Population Pharmacokinetic-Pharmacodynamic Model of the Vascular-Disrupting Agent 5,6-Dimethylxanthenone-4-Acetic Acid in Cancer Patients

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Abstract

Purpose: To develop a population pharmacokinetic-pharmacodynamic (PK-PD) model that defines the dose-concentration-effect relationship of 5,6-dimethylxanthenone-4-acetic acid (DMXAA), using plasma 5-hydroxyindole-3-acetic acid (5-HIAA) as a biomarker for the antivascular effect of DMXAA.

Experimental Design: The plasma DMXAA and 5-HIAA concentration data were obtained from 124 patients receiving DMXAA monotherapy as a 20-minute i.v. infusion weekly or every 3 weeks at doses of 6 to 4,900 mg/m². The PK and PD data were analyzed by nonlinear mixed effects modeling with NONMEM version 5.

Results: DMXAA concentration-time profiles were well described by a three-compartment model with saturable elimination (Michaelis-Menten kinetics). Body surface area (BSA) and sex were significant covariates on the volume of distribution of the central compartment ($V_c$) and the maximum elimination rate ($V_m$), respectively. Population estimates for $V_m$, $K_m$ (concentration at which half $V_m$ is achieved), and $V_t$, were 112[1 + 0.474(2 - sex)] μmol/L/h, 102 μmol/L, and 8.19(BSA/1.8)$^{0.857}$ liters, respectively (sex in $V_m$ is equal to 1 for males and equal to 2 for females). The effect of DMXAA on plasma 5-HIAA was described by the stimulatory $E_{max}$ model, where population estimates for baseline, $E_{max}$, and $EC_{50}$ were 46.3 μmol/L, 2.62-fold increase of the baseline value, and 631 μmol/L, respectively.

Conclusions: DMXAA plasma disposition is characterized by a saturable elimination process. BSA-guided dosing is important. The present PK-PD model, with 5-HIAA as a biomarker, supports the use of DMXAA doses of 1,000 to 2,000 mg/m² in phase II studies, and provides an example of how PK-PD models can be used to aid in selection of drug doses for phase II evaluation.

5,6-Dimethylxanthenone-4-acetic acid (DMXAA) is a small-molecule vascular-disrupting agent that selectively disrupts established tumor blood vessels. It has been evaluated as monotherapy in phase I trials (1–3) and is being assessed in phase II trials in combination with cytotoxic agents. Recent data from a phase II study have shown that the combination of DMXAA with carboplatin and paclitaxel increased survival by 5.2 months in patients with non–small cell lung cancer compared with treatment with carboplatin and paclitaxel alone (median survival 14 versus 8.8 months; ref. 4).

DMXAA induces rapid vascular collapse and subsequent tumor hemorrhagic necrosis through induction of apoptosis in tumor vascular endothelial cells (5) and indirect vascular effects involving various cytokines, in particular, tumor necrosis factor-α, serotonin, and nitric oxide (6–10). DMXAA has also shown an antiangiogenesis effect in vivo (11). The degree of DMXAA-induced tumor hemorrhagic necrosis is significantly correlated with increased tumor vascular permeability, decreased functioning tumor blood vessels, and increased plasma 5-hydroxyindoleacetic acid (5-HIAA) concentrations (as a measure of serotonin release; ref. 7). Serotonin, a naturally occurring vasoactive substance that is released from platelets into plasma under various pathologic conditions, is involved in the antitumor and host effects of DMXAA and also mediates the effects of tumor necrosis factor-α (9). Serotonin plasma concentrations increase in response to DMXAA in mice (8), suggesting that it might be a surrogate marker for antivascular effects. However, serotonin is unstable during blood processing and storage, whereas its primary metabolite, 5-HIAA, has greater stability and accumulates in plasma following serotonin release (12). Phase I studies have shown that 5-HIAA plasma concentrations increased in a dose-dependent manner at DMXAA doses of $\geq$650 mg/m² (2, 3). Therefore, 5-HIAA, providing an index of vascular damage, could be a suitable surrogate biomarker for the antivascular and/or antitumor effects of DMXAA.

One challenge in the clinical development of molecularly targeted therapies is to identify an optimal biological and...
efficacious dose that may be below the maximum tolerated dose. Defining dose-concentration-effect relationships could aid in this process. Population pharmacokinetic-pharmacodynamic (PK-PD) analysis using nonlinear mixed effects modeling is particularly useful to characterize the dose-concentration-effect relationship for new chemical entities or established drugs because it allows simultaneous analysis of PK (concentration) and PD (effect) data from all complete or incomplete individual data sets as well as identification of sources of interindividual PK-PD variability.

The purpose of this study was to develop a population PK-PD model to (a) characterize DMXAA plasma concentration-time profiles following single- and multiple-dose administration in cancer patients treated with single-agent DMXAA, (b) relate the plasma concentration-time profiles to changes in 5-HIAA concentrations as a biomarker for antivascular effect, and (c) identify clinical covariates that influence DMXAA pharmacokinetics and pharmacodynamics. In addition, simulations were done to show how the model can be used to predict the antivascular effect and guide dose selection for future studies of DMXAA.

Materials and Methods

Clinical studies and patients. The clinical and pharmacokinetic data used in the analysis were obtained from three phase I studies involving a total of 124 cancer patients treated with single-agent DMXAA as previously described (1–3). Patients with histologically confirmed solid tumors that were not amenable to any standard therapy or were refractory to conventional therapy were eligible for participation in the trials. Patient demographic and baseline clinical characteristics including sex, age, weight, height, body surface area (BSA), WHO performance status, liver function (i.e., total bilirubin, alanine aminotransferase, and alkaline phosphatase), kidney function (i.e., serum creatinine), serum albumin, and hemoglobin are listed in Table 1. There were no clinically significant differences in patient demographic and clinical characteristics across the three studies. The study protocols were approved by the medical ethics committee in all study centers, and written informed consent was obtained from all the patients.

Two parallel phase I dose escalation studies composed of (a) a two-center study in the United Kingdom with patients receiving DMXAA once weekly and (b) a single-center study in New Zealand with patients receiving DMXAA once every 3 weeks. The drug was given as 20-minute i.v. infusion in both studies. The starting dose was 6 mg/m², and at least three patients were treated at each dose level. In the second study, 19 dose levels (i.e., 6, 10.2, 20.4, 40.8, 81.6, 160, 240, 320, 500, 650, 850, 1,100, 1,375, 1,650, 2,000, 2,600, 3,100, 3,700, and 4,900 mg/m²) were evaluated in 46 patients. One course of treatment consisted of 6 weekly infusions, and a maximum of two courses (12 infusions) was permitted. In the New Zealand trial, 19 dose levels (i.e., 6, 10.2, 20.4, 40.8, 81.6, 160, 240, 320, 500, 650, 850, 1,100, 1,375, 1,650, 2,000, 2,600, 3,100, 3,700, and 4,900 mg/m²) were evaluated in 46 patients. One course of treatment consisted of one infusion of DMXAA every 3 weeks, with a maximum of six courses (six infusions) in patients with stable disease, but where a tumor response occurred, a patient could continue for two courses following maximum response. The third trial, the DART study, was a phase I multicenter double-blind randomized six-way intrapatient dose-ranging crossover safety study to further characterize the effect of DMXAA on QTc, interval, ophthalmic safety, and pharmacodynamic

### Table 1. Patient demographic and baseline characteristics

<table>
<thead>
<tr>
<th></th>
<th>United Kingdom study</th>
<th>New Zealand study</th>
<th>DART study</th>
<th>All studies</th>
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<td>21/42</td>
<td>12/3</td>
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<td>73.1 (39.7-131.5)</td>
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<td>70 (40-131)</td>
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<td>1.65 (1.48-1.86)</td>
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<td>BSA (m²)</td>
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<td>1.85 (1.40-2.30)</td>
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<tr>
<td>WHO performance status †</td>
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<td></td>
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<td>0</td>
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<td>Total bilirubin (µmol/L)</td>
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<td>22 (6-322)</td>
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<td>AST (µmol/L)</td>
<td>31 (16-143)</td>
<td>26 (5-295)</td>
<td>26 (12-75)</td>
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<td>107 (13-948)</td>
<td>100 (37-1125)</td>
<td>107 (13-1191)</td>
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<td>119 (95-150)</td>
<td>114 (88-151)</td>
<td>124 (94-151)</td>
<td>116 (88-151)</td>
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</table>

NOTE: Data are presented as median (range).
Abbreviations: ALT, alanine aminotransferase; AST, aspartase aminotransferase; Alk phos, alkaline phosphatase.
*(Values are patient numbers).
† Values are shown as patient number (probability).
effects on tumor blood flow that was assessed by using dynamic contrast enhanced magnetic resonance imaging. In this study, 15 patients were randomly allocated to a treatment sequence according to multiple six-by-six Latin squares and were blinded to receive each of six doses (i.e., 300, 600, 1,200, 1,800, 2,400, and 3,000 mg/m²) at weekly intervals. Patients who derived clinical benefit from the course of six infusions were offered additional courses of weekly infusion for 6 weeks at 1,200 mg/m² and allowed to remain on treatment until they no longer had clinical benefit or disease progression, or until toxicity led to patient withdrawal.

PK and PD data. A total of 3,050 DMXAA plasma concentration-time data were obtained from 124 patients treated with single-agent DMXAA in the three studies. In the United Kingdom study, blood samples were collected before the infusion; immediately at the end of infusion; and at 0.25, 0.5, 0.75, 1, 1.5, 2, 4, 8, 12, and 24 h after the first infusion on week 1. Additional blood samples were drawn before the infusion, immediately at the end of infusion, and 4 h after the infusion in the subsequent treatment on week 2. In the New Zealand study, blood samples were collected before the infusion; 10 min before the end of infusion; immediately at the end of infusion; and at 0.25, 0.5, 0.75, 1, 1.5, 2, 4, 8, 12, and 24 h after the first infusion on week 1 (and, in one patient at each dose level, repeated with the second infusion on week 4), and additional samples were obtained before the infusion, at the end of infusion, and 4 h after the infusion in the subsequent treatment. In the DART study, blood samples were taken before the infusion; immediately at the end of infusion; and at 1, 2, 4, 8, 12, and 24 h after each infusion.

All blood samples were centrifuged at 1,300 × g for 10 min (for the United Kingdom and New Zealand studies) or at 1,200 × g for 5 min (for the DART study), and aliquots of plasma were stored at -70°C (United Kingdom and DART studies) or -20°C (New Zealand study) until analysis. DMXAA plasma concentrations were determined using validated high-performance liquid chromatography methods with fluorescence detection (for United Kingdom and New Zealand studies; ref. 2) or with tandem mass spectrometric detection (liquid chromatography/tandem mass spectrometry, for the DART study; ref. 3).

Blood samples for 5-HIAA plasma concentrations were taken at the same time as above for DMXAA concentrations, and 5-HIAA concentrations were determined by high-performance liquid chromatography with electrochemical detection (12).

Population PK analysis. DMXAA and 5-HIAA concentration-time data were analyzed using the nonlinear mixed effects modeling approach as implemented in NONMEM (version 5; University of California, San Francisco, CA). All analyses were done with the first-order approximation method. Xpose 3.1/S-PLUS 6.0 was used for graphical diagnostics and covariate screen. The population PK model of DMXAA was developed in two steps: (a) basic (structural) model development and (b) covariate model development.

Fig. 1. Observed DMXAA plasma concentration-time profiles following single-dose administration in the (A) United Kingdom, (B) New Zealand, and (C) DART studies. D, observed plasma 5-HIAA concentration-time profiles following each i.v. infusion of DMXAA in individual patients from all three studies.
The structural model was built to fit multiple-dose DMXAA plasma concentration-time profiles from all 124 patients simultaneously. DMXAA concentration data were log transformed. Three-compartment models with first-order elimination or nonlinear elimination characterized by Michaelis-Menten kinetics (Eq. A) were tested to fit DMXAA plasma concentration data.

\[
R = \frac{V_m \times C}{K_m + C}
\]  

where \( R \) is the elimination rate, \( V_m \) is the maximum elimination rate, \( K_m \) is the concentration at which half-maximum elimination rate is achieved, and \( C \) is DMXAA plasma concentration.

Mean population PK variables, interindividual variability, and residual error (intraindividual variability) were assessed in the model. Interindividual variability for each PK variable was modeled with an exponential function. Residual error was modeled with a combination method, including an additive and a proportional part, each of which could be excluded if it was estimated to be negligible. Individual PK variables were obtained by posterior Bayesian estimation.

Model selection for nonhierarchical models (i.e., linear and nonlinear elimination models) was guided by graphical goodness-of-fit plots (e.g., observed versus predicted plasma concentrations, residuals versus predicted concentrations, and weighted residuals versus predicted concentration). Akaike information criterion (AIC), and precision of estimates. AIC was calculated as \( AIC = (-2LL) + 2 \times p \), where -2LL is the NONMEM objective function value (OFV, -2 log likelihood) and \( p \) is the number of PK variables. The model was chosen on the basis of smaller values of AIC, better precision of estimates, and superior goodness-of-fit plots.

The covariate model building was a stepwise process. A screen for potential significant covariates was done using S-PLUS 6.2/Xpose 3.1 software (Uppsala University, Uppsala, Sweden) with generalized additive model (GAM; ref. 13). The potential covariates, as listed in Table 1, as well as the treatment cycle and bioanalysis assay were tested for influence on the structural PK variables. When a covariate value was missing in a patient, this covariate value was replaced by the population median value. The frequency of missing covariate was <10%. The potential significant covariates selected from GAM analysis were introduced into the covariate model as linear, exponential, or power function, and assessed in the PPK models. Discrimination between hierarchical models was made by comparison of the NONMEM OFV and by visual inspection of the goodness-of-fit plots. The difference between the OFV for two hierarchical models is approximately \( \chi^2 \) distributed, where the degrees of freedom are based on the difference between the numbers of estimated variables. A significant covariate was selected to be retained in the final model if addition of the covariate resulted in a decrease in OFV >3.875 (\( P < 0.05 \)) during the forward full covariate model building, and removal of the covariate resulted in an increase in OFV >10.828 (\( P < 0.001 \)) during the stepwise backward model reduction. In addition, the increase in precision of the variable estimate (% relative SE of prediction) and reduction in interindividual variability were used as another indicator of the improvement of the goodness of fit.

Fig. 2. Basic goodness-of-fit plots from the final PPK three-compartment model with Michaelis-Menten elimination and with sex and BSA being introduced as a significant covariate on \( V_m \) and \( V_1 \), respectively. Drug input is actual dose (mg/h). DV, observed concentrations (log transformed); PRED, population predicted concentrations (log transformed); IPRED, individual predicted concentrations (log transformed); IWRES, individual weighted residues; WRES, weighted residues.
Population PK-PD analysis. Plasma 5-HIAA was used as a biomarker for the antivascular effect of DMXAA. Because the increase in 5-HIAA plasma concentration in each patient lagged behind the plasma concentration time course of DMXAA, an effect compartment model, which assumes the rate of drug distribution to and from the hypothetical effect site will determine the rate of onset of effect, was implemented to relate DMXAA concentrations to the change of plasma 5-HIAA (Scheme 1; refs. 14, 15). In such a model, the time course of the effect site concentration \(C_e\) is described by the rate constant \(k_{01}\) from the effect compartment; following the dose of the drug, there is a slow absorption of the drug response that is governed by the inhibition or stimulation of factors controlling this response. This can produce either a decrease or increase in the observed response variable (such as a biomarker) depending on whether the input or disposition process is inhibited or stimulated (14). The vascular-disrupting effect of DMXAA resulted in an increase of 5-HIAA, and therefore, a stimulatory effect compartment model was chosen and different functional forms (i.e., linear, \(E_{\text{max}}\) and sigmoidal \(E_{\text{max}}\)) were tested. A stimulatory \(E_{\text{max}}\) model related to baseline (Eq. B) was found to best describe the relationship between the effect site concentration of DMXAA and 5-HIAA.

\[
E = E_{01}(1 + \frac{E_{\text{max}} 	imes C_e}{E_{\text{50}} + C_e})
\]

where \(E\) is the observed effect at the corresponding effect-site concentration \((C_e)\); and \(E_{01}, E_{\text{max}}\) and \(E_{\text{50}}\) represent baseline, maximum effect (times of baseline), and concentration at half-maximum effect, respectively.

DMXAA and 5-HIAA concentration-time data were fitted with the model simultaneously using the first-order approximation method as implemented in NONMEM, with population PK variables fixed to the previously obtained values from the final PK model. Interindividual variability for each PK and PD variable was modeled as an exponential function. An additive error model was used because DMXAA concentration and 5-HIAA concentration data were log transformed in the data file, enabling an additive error model, which approximates to a proportional error model on untransformed data. The covariate PD model building was done as described above.

Simulations. To illustrate the use of the model for dose selection, the concentration-time courses of plasma DMXAA and 5-HIAA, as predicted by the final PK-PD model, were simulated under different dose levels ranging from 50 to 4,400 mg/m² in a typical subject. This is a hypothetical individual who had PK and PD values equal to population median values and covariate values equal to the median values of the studied group, unless otherwise stated.

Results

Structural PK model. Figure 1 shows the observed individual DMXAA plasma concentration-time profiles following single- and multiple-dose administration in all patients enrolled to the three trials. DMXAA plasma concentration-time profiles following a 20-minute i.v. infusion were characterized by a triexponential decline by visual inspection (Fig. 1). Therefore, a three-compartment model with a first-order elimination process from the central compartment was first tested to fit DMXAA plasma concentration data. However, the linear-elimination model generally overpredicted the observed concentrations at lower drug dose levels (e.g., dose <100 mg/m²), whereas it underpredicted the observed concentrations at higher dose level (e.g., dose >1,750 mg/m²), as shown in the goodness-of-fit plots and individual observed and predicted DMXAA concentration-time curves (data not shown).

The prediction bias suggested that DMXAA disposition might exhibit a nonlinear PK behavior over the studied dose levels (from 6 to 4,900 mg/m²), and therefore, a three-compartment model with a nonlinear elimination process characterized by Michaelis-Menten kinetics (Eq. A) from the central compartment was tested to fit DMXAA concentration data. The nonlinear model resulted in a reduction in the NONMEM OFV by 1,275 and a corresponding drop in the AIC by 1,269, compared with the linear model. In addition, the goodness-of-fit plots and individual observed and predicted concentration-time curves (data not shown) did not show any prediction bias.

![Table 2. Population PK and PD variables of DMXAA, expressed as the estimate (with % relative SE of estimation), and the corresponding interindividual variability](http://www.aacrjournals.org)
from the nonlinear model. Therefore, the three-compartment model with Michaelis-Menten elimination from the central compartment was chosen as the structural PK model.

**Covariate PK model.** The following potentially significant covariates that were selected from GAM analysis were included and tested during the covariate model building: sex on \(V_m\), BSA on \(V_1\), sex on \(V_2\), BSA and assay on \(V_3\), height on \(Q_2\), and age on \(K_m\). Sex and BSA were identified as two significant covariates. With sex being added as the significant covariate on \(V_m\), the OFV was decreased by 49 (\(P < 0.001\)) from the structural model, and the further addition of BSA on \(V_1\) resulted in a further decrease of OFV by 100 (\(P < 0.001\)). Further additions of sex on \(V_2\), BSA and assay on \(V_3\), BSA on \(Q_2\), height on \(Q_3\), or age on \(K_m\) either resulted in a negligible improvement in the fitting (with the decrease in OFV <3.875, \(P > 0.05\)) or extreme difficulty in the model convergence. Therefore, the final PK model consisted of a three-compartment model with nonlinear elimination, with sex and BSA being included as a significant covariate on \(V_m\) and \(V_1\), respectively. Figure 2 shows the goodness-of-fit plots from the final PK model. The population-predicted and posterior Bayes (individual)–predicted DMXAA concentrations were highly correlated with the observed values, with correlation coefficients (\(r^2\)) of 0.9455 and 0.9825, respectively (Fig. 2); IWRES versus IPRED plots and individual observed and predicted concentration-time profiles (data not shown) indicated no prediction bias.

The typical population PK variables and respective interindividual variability estimated from the structural and final PK models are presented in Table 2. From the structural to final model, the value of OFV decreased by 149; however, the addition of covariates sex on \(V_m\) [as \(V_m (\mu\text{mol}/\text{L}/\text{h}) = 112(1+0.474(2-\text{sex}))\), male: \text{sex} = 1, female: \text{sex} = 2] and BSA on \(V_1\) [as \(V_1 (\text{liters}) = 8.19(\text{BSA}/1.8)^{0.857}\)] accounted for small proportion of unexplained interindividual variability on \(V_m\) and \(V_1\) (Table 2), suggesting that other clinical covariates that were not tested in the present study may account for a larger proportion of the interindividual variability. For a male with BSA of 1.8 \(m^2\), the typical population PK variables were estimated from the final model as \(V_m 165 \mu\text{mol}/\text{L}/\text{h}; K_m 102 \mu\text{mol}/\text{L}; V_1 8.19 \text{ liters}; V_2 13.2 \text{ liters}; V_3 52.4 \text{ liters}; Q_2 5.15 \text{ L/h}; \text{and} Q_3 2.33 \text{ L/h} \) (Table 2).

**PK-PD model.** The observed plasma 5-HIAA concentration-time profiles following each i.v. infusion of DMXAA in the individual patients are depicted in Fig. 1D. 5-HIAA peak concentrations, in general, appeared between 4 and 12 hours following DMXAA i.v. infusion (Fig. 1D), whereas DMXAA peak concentrations occurred at the end of infusion (20 minutes after the start of infusion). The effect delay was characterized by a first-order rate constant (\(k_{\text{el}}\)) of 0.331 per hour (corresponding to a lag time of 3 hours) from the central plasma compartment to an effect compartment. The goodness-of-fit plots for the analysis of plasma 5-HIAA concentration...
(PD) data are presented in Fig. 3. The typical population PD variables and their interindividual variability, with percent relative SE of estimation, are shown in Table 2. The population estimates for baseline 5-HIAA plasma concentration ($E_0$), maximum increase of 5-HIAA ($E_{\text{max}}$), DMXAA concentration resulting in half-maximum effect ($EC_{50}$) were 46.3 μmol/L, 2.62-fold of the baseline value, and 631 μmol/L, respectively; the respective interindividual variability was estimated as 51%, 30%, and 59% (Table 2). GAM analyses indicated no significant correlations between any of clinical covariates (listed in Table 1) and PD variables.

**Predictions from the final model.** The simulated concentration-time curves of DMXAA and 5-HIAA are shown in Fig. 4A and B. With a dose increase from 50 to 4,400 mg/m², the predicted peak plasma DMXAA concentration ($C_{\text{max}}$) at the end of infusion was increased from 27.3 to 2,902 μmol/L, which resulted in a peak plasma 5-HIAA concentration increasing from 1.02- to 2.42-fold of the baseline value. The plots of DMXAA dose against peak plasma DMXAA and 5-HIAA concentrations are shown in Fig. 4C. The $EC_{50}$ of DMXAA could be achieved at a dose of ~1,000 mg/m².

**Discussion**

Preliminary analysis from the phase I study data suggested that DMXXA exhibited dose-dependent PK behavior, which was characterized by a nonlinear increase of the area under the concentration-time curve (AUC) and maximum plasma concentration ($C_{\text{max}}$) and decrease of system clearance, with an increase of DMXAA dose in cancer patients (2, 16). The nonlinear PK profile of DMXAA was confirmed and well described by the present population PK model incorporating a three-compartment model with saturable elimination characterized by Michaelis-Menten kinetics.

The nonlinear elimination of DMXAA could be attributable to its saturable metabolism. DMXAA is extensively metabolized, mainly through acyl glucuronidation catalyzed by UDP-glucuronosyltransferase 1A9 and 2B7 (17) and through 6-methylhydroxylation by cytochrome P450 1A2 (CYP1A2; refs. 18–21). In vitro enzyme kinetic studies have shown that in human liver microsomes, the mean apparent $K_m$ values are 143 and 21 μmol/L, respectively, for acyl glucuronidation and 6-methylhydroxylation, and the respective mean apparent $V_m$ values are 0.71 and 0.04 nmol/min/mg (22). The $K_m$ value of 102 μmol/L estimated from the population PK model was in line with the $K_m$ value of 143 μmol/L for DMXAA glucuronidation determined in human microsomes, indicating that glucuronidation was a predominant and rate-limited pathway for DMXAA elimination from the body.

The population PK model suggested that the maximum elimination rate ($V_m$) of DMXAA was sex dependent: Males had higher $V_m$ than females. This result was consistent with previous animal studies that showed that DMXAA exhibited a higher system clearance in male rats than in female rats (23). The sex-dependent DMXAA elimination could be due to gender-dependent expression and/or activity of metabolic enzymes responsible for the metabolism of DMXAA, in particular CYP1A2 and UGT1A9 and 2B7. It has been reported that CYP1A2 activity is greater in males than females (24) and that mRNA expression and activity of UDP-glucuronosyltransferases can be regulated by sex hormones (25, 26).

The combined addition of sex and BSA as significant covariates in the PK model reduced the NONMEM OFV by 149 ($P < 0.001$); nevertheless, they accounted for a small fraction of the interindividual variability in DMXAA pharmacokinetics (Table 2), suggesting that other covariates that were not evaluated in this study may make a larger contribution to the PK variability and should be further investigated. Factors influencing DMXAA metabolic pathways could alter the total elimination rate and contribute significantly to the interindividual PK-PD variability of the drug. Genetic polymorphisms in the metabolic enzymes have potential clinical significance. UGT1A9, UGT2B7, and CYP1A2 polymorphisms have been
repeatedly reported to be associated with altered pharmacokinetics and/or toxicity of a number of compounds such as irinotecan, SN-38, carvedilol, and caffeine (27–31). The evaluation of polymorphisms in UGT1A9, 2B7, and CYP1A2 may account for a significant fraction of the interinddividual PK variability of DMXAA. In addition, UGT1A9 and UGT2B7 are inducible by various compounds such as BNF (32), t-butylhydroquinone, and 2,3,7,8-tetrachlorodibenzo-p-dioxin (33), whereas CYP1A2 can be induced by smoking (34) and certain antitumor agents such as aminoflavone (35). Potential enzyme induction or inhibition by concomitant drugs may alter the pharmacokinetics of DMXAA when it is combined with other chemotherapeutic agents and should be further investigated in the future phase II/III combination studies.

DMXAA-induced increase of plasma 5-HIAA concentrations has been shown to be a result of its vascular effects involving the release of serotonin by platelets, and therefore, 5-HIAA was used as a biomarker of DMXAA treatment. The time courses of plasma DMXAA and 5-HIAA concentrations were simultaneously fitted by the effect compartment PK-PD model with a stimulatory $E_{\text{max}}$ model related to baseline describing the antivascular effect of DMXAA. Because of the nonlinearity of DMXAA pharmacokinetics and its $E_{\text{max}}$ effect on plasma 5-HIAA, the increase of 5-HIAA plasma concentration was not proportional to the increase of DMXAA dose. As shown by simulations in a typical individual (Fig. 4C), at a dose of ~1,000 mg/m², the EC₅₀ (typical population value, 631 μmol/L) of DMXAA was achieved at the end of infusion, which resulted in a peak plasma 5-HIAA concentration at ~1.45-fold of the baseline value. When the dose was increased to 2,000 mg/m², DMXAA peak plasma concentration was ~1,300 μmol/L and the achieved 5-HIAA peak concentration was ~1.87-fold of the baseline value (Fig. 4). With further increases of the doses beyond 2,000 mg/m², both DMXAA and 5-HIAA peak plasma concentrations gradually increased and reached the plateau (Fig. 4C), whereas DMXAA-induced toxicities (grades 1–4) increased significantly (1, 2). Transient prolongation of the corrected cardiac Q₁ interval was observed in all 13 patients evaluated at doses of 2,000 mg/m² and above in the New Zealand study (1) and the maximum tolerated dose of DMXAA was established at 3,700 mg/m² (2). Therefore, combining the pharmacodynamic effect based on the present PK-PD model and toxicity data, a DMXAA dose ranging from 1,000 to 2,000 mg/m² is recommended for future studies, supporting the DMXAA doses of 1,200 and 1,800 mg/m² currently under evaluation in combination with chemotherapy in phase II studies.

In conclusion, DMXAA plasma disposition was characterized by a nonlinear elimination process. BSA was a significant covariate in the PK model, supporting the current use of BSA-guided dosing. The PK-PD model, with 5-HIAA as a biomarker, supports the use of DMXAA dose of 1,000 to 2,000 mg/m² in phase II studies, and provides an example of how PK-PD models can be used to aid in selection of drug doses for phase II evaluation.

References

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Population Pharmacokinetic-Pharmacodynamic Model of the Vascular-Disrupting Agent 5,6-Dimethylxanthenone-4-Acetic Acid in Cancer Patients

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