Integrating Biopharmaceutic Data and Gastrointestinal Physiology Using Mechanistic Modeling

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### **PBPK-IVIVE linked models**

#### • Estimate fundamental parameters (deconvolution)

Process	<i>In vitro</i> system	Models	Fundamental parameters
Metabolism	Hepatocytes, human liver microsomes or recombinant enzymes	Michaelis-Menten model	K <sub>m</sub> and V <sub>max</sub>
Uptake Transport	Overexpressing cell lines suspended or plated	Mechanistic compartmental uptake models	$CL_{diff}$ , $f_{u,cell}$ , $K_m$ and $V_{max}$
Dissolution / Precipitation	One- or multi-stage dissolution apparatuses	Diffusion layer models, Z-factor, Mooney model, biphasic dissolution model, transfer model, transmembrane flux model, etc.	Dependent on apparatus and model choice

- Reassemble the process using PBPK modeling (convolution)
  - Integrate drug- and formulation-specific parameters with physiology



- Reverse translation: identifying a biopredictive dissolution method
  - Ibuprofen (BCS II weak acid): free acid vs sodium salt



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Modified from Legg et al, 2014. *Drugs RD* 14(4):283-290; Dewland et al, 2009. *BMC Clin Pharmacol* 9:19; Modified from Cristofoletti et al, 2017. *J Pharm Sci* 106(1):92-99.

- Model-based analysis of in vitro dissolution data (deconvolution)
  - Simultaneous fitting of a DLM to multiple *in vitro* dissolution profiles
  - Deriving a product-specific particle size distribution (P-PSD)



Lines: predicted in vitro dissolution of ibuprofen



- Integration drug and formulation parameters with physiology (convolution)
  - Simulating in vivo dissolution profiles





1.2

#### • Virtual BE trials

- Reference, test BE and test non-BE formulations containing ibuprofen free acid
- Post-hoc assignment of WSV to BE metrics ("Fixed subjects")



OBS 90% CI of C<sub>max</sub>: 1.06 – 1.21 *Post-hoc* 90% CI of C<sub>max</sub>: 0.97 – 1.13



OBS 90% CI of C<sub>max</sub>: 1.18 – 1.35 Post-hoc 90% CI of C<sub>max</sub>: 1.04 – 1.26



• Model-based analysis of *in vitro* precipitation data from different systems

Aqueous SIM

Organic SIM

Aqueous OBS

Organic OBS

210

240

150

180

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• Ketoconazole (BCS II weak base)



Kambayashi et al, 2016. Eur J Pharm Biopharm 103:95-103; O'Dwyer et al, 2020. Pharmaceutics 12:272.

#### IVIVE-PBPK modeling

- Ketoconazole solution 200 mg fasted
  - Different *in vitro* systems, different results



100

80

• OBS f preciptiated 100 mg

Dumping

OBS f precipitated 300 mg

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Psachoulias et al, 2011. Pharm Res 28:3145-3158; Huang et al, 1986. Antimicrob Agents Chemother 30(2):206-210; Pathak et al, 2017. Mol Pharm 14(12):4305-4320; Cristofoletti et al. 2017 J Pharm Sci 106(2):560-569.

#### IVIVE-PBPK modeling

- Ketoconazole solution 400 and 800 mg fasted
  - Dose-dependent first-order precipitation rate





- Model-based analysis of *in vitro* data is helpful to derive fundamental input parameters for PBPK models, however navigating between different *in vitro* models might be challenging;
- Generalization of first-order precipitation rate across different doses is not straightforward;
- Further research is needed to optimize propagation of WSV through simulations.



#### Thank you!

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