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A Neonatal Amikacin Covariate Model Can Be Used to Predict Ontogeny of Other Drugs Eliminated Through Glomerular Filtration in Neonates

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ABSTRACT

Purpose Recently, a covariate model characterizing developmental changes in clearance of amikacin in neonates has been developed using birth bodyweight and postnatal age. The aim of this study was to evaluate whether this covariate model can be used to predict maturation in clearance of other renally excreted drugs.

Methods Five different neonatal datasets were available on netilmicin, vancomycin, tobramycin and gentamicin. The extensively validated covariate model for amikacin clearance was used to predict clearance of these drugs. In addition, independent reference models were developed based on a systematic covariate analysis.

Results The descriptive and predictive properties of the models developed using the amikacin covariate model were good, and fairly similar to the independent reference models (goodness-of-fit plots, NPDE). Moreover, similar clearance values were obtained for both approaches. Finally, the same covariates as in the covariate model of amikacin, i.e. birth bodyweight and postnatal age, were identified on clearance in the independent reference models.

Conclusions This study shows that pediatric covariate models may contain physiological information since information derived from one drug can be used to describe other drugs. This semi-physiological approach may be used to optimize sparse data analysis and to derive individualized dosing algorithms for drugs in children.

KEY WORDS developmental changes · extrapolation · glomerular filtration rate · neonates · pharmacokinetics

ABBREVIATIONS

bBW	Birth bodyweight
cBW	Current bodyweight
GFR	Glomerular filtration rate
NPDE	Normalized prediction distribution error method
PD	Pharmacodynamics
PK	Pharmacokinetics
PNA	Postnatal age

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INTRODUCTION

Although regulations like the Pediatric Rule (FDA) and the Pediatric Regulation (EMA), encourage pharmaceutical companies to perform research in the pediatric age range when new drugs are developed, to date, drugs in pediatrics are often administered in an off-label or unlicensed manner (1–3). Because of practical, ethical and economical reasons, it remains very challenging to perform pharmacokinetic and pharmacodynamic studies in the pediatric population with the ultimate aim to develop rational dosing regimens (4,5). One of the preferred approaches to facilitate the knowledge on the pharmacokinetics and pharmacodynamics in pediatrics is by applying population pharmacokinetic/pharmacodynamic (PK/PD) modeling (4,6–8). This approach is based on a simultaneous analysis of all data of the entire population while still taking into account that different observations are derived from different patients. Consequently, this population approach allows for the analysis of sparse and unbalanced data, which often applies to pediatric clinical studies. Moreover, the application of the population approach may lead, besides a reduction in invasiveness and burden for the patients, to considerably reduced costs.

However, to avoid that for each new or existing drug a systematic and time-consuming pharmacokinetic and/or pharmacodynamic analysis needs to be conducted (9,10), new approaches are required. One approach, which is gaining more attention in industry, academia and regulatory agencies, is to develop evidence-based dosing regimens in children by PK/PD modeling and simulation in which extrapolations are performed between populations that vary in age (10–12) (bridging). Another recently proposed approach is the use of information obtained from one drug for extrapolation to other drugs that are eliminated through the same route (13). This implicates that pediatric covariate models also contain biological system-specific information reflecting underlying physiological changes that can be used between drugs (13–15).

In a previous analysis, the developmental changes in amikacin clearance were characterized in more than 800 (pre)term neonates with varying gestational ages, birth bodyweights and postnatal ages, on the basis of birth bodyweight and postnatal age as covariates representing antenatal and postnatal maturation of the kidney, respectively (16). The aim of this study was to evaluate whether this internally and externally validated covariate model of amikacin in (pre)term neonates contains system-specific information on the developmental changes in glomerular filtration and that therefore the covariate model can be extrapolated to other drugs eliminated through glomerular filtration. In this study the amikacin covariate model was primarily extrapolated to netilmicin, tobramycin, vancomycin and gentamicin,

drugs which were used as paradigm compounds as they are all almost entirely eliminated through GFR and with similar physicochemical properties compared to amikacin.

MATERIALS AND METHODS

Patients and Data

For this analysis, data of renally excreted antibiotics in neonates were obtained from 5 different (in part) previously published studies (17–21). Since the amikacin covariate model (16) which was based on data from 874 neonates varying in postnatal age between 1 and 30 days, was used to describe the data of the other renally excreted drugs, only neonates with a postnatal age until 30 days were included from these datasets. Besides trough and peak samples taken before and at 1 h after initiation of the dose, respectively, samples at varying time points were available in all datasets (17,19–21), except for the tobramycin dataset (18). An overview of the patient characteristics of the different datasets is given in Table I. The different datasets are discussed briefly here, while more details on the studies can be found in the original articles (17–21).

Amikacin (16)

A dataset of amikacin containing 2,186 concentrations from 874 (pre)term neonates (birth bodyweight (bBW) 385–4,650 g, postnatal age (PNA) 1–30 days) was used to obtain the amikacin covariate model. Patients were enrolled in the study when at least one peak and trough concentration was available for each patient.

Netilmicin (17)

This dataset contained 267 netilmicin concentrations, collected in 88 (pre)term neonates (bBW 470–3,000 g, PNA 3–30 days). Concentrations were taken at the administration of the third dose or after a change in dose or dosing interval.

Tobramycin (18)

Four-hundred and seventy (pre)term neonates (bBW 485–5,245 g, PNA 1–4 days) were included in this dataset of which only paired peak and trough concentrations were available (taken after and before the fourth dose) resulting in 940 tobramycin concentrations.

Vancomycin (19)

This dataset contained 689 vancomycin concentrations collected in 273 preterm neonates (bBW 385–2,550 g, PNA 1–

Table 1 Overview of the Patient Characteristics of the Model Developed for Amikacin Applied in the Models Using the Amikacin Covariate Model and of the Different Datasets Used in the Current Analysis as Basis for the Models Using the Amikacin Covariate Model and the Independent Reference Models

Dataset	Datasets used in this analysis					
	Amikacin dataset (16)	Netilmicin (17)	Tobramycin (18)	Vancomycin (19)	Gentamicin A (20)	Gentamicin B (21)
Number of patients	874	88	470	273	673	59
Gestational age (weeks)	32 (24–43)	28 (23–41)	32 (24–43)	29 (23–34)	34 (23–43)	29 (23–42)
Postmenstrual age (weeks)	33 (24–43)	30 (23–44)	32 (24–43)	30 (24–38)	36 (23–44)	30 (23–42)
Postnatal age (days)	2 (1–30)	15 (3–30)	2 (1–4)	14 (1–28)	3 (1–30)	6 (1–30)
Birth bodyweight (g)	1,750 (385–4,650)	1,000 (470–3,000)	1,530 (485–5,245)	1,140 (385–2,550)	2,350 (440–5,240)	1,279 (520–4,950)
Current bodyweight (g)	1,760 (385–4,760)	1,115 (470–3,592)	–	1,170 (415–2,630)	2,550 (440–5,420)	1,009 (480–5,315)
Co-administration of ibuprofen or indomethacin (n(%))	118 (13.5)	–	45 (9.6)	23 (8.4)	70 (10.4)	6 (10.2)

Values are expressed as median (range)

28 days). Concentrations were taken around the second or third dose infusion of vancomycin.

Gentamicin (20,21)

For this drug two different datasets were available.

The first dataset (Gentamicin A), was obtained after combining previously published data (20) with more recently obtained data, resulting in a total of 1,531 concentrations from 673 (pre)term neonates (bBW 440–5,240 g, PNA 1–30 days).

In the second dataset (Gentamicin B) (21), 796 gentamicin concentrations were available of 59 (pre)term neonates (bBW 520–4,950 g, PNA 1–30). In this study several concentrations taken at different time points (e.g. 15 min or 4–8 h after the end of the infusion), besides peak and trough, were available.

Pharmacokinetic Modeling

Model Development

Non-linear mixed effect modeling was used to analyze the pharmacokinetic data. The first-order conditional estimation method with interaction option was used in NONMEM 6.2. (ICON Development solutions, Hanover, MD, USA). The following tools were used to visualize and evaluate the model: S-Plus version 6.2.1 (Insightful software, Seattle, WA) with NM.SP.interface version 05.03.01 (© by LAP&P Consultants BV, Leiden, The Netherlands), PsN and R (version 2.10.1). To test the hypothesis of between-drug extrapolation of covariate models, two different population pharmacokinetic models were developed for each dataset (14): 1) Models using the amikacin covariate model (16) and 2) Independent reference models based on a systematic covariate analysis. More information on both approaches can be found below under Covariate model. Model development was performed in four different steps: (i) choice of the structural model, (ii) choice of the statistical sub-model, (iii) choice of the covariate model, (iv) model evaluation. Discrimination between models was based on different diagnostic tools (22). A difference in objective function value (OFV) of 3.9 points or more was considered as statistically significant ($p < 0.05$ based on X^2 distribution). Finally, the goodness-of-fit plots, the total number of parameters, visual improvement of individual plots, correlation matrix, confidence intervals of parameter estimates, ill-conditioning (23) and shrinkage (24) were assessed. The ill-conditioning was assessed by taking the ratio of the largest and smallest eigenvalue of the covariance matrix of the estimate from the NONMEM output.

Structural Model

For the structural model, both one-, two and three-compartment models were tested. A two compartment model parameterized in

terms of clearance (CL), inter-compartmental clearance (Q), volume of distribution of the central compartment (V1) and the peripheral compartment (V2) was found to best describe the different datasets for both the models using the amikacin covariate models as the reference models. Only for the reference model of tobramycin, a two compartment model could not be supported as only peak and trough samples were available. Therefore, a one compartment model was preferred for the tobramycin reference model. For some of the models no covariance step could be given or the bootstrap failed meaning that some of the models were possibly overparameterized. As a result these models were simplified by equalizing V2 to V1 or Q to CL or by estimating Q as a fraction of clearance. These assumptions did not influence the estimate of the parameters of primary focus (CL and V1) with changes in parameter estimates being less than 5%.

Statistical Submodel

The interindividual variability was tested assuming a log-normal distribution in an individual *i* (*post hoc* value) and is given by the following equation:

$$\theta_i = \theta_{TV} \times e^{\eta_i} \tag{1}$$

in which θ_{TV} is the typical value of the parameter and η_i is assumed to be a random variable with mean value zero and variance ω^2 . For the intra-individual variability and residual error (statistical submodel), proportional, additive and combination error models were tested. In this analysis, the interindividual variability was only estimated on clearance since the interindividual variability on the other parameters (V1, V2 and Q) could not be estimated and was therefore fixed to zero for all models. For the intra-individual variability and residual error a proportional error model (Eq. 2) was chosen for all the models:

$$Y_{ij} = C_{pred,ij} \times (1 + \varepsilon_{ij}) \tag{2}$$

where Y_{ij} is the *j*th observation in the *i*th individual, $C_{pred,ij}$ is the predicted concentration and ε_{ij} is a random variable from a normal distribution with a mean of zero and estimated variance of σ^2 .

Covariate Model

For each dataset two population pharmacokinetic models were developed as proposed in the analysis of Krekels *et al.* (14)

Models Using the Amikacin Covariate Model (16). In these models, the internally and externally validated covariate model for amikacin (16) (Fig. 1), was directly incorporated into the

pharmacokinetic model that was developed for each dataset. This implicates that birth bodyweight was implemented as a covariate on clearance using a power function with an exponent of 1.34 as well as postnatal age using a linear function with a slope of 0.213. In the original covariate model of amikacin, co-administration of ibuprofen was identified as a third covariate on clearance, causing a 16.2% decrease in clearance of amikacin. This decrease in clearance was also implemented in the current analysis when ibuprofen or indomethacin was co-administered. Although the decrease in glomerular filtration was reported to be more pronounced after the administration of indomethacin compared to ibuprofen (25), in this analysis the 16.2% decrease in clearance seen for ibuprofen was also applied for indomethacin.

Current bodyweight was implemented on volume of distribution using a power function with 0.919 as exponent. While the pediatric covariate model is considered to describe the developmental changes in clearance and volume of distribution, the population values of these parameters were still estimated by NONMEM since they are considered drug specific properties (14) (Eq. 3):

$$CL_i = CL_p \times \left(\frac{bBW}{bBW_{Median}}\right)^{1.34} \times \left(1 + \left(0.213 \times \frac{PNA}{PNA_{Median}}\right)\right) \times 0.838_{ibuprofen}$$

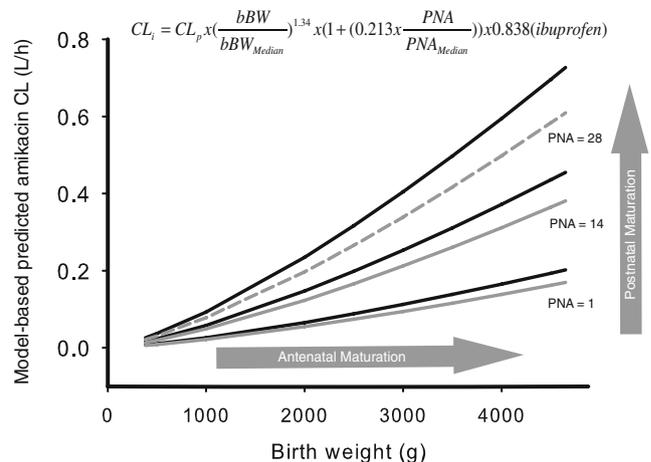
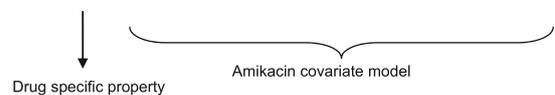


Fig. 1 Covariate model of amikacin (16) which was applied to the other renally excreted drugs. The figure illustrates the model-based predicted amikacin clearance (CL) values versus birth bodyweight (bBW) for postnatal age of 0, 14 and 28 days with (grey) and without (black) co-administration of ibuprofen. Birth bodyweight reflects the antenatal maturation of the kidney, postnatal age is reflecting the postnatal maturation. Reproduced from [De Cock RF, Allegaert K, Schreuder MF, *et al.* Maturation of the glomerular filtration rate in neonates, as reflected by amikacin clearance. Clin Pharmacokin 2012 Feb 1;51 (2): 105–17] with permission from Adis (© Springer International Publishing AG [2012]. All rights reserved.).

where CL_i represents the clearance in the i th individual, CL_p represents the population value of clearance and is estimated separately for each drug since it is considered to be a drug specific property, and the amikacin covariate model with birth bodyweight (bBW), postnatal age (PNA) and co-administration of ibuprofen is considered to describe the developmental changes in clearance through glomerular filtration.

Independent Reference Models (14). For these models a systematic covariate analysis (22) was performed in which the following covariates were tested for significance: birth bodyweight (weight at day of birth), current bodyweight (weight at day of blood sampling), gestational age, postmenstrual age, postnatal age, serum creatinine, co-administration of ibuprofen or indomethacin. Covariates were tested using a linear or power function. For serum creatinine, linear or power functions were tested in the denominator since a negative relationship was seen between serum creatinine concentrations and clearance. Previously, it has been shown that serum creatinine values in the first days of life are derived from the mother reflecting maternal renal function instead of neonatal renal function (26,27). Additionally, a progressive increase in serum creatinine concentrations has been reported with maximum serum creatinine concentrations at day 3–4 after birth followed by a subsequent decrease. This trend may be caused by differences in duration and extent of passive tubular back leak (28). As a consequence, serum creatinine values in the first 5 days of life were not taken into account in this analysis.

The significance of a covariate was statistically evaluated by the use of the objective function value. In the forward inclusion a p value <0.005 was considered as statistically significant while a more stringent p value <0.001 was used in the backward deletion. When two or more covariates were found to significantly improve the model, the covariate that reduces the objective function value the most was retained into the model and served as a basis for subsequent inclusion of additional covariates. In addition, the individual and population predicted parameters were plotted against the most predictive covariate to evaluate whether the individual predicted parameters were equally distributed around the population predicted parameters (22). Finally the covariate model was evaluated as mentioned previously under Model development, whereby the results of the Model validation were also considered.

Model Validation

The models using the amikacin covariate model as well as the independent reference models were internally validated using two different methods (22).

To evaluate parameter precision and stability a non stratified bootstrap analysis was performed in which 1,000 replicate

datasets of the same size as the original data analysis but with a different combination of individuals were generated. The parameter estimates obtained with the bootstrap were compared to the parameter estimates of the final models.

To evaluate the predictive properties of the models using the amikacin covariate model and reference models, the normalized prediction distribution error method (NPDE) was used, which is a Monte-Carlo simulation-based diagnostic in which the random effects were included (29,30) The dataset was simulated 1,000 times in NONMEM, each observed concentration was subsequently compared to the simulated reference distribution using the NPDE add-on package in R. A histogram of the NPDE distribution in the total dataset and plots of NPDE *versus* individual predicted concentrations and *versus* time were used to evaluate the final model.

Comparison of the Models Using the Amikacin Covariate Model and Independent Reference Models

The descriptive and predictive performance of the models using the amikacin covariate model and the independent reference models was compared by different diagnostic tools (14,22). The goodness-of-fit plots were compared to visually evaluate the descriptive performance. Secondly, individual and population clearance values obtained in the models using the amikacin covariate model were compared with the values obtained in the independent reference models (14). To evaluate the difference in clearance values more closely between both models, the population clearance values were plotted for both approaches *versus* birth bodyweight for PNA 1, 14 and 28 days. Furthermore, the individual and population predicted parameters were plotted against the most predictive covariate for both approaches to evaluate whether the individual predicted parameters were equally distributed around the population parameters (22). Additionally, the objective function values were evaluated as the models developed using both approaches are based on the same datasets. Finally, the results of the model validation (bootstrap analysis) as well as ill-conditioning and shrinkage were assessed. The predictive performance of the models using the amikacin covariate model and reference models was evaluated by comparison of the NPDE-results.

RESULTS

Pharmacokinetic Modeling

Models Using the Amikacin Covariate Model (16)

In these models, the previously published amikacin covariate model (Fig. 1) was directly incorporated in the pharmacokinetic models of the different drugs. The parameter estimates obtained for the models using the amikacin

Table II Final Parameter Estimates and Their Corresponding Coefficients of Variation (CV%) of the Model Developed for Amikacin Applied in the Models Using the Amikacin Covariate Model and of the Models Derived in the Current Study Using the Amikacin Covariate Model for Netilmicin, Tobramycin, Vancomycin, Gentamicin Dataset A and Gentamicin Dataset B

Parameter	Amikacin (16)	Netilmicin (17)	Tobramycin (18)	Vancomycin (19)	Gentamicin A (20)	Gentamicin B (21)
Objective function value	7738.145	278.771	970.81	2763.631	1824.456	570.064
Fixed effects						
CL _p in CL = CL _p × (bBW/median) ^m × (1 + n × (PNA/median)) × o (ibuprofen)	0.049 (2.21)	0.051 (5.22)	0.062 (2.06)	0.053 (2.74)	0.049 (1.47)	0.047 (3.12)
m	1.34 (2.04)	1.34	1.34	1.34	1.34	1.34
n	0.213 (9.81)	0.213	0.213	0.213	0.213	0.213
o	0.838 (3.88)	–	0.838	0.838	0.838	0.838
V _p in V ₁ = V _p × (cBW/median) ^p	0.833 (1.34)	0.995 (7.64)	1.03 (1.42)	0.913 (2.69)	0.762 (1.9)	0.731 (3.35)
p	0.919 (2.46)	0.919	0.919	0.919	0.919	0.919
Q = r × CL	0.415 (12.3)	–	–	0.904 (10.4)	–	1.47 (14.3)
Q = CL	–	Q = CL	Q = CL	–	Q = CL	–
V ₂ = V ₁	V ₂ = V ₁	V ₂ = V ₁	V ₂ = V ₁	V ₂ = V ₁	V ₂ = V ₁	V ₂ = V ₁
Interindividual variability						
ω ² (CL)	0.0899 (14.9)	0.186 (32.4)	0.15 (10.7)	0.11 (12.1)	0.106 (13.4)	0.0536 (21.8)
Residual variability						
σ ² (proportional)	0.0614 (8.2)	0.117 (18.2)	0.044 (9.62)	0.095 (7.89)	0.0804 (9.37)	0.0483 (12.1)
σ ² (additive)	0.267 (27.2)	–	–	–	–	–

CL_p population value for clearance, V_p population value for volume of distribution of the central compartment, bBW bodyweight at birth, cBW current bodyweight, PNA postnatal age, Q intercompartmental clearance, V₂ volume of distribution of the peripheral compartment

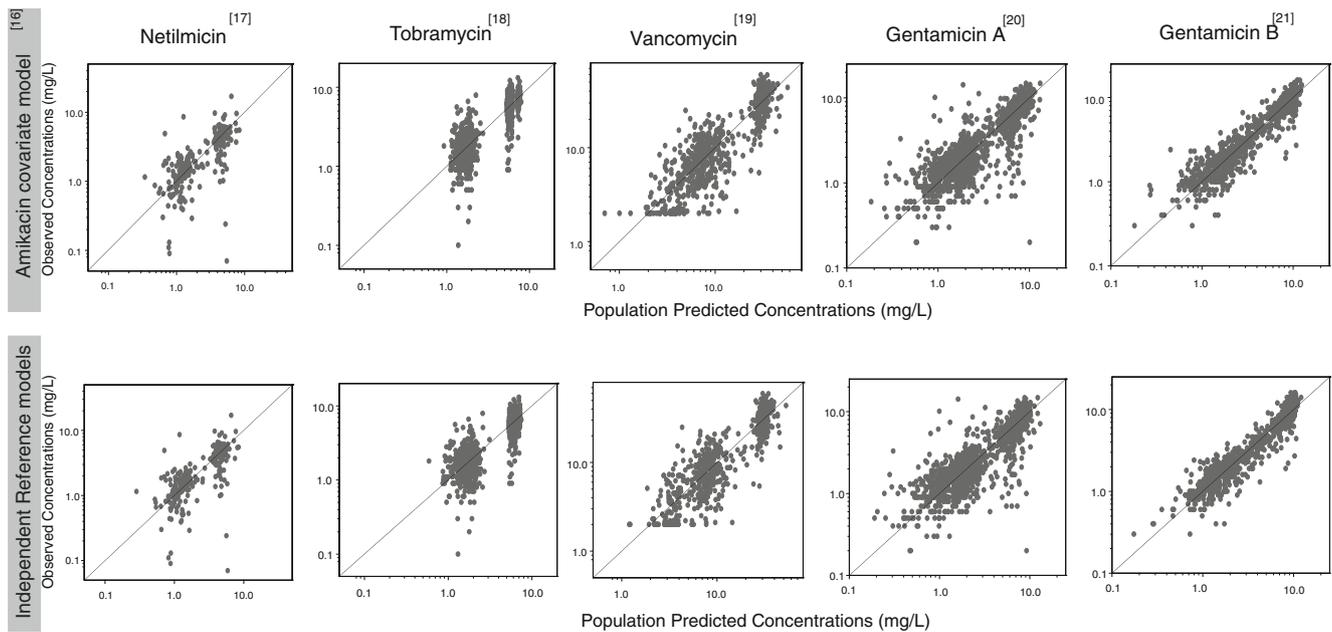


Fig. 2 Observed versus population predicted concentrations for the models using the amikacin covariate model (top) and the independent reference models (bottom) of netilmicin, vancomycin, tobramycin, gentamicin A and gentamicin B.

covariate model are shown in Table II together with the parameter estimates obtained in the final model for amikacin (16). As illustrated in Fig. 2 (top panels), the models using the amikacin covariate model described the observed concentrations without bias. The individual *post hoc* clearances and population predicted clearances versus the most predictive covariate (birth bodyweight) are given

in Fig. 3, showing that the population predicted clearance values are describing the individual *post hoc* clearances without bias. Furthermore, the results of the NPDE analysis in Fig. 4 show that the models can predict the median concentrations in the different datasets accurately. Finally, no trend was seen in the plots of the NPDE versus time and predicted concentrations (Fig. 4).

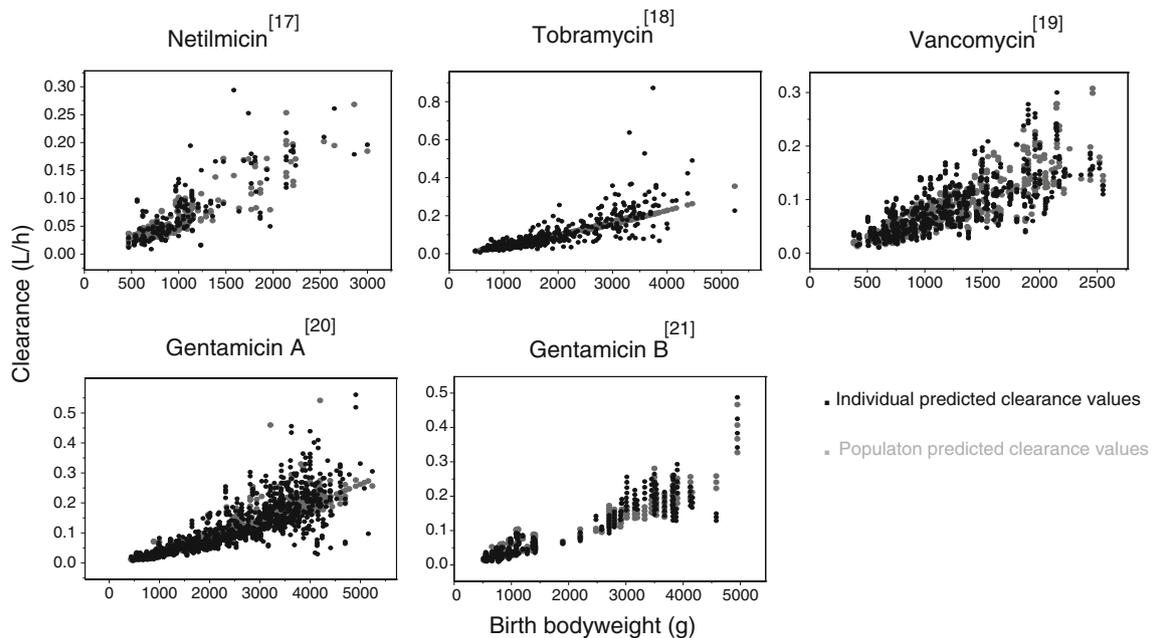
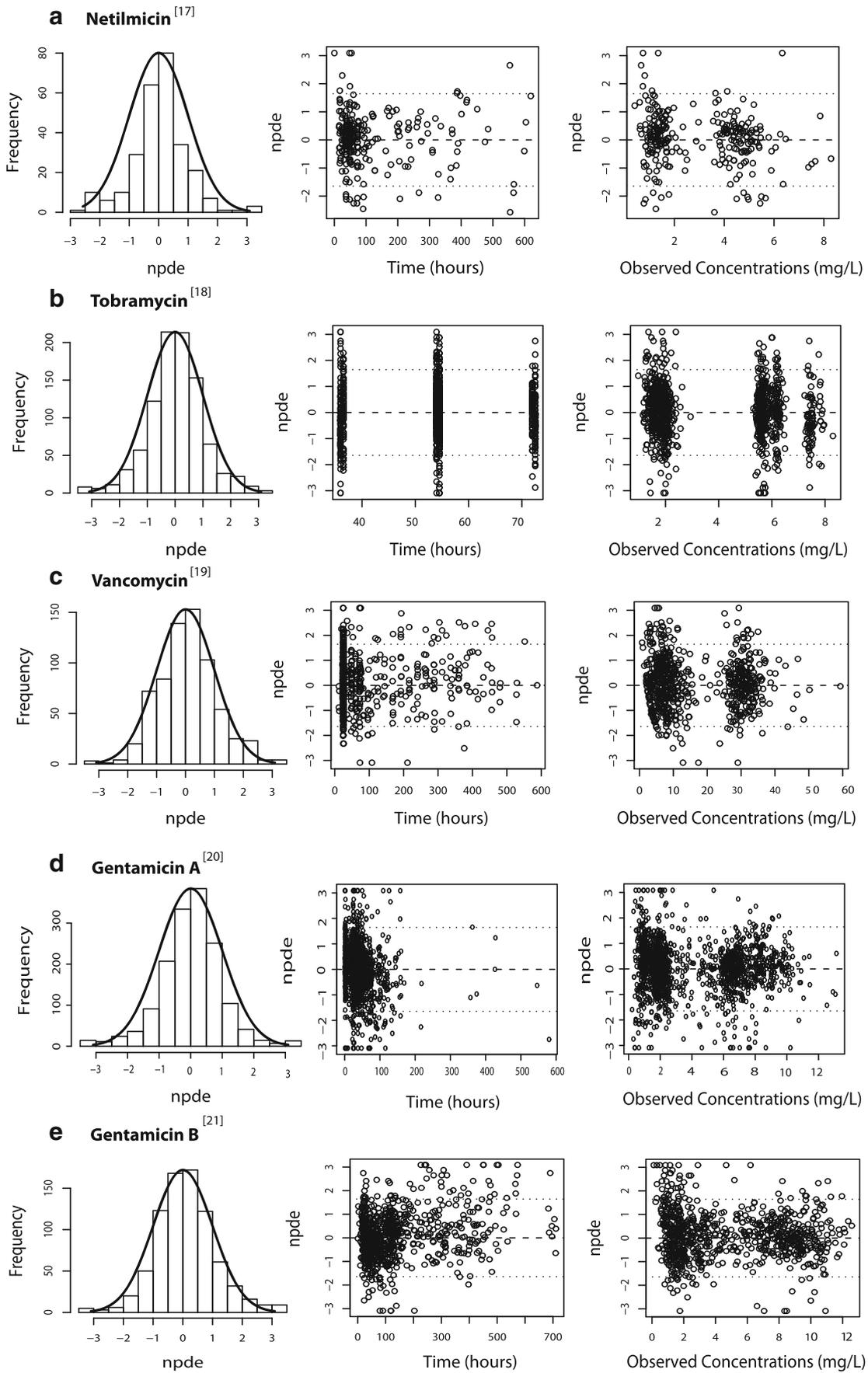


Fig. 3 Individual *post hoc* (black) and population predicted (grey) clearance values (L/h) versus the most predictive covariate, birth bodyweight (g), for the models using the amikacin covariate model.



◀ **Fig. 4** Results of the NPDE analysis for the models using the amikacin covariate model of (a) netilmicin, (b) tobramycin, (c) vancomycin, (d) gentamicin A and (e) gentamicin B. Left panel: Histograms of the NPDE distribution with the solid line representing a normal distribution as a reference, Middle panel: NPDE versus time (hours); Right panel: NPDE versus observed concentrations (mg/L).

Independent Reference Models

In the independent reference models of netilmicin, vancomycin and gentamicin datasets A and B, birth bodyweight and postnatal age were identified as the most important covariates to describe clearance. Current bodyweight was found as most important covariate to describe volume of distribution. Birth bodyweight and current bodyweight were implemented on clearance and volume of distribution of the central compartment, respectively, using a power function. Postnatal age was implemented using a power function (netilmicin, gentamicin A and gentamicin B datasets) or linear function (vancomycin dataset) depending on the dataset. For tobramycin, birth bodyweight was implemented on both clearance and volume of distribution using a power function. Based on the statistical criteria, postnatal age was not identified as a covariate on clearance. This may be explained by the fact that data of tobramycin were only available for the first 4 days after birth. In Fig. 2 (bottom panels) the observed *versus* population predicted concentrations are illustrated for the independent reference models. In Table III, the different parameter estimates are given for the reference models of the 5 different neonatal datasets. In the various independent reference models, serum creatinine was not found as a covariate to describe clearance. Furthermore when plotting the individual and population predicted clearance values *versus* birth bodyweight, it was seen that the individual *post hoc* clearances were randomly scattered around the population predicted clearances (figures not shown). Finally the results of the NPDE analyzes showed that the independent reference models were able to adequately predict the median concentrations of the different datasets (figures not shown).

Comparison of the Models Using the Amikacin Covariate Model and Independent Reference Models

In Fig. 2, observed *versus* population predicted concentrations are shown for the models using the amikacin covariate model as well as the independent reference models. Visual examination of the plots shows that both the models using the amikacin covariate model as well as the independent reference models are able to predict the observed concentrations and that the difference in performance of the two approaches is negligible. In Fig. 5 the individual and population clearance values for

the models using the amikacin covariate model are plotted *versus* those of the independent reference models of the different datasets. While both approaches estimate similar individual and population clearance values for netilmicin, tobramycin, gentamicin A and gentamicin B, a slight difference in population clearance values is seen for vancomycin, a drug with slightly different physicochemical and pharmacokinetic drug properties compared to amikacin and the other drugs. In Fig. 6, the population clearance values obtained using both approaches are plotted *versus* birth bodyweight for PNA 1, 14 and 28 days. To obtain the clearance values for the models using the amikacin covariate model, the full study range of the amikacin dataset was used while for the independent reference models, the study range available for that particular dataset was applied, explaining the differences seen in the length of both lines illustrating the population clearance values using both approaches. Based on this figure, it was concluded that at day of birth (day 1) and 14 days similar clearance values are obtained for both approaches for all drugs, while at day 28, a slight difference is seen for vancomycin and gentamicin B. For tobramycin it should be noted that no population clearance values are illustrated following the independent reference model for day 14 and 28 since this model is based on the original dataset which only included data during the first 4 days after birth. When plotting the individual and population predicted parameters against the most predictive covariate for both approaches, it was observed that the individual predicted parameters were equally distributed around the population parameters. Finally, when considering the differences in objective function values between the models using the amikacin covariate model and the reference models (Tables II and III), it was seen that the reference models of netilmicin, vancomycin, and gentamicin A and B had a lower objective function value (Δ objective function value: netilmicin 5 points, vancomycin 23 points, gentamicin A 67 points, gentamicin B 59 points) as compared to the models using the amikacin covariate model. For tobramycin, the objective function value of the reference model was 43 points higher compared to the model using the amikacin covariate model, which can be explained by the use of a one compartment reference model *versus* a two compartment model using the amikacin covariate model. Furthermore, Tables II and III show that the coefficients of variation of both fixed and random effects are well below 50% indicating that both approaches are able to estimate the parameters with high precision. Moreover, no ill-conditioning was detected in the models using both approaches since the condition number of the final pharmacokinetic models (range 2.23–64.44) was far below the critical value of 1,000. Finally, η -shrinkage expressed as a percentage was identified to be below 20% for all final pharmacokinetic models using both approaches.

Results of the bootstrap analysis showed that the median estimated values based on re-sampled data were close (<20%)

Table III Final Parameter Estimates and Their Corresponding Coefficients of Variation (CV%) of the Independent Reference Models Derived in the Current Study for Netilmicin, Tobramycin, Vancomycin, Gentamicin Dataset A and Gentamicin Dataset B

	Netilmicin (17)	Tobramycin (18)	Vancomycin (19)	Gentamicin A (20)	Gentamicin B (21)
Parameter					
Objective function value	273.659	1013.684	2740.457	1757.241	511.263
Fixed effects					
CL _p in CL = CL _p × (bBW/median) ^m × (1 + n × (PNA/median))	–	0.067 (1.43)	0.038 (9.24)	–	–
m	–	1.31 (2.43)	1.1 (5.72)	–	–
n	–	–	0.955 (19.7)	–	–
CL _p in CL = CL _p × (bBW/median) ^o × (PNA/median) ^p	0.063 (5.92)	–	–	0.097 (1.5)	0.046 (4.04)
o	1.44 (7.57)	–	–	1.36 (2.12)	1.41 (4.42)
p	0.481 (18.5)	–	–	0.458 (8.78)	0.371 (8.36)
V _p in V _I = V _p × (cBW/median) ^q	0.65 (7.14)	0.926 (1.45)	0.618 (2.85)	1.07 (2.3)	0.508 (1.74)
q	1 (13.4)	0.859 (3.41)	0.952 (8.21)	0.807 (5.29)	0.848 (3.27)
Q = r × CL	–	–	–	–	0.688 (12.1)
Q = CL	Q = CL	–	Q = CL	Q = CL	–
V ₂ = s	–	–	–	–	0.846 (25.9)
V ₂ = V _I	V ₂ = V _I	–	V ₂ = V _I	V ₂ = V _I	–
Interindividual variability					
ω ² (CL)	0.188 (32.1)	0.065(10.4)	0.103 (12.5)	0.102 (13.4)	0.0357 (22.1)
Residual variability					
σ ² (proportional)	0.114 (17.9)	0.0439 (9.16)	0.0938 (7.64)	0.0776 (9.27)	0.0465 (8.65)

CL_p population value for clearance, V_p population value for volume of distribution of the central compartment, bBW bodyweight at birth, cBW current bodyweight, PNA postnatal age, Q intercompartmental clearance, V₂ volume of distribution of the peripheral compartment

to the estimated values of the final models using the amikacin covariate model and independent reference models. This suggests that the final models using the amikacin covariate model and the independent reference models are stable and that the estimated parameter values are precise.

Considering the predictive performance, both the models using the amikacin covariate model as well as the independent reference models perform similar since both approaches can accurately predict the overall median concentrations. In addition, there was no bias in the normalized prediction distribution errors over time, nor over the predicted concentration range.

DISCUSSION

To facilitate the development and availability of drugs in children and to avoid the development and validation of PK/PD models for each new or existing drug, new approaches are needed. Therefore the aim of the current study was to evaluate whether the internally and externally validated covariate model of amikacin in (pre)term neonates (16) can be extrapolated to other drugs eliminated through glomerular filtration in neonates. This implicates that pediatric covariate models also contain biological system-specific information reflecting underlying physiological changes (13–15). To test this hypothesis the

covariate model of amikacin was directly incorporated in the pharmacokinetic model for netilmicin, vancomycin, tobramycin, gentamicin A and gentamicin B, drugs that, like amikacin are almost entirely eliminated through glomerular filtration. Using this approach a distinction is being made between drug-specific and system-specific information as explained in the “Materials and Methods” section in which the pediatric covariate model is considered system-specific while the population values are considered to be drug-specific. Subsequently the descriptive and predictive performance of models using the amikacin covariate model was compared to the independent reference models in which the covariate model was identified using a systematic covariate analysis (14,22).

To extrapolate information from one drug to another a few requirements need to be met (14). First of all it is a prerequisite that the covariate models, which are assumed to contain system-specific information, are extensively validated. In this analysis the covariate model, developed to describe the pharmacokinetics of amikacin, was based on the analysis of 2,186 amikacin samples in 874 (pre)term neonates. The covariate model was both internally and externally validated (16). Furthermore, it is important that the covariate models which are extrapolated to other drugs, are based on a considerable number of samples from a large patient cohort with varying characteristics such as gestational age, birth bodyweight and

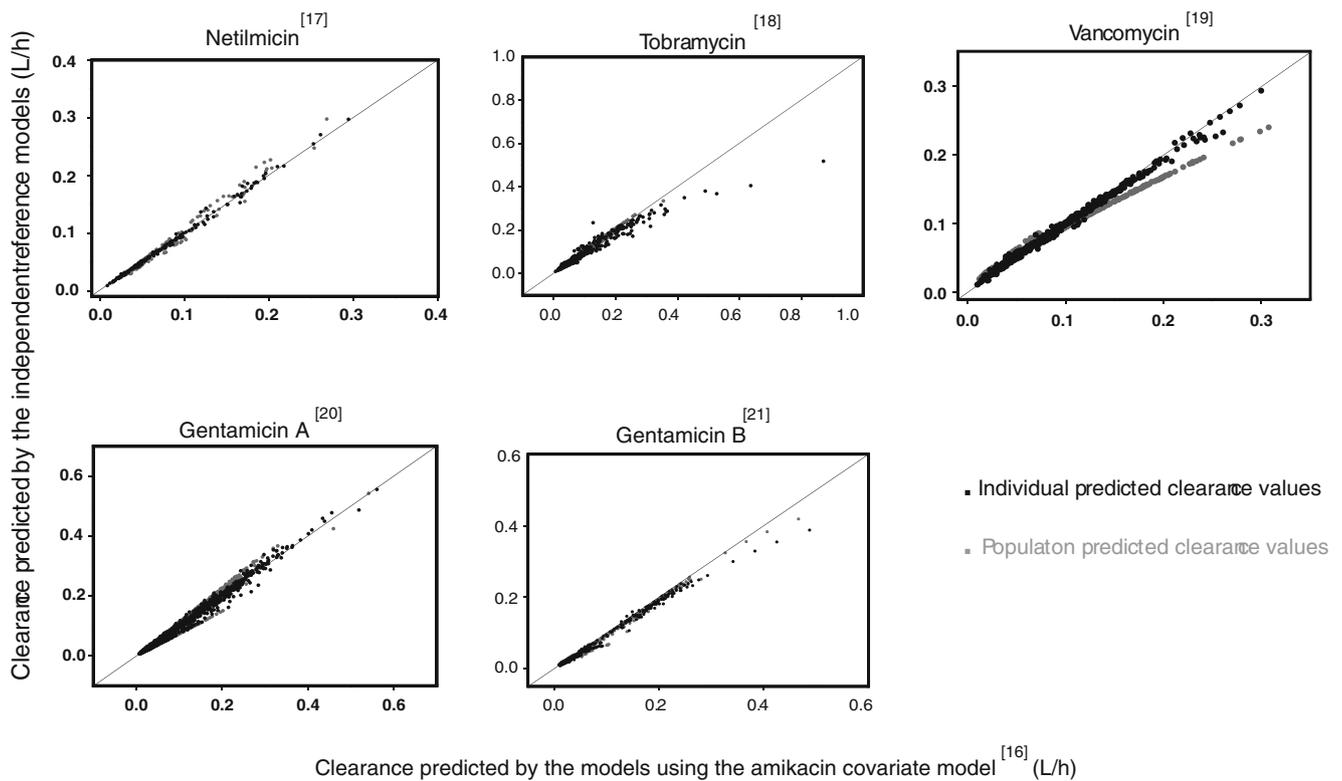


Fig. 5 Individual and population predicted clearances for the models using the amikacin covariate model versus the independent reference models for netilmicin, vancomycin, tobramycin, gentamicin dataset A and gentamicin dataset B.

postnatal age. In addition, it should be emphasized that the covariate models can only be extrapolated to populations with clinical characteristics that are within the studied range of the applied covariate model. In this analysis the amikacin covariate model developed for (pre)term neonates between 1 and 30 days was extrapolated to five other datasets in which the clinical characteristics are similar compared to the amikacin dataset considering bodyweight and age range (Table I). Finally, a similar disease status was seen between the patients used for development of amikacin covariate model and the patients collected for the analysis of the unstudied drugs since all patients were admitted to the neonatal intensive care unit. When all the mentioned requirements are fulfilled, model development of the unstudied drugs (netilmicin, vancomycin, tobramycin and gentamicin) may be based on an even more limited number of data which is by all means a major advantage in the design and sampling strategy of (pediatric) clinical trials since the number of patients and the burden for patients participating in the trial can be reduced. However a limited amount of data still needs to be available to estimate the population parameter values for each drug as these are considered to be drug-specific parameters (see “Covariate Model” section under “Materials and Methods”). When all the above mentioned requirements are fulfilled, an advantage of utmost importance is seen in the time required to develop and validate models using a covariate model which already has been extensively validated (weeks) compared to reference models (months).

The descriptive and predictive performance of the models using the amikacin covariate model was confirmed by Figs. 2, 3 and 4. This suggests that the covariate model of amikacin may contain system-specific information on the developmental changes in glomerular filtration. In an analysis of Krekels *et al.* (14), the same concept was applicable since it was illustrated that the covariate model for the glucuronidation of morphine in (pre)term neonates to children up to 3 years of age was able to describe the developmental changes in the glucuronidation of zidovudine in term neonates and infants. Although similar individual and population clearance values are predicted by both approaches as seen in Fig. 5, a slight difference is seen in population clearance values of vancomycin and gentamicin B between the reference model and the model using the amikacin covariate model for the higher population values for clearance. This is also reflected in Fig. 6 in which this slight difference was seen between the clearance values for vancomycin and gentamicin B following both approaches at day 28. It is however unknown which of the two approaches best reflects the true clearance of these drugs. For gentamicin B, it should be noted that this difference seen in the higher clearance values of both approaches is only based on a limited number of data. Considering vancomycin, it is possible that the population clearance values predicted by the reference model are slightly different because this model is only based on vancomycin data from preterm neonates (Table I). This limitation in information is less important in the model using the amikacin covariate model in

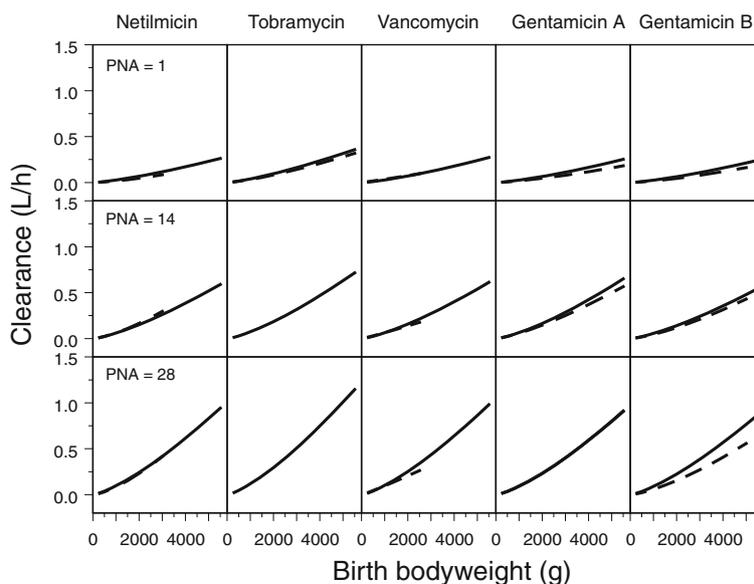


Fig. 6 Population clearance values for the models using the amikacin covariate model (black) and the independent reference models (dotted line) versus birth bodyweight for PNA 1 (above), 14 (middle) and 28 days (below) for netilmicin, vancomycin, tobramycin, gentamicin A and gentamicin B. For the clearance values obtained with the amikacin covariate model, the full study range of the amikacin dataset is used while for the independent reference models, the study range available for that particular dataset is used causing the differences seen in the length of both lines illustrating the population clearance values using both approaches. Moreover for tobramycin, no clearance values for to the independent reference model could be illustrated for PNA 14 and 28 days since no data were available.

which missing information is supplemented by information gained from the amikacin covariate model. The slight difference in population clearance values between the two approaches can also indicate that although vancomycin is mainly eliminated by glomerular filtration, the elimination of vancomycin may be modified by the presence of tubular processes (secretion or reabsorption) (31), which is not captured by the amikacin model. Finally, the difference may also be due to the different physicochemical properties of vancomycin compared to the other drugs, because in contrast to netilmicin, tobramycin and gentamicin, drugs that belong to the same class as amikacin, namely the aminoglycosides, vancomycin is a tricyclic glycopeptide. Besides the large difference in molecular mass of vancomycin (1449.3 g/mol) compared to amikacin (585.603 g/mol), netilmicin (475.58 g/mol), tobramycin (467.515 g/mol) and gentamicin (477.596 g/mol), the difference between the two drugs classes is also reflected in the protein binding. For the aminoglycosides the protein binding is below 10% in adults while this is much higher (approximately 55%) for the glycopeptide vancomycin. For antibiotics with a higher protein binding a lower renal clearance is often seen since only free drug is eliminated through the renal function (32).

In this analysis, the amikacin covariate model was in a first step extrapolated to drugs which are also almost entirely eliminated through GFR and with similar physicochemical properties compared to amikacin. However the majority of the drugs is eliminated by different elimination routes (hepatic and renal elimination). Therefore in a future analysis, the extension of the amikacin covariate model will be evaluated as well as the exact influence of differences in physicochemical and

pharmacokinetic drug properties on the extrapolation of the amikacin covariate model to other drugs which are eliminated by different routes. To analyze this, a future analysis needs to be performed as done by Krekels *et al.* (15). In that analysis the exact influence of differences in physiochemical properties on the extrapolation potential of the glucuronidation function was examined by using a physiological based (PBPK) modeling approach. Finally it will also be evaluated whether it is possible to characterize developmental changes in tubular processes in preterm and term based on the amikacin covariate model describing the developmental changes in GFR. A combination of all these different strategies (extrapolation to other drugs, adult data or non-clinical data) (33–35) will result in an approach focusing on the underlying system instead of focusing on the drugs and may facilitate development of pharmacokinetic models and evidence-based dosing regimens in the pediatric population.

CONCLUSIONS

In this study it was demonstrated that the descriptive and predictive performance of the models using the amikacin covariate model was similar to the independent reference models. This indicates that the use of system-specific information from one drug to other drugs may lead to optimization of sparse data analysis in children and that the covariate model, which in this case is describing the developmental changes in GFR, can be used to evaluate and optimize study and sampling design. As a consequence, the covariate model may play an important role

to determine first-in-child dosing strategies and evidence-based dosing regimens of new and existing drugs.

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