

Assessment of food effects during clinical development

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20 **Abstract**

21 Food-drug interactions frequently hamper oral drug development due to various physicochemical,
22 physiological and formulation-dependent mechanisms. This has stimulated the development of a
23 range of promising biopharmaceutical assessment tools which, however, lack standardized settings
24 and protocols. Hence, this manuscript aims to provide an overview of the general approach and
25 the methodology used in food effect assessment and prediction. For *in vitro* dissolution-based
26 predictions, the expected food effect mechanism should be carefully considered when selecting
27 the level of complexity of the model, together with its drawbacks and advantages. Typically, *in*
28 *vitro* dissolution profiles are then incorporated into physiologically based pharmacokinetic
29 models, which can estimate the impact of food-drug interactions on bioavailability within 2-fold
30 prediction error, at least. Positive food effects related to drug solubilization in the GI tract are
31 easier to predict than negative food effects. Preclinical animal models also provide a good level of
32 food effect prediction, with beagle dogs remaining the gold standard. When solubility-related
33 food-drug interactions have large clinical impact, advanced formulation approaches can be used
34 to improve fasted state pharmacokinetics, hence decreasing the fasted/fed difference in oral
35 bioavailability. Finally, the knowledge from all studies should be combined to secure regulatory
36 approval of the labelling instructions.

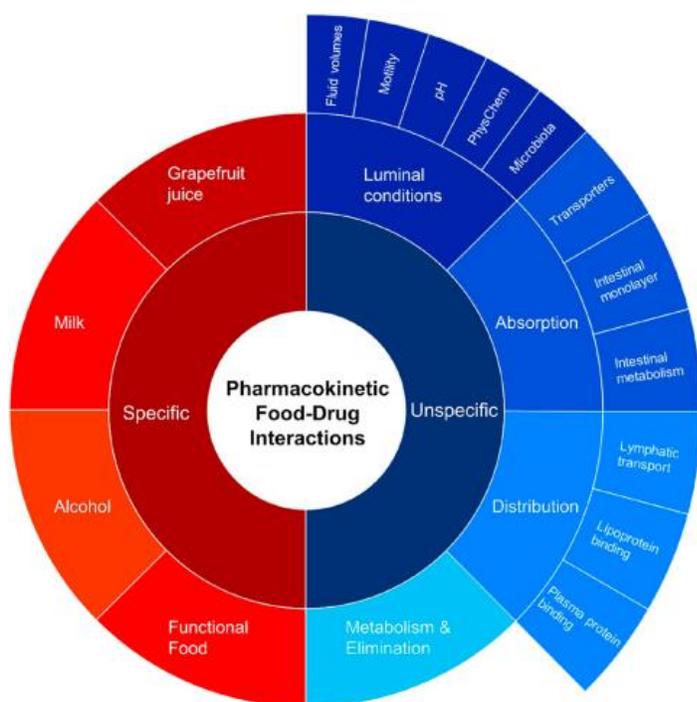
37 **Keywords**

38 Food-drug interactions; *in vitro*; *in silico*; *in vivo*; formulation

39

40 **Introduction**

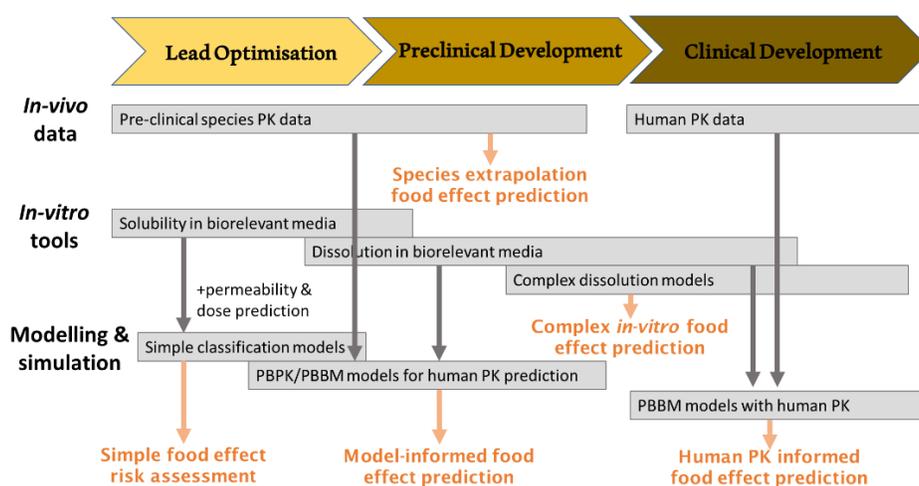
41 Food-drug interactions often present a significant challenge during the development of oral
42 medicines, due to their influence on drug pharmacodynamics and pharmacokinetics (PK). In
43 particular, food may have a substantial impact on drug absorption and metabolism, which will be
44 reflected in the measured PK parameters. The high degree of complexity when dealing with food
45 effects on oral bioavailability arises from the diversity of underlying mechanisms (**Figure 1**),
46 which can originate from the drug physicochemical properties, the formulation technology or the
47 physiology (for details see the review of Koziolok *et al.*, 2019a) and the difficulties in predicting
48 such food effects at the pre-clinical stage (Bennett-Lenane *et al.*, 2022; Koziolok *et al.*, 2019a).



49
50 **Figure 1.** Summary of specific and unspecific pharmacokinetic food-drug interactions. Reprinted
51 from Koziolok *et al.* 2019a, Creative Commons CC-BY license.

52

53 As a result, regulatory agencies generally require submission of pharmacokinetic data after food
 54 intake from the pharmaceutical industry to support labelling instructions (FDA, 2002, 2022).
 55 Hence, the study of food effects, their mechanisms and their impact on drug safety and efficacy
 56 has attracted considerable interest. A wide variety of *in silico*, *in vitro* and *in vivo* methods
 57 (Figure 2) have been developed to assess the various mechanisms and implications of food effects
 58 (Chen *et al.*, 2018; Koziolok *et al.*, 2019a; Koziolok *et al.*, 2018; Veerman *et al.*, 2020). Some of
 59 those methods have been described in a recent review (Wilson *et al.*, 2022).



60
 61 **Figure 2.** Food effect prediction workflow in pharmaceutical development.

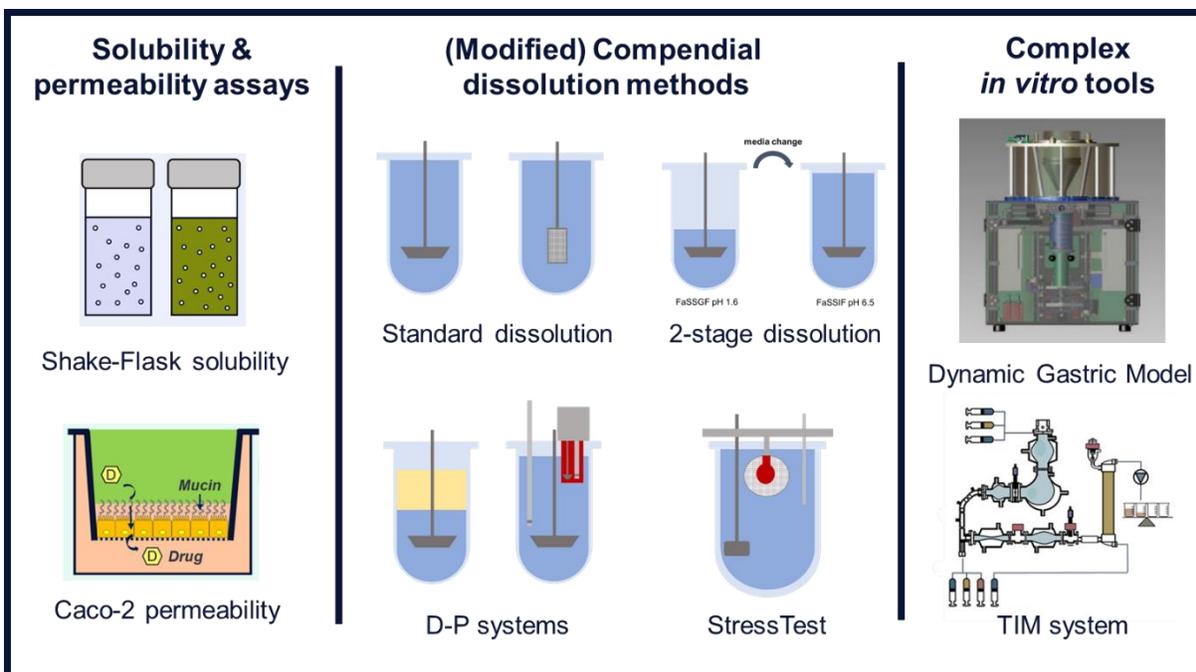
62 At the same time, method selection depends on the goal of food effect evaluation and on the
 63 stage of drug development: for example, early assessment protocols serve to estimate the risk of
 64 significant food effects in the clinic, largely based on drug properties alone. Recently,
 65 physiologically based pharmacokinetic (PBPK) modelling has gained larger attraction also for
 66 food effect prediction at preclinical stages. As a project approaches first-in-human dosing, pre-
 67 clinical *in vivo* data and formulation specific *in vitro* data can be used to attempt to prospectively
 68 predict clinically relevant effects of food intake on drug PK in humans. Finally, once clinical PK
 69 data is available, this can be used to guide further formulation development (*e.g.* to develop a

70 formulation with a reduced food effect) and to further refine *in silico* and *in vitro* methods (Figure
71 1).

72 Although the recently published Food and Drug Administration (FDA) guidance for assessing
73 the food effects provides an updated regulatory perspective on the topic (FDA, 2022), it does not
74 include an overview of the various methodologies that are actually being used to assess the impact
75 of food by the pharmaceutical industry. Hence, this review aims to describe the current practices
76 in the application of *in vitro*, *in vivo* and *in silico* tools for food effect assessment in the context of
77 the drug development stage and to provide an overview of the respective regulatory and clinical
78 development considerations.

79 ***In vitro* prediction tools**

80 *In vitro* prediction tools can be used to predict the *in vivo* performance of a drug product in humans
81 after administration of food, relative to fasted state, especially when the dissolution of the drug in
82 the gastrointestinal (GI) lumen is the primary driver for a food effect. In practice, this means that
83 food effect prediction via *in vitro* tools commonly focuses on drugs with poor aqueous solubility,
84 which often display positive food effects on oral drug bioavailability. Such drugs belong to class
85 2 or 4 of the biopharmaceutical classification systems (BCS). This area of focus is logical as poorly
86 water-soluble drugs are very common in modern pharmaceutical company portfolios, and as they
87 are also more likely to display clinically significant food effects. For BCS 1 and 3 drugs, clinically
88 significant food effects are somewhat less frequently encountered, and due to high drug solubility,
89 may be related to the impact of the fed state environment on aspects beyond the dissolution of the
90 drug product. In the following sections, we will address the some of the most frequently used in
91 vitro tools, which can vary greatly in their complexity and ability to mimic the real situation in the
92 human gastrointestinal tract, see **Figure 3**.



93

94 **Figure 3.** Schematic representation of the types of in vitro models used to study food effects.

95 D-P denotes “dissolution-permeation” and TIM denotes “TNO Gastro-Intestinal Model”. The

96 sketch of the Caco-2 permeability setup was obtained from Ye *et al.* 2022, the Dynamic Gastric

97 Model sketch was obtained from Mann and Pygall 2014 and the TIM sketch was obtained from

98

López Mármol *et al.* 2022.

99 ***Simple solubility- and permeability-based models for food effect prediction***

100 Solubility in biorelevant media is often used as a starting point for food effect prediction for poorly

101 water-soluble drugs when new drug candidates are identified. Solubility in fasted and fed state

102 simulated intestinal fluids (FaSSIF/FeSSIF) has been shown to reflect that observed in human

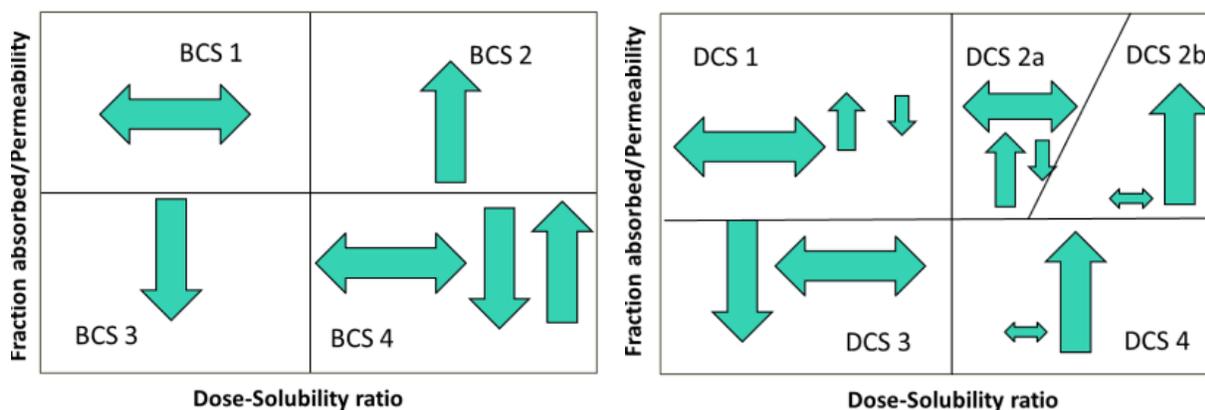
103 aspirates reasonably well, in both the fasted and fed state (Augustijns *et al.*, 2014). However, it is

104 only if a drug’s absorption is incomplete (due to low solubility and/or slow dissolution) when

105 differences in FaSSIF/FeSSIF solubility potentially translate to a meaningful difference in

106 bioavailability. The BCS (Fleisher *et al.*, 1999; Ku, 2008; O’Shea *et al.*, 2019) and the related

107 Biopharmaceutics Drug Disposition Classification System (BDDCS) (Benet, 2013) have been
 108 proposed as tools for use in the prediction of food effects. The typical assumptions for how food
 109 effects vary with BCS class are shown in **Figure 4A**.



110
 111 **Figure 4.** (A) Postulated direction of food effect (fed/fasted ratio) on the bioavailability of orally
 112 administered drugs based on the Biopharmaceutical Classification System (BCS). (B) Postulated
 113 direction of food effect on the bioavailability of orally administered drugs based on the
 114 Developability Classification System (DCS). The size of the arrows represents the approximate
 115 frequency of a positive, negative, or no food effect being observed based upon a set of 131 oral
 116 drugs approved by the FDA between 2011 and 2017. A significant food effect was classified as a
 117 change in AUC of 15% or greater, irrespective of whether this was deemed a clinically
 118 significant difference.

119
 120 However, as BCS is primarily designed to identify risks of bio-inequivalence in a regulatory
 121 setting, it is therefore by nature conservative when determining if an actual *in vivo* effect is likely.
 122 For instance, the common assumption that BCS 2 drugs are likely to have positive food effects
 123 does not necessarily hold true, as many BCS 2 drugs can be formulated in a manner that allows
 124 almost completely absorption even in fasted state, thus eliminating the potential for a solubility-
 125 related food effect.
 126 The Developability Classification System (DCS) system (Butler and Dressman, 2010), which was
 127 developed with early development biopharmaceutics questions in mind, including the propensity
 128 for food effects, is a more discriminative tool than BCS in predicting solubility-related food

129 effects. It uses solubility in FaSSIF as the arbiter of whether a drug is high or low solubility, and
130 subdivides BCS class 2 drugs into class 2a (dissolution rate-limited) and class 2b (solubility-
131 limited) drugs. As shown in **Figure 4B**, the solubility-limited drugs (DCS class 2b and 4) have the
132 highest propensity to show positive food effects.

133 The true picture of how food effect relates to BCS/DCS class is complex, due to the multiple, and
134 sometimes poorly understood factors involved, some of which are inadequately captured in a
135 simple solubility/permeability framework. It is worth noting that whilst BCS/DCS class 3 drugs
136 have a greater risk of negative food effects, they are equally likely to display no significant food
137 effect. As could be expected, BCS/DCS class 1 drugs rarely show meaningful food effects.

138 ***Compendial dissolution methods to predict food effects for poorly water-soluble compounds***

139 When evaluating formulations for food effects, comparative dissolution generated in a compendial
140 apparatus, such as the paddle (USP apparatus 2) method in FaSSIF/FeSSIF can be used at initial
141 stages. The dissolution profiles can be used directly to indicate a food effect by the difference
142 between the fasted and fed states. Alternatively, the dissolution profiles may be incorporated into
143 a PBPK or a physiologically-based biopharmaceutics model (PBBM) to account for other factors
144 potentially influencing the actual food effect. Working with the first widely applied versions of
145 bile salt micelle-containing biorelevant media, Galia *et al.* demonstrated that dissolution in
146 FaSSIF/FeSSIF (version 1) could broadly predict the observed food effect in humans for the
147 neutral, low solubility drug danazol (Galia *et al.*, 1998), whilst Nicolaidis *et al.* demonstrated that
148 differences in human bioavailability in fasted/fed state for four low solubility neutral/weak acid
149 drugs were also predicted from the *in vitro* data (Nicolaidis *et al.*, 1999). In addition, human
150 pharmacokinetic data in the fasted and fed state has been shown to be reasonably well correlated

151 to FaSSIF/FeSSIF dissolution profiles for a wider set of poorly water soluble compounds (Mathias
152 *et al.*, 2015).

153 Since the publication of the original biorelevant media recipes in the late 1990's, modified
154 intestinal media (version 2), plus media for the fed state gastric environment (Jantratid *et al.*, 2008)
155 were proposed. In addition, newer versions incorporate the products of lipid digestion into
156 simulated intestinal media (Fuchs *et al.*, 2015; Jantratid *et al.*, 2008). Subsequent to the
157 introduction of biorelevant dissolution media, the incorporation of dissolution data into PBPK
158 models has been demonstrated to be an invaluable approach with numerous publications
159 advocating their use (Kushwah *et al.*, 2021; Otsuka *et al.*, 2013; Shono *et al.*, 2009; Shono *et al.*,
160 2010).

161 For modified and extended-release oral products, attempts have been made to predict fasted and
162 fed state performance using flow-through (USP apparatus 4) and reciprocating cylinder (USP
163 apparatus 3) set ups. Both set ups allow multiple biorelevant media changes to mimic the transit
164 of a dosage form through the GI tract. Andreas *et al.* demonstrated that for two nifedipine ER
165 formulations, the reciprocating cylinder method was shown to qualitatively predict the positive
166 food effect, although the flow-through method was less predictive (Andreas *et al.*, 2016). Both
167 these compendial set ups have also been used with success to predict the impact of food on
168 mesalamine formulations (Andreas *et al.*, 2015). As well as being used for extended-release
169 formulations, the flow-through apparatus with biorelevant media has also been shown to predict
170 the food effect of immediate release formulations (Kushwah *et al.*, 2021; Sunesen *et al.*, 2005).
171 However, these compendial methods, even with multiple media changes, miss many motility-
172 related events *in vivo*, especially the strong peristaltic movements associated with gastric emptying
173 of residual solids and meal components (Koziolek *et al.*, 2018).

174

175

176 ***Modified compendial set ups***

177 Whilst FaSSIF/FeSSIF dissolution comparisons may be useful, and certainly add physiological
178 relevance in terms of micellar solubilization over simple buffer solutions, there are caveats in their
179 use which can lead to under- or over-prediction of an *in vivo* food effect, especially if the fasted/fed
180 ratio is estimated directly from the *in vitro* data. These include:

- 181 a) Differences in dissolution rate and/or solubility *in vitro* in FaSSIF/FeSSIF will not translate
182 directly into *in vivo* differences for drugs where suitable formulation and size control
183 strategies have been employed to ensure close to complete absorption in the fasted state.
184 For some poorly water-soluble compounds, adequate control of particle size can therefore
185 lead to the elimination of food effects (Butler and Dressman, 2010; O'Shea *et al.*, 2019)
- 186 b) For drugs, which supersaturate *in vivo* such as some low solubility weak bases, and for
187 formulations which utilize supersaturation as a bio-enabling strategy, simple dissolution
188 experiments directly in FaSSIF/FeSSIF will not capture the potentially critical gastric
189 dissolution process, nor adequately reflect gastric emptying kinetics or any subsequent
190 saturation/precipitation.
- 191 c) The micellar components in food (and in the *in vitro* set ups), whilst typically increasing
192 bulk drug concentration in solution, may entrap dissolved drug in the small intestinal
193 lumen, reducing the free drug concentrations, and therefore reducing the availability of
194 drug for absorption at the gut wall (Miller *et al.*, 2011).
- 195 d) *In vivo* impact of food intake that is unrelated to drug dissolution and solubility, such as
196 the impact of binding to specific food components like trypsin (Lee *et al.*, 2016), the

197 influence of food on pre-systemic drug metabolism (Melander et al., 1988), or the impact
198 on efflux transporters (Sharma and Prasad, 2021) will clearly not be accounted for in a
199 typical dissolution-based *in vitro* model.

200 To overcome some of these limitations, modifications to compendial paddle methods have been
201 proposed in recent years to improve biorelevance. These include:

202 1) Adding an absorption stage to the dissolution test, to mimic permeation across the gut wall,
203 which is thought to be primarily accessible to the free drug, rather than to strongly micellar
204 bound drug. There are several different methods reported in the literature to modify
205 compendial set ups to achieve this. One approach is to use an immiscible organic liquid
206 layer such as octanol (Frank *et al.*, 2014; Mudie *et al.*, 2012; Xu *et al.*, 2017), in the
207 compendial apparatus. However, these biphasic methods need to be used with caution with
208 micelle-containing media (due to possible emulsification of octanol), so their application
209 to food effect prediction may be limited. Even so, their use with biorelevant media in food
210 effect prediction has been reported (Xu *et al.*, 2017). Alternatively, a semi-permeable
211 membrane that only allows the permeation of free drug, rather than micelle bound drug can
212 be used. A range of set ups have been proposed for potential use in combination with
213 compendial dissolution apparatus (Berben *et al.*, 2018a; Berben *et al.*, 2018b; Borbás *et*
214 *al.*, 2019; Borbas *et al.*, 2018; Hens *et al.*, 2015). In this case, the surface-to-volume ratio
215 of the respective permeation method should be considered, as it often limits the transfer of
216 the drug to the acceptor compartment (complete transfer to the acceptor is usually not
217 achieved). A detailed review of the best practices in drug permeation assessment has
218 recently been published (O'Shea *et al.*, 2022).

219 2) Use of two-stage biorelevant dissolution in which the gastric and intestinal environments
220 are mimicked in sequence. This may be done with a simple transfer model (Kostewicz *et*
221 *al.*, 2004; Wagner *et al.*, 2012) in which drug is pre-dissolved in a simulated gastric media
222 and supersaturation/precipitation measured upon controlled transfer at a fixed rate to
223 intestinal media, with mixing in the intestinal media provided by the stirring action in a
224 standard paddle apparatus. The biorelevant media used, and the transfer rate can be altered
225 to represent that likely to be seen *in vivo*, including that observed in the fasted and fed
226 states (Litou *et al.*, 2020; Ruff *et al.*, 2017). Alternatively, a two-stage dissolution test set
227 up in which a second media is added to mimic the change from a gastric environment to an
228 intestinal environment may be used (Berben *et al.*, 2019; Mann *et al.*, 2017). Using a
229 methodology which combines both two-stage biorelevant dissolution, and the use of a
230 permeation bag to mimic the permeation barrier, Hens *et al.* determined the free drug
231 concentrations available for absorption for two formulations of fenofibrate, in both the
232 fasted and fed state (Hens *et al.*, 2015). This work demonstrated that it was the free drug
233 concentrations that were key to predicting the actual food effects observed *in vivo* with the
234 two formulations. One potential disadvantage with two-stage methods is that typically, an
235 intestinal medium is added to the gastric media rapidly at an uncontrolled rate. This rapid
236 addition of a second medium contrasts with comparatively slower gastric emptying *in vivo*,
237 especially in the fed state.

238 3) Replacement of the paddle or basket for agitation with pressure application devices to
239 simulate the forces associated with gastrointestinal motility and transit. This has been
240 explored through the use of the Stress Test apparatus, developed at the University of
241 Greifswald (Garbacz *et al.*, 2010). In terms of food effect prediction, this apparatus has

242 been shown to be especially advantageous in the assessment of extended-release matrix
243 tablets (Garbacz *et al.*, 2009; Garbacz *et al.*, 2008; Koziolok *et al.*, 2013).

244 Ultimately, although compendial based set ups can provide useful insights - provided appropriate
245 biorelevant media are used - the design of the currently available compendial apparatus restricts
246 the opportunities for adequate simulation of the highly dynamic GI environments *in vivo*, meaning
247 more complex *in vitro* tools and/or the incorporation of dissolution data into a PBPK model which
248 can account for these other factors may be required for reliable food effect prediction.

249 ***Complex in vitro tools to predict food effects for poorly water-soluble compounds***

250 Complex *in vitro* tools that have shown benefit in the prediction of food effects for drug products
251 include the TIM-1 / tiny-TIM systems (Verwei *et al.*, 2016), as well as the Dynamic Gastric Model
252 (DGM) / Model Gut system (Thuenemann *et al.*, 2015). Typically, these systems were developed
253 for understanding of the interplay between GI motility, food digestion and nutrient dissolution. In
254 addition to the TIM and DGM systems discussed below, there are a wide range of other complex
255 *in vitro* tools applied in food science that could theoretically be used to understand and predict
256 food effects of oral drug products. Several comprehensive reviews of these systems are available
257 (Dupont *et al.*, 2019; Li and Kong, 2022). It's also worth noting that based on the ability of TIM
258 systems to predict relative pharmacokinetic performance of different formulations, their
259 application to completely replace pre-clinical models for formulation performance evaluation has
260 been proposed and adopted by some pharmaceutical companies (Dickinson *et al.*, 2012; Barker *et*
261 *al.*, 2014).

262 The TIM systems and the DGM model are designed to mimic the dynamic situation resulting from
263 secretions, digestion, transfer of material and motility in the human GI tract. Originally developed
264 with applications to the food industry in mind, these systems have the capability to test drug

265 products in the presence of the exact meal used in any clinical study, with the meal being added to
 266 the model after being homogenized, or by actual chewing by the operator during the experiment
 267 set up. A summary table of TIM model applications to predict food effects is shown in Table 1.
 268 As can be seen from the table, Verwei *et al.* showed that TIM-1 and tiny-TIM models correctly
 269 predicted the positive food effect for a posaconazole suspension, and the lack of a food effect for
 270 an immediate release ciprofloxacin tablet formulation. However, both systems overpredicted the
 271 positive food effect of the Noxafil® suspension. This discrepancy between the *in vitro* and *in vivo*
 272 data might be explained by the high permeability of posaconazole, which partially compensates
 273 the poor solubility in fasted state human intestinal fluids. Ojala *et al.* demonstrated for immediate
 274 release formulations of a poorly water-soluble, weakly basic drug that the TIM-1 model was a
 275 more reliable predictor of fasted/fed pharmacokinetics than simpler compendial set-ups with
 276 biorelevant media (Ojala *et al.*, 2020). In addition, Lloyd *et al.* were able to show that the TIM-1
 277 model could be predictive of a negative food effect observed for the low solubility, zwitterionic
 278 drug danirixin (Lloyd *et al.*, 2020).

279 **Table 1.** Prediction of food effects using TIM systems.

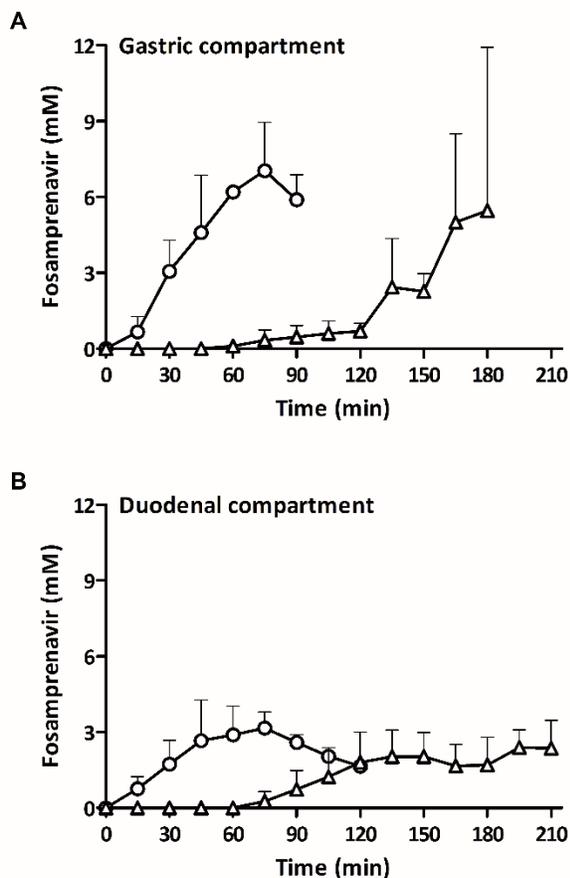
API	Formulation	Meal type	<i>In vivo</i> fed/fasted ratio	TIM <i>in vitro</i> fed/fasted ratio	Publication TIM data
Danirixin	DNX HBr	High fat meal	0.6 (AUC _{0-inf})	0.6 (TIM-1)	(Lloyd <i>et al.</i> , 2020)
Diclofenac	Cataflam IR	Ensure Plus	1.0 (AUC _{0-8h})	1.0 (TIM-1)	(Van Den Abeele <i>et al.</i> , 2017)
Ciprofloxacin	Ciproxin ER	High fat meal	1.0 (AUC)	1.2 (TIM-1) 1.0 (tiny-TIM)	(Verwei <i>et al.</i> , 2016)
Acetaminophen	Paracetamol IR	High caloric meal	0.9 (AUC _{0-inf})	1 (TIM-1)	(Souliman <i>et al.</i> , 2006)
Acetaminophen	Sinaspril *crushed	Infant formula	No food effect	No food effect (tiny-TIM _{pediatrics})	(Havenaar <i>et al.</i> , 2013)

Fosamprenavir	Telzir IR	Scandi-shake Mix	No food effect AUC Effect on disintegration	No food effect bioacc. Effect on disintegration (TIM-1)	(Brouwers <i>et al.</i> , 2011)
Celecoxib	Celebrex	High fat meal	1.6 (AUC _{0-inf})	2.0 (TIM-1)	(Lyng <i>et al.</i> , 2016)
Nifedipine	Adalat XL MR	High fat meal	1.7 (AUC _{0-9h})	3.5 (TIM-1) 3.6 (tiny-TIM)	(Verwei <i>et al.</i> , 2016)
Posaconazole	Noxafil Suspension	High fat meal	4 (AUC _{0-72h})	13.8 (TIM-1) 12.9 (tiny-TIM)	(Verwei <i>et al.</i> , 2016)
Undisclosed investigational drug	Tablets: doses 10-80mg	High fat meal	2.2 (AUC _{0-t}) at 10mg* 3.2 (AUC _{0-t}) at 80mg*	2.9 (tiny-TIM) at 10mg 2.7 (tiny-TIM) at 80mg	(Luo <i>et al.</i> , 2022)
Ibuprofen	Advil FR and Advil LG	High fat meal	0.9 (AUC Advil FR)* 0.9 (AUC Advil LG)*	No food effect (tinyTIM Advil FR) No food effect (tinyTIM Advil LG)	(Chiang <i>et al.</i> , 2022)

280 *TIM data incorporated into a PBPK model to optimally predict AUC

281 The data in the table demonstrates that human food effects can be adequately predicted by the TIM
282 models. Even so, some caution is needed – the magnitude of the food effect for pozaconazole was
283 overpredicted, whilst not all the mechanisms leading to negative food effects are likely to be
284 captured by the model.

285 A specific advantage of using these predictive complex *in vitro* tools is that the mechanisms behind
286 specific food effects can be investigated and then confirmed by simpler *in vitro* methods. Lyng *et*
287 *al.* used the TIM-1 model to show that bile salt driven micellar solubilization was the primary
288 reason for the positive food effect for a celecoxib immediate release capsule (Lyng *et al.*, 2016).
289 Brouwers *et al.* used a combination of the TIM-1 model and separate imaging of disintegration by
290 MRI to show that differences in onset in the fasted and fed state for fosamprenavir tablets could
291 be linked to delays in tablet disintegration in the fed state, see **Figure 5** (Brouwers *et al.*, 2011).
292 Further scientific efforts will be needed to integrate information from complex *in vitro* systems
293 into PBPK models.



294
 295 **Figure 5.** Fosamprenavir concentration–time profiles in the stomach (A) and duodenum
 296 (B) compartment of TIM-1, simulating the fasted (open circles) and fed (open triangles) state.
 297 Results are expressed as mean \pm sd (n = 3). Reprinted from European Journal of Pharmaceutics
 298 and Biopharmaceutics, 77, Brouwers, J., Anneveld, B., Goudappel, G.-J., Duchateau, G.,
 299 Annaert, P., Augustijns, P., Zeijdner, E. “Food-dependent disintegration of immediate release
 300 fosamprenavir tablets: In vitro evaluation using magnetic resonance imaging and a dynamic
 301 gastrointestinal system”, 313-319, Copyright (2011), with permission from Elsevier.
 302
 303 Often, the simulation of GI physiology in the *in vitro* system and the *in silico* model are different,
 304 which makes direct integration of data very challenging. For instance, data from Tiny-TIM and

305 TIM-1 are used to verify predictions from PBPK modelling, but the information are typically not
306 used as direct inputs. To derive parameters such as dissolution rate or precipitation rate from the
307 complex in vitro experiments, in silico models must be developed, in which the in vitro experiment
308 is simulated.

309 Using the Dynamic Gastric Model (DGM), Vardakou *et al.* demonstrated that antral grinding
310 forces could be mimicked with much greater accuracy than using compendial dissolution apparatus
311 (Vardakou *et al.*, 2011a). Investigational work also showed that the model could predict the
312 differing drug release properties of various immediate release capsules in the fed and fasted state
313 (Vardakou *et al.*, 2011b). In addition, *in vitro* work on the DGM model has been used to show that
314 this system is likely to have specific advantages for investigating the dissolution properties of
315 extended-release matrices in the fed state, compared to fasted (Chessa *et al.*, 2014; Mason *et al.*,
316 2016).

317 One specific concern regarding the impact of food on the performance of oral dosage forms is that
318 of the impact on extended release matrices, where the influence of GI motility can play a critical
319 role in formulation robustness and drug release, sometimes leading to so called “dose dumping”
320 events, where a large proportion of the dose is released rapidly, circumventing the extended release
321 design of the product. In addition to the Stress Test apparatus mentioned in the previous section
322 on modified compendial apparatus, more complex tools such as TIM-1, TinyTIM and DGM which
323 are more commonly used to predict immediate release formulation performance in the presence of
324 food, may also be applied to understanding the *in vivo* behavior of extended release products
325 (Chessa *et al.*, 2014; Mason *et al.*, 2016). Note that *in vitro* tools to study the impact of food on
326 extended release formulations, have previously been reviewed in detail (Koziolok *et al.*, 2018),

327 whilst *in vitro* tools to study the impact of food on immediate release formulations have also been
328 the topic of a recent review article (Lex *et al.*, 2022).

329 ***In vivo* models for food effect predictions**

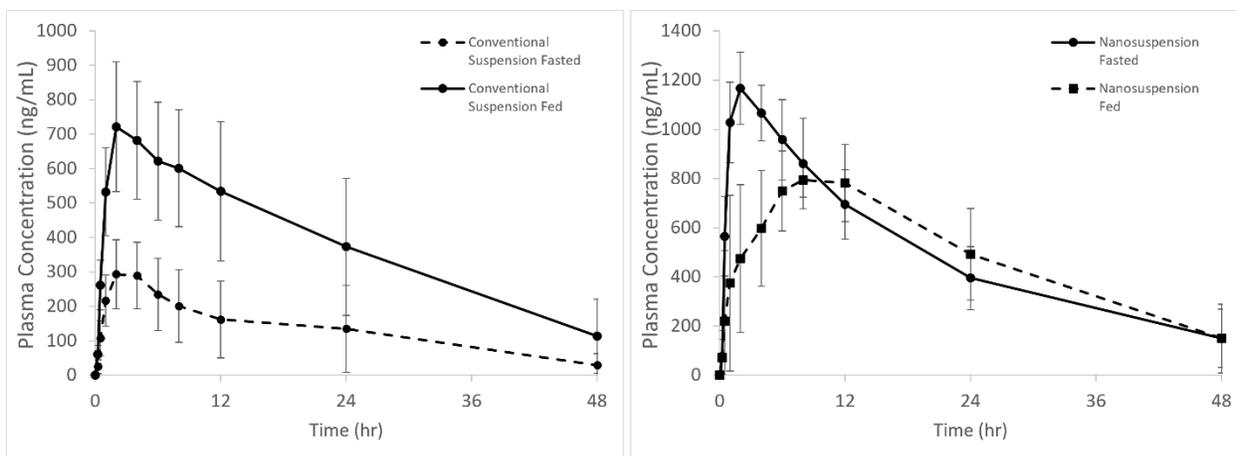
330 As highlighted in the previous sections, food effects on drug bioavailability are the result of the
331 complex interplay of different physiological factors that change after the intake of food (Koziolek
332 *et al.*, 2019a). Before complex and powerful *in vitro* tools (*e.g.* TIM-1, DGM) and *in silico* models
333 (*e.g.* SimCYP, GastroPlus) were made commercially available, food effect prediction was
334 primarily performed in animal models. Theoretically, different animal models such as mice, rats,
335 dogs, pigs or monkeys may be used for this purpose as they are available in pharmaceutical R&D
336 units. However, for the selection of the most suitable animal model, pharmaceutical scientists need
337 to take a deeper look at the following requirements:

- 338 1. The animal model should be able to simulate the conditions of the human GI tract in
339 both fasted and fed state. One of the major challenges is not only to simulate fed state
340 conditions in a way that is comparable to the human situation, but also to enable a
341 realistic assessment of drug product performance in fasted state. Only if both, fasted
342 and fed state, are simulated correctly, a food effect on oral bioavailability can be
343 predicted.
- 344 2. The formulation plays an important role in the occurrence of food effects. It is therefore
345 not enough to simply administer neat API or simple suspensions/solutions to the
346 animal. Ideally, the finished drug product can be administered to the animal to make a
347 realistic food effect assessment. Moreover, a suitable protocol must be taken into place
348 to adequately simulate food effect studies in humans (FDA, 2002, 2022).

349 3. The animal GI tract can differ in various aspects from the human GI tract. Based on the
350 pharmacokinetic, pharmacological and physicochemical properties of the drug product,
351 certain mechanisms leading to food effects can be expected (Hatton *et al.*, 2015;
352 Koziolk *et al.*, 2019a; Sjogren *et al.*, 2014). Based on this expectation, some models
353 may be more relevant than others.

354 For mice and rats, which are used broadly during drug discovery and also at preclinical stages,
355 their GI anatomy and physiology (including the digestive enzymes) is highly different from the
356 human GI tract (Hatton *et al.*, 2015; Koziolk *et al.*, 2019a). Moreover, larger formulations cannot
357 be administered to these animals. Therefore, they may be used to elucidate certain mechanisms
358 potentially leading to food effects (Holmstock *et al.*, 2013), but they do not represent ideal models
359 for an accurate prediction of food effects on oral bioavailability. On the other hand, for monkeys,
360 which are considered to be the best model for oral bioavailability prediction in humans (Muster
361 *et al.*, 2014), there is very limited experience with food effect prediction. Although the
362 physiological conditions in fed cynomolgus monkeys have been characterized and compared to
363 the human situation in two studies by Kondo and colleagues (Kondo *et al.*, 2003a; Kondo *et al.*,
364 2003b), a standard protocol on how to simulate fed conditions in monkeys has not been established
365 yet. Moreover, due to the small size of the cynomolgus monkeys (< 10 kg), it is probably difficult
366 to administer larger formulations. Therefore, monkeys are typically not used for food effect
367 predictions. Instead, the Beagle dog represents the most widely used animal model for human food
368 effect prediction. In the last years, some groups also reported on the use of pigs for food effect
369 prediction. In the following text, we will therefore focus on these two animal models and discuss
370 their potential application based on selected case examples.

371 In many pharmaceutical companies, the Beagle dog is the primary animal model to predict food
372 effects on oral bioavailability. First studies on the application of this model for simulation of drug
373 product performance in fed state have been published more almost 40 years ago (Cox *et al.*, 1985;
374 Shiu *et al.*, 1989). Therefore, there is large experience within the pharmaceutical industry on the
375 application of this animal model. However, whereas various guidance documents were issued by
376 regulatory authorities on food effect studies in humans (EMA, 2012; FDA, 2002), there is still no
377 standard protocol in terms of pre-treatment, type and timing of food intake, fluid intake during
378 administration as well as subsequent food or liquid intake for food effect studies in dogs. Studies
379 in which the dog model was successfully applied to predict drug product performance in presence
380 of food, often have anecdotal character and can hardly be compared to other food effect studies in
381 dogs. Nonetheless, the dog model can provide useful insights into drug product performance in
382 fed state. For instance, Wu and colleagues nicely illustrated how a dog model was used to support
383 the development of a nanocrystalline formulation of MK-0869 (aprepitant). Canine data could
384 demonstrate that this formulation has a reduced food effect as compared to a conventional
385 suspension, see **Figure 6** (Wu *et al.*, 2004). However, only few systematic studies on the use of
386 dogs for food effect prediction have so far been performed (Lentz *et al.*, 2007; Mathias *et al.*, 2015;
387 Zane *et al.*, 2014). In this context, one of the most relevant articles was published in 2007 by Lentz
388 and colleagues, who studied the impact of the study protocol and investigated the correlation
389 between food effect in dogs and humans (Lentz *et al.*, 2007). Based on two model compounds
390 (atazanavir and pravastatin), it was first shown that, to achieve the best correlation to human data,
391 a 50 g aliquot of the FDA meal should be used and that dogs should be pretreated with pentagastrin
392 to stimulate gastric acid secretion in fasted state.
393



394
 395 **Figure 6.** Assessment of food effect for conventional (left) and nanosized (right) suspensions in
 396 dogs. Based on data from Wu et al, Int J Pharm, 285 (2004), 135-146.

397 The optimized protocol was then applied in three Beagle dogs, who received nine different drug
 398 products with different types of food effect (*i.e.* negative, positive or no food effect) in a cross-
 399 over design. This dog model was able to capture positive food effects for drugs which also showed
 400 positive food effects in humans. Also, for drugs with negative food effects, it indicated the correct
 401 direction of the food effect. However, there was a slight tendency to overestimate drug product
 402 performance in fed state and therefore, for two out of three drugs, which showed no food effects
 403 in humans, a positive food effect was seen in dogs. This study was one of the first to provide a
 404 scientific basis for the application of a preclinical dog model, but the small sample size is a major
 405 limitation, especially if the huge variability is considered that is often seen in dog studies.

406 In a follow-up study by Mathias, 15 different compounds were studied in dogs and PK data were
 407 again compared to human data (Mathias *et al.*, 2015). Here, the food effect ratio in dogs correlated
 408 linearly with the food effect ratio in humans ($R^2 = 0.74$). Again, the dog model was able to predict
 409 the direction of food effects in most cases, whereas the extent was not always predicted correctly.
 410 Another interesting study was published by Zane and colleagues in 2014, who used the dog model
 411 to study the performance of different formulations of four drugs (Zane *et al.*, 2014). This study

412 was performed in a cross-over design with eight Beagle dogs that were pretreated with
413 pentagastrin. Despite the fact that very different formulation concepts were compared to each other
414 (*e.g.*, capsules vs. tablets, salt vs. lipid based formulations), the authors found a clear relationship
415 between canine and human data. In each case, the dog model was able to predict the direction of
416 food effects. However, it was not able to adequately predict the extent of the food effect seen in
417 humans for the different formulations tested.

418 A correct prediction of the food effect on oral bioavailability is often impeded by certain
419 differences in terms of canine GI anatomy and physiology as compared to humans. Recently,
420 Koziolk and colleagues used the SmartPill to further study the physiological conditions in dogs
421 under different prandial conditions as well as after different pretreatments (pentagastrin and
422 famotidine) (Koziolk *et al.*, 2019b). The data could be directly compared to similar data obtained
423 in humans that were generated earlier by the same authors. Interestingly, canine and human GI
424 physiology were comparable in various aspects such as gastric or intestinal pH. However, some
425 important differences were noted in terms of gastric transit time in fed state, small intestinal transit
426 time as well as in gastrointestinal pressures. All these parameters can play an important role for
427 oral drug delivery and thus, they may affect the prediction of food effects. It should be noted that
428 parameters such as gastric pH or gastric residence time highly depend on the type of meal used in
429 these studies. Therefore, the protocol can be of major importance for the outcome of food effect
430 predictions. Unlike in humans, where the FDA has issued a guidance on how to perform food
431 effect studies, the protocols used in the pharmaceutical industry differ among the different
432 companies. For instance, different meals such as dog food or shredded FDA meal are used
433 depending on the individual protocol. In addition, there are further differences between human and
434 dogs in terms of paracellular absorption as well as in terms of enzyme and transporter expressions

435 (Martinez *et al.*, 2019). Thus, data from dog studies should always be interpreted with care and
436 further data from *in vitro* and *in silico* models should confirm the findings.

437 Another animal model that may be useful for food effect prediction is the pig. This animal model
438 is widely used by food scientists to simulate digestive processes but also to model certain diseases.
439 However, its application in pharmaceutical R&D is rather limited. In recent years, Brendan Griffin
440 and team were studying the suitability of the pig model for food effect predictions. Despite the fact
441 that the simulation of fasted state conditions is complex in pigs due to slow gastric emptying of
442 digesta and in particular large objects (Henze *et al.*, 2021; Henze *et al.*, 2019), which limits the
443 application of this model for slowly or non-disintegrating monolithic dosage forms, the model may
444 be valuable for the prediction of food effects for immediate release formulations of poorly water-
445 soluble drugs as was shown recently for fenofibrate (Henze *et al.*, 2019). It will be interesting to
446 see if further studies will confirm this hypothesis and if this model will receive broader attention
447 for food effect prediction in case of drugs with poor aqueous solubility.

448 In conclusion, animal models such as the Beagle dog have been and still are valuable tools for
449 prediction of the direction of food effects on oral bioavailability and the assessment of formulation
450 performance in fasted/fed state. However, various physiological parameters differ significantly
451 between humans and laboratory animals commonly used for food effect prediction, which may
452 impair their predictive power. Generally, like in humans, the study protocol has huge impact on
453 the outcome of food effect studies in animals. In light of the 3R approach to reduce, replace and
454 refine the use of animal in pharmaceutical R&D, some companies have stopped using animal
455 models to support formulation development and food effect assessment. Apart from ethical
456 reasons, the relatively high costs associated with animal studies, the high variability often seen in
457 PK studies as well as the limited predictability with respect to human PK have been important

458 reasons for this decision. With further improvement of the various *in vitro* and *in silico* tools and
 459 their predictive power, the number of animal studies will most probably further decline in the
 460 coming years.

461 **Physiologically Based Pharmacokinetic modeling**

462 PBPK models have been historically utilized in the pharmaceutical industry primarily for first-in-
 463 human (FIH) dose predictions and for predicting drug-drug interactions (DDIs). With the
 464 expansion of PBPK models to modeling of oral absorption processes and guiding formulation
 465 development, there has been increased interest to the application of these models for food effect
 466 predictions, see Table 2. Since 2009, approximately 20 manuscripts have been published
 467 specifically discussing case studies of PBPK models applied to food effect
 468 prediction/characterization, covering more than 30, primarily BCS/BDDCS class 2 and 4 drugs.
 469 The principles and limitations of published PBPK models have been reviewed elsewhere
 470 (Kesisoglou, 2020; Li *et al.*, 2018).

471 **Table 2.** Summary of publications with PBPK models for food effect, listed chronologically
 472 (modified from Kesisoglou (Kesisoglou, 2020))

Publication	Compound	BCS	Food effect (AUC as primary endpoint)
(Parrott <i>et al.</i> , 2009)	Theophylline (CR)	I	None
	aprepitant	II	positive (micronized tablet), no (nanosuspension)
(Shono <i>et al.</i> , 2009)	Celecoxib	II	Positive
(Shono <i>et al.</i> , 2010)	Aprepitant	II	Positive/None (micron/nano - sized)
(Heimbach <i>et al.</i> , 2013)	Proprietary Compound (NVS732)	I	None
	Proprietary Compound (NVS406)	II	Positive
	Proprietary Compound (NVS701)	II	Positive
	Proprietary Compound (NVS113)	II	Negative

(Xia <i>et al.</i> , 2013)	Proprietary Compound (NVS123)	II	Positive
	Proprietary Compound (NVS169)	IV	None
	Proprietary Compound (NVS562)	II or IV	Positive
(Zhang <i>et al.</i> , 2014)	Proprietary Compound	II or IV	Positive
(Cristofolletti <i>et al.</i> , 2016)	Ketoconazole	II	Positive
	Posaconazole	II	Positive
(Parrott <i>et al.</i> , 2016)	Alectinib	II	Positive
(Sutton <i>et al.</i> , 2017)	Ziprasidone	II	Positive
(Rose <i>et al.</i> , 2017)	Propranolol	II	Positive
	Ibrutinib	II	Positive
(Andreas <i>et al.</i> , 2017)	Zolpidem MR	I	Negative
(Emami Riedmaier <i>et al.</i> , 2018)	Venetoclax	IV	Positive
(Tistaert <i>et al.</i> , 2019)	Proprietary Compound	I	None
	Mebendazole	II	Positive
	Bitopertin	II	Positive
	Proprietary Compound	II	None
(Radwan <i>et al.</i> , 2019)	Clarithromycin	II	None
(Gajewska <i>et al.</i> , 2020)	alpelisib	II	positive
(Lloyd <i>et al.</i> , 2020)	Danirixin HBr	IV	negative
(Arora <i>et al.</i> , 2020)	Ritonavir	IV	negative
	Ribociclib	II or IV	None
(Pepin <i>et al.</i> , 2021)	nefazodone-HCl	I	negative
	furosemide	IV	negative
	Aprepitant	II	Positive/None (micron/nano - sized)
(Wagner <i>et al.</i> , 2021)	pazopanib-HCl	II	positive
	ziprasidone-HCl	II	positive
	trospium-Cl	III	negative
(Kushwah <i>et al.</i> , 2021)	rivaroxaban	II	positive
(Jeong <i>et al.</i> , 2022)	tegoprazan	II	none
(Pepin <i>et al.</i> , 2022)	selumetinib	IV	negative

473

474 Evolution of the models over the years reflects the increased utilization of more complex *in vitro*

475 methodologies discussed earlier in this manuscript; while initial models largely focused on the

476 solubility differential in biorelevant media such as FeSSIF and FaSSIF, data from multi-

477 compartment systems to characterize dissolution and precipitation are now more commonly
478 utilized.

479 Models are typically applied first in the preclinical, pre-FIH stage, to assess the possibility of food
480 effect and inform formulation optimization or dosing instructions in the FIH study (Xia *et al.*,
481 2013). At this stage in the absence of clinical model validation, the primary focus is on prediction
482 of relatively large food effect differences (>2-fold) and especially for positive food effect, to
483 inform whether a different formulation approach should be implemented. The PBPK models are
484 typically used as orthogonal to studies in preclinical/dissolution models to drive a decision based
485 on totality of evidence. Once clinical food effect data are available, the model is refined for
486 application to provide further mechanistic insights to the observed food effect and inform
487 subsequent formulation efforts (Emami Riedmaier *et al.*, 2018; Tistaert *et al.*, 2019; Zhang *et al.*,
488 2014). Available clinical data allows for validation of the model and a decision whether the food
489 effect mechanism can be captured. Based on experience across several pharmaceutical companies,
490 Tistaert *et al.* recently proposed a workflow for implementation of food effect PBPK models
491 during preclinical development (Tistaert *et al.*, 2019). Given that not all food effect mechanisms
492 can be readily predicted, the authors recommended that model application focuses on
493 BCS/BDDCS class 2 drug formulated in IR drug products, with linear pharmacokinetics without
494 significant gut transporter involvement, where the major mechanisms for food effect is related to
495 luminal solubilization (*e.g.*, increase in bile salts and presence of fatty acids with meal) and/or
496 delay in gastric emptying. These recommendations are largely in agreement with a more recent
497 analysis published by Riedmaier *et al.* where authors, as part of an IQ Consortium effort, assessed
498 predictability of PBPK models in relation to the food effect mechanism and also concluded that

499 successful predictions were associated with changes in the gastrointestinal luminal fluids or
500 physiology (Riedmaier *et al.*, 2020).

501 At later stages of development, the desire is to use PBPK models for regulatory interactions, such
502 as replacing clinical studies. However, despite the numerous successful examples in the literature,
503 best practice and regulatory acceptance of PBPK models for food effect predictions are still
504 evolving. As a result, confidence in the models by regulators is still low (Li *et al.*, 2018).
505 Development of standardized input and model development workflows have been recently
506 proposed (Riedmaier *et al.*, 2020) as a step towards that direction. In practice, validation of the
507 prediction against early-stage clinical food effect data before use of the model for *a priori*
508 predictions, as recommended by Tistaert *et al.* and Kesisoglou (Kesisoglou, 2020; Tistaert *et al.*,
509 2019), is likely going to be a prerequisite for model application at later development stages and in
510 a regulatory setting.

511 **Clinical Development and Regulatory Considerations**

512 Evaluation of the effect of food on drug bioavailability is a core component of the Clinical
513 Pharmacology/Biopharmaceutics program during development of a new chemical entity. Barring
514 any specific dosing restrictions informed by specific drug, formulation and target patient
515 population characteristics (*e.g.*, if very low bioavailability is expected in the fasted state, one may
516 decide to conduct early studies with dosing with a meal), food effect is often evaluated early in
517 clinical development, comparing fasted and fed administration, as part of the first-in-human single-
518 ascending or multiple-ascending dose studies. These studies, typically conducted with healthy
519 volunteers using standardized dosing conditions, such as a high-fat/high-caloric breakfast
520 described in the US FDA guidance (FDA, 2022), serve as the basis to inform dosing in subsequent
521 clinical trials when studies expand to larger number of patients. Even for indications such as

522 oncology where first-in-human dosing may be in patients, it is generally recommended that the
523 effect of food is explored early on. In many cases, food effect studies may be repeated later in
524 development to test food effect for new formulations, to assess different meal types or when the
525 program expands to a new population (*e.g.*, pediatrics). For post-approval of significant
526 formulation changes and for generic drug products, fed bioequivalence studies may be required
527 depending on the drug product label and the type of formulation used (FDA, 2021).

528 Assessment of food-drug interactions is covered by guidelines by all major health authorities for
529 both new chemical entities (EMA, 2012; FDA, 2022; HealthCanada, 2018) and generic drug
530 products (EMA, 2010; FDA, 2021; NIHS-Japan, 2012). The available guidelines provide
531 recommendations on study design, meals to be evaluated and interpretation of the results. Based
532 on current regulatory guidelines the presence of a food effect is established based on
533 pharmacokinetic bioequivalence bounds (*i.e.*, if the 90% confidence interval for the geometric
534 mean ratio for AUC and C_{max} between fed and fasted dosing meets the limits of 80%-125%).
535 Nevertheless, during clinical development, decisions on dosing instructions for clinical studies and
536 eventually for drug labeling are typically more flexible and take into account safety and efficacy
537 margins to define the clinical relevance of the food effect. In early clinical studies with smaller
538 populations before food effect has been thoroughly evaluated, or when a fit-for-purpose
539 formulation is used, it is often feasible to adopt more prescriptive dosing instructions such as fasted
540 administration. However, as dosing expands to larger populations in Phase 2 trials and beyond,
541 especially in pivotal studies, it is generally desirable to be able to dose medications without regard
542 to food, as compliance to more strict dosing regimens can be an issue and is difficult to track. The
543 dosing regimen implemented in late-stage pivotal trials is usually very similar to that on the drug
544 prescribing information.

545 If the physicochemical and metabolic properties of the compound are not inherently supportive of
546 comparable bioavailability in fasted and fed state, formulation interventions may be considered as
547 discussed later in the following section. In cases where a formulation solution is not implemented,
548 dosing instructions for administration with or without food may be also considered as long as they
549 are supported by the established clinically relevant bounds. For example, for products with a
550 positive food effect, that require administration with food to achieve adequate bioavailability, it is
551 highly desirable that, at minimum, dosing instructions are not prescriptive of the type of meal
552 required. Thus, whether administration with lighter meals is feasible is commonly evaluated to
553 provide more flexibility to patients. This is the case for example for vericiguat or venetoclax where
554 for the former the tablets are recommended to be taken with food, but high-fat, high-calorie or
555 low-fat, low-calorie meals are both acceptable as they result in similar pharmacokinetics
556 (VERQUVO® prescribing information (Merck & Co., Inc., Rahway, NJ, USA, 2021)), whilst the
557 latter can be taken with either a low fat and a high-fat meal, even though the magnitude of the food
558 effect is affected by fat content, as both result in sufficient, and much improved over fasted state
559 bioavailability (VENCLEXTA® prescribing information (Abbvie, 2021)). However sometimes
560 the exposure differences between meals are significant, as was the case with telaprevir
561 (INCIVEK™), where systemic exposure increase was approximately 117% and 330% with low-
562 fat and high-fat meal respectively. For INCIVEK, administration with food (not low fat) is
563 prescribed in the label. A positive food effect may also result in different dose recommendation in
564 the fed and fasted state. This is the case for ceritinib, where the recommended administration is a
565 450 mg dose with food, but 750 mg fasted may be used for patients unable to take drug with food
566 (ZYKADIA EPAR-Product Information (Novartis, 2021)). If the increase in bioavailability with
567 food, or specific types of food, raises safety concerns, specific wording may be included in the

568 prescribing information, such as is the case with ibrutinib where patients are advised not to take
569 the drug with grapefruit or Seville oranges (IMBRUVICA EPAR-Product Information (Janssen,
570 2021)). For compounds with significantly negative food effect, one could consider staggering food
571 intake with compound administration as is the case for semaglutide. According to the Rybelsus®
572 label, it is recommended that the drug is taken “at least 30 minutes prior to the first food, beverage
573 or other oral medications of the day with no more than 4 oz of plain water only” (RYBELSUS®
574 prescribing information (NovoNordisk, 2021)).

575 **Mitigation of food effects by formulations**

576 Depending on the root cause of the food effect, drug formulation can have a huge impact on the
577 direction and the extent of food effects. For instance, itraconazole, a poorly water soluble but
578 highly permeable drug (BCS class II), shows a positive food effect if formulated as pellets based
579 on an amorphous solid dispersion (Barone *et al.*, 1993). Due to longer residence times in the
580 stomach and higher bile salts levels in the small intestine, the intake together with food provides
581 improved conditions for dissolution in luminal fluids, which ultimately leads to higher oral
582 bioavailability in fed state. However, the oral solution formulation based on cyclodextrins shows
583 a negative food effect (Barone *et al.*, 1998). Here, the higher bile salt levels potentially lead to the
584 displacement of the drug from the apolar cavity of the cyclodextrins, which results in precipitation
585 (Stappaerts and Augustijns, 2016). Another prominent example was published by Wu and
586 colleagues, who could show in a Beagle dog model that food effect for MK-0869 (aprepitant)
587 could be reduced if the formulation was changed from a conventional oral suspension to a
588 nanocrystalline formulation (Wu *et al.*, 2004). Therefore, the commercial formulation (EMEND)
589 can be taken irrespective of food intake (Shadle *et al.*, 2012).

590 These examples nicely illustrate that by optimization of the formulation, food effects on oral
591 bioavailability can be reduced. This topic was specifically highlighted for oral anticancer drugs in
592 a recent article by Herbrink and colleagues, who stated that for 16 out of 28 drug products low
593 bioavailability and high variability is observed (Herbrink *et al.*, 2017). Since they regard those
594 “creaky formulations” as inadequate, they call for an improvement of the formulations. Although
595 this call is comprehensible, one should first take a deeper look at the current possibilities for
596 pharmaceutical industry in terms of this question. In this regard, O’Shea and colleagues
597 summarized existing literature on this topic in an excellent review (O’Shea *et al.*, 2019). They
598 showed that if the oral bioavailability is mainly limited by solubility of the drug in luminal fluids,
599 the use of bio-enabling formulation techniques such as amorphous solid dispersions, lipid-based
600 formulations or cyclodextrins presents a valid strategy for food effect reduction. Thereby, any
601 strategy for reduction of the food effect should aim to enhance the oral bioavailability in fasted
602 state, rather than reducing the oral bioavailability in fed state. In addition, it must be considered
603 that bioavailability is only one of the key design requirements in drug product development.
604 Stability and manufacturability must also be considered and sometimes represent major roadblocks
605 to the development of certain formulations even if bioavailability is improved. Moreover, the
606 demand for a short time to market for highly potent drugs often represents another obstacle to
607 formulation optimization in later clinical stages. Best practice is to address food effects already at
608 preclinical or early clinical stages in order to study the potential of a novel drug in terms of oral
609 bioavailability and to enable the early development of a formulation with reduced food effect.

610 In a recent work by Pandey *et al.*, it was nicely shown how a large positive food effect identified
611 in early clinical studies was addressed by formulation optimization and accompanied by the
612 application of proper *in vivo*, *in vitro* and *in silico* methods (Pandey *et al.*, 2014). In general, a food

613 effect can only be reduced by formulation optimization if adequately reliable *in vivo* (e.g., dog
614 model), *in vitro* (e.g., Dynamic Gastric Model, TIM-1 system) and/or *in silico* tools (e.g., SimCYP,
615 GastroPlus) are available. If applied in a meaningful manner as presented in Figure 1, these can
616 provide mechanistic insights into the potential root causes of the food effect and by this, can guide
617 the formulation activities during drug product development.

618 However, the optimization of an oral formulation in terms of drug release does not necessarily
619 result in a reduction of food effects. If the food-induced changes of oral bioavailability are
620 associated with food effects on drug absorption or subsequent events such as splanchnic blood
621 flow, metabolism or elimination, it will be difficult, often impossible, to reduce the food effect
622 simply by formulation changes. In particular, negative food effects which are often associated with
623 how food affects drug absorption or metabolism, are difficult to formulate away (O'Shea *et al.*,
624 2019).

625 **Summary and outlook**

626 The assessment of food effects remains a complex issue, best addressed early on in the drug
627 development cycle by a variety of techniques spanning from simple solubility studies and complex
628 dissolution/permeation assays to animal models and software-based modelling tools. The
629 combination of these *in vitro*, *in vivo* and *in silico* methods is a necessary requirement to
630 understand the food effect mechanisms and, on this basis, to develop a strategy for their control or
631 mitigation, usually via changes in the formulation. It is important to emphasize that due to the lack
632 of standardization of the various tools, this current approach for food effect assessment can only
633 be successfully implemented by the careful collaboration of scientists with sufficient knowledge
634 in the methods that are being employed, including experts in biopharmaceutics and in clinical

635 pharmacokinetics. Hence, continued efforts to develop a unified, standard approach in dealing with
636 food effects are required, to decrease food-effect driven risks in oral drug development.

637 **Credit author statement**

638 All authors contributed equally to this review. In addition, Zahari Vinarov and Patrick Augustijns
639 were responsible for putting the individual parts together and revising the manuscript.

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