

Choosing the best models, lessons from diverse complex diseases: Breast Cancer

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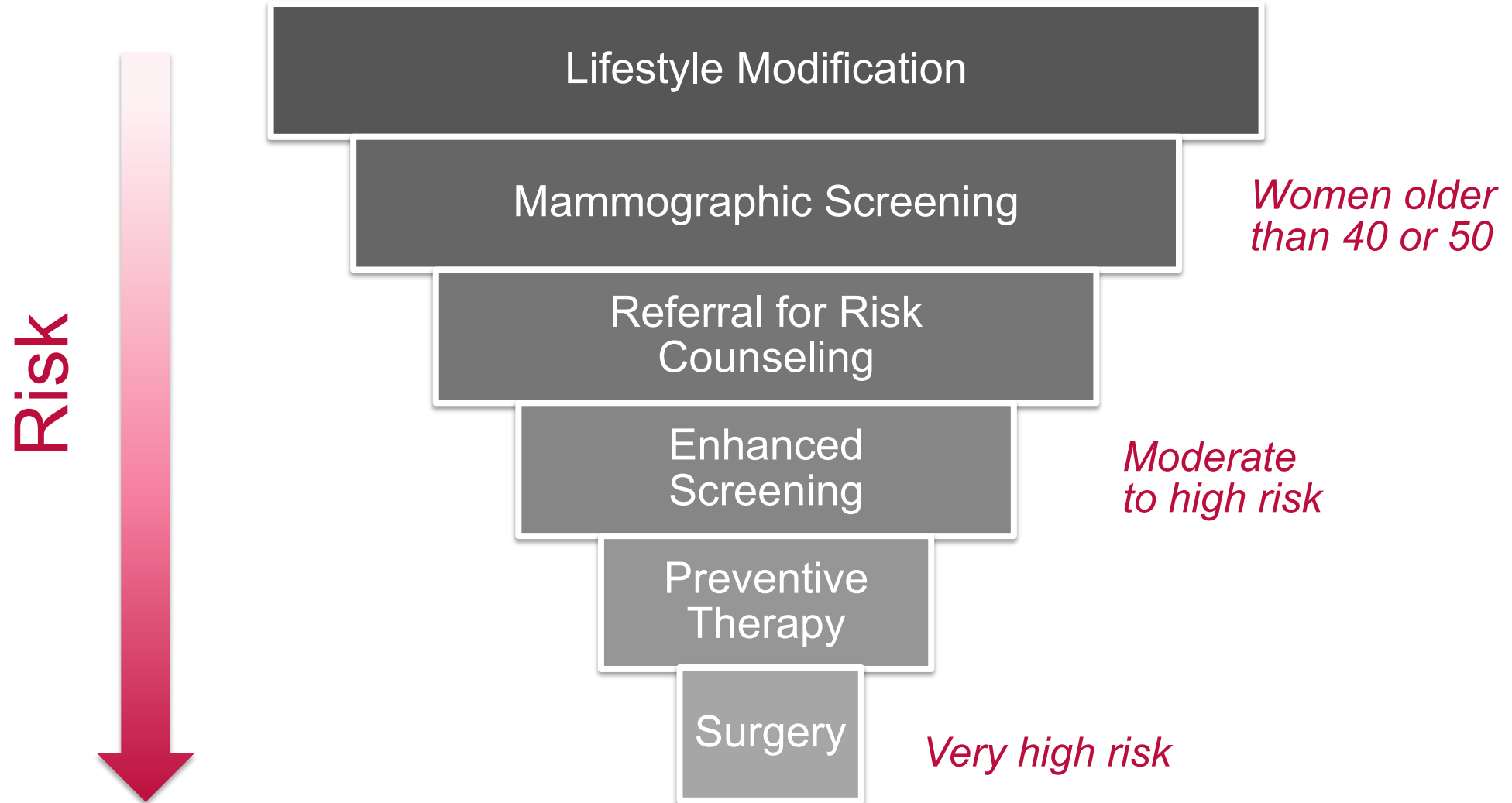
Deputy Director

Division of Cancer Epidemiology and Genetics

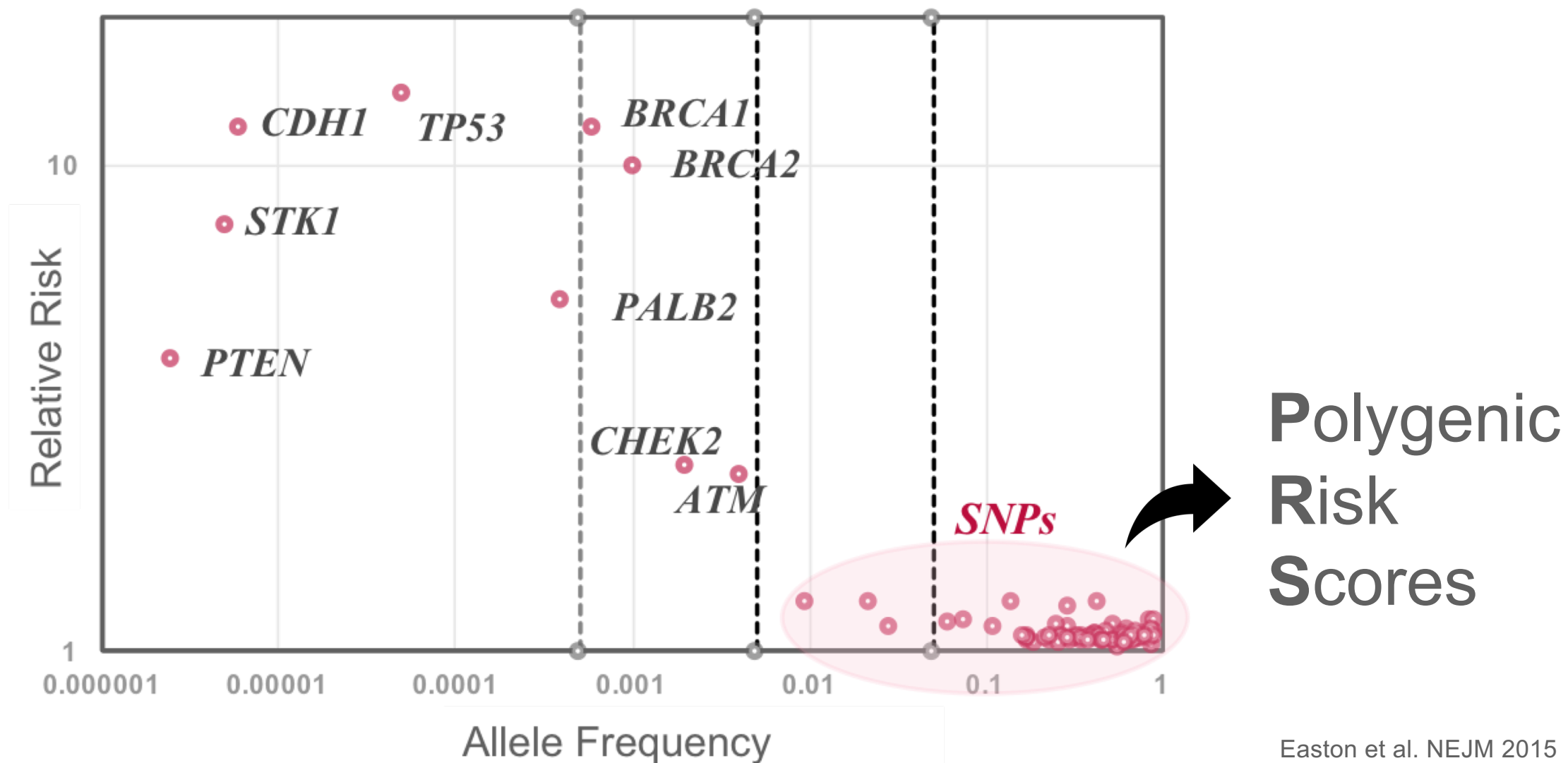
Outline

- Polygenic risk scores (PRS) for breast cancer:
 - How good are they now ?
 - How much better can they get ?
- Integrated breast cancer risk models incorporating PRS
- Future directions

Breast cancer risk reducing strategies tailored to risk



Genetic Architecture of Breast Cancer



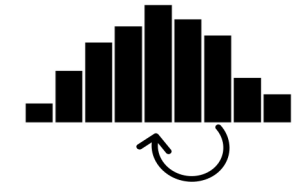
Development of polygenic risk score (PRS)

Breast Cancer Association Consortium (BCAC), European ancestry

PRS	Studies	Cases	Controls
Development	69	94,075	75,017
Validation	10	11,428	18,323

Genome-wide arrays plus imputation (~7 million SNPs)

PRS and breast cancer risk



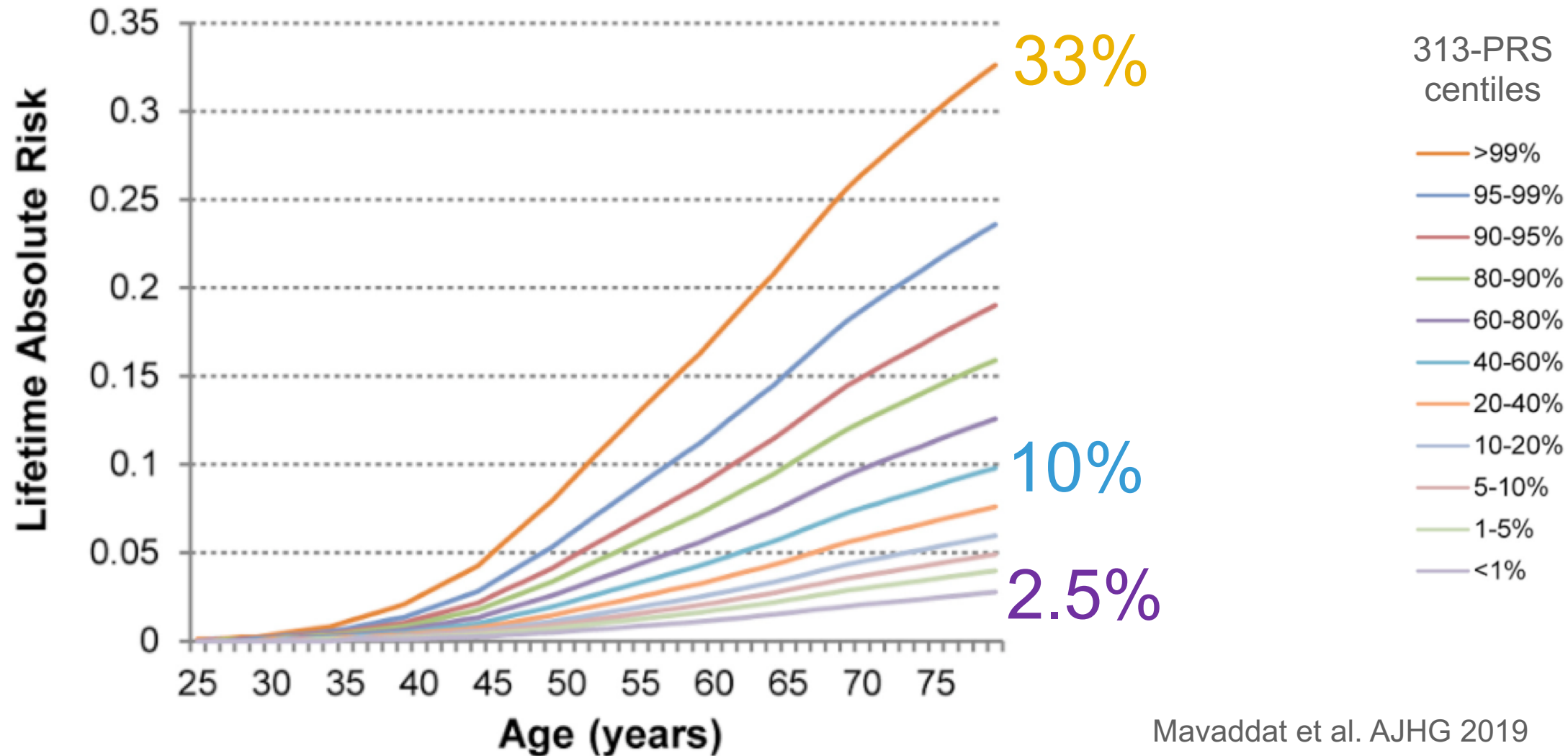
PRS	Relative Risk per SD	95% CI	AUC (%)
77-SNP*	1.46	(1.42-1.49)	60.3
313-SNP	1.61	(1.57-1.65)	63.0
3,820-SNP	1.66	(1.61-1.70)	63.6

*Mavaddat et al. JNCI 2015

313-PRS by ER status of breast tumors

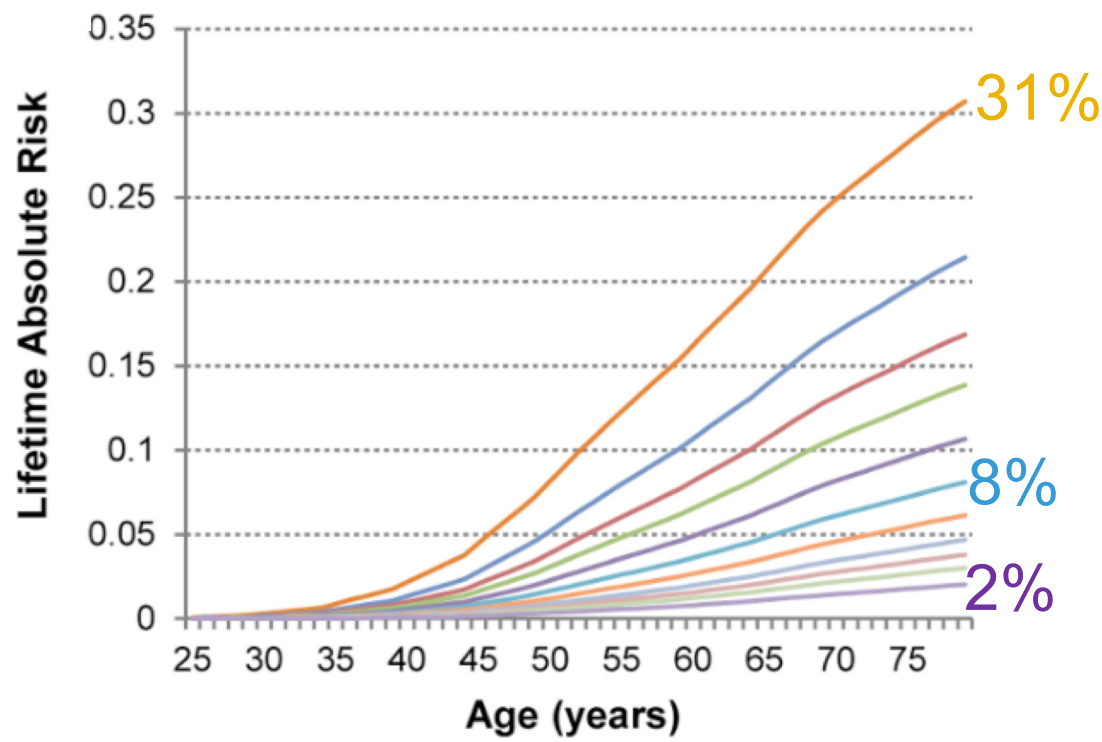
PRS	Overall		ER-Positive		ER-negative	
	OR	95% CI	OR	95% CI	OR	95% CI
BCAC (European)	1.61	1.57 – 1.65	1.68	1.63 – 1.73	1.45	1.37 - 1.53

Absolute risk stratification of 313-SNP PRS – UK Women

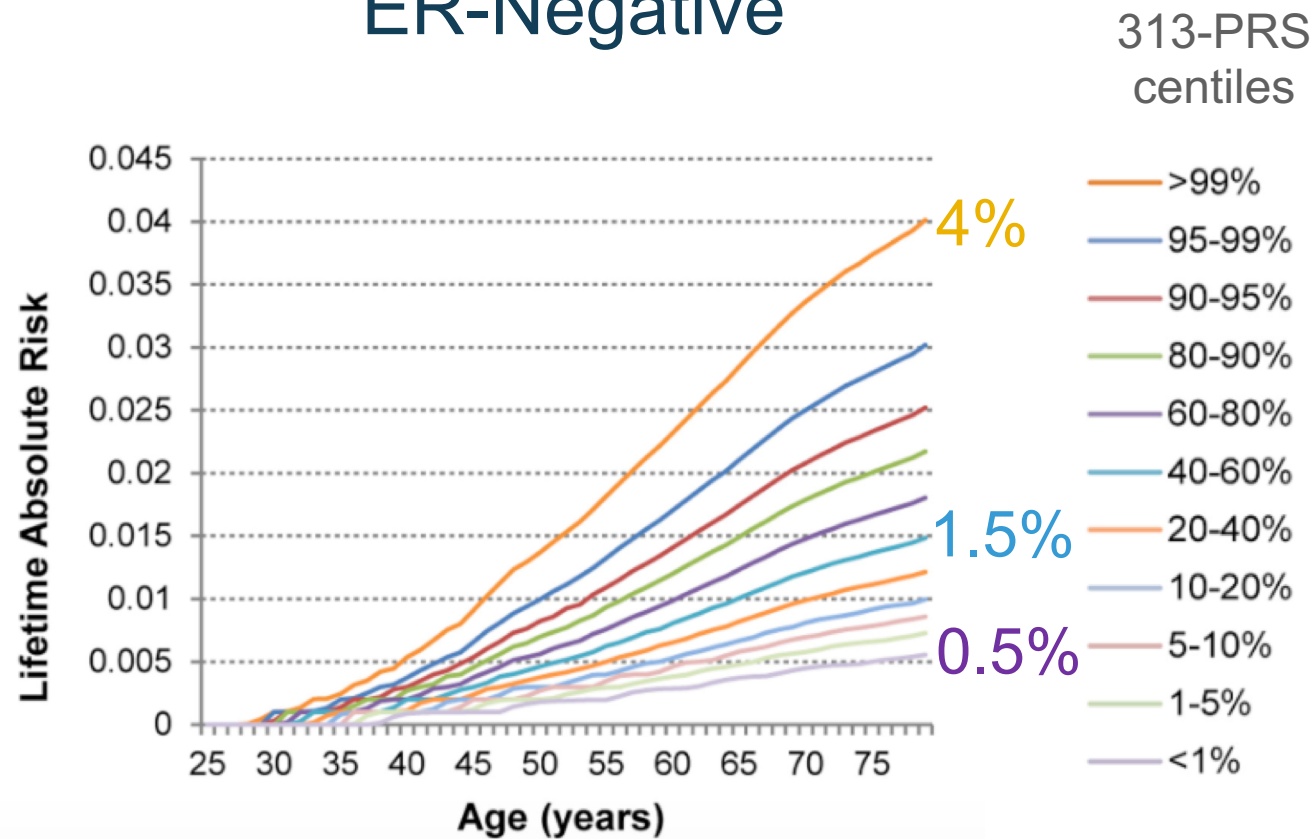


Absolute risk stratification of 313-SNP PRS – UK Women

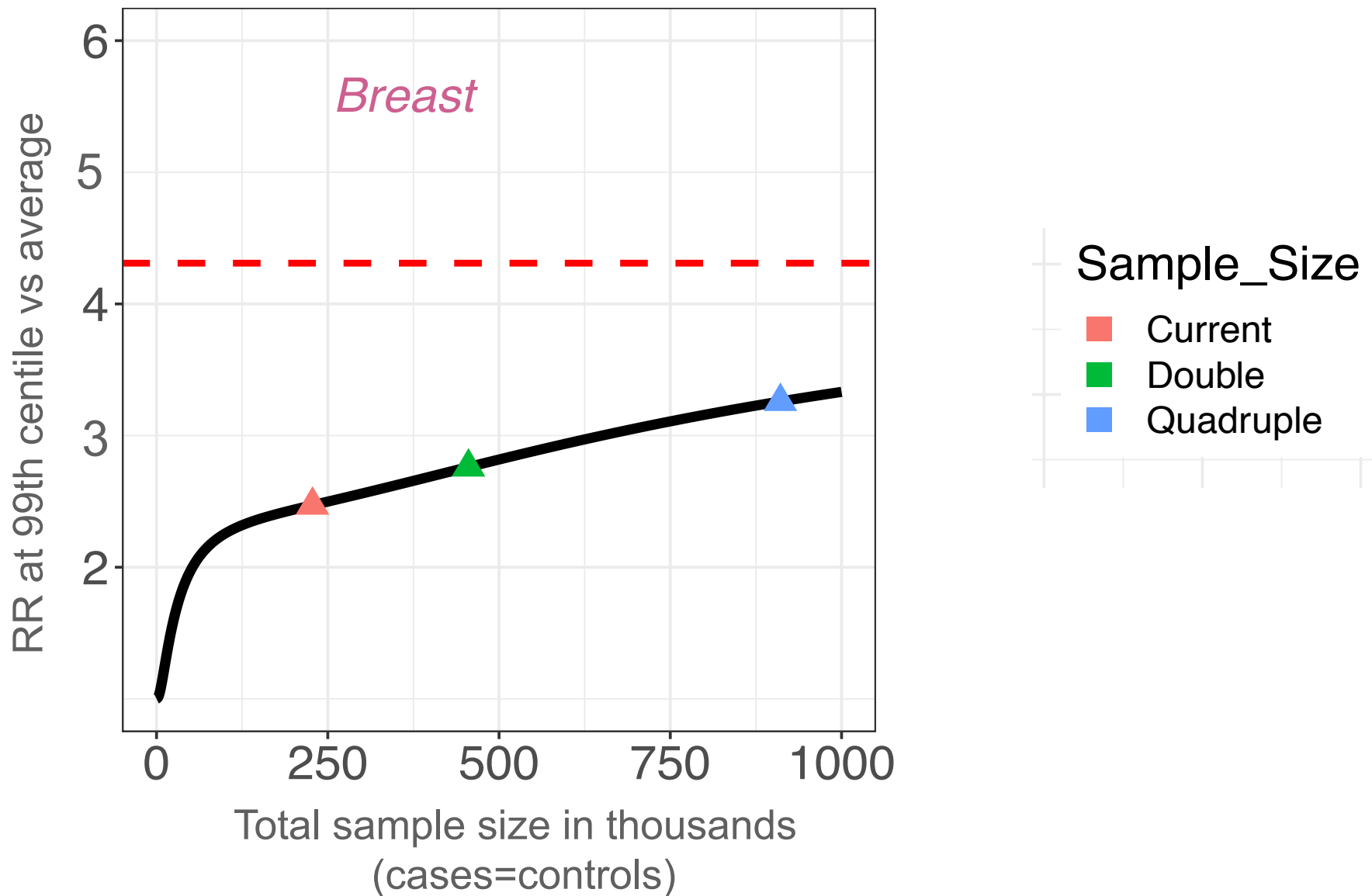
ER-Positive



ER-Negative

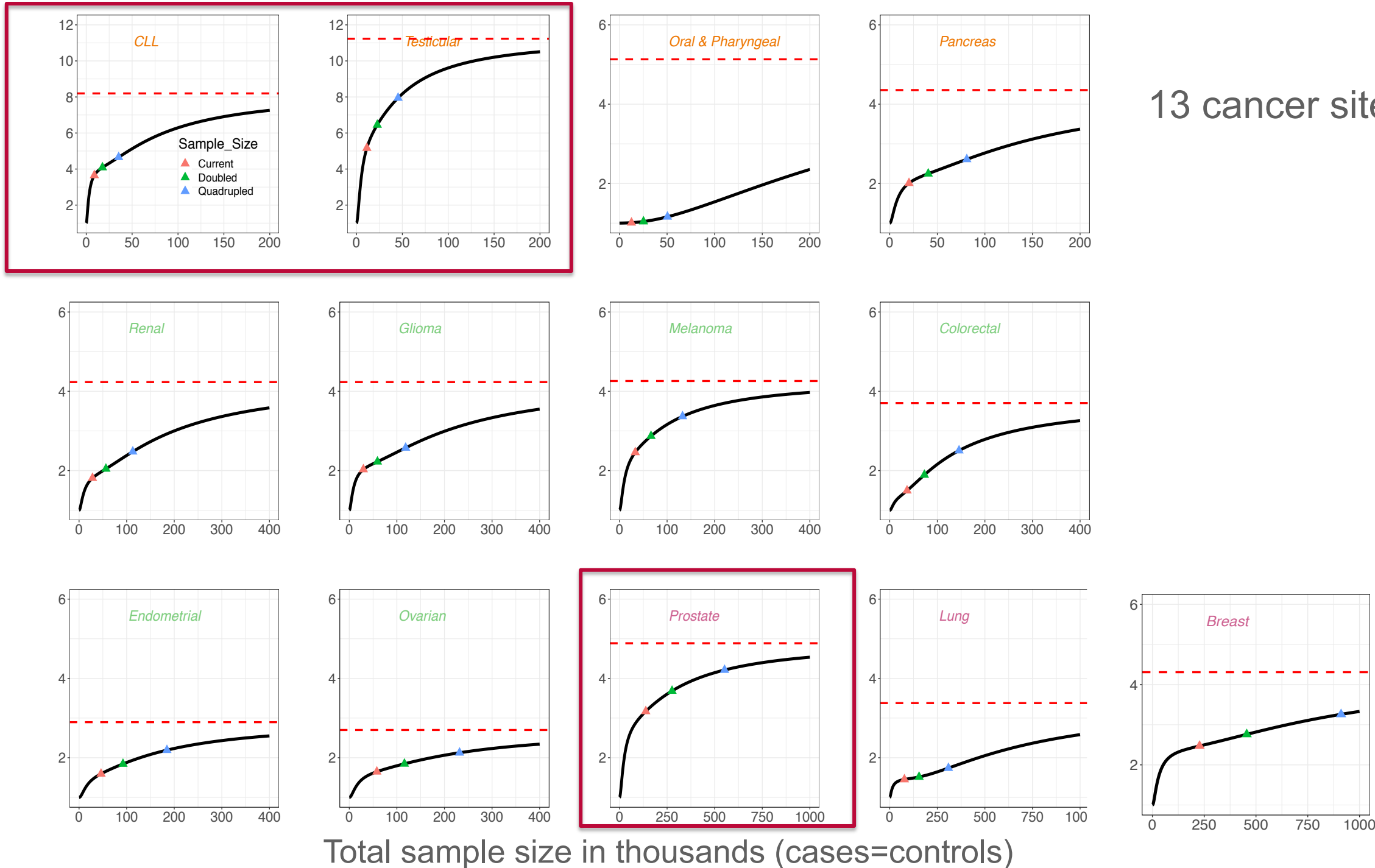


Projected gains in **risk stratification** with increasing GWAS size

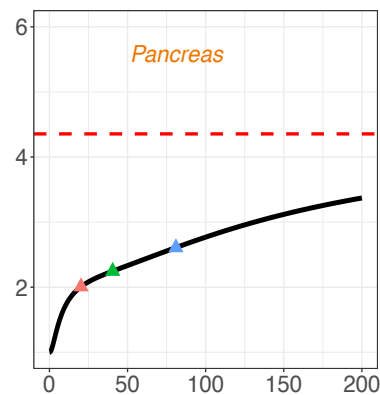
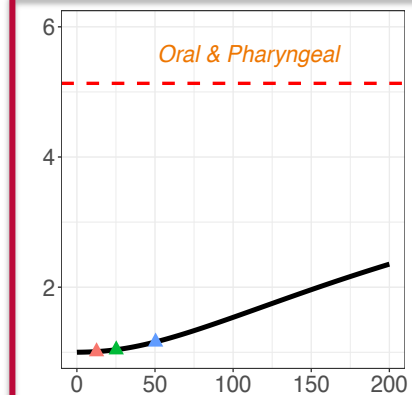
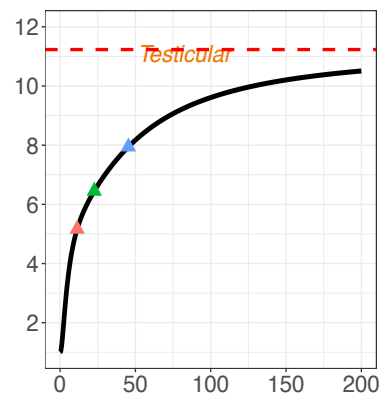
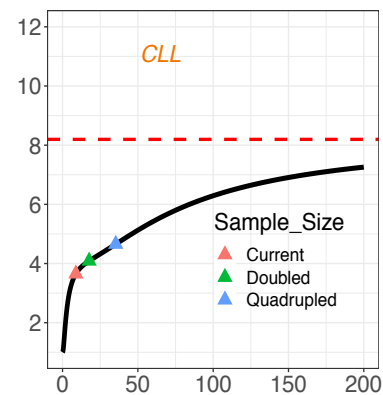


RR at 99th centile vs average

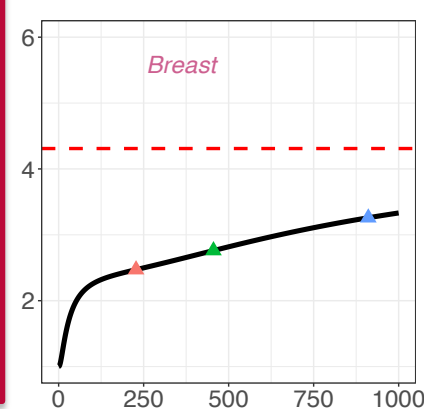
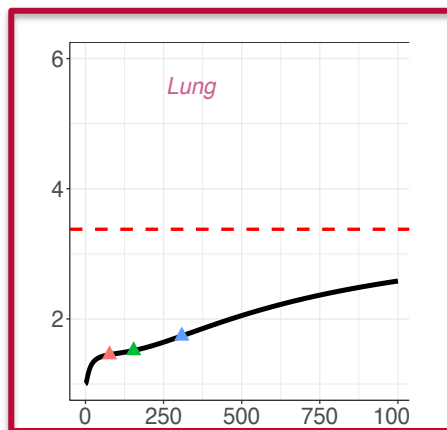
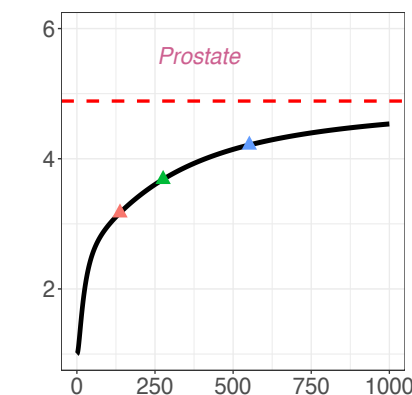
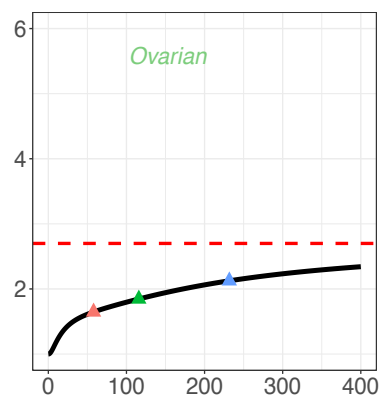
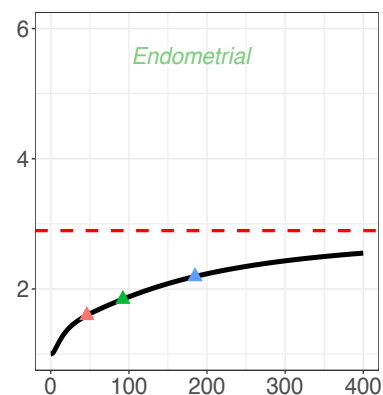
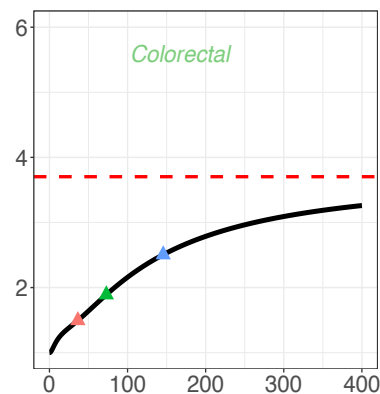
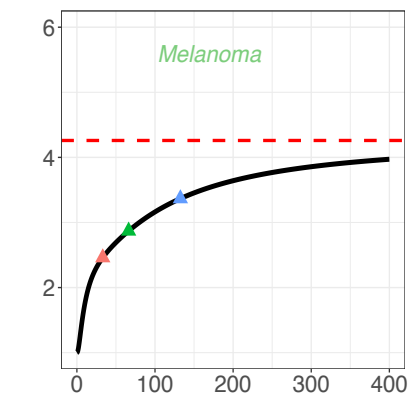
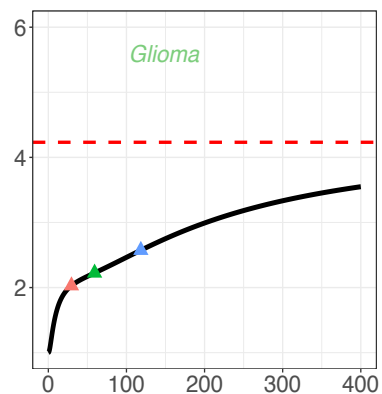
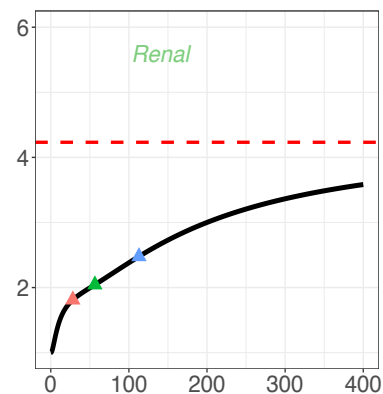
13 cancer sites



RR at 99th centile vs average

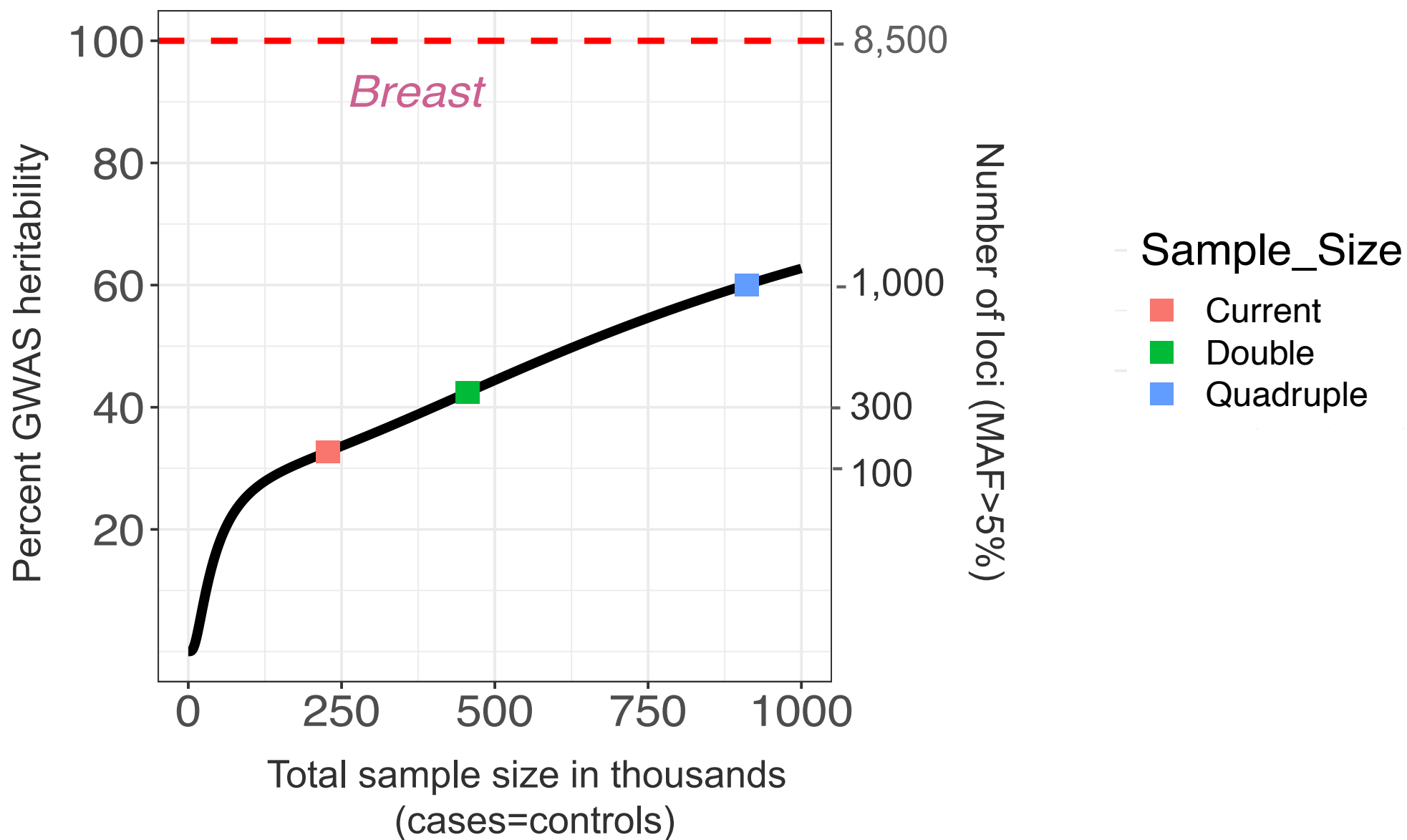


13 cancer sites



Total sample size in thousands (cases=controls)

Projected gains in number of loci and percent GWAS heritability





Confluence

Uncovering breast cancer genetics

More than doubling the size of
current breast cancer GWAS



>300,000
Breast Cancer
Patients



>300,000
Controls



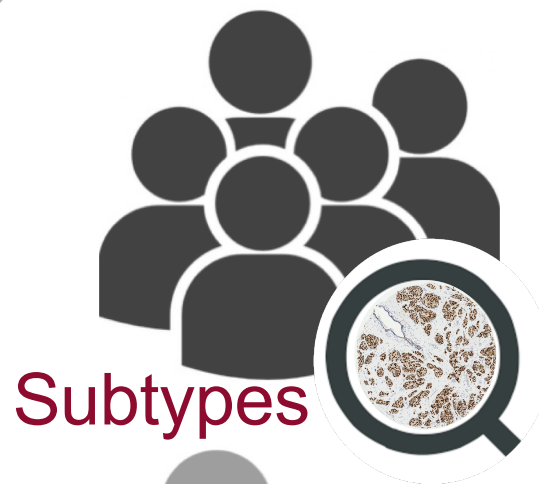
International
Multi-Ancestry



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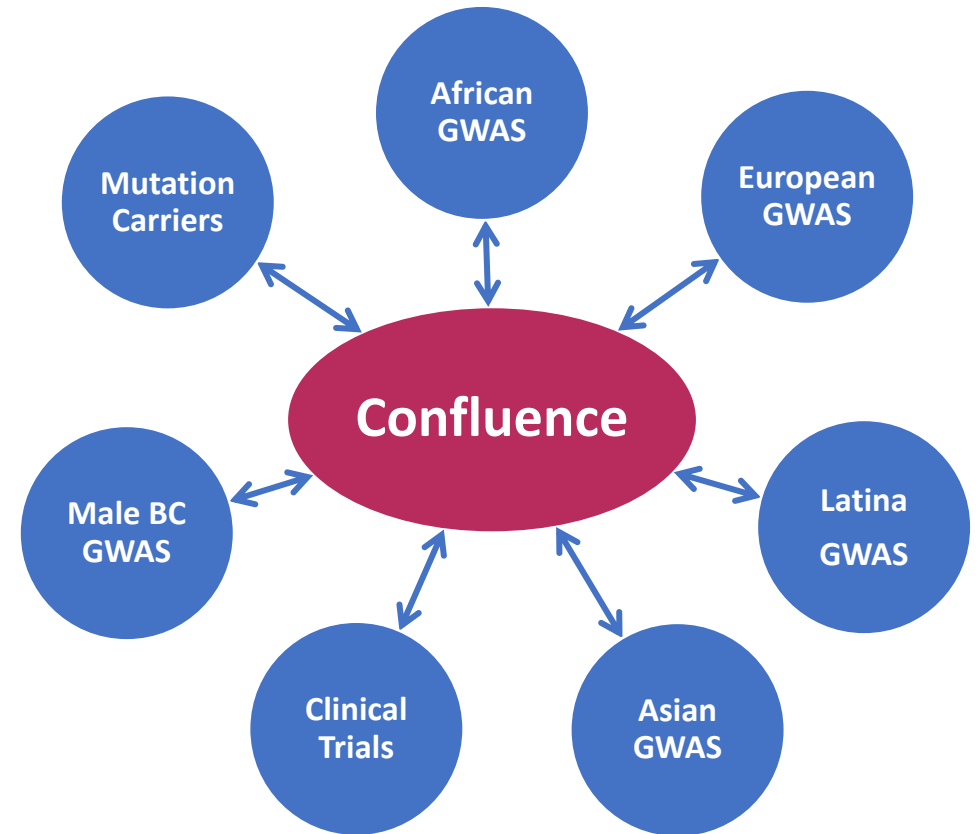
>300,000
Controls



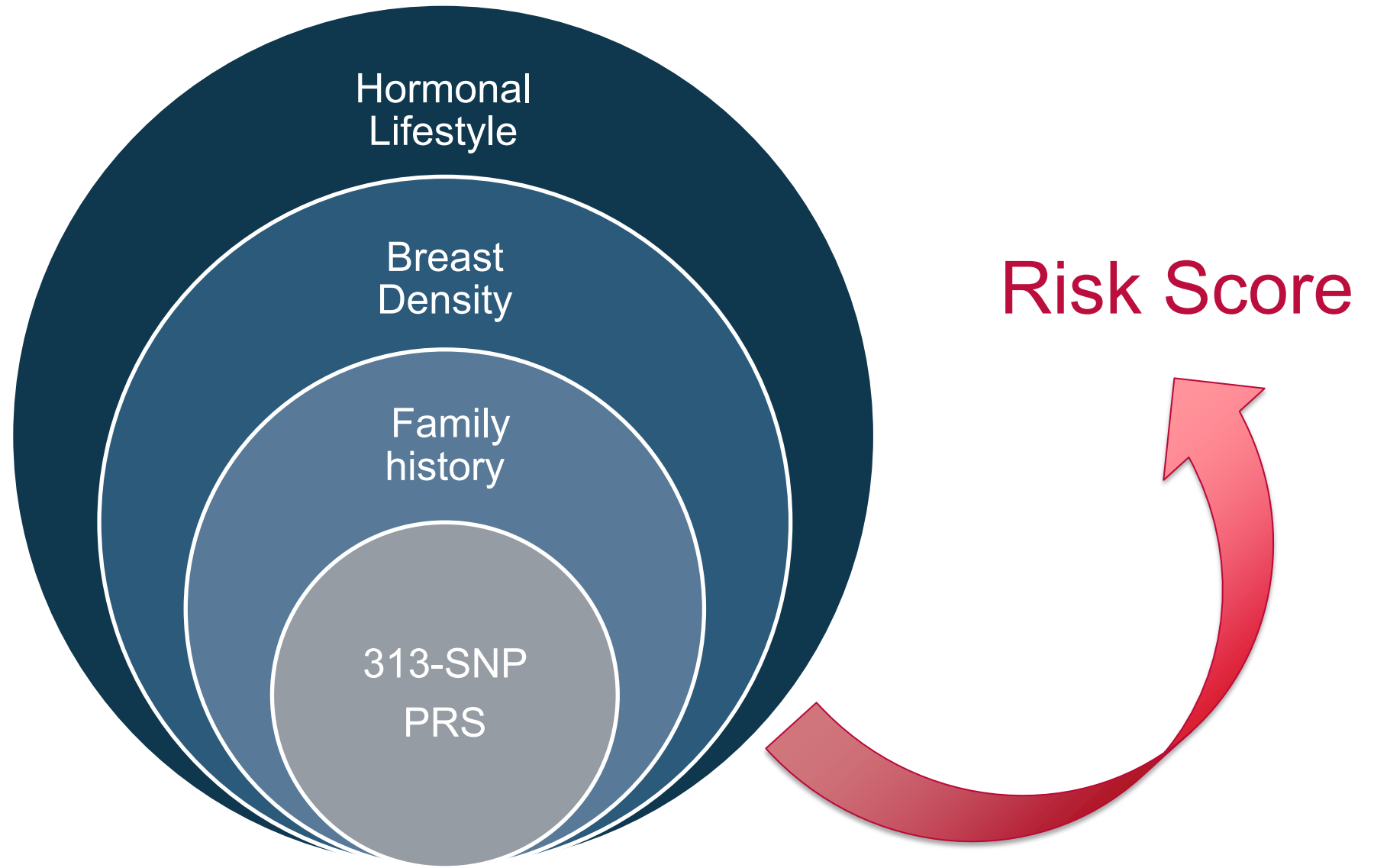
International
Multi-Ancestry

Consortium of consortia: confluence of research resources

- Breast Cancer Association Consortium (BCAC)
- Consortium of Investigators of Modifiers of BRCA1/2 (CIMBA)
- African-Ancestry Breast Cancer Genetic Study (AABCGS)
- Asia Breast Cancer Consortium (ABCC)
- Latin America Genomics Breast Cancer Consortium (LAGENO- BC)

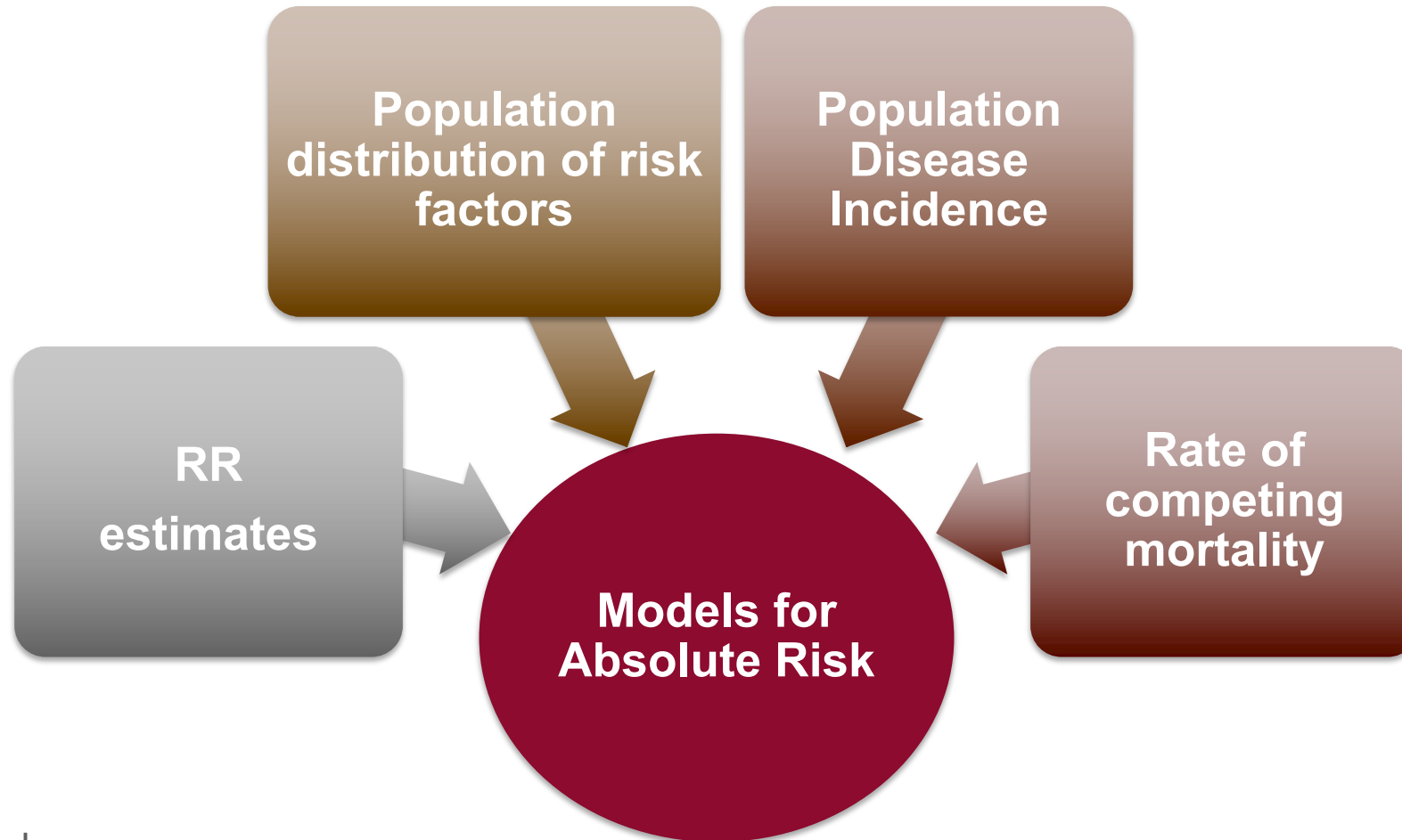


Building integrated breast cancer risk models



Flexible tool to develop and validate absolute risk models

Individualized Coherent Absolute Risk Estimator (iCARE)



Performance of risk prediction models



- **Calibration**

- How close is predicted to actual risk ?



- **Discrimination**

- How well can predicted risk stratify women?



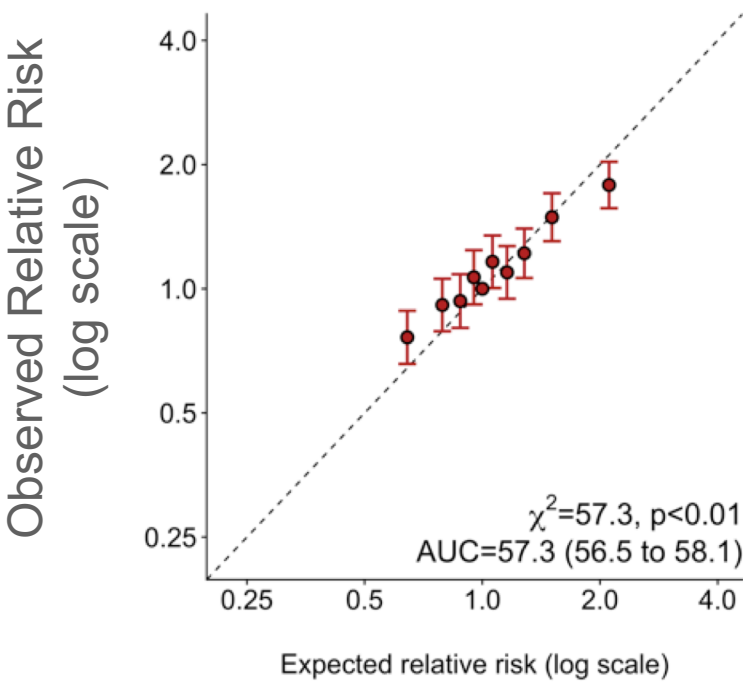
- **Personal and clinical utility**

- How useful is the prediction to make decisions ?

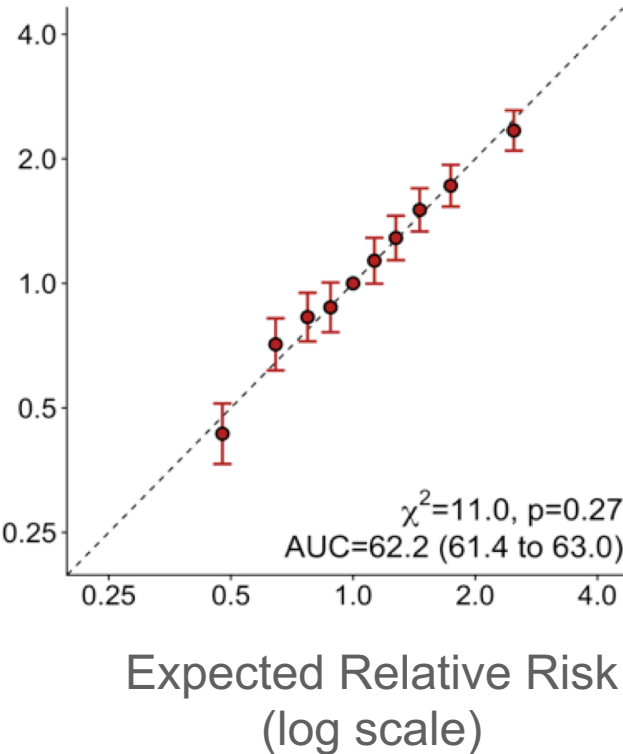
Relative Risk Calibration Risk Models

Meta-analysis for 11 cohorts with 232,743 women, 5,920 breast cancers
Age ≥ 50 years

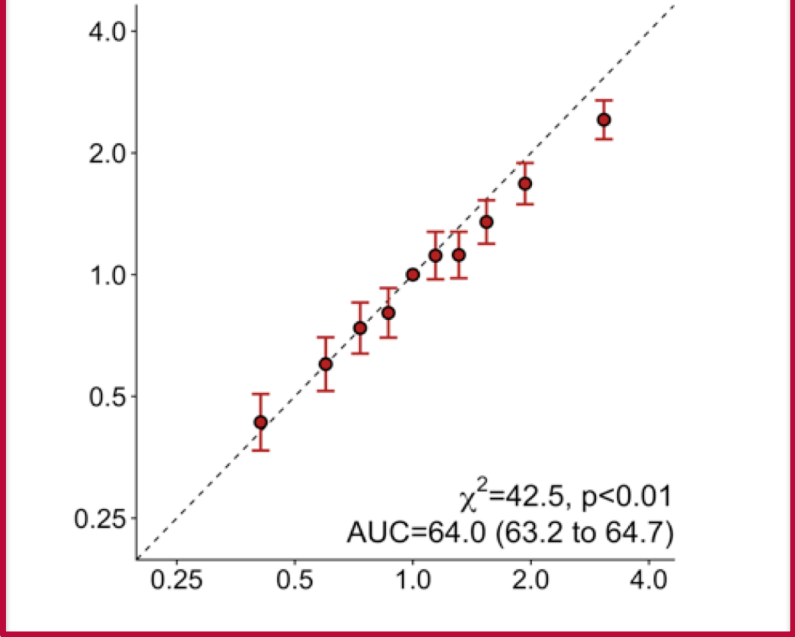
Classical Risk Factors



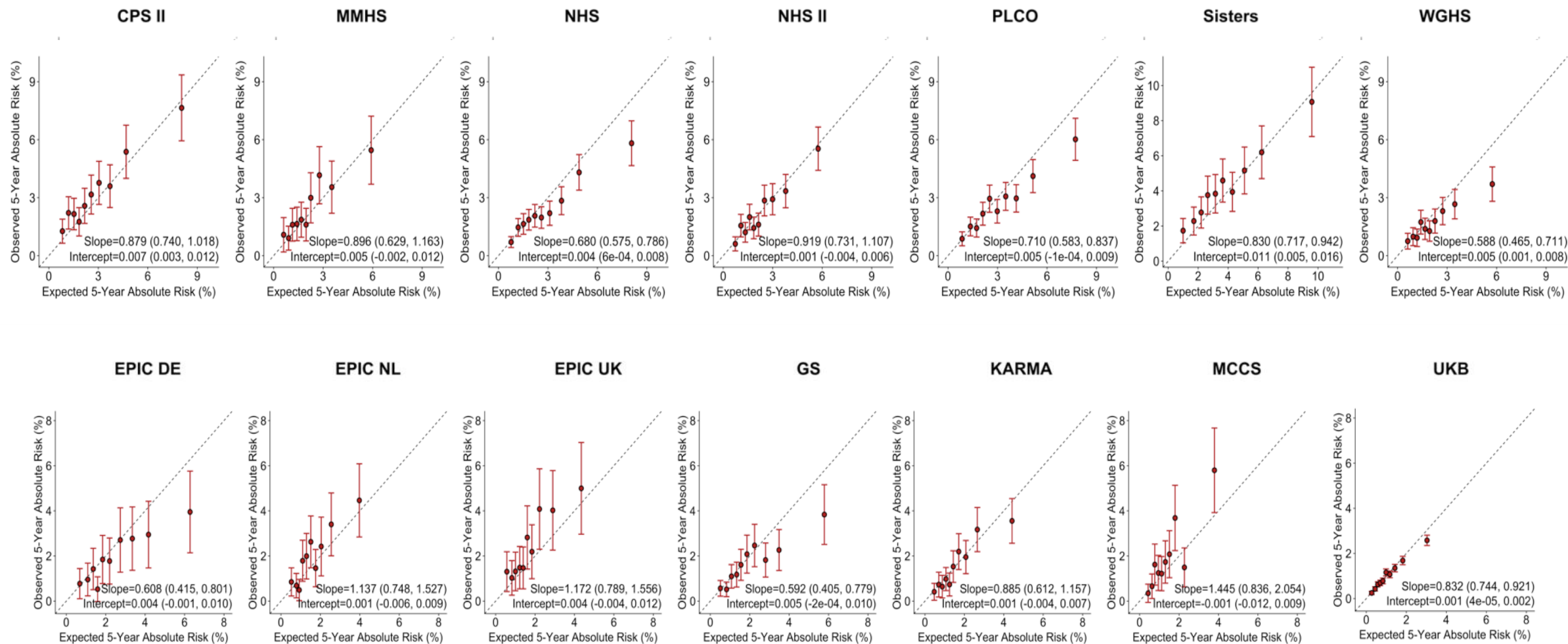
313-SNP PRS



Classical + PRS



Absolute risk calibration across multiple study populations



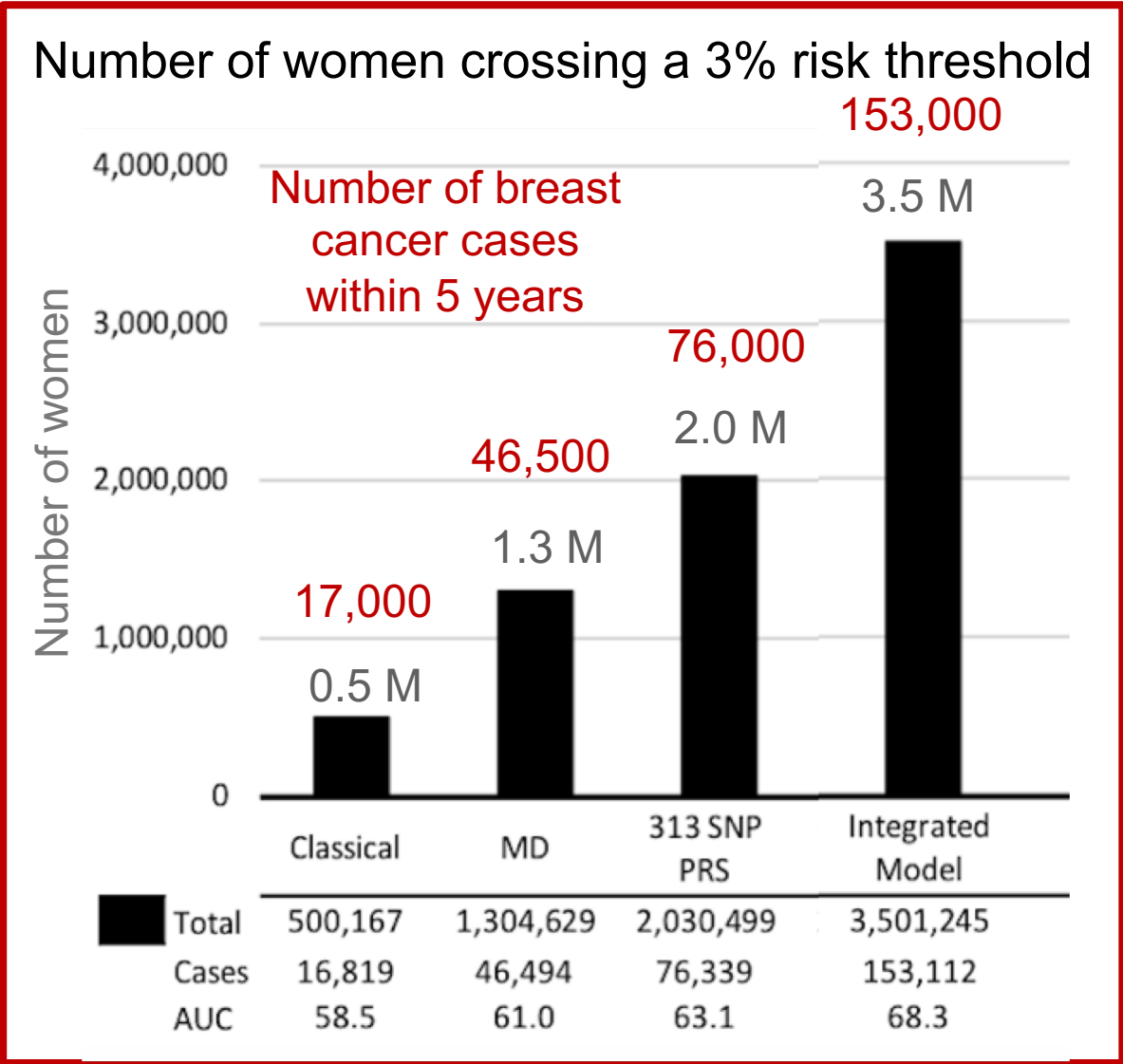
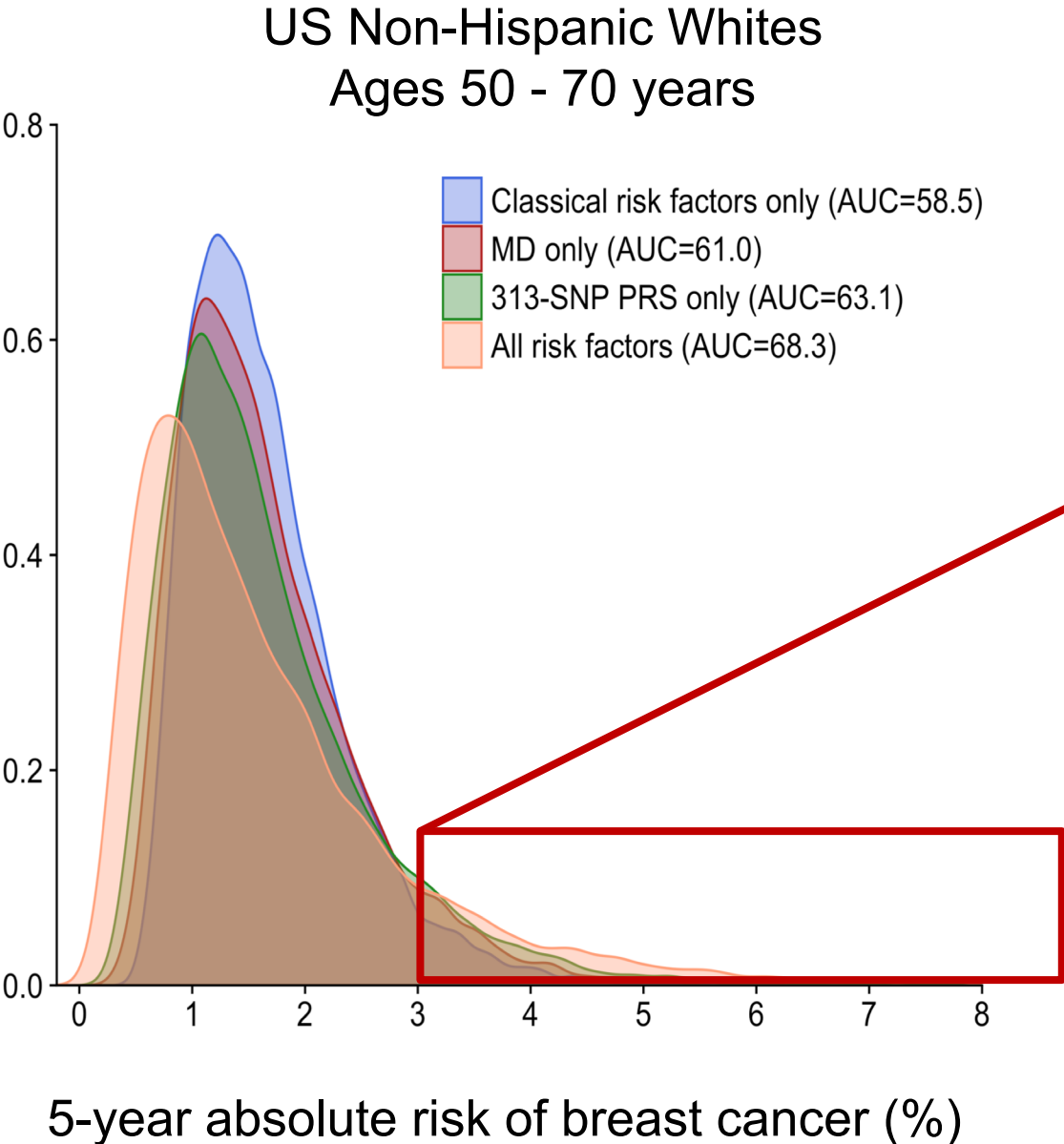


connect

for **cancer** prevention | ADVANCING CANCER
RESEARCH TOGETHER

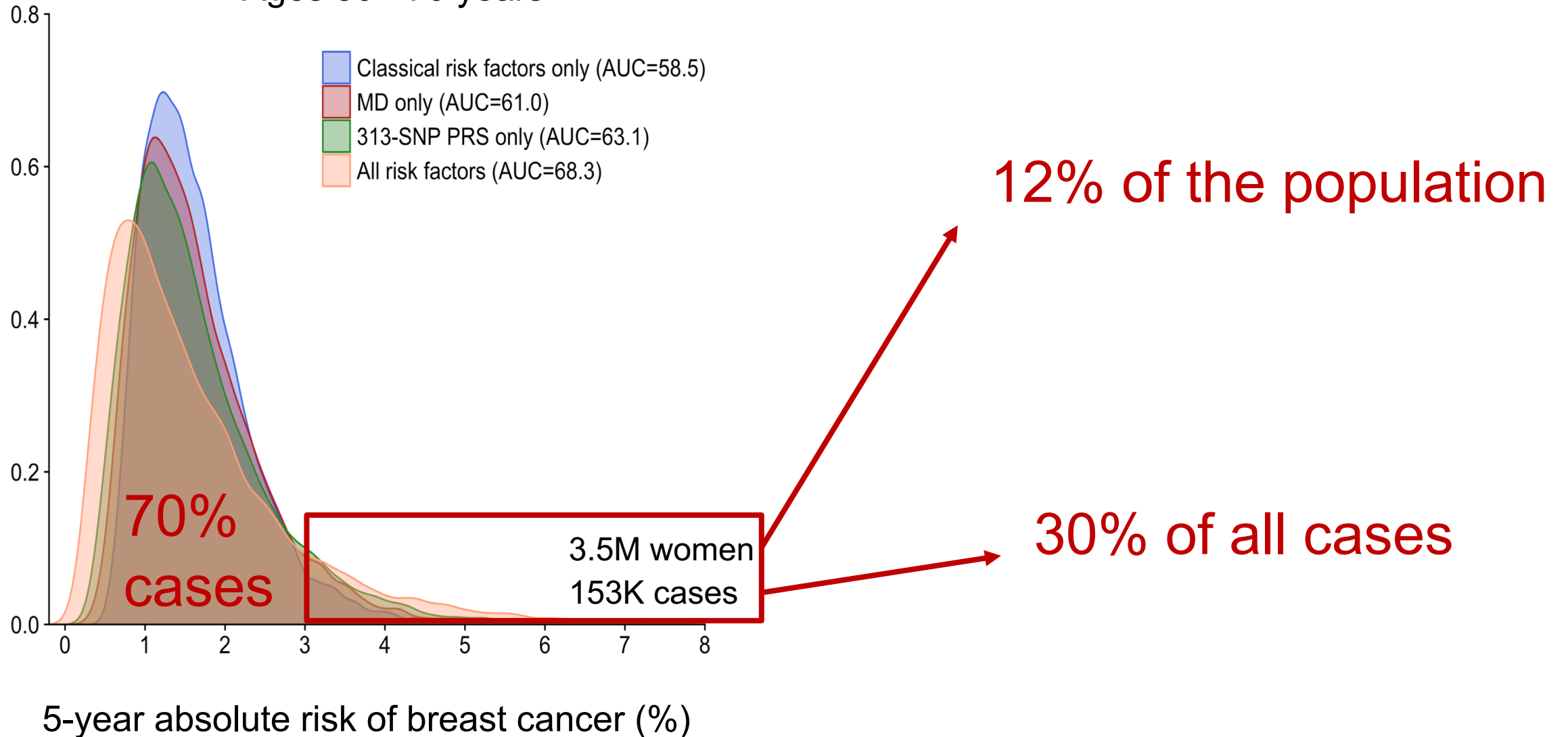
- 200,000 adults
- Long-term follow up
- Serial assessments
- Comprehensive
- Tissue collections
- Case ascertainment

Breast cancer risk stratification by different risk factors

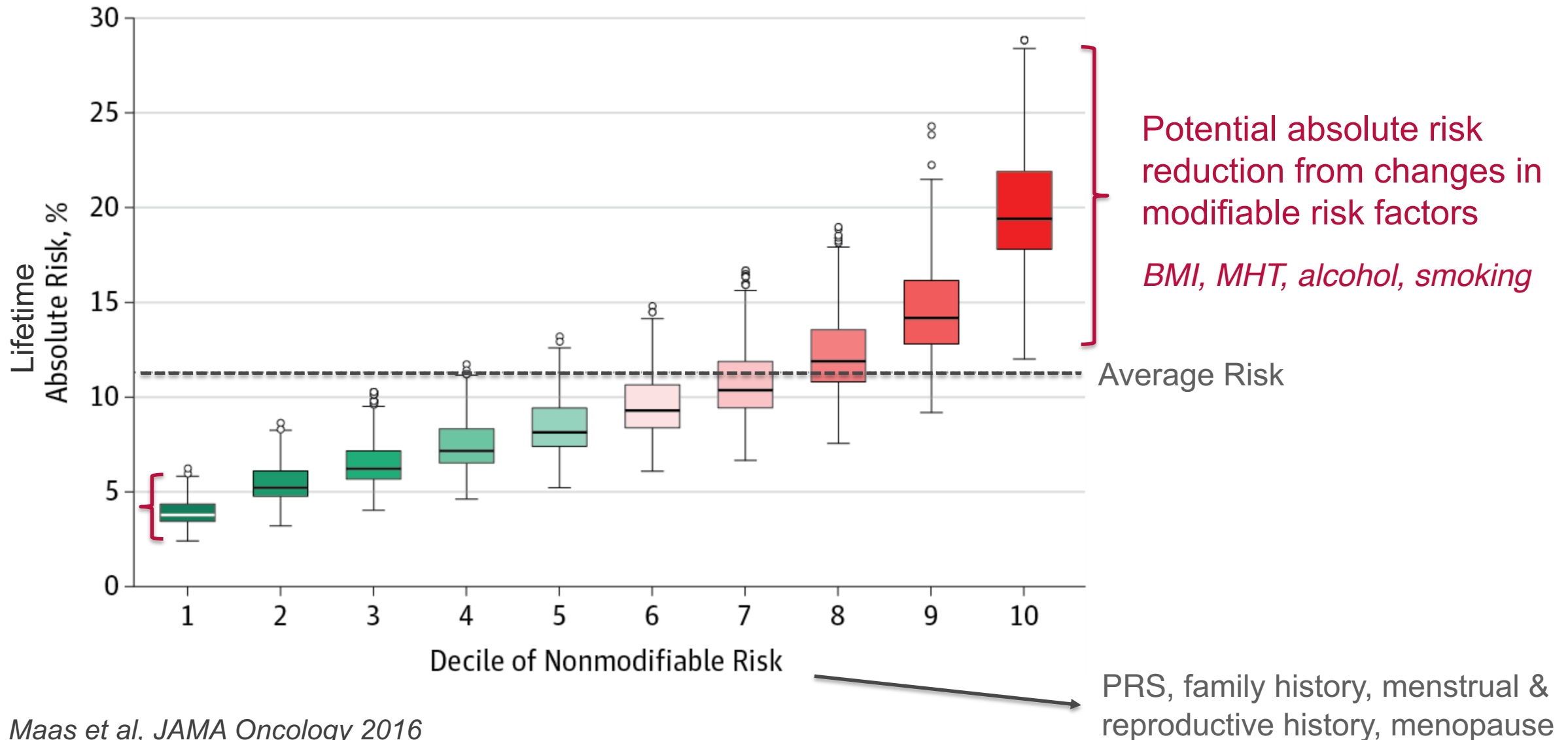


Breast cancer risk stratification by different risk factors

US Non-Hispanic Whites
Ages 50 - 70 years



Distribution of modifiable risk by deciles of non-modifiable risk



Conclusions

- PRS adds substantially to risk stratification, with good model calibration
- Integrated models can identify larger numbers of women at elevated risk risk -> targeted strategies
- Most cases will still occur outside “high-risk” groups -> population-wide strategies

Future directions

- Improve PRS, subtype, ancestry-specific
- New risk factors/biomarkers needed to further improve risk stratification
- Better preventive and early detection strategies needed to control the disease

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Jenny Chang-Claude

University of Melbourne

Roger Milne

Karolinska Institute

Per Hall

Laval University

Jaques Simard

University of Queensland

Georgia Chenevix-Trench

Prospective Cohorts for Breast Cancer Risk Modelling

GS: Swerdlow, Brook

CPS II: Brian Carter, Mia Gaudet

EPIC: Anika Husing, Myrto Barrdahl

KARMA: Mikael Eriksson, Per Hall

MCCS: Kara Martin, Roger Milne

MMHS: Chris Scott, Celine Vachon

NHSI/NHSII, WGHS: Chi Gao, Pete Kraft

PLCO/UKBiobank: Garcia-Closas, Chatterjee

PROCAS: Elaine Harkness, Gareth Evans

SISTERS: Min Shi, Clarise Weinberg, Dale Sandler