

High Throughput Pharmacokinetics for Environmental Chemicals

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Pharmacokinetic (PK) models are critical to determine whether chemical exposures produce potentially hazardous tissue concentrations. For bioactivity identified *in vitro* (e.g. ToxCast) – hazardous or not – PK models can forecast exposure thresholds, below which no significant bioactivity is expected. Successful methods have been developed for pharmaceutical compounds to determine PK from limited *in vitro* measurements and chemical structure-derived property predictions. These high throughput (HT) PK methods provide a more rapid and less resource-intensive alternative to traditional PK model development. Unfortunately, predictions from HTPK approaches have demonstrated mixed success for environmental chemicals when compared to predictions made by PK models developed with extensive *in vivo* data. Here we tested assumptions of previous HTPK approaches using a simple physiologically-based PK (PBPK) model and *in vitro* data for 232 chemicals in human and 39 chemicals in rat. We then analyzed the discrepancy between the predictions of HTPK and *in vivo* literature PK data for 44, mostly pharmaceutical, chemicals, using the method of best subsets to identify those properties that correlate with poor predictive ability (e.g., *in vitro* HTPK data, physico-chemical descriptors, chemical structure, and predicted transporter affinities). We propose a framework for PK triage in stages: First, *in vitro* measurements and *in silico* predictions determine whether the simplest HTPK approaches are likely to be sufficient. Then identify and collected any additional, targeted *in vitro* data that is needed. Finally, identify those chemicals most likely to require traditional, *in vivo* PK methods. This methodology allows prioritization of PK resources and characterizes the confidence in HTPK model predictions for potentially thousands of environmental chemicals that currently have no PK data. *This abstract does not necessarily reflect EPA policy.*