Expression of Hepatic Drug-Metabolizing Cytochrome P450 Enzymes and Their Inter-Correlations: A Meta-Analysis

Brahim Achour, Jill Barber, Amin Rostami-Hodjegan

Manchester Pharmacy School, University of Manchester, Manchester, United Kingdom (B.A., J.B., A.R-H.), Simcyp limited, a Certara Company, Sheffield, United Kingdom (A.R-H.)
Running title: Meta-analysis of P450 abundances and correlations

Corresponding author: Professor A. Rostami-Hodjegan

Manchester Pharmacy School,
University of Manchester, Stopford Building
Oxford Road, Manchester, M13 9PT, UK
Tel: (+) 44 161 306 0634
Fax: (+) 44 161 275 8349
Email: amin.rostami@manchester.ac.uk

Pages: 40.
Tables: 2.
Figures: 3.
References: 77 references.
Text: 3512 words.
Abstract: 187 words.
Introduction: 483 words.
Discussion: 1379 words.

Abbreviations: P450, cytochrome P450; HLM, human liver microsomes; IVIVE, \textit{in vitro-in vivo} extrapolation; LC-MS, liquid chromatography in conjunction with mass spectrometry; MPPGL, microsomal protein per gram liver; PBPK, physiologically-based pharmacokinetics; ELISA, enzyme-linked immunosorbent assay.
Abstract

Cytochrome P450 is a family of enzymes that catalyze reactions involved in the metabolism of drugs and other xenobiotics. These enzymes are, therefore, important in pharmacological and toxicological studies and information on their abundances is of value in the process of scaling in vitro data to in vivo metabolic parameters. A meta-analysis was applied to data on abundance of human hepatic cytochrome P450 enzymes in Caucasian adult livers (50 studies). In spite of variation in methodologies used to measure the abundance of enzymes, agreement between the studies in 26 different laboratories was generally good. Nonetheless, some heterogeneity was detected (Higgins and Thompson heterogeneity test). More importantly, large inter-individual variability was observed in the collated data. Positive correlations between the expression levels of some cytochrome P450 enzymes were found in the abundance data including the pairs: CYP3A4/CYP3A5*1/*3 (Rs = 0.70, P < 0.0001, n = 52), CYP3A4/CYP2C8 (Rs = 0.68, P < 0.0001, n = 134), CYP3A4/CYP2C9 (Rs = 0.55, P < 0.0001, n = 71) and CYP2C8/CYP2C9 (Rs = 0.55, P < 0.0001, n = 99). These correlations can be used to demonstrate common genetic transcriptional mechanisms.
Introduction

Cytochrome P450 enzymes make up the main family of enzymes responsible for phase I metabolism reactions (Al Omari and Murry, 2007). The survey by Wienkers and Heath (2005) indicated the involvement of these enzymes in the metabolic clearance of nearly 75% of the 200 most prescribed therapeutic drugs in the US in 2002. A recent report (Zanger et al., 2014) described the contribution of each drug-metabolizing cytochrome P450 enzyme to drug metabolism pathways as well as the various genetic and non-genetic factors that affect variability in enzymatic activity, and therefore the clinical outcome of therapy.

Many studies attempted to quantify the abundance of cytochrome P450 enzymes and to describe genetic and epigenetic factors that affect the expression of these enzymes. As indicated by Proctor et al. (2004) abundance data of cytochrome P450 can be used for extrapolation from in vitro data to in vivo pharmacokinetic parameters, particularly when constructing the population variability is of interest. Hence, these data can be used for simulation experiments involving virtual populations for the pharmacological and toxicological assessment of drugs (Rostami-Hodjegan and Tucker, 2007). Correlations of expression between enzymes have been reported in the literature (Jover et al., 2009), and although in the current framework of Monte-Carlo simulations for creating virtual populations (Rostami-Hodjegan and Tucker, 2007) these correlations are not considered, such correlations, if established, can assist in producing more reliable scenarios.

There have been many efforts to quantify cytochrome P450 enzymes and some of these studies attempted to assess correlations of expression. These quantitative data were
obtained using a variety of methods (Western blotting, ELISA, or LC-MS) and using different numbers of subjects from different age groups and ethnic backgrounds.

Meta-analysis is a process of combining different studies in order to obtain an overall figure for a measured variable, detect the degree of heterogeneity between the individual studies, and to describe inconsistency and its possible sources (Sánchez-Meca and Martín-Martínez, 1997). With the growing number of studies that quantify pharmacokinetically relevant enzymes, there is a pressing need for a meta-analysis of these data; identifying weighted mean abundances of the enzymes, the degree of variability in these abundance data, and the degree of heterogeneity between these studies. The most recent meta-analysis of abundance data of drug metabolizing cytochrome P450 enzymes was published in 2004 and analyzed data from 19 studies. This was a comprehensive and influential analysis even though it did not distinguish between CYP3A isoforms (Rowland-Yeo et al., 2004) (see Supplemental Figure 1). Since 2004, a further 31 important studies of cytochrome P450 abundances have been published; for this reason, we now present a meta-analysis of the literature on cytochrome P450 abundance published up to the beginning of 2014 and provide correlations of expression where data are available. This will assist different groups that work with physiologically-based pharmacokinetic (PBPK) models to incorporate more representative values for the system parameters related to the expression levels of cytochrome P450 enzymes.
Materials and Methods

Collection of data

Two electronic databases, Medline (http://www.nlm.nih.gov/bsd/pmresources.html) and Web of Knowledge (http://wok.mimas.ac.uk/) (between the years 1980-2014), were searched for relevant literature on abundance of enzymes using appropriate keywords (hepatic / liver cytochrome P450 abundance, hepatic / liver CYP abundance, correlation of expression). Other keywords were also used, e.g. quantity / concentration / content, quantification / measurement, instead of abundance to widen the search scope. Resources cited by the collected papers were also inspected to locate further literature that could be used. A set of pre-defined criteria were employed for inclusion of identified studies. The selected studies were those that were conducted on individual liver microsomal samples (rather than pooled samples or cell lines), where absolute protein abundance was measured in Caucasian adults. In addition, studies quoting abundance data in arbitrary, relative, or non-standard units (other than pmol mg\(^{-1}\)) were excluded. Finally, studies that quantified mRNA levels or enzyme activities only were also excluded. The sources of data were identified to ensure that none of the data used were duplicated in the analysis. Where data were not quoted and graphs contained data points, GetData Graph Digitizer version 2.25 was used to obtain abundance values.

Calculation of weighted means and coefficients of variation

For each enzyme, abundance values were tested for heterogeneity and normality of distribution. The Higgins and Thompson heterogeneity test (Cochran, 1954; Higgins and Thompson, 2002; Higgins et al., 2003) determines whether the 50 studies are consistent
with one another, whereas the normality test determines whether the combined data are normally distributed. The normality test was performed according to the method of Kolmogorov and Smirnov.

The data from the individual studies were combined. The weighted means and the weighted coefficients of variation of the abundances of the different enzymes from the collected studies were calculated using Equations 1 and 2 (Armitage et al., 2001), respectively.

\[
W\bar{X} = \frac{\sum_{j=1}^{J} n_j \bar{X}_j}{\sum_{j=1}^{J} n_j} \tag{1}
\]

\[
WCV = \frac{\sum_{j=1}^{J} n_j CV_j}{\sum_{j=1}^{J} n_j} \tag{2}
\]

Where \( W\bar{X} \) and \( WCV \) represent the weighted mean and weighted coefficient of variation, respectively. Subscript \( j \) indicates the study, \( n_j \) the number of samples in study \( j \), and \( \bar{x}_j \) the mean abundance of a particular enzyme in study \( j \).

**Assessment of heterogeneity between studies**

To assess homogeneity between the means and coefficients of variation of individual studies and the overall mean and variability of the collated data, Equations 3, 4, and 5 were used.
\[
VarW\bar{X} = \frac{\sum_{j=1}^{J} w_j \bar{X}_j}{\sum_{j=1}^{J} w_j} \]

(3)

\[
w_j = \frac{1}{(sd_j)^2} \]

(4)

\[
Q = \sum_{j=1}^{J} \left( w_j (\bar{X}_j - VarW\bar{X})^2 \right) \]

(5)

Where \( w_j \) is the weight of study \( j \) expressed as the variance and calculated using Equation 4 (the inverse of standard deviation, \( sd_j \), squared), \( VarW\bar{X} \) is the variance in the weighted mean of the data from all the studies. \( Q \) is the coefficient of heterogeneity (Equation 5) of the collated data (Cochran’s \( Q \) test (Cochran, 1954)) expressed as the collective weighted squared differences between the mean of each study and the variance in the weighted mean. A higher value of \( Q \) indicates greater heterogeneity.

The degree of heterogeneity can be assessed using the \( I^2 \) index, Equation 6, proposed by Higgins and Thompson (2002). This provides a percentage of overall heterogeneity that can be interpreted as proposed by Higgins et al. (2003) as follows: around 0%, no heterogeneity; around 25%, low heterogeneity; around 50%, moderate heterogeneity; and around 75%, high heterogeneity.

\[
I^2 = 100 \times \frac{[Q - (k - 1)]}{Q} \]

(6)

Where \( I^2 \) is the index of heterogeneity, \( Q \) is Cochran’s heterogeneity coefficient, and \( (k - 1) \) is the number of degrees of freedom defined as the number of studies, \( k \), minus one. Where \( I^2 \) is negative, it is set to zero. The probability, \( P \), of the test can be quoted by
comparing the $Q$ value to a $\chi^2$ distribution with the same number of degrees of freedom (Higgins et al., 2003).

**Assessing correlations between the abundances of drug-metabolizing enzymes**

The Spearman rank test (Armitage et al., 2001) was used to assess correlations of the expression levels of the enzymes. Before correlation calculations, collated data were normalized using mean values of reference studies (Equation 7).

\[ x_{j,\text{normalized}} = \frac{x_j - \bar{x}_{\text{reference}}}{\bar{x}_j} \]

Where $x_{j,\text{normalized}}$ is the normalized abundance value of the abundance measurement $x$ in study $j$, $\bar{x}_j$ is the mean abundance value of study $j$, and $\bar{x}_{\text{reference}}$ is the mean abundance of the same enzyme in the reference study used for the normalization process. The reference abundance values used for normalization were those obtained in this meta-analysis (Table 1).

Calculations of the Spearman correlation coefficients, $R_s$, and probabilities, $P$, were carried using a $t$-Student distribution. A Bonferroni correction was used to define the $P$-value for statistical significance for the number of tests applied on each enzyme (a maximum of 14 tests for each enzyme).

**Assessment of gender and age related differences in expression**

Differences between expression levels in Caucasian adult liver cytochrome P450 enzymes in male and female subjects and in different age groups were investigated where data were
available. Gender-related differences were investigated using the Mann-Whitney rank order $U$-test and correlation of expression with age was examined using the Spearman rank correlation test.
Results

Studies used for the meta-analysis

The number of studies used in the meta-analysis was 50, out of 94 studies collected using the search strategy. There were several criteria employed to select the studies to be used for the meta-analysis. Studies were excluded where the samples were not individual liver microsomal samples (either recombinantly expressed systems, cell lines or pooled samples, 13 studies), where enzymatic activity or mRNA levels of expression were the only quantification data (19 studies), and where all the study subjects were from a non-Caucasian ethnic group (6 studies) or from a paediatric age group (4 studies). Additionally, studies quoting abundance data in arbitrary, relative, or non-standard units (other than pmol mg\(^{-1}\)) were also excluded (7 studies). Some studies were excluded for more than one of these reasons.

The data in the selected studies for the meta-analysis were obtained in 26 different laboratories worldwide, using two different types of experimental method for abundance measurement: immunoquantification (Western blotting or ELISA) and LC-MS. Most of the studies included in the meta-analysis used Western blotting (44 studies), whereas ELISA was used in 3 studies (Barter et al., 2010; Snawder and Lipscomb, 2000; Stresser and Kupfer, 1999) and LC-MS was used in 5 studies (Achour et al., 2014a; Langenfeld et al., 2009; Ohtsuki et al., 2012; Seibert et al., 2009; Wang et al., 2008). Langenfeld et al. (2009) and Wang et al. (2008) used both immunoquantification and LC-MS. The range of enzymes quantified and the number of livers used in the studies were different for each enzyme (Table 1).
For correlation analysis, additional studies conducted on samples from non-Caucasian (Kawakami et al., 2011) and paediatric populations (Hines, 2007; Koukouritaki et al., 2004) were also included. Studies expressing abundance in relative units (Wrighton et al., 1987; Wrighton et al., 1993) were also used.

**Meta-analysis of cytochrome P450 enzyme abundances**

The data from available literature were collated and tested for normality and heterogeneity. Data did not show normal distribution. The data from different studies did not show extensive heterogeneity except for the CYP3A43 dataset, which came from only two independent studies. Table 1 shows the results of the heterogeneity tests and the weighted means ($\bar{X}_W$) and weighted coefficients of variation ($WCV$). Figure 1 shows the results of the meta-analysis for the 15 drug-metabolizing cytochrome P450 enzymes and a pie chart of the distribution of expression of these enzymes in hepatic tissue. The bar graph in Figure 1 was divided into panels A and B to make it possible to clearly visualize the low abundance enzymes (CYP2C18, 2J2 and 3A43) using two different scales. Cytochrome P450 protein abundance pie charts from this meta-analysis and previous studies (Rowland-Yeo et al., 2004; Shimada et al., 1994) are shown in Supplemental Figure 1.

**Correlation analysis of the abundances of cytochrome P450 enzymes**

Where more than one enzyme was quantified in the same samples, the data were used for correlation analysis. This analysis included data from other ethnic backgrounds and from paediatric samples. Since the data were not normally distributed, the Spearman rank
correlation test was used after normalizing the collated data using mean values obtained in this meta-analysis (Table 1). Table 2 shows the correlation matrix obtained using the correlation analysis and Figure 2 shows examples of statistically significant weak and strong correlations.

**Gender and age related differences of expression**

There were no statistically significant differences in cytochrome P450 abundances between male and female Caucasian subjects (Mann-Whitney *U*-test, *P* > 0.05) for all datasets except for CYP3A4, which was found to be higher in female (n = 105) than male (n = 114) livers (Mann-Whitney *U*-test, *P* < 0.0001) (Figure 3). In addition, no correlation between hepatic levels of enzyme expression and age was found in Caucasian adults (Supplemental Figure 2) except in the case of CYP2C9 (Spearman correlation test, *P* < 0.05, n = 60).
Discussion

Abundances and activities of drug metabolizing enzymes play a major role in determining the fate of drugs and other xenobiotics in the body. In addition, pharmacokinetic evaluation of a new therapeutic entity, in the process of drug development, has become increasingly reliant on the use of in vitro systems that are dependent on scaling factors, which include abundance data of enzymes involved in metabolism pathways relevant to the particular drug (Barter et al., 2007; Rostami-Hodjegan and Tucker, 2007).

This study aimed to analyze the published literature on the abundance of cytochrome P450 enzymes in adult Caucasian livers. Using the defined criteria, 50 individual studies from 26 different laboratories were selected for analysis. The number of livers used to quantify each enzyme ranged from 23 to 713 livers, with the CYP3A4 dataset having the highest number of livers (n = 713) with the largest number of studies (k = 31). There was some heterogeneity between data from different studies especially in the case of CYP3A43; however, in the absence of information on inter-laboratory consistency of assays, it is difficult to assign these differences to heterogeneity between samples. Nevertheless, it was clear that, when Western blotting was the method of quantification, there were significant differences between abundance levels when different standards were used. Heterogeneity also tended to become less noticeable between datasets generated using the same quantitative method (immunoquantification or LC-MS). Datasets collected from a small number of studies tended to show higher levels of heterogeneity (especially in the case of CYP3A43), which should be interpreted with caution as the decreased numbers of degrees of freedom in these cases significantly
diminish the power of the heterogeneity test (Higgins and Thompson, 2002; Higgins et al., 2003). Other sources of heterogeneity may include different protocols of the same method (especially in the case of Western blotting) and low numbers of samples in some studies, which can cause the effect of outliers to be magnified.

The data obtained in this study can be used as scaling factors for in vitro - in vivo extrapolation (IVIVE) of measurements from in vitro recombinant enzyme systems and in simulations of drug trials and pharmacokinetic experiments (Table 1) (Barter et al., 2007; Rostami-Hodjegan and Tucker, 2007). The meta-analysis data revealed large inter-individual variation across cytochrome P450 enzymes (5-1300 fold difference), highlighting the importance of taking variability into account in the process of in vitro - in vivo extrapolation (Rostami-Hodjegan and Tucker, 2007). Another aspect that can be incorporated in the process of pharmacokinetic simulation and the building of virtual populations are the correlations of expression between pharmacokinetically relevant proteins (Table 2). The strongest correlation was seen between levels of CYP3A4 and CYP3A5*1/*3 genetic variant (Rs = 0.70, P < 0.0001, n = 52, 9 studies). This correlation was not strong when the CYP3A5 genotype was ignored (Rs = 0.06, P = 0.37, n = 218), a consistent observation with the literature (Achour et al., 2014a; Barter et al., 2010; Lin et al., 2002). Strong and statistically significant correlations between cytochrome P450 enzymes identified also include the pairs: CYP3A4/CYP2C8 (Rs = 0.68, P < 0.0001, n = 134), CYP3A4/CYP2C9 (Rs = 0.55, P < 0.0001, n = 71), CYP2C8/CYP2C9 (Rs = 0.55, P < 0.0001, n = 99), and CYP1A2/CYP2C9 (Rs = 0.56, P < 0.0001, n = 53). Although a relatively strong correlation between CYP2B6 and CYP3A4 expression levels
was reported in individual studies in the literature (Rs > 0.5) (Achour et al., 2014a; Mimura et al., 1993; Totah et al., 2008), this correlation was weaker upon analysis of all the collected data (Rs = 0.46, P < 0.0001, n = 124, 5 studies). Correlations of expression have also been reported at the mRNA level (Wortham et al., 2007), which further supports reports of common genetic regulation of the expression of cytochrome P450 enzymes (Jover et al., 2009).

Although some of the correlations demonstrated by this meta-analysis are shown to be strong, it is necessary to exercise caution in interpreting these relationships. All the correlations observed were positive, and while this would be expected (correlation arising from common regulatory receptors and pathways (Jover et al., 2009; Wortham et al., 2007)), there appears to be a background positive correlation between all the cytochrome P450 enzymes under study. The background correlation is a necessary artefact of measuring abundances in units of pmol mg⁻¹. A Bonferroni correction of the P-value made the test stricter and provided more confidence in the strong correlations observed.

Recently, it has been proposed that abundances would be more usefully measured in units of μg per gram of tissue in the case of uridine 5′-glucuronosyltransferase (UGT) enzymes (Milne et al., 2011), and the findings reported here may lend support to this suggestion.

Aging in adults (18 years onward) seemed to correlate with an overall apparent decline in expression of cytochrome P450 enzymes; however, correlation between age and levels of cytochrome P450 was not statistically significant (Spearman rank test, P > 0.05, Supplemental Figure 2) for all the datasets except one (CYP2C9). In adult subjects, age
was previously reported to have minimal effect on the abundance and activity of cytochrome P450 enzymes per mass unit of liver (Galetin et al., 2004; King et al., 2003; Wolbold et al., 2003). However, analysis of reported metabolic ratios (which were constant with age) indicated a decrease in the liver metabolic activity mediated by cytochrome P450 enzymes that is paralleled by the decline in renal function with age (Rostami-Hodjegan et al., 1999). Two contributing factors to these seemingly contradictory observations (i.e. no change in abundance per unit mass with age and reduced total metabolic activity) can be liver shrinkage that occurs when body size decreases with age (Johnson et al., 2005) and the decrease in the amount of total microsomal protein per gram liver (MPPGL) (Barter et al., 2008). Therefore, although the abundances of expressed cytochrome P450 enzymes seem not to be affected by aging to a significant extent, the literature suggests that overall activity may be affected. The results of a recent study by Polasek et al. (2013) are also supportive of this view although the area requires further investigation.

Gender differences in expression of cytochrome P450 were not shown to be statistically significant except in the case of CYP3A4 in line with literature on CYP3A4 abundance (Wolbold et al., 2003). This is supported by reports of more efficient clearance of CYP3A substrates, such as nifedipine (Krecic-Shepard et al., 2000), verapamil (Wolbold et al., 2003) and cyclosporine (Kahan et al., 1986), in female subjects. CYP3A4 is highly inducible, and only one study (Wolbold et al., 2003) has distinguished between inducer-exposed and control subjects. In the controls, CYP3A4 levels were found to be on average 2.6 times higher in female than male subjects (n = 39), whereas in
inducer-exposed patients the levels were much closer (1.7 fold difference in the mean, n = 55). The headline numbers would suggest that gender might influence clinical practice in prescribing drugs metabolized by CYP3A4; the reality, however, is that there is overlap between males and females in CYP3A4 expression, and that exposure to inducers is much more significant than gender.

It is interesting to compare the present work with the single study (Shimada et al., 1994) published in 1994 and with the first meta-analysis (Rowland-Yeo et al., 2004) published in 2004 (see Supplemental Figure 1). When restricted to the enzymes detectable in 1994 (CYP1A2, 2A6, 2B6, 2C, 2D6, 2E1, 3A) the pie charts are very similar, with the most striking difference being the growth of the relative abundance of CYP2B6 at the expense of CYP3A. We attribute this difference to the cross-reactivity of antibodies (early antibodies to CYP3A4 were rather non-specific). The major difference between this work and earlier work, however, is the sheer number of individual enzymes detected; 15 individual enzymes have now been quantified, allowing a much more detailed human liver pie to be created and complementing our recent UGT liver pie (Achour et al., 2014b).

We look forward to extending this work to describing expression profiles of cytochrome P450 in different ethnic groups and in paediatric samples in addition to investigating correlations of expression between different proteins (enzymes and transporters) that govern the processes of metabolism, disposition and elimination of drugs in different organs including the liver, the intestine and the kidney.
Acknowledgments

The authors thank Eleanor Savill for assistance in preparing the manuscript.

Authorship Contribution

Participated in research design: Achour, Barber, and Rostami-Hodjegan.

Performed data analysis: Achour.

Wrote or contributed to the writing of the manuscript: Achour, Barber, and Rostami-Hodjegan.
References


2405-2409.


Kawakami H, Ohtsuki S, Kamiie J, Suzuki T, Abe T, and Terasaki T (2011) Simultaneous quantification of 11 cytochrome P450 isoforms in human liver microsomes by liquid


Carlo comparison of statistical power and Type I error. *Qual Quant* **31**: 385-399.


Figure legends

**Figure 1.** Bar graph (A, B) and pie chart (C) of weighted mean abundances of cytochrome P450 enzymes in adult Caucasian livers. Error bars represent weighted standard deviation values. n, the number of livers.

**Figure 2.** Plots of examples of significant correlations of hepatic cytochrome P450 expression levels. Correlations between CYP2B6 and CYP3A4 (A) and between CYP2C9 and CYP2C19 (E) are weak, whereas correlations between CYP2C8 and CYP2C9 (B), CYP2C8 and CYP3A4 (C), and CYP2C9 and CYP3A4 (D) are strong. The strongest correlation was found between CYP3A4 and CYP3A5*1/*3 (F). Plots show normalized collated data.

**Figure 3.** Expression of cytochrome P450 in male and female Caucasian subjects. No sex differences were detected ($U$-test, $P > 0.05$) in all cases, except between levels of CYP3A4 in male and female livers ($U$-test, $P < 0.0001$). F, female; M, male; n, number of subjects in each set.
**Table 1.** The weighted means, coefficients of variation (CV), ranges and heterogeneity analysis of the analyzed hepatic cytochrome P450 enzyme abundance data from 50 studies. $Q$, Cochran’s heterogeneity coefficient; $I^2$, Higgins and Thompson’s heterogeneity index. * Ranges are included where available.

<table>
<thead>
<tr>
<th>Enzyme</th>
<th>Mean (pmol mg$^{-1}$)</th>
<th>CV (%)</th>
<th>Range* (pmol mg$^{-1}$)</th>
<th>no. livers</th>
<th>$Q$</th>
<th>$I^2$ (%)</th>
<th>Heterogeneity</th>
<th>Studies</th>
</tr>
</thead>
<tbody>
<tr>
<td>CYP1A2</td>
<td>39</td>
<td>78</td>
<td>1 – 263</td>
<td>148</td>
<td>19.54</td>
<td>54</td>
<td>Medium</td>
<td>[1-10]</td>
</tr>
<tr>
<td>CYP2A6</td>
<td>27</td>
<td>86</td>
<td>0 – 191</td>
<td>180</td>
<td>2.44</td>
<td>0</td>
<td>None</td>
<td>[3, 6, 7, 9, 11]</td>
</tr>
<tr>
<td>CYP2B6</td>
<td>16</td>
<td>125</td>
<td>0 – 180</td>
<td>504</td>
<td>14.55</td>
<td>0</td>
<td>None</td>
<td>[3, 6, 7, 9,10,12-23]</td>
</tr>
<tr>
<td>CYP2C8</td>
<td>22.4</td>
<td>68</td>
<td>0 – 85</td>
<td>144</td>
<td>3.72</td>
<td>0</td>
<td>None</td>
<td>[6, 7, 9, 24-28]</td>
</tr>
<tr>
<td>CYP2C9</td>
<td>61</td>
<td>54</td>
<td>0 – 277</td>
<td>120</td>
<td>14.36</td>
<td>30</td>
<td>Low</td>
<td>[1, 4, 6, 7, 9, 25-27, 29-31]</td>
</tr>
<tr>
<td>CYP2C18</td>
<td>0.4</td>
<td>39</td>
<td>0.2 – 0.7</td>
<td>23</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>[9]</td>
</tr>
<tr>
<td>CYP2C19</td>
<td>11</td>
<td>82</td>
<td>0 – 67</td>
<td>76</td>
<td>9.81</td>
<td>18</td>
<td>Low</td>
<td>[4, 6, 10, 25-27, 29, 30, 32]</td>
</tr>
<tr>
<td>CYP2D6</td>
<td>12.6</td>
<td>74</td>
<td>0 – 75</td>
<td>206</td>
<td>2.55</td>
<td>0</td>
<td>None</td>
<td>[1-4, 6, 7, 9, 10,33, 34]</td>
</tr>
<tr>
<td>CYP2E1</td>
<td>64.5</td>
<td>53</td>
<td>2 – 201</td>
<td>145</td>
<td>9.61</td>
<td>37</td>
<td>Low</td>
<td>[1, 3, 5-7, 8, 20]</td>
</tr>
<tr>
<td>CYP2J2</td>
<td>1.2</td>
<td>58</td>
<td>0 – 3</td>
<td>23</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>[9]</td>
</tr>
<tr>
<td>CYP3A4</td>
<td>93</td>
<td>81</td>
<td>0 – 601</td>
<td>713</td>
<td>46.78</td>
<td>36</td>
<td>Low</td>
<td>[2-4, 6, 7, 8, 9, 12, 13, 23, 24, 28, 30-32, 35-47, 48-50]</td>
</tr>
<tr>
<td>CYP3A5</td>
<td>17</td>
<td>185</td>
<td>0 – 291</td>
<td>250</td>
<td>26.16</td>
<td>46</td>
<td>Medium</td>
<td>[6, 9, 35-44, 47, 49, 50]</td>
</tr>
<tr>
<td>CYP3A7</td>
<td>9</td>
<td>154</td>
<td>0 – 90</td>
<td>54</td>
<td>0.68</td>
<td>0</td>
<td>None</td>
<td>[6, 9, 35, 37]</td>
</tr>
<tr>
<td>CYP3A43</td>
<td>2</td>
<td>35</td>
<td>0 – 6</td>
<td>35</td>
<td>72.96</td>
<td>99</td>
<td>High</td>
<td>[6, 9]</td>
</tr>
<tr>
<td>CYP4F2</td>
<td>11.4</td>
<td>45</td>
<td>1 – 24</td>
<td>23</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>[9]</td>
</tr>
</tbody>
</table>

**Studies:**

Table 2. Correlations matrix of the abundance data of 15 cytochrome P450 enzymes from 32 studies. Spearman correlation (Rs) analysis was used. Significant strong correlations are in bold font and significant very strong correlations are shown in italic bold font. Non-significant correlations are not included (-). Rs, the Spearman correlation coefficient; P, probability value; n, number of subjects.

<table>
<thead>
<tr>
<th>Enzyme 1</th>
<th>Enzyme 2</th>
<th>Rs</th>
<th>P</th>
<th>n</th>
<th>Rs</th>
<th>P</th>
<th>n</th>
<th>Rs</th>
<th>P</th>
<th>n</th>
<th>Rs</th>
<th>P</th>
<th>n</th>
</tr>
</thead>
<tbody>
<tr>
<td>CYP1A2</td>
<td>CYP1A2</td>
<td>0.50</td>
<td>0.0002</td>
<td>51</td>
<td>0.34</td>
<td>0.003</td>
<td>72</td>
<td>0.56</td>
<td>&lt;0.0001</td>
<td>53</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>CYP2A6</td>
<td>CYP2A6</td>
<td>0.47</td>
<td>0.005</td>
<td>51</td>
<td>0.64</td>
<td>&lt;0.0001</td>
<td>51</td>
<td>0.52</td>
<td>&lt;0.0001</td>
<td>50</td>
<td>0.46</td>
<td>&lt;0.0001</td>
<td>124</td>
</tr>
<tr>
<td>CYP2B6</td>
<td>CYP2B6</td>
<td>0.40</td>
<td>0.0036</td>
<td>51</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>CYP2C8</td>
<td>CYP2C8</td>
<td>0.55</td>
<td>&lt;0.0001</td>
<td>99</td>
<td>0.37</td>
<td>0.001</td>
<td>61</td>
<td>0.61</td>
<td>&lt;0.0001</td>
<td>51</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>CYP2C9</td>
<td>CYP2C9</td>
<td>0.50</td>
<td>&lt;0.0001</td>
<td>23</td>
<td>0.40</td>
<td>&lt;0.0001</td>
<td>419</td>
<td>0.52</td>
<td>&lt;0.0001</td>
<td>71</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>CYP2C18</td>
<td>CYP2C18</td>
<td>0.53</td>
<td>0.005</td>
<td>23</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>CYP2C19</td>
<td>CYP2C19</td>
<td>0.53</td>
<td>0.005</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

This article has not been copyedited and formatted. The final version may differ from this version.
<table>
<thead>
<tr>
<th></th>
<th>CYP2D6</th>
<th>CYP2E1</th>
<th>CYP2J2</th>
<th>CYP3A4</th>
<th>CYP3A5</th>
<th>CYP3A7</th>
<th>CYP3A43</th>
<th>CYP4F2</th>
<th>n = 27</th>
<th>Rs = 0.51</th>
<th>P = 0.002</th>
<th>n = 36</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1</td>
<td></td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td>-</td>
<td></td>
<td>Rs = 0.70*</td>
<td>P &lt; 0.0001*</td>
<td>n = 52*</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1</td>
<td></td>
<td></td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1</td>
<td></td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

*Correlation between CYP3A4 and CYP3A5*1/*3 abundance data from 9 studies.
Figure 2

A

Rs = 0.46
P < 0.0001
n = 124

CYP3A4 vs CYP2B6

B

Rs = 0.55
P < 0.0001
n = 99

CYP2C9 vs CYP2C8

C

Rs = 0.68
P < 0.0001
n = 134

CYP3A4 vs CYP2C8

D

Rs = 0.55
P < 0.0001
n = 71

CYP3A4 vs CYP2C9

E

Rs = 0.40
P < 0.0001
n = 419

CYP2C19 vs CYP2C9

F

Rs = 0.70
P < 0.0001
n = 52

CYP3A5 vs CYP3A4
Drug Metabolism and Disposition

Expression of Hepatic Drug-Metabolizing Cytochrome P450 Enzymes and Their Inter-Correlations: A Meta-Analysis

Brahim Achour, Jill Barber, Amin Rostami-Hodjegan
**Supplemental Figure 1.** (A) Comparison between cytochrome P450 distribution pie charts from Shimada et al. (1994), Rowland-Yeo et al. (2004) and the present study (2014). (B) Detailed abundance distribution of hepatic cytochrome P450 enzymes from the present study.
Supplemental Figure 2. Example plots of trend analysis of the effect of age on the expression of cytochrome P450 2A6 (A), 2C9 (B), 2C19 (C), 2D6 (D), 3A4 (E) and 2E1 (F). Very weak correlations with no statistical significance are identified ($P > 0.05$) except in the case of CYP2C9 (B), which showed weak but statistically significant correlation.