

# Development of User-friendly Bayesian Predictive Platform for Blood Boron-10 Pharmacokinetics following Intravenous Infusion of [<sup>10</sup>B] L-4-BORONOPHENYALANINE

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## Introduction/Objectives

- Boron neutron capture therapy (BNCT) allows high-precision radiotherapy against tumor using boron-10 (<sup>10</sup>B) with tumor-localizing characteristics and strong tendency to capture thermal neutrons.
- It is important to accurately predict the blood <sup>10</sup>B concentration during the neutron irradiation to deliver the prescribed dose as planned.
- This study was performed to develop user-friendly Bayesian predictive platform for pharmacokinetics (PK) of <sup>10</sup>B which is clinically applicable in BNCT.

## Methods

- Population PK model for <sup>10</sup>B was constructed using blood <sup>10</sup>B concentration over time data following intravenous infusion of boronophenylalanine (BPA) which were digitized from previous study results, which was used as a prior distribution model for the Bayesian prediction.
- The predictive model was implemented in NONMEM® 7.4 (ICON Development Solutions, USA), and NONMEM was executed using R (version 4.03) with user-friendly interface provided by Shiny package.
- Simulation and sensitivity analyses were conducted to evaluate the predictive performance of the platform and identify optimal PK sampling time for blood <sup>10</sup>B.

## Results (PK model)

- Elimination rate constant from central compartment value ( $k_{10}$ ) was estimated as 0.006(1/min). Volume of distribution of central compartment( $V_1$ ) was estimated as 0.252(L).

## Conclusion

- This predicted platform has the potential to play a crucial role in clinical trials aiming at evaluating the effect of BNCT, as it assess the treatment effect in a user-friendly and reliable way. By enabling a more precise prediction of the therapeutic effect, the platform can help clinicians tailor treatment plans to individual patients, ultimately improving the overall quality of care in BNCT.

## References

[1] Palmer, M. R., Goorley, J. T., Kiger, W. S., Busse, P. M., Riley, K. J., Harling, O. K., & Zamenhof, R. G. (2002). Treatment planning and dosimetry for the Harvard-MIT Phase I clinical trial of cranial neutron capture therapy. *International Journal of Radiation Oncology\*Biophysics*, 53(5), 1361–1379.

[2] Chanana, Capala, Chadha, et al. Boron neutron capture therapy for glioblastoma multiforme: interim results from the phase I/II dose-escalation studies. *Neurosurgery*. 1999;44(6):1182-1192; discussion 1192-3.

## Data processing

### Import PK Dataset

- Making NONMEM dataset using R shiny
- Input patient's specific information (total dose, rate and also boron concentration acquired from blood samples for its estimation during irradiation)

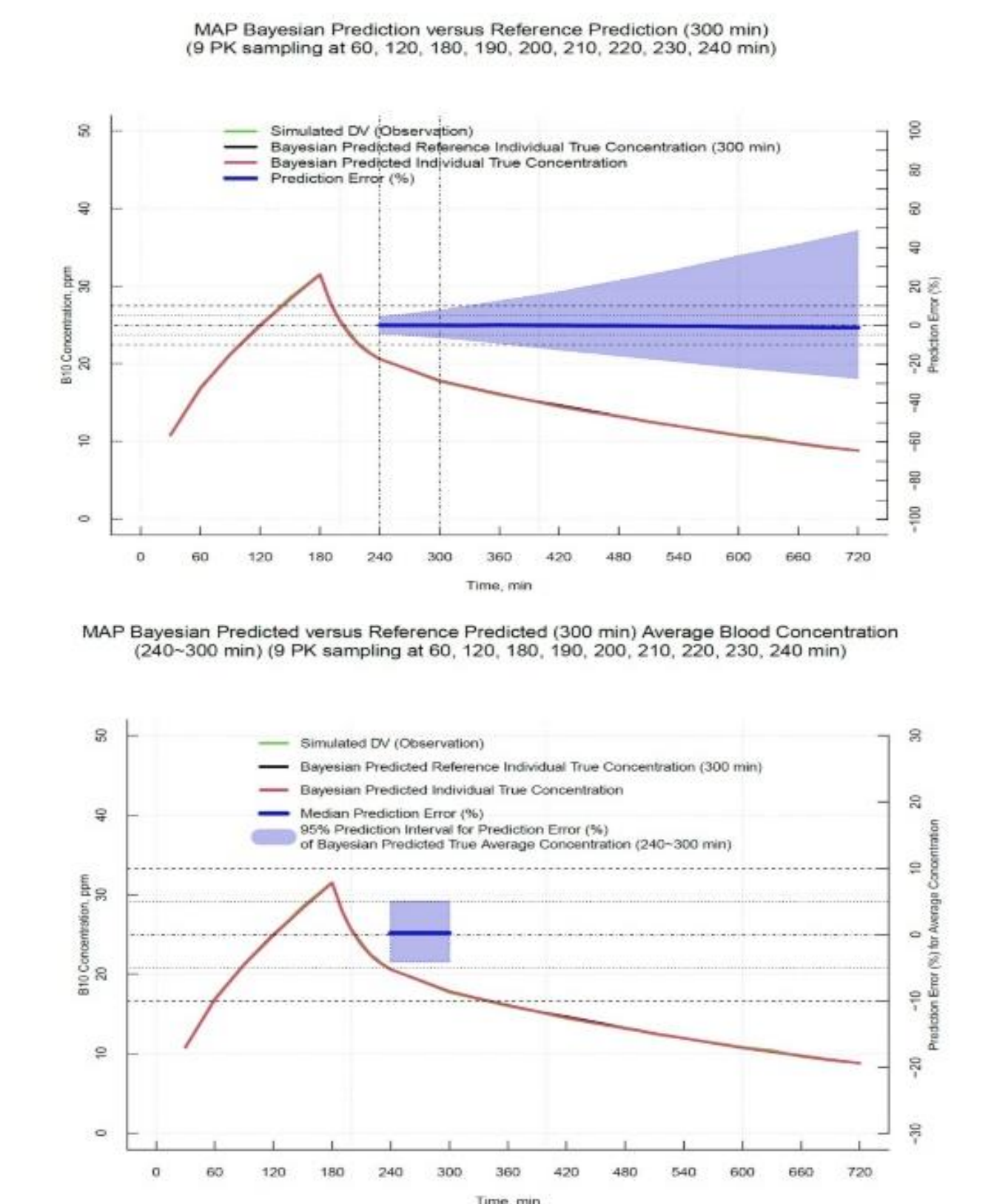
Individual Dataset for EBE

Subject ID	ID	TIME	AMT	RATE	DV	MDV	ADDL	WT
20221021	0	0	500	0.1229002764407	1	0	0	60
20221021	0	60	0	0	1	0	0	60
20221021	0	120	0	0	1	0	0	60
20221021	0	180	0	0	1	0	0	60
20221021	0	240	0	0	1	0	0	60
20221021	1	0	0	0	1	0	0	60
20221021	1	60	0	0	1	0	0	60
20221021	1	120	0	0	1	0	0	60
20221021	1	180	0	0	1	0	0	60
20221021	1	240	0	0	1	0	0	60
20221021	2	0	0	0	1	0	0	60
20221021	2	60	0	0	1	0	0	60
20221021	2	120	0	0	1	0	0	60
20221021	2	180	0	0	1	0	0	60
20221021	2	240	0	0	1	0	0	60
20221021	3	0	0	0	1	0	0	60
20221021	3	60	0	0	1	0	0	60
20221021	3	120	0	0	1	0	0	60
20221021	3	180	0	0	1	0	0	60
20221021	3	240	0	0	1	0	0	60
20221021	4	0	0	0	1	0	0	60
20221021	4	60	0	0	1	0	0	60
20221021	4	120	0	0	1	0	0	60
20221021	4	180	0	0	1	0	0	60
20221021	4	240	0	0	1	0	0	60
20221021	5	0	0	0	1	0	0	60
20221021	5	60	0	0	1	0	0	60
20221021	5	120	0	0	1	0	0	60
20221021	5	180	0	0	1	0	0	60
20221021	5	240	0	0	1	0	0	60
20221021	6	0	0	0	1	0	0	60
20221021	6	60	0	0	1	0	0	60
20221021	6	120	0	0	1	0	0	60
20221021	6	180	0	0	1	0	0	60
20221021	6	240	0	0	1	0	0	60
20221021	7	0	0	0	1	0	0	60
20221021	7	60	0	0	1	0	0	60
20221021	7	120	0	0	1	0	0	60
20221021	7	180	0	0	1	0	0	60
20221021	7	240	0	0	1	0	0	60
20221021	8	0	0	0	1	0	0	60
20221021	8	60	0	0	1	0	0	60
20221021	8	120	0	0	1	0	0	60
20221021	8	180	0	0	1	0	0	60
20221021	8	240	0	0	1	0	0	60
20221021	9	0	0	0	1	0	0	60
20221021	9	60	0	0	1	0	0	60
20221021	9	120	0	0	1	0	0	60
20221021	9	180	0	0	1	0	0	60
20221021	9	240	0	0	1	0	0	60

## Bayesian Estimation

### Bayesian Estimation

- Using previous built population PK model for BPA and using patient's blood BPA concentration dataset.
- Bayesian prediction comparison between full dataset ( irradiation vs Pre-Irradiation set)
- Figure show real time estimation accuracy and its error.



## Output

### Configure the Draft Report in an Editable Format(PDF)

- Numerical values reported in the main text are automatically updated according to the various input datasets.
- Figure and table captions are automatically generated.

### Finalize the Report

- The final report is completed with minimal edits

