



Mean Squared Error as Criterion for Sampling Schedule Optimization for Individual Dose Targeting in Busulfan

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RFPK

Introduction:

Individual MAP Estimator
(uses population prior)

Known to introduce **bias**
(shrinkage towards parameter mean)
in the estimate for an individual

Optimal Individual MAP Schedule?

Proposed Design Criterion:
Mean Squared Error

$$MSE = \frac{1}{M} \sum_{j=1}^M (\hat{\theta}_j - \bar{\theta})^2$$

$\bar{\theta}$ = true individual parameter value
 $\hat{\theta}_j$ = individual estimate for error replicate j

$$MSE = \underbrace{\left(\frac{1}{M} \sum_{j=1}^M (\hat{\theta}_j - \bar{\theta}) \right)^2}_{me^2} + \underbrace{\frac{1}{M} \sum_{j=1}^M \left((\hat{\theta}_j - \bar{\theta}) - \frac{1}{M} \sum_{j=1}^M (\hat{\theta}_j - \bar{\theta}) \right)^2}_{pev}$$

Mean prediction error (me)
A measure of "bias"

Prediction error variance (pev)
A measure of "precision"

D-optimality criterion also a measure of precision

Proposed individual estimation design criterion: MSE – accounts for precision and bias

Most optimal design software for both individual (ADAPT II) and population (PFIM, WinPop, popED) experimental design is based on D-optimality design criterion (determinant of Fischer information matrix; a measure of precision only).

MAP = maximum a posteriori estimation (i.e. MLE with population prior)

MSE; see e.g. [1] and [2]

Objective 1: Illustration of MSE-optimality vs. D-optimality:

Single time point schedule
(in 0 – 6 hrs)

Kinetic and statistical Model:
Abstract IV Busulfan kinetics Model

Optimal Individual MAP Schedule?

CL – LN(12.2 L/h, 10% CV)
Vd – LN(49.3 L, 30% CV)
Corr(CL,Vd) = 0.2
RUV: 0.06 SD & 8.3% CV

Compare criteria:
Plotted as:
– root mean MSE
– root me² (bias)
– root pev (precision)
– det(FIM) (D-Opt) (inverted scale on rhs)

Optimality Criteria Comparison

D-Optimality aligns with precision
Bias and precision contribute –equally to MSE

Optimality Criteria - by Parameter

CL (with only 10% CV) is well estimated
Vd (with 50% CV) dominates

Two simulated individuals

Single Individual: 25 Replicate Error Installations: Simulated Data and Fitted Curves (MAP)

Single Individual: 25 Replicate Error Installations: Simulated Data and Fitted Curves (MAP)

Bias is visible Bias is obvious

Modeling and Simulation:

Population mixed-effect modeling of IV Busulfan
First order kinetics with three hour infusion
42 individuals

CL – LN(12.2 L/h, 31.1% CV)
Vd – LN(49.3 L, 29.5% CV)
Corr(CL,Vd) = 0.9
RUV: 0.06 SD & 8.3% CV

Simulation:
N=1000-2500 simulated subjects
M=15-25 error replicates

Re-estimation (MAP):
nlx function in R implemented with
– population prior and
– rotation to incorporate full covariance [as in 3]

Optimal Individual MAP Schedule?

Bias (all curves above true time course) due to parameter "shrinkage" toward population values

MSE criterion (precision and bias):
MSE computed:
– for each parameter (CL and Vd)
– scaled ($MSE/\bar{\theta}^2$) and averaged
– for each individual
– on each potential time schedule

D-optimality criterion (precision):
Determinant of the Fisher Information Matrix (FIM)
Computed in R as $1/\det(vcov)$
– for each replicate
– for each individual (as mean over replicates)
– on each potential time schedule
(vcov = covariance matrix of model parameters returned from the nlx)

Figure of merit:
Mean MSE (over all individuals)

Figure of merit:
Median det(FIM) (over all individuals)

Objective 2: Rich sampling optimal outpatient schedule for Individual Busulfan TDM

Outpatient sampling <6 hrs from infusion initiation

Rich sampling: 6 time points (0-6 hrs)
– Samples at 3, 3.25, 6 hrs fixed (by clinician)
– Sampling on 1/4, 1/2, 3/4 and full hour only
– Single sample per time point only

Optimal Individual MAP Schedule?

Rich sampling:
MSE and D –Optimal schedules similar (as expected)

Typical subject with MSE-Optimal and D-Optimal sampling schedules

Objective 3: Comparison of Outpatient schedule with Standard of Care (Inpatient)

Proposed Outpatient Schedule with MAP Estimation
vs.
Inpatient Schedule (Standard of Care) with MLE Estimation

Optimal Individual MAP or MLE Schedule?

Comparison of Proposed Outpatient (OUT) and Standard of Care Inpatient (SoC) sampling Schedules (with and without Prior):
Schedules being considered:
– OUT w/MAP (red square)
– OUT w/MLE (blue circle)
– SoC w/MAP (green triangle)
– SoC w/MLE (magenta diamond)

Typical subject with standard-of-care and proposed sampling schedules

• Estimation "with prior" (MAP) increased bias but improved precision.
• Proposed **Outpatient** schedule with MAP had lowest MSE
• Standard of care schedule with MLE had the highest MSE
• Conclusion: **Outpatient sampling is feasible**

Bonus Example: Fludarabine Sampling Schedule Optimization

Population Model:

CL – LN(11.3 L/h, 39% CV) CL V1 Q
V1 – LN(58.1 L, 40% CV) 0.8
Q – LN(16.2 L/h, 46% CV) 0^a 0.3^b
V2 – LN(85.2, 47% CV) 0^a 0.3^b 1^a

(a – correlations assumed by construction)
(b – correlations equal by construction)

RUV: 0.016 SD & 23.7% CV

Optimal Individual MAP Schedule?

Compare criteria:
– for each parameter
– All params. combined
– 24 hour AUC
– 120 hour AUC
Plotted as:
– root mean MSE
– root me² (bias)
– root pev (precision)

MSE

Bias

Precision

Bias, precision and MSE of AUC follows similar trend as for combined parameters

Conclusions: MSE-optimality vs. D-optimality Criteria:

Individual MAP estimator bias:
Extent of bias determined by number of samples (well known) as well as sampling schedule

Optimal Individual MAP Schedule?

MSE as optimality criterion:
– Accounts for bias and precision
– Requires simulation/re-estimation

D-optimality criterion:
– Accounts for precision only
– Does not require simulation

References:

[1] Sheiner LB and Beal SL (1981) Some suggestions for measuring predictive performance. J. Pharmacokinetics and Biopharmaceutics 9:4-503-512
[2] Monteleone JPR and Duffull SB. (2003) Choice of best design. In Kimko HC and Duffull SB Eds. Simulation for designing clinical trials. Taylor and Francis Group, NY, New York.
[3]. Callegari T, Caumo A, and Cobelli C (2002) Generalization of Map Estimation in SAAM II: Validation Against ADAPT II in a Glucose Model Case Study Annals of Biomed Eng