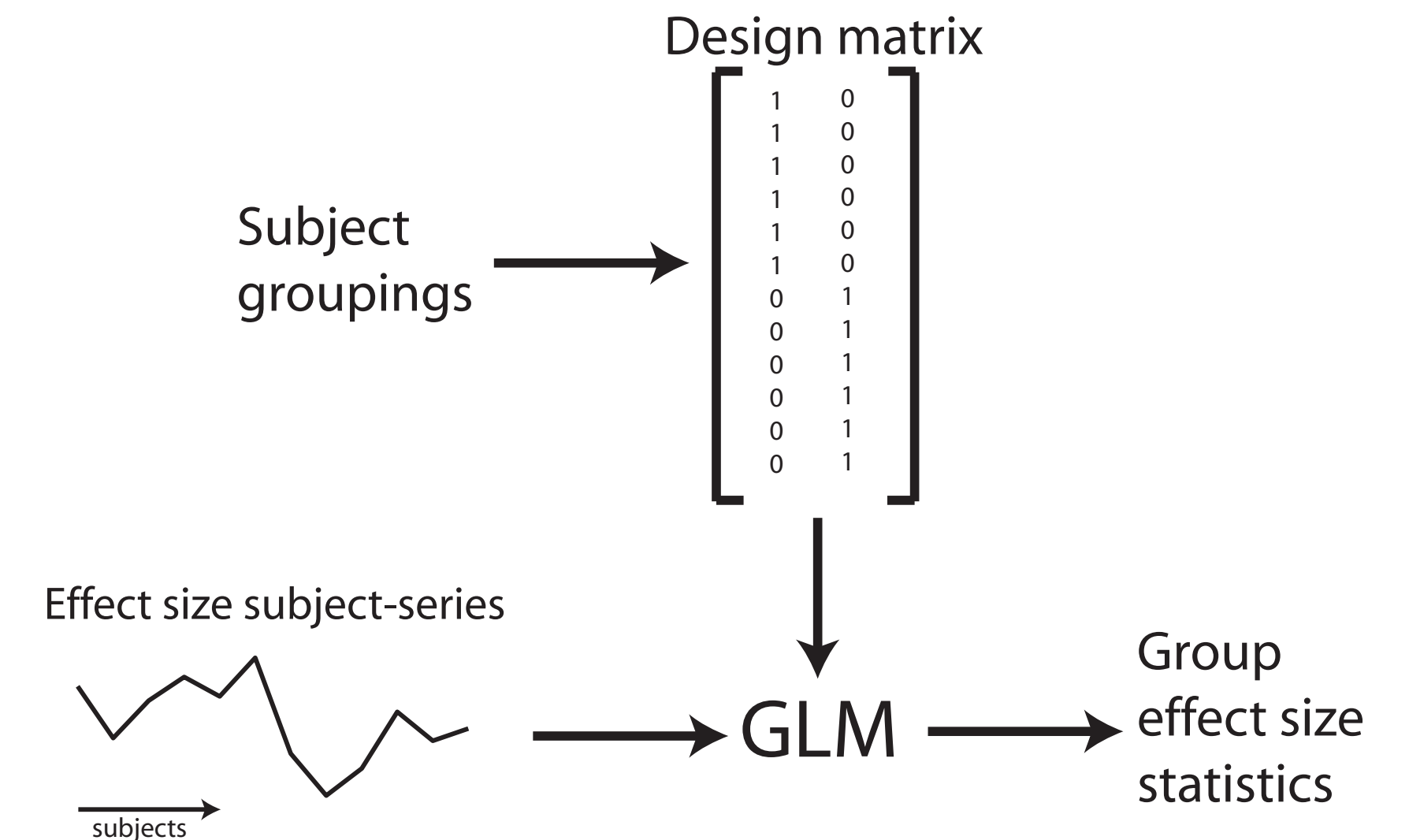
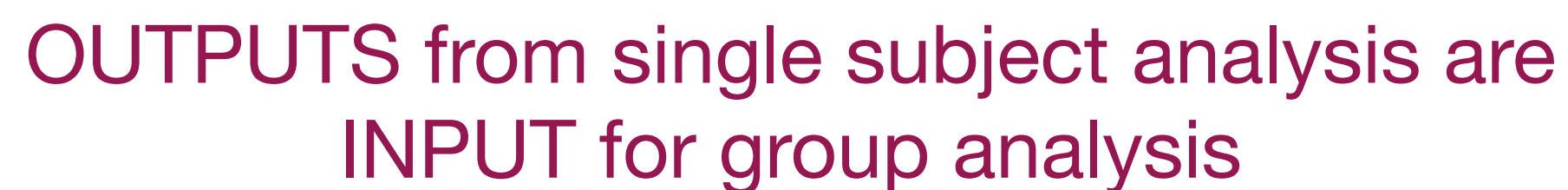


FMRI group analysis

- Overview
- Fixed versus mixed effects
- Multiple sessions per subject





Regressors,
Explanatory Variable (EV)

Regression parameters,
Effect sizes

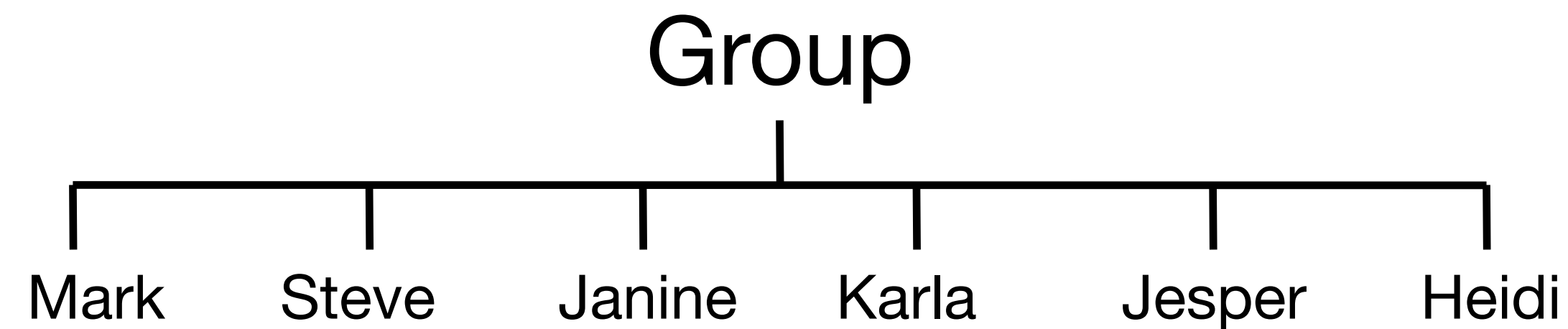
subjects

$\mathbf{y} = \mathbf{X} \boldsymbol{\beta} + \mathbf{e}$

COPEs from
a voxel across
subjects

Design Matrix

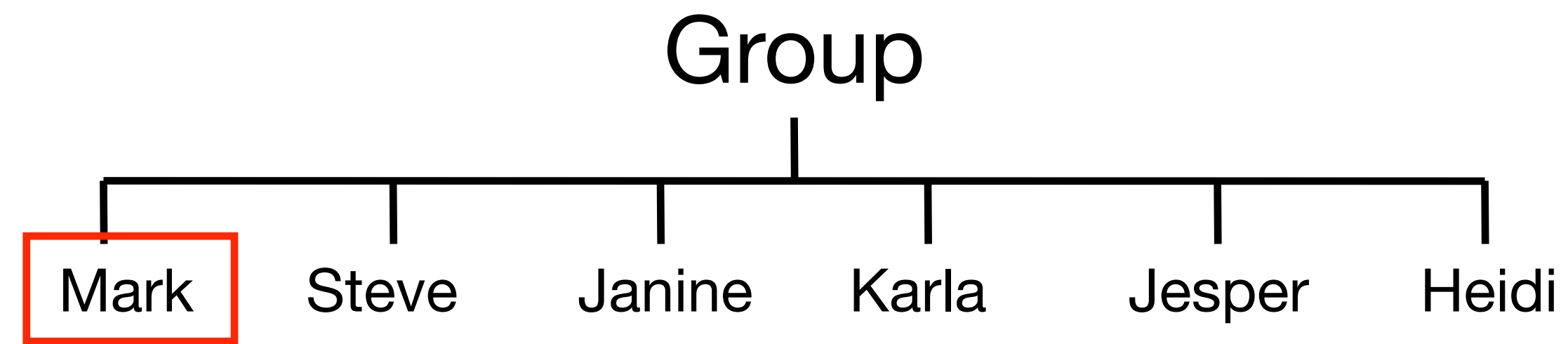
A simple example



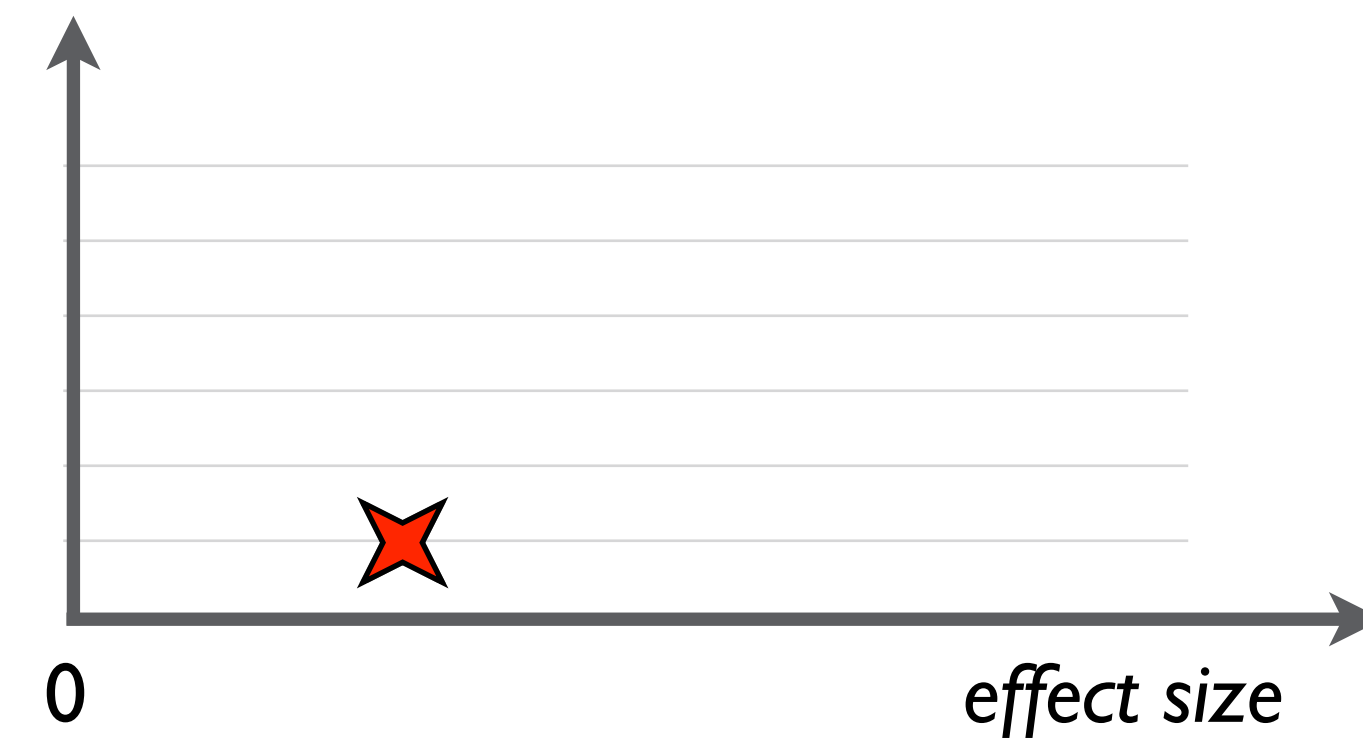
Does the group activate on average?

For a specific PE/contrast from the first level analysis,
which part of the brain were significantly activated across all subjects?

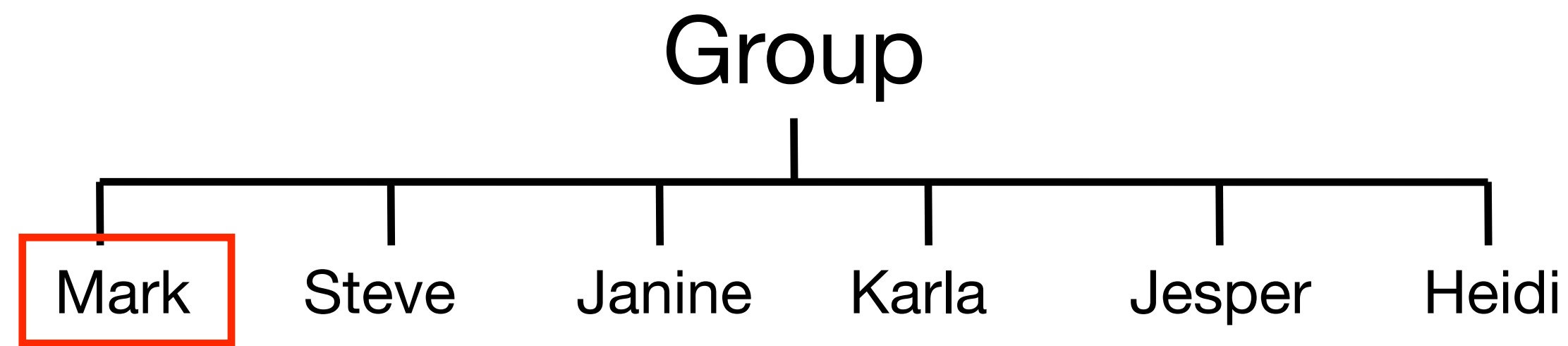
A simple example



Does the group activate on average?



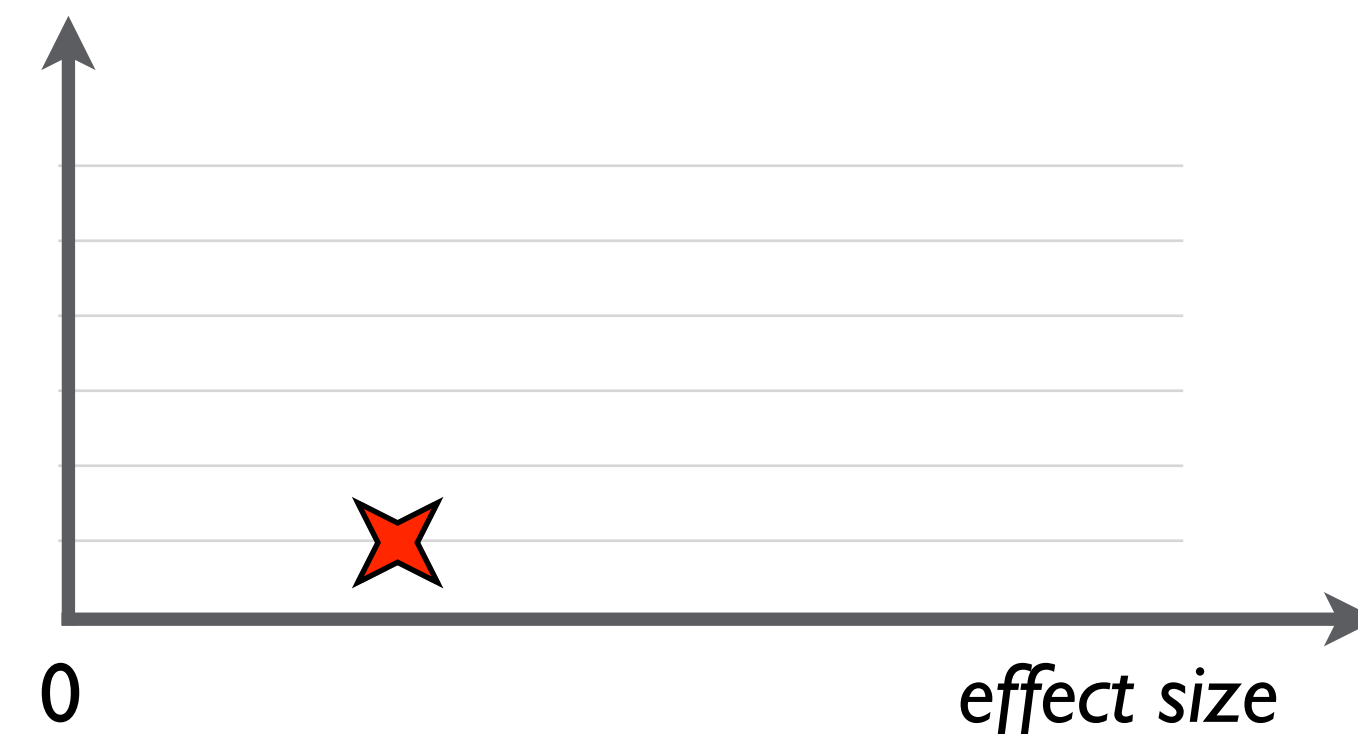
A simple example



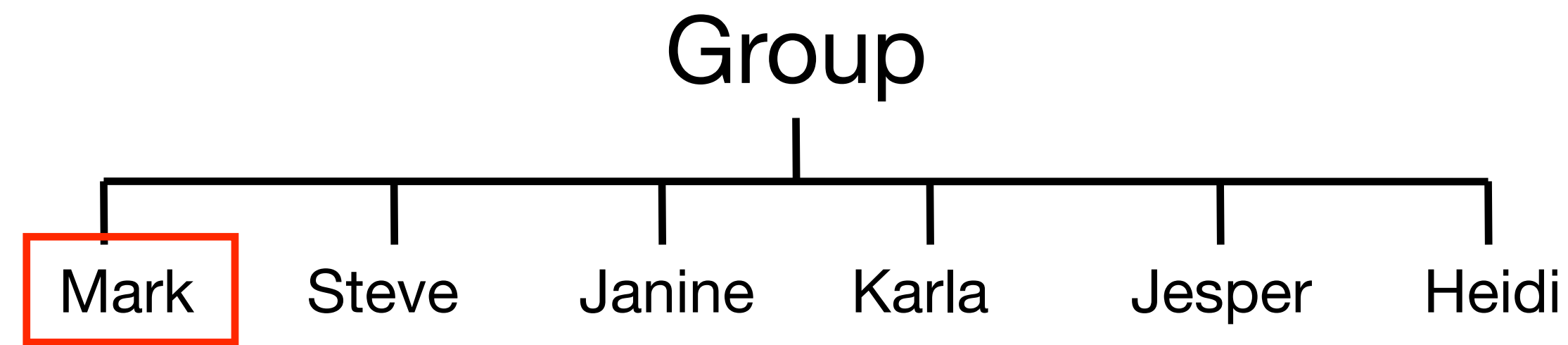
Does the group activate on average?

$$Y_k = X_k \beta_k + \epsilon_k$$

First-level GLM
on Mark's 4D FMRI
data set

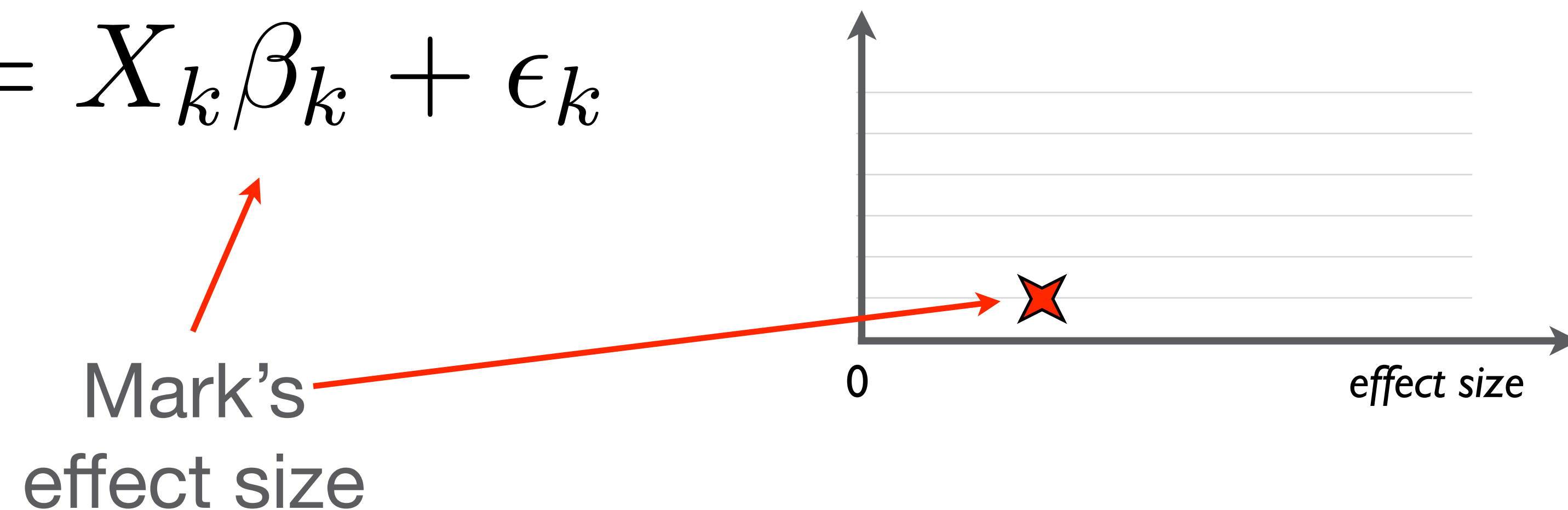


A simple example

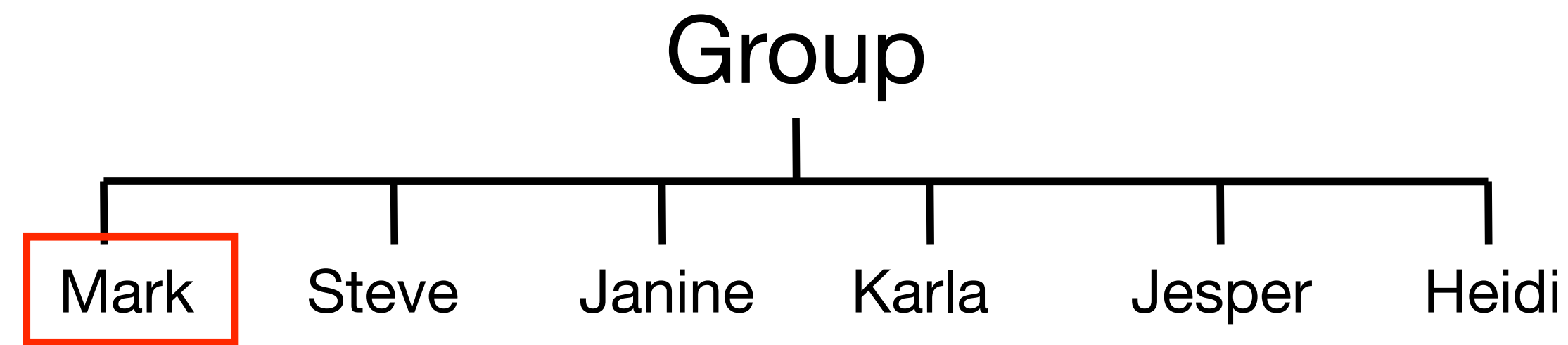


Does the group activate on average?

$$Y_k = X_k \beta_k + \epsilon_k$$

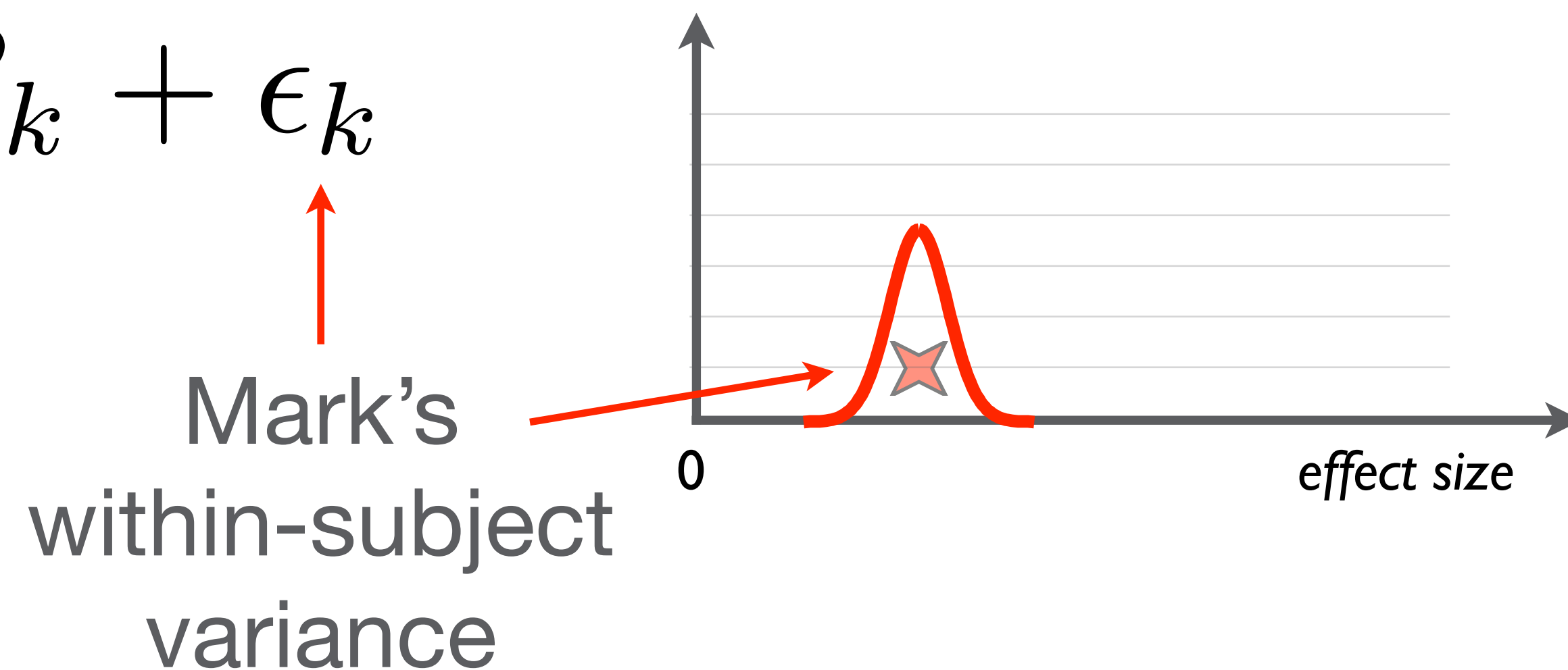


A simple example

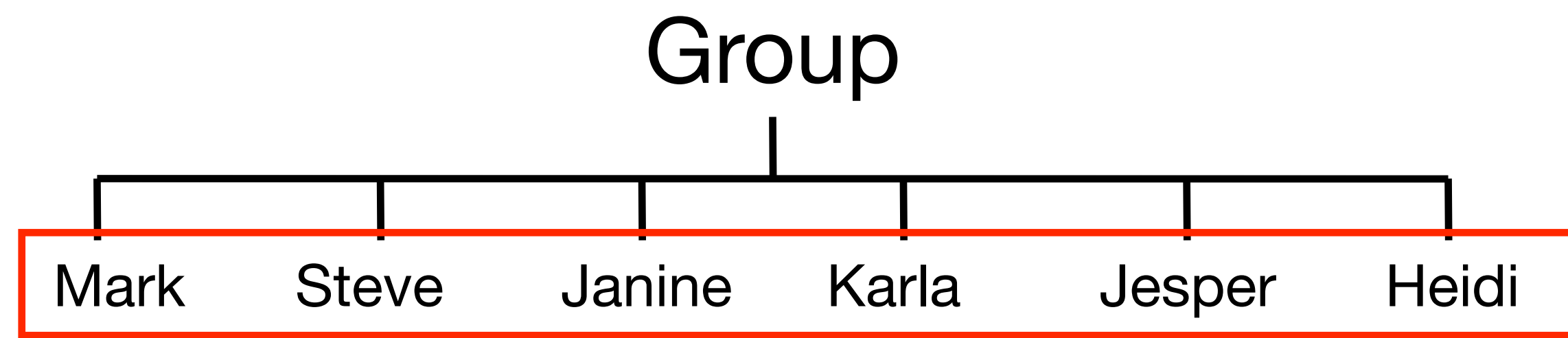


Does the group activate on average?

$$Y_k = X_k \beta_k + \epsilon_k$$



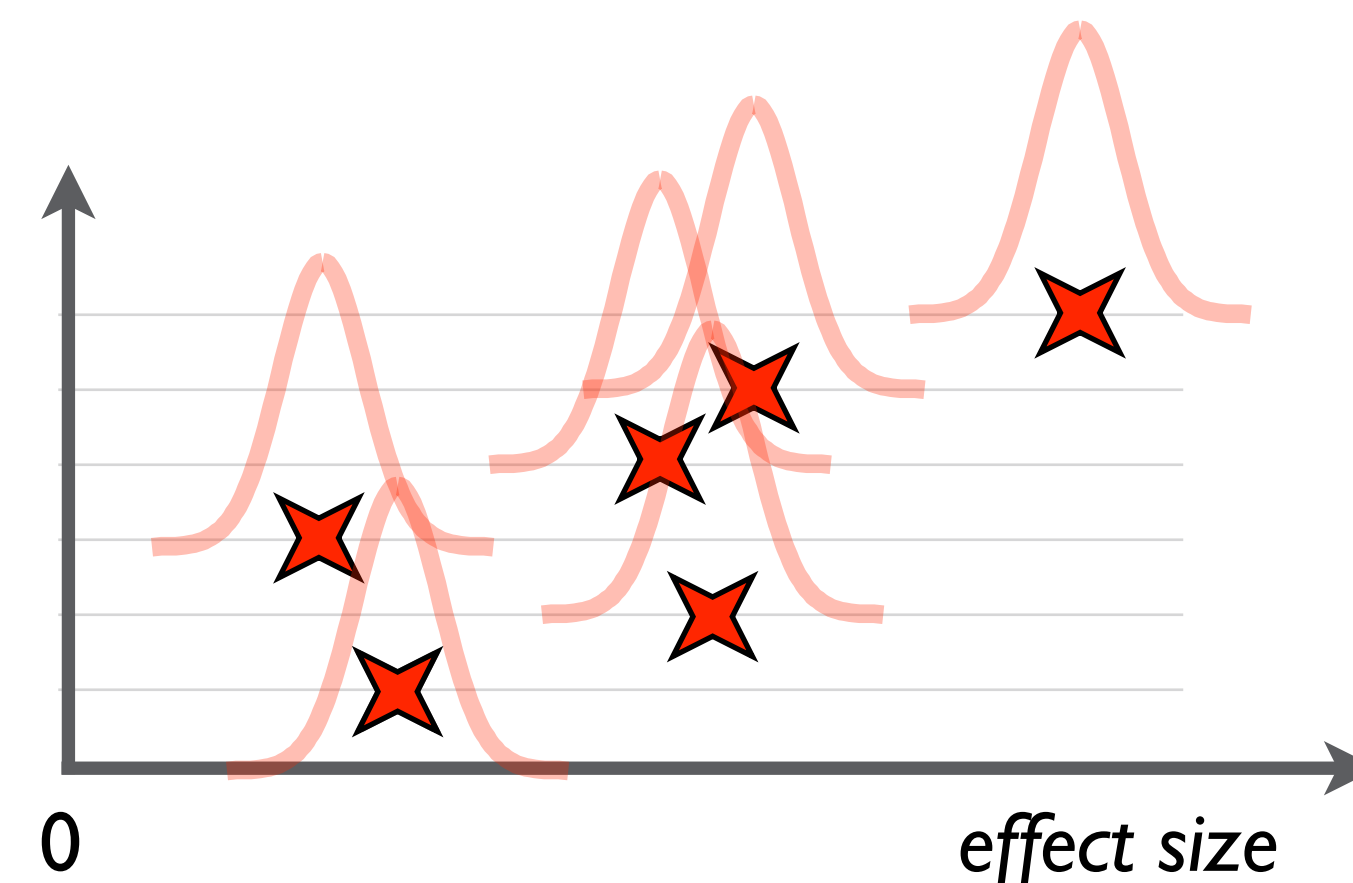
A simple example



Does the group activate on average?

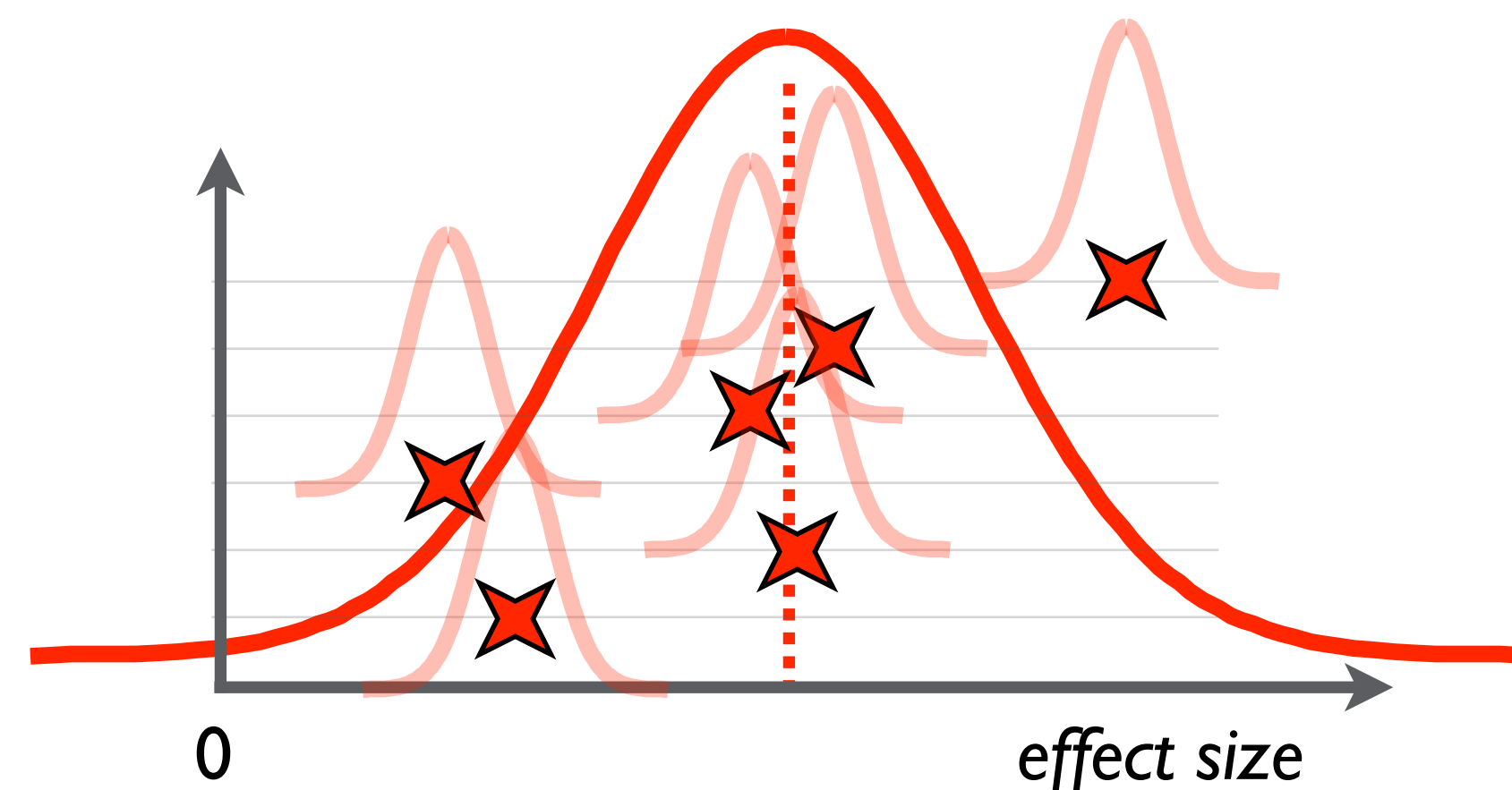
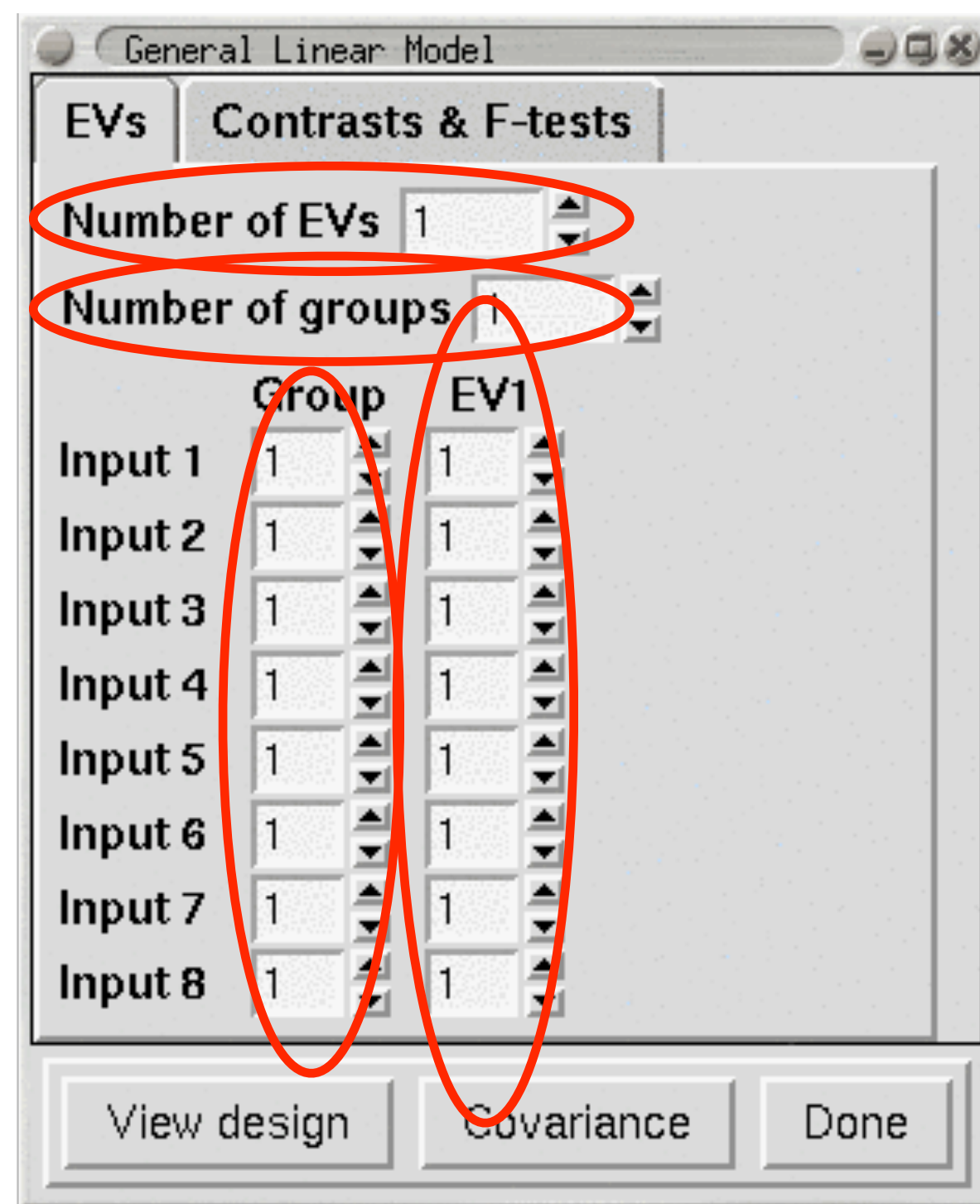
$$Y_k = X_k \beta_k + \epsilon_k$$

All first-level GLMs
on 6 FMRI data set



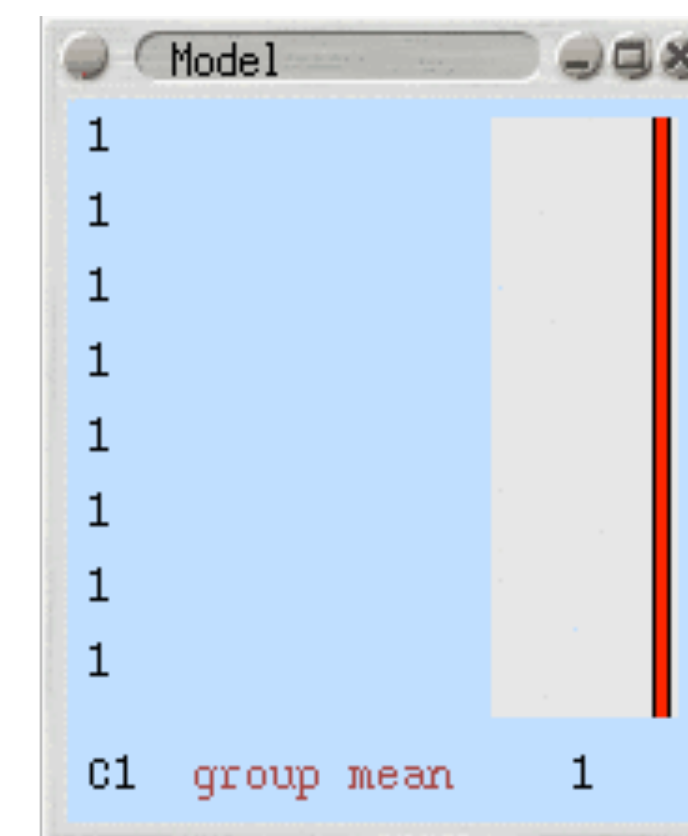
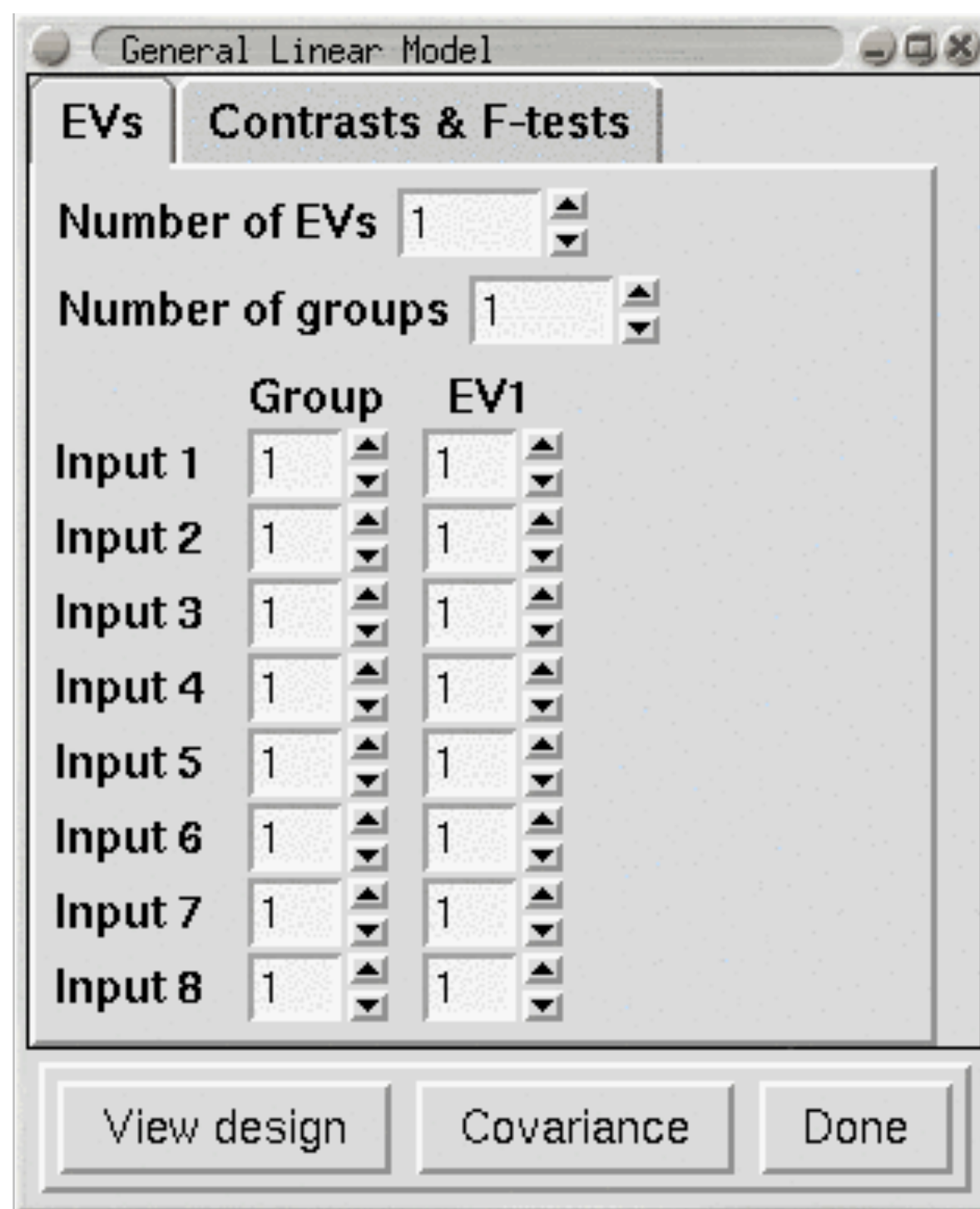
Single Group Average

Does the group activate on average?



Single Group Average

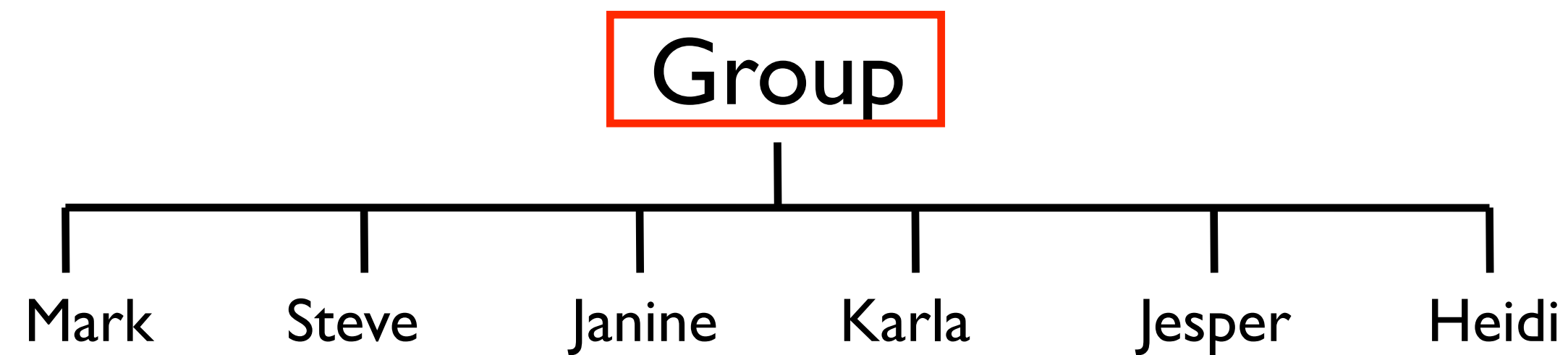
Does the group activate on average?



Fixed versus mixed effects

Fixed-Effects Analysis

Do these exact 6 subjects activate on average?

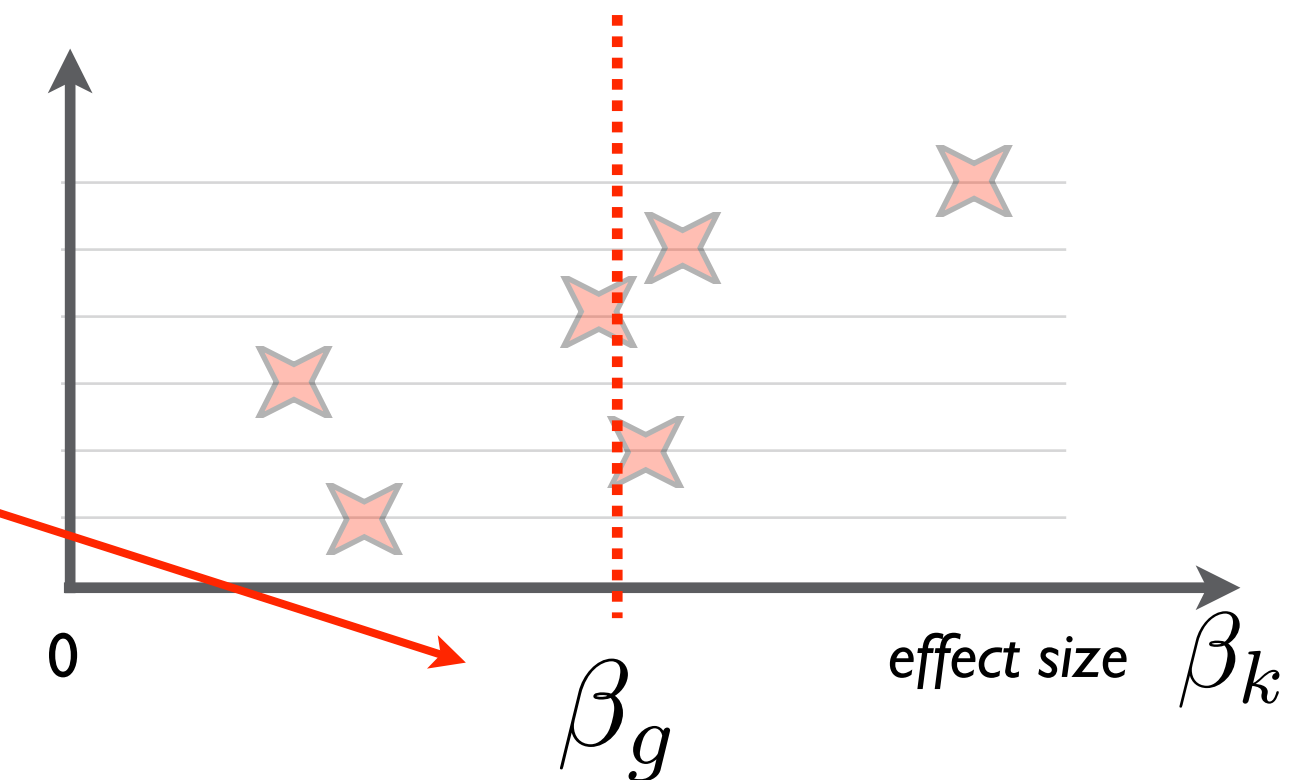


$$Y_K = X_K \beta_K + \epsilon_K$$

$$\beta_K = X_g \beta_g$$

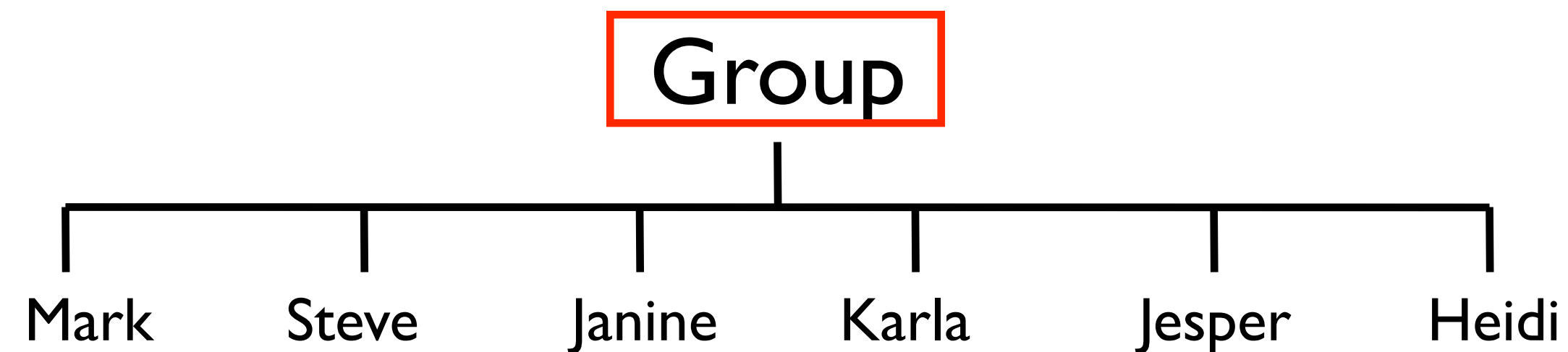
Fixed Effects Analysis:

- Consider only these 6 subjects
 - estimate the mean across these subject
 - only variance is within-subject variance



Mixed-Effects Analysis

Does the population activate on average?

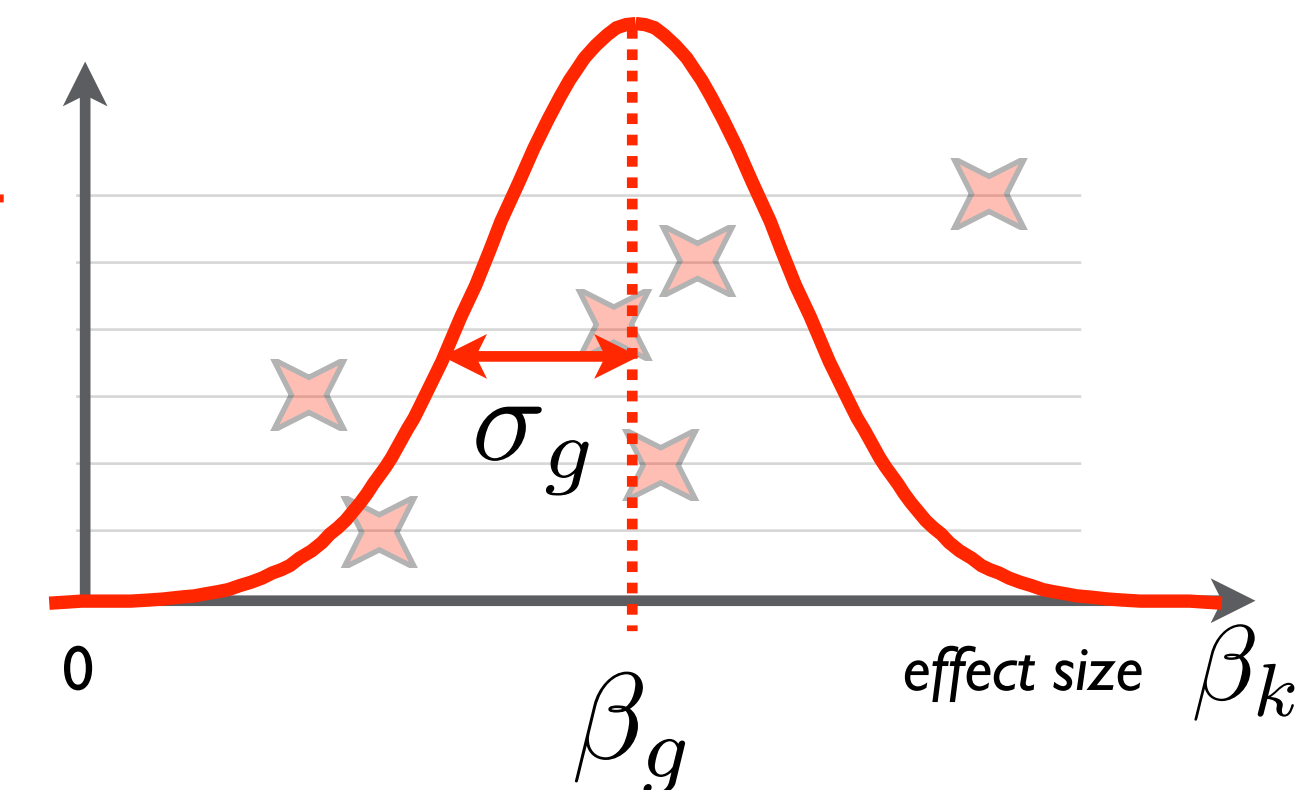


$$Y_K = X_K \beta_K + \epsilon_K$$

$$\beta_K = X_g \beta_g + \epsilon_g$$

Mixed-Effects Analysis:

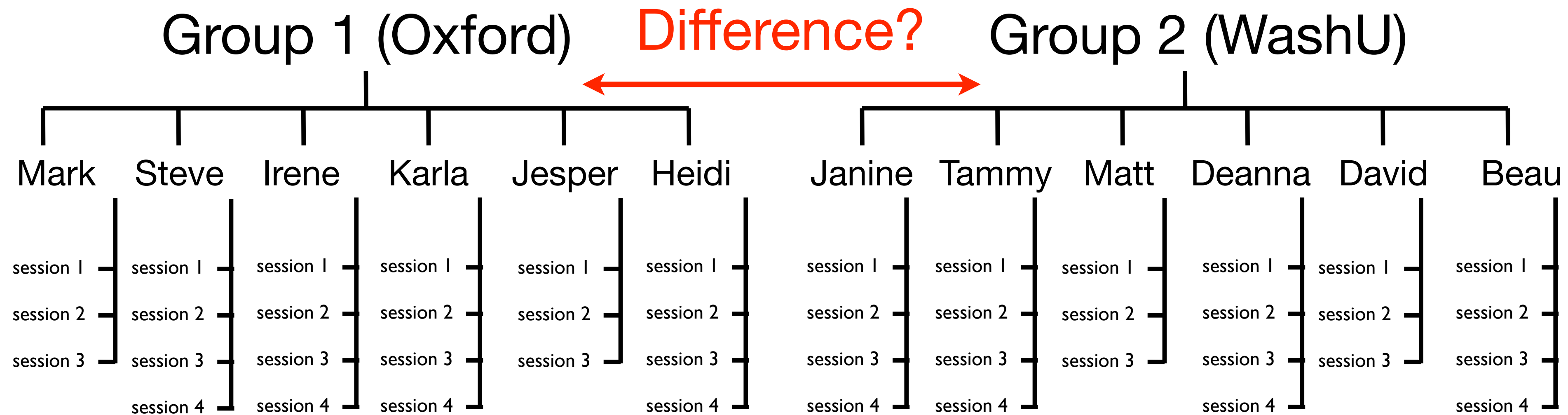
- Consider the 6 subjects as samples from a wider population
- estimate the mean across the population
- between-subject variance accounts for random sampling





Multiple sessions per subject

All-in-One Approach



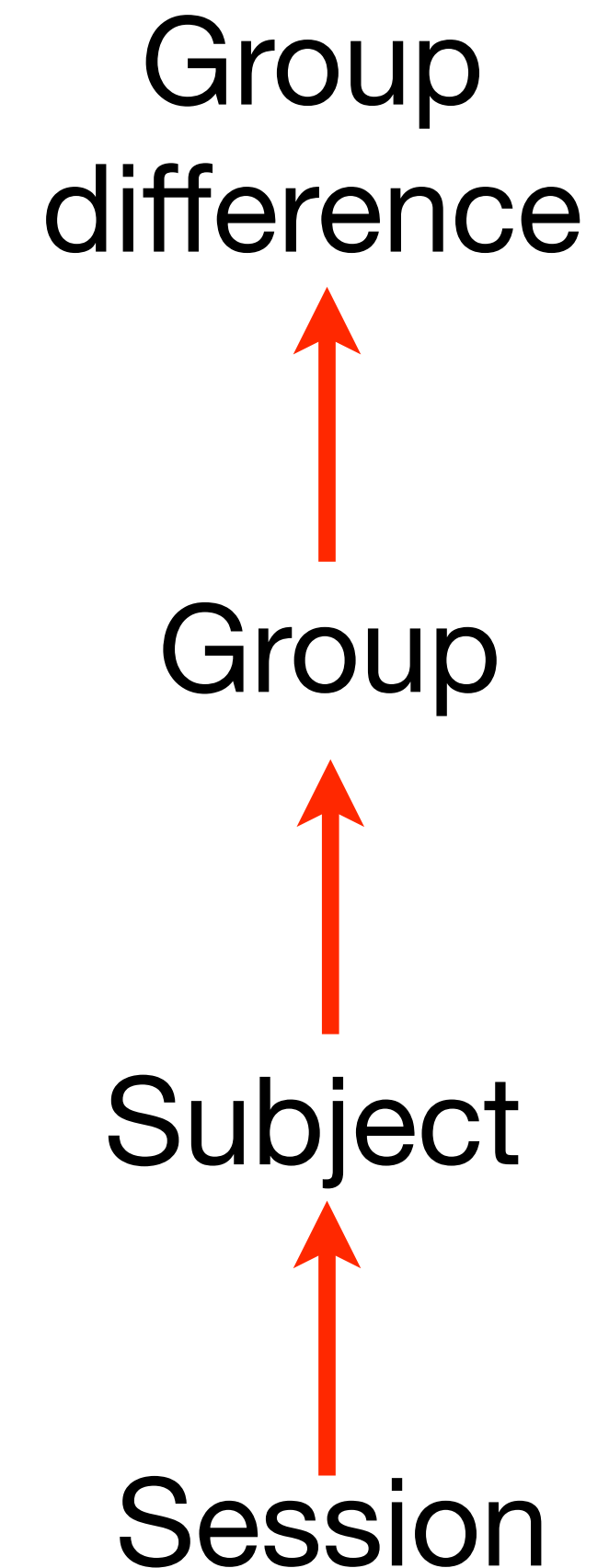
- Could use one (huge) GLM to infer group difference
 - difficult to ask sub-questions in isolation
 - computationally demanding
 - need to process again when new data is acquired



Summary Statistics Approach

In FEAT estimate levels one stage at a time

- At each level:
 - Inputs are summary stats from levels below (or FMRI data at the lowest level)
 - Outputs are summary stats or statistic maps for inference
- Need to ensure formal equivalence between different approaches!

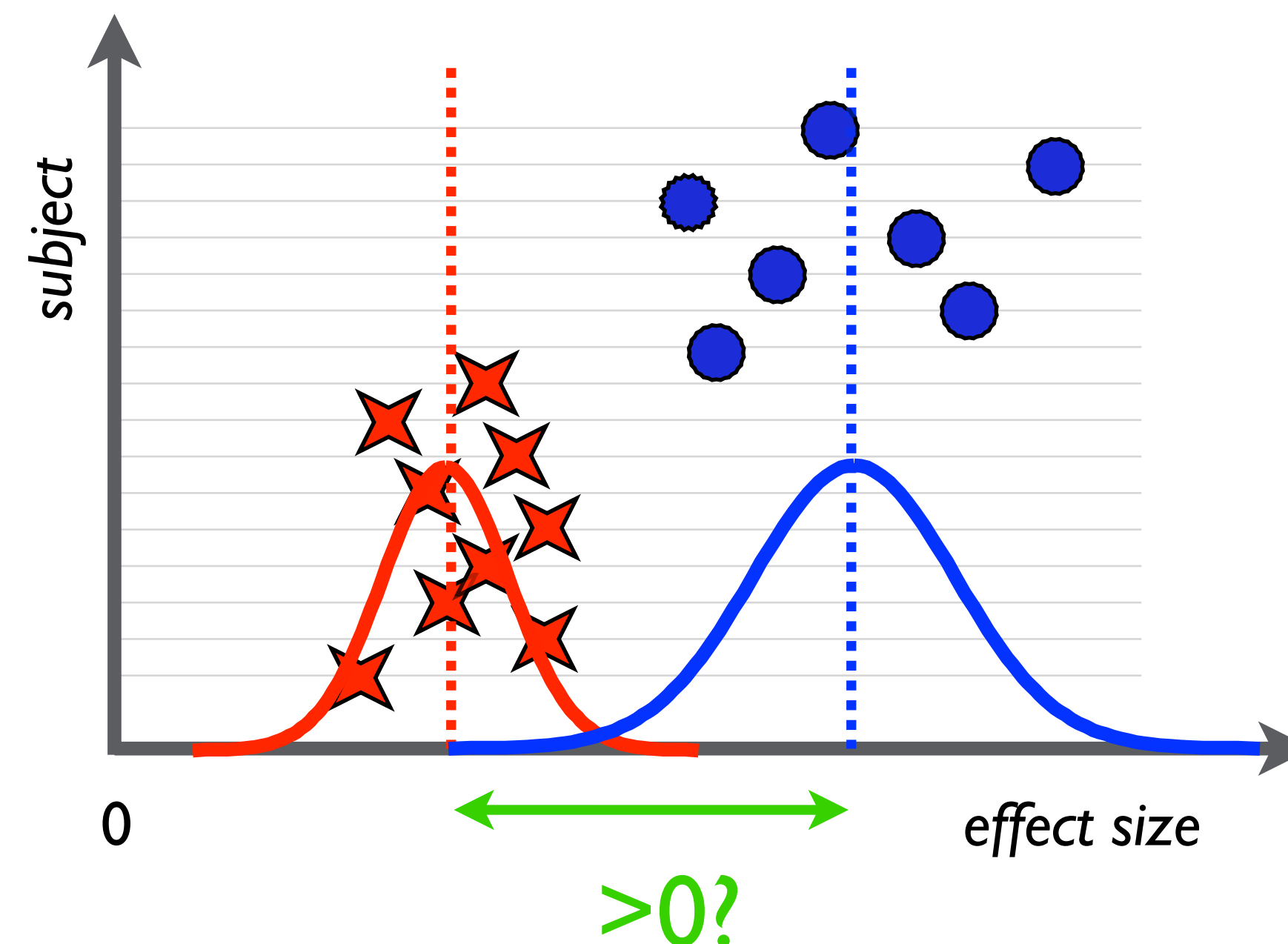


Unpaired Two-Group Difference

- We have two groups (e.g. 9 WashU, 7 Oxford) with different between-subject variance

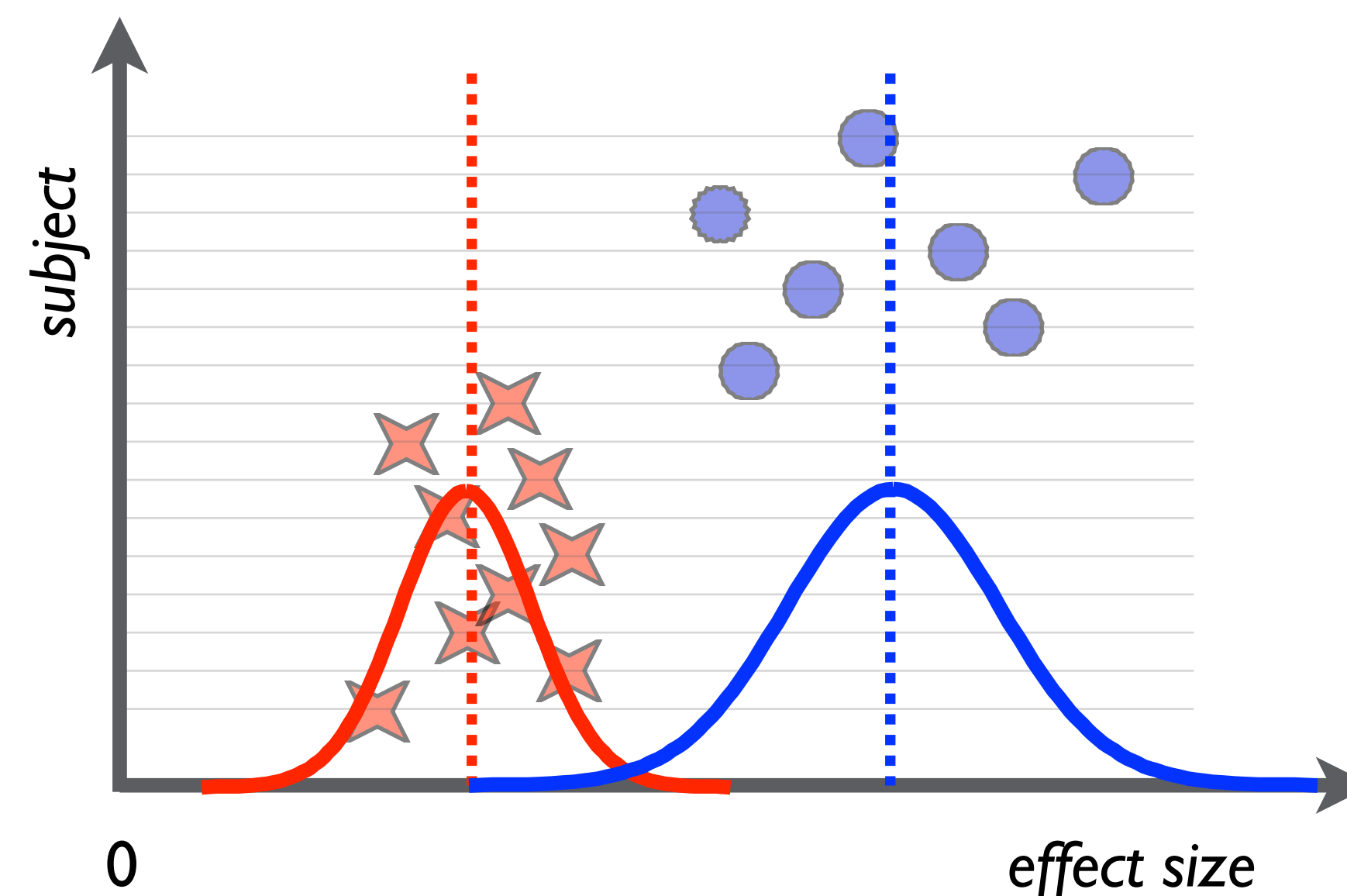
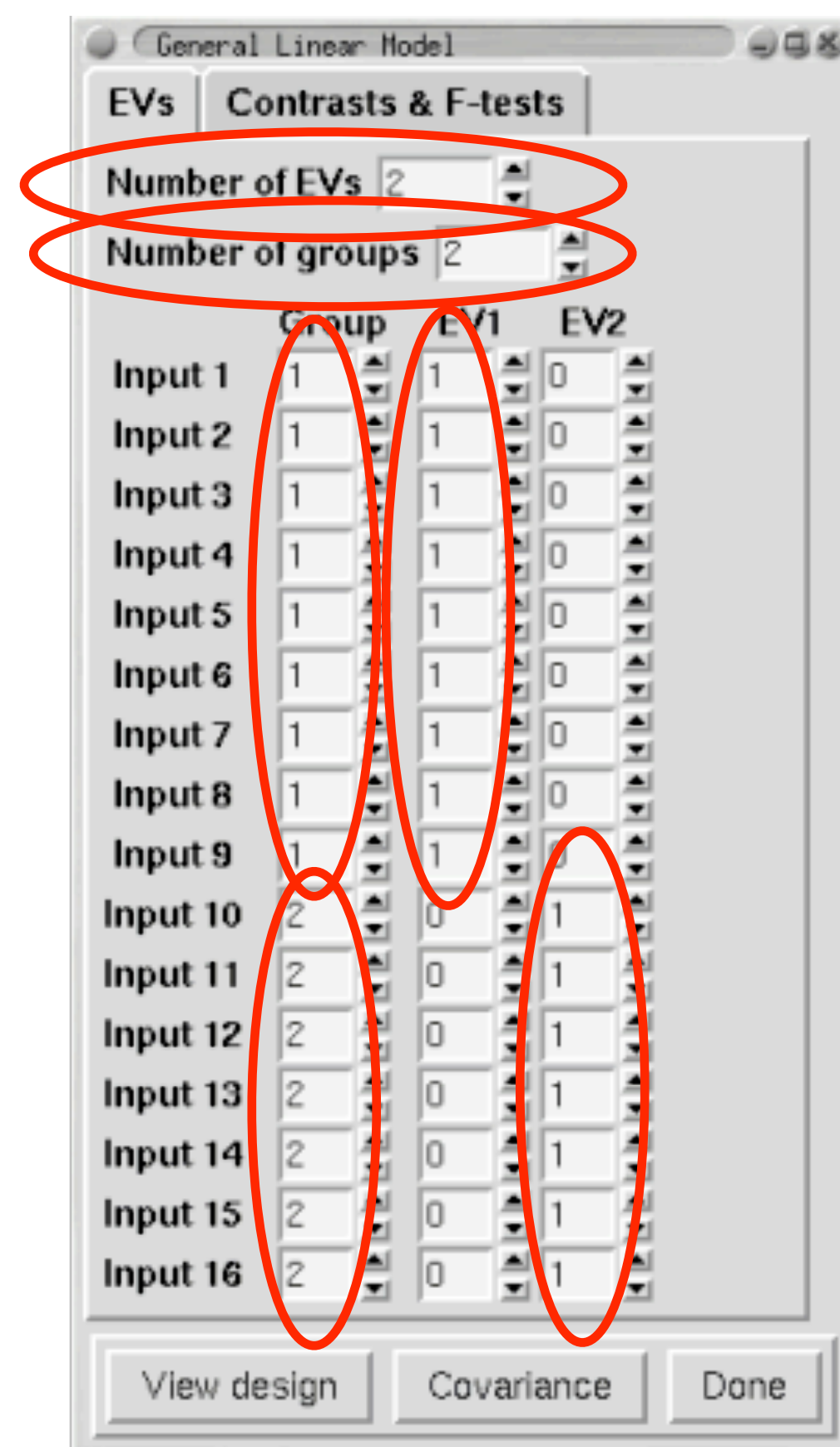
Is there a significant group difference?

- estimate means
- estimate std-errors (FE or ME)
- test significance of difference in means



Unpaired Two-Group Difference

Is there a significant group difference?



Unpaired Two-Group Difference

Is there a significant group difference?

General Linear Model

EVs Contrasts & F-tests

Number of EVs 2

Number of groups 2

	Group	EV1	EV2
Input 1	1	1	0
Input 2	1	1	0
Input 3	1	1	0
Input 4	1	1	0
Input 5	1	1	0
Input 6	1	1	0
Input 7	1	1	0
Input 8	1	1	0
Input 9	1	1	0
Input 10	2	0	1
Input 11	2	0	1
Input 12	2	0	1
Input 13	2	0	1
Input 14	2	0	1
Input 15	2	0	1
Input 16	2	0	1

View design Covariance Done

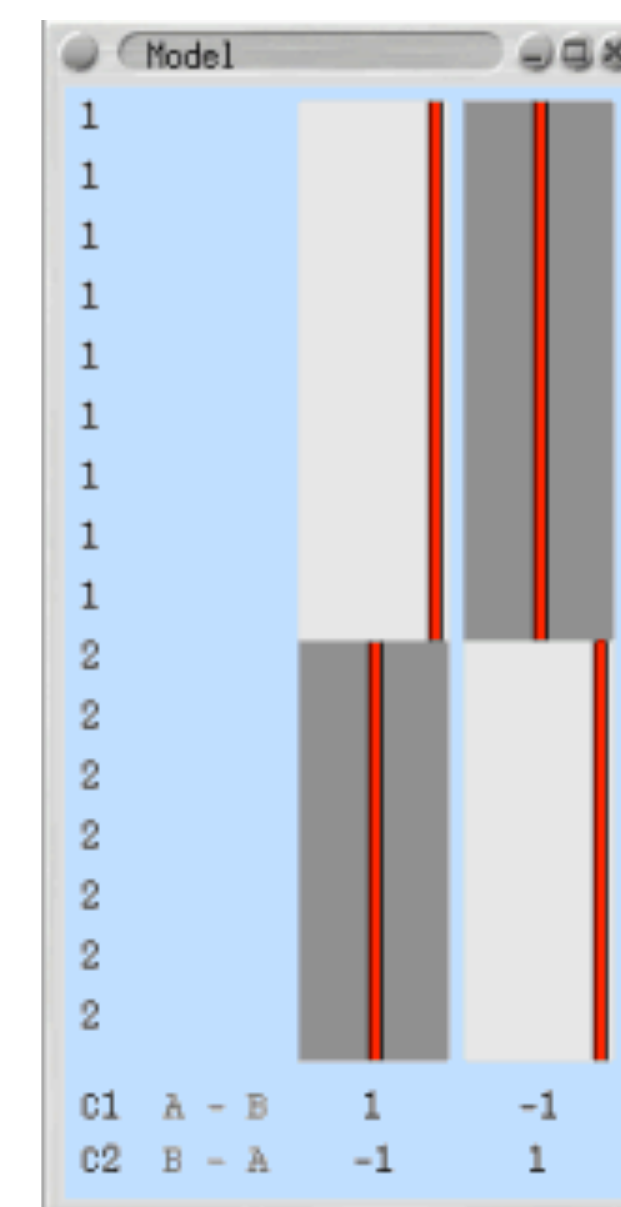
General Linear Model

EVs Contrasts & F-tests

Contrasts 2 F-tests 0

	Title	EV1	EV2
C1	A - B	1	-1
C2	B - A	-1	1

View design Covariance Done

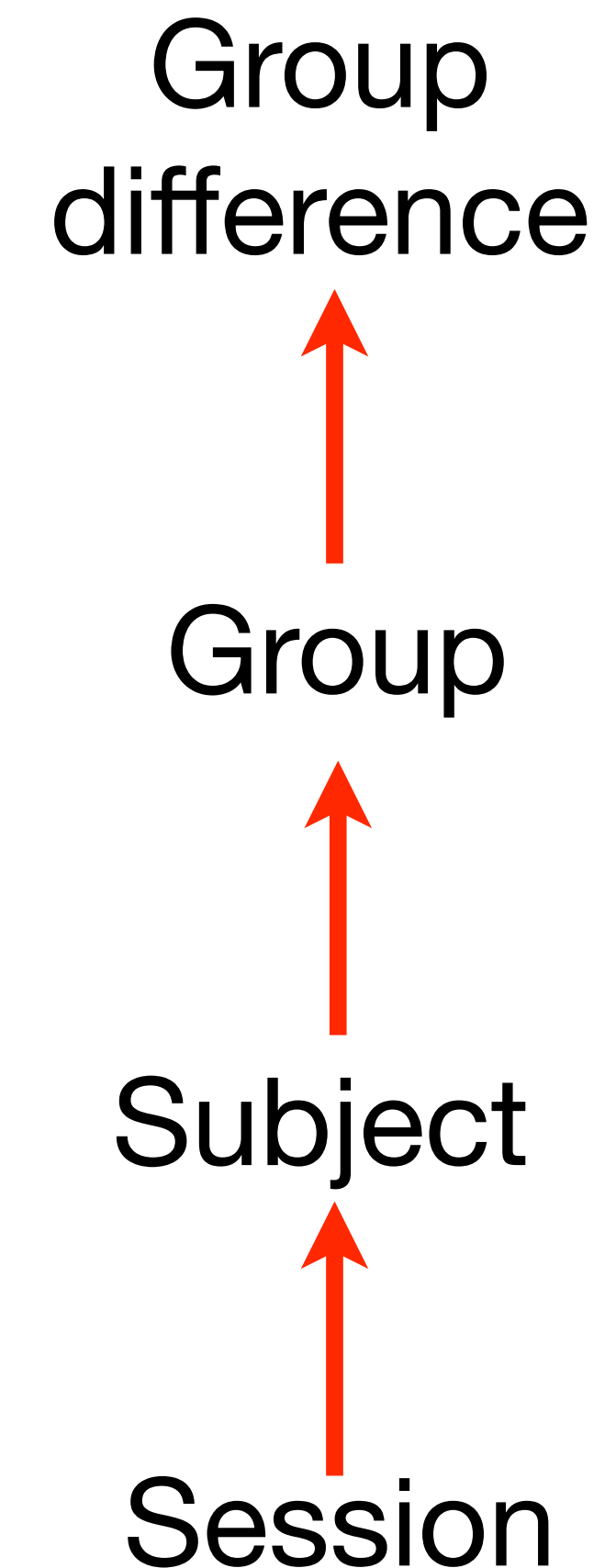




FLAME

FMRIB's Local Analysis of Mixed Effects

- Fully Bayesian framework
 - Input COPES, VARCOPEs & DOFs from lower-level
 - estimate COPES, VARCOPEs & DOFs at current level
 - pass these up
- Infer and threshold at top level (Z-stat)
- Equivalent to All-in-One approach





FLAME Inference

- Default is:
 - FLAME1: fast approximation for all voxels
- Optional slower, slightly more accurate approach:
 - FLAME1+2:
 - FLAME1 for all voxels, FLAME2 for voxels close to threshold
 - FLAME2: MCMC sampling technique



Choosing Inference Approach

1. Fixed Effects

Use for intermediate/top levels

2. Mixed Effects - OLS

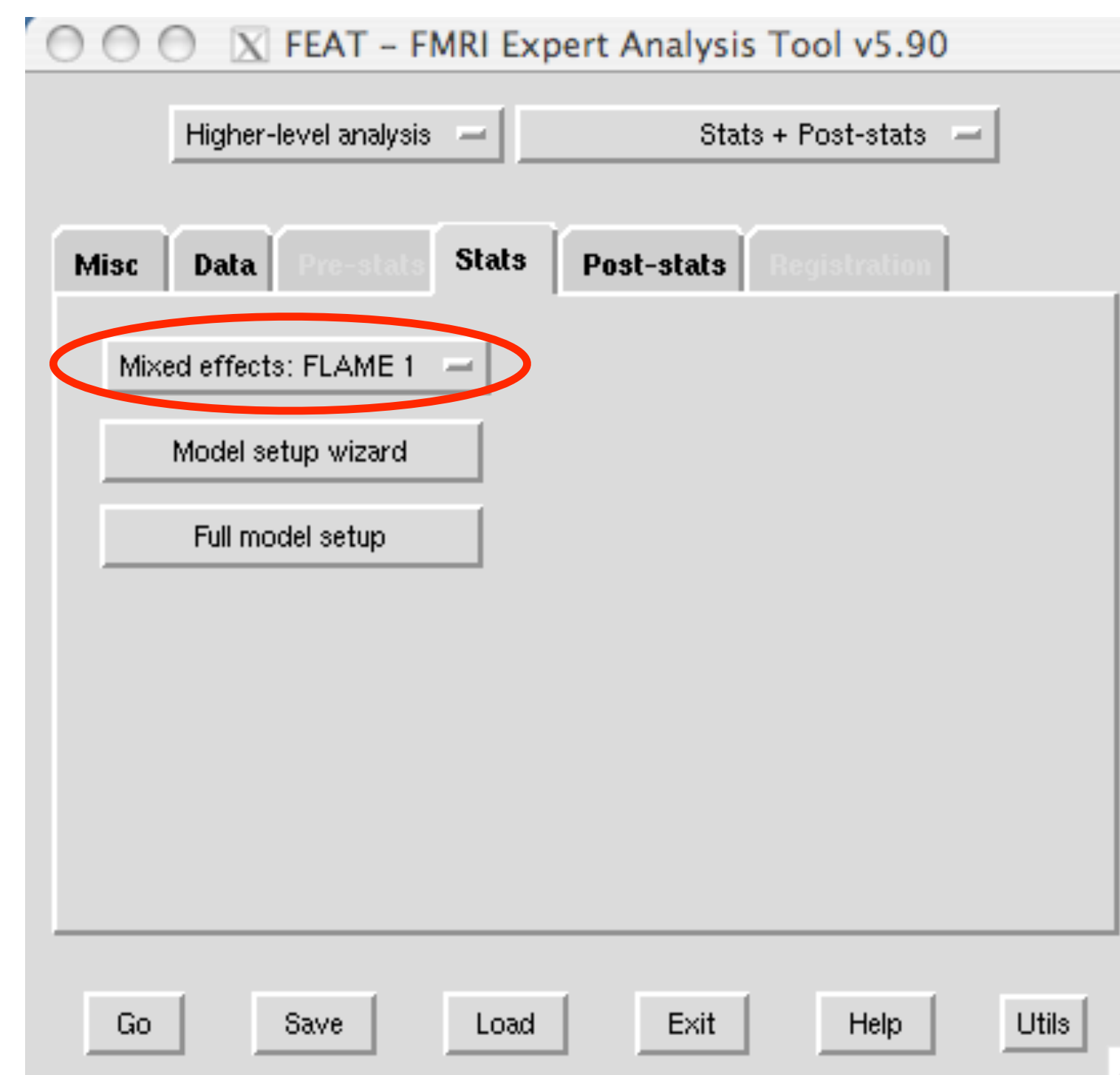
Use at top level: quick and less accurate

3. Mixed Effects - FLAME 1

Use at top level: less quick but more accurate

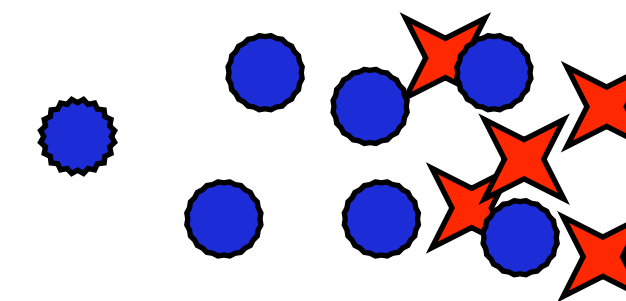
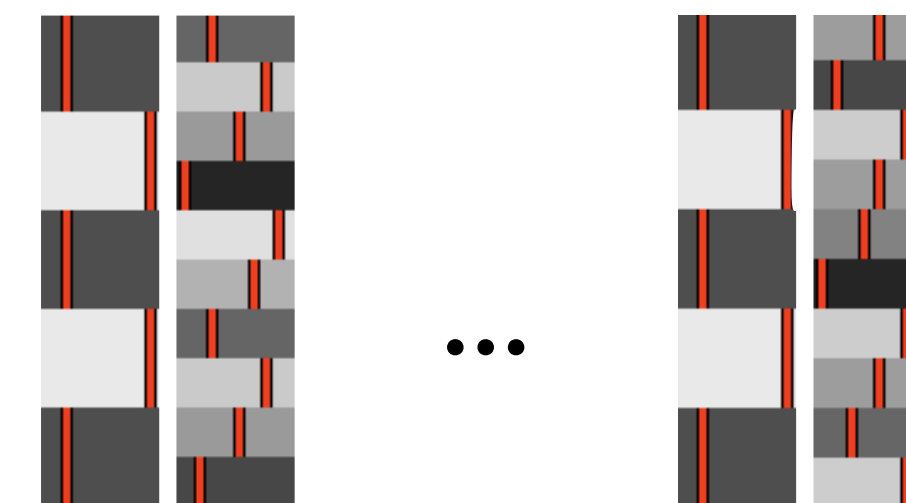
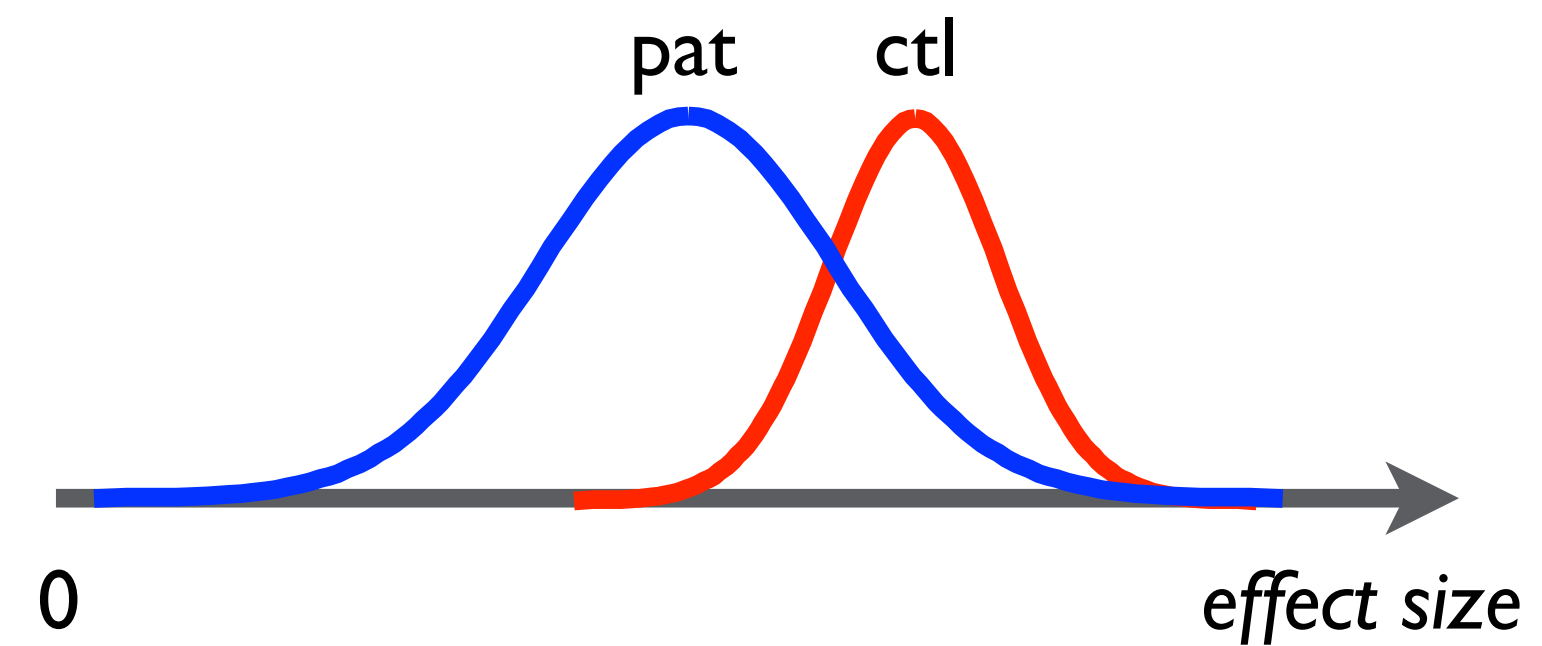
4. Mixed Effects - FLAME 1+2

Use at top level: slow but even more accurate



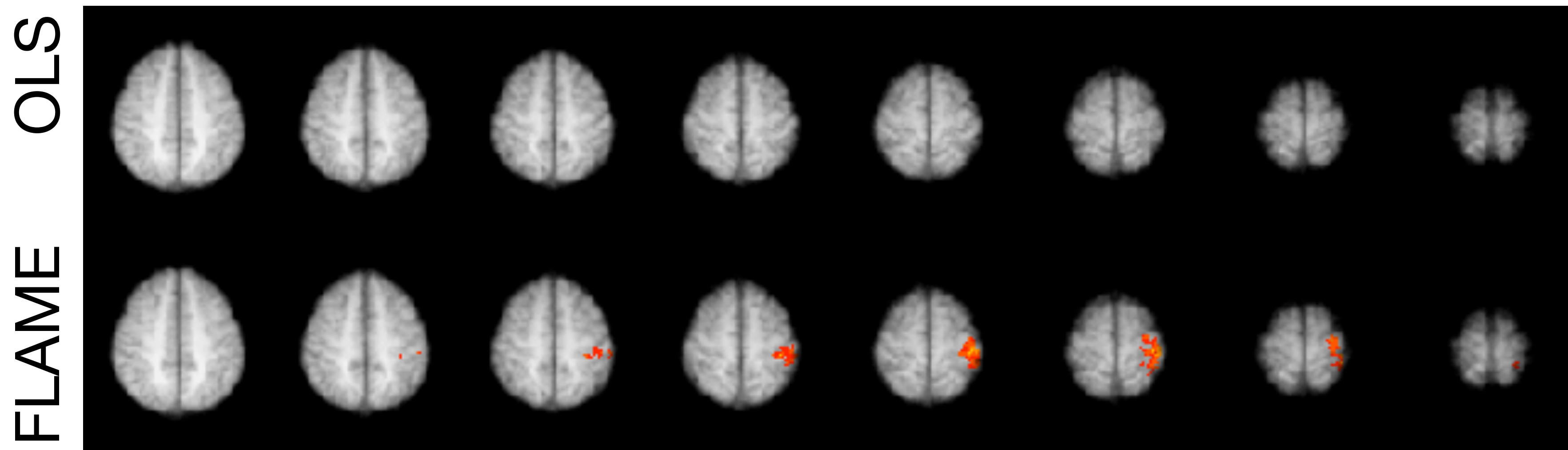
FLAME vs. OLS

- allow different within-level variances (e.g. patients vs. controls)
- allow non-balanced designs (e.g. containing behavioral scores)
- allow un-equal group sizes



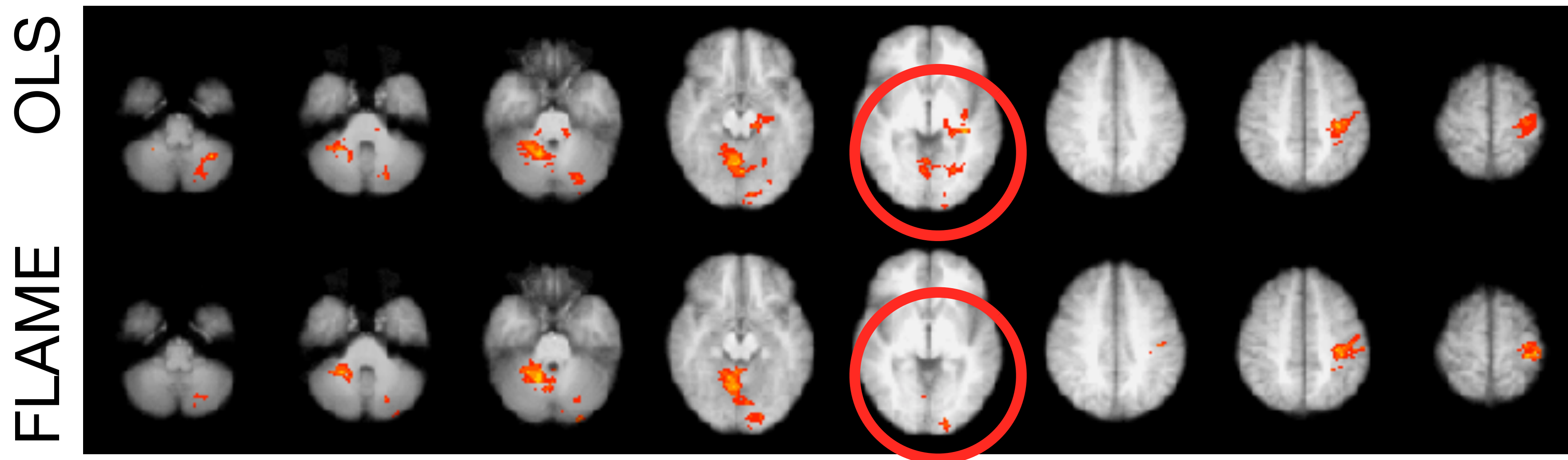
FLAME vs. OLS

- Two ways in which FLAME can give different Z-stats compared to OLS:
 - *higher Z* due to increased efficiency from using lower-level variance heterogeneity

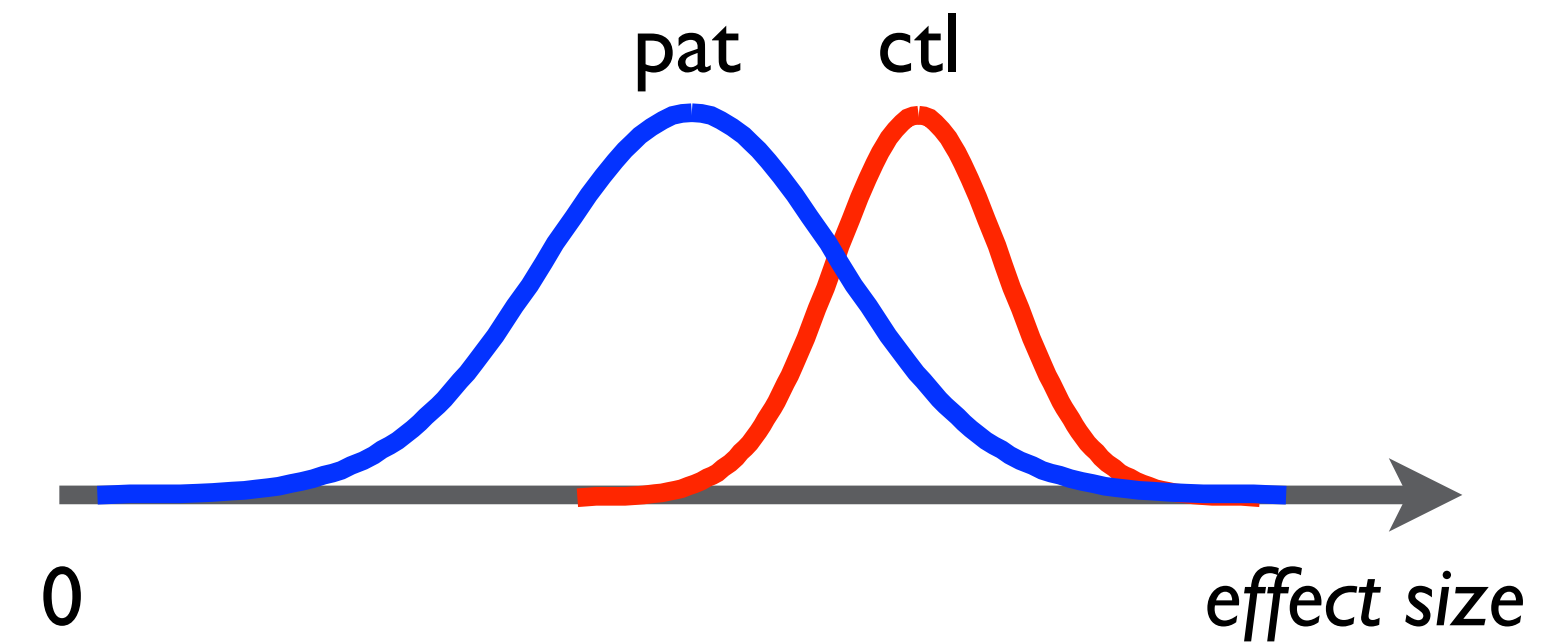


FLAME vs. OLS

- Two ways in which FLAME can give different Z-stats compared to OLS:
 - *Lower Z* due to higher-level variance being constrained to be positive (i.e. solve the implied negative variance problem)



Multiple Group Variances



- can deal with multiple group variances
- separate variance will be estimated for each variance group (be aware of #observations for each estimate, though!)
- EVs can only have non-zero values for a single group

1	1.0	0
1	1.0	0
1	1.0	0
2	0	1.0
2	0	1.0
2	0	1.0

✓

1	1.0	1.0
1	1.0	1.0
1	1.0	1.0
2	1.0	-1.0
2	1.0	-1.0
2	1.0	-1.0

✗

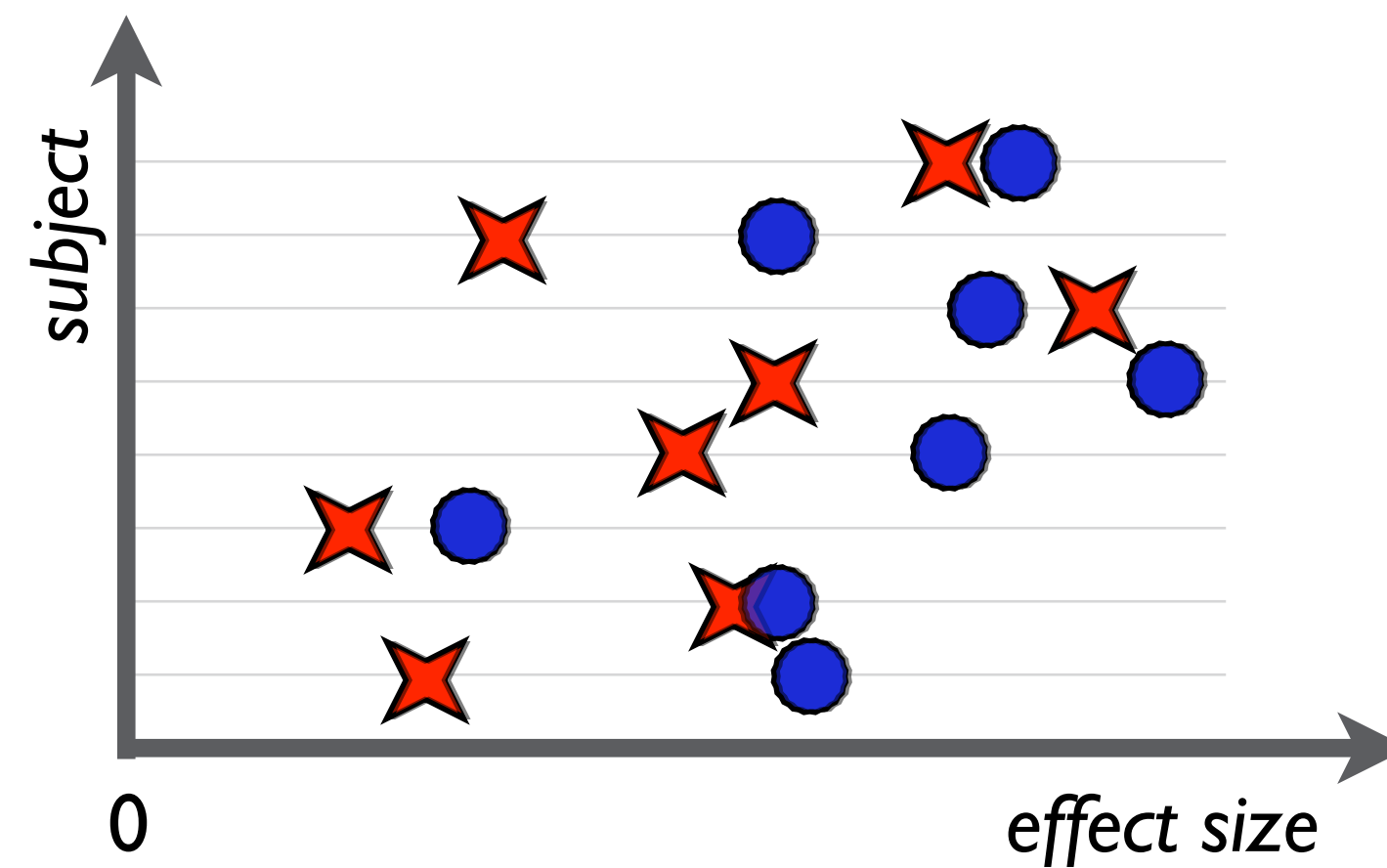


Paired T-test

Paired T-Test

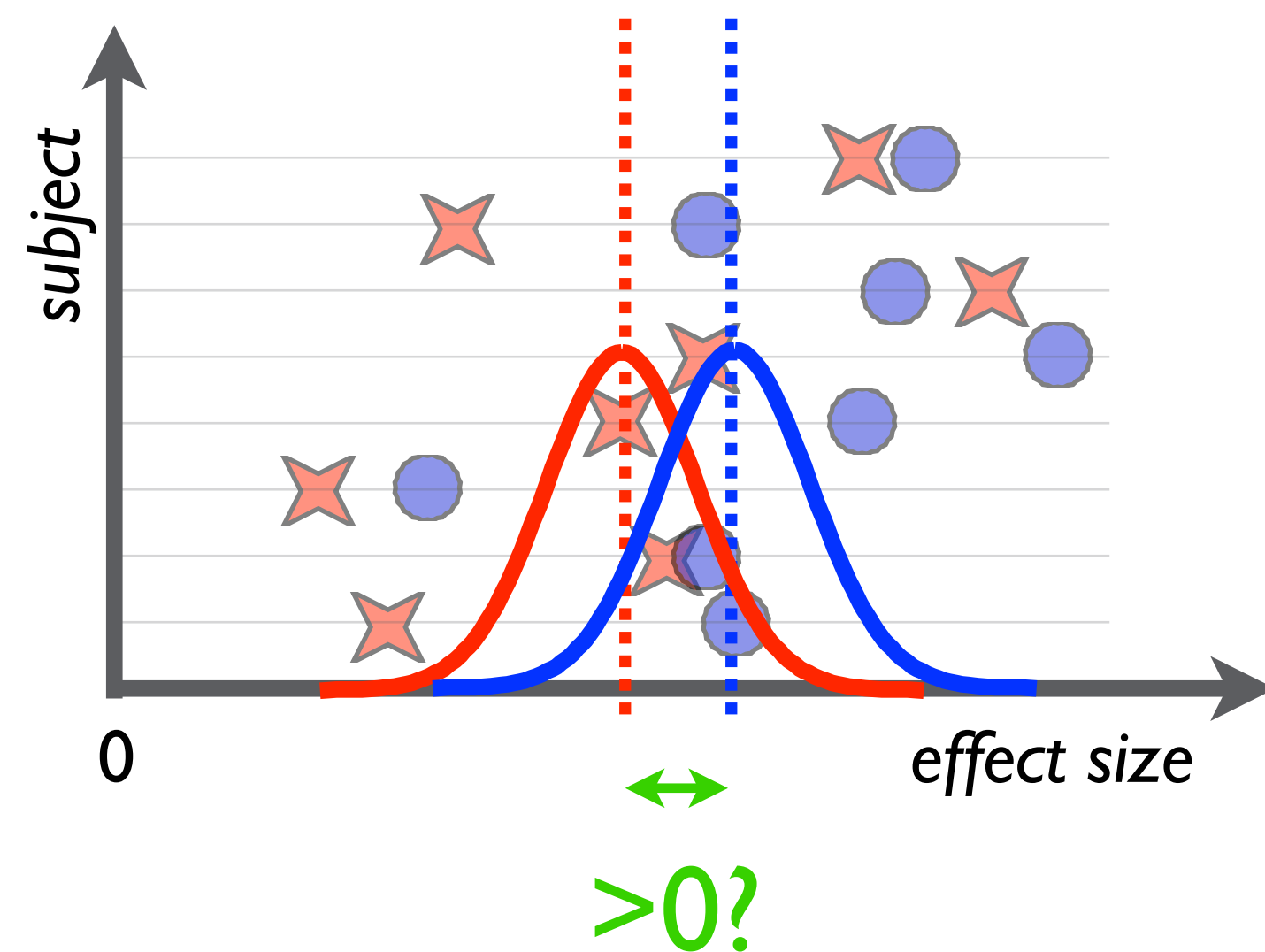
- 8 subjects scanned under 2 conditions (A,B)

Is there a significant difference between conditions?



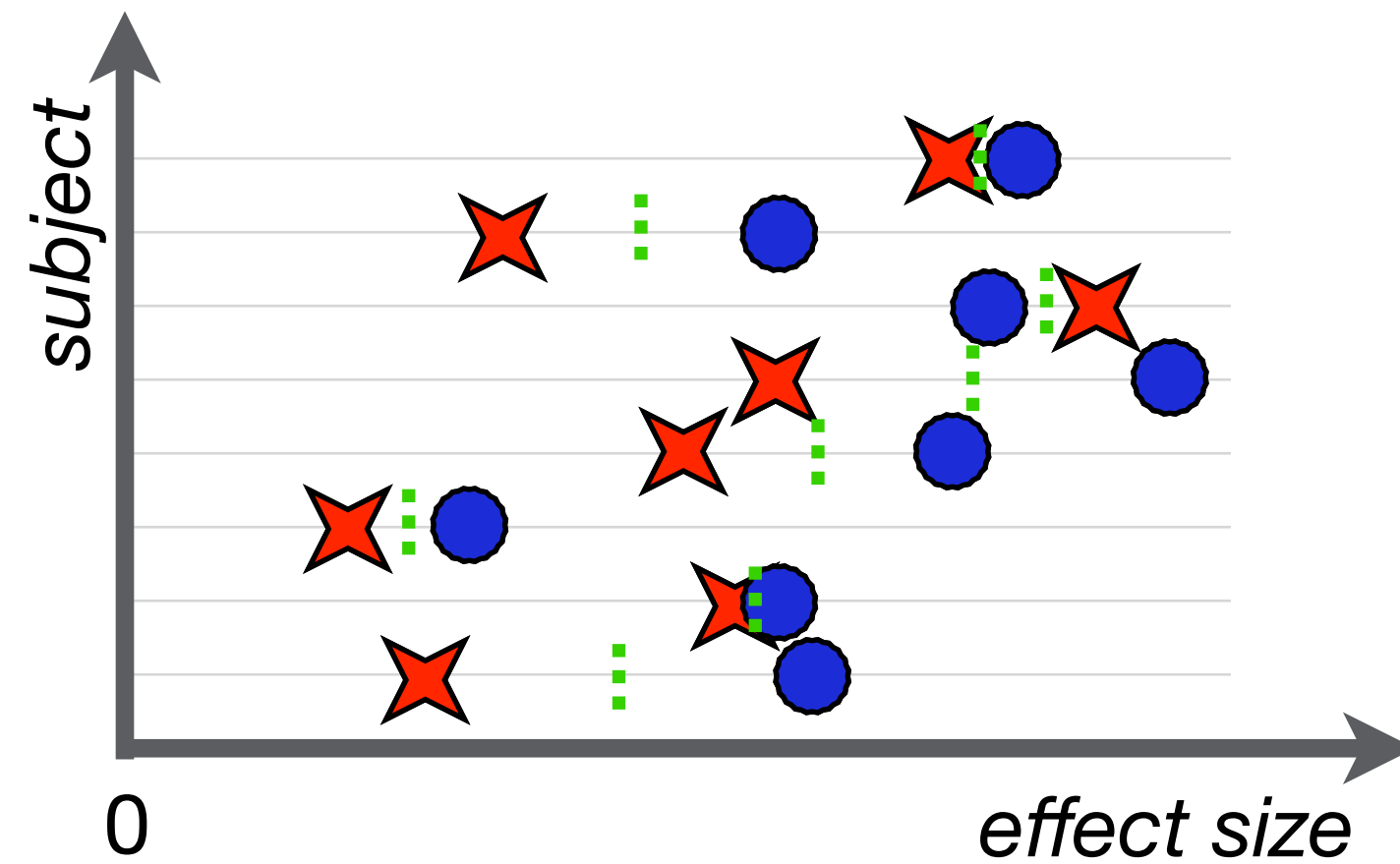
Paired T-Test

First, let's try an unpaired T-test

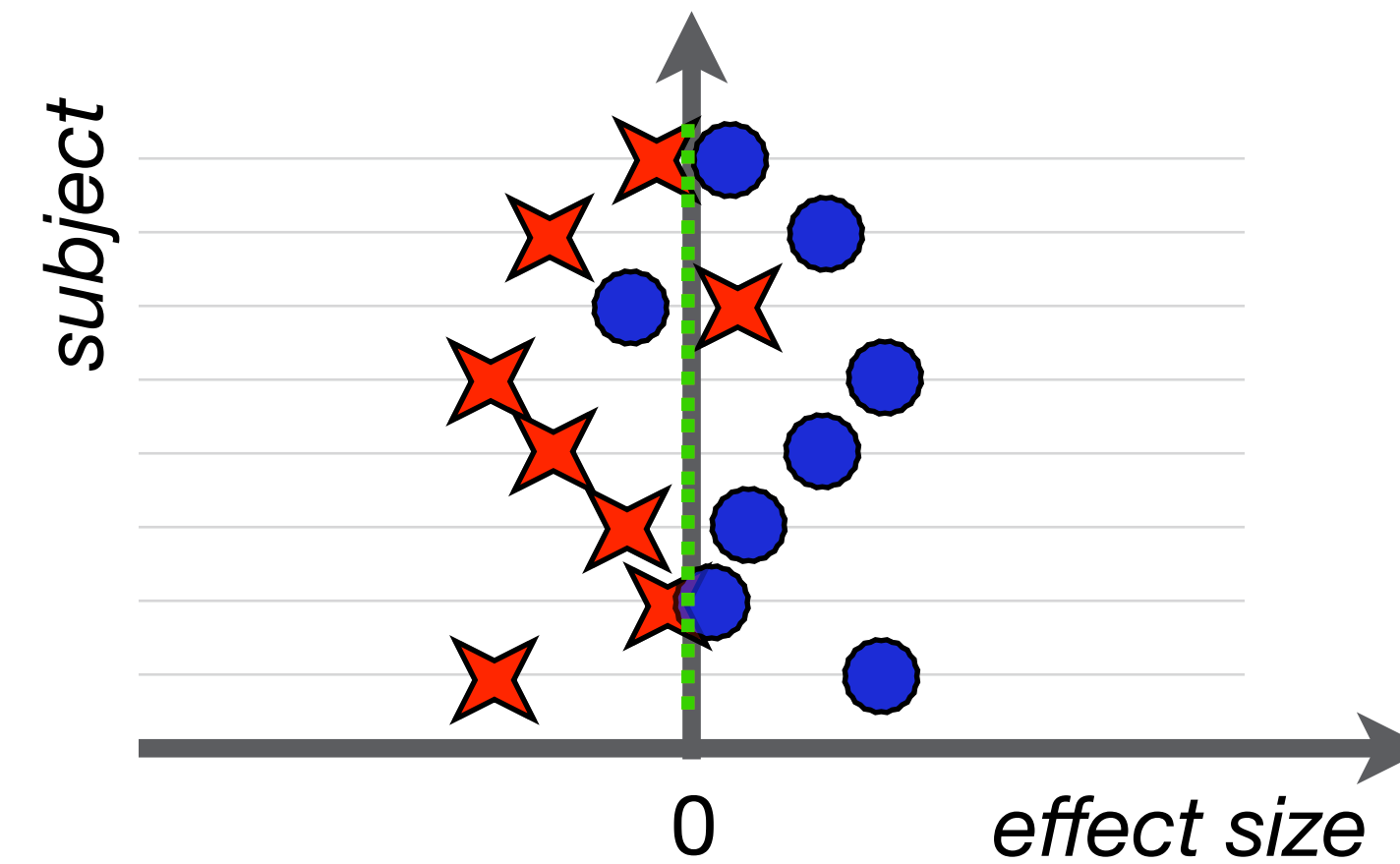


Paired T-Test

data



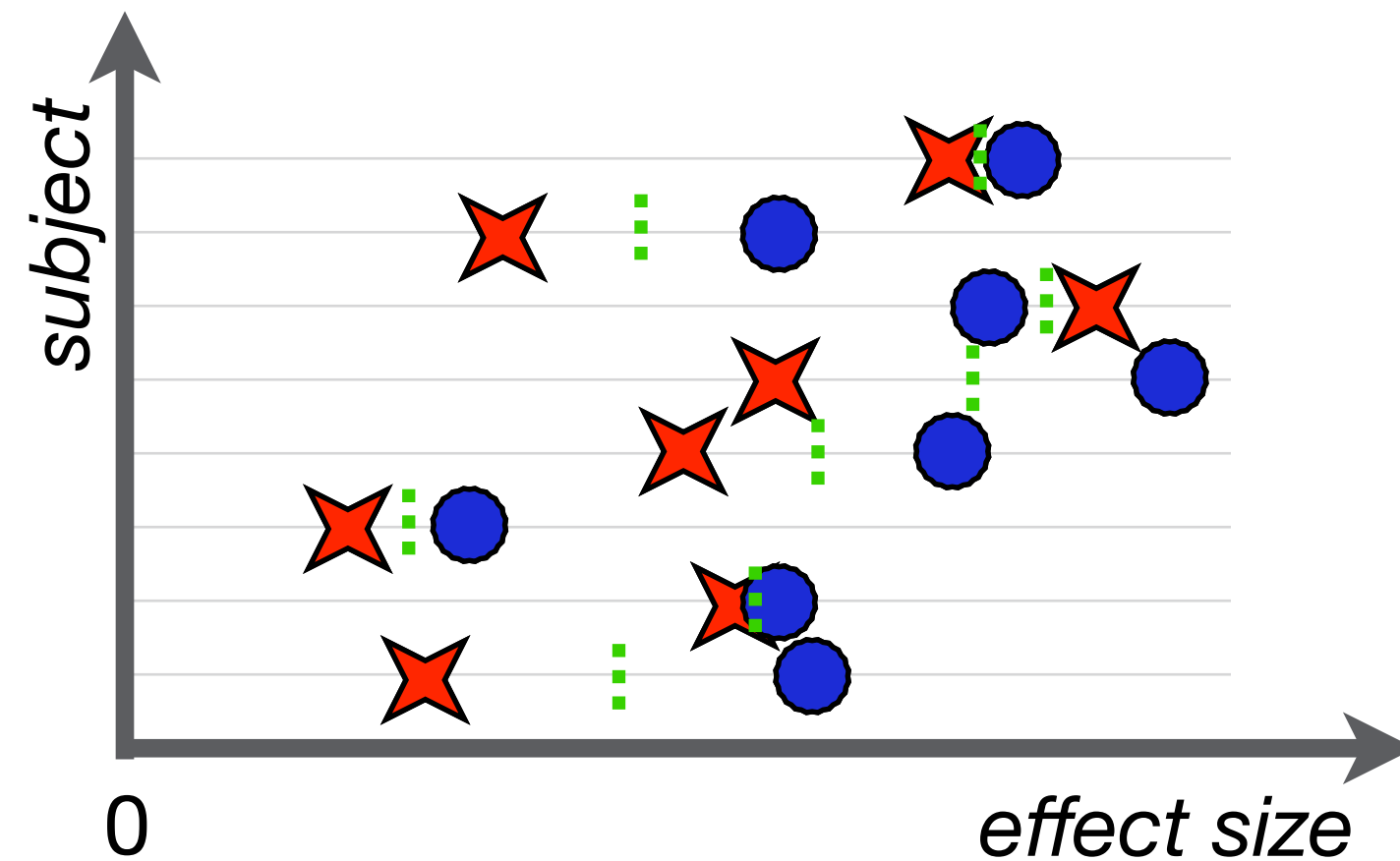
de-meaned data



subject mean
accounts for large prop.
of the overall variance

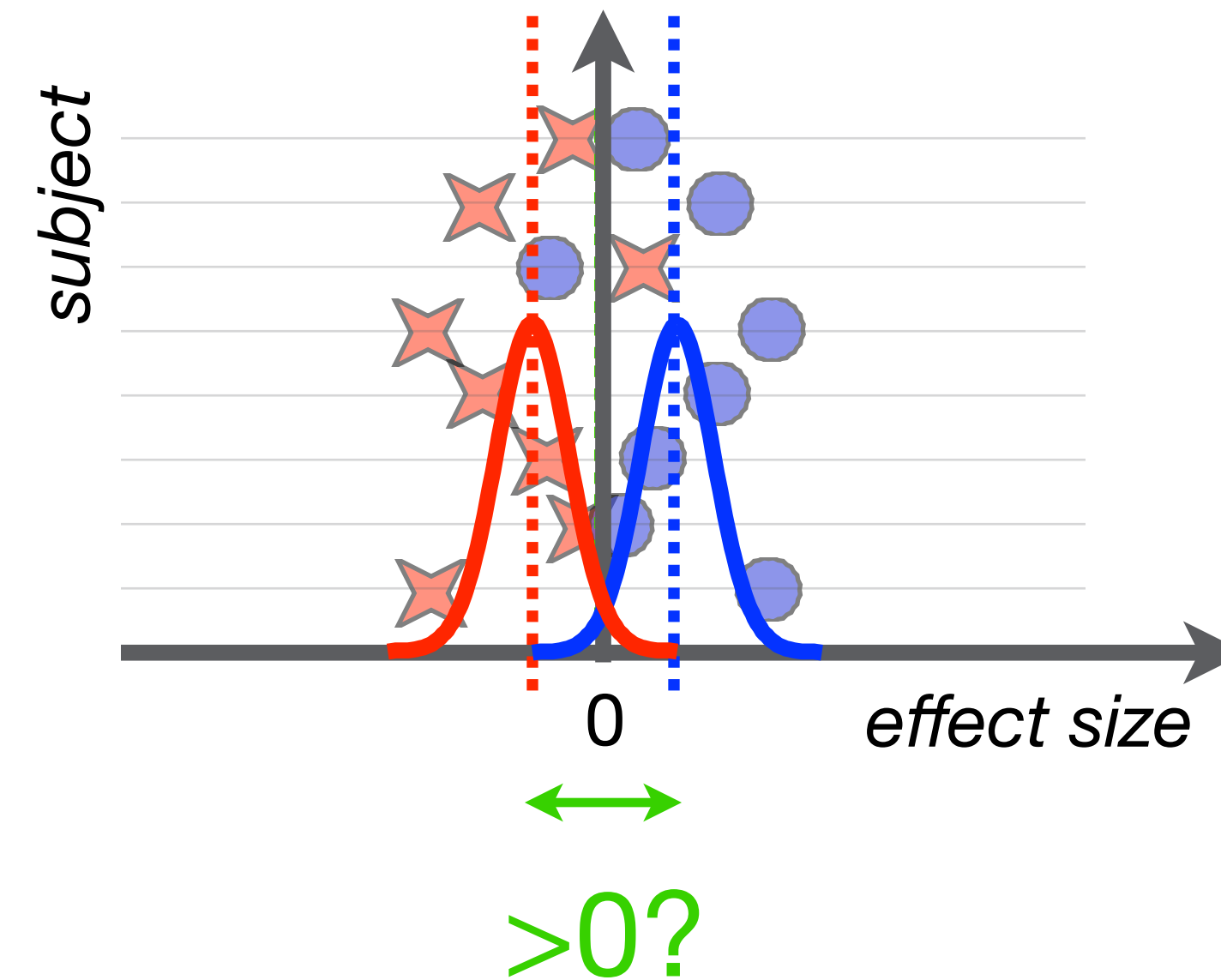
Paired T-Test

data

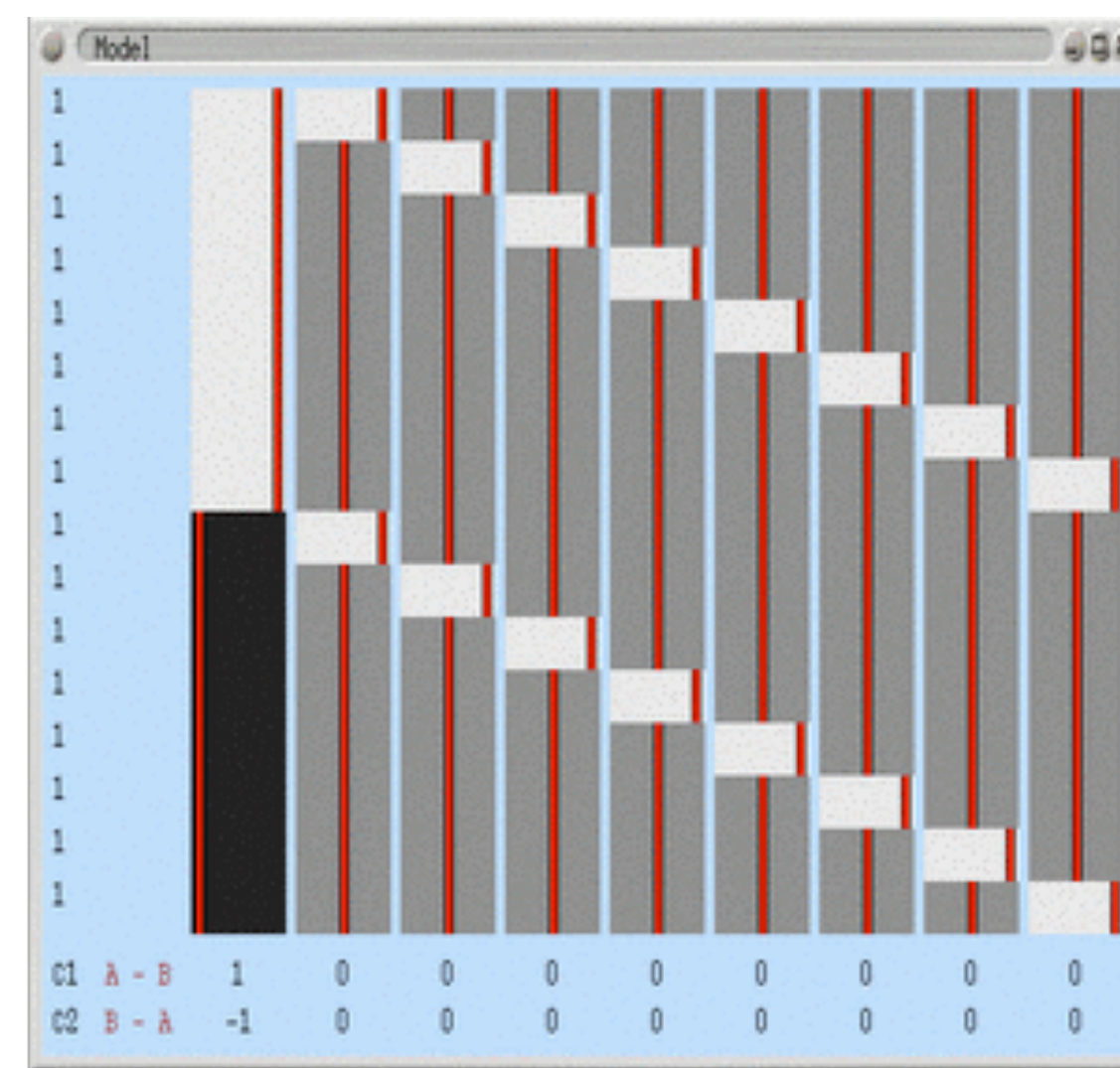


subject mean
accounts for large prop.
of the overall variance

de-meaned data



Model out each subject's mean

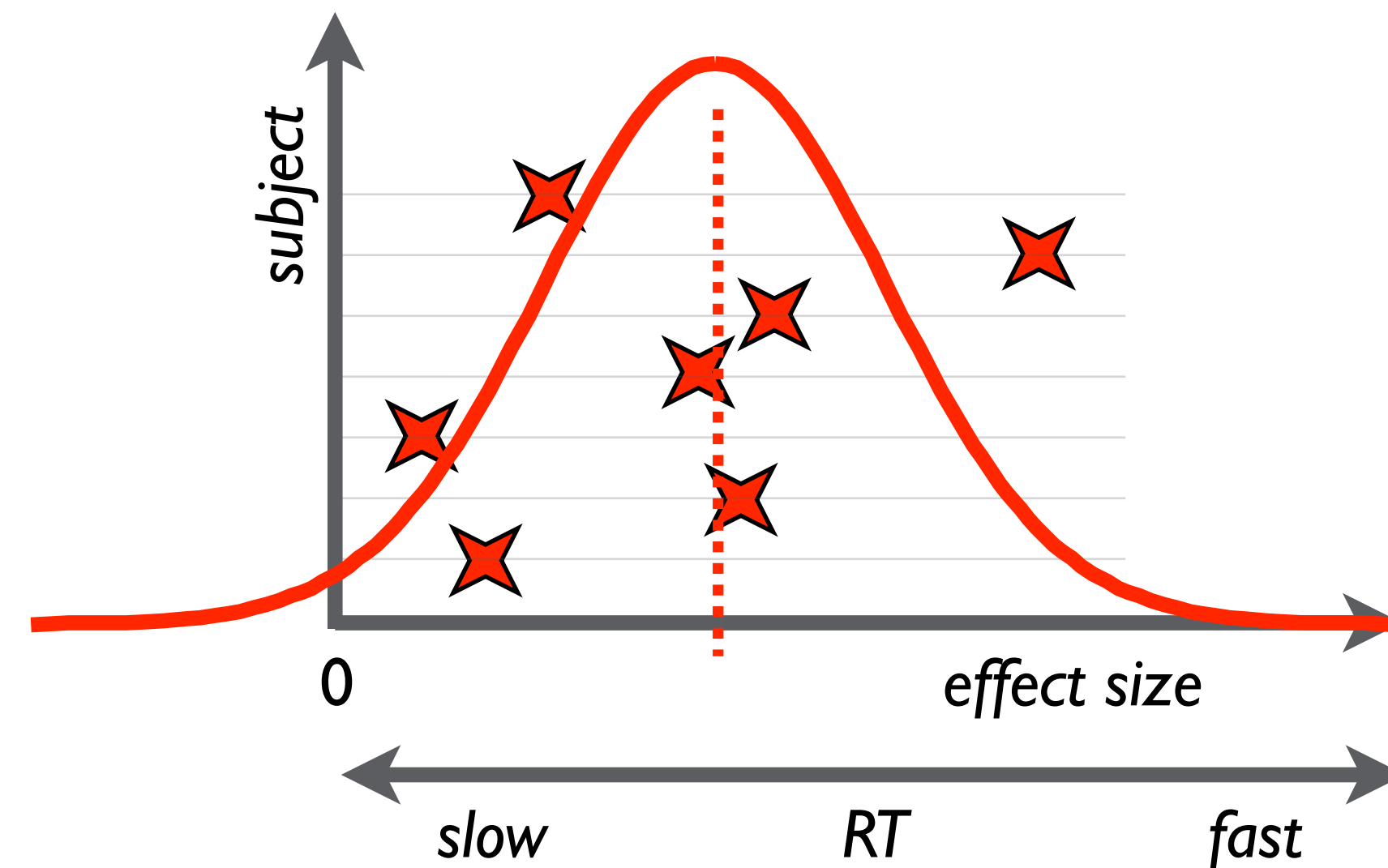




Group average with covariate

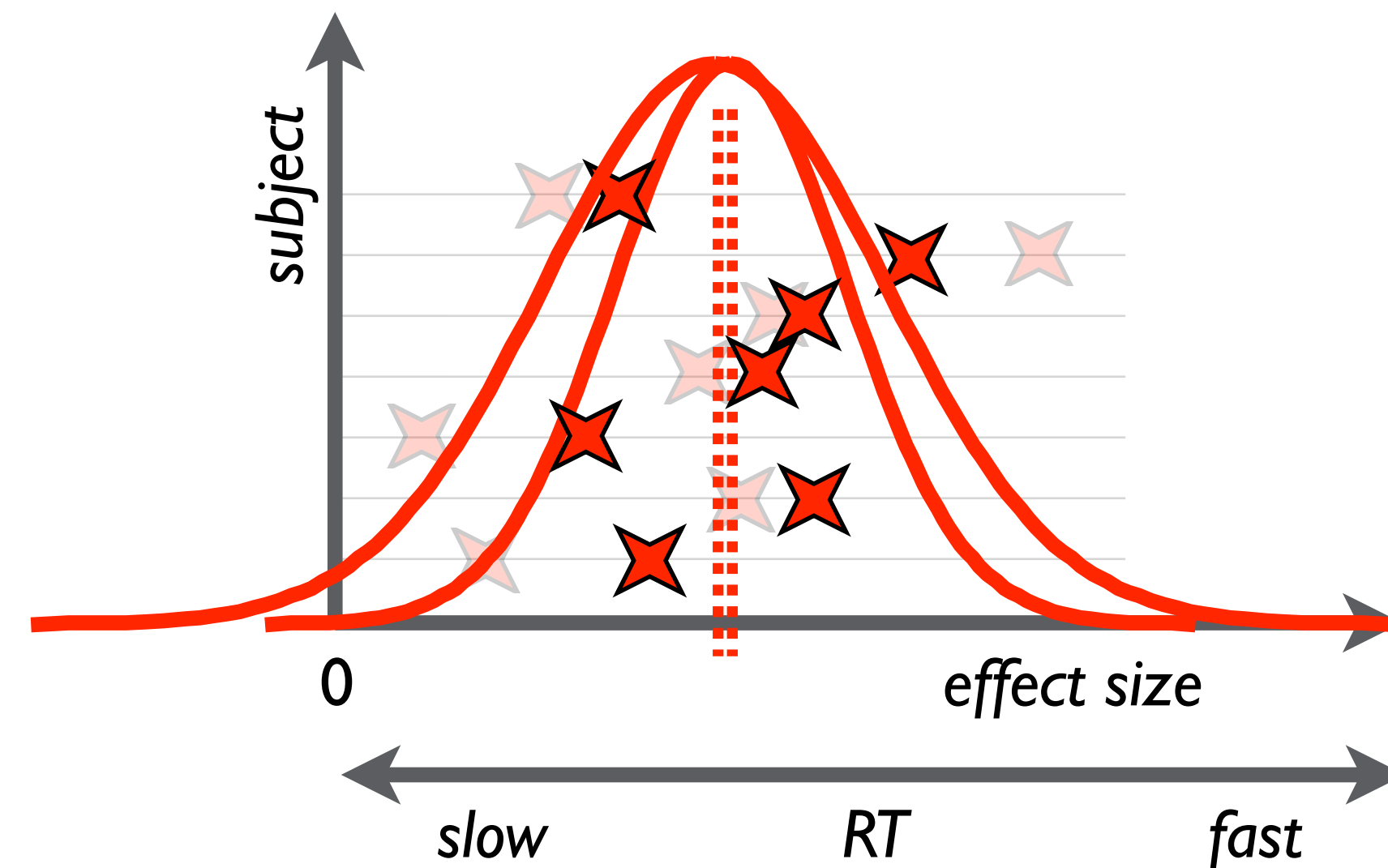
Group average with covariate

- Additional measurements (e.g. age; disability score; behavioral measures like reaction times)
- use covariates to 'explain' variation



Group average with covariate

- Additional measurements (e.g. age; disability score; behavioral measures like reaction times)
- use covariates to 'explain' variation



Group average with covariate

Need to demean covariates

General Linear Model

EVs Contrasts & F-tests

Number of EVs 2

Number of groups 1

	Group	EV1	EV2
Input 1	1	1	24
Input 2	1	1	-18
Input 3	1	1	-7
Input 4	1	1	5
Input 5	1	1	-4
Input 6	1	1	6
Input 7	1	1	-6

View design Covariance Done

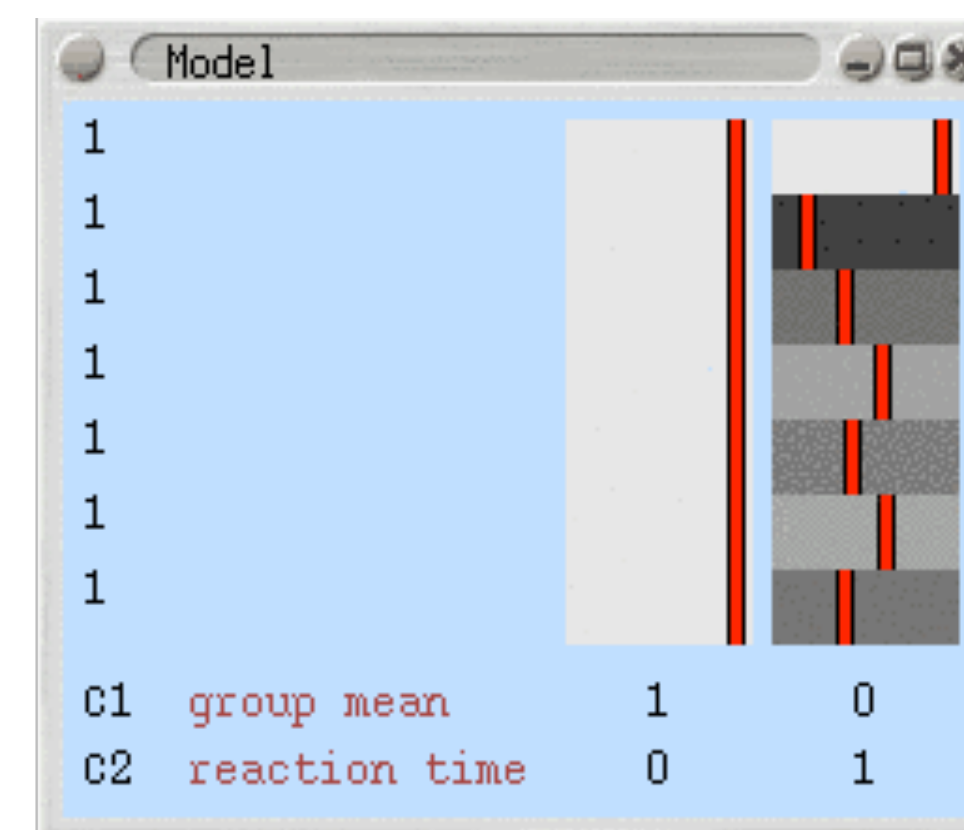
General Linear Model

EVs Contrasts & F-tests

Contrasts 2 F-tests 0

	Title	EV1	EV2
C1	group mean	1	0
C2	reaction time	0	1

View design Covariance Done



Break Time!

