

Context

- Chronic hepatitis C virus (HCV) infection**
 - one of the most common causes of chronic liver disease
 - standard of care : weekly injections of pegylated interferon (PEG-IFN)+ daily oral ribavirin
- Mathematical modeling of HCV RNA (viral load) decay after treatment initiation**
 - complex system of non-linear Ordinary Differential Equations (ODE)
 - critical insights for the understanding of the virus pathogenesis
 - parameters crucial for early predicting treatment outcome
- Population designs evaluation and optimization for multiple response models**
 - methodology based on the Fisher Information matrix (M_F) [1]
 - implementation in PFIM 3.0 [2, 3]
- Optimization with PFIM 3.0**
 - D-optimality criterion ($\det(M_F)$)
 - Fedorov-Wynn algorithm (statistical design optimization)
 - optimization of the sampling times in a given set specified by users

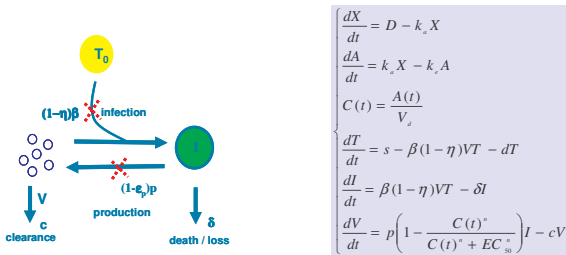
Objectives

- To show the relevance of PFIM with model described by ODE system
- To evaluate and optimize designs for the estimation of viral parameters

Viral dynamics modeling

- Viral dynamics model with the drug effectiveness under PEG-IFN**
 - Neumann et al. viral dynamics model [4]
 - target cells (I), infected cells (I) and free virus (V)

Figure 1. HCV cells infection dynamics with the ODE system



- pharmacokinetic model of concentrations of PEG-IFN [5]
 - first order absorption and elimination
 - D : dose of 180 μg of PEG-IFN by injection (weekly basis)
- no closed-form solution to this system
- as only concentrations of PEG-IFN and viral load are measured
 - some parameters are fixed [6] : $p=10$, $s=20000 \text{ mL}^{-1} \cdot \text{d}^{-1}$, $d=0.001 \text{ d}^{-1}$, $b=10^{-7} \text{ mL} \cdot \text{d}^{-1}$, $F=1$, $\eta=0$

Population model

- Population parameters
 - values of fixed effects
 - exponential model for random effects (CV = 50%)
 - additive error model for concentrations and \log_{10} (viral load)

EC_{50} ($\mu\text{g} \cdot \text{L}^{-1}$)	n	δ (d^{-1})	C (d^{-1})	k_x (d^{-1})	k_s (d^{-1})	V_s (L)
0.20	0.12	0.10	0.13	0.12	0.10	0.10

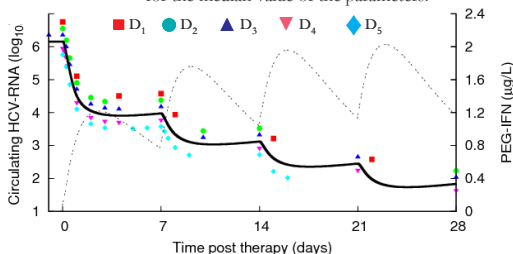
Population designs

- five popular designs of the literature [7, 8, 9, 10, 5]

Table 1. Design used in the five studies of viral dynamics.

Design	Reference	Measurement times (in days after first injection)	Number of samples
D ₁	Zeuzem (2005)	{0, 1, 4, 7, 8, 15, 22, 29}	8
D ₂	Sherman (2005)	{0, 0.25, 0.5, 1, 2, 3, 7, 10, 14, 28}	10
D ₃	Herrmann (2003)	{0, 0.25, 0.5, 1, 2, 3, 4, 7, 10, 14, 21, 28}	12
D ₄	Zeuzem (2001)	{0, 0.040, 0.080, 0.12, 0.20, 0.33, 1, 2, 3, 4, 7, 14, 21, 28}	14
D ₅	Talal (2006)	{0, 0.25, 0.5, 1, 2, 3, 5, 6, 7, 7.25, 7.5, 8, 9, 14, 15, 16}	16

Figure 2. Simulation of the viral dynamics (black) and PK profile (dashed) for the median value of the parameters.

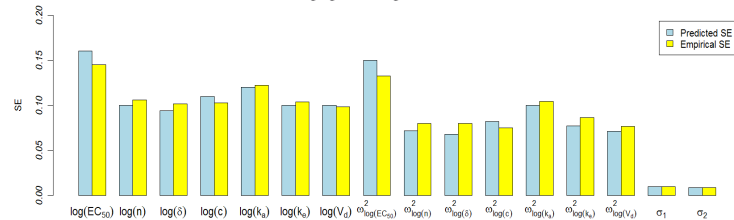


Methods

- Comparison between the standard error (SE) given by PFIM and the empirical SE**
 - simulation of 500 data sets (R software) using D3 design
 - estimation of the population parameters with MONOLIX (SAEM)
 - computation of the empirical SE defined as the standard deviation on the 1000 estimates of each parameter
- Evaluation of five designs [Table 1]**
 - 30 subjects
- Design optimisation using PFIM 3.0**
 - total number of samples : 240
 - potential sampling times are {D1-D5}
 - different number of sampling times per subject: 3, 4, 5, 6, 7

Results

- Figure 3. Barplot of the SE computed by PFIM and the empirical SE for all population parameters**



→ SE predicted by PFIM close to the empirical one

→ Relevance of PFIM for models described by ODE system

- Table 2. SE obtained for the fixed effects with the five designs included 30 patients**

Design	Number of sampling times per patient	$\log(EC_{50})$	$\log(n)$	$\log(\delta)$	$\log(c)$	$\log(k_x)$	$\log(k_s)$	$\log(V_s)$
D ₁	8	0.20	0.12	0.10	0.13	0.12	0.10	0.10
D ₂	10	0.16	0.10	0.095	0.11	0.13	0.11	0.10
D ₃	12	0.16	0.10	0.094	0.11	0.12	0.10	0.10
D ₄	14	0.16	0.11	0.094	0.10	0.12	0.10	0.10
D ₅	16	0.14	0.10	0.10	0.11	0.11	0.10	0.10

→ Similar SE for all pharmacokinetic (PK) parameters

→ D₅ can precisely estimate IFN effectiveness (EC_{50} & n)

- Table 3. Optimal designs for N = 240 sampling times per response according to the number of samples per patient**

Number of samples	N	Optimal Design (sampling times, n)	$\log(EC_{50})$	$\log(n)$	$\log(\delta)$	$\log(c)$	$\log(k_x)$	$\log(k_s)$	$\log(V_s)$	Information criterion
3	80	{(0.7, 9), 3 (0.10, 28), 11 (0.1, 28), 16 (0.4, 29), 19 (0.1, 4), 37}	0.21	0.12	0.081	0.096	0.14	0.11	0.084	193.2
4	60	{(0.1, 4, 28), 38 (0.1, 10, 28), 22}	0.17	0.090	0.070	0.090	0.12	0.083	0.08	230.2
5	48	{(0.1, 4, 16, 28), 14 (0.1, 7, 10, 29), 34}	0.14	0.061	0.057	0.075	0.087	0.068	0.061	224.0
6	40	{(0.1, 4, 7, 16, 28), 40}	0.15	0.095	0.084	0.10	0.11	0.095	0.090	208.3
7	34	{(0.040, 0.1, 4, 7, 9, 29), 4 (0.1, 4, 7, 9, 28, 29), 10 (0.040, 0.1, 4, 7, 16, 28), 20}	0.15	0.070	0.065	0.081	0.094	0.075	0.070	193.0

→ Best design with a number of samples per patient of four

→ Close SE as D₅ with a reduction by two of the number of samples

→ Importance of sampling times at four weeks

Conclusion

- Good approximation of M_F in PFIM for ODE systems
 - negligible computation burden to evaluate / optimise designs
- Total number of sampling measurements reduced by half with an appropriate design
- Design should not neglect long-term kinetics

References

- Bazzoli et al. *Statistics in Medicine*. 2009, [2] Bazzoli et al. *Computer Methods and Program in Biomedicine*. 2009, [3] www.pfim.biostat.fr, [4] Neuman et al. *Science*. 1998, [5] Talal et al. *Hepatology*. 2006, [6] Guedj et al. *Bulletin of Mathematical Biology*. 2007, [7] Zeuzem et al. *Journal of Hepatology*. 2005, [8] Sherman et al. *Gastroenterology*. 2005, [9] Herrmann et al. *Hepatology*. 2003, [10] Zeuzem et al. *Gastroenterology*. 2001.