

# Outline 2

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## 0.1 Step1:

To calculate the max purity function, first, fit the linear mixed model for the outcome  $y$  and time  $\mathbf{X}$ , with baseline covariates  $\mathbf{x}$ :

$$\mathbf{y} = \mathbf{X}(\boldsymbol{\beta} + \mathbf{b} + \boldsymbol{\Gamma}(\boldsymbol{\alpha}'\mathbf{x})) + \boldsymbol{\epsilon}.$$

We can define the covariate matrix of  $\mathbf{X}$  as  $\mathbf{z}$ . The  $\mathbf{z}$  contains both fixed effects and random effects.

$$\mathbf{z} = \boldsymbol{\beta} + \mathbf{b} + \boldsymbol{\Gamma}\mathbf{x}$$

We can also write the above equation in the matrix version:

$$Y = \begin{bmatrix} 1 & t_1 & t_1^2 \\ 1 & t_2 & t_2^2 \\ 1 & t_3 & t_3^2 \\ 1 & t_4 & t_4^2 \end{bmatrix} \left[ \begin{pmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{pmatrix} + \begin{pmatrix} b_0 \\ b_1 \\ b_2 \end{pmatrix} + \begin{pmatrix} \gamma_0 \\ \gamma_1 \\ \gamma_2 \end{pmatrix} x \right] + \boldsymbol{\epsilon}$$

where

- \*  $\boldsymbol{\beta}$  is the vector of covariates for fixed effects of  $\mathbf{X}$ .
- \*  $\mathbf{b}$  is the vector of random effects
- \*  $\boldsymbol{\Gamma}\mathbf{x}$  is the vector of fixed effects of the baseline covariates.
- \*  $x$  is the combination of the input baseline covariates.  $x = \boldsymbol{\alpha}'\mathbf{x}$  and  $\mathbf{x}$  is the covariates vector.

## 0.2 Step2:

Then we need to estimate the distribution of  $\underline{z}$  for drug group and placebo, separately.

$$\begin{aligned} f(z) &= \int_x f(z, x) dx \\ &= \int_x f(z|x) g(x) dx \end{aligned}$$

where,

\* the conditional distribution  $f(z|x) \sim MVN(\boldsymbol{\beta} + \boldsymbol{\Gamma}x, \boldsymbol{D})$

\*  $g(x)$  is the distribution of the covariates combination. For example, if the covariates combination only contains "sex", which is binary, then the integral becomes summation.

\*  $\boldsymbol{D}$  is the covariates matrix of random effects  $\boldsymbol{b}$ .

We could then fit the  $f(\cdot)$  for drug group and placebo group separately, i.e.  $f_1(z)$  and  $f_2(z)$ .

## 0.3 Step3:

Define the purity for a combination of covariates  $x$  as

$$f_1(z|x) = |2\pi D|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(\boldsymbol{z} - \hat{\boldsymbol{\beta}}_1 - \hat{\boldsymbol{\Gamma}}_1 x)^T D^{-1}(\boldsymbol{z} - \hat{\boldsymbol{\beta}}_1 - \hat{\boldsymbol{\Gamma}}_1 x)\right)$$

$$f_2(z|x) = |2\pi D|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(\boldsymbol{z} - \hat{\boldsymbol{\beta}}_1 - \hat{\boldsymbol{\Gamma}}_1 x)^T D^{-1}(\boldsymbol{z} - \hat{\boldsymbol{\beta}}_1 - \hat{\boldsymbol{\Gamma}}_1 x)\right)$$

$$\int_x \left( \frac{\pi_1 f_1(z|x) - \pi_2 f_2(z|x)}{\pi_1 f_1(z|x) + \pi_2 f_2(z|x)} \right)^2 g(x) dx$$

which can be also approximated as the summation form:

$$\sum_x \left( \frac{\pi_1 f_1(z|x) - \pi_2 f_2(z|x)}{\pi_1 f_1(z|x) + \pi_2 f_2(z|x)} \right)^2 g(x)$$

This is one purity value based on one combination of baseline covariates. Define the purity as  $P_x$

We would like to find the  $\boldsymbol{\alpha}$  that max the  $P_x$