Exploring and explaining variability in tamoxifen and endoxifen pharmacokinetics in breast cancer patients: A pooled analysis

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Background and Objectives

To improve tamoxifen treatment, it is crucial to better understand the complex pharmacokinetics (PK) of tamoxifen and its major metabolite endoxifen, which is influenced by many internal (e.g. CYP polymorphisms) and external factors (e.g. drug-drug interactions).

By combining data from different clinical studies we enriched the single database for analysis, i.e. increased the power to detect covariate relationships. This study aimed to explore and explain different levels of variability in tamoxifen and endoxifen PK.

		Table 1. Study characteristics of six pooled tamoxifen studies [1-6].									
Methods			1 dbr	udy 2	udy 3	udy 4	ldy 5	ldy 6	otal		
Data	abase		Sti	Sti	Sti	Sti	Sti	Sti	F	explain	
• P	lasma concentration	N _{patients}	247	128	40	7	15	31	468	covaria	
da	ata of tamoxifen and	N _{samples/pts}	≤4	1	≤9	≤18	≤18	≤27	-	Drua-dru	
e	ndoxifen from 468	N _{occasion}	≤4	1	1	≤3 100	≤3 ≂°	≤3 ∝=	-		
bi	reast cancer patients	20 mg QD , %pts	100	100	70	100	73	97	96		
w	ere pooled (Tab. 1):	40 mg QD , %pts	0	0	30	0	27	3	4		
	2 large studies [1 2]	Setting (%pts) 1: adjuvant		1 (100)	n.r. (5) 1 (65)	1 (100)	n.r. (33) 1 (67)	n.r. (3) 1 (97)	n.r. (2) 1 (43)	569% IN	
	sparse sampling	2: neo-adjuvant 3: metastatic	2 (25) 3 (75)	. (,	3 (30)	. (,	. ()	. (0.)	2 (13) 3 (42)	CL23/F	
$ \rightarrow$	4 smaller studies [3–	Age [years],	72	61	52	51	51	51	64	inhibitor	
	6]: rich sampling	median (range)	(48-95)	(41-80)	(25-70)	(29-60)	(38-65)	(27-68)	(25-95)	64 6% r	
• A	patient's CYP2D6	CYP2D6, %pts	0	Б	0	0	0	0	1	Table 2. Para	
pl	henotype was	gNM	82	69	83	14	47	87	76	Parameter	
pi	redicted from genotype	gIM	10	12	10	0	7	0	10		
a	ccording to CPIC	gPM n r	5	7 8	3	14 72	0 47	10	6 7		
g	uidelines [7].	SSRI, %pts	2.0	3.9	2.5	0	80	0	4.9	CL20/F (L/h)	
		RIF , %pts	0	0	0	70	0	0	0.01	$L = \frac{V^2}{F} (L)$	
Mod	Modelling approach Selective serotonin reuptake inhibitors): <i>RIF</i> : concomitant use of rifempicin							oliser; <i>n.r.</i> : paroxetine	1) 12 k12 (1/h) t _{lag} (h)		
• A	ioint parent-metabolite	model includi	na kev	[,] covar	iate re	lations	hips ba	ased o	n prior	CL30/F (L/h)	
kr	nowledge was develope	d (Fia. 1) usin	a NON	MEM (v. 7.3).					CL20_RIF	
• To account for intoretudy variability (IGV) three implementation strategies were 0										CL20_SIDY	
in	investigated in NONMEM (a: as covariate, CL20_STDY; b: as variability nested into interindividual variability; c: as additional variability on CL20/E										
	a diatinguiah hatwaan ar					viohili.	, inte			CL20 AGE	
• To distinguish between and quantify different levels of variability –interstudy (ISV), CL23 interindividual (IIV) and interoccasion variability (IOV)– three models were IV CL											
CC	ompared:	Мс	odel 1:	Witho	ut IOV.	ISV ar	nd cova	riates.			
		N/ c				landy	vithout	oovorid	otoo		

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/ and ISV were partly ed by the investigated



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ates (Tab. 3):

ug interactions

↑↑: Rifampicin (potent 4 inducer; RIF) caused crease.

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 \checkmark : Fluoxetine and ine (Strong CYP2D6 rs; SSRI) responsible for eduction.

meter estimates of model 1-3. Model 1 Model 2

neter	Mod	el 1	Mode	el 2	Model 3		
OFV	-3471		-6511		-7087		
OFV to model 1	0		-3039		-3616		
	Estimate	RSE	Estimate	RSE	Estimate	RSE	
20/F (L/h)	5.98	2.60%	5.13	2.20%	5.49	3.90%	
/F (L)	285	13.1%	286	15.5%	282	15.1%	
23/F (L/h)	0.363	2.90%	0.361	3%	0.414	3.50%	
2 (1/h)	0.225	(FIX)	0.225	(FIX)	0.225	(FIX)	
(h)	0.295	(FIX)	0.295	(FIX)	0.295	(FIX)	
= (L/h)	71.1	(FIX)	71.1	(FIX)	71.1	(FIX)	
30/F (L/h)	5.10	(FIX)	5.10	(FIX)	5.10	(FIX)	
20_RIF					5.87	8.80%	
20_STDY			0.873	11.9%	0.569	17.2%	
23_PM					-0.675	4.30%	
23_IM					-0.433	16.9%	
23_UM					0.764	26.3%	
23_SSRI					-0.646	7.60%	
20 AGE					-0.461	28.4%	
23 RIF					1.25	38.4%	
CL20, CV	46.8%	5.20%	35.2%	7.30%	35.2%	6.20%	
CL23, CV	58.4%	3.80%	56.9%	4.30%	47.6%	7.20%	
/ CL20, CV			26.9%	14.5%	14.8%	20.6%	
V CL23. CV			25.8%	10.8%	17.0%	16.6%	
	30 6%	16 30/	15 6%	5 50%	15 7%	5 20%	



Figure 4. Predictive performance of Model 3. Predicted (boxes) and observed (dots) minimum concentrations at steadystate (Css) of tamoxifen (top) and endoxifen (bottom) dose-normalised and stratified by 10 subgroups. *Boxes:* simulations (n_{samples}=1000); dots: observations (Study 1-6). gNM (Ref.), gUM, gIM, gPM: genotype-predicted normal (reference, darkblue box), ultrarapid (orange box), intermediate (lightblue box) and poor metaboliser (red box); *RIFgX, SSRI_gX:* rifampicin or paroxetine/fluoxetine concomitant use by respective CYP2D6 phenotype.

CYP2D6 phenotype

CL23/F \downarrow : Poor and intermediate metabolisers showed 67.5% and 43.3% reduction, respectively (reference group: normal metabolisers). **CL23/F** ↑: ultrarapid metabolisers

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Although the database was heterogeneous and unbalanced (representing "real world data")



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Figure 3. Goodness of fit plots (a) of Model 1 and (b) Model 3. Obs: observations; IPRED: individual predictions; *PRED*: population predictions; numbers & colours: respective study, see Tab 1; unit: ng/mL.

- Nonlinear mixed-effects approach was able to differentiate and quantify several levels of variability;
- Influential factors were strong and reliably identified.
- However, **unexpectedly large differences between** the pooled **studies** were lacksquareobserved (remaining unexplained ISV):
 - \rightarrow Study design, bioanalytical methods or factors that had not been reported, such as adherence or tamoxifen formulation, might cause these differences [8].
- To avoid subtherapeutic concentrations we need to identify and control these factors by investigating high-quality clinical studies which will inform PK models and thus dose individualisation strategies such as model-based therapeutic drug monitoring.



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[1] P. Neven. CYPTAMBRUT-2: http://clinicaltrials.gov/show/NCT00965939. Accessed 20 Feb 2017. [2] P. Neven. CYPTAM-BRUT 3: http://clinicaltrials.gov/show/NCT00966043. Accessed 20 Feb 2017. [3] A.J.M. de Graan, S.F. Teunissen, F.Y.F.L. de Vos, et al. J. Clin. Oncol., 29.: 3240-6 (2011). [4] L. Binkhorst, T. van Gelder, W.J. Loos, et al. Clin. Pharmacol. Ther., 92.: 62–67 (2012). [5] L. Binkhorst, M. Bannink, P. de Bruijn, et al. Clin. Pharmacokinet., 55.: 249–255 (2016). [6] L. Binkhorst, J.S.L. Kloth, A.S. de Wit, et al. Breast Cancer Res. Treat., 152.: 119-28 (2015). [7] A. Gaedigk, K. Sangkuhl, M. Whirl-Carrillo, et al. Genet. Med., 19.: 69–76 (2017). [8] L. Klopp-Schulze et al. Clin. Pharmacokinet. [Epub ahead of print] (2017).



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