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의학박사 학위논문

The effect of solute carrier organic transporter family member 2B1 genotype and cytochrome P450 3A activity on the pharmacokinetic characteristics of voriconazole

Solute carrier organic transporter family member 2B1 유전형 및 CYP3A 대사능이 voriconazole 의 약동학적 특성에 미치는 영향 평가 연구

2020년 08월

서울대학교 의과대학 대학원 협동과정 임상약리학전공 이 상 워 Solute carrier organic transporter family member 2B1 유전형 및 CYP3A 대사능이 voriconazole 의 약동학적 특성에 미치는 영향 평가 연구

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Department of Clinical Pharmacology and Therapeutics, Seoul National University College of Medicine Sang Won Lee Solute carrier organic transporter family member 2B1 유전형 및 CYP3A 대사능이 voriconazole 의 약동학적 특성에 미치는 영향 평가 연구

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The effect of solute carrier organic transporter family member 2B1 genotype and cytochrome P450 3A activity on the pharmacokinetic characteristics of voriconazole

by Sang Won Lee

A thesis submitted to the Department of Clinical Pharmacology and Therapeutics in partial fulfillment of the requirements for the Degree of Doctor of Philosophy in Medicine at Seoul National University College of Medicine

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ABSTRACT

Introduction: High pharmacokinetic variability of voriconazole is mainly explained by CYP2C19 phenotype, but there are still unknown factors affecting the variability.

Methods: In this study, the effect of solute carrier organic anion transporter family member 2B1 (SLCO2B1) genotype on the pharmacokinetics (PKs) of voriconazole was evaluated in 12 healthy CYP2C19 poor metabolizers after a single administration of voriconazole 200 mg intravenously and orally. A model—based simulation was done to further explore the effect of SLCO2B1 genotype on the pharmacokinetics of voriconazole. In addition, the influence of CYP3A4 enzyme activity was also explored.

Results: The oral absorption of voriconazole was decreased and delayed in the subjects with the SLCO2B1 c.*396T>C variant compared to the subjects with wild-type. However, the CYP3A activity markers measured in this study did not show significant association with metabolism of voriconazole. SLCO2B1 genotype did not seem to have a significant effect on the probability of reaching therapeutic range after a standard voriconazole dosing regimen.

Conclusions: The results suggest that the SLCO2B1 c.*396T>C

may be associated with the decreased function of intestinal OATP2B1, and it could contribute to inter-individual PK variability of voriconazole. The clinical relevance of these findings should be confirmed through further investigations.

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Keywords: Voriconazole, drug transporter, SLCO2B1, pharmacokinetics, pharmacogenomics

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Area under the concentration—time curve from 0

 AUC_{last}

to last measurable time-point

Area under the concentration—time curve from 0

AUCinf

to infinity

Area under the glucose concentration—time curve

 AUG_{0-4h}

for 4 hours

BMI Body mass index

CI Confidence interval

CL/F Total apparent clearance

CL_{Cr} Creatinine clearance

CL_R Renal clearance

C_{max} Maximum concentration of metformin

F Bioavailability

 F_{0-4h} Bioavailability from 0 to 4 hours post-dose

GMR Geometric mean ratio

IR Immediate release formulation

OCT Organic cation transporter

PCR Polymerase chain reaction

PD Pharmacodynamic

PK Pharmacokinetic

SNP Single nucleotide polymorphism

SST Serum separating tube

 $t_{1/2}$ Terminal half-life

Time to reach maximum concentration of

 T_{max}

metformin

 V_d/F Apparent volume of distribution

 λ_z Terminal elimination rate constant

INTRODUCTION

Invasive fungal infections, including aspergillosis, have caused significant medical problems, and are one of the leading causes of mortality in immunocompromised patients [1]. Voriconazole, a second—generation triazole, is one of the widely used antifungal agents for the prevention and treatment of invasive fungal infections [2, 3]. The use of voriconazole is, however, limited in some patients due to its adverse drug reactions, such as hepatotoxicity and central nervous system toxicity (visual disturbance, hallucinations, and altered mental status). These adverse reactions of voriconazole are closely related to its plasma concentrations [4]. The target trough concentration of voriconazole in plasma as recommended by British Society for Medical Mycology for the treatment of invasive fungal infections is between 1.0 and 5.5 mg/L [4, 5]. The plasma concentration of voriconazole, however, can widely vary among the individuals, and therefore, its efficient and safe use is often difficult [4, 6].

Voriconazole is primarily metabolized in the liver by the cytochrome P450 (CYP) enzymes (including CYP2C19, CYP2C9, and CYP3A4 isozymes). Genetic polymorphisms in CYP2C19 enzyme are one of the key factors that attribute to the inter-

individual variability of systemic voriconazole exposures [7-9]. CYP2C19 is the major metabolic enzyme that metabolizes voriconazole into its major inactive metabolite, voriconazole Noxide. The CYP2C19 metabolizers (PMs) poor have demonstrated higher voriconazole systemic exposures than the intermediate metabolizers (IMs) or the extensive metabolizers (EMs) [9, 10]. The genetic polymorphisms in the CYP2C19 gene, however, can only explain 30-50% of the inter-individual variability in the systemic exposure of voriconazole, and many other contributing factors are still unexplained [11].

To identify the novel genetic variants affecting the pharmacokinetics (PK) of voriconazole, we have performed a pharmacogenomic analysis using the blood samples of the CYP2C19 PMs obtained from a previously performed clinical study (NCT01657201) [12], using a targeted Next-Generation Sequencing (NGS) pharmacogenomic panel [13]. From the NGS analysis, a total of 2,336 single nucleotide polymorphisms (SNPs) were identified. Among those SNPs, one (c.*396T>C, rs3781727) is located in the three-prime untranslated region (3'-UTR) of the solute carrier organic anion transporter family member 2B1 (SLCO2B1) gene. Its minor allele frequency (MAF) in the

Koreans is 29.3% and is significantly associated with the systemic voriconazole exposure [14], although this data needs to be validated in further in-vitro or in-vivo studies.

As voriconazole is also metabolized by the CYP3A4 enzyme, the functional variability of the CYP3A4 can be one of the factors attributing to the PK variability of voriconazole [8, 15]. Most of the CYP3A4 SNPs except for CYP3A4*22 are, however, known to have no clinically significant effect on the enzyme function. Further, the frequency of CYP3A4*22 is very rare in Asians [16]. Previous studies have indicated that the blood CYP3A4 messenger ribonucleic acid (mRNA) expression levels or urinary endogenous CYP3A metabolic biomarkers (6ß-OH-cortisol/cortisol and 6β-OH-cortisone/cortisone) related to the CYP3A enzyme activity [17-25]. It has been also demonstrated that the CYP3A5 enzyme function is closely related to the CYP3A4 enzyme function, whereas, the CYP3A5 SNP (CYP3A5*3) is not related to the PK of voriconazole [9, 16].

Based on these findings, the present study aimed to confirm the effects of SLCO2B1 c.*396T>C variant on the PK of voriconazole in the CYP2C19 PMs. Furthermore, the influence of the CYP3A4 enzyme activity on the PK of voriconazole was also

explored using CYP3A5 genotype, blood CYP3A4 mRNA expression levels, and urinary endogenous metabolic biomarkers. Additionally, we developed a population pharmacokinetic model of voriconazole based on our findings to further explore the influence of SLCO2B1 genotype on the pharmacokinetics of voriconazole after multiple administration.

METHODS

Subjects

Twelve healthy male subjects, aged between 20 and 45 years, with a body mass index in the range of 18.0 – 27.0 kg/m2, and all CYP2C19 PMs were enrolled in this study based on their SLCO2B1 c.*396T>C status [SLCO2B1 c.*396T>C wild-type and variant type (n = 6 or each type)]. To control the genetic factors except for SLCO2B1, only subjects who were CYP2C9 EMs and CYP3A4*22 non-carriers were enrolled to the study. Subjects with history of metabolic, hepatic, renal, hematological, pulmonary, cardiovascular, gastrointestinal, urological, endocrine, neurological, or psychiatric disorders that could have a significant effect on the PK of voriconazole were excluded from this study.

Clinical study design

This clinical study (NCT02906176) was approved by the Institutional Review Board of the Seoul National University Hospital, Seoul, Republic of Korea. The study was conducted in accordance to the principles of the Declaration of Helsinki, and

International Conference on Harmonisation of Technical Requirements for Pharmaceuticals for Human Use (ICH) Good Clinical Practice.

An open-label, single-dose, one-sequence, two-period, two-treatment, crossover study was performed (Figure 1). For genotyping of CYP2C19, CYP2C9, CYP3A4, CYP3A5, and SLCO2B1, the blood samples were obtained during a screening visit. The enrolled subjects received a single dose of voriconazole 200 mg intravenously (IV, infusion over 90 minutes) during the first period, and orally (with 150 mL of water) during the second period with a one-week washout between each dose.

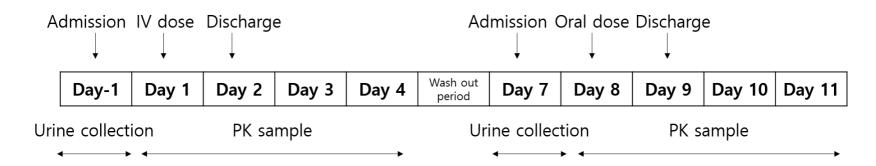


Figure 1. Study design. IV dose: Voriconazole 200 mg, Oral dose: Voriconazole 200 mg

Blood and urine collection

For voriconazole PK analysis, serial blood samples were collected in ethylenediaminetetraacetic acid tubes at 0 (i.e., predose), 0.25, 0.5, 1, 1.5, 2, 3, 4, 6, 8, 12, 24, 48, and 72 h postdose in each period. For measuring the CYP3A mRNA expression levels, whole blood samples were collected at predose in each period. The interval urine samples were collected for 12 h before dosing in each period to analyse the endogenous CYP3A metabolic biomarkers.

Determination of the plasma concentration of voriconazole and voriconazole N-oxide

The quantitative analysis of voriconazole and voriconazole N-oxide was performed simultaneously using a high-performance liquid chromatography (Agilent 1260, Agilent Technologies, Santa Clara, CA, USA) and tandem mass spectrometry (API4000, SCIEX, Framingham, MA, USA) operated in a positive ionization mode. A Luna column (5 µm, C18, 2.0 mm x 100 mm, Phenomenex, Torrance, CA, USA) was used as a stationary phase, while the mobile phase consisted of 0.1% formic acid in

distilled water, and 0.1% formic acid in methanol (10:90, v/v) under an isocratic condition with a flow rate of 0.3 mL/min. The samples were extracted using 100% plasma Voriconazole-d3 was used as an internal standard. The concentration of voriconazole (or voriconazole N-oxide) was determined by computing the peak-area ratios of voriconazole (or voriconazole N-oxide) to voriconazole-d3 analysed in a reaction monitoring multiple (MRM) mode (m/z) $350.216 \rightarrow 281.200$ for voriconazole. m/z: $366.128 \rightarrow 143.200$ for voriconazole N-oxide. and m/z: $353.268 \rightarrow 284.200$ voriconazole-d3). Linear calibration curves were established between the concentration range of 15 and 7500 ng/mL for both voriconazole and voriconazole N-oxide. The correlation coefficients (r) of the calibration curves were > 0.9966 for voriconazole and 0.9976 for voriconazole N-oxide. The intraand inter-day precision results of the quality control samples for voriconazole were < 7.8% and 4.1%, respectively, and the intraand inter-day accuracy results were approximately between 98.6-107.3% and 98.2-104.3%, respectively. The intra- and inter-day precision results of the quality control samples for voriconazole N-oxide were < 5.7% and 9.9%, respectively, and

the intra- and inter-day accuracy results were approximately between 99.3-108.9% and 95.9-100.5%, respectively.

Genotyping of CYP2C19, CYP2C9, CYP3A4, CYP3A5, and SLCO2B1

The DNA for genotyping was extracted from 200 µL of peripheral whole blood using a QIAamp DNA Blood Mini Kit (QIAGEN GmbH, Germany). The genotyping was performed using TagMan allelic discrimination assays on a real time-polymerase chain reaction (PCR) System (Applied Biosystems®, Foster City, CA, USA). Ten microliters of PCR reaction mixture was prepared with 5 µL of 2X TaqMan Genotyping Master Mix, 0.5 µL of 20X Drug Metabolism Genotyping Assay Mix, 3.5 μL of DNase-free water, and 1 µL of DNA. The genotyping for the CYP2C19*2 C_25986767_70), (rs4244285, assay ID: CYP2C19*3 (rs4986893, assay ID: C_27861809_10), CYP2C9*3 (rs1057910, ID: C_27104892_10), CYP3A4*22 assay (rs35599367, assay ID: C_59013445_10), CYP3A5*3 (rs776746, assay ID: C_26201809_30), and SLCO2B1 (rs3781727, assay ID: C_27500327_10) SNPs was performed with validated TaqMan Genotyping Assays purchased from

Applied Biosystems. The PCR reactions were carried out as follows: initial denaturation at 95°C for 10 min, 50 cycles for denaturation at 92°C for 15 s, and anneal/extension at 60°C for 1 min. The allelic discrimination results were determined after the amplification by performing an end-point read using a 7500 Real-Time PCR System software version 2.0.1 (Applied Biosystems, Foster City, CA, USA).

Estimation of CYP3A4 mRNA expressions in blood

The total RNA was extracted from the whole blood using a PAXgene Blood RNA Kit (PreAnalytiX). The complementary DNA (cDNA) was synthesized by the reverse transcription of 1 µg of total RNA using a QuantiNova Reverse Transcription Kit (Qiagen). A total of 15 µL of genomic DNA removal reaction was prepared with 2 µL of gDNA removal mix, 1 µL of internal control RNA, and 1 µg of RNA and RNase—free water. After incubation for 2 min at 45°C, 5 µL of a reverse—transcription Master Mix containing reverse transcription enzyme, buffer, and deoxynucleotides (dNTPs) were added to the genomic DNA removal reaction. The mixture was incubated for 20 min at 45°C, and the reverse transcriptase was inactivated for 5 min at 85°C.

A total of 20 μL of PCR reaction mixture was prepared with 10 μL of 2X QuantiNova SYBR Green PCR Master Mix (Qiagen), 2 μL QuantiTect Primer Assay (Qiagen), 6 μL of nuclease-free water, and 2 μL of complementary DNA. A real-time PCR was performed using CFX Connect Real-time PCR instrument (Bio-Rad, Hercules, CA, USA) as follows: initial denaturation at 95°C for 2 min, 45 cycles for denaturation at 95°C for 5 s, and anneal/extension at 60°C for 10 sec. Glyceraldehyde 3-phosphate dehydrogenase (GAPDH) expression was used to normalize the CYP3A4 mRNA expression.

Pharmacokinetics data analysis

The PK parameters of voriconazole were determined by a non-compartmental analysis (NCA) using Phoenix WinNonlin software version 7.0 (Certara, St Louis, MO, USA). The area under the plasma concentration versus time curve (AUC) from 0 hour to the last measurable time (AUC $_{last}$) and AUC from 0 hour to 4 hour (AUC $_{0-4h}$) were determined using the linear trapezoidal method for the ascending concentrations and the log trapezoidal method for the descending concentration. The AUC $_{0-4h}$ was used in this study to evaluate the PK change during the oral absorption

phase. The maximum concentration (C_{max}) and the time to reach C_{max} (T_{max}) were obtained from the observed values. The elimination rate constant (λ_z) was calculated using a linear regression of the log-linear portion of concentration—time curve, and the elimination half—life $(t_{1/2})$ was calculated as $\ln (2)/\lambda_z$. The total clearance (CL) and apparent clearance (CL/F) were calculated using the ratio of the dose to the AUC. The volume of distribution (V_d) and apparent volume of distribution (V_d/F) were calculated as the CL (or CL/F) divided by $\lambda_z.$ The metabolic ratios were calculated as AUC_{last} of voriconazole N-oxide/AUC_{last} of voriconazole. The AUC from time zero to infinity (AUC_{inf}) was calculated as AUC_{last} + Ct/ λ_z , where Ct is the last measurable concentration. Bioavailability (F) was calculated as AUCinf after oral dose/AUCinf after IV dose, and F from 0 to 4 hours post-dose (F_{0-4h}) was calculated as AUC_{0-4h} 4h after oral dose/AUC_{0-4h} after IV dose. The PK analysis set included all the subjects who completed the visit and received the study drug and have at least one evaluable post-dose PK concentration data. Actual sampling time was used to derive the PK parameters. Concentrations that were below the limit of quantification (BLQ) before T_{max} were treated as zero. However, any concentrations that were BLQ after T_{max} were treated as a missing value. Missing values were excluded from the calculations.

Analysis of endogenous CYP3A metabolic biomarkers

The urinary concentrations of four steroids (6β-OH-cortisol, cortisol, 6β-OH-cortisone, and cortisone) were quantified using a 7890B series gas chromatography (Agilent Technologies, Santa Clara, CA, USA) coupled with a 7000B series triple quadrupole mass spectrometry (Agilent Technologies, GC-MS) following a previously described method [24]. Briefly, the four steroids were extracted from 2 mL of urine followed by solidphase extraction, hydrolysis with β-glucuronidase, and liquidliquid extraction. The extracted samples were evaporated using an N₂ evaporator, and the dried samples were derivatized with 40 μL of N-methyl-N-(trimethylsilyl) trifluoroacetamide. Finally, 3 μL of each sample was injected into the GC-MS for quantification. The CYP3A activities were estimated by the ratio of 6β-OH-cortisol/cortisol and 6β-OH-cortisone/cortisone [23-25].

Statistical analysis

All the demographic characteristics and PK parameters are expressed as the arithmetic mean and standard deviation (SD). Geometric mean ratios (GMRs) of the PK parameters and their 90% confidence intervals (CIs) for SLCO2B1 c.*396T>C carriers (variant type) over non-carriers (wild type) were estimated using a general linear model. In this model, log-transformed PK parameters were used as dependent variables and genotype was used as a covariate. A non-parametric Mann-Whitney test was used to compare the PK parameters. A P value of < 0.05 was considered as statistically significant. All statistical analyses were performed using SAS software version 9.4 (SAS Institute Inc., Cary, NC, USA).

Population pharmacokinetic analysis

We used a logarithmically transformed voriconazole concentration data from the clinical trial to perform a population pharmacokinetic analysis using a non-linear mixed effect modelling approach with NONMEM (version 7.3.0, Icon Development Solutions, Ellicott City, MD, USA). To estimate the

pharmacokinetic parameters and their variabilities, a first-order conditional estimation method with the interaction option was employed.

The structure model was based on a known three-compartment models with a hypothetical inhibition compartment and a linear and/or non-linear elimination (i.e. Michaelis-Menten) model [26]. The absorption profile of oral voriconazole was described by a first-order process with a lag time, and the absolute bioavailability (F) was estimated using a logit model. The inter-individual variability for each pharmacokinetic parameter was evaluated using an exponential error model. An additive residual error model was used to describe the residual unexplained variability.

The effect of *SLCO2B1* genotype on the pharmacokinetics of voriconazole were tested using separate categories of VT referenced to WT. It was considered to be statistically significant when the objective function value (OFV) decreased > 3.84 (p < 0.05, χ^2 distribution with 1 degree of freedom). Only the biologically plausible parameter—covariate relationships were considered and included in the final model.

Model selection and validation

Throughout the model development process, model selection was evaluated based on the goodness of fit plots, the estimates, precision of parameters, and the decrease in OFV. The goodness of fit plots consisted of four plots as follows: observations versus predictions, observations population versus individual predictions, conditional weighted residual versus population predictions, and conditional weighted residuals versus time. The predictive performance of the model was assessed graphically by visual predictive checks (VPCs) performed by stratification of the SLCO2B1 genotype (WT and VT) and rout of administration (IV and oral). The adequacy of the model was demonstrated by plotting the time course of the observations along with the prediction interval for the simulated values.

Model-based simulation

Based on the final parameter estimates of the developed model. model-based simulations were performed to predict the concentration profiles of voriconazole according to the SLCO2B1 genotypes after multiple oral doses of different dosing regimens for CYP2C19 EM and PM subjects. For each CYP2C19 phenotype (EM and PM), 500 individual concentration-time profiles were simulated (A total of 1,000 individual concentration—time profiles). Each individual was randomly assigned to be either SLCO2B1 WT or VT. All parameters were fixed as the estimated values from the final model, including the inter-individual variabilities. For CYP2C19 PM subjects, e^{-0.746} was multiplied to clearance (CL), and $e^{-0.44}$ was multiplied to fraction of clearance which cannot be explained (RCLP). For SLCO2B1 VT subjects, $e^{0.154}$, $e^{0.062}$, $e^{-0.0824}$, and $e^{0.0289}$ was multiplied to central volume of distribution (V2), clearance (CL), absorption rate constant (K_a) , and bioavailability (F_1) , respectively. The simulated dosing regimen was a standard oral dose (400 mg every 12 h on the first day followed by 200 mg every 12 h). The simulation was done for 7 days, which is considered sufficient to achieve the theoretical steady-state. Using the simulated voriconazole concentration according to the *SLCO2B1* genotypes, the probability of attainment of the therapeutic target was calculated, where the target voriconazole trough concentration was predefines as the currently used therapeutic range of 2.0–5.5 mg/L [27, 28]. All simulation processes were carried out using NONMEM (version 7.3.0, Icon Development Solutions, Ellicott City, MD, USA).

RESULTS

Genotypes and demographic characteristics of the subjects

Twelve healthy male CYP2C19 PMs were enrolled and completed the study. Among them, six subjects were of the *SLCO2B1* wild—type (c.*396 TT) and the other six subjects were of the variant types [one with homozygous variant (c.*396 CC) and five with heterozygous variant (c.*396 TC)]. All the subjects were CYP2C9 EMs and were not carrying CYP3A4*22. One subject was CYP3A5 EM (*1/*1), five subjects were CYP3A5 IMs (*1/*3), and six subjects were CYP3A5 PMs (*3/*3) (Table 1).

The mean \pm standard deviation (SD) of the different demographic characteristics of the SLCO2B1 wild-type and variant types were as follows: 32 ± 6 vs. 27 ± 5 years (age), 178 \pm 2 vs. 179 ± 7 cm (height), 71.4 ± 5.8 vs. 73.6 ± 8.5 kg (body weight), and 22.7 ± 1.8 vs. 23.1 ± 2.6 kg/m2 (body mass index), respectively. The demographic characteristics were not significantly different (p > 0.05) between the two SLCO2B1 genotype groups (Table 2).

Table 1. Subjects' genetic status

Subject Number	SLCO2B1	CYP2C19	CYP2C9	CYP3A4*22	CYP3A5
AN101	c.*396TT	PM	EM	non-carrier	*1/*1
AN102	c.*396TT	PM	EM	non-carrier	*3/*3
AN103	c.*396TT	PM	EM	non-carrier	*1/*3
AN104	c.*396TT	PM	EM	non-carrier	*1/*3
AN105	c.*396TT	PM	EM	non-carrier	*1/*3
AN106	c.*396TT	PM	EM	non-carrier	*1/*3
AN201	c.*396TC	PM	EM	non-carrier	*3/*3
AN202	c.*396TC	PM	EM	non-carrier	*3/*3
AN203	c.*396TC	PM	EM	non-carrier	*3/*3
AN204	c.*396TC	PM	EM	non-carrier	*1/*3
AN205	c.*396CC	PM	EM	non-carrier	*3/*3
AN206	c.*396TC	PM	EM	non-carrier	*3/*3

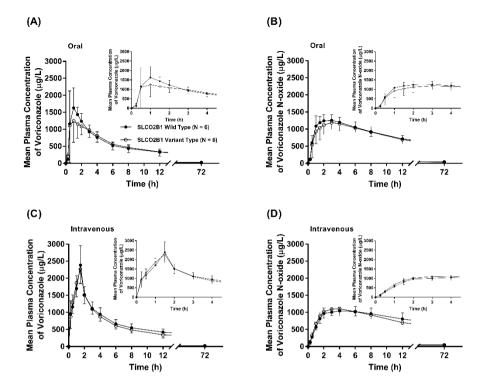
EM, Extensive metabolizer; IM, Intermediate metabolizer; PM, poor metabolizer

Table 2. Demographic characteristics

		SLCO2B1 Wild Type	SLCO2B1 Variant Type	Total
		(N = 6)	(N = 6)	(N = 12)
Age	Mean	32	26.83	29.42
(years)	SD	5.97	4.71	5.79
	Range	24 - 41	20 - 31	20 - 41
BMI	Mean	22.67	23.1	22.88
(kg/m2)	SD	1.77	2.58	2.12
	Range	19.6 - 24.1	19.2 - 25.7	19.2 - 25.7
Height	Mean	1.78	1.79	1.78
(m)	SD	0.02	0.07	0.05
	Range	1.74 - 1.79	1.69 - 1.86	1.69 - 1.86
Weight	Mean	71.37	73.58	72.48
(kg)	SD	5.82	8.52	7.05
	Range	60.7 - 76.3	57.6 - 82.5	57.6 - 82.5

Effects of the SLCO2B1 genotype on the PK of voriconazole

systemic exposure of voriconazole after its administration was lower in the *SLCO2B1* variant type group than the wild-type group (Table 3 and Figure 2-4). The geometric mean (90% confidence interval) ratio [GMR (90% CI)] of the variant group to the wild-type group for the C_{max}, AUC_{last}, and AUC_{0-4h} of voriconazole were 0.76 (0.53-1.08), 0.87 (0.61-1.25), and 0.89 (0.69-1.15), respectively. Furthermore, the inter-individual variations of these PK parameters were higher in the variant group (Table 3 and Figure 4). The T_{max} was prolonged in the variant group as compared to the wild-type group while the $t_{1/2}$ did not show significant difference across the (Table 3 and Figure 4). After intravenous genotypes administration of voriconazole, the AUC_{last} of voriconazole was also lower in the variant group as compared to the wild-type group, while the C_{max} and AUC_{0-4h} did not show much difference across the genotypes (Figure 2, 3, and Table 3). While the metabolic ratios were higher (p = 0.0130) after oral administration compared to IV administration, they were similar across the genotypes regardless of the route of administration (Table 3, Figure 5).



Mean Figure 2. plasma concentration-time profiles voriconazole and voriconazole N-oxide administration of voriconazole 200 mg up to 72 h post-dose (insets: up to 4 h post-dose) in the CYP2C19 poor metabolizers with SLCO2B1 c.*396T>C (rs3781727) wild (filled circle) or variant (empty circle) genotype. The error bars denote the standard deviations. A and B: oral route, and C and D: intravenous route

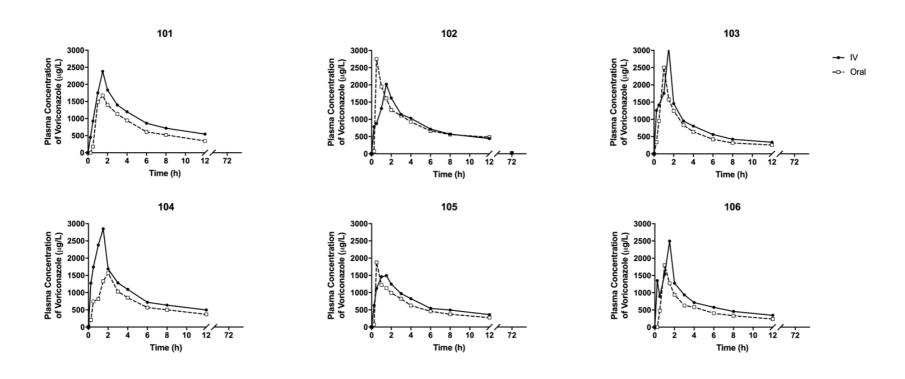


Figure 3-1. Individual plasma concentration—time profiles of voriconazole after a single administration of voriconazole 200 mg up to 72 h post—dose in the CYP2C19 poor metabolizers with SLCO2B1 c.*396T>C (rs3781727) wild (101—106) or variant (201—206) genotype. filled circle/solid—line: intravenous route, empty square/dashed—line: oral route.

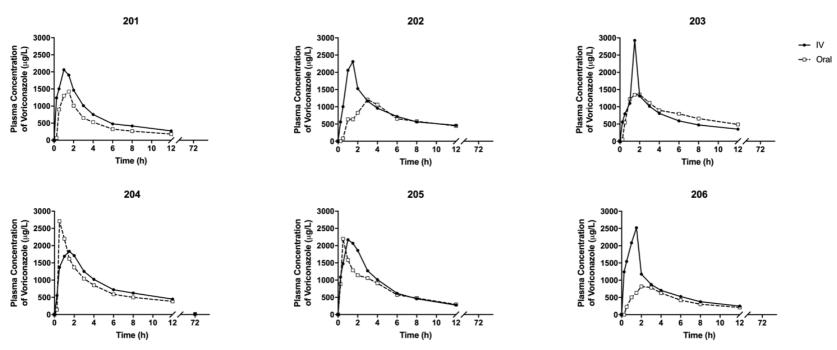


Figure 3-2. Individual plasma concentration—time profiles of voriconazole after a single administration of voriconazole 200 mg up to 72 h post—dose in the CYP2C19 poor metabolizers with SLCO2B1 c.*396T>C (rs3781727) wild (101—106) or variant (201—206) genotype. filled circle/solid—line: intravenous route, empty square/dashed—line: oral route.

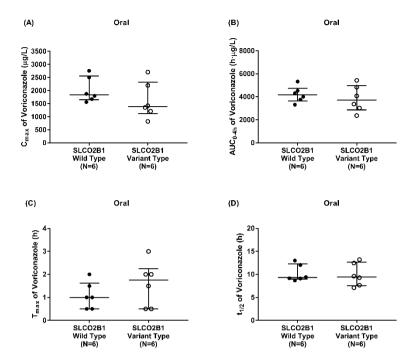
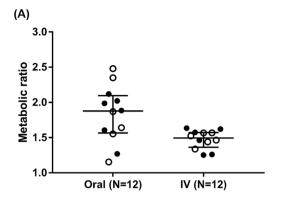


Figure 4. Individual pharmacokinetic parameters (C_{max} , AUC_{0-4h} , T_{max} , and $t_{1/2}$) after a single oral administration of voriconazole 200 mg in CYP2C19 poor metabolizers with SLCO2B1 c.*396T>C (rs3781727) wild (filled circle) or variant (empty circle) genotype. The horizontal lines in the middle, top, and bottom represents the median, the 75^{th} percentile, and the 25^{th} percentile, respectively. (A) C_{max} by oral route. (B) AUC_{0-4h} by oral route. (C) T_{max} by oral route. (D) $t_{1/2}$ by oral route.



- SLCO2B1 Wild Type
- SLCO2B1 Variant Type

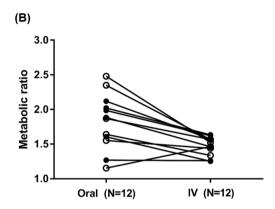


Figure 5. Individual metabolic ratio by rout of administration after administration of voriconazole 200 mg in CYP2C19 poor metabolizers with *SLCO2B1* c.*396T>C (rs3781727) wild (filled circle) or variant (empty circle) genotype. The horizontal lines in the middle, top, and bottom represents the median, the 75th percentile, and the 25th percentile, respectively. IV: intravenous

Table 3. The pharmacokinetic parameters of voriconazole and voriconazole N-oxide after a single oral and intravenous administration of voriconazole 200 mg in the cytochrome P450 2C19 poor metabolizers with *SLCO2B1* c.*396T>C (rs3781727) wild or variant genotypes

Route of administration	Parameters	Voriconazole		Voriconazole N-oxide			
		SLCO2B1 wild-type (n = 6)	SLCO2B1 variant type (n = 6)	GMR [90% CI]*	SLCO2B1 wild-type (n = 6)	SLCO2B1 variant type (n = 6)	GMR [90% CI]*
Intravenous	T _{max} (h)	1.50 [1.50 – 1.50]	1.50 [1.00 – 1.50]	-	4.00 [2.00 – 6.00]	4.00 [3.00 – 4.00]	-
	C _{max} (µg/L)	2384.50 ± 570.85 [23.9]	2303.83 ± 382.26 [16.6]	0.98 [0.78 – 1.23]	1085.85 ± 154.94 [14.3]	1116.53 ± 197.87 [17.7]	1.02 [0.87 - 1.20]
	AUClast	16905.48 ± 3169.53	14930.75 ± 4168.27	0.86	24509.16 ± 3520.37	21872.13 ± 4967.61	0.88
	(h·μg/L)	[18.7]	[27.9]	[0.68 - 1.11]	[14.4]	[22.7]	[0.73 - 1.07]
	AUC _{0-4h}	5411.20 ± 812.04	5396.38 ± 421.90	1.00	3077.30 ± 474.50	3341.46 ± 659.06	1.08
	(h·μg/L)	[15.0]	[7.8]	[0.89 - 1.14]	[15.4]	[19.7]	[0.91 - 1.28]
	t _{1/2} (h)	11.04 ± 2.27 [20.6]	10.23 ± 2.87 [28.1]	-	12.20 ± 2.20 [18.0]	10.71 ± 2.55 [23.8]	-
	CL (L/h)	11.78 ± 2.18 [18.5]	13.42 ± 3.23 [24.1]	-	-	-	-
	Metabolic ratio**	-	_	-	1.47 ± 0.17 [11.6]	1.48 ± 0.09 [6.1]	-
Oral	T _{max} (h)	1.00 [0.50 – 2.00]	1.75 [0.50 – 3.00]	-	2.00 [1.00 – 3.00]	3.50 [1.50 – 4.00]	-
	C _{max} (µg/L)	2027.67 ± 483.08 [23.8]	1619.27 ± 698.46 [43.1]	0.76 [0.53 – 1.08]	1314.83 ± 185.85 [14.1]	1231.92 ± 223.39 [18.1]	0.93 [0.78 – 1.11]
	AUC _{last} (h·μg/L)	13373.42 ± 4220.17	11982.11 ± 4508.32	0.87	23250.79 ± 3011.77	20570.15 ± 5314.13	0.87
		[31.6]	[37.6]	[0.60 - 1.25]	[13.0]	[25.8]	[0.70 - 1.07]
	AUC _{0-4h} (h·μg/L)	4209.45 ± 695.11	3849.54 ± 1150.12	0.89	4176.52 ± 541.19	3826.55 ± 1007.10	0.90
		[16.5]	[29.9]	[0.69 - 1.15]	[13.0]	[26.3]	[0.72 - 1.11]
	t _{1/2} (h)	10.17 ± 1.86 [18.3]	9.90 ± 2.50 [25.3]	-	11.16 ± 2.27 [20.3]	10.78 ± 2.97 [27.6]	_
	CL/F (L/h)	15.58 ± 4.05	16.90 ± 6.08	_	_	_	_

	[26.0]	[35.9]				
F (%)	77.9 ± 13.6 [17.4]	82.5 ± 14.3 [17.3]	-	-	-	-
F _{0-4h} (%)	78.9 ± 15.9 [20.2]	71.3 ± 20.3 [28.5]	-	-	-	_
Metabolic ratio**	-	-	-	1.81 ± 0.32 [17.7]	1.84 ± 0.50 [27.2]	-

Data presented as mean \pm standard deviation [coefficient of variation %] except for T_{max} , which is presented as median [minimum – maximum].

^{*}Geometric mean ratio (GMR) and 90% confidence intervals (CI) of the SLCO2B1 variant type to the SLCO2B1 wild-type for C_{max} , AUC_{last} , and AUC_{0-4h}

^{**}Metabolic ratio was calculated as the AUC_{last} of N-oxide voriconazole divided by the AUC_{last} of voriconazole

Effects of CYP3A activity on the PK of voriconazole

The CYP3A activity measured in this study did not show any significant association with the N-oxide metabolism of voriconazole. The blood CYP3A4 mRNA expressions and the endogenous CYP3A activity biomarkers (6 β -OH-cortisol/cortisol and 6 β -OH-cortisone/cortisone) exhibited no statistically significant correlation with the metabolic ratio of voriconazole (Figure 6). The metabolic ratios were also not significantly different across the CYP3A5 genotypes (Figure 6).

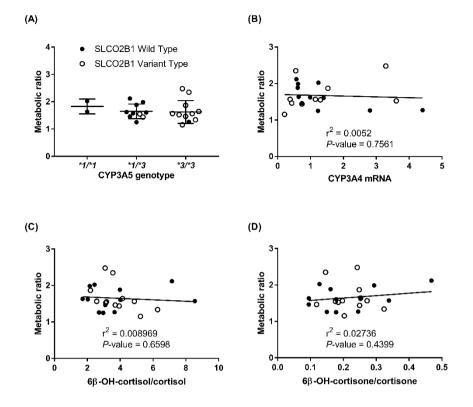


Figure 6. Relationship between the individual metabolic ratio and CYP3A4 activity biomarkers after a single (intravenous or oral) administration of voriconazole 200 mg in CYP2C19 poor metabolizers with SLCO2B1 c.*396T>C (rs3781727) wild (filled circle) or variant (empty circle) genotype. (A) relationship between the individual metabolic ratio and CYP3A5 genotype. (B) relationship between the individual metabolic ratio and CYP3A4 relative baseline (trough) mRNA expression. (C) relationship between the individual metabolic ratio and baseline (trough) 6β -OH-cortisol/cortisol. (D) relationship between the individual 6 *β* −OH− metabolic ratio and baseline (trough) cortisone/cortisone.

Population pharmacokinetic model

A three-compartment model with a first-order oral absorption, an absorption lag time, and elimination along with an inhibition compartment model appropriately described the time-concentration profile of voriconazole, showing distinct non-linear pharmacokinetic behavior.

Since our data was based on 12 healthy subjects, while the original model was developed using approximately 200 subject data (patients and healthy subjects), the model was overcomplicated for our data. Therefore, some of the parameters were fixed with known values from a previous report [26]. The SLCO2B1 genotype did not turn out to be a statistically significant covariate for the PK parameters. However, this was somewhat expected since SLCO2B1 genotype did not show statistically significant effect on the PK of voriconazole based on the NCA from the clinical trial. For exploratory reasons, we have incorporated fixed values (estimated values from covariate analysis) to central volume of distribution (V2), clearance (CL), absorption rate constant (K_a), and bioavailability (F1). The NONMEM control code for the final model was as follows:

\$SUBROUTINES ADVAN6 TOL=5

\$MODEL

COMP = (DEPOT)

COMP=(CENTRAL)

COMP=(PERIPH1)

COMP=(PERIPH2)

COMP=(INHIBIT)

\$PK

IF (GENO.EQ.0) THEN; RM/EM

GENOF=0

GENOF1=0

ENDIF

IF (GENO.EQ.1) THEN; IM

GENOF = THETA(17)

GENOF1=THETA(19)

ENDIF

IF (GENO.EQ.2) THEN; PM

GENOF=THETA(18)

GENOF1 = THETA(20)

ENDIF

IF (GRADE.GE.3) THEN; GRADE 3&4

LFF = THETA(24)

ELSE

LFF=0

ENDIF

IF (SLCO.EQ.0) THEN; WT

SLCOV = 0

SLCOC=0

SLCOK=0

SLCOF=0

ENDIF

IF(SLCO.EQ.1) THEN; VT

```
SLCOV=THETA(25)
SLCOC=THETA(26)
SLCOK=THETA(27)
SLCOF=THETA(28)
ENDIF
TVV2 = THETA(1) * EXP(SLCOV) ; SLCO~V
TVCL = THETA(2) * (WT/70) **THETA(16) * EXP(GENOF) *
EXP(LFF) * EXP(SLCOC) ;SLCO~C
TVV3 = THETA(3) * (WT/70) **THETA(15)
TVQ2 = THETA(4) * (WT/70) **THETA(21)
TVV4 = THETA(5)
TVQ3 = THETA(6)
TVKA = THETA(7) * EXP(SLCOK) ; SLCO~K
TVF1 = THETA(8) * EXP(SLCOF) ; SLCO~F
V2 = TVV2 * EXP(ETA(1))
CL = TVCL * EXP(ETA(2))
V3 = TVV3 * EXP(ETA(3))
Q2 = TVQ2 * EXP(ETA(4))
V4 = TVV4
Q3 = TVQ3
KA = TVKA * EXP(ETA(5))
LGF1 = LOG(TVF1/(1-TVF1)) + ETA(6)
F1 = EXP(LGF1) / (1 + EXP(LGF1))
ALAG1 = THETA(9) * EXP(ETA(7))
RCLF = THETA(10) * EXP(GENOF1) * EXP(ETA(8))
IC50 = THETA(11)
KIC = THETA(12)
KE = CL/V2
K23 = Q2/V2
K32 = Q2/V3
K24 = Q3/V2
K42 = Q3/V4
S2 = V2/1000
$DES
CONC = A(2)/V2
```

```
INH = RCLF + (1-RCLF)*(1 - A(5)/(IC50 + A(5)))
DADT(1) = -KA*A(1)
DADT(2) = KA*A(1) - KE*A(2)*INH - K23*A(2) +
K32*A(3) - K24*A(2) + K42*A(4)
DADT(3) = K23*A(2) - K32*A(3)
DADT(4) = K24*A(2) - K42*A(4)
DADT(5) = KIC*(CONC-A(5))
$ERROR
IPRED=0
IF(F.GT.0) IPRED=LOG(F)
IF (PART.EQ.5) THEN
W = SQRT(THETA(22)**2*IPRED**2 + THETA(23)**2)
ELSE
W = SQRT(THETA(13)**2*IPRED**2 + THETA(14)**2)
ENDIF
Y = IPRED + W*EPS(1)
IRES = DV - IPRED
IWRES = IRES/W
$THETA
(0, 35.6); V2
(0.45.2); CL
(0.56.6); V3
(0, 10.3); Q2
(25.4) FIX; V4
(54.6) FIX; Q3
(0, 1.24); KA
(0, 0.859,1); F1
(0, 0.235); ALAG1
(0.162) FIX;RCLF
(0.01) FIX; IC50
(0.002) FIX;KIC
(0.0001) FIX; Prop.RE (sd)
(0, 0.31); Add.RE (sd)
(0. 2.2) FIX; WT~V3
(0, 0.595) FIX; WT~CL
(-0.186) \text{ FIX} ; \text{IM} \sim \text{CL}
(-0.746) \text{ FIX} ; PM~CL
```

(-0.51) FIX; IM~RCLF

(-0.44) FIX; PM~RCLF (0, 2.56) FIX; WT~Q2

(0.0001) FIX; Prop.RE (sd) PART5 (0.799) FIX; Add.RE (sd) PART5 (-0.75) FIX; GRADE>=3~CL

(0.154) FIX; SLCO~V2 (0.062) FIX; SLCO~CL (-0.0824) FIX; SLCO~Ka (0.0289) FIX; SLCO~F1

\$OMEGA BLOCK(2)

0.109; IIV V2

0.0346 0.128; IIV CL

\$OMEGA

0 FIX; IIV V3 0 FIX; IIV Q2 0.65; IIV KA 0.179; IIV F1 0 FIX; IIV ALAG1 0.4 FIX; IIV RCLF

\$SIGMA 1 FIX

\$EST METHOD=1 INTER MAXEVAL=9999 NOABORT SIG=2 PRINT=1 POSTHOC

\$COV

\$TABLE ID NTIME TIME NTAD TAD AMT RATE DV MDV EVID CMT PRD TRT PART IPRED IWRES CWRES ONEHEADER NOPRINT FILE=sdtab1045

\$TABLE V2 CL V3 Q2 KA F1 ALAG1 V4 Q3 RCLF IC50 KIC ETA1 ETA2 ETA3 ETA4 ETA5 ETA6 ETA7 ETA8 ONEHEADER NOPRINT FILE=patab1045

\$TABLE ID AGE WT HT BMI ALP AST ALT GGT CRE GFR ONEHEADER NOPRINT FILE=cotab1045

\$TABLE ID GENO SEX VPC TDM CON PPI USE LFT GRADE
SLCO ONEHEADER NOPRINT FILE=catab1045

The absorption rate constant of $1.22h^{-1}$ and the lag time of 0.236h appropriately describe the absorption phase of orally administered voriconazole. The absorption oral bioavailability of voriconazole was estimated to be 86%. The estimated typical CL of voriconazole was 48.1 L/h. Most of the typical parameter values were estimated with a good precision (Table 4).

 Table 4. Parameter estimates of the final population

 pharmacokinetic model.

Parameters	Estimates	RSE (%)
Structural model		
V2; central volume of distribution (L)	27.8	19
CL; clearance (L/h)	48.1	4
V3; peripheral 1 volume of distribution (L)	58.1	7
Q2; inter-compartmental clearance between central and peripheral 1 compartment (L/h)	16.3	12
V4; peripheral 2 volume of distribution (L)	25.4 FIX	NA
Q3; inter-compartmental clearance between central and peripheral 2 compartment (L/h)	54.6 FIX	NA
K _a ; absorption rate constant (h ⁻¹)	1.22	33
F ₁ ; bioavailability	0.86	4
ALAG ₁ ; absorption lag-time (h)	0.236	2
RCLP; fraction of clearance which cannot be inhibited	0.162 FIX	NA
IC ₅₀ ; concentration in the inhibition compartment yielding 50% of maximum clearance inhibition	0.01 FIX	NA
K_{IC} ; rate constant into inhibition compartment	0.002 FIX	NA
Inter-individual variability (IIV)		
IIV for V2 (% CV)	44.1	36^{a}
IIV for CL (% CV)	12.1	14ª
IIV for V3 (% CV)	0 FIX	NAª
IIV for Q2 (% CV)	0 FIX	NAª
IIV for Ka (% CV)	114.1	28ª
IIV for F1 (% CV)	83.4	39ª
IIV for ALAG1 (% CV)	0 FIX	NA^{a}
IIV for RCLF (% CV)	70.1 FIX	NA^a

Residual variability

Additive error for healthy subjects (mg/L)	0.224	8	
Additive error for patients (mg/L)	0.799 FIX	NA	_

^a Standard error given on the variance scale; ^b Standard error of the covariance estimate.

Model validation

The basic goodness of fit (Figure 7) and VPC plots (Figure 8) showed a good predictive performance of the developed model and indicated that the model appropriately described the observed voriconazole concentration in accordance with the *SLCO2B1* genotypes.

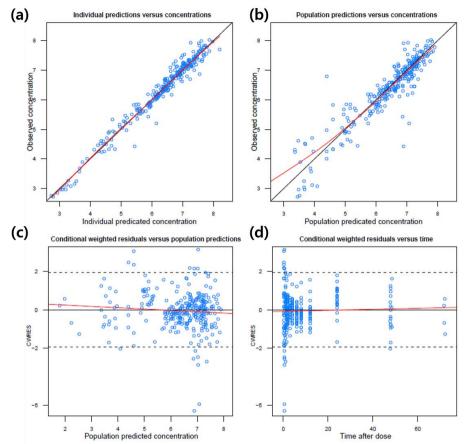


Figure 7. Basic goodness-of-fit plots of final population pharmacokinetic model for voriconazole. (a) observations versus population predicted concentration; (b) observations versus individual predicted concentration; (c) conditional weighted residuals (CWRES) versus population predicted concentration; (d) conditional weighted residuals (CWRES) versus time after dose

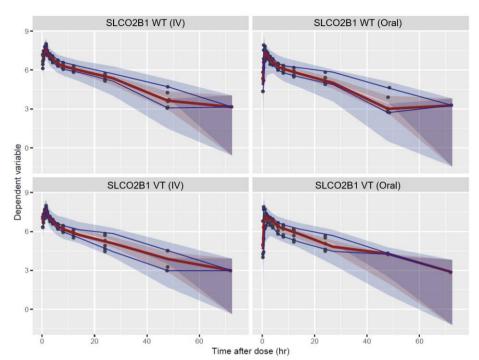


Figure 8. Visual predictive check for SLCO2B1 wild types (WT) and SLO2B1 variant types (VT) in the final pharmacokinetic model by rout of administration. The circles represent the observed concentrations. The lines represent the median (red) and the $5^{\rm th}$ and $95^{\rm th}$ percentiles (blue) of the observed concentration. The areas represent the 95% confidence intervals for the median (red) and 90% prediction interval (blue) of the simulated concentrations.

Simulation of standard dosing regimen according to the SLCO2B1 genotype in CYP2C19 EMs and PMs

Based on the final population pharmacokinetic model, the concentration—time profiles of voriconazole after 7—day multiple oral doses of standard dosing regimen (400 mg twice daily for two doses followed by 200 mg twice daily) were simulated according to the *SLCO2B1* genotypes in CYP2C19 EMs and PMs (Figure 9). When classified according to the *SLCO2B1* genotypes, the trough concentrations reached mostly below the target range (2.0—5.5 mg/L) for both *SLCO2B1* WT and VT in CYP2C19 EMs. In CYP2C19 PMs, the trough concentrations were within the target range regardless of the *SLCO2B1* genotype.

The probabilities of reaching subtherapeutic concentration were higher in *SLCO2B1* VT (73%) compared to WT (68%) in CYP2C19 EMs. Meanwhile, the probabilities of reaching therapeutic concentration (*SLCO2B1* WT: 24%, *SLCO2B1* VT: 22%) and toxic concentrations (*SLCO2B1* WT: 7%, *SLCO2B1* VT: 5%) were higher in *SLCO2B1* WT compared to *SLCO2B1* VT in CYP2C19 EMs. (Figure 10).

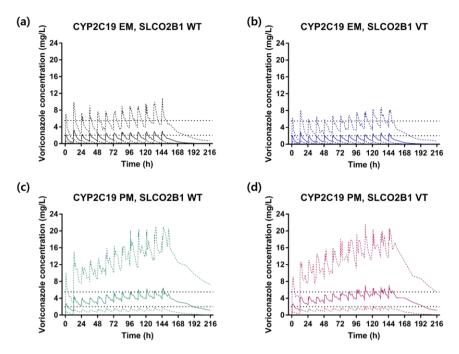
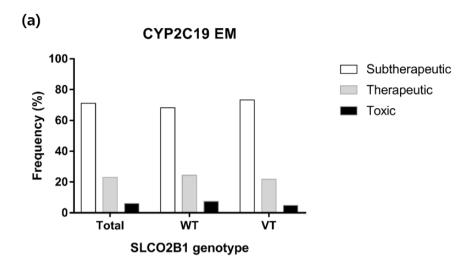


Figure 9. Predicted median concentration—time profile over the first 7 days of treatment of total 1,000 simulated patients. (a) CYP2C19 EM and SLCO2B1 WT; (b) CYP2C19 EM and SLCO2B1 VT; (c) CYP2C19 PM and SLCO2B1 WT; (d) CYP2C19 PM and SLCO2B1 VT. Standard oral dosing (400 mg twice daily for two doses followed by 200 mg twice daily) was used. The solid lines represent the median, with the dotted lines representing the 10th and 90th percentiles. The horizontal dotted lines represent the therapeutic target range for voriconazole trough plasma concentration of 2.0 to 5.5 mg/L.



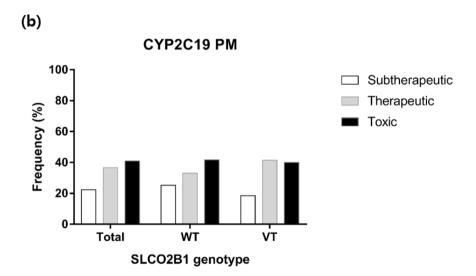


Figure 10. Probability of voriconazole therapeutic target attainment from model-based simulations of voriconazole pharmacokinetic profiles after the following voriconazole standard oral dosing regimens on day 7 (400 mg twice daily for two doses followed by 200 mg twice daily); (a) CYP2C19 EM; (b) CYP2C19 PM. Therapeutic target range for voriconazole trough plasma concentration was from 2.0 to 5.5 mg/L.

DISCUSSION

To maximize the influence of SLCO2B1 c.*396T>C variant on the PK of voriconazole, only the CYP2C19 PMs were included in this study. Several confounding factors apart from the genetic status of CYP2C19, such as disease status, co-medications (CYP inducers and inhibitors), foods, and herbal medicines (e.g., St John's-wort), however, can increase the inter-patient PK variability of voriconazole [2, 29, 30]. Genetic polymorphisms in the metabolizing enzymes apart from CYP2C19, such as CYP2C9 and CYP3A4 or in the drug transporters may also add to the PK variability of voriconazole. To control these possible confounding factors, this study was performed only with the healthy subjects who were CYP2C9 EMs and carrying a CYP3A4*22 wild-type, and the use of co-medications was strictly controlled.

In this study, the systemic exposure of voriconazole were decreased in the subjects with a SLCO2B1 c.*396T>C variant than the subjects carrying a wild-type, although it failed to reach a statistical significance. The C_{max} , AUC_{last} , and AUC_{0-4h} of voriconazole were lower in the variant group compared to the wild-type group with higher inter-individual variations.

Furthermore, the T_{max} of voriconazole was prolonged in the variant group compared to the wild-type group, while the $t_{1/2}$ did not show much difference (Table 3 and Figure 3). It suggests that the rate and extent of oral absorption of voriconazole was decreased in the variant group than the wild-type group.

An influx transporter, organic anion transporting polypeptide 2B1 (OATP2B1) is encoded by the SLCO2B1 gene and is expressed in various tissues in the human body including placenta, heart, skeletal muscle, liver, and intestine [31, 32]. To the best of our knowledge, studies evaluating the clinical implications of SLCO2B1 c.*396T>C SNP are sparse. It has been revealed that the OATP2B1 has an important role in the oral absorption of statins, and the inhibition of OATP2B1 can significantly reduce the systemic exposure of rosuvastatin in human [33]. Moreover, several clinical studies have reported the associations of SLCO2B1 SNP with the PK and response of its substrate drugs. The SLCO2B1 c.1457C>T SNP was shown to decrease the exposure of fexofenadine, and c.935G>A SNP was associated with a decreased montelukast plasma concentration leading to a poor drug response [34, 35]. The SLCO2B1 c.*396T>C SNP is located in the 3'-UTR region of the SLCO2B1 gene, and this SNP is capable of altering the OATP2B1 transporter functions in human. Specifically, the SLCO2B1 c.*396T>C SNP may decrease the OATP2B1 functions in the human intestine, thereby reducing the systemic absorption of voriconazole in the individuals carrying the variant type. Although the difference in AUC_{last} was greater than that of AUC_{0-} 4h between SLCO2B1 wild-type and variant-type group, the differences in C_{max} and AUC_{0-4h} of voriconazole between the two groups were greater when voriconazole was administered orally compared to an intravenous route (Table 3 and Figure 4). Therefore, the claim that the absorption of voriconazole may have been affected by SLCO2B1 genotype cannot be dismissed. Additionally, this study was conducted in subjects with the same CYP2C19, CYP2C9, and CYP3A4*22 genotypes, implying that the SLCO2B1 genotype is most likely responsible for these results.

The F_{0-4h} of voriconazole was reduced in SLCO2B1 variant-type compared to wild-type group, while it was not the case for F. This suggests that while SLCO2B1 genotype may affect the absorption of voriconazole initially, its effect may be reduced over time. This could be explained by the fact that

voriconazole could potentially cross cell membranes via passive diffusion since it is a lipophilic compound of moderate size with no net charge at physiological pH values [3, 39].

The CYP3A4 activity biomarkers evaluated in this study could not explain the additional PK variabilities of voriconazole beside the CYP2C19 genetic status (Figure 6). The CYP3A4 enzyme activity is highly variable among the individuals and its variability cannot be explained by the common genetic variations [36]. The CYP3A4*22 is a well-known allele that causes loss of function of the CYP3A4 enzyme, but its frequency is rare in the general populations as well as in Asians [16]. The CYP3A5*3 is another important SNP that affects the total CYP3A subfamily enzyme function [16]. The CYP3A5*3 was, however, not related to the voriconazole plasma concentration in this study. This finding is consistent with other previous reports [9, 16]. In addition, based on some earlier studies, we expected that the CYP3A4 mRNA expression level in blood or the urinary endogenous steroid ratios could represent the CYP3A4 activity in human [17-25]. Our study result, however, may indicate that the variability of the CYP3A4 activity dose not significantly affect voriconazole N-oxidation when the other metabolizing pathway (e.g. CYP2C9) is still functional.

The metabolic ratio of voriconazole was increased when administered orally, regardless of the SLCO2B1 genotype. It is reasonable to assume that the increased metabolic ratio is due to metabolism of voriconazole. the first-pass First-pass metabolism may occur in the liver or intestine, and CYP3A4 is the predominant enzyme (approximately 80%) expressed in the small intestine [37]. Therefore, if metabolism in the intestine was a significant factor in the first-pass metabolism, CYP3A4 would have been the major source of the metabolism. However, in our study, the CYP3A4 activity did not affect the metabolism of voriconazole which implies that the intestinal CYP3A4 did not play a significant role in the increased metabolic ratio after oral administration. This is consistent with a previous physiologically based pharmacokinetic model study that suggested voriconazole is not subject to intestinal first-pass metabolism in adults [38].

This study had some major limitations. First, due to the exploratory nature, the study was conducted in a limited number of subjects. As a result, we could not show statistically significant difference in PK parameters between the two

genotype groups, and the simulation results based on the final population PK model did not show *SLCO2B1* genotype having a drastic effect on the probability of reaching therapeutic range in CYP2C19 EMs. Secondly, only CYP2C19 PM males were enrolled in the clinical study, while the effects of *SLCO2B1* genotype on the PK of voriconazole in CYP2C19 EMs and female subjects are unknown. Consequently, the simulation results may not properly predict the time—concentration profile of voriconazole in CYP2C19 EMs and/or female subjects. Therefore, further studies with greater sample size including variety of subjects (e.g. CYP2C19 EM, female) are warranted to confirm our hypothesis. However, we believe our clinical study could serve as an early proof of concept study.

CONCLUSION

In conclusion, the oral absorption of voriconazole was decreased and delayed in the subjects with a *SLCO2B1* c.*396T>C variant compared to the wild-type. This finding suggests that the *SLCO2B1* c.*396T>C may be associated with a reduced function of the intestinal OATP2B1, and thereby attributing to the inter-individual PK variability of voriconazole. Further studies in a bigger pool of patients will be helpful to evaluate the clinical importance of the *SLCO2B1* c.*396T>C genetic polymorphism with regard to the safety and efficacy of voriconazole.

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APPENDICES

1. R script for goodness of fit (GOF) plots

```
if(Sys.getenv("R_LIB") != "") { .libPaths(c(Sys.getenv("R_LIB"), .libPaths())) }
library (lattice)
library (grid)
reg.fields <- c("DV", "PRED", "IPRED", "TIME", "CWRES")
models <- list (
  "run1045" = list (
   modelfile
                  = "run1045.ctl",
   description
                  = "RE_SLCO~all",
   reference_model = "run1027".
   working dir = "C:/NM/phd/run".
   data file
                  = "..//PopPK_phd_2.csv",
   output_file
                 = "run1045.lst".
                  = c ("sdtab1045","patab1045","cotab1045","catab1045"),
   tables
                  = list(
   estim
                 = c(3.02E+01, 5.01E+01, 5.76E+01, 1.61E+01, 2.54E+01, 5.46E+01, 1.17E+00, 8.72E-01,
     th
2.36E-01, 1.62E-01, 1.00E-02, 2.00E-03, 1.00E-04, 2.25E-01, 2.20E+00, 5.95E-01, -1.86E-01, -7.46E-01,
-5.10E-01, -4.40E-01, 2.56E+00, 1.00E-04, 7.99E-01, -7.50E-01),
```

```
th se
               = c(5.97E+00, 1.86E+00, 4.48E+00, 2.02E+00, NA, NA, 3.81E-01, 3.38E-02, 4.48E-03, NA,
= c(0.16, 0.014, 0, 0, 0.825, 0.526, 0, 0.259),
     om
     om_se
               = c(0.132,0.00466,0.0.0.434,0.391,0.0)
     om_block
               -0.0212, 0.014, 0, 0, 0, 0, 0, 0, 0
                         0, 0, 0, 0, 0, 0, 0, 0,
                         0, 0, 0, 0, 0, 0, 0, 0,
                         0, 0, 0, 0, 0.825, 0, 0, 0,
                         0, 0, 0, 0, 0, 0.526, 0, 0,
                         0, 0, 0, 0, 0, 0, 0, 0,
                         0, 0, 0, 0, 0, 0, 0, 0.259, ncol=8,
                = c(0.132,0.00466,0,0,0.434,0.391,0,0),
     om se
               = c(1),
     si
     si_se
               = c(0).
               = matrix(c(1), ncol=1),
     si_block
     si_se_block = matrix(c(0), ncol=1)
run_from <- list(software = "pirana", version = "2.9.7")
open_res <- 1
setwd('C:/NM/phd/run')
```

```
model names <- names (models)
if (!file.exists("pirana_reports")) {dir.create ("pirana_reports")}
for (i in 1:length(model_names)) {
    fname <- paste ("pirana_reports/",names (models) [i]."_GOF.pdf", sep="")
    header<- paste ("Goodness of fit model", names (models) [i], sep=" ")
   if (file.exists (fname)) {
        file.remove (fname)
              <- models[[model names[i]]]</pre>
    mod
    tab_file <- c(mod$tables[grep("sdtab", mod$tables)])[1]
    if (is.na(tab file)) { # then assume the first table has the gof variables
        tab file <- mod$tables[1]
   if (file.exists (tab file)) {
        tab <- read.table (tab_file, skip=1, header=T) # NONMEM table with ONEHEADER option
        m <- match (names (models [[model_names [i]]] $input_trans), colnames (tab))
                                                                                            # tranlation of $INPUT
variables
        colnames(tab)[m] <- unlist (models[[model_names[i]]]$input_trans, use.names=F) # tranlation of $INPUT
variables
        if ("MDV" %in% names(tab)) { tab <- tab[tab$MDV==0,] }
                                                                                          # check if MDV present
       if ("EVID" %in% names(tab)) { tab <- tab [tab$EVID==0,] }
                                                                                         # check if EVID present
        not.found <- req.fields[is.na(match(req.fields, colnames(tab)))]</pre>
        if (length (not.found) > 0) {
          cat (paste("The variable(s)",not.found,"were not found. Please check your output tables.\nStopping R
```

```
execution."))
         quit()
# R code for plotting starts here
       pan2 \leftarrow function(x,y,...)
           panel.xyplot(x,y,...)
           panel.abline(0,1,col="black")
           panel.loess(x,y,col="red", lwd=1)
       pan3 \leftarrow function(x,y,...)
           panel.xyplot(x,y,...)
           panel.abline (0,0,col="black")
           panel.abline (1.96,0,col="black",lty=2)
           panel.abline (-1.96,0,\text{col}=\text{"black"},\text{lty}=2)
           panel.lmline (x,y,col="red",lwt=2)
       ipredlim
                   \leftarrow range (tab$IPRED) * c(0.95, 1.05)
                   \leftarrow range (tab$DV) * c(0.95, 1.05)
       dvlim
                   <- range(c(ipredlim, dvlim))
       lim 1
       reslim <- range(tab$CWRES) * 1.05
       tick \langle -c(-5, -2, 0, 2, 5) \rangle
```

```
pl1 <-
             xyplot (DV~IPRED, data=tab,
              panel=pan2,
              type="p", cex=0.5,
              ylab=list("Observed concentration", cex=0.7, just=c(0.5,2.5)),
              xlab=list("Individual predicated concentration", cex=0.7, just=c(0.5,-1)),
               scales=list (tck=c(1, 0), x=list(cex=0.5), y=list(cex=0.5)),
              ylim=lim1,
              xlim=lim1,
              main=list ("Individual predictions versus concentrations", cex=0.55, just=c(0.43,4))
pl2 <-
             xyplot(DV~PRED, data=tab,
              panel=pan2,
              type="p", cex=0.5,
              ylab=list("Observed concentration", cex=0.7, just=c(0.5,2.5)),
              xlab=list("Population predicated concentration", cex=0.7, just=c(0.5,-1)),
              ylim=lim1,
              xlim=lim1,
               scales=list (tck=c(1, 0), x=list(cex=0.5), y=list(cex=0.5)),
              main=list ("Population predictions versus concentrations", cex=0.55, just=c(0.43,4))
pl3 <-
             xyplot(CWRES~PRED, data=tab,
```

```
ylim=reslim,
                    panel=pan3.
                    type="p", cex=0.5,
                    vlab=list("CWRES", cex=0.5, just=c(0.7.3)).
                    xlab=list(" Population predicted concentration", cex=0.7, just=c(0.5,-1)),
                    scales=list (tck=c(1, 0),x=list(cex=0.5),y= list(cex=0.5, at=tick)),
                    main=list("Conditional
                                              weighted
                                                              residuals
                                                                                           population
                                                                              versus
predictions", cex=0.55, just=c(0.43,4))
      pl4 <-
                   xyplot(CWRES~TAD, data=tab,
                    vlim=reslim.
                    panel=pan3,
                    type="p", cex=0.5,
                    vlab=list("CWRES", cex=0.5, just=c(0.7.3)).
                    xlab=list("Time after dose", cex=0.7, just=c(0.5,-1)).
                    scales=list (tck=c(1, 0),x=list(cex=0.5),y= list(cex=0.5, at=tick)),
                    main=list("Conditional weighted residuals versus time", cex=0.55, just=c(0.43,4))
} else {
       cat (paste ("The table file ",tab_file," was not found. Please check your output tables.\nStopping R execution."))
      quit()
```

```
pdf (file = fname)
   print (pl3, position=c(0,-0.01, 0.52,0.51), more=TRUE)
   print (pl4, position=c(0.48, -0.01, 1, 0.51)
                                                , more=TRUE)
   print (pl1, position=c(0,0.45,
                                    0.52, 0.99
                                                , more=TRUE)
    print (pl2, position=c(0.48,0.45, 1,0.99)
                                                , more=TRUE)
    grid.text(header,y = 0.98)
    dev.off()
    # open created file
    cat (paste ("OUTPUT: ", fname, sep=""))
   if (file.exists(fname) && open_res) {
       if (Sys.info() ['sysname'] == 'Windows') { shell.exec(paste(getwd(),"/",fname,sep="")) } # windows
       else if (Sys.info() ['sysname'] == 'Darwin') { system(paste ("open ",fname, sep="")) } # mac
      else { system(paste("xdg-open ", fname, sep=""), ignore.stdout=TRUE, ignore.stderr=TRUE, wait=FALSE) }
# linux
quit()
```

2. R script for visual predictive check (VPC)

```
if(Sys.getenv("R_LIB") != "") { .libPaths(c(Sys.getenv("R_LIB"), .libPaths())) }
library (xpose4)
library (ggplot2)
library (reshape)
models <- list (
 "run1045" = list (
   modelfile
               = "run1045.ctl".
   description
               = "RE SLCO~all".
   reference model = "run1027".
   working_dir
             = "C:/NM/phd/run",
   data_file
               = "..//PopPK_phd_2.csv",
              = "run1045.lst",
   output_file
               = c ("sdtab1045", "patab1045", "cotab1045", "catab1045"),
   tables
               = list(
   estim
              = c(3.02E+01, 5.01E+01, 5.76E+01, 1.61E+01, 2.54E+01, 5.46E+01, 1.17E+00, 8.72E-01,
    th
2.36E-01, 1.62E-01, 1.00E-02, 2.00E-03, 1.00E-04, 2.25E-01, 2.20E+00, 5.95E-01, -1.86E-01, -7.46E-01,
-5.10E-01. -4.40E-01. 2.56E+00. 1.00E-04. 7.99E-01. -7.50E-01).
              = c(0.16, 0.014, 0, 0, 0.825, 0.526, 0, 0.259),
    om
```

```
= c(0.132,0.00466,0,0,0.434,0.391,0,0),
     om_se
                 om block
                           -0.0212, 0.014, 0, 0, 0, 0, 0, 0, 0
                            0, 0, 0, 0, 0, 0, 0, 0,
                            0, 0, 0, 0, 0, 0, 0, 0,
                            0, 0, 0, 0, 0.825, 0, 0, 0,
                            0, 0, 0, 0, 0, 0.526, 0, 0,
                            0, 0, 0, 0, 0, 0, 0, 0,
                            0, 0, 0, 0, 0, 0, 0, 0.259, ncol=8),
                  = c(0.132,0.00466,0,0,0.434,0.391,0,0),
     om_se
     si
                = c(1).
     si se
                 = c(0),
                = matrix(c(1), ncol=1),
     si_block
     si_se_block = matrix(c(0), ncol=1)
run_from <- list(software = "pirana", version = "2.9.7")
open_res <- 1
setwd('C:/NM/phd/run')
folder <- "vpc_run1045"
command <- readLines(paste(folder, "/command.txt", sep=""))
```

```
args <- strsplit (command, " ") [[1]] [-1]
for (i in seq(args)) { if(substr(args[i],0,1) !="-") { modfile \leftarrow args[i] } }
runno <- strsplit (modfile, "\\.") [[1]][1]
csv_file <- dir(folder, pattern="raw_result") [1]
fname <- paste("pirana_reports/vpc_", folder, ".pdf", sep="")</pre>
if (runno != "") {
  fname <- paste ("pirana_reports/",runno,"_vpc_", folder, ".pdf", sep="")
if (!file.exists("pirana_reports")) {dir.create ("pirana_reports")}
if (file.exists (paste ("pirana_reports/vpc_",folder,".pdf", sep=""))) {
    file.remove (paste ("pirana reports/vpc ".folder.".pdf", sep=""))
}
vpc.info <- paste(folder,"/vpc_results.csv", sep="")</pre>
vpctab <- paste(folder, "/", dir(folder, pattern = "^vpctab") [1], sep="")</pre>
if (!file.exists(vpctab)|!(file.exists(vpc.info))) {
  cat ("VPC output not found. The vpc in PsN probably failed.")
  quit()
tab <- read.vpctab(vpctab)
vpc <- read.npc.vpc.results(vpc.info)</pre>
vpc_tmp <- data.frame(vpc$result.tables[-length(vpc$result.tables)])
tab_dat <- tab@Data
```

```
## handle stratification (if present)
strat_id <- grep("lower", names(vpc_tmp))</pre>
if (length(strat_id) > 1) {
  strat_id <- c(grep("lower", names(vpc_tmp)), (length(vpc_tmp[1,])+1))
  vpc_dat < -c()
  for (i in 1: (length(strat_id)-1)) {
    tmp \leftarrow cbind(vpc\_tmp[,c(strat\_id[i]:(strat\_id[i+1]-1))], strata\_no = i)
    if (i>1) { colnames (tmp) \leftarrow colnames (vpc_dat) }
    vpc_dat <- rbind(vpc_dat, tmp)</pre>
  vpc_dat$strata_name <- vpc$strata.names[vpc_dat$strata_no]</pre>
  tab_dat$strata_name <- vpc$strata.names[tab_dat$strata]
  pl <- ggplot (tab_dat, aes(group=factor(as.character(strata_name)))) + facet_wrap ( ~ strata_name)
} else {
  vpc_dat <- vpc_tmp
  pl <- ggplot (tab_dat)
# enlarge limits to allow plotting of confidence intervals
xlim <- c(min(c(vpc_dat$lower, tab_dat$TIME), na.rm=TRUE), max(c(vpc_dat$upper, tab_dat$TIME),
na.rm=TRUE))
ylim <- c(min(c(vpc_dat[["X95.CI.for.5.from"]], tab_dat$DV), na.rm=TRUE), max(c(vpc_dat[["X95.CI.for.95.to"]],
tab_dat$DV), na.rm=TRUE))
```

```
# for unbinned VPCs
if (is.na(vpc_dat$lower[1])) {
  vpc_dat$lower <- vpc_dat$upper - min(diff(vpc_dat$upper))/2
  vpc_dat$upper <- vpc_dat$upper + min(diff(vpc_dat$upper))/2
 x\lim[2] < -x\lim[2] *1.02
# plot all layers
pl \leftarrow pl +
                                        aes(x=(upper+lower)/2.
                                                                           vmin=vpc dat[["X95.CI.for.5.from"]].
   geom_ribbon(data=vpc_dat,
ymax=vpc_dat[["X95.CI.for.5.to"]]), linetype=0, alpha=0.2, fill="#002288") +
                                                                         ymin=vpc_dat[["X95.CI.for.95.from"]].
    geom_ribbon(data=vpc_dat,
                                        aes(x=(upper+lower)/2.
vmax=vpc_dat[["X95.CI.for.95.to"]]), linetype=0, alpha=0.2, fill="#002288") +
    geom_ribbon(data=vpc_dat,
                                        aes(x=(upper+lower)/2,
                                                                         vmin=vpc_dat[["X95.CI.for.50.from"]].
vmax=vpc_dat[["X95.CI.for.50.to"]]), linetype=0, alpha=0.2, fill="#882222") +
    geom_point(data=tab_dat, aes(TAD, DV), colour="#444444") +
                                                                            vpc_dat[["X50.real"]]),
                                       (vpc_dat$upper+vpc_dat$lower)/2,
   geom_line(data=vpc_dat,
                               aes (
                                                                                                      size=1.5.
colour="#882222") +
   geom_line(data=vpc_dat,
                                        (vpc_dat$upper+vpc_dat$lower)/2.
                                                                              vpc_dat[["X5.real"]]).
                                                                                                       size = .5,
                               aes(
linetype="solid". colour="#222288") +
    geom line (data=vpc dat.
                                        (vpc_dat$upper+vpc_dat$lower)/2,
                                                                             vpc dat[["X95.real"]]).
                               aes(
                                                                                                       size = .5.
linetype="solid", colour="#222288") +
   xlab ("Time after dose (hr)") +
```

```
ylab ("Dependent variable") +
xlim (xlim) +
ylim (ylim)

pdf(file=fname, height=6, width=8)
print (pl)
dev.off()

## open created file
cat (paste("OUTPUT: ", fname, sep=""))
if (file.exists(fname) && open_res) {
    if (Sys.info() ['sysname'] == 'Windows') { shell.exec(paste(getwd(),"/",fname,sep="")) } # windows
    else if (Sys.info() ['sysname'] == 'Darwin') { system(paste ("open ",fname, sep="")) } # mac
    else { system(paste("xdg-open ", fname, sep=""), ignore.stdout=TRUE, ignore.stderr=TRUE, wait=FALSE) }
# linux
}

quit()
```

국문 초록

서론: Voriconazole 의 높은 약동학적 변동성은 주로 CYP2C19 의유전형에 의해 설명할 수 있지만, 아직 변동성에 기여하는 인자 중밝혀지지 않은 것들이 많다.

방법: 본 연구에서는 solute carrier organic anion transporter family member 2B1 (SLCO2B1)의 유전형이 voriconazole의 약동학에 미치는 영향을 탐색하고자 12 명의 건강한 CYP2C19 poor metabolizer를 대상으로 voriconazole 200 mg을 단회 정맥 또는 경구 투여하였다. 모델 기반 시뮬레이션을 통해 voriconazole 반복투여 시 SLCO2B1 유전형이 약동학에 미치는 영향을 탐색하였다. 추가로, CYP3A4 효소의 활성도가 voriconazole의 약동학에 미치는 영향도 탐색하였다.

결과: Voriconazole 경구 투여 시 SLCO2B1 c.*396T>C variant type 대상자들은 wild type 대상자들에 비해 voriconazole의 흡수가 감소하는 양상이 관찰되었다. 하지만, 본 연구에서 측정된 CYP3A 활성도 표지자들은 voriconazole의 대사와 유의한 상관관계를 보이지 못했다. 모델 기반 시뮬레이션 결과, SLCO2B1 유전형은 voriconazole 표준 용법 반복 투여 시 치료적 범위에 도달하는 확률에 유의한 영향을 미치지 않았다.

결론: 본 연구의 결과는 SLCO2B1 c.*396T>C 이 장내 OATP2B1의 기능 저하와 연관되었을 가능성과 voriconazole의 개인간 약동학적 변이에 기여할 수 있음을 시사한다. 추가적인 연구를 통해 이러한 결과의 임상적 유의성을 확인할 필요가 있다.

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주요어: 보리코나졸, 약물 수송체, SLCO2B1, 약동학, 약물유전체학

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