



# Developing predictive *in silico* models for liver toxicity endpoints from *in vivo* histopathology data

Pinto-Gil K, Gregori-Puigjane E, Pasamontes-Funez I, Gómez-Tamayo JC, Stela B, Sanz F and Pastor M.

Research Programme on Biomedical Informatics (GRIB), FIMIM, Barcelona, Spain.

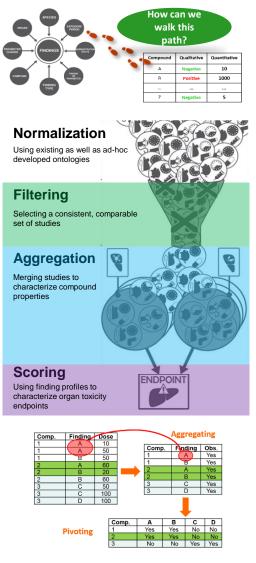
#### Facts & Figures

| Start date:                                       | 01/01/2010         |  |  |  |
|---|--------------------|--|--|--|
| End date:   | 31/12/2016         |  |  |  |
| Contributions                                     |                    |  |  |  |
| IMI funding:                                      | 6,910,018€         |  |  |  |
| EFPIA in kind:                                    | 10,157,590€        |  |  |  |
| Other:  | 1 719 500 €        |  |  |  |
| Total Cost:                                       | 18,877,109€        |  |  |  |
| Project website:                                  | www.etoxproject.eu |  |  |  |
| Social media: www.youtube.com/watch?v=fMWpDUaOdNc |                    |  |  |  |
|   |                    |  |  |  |

#### Challenge

- The eTOX project compiled a collection of nearly 9000 of such studies.
- In this work we describe how these collected data can be exploited for the development of predictive models for *in vivo* toxicity endpoints.

# Approach & Methodology

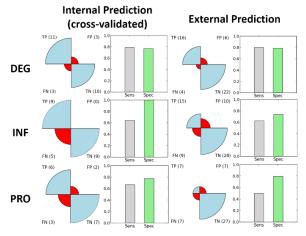


## Results

Distribution of qualitative hepatotoxicity scorings.

| Liver in vivo Endpoint                        | Positive<br>compounds | Negative compounds |
|---|-----------------------|--------------------|
| Degeneration (DEG)                            | 164                   | 168                |
| Inflamatory (INF)                             | 94                    | 164                |
| Non-neoplasic Proliferative<br>lesions ( PRO) | 82                    | 164                |

Representation of the confusion matrix, sensitivity and specificity for the best qualitative models obtained using conformal Random Forest Classifier.



#### Quality parameters of conformal model predictions.

| Fachariat | Internal prediction (cross-validation) |      | External prediction |      |
|-----------|--|------|---------------------|------|
| Endpoint  | Coverage                               | мсс  | Coverage            | мсс  |
| DEG       | 0.47                                   | 0.55 | 0.48                | 0.58 |
| INF       | 0.74                                   | 0.64 | 0.81                | 0.44 |
| PRO       | 0.64                                   | 0.45 | 0.65                | 0.29 |

In general, the quality of the models is acceptable, if we consider the complexity of the *in vivo* endpoints, representing many different mechanisms of liver toxicity.

# Value of IMI collaboration

- The results shown here are only an example of how the results of the eTOX project (the eTOX database and eTOXsys) are being exploited for practical purposes.
- The **eTOX** database represents a successful example of precompetitive sharing of information between 13 pharmaceutical companies, made possible thanks to the public-private partnership IMI project.

### Impact & take home message

- The information present in RDT reports generated for regulatory purposes can be transformed to obtain data amenable for developing predictive models.
- These models exemplify the practical use of the *in vivo data* collected by the **eTOX** project.
- Precompetitive collaboration can produce valuables resources which can be exploited to obtaining safer and more effective drugs, in a faster way.

innovative medicines initiative efpta This work has received support from the EU/EFPIA/Innovative Medicines Initiative Joint Undertaking (eTOX grant no 15002).