## My Submission

Daniel González Juclà

## Code structure

- 1. **data\_gathering\_and\_integration.ipynb:** Contains the gathering and the integration of the data from the different sources. PDF data was not considered because only a proportion could be parsed correctly in an automated way, so as it only represents a 1% of the data it has been ignored.
- 2. **data\_preprocessing.ipynb**: This notebook contains the exploration and preprocessing of the data, to form the final training and testing datasets. The preprocessing consists on:
  - 2.1. **Drop** those columns that are not consider relevant
  - 2.2. Encode time as day of the year and year
  - 2.3. Scale the numerical variables in the [0,1] range
  - 2.4. One-hot encoding of the categorical variables
  - 2.5. Dimensionality reduction of the one-hot encoded columns with PCA.
- 3. **xgboost\_modeling\_pca.ipynb**: The final modeling has been carried out with an XGBoost model, with has performed better than another modeled MLP. This XGBoost modeling is explained next.

## XGBoost modeling

The final modeling has been done by fine-tuning an XGBoost model, following this guide.

More precisely, the model has been fine-tuned with Cross-validation and the following steps:

- 1. Choose a relatively **high learning rate** and determine the **optimum number of trees** for this learning rate.
- **2. Tune tree-specific parameters** (max\_depth, min\_child\_weight, gamma, subsample, colsample\_bytree) for decided learning rate and number of trees.
- **3. Tune regularization parameters** (lambda, alpha) for xgboost which can help reduce model complexity and enhance performance.
- **4. Lower the learning rate** and decide the optimal parameters .