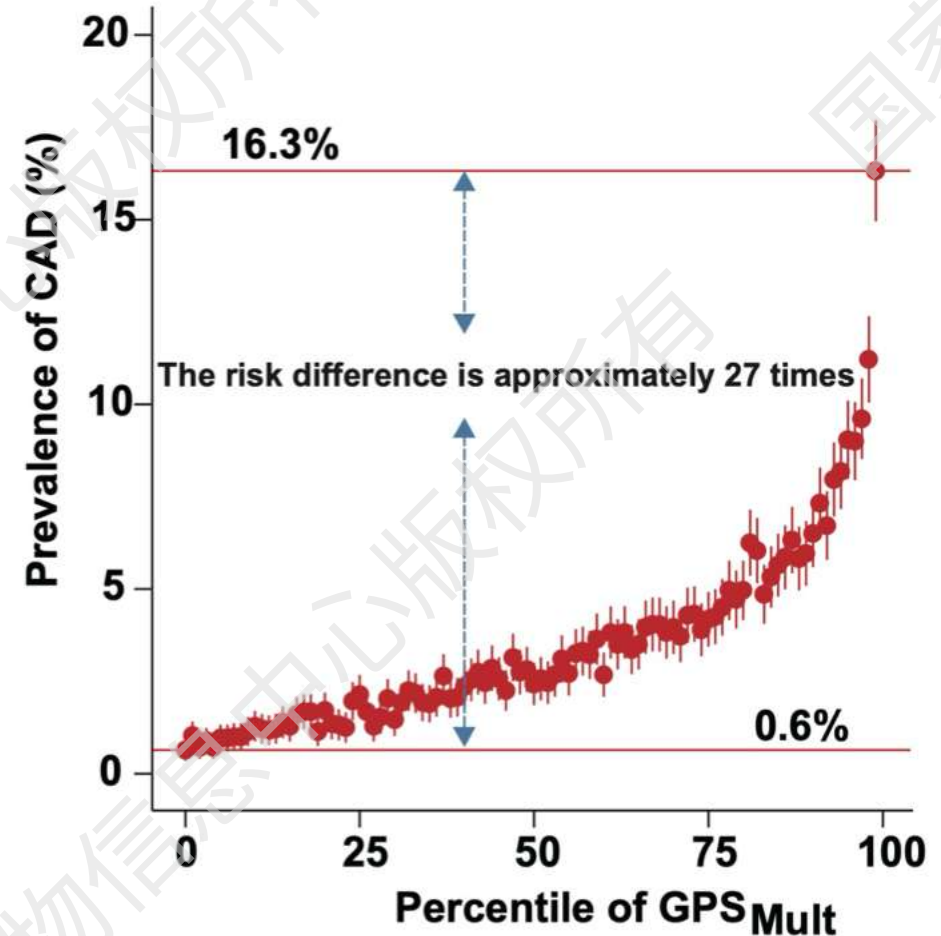
27x

163%



GPSmult 100 bins

GPSmult (CAD-GPS2.0)



Polygenic Risk Scores and Disease Risk Stratification

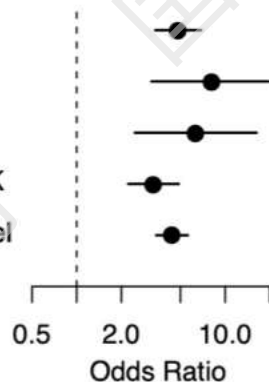


LDLR | 41k CAD cases and 220k controls

- Odds ratio 4.39
- $P = 1.74 \times 10^{-32}$
- 0.91% cases vs 0.34% controls

DATA

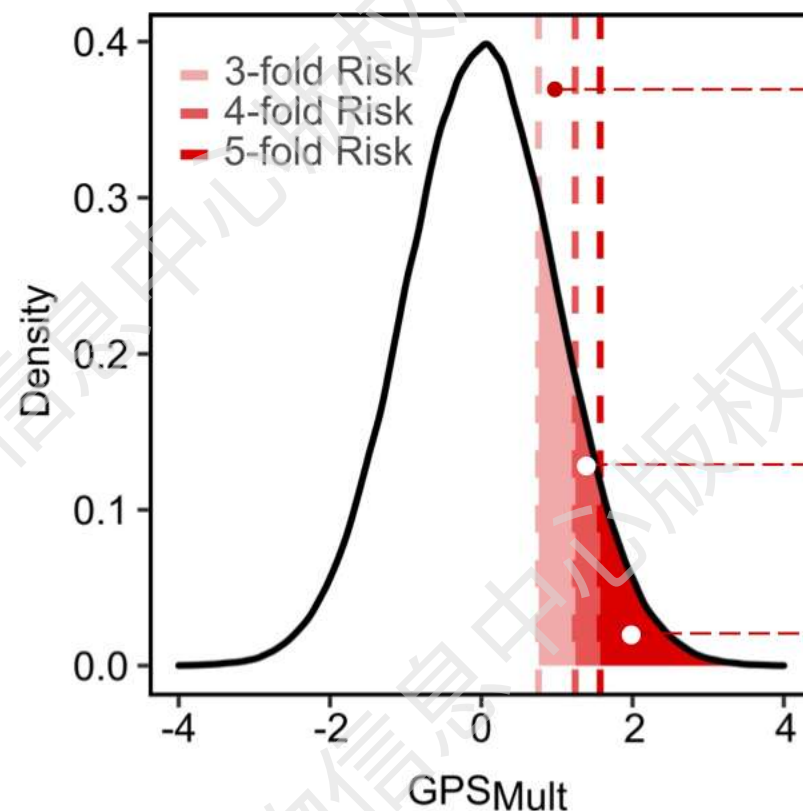
MIGen ExSeq
MIGen WGSeq
UK Biobank 13K
UK Biobank 200K
Fixed effect model



Odds ratio [95CI] P Value

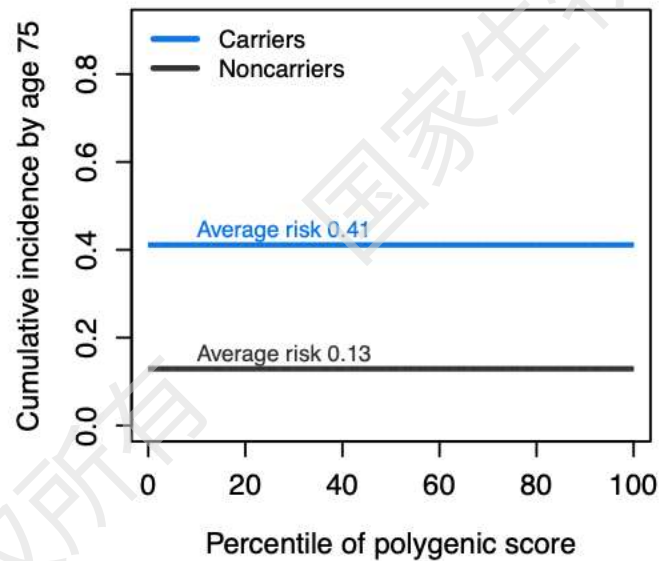
4.83 [3.39–6.88] 3.12^{-18}
8.12 [3.20–20.65] 1.08^{-05}
6.33 [2.46–16.29] 1.28^{-04}
3.29 [2.23–4.87] 2.43^{-09}
4.39 [3.44–5.60] 1.74^{-32}

汪敏先*等,
Circ Genomic Precis Med. 2022



汪敏先*等,
Nature Medicine. 2023

Personalized Genetic Risk Assessment V2.0



?
**Monogenic
mutation**

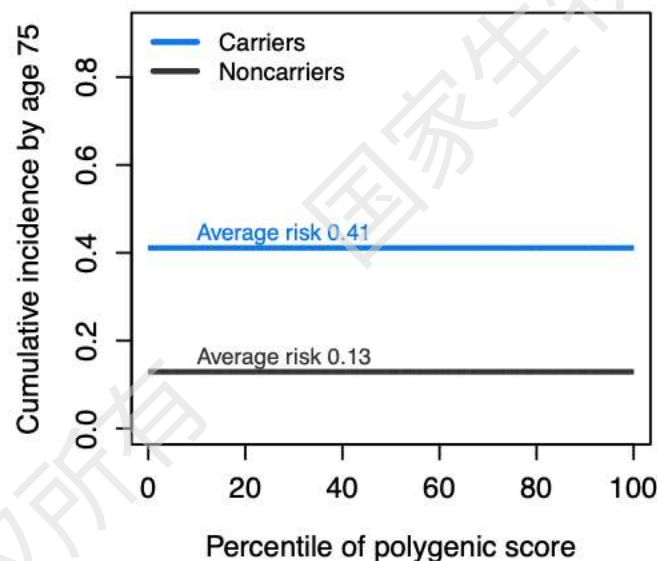


**High risk
carriers**

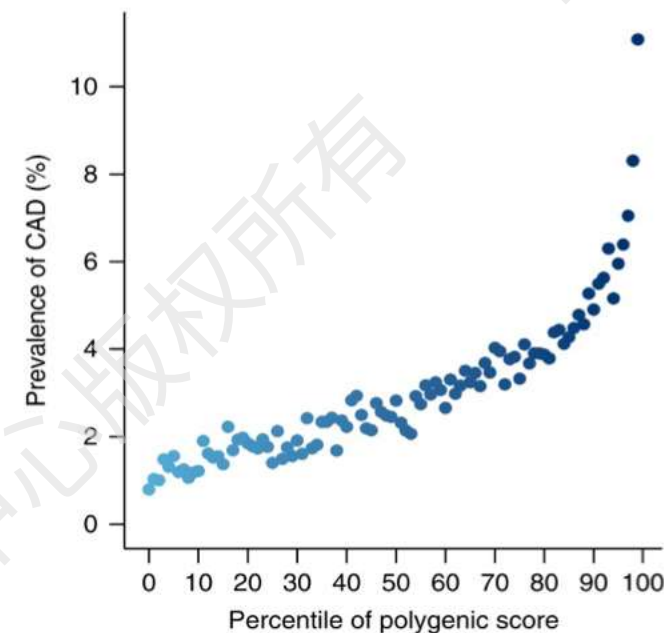
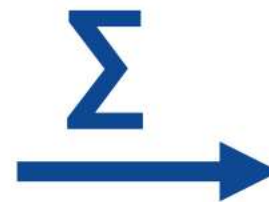


**Average risk
non-carriers**

Personalized Genetic Risk Assessment V2.0



?
**Monogenic
mutation**



**High risk
carrier**



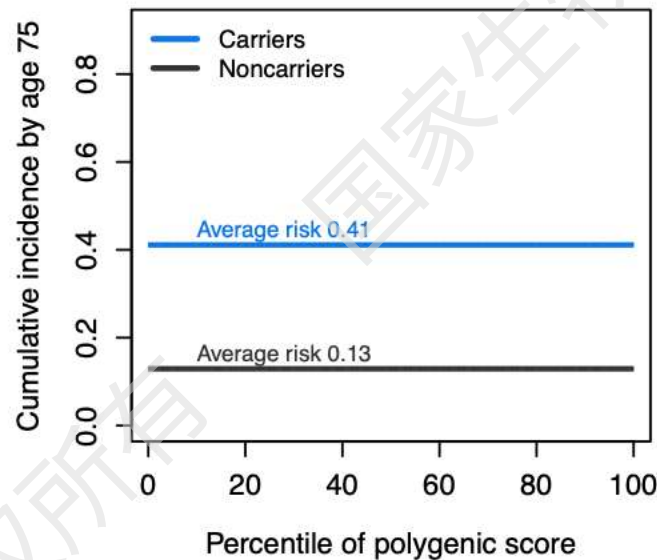
**Average risk
non-carriers**

**Polygenic risk
summarize small effects**

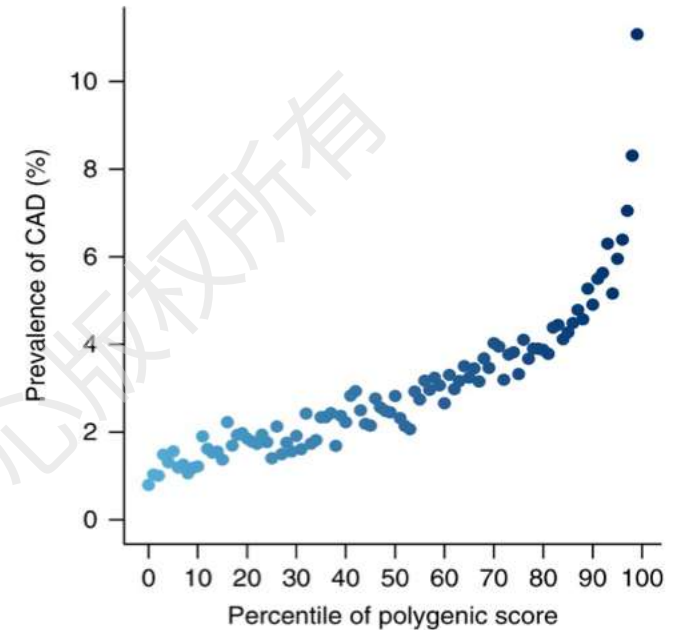
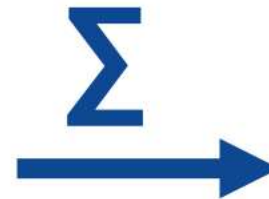
Personalized Genetic Risk Assessment V2.0



? Interplay of multiple effects?



**?
Monogenic
mutation**



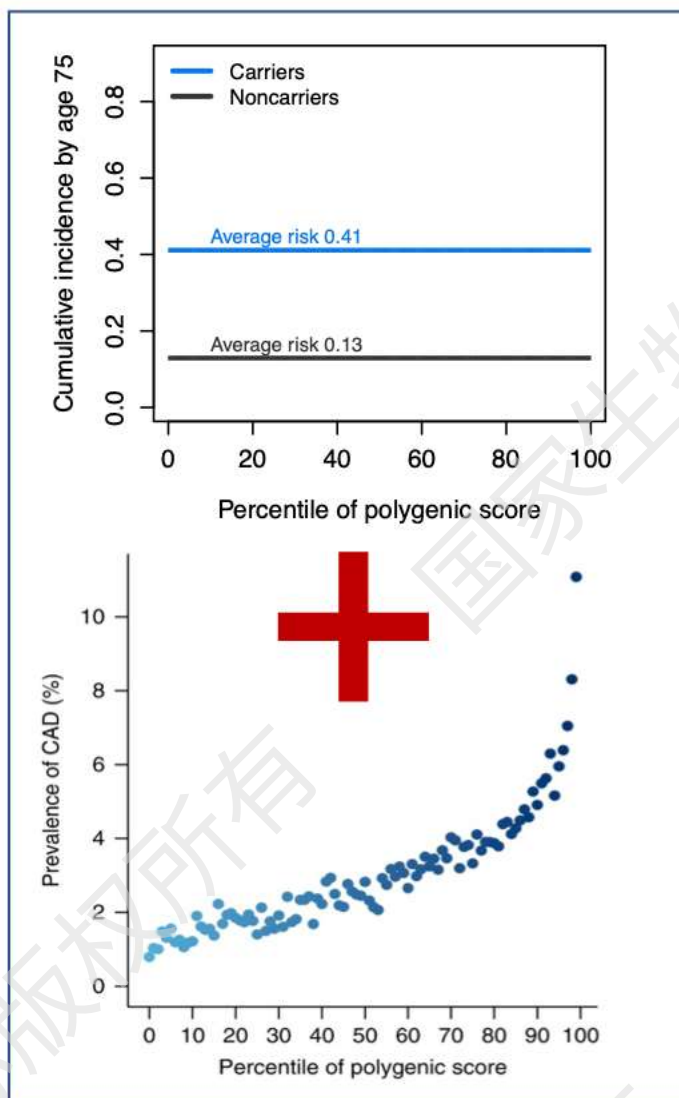
**High risk
carrier**



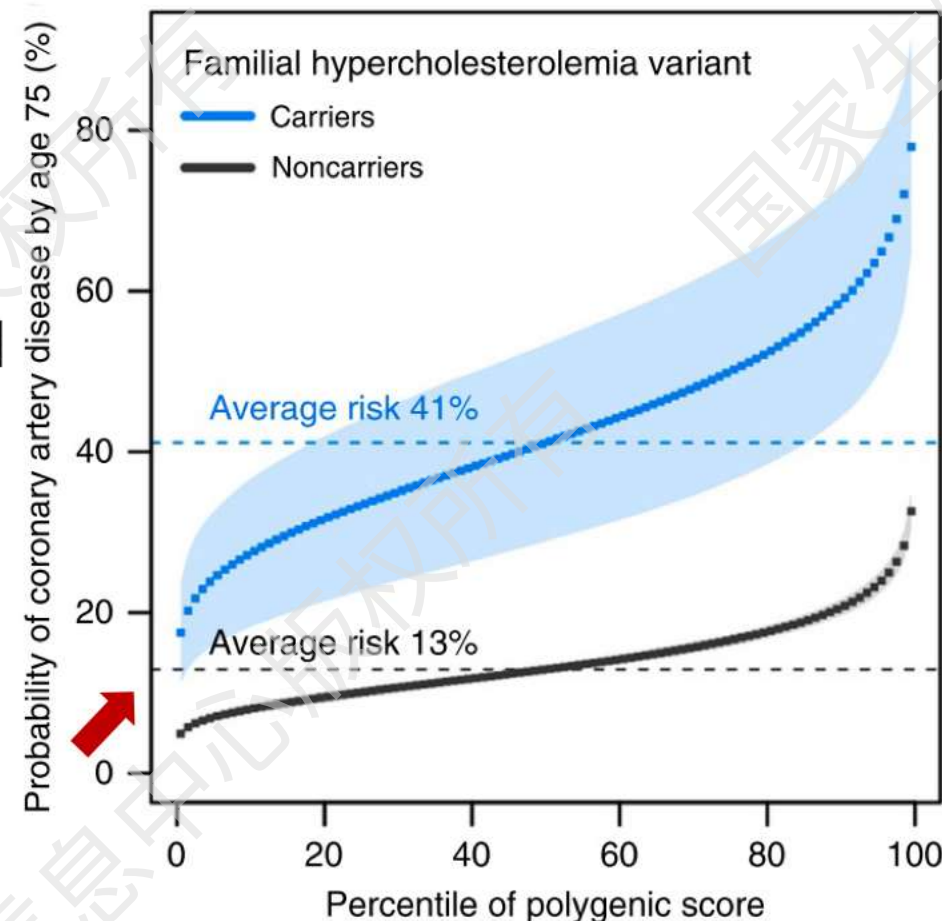
**Average risk
non-carriers**

**Polygenic risk
small effects size accumulation**

Personalized Genetic Risk Assessment For CAD



**Joint modeling of
monogenic variants and
polygenic risk**



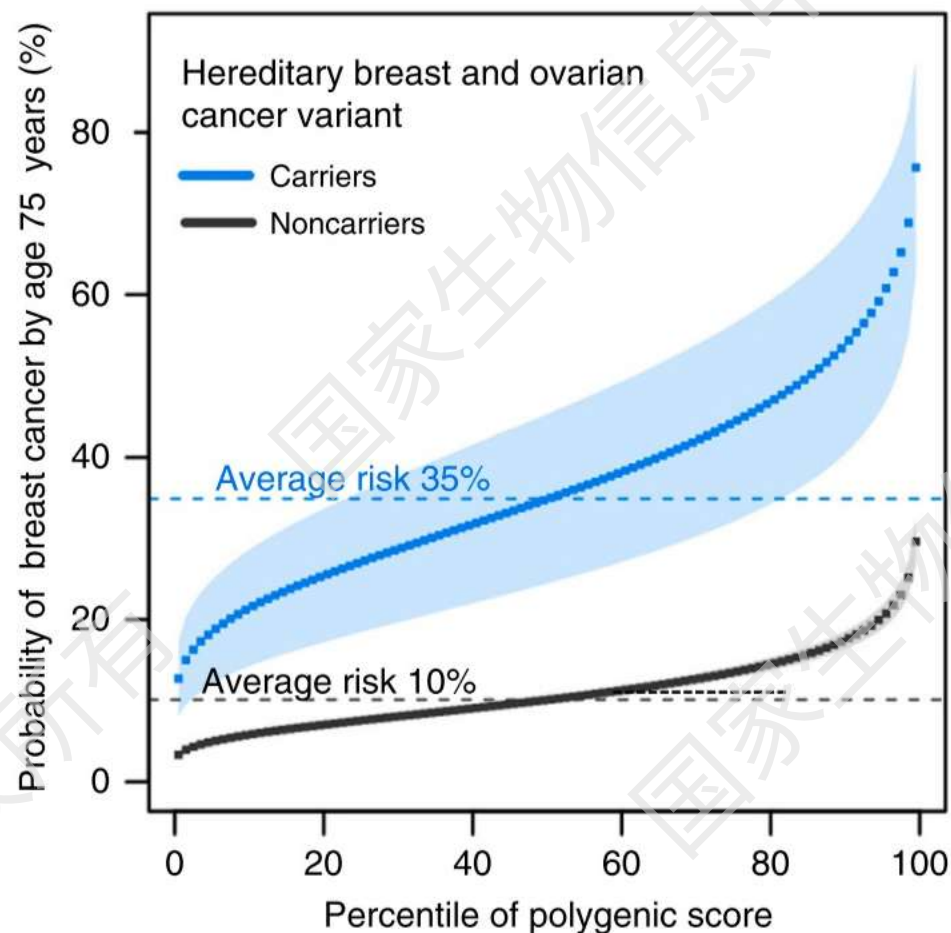
ACMG Variation	Avg	Low	High
+	41%	17%	78%
-	13%	4.9%	33%

Nature Communications. 2020

V2.0: Monogenic + Genome-Wide Background Susceptibility

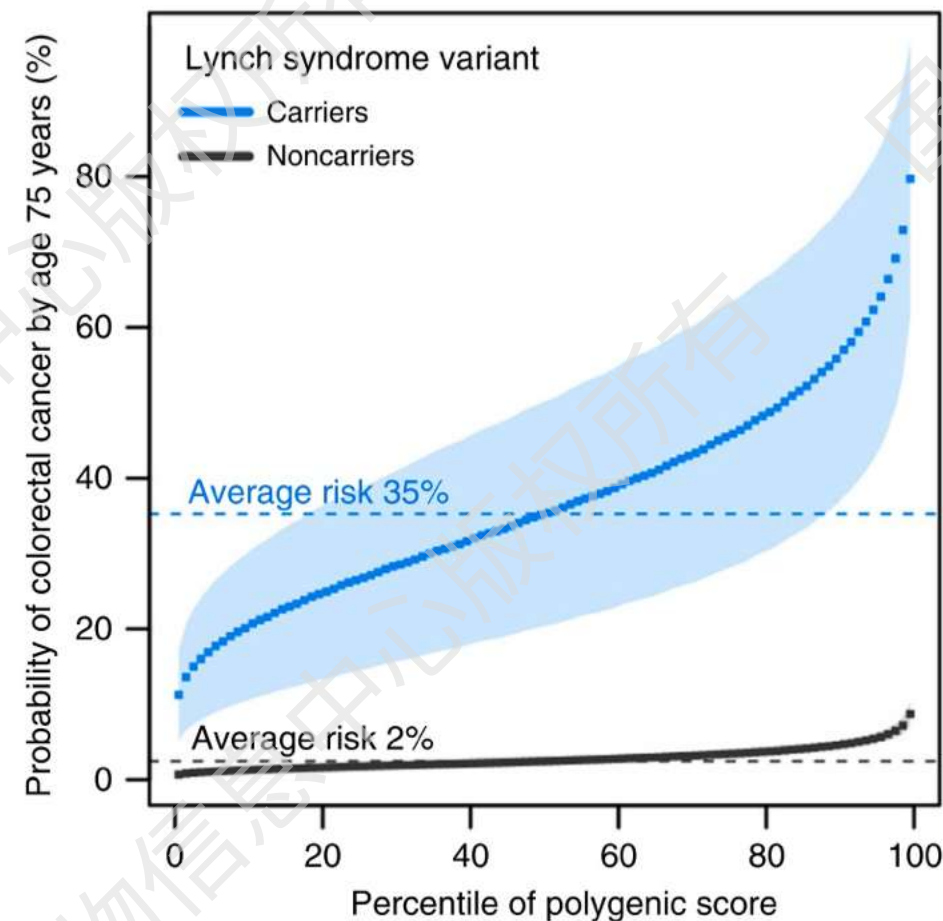


Breast cancer absolute disease risk



564 Cases + 25,260 Controls, 8 Years follow up.

Colorectal cancer



361 Cases + 48,812 Controls, 8 Years follow up.

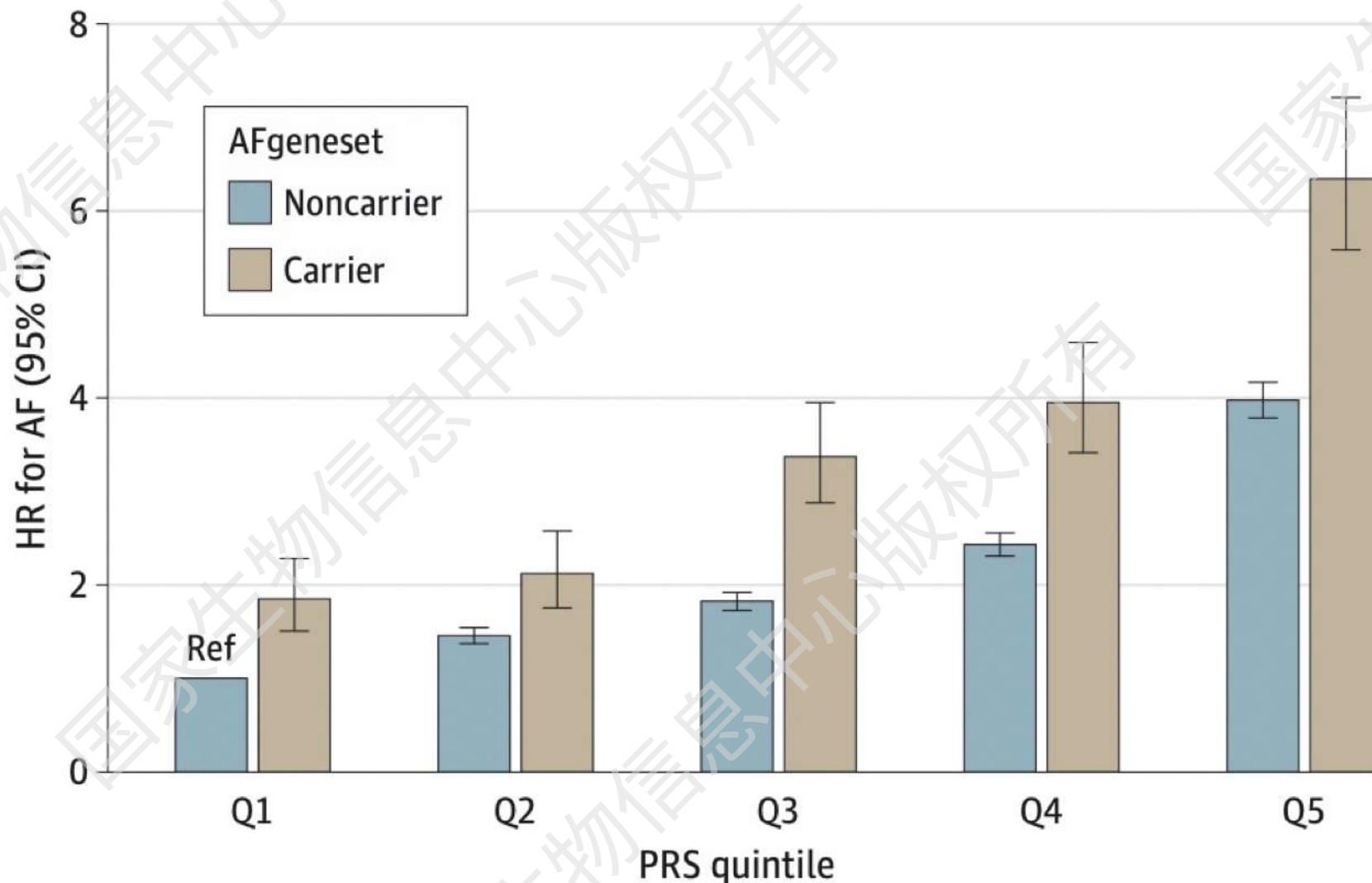
Monogenic + Genome-Wide Background Susceptibility (AF)



biobank^{uk}

AF = 30 797

N = 41 6085



➤ Highlighted by Science

➤ Top 10 clinical genetics advances of 2020 (U.S. NHGRI Advisory Council)

YEAR IN REVIEW

Genomic Medicine Year in Review: 2020

Teri A. Manolio,^{1,*} Carol J. Bult,² Rex L. Chisholm,³ Patricia A. Deverka,⁴ Geoffrey S. Ginsburg,⁵ Madison Goldrich,¹ Gail P. Jarvik,⁶ George A. Mensah,⁷ Mary V. Relling,⁸ Dan M. Roden,⁹ Robb Rowley,¹ Cecelia Tamburro,¹ Marc S. Williams,¹⁰ and Eric D. Green¹

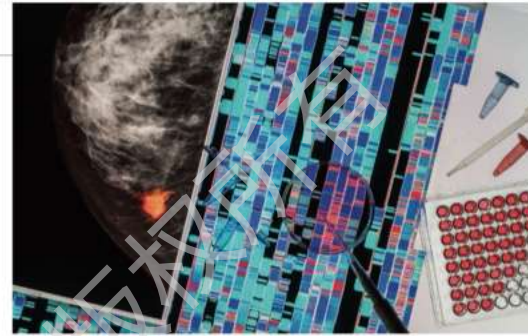
Introduction

In December 2019, *The American Journal of Human Genetics* introduced a new annual feature¹ designed to chronicle **ten key advances in applying genomic information to clinical care in the previous twelve months** of published literature. The Genomic Medicine Working Group of the **National Advisory Council for Human Genome Research of the National Human Genome Research Institute (NHGRI)**

and understandable trend toward fewer citations of those published later in the year.

This year the Working Group has again selected its ten most significant advances among the 45 recognized accomplishments and resources published during the 12 months ending August 31, 2020. These papers clustered into 13 broad categories, two of which—"Genomic Medicine Resource" and "Pilot Implementation"—were added for the 2020 review

to genotype-guided treatment has not been studied. Two randomized trials compared outcomes of rapid genotype-guided versus standard antiplatelet therapy in patients undergoing percutaneous coronary intervention (PCI). Both trended in favor of genotyping but came to different stated conclusions based on prespecified statistical analysis plans. Neither trial was powered to study outcomes in carriers of two LOF alleles, the group unable to bioactivate clopidogrel at all.



BIOMEDICINE

'Polygenic' analyses may sharpen disease risk predictions

Genetic background modifies impact of single disease genes

By Jocelyn Kaiser

Women who learn that they have a mutation in the breast cancer gene *BRCA1* face a wrenching decision. Their doctor or genetic counselor will likely tell them that women with such mutations have, on average, a 72% lifetime risk of breast cancer and a 44% risk of ovarian cancer. Given that, up to half decide to have prophylactic mastectomies, and many have ovaries removed, too.

But recent studies show a woman could receive a more individualized, accurate cancer risk estimate by factoring in other gene variants. A preprint posted last month finds that a person's "polygenic" background influences not only the disease risk conferred by a *BRCA1* defect, but also risks from single gene mutations linked to colorectal cancer and heart disease. Some individuals were very likely to develop cancer or heart disease by age 75, the analysis showed, whereas in others the risk was not much greater than in a person without the high-risk mutation.

"It's pretty striking," says cardiologist and geneticist Amit Khara of Massachusetts General Hospital (MGH) in Boston, leader of the study, which is on the medRxiv preprint server. "It's become clear that there are both monogenic and polygenic [disease] drivers. The future is to assess both."

"The message is a very important one for patients and clinicians," says Teri Manolio of the National Human Genome Research Institute in Bethesda, Maryland. "Carriers of *BRCA1* mutations or other pathogenic vari-

ants don't invariably develop disease, and genomics can be used to help parse carriers who are at lower risk." Others caution, however, that risk scores summing how dozens to thousands of other genetic variants interact with a single major disease gene aren't yet accurate enough to use in the clinic. The new paper "is teasing at the possibility, but there's a lot of work to be done," says Harvard University epidemiologist Peter Kraft.

MGH cardiologist fellow Aid Fayed and others in Khara's group explored polygenic influences on the three important single-gene disorders in the United States: familial hypercholesterolemia, which leads to sky-high cholesterol levels and dramatically elevates risk of heart disease; Lynch syndrome, a flaw in DNA repair that brings a lifetime risk of colorectal cancer of about 60%; and inherited breast cancer, caused by variants in *BRCA1* or *BRCA2*. They took advantage of databases that combine medical and genomic information from thousands of people, enabling researchers to tally how the many genetic variants with subtle effects modify disease risks and complex traits such as height.

Drawing on some 50,000 participants in the UK Biobank and 19,000 women tested for *BRCA1* genes by the company Color Genomics, the team found that polygenic background strongly modified the risk of carrying a mutation in the key genes for the three disorders. For a small proportion of major disease gene carriers, other genetic variants boosted their overall risk of cancer or heart disease to about 80%, well above the average of 30% to 40% that Khara's group estimated for its

A woman's genetic background can powerfully modify her cancer risk from a *BRCA1* mutation.

study populations based on just the single disease gene mutations. (The team's monogenic disease risk predictions are lower than many other estimates for several possible reasons, Khara notes, including that the UK Biobank participants are healthier than the general population.) At the other extreme, the polygenic analysis suggested that a few people with those mutations have much lower risks than predicted by their single mutation alone, as low as 11% for colon cancer, 13% for breast cancer, and 17% for heart disease—not much higher than other people in general.

Khara's group says adding polygenic data to single-gene tests could help people decide whether to take aggressive steps to head off disease—mastectomy or removal of the ovaries for women carrying *BRCA1* mutations or frequent colonoscopies for people with Lynch syndrome. But the new study does not include enough data for clinical decisions, says genetic epidemiologist Antonia Antoniou of the University of Cambridge in the United Kingdom. Only 116 women in the UK Biobank sample had *BRCA1* mutations, which he notes "is an extremely small number to make inferences about risks."

Two years ago, Antoniou led a study that reported on how polygenic scores influence risks in 25,000 carriers of *BRCA1* mutations and found nearly as wide a range of overall cancer risks. His team has incorporated those data into a breast cancer risk estimator along with factors such as family history.

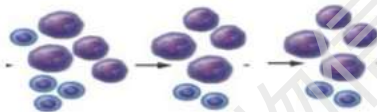
The MGH study is "an important and exciting paper" that complements other work, says David Ledbetter, chief scientific officer for the Geisinger Health System in Danville, Pennsylvania. His team recently looked at 92,000 participants in an ongoing genomic medicine study called MyCode, focusing on those who carried mutations predisposing them to 11 rare disorders that affect traits such as height, weight, and cholesterol levels. Incorporating polygenic scores helped predict those traits, the group reported on 25 October in *Nature Communications*.

It may be a while before physicians are comfortable telling patients how genetic backgrounds modify the risk posed by a major disease gene mutation. But some companies already offer polygenic scores for cancer and other diseases, and tests that combine both kinds of information are imminent. Before insurance companies agree to pay for such tests, Ledbetter cautions, "They're going to want to see much more clinical validation"—including for minorities, because current polygenic analyses draw on data primarily from people of European ancestry. ■

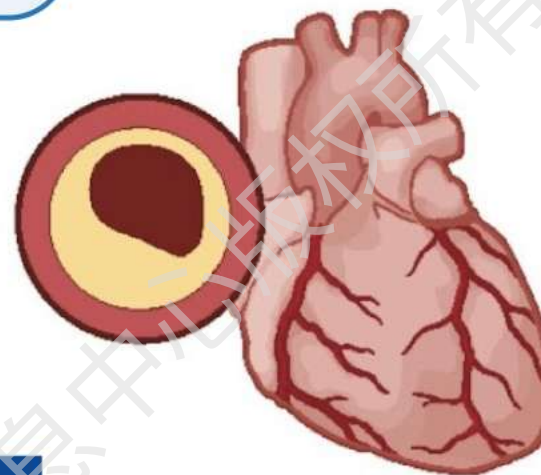
Genetic Risk Factors for Cardiovascular Disease



Somatic Mutations
CHIP



Telomere Length
LTL



Genetic Factors

Clonal Hematopoietic Somatic Mutations and CAD risk

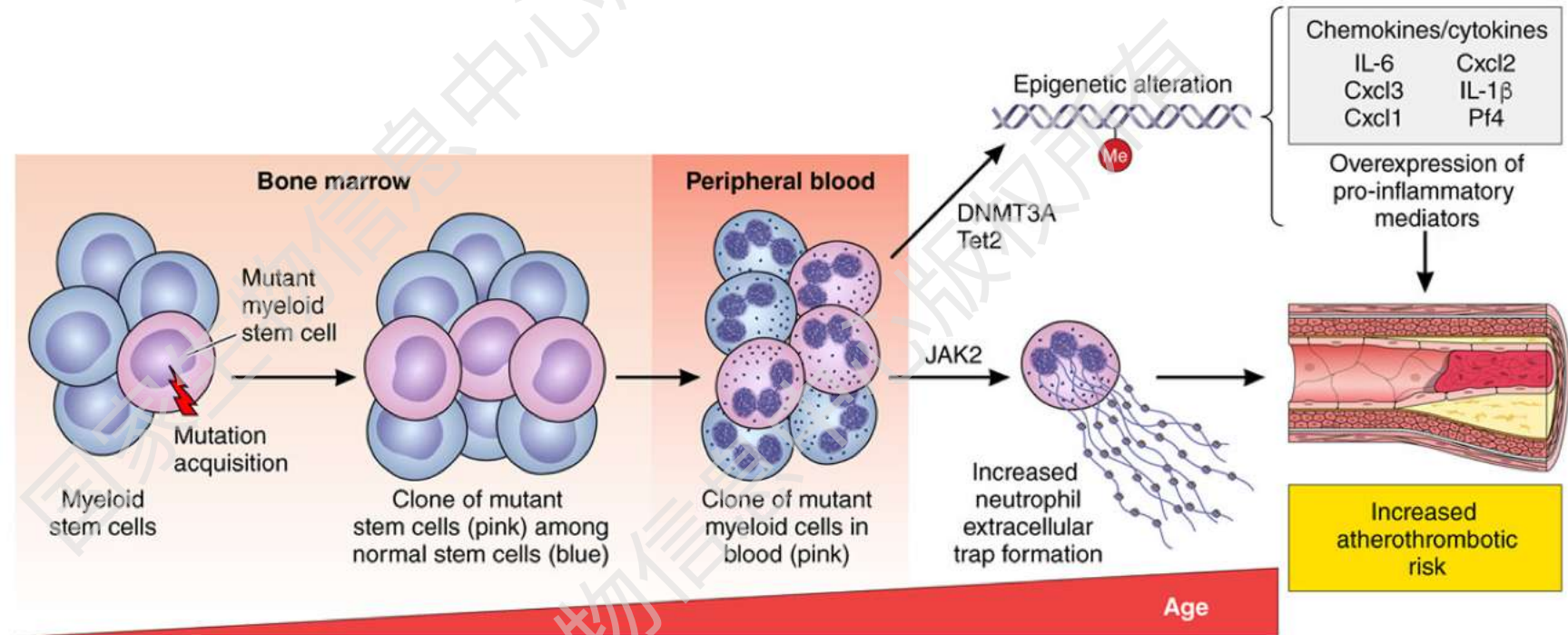
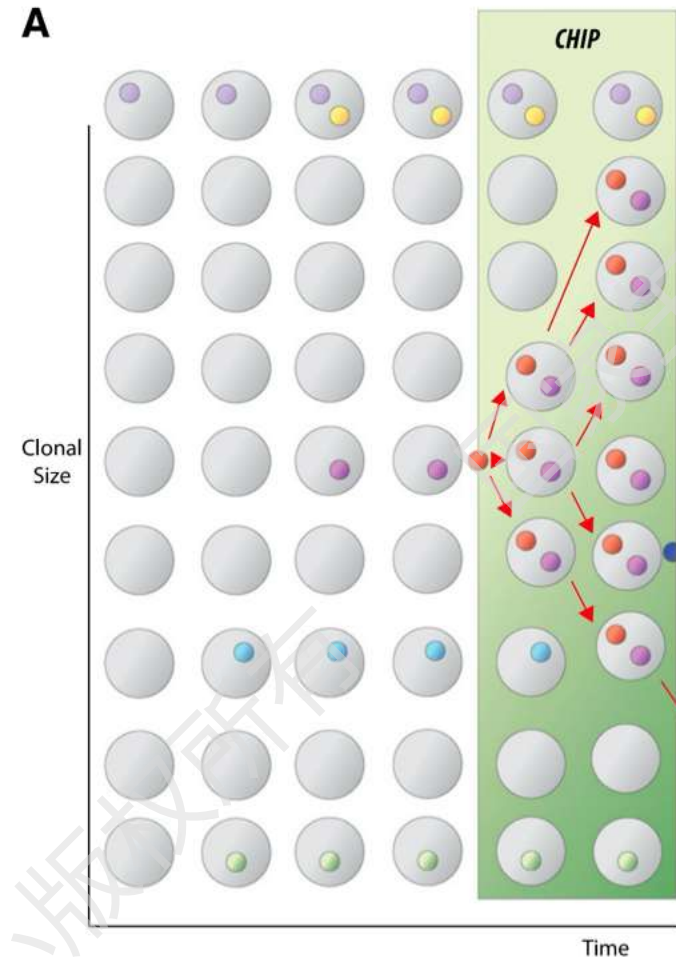


➤ Clonal hematopoiesis of indeterminate potential (CHIP)

> Age > 65 ~ 10%-15%

> Age > 80 ~ 30%

Clonal hematopoiesis of indeterminate potential (CHIP), also known as age-related clonal hematopoiesis (ARCH)



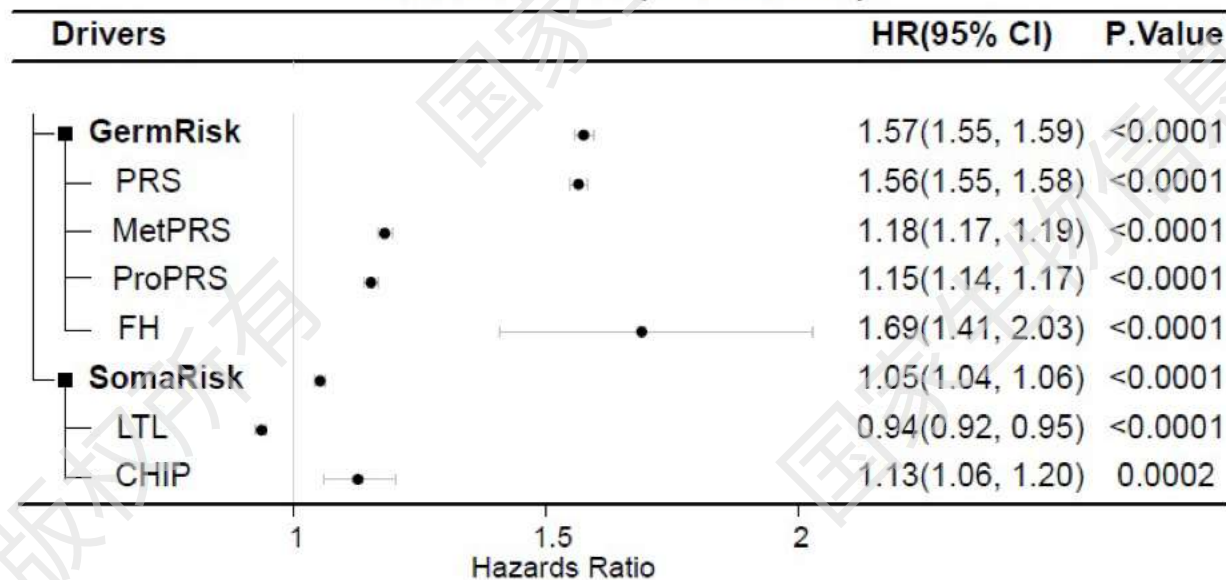
Contribution of Somatic Mutations and Telomere Length to Cardiovascular Disease Risk



biobank^{uk}

CAD = 42 602
N = 391,536

Hazards Ratio (UK Biobank)

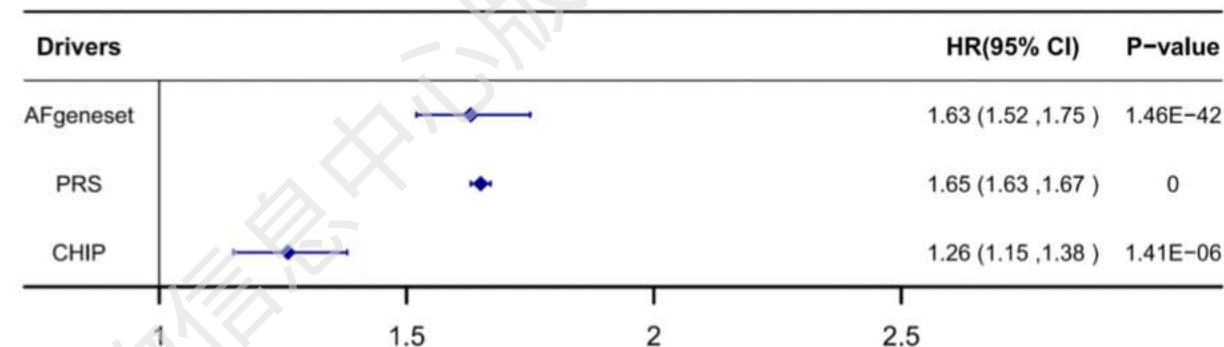


Xiong Yang ... Minxian Wang#, Akl Fahed#
Nat. commun. (in press)

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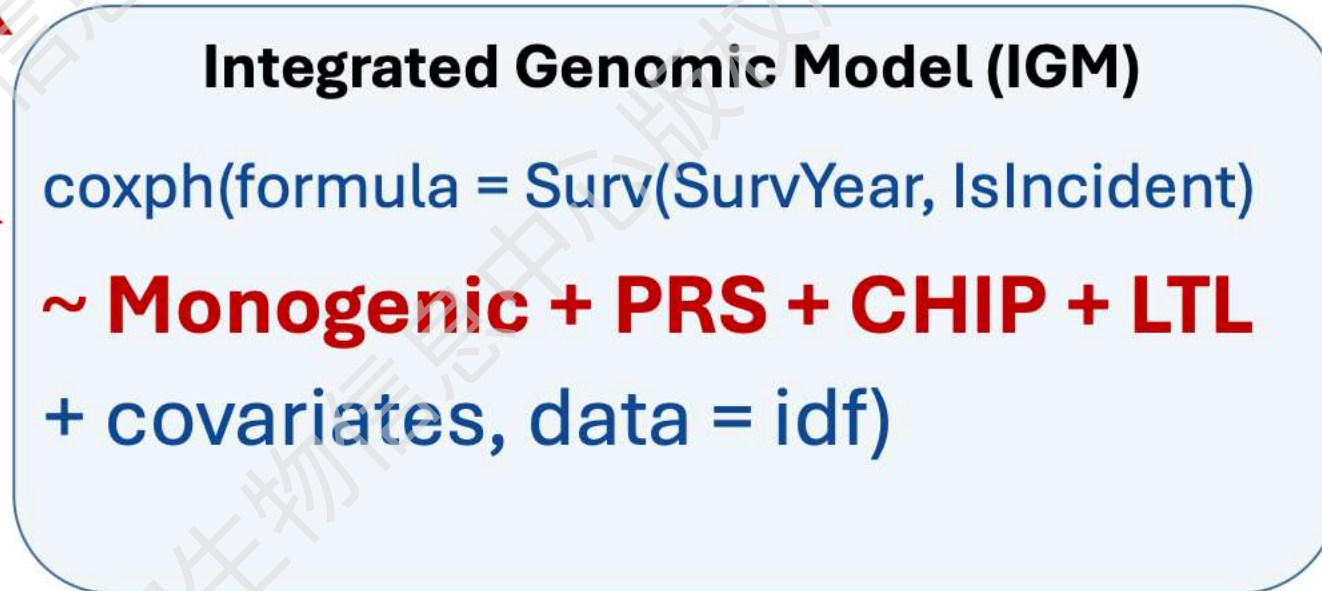
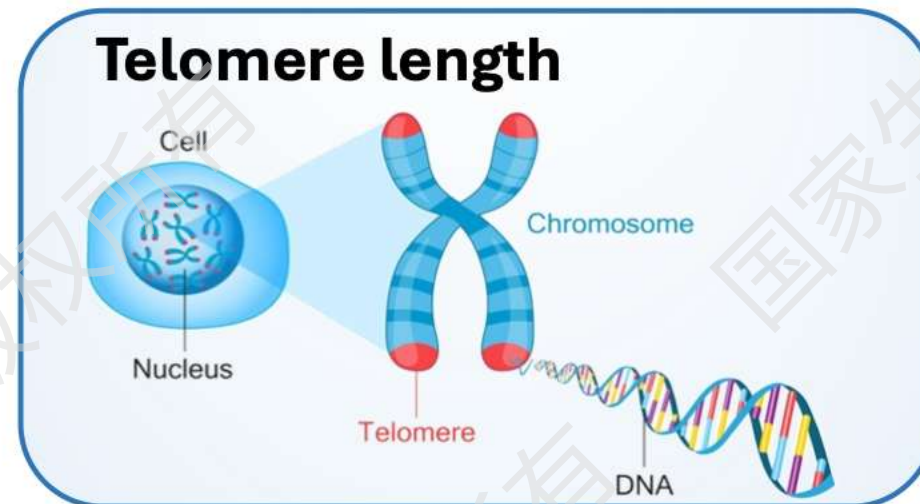
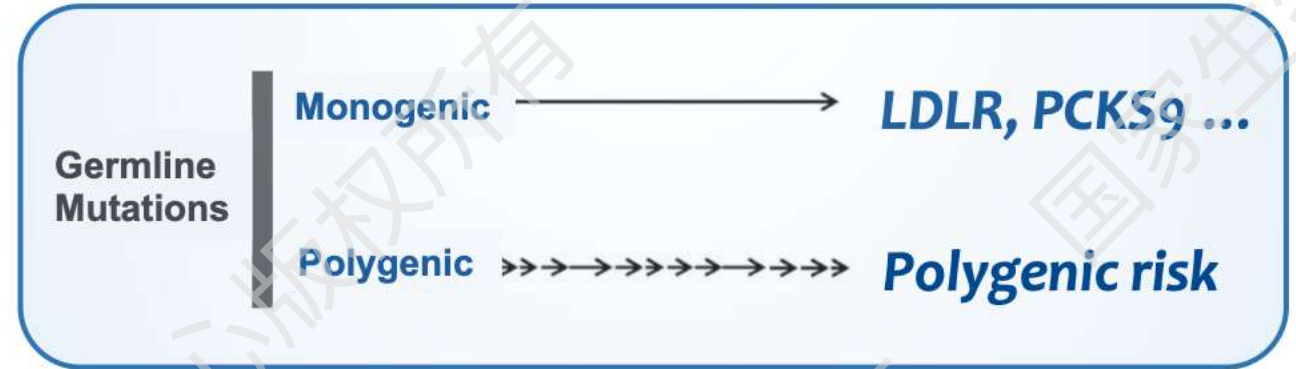
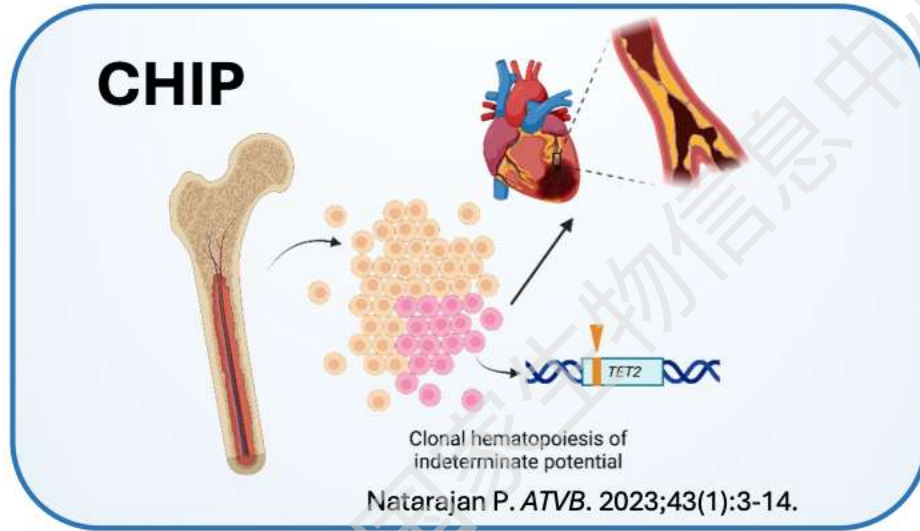
AF = 30 797
N = 41 6085

Hazard Ratio

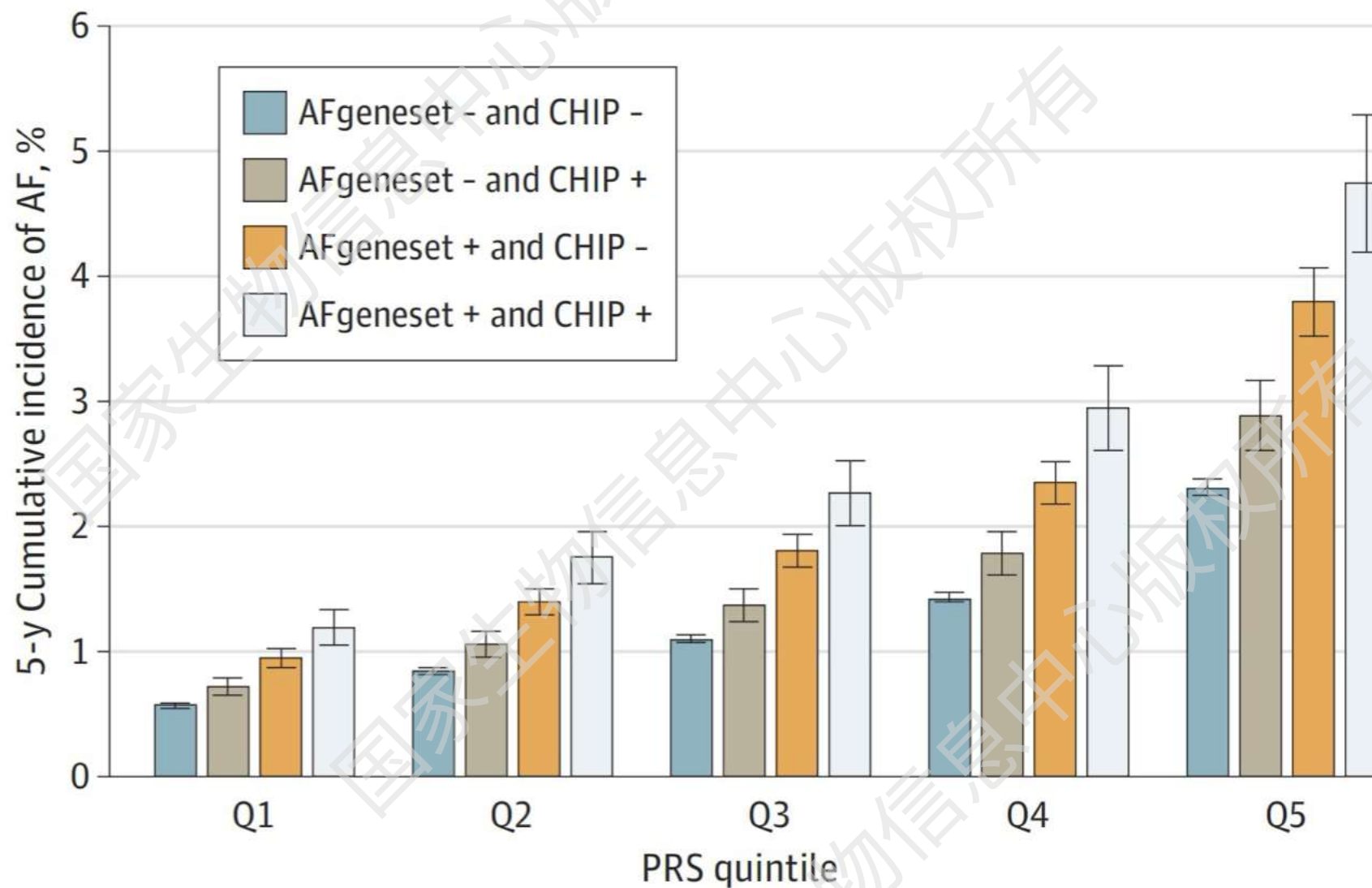


Rufan Zhang ... Minxian Wang#, Akl Fahed#
JAMA cardiology. 2025

V3.0: Monogenic + PRS + Somatic Mutations



V3.0: Monogenic + PRS + Somatic Mutations (AF)

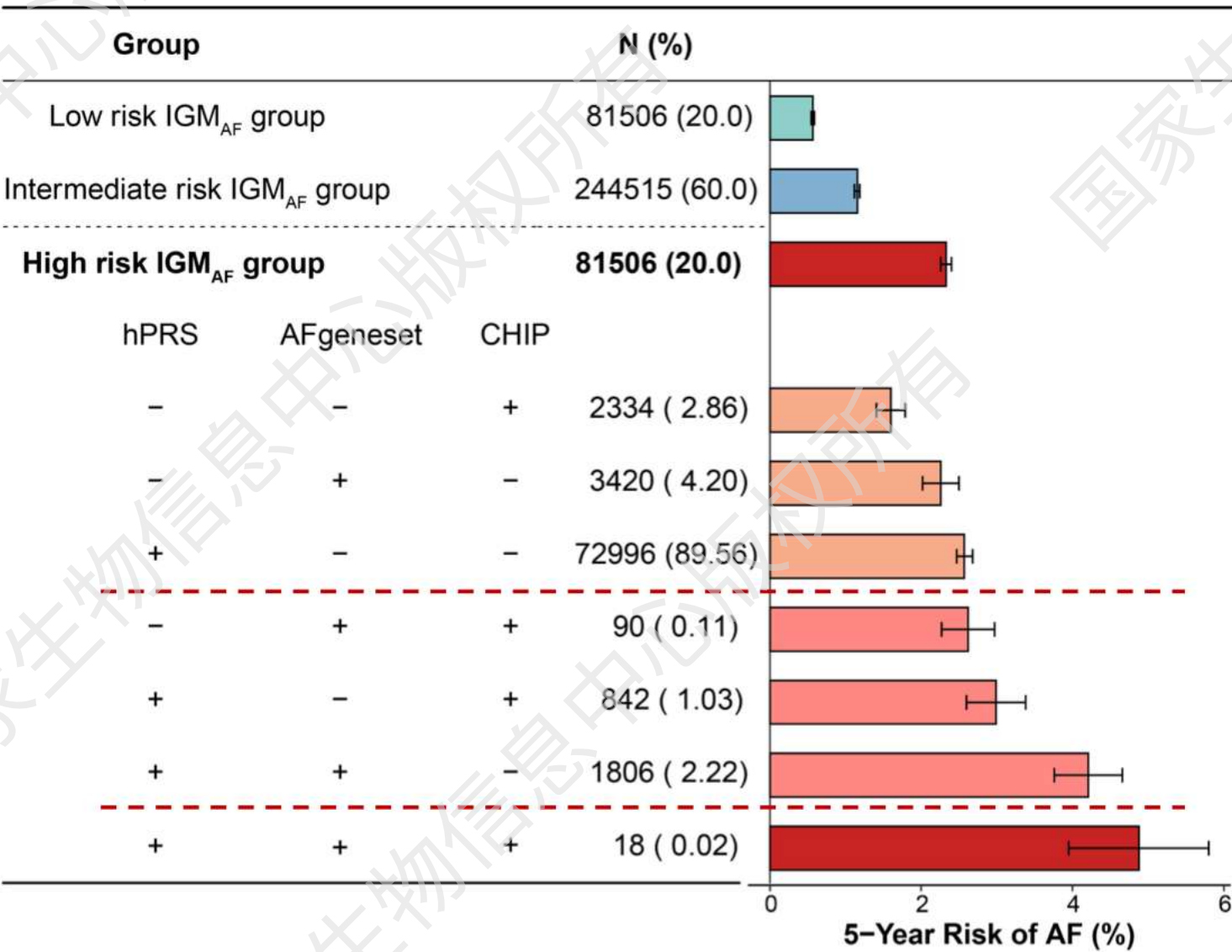


V3.0: Monogenic + PRS + Somatic Mutations (AF)



AF = 30 797
N = 41 6085

Rufan Zhang ... Minxian Wang#, Akl Fahed#
JAMA cardiology. 2025 (in press, IF = 15)



V3.0: Monogenic + PRS + Somatic Mutations (CAD)

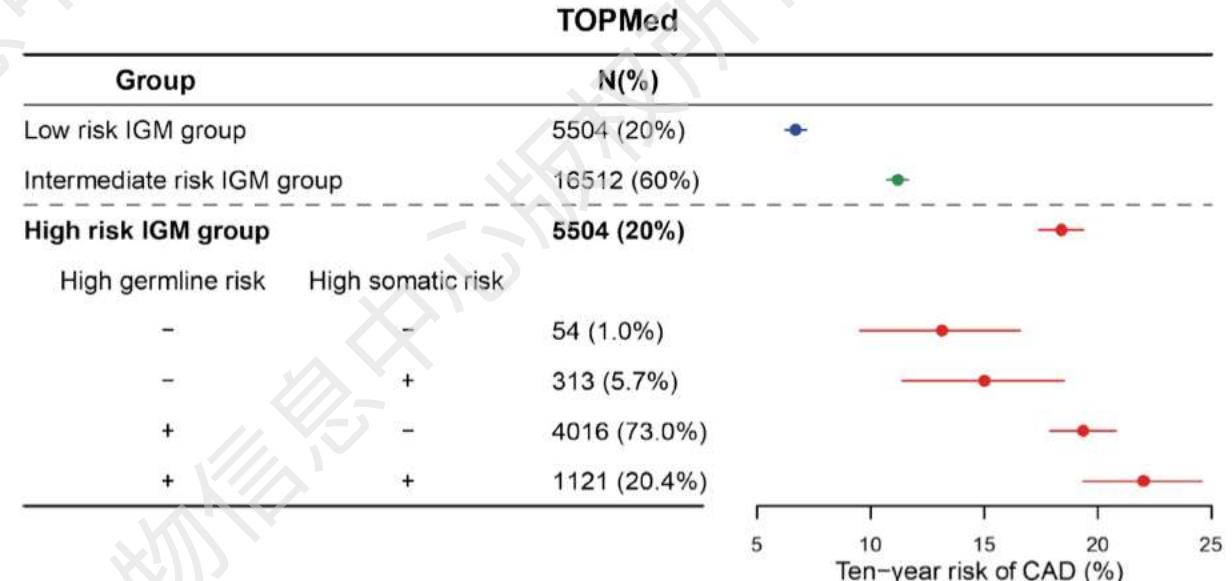
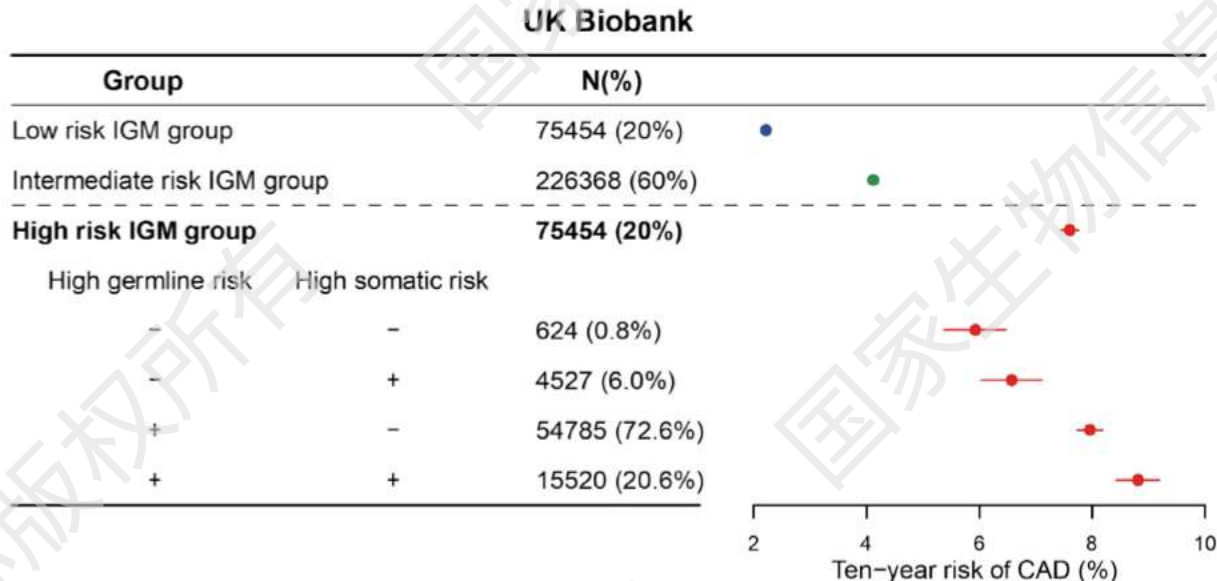


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CAD = 42 602
N = 391,536



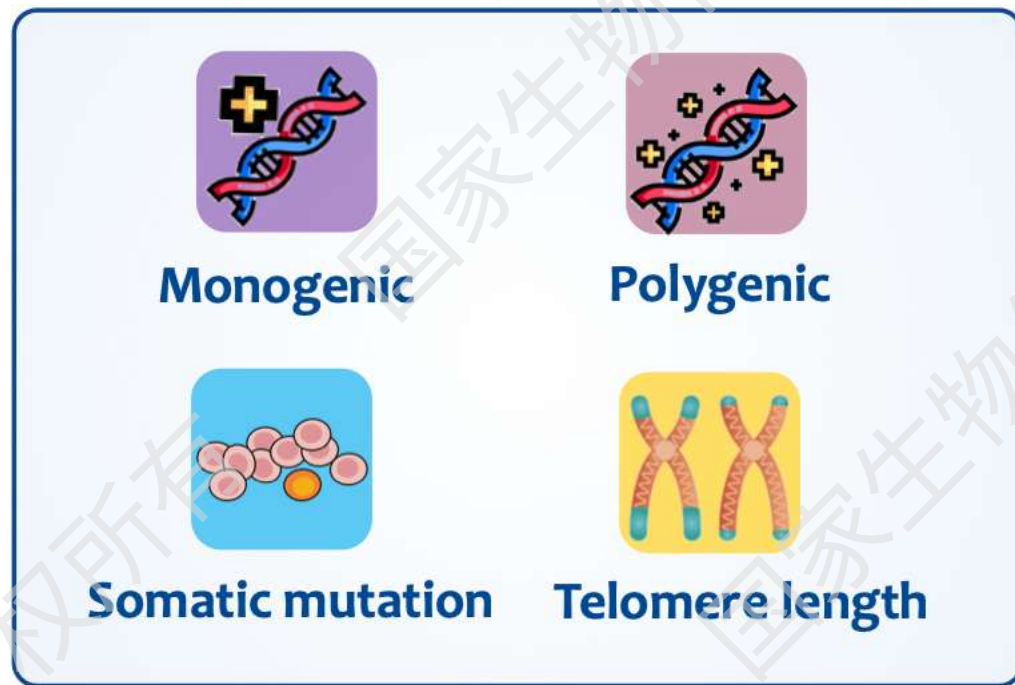
CAD = 4 326
N = 34 177



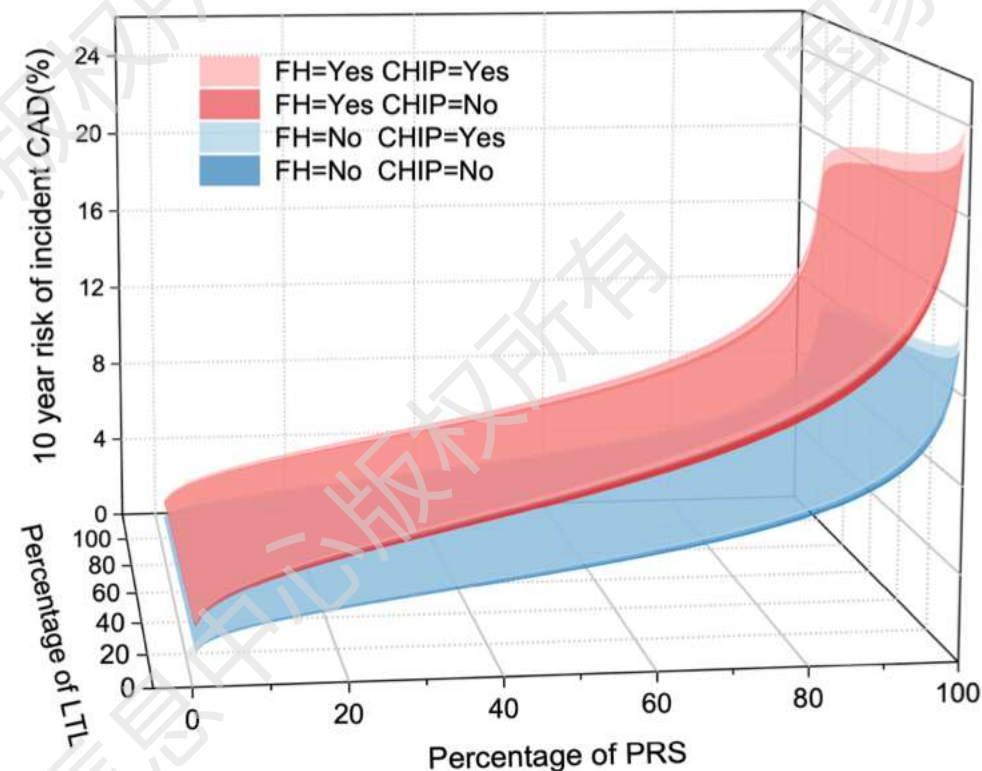
Personalized risk prediction V3.0

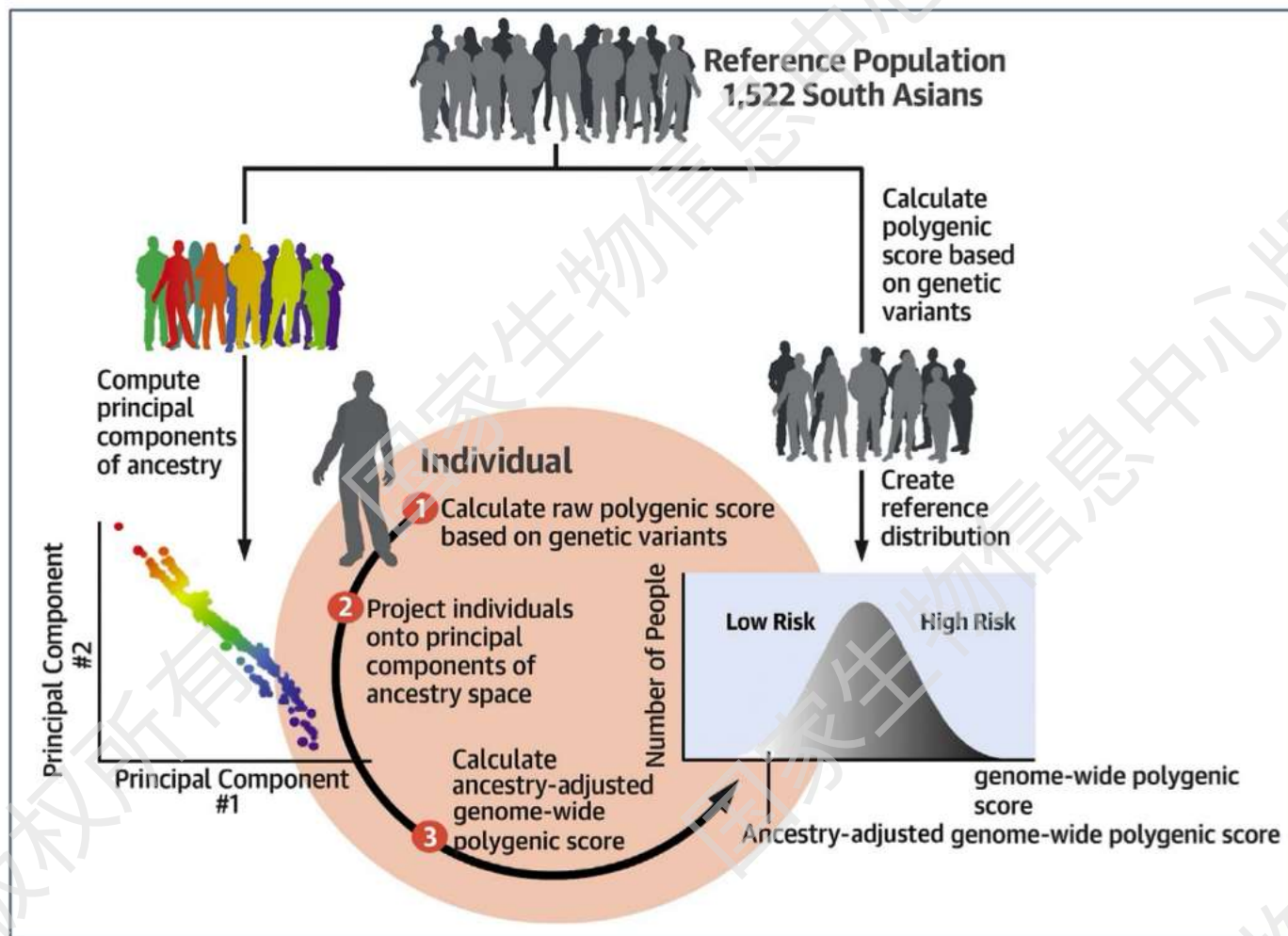


**multiple genetic risk
components joint modeling**

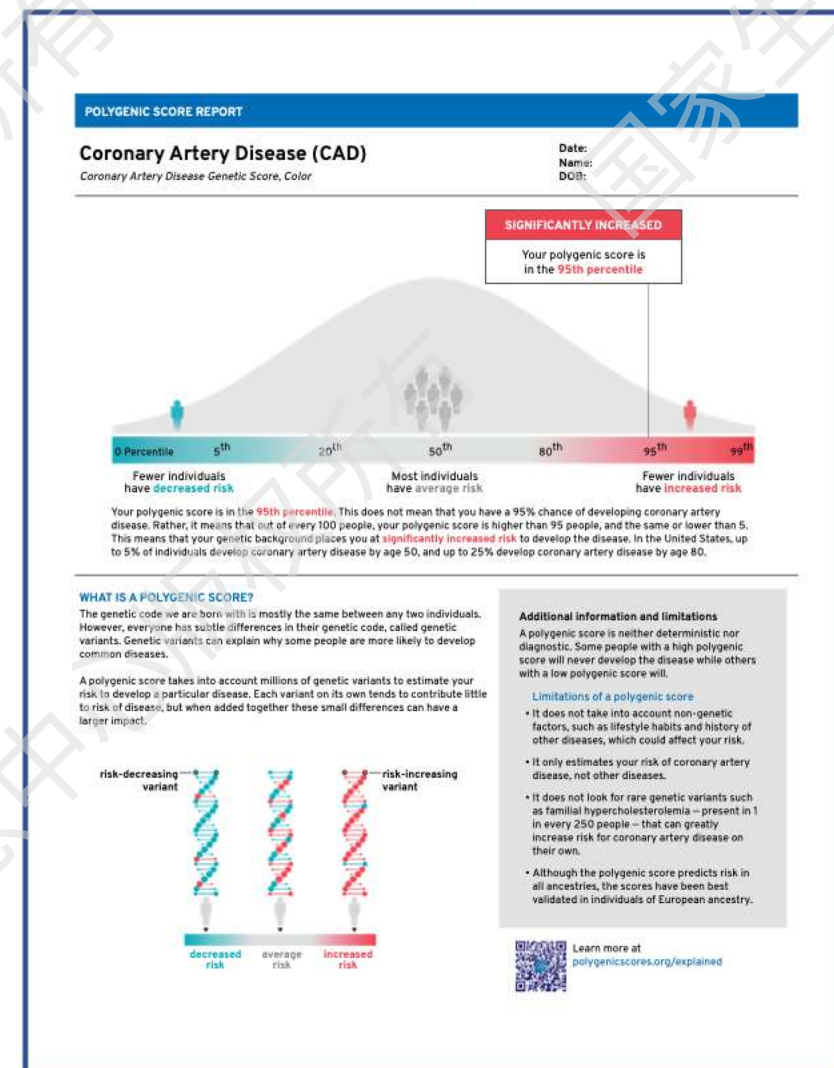


UK Biobank 400K individuals

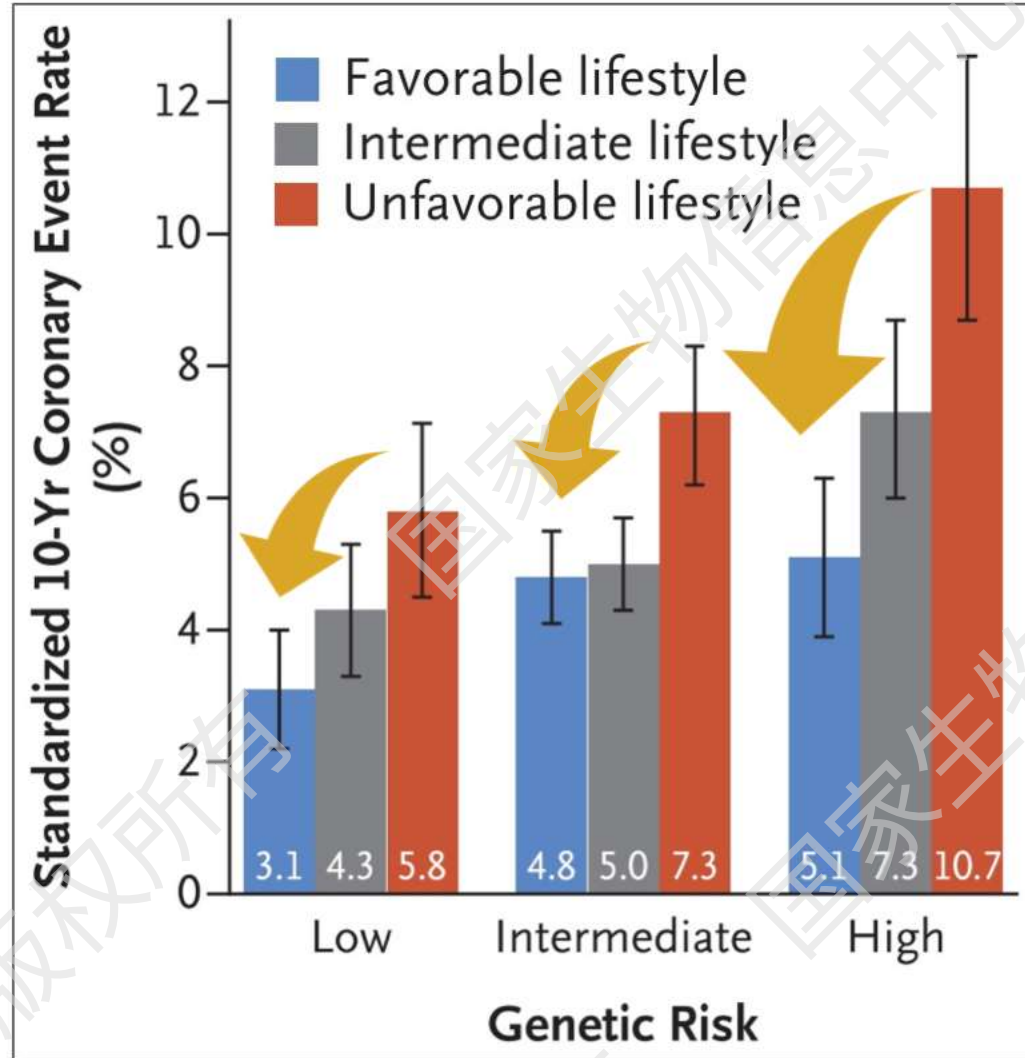




Risk Report



DNA is not destiny !

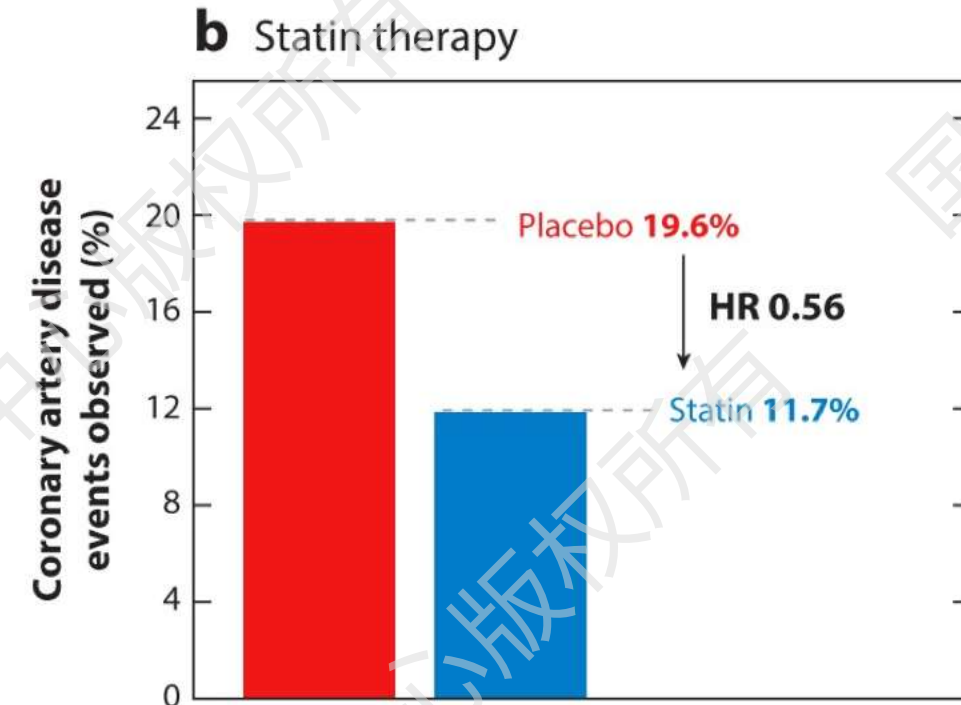
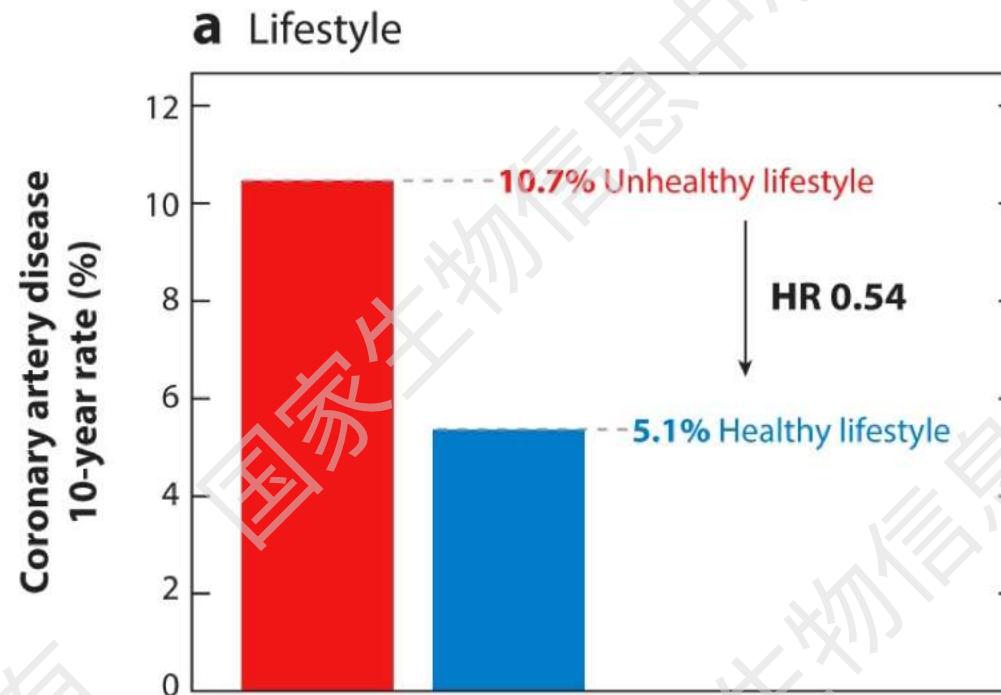


Khera, A. V. et al. *N. Engl. J. Med.* 2016.



- No smoking, no drinking
- Good exercise
- Healthy diet
- Keep a low BMI

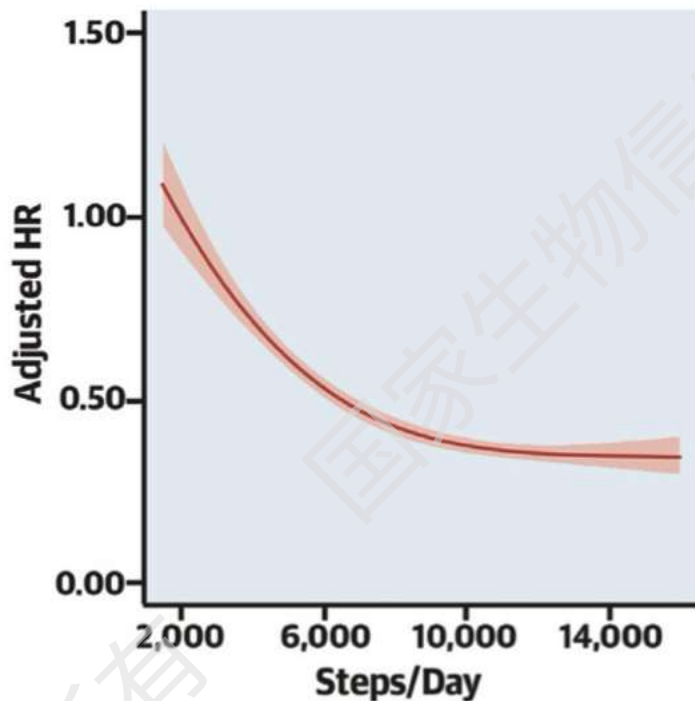
DNA is not destiny !



DNA is not destiny -- Offset disease risk is easy!!!

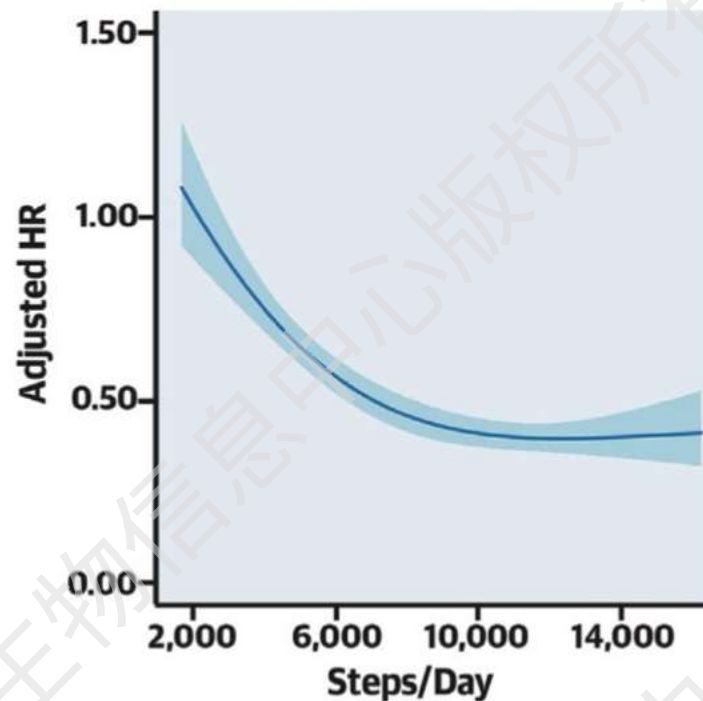


All-cause mortality



Steps	Lower risk
2517	8%
8763	60%
16,000	65%

Cardiovascular disease risk



Steps	Lower risk
2735	11%
7126	51%
16,000	58%

12 cohorts
110k individuals
Meta-analysis

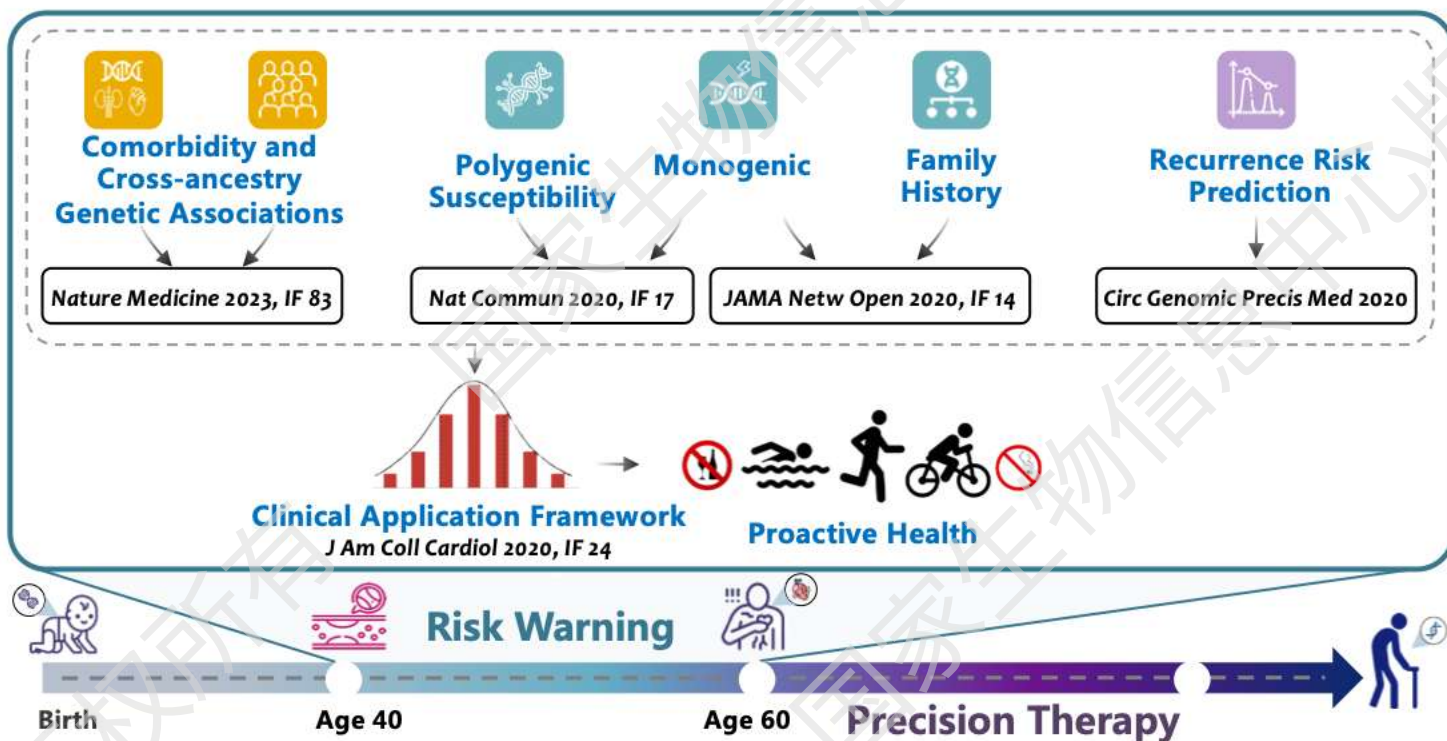
Optimal steps:
8K per day



Integrated Multi-Factor Disease Risk Prediction Model



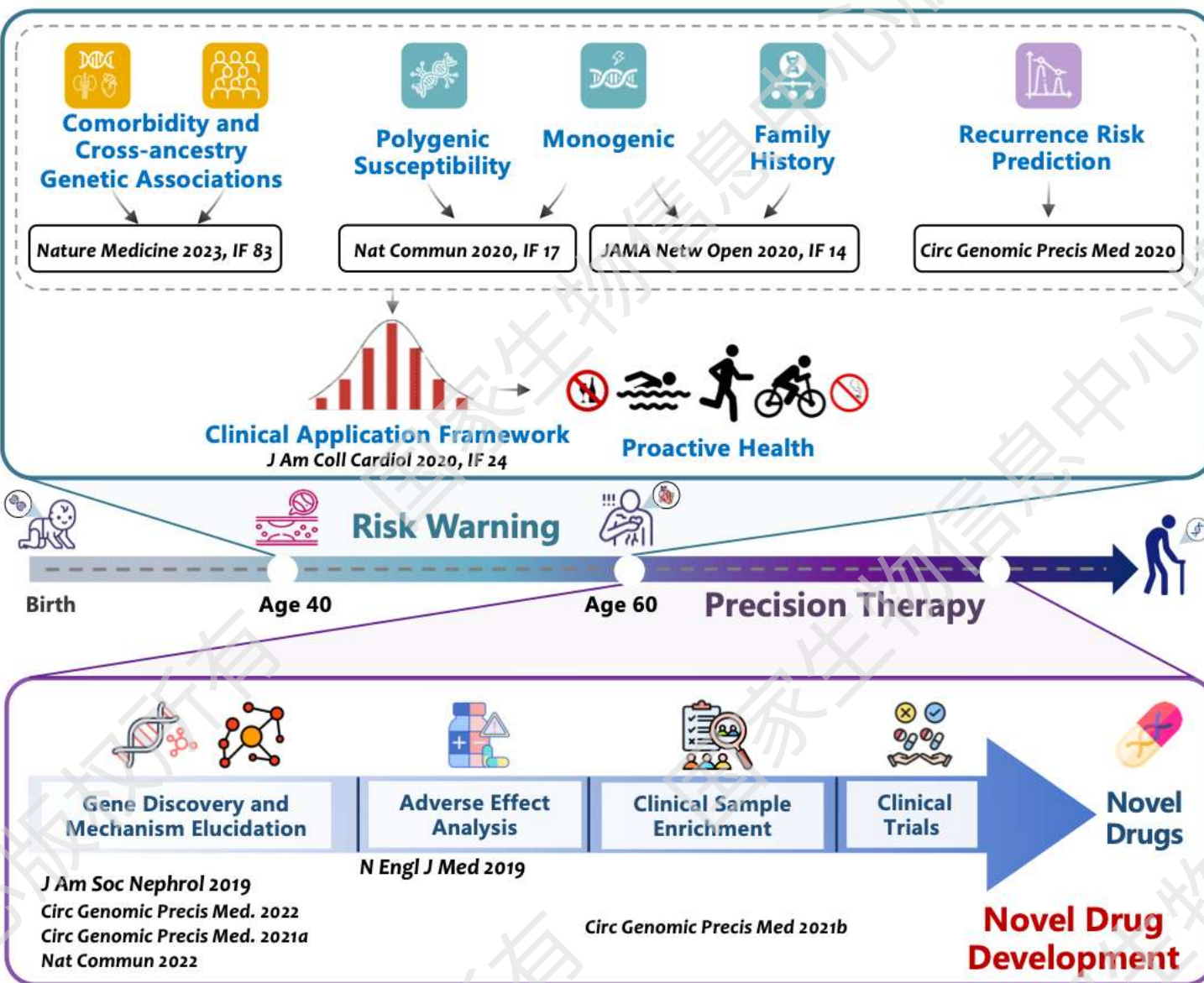
Chronic Disease Risk Management



Cardiovascular and metabolic diseases are influenced by complex interactions between genetic and environmental factors

- **Systematically integrate** multiple risk factors for the occurrence and development of cardiovascular and metabolic diseases, develop personalized risk assessment and early warning systems, promote proactive health, and reduce disease incidence at the population level

Cohort Genomics and Precision Medicine



Cohort Genomics → Precision Medicine

Integrate multi-factor risks → personalized risk assessment & early warning. Promote proactive health, reduce population-level disease incidence

Leveraging large-scale cohort genomic data to **streamline multiple key steps in genetic target discovery**, facilitating innovative drug development for coronary artery disease.

中华心血管病杂志《队列基因组学与冠心病精准医学》

Acknowledgement



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- Amit Khera lab
- Akl Fahed lab
- Aniruddh Patel
- Krishna Aragam
- Saaket Agrawal



INNOVATION TOGETHER

Wang lab

A data-driven
precision medicine lab



Mass General
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BIOBANK JAPAN



WangLab@CNCB
for the prevention and
cure of cardiovascular
diseases

Thank you!

