A Model- and Systems-Based Approach to Efficacy and Safety Questions Related to Generic Substitution Lesko, Lawrence J.¹; Biliouris, Konstantinos¹; Samant, Tanay S.¹; Combes, Francois P.¹; Fang, Lanyan²; Schmidt, Stephan¹; Trame, Mirjam N.¹ **College of Pharmacy** 1. Department of Pharmaceutics, University of Florida, Orlando, FL, United States, 2. Office of Generic Drugs, Food and Drug Administration, Silver Springs, MD, United States **UNIVERSITY** of FLORIDA

Introduction

- > The Office of Generic Drugs (OGD) at FDA occasionally receives reports of potential product failure after switching from brand to generic product resulting in failure of efficacy or an adverse event. It is difficult to interpret these reports and confirm them > Bioequivalence in PK between drug products is typically evaluated using bioequivalence (BE) criteria [90% Confidence Interval (CI) for Area Under the Curve (AUC) and maximal concentration (C_{max})] as a surrogate for therapeutic equivalence¹
- \succ The objective of our research is to develop a scientific platform that allows investigators at FDA (or, industry and academia) to conduct a 3-step mechanistic investigation to determine whether or not the aforementioned generic drug report is possible or not. This poster demonstrates the process of using the scientific platform for the anti-epileptic drug (AED), levetiracetam (LVT). LVT is available as both branded and generic products. This hypothetical example as there are no known bioequivalence or therapeutic equivalence issues with LVT

Methods

The 3-step approach consisted of : I) FAERS data mining and systems pharmacology, II) physiological based pharmacokinetics (PBPK) modeling and III) population PK/PD modeling

- Data mining and systems pharmacology to identify the frequency, nature and patient outcome of LVT Adverse Events (AE) using data from the FDA Adverse Event Reporting System (FEARS)².
 - 1. To elucidate the top 20 molecular targets and pathways (CYP enzymes, transporters and pharmacological receptors) of LVT using the Molecular Analysis of Adverse Events (MASE) software platform³
 - 2. To dissect the molecular link between AE and molecular targets/pathways for

II. PBPK Modeling

- 1. To develop and qualify a PBPK model for Immediate and extended release (IR and XR) LVT based on *in vitro* release kinetics and system properties
- 2. To perform a sensitivity analysis on formulations and drug-dependent quality attributes to identify parameters which might possibly cause bioinequivalence
- 3. To simulate a range of PK profiles using the PBPK model to determine the range of effects of quality attributes on 90% CI criteria for generic products

III. PK/PD Modeling

- 1. To estimate basic PK parameters for brand and generic formulations based on available literature and from PBPK modeling and simulation
- 2. To simulate and to compare PK profiles or their BE metrics (AUC or C_{max}) based on the different simulated PK profiles

Characteristics of LVT Relevant to PBPK and PK/PD modeling:

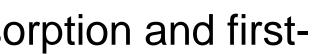
- \succ BCS class I drug: High solubility and high permeability
- \succ Rapid and complete absorption (T_{max} around 1 h)
- Oral administration: Half-life of 7 h
- > Dose: 500 5000 mg/day
- Steady state after 2 days
- \succ Therapeutic concentrations: 12-40 µg/mL
- IR and XR formulations available
- > Population PK model: One compartment with first-order absorption and firstorder elimination⁴

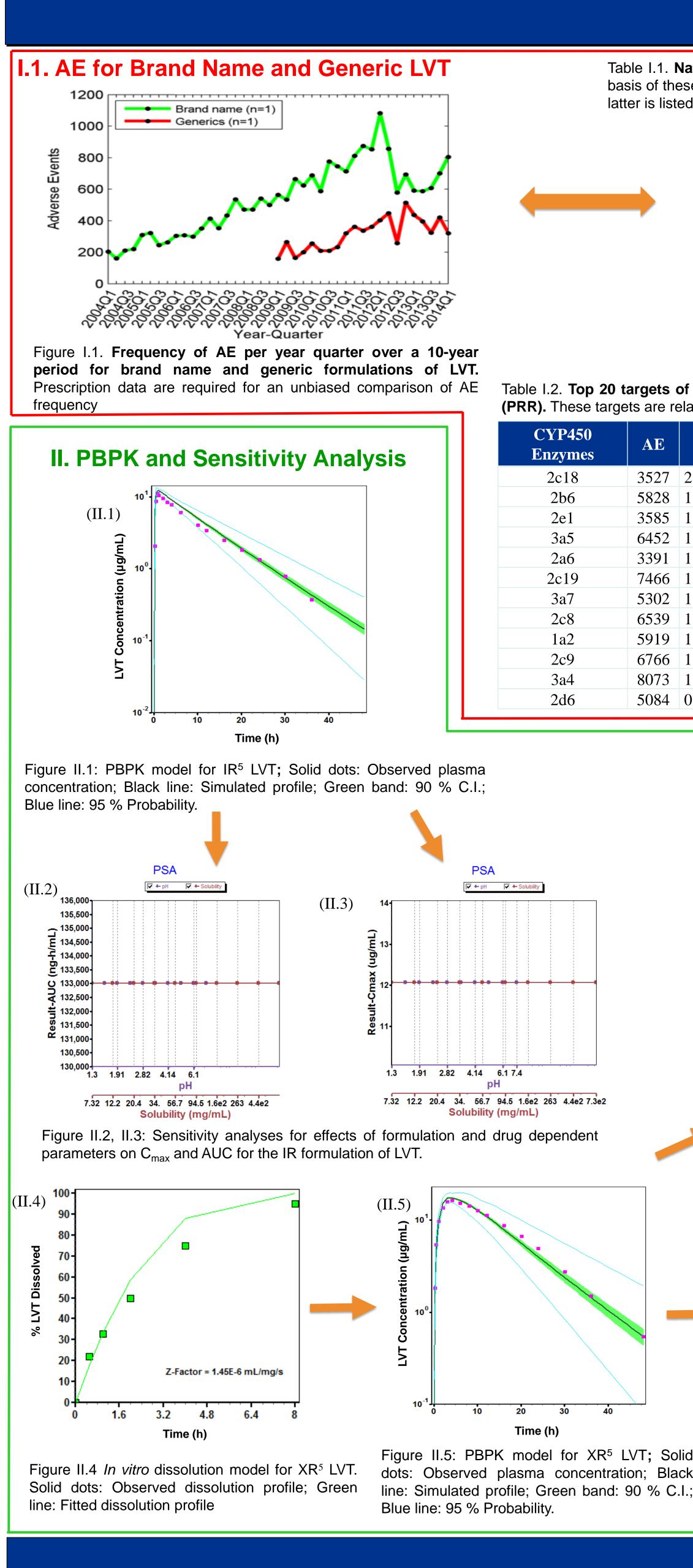
REFERENCES

- 1. Food and Drug Administration. Guidance for Industry Final guidance Statistical approaches to establishing bioequivalence
- 2. FAERS database: http://www.fda.gov/Drugs/
- MASE software: https://mase.molecularhealth.com Pigeolet E, et al, Clin Pharmacokinetics, 2007;46(6):503-512

5. Levetiracetam (Keppra XR), Clinical pharmacology/Biopharmaceutics review, FDA

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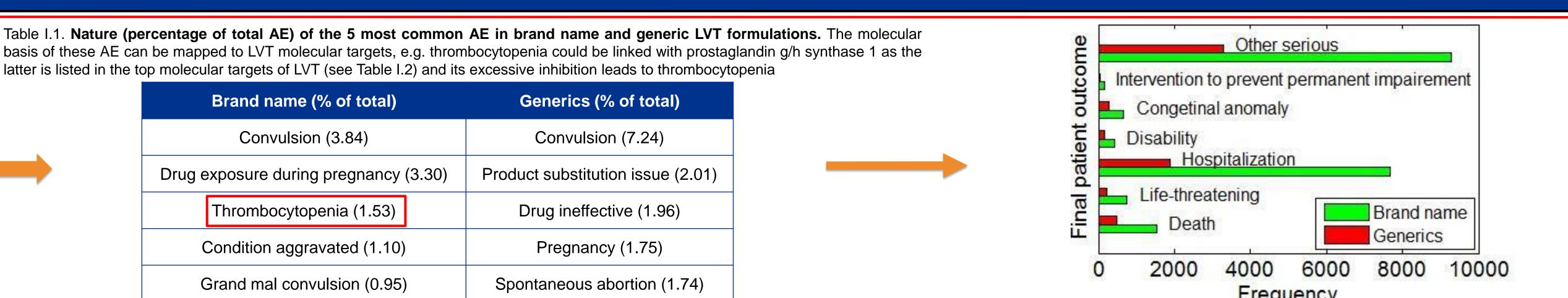
Data-mining & Systems Pharmacology:

 \succ Data mining in FAERS allowed for comparing the AE \succ frequency, AE nature and final patient outcome of brand vs generic formulations of LVT. However, prescription data (pending request) for each product is required for an unbiased comparison

> Systems pharmacology tools enabled mapping molecular targets of LVT to purported AE in FAERS

Results

latter is listed in the top molecular targets of LVT (see Table I.2) and its excessive inhibition leads to thrombocytopenia



I.2. Top 20 Molecular Targets of LVT

Table 1.2. Top 20 targets of LVT separated into CYP enzymes, transporters and other molecular targets and ranked according to proportional reporting ratio (PRR). These targets are related either to the PK effect (CYP enzymes, Transporters) or to the pharmacodynamics effect (Molecular targets)

CYP450 Enzymes	AE	PRR (CI PRR)	Transporters	AE	PRR (CI PRR)	Molecular Targets	AE	PRR (CI PRR)
2c18 2b6 2e1 3a5	3527 5828 3585 6452	2.67 (2.60-2.74) 1.90 (1.87-1.93) 1.59 (1.55-1.63) 1.56 (1.54-1.59)	Canalicular multispecific organic anion transporter 1	11404	6.17 (6.15-6.19)	Voltage-dependent n-type calcium channel subunit alpha 1b	11404	22.32 (22.19-22.46)
2a6 2c19 3a7	3391	1.52 (1.48-1.57) 1.51 (1.49-1.53)	Multidrug resistance protein 1 Serum albumin	11404 3908	2.00 (1.99-2.00) 1.49	Gamma-aminobutyric-acid receptor subunit alpha-1	3347	3.47 (3.37-3.57)
2c8 1a2	6539 5919	· · · · · · · · · · · · · · · · · · ·	Solute carrier family 22 member		(1.46-1.53)	Prostaglandin g/h synthase 1	3229	1.21 (1.17-1.24)
2c9 3a4 2d6	6766 8073 5084	1.19 (1.17-1.21) 1.14 (1.13-1.15) 0.96 (0.94-0.98)	6	5015	(1.28-1.35)	Synaptic vesicle glycoprotein 2a	11404	n/a

Virtual trial (1000

data from PBPK

analysis or from a

literature researc

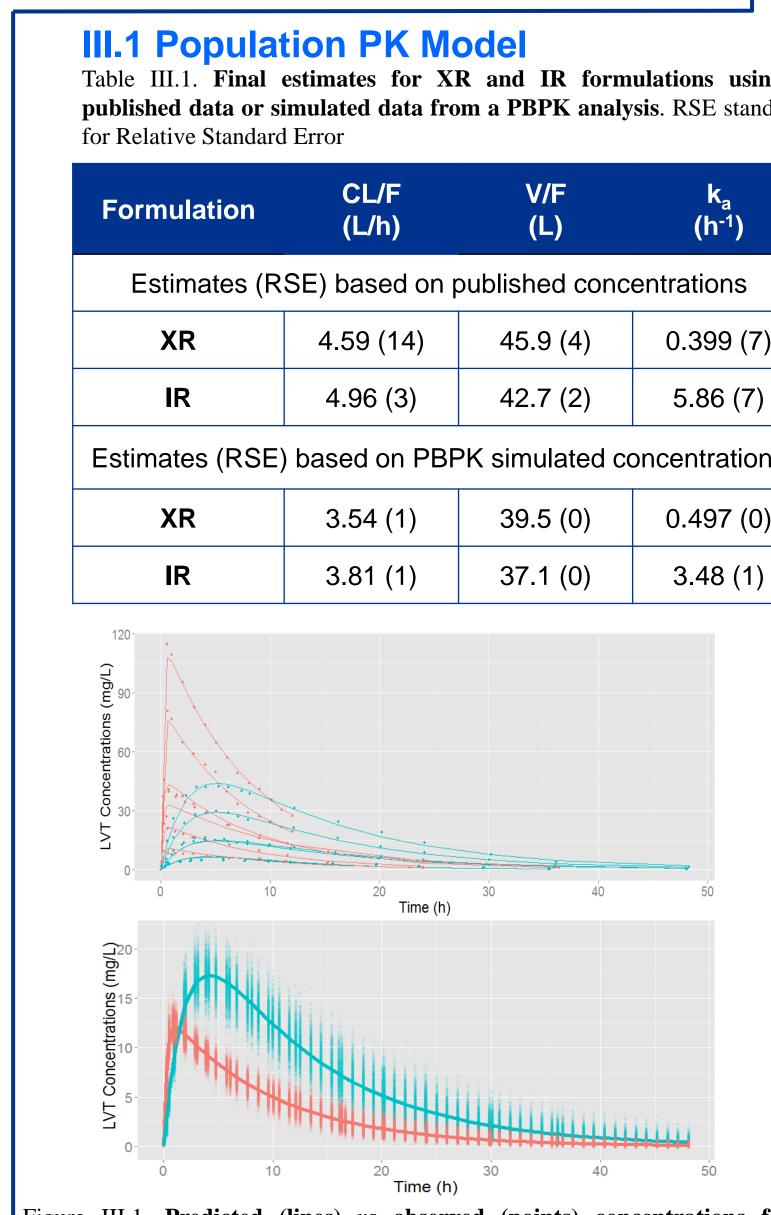
using Non-Linear

were analyzed

Mixed Effects

(NONMEM[®]7.2

subjects) simulated



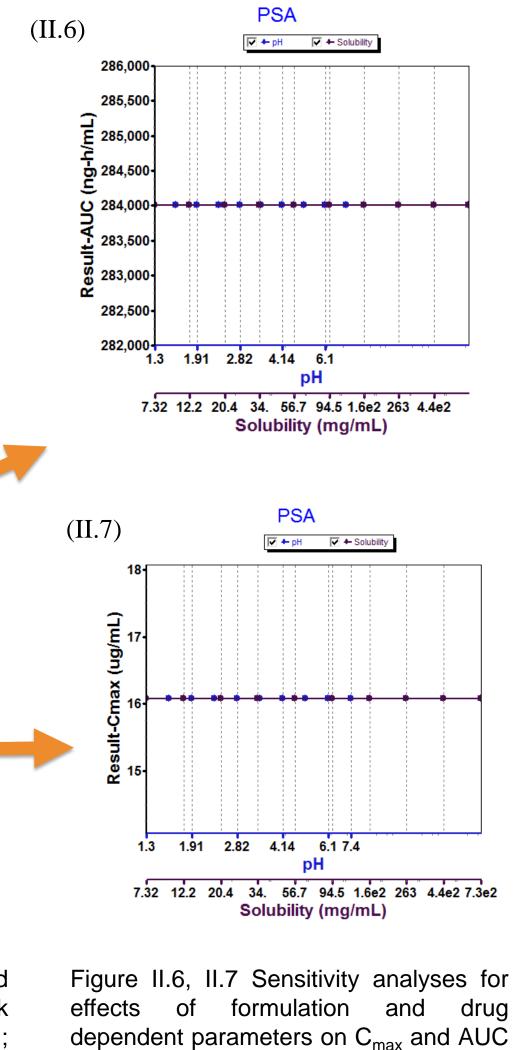
mulated switchability between IR (2 upper graphs), or XR (two lower graphs) formulations, when the subjects start the atment with the published (upper panel) or the PBPK (lower panel) Predicted (lines) vs observed formulation. Lines represent the median (black) and 9-95% percentiles ublished data (upper panel) and and PBPK simulated data (lower panel for both IR (red) and XR datas (blue). Several doses were used for published (green) of the predictions, and the upper and lower therapeutic data, when PBPK data were simulated for 500mg (IR) and 1000mg (XR) concentrations (Red)

Conclusion

II. PBPK

for the XR formulation of LVT

- Based on the sensitivity analysis, none of the selected \succ formulation or drug dependent parameters were likely to lead to bioinequivalence due to a significant change in the AUC or C_{max}
- The sensitivity analysis results can be used for hypothesis testing of BE and compared with probable therapeutic inequivalence between brand and generic LVT from FAERS data mining





Frequency

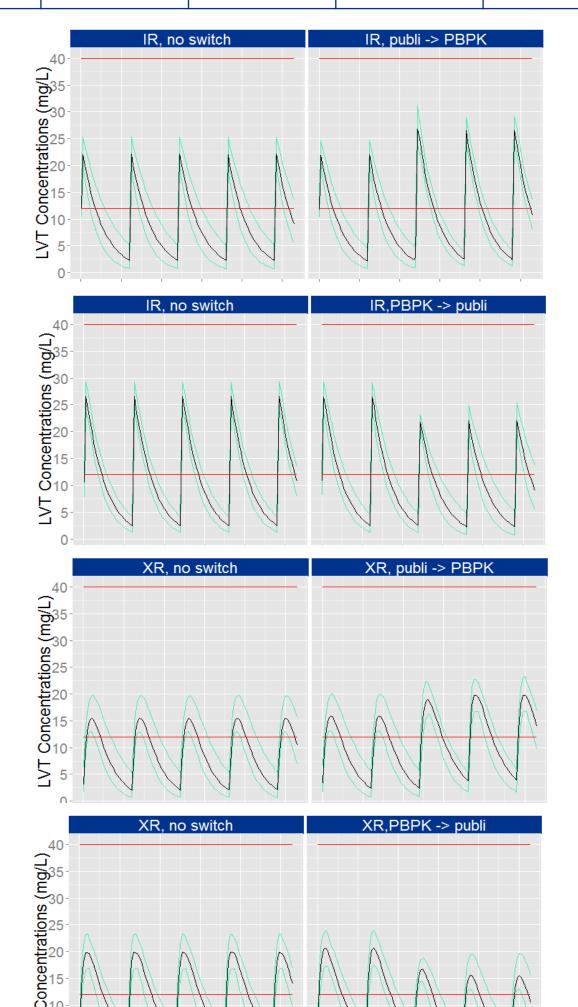
Figure I.2. Outcome of AE in brand name and generic LVT formulations. The majority o AE resulted in Serious outcomes or Hospitalization for both brand name and generics

'n	CL/F (L/h)	V/F (L)	k _a (h⁻¹)			
s (R	SE) based on p	oublished conc	entrations			
	4.59 (14)	45.9 (4)	0.399 (7)			
	4.96 (3)	42.7 (2)	5.86 (7)			
RSE) based on PBPK simulated concentrations						
	3.54 (1)	39.5 (0)	0.497 (0)			
	3.81 (1)	37.1 (0)	3.48 (1)			

III.2. Bioequivalence outcomes

Table III.2. PK metrics (AUC and C_{max}) along with ratio of the logarithmic mean of PBPK vs published computed using simulated BE trial. The four formulations from III.1 were used for simulation

Estimate	Published IR	PBPK IR	Published IR	PBPK IR
$\frac{\text{AUC}_{0 \to \infty}}{(\text{mg.h/L})} / F$	219.6	258.8	200.7	273.9
Ratio <i>vs</i> published	1	1.18	1	1.36
C _{max} (mg/L)	19.26	21.85	12.59	15.02
Ratio vs published	1	1.21	1	1.22



III. PK/PD

Estimation of PK parameters from PBPK simulated data gives results similar to those found in the literature or those estimated using published data

This methodology allows the simulation of PK (plus PD) of brand vs generic drug products to test the relative risk of a generic product being potentially bioinequivalent