Simulation report – prediction of survival probabilities

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September 10, 2015

1 Objective

To compare the predictive capabilities of survival probabilities between QRJM and JM using linear mixed model (LMJM) for data from different distribution features.

2 Simulation procedure

- 1. Define different simulation scenarios in terms of the distribution of random error and simulate 100 data sets for each scenario (see below for the specification of scenarios). Each simulated data set has 600 subjects, 500 out of which will be used to fit the model for inference purpose and the rest 100 will be used to make predictions and validation.
- Fit the data using QRJM and LMJM respectively and save the posterior samples of the model parameters.
- 3. Validation data preparation: choose a time t so that all the patients selected to for prediction will only have longitudinal measurements up to this time t.
- 4. Make predictions of subject-specific random effects: use saved posterior samples in step 2 and longitudinal measurements from step 3 to predict subject-specific random effects for every subject in the validation samples.
- 5. Calculate the predictions of survival probabilities for all the subjects in validation data for some time $u = t + \Delta t \ (\Delta t > 0)$.
- 6. Summarize the result: make Bland-Altman plots and calculate the MSE and bias for our predictions versus the gold standard, which is calculated from the true simulated values (i.e. the random effects and the parameters).

3 Simulation scenarios and results

$$\begin{cases}
Y_{it} = \boldsymbol{X}_{it}^{\top} \boldsymbol{\beta} + \boldsymbol{H}_{it}^{\top} \boldsymbol{\delta} + \boldsymbol{Z}_{it}^{\top} \boldsymbol{u}_{i} + \varepsilon_{it}, \varepsilon_{it} \sim ALD(0, \sigma, \tau) \\
h(T_{i}|\mathcal{T}_{iT_{i}}, \boldsymbol{W}_{i}; \boldsymbol{\gamma}, \alpha_{1}, \alpha_{2}) = h_{0}(T_{i}) \exp(\boldsymbol{W}_{i}^{\top} \boldsymbol{\gamma} + \alpha(\boldsymbol{H}_{iT_{i}}^{\top} \boldsymbol{\delta} + \boldsymbol{Z}_{iT_{i}}^{\top} \boldsymbol{u}_{i}))
\end{cases}$$
(1)

There are three scenarios in the simulation study:

1. Scenarios One: data are generated using Model (1). Choose the $\tau = 0.25$ for the ALD distribution.

- 2. Scenarios Two: data are generated using Model (1). Choose $\tau = 0.5$ for the ALD distribution.
- 3. Scenario Three: data are generated using Model (1), but the random error follows standard normal distribution instead of ALD.

3.1 Inference results

Table 1: Inference result for Scenario One

| | QRJM ($\tau = 0.25$), true model | | | QRJM $(\tau = 0.5)$ | | | | LMJM | | |
|---------------------------|------------------------------------|-------|-------|---------------------|-------|-------|--------|-------|-------|--|
| | bias | se | MSE | bias | se | MSE | bias | se | MSE | |
| alpha | 0.008 | 0.110 | 0.012 | -0.003 | 0.131 | 0.017 | -0.012 | 0.138 | 0.019 | |
| beta[1] | -0.001 | 0.101 | 0.010 | 1.697 | 0.156 | 2.904 | 2.683 | 0.174 | 7.227 | |
| beta[2] | -0.005 | 0.096 | 0.009 | 0.002 | 0.132 | 0.017 | 0.014 | 0.150 | 0.023 | |
| $^{\mathrm{c}}$ | -0.007 | 0.090 | 0.008 | -0.004 | 0.091 | 0.008 | -0.007 | 0.090 | 0.008 | |
| delta[1] | 0.002 | 0.085 | 0.007 | 0.019 | 0.113 | 0.013 | 0.029 | 0.129 | 0.018 | |
| delta[2] | 0.009 | 0.092 | 0.009 | 0.018 | 0.110 | 0.012 | 0.038 | 0.128 | 0.018 | |
| $\operatorname{gamma}[1]$ | 0.007 | 0.083 | 0.007 | 0.012 | 0.086 | 0.007 | 0.007 | 0.085 | 0.007 | |
| $\operatorname{gamma}[2]$ | -0.001 | 0.087 | 0.008 | 0.005 | 0.090 | 0.008 | 0.000 | 0.089 | 0.008 | |
| sigma | -0.001 | 0.034 | 0.001 | -0.319 | 0.025 | 0.103 | - | - | - | |

Table 2: Inference result for Scenario Two

| | QRJM | $(\tau = 0.5$ |), true model | | LMJM | | | |
|---------------------------|--------|---------------|---------------|--------|-------|-------|--|--|
| | bias | se | MSE | bias | se | MSE | | |
| alpha | 0.013 | 0.094 | 0.009 | 0.013 | 0.106 | 0.011 | | |
| beta[1] | -0.007 | 0.089 | 0.008 | -0.009 | 0.112 | 0.013 | | |
| beta[2] | 0.011 | 0.080 | 0.007 | 0.013 | 0.103 | 0.011 | | |
| $^{\mathrm{c}}$ | 0.002 | 0.084 | 0.007 | -0.001 | 0.086 | 0.007 | | |
| delta[1] | -0.009 | 0.075 | 0.006 | -0.006 | 0.092 | 0.009 | | |
| delta[2] | 0.002 | 0.082 | 0.007 | 0.006 | 0.097 | 0.009 | | |
| $\operatorname{gamma}[1]$ | 0.009 | 0.090 | 0.008 | 0.009 | 0.090 | 0.008 | | |
| $\operatorname{gamma}[2]$ | 0.001 | 0.086 | 0.007 | 0.002 | 0.087 | 0.008 | | |
| $_{ m sigma}$ | 0.003 | 0.037 | 0.001 | - | - | - | | |

Table 3: Inference result for Scenario Three

| | QRJM ($\tau = 0.5$) | | | | LMJM, true model | | | |
|---------------------------|-----------------------|-------|-------|--|------------------|-------|-------|--|
| | bias | se | MSE | | bias | se | MSE | |
| alpha | -0.012 | 0.075 | 0.006 | | 0.004 | 0.076 | 0.006 | |
| beta[1] | 0.013 | 0.050 | 0.003 | | 0.003 | 0.046 | 0.002 | |
| beta[2] | 0.001 | 0.045 | 0.002 | | 0.002 | 0.043 | 0.002 | |
| $^{\mathrm{c}}$ | 0.001 | 0.082 | 0.007 | | 0.005 | 0.081 | 0.007 | |
| delta[1] | 0.007 | 0.045 | 0.002 | | 0.001 | 0.041 | 0.002 | |
| delta[2] | 0.008 | 0.058 | 0.003 | | 0.000 | 0.055 | 0.003 | |
| $\operatorname{gamma}[1]$ | -0.003 | 0.081 | 0.007 | | -0.007 | 0.080 | 0.006 | |
| gamma[2] | -0.002 | 0.085 | 0.007 | | -0.006 | 0.084 | 0.007 | |
| $_{ m sigma}$ | - | - | - | | 0.001 | 0.029 | 0.001 | |

3.2 BA plots of predictions

Scenario One

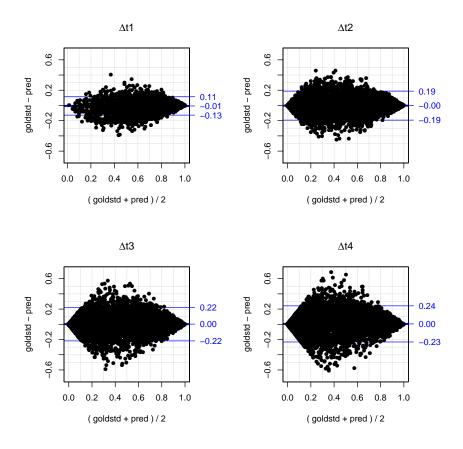


Figure 1: BA plot: data fitted using QRJM model with $\tau=0.25$

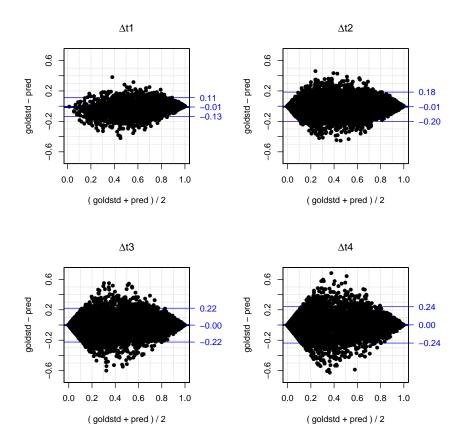


Figure 2: BA plot: data fitted using QRJM model with $\tau=0.50$

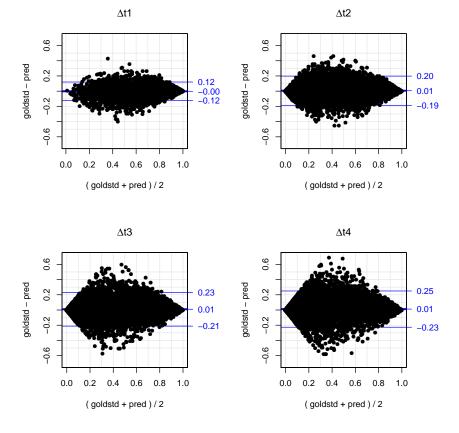


Figure 3: BA plot: data fitted using LMJM $\,$

Scenario Two

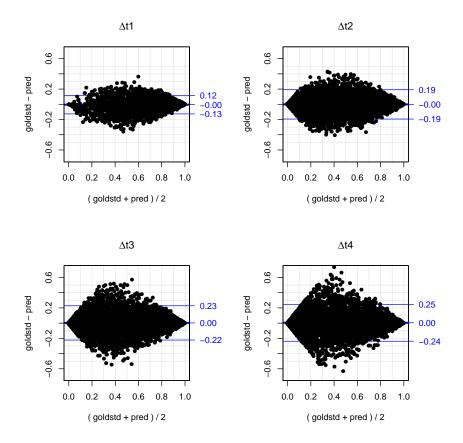


Figure 4: BA plot: data fitted with JM using QR model with $\tau=0.5$

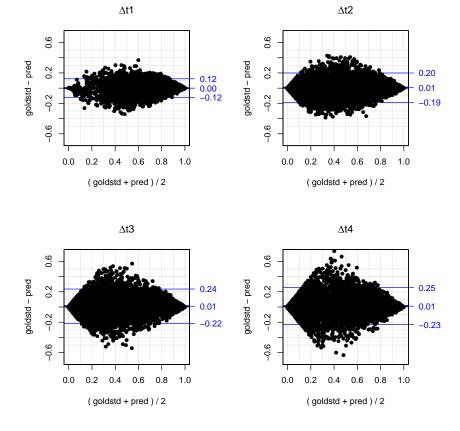


Figure 5: BA plot: data fitted using LMJM $\,$

Scenario Three

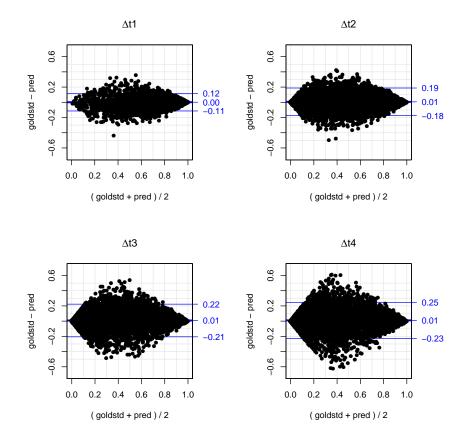


Figure 6: BA plot: data fitted using LMJM $\,$

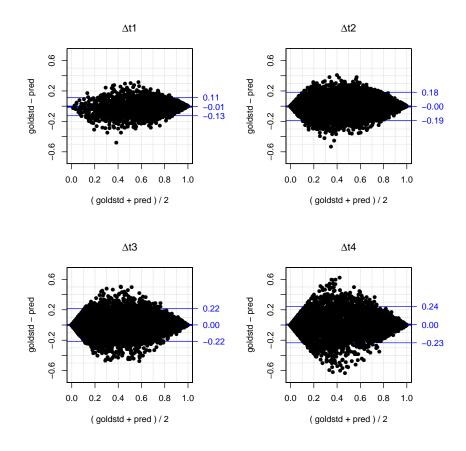


Figure 7: BA plot: data fitted with JM using QR model with $\tau=0.5$

${\bf 3.3}\quad {\bf Summary\ table-predictions}$

Table 4: MSE and bias of the predictions of survival probabilities from two models ($\Delta t_1 < \Delta t_2 < \Delta t_3 < \Delta t_4$)

| Scenario | | QRJM ($\tau = 0.25$) | | QRJM | QRJM $(\tau = 0.5)$ | | LMJM | |
|----------|--------------|------------------------|--------|-------|---------------------|--------|--------|--|
| | | MSE | Bias | MSE | Bias | MSE | Bias | |
| | Δt_1 | 0.004 | 0.006 | 0.004 | 0.011 | 0.004 | 0.001 | |
| 1 | Δt_2 | 0.009 | 0.002 | 0.009 | 0.007 | 0.009 | -0.006 | |
| | Δt_3 | 0.012 | -0.002 | 0.012 | 0.002 | 0.012 | -0.010 | |
| | Δt_4 | 0.014 | -0.005 | 0.014 | -0.001 | 0.014 | -0.012 | |
| | Δt_1 | | | 0.004 | 0.004 | 0.004 | 0.000 | |
| 2 | Δt_2 | - | - | 0.009 | 0.001 | 0.009 | -0.006 | |
| | Δt_3 | - | - | 0.013 | -0.003 | 0.013 | -0.010 | |
| | Δt_4 | - | - | 0.015 | -0.005 | 0.015 | -0.011 | |
| | Δt_1 | | | 0.004 | 0.007 | 0.003 | -0.001 | |
| 3 | Δt_2 | - | - | 0.009 | 0.002 | 0.004 | -0.006 | |
| | Δt_3 | - | - | 0.012 | -0.012 | -0.001 | -0.008 | |
| | Δt_4 | - | - | 0.014 | -0.012 | -0.005 | -0.010 | |
| | 04 | | | 0.011 | 0.012 | 0.000 | | |

3.4 Comments

- 1. In both model settings, the accuracy (in terms of MSE and the BA plots) of predictions decreases as Δt increases, which makes sense as it's more difficult to accurately predict the survival for longer time in the future since there are more variabilities and uncertainty.
- 2. QRJM and LMJM perform closely in predicting the survival probabilities in all three scenarios although the inference results from the true model is better than others. I have no formal interpretation yet but thinking that probably it is because in the prediction part we only use parameters α , δ and γ to calculate the survival probabilities (But we used all the model parameters in predicting random effects), whose estimation from both model settings are pretty accurate and close to truth.