

Likelihood comparison

If the true model is

$$\begin{aligned} y_1 &= X(\beta_1 + \Gamma_1(\alpha'_1 x)) + Xb_1 + \epsilon \\ y_2 &= X(\beta_2 + \Gamma_2(\alpha'_2 x)) + Xb_2 + \epsilon \end{aligned} \quad (1)$$

If we assume they share the same α , the two models are

$$\begin{aligned} y_1 &= X(\beta_1 + \Gamma_1(\alpha' x)) + Xb_1 + \epsilon \\ y_2 &= X(\beta_2 + \Gamma_2(\alpha' x)) + Xb_2 + \epsilon \end{aligned} \quad (2)$$

df = 3

Let's compare the log likelihood calculated from the LME, by using the $\hat{\alpha}$ estimated from

- purity method 1 (coefficient),
- purity method 2 (outcome),
- likelihood method 1 ($f(\theta; y|x)$),
- likelihood method 2 ($f(\theta; y)$),
- likelihood method 3 ($E(\theta)$)

Besides, if we simulate the drug and placebo dataset with different α , let's test whether there are significant difference between estimating with one shared α or not.

```
fit_pbo_est_full = lmer(y ~ tt + I(tt^2) + w_pbo + w_pbo * tt + w_pbo * I(tt^2) +
  (tt + I(tt^2) |subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_full = lmer(y ~ tt + I(tt^2) + w_drg + w_drg * tt + w_drg * I(tt^2) +
  (tt + I(tt^2) |subj),
  data = dat_drg, REML = FALSE)
```

```
l0 = summary(fit_pbo_est_full)$logLik + summary(fit_drg_est_full)$logLik
l0
```

```
## 'log Lik.' -6094.255 (df=13)
```

```
fit_pbo_est_0 = lmer(y ~ tt + I(tt^2) + w0 + w0 * tt + w0 * I(tt^2) +
  (tt + I(tt^2) |subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_0 = lmer(y ~ tt + I(tt^2) + w0 + w0 * tt + w0 * I(tt^2) +
  (tt + I(tt^2) |subj),
  data = dat_drg, REML = FALSE)
```

```
l00 = summary(fit_pbo_est_0)$logLik + summary(fit_drg_est_0)$logLik
l00
```

```
## 'log Lik.' -6098.291 (df=13)
```

```
fit_pbo_est_1 = lmer(y ~ tt + I(tt^2) + W1 + W1 * tt + W1 * I(tt^2) +
  (tt + I(tt^2) |subj),
  data = dat_pbo, REML = FALSE)
```

```

fit_drg_est_1 = lmer(y ~ tt + I(tt^2) + W1 + W1 * tt + W1 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_drg, REML = FALSE)
fit_pbo_est_2 = lmer(y ~ tt + I(tt^2) + W2 + W2 * tt + W2 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_2 = lmer(y ~ tt + I(tt^2) + W2 + W2 * tt + W2 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_drg, REML = FALSE)

## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00203211 (tol = 0.002, component 1)
fit_pbo_est_3 = lmer(y ~ tt + I(tt^2) + W3 + W3 * tt + W3 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_3 = lmer(y ~ tt + I(tt^2) + W3 + W3 * tt + W3 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_drg, REML = FALSE)
fit_pbo_est_4 = lmer(y ~ tt + I(tt^2) + W4 + W4 * tt + W4 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_4 = lmer(y ~ tt + I(tt^2) + W4 + W4 * tt + W4 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_drg, REML = FALSE)
fit_pbo_est_5 = lmer(y ~ tt + I(tt^2) + W5 + W5 * tt + W5 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_5 = lmer(y ~ tt + I(tt^2) + W5 + W5 * tt + W5 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_drg, REML = FALSE)

## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## unable to evaluate scaled gradient

## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge: degenerate Hessian with 1 negative eigenvalues
fit_pbo_est_6 = lmer(y ~ tt + I(tt^2) + W6 + W6 * tt + W6 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_pbo, REML = FALSE)
fit_drg_est_6 = lmer(y ~ tt + I(tt^2) + W6 + W6 * tt + W6 * I(tt^2) +
  (tt + I(tt^2)|subj),
  data = dat_drg, REML = FALSE)

l1 = summary(fit_pbo_est_1)$logLik + summary(fit_drg_est_1)$logLik
l2 = summary(fit_pbo_est_2)$logLik + summary(fit_drg_est_2)$logLik
l3 = summary(fit_pbo_est_3)$logLik + summary(fit_drg_est_3)$logLik
l4 = summary(fit_pbo_est_4)$logLik + summary(fit_drg_est_4)$logLik
l5 = summary(fit_pbo_est_5)$logLik + summary(fit_drg_est_5)$logLik
l6 = summary(fit_pbo_est_6)$logLik + summary(fit_drg_est_6)$logLik

likes = c(10,11,12,13,14,15, 16)

```

Method	Purity (coefficient)	Purity (outcome)	loglikelihood (condition)	loglikelihood (outcome)	loglikelihood (mean)	Random
loglikelihood	-6094.255	-6097.163	-6094.360	-6131.375	-6132.756	-6109.696
p-value		0.121	0.975	0.991	5.55e-16	1.11e-16