In Situ, In Vitro, and In Silico Permeability Values as Inputs for Blood-Brain Barrier Penetration Prediction: Impact on Brain Exposure for Passively Diffusing Compounds, with Ethanol as a Case Study

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BACKGROUND

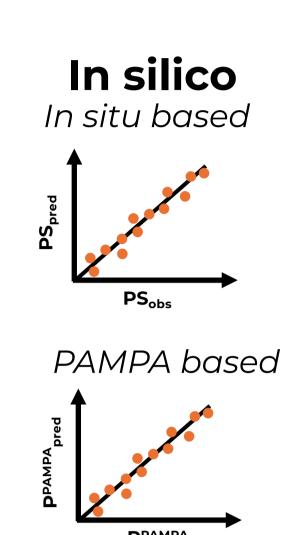
- Brain drug levels often differ from plasma due to the blood-brain barrier (BBB)
- Physiologically based pharmacokinetic models (PBPK) are used to predict brain exposure of central nervous system (CNS) active drugs^[1, 2]
- Permeability * surface area of BBB (PS_{BBB}) is one of the key input parameters in PBPK CNS models, but there is no consensus in the literature on its selection^[3,4]
- Investigate variability of **PS_{BBB}** values from in situ, in vitro and in silico sources

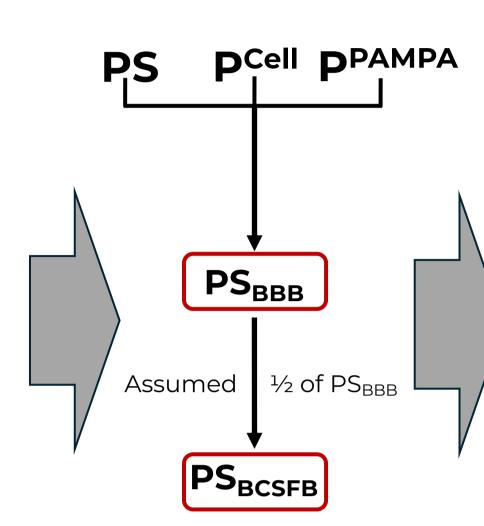
OBJECTIVES

- Demonstrate **PS_{BBB} impact** on human brain exposure using ethanol as a case study
- Permeability data collection^[5, 6]:
 - In situ
 - In vitro (cell-based, PAMPA)
 - In silico^[7,8]
 - $PS = 10^{-2.06+0.448*logP-0.366*MW/100}$
 - $P^{PAMPA} = 10^{0.939 * log P 6.210}$
 - **Translation** permeability to PS_{BBB}
 - In situ: $PS_{BBB} = PS \times BrainWT^{[9]}$ • In vitro: $PS_{BBB} = P \times SA_{BBB}$ [10]
- Focus on compounds crossing BBB via passive diffusion $(0.3 < K_{p,uu,brain} < 3)$
- Brain exposure prediction for ethanol using different PS_{BBB} values
- Validation of ethanol brain exposure prediction using human data[11-14] and 2-Fold Error (FE) as criterion

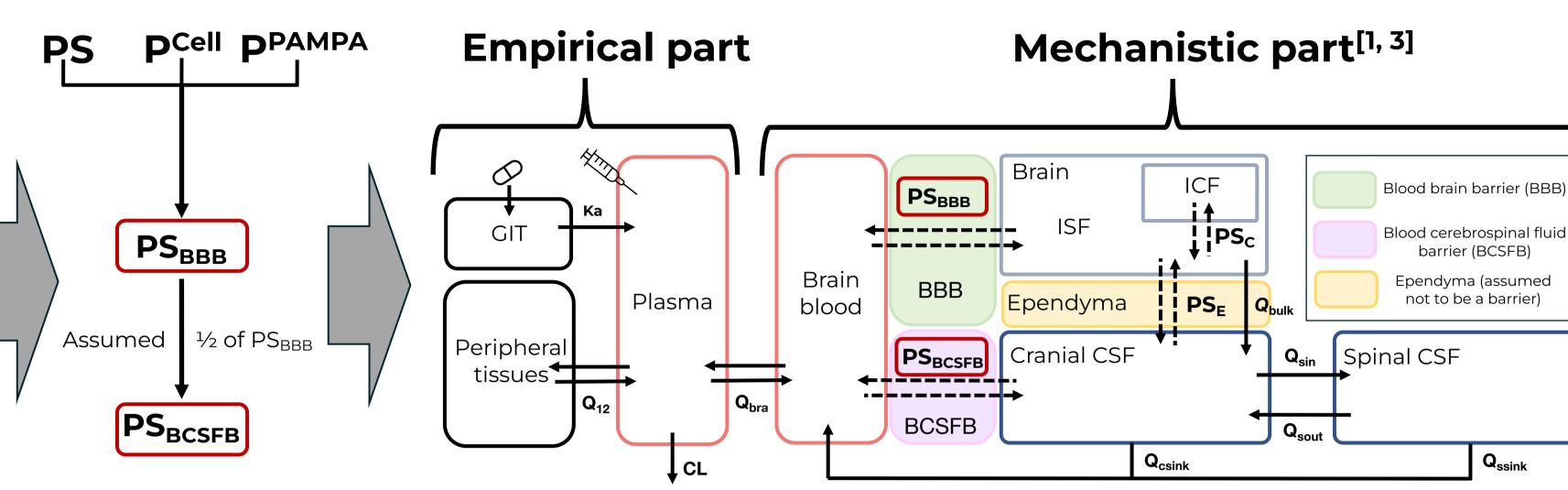
Sources of permeability data

In situ In vitro Cell-based P^{Cell} [cm/s] PAMPA PPAMPA [cm/s] PS [mL/s/g]





Translation



METHODS

PBPK CNS model

RESULTS

Permeability variability

- Permeability data were collected and translated to PS_{BBB} values for 10 CNS-active compounds
- PS_{BBB} values differed from 3-fold for lidocaine to 1385-fold for ethanol with an average difference of 230-fold
- For half the compounds, in situ perfusion gave the highest PS_{BBB} values, while PAMPA gave the lowest. In vitro cell-based showed intermediate PS_{BBB} values with the least variability.
- Maximum PS_{BBB} values among the 10 selected compounds were 2357 L/h (PAMPA), 522 L/h (in situ) and 40 L/h (cell-based)

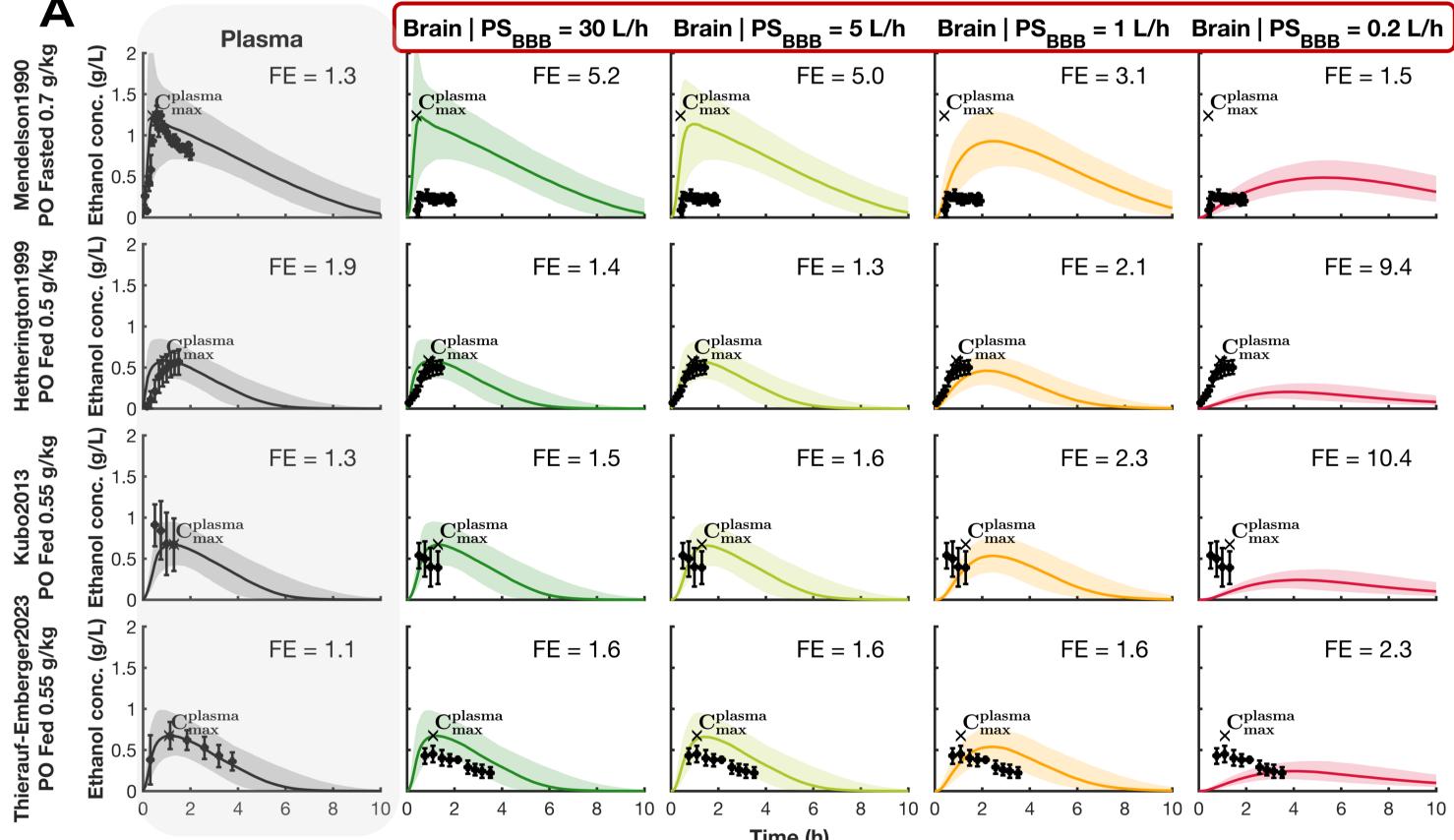
Permeability by source Permeability by compound 10² 10⁰ In situ In vitro In vitro

Figure 1. Variability in permeability of 10 compounds by source (left) and by compound (right).

cell-based

PAMPA

PS_{BBB} impact on brain exposure for Ethanol



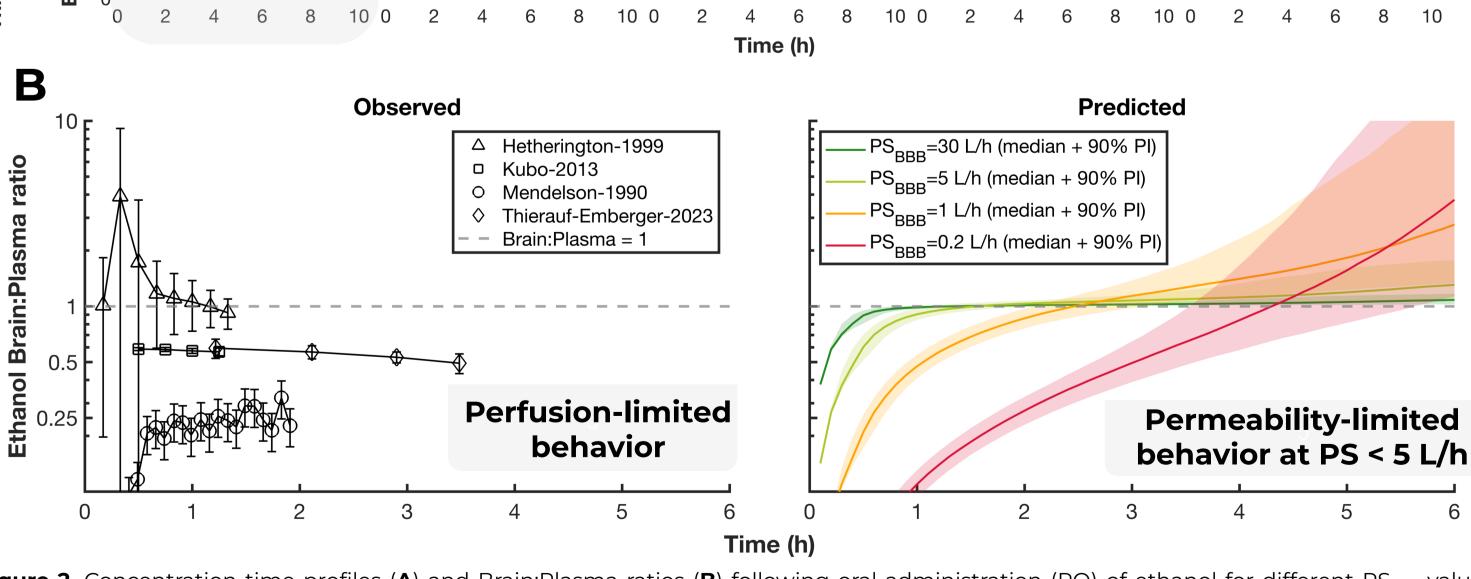


Figure 2. Concentration-time profiles (A) and Brain:Plasma ratios (B) following oral administration (PO) of ethanol for different PS_{BB} values. Experimental data were obtained from 4 human studies using magnetic resonance spectroscopy and are presented as mean ± SD. Simulation results are shown as median and 90% prediction interval (PI) based on a virtual population of 1000 subjects. Each color corresponds to a specific PS_{BBB} value. FE is the absolute average error of the prediction based on the averaged data.

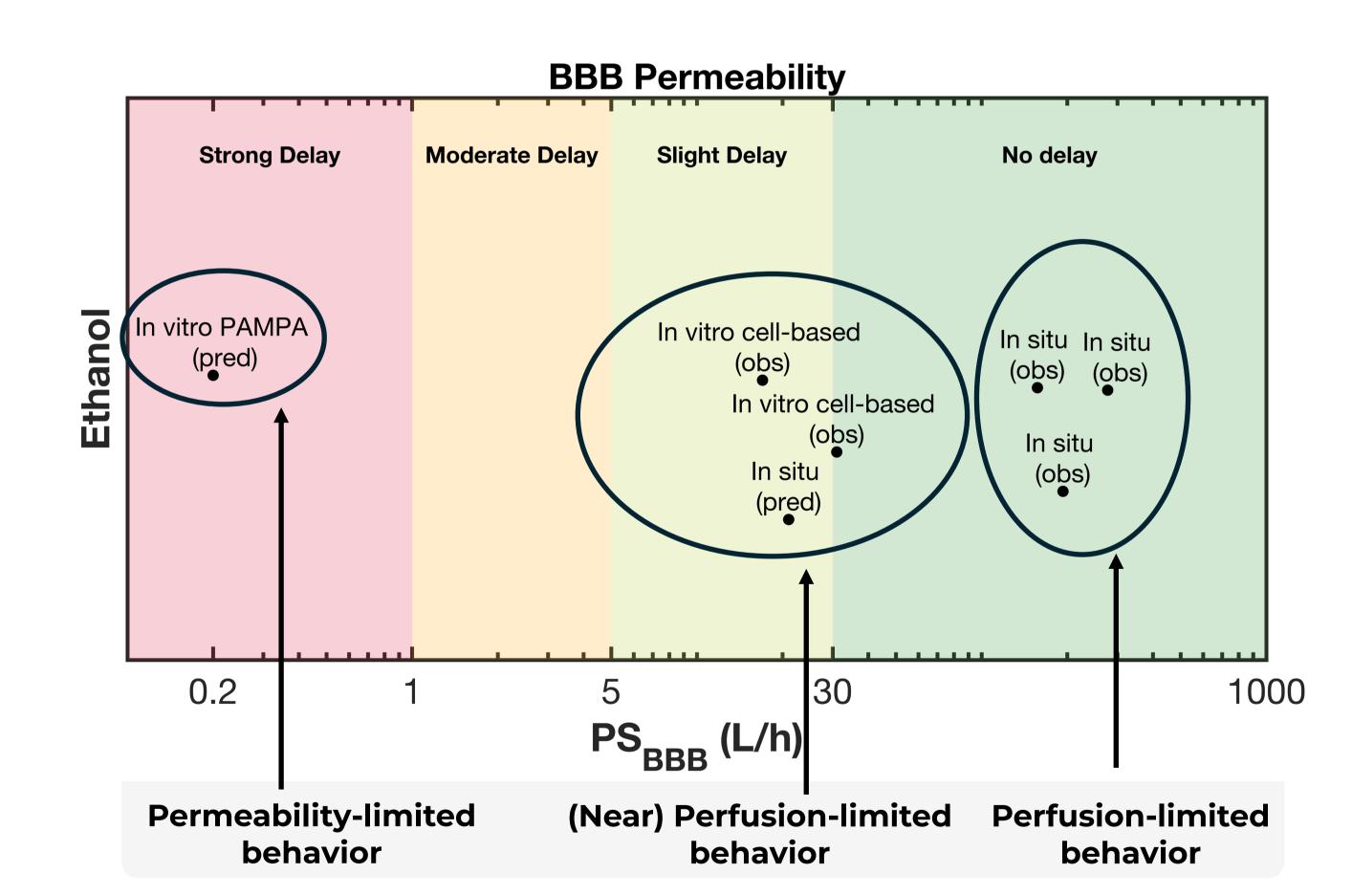


Figure 3. Comparison of the impact of PS_{BBB} values derived from different sources on ethanol brain penetration kinetics.

- Despite high PS_{BBB} variability, 3 out of 4 brain profiles were **predicted within 2-FE**
- **Permeability impact** on ethanol brain exposure was evident only at $PS_{BBB} < 5 L/h$, while experimental values ranged from 17 to 277 L/h
- Only one PS_{BBB} value, predicted by in silico PAMPA based model, resulted in a strong brain-plasma delay and large prediction error (up to 10-fold)

CONCLUSIONS

- PS values for BBB passive compounds can vary by up to 4 orders of magnitude
- For ethanol, brain exposure predictions were insensitive to large PS_{BBB} differences due to perfusion-limited distribution
- In silico PAMPA based method yielded an unrealistic PS_{BBB} prediction for **ethanol**, resulting in **permeability-limited distribution**

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