A Novel Mechanistic Approach to Predict the Steady State Volume of Distribution (V_{ss}) using the Fick-Nernst-Planck Equation



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Background

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Implementing Translational Science

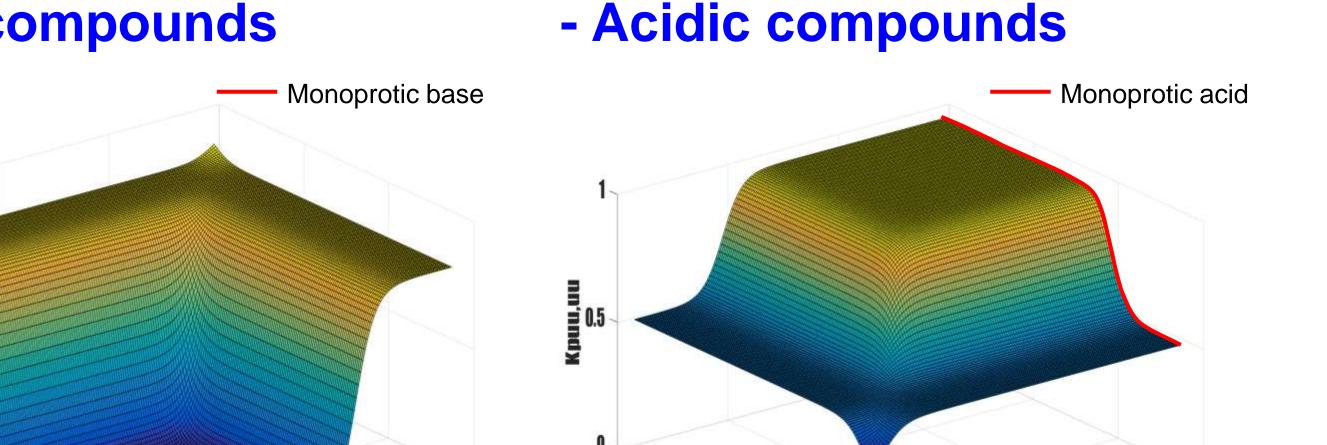
- Rodgers and Rowland method (R&R) for predicting tissue:plasma partition coefficient (Kp) and volume of distribution (V_{ss}) at steady state assumes the unbound unionized drug concentrations are in equilibrium between intracellular and extracellular water, expressed as Kp_{uu,uu}=1 [1, 2].
- Ionised compounds can passively diffuse across cell membranes dependent on the degree of ionisation and the potential difference across the membrane that is maintained by the cell, resulting in Kp_{uu,uu}≠1 for ionisable drugs.
- The impact of incorporating membrane potential-dependent Kp_{uu,uu} within the R&R method on Kp and V_{ss} prediction is evaluated in a big dataset.

Results

1.5

Kp_{uu.uu} prediction

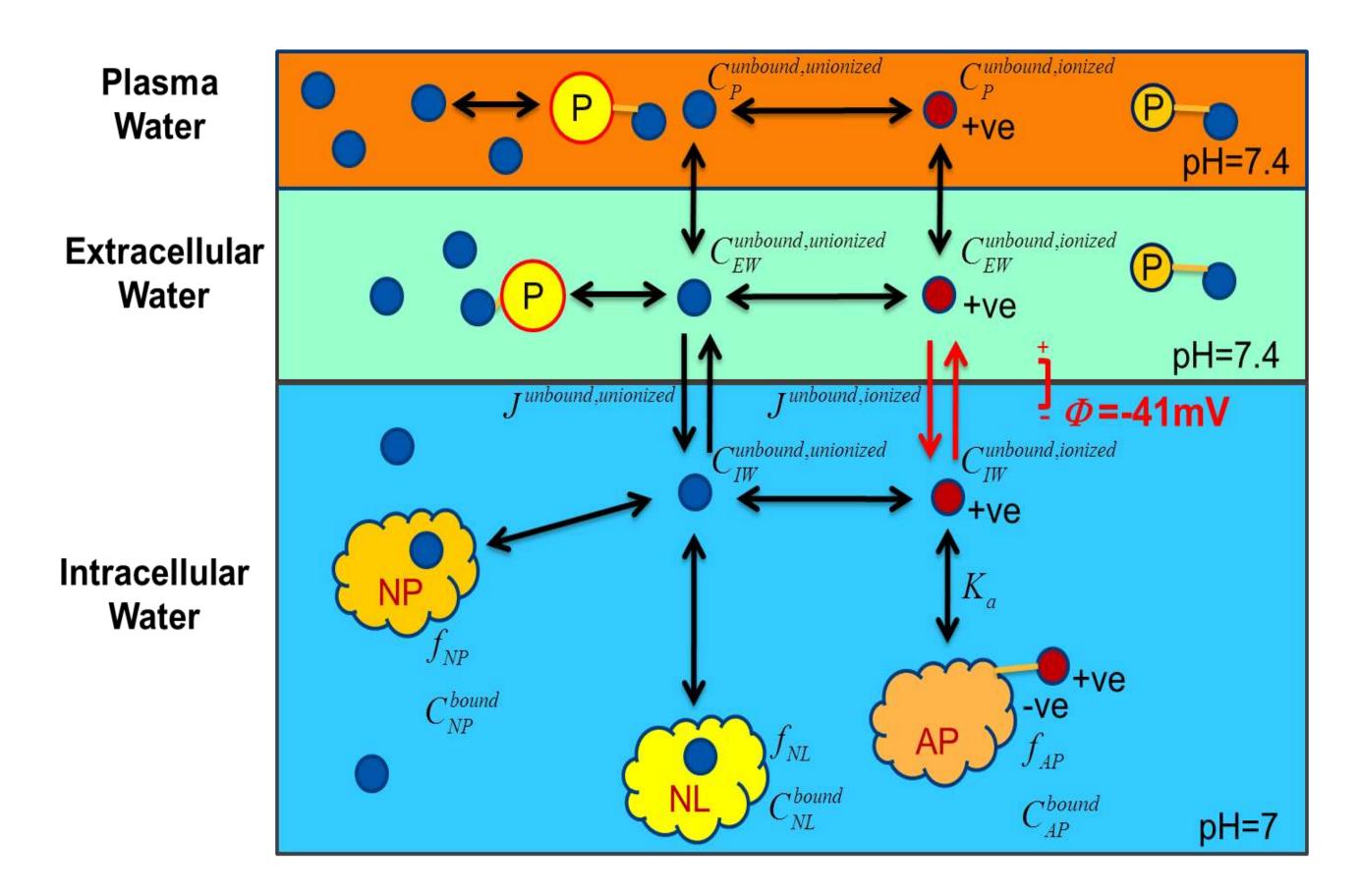
- Basic compounds

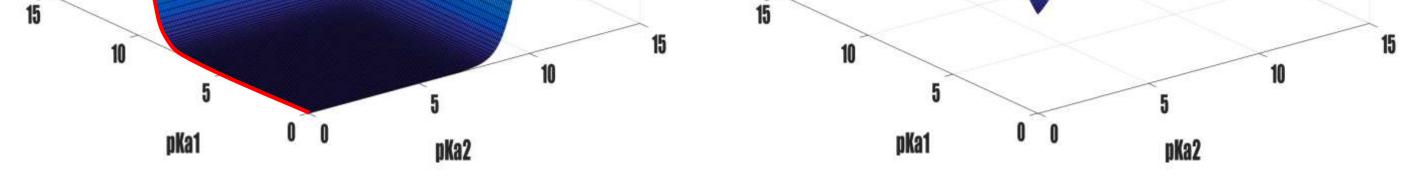


Method

Model structure

The R&R model for prediction of Kp and V_{ss} are expanded to account for the membrane potential and the passive permeation of both neutral molecules (Fick's law) and ionized molecules (Nernst-Planck equation) across the cell membranes [3].



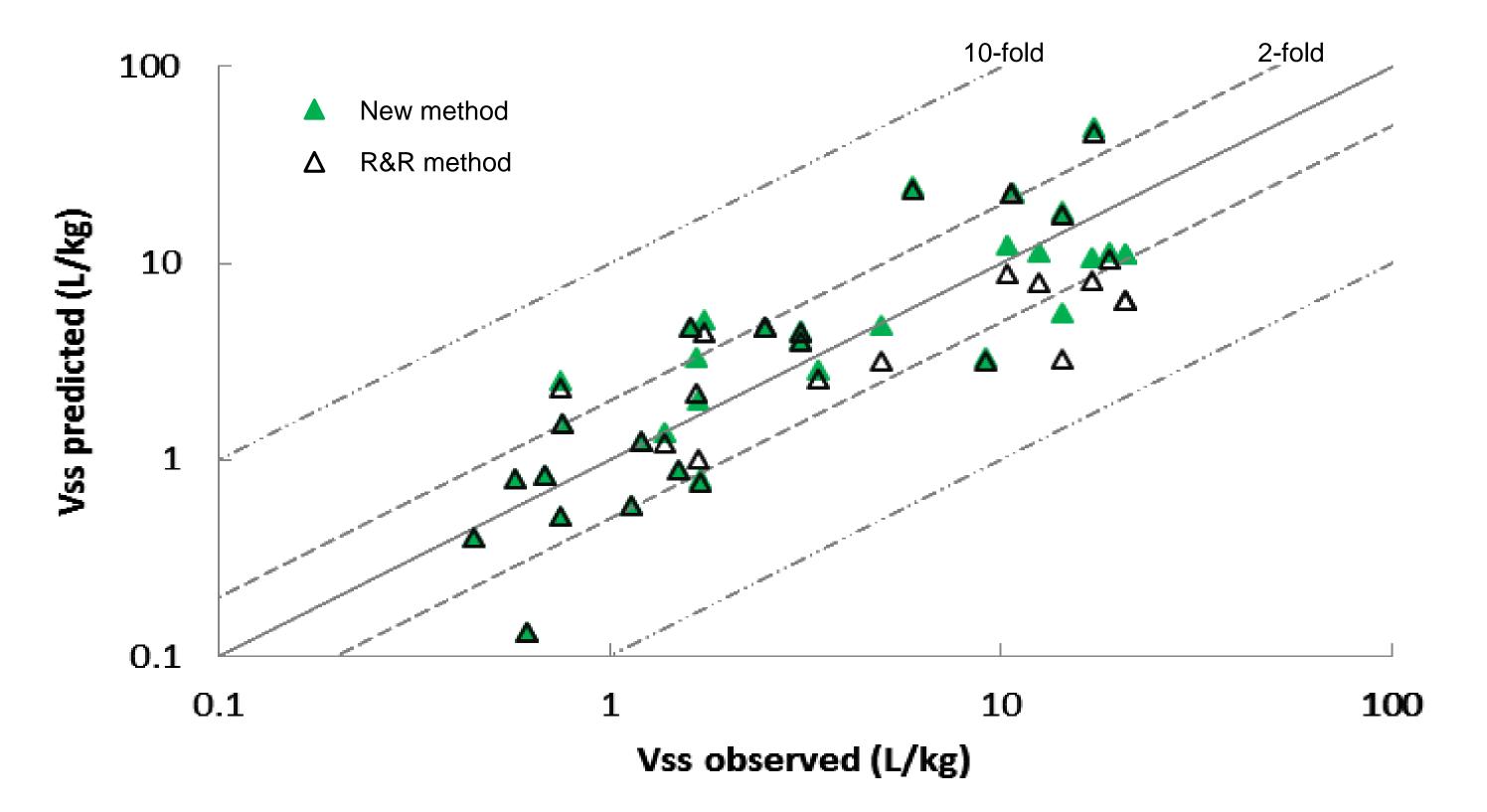


V_{ss} prediction

Compared to the R&R method, the new method predicted higher V_{ss} for basic compounds, but lower V_{ss} for acidic compounds.

The new method improved V_{ss} prediction for strong bases (pKa \ge 7; n = 22). V_{ss} was predicted within 2 or 3 -fold of observed values for 15 and 20 compounds, respectively.

The original R&R method predicted V_{ss} within 2 or 3 -fold of observed values for 13 and 18 compounds, respectively.



Fick-Nernst-Planck equation

$$\begin{aligned} V &= J^{unbound, unionized} + J^{unbound, ionized} \\ &= P^{unbound, unionized} A \Big(C_{EW}^{unbound, unionized} - C_{IW}^{unbound, unionized} \Big) \\ &+ \sum_{i=1}^{2} \left(P^{unbound, ionized_{-i}} A \frac{N_{i}}{e^{N_{i}} - 1} \begin{pmatrix} C_{EW}^{unbound, ionized_{-i}} \\ - C_{IW}^{unbound, ionized_{-i}} e^{N_{i}} \end{pmatrix} \right) \end{aligned}$$

V_{ss} prediction

$$V_{ss} = V_p + V_{RBC} \cdot EP + \sum \left(V_{tissue} \cdot Kp_{tissue} \right)$$

Model parameters

- Physiologically relevant membrane potential (-10 mV for red blood cells and -41 mV for all tissue cells) and passive permeability for electrolytes (2-4 logunit lower than neutral molecules) were assumed.
- Reference volumes for plasma, red blood cells and various tissues in an

Discussions

- For neutral compounds where all molecules are unionized and Kp_{uu,uu}=1 the new method performance is the same as the R&R model.
- If the cell membrane potential is zero, the new method results in Kp_{uu,uu}=1.
- For weak basic or weak acidic compounds, membrane potential has less impacts on Kp_{uu,uu}.
- Membrane potential values are essential system parameters that affect the model performance. However, tissue-specific membrane potential is lacking in the literature.
- Permeability parameters generated from *in vitro* systems generally represents a total effect of passive penetration of neutral and ionized drugs across the membrane.
- Modelling such the *in vitro* systems using the FNP equation allows defining the permeability ratio of ionized molecules vs. neutral molecules.
- Impact of membrane potential-dependent $Kp_{uu,uu}$ on drug distribution (V_{ss}) has been evaluated in a dataset of compounds. Its impacts on metabolism, drug-

adult male (20 years old and 81 kg) in Simcyp Healthy volunteer population with default plasma pH, tissue composition and volume were used.

- A library of compounds (n=72, including 7 neutral compounds, 35 monoprotic bases, 13 monoprotic acids, 6 diprotic bases and 11 ampholytes) were tested.
- The performance of the new method was evaluated by comparing the predicted V_{ss} to those predicted by R&R method and observed.

J: flux; N: Nernst constant. P: passive permeability; A: membrane surface area; ω : ionization parameter; fu: unbound fraction in plasma; f_{EW}: fraction of extracellular water; f_{IW}: fraction of intracellular water; f_{NL}: fraction of neutral lipids: f_{NP}: fraction of neutral phospholipids; [AP-]: negative charged acid phospholipids; P_{o:w}: octanol: water partition coefficient; K_{aPR}: protein binding constant; [PR]_{tissue}: binding protein concentration in tissue; K_{aAP}: Acid phospholipid binding constant; V_p: plasma volume; V_{RBC}: red blood cell volume; EP: red blood cell partition coefficient; V_{tissue}: tissue volume. drug interaction and the drug efficacy and toxicity can be further explored if the tissue models are modified accordingly within the dynamic PBPK models.

Conclusion

- A novel semi-mechanistic approach to predict the Kp and V_{ss} using the Fick-Nernst-Planck equation is developed.
- The results show accounting for electrolyte passive permeation has an impact on the prediction of $\rm V_{ss}.$
- The V_{ss} predictions for the strongly basic compounds (pKa ≥ 7) investigated were improved, compared to R&R method.
- Model performance depends on tissue-specific membrane potential and the ratio of passive permeability of ions vs. neutral molecules.

References

[1] Rodgers T, et al., J Pharm Sci, 2005, 94:1259-1276.
[2] Rodgers T & Rowland M, J Pharm Sci , 2006, 95:1238-1257.
[3] Trapp S, et al., European Biophysics Journal, 2008. 37:1317-1328.

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