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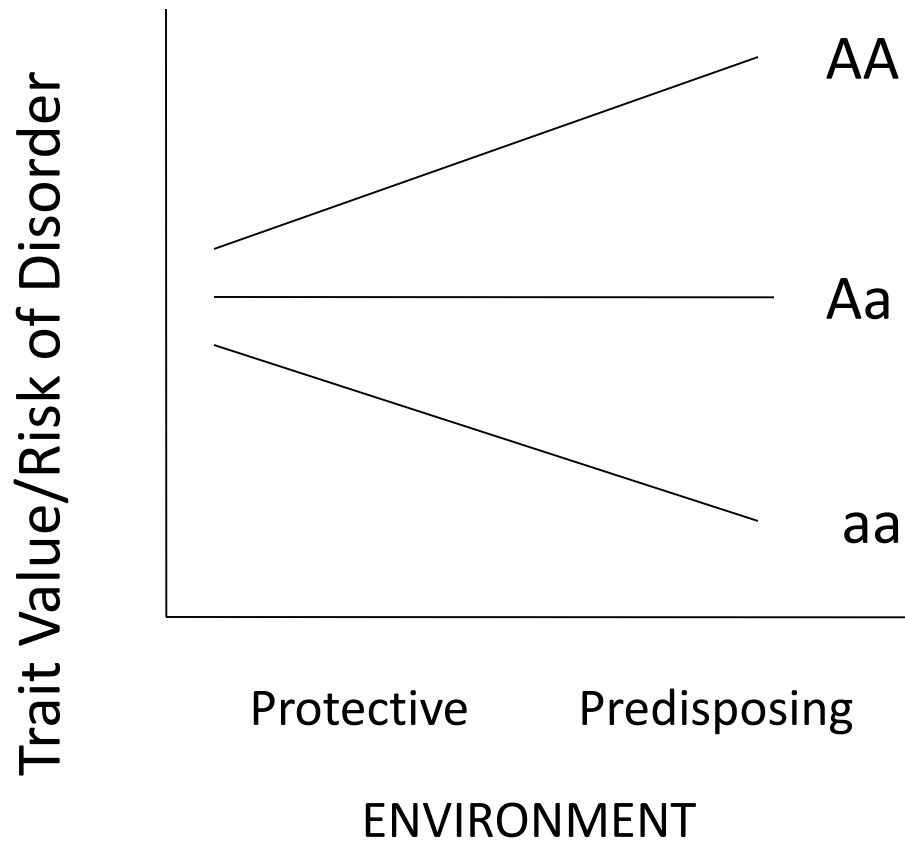
# Interplay between genes and environment

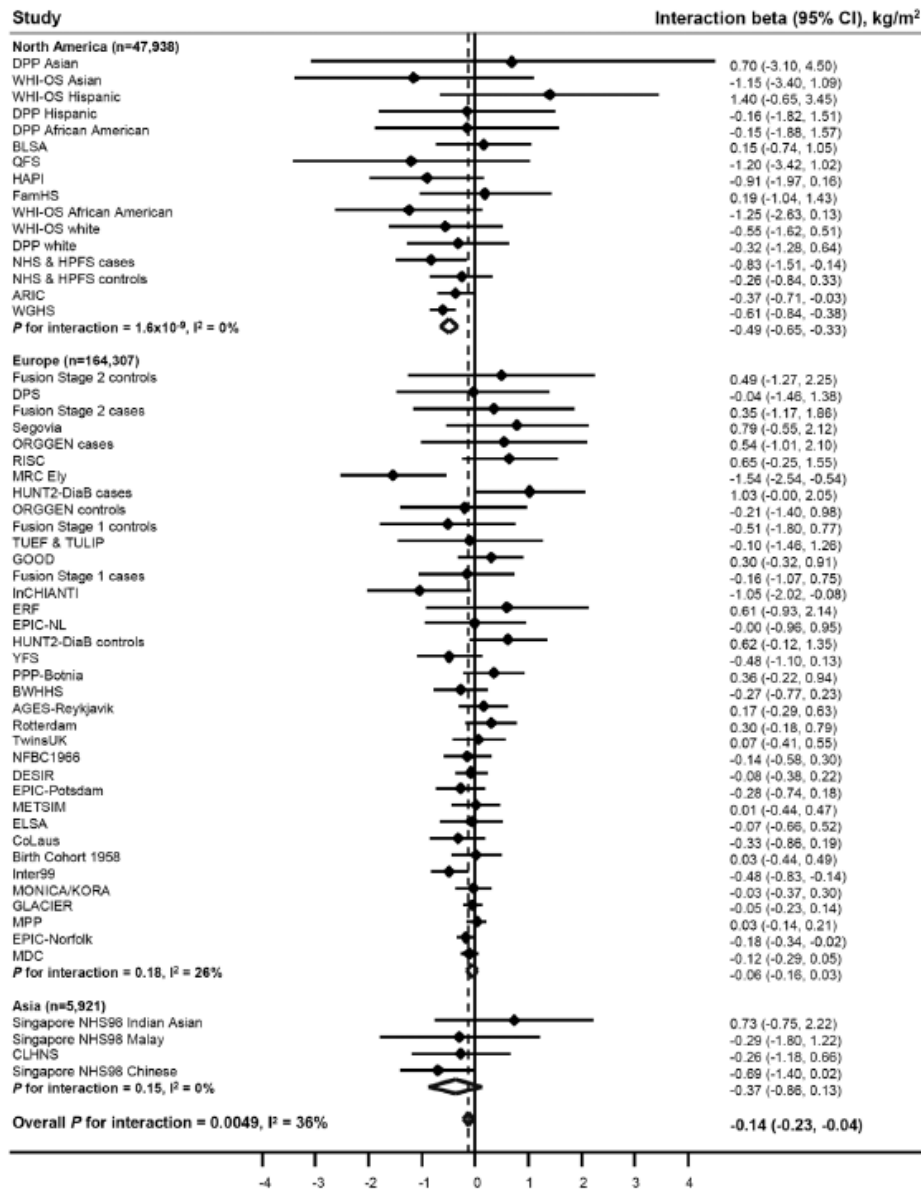
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# Interplay between genes and environment

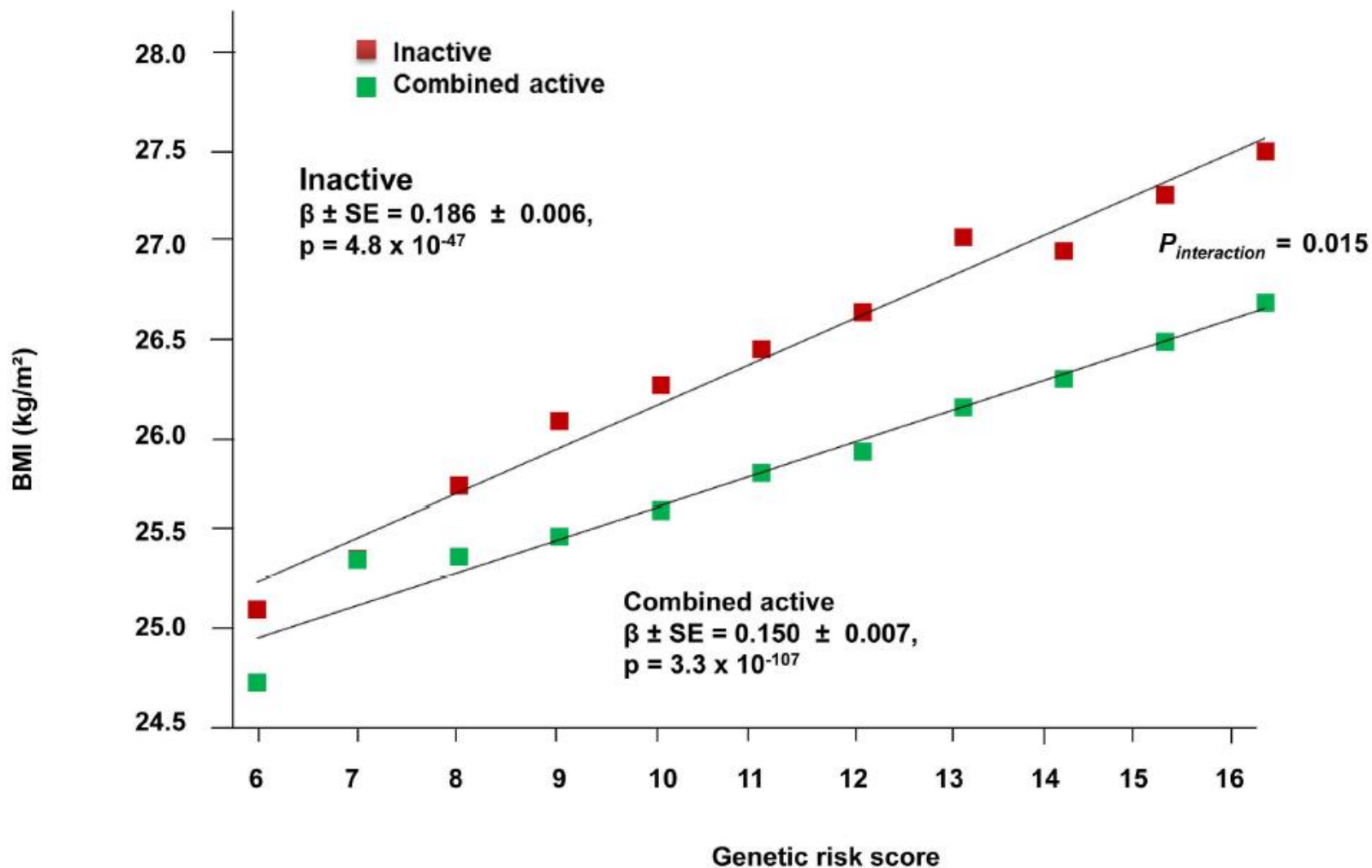
- ▶ Until recently the effects of genes and environment are usually treated independently in human genetics
- ▶ It is, however, very likely that genes and environment do not operate independently but rather interact with each other
- ▶ In gene–environment interactions, the effect of same gene differs between different environments or genes modify the effect of environmental exposure
  - In practice, these two mechanisms cannot usually be distinguished
- ▶ In recent years, interest in gene–environment interactions has dramatically increased because of pharmacogenetics
  - The responses of drugs may be modified by genotype
  - Understanding these interactions may lead to personalized medicine in the future

# Conceptualizing G\*E interaction in the case of single gene with two alleles





Interaction between physical activity and rs9939609 FTO variant when predicting BMI : a meta-analysis of 218,166 adults

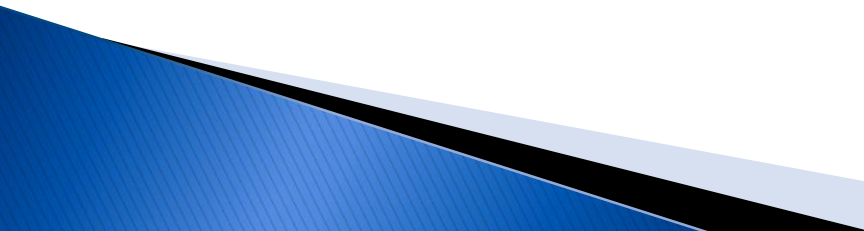


**Figure 2. Association between the GRS and BMI in the inactive and 'combined active' groups (N=111,421).** Physical activity was estimated according to the Cambridge Physical Activity Index (CPAI), where the inactive group is defined as individuals with a CPAI of 1 and the 'combined active' group as individuals with a CPAI of 2–4.

doi:10.1371/journal.pgen.1003607.g002

# **Twin studies in analyzing gene–environment interactions**

# Opportunities of twin studies

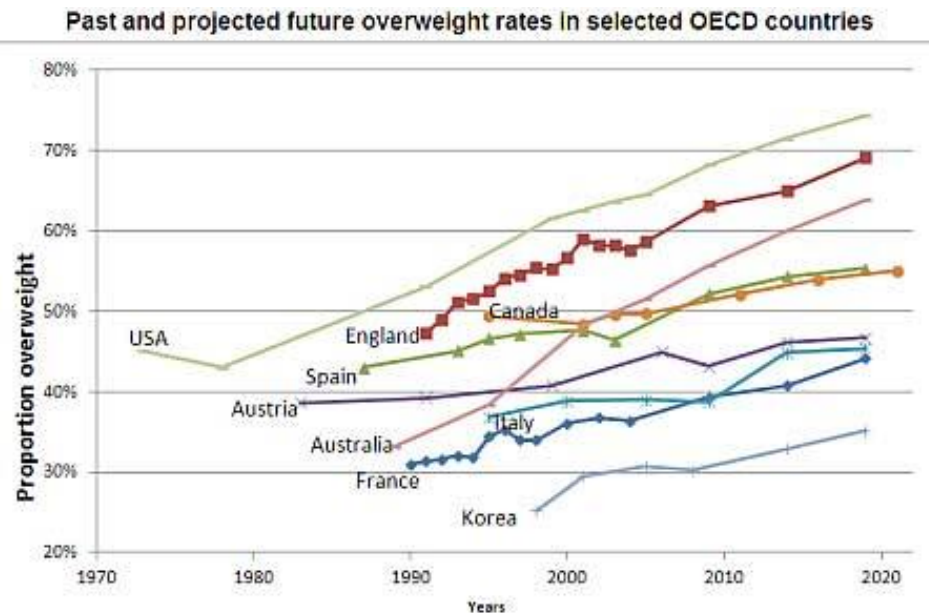
- ▶ Even when it is possible to analyze how environment modifies the effect of one gene or genetic risk score on a phenotype, twin studies of G–E interaction are still useful
  - ▶ Twin design allows to analyse total genetic variation whereas GRS explain only a few percent of the variation
  - ▶ For many traits, GRS or strong candidate genes are not available
  - ▶ Twin data is also much easier and cheaper to collect and are available for a long period of time
  - ▶ It is also possible to analyze environmental part of variation
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# G\*E interaction based on multiple group analysis

- ▶ A simple way to analyze G–E interactions is to stratify the data by the environmental exposure
  - Run genetic twin model in several groups and study differences between the variance components
- ▶ Thus, we can simply utilize multiple group comparison using univariate models
- ▶ Significant differences in genetic and/or environmental variance components across the categories indicate the existence of G–E interaction
  - Can be tested by  $\chi^2$ –statistics
- ▶ Shows how simple methods can sometimes be used to answer complex research questions



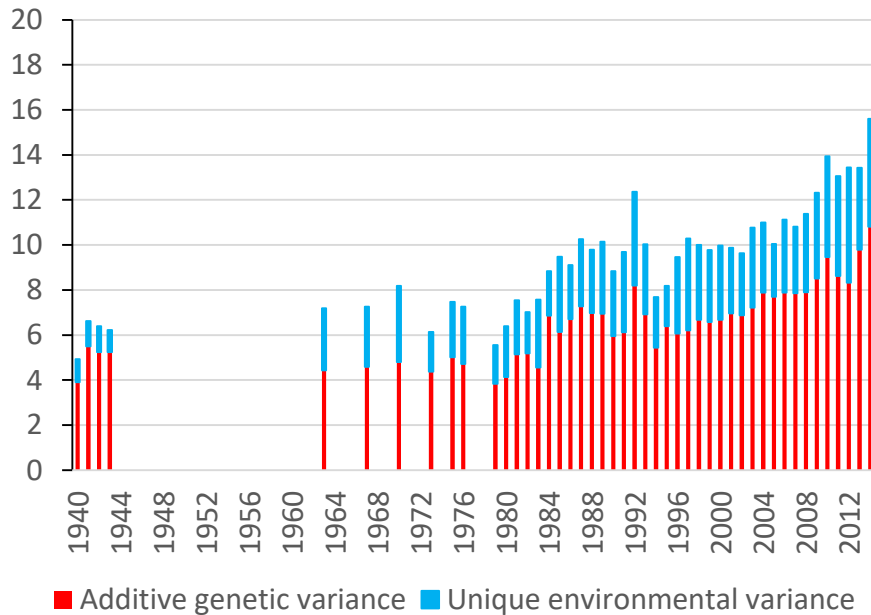
# Trends of overweight in the world



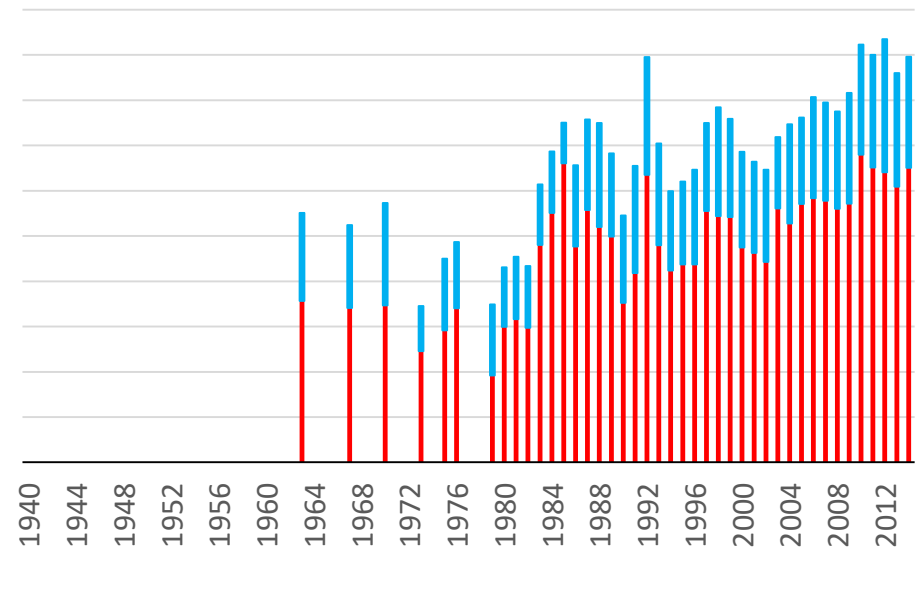
- Similar trends in overweight and obesity have been found all over the world
- However, also remarkable differences between countries
- Not directly associated with standard of living
- Obesity levels are highest in USA and lowest in East-Asia
- May indicate cultural factors such as traditional diets
- Consumption of meat vs. vegetables

# Additive genetic and unique environmental variance of BMI by measurement year

## Males



## Females

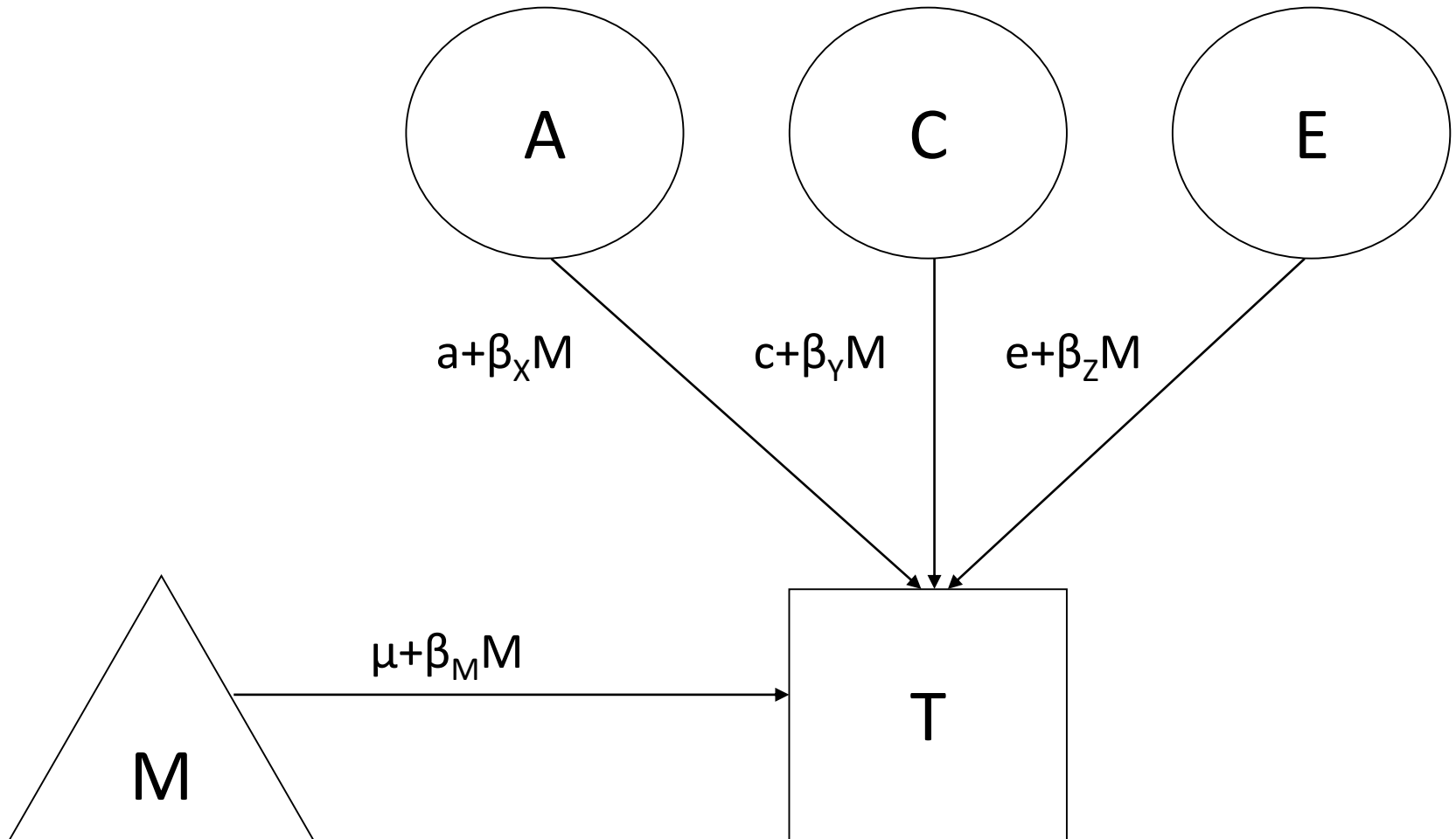


# Problems in multiple group comparisons

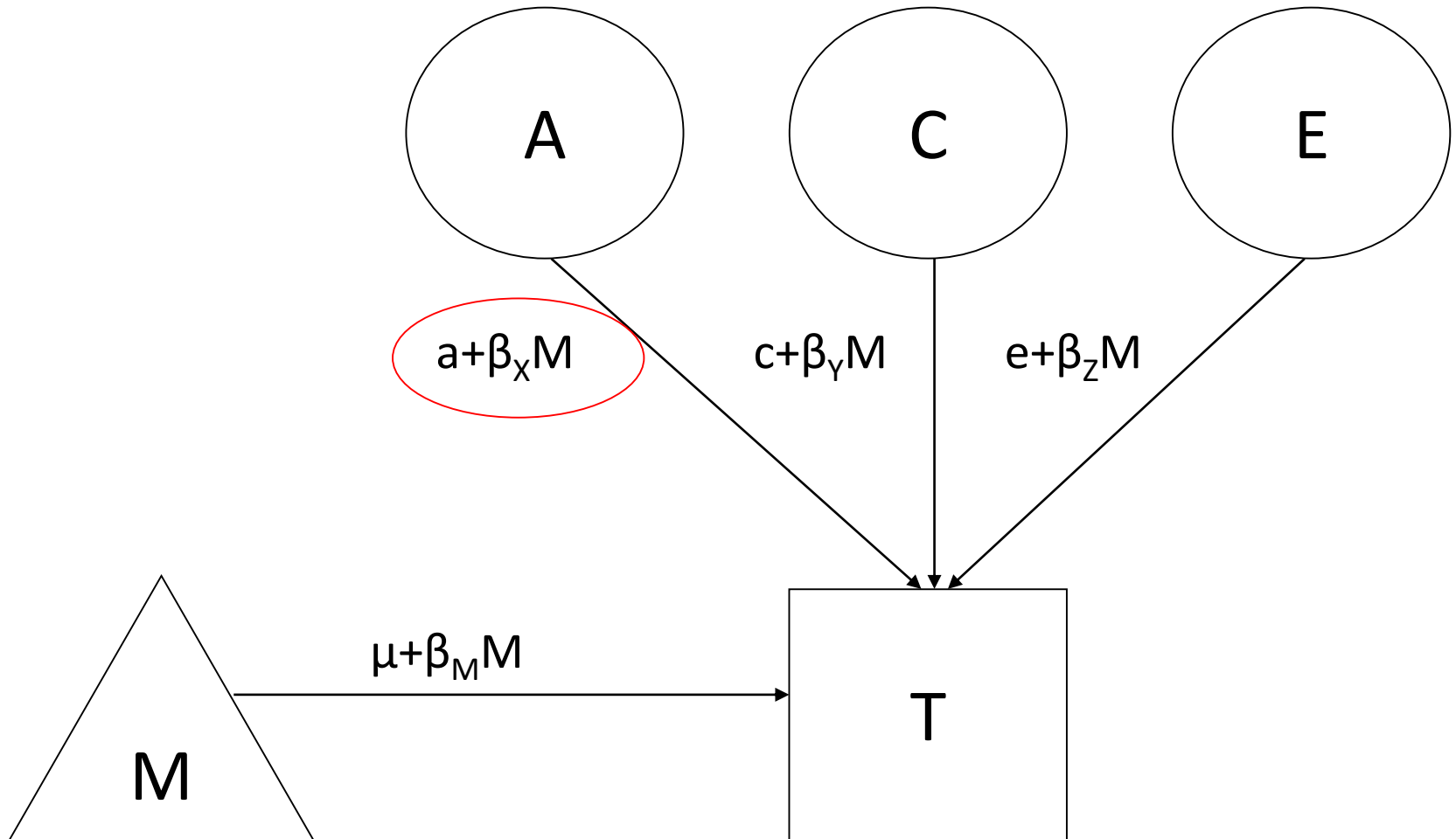
- ▶ Multiple group comparisons have limitations, which make them unsuitable to many situations
- ▶ Environmental exposure needs to be same for both co-twins
  - Such as birth cohort or place of residence
- ▶ If environmental exposure is continuous, categorizing it loses a lot of information if the associations are linear
- ▶ However if this kind of limitations are not a problem, multiple group comparison is a good alternative to more sophisticated G-E interaction models
  - Interpretation of the results is very straightforward
  - Possible non-linearity is not a problem
  - We can accept heterogeneity between the categories
  - Models are very robust whereas more complex models can be sensitive to starting values
- ▶ Especially if the modifying effect does not follow any function, this approach is a good alternative
  - However, there are also statistical methods to model this type of interaction effects

# **Gene–environment interaction model for twins**

# G-E interaction model



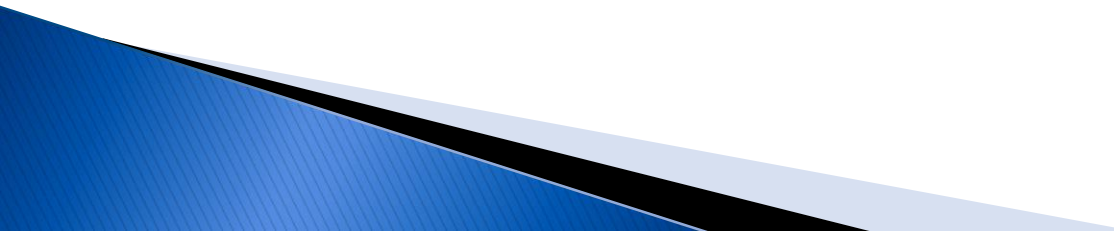
# G-E interaction model



# Matrix algebra for G-E interactions

- ▶ The equation  $a + \beta_X M$  is a linear function
  - Why this can be used to analyze interactions?
- ▶ We are interested in the variance component  $a^2$  instead of the path coefficient  $a$
- ▶ Thus  $(a + \beta_X M)^2 = a^2 + 2 * a * \beta_X M + (\beta_X M)^2$
- ▶ This can be easily generalized to multivariate case using matrix algebra rules

# Practical

- ▶ We will analyze how age affects the variation of BMI in Finnish Twin Cohort
  - ▶ Human BMI increases by age
  - ▶ Is this affected by increased environmental or genetic variation?
  - ▶ Script GE model.R
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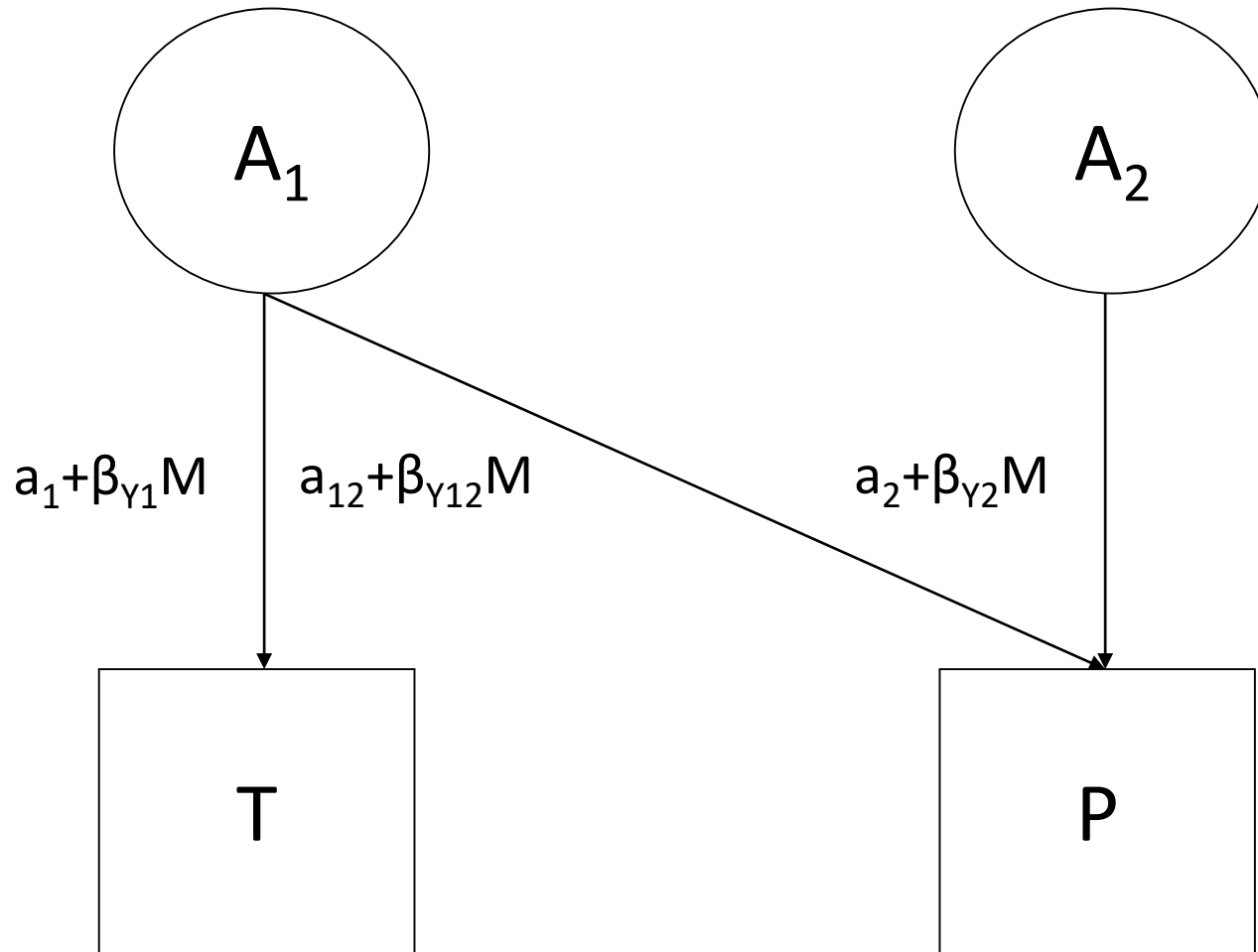
# Starting values in GE models

- ▶ GE models can be quite sensitive to starting values
  - This is a general problem in OpenMx but it is especially difficult to GE models
- ▶ This is understandable because quite similar total variance can be found by different combinations of intercept (variance in moderator value 0) and slope (change of variance as a function of moderator)
- ▶ Sometimes the model does not fit
- ▶ However, even more problematic is that the model finds a local maximum of likelihood function
- ▶ In that case, the results may change when changing the starting values
- ▶ Need to try different starting values and compare  $-2LL$  values

# **More complex gene-environment interaction models**



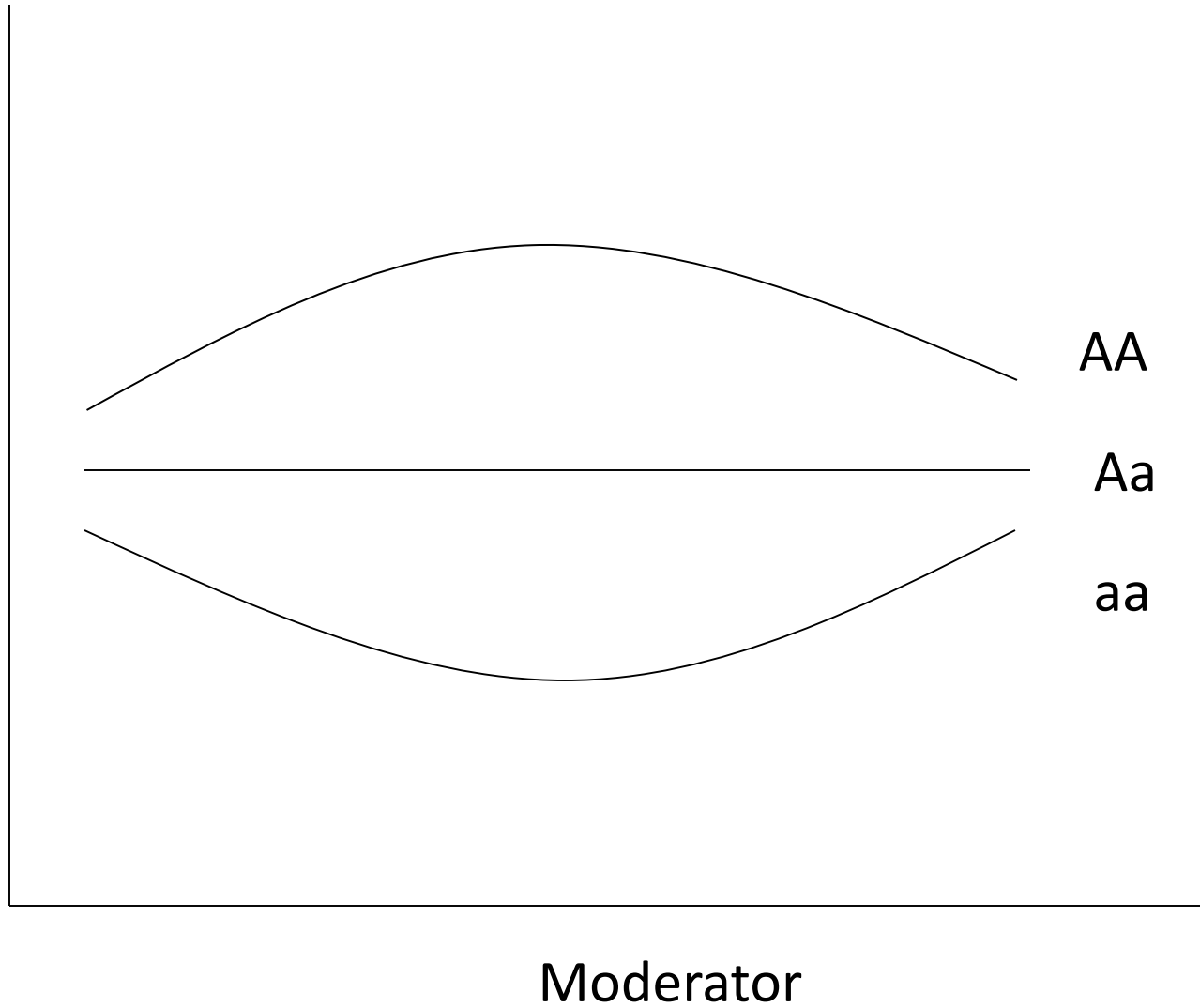
# Multivariate G-E interaction model



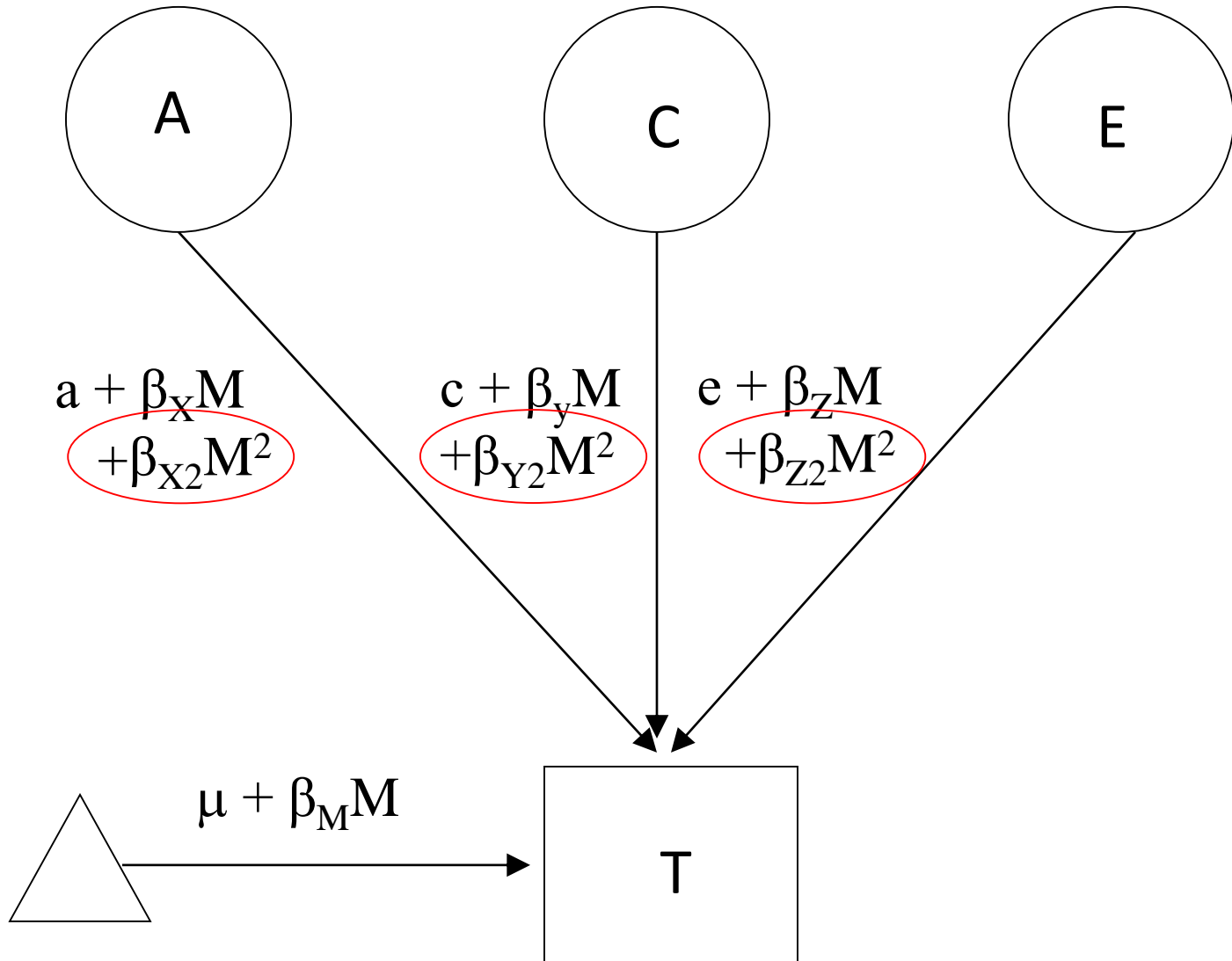
# Non-linear interaction effects

- ▶ It is also possible that the effect of environmental exposure is not linear but curvilinear
- ▶ For example, genetic variation may be low both at low and high level of environmental exposure
- ▶ This can be modeled simply by including a new moderator term in the model
- ▶ Even when curvilinear effects are not difficult to model, power may be a problem
- ▶ Also the extreme ends of environmental exposures may be problematic
  - Reporting errors etc.
- ▶ Before analyzing curvilinear associations, there should be clear theoretical justification why we expect this kind of associations
- ▶ Sample size should also be large and the measurement of environment high quality

# Nonlinear Moderation



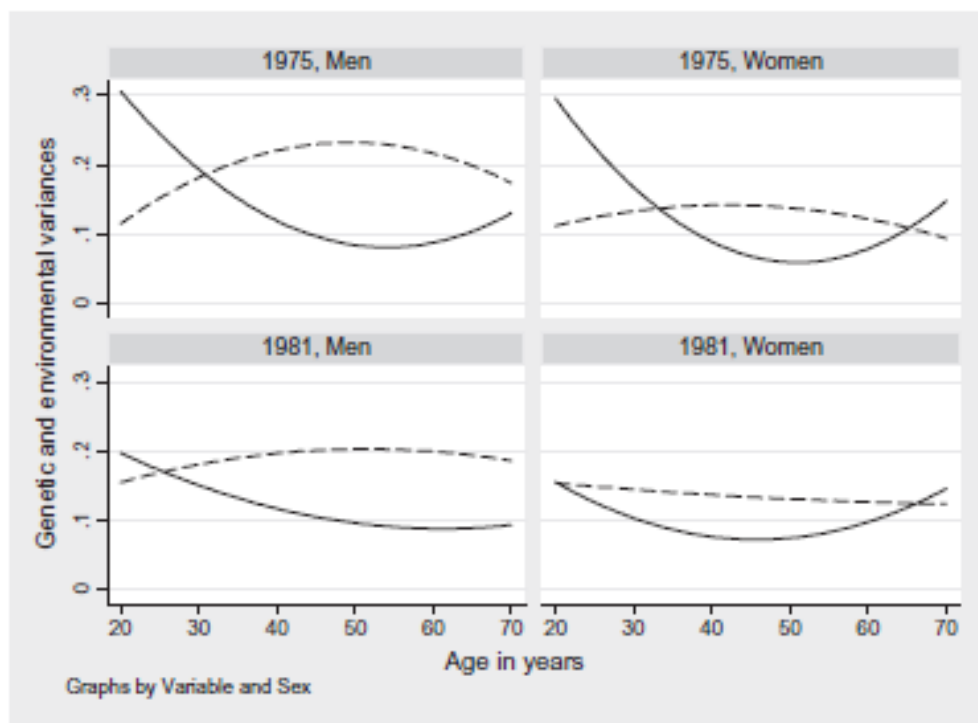
# Nonlinear Moderation can be modeled with the Addition of a quadratic term



## Genetics of coffee consumption and its stability

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**Figure 1** Changes of additive genetic (continuous line) and unique environmental (dashed line) variance with increasing age in quadratic gene-environment interaction model

# Genetics of coffee consumption and Finnish culture

- ▶ Coffee consumption is highest in Finland in the world (9.9 kg/person/year)
  - This is quite a lot especially when considering that usually only adults drink coffee in Finland
- ▶ In Finland, it is very common that in work contracts it is said that employees can have two 12 minutes coffee breaks (this term is used) payed by the employer
  - Tea is also consumed but much less than coffee and traditionally it is consumed at evening
- ▶ Thus, coffee drinking is an important social event in many work places
- ▶ This may explain the lower genetic variation of coffee consumption in working age population
- ▶ Many persons who do not initially like coffee, may start to drink it at work because of social reasons
- ▶ The effect may have declined from 1971 to 1982 when tea has become a more common alternative to coffee also at daytime



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