

PREDICTING HUMAN DRUG GLUCURONIDATION PARAMETERS: Application of In Vitro and In Silico Modeling Approaches

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■ **Abstract** Cytochrome P450 (CYP) and UDP-glucuronosyltransferase (UGT), which both exist as enzyme “superfamilies,” are together responsible for the metabolism of most hepatically cleared drugs. There is currently intense interest in the development of techniques that permit identification of the CYP and UGT isoform(s) involved in the metabolism of a newly discovered drug, and hence prediction of factors likely to alter elimination in vivo. In addition, the quantitative scaling of kinetic parameters for a metabolic pathway assumes importance for identifying newly discovered drugs with undesirable in vivo pharmacokinetic properties. Although qualitative and quantitative in vitro–in vivo correlation based on data generated using human liver tissue or recombinant enzymes have been applied successfully to many drugs eliminated by CYP, these strategies have proved less definitive for glucuronidated compounds. Computational (in silico) modeling techniques that potentially provide a facile and economic alternative to the in vitro methods are now emerging. This review assesses the utility of in vitro and in silico approaches for the qualitative and quantitative prediction of drug glucuronidation parameters and the challenges facing the development of generalizable models.

INTRODUCTION

Hepatic metabolism is the principal elimination mechanism for the majority of drugs in humans. Hence, knowledge of hepatic extraction ratio (E_H), hepatic clearance (CL_H), and factors that modulate these parameters is of fundamental importance given their impact on the efficacy and safety of drug treatment. Impairment of drug metabolism may result in toxicity as a result of decreased CL_H

and/or increased bioavailability. Conversely, induction of metabolism may result in loss of efficacy owing to increased CL_H and/or decreased bioavailability. Given these characteristics of hepatically cleared drugs, the development of in vitro techniques to predict aspects of human drug metabolism and pharmacokinetics in vivo (in vitro–in vivo correlation) has attracted enormous interest over the past decade. Indeed, there is now widespread acceptance of in vitro–in vivo correlation, particularly for compounds eliminated by cytochrome P450 (CYP)-catalyzed hepatic biotransformation. More recently, computational (in silico) approaches have also been applied to prediction of the involvement of specific enzymes in the metabolism of particular drugs.

The economic imperatives associated with drug discovery and greater promotion of the rational use of drugs (rational therapeutics) have been the principal factors driving the development of in vitro and in silico methods to predict drug metabolism parameters. Drug discovery and development is immensely expensive and time-consuming. The success rate of new chemical entities selected for clinical development is approximately 20% (1), with most failures attributed to unacceptable pharmacokinetic properties (2). Undesirable properties, such as poor absorption, high E_H and CL_H (leading to low and variable bioavailability), drug interactions, and metabolism by a polymorphic enzyme, may be predicted from in vitro (and possibly in silico) data, thus facilitating selection of the most appropriate lead compound and decreasing attrition during clinical development. Moreover, selecting for development those compounds that are affected to the least extent by genetic polymorphism and drug interactions optimizes the clinical utility and market success of newly approved drugs. Knowledge of genetic polymorphism, drug interactions, and other factors altering CL_H is similarly important for rationalizing and optimizing dosage regimens of established drugs, thereby improving therapeutic outcome.

In vitro–in vivo correlation allows the prediction of drug metabolism parameters at both the qualitative and quantitative levels. The principal drug metabolizing enzymes CYP and UDP-glucuronosyltransferase (UGT) exist as gene “superfamilies” (3, 4). The individual CYP and UGT proteins (isoforms) tend to differ in terms of substrate and inhibitor selectivities, regulation, and patterns of drug interactions. Thus, identification of the isoform(s) responsible for the metabolism of any given drug, a process referred to as reaction phenotyping, together with an understanding of isoform regulation and drug interactions allows prediction of those factors likely to alter CL_H and clinical response. Procedures for CYP reaction phenotyping are well established and normally involve the integration of data from human liver microsomes (although hepatocyte suspensions may also be used) and recombinant human CYP isoforms (5–9). An alternative approach involves normalizing kinetic data generated with recombinant enzymes for relative isoform expression in human liver or by use of the relative activity factor (6, 7, 9).

Quantitative prediction most commonly involves the scaling of in vitro CL_{int} , calculated from the kinetic parameters (K_m , V_{max}) for the formation of a metabolite by human liver microsomes or hepatocytes, to in vivo CL_H and E_H using a mathematical model of hepatic clearance (10–15). Typically, this requires correction

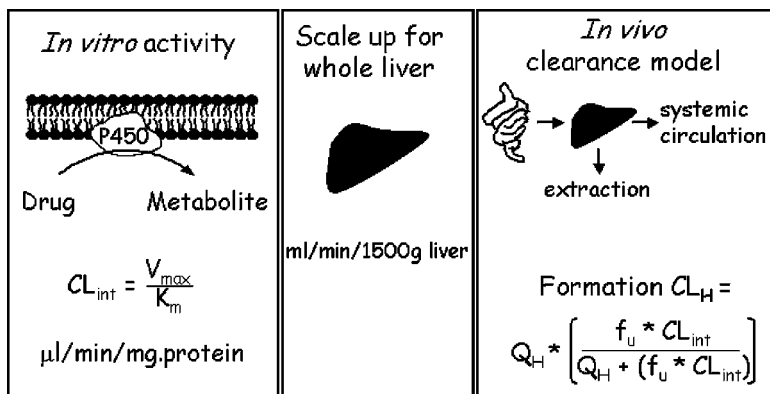


Figure 1 Approach for the calculation of *in vivo* hepatic clearance (CL_H) from kinetic constants determined for metabolite formation by human liver microsomes using the expression for the well-stirred model of hepatic clearance. Abbreviations: K_m , Michaelis constant; V_{max} , maximal velocity; CL_{int} , intrinsic clearance; Q_H , liver blood flow; f_u , fraction unbound in blood. Reprinted with permission from Miners JO. 2002. Annual Scientific Meeting of ASCEPT, 2001. Special article: The evolution of drug metabolism: hitchhiking the technology bandwagon. *Clin. Exp. Pharmacol. Physiol.* 29:1040–44

of human liver microsomal CL_{int} for microsome yield (milligrams per gram of human liver) and liver weight (normally assumed as 1.5 kg) to obtain a “whole organ” CL_{int} , which subsequently is substituted in the expression for the well-stirred, parallel-tube, or dispersion models (Figure 1). In general, this approach has proved valuable for predicting the *in vivo* CL_H and E_H of drugs metabolized by CYP (14, 15). The accuracy of prediction may be improved further by correcting the K_m for nonspecific binding of substrate to the microsomal membrane where this occurs (16, 17) or by applying scaling factors derived from animal data to the *in vitro* CL_{int} (18). An inhibitor constant (K_i) calculated for inhibition of metabolite formation *in vitro* may also be used to forecast *in vivo* decrement in CL_H arising from coadministration of an inhibitor (15, 19, 20). Although *in vitro* kinetic data have provided reasonably accurate prediction of some interactions involving CYP substrates, further refinement and validation is necessary before this approach may be applied more widely.

More recently, *in silico* approaches have been adopted to predict the absorption, distribution, metabolism, excretion, and toxicity (ADMET) parameters of new drug entities (21). Potentially, *in silico* modeling may be used to evaluate millions of compounds for their ADMET and pharmacological properties, thereby minimizing the need to experimentally characterize large numbers of molecules. In particular, homology- and pharmacophore-models and two- and three-dimensional quantitative structure activity relationships (2D- and 3D-QSAR) have been generated for substrates and inhibitors of CYP isoforms (22). Such quantitative structure

metabolism relationships (QSMR) provide insights into the structural features of drugs that confer isoform selectivity together with an estimate of binding affinity (K_m or K_i) (in silico–in vitro correlation).

As alluded to previously, CYP and UGT are quantitatively the most important drug metabolizing enzymes. Together, these enzymes are responsible for the elimination of more than 90% of hepatically cleared drugs. Despite the acceptance of in vitro–in vivo correlation and the promise of in silico–in vitro correlation for drugs eliminated by CYP, the application of these approaches to drug glucuronidation has proved challenging. Here we explore the utility of extrapolating in silico and in vitro metabolism data for drugs and other compounds cleared by UGT.

UDP-GLUCURONOSYLTRANSFERASE

Glucuronidation

UGT catalyzes the covalent linkage (conjugation) of glucuronic acid, derived from the cofactor UDP-glucuronic acid (UDPGA), to a substrate bearing a suitable functional group according to a second-order nucleophilic substitution mechanism (4). Glucuronides form via conjugation of a carboxyl, hydroxyl (phenol or aliphatic alcohol), amino, acidic carbon, or sulfuryl moiety present on a typically lipophilic substrate. It is therefore not surprising that glucuronidation serves as an essential clearance mechanism for a myriad of compounds, including drugs from all therapeutic classes, dietary chemicals, environmental pollutants, and endogenous compounds (e.g., bilirubin, bile acids, hydroxysteroids) (4, 23–25). In addition, glucuronidation facilitates excretion of these compounds and the products of phase I metabolism in urine and bile as their hydrophilic conjugates and generally results in detoxification, although a limited number of glucuronides possess biological activity (26).

UGT Heterogeneity

Consistent with its broad substrate profile, UGT exists as an enzyme superfamily (4, 25). UGTs are evolutionarily related to enzyme families in bacteria and plants, which also use activated sugar nucleotides to donate monosaccharides to a substrate [(24); see also UGT homepage <http://som.flinders.edu.au/FUSA/ClinPharm/UGT>]. The UGTs all contain a conserved 29-residue carboxy terminus “signature sequence,” which probably contributes to the recognition and binding of the UDP-sugar (27). To date, nucleotide sequences encoding 18 UGT proteins of approximately 530 amino acids have been identified (Figure 2) (<http://som.flinders.edu.au/FUSA/ClinPharm/UGT>). Based on sequence identity, UGTs can be divided into two gene families: *UGT1* on human chromosome 2q37 and *UGT2* on human chromosome 4q13. The human UGT1 protein family consists of nine members with different amino-terminal domains but identical carboxyl termini (28). Each UGT1 protein is encoded by a transcript that is formed by the splicing of a distinct first

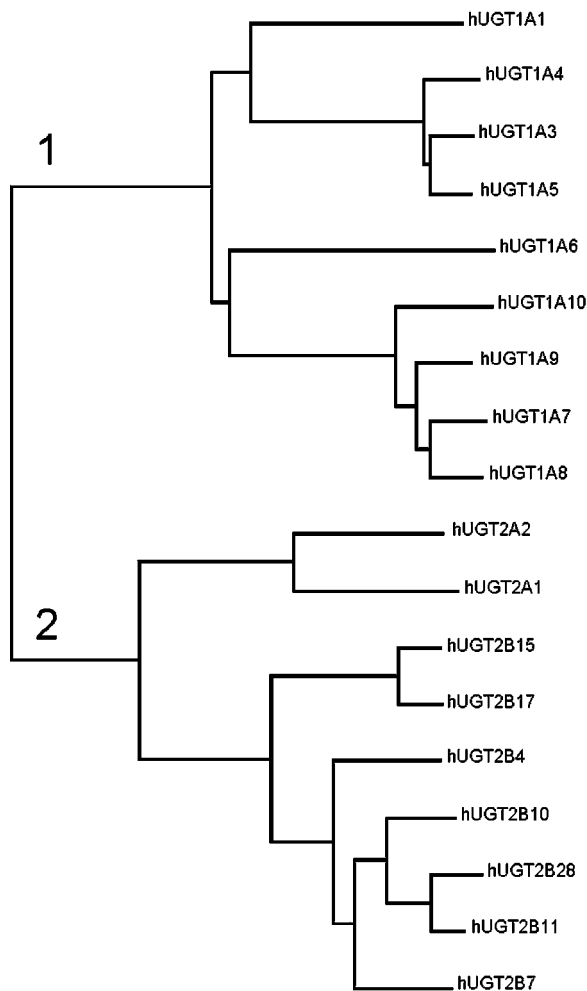


Figure 2 Phylogram of human UGT proteins.

exon (A1, A3, A4, A5, A6, A7, A8, A9, and A10) to a set of four downstream exons, designated 2–5. Four other first exons on the *UGT1A* locus (A2, A11, A12, and A13) lack open reading frames and are designated as pseudogenes (29). In contrast to the UGT1 family, the UGT2 enzymes do not share a common carboxyl-terminal domain and are encoded by separate genes comprising six exons (27). The human UGT2 family is further subdivided into two subfamilies; UGT2A and UGT2B, which contain two (2A1, 2A2) and seven (2B4, 2B7, 2B10, 2B11, 2B15, 2B17, and 2B28) members, respectively. Five *UGT2B* pseudogenes (24P–28P) have additionally been identified.

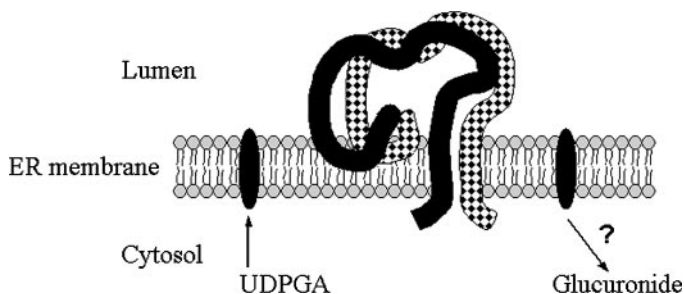


Figure 3 Hypothetical UGT topology model depicting UGT as a dimer, with the active site localized on the luminal aspect of the endoplasmic reticulum, and a UDP-glucuronic acid (UDPGA) transporter. Adapted from Reference 4.

UGT Membrane Localization

The UGTs are resident in the endoplasmic reticulum (ER) and nuclear compartment of cells (4). Each enzyme is synthesized as a precursor of approximately 530 residues containing an amino terminal signal peptide that is removed during insertion of the protein into the ER membrane (Figure 3) (30). The mature protein of approximately 500 residues is oriented on the luminal aspect of the ER membrane and contains an amino-terminal domain responsible for substrate selection and a carboxyl-terminal domain that binds UDPGA. Experimental evidence in support of this topology has been provided by studies with antibodies, proteases, detergents, and photoaffinity probes on microsomal UGTs and on UGTs synthesized *in vitro* in the presence and absence of membranes (reviewed in Reference 4).

Accumulating evidence strongly suggests that the active UGT enzyme consists of dimers of two UGT polypeptide chains (Figure 3). Catalytically active homodimers of the rat UGT2B1 that appear to interact through their amino-terminal regions have been detected (31). Homodimers of rat UGT1A6 (32), human UGT1A1 (33), and human UGT1A9 (34), and heterodimers of UGT2B1 and UGT1A6 (32) have also been detected by chemical cross-linking and coimmunopurification. Furthermore, it has been postulated that the stability of the interaction or the rates of dimerization may be governed by the specific UGT monomers involved and/or by interaction with substrate (4). It is possible that dimerization may impact on the kinetics of UGT-catalyzed reactions (discussed below), and incorporation of these possibilities into predictive models metabolism may prove necessary.

Regulation of UGT Expression

The majority of UGT isoforms (including 1A1, 1A3, 1A4, 1A6, 1A9, 2B4, 2B7, 2B15, 2B17, and 2B28) are expressed in the liver. However, UGT1A and UGT2B isoforms are also differentially expressed in a range of other tissues, including the kidney, olfactory epithelium, ovary, lung, small intestine, mammary gland,

testis, and prostate (reviewed in Reference 25). Indeed, UGT1A8 and UGT1A10 are expressed exclusively in the gastrointestinal tract (35). In addition, current evidence suggests that the UGT2A forms are predominantly expressed in the nasal epithelium and may be involved in odorant signal termination (36). All UGTs investigated to date display marked interindividual variation in levels within a tissue. The mechanisms that determine UGT tissue distribution and content are largely unknown, although factors that regulate their levels in the liver are beginning to be elucidated (37, 38).

UGT Isoform Substrate Selectivity

The majority of UGT isoforms exhibit distinct, but overlapping, substrate selectivities (4, 23, 25). For example, only UGT1A3 and UGT1A4 apparently conjugate tertiary amines, although selectivity towards other chemical classes differs. The exception appears to be UGT2B4, which glucuronidates the same substrates as UGT2B7 but with markedly lower activity (39). Despite considerable effort, substrates for UGT2B10 and UGT2B11 have not been identified to date. Although UGTs have been classified in the past according to the chemical characteristics of substrates, this approach is simplistic because most isoforms have the capacity to glucuronidate structurally diverse compounds. Studies of UGT isoform substrate selectivity have also tended to utilize the same “pool” of substrates (e.g., commercially available phenols and aliphatic alcohols), and only recently has the diversity of chemical space screened been increased to allow identification of isoform selective substrates.

Factors Affecting UGT Activity In Vivo

Numerous factors are known to alter human UGT activity in vivo, including age (particularly the neonatal period), diet, disease states, drug-drug interactions (induction and inhibition), ethnicity, genetic polymorphism, and hormonal effects (23, 40). Apart from genetic polymorphism, however, the isoform selectivity of these influences is generally poorly understood owing to the unavailability of isoform-selective substrates for the investigation of drug glucuronidation in vivo. Nevertheless, available evidence (see below) and experience with CYP (8) suggests selectivity is highly likely. Genetic polymorphism has been reported for *UGT1A1*, *-1A6*, *-1A7*, *-1A8*, *-2B4*, *-2B7*, and *-2B15* (25, 40–45), although there is evidence for allelic variation in most *UGTs* (<http://som.flinders.edu.au/FUSA/ClinPharm/UGT>). Mutations in *UGT1A1*, *-1A7*, *-1A8*, and *-2B15* have been implicated as risk factors for certain cancers (41, 44, 46, 47). More than 50 mutations in *UGT1A1* may give rise to inherited disorders of bilirubin glucuronidation (Gilbert and Crigler-Najjar syndromes type I and II) and, in some instances, impaired elimination of xenobiotics (25, 44, 48). Patients with variant *UGT1A1* genotypes are overrepresented among those experiencing severe toxicity to the anticancer drug irinotecan owing to impaired glucuronidation of the active metabolite SN-38 (49). In addition, indinavir, which appears to be a *UGT1A1* substrate, may precipitate jaundice in

patients with Gilbert syndrome variant alleles as a result of competitive inhibition of bilirubin conjugation (50). Thus, available evidence suggests that knowledge of the UGT isoform selectivity (substrate and inhibitor), particularly involvement of UGT1A1, is likely to be of benefit in assessing the potential effects of genetic polymorphism and drug-drug/endobiotic interactions on the elimination and toxicity of new chemical entities.

REACTION PHENOTYPING IN VITRO

Reaction Phenotyping Approaches

Approaches for CYP reaction phenotyping are based largely on the use of human liver microsomes as the enzyme source, although hepatocyte suspensions have also been used and include (5–9): (a) characterization of the effects of CYP isoform-selective inhibitors on the metabolism of the compound; (b) investigation of correlations between rates of metabolism of the compound and immunoreactive CYP isoform contents or prototypic isoform-selective activities in a “panel” of human liver microsomes; (c) competitive inhibition of the metabolism of isoform-selective substrates by the drug, with K_m matching K_i ; and (d) comparative metabolism by recombinant human CYP isoforms. Taken together, these procedures allow identification of the CYP isoform(s) responsible for the metabolism of a drug with a high degree of certainty, although inhibition by isoform-selective inhibitors alone is often considered diagnostic, providing experimental conditions for inhibitor selectivity are well established (5, 51). Thus, the availability of isoform selective substrates and inhibitors is pivotal to reaction phenotyping in vitro. Examples of CYP isoform selective substrates and inhibitors include phenacetin and furafylline for CYP1A2 and tolbutamide and sulfaphenazole for CYP2C9 (5). Moreover, substrates that may be safely administered to humans are necessary for the characterization of isoform regulation and drug-drug interactions in vivo. Thus, such probes are most commonly clinically used drugs.

Reaction Phenotyping of UGT Substrates

Although numerous compounds are known to inhibit human UGT activity in vitro and in vivo (52), no UGT isoform selective inhibitors have been identified to date. However, reasonably selective substrates are now available for a number of isoforms, including the more important drug metabolizing hepatic UGTs. Clearly, these may also be used as isoform selective inhibitors in vitro.

It is well established that bilirubin is glucuronidated predominantly, if not solely, by UGT1A1, with a K_m of 24 μM (48, 53). Impaired UGT1A1 activity owing to genetic polymorphism or drug interactions may give rise to clinically significant hyperbilirubinemia (48–50). In addition, it has been proposed that ethinylestradiol and estradiol are selective substrates for UGT1A1 (54–56). Although estradiol 3-glucuronidation appears to be mainly catalyzed by UGT1A1, other isoforms

(e.g., UGT1A3, -1A9, and -2B7) may also metabolize these compounds, and the absolute contribution of UGT1A1 to human hepatic ethinylestradiol and estradiol glucuronidation requires further characterization. As noted previously, both UGT1A3 and UGT1A4 have the capacity to glucuronidate tertiary amines. However, only UGT1A4 appears to catalyze the N-glucuronidation of imipramine and trifluoperazine (57, 58). A later study further demonstrated that imipramine is not a substrate for other UGT1A family isoforms or for UGT2B7 and UGT2B15 (59), but a role for additional isoforms in trifluoperazine glucuronidation remains to be discounted. Although acetaminophen (paracetamol) glucuronidation has been used as a probe for UGT1A6 activity, multiple isoforms probably contribute to this pathway (60). More recently, it has been shown that serotonin (5-hydroxytryptamine) is glucuronidated only by UGT1A6, and hence, this compound represents a selective, albeit low-affinity (K_m approximately 5 mM), substrate for this isoform (61). Early studies established that propofol was a substrate of UGT1A9 (K_m 170 μ M), but not UGT1A1, -1A4, or -1A6 (54). Subsequent studies have also excluded UGT1A3, -1A10, and -2B15 as catalysts of propofol glucuronidation. Although propofol is a good substrate for UGT1A8 (K_m 412 μ M) (62), this isoform is not expressed in human liver.

Within the UGT2B subfamily, UGT2B7 is the only isoform for which selective substrates have been identified. UGT2B7 is apparently the only member of this subfamily that glucuronidates zidovudine (K_m 82 to 91 μ M) (63). Recent studies in this laboratory have also excluded metabolism of zidovudine by UGT1A family enzymes and indicate that the K_m for glucuronidation by UGT2B7 is higher (i.e., 400 μ M) than originally reported (J.O. Miners & P.I. Mackenzie, unpublished data). UGT2B7 also appears to be the principal hepatically expressed UGT involved in the glucuronidation of epirubicin (4-epi-doxorubicin) (64), although the involvement of UGT1A8 and -1A10 cannot be excluded in other tissues. Although commonly used as a UGT2B7 selective substrate, morphine is glucuronidated in the 3-position by other isoforms, including UGT1A1, -1A3, -1A6, -1A9, and -1A10, and there is some evidence to suggest the involvement of multiple UGTs in human liver microsomal morphine 3-glucuronidation (65). However, morphine 6-glucuronidation is catalyzed solely by UGT2B7 (65). Hyodeoxycholic acid has been proposed as a selective substrate for UGT2B4, but this compound is glucuronidated more efficiently by UGT2B7 (66). Indeed, UGT2B4 and UGT2B7 have similar substrate selectivities, but turnover by the latter is generally at least one to two orders of magnitude higher (39, 66). Similarly, recent experience suggests that many reported UGT2B15 substrates (67) are also effectively glucuronidated by UGT2B7 [(68); J.O. Miners & P.I. Mackenzie, unpublished data].

It is apparent that the experimental tools necessary for reaction phenotyping UGT substrates are becoming increasingly available in parallel with increasing awareness of the importance of this enzyme in drug metabolism. However, significant problems remain. The use of alternate substrates as inhibitors requires careful selection of the concentration used in incubations of human liver microsomes or

other tissue preparations because the degree of inhibition will be dependent on the concentrations of substrate and inhibitor relative to their respective K_m values. Indeed, the successful reaction phenotyping of CYP substrates is critically dependent on the careful selection of substrate/inhibitor concentration and incubation conditions (5). Furthermore, it is not possible to discount at this stage that the selective substrates identified to date act as inhibitors of other UGTs, as is the case with quinidine (a CYP3A substrate) inhibition of CYP2D6. The lack of a convenient substrate/inhibitor probe for UGT1A1, which appears to have an important role in drug metabolism, remains a problem. Bilirubin has only moderate stability, and assays for the measurement of bilirubin glucuronidation *in vitro* are difficult. It should also be recognized that many isoform "selective" substrates have not been screened with all known UGTs. In particular, the xenobiotic substrate selectivity of UGT2B28 is largely unknown (68) and the involvement of UGT1A5, -1A7, -1A8, -1A10, -2B4, and -2B17 in drug and xenobiotic metabolism is not commonly investigated. With the exception of UGT2B17 and -2B28, however, these forms are expressed in extrahepatic tissues and/or exhibit low activity toward xenobiotics, and a significant role in hepatic drug clearance can generally be discounted. It is also apparent that the selective substrates referred to previously are of limited value for characterizing isoform regulation *in vivo*. These compounds are either endobiotics or they are drugs that are unsuitable for administration to healthy volunteers (owing to their inherent pharmacological properties and/or toxicity) and which have undesirable pharmacokinetic properties or low clearance by glucuronidation *in vivo* (e.g., imipramine).

Given the limited availability of isoform selective substrates and inhibitors, reaction phenotyping of UGT substrates has most commonly involved screening for activity by recombinant isoforms, for example, References 69, 70. Although this may yield useful results where only a single isoform is identified, interpretation is difficult where multiple UGTs are involved (71). Whereas CYP expression *in vitro* may be quantified spectrophotometrically, assessment of the relative levels of UGTs in expression systems remains problematic (being dependent on the interpretation of Western blots). In addition, the relative expression of UGT isoforms *in vivo* is currently unknown. Hence, the use of a relative activity factor or scaling for isoform expression in human liver, approaches used for CYP-catalyzed reactions (6, 7), are currently not feasible for UGT reaction phenotyping.

IN VITRO–IN VIVO CORRELATION

The Predicability of In Vitro Drug Glucuronidation Kinetic Data

As described previously, *in vitro*–*in vivo* correlation most commonly involves scaling of the CL_{int} value determined for a reaction in human liver microsomes to *in vivo* E_H and CL_H (Figure 1). Although this approach has proved useful for many CYP-catalyzed reactions, recent studies have demonstrated that extrapolated CL_H

consistently underestimates known clearance by glucuronidation in vivo (72–74). The relationship between in vivo blood clearance by glucuronidation and predicted CL_H (from the scaling of human liver microsomal CL_{int} using the dispersion model) for 14 glucuronidated drugs is shown in Figure 4. Predicted and known in vivo CL_H are significantly correlated ($r^2 = 0.53$, $p < 0.01$), but predicted CL_H underestimates known glucuronidation clearance by a factor of 11.2. Exclusion of the data for propofol and naloxone (predicted CL_H values 1.4 and 3.4 l/h, respectively) improves the correlation markedly (slope = 0.136, $r^2 = 0.95$, $p < 0.001$). Mistry & Houston previously reported similar findings for glucuronidated opioids in rat, where hepatic microsomal CL_{int} values were 20- to 30-fold lower than their in vivo counterparts (75). The data shown in Figure 4 suggest that it may be possible to apply a scaling factor to the extrapolated CL_H in order to predict in vivo glucuronidation clearance, but further validation of this approach is clearly required. Interestingly, a linear relationship was reported for the rate of glucuronidation of a series of 5-lipoxygenase inhibitors by cynomolgus monkey microsomes and

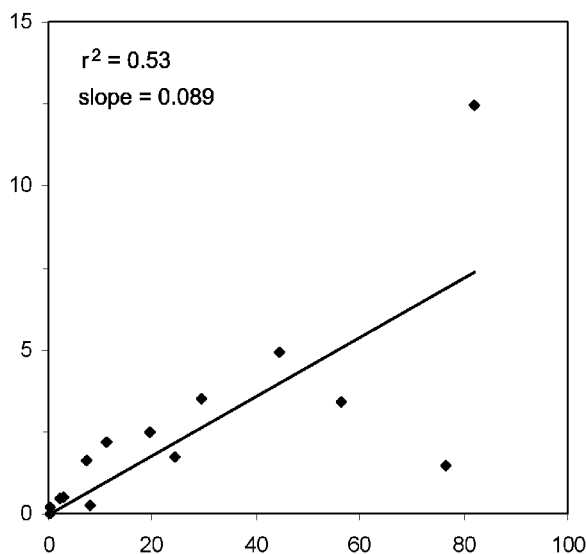


Figure 4 Correlation between the in vivo blood clearances by glucuronidation and hepatic clearances predicted from published human liver microsomal CL_{int} values (using the dispersion model) for amitriptyline, clofibric acid, 5,6-dimethylxanthene-4-acetic acid, lamotrigine, morphine, naloxone, naproxen, propofol, valproic acid, zidovudine (as reported in Reference 72), codeine, ethinylestradiol, gemfibrozil (as reported in Reference 73), and dihydroartemisinin (reported in Reference 74). K_m values used for the calculation of the CL_{int} values for amitriptyline, ethinylestradiol, gemfibrozil, and propofol were corrected for nonspecific binding to human liver microsomes using literature values of $f_{u_{inc}}$ (16, 73).

in vivo plasma clearance in this species (76), which provided the basis for selection of an analogue with improved metabolic stability.

Factors Influencing Drug Glucuronidation Kinetics In Vitro

The reason(s) for the underestimation of in vivo CL_H from in vitro kinetic data remain unclear. The possibility remains that the assumptions underpinning the mathematical models of CL_H are not applicable to drug glucuronidation by human liver microsomes. As noted earlier, UGT is localized on the luminal side of the microsomal membrane and this may give rise to “diffusional” barriers. Indeed, Soars et al. reported that CL_{int} values generated using human hepatocytes as the enzyme source predicted glucuronidation clearance in vivo reasonably well and speculated that transport of drug substrates may be more limited in microsomes than in intact hepatocytes (73). It has also been suggested that renal glucuronidation may be a major contributor to drug glucuronidation in vivo, but this seems unlikely (discussed below). Nonspecific microsomal binding presents another confounding factor in the calculation of kinetic constants for human liver microsomal drug metabolism reactions, leading to overestimation of K_m and, hence, underprediction of CL_{int} (16). However, nonspecific binding is minor for most of the drugs reported in Figure 4 and was accounted for when known to be significant (e.g., propofol).

The quality of the in vitro kinetic data used for in vitro–in vivo extrapolation warrants special consideration. The success of predictions of in vivo CL_H is critically dependent on in vitro CL_{int} and how closely the kinetic parameters (K_m , V_{max}) used to derive this parameter reflect enzyme activity in vivo. Incubation components are known to modulate microsomal UGT activity, and the kinetics of drug glucuronide formation therefore varies with experimental conditions. In particular, we have demonstrated that the kinetics of human liver microsomal zidovudine glucuronidation are dependent on buffer type, pH, and ionic strength, and on the presence of Mg^{2+} , detergent, alamethacin, and the endogenous activator UDP-N-acetylglucosamine (72). Depending on incubation conditions, in vitro CL_{int} varied almost sixfold. Because investigations of UGT activity in vitro often employ widely differing reaction conditions (77), comparison of drug glucuronidation kinetic data between laboratories is frequently not meaningful.

Drug Glucuronidation Kinetics In Vitro

The interpretation and analysis of drug glucuronidation kinetic data further impacts on the reliability and predictive value of in vitro CL_{int} . Until recently, kinetic plots of drug glucuronidation by human liver microsomes or recombinant isoforms were published infrequently and data were generally uncritically fitted to the Michaelis-Menten equation to obtain estimates of K_m and V_{max} . However, there is increasing evidence demonstrating that drug glucuronidation reactions in vitro commonly exhibit atypical kinetic behavior. Estradiol 3-glucuronidation by human

liver microsomes (believed to involve UGT1A1; see previous discussion) and 1-naphthol glucuronidation by recombinant UGT1A1 follow sigmoidal kinetics characteristic of autoactivation (56, 78, 79). However, UGT1A1-catalyzed 4-methylumbelliferone glucuronidation follows hyperbolic (Michaelis-Menten) kinetics (79). Consistent with these observations, alternate UGT1A1 substrates variably caused activation or inhibition of human liver microsomal estradiol 3-glucuronidation (78). Recent studies in this laboratory have also provided evidence for multiple kinetic mechanisms in UGT2B7-catalyzed xenobiotic glucuronidation. Morphine 3- and 6-glucuronidation by recombinant UGT2B7 exhibit biphasic kinetics suggestive of “negative cooperativity” (65), whereas 4-methylumbelliferone and zidovudine glucuronidation by this enzyme follow sigmoidal and hyperbolic kinetics, respectively [(63); J.O. Miners & P.I. Mackenzie, unpublished data]. However, not only may glucuronidation kinetic models vary between substrates for the same isoform, but kinetic behavior varies between isoforms for the glucuronidation of a common substrate. For example, we have observed hyperbolic (Michaelis-Menten), substrate inhibition and sigmoidal (homotropic positive cooperativity) kinetics for the glucuronidation of the nonselective substrate 4-methylumbelliferone by UGT1A1, UGT1A3, and UGT2B7, respectively (Figure 5) [(79); J.O. Miners & P.I. Mackenzie, unpublished data].

These observations are consistent with the existence of allosteric effector sites or the simultaneous binding of two substrate molecules to the active site, mechanisms that have been proposed for CYP3A4-catalyzed reactions. Alternatively, UGTs may act as cooperative ligand-binding multisubunit enzymes because there is evidence indicating at least some UGT isoforms exist as dimers (see previous discussion). Irrespective of the mechanism, it is clear that model-fitting is essential for the description and kinetic analysis of UGT-catalyzed reactions, and multisite models developed for CYP3A4 (80, 81) are likely to prove useful in this regard. A

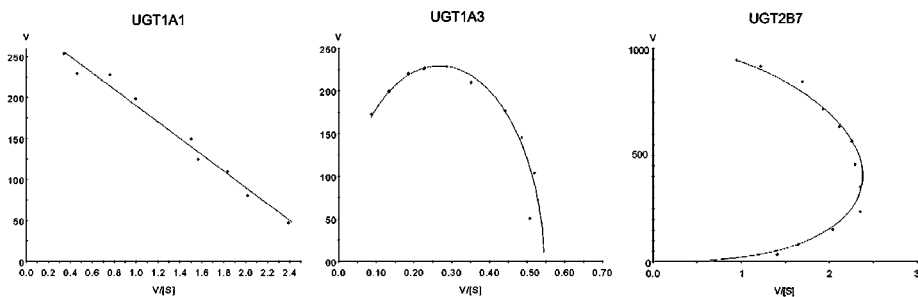


Figure 5 Eadie-Hofstee plots for 4-methylumbelliferone glucuronidation by UGT1A1, UGT1A3, and UGT2B7. Points show experimentally derived values. Curves of best fit were generated using expressions for hyperbolic (UGT1A1), substrate inhibition (UGT1A3), and homotropic positive cooperativity (UGT2B7) models. Units: rates of reaction (v) as pmol glucuronide formed/min mg and substrate concentration ($[S]$) as μM .

typical glucuronidation kinetics further complicate in vitro–in vivo extrapolation as the in vivo correlate of nonhyperbolic kinetics in vitro remains obscure (80).

Extrahepatic Drug Glucuronidation

UGTs are widely distributed throughout the body. Apart from the liver, UGT1A3, -1A9, and -2B7 are expressed in kidney and multiple isoforms are expressed throughout the gastrointestinal tract (25, 42). Not surprisingly, UGT activity is readily measurable in human kidney and gastrointestinal tract and this has led to the proposal that these organs contribute significantly to the systemic clearance and first-pass extraction, respectively, of glucuronidated drugs (73, 82–84). In particular, the rates of glucuronidation of a number of drugs by human kidney microsomes are at least comparable to those of human liver microsomes (73, 82, 84). However, in proposing a major role for the kidney in systemic drug clearance, such studies have failed to recognize that scaling factors for renal and hepatic kinetic parameters differ, and when relative organ weights and microsome yields are taken into account, metabolic clearance by the kidney appears to be low (85). Interpretation of drug glucuronidation kinetic data by human small intestine similarly requires careful interpretation because UGT activity and microsome yield vary between duodenum, jejunum, and ileum (42, 86). Nevertheless, comparison of CL_{int} values derived using liver, kidney, and intestinal microsomes may provide important insights into the relative contribution of these organs to CL_H and/or first-pass extraction (85) and the role of glucuronidation as a local detoxification mechanism.

IN SILICO REACTION PHENOTYPING

Computational methods potentially provide a facile and economic alternative to in vitro reaction phenotyping approaches. In silico reaction phenotyping utilizes pattern recognition techniques to elucidate a set of chemical properties (descriptors) associated with the binding and metabolism of substrates by an enzyme. The two- and three-dimensional (2D and 3D) descriptors employed represent chemical features extracted by a well-defined algorithm from a molecular representation of a complex system (87).

Implicit Classification

If a set of descriptors is common to the known substrates of an enzyme, the presence of these descriptors in other compounds can be used to classify them as substrates of that enzyme by implication. Descriptors used to imply classification are commonly found using pharmacophore elucidation or the development of 2D-QSAR. Pharmacophores, which represent a configuration of common structural features associated with biological activity (in this case, metabolism by an isoform), provide one of the most intuitive 3D descriptors used for reaction

phenotyping. Ideally, a pharmacophore should be unique for a particular isoform. Adoption of *in silico* reaction phenotyping for compounds metabolized by CYP has progressed in parallel with the increasing availability of isoform-selective substrates and inhibitors, which are necessary for pharmacophore generation, and has additionally been aided by the availability of homology models based on the X-ray crystal structures of bacterial and mammalian P450s (21, 22, 88, 89). As discussed previously, the substrate selectivities of UGT isoforms have begun to approach interpretable levels and, hence, a basis now exists for pharmacophore and QSAR modeling of UGT substrates. Indeed, such approaches assume particular relevance in the absence of a UGT X-ray crystal structure.

Recent studies in this laboratory resulted in development of the first 2D-QSAR, 3D-QSAR, and pharmacophore models for substrates of human UGT isoforms (79, 90, 91). In generating pharmacophores for UGTs, it became apparent that it was essential to overlay the glucuronidation sites of the individual substrates to obtain catalytically sensible alignments. The benefits of incorporating sites of metabolism have similarly been highlighted in a recent evaluation of pharmacophores for substrates of CYP2B6 (92). Common features pharmacophores, which included an essential "glucuronidation feature," have been developed for UGT1A1, UGT1A4, and UGT1A9 (79, 90, 91), and these are shown in Figure 6. Owing to their similarity, the utility of the pharmacophores for reaction phenotyping substrates of the individual isoforms is probably limited, although the pharmacophores do provide important insights into the binding requirements of UGT1A family enzymes more generally. The site of glucuronidation is invariably adjacent to a hydrophobic region, with another hydrophobic domain located 6 to 8 Å from the site of conjugation. A hydrogen-bond acceptor near the distal hydrophobe differentiates UGT1A9 substrates, but high-affinity substrates for UGT1A4, and possibly UGT1A1, may additionally benefit from hydrogen-bonding interactions (90). The common "core" features associated with each pharmacophore represent a molecular basis for the overlapping substrate selectivities characteristic of UGT1A isoforms.

There are two important limitations of pharmacophores that impact on their application to UGTs. First, it is assumed that the substrates used to define the pharmacophore share the same binding mode (93). The atypical glucuronidation kinetics observed for some UGT substrates and known ability of UGT isoforms to glucuronidate poly-functional substrates at multiple sites [for example, clozapine; (94)] indicate that this may not always be the case. Second, the presence or absence of pharmacophoric features in nonsubstrates is seldom investigated, and, hence, it remains unclear whether this approach can differentiate substrates from nonsubstrates.

Not uncommonly, descriptors characteristic of substrates are extracted from a QSAR, which is a quantitative relationship between descriptors and an activity measurement. Two-dimensional descriptors characteristic of human CYP isoforms have been identified from 2D-QSAR developed using kinetic constants, such as K_m and K_i (95), allowing differentiation of CYP1, -2A, -2B, -2C, -2D, -2E, and

-3 substrates on the basis of physicochemically interpretable descriptors, such as pK_a , logP, and collision diameter. Again, this approach does not necessarily aid the distinction of substrates and nonsubstrates.

Explicit Classification

In combination with known isoform substrate selectivities, carefully selected and relevant nonsubstrate data provide valuable information when classification algorithms are used to probe explicit differences between the two sets, and this represents another approach to *in silico* reaction phenotyping. Afzelius et al. reported the use of partial least squares discriminant analysis (PLS-DA) to classify inhibitors and noninhibitors of CYP2C9, with an accuracy of 74% (96). Nonlinear classification algorithms potentially provide greater flexibility and generalization performance (97). For example, an artificial neural network (ANN) based on 2-D Unity “fingerprints” has been developed that recognized CYP3A4 inhibitors with 89% accuracy (98). Nonlinear Support Vector Machine (SVM) algorithms, which have been successfully applied to classifying drug CNS-permeability (99), provide another option for reaction phenotyping.

A comprehensive database of all reported substrates and nonsubstrates of human UGT isoforms has been compiled in this laboratory to investigate the utility of various classification techniques for UGT reaction phenotyping. PLS-DA, ANN, and SVM were compared using 2D chemical descriptors generated for substrates and nonsubstrates of twelve human UGT isoforms (100). Using SVM, predictability was excellent (>80% accuracy) for five isoforms and good (63%–80% accuracy) for the other seven, confirming the potential value of this approach. The variability between isoforms probably reflects differences in the size and structural diversity of the datasets.

As noted previously, there is evidence to suggest multiple binding modes for UGT substrates. Multiple pharmacophores have been utilized to characterize molecular recognition by a number of xenobiotic binding proteins (101–103) and recent studies in this laboratory investigated whether an “ensemble” of pharmacophores, which define the chemical features relevant to each of the possible binding modes, would prove useful for UGT substrate reaction phenotyping (M.J. Sorich, J.O. Miners, R.A. McKinnon & P.A. Smith, manuscript in preparation). Employing concepts developed for pharmacophore fingerprinting (104, 105) and using pattern-recognition techniques to select subsets of pharmacophores associated with substrates and nonsubstrates, models that were more intuitive but marginally less predictive than classification using 2D descriptors were generated. In addition, it was observed that a number of the pharmacophores selected as important included simple chemical features. Further analysis demonstrated significant isoform-related differences in the prevalence of nucleophilic functional groups (e.g., phenol, hydroxyl, carboxyl, imidazole, and primary, secondary, and tertiary amine function) in substrates. These simple, intuitive features could be used for classification with equal or better accuracy than approaches using 2D descriptors or pharmacophore fingerprints. This method is likely to prove most valuable when

conjugative regioselectivity is included. In this regard, nucleophile preferences between isoforms may well correlate with quantum chemical descriptors. Importantly, advances in the prediction of CYP regioselectivity using quantum chemical descriptors (106, 107) is in no small part due to the availability of training data derived from thorough characterization of metabolite regioselectivity. Diligence in the characterization of conjugative regioselectivity is required in order to realize similar advances with *in silico* UGT reaction phenotyping.

IN SILICO–IN VITRO PREDICTION

As discussed previously, the kinetic parameters K_m , V_{max} , and CL_{int} underpin the *in vitro*–*in vivo* extrapolation of kinetic data for drugs eliminated by hepatic metabolism. Kinetic parameters generated *in silico* may potentially be used as surrogates for experimentally derived values, thereby precluding the requirement for time-consuming and expensive *in vitro* studies using human tissues and/or recombinant enzymes. Both 2D- and 3D-QSARs have been generated for numerous CYP isoforms that allow prediction of K_m (or K_i in the case of inhibitors) (21, 22, 95). However, in most instances, predictability varies by up to 1 log order.

UGT QSMR were developed initially from activity data using liver microsomes from various species as the enzyme source, and include the pioneering studies of Bray et al. (108) and Hansch et al. (109) through to later mechanistic (110) and kinetic investigations (111). More recently, studies have been conducted with recombinant UGTs, although these have tended to investigate QSMR for compounds from the same chemical class (for example, References 112, 113). Reports from this laboratory represent the first attempts to develop UGT QSMR from structurally diverse compounds. Linear 2D-QSAR approaches were found to outperform pharmacophore and molecular field-based 3D-QSAR alignment methods for UGT1A1 and UGT1A4 (79, 90), with the 2D-QSAR generally predicting the K_m for substrates of these enzymes within 0.5 log order. Interestingly, although the generation of predictive models was possible with UGT1A1 and UGT1A4, attempts to develop predictive QSMR for substrates of UGT1A9 were unsuccessful (91).

Further development of QSMR capable of predicting the kinetic parameters of glucuronidated compounds is problematic and complicated by several factors. Data sets available for modeling are generally limited in size and vary in quality. As discussed above, kinetic constants published by different laboratories are frequently not comparable owing to the use of differing incubation conditions. The atypical kinetic behavior of many glucuronidation reactions further confounds data generation and selection for QSMR modeling. Indeed, classification of kinetic mechanism would appear to be a necessary precursor to the modeling of K_m and other parameters. Efforts to date have focused solely on prediction of K_m , but V_{max} is also required for calculation of CL_{int} . Development of a 2D-QSAR for prediction of V_{max} has been reported recently for the CYP2E1-catalyzed metabolism of a series of alkylbenzene derivatives (i.e., a “local” model suitable only for prediction of V_{max} values for structurally related molecules) (114). The limited

UGT kinetic data currently available similarly lends itself only to the development of local models. Increasingly, however, there is a requirement for “global” or generalizable ADMET models (115, 116). Given the inherent complexity of the physiological processes involved and the vastness of the chemical space to be investigated, sophisticated pattern recognition techniques and large, quality data sets will undoubtedly be required to generate globally predictive models in future.

CONCLUSIONS

The UGT reaction phenotyping of glucuronidated drugs and other compounds using human liver microsomal preparations and possibly isolated hepatocytes is feasible, although currently limited in scope by the relatively few isoform selective “probes” available. However, the number of isoform-selective substrates and inhibitors will undoubtedly increase as increasing numbers of structurally diverse UGT substrates are identified and characterized. Isoform-specific inhibitory antibodies may provide another option in this regard. Quantitative characterization of the expression of isoforms in liver and other tissues will similarly improve the interpretation of relative activity measurements obtained using recombinant UGT isoforms. Available evidence suggests that quantitative prediction of CL_H and E_H from human liver microsomal kinetic data is unlikely to be successful, although the application of a scaling factor may at least differentiate high- and low- CL_H glucuronidated drugs. Further exploration of this option is warranted, as is the utility of isolated human hepatocytes for the generation of kinetic parameters that may be accurately extrapolated. Irrespective of the enzyme source, however, atypical kinetic behavior may confound meaningful *in vitro*–*in vivo* correlation for many glucuronidated compounds. Apart from *in vitro* approaches, *in silico* modeling shows promise for the reaction phenotyping of UGT substrates. In particular, substrates and nonsubstrates of individual UGTs may be characterized using 2-D descriptors and combinations of pharmacophores, the latter incorporating “site of conjugation” features, which account for multiple, catalytically meaningful binding modes and facilitate model interpretability. Flexible nonlinear classification algorithms, such as SVM, are most suited to delineating the complex relationships between chemical structure and glucuronidation (or absence thereof) by an isoform. The UGT QSMR field is in its infancy. Although the recent generation of 2D- and 3D-QSAR, which predict the K_m values of substrates of UGT1A1 and UGT1A4, demonstrates the potential utility of this approach, the development of global UGT QSMR will ultimately require large data sets that encompass the multiple processes associated with metabolism by UGT.

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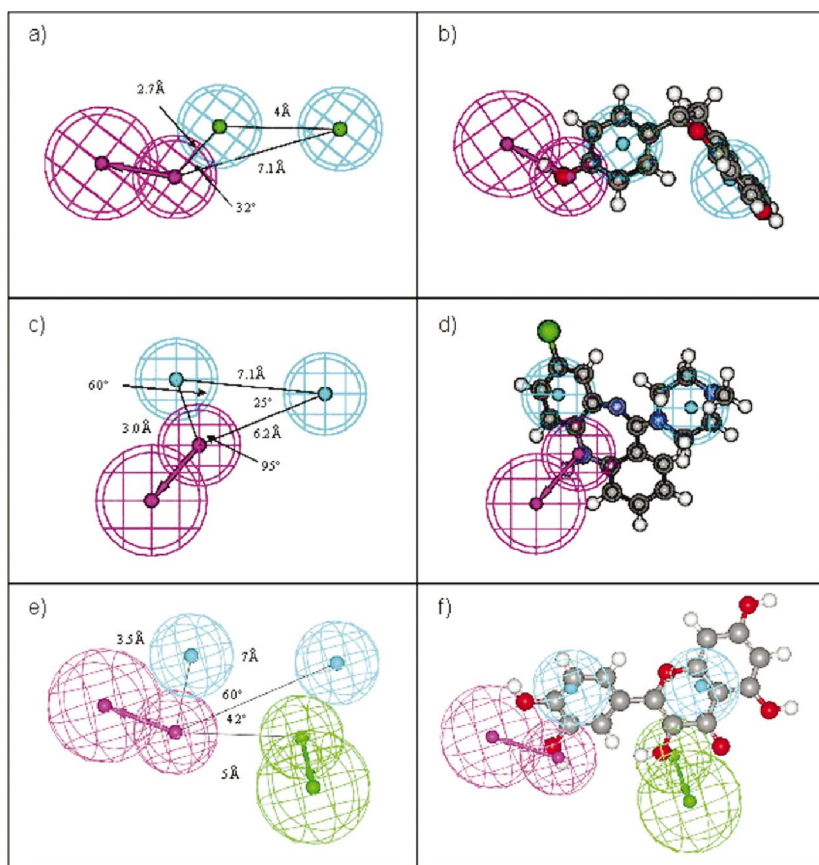


Figure 6 UGT1A1 (*panel a*), UGT1A4 (*panel c*), and UGT1A9 (*panel e*) common features pharmacophores. The cyan, green, and purple spheres represent a hydrophobic feature, hydrogen bond acceptor, and glucuronidation feature, respectively. Bold arrows show the direction of lone-pair electron donation. Panels *b*, *d*, and *f* show naringenin, clozapine, and quercetin mapped to the respective UGT1A1, UGT1A4, and UGT1A9 pharmacophores.