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Matrix

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PURPOSE

When data are sparse, parameters derived from a non-linear mixed effects model analysis can shrink to the mean and can be misleading. The objective of this project was to predict the shrinkage on parameters using Bayesian methodology and test whether the results of a published 1 compartment model example by Combes et al., are applicable to more complex models.

METHOD

- Shrinkage values were predicted using the Bayesian FIM of PFIM (BPFIM) and compared to values obtained from NONMEM.
- Step 1: Initially, the work completed by Combes et al. involving a 1 compartment IV model was replicated.
- Step 2: Utilized the methods to predict and determine shrinkage on individual parameters in more complex models; 1 and 2 compartment oral and IV models.
- Various scenarios ranging from high (53.29%CV) to low (2.5%CV) variance and utilizing combined and additive error models were examined.
- Sector Each scenario was evaluated with a range of sampling times at optimized time points which were obtained using PFIM with 2-8 points per profile.
- Observed shrinkages were calculated using conditional estimation with interaction and MAXEVAL=0, using NONMEM 7.4.
- The predicted and observed shrinkage values were then compared.

RESULTS

- Observed and predicted shrinkage values obtained from all the scenarios were plotted on x and y axis respectively.
- Predicted shrinkage values for combined error models were greater than predicted shrinkage for additive error models.
- There was a larger degree of deviation between BPFIM predicted and NONMEM estimated shrinkage for additive error models with larger variances compared to smaller variances.
- Combined error models had more accurate shrinkage predictions compared to NONMEM estimated values.
- Shrinkage is dependent on number of samples collected per subject and is inversely proportional to parameter variance (omega) and directly proportional to residual variability (sigma).
- The largest difference in predicted and observed shrinkages was seen in the KA and Q parameters of the two compartment oral absorption model.



Predicting Shrinkage of Individual Parameters in More Complex NLME Models Using Bayesian Fisher Information

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Table 1: NONMEM	versus BPFIM re	ported shrinkage valu	es for various mod	els and parameters	Table 2: Median diffe	erence (Obs	erved minus Pr	edicted) Shrinkage	value in differe	ent models
Model	1_CMT_IV	1_CMT_PO NONMEM/	2_CMT_IV	2_CMT_PO	Parameters	Value	1_CMT_IV	1_CMT_PO	2_CMT_IV	2_CMT_P
Parameters	NONMEM/ BPFIM %	BPFIM %	NONMEM/ BPFIM %	NONMEM/ BPFIM %	Clearance	0.6	2.864	1.812	-0.16	-0.451
					Central Volume	8	-0.001	0.484	2.481	1.303
					Ка	1.5		1.205		4.927
Clearance	0-57.628 /0.001- 57.852	0-56.979 /0.0008-57.699	4.185-92.896 /0.719-93.06	12-91.8 / 1.54-92.56	Peripheral Volume	40			2.603	1.872
Central Volume	0-74.432 /0.00015- 73.850	0-74.432 /0.00015-73.850	0-83.323/ 0.0005-81.613	2.47-82.6 / 0.141-82.80	Inter- Compartmental Clearance	3.5			6.19	4.953
Ка		0-88.040 /0.004-87.131		7.73-85.94 /	Table 3: Random effe	ects list – fo	r simulations			
				0.181-86.065	Omega List	0.1 (O1)	0.025 (O2)	0.05 (O3)	0.25 (O4)	0.5 (O5)
Peripheral			2.444-79.599	1.79-77.18/	ons					
Volume			/0.231-79.977	0.426-79.113	Sigma (inter, 0.15 slope) (F	, 0.15 0.: R1)	30, 0.15 0.30 (R2) (0, 0.30 0.15, 0 R3) (R4)	0.3, 0 (R5)	0.5, 0 (R6)
Inter-			0-86.554 /	0-79.1/						
Compartmental Clearance			0.012-85.92	0.0208-79.5972						

2 CMT IV - Parameters in Table 1







CONCLUSIONS

- Results demonstrate that there is a correlation between BPFIM predicted and NONMEM estimated shrinkage values even for more complex models.
- The observed and predicted shrinkage values are roughly centered on the line of identity for most models, though not as well as the IV 1 compartment model presented by Combes et al.
- BFIM as implemented in PFIM is a useful method to predict shrinkage, especially for combined error models.

REFERENCES

- Combes, F.P., Retout, S., Frey, N. et al. Pharm Res (2013) 30: 2355. <u>https://doi.org/10.1007/s11095-013-</u> 1079-3
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