Clustering of environmental compounds based on structure and toxicokinetic properties Michael Lawless, Rafał A. Bachorz, and Priyata Kalra **St Simulations Plus** 4070/K569 Simulations Plus, Inc. Research Triangle Park NC

INTRODUCTION

TK Analogues?



Traditional toxicokinetic (TK) models rely heavily on in vivo data, necessitating animal testing. At the same time, the scientific toolbox is expanding with new approach methodologies (NAMs) that do not rely on TK studies. One promising strategy (functional similarity¹, TK-equivalence, or read-across²) involves leveraging known in vivo data of a reference compound to develop and validate a TK model for a *target* compound without additional animal data. Of course, this assumes the absence of activity cliffs. Here, we compare three methods (structure, TK inputs, and TK outputs) to find a reference compound (nearest neighbor (NN)) for the target compound. The similarity of the Cp-time curves of each pair, defined as a distance, is used to judge success.



calculated with ADMET Predictor. There are scaled from 0 to 1 and distance is computed as the square root of the sum of squared scaled differences. The time scale of the Cp-time curves is standardized and input to an RNN-based autoencoder. The distance between two curves is calculated in the latent space.



Figure 2 – ADMET Predictor HTPK simulations. A 70 mg immediate release tablet is dosed to a 70 kg human subject (1 mg/kg). The ACAT model is used to simulate the fraction absorbed (%Fa). First pass extraction (FPE) is the percent of compound metabolized during its initial passage through the liver. The oral bioavailability (%Fb) is the percent of the dose reaching systemic circulation. The portion that enters systemic circulation distributes into organs and tissues. Systemic clearance combines renal and liver clearance.

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Figure 3 – t-SNE plots of various parameter spaces. The points are colored by the log of Cmax. TK outputs nicely clusters Cmax, the clusters are a little more scattered for TK inputs, and even worse for descriptors.



Figure 4 – The target's NN in **descriptor space** is 0.123. The Cp-time curves distance is 0.072. The %Fb, Cmax, and AUC are slightly different for the molecules.



Figure 5 – The target's NN in **TK input space** is 0.0595. The native solubility pH and first pass extraction are the two largest contributors to the distance. The Cp-time curve distance is 0.117, so it is less similar to the target than the descriptor NN.







Figure 6 – The target's NN in **TK output space** is 0.0092. The Cp-time curve distance is 0.077. This is slightly higher than descriptor NN's Dcp.



Figure 7 – The target's NN in **Cp-time space** is 0.015.



Figure 8 – Cp-time curves for the target and various neighbors. The y-axis is linear in the plot on the left and natural log on the right. As expected, the Cp-time NN (green line) overlays with the target curve the best. The descriptor NN is the next best match.









higher results



Developing TK models in the absence of *in vivo* data is extremely challenging. We compared different methods that find the nearest neighbor of a target compound. The similarity of the Cp-time curves from each method was calculated from an RNN-based autoencoder. In the first and last examples in this poster, the descriptor NN has similar Cp-times to the target compound. In the second example, the descriptor NN was very similar structurally to the target, but the Cp-time curves were quite different. This is a structural "activity cliff", i.e., visually similar compounds have different Cp-time curves. The simulations allow one to determine which inputs to the TK simulations are responsible for differences in the Cp-time curves. Ideally, these inputs would be from *in vitro* experiments.

Ellison C. Regulatory Toxicology and Pharmacology, 2018, 99, 61-77. ² Escher SE et al. (2021). *Read-Across Methodology in Toxicological Risk* Assessment. In: Reichl, FX., Schwenk, M. (eds) Regulatory Toxicology. Springer, Berlin, Heidelberg.

³ Pearce RG et al. *J. of Statistical Software*, **2017**, 79(4), 1-25. ⁴ ADMET Predictor version 12, Simulations Plus, Inc. Durham NC 27709

SUMMARY

REFERENCES