

### Optimal Design To Estimate the Time Varying Receptor Occupancy Relationship in a PET Experiment

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# Why Study PK/RO?

- Positron emission tomography (PET) is one of the most effective imaging in vivo techniques to estimate RO
- The assessment of the RO-time profile is critical to predict the time course of pharmacological response



# Why Study PK/RO?

- To characterise the shape and location of the concentration-RO curve
- To explain variability in response
- To rationalise dose/dosing regimen selection
- To predict the time course of pharmacological response (therapeutic & toxic)
- To understand complex relations (tolerance, sensitisation)





# **Experimental design issues in a PET study**

- Cost and ethical reasons limit the total number of subjects (usually n < 20) and the number of PET scans per individual (≤ 3 scans)
- Uncertainty in the structure of the mechanistic model relating RO and PK (Equilibration delay, Mechanistic delay, Tolerance)
- Inter- and intra-subject variability in PK and in drug-toreceptor binding resulting in an overall inflation in variability
- Need to estimate typical exposure/RO link in a target patient population (fraction of subjects achieving an 'effective' RO in a chronic treatment)





#### **Time course RO – Mechanistic models**

- Typically a direct link between plasma concentrations and RO is assumed
- To account for delay/hysteresis between occupancy levels and plasma concentration the following models\* can be applied:

Effect Compartment Model

**Receptor Association-Dissociation Model** 

$$\frac{dRO}{dt} = k_{on} \cdot C_P \cdot (R_T - RO) - k_{off} \cdot RO$$

 $R_T$  = maximum receptor occupancy (fixed at 100%) RO = observed receptor occupancy time course data

Combined Biophase / Receptor Association-Dissociation Model



predict and confirm

\*Yassen et al,. Anesthesiology. 2006; 104-1232-42



#### **Optimal Population Design**

$$CP_{i} = f(\theta_{CP_{i}}, \xi_{CP_{i}}) + \varepsilon_{CP_{i}}$$
$$RO_{i} = g(CP_{i}, \theta_{RO_{i}}, \xi_{RO_{i}}) + \varepsilon_{RO_{i}}$$
$$\theta_{i} \in N(\Phi, \Omega)$$

Population PK model (fixed effect)

Population PK/RO model (fixed effect)

Population PK/RO parameters (random effect)

 $\Xi$  Population design vector of PK and RO sampling times ( $\xi$ )

 $\Psi$  Vector of population fixed ( $\Phi$ ) and random effect ( $\Omega$ ) parameters

For a given individual with design  $\xi$ 

 $M_F(\Psi,\xi) = E\left(-\frac{\partial^2 l(\Psi;y)}{\partial \Psi \partial \Psi^T}\right)$ 

**CDKMS** 

 $l(\Psi; y)$  is the log-likelihood of vector of observation y of individual for the population parameters

Population Fisher information matrix (PFIM):

$$M_F(\Psi, \Xi) = \sum_{i=1}^N M_F(\Psi, \xi_i)$$

Designs are optimized maximizing the determinant of the PFIM with respect to  $\Xi$  assuming known  $\Psi$ 

$$\det[M_F(\Psi, \Xi)]^{\dim(\Psi)}$$

Retout S, Duffull S, Mentre F. Comput. Meth. Prog. Biomed. 2001; Star

### Case study: Optimal study design

- PK/PD model was defined using pre-clinical data (kon/koff) and human PK data (considered fixed)
- Optimal design was performed to allocate PET scans at the most informative time-points
- Various designs exploring the influence of optimizing the PET scan time allocation, the number of subjects to elementary design and the number of dose levels were considered
- Optimisation was performed using WinPOPT<sup>®\*</sup>

predict and confirm

 Study designs: 16 subjects in total, up to 4 doses (1.5, 2.5, 4 and 6 mg), and 3 PET scans per subject





### Modelling time-course RO data

 A mechanistic model was developed using pre-clinical data to estimate the RO time-course and the time-varying link between PK and RO based on k<sub>on</sub> and k<sub>off</sub> parameters



2) Receptor Association-Dissociation Model

$$\frac{dRO}{dt} = k_{on} \cdot C_P \cdot (R_T - RO) - k_{off} \cdot RO$$

 $C_P$  = predicted plasma concentrations (input function)  $R_T$  = maximum receptor occupancy (fixed at 100%) RO = observed receptor occupancy time course data kon = association binding rate constant koff = dissociation binding rate constant.





#### **Preclinical data - Results**



<b>PK/RO</b> parameters estimetes					
Receptor	kon koff Ko		Kd=k <sub>off</sub> /k <sub>on</sub>		
	$(ml.ng^{-1}.h^{-1})$	$(h^{-1})$	(ng/ml)		
5-HT <sub>1A</sub>	0.088	0.221	2.5		
5-HT <sub>1B</sub>	0.088	0.183	2.1		





#### **Population PK/PD model**

 Optimization was performed on the PK/PD model based on preclinical data (kon=0.088 and koff=0.221, variability=30%) and human PK data (first order absorption with 2 comp. model)

Parameter (units)	Fixed effects	Between-subject variance	Population distribution
CL (L/h)	34.9 Fixed	0.112 Fixed	Log normal
VSS (L)	1200 Fixed	0.134 Fixed	Log normal
<b>Q</b> (L/h)	21.7 Fixed	0.179 Fixed	Log normal
FVC	0.67 Fixed	0 Fixed	Normal
<b>Ka</b> (h⁻¹)	0.605 Fixed	0.97 Fixed	Log normal
<b>Kon</b> (h <sup>-1</sup> )	0.088	0.1	Log normal
<b>Koff</b> (h⁻¹)	0.221	0.1	Log normal
σ²	0.1	-	Proportional





## **Optimality criteria improved trial efficiency**

Number of doses	Number of groups (n. groups x dose)	Sampling times	Efficiency (%)
4 (2.5, 4, 6, 1.5)	4 (1, 1, 1, 1)	{6, 48}; {6, 48}; {6, 48}; {6, 48} Empirical design (1)	50
4 (2.5, 4, 6, 1.5)	4 (1, 1, 1, 1)	{6, 48}; {6, 24}; {3, 36}; {6, 48} Empirical design (2)	100 (crit = 73.57)
4 (2.5, 4, 6, 1.5)	4 (1, 1, 1, 1)	{5.21 13.9}; {3.9 13.9}; {9.83 13.9}; {5.73 15.1}	558
2 (1.5,6)	8 (4, 4)	{5.73, 15.1} {5.73, 15.1} {5.73, 15.1} {5.73, 15.1}; {10.7, 15.1} {7.54, 11.7} {4.74, 9.01} {9.83, 13.9}	563
4 (1.5, 2.5 ,4 ,6)	8 (2, 2, 2, 2)	{5.73, 15.1} {5.73, 15.1}; {4.74, 13.9} {4.74, 13.9}; {3.90, 13.9} {10.7, 15.1}; {8.24, 12.7} {10.7, 15.1}	554
3 (2.5, 4, 6)	8 (3, 2, 3)	{4.74, 13.9} {4.74, 13.9} {4.74, 13.9}; {3.9, 13.9} {3.9, 13.9}; {9.83, 13.9} {9.83, 13.9} {9.83, 13.9}	549

• number of groups and number of doses appear not to be relevant

sampling time has been identified as the key driver to optimize the design





# Simulation approach to assess the design performance

- Simulate RO time course with <u>optimal</u> and <u>empirical</u> timepoints (~ Cmax and trough levels)
- Study design: 4 doses (1.5, 2.5, 4 and 6 mg); 4 subjects per dose (16 subjects in total)
- RO were sampled at two time-points
  - Empirical design: {6, 48} {6, 24} {3, 36} {6, 48}
  - Optimal design (at least 4 hours between two PET scans): {5.21 13.9} {3.9 13.9} {9.83 13.9} {5.73 15.1}
- Test the performance of the two approaches (bias, precision and accuracy)





### PK and RO profile (typical dose)

- PK/PD model defined using pre-clinical data (kon=0.088 and koff=0.221, variability=30%) and human PK data (first order absorption with 2 comp. model)
- N=1600 subjects were generated from the PK/RO model



#### **Simulation – Results**

- 100 trials (with 16 subject/trial) where generated using a Monte Carlo simulation approach
- A non-linear mixed-effect methodology was used to estimate the kon and koff parameters with their inter-individual variability
- Performances were measured as bias, precision and accuracy on kon and koff estimates





ME = mean error, Var = variance, SD = standard deviation, MSE = mean square error, SME = scaled mean error, CV = coefficient of variation

#### As expected the optimal study design approach provided more accurate and reliable model parameter estimates



### **Design of sequential PET experiments based on Optimality Criteria**







#### Conclusions

- A mechanistic model has been proposed to estimate the RO time-course and the time-varying link between PK and RO
- Optimal study design approach provided more accurate and reliable model parameter estimates
- The use of sequential model design could increase the efficiency of the study for the PK/RO assessment especially when the human receptor kinetic model cannot be properly inferred from animal estimates
- Further work to develop a standardized methodology for the use of sequential optimal design approach is in progress



