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5                                   **Assessment of food effects during clinical development**

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24 **Abstract**

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

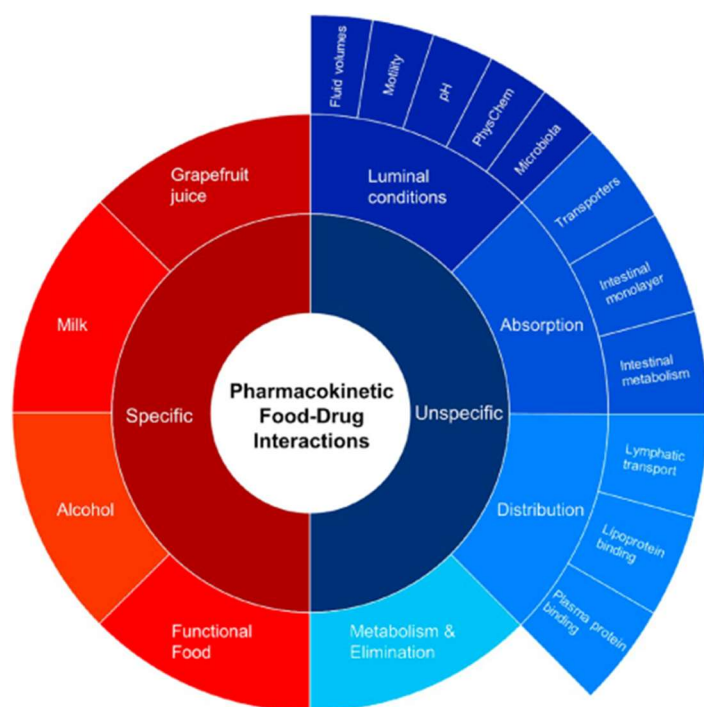
41 **Keywords**

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

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44 **Introduction**

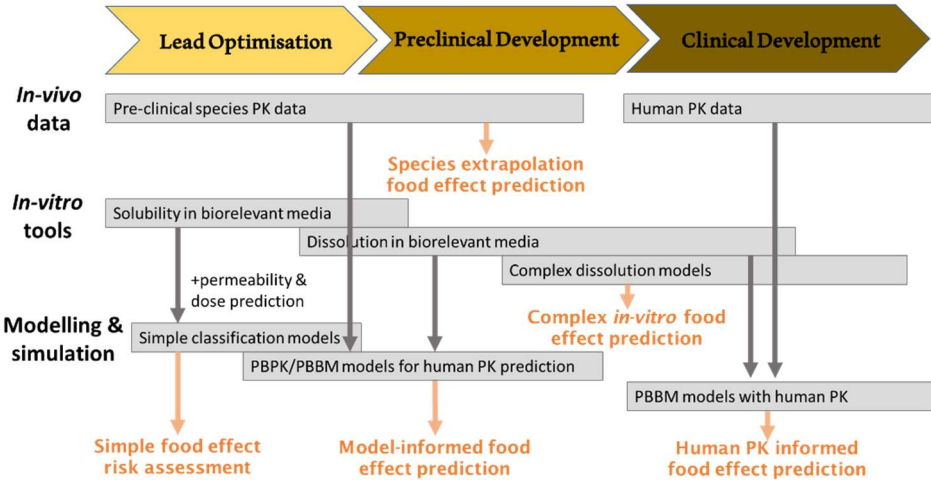
45 Food-drug interactions often present a significant challenge during the development of oral  
46 medicines, due to their influence on drug pharmacodynamics and pharmacokinetics (PK). In  
47 particular, food may have a substantial impact on drug absorption and metabolism, which will be  
48 reflected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



53  
54 **Figure 1.** Summary of specific and unspecific pharmacokinetic food-drug interactions. Reprinted  
55 from Koziolok *et al.* 2019a, Creative Commons CC-BY license.

56

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



64  
 65 **Figure 2.** Food effect prediction workflow in pharmaceutical development.

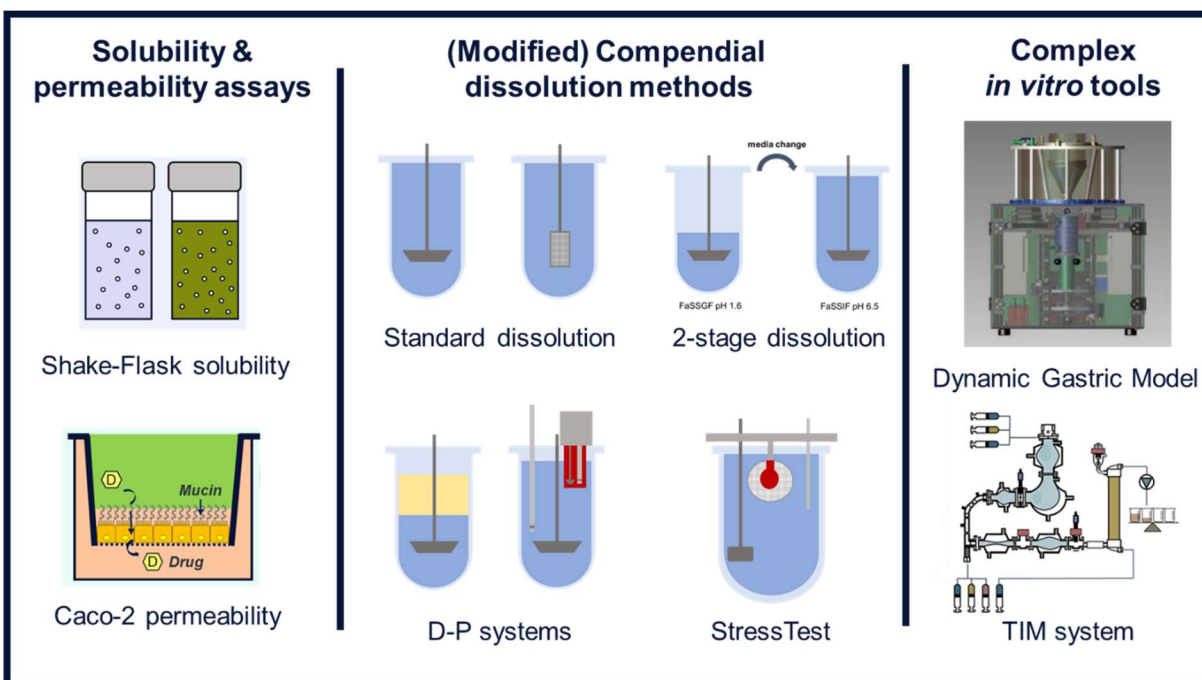
66 At the same time, method selection depends on the goal of food effect evaluation and on the  
 67 stage of drug development: for example, early assessment protocols serve to estimate the risk of  
 68 significant food effects in the clinic, largely based on drug properties alone. Recently,  
 69 physiologically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

74 formulation with a reduced food effect) and to further refine *in silico* and *in vitro* methods (Figure  
75 1).

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

### 83 ***In vitro* prediction tools**

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

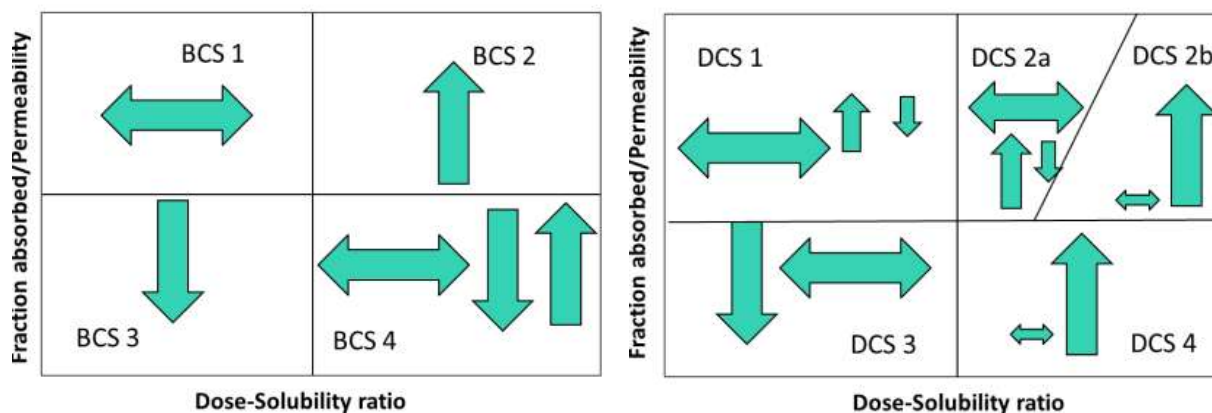


97  
 98 **Figure 3.** Schematic representation of the types of in vitro models used to study food effects.  
 99 D-P denotes “dissolution-permeation” and TIM denotes “TNO Gastro-Intestinal Model”. The  
 100 sketch of the Caco-2 permeability setup was obtained from Ye *et al.* 2022, the Dynamic Gastric  
 101 Model sketch was obtained from Mann and Pygall 2014 and the TIM sketch was obtained from  
 102 López Mármol *et al.* 2022.

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

104 Solubility in biorelevant media is often used as a starting point for food effect prediction for poorly  
 105 water-soluble drugs when new drug candidates are identified. Solubility in fasted and fed state  
 106 simulated intestinal fluids (FaSSIF/FeSSIF) has been shown to reflect that observed in human  
 107 aspirates reasonably well, in both the fasted and fed state (Augustijns *et al.*, 2014). However, it is  
 108 only if a drug’s absorption is incomplete (due to low solubility and/or slow dissolution) when  
 109 differences in FaSSIF/FeSSIF solubility potentially translate to a meaningful difference in  
 110 bioavailability. The BCS (Fleisher *et al.*, 1999; Ku, 2008; O’Shea *et al.*, 2019) and the related

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



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

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

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

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

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

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



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

157 Since the publication of the original biorelevant media recipes in the late 1990's, modified  
158 intestinal media (version 2), plus media for the fed state gastric environment (Jantratid *et al.*, 2008)  
159 were proposed. In addition, newer versions incorporate the products of lipid digestion into  
160 simulated intestinal media (Fuchs *et al.*, 2015; Jantratid *et al.*, 2008). Subsequent to the  
161 introduction of biorelevant dissolution media, the incorporation of dissolution data into PBPK  
162 models has been demonstrated to be an invaluable approach with numerous publications  
163 advocating their use (Kushwah *et al.*, 2021; Otsuka *et al.*, 2013; Shono *et al.*, 2009; Shono *et al.*,  
164 2010).

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

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***Modified compendial set ups***

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

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

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

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

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

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

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

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

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

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

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

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

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

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

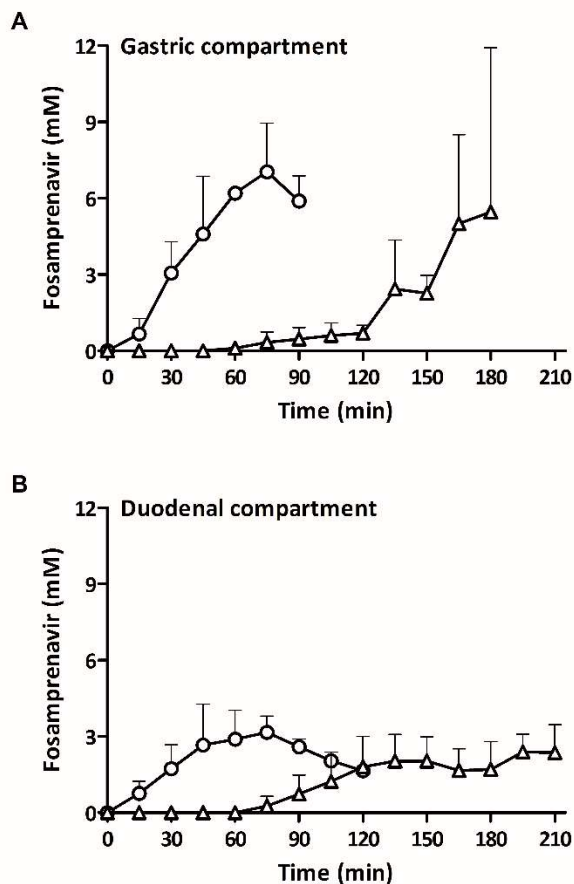
API	Formulation	Meal type	<i>In vivo</i> fed/fasted ratio	TIM <i>in vitro</i> fed/fasted ratio	Publication TIM data
Danirixin	DNX HBr	High fat meal	0.6 (AUC <sub>0-inf</sub> )	0.6 (TIM-1)	(Lloyd <i>et al.</i> , 2020)
Diclofenac	Cataflam IR	Ensure Plus	1.0 (AUC <sub>0-8h</sub> )	1.0 (TIM-1)	(Van Den Abeele <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 <sub>0-inf</sub> )	1 (TIM-1)	(Souliman <i>et al.</i> , 2006)
Acetaminophen	Sinaspril <i>*crushed</i>	Infant formula	No food effect	No food effect (tiny-TIM <sub>pediatrics</sub> )	(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 <sub>0-inf</sub> )	2.0 (TIM-1)	(Lyng <i>et al.</i> , 2016)
Nifedipine	Adalat XL MR	High fat meal	1.7 (AUC <sub>0-9h</sub> )	3.5 (TIM-1) 3.6 (tiny-TIM)	(Verwei <i>et al.</i> , 2016)
Posaconazole	Noxafil Suspension	High fat meal	4 (AUC <sub>0-72h</sub> )	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 <sub>0-t</sub> ) at 10mg* 3.2 (AUC <sub>0-t</sub> ) 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)

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

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

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



298  
 299 **Figure 5.** Fosamprenavir concentration–time profiles in the stomach (A) and duodenum  
 300 (B) compartment of TIM-1, simulating the fasted (open circles) and fed (open triangles) state.  
 301 Results are expressed as mean  $\pm$  sd (n = 3). Reprinted from European Journal of Pharmaceutics  
 302 and Biopharmaceutics, 77, Brouwers, J., Anneveld, B., Goudappel, G.-J., Duchateau, G.,  
 303 Annaert, P., Augustijns, P., Zeijdner, E. “Food-dependent disintegration of immediate release  
 304 fosamprenavir tablets: In vitro evaluation using magnetic resonance imaging and a dynamic  
 305 gastrointestinal system”, 313-319, Copyright (2011), with permission from Elsevier.  
 306  
 307 Often, the simulation of GI physiology in the *in vitro* system and the *in silico* model are different,  
 308 which makes direct integration of data very challenging. For instance, data from Tiny-TIM and



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

313 Using the Dynamic Gastric Model (DGM), Vardakou *et al.* demonstrated that antral grinding  
314 forces could be mimicked with much greater accuracy than using compendial dissolution apparatus  
315 (Vardakou *et al.*, 2011a). Investigational work also showed that the model could predict the  
316 differing drug release properties of various immediate release capsules in the fed and fasted state  
317 (Vardakou *et al.*, 2011b). In addition, *in vitro* work on the DGM model has been used to show that  
318 this system is likely to have specific advantages for investigating the dissolution properties of  
319 extended-release matrices in the fed state, compared to fasted (Chessa *et al.*, 2014; Mason *et al.*,  
320 2016).

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

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

### 333 ***In vivo* models for food effect predictions**

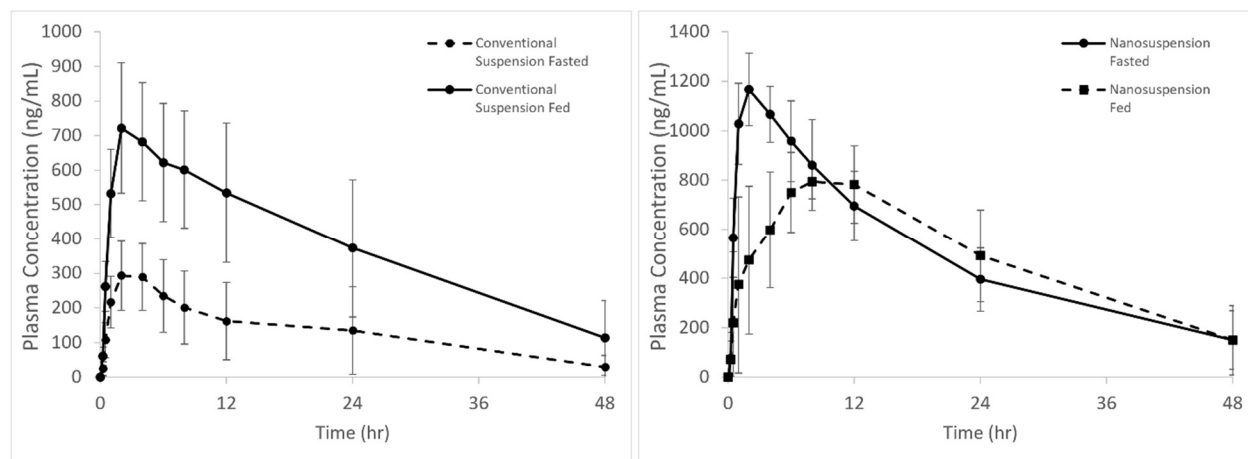
334 As highlighted in the previous sections, food effects on drug bioavailability are the result of the  
335 complex interplay of different physiological factors that change after the intake of food (Koziolek  
336 *et al.*, 2019a). Before complex and powerful *in vitro* tools (*e.g.* TIM-1, DGM) and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

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

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

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

375 In many pharmaceutical companies, the Beagle dog is the primary animal model to predict food  
376 effects on oral bioavailability. First studies on the application of this model for simulation of drug  
377 product performance in fed state have been published more almost 40 years ago (Cox *et al.*, 1985;  
378 Shiu *et al.*, 1989). Therefore, there is large experience within the pharmaceutical industry on the  
379 application of this animal model. However, whereas various guidance documents were issued by  
380 regulatory authorities on food effect studies in humans (EMA, 2012; FDA, 2002), there is still no  
381 standard protocol in terms of pre-treatment, type and timing of food intake, fluid intake during  
382 administration as well as subsequent food or liquid intake for food effect studies in dogs. Studies  
383 in which the dog model was successfully applied to predict drug product performance in presence  
384 of food, often have anecdotal character and can hardly be compared to other food effect studies in  
385 dogs. Nonetheless, the dog model can provide useful insights into drug product performance in  
386 fed state. For instance, Wu and colleagues nicely illustrated how a dog model was used to support  
387 the development of a nanocrystalline formulation of MK-0869 (aprepitant). Canine data could  
388 demonstrate that this formulation has a reduced food effect as compared to a conventional  
389 suspension, see **Figure 6** (Wu *et al.*, 2004). However, only few systematic studies on the use of  
390 dogs for food effect prediction have so far been performed (Lentz *et al.*, 2007; Mathias *et al.*, 2015;  
391 Zane *et al.*, 2014). In this context, one of the most relevant articles was published in 2007 by Lentz  
392 and colleagues, who studied the impact of the study protocol and investigated the correlation  
393 between food effect in dogs and humans (Lentz *et al.*, 2007). Based on two model compounds  
394 (atazanavir and pravastatin), it was first shown that, to achieve the best correlation to human data,  
395 a 50 g aliquot of the FDA meal should be used and that dogs should be pretreated with pentagastrin  
396 to stimulate gastric acid secretion in fasted state.  
397



398  
 399 **Figure 6.** Assessment of food effect for conventional (left) and nanosized (right) suspensions in  
 400 dogs. Based on data from Wu et al, Int J Pharm, 285 (2004), 135-146.

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

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

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

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

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

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

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

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

#### 465 **Physiologically Based Pharmacokinetic modeling**

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

475 **Table 2.** Summary of publications with PBPK models for food effect, listed chronologically  
 476 (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

477

478 Evolution of the models over the years reflects the increased utilization of more complex *in vitro*  
479 methodologies discussed earlier in this manuscript; while initial models largely focused on the  
480 solubility differential in biorelevant media such as FeSSIF and FaSSIF, data from multi-

481 compartment systems to characterize dissolution and precipitation are now more commonly  
482 utilized.

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

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

505 At later stages of development, the desire is to use PBPK models for regulatory interactions, such  
506 as replacing clinical studies. However, despite the numerous successful examples in the literature,  
507 best practice and regulatory acceptance of PBPK models for food effect predictions are still  
508 evolving. As a result, confidence in the models by regulators is still low (Li *et al.*, 2018).  
509 Development of standardized input and model development workflows have been recently  
510 proposed (Riedmaier *et al.*, 2020) as a step towards that direction. In practice, validation of the  
511 prediction against early-stage clinical food effect data before use of the model for *a priori*  
512 predictions, as recommended by Tistaert *et al.* and Kesisoglou (Kesisoglou, 2020; Tistaert *et al.*,  
513 2019), is likely going to be a prerequisite for model application at later development stages and in  
514 a regulatory setting.

### 515 **Clinical Development and Regulatory Considerations**

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

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

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

549 If the physicochemical and metabolic properties of the compound are not inherently supportive of  
550 comparable bioavailability in fasted and fed state, formulation interventions may be considered as  
551 discussed later in the following section. In cases where a formulation solution is not implemented,  
552 dosing instructions for administration with or without food may be also considered as long as they  
553 are supported by the established clinically relevant bounds. For example, for products with a  
554 positive food effect, that require administration with food to achieve adequate bioavailability, it is  
555 highly desirable that, at minimum, dosing instructions are not prescriptive of the type of meal  
556 required. Thus, whether administration with lighter meals is feasible is commonly evaluated to  
557 provide more flexibility to patients. This is the case for example for vericiguat or venetoclax where  
558 for the former the tablets are recommended to be taken with food, but high-fat, high-calorie or  
559 low-fat, low-calorie meals are both acceptable as they result in similar pharmacokinetics  
560 (VERQUVO® prescribing information ( Merck & Co., Inc., Rahway, NJ, USA, 2021)), whilst the  
561 latter can be taken with either a low fat and a high-fat meal, even though the magnitude of the food  
562 effect is affected by fat content, as both result in sufficient, and much improved over fasted state  
563 bioavailability (VENCLEXTA® prescribing information (Abbvie, 2021)). However sometimes  
564 the exposure differences between meals are significant, as was the case with telaprevir  
565 (INCIVEK™), where systemic exposure increase was approximately 117% and 330% with low-  
566 fat and high-fat meal respectively. For INCIVEK, administration with food (not low fat) is  
567 prescribed in the label. A positive food effect may also result in different dose recommendation in  
568 the fed and fasted state. This is the case for ceritinib, where the recommended administration is a  
569 450 mg dose with food, but 750 mg fasted may be used for patients unable to take drug with food  
570 (ZYKADIA EPAR-Product Information (Novartis, 2021)). If the increase in bioavailability with  
571 food, or specific types of food, raises safety concerns, specific wording may be included in the

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

### 579 **Mitigation of food effects by formulations**

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

594 These examples nicely illustrate that by optimization of the formulation, food effects on oral  
595 bioavailability can be reduced. This topic was specifically highlighted for oral anticancer drugs in  
596 a recent article by Herbrink and colleagues, who stated that for 16 out of 28 drug products low  
597 bioavailability and high variability is observed (Herbrink *et al.*, 2017). Since they regard those  
598 “creaky formulations” as inadequate, they call for an improvement of the formulations. Although  
599 this call is comprehensible, one should first take a deeper look at the current possibilities for  
600 pharmaceutical industry in terms of this question. In this regard, O’Shea and colleagues  
601 summarized existing literature on this topic in an excellent review (O’Shea *et al.*, 2019). They  
602 showed that if the oral bioavailability is mainly limited by solubility of the drug in luminal fluids,  
603 the use of bio-enabling formulation techniques such as amorphous solid dispersions, lipid-based  
604 formulations or cyclodextrins presents a valid strategy for food effect reduction. Thereby, any  
605 strategy for reduction of the food effect should aim to enhance the oral bioavailability in fasted  
606 state, rather than reducing the oral bioavailability in fed state. In addition, it must be considered  
607 that bioavailability is only one of the key design requirements in drug product development.  
608 Stability and manufacturability must also be considered and sometimes represent major roadblocks  
609 to the development of certain formulations even if bioavailability is improved. Moreover, the  
610 demand for a short time to market for highly potent drugs often represents another obstacle to  
611 formulation optimization in later clinical stages. Best practice is to address food effects already at  
612 preclinical or early clinical stages in order to study the potential of a novel drug in terms of oral  
613 bioavailability and to enable the early development of a formulation with reduced food effect.  
614 In a recent work by Pandey *et al.*, it was nicely shown how a large positive food effect identified  
615 in early clinical studies was addressed by formulation optimization and accompanied by the  
616 application of proper *in vivo*, *in vitro* and *in silico* methods (Pandey *et al.*, 2014). In general, a food

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

622 However, the optimization of an oral formulation in terms of drug release does not necessarily  
623 result in a reduction of food effects. If the food-induced changes of oral bioavailability are  
624 associated with food effects on drug absorption or subsequent events such as splanchnic blood  
625 flow, metabolism or elimination, it will be difficult, often impossible, to reduce the food effect  
626 simply by formulation changes. In particular, negative food effects which are often associated with  
627 how food affects drug absorption or metabolism, are difficult to formulate away (O'Shea *et al.*,  
628 2019).

## 629 **Summary and outlook**

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



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

#### 641 **Credit author statement**

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

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