

# Lecture 4

## Meta-Analysis in GWAS: An Overview and Applications

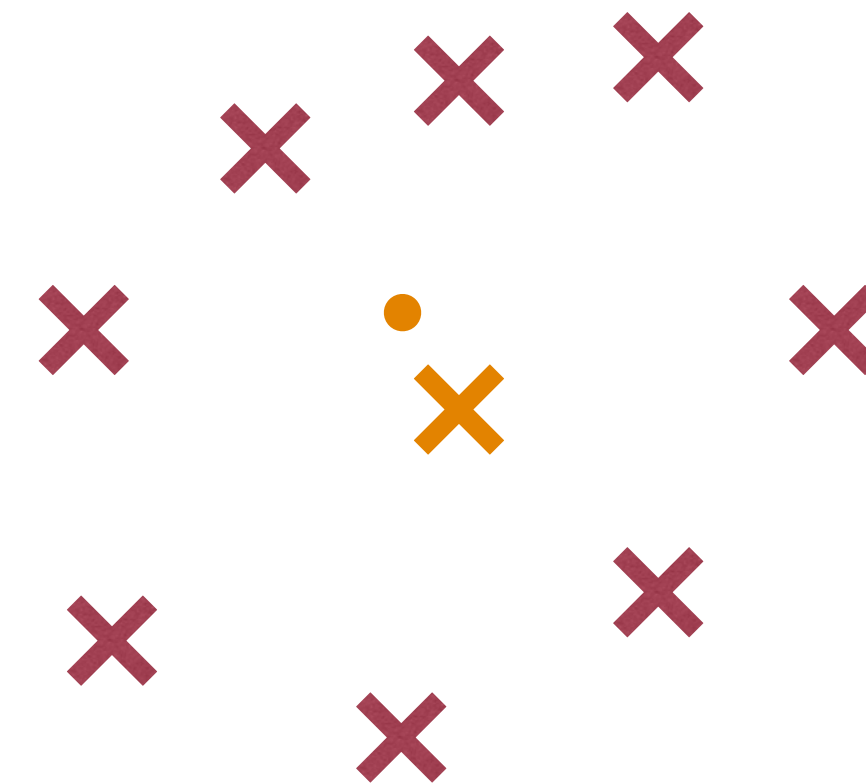
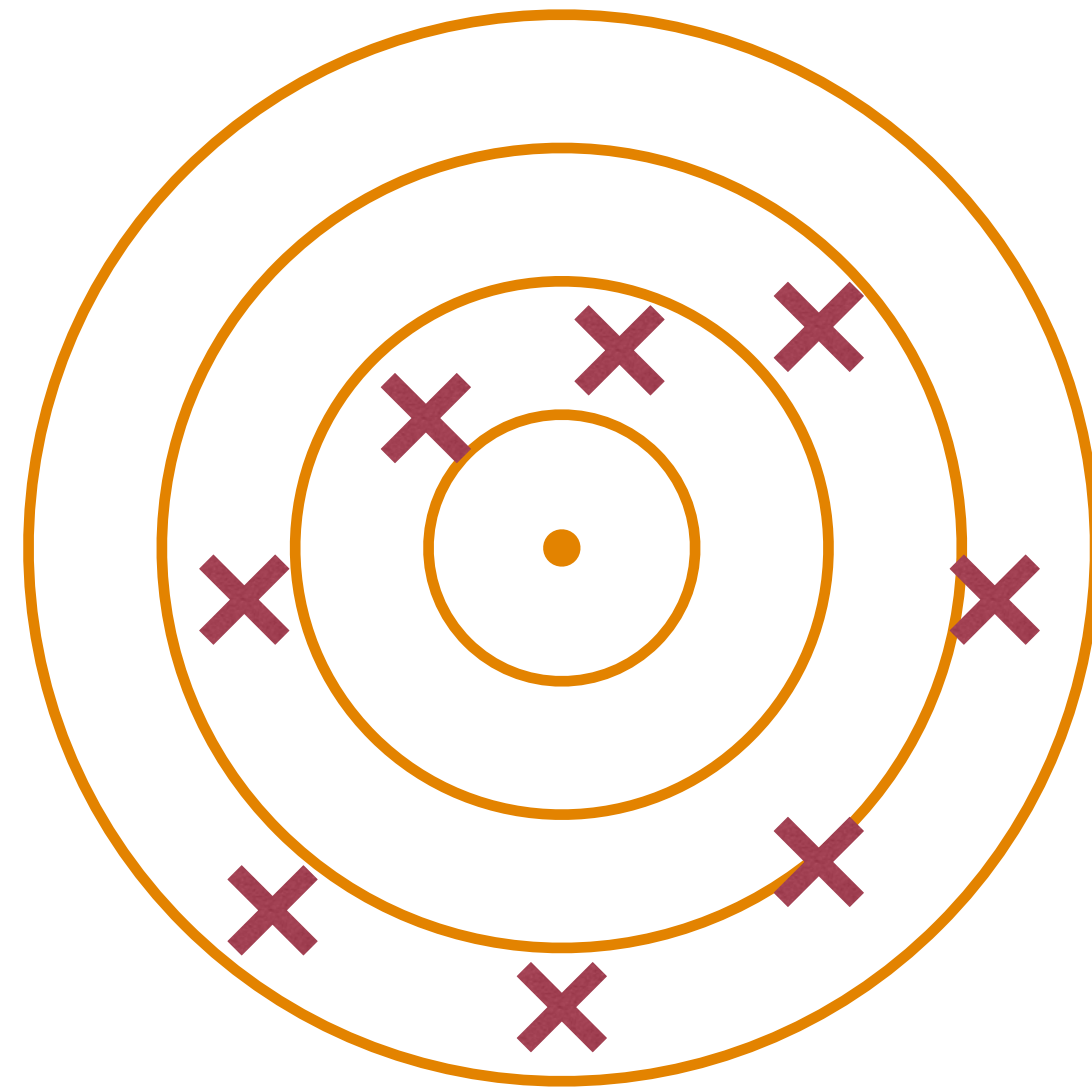
by Dr. Mustafa İsmail Özkaraca

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1. What is Meta-Analysis?
2. Why Meta-Analyse in GWAS?
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# What is Meta-Analysis?

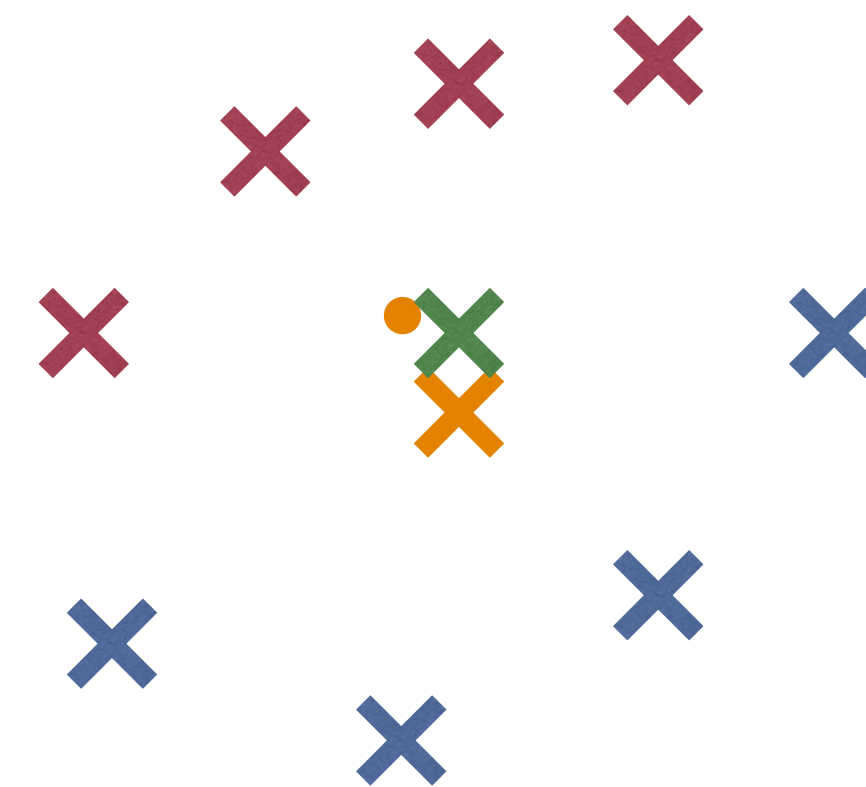
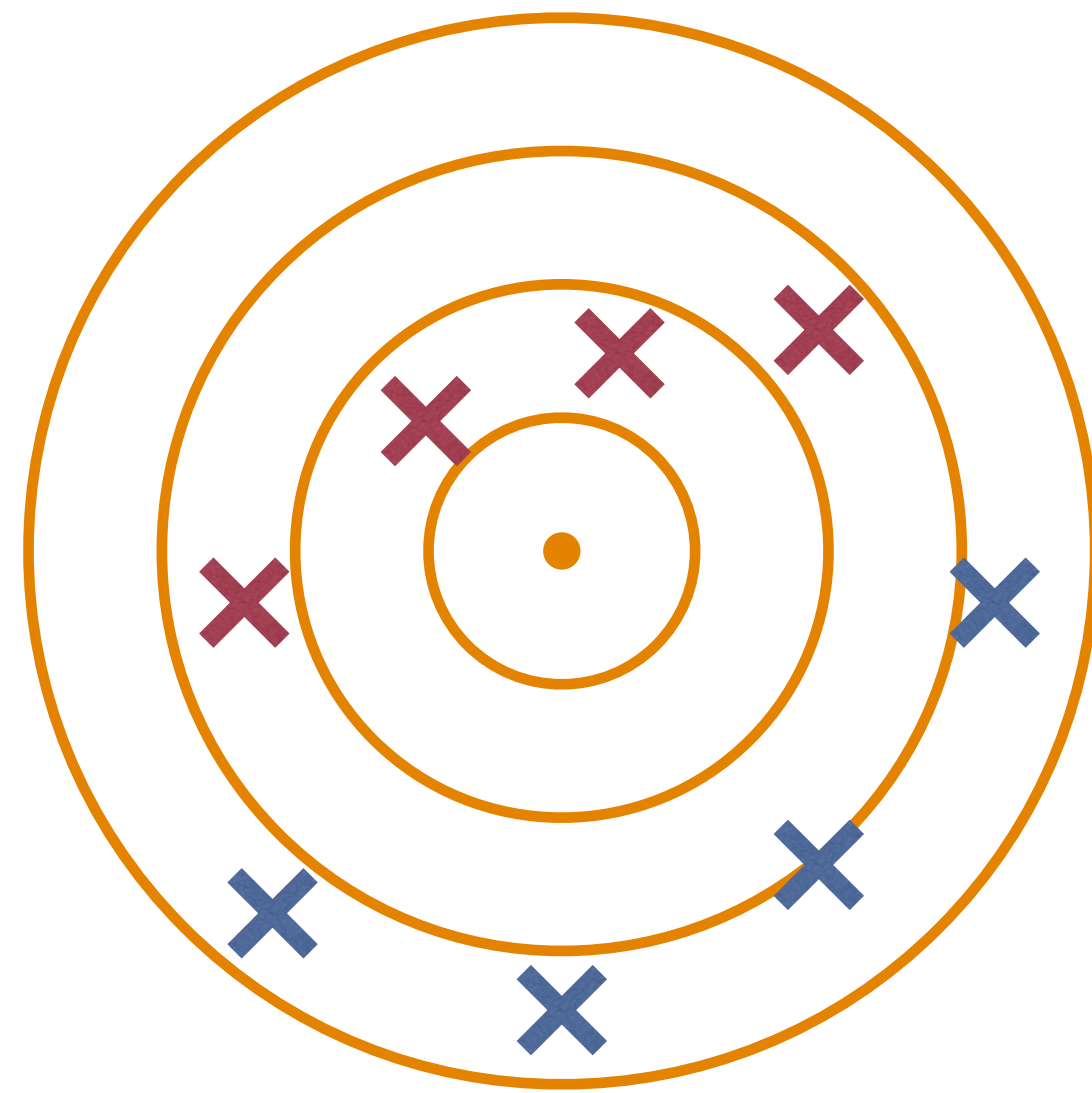
Q: Where is the **centre** of the board?



- ✕ Single estimates
- ✕ Center of 8 estimates (**Meta Estimate**)
- True center

# What is Meta-Analysis?

Q: Where is the **centre** of the board?



× Single estimates (Worse)  
× **Informed Meta Estimate**

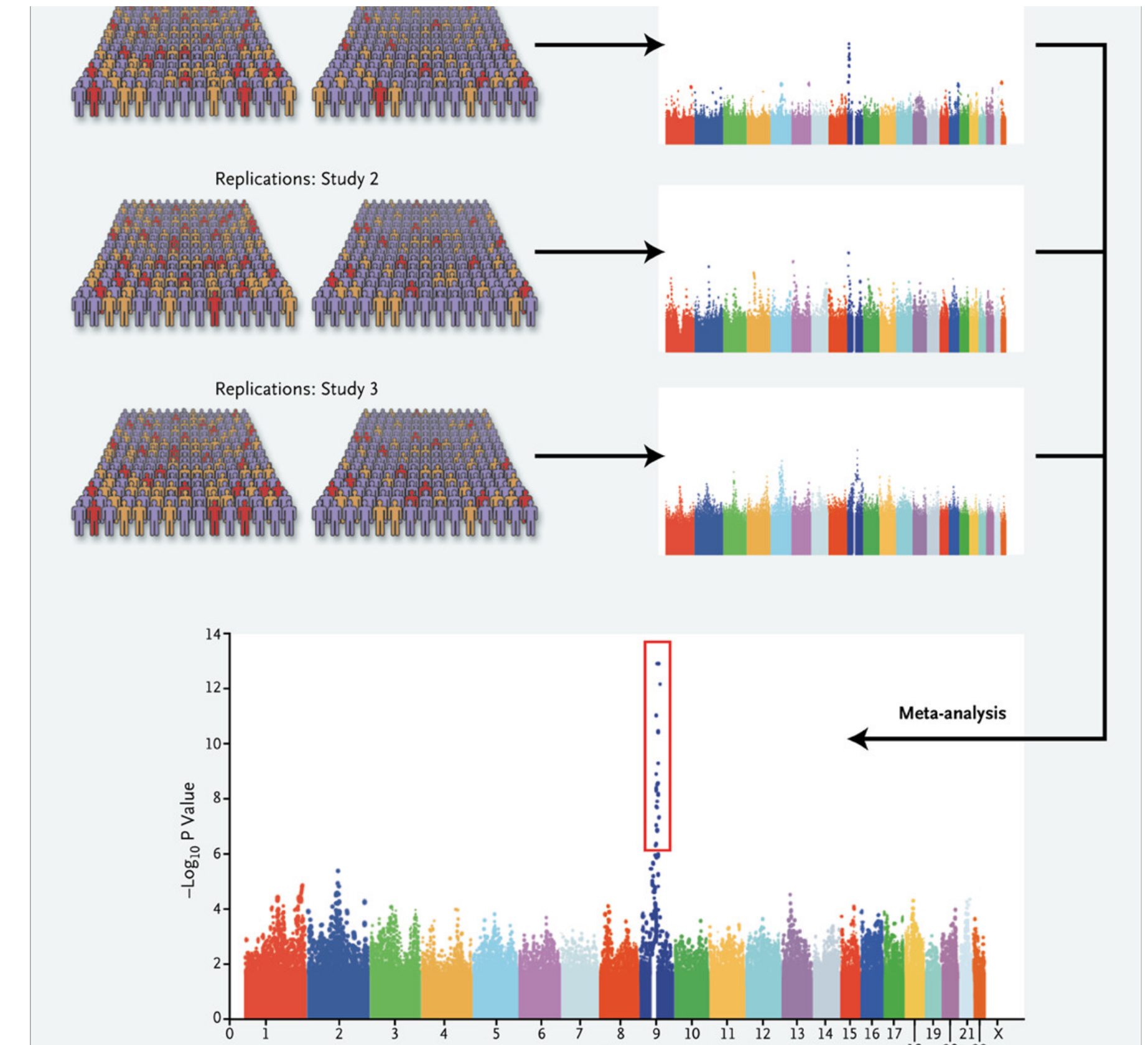
× Single estimates (Better)  
× Center of 8 estimates (**Meta Estimate**)  
• True center

“Meta-analysis is the art of **harmonising multiple results** in order to arrive **more accurate** conclusions”. (MIO)

# Why Meta-Analyse in GWAS?

RECALL: (GWAS Challenges)

- ▶ Multiple Testing Burden
- ▶ GWAS requires large-scale datasets
- ▶ GWAS software capable handling large-scale datasets



[1] Figures From:

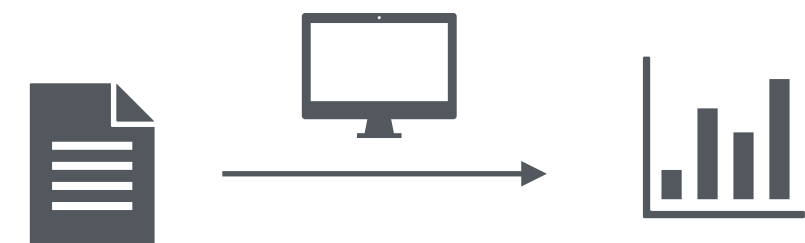
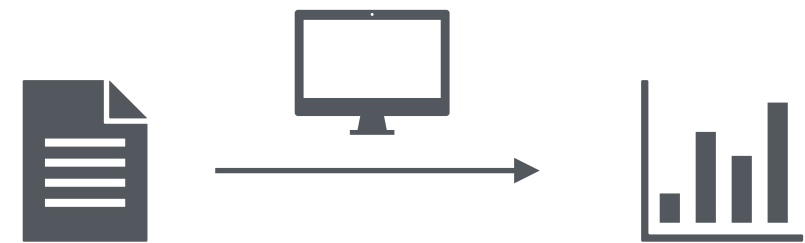
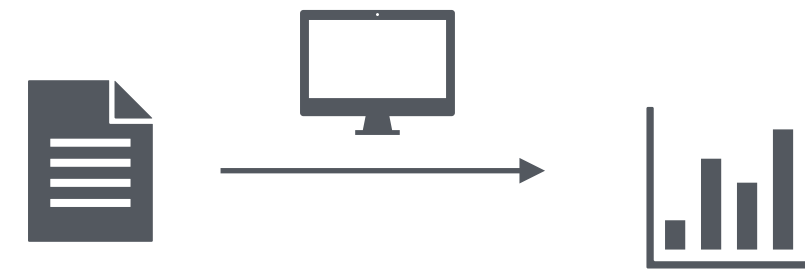
Manolio, Teri A.

"Genomewide association studies and assessment of the risk of disease."

*New England journal of medicine* 363.2 (2010): 166-176.



# Types of Meta-Analysis in GWAS



## Individual Participant Data Meta-Analysis



## Summary Data Meta-Analysis



# Types of Meta-Analysis in GWAS

## PART 1: pvalue based Methods

Fisher's method:

$$\chi^2 = 2 \sum_{i=1}^K \log(pval_i)$$

Under the null (i.e. homogeneity):  $\chi^2 \sim \chi^2_{2K}$

Z-scores method:

$$z = 2 \frac{\sum_{i=1}^K z_i \sqrt{N_i}}{\sqrt{\sum_{i=1}^K N_i}}$$

$N_i$  is sample size,

$z_i = \Phi^{-1}(1 - p_i/2) * (\text{direction of study } i)$

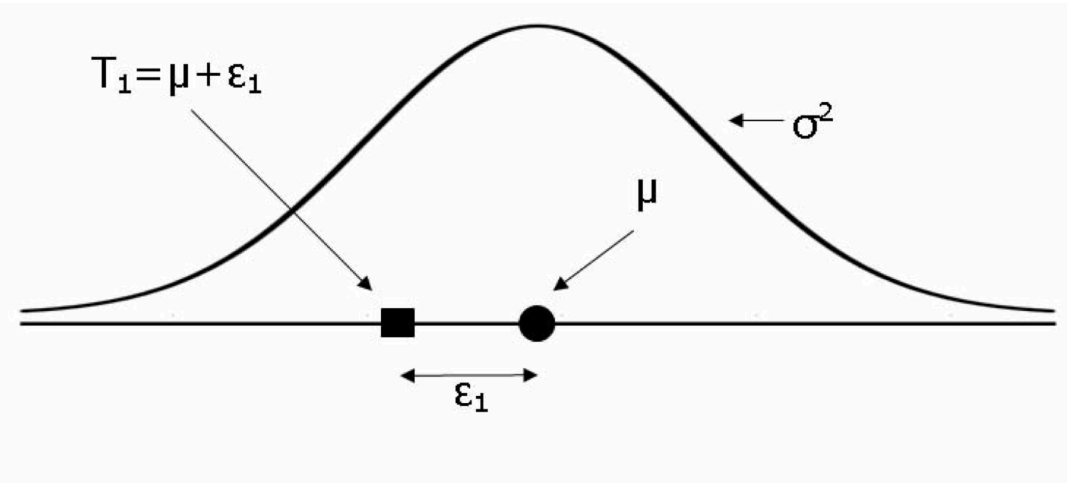
### Disadvantages:

- A. No information about the size of the effects,
- B. Assumes homogeneity,
- C. No direction of effect (for Fisher's method)

# Types of Meta-Analysis in GWAS

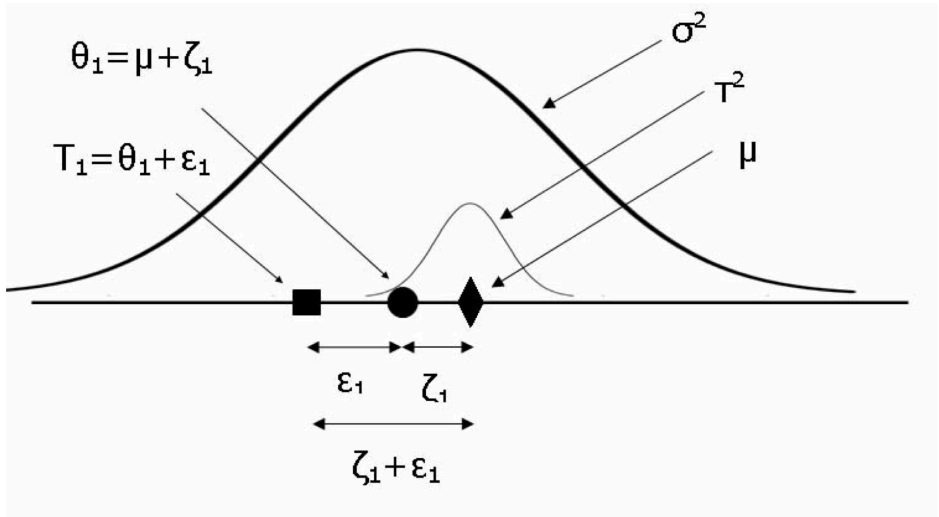
## PART 2: Effect size based Methods

### Fixed Effect Model [1]



Fixed effect model. The observed effects are sampled from a distribution with true effect  $\mu$ , and variance  $\sigma^2$ . The observed effect  $T_1$  is equal to  $\mu + \varepsilon_1$ .

### Random Effects Model [1]



Random effects model. The observed effect  $T_1$  (box) is sampled from a distribution with true effect  $\theta_1$ , and variance  $\sigma^2$ . This true effect  $\theta_1$ , in turn, is sampled from a distribution with mean  $\mu$  and variance  $\tau^2$ .

[1] Figures are from:

[https://meta-analysis.com/download/Meta-analysis\\_fixed\\_effect\\_vs\\_random\\_effects\\_072607.pdf](https://meta-analysis.com/download/Meta-analysis_fixed_effect_vs_random_effects_072607.pdf)

Borenstein, Michael, Larry Hedges, and Hannah Rothstein.  
"Meta-analysis: Fixed effect vs. random effects."  
*Meta-analysis.com* (2007): 1-162.

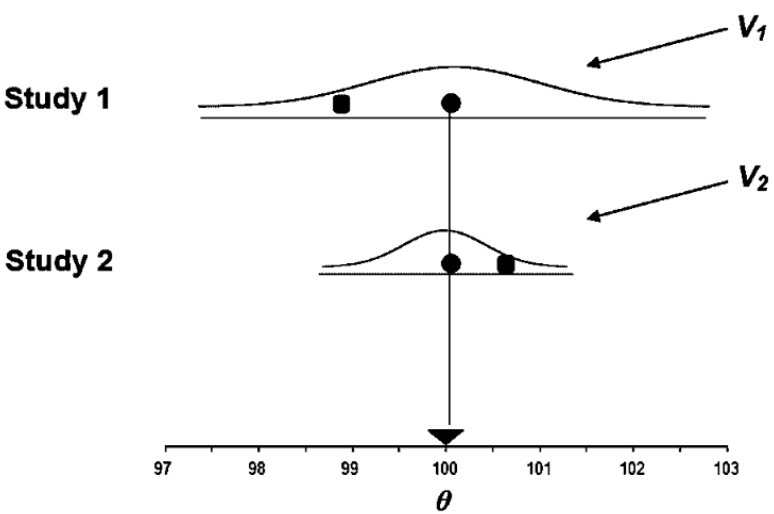


Figure 3. Schematic of the fixed-effect model.

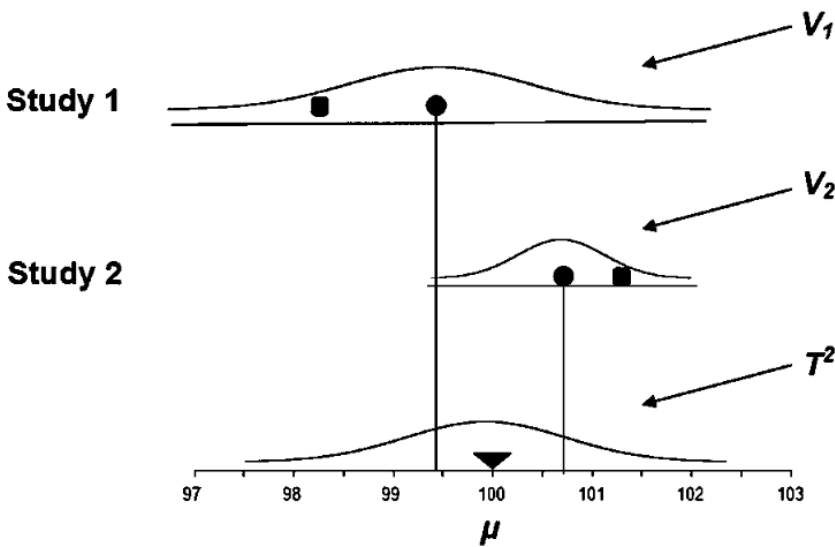


Figure 4. Schematic of the random-effects model.

Figure 3 and Figure 4 are from: Borenstein, Michael, et al. "A basic introduction to fixed-effect and random-effects models for meta-analysis." (2010).



# Types of Meta-Analysis in GWAS

## PART 2: Effect size based Methods

Fixed Effect (Inverse Variance Method - IVM [2]):

$$\widehat{\beta}_{meta} = \sum_{i=1}^K \omega_i \hat{\beta}_i$$

$$\text{where } \omega_i = \frac{V_i^{-1}}{\sum_{i=1}^K V_i^{-1}},$$

$V_i$  is the variance for study  $i$ .

Random Effects (DerSimonian and Laird Estimator [3]):

$$\widehat{\beta}_{meta} = \sum_{i=1}^K \omega_i \hat{\beta}_i$$

$$\text{where } \omega_i = \frac{(V_i + T^2)^{-1}}{\sum_{i=1}^K (V_i + T^2)^{-1}},$$

$V_i$  is the variance for study  $i$ ,

$T^2$  is the between-study variance.

[2] Borenstein, Michael, et al. "A basic introduction to fixed-effect and random-effects models for meta-analysis." (2010).

[3] DerSimonian, Rebecca, and Nan Laird. "Meta-analysis in clinical trials." (1986).

# Types of Meta-Analysis in GWAS

## PART 2: Effect size based Methods (When to use Random Effects Model?)

### 1. Cochran's Q [4]

$$Q = \sum_{i=1}^K \omega_i (\hat{\beta}_i - \hat{\beta}_{IVM})^2, \text{ under the null } Q \sim \chi_{K-1}^2.$$

### 2. $I^2$ [4]

$$I^2 = 100 * \left( 1 - \frac{K-1}{Q} \right)$$

### 3. $T^2$ [4]

$$T^2 = \frac{Q - K + 1}{A - \frac{B}{A}} \text{ where } A = \sum_{i=1}^K \omega_i \text{ and } B = \sum_{i=1}^K \omega_i^2$$

**"The strategy of starting with a fixed-effect model and then moving to a random-effects model if the test for heterogeneity is significant relies on a flawed logic and should be strongly discouraged."** [5]

"Fixed-effects calculations avoided power deserts and maximized discovery of association signals at the expense of much higher false-positive rates. Therefore, random- and fixed-effects models are preferable for different purposes (**fixed effects for initial screenings, random effect is for generalizability applications**)."

 [6]

**"Random effects models are not used in discovery efforts** owing to far more limited power than fixed effects models; however, random effects models are more appropriate than fixed effects models when the aim is to consider the generalizability of the observed association and estimate the average effect size of the associated variant and its uncertainty across different populations: for example, for predictive purposes<sup>28,33</sup>."

 [7]

"The conventional wisdom in the statistical literature is that **when heterogeneity is present** or even likely, **the random effects model is more appropriate than the fixed effects model. We suggest that this might not be the right approach for GWAS**, because (i) the number of studies being combined is often not very large (leading to an imprecise heterogeneity estimate) and (ii) the form of the heterogeneity typically does not fit a Gaussian random effects model."

 [8]

[4] Borenstein, Michael, et al. *Introduction to meta-analysis*. John Wiley & Sons, 2009.

[5] Borenstein, Michael, et al. "A basic introduction to fixed-effect and random-effects models for meta-analysis." (2010).

[6] Pereira, Tiago V., et al. "Discovery properties of genome-wide association signals from cumulatively combined data sets." (2009).

[7] Evangelou, Evangelos, and John PA Ioannidis. "Meta-analysis methods for genome-wide association studies and beyond." (2013).

[8] Begum, Ferdouse, et al. "Comprehensive literature review and statistical considerations for GWAS meta-analysis." (2012)



# Types of Meta-Analysis in GWAS

## PART 2: Effect size based Methods (When to use Random Effects Model?)

“A popular way out of this dilemma is **to start with a fixed effects meta-analysis but to report the random effects meta-analysis** when heterogeneity is found in a top hit.” [9]

**Example Study:** GWAS on Height [10]

A. 5.4 million individuals, 5 ancestry groups.

B. Meta-Analysis of Meta-analyses.

C. IVM applied within each ancestry groups.

Then IVM applied between ancestry groups.

**Heterogeneity in meta analysis of GWAS:**

“Using data from a meta-analysis of seven GWA studies on obesity, we concluded that moderate or larger heterogeneity was common in meta-analysis of GWA studies.” [11]

[9] Thompson, John R., John Attia, and Cosetta Minelli. “The meta-analysis of genome-wide association studies.” (2011).

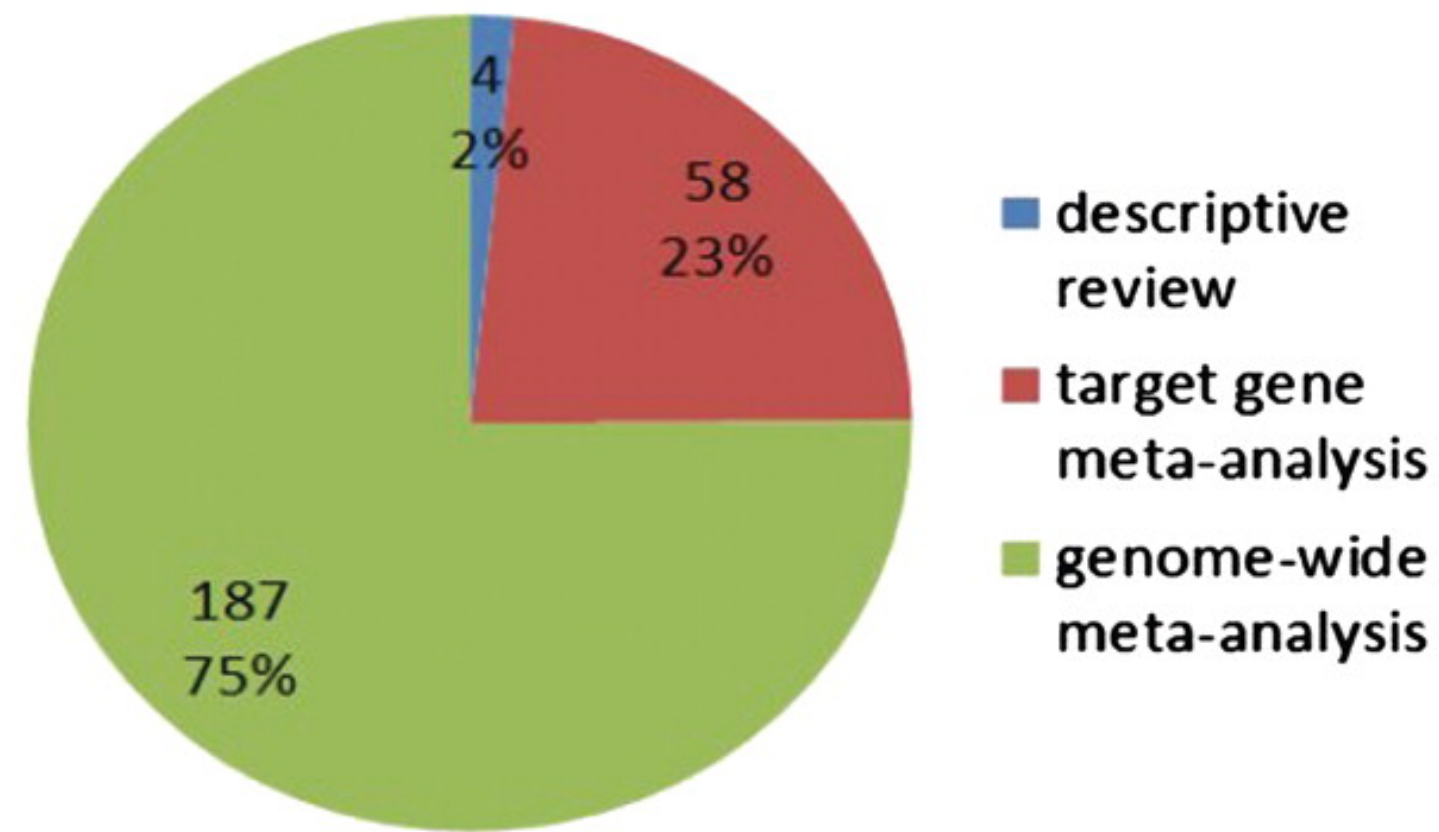
[10] Yengo, L., Vedantam, S., Marouli, E. et al. A saturated map of common genetic variants associated with human height. *Nature* 610, 704–712 (2022).

[11] Pei, Yu-Fang, et al. “Exploring the major sources and extent of heterogeneity in a genome-wide association meta-analysis.” (2016).

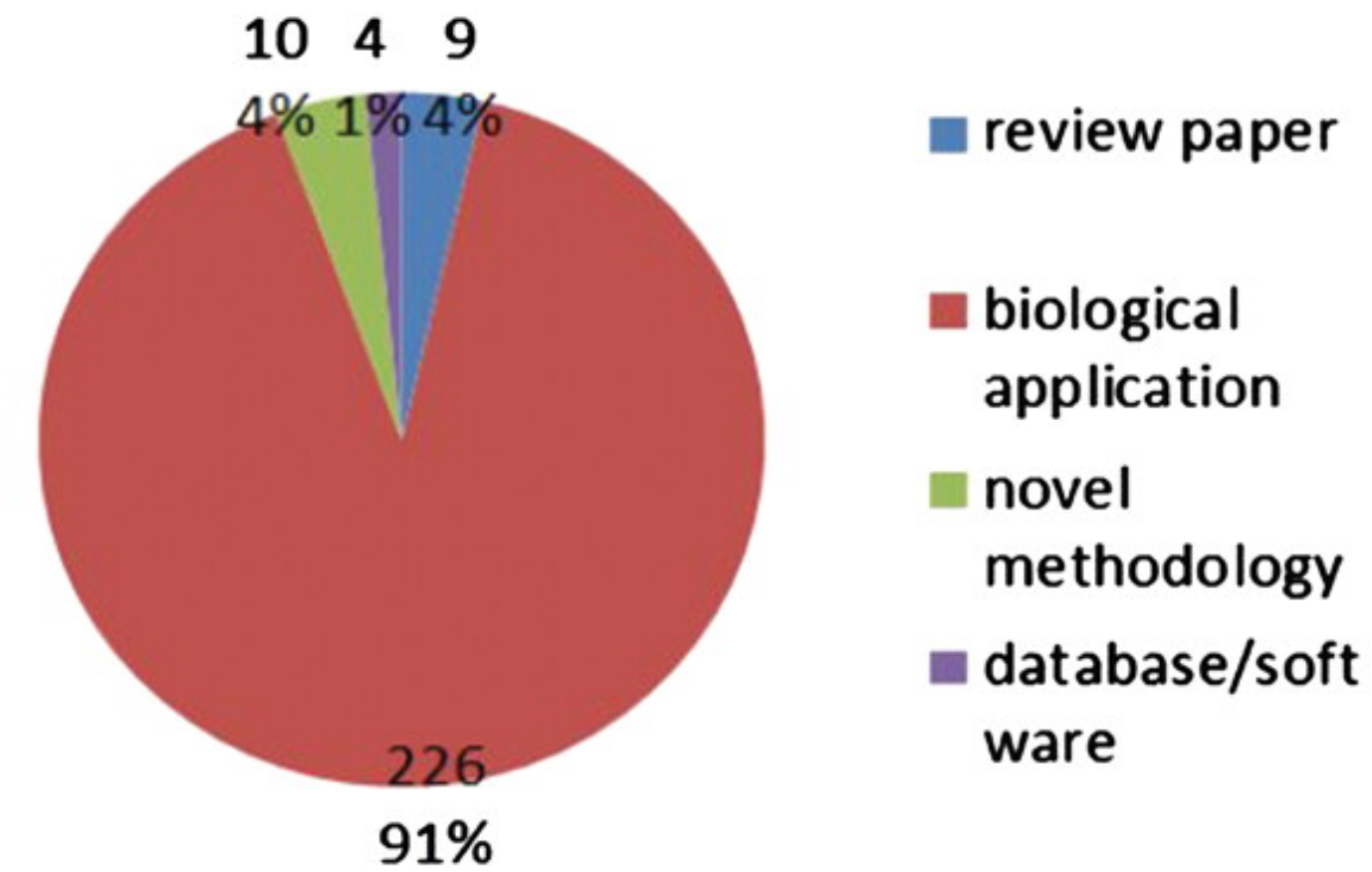


# Types of Meta-Analysis in GWAS

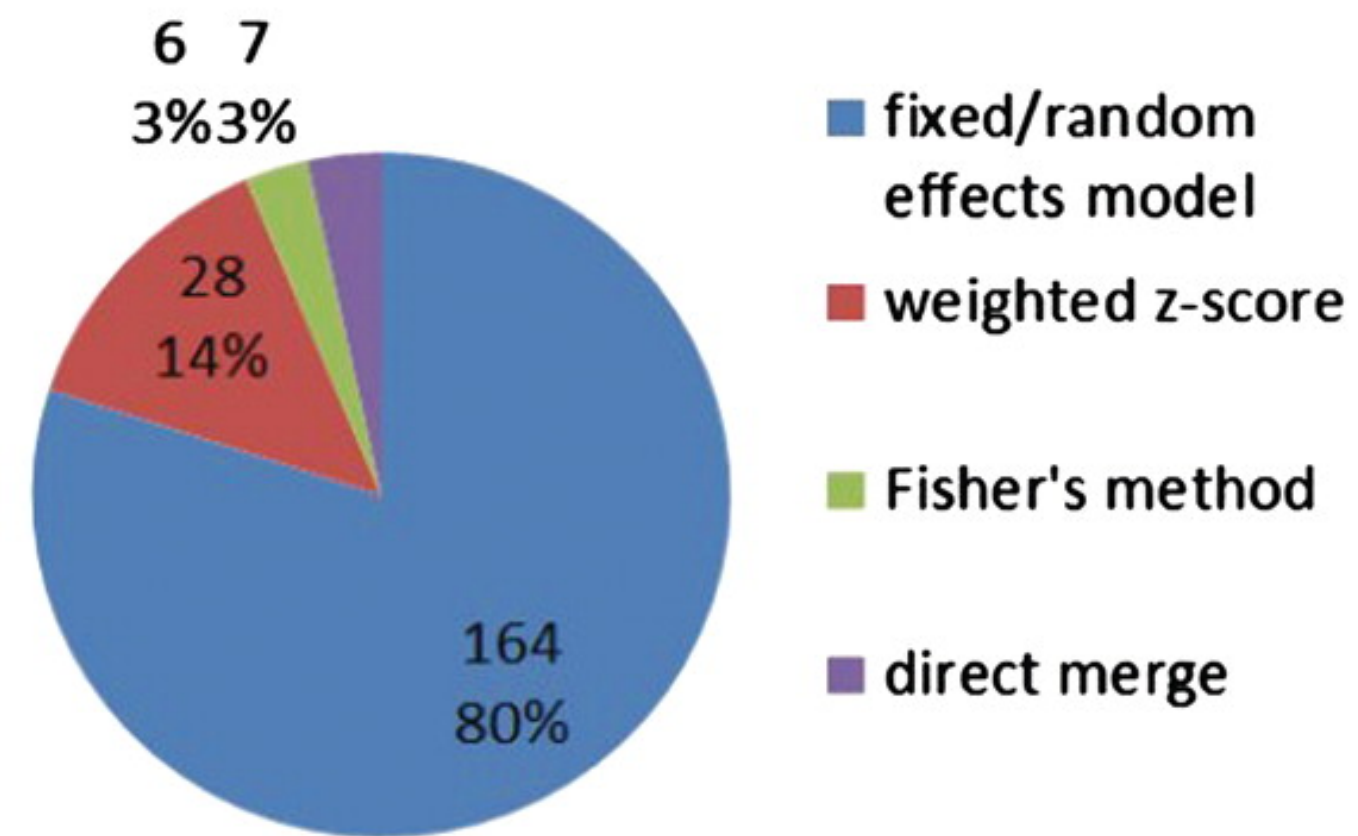
**A Type of meta-analysis**



**B Type of paper**



**C Type of meta-analysis methods used**



**D Software used**

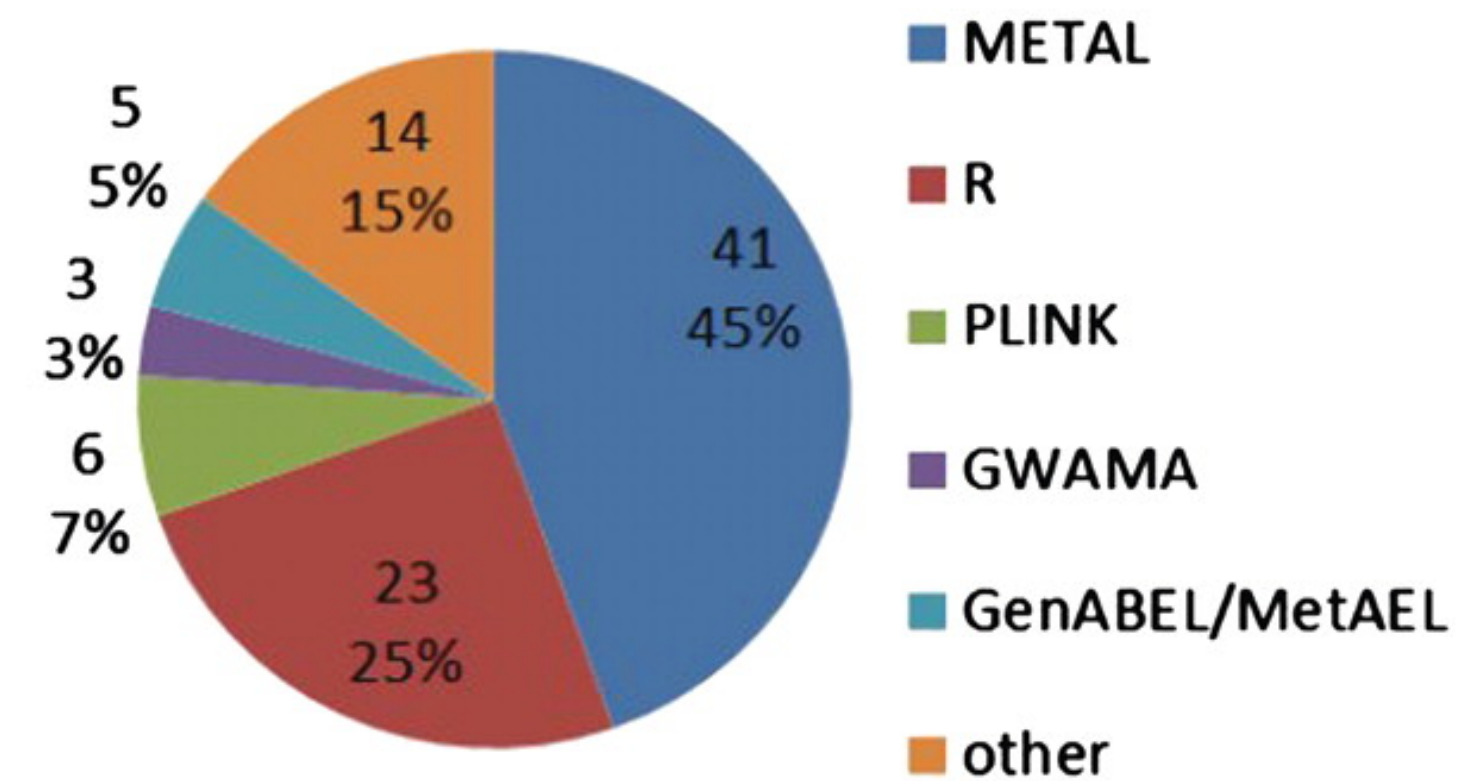


Figure from: Begum, Ferdouse, et al. "Comprehensive literature review and statistical considerations for GWAS meta-analysis." *Nucleic acids research* 40.9 (2012): 3777-3784.

# What's Next

1. Understanding Polygenic Risk Score (PRS)?
2. Calculating PRS
3. Population Bias