Assessment of activity levels for CYP2D6*1, CYP2D6*2, and CYP2D6*41 genes by population pharmacokinetics of dextromethorphan

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Abstract: The pharmacokinetics of dextromethorphan (DM) is markedly influenced by cytochrome P450 2D6 (CYP2D6) enzyme polymorphisms. The aim of this study was to quantify the effects of the CYP2D6*1, *2, and *41 variants on DM metabolism in vivo and to identify other sources of pharmacokinetic variability. Concentrations of DM and dextrorphan (DO) in plasma and urine were evaluated in 36 healthy Caucasian men. These volunteers participated in three clinical studies and received a single oral dose of 30 mg DM-HBr. Data were modeled simultaneously using the population pharmacokinetics NONMEM software. A five-compartment model adequately described the data. The activity levels of the alleles assessed differed significantly. The clearance attributable to an individual CYP2D6*1 copy was 2.5-fold higher as compared with CYP2D6*2 (5,010 vs. 2,020 l/h), whereas the metabolic activity of CYP2D6*41 was very low (85 l/h). Urinary pH was confirmed as a significant covariate for DM renal clearance. These results refine genotype-based predictions of pharmacokinetics for DM and presumably for other CYP2D6 substrates as well.

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ACTIVITY ASSESSMENT FOR CYP2D6*1, CYP2D6*2 AND CYP2D6*41 GENES BY POPULATION PHARMACOKINETICS OF DEXTROMETHORPHAN

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Abstract

Dextromethorphan pharmacokinetics is markedly influenced by CYP2D6 enzyme polymorphisms. The aim of this study was to quantify the effect of the *CYP2D6* *1, *2* and *4I* variants on dextromethorphan metabolism in vivo and to identify other sources of pharmacokinetic variability. Plasma and urine concentrations of dextromethorphan and dextrorphan were evaluated from 36 healthy Caucasian males. These volunteers participated in three clinical studies and received a single oral dose of 30 mg dextromethorphan-HBr. Data were modelled simultaneously using the population pharmacokinetic NONMEM software. A five-compartment model adequately described the data. Activities of the alleles assessed differed significantly. The clearance attributable to an individual *CYP2D6* *1* copy was 2.5-fold higher compared to *CYP2D6* *2* (5010 vs 2020 L/h) while the metabolic activity for *CYP2D6* *4I* was very low (85 L/h). Urinary pH was confirmed as a significant covariate for dextromethorphan renal clearance. These results refine genotype-based prediction of pharmacokinetics for dextromethorphan and presumably other CYP2D6 substrates.
Introduction

Dextromethorphan (DM) is an antitussive drug with rapid absorption from the gastrointestinal tract ($k_a = 2.6 +/- 0.9 \text{ h}^{-1}$) after oral administration and a large central volume of distribution 961 L ranging from 585 to 1292 litres. (1) DM undergoes rapid and extensive first-pass metabolism mainly by CYP2D6 with an oral bioavailability ranging from 1% to 2% and about 80% in extensive and poor CYP2D6 metabolizer subjects, respectively (2) Its intrinsic hepatic clearance was about 22L/h and 4737 L/h in respective CYP2D6 metabolizer groups. (2) In another study a value of 1280 ± 483 L/h for its apparent clearance has been reported for CYP2D6 extensive metabolizers. (3) Primary metabolic steps include formation of dextrorphan (DO) mainly by CYP2D6 and of 3-methoxymorphinan by CYP3A enzymes. Both are further metabolized to 3-hydroxymorphinan by CYP3A4 and CYP2D6, respectively. (4, 5) The dominating metabolic pathway in the majority of the population is the conversion of DM to DO by CYP2D6, contributing to more than 80% to the formation of DO. (1, 6) The selectivity of DO formation via CYP2D6, in addition to its favourable safety and availability, has made DM a probe of choice for CYP2D6 phenotyping. (7)

CYP2D6 is a highly polymorphic gene with more than 100 known allelic variants (March 15, 2010; http://www.cypalleles.ki.se/cyp2d6.htm). Because the overall disposition of DM is highly dependent on CYP2D6 activity, these polymorphisms are the main source for the wide interindividual variation in its plasma levels and response. (8, 9) For other CYP2D6 substrates such as antidepressants and antipsychotic drugs such variability can lead to failure of treatment in carriers of alleles encoding very high CYP2D6 activity or may expose individuals carrying alleles conferring no or low activity to a high risk of toxicity. (10, 11) Conversely, for some opioids such as dihydrocodeine, codeine and tramadol, where a highly active metabolite is formed by CYP2D6, therapeutic failure may be associated with low activity genotypes and toxicity with genotypes resulting in high enzymatic activity. (12, 13)

As a prerequisite to personalize therapy with CYP2D6 substrates, different proposals have been made to establish simple useful systems for the prediction of CYP2D6 phenotype based on CYP2D6 genotype. (14) Such predictions suffer from potential pitfalls depending on the correctness of underlying assumptions. The classical procedure divided populations based on the urinary metabolic ratio of DO over DM and an arbitrary “cut-off” value, i.e. antimode, into poor and extensive metabolizers. (15) While subjects with two non-functional alleles represent the poor metabolizer group, it is difficult to precisely predict the phenotype of individuals carrying combinations of functional and reduced function alleles. Furthermore,
wide variability in CYP2D6 activity is not only observed between subjects with different functional genotypes, but is also rather the rule than the exception among subjects within a genotype group.(16) Attempts to better classify especially those subjects carrying gene duplications and reduced function alleles, ultra-rapid and intermediate metabolizers subgroups have been introduced creating a polymodal population distribution.(17)

Another system called semi-quantitative scoring system defining categories for CYP2D6 alleles has been developed(18) which assigns scores to each allele (1 for fully functional alleles, 0.5 for reduced activity functional alleles, and zero for non-functional alleles) and uses these ‘gene doses’ as an indicator for phenotype. Another recent scoring system, the so-called activity score system, also assigns values to individual CYP2D6 alleles and utilizes their sum as a surrogate to predict phenotype.(8) Because the model did not improve when assigning values of 0.25 and 0.75 to alleles perceived to confer slight and substantial reduction in activity, the simplest model using values of 1 for fully functional, 0.5 for all reduced activity, and 0 for non-functional alleles, respectively, was chosen. In an effort to better describe the alleles carrying functional gene duplications such as CYP2D6*1xN or *2xN, the value of the allele was doubled, i.e. a value of 2 was assigned to CYP2D6*1xN. Although these classification systems have improved phenotype prediction, considerable variability within score groups persists.

By definition, these scoring systems are valid for non-functional alleles such as CYP2D6*3 and*4, but the simplifying assumption that fully functional alleles such as CYP2D6*1 and *2 have identical activity enzymes can be challenged, as well as the allocation of a value of 0.5 to any low activity allele. Indeed, several mechanisms including differences in splicing and other reasons for different expression levels, enzyme stability, and/or substrate specific enzyme kinetics may cause net differences for in vivo substrate turnover between individual functional CYP2D6 variants.

Thus, there is a need to precisely quantify the in vivo activity associated with CYP2D6 allelic variants and genotypes to better predict phenotype from genotype data. Such information will ultimately be useful for clinical decision making in respect to drug selection and dose adjustment. In the present study we assessed the activity of individual CYP2D6 alleles with regard to the clearance of DM to DO, using a population pharmacokinetic approach.

Results
A total number of 537 plasma samples and 136 urine samples from 36 healthy Caucasian men obtained after oral administration of a 30 mg DM-HBr dose were analysed for DM and DO concentrations in this study. Demographic and CYP2D6 genetic characteristics of these volunteers are given in Table 1 and 2, respectively.

As expected, we observed large interindividual differences in the time-concentration profiles within the data set, with the highest variability for DM. DM initial plasma concentrations decreased rapidly in many, but not all, individuals (within the first 5 hours after dosing) reflecting the known variability in drug clearance. DO plasma concentrations appeared in many individuals earlier than the parent drug, supporting the importance of first-pass metabolism.

Modelling all data with the basic structure 4-compartment model (see Methods section) was not sufficient to describe the data, with misspecifications especially for parent concentrations in plasma. Addition of a peripheral compartment for DM was associated with a significant drop in the objective function value ($\Delta = -114$). Based on individual and population predictions, this model was still unable to account for the early appearance of DO in plasma. Therefore, other models (19, 20) that take the first pass metabolism into account were tested. These models were stepwise simplified or extended for modelling the available data. A hypothetical metabolism compartment was added assumed to be in rapid equilibrium with the central compartment of the parent drug, with all metabolic steps taking place in this compartment. After administration of the drug, it was assumed that the drug goes from the absorption site to the metabolism compartment before reaching the plasma (for specifications, see appendix A). This model was associated with a profound drop in the objective function value ($\Delta = -324$). Inclusion of a lag-time parameter led also to further significant drop in the objective function ($\Delta = -45$).

With respect to identification of covariates, as expected CYP2D6 genotype had a major impact on the metabolic clearance ($\Delta = -532$). About 55% of the interindividual variability (CV %) in DM metabolic clearance, CL23, was explained by addition of CYP2D6 genotype as a covariate. Inclusion of urine pH values as a covariate on renal clearance of DM also resulted in a significant drop in the objective function value ($\Delta = -105$). The best form of this relationship was that included in the final model and is described by the following empirical equation:

$$0_{PV} = 0_{TV} \cdot (5.7/UpH)^{0s}$$
where $\theta_{PV}$ is the individual value of renal clearance, $\theta_{TV}$ is the population value of renal clearance, the value 5.7 represent a published mean of urine pH in humans(21), $U_{\text{pH}}$ is the measured individual urine pH at each collection period, and $\theta_{s}$ is the shape parameter which explains the change in renal clearance in terms of change in urine pH. Using the non-ionized fraction of DM calculated according the Henderson-Hasselbalch equation as a covariate was inferior to this empirical equation. Including an effect of urine pH on renal clearance of DO did neither improve model fitting nor led to a decrease in the OFV. The model also identified age as important covariate. Age contributed to interindividual variability in the apparent volume of distribution of DO ($\Delta = -21.9$) and the clearance of DO to other species, CL$_{30}$, ($\Delta = -5.8$). This covariate was modelled according to the relationship:

$$
\theta_{PV} = \theta_{TV} \cdot \exp (\theta_{AGE} \cdot (\text{AGE} - 27))
$$

where $\theta_{PV}$ is the population value of the model parameter, $\theta_{TV}$ is the typical value in an individual with age of 27 years old (the median age in this study), $\theta_{AGE}$ is the fractional change in $\theta_{PV}$ per year different from 27 years of age. None of other apparent volumes of distribution, clearance, or inter-compartmental clearance parameters has been found to be influenced by age in this study. Neither body weight nor height was found to significantly contribute to model disposition parameters. Plasma data were best fitted if combined additive and proportional error terms were used, while urine data were best fitted with proportional error terms. The blueprint of the final model is shown in Figure 1.

Final pharmacokinetic parameter estimates with 95% confidence intervals are shown in Table 3. The apparent metabolic clearance (CL$_{23}$) estimate ranged from 10 to 10030 L/h in this population. The fraction of the clearance attributable to a CYP2D6*1 copy was on average 2.5-fold greater than that for CYP2D6*2. CYP2D6*41 related clearance was lowest: the point estimates per copy were 5010 L/h, 2020 L/h and 85 L/h for CYP2D6*1, *2 and *41, respectively.

Diagnostic plots of the final model are given in Figure 2, indicating model adequacy. The 5th, 50th and 95th percentiles from simulated data, based on the final model estimates, are shown in Figure 3. Simulated time courses of plasma concentrations and cumulative amounts excreted in urine for subjects who are homozygous for a given allele, i.e. CYP2D6*1/*1, *2/*2, etc, are given in figure 4. The expected values of standard CYP2D6 phenotypic metrics, i.e. metabolic ratios of DM over DO at 3h in plasma and 0-8h in urine, were 0.12 and 0.015 for CYP2D6*1/*1, 0.28 and 0.04 for CYP2D6*2/*2, 5.0 and 0.079 for CYP2D6*41/*41, and 90.9 and 14.3 for CYP2D6*4/*4 carriers (simulated for a 27-years old male having a urine pH of 5.7).
Discussion

The main objective of this study was to quantify the metabolic activity attributable to frequent individual CYP2D6 alleles using DM as a probe drug. The mean CYP2D6 mediated clearance $\text{CL}_{\text{CYP2D6}}$ is estimated as 15030 L/h in subjects genotyped as $CYP2D6^{*1/*1}x2$, whereas it is 170 L/h in subjects who carry two $CYP2D6^{*41}$ alleles, and zero for subjects with genotypes comprising two non-functional alleles. The results suggest that quantitative differences in DM based phenotyping metrics between CYP2D6 genotypes are similar to the differences in enzyme activity, despite additional processes with impact on DM pharmacokinetics (see Table 3).

Based on the final model estimates, the activity for the $CYP2D6^{*1}$ allele with respect to formation of DO in vivo is about 2.5-fold higher compared to that seen for the $CYP2D6^{*2}$ allele. The CYP2D6.2 protein differs from the CYP2D6.1 protein in two amino acids (R296C and S486T) (http://www.cypalleles.ki.se/cyp2d6.htm), while there is no evidence that the $CYP2D6^{*2}$ allele differs from the $^{*1}$ allele with respect to expression levels. Key substrate binding sites of CYP2D6 include D301, F120, F483 and F481. However, both positions 296 and 486 are supposed to be part of the substrate recognition site regions, and especially replacement of the large positively charged arginine by cysteine close to position 301 may well modify substrate binding but not profoundly change catalytic properties. Indeed, early reports by Sachse et al(17) showed that presence of the $CYP2D6^{*2}$ allele was related to a slightly reduced enzyme activity as determined by urinary DM and/or debrisoquine phenotyping metrics. Both DM and pinoline intrinsic clearance for the respective O-demethylation in enzyme variants purified after expression using a baculovirus-mediated system were only a fifth for CYP2D6.2 compared to CYP2D6.1.(24, 25) Differences were smaller but in the same direction for other reactions: DM N-demethylation, 0.77-fold; codeine O-demethylation, 0.35-fold; and fluoxetine N-demethylation, 0.26-fold.(24) Intrinsic clearance for CYP2D6.2 expressed in Saccharomyces cerevisiae compared to CYP2D6.1 was also about 0.5-fold lower for several progesterone hydroxylation pathways and 0.23-fold lower for dopamine formation from D-tyramine.(26) The respective activity ratio of CYP2D6.2 over CYP2D6.1 expressed in yeast was 0.44-fold for DM O-demethylation and 0.74-fold lower for bufurolol 1’-hydroxylation, while there were no differences with regard to metabolism of debrisoquine or metoprolol.(27) In a COS-7 expression system, CYP2D6.2 reached 0.71-fold and 0.72-fold activity relative to CYP2D6.1 for DM O-demethylation and bufuralol hydroxylation.(28) In summary, on an average CYP2D6.2 is approximately half active than CYP2D6.1. Whether the reported differences in
this ratio between substrates in the in vitro assays are real or rather reflect the notorious problems in reproducibility of in vitro enzyme kinetic parameters(29) remains to be assessed in detail by clinical studies. Indeed, differences in urinary DM/DO metabolic ratios in a Gabonese population(30) showed the same difference between CYP2D6*1 and *2 carriers as observed here.

Furthermore, this study shows that the metabolic clearance of DM conveyed by CYP2D6*41 is markedly less than that of CYP2D6*1. It is well-known that the activity for CYP2D6*41 is lower when compared to CYP2D6*1 and *2. CYP2D6*41 codes for the lower activity CYP2D6.2 protein and in addition has a lower level of expression, as shown by 2- to 5-fold lower levels of CYP2D6 mRNA containing exon 6, apoprotein and enzyme activity in mechanistic studies.(31) When comparing to individuals with only one functional *1 allele, there was a 0.27-fold lower mean enzyme activity in individuals with one functional *41 allele as quantified by the sparteine metabolic ratio (data taken from graph) (32). This is the first quantitative estimate for activity attributable to CYP2D6*41 in vivo based on DM O-demethylation. The extent of the difference (0.017-fold) is approximately one order of magnitude higher than expected from previous information and needs to be confirmed in larger groups with this genotype.

The finding that the CYP2D6*1 allele is associated with an about 2.5-fold higher activity with dextromethorphan compared to CYP2D6*2 would suggest that different activity values need to be assigned to these alleles in scoring systems. Both the activity score(8) and semi-quantitative scoring(18) systems assumed similar activity for the CYP2D6*1 and *2 alleles. To reflect reduced activity, these systems(8, 18) assigned a value of 0.5 to the CYP2D6*41 allele. Based on the results presented here, clearance for CYP2D6*41 is approximately 60-fold reduced in comparison to CYP2D6*1. Consequently scoring values of 1, 0.4 and 0.017 should be assigned to CYP2D6*1, *2 and *41, respectively to more accurately reflect their activity. Using the respective final parameter estimates, DM pharmacokinetics for various genotypes were simulated (Figure 4). Similar genotype specific predictions can be made for other CYP2D6 substrates. Such predictions, however, may need to be modified, since previous studies demonstrated that protein specific activity of cytochrome P450 enzymes including CYP2D6 is in part substrate dependent.(11, 33, 34)

The final pharmacokinetic model describes the pharmacokinetic profiles of DM and its major metabolite adequately. In this model, inclusion of the hypothetical metabolism compartment
was able to explain the appearance of DO in plasma before DM. The presented model shares some features with previously reported models that have been successfully applied to describe first-pass metabolism.(1, 20) The specific features in our model are that, (i) activity of active alleles of CYP2D6 were estimated separately, (ii) urinary data of both parent and metabolite were included in the model and simultaneously evaluated with plasma data; (iii) the impact of urine pH on DM clearance was quantified; (iv) it can be extended to estimate the activity of other additional alleles such as *9, *10, and *17 alleles in other data sets. These features increase applicability of the model, while its clinical implications remain to be ascertained.

Our findings are in line with published data assuming a two-compartment model for DM (1, 35) and one-compartment models for DO.(1) The absorption rate constant (h⁻¹) estimate is about 0.25, which lies between previously reported values of 2.6 h⁻¹ and 0.1 h⁻¹.(1, 36) The lag time was estimated to be 0.31 h, which also lies between previously reported values 0.8 h and 0.087 h.(1, 35) The model estimates for the apparent central and peripheral volumes of distribution of DM in this study are smaller at 648 L and 1560 L compared with the previously reported values of 961 L and 1951 L.(1) The volume of distribution of DO with 419 L is smaller than the reported value of 650 L (37) and 3776 L (1), but higher than 109 L reported by others. (38) Urine pH plays a significant role as a covariate and explains part of the variability seen in DM metabolic ratios. This is in agreement with a previous study describing an approximate 20-fold variation in the urinary metabolic ratio due to urinary pH.(39) Taken together, these comparisons suggest that our model is valid.

The goodness-of-fit plots (Figure 2) show a possible overestimation of some DM and DO concentrations in plasma/urine, both due to model misspecification which is a semi-mechanistic rather than a fully mechanistic model. This may be explained by the existence of other sources of variability that were not taken into account in the final model. Possible sources could be the interindividual variability in CYP3A activity, incomplete description of the first pass metabolism such as the formation of methoxymorphinan from DM, the lack of modeling the metabolic fate of DO, the possible contribution of glucuronidation, and inter-study variability and lastly, the possibility that further sequence variations were present in addition to those covered in the genotype analysis. The interaction between the final model and the study design can be seen from the simulated predictive check plots. Generally inadequately fitted concentrations form only a negligible fraction of the whole data set.
**In conclusion,** the final population model adequately described DM pharmacokinetics in healthy Caucasian men. The results confirmed that individual CYP2D6 genotype and to a lesser degree urine pH contribute to the variability in pharmacokinetic profiles of DM. The estimated quantitative differences in the metabolic capacity attributable to copies of individual CYP2D6 genes including *1, *2 and *4t suggest that existing scoring systems to predict CYP2D6 activity from CYP2D6 genotype need to be refined. To which extent the observed activity differences between functional alleles for DM as a substrate are the same for other substrates remains to be assessed. (40)
Methods

Study design

Data derived from 36 healthy Caucasian male volunteers who participated in 3 cocktail interaction studies were used for this evaluation; n=15 from study A, n=10 from study B, and n=11 from study C.(41, 42) The data evaluated in this study were taken from the respective reference period where each subject received a single oral dose of 30 mg DM hydrobromide (one capsule of Hustenstiller-ratiopharm, Ratiopharm, Ulm, Germany). Subjects eligible for inclusion in the analysis were carriers of one, two or a combination of the following allelic variants: CYP2D6*1, *1xN, *2, *2xN, *3, *4, *4xN, *6, *41 or *41xN. Details of demographic data of the 36 subjects are shown in Table 1.

Blood samples were collected for quantification of DM and DO as follows: study A immediately before administration and 0.17, 0.5, 0.75, 1.0, 1.33, 2.0, 3.15, 4.25, 5.25, 6.0, 8.0, 10.0, 11.95, 14.0, 24.0 h after dosing; study B immediately before administration and 0.17, 0.33, 0.50, 1.0, 1.33, 1.67, 2.0, 2.5, 3.25, 3.95, 4.5, 5.25, 7.00, 10.00, 12.0, 16.0, 24.0 h after dosing; study C immediately before administration and 0.17, 0.33, 0.50, 0.75, 1.00, 1.33, 1.67, 2.0, 2.5, 3.25, 3.95, 4.13, 4.25, 4.5, 4.75, 5.0, 5.5, 6.5, 8.0, 10.0, 12.0 h after dosing.

To determine the urinary metabolic ratio of DM to DO, urine samples were collected periodically and their volumes and pH values measured of each collection period. Urine sampling schedules were: study A immediately before administration and 0-2, 2-4, 4-6, 6-8, 8-12, 12-16 und 16-24 h after dosing; study B immediately before administration and, 0-6 und 6-12 h after dosing; study C immediately before administration and, 0-2, 2-4, 4-6, 6-8, 8-12 h after dosing.

Quantification of DM and DO

Plasma and urine samples were analyzed according to a previously published validated LC-MS/MS method.(43) Urine samples were treated with β-glucuronidase for cleavage of the DO glucuronides before measurement. The lower limits of quantification were 0.103ng/mL for DM and 0.101ng/mL for DO. Precision ranged from 3.2 to 7.8% for DM, and from 4.7 to 9.2% for DO, while accuracy was 101.8–102.9%, and 97.4–99.5% for DM and DO, respectively.

Genotyping

Genotyping was performed essentially as described earlier in detail.(8) Briefly, a 6.6 kb long CYP2D6-specific PCR product was generated and subsequently used as a template to
determine the presence of allele-identifying SNPs. Gene duplications and the CYP2D6*5 gene deletion were also determined by long-range PCR (44)

*Population pharmacokinetic modelling*

All plasma and urine-derived data points for DM and DO were analysed simultaneously using the nonlinear mixed-effect software in NONMEM version VI (Globomax, LLC, Hanover, MD, USA). The analysis was performed using the first-order conditional estimation method with interactions (FOCE INTER). Models were specified in set of differential equations using the ADVAN6 subroutine and were parameterized for disposition kinetics using apparent clearance and volume of distribution terms.

The activity of the non-functional CYP2D6*4, *3 and *6, or null enzyme variants was set to zero as these alleles carry inactivating sequence variations (http://www.cypalleles.ki.se/cyp2d6.htm). Clearance of DM in individuals with CYP2D6*null/*null genotypes was assumed to reflect non-CYP2D6-mediated clearance. Hence, the study evaluated the activity of alleles containing CYP2D6*1, *2 and *41 gene copies, which have been recognized as true variants influencing DM metabolism activity in vivo.(17, 32) In addition to CYP2D6 polymorphisms, age, urine pH, and body weight were considered as covariates during model building. An allometric model was tested for weight, while exponential and combined models were tested to centre the age effect to the mean or the median age values.(45) Different relationships were tested to explore the effect of urine pH on renal clearances of DM and DO. Covariates were included in the model in a forward stepwise manner and backwards removed starting with the covariate that resulted in most reduction in the objective function value (OFV). This procedure was repeated until no significant drop was obtained in OFV.

As a starting point for the model building, a four-compartment open model was tested to describe the entire data set. Each compartment represents each site of measurement (i.e., two compartments each [plasma and urine] for DM data and two compartments for DO data). A first-order absorption process with and without a lag time was explored. Disposition processes for both substances were assumed to follow linear kinetics. For the purpose of the semimechanistic model, DM was assumed to be converted entirely to DO. This metabolic clearance was described as follows: \( CL_m = CL_b + CL_{CYP2D6} \) where \( CL_m \) is the total apparent clearance of DM describing the biotransformation of DM to DO, \( CL_b \) is the basic value for DM metabolic clearance that is not subject to CYP2D6 activity, \( CL_{CYP2D6} \) is the metabolic
clearance due to CYP2D6 activity. \( CL_{CYP2D6}=n_{1} \times CL_{1} + n_{2} \times CL_{2} + n_{41} \times CL_{41} \), where \( n_{x} \) and \( CL_{x} \) are the number of gene copies observed and metabolic clearance values attributable to the respective gene. This model was selected as a basic structural model for further model building activities.

An exponential interindividual variability model was included for all model parameters forward and backward step by step interchangeably. Four residual-error models were added, one for each data type (i.e., a separate residual-error model for plasma parent data, another for plasma metabolite data etc.). Each residual error model was built up using an additive model, a proportional model, or a combination of both.

Model derivation and justification was guided by (i) the difference in the objective function value (\( \Delta OFV \)) generated via NONMEM, a \( \Delta OFV \) of 3.84 (approximate \( \chi^2 \)-distribution) for an additional parameter was used to determine statistical significance \((P < 0.05)\) of the difference between any two models, (ii) visual inspection of goodness-of-fit plots, as well as (iii) physiological plausibility and statistical precision of model parameter estimates. The final model performance was also justified by simulating plasma and urine concentration time profiles for 1000 subjects having the observed \( CYP2D6 \) allele combinations and visualized using the Active Perl program (Active Perl, version 5.10.0; ActiveState, Vancouver, Canada).
Appendix

The differential equations specifying the model shown in Figure 1 for a generic individual are:

\[ \frac{dA_1}{dt} = -k_{12} A_1 \]

\[ C_z = k_{12} A_1 + Q_1 C_2 / (Q_1 + CL_{23}) \]

\[ \frac{dA_2}{dt} = Q_1 C_z - Q_1 C_2 - Q_2 C_2 + Q_2 C_6 - CL_{24} C_2 \]

\[ \frac{dA_3}{dt} = CL_{23} C_z - CL_{30} C_3 - CL_{35} C_3 \]

\[ \frac{dA_4}{dt} = CL_{24} C_2 \]

\[ \frac{dA_5}{dt} = CL_{24} C_2 \]

\[ \frac{dA_6}{dt} = Q_2 C_2 - Q_2 C_6 \]

The differential equation for the hypothetical enzyme compartment can be written as follows:

\[ \frac{dC_z}{dt} = k_{12} A_1 + Q_1 C_2 - Q_1 C_z - CL_{23} C_z \]

where \( C_z \) is the concentration in the metabolizing compartment. The model assumes rapid equilibrium between this compartment and the DM central compartment.

\[ 0 = k_{12} A_1 + Q_1 C_2 - (Q_1 + CL_{23}) C_z \]

This equation can be rearranged to get the following expression for \( C_z \)

\[ C_z = k_{12} A_1 + Q_1 C_2 / (Q_1 + CL_{23}) \]

The differential expression for DM in the central compartment (\( A_2 \)) is:

\[ \frac{dA_2}{dt} = Q_1 C_z - Q_1 C_2 - Q_2 C_2 + Q_2 C_6 - CL_{24} C_2 \]

After substituting \( C_z \)

\[ \frac{dA_2}{dt} = Q_1 (k_{12} A_1 + Q_1 C_2/( Q_1 + CL_{23})) - Q_1 C_2 - Q_2 C_2 + Q_2 C_6 - CL_{24} C_2 \]

Similarly for the DO plasma compartment:

\[ \frac{dA_3}{dt} = CL_{23} C_z - CL_{30} C_3 - CL_{35} C_3 \]

After substituting \( C_z \):

\[ \frac{dA_3}{dt} = CL_{23} ((k_{12} A_1 + Q_1 C_2)/( Q_1 + CL_{23})) - CL_{30} C_3 - CL_{35} C_3 \]

where \( A_i \) and \( C_i \) denote the amount and concentration of the species associated with the \( i \)th compartment in Figure 1.
**Figure Legends**

**Figure 1** Blueprint of the final semi-mechanistic population model of dextromethorphan (DM) and dextrorphan (DO) in plasma and urine after administration of single oral doses of 30 mg DM hydrobromide. The absorption phase is described with a rate constant (ka) and lag time (t_{lag}). C_p and C_u are plasma and urine concentration of DM or DO in the corresponding compartments. C_z is the concentration of DM in the hypothetical metabolizing enzyme compartment, which is in rapid equilibration with that in plasma (C_p). Q_1 is the intercompartmental clearance between the central compartment of DM and the metabolizing enzyme compartment. CL_{23} is the apparent systemic metabolic clearance of DM to DO, which is the sum of the basic (CL_b) and CYP2D6-mediated (CL_{CYP2D6}) components. Q_2 is the intercompartmental clearance between central and peripheral compartments of DM, CL_{24} is the renal clearance of DM under the influence of urine pH (UpH), CL_{35} represents DO renal clearance, and CL_{30} is the clearance of DO to other species.

**Figure 2** Goodness-of-fit plots for the final model; **a** Goodness-of-fit plots for plasma concentrations of dextromethorphan (left) and dextrorphan (right); **b** Goodness-of-fit plots for amounts excreted in urine of dextromethorphan (left) and dextrorphan (right).

**Figure 3** Visual predictive check plots showing adequacy of the developed model to describe the actual data (filled black circles=dextromethorphan data, unfilled circles=dextrorphan) as they are almost equally dispersed on the both side of simulated median (50th percentile). Plasma observations are shown on the upper panel.

**Figure 4** Simulated typical time courses in plasma (upper panel) and urine (lower panel) for individuals with different genotypes.
References


(13) Kirchheiner, J., Schmidt, H., Tzetkov, M., Keulen, J.T., Lotsch, J., Roots, I. & Brockmoller, J. Pharmacokinetics of codeine and its metabolite morphine in ultra-
rapid metabolizers due to CYP2D6 duplication. *Pharmacogenomics J* 7, 257-65 (2007).


(22) Costache, A.D., Trawick, D., Bohl, D. & Sem, D.S. AmineDB: large scale docking of amines with CYP2D6 and scoring for druglike properties--towards defining the scope of the chemical defense against foreign amines in humans. *Xenobiotica* 37, 221-45 (2007).


## Tables

**Table 1** Demographic data given as mean ± standard deviation (range).

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<th>Characteristics</th>
<th>Values</th>
<th>Study A</th>
<th>Study B</th>
<th>Study C</th>
<th>Total</th>
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<td>Number of subjects used in the evaluation</td>
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<td>Age (years)</td>
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<td>28.5 ± 4.70</td>
<td>29.6 ± 6.5</td>
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<td>(23 – 38)</td>
<td>(22 – 43)</td>
<td>(18 – 48)</td>
<td>(18 – 49)</td>
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<td>Body weight (kg)</td>
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<td>78.1 ± 10.1</td>
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<td>Height (cm)</td>
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<td>182.3 ± 7.0</td>
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### Table 3: Pharmacokinetic parameter estimates for the population

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<th>Model parameter§</th>
<th>Point estimate</th>
<th>95% CI</th>
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<td>Absorption rate constant, $k_a$ (h$^{-1}$)</td>
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<td>0.23 – 0.27</td>
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<td>Lag time (h)</td>
<td>0.31</td>
<td>0.30 – 0.32</td>
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<td>Apparent volume of distribution of DM in central compartment, $V_{c, cex}$ (l)</td>
<td>648</td>
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<td>Intercompartmental clearance $Q_1$ (l/h)</td>
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<td>295 – 825</td>
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<td>Apparent volume of distribution of DM in peripheral compartment $V_{p, cex}$ (l)</td>
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<td>1092 – 2028</td>
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<td>Intercompartmental clearance, $Q_2$ (l/h)</td>
<td>173</td>
<td>129 – 217</td>
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<td>Apparent renal clearance of DM, $CL_{24}$ (l/h)</td>
<td>6.52</td>
<td>4.84 – 8.20</td>
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<td>Urine pH effect on DM renal clearance in a man with a standard pH value of 5.7</td>
<td>3.93</td>
<td>2.74 – 5.12</td>
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<td>Components of apparent metabolic clearance $CL_{23}$</td>
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<td>Basic (l/h)</td>
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<td>6.38 – 13.82</td>
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<td>$CYP2D6*1$ (l/h, per gene copy)</td>
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<td>$CYP2D6*41$ (l/h, per gene copy)</td>
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<td>Apparent volume of distribution of DO $V_{m, dor}$ (l)</td>
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<td>Age effect on $V_{m, dor}$ (%change/ year)**</td>
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<td>Apparent renal clearance of DO $CL_{35}$ (l/h)</td>
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<td>Apparent clearance of DO to other species $CL_{30}$ (l/h)</td>
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<td>Age effect on $CL_{30}$ (%change/ year)**</td>
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<td>Between subjects variability (BSV)</td>
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<td>DO</td>
<td>Proportional error (%CV)</td>
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§see Fig. 1 for individual parameters; *= percentage of relative standard error

** centered around 27 years