



"Okay yeah, that explains a lot."



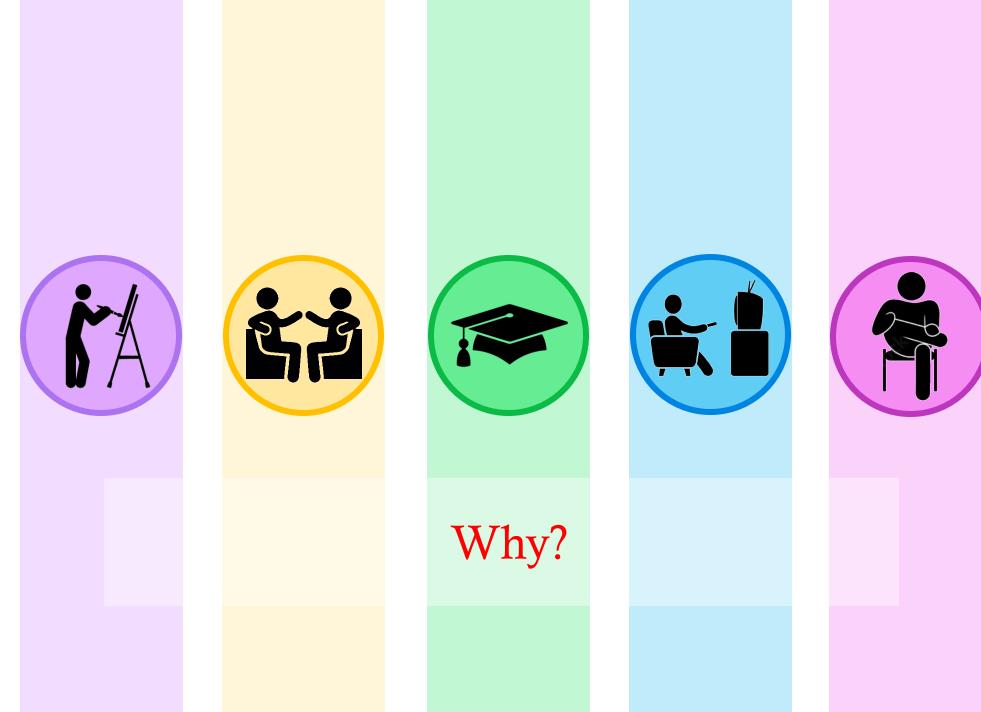








There are so many ways that parents and offspring resemble one another.

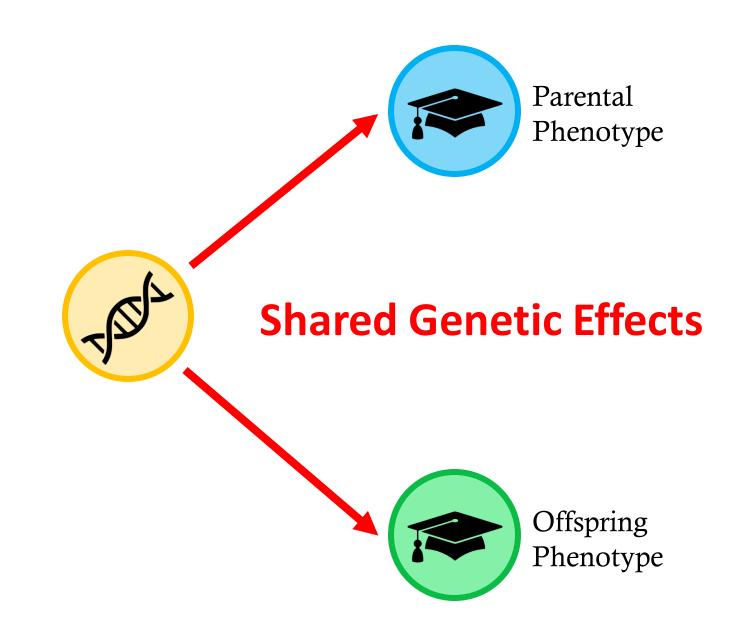


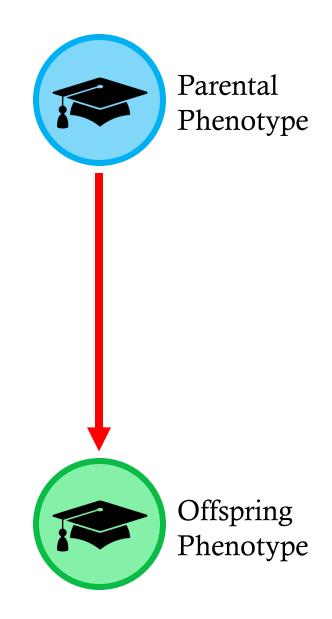


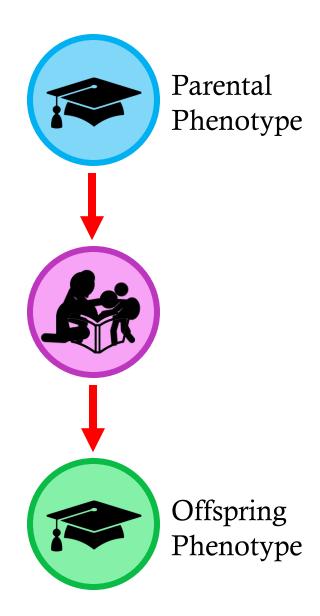


Education



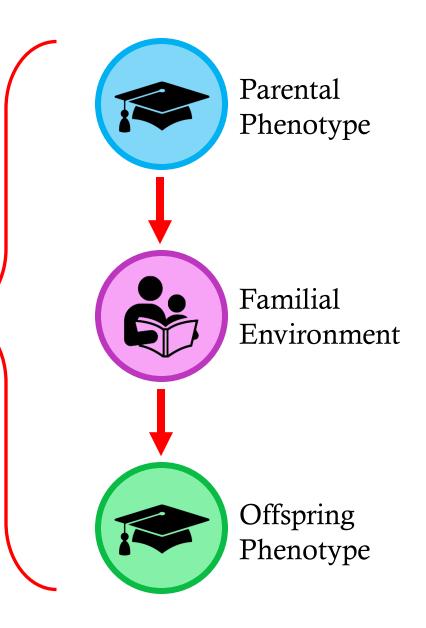


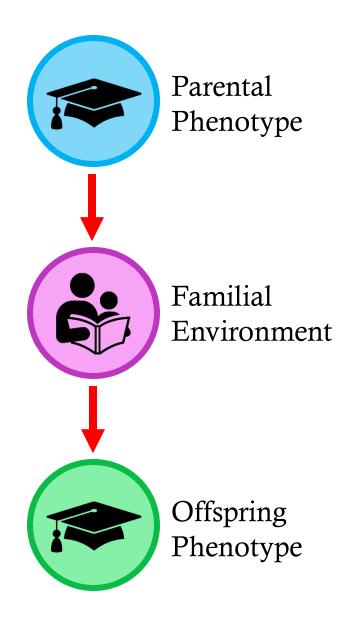


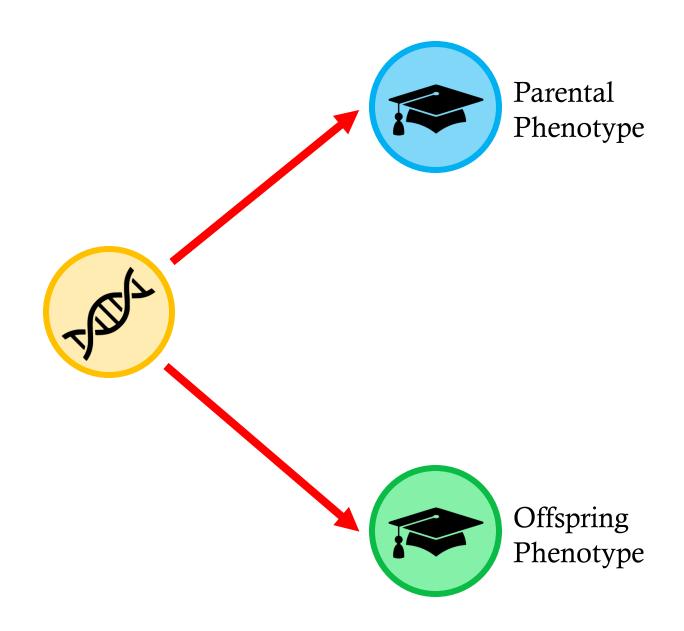


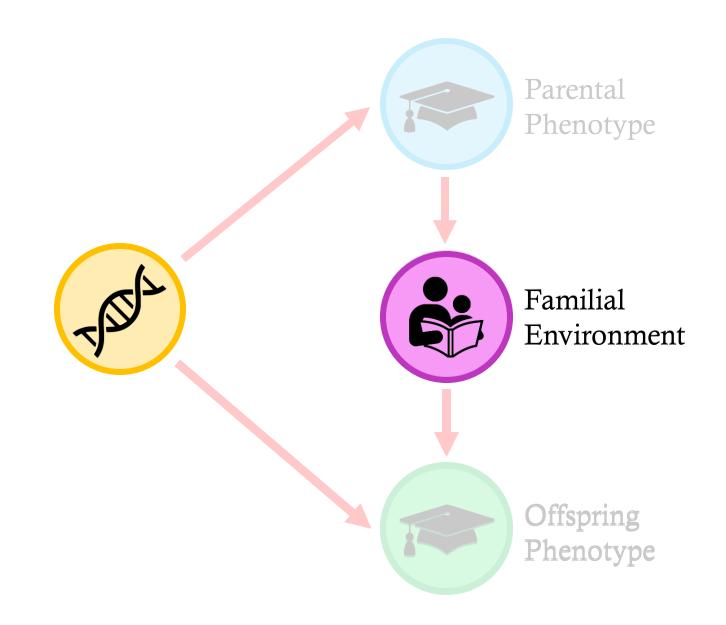
Vertical Transmission

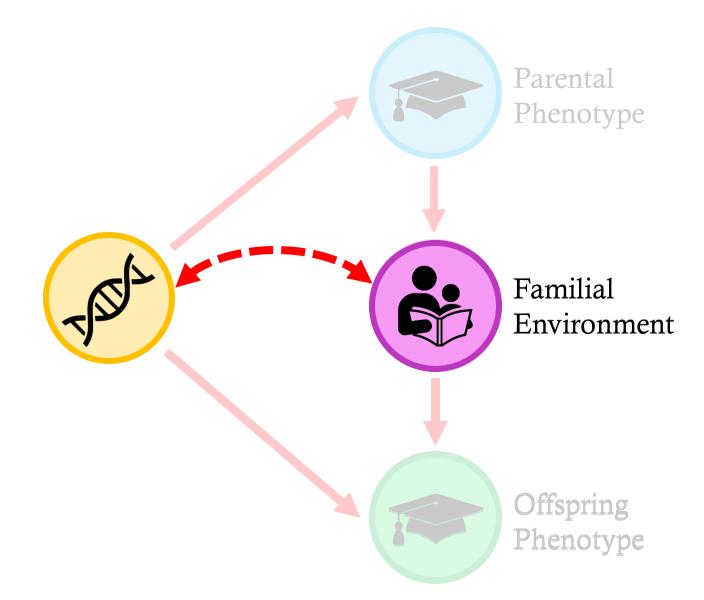
(aka "Cultural Transmission")





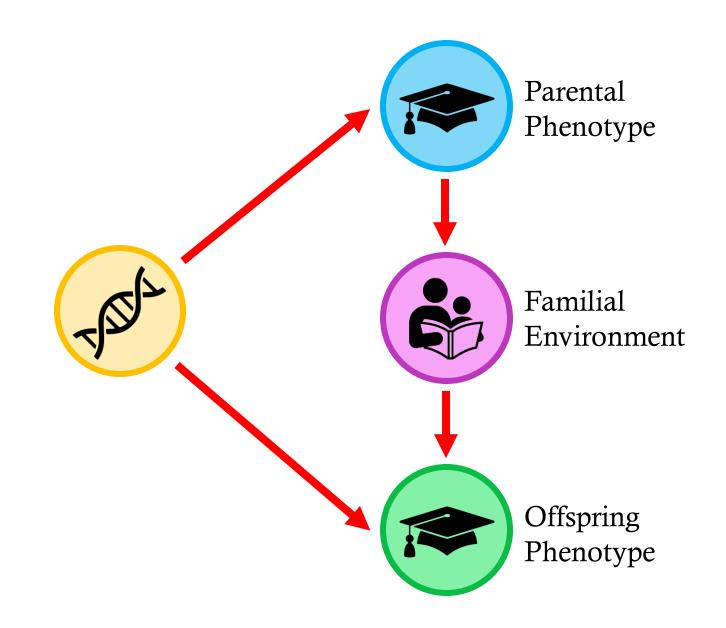


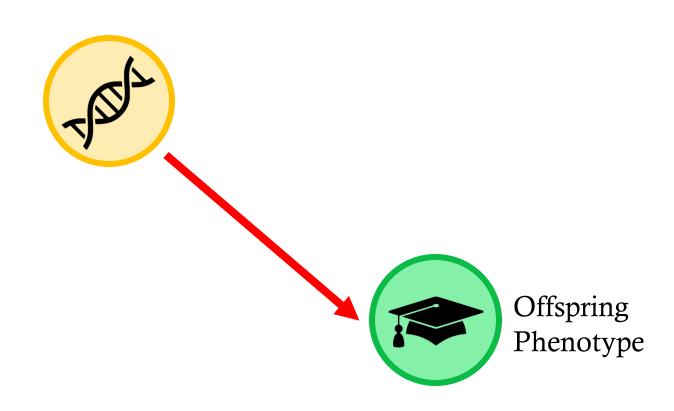


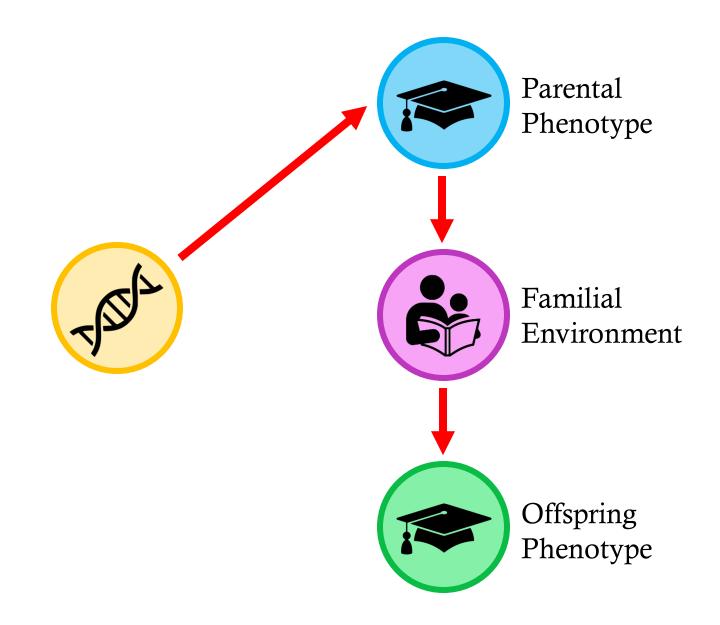


Genetic Nurture

aka "Passive Gene-Environment Covariance"

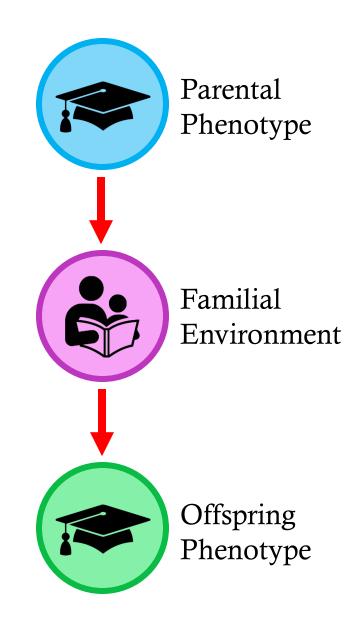


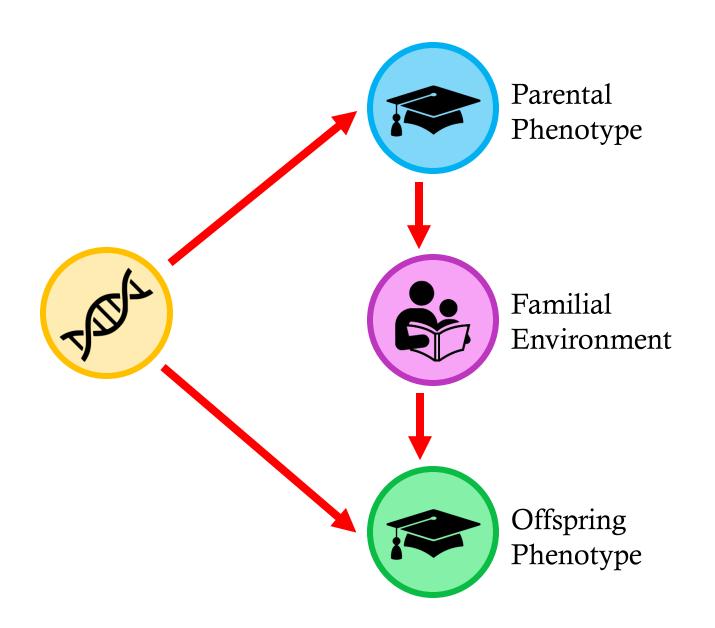




If unaccounted for, vertical transmission:

- Will upwardly bias estimates of SNP effects from GWAS
- Will upwardly bias estimates of shared environmental effects (C) in twin ACE models
- Will upwardly bias estimates of additive genetic effects (A) in twin AE and ADE models







International Journal of Epidemiology, 2018, 1229–1241 doi: 10.1093/ije/dyy015 Advance Access Publication Date: 13 February 2018 Original article **SEA**

 $International\ Journal\ of\ Epidemiology,\ 2019,\ 861-875$

loi: 10.1093/ije/dyz019

Advance Access Publication Date: 27 February 2019
Original article

Methods

Using structural equation modelling to jointly estimate maternal and fetal effects on birthweight in the UK Biobank

Nicole M Warrington, 1 Rachel M Freathy, 2,3 Michael C Neale and David M Evans 1,3,5 *



RESEARCH ARTICLE

Assessing the Causal Relationship of Maternal Height on Birth Size and Gestational Age at Birth: A Mendelian Randomization Analysis

Ge Zhang^{1,2}*, Jonas Bacelis³, Candice Lengyel², Kari Teramo⁴, Mikko Hallman⁵, Øyvind Helgeland⁶, Stefan Johansson^{6,7}, Ronny Myhre⁸, Verena Sengpiel³, Pål Rasmus Njølstad^{6,9}, Bo Jacobsson^{8,10}, Louis Muglia²*

Mendelian Randomization

Elucidating the role of maternal environmental exposures on offspring health and disease using two-sample Mendelian randomization

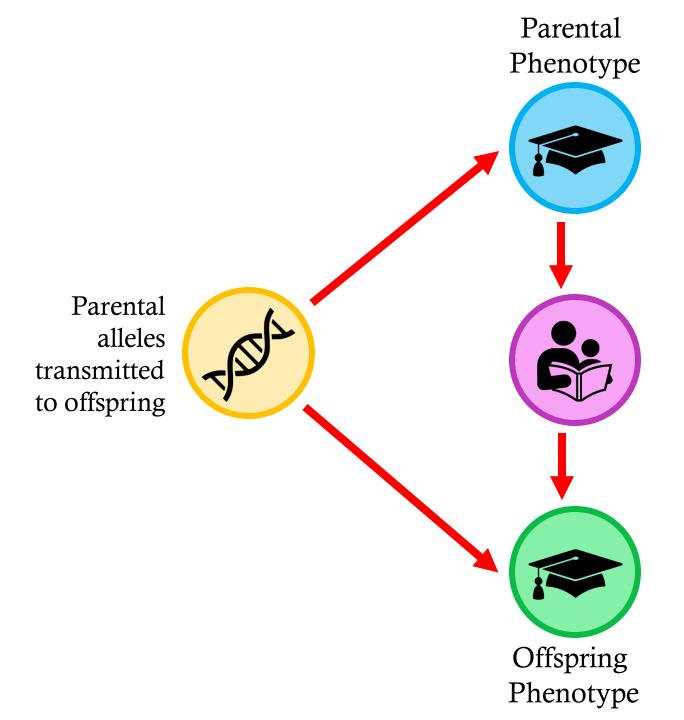
David M Evans , , , , , , , Gunn-Helen Moen, , Liang-Dar Hwang, Debbie A Lawlor^{2,3,6} and Nicole M Warrington ,

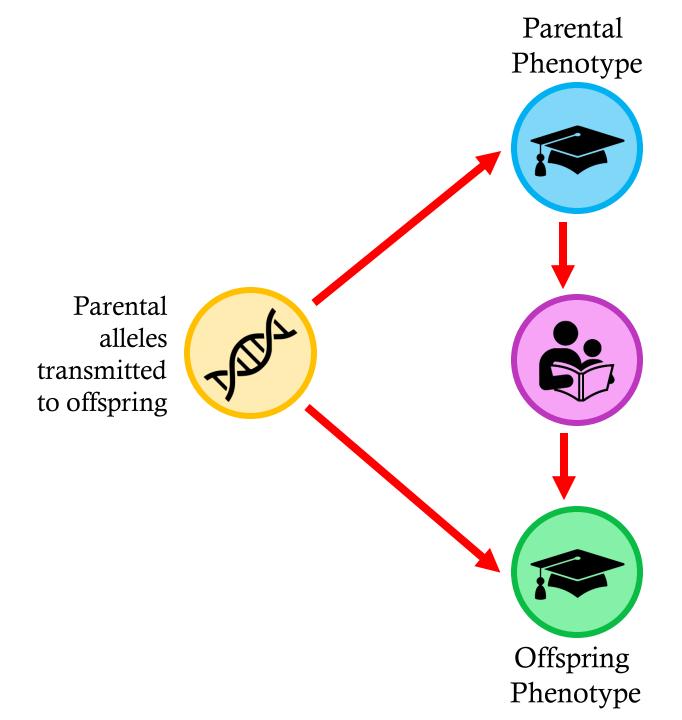
RESEARCH

HUMAN GENOMICS

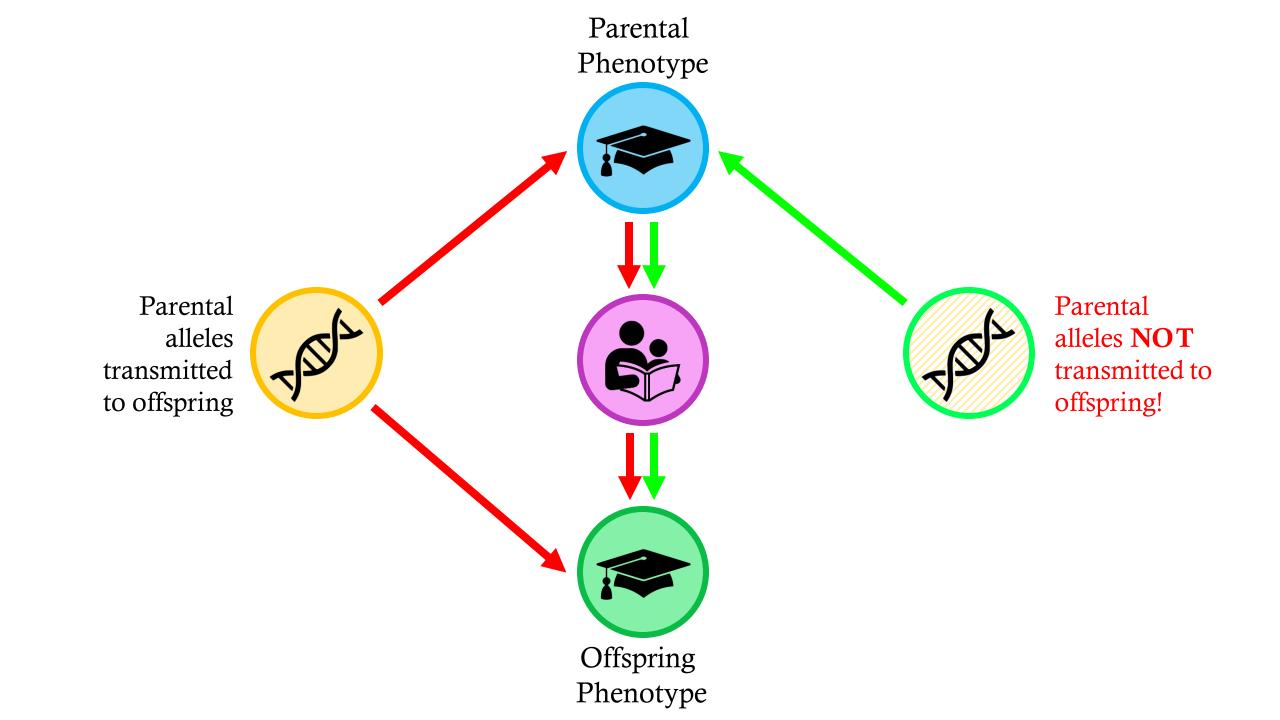
The nature of nurture: Effects of parental genotypes

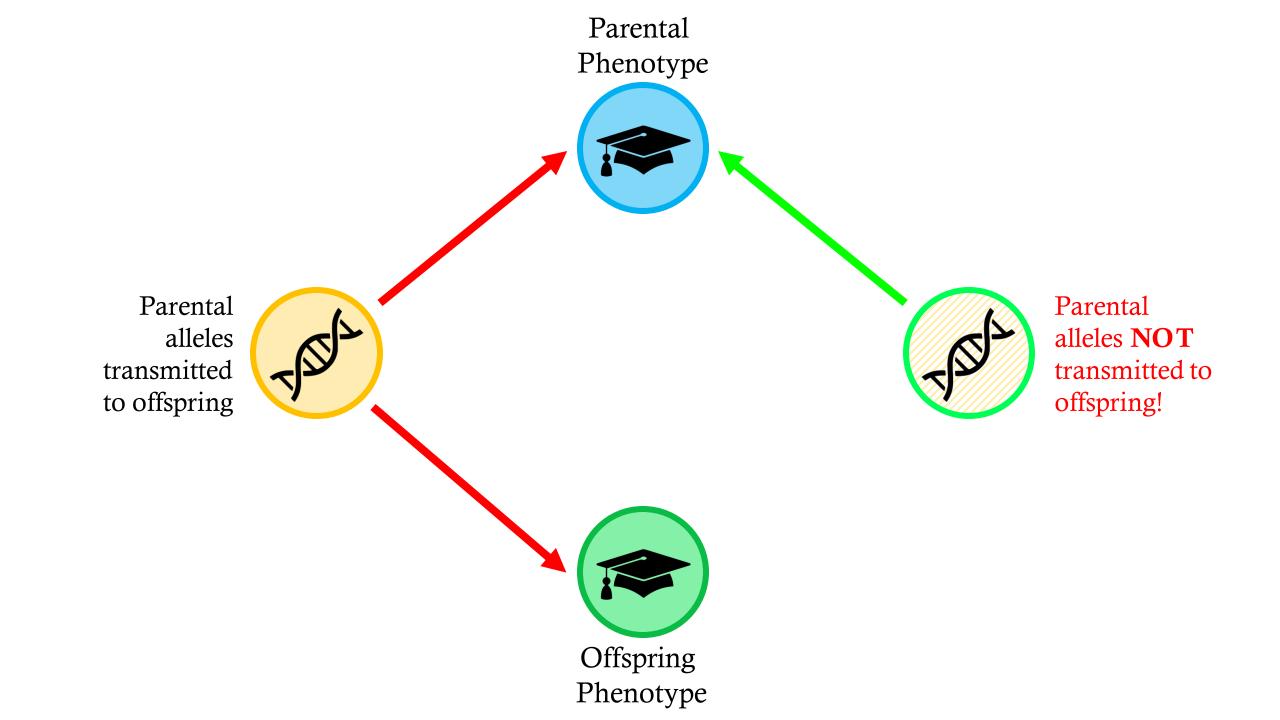
Augustine Kong,^{1,2,3*} Gudmar Thorleifsson,¹ Michael L. Frigge,¹ Bjarni J. Vilhjalmsson,^{4,5} Alexander I. Young,^{1,2,6} Thorgeir E. Thorgeirsson,¹ Stefania Benonisdottir,¹ Asmundur Oddsson,¹ Bjarni V. Halldorsson,¹ Gisli Masson,¹ Daniel F. Gudbjartsson,^{1,3} Agnar Helgason,^{1,7} Gyda Bjornsdottir,¹ Unnur Thorsteinsdottir,^{1,8} Kari Stefansson^{1,8*}

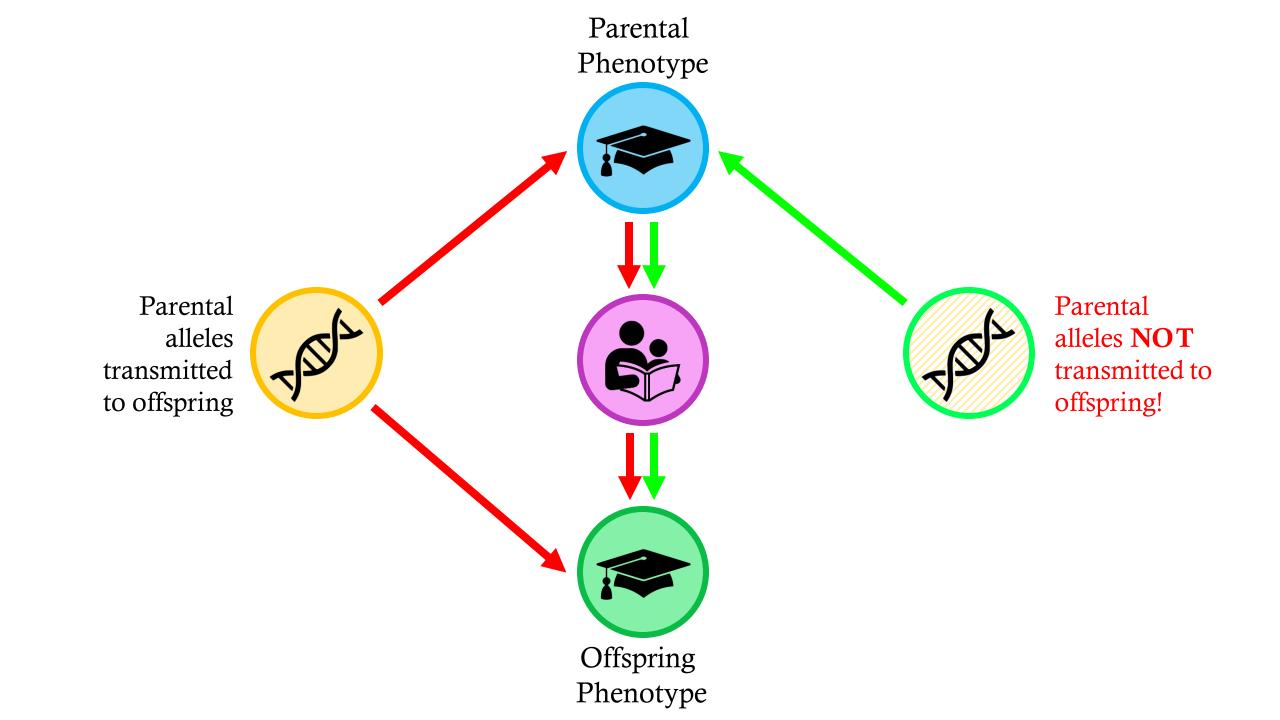


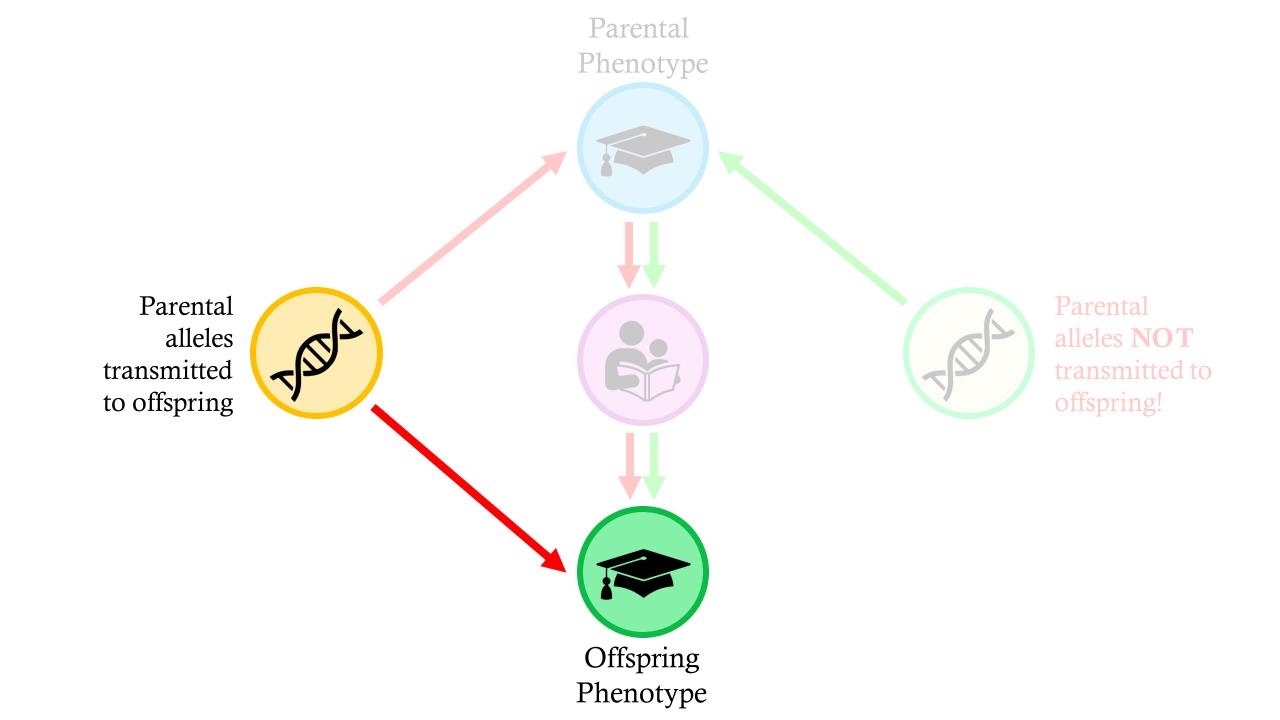


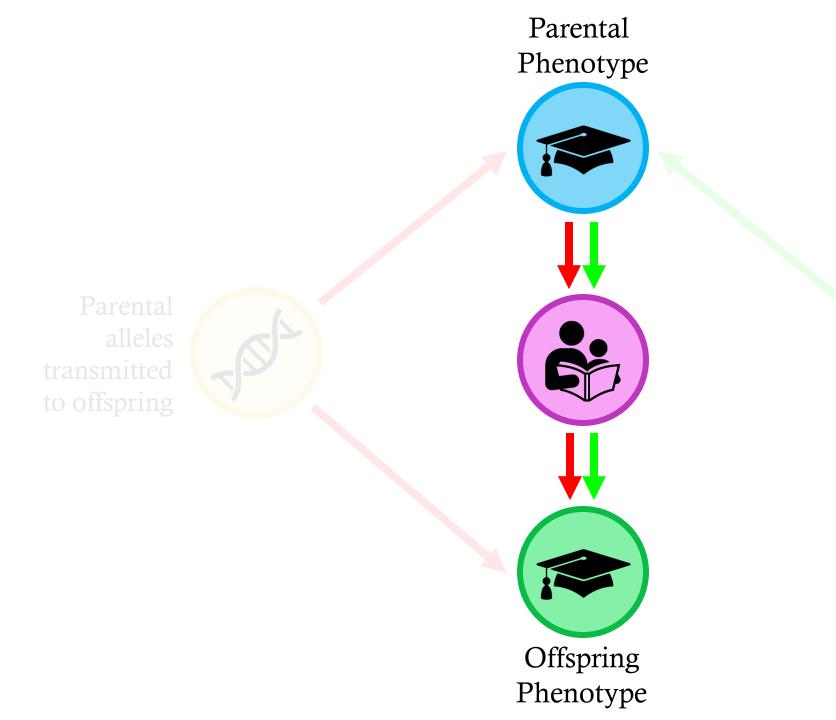




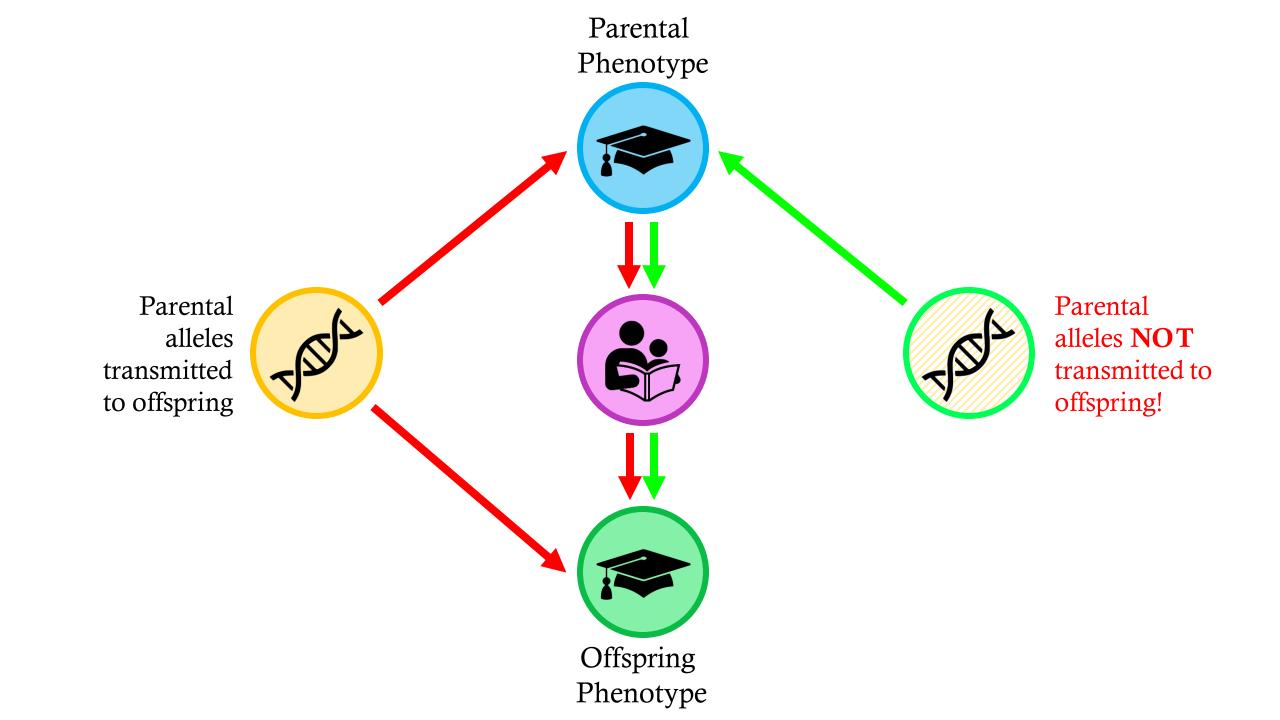












ORIGINAL RESEARCH

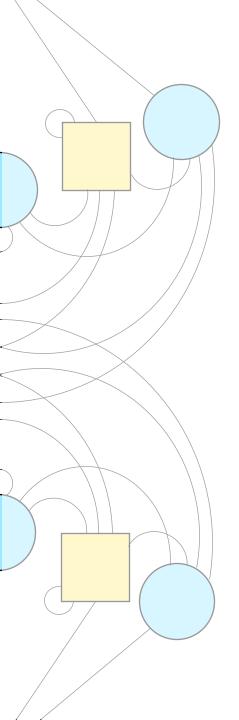


Estimation of Parental Effects Using Polygenic Scores

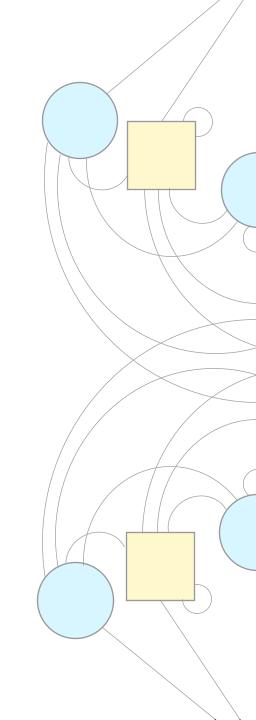
Jared V. Balbona^{1,2} · Yongkang Kim¹ · Matthew C. Keller^{1,2}

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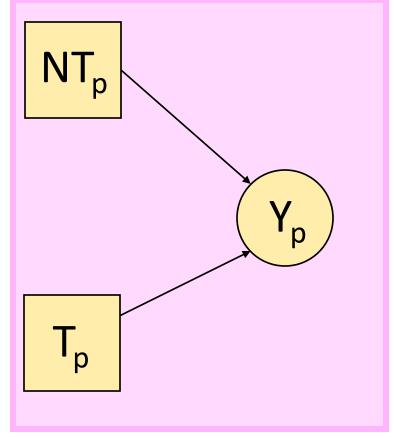


SEM-PGS

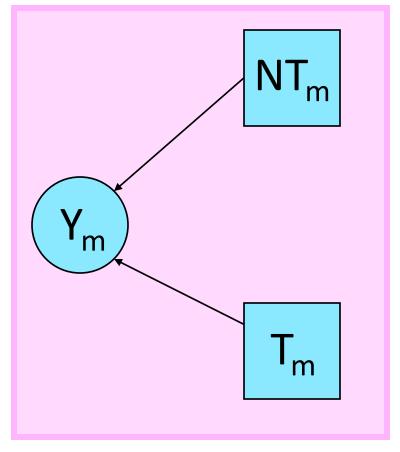


SEM-PGS

- 1. Unbiased estimates of h^2 , vertical transmission, and genetic nurture— Even when the PGS being used is not very predictive
- 2. Properly accounts for various types of assortative mating at both equilibrium and disequilibrium
- 3. In principle, is not biased by missing data-- Thus, it does not require complete trios
- 4. Is very easy to use, relying on relatively simple math
- 5. Allows for a vast number of extensions to best fit the trait and data being worked with



Paternal Genes and Phenotype

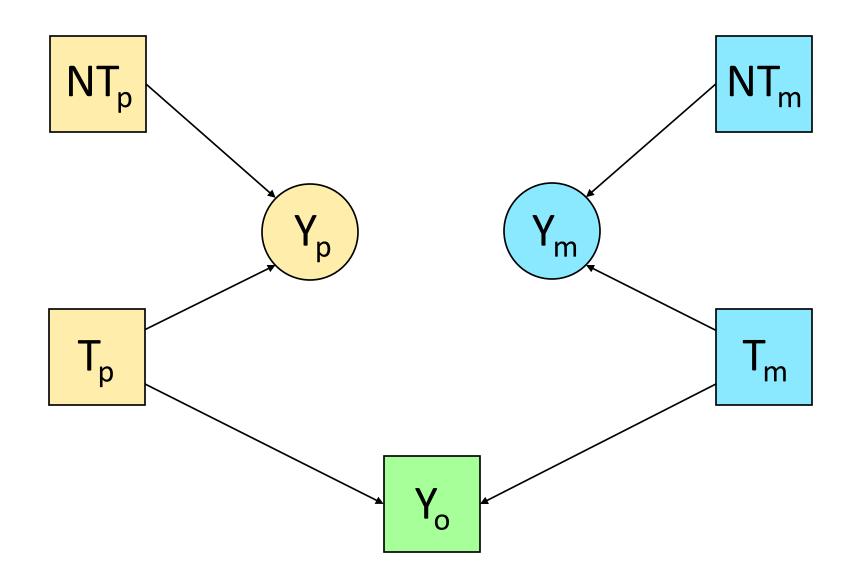


Maternal Genes and Phenotype

Y: Phenotypic Value

T: PGS made from the parent's *transmitted* alleles

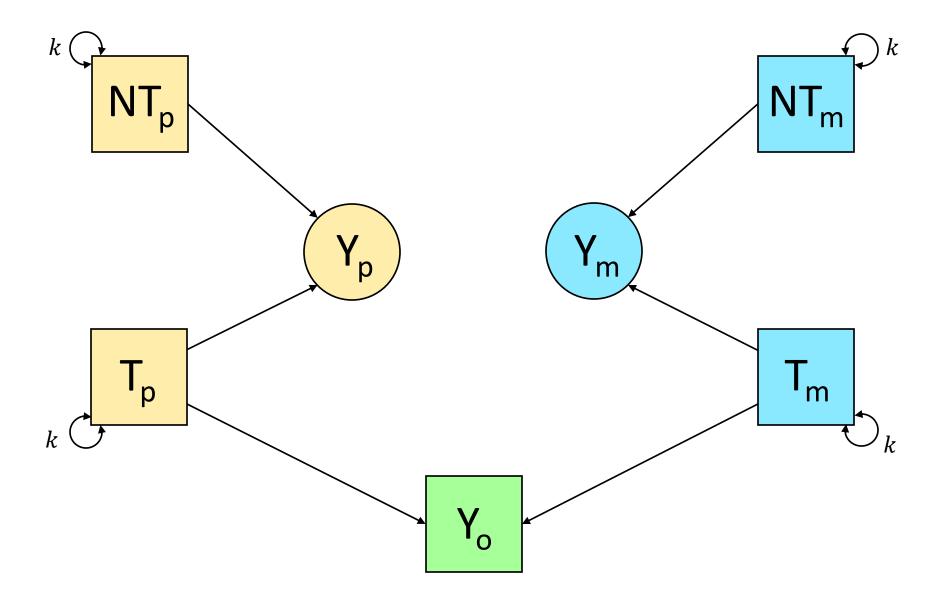
NT: PGS made from the parent's non-transmitted alleles



Y: Phenotypic Value

T: PGS made from the parent's transmitted alleles

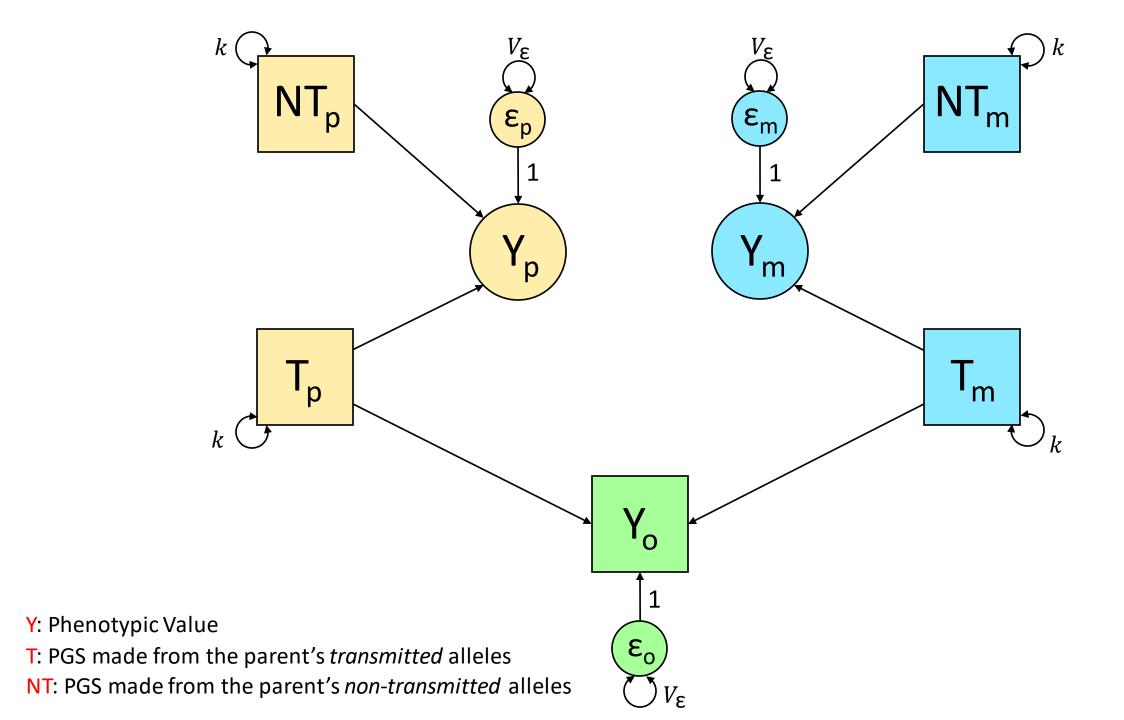
NT: PGS made from the parent's non-transmitted alleles

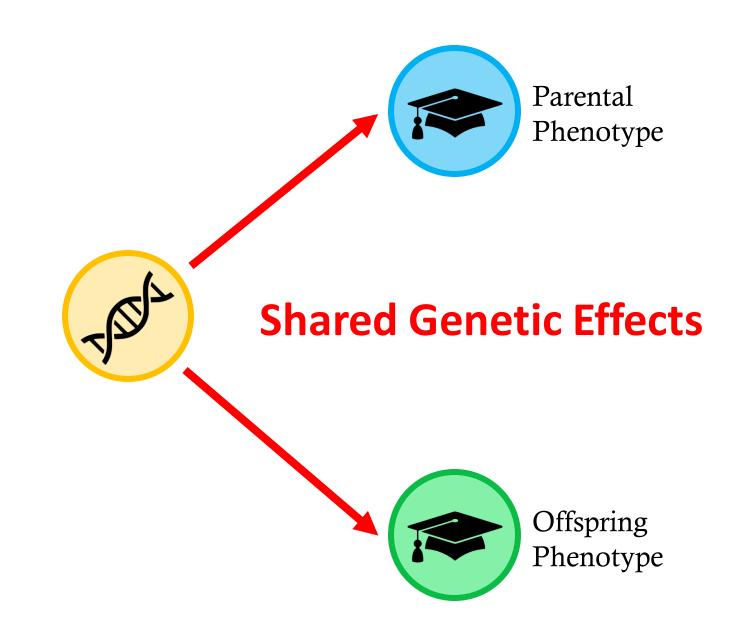


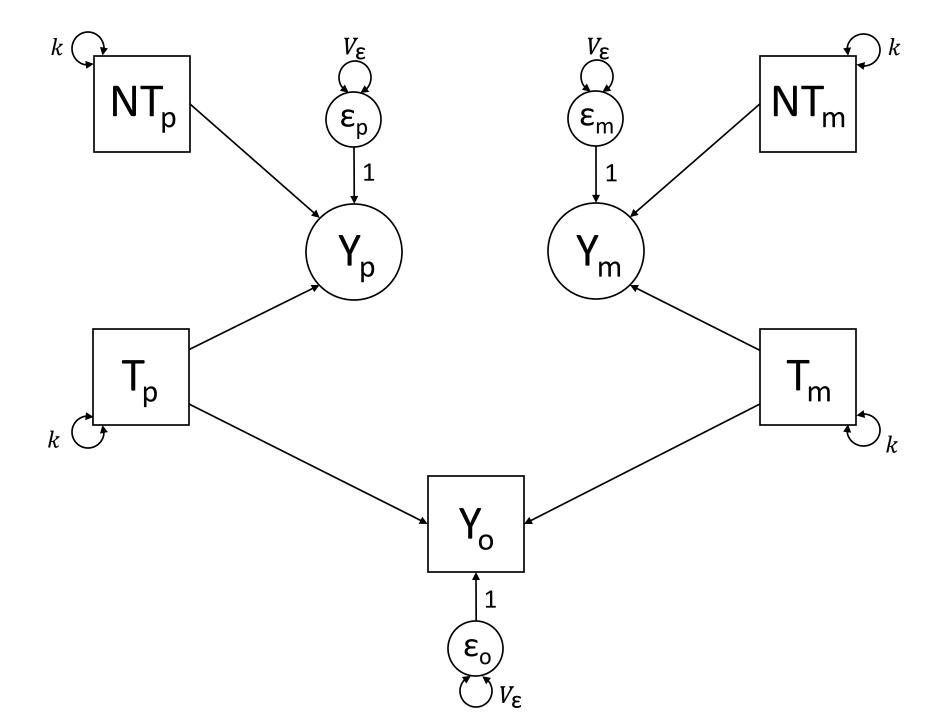
Y: Phenotypic Value

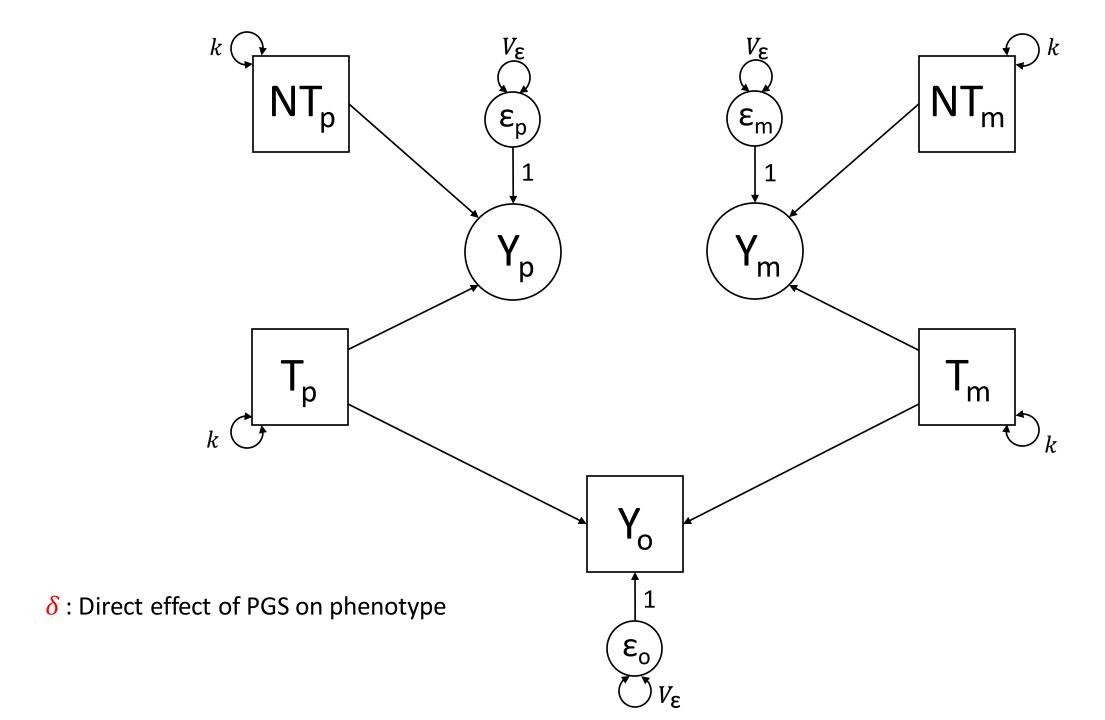
T: PGS made from the parent's transmitted alleles

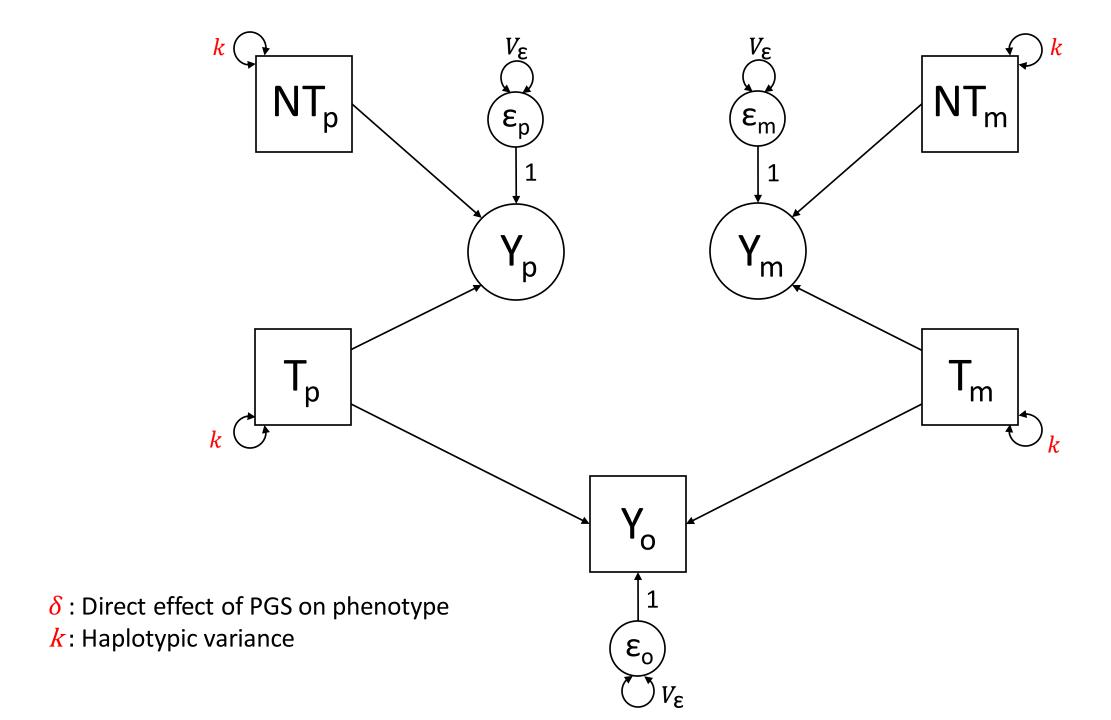
NT: PGS made from the parent's non-transmitted alleles

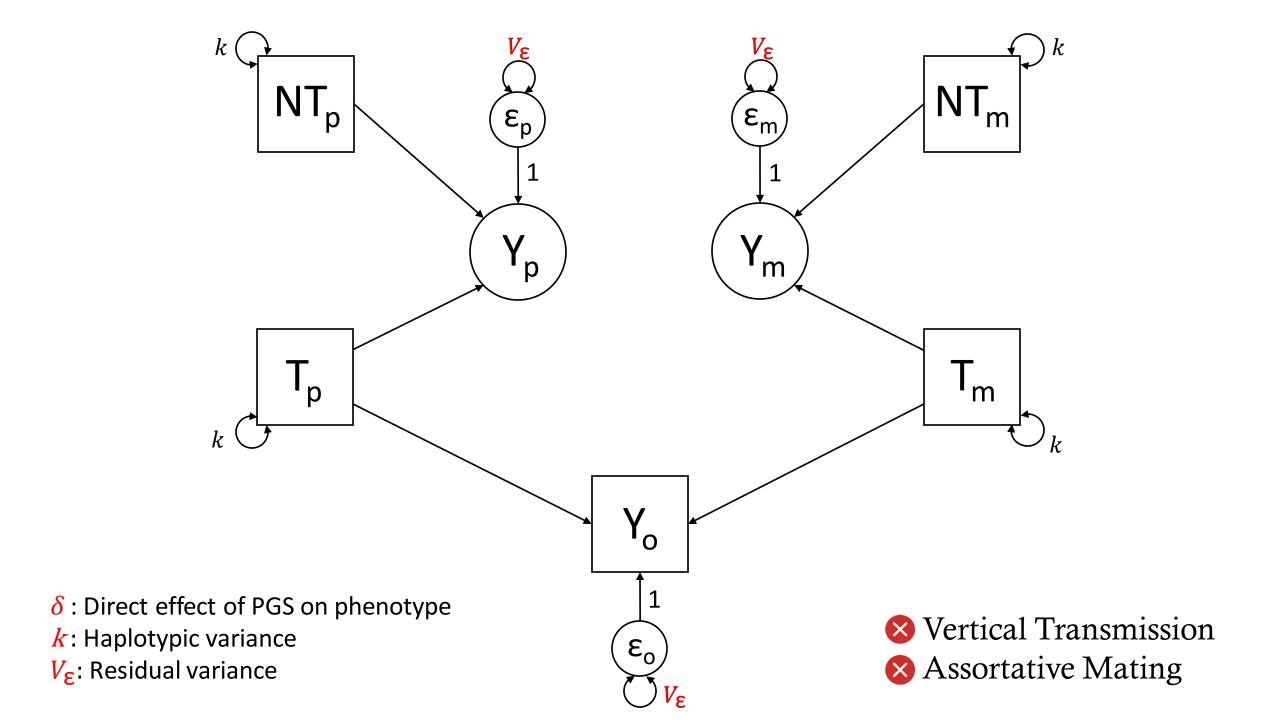


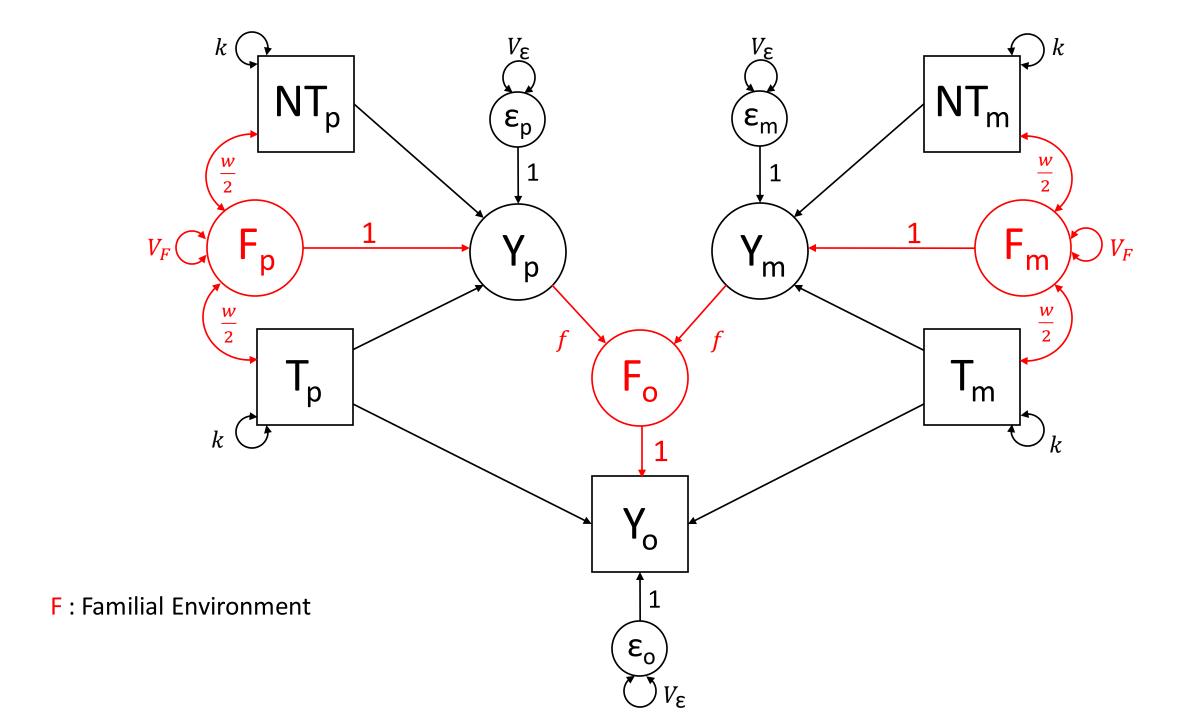


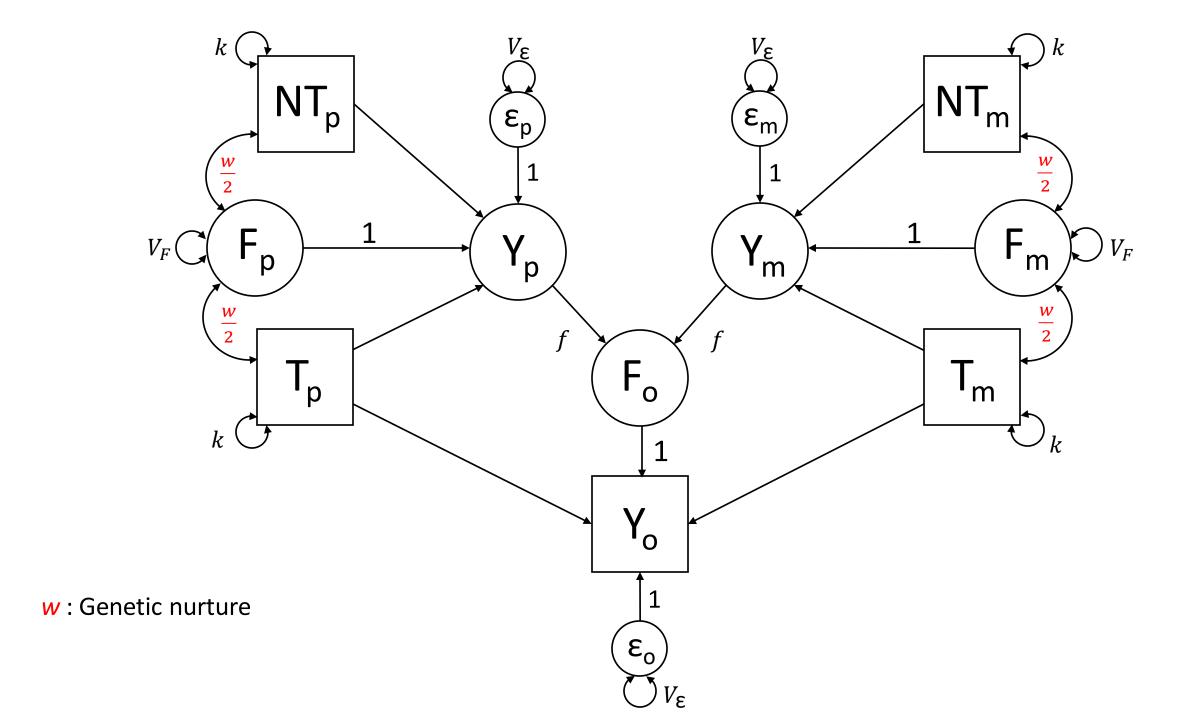


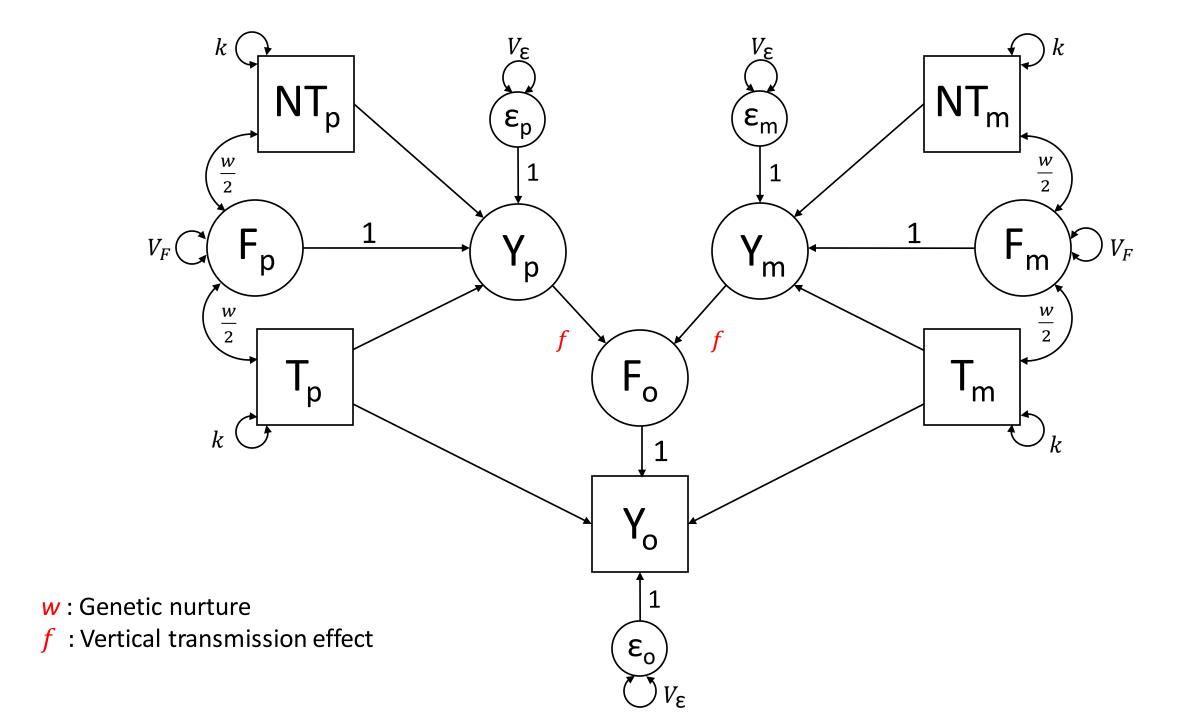


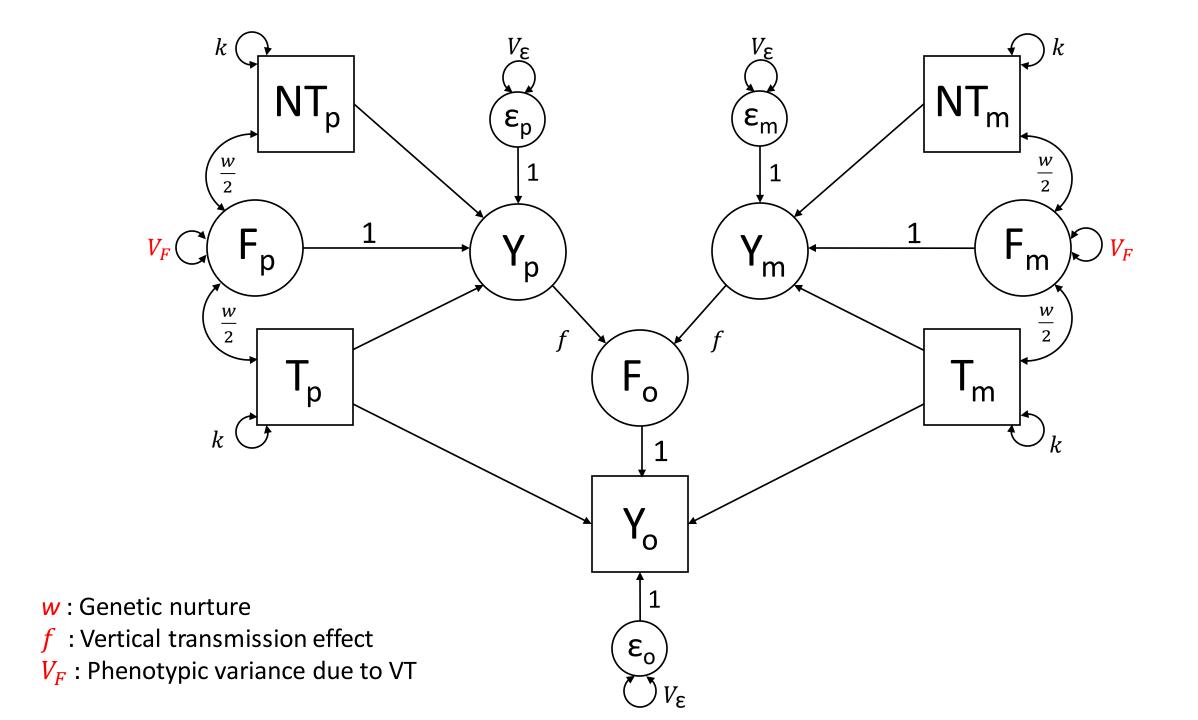


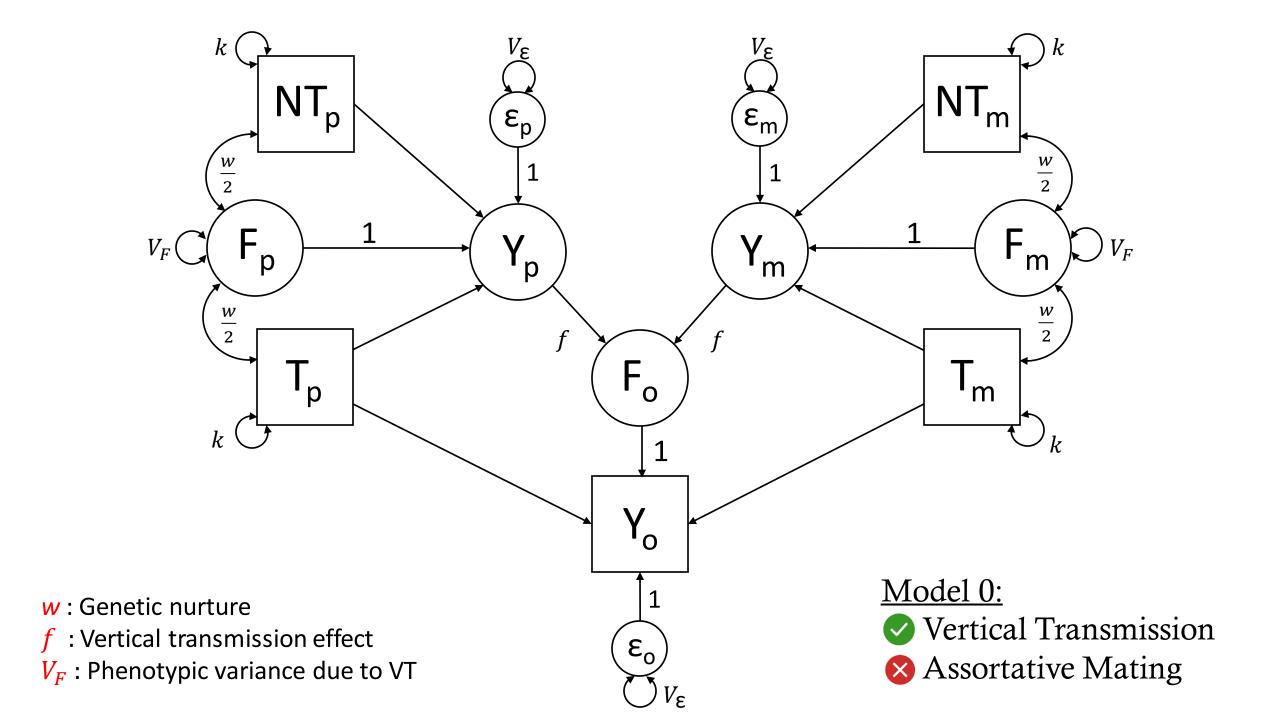


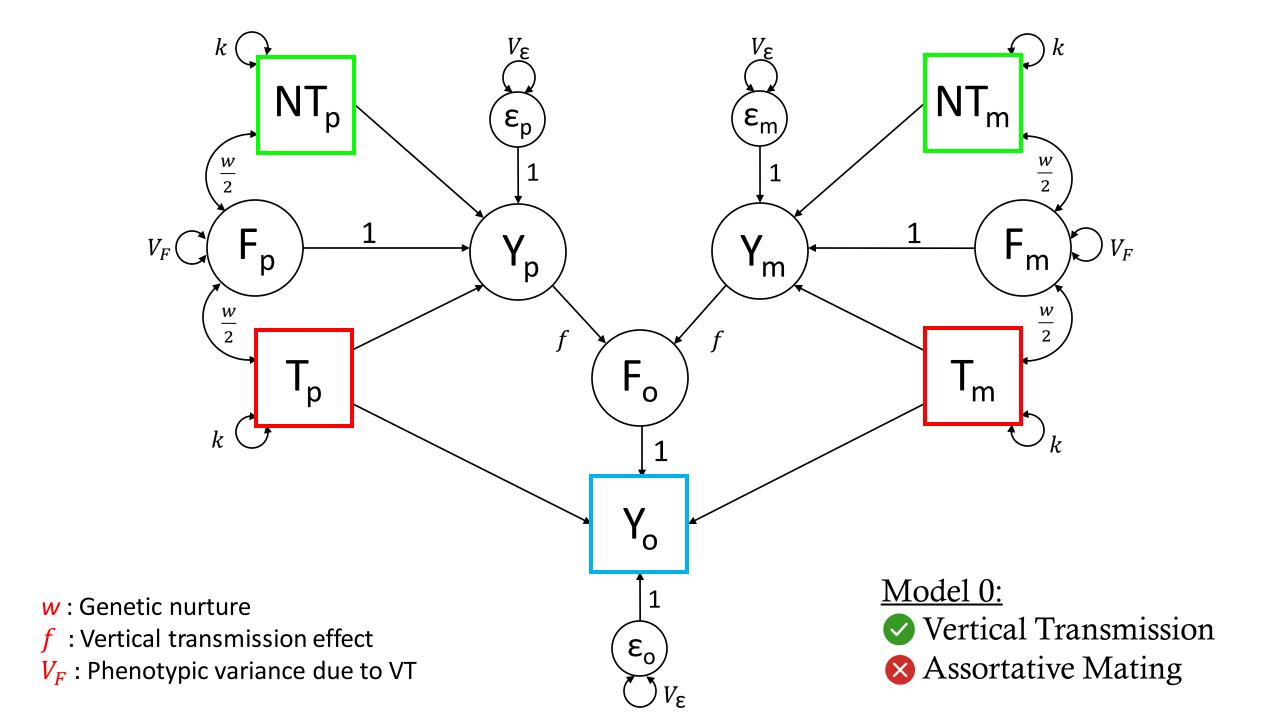


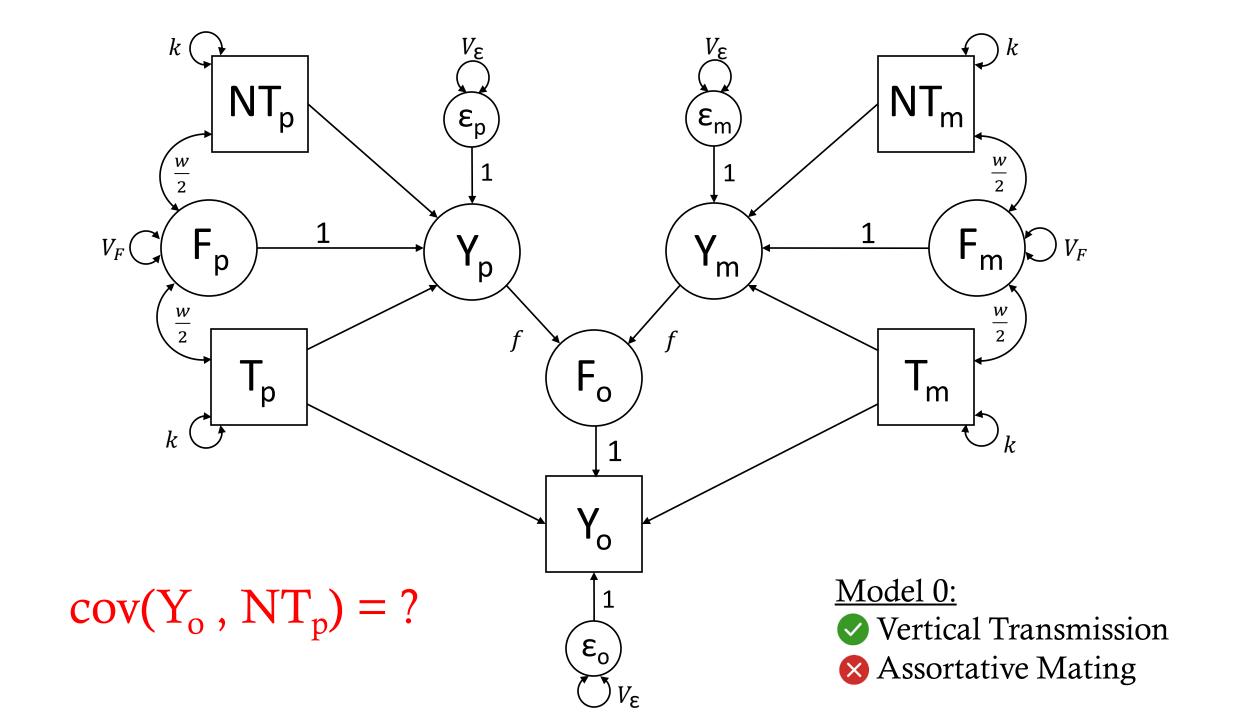


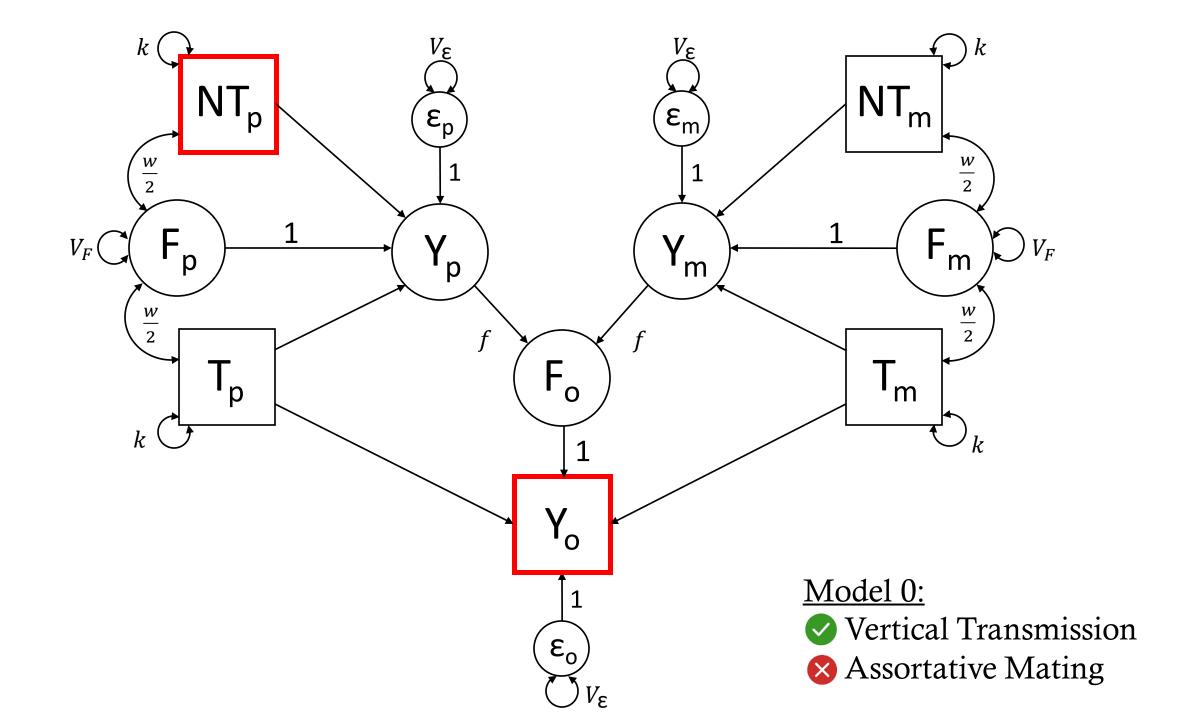






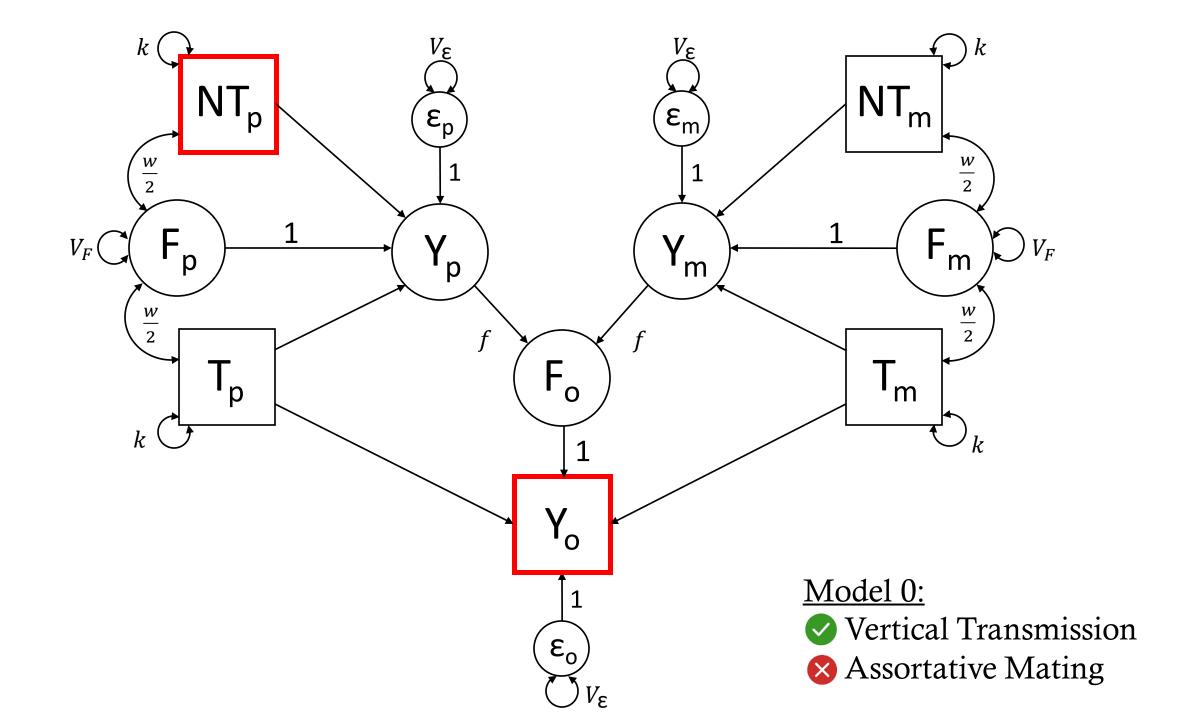






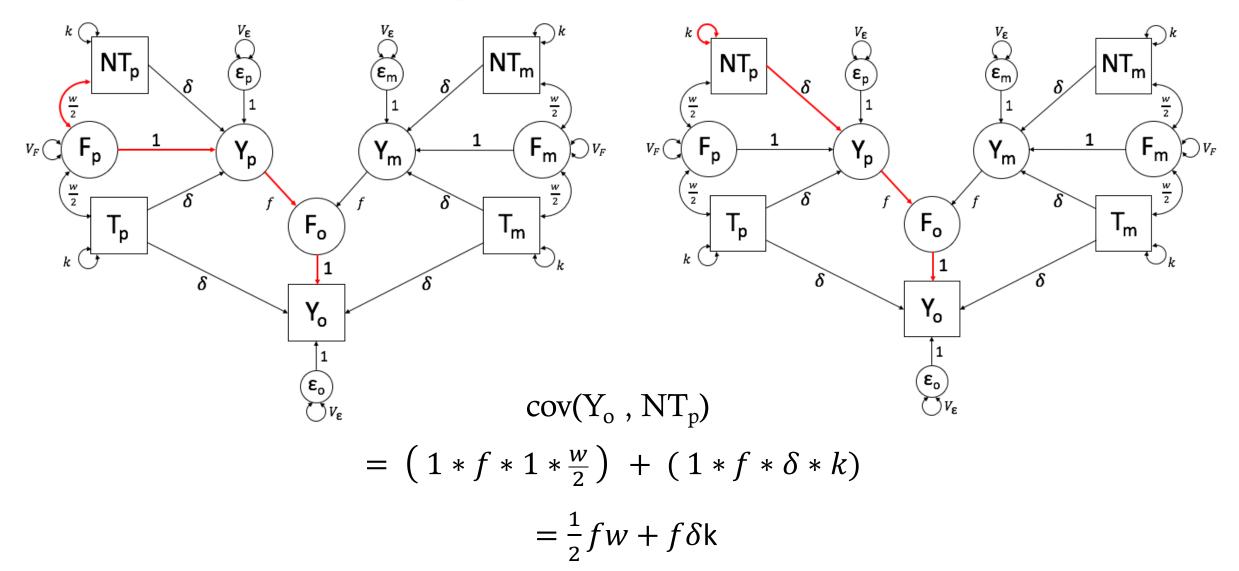
Path Tracing Rules (described in Balbona et al., 2021)

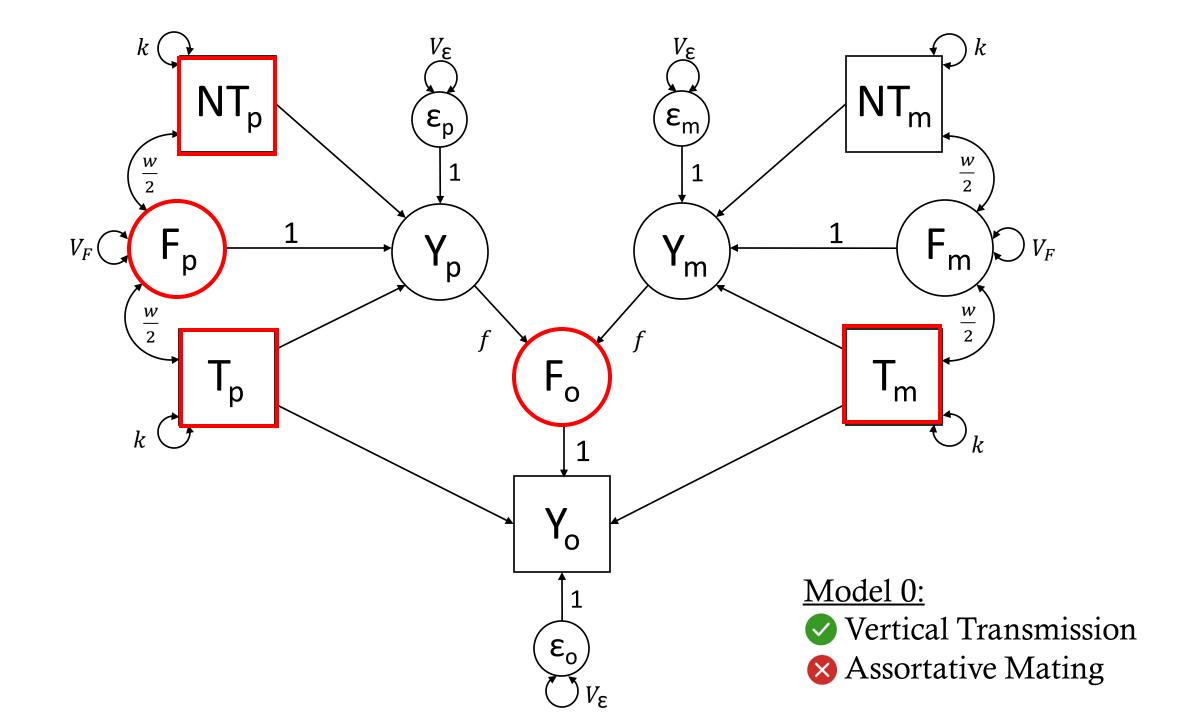
- 1. A chain begins by travelling backwards against the direction of a single or double-headed arrow (from the arrow's head to its tail). However, once a double-headed arrow has been traversed, the direction reverses such that the chain now travels forwards, in the direction of the arrows.
- 2. A chain must include exactly one double-headed arrow (a variance or a covariance term), which is equivalent to stating that a chain must change directions exactly once. This is necessary because double-headed arrows provide the proper scaling for the coefficients in each chain.
- 3. All chains must be counted exactly once and each must be unique. However, the order of the links in the chains matters. For example, despite being algebraically equivalent, the chain $Y_p \to NT_p \to T_p \to Y_p$ is distinct from the chain $Y_p \to T_p \to NT_p \to Y_p$ in Figure 1. Both are unique and both must be counted in determining the variance of Y_p .
- 4. Co-paths may only be traversed once in a given chain, and a chain must be legitimate before traversing the co-path. However, once the co-path is crossed, the first two rules above reset. A chain must therefore contain exactly one double-headed arrow before traversing the co-path, and one double-headed arrow after traversing the co-path. Thus, co-paths connect two legitimate chains to create a single, longer chain.

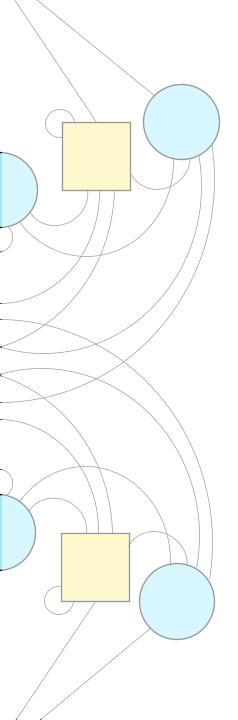


Model 0:

- ✓ Vertical Transmission,
- Assortative Mating







Thank you!

