#### **Hackathon Instructions**

26 October 2019

#### **Data files**

- 1) songs train.csv (training data)
- 2) songs test without target.csv (test data)
- 3) songs test target predictions example.csv (format of the results file)

# **Data description**

Each row contains data related to a song, including its musical characteristics and information about its original performer (singer).

# What you need to do

All your work must be done in Jupyter Notebook.

- 1) Build a regression model that predicts the **song\_rating** attribute from the other song's attributes, using the data in file "**songs\_train.csv**" and **mean absolute error** (MAE) as a primary performance metric to evaluate your model. You can only use the models imported below.
- 2) Use your model to predict the song rating of the songs in file "songs test without target.csv".
- 3) Submit the .ipynb file(s) containing the source code of ALL your experimental work.
- 4) Submit a file called "songs\_test\_target\_predictions.csv" that contains ONE column of the predictions of the test file. You can do this using the code below.

#### Sample code:

```
df = pd.Series(test_predictions)
df.name = 'song_rating'
df.to csv('./songs test target predictions.csv', index=False,float format="%.3f", header=True)
```

(where test predictions is a 1-D array variable that contains your predicted values)

# Helpful imports

from sklearn.neighbors import KNeighborsRegressor from sklearn.linear\_model import LinearRegression from sklearn.ensemble import RandomForestRegressor, AdaBoostRegressor, GradientBoostingRegressor

from sklearn.metrics import mean\_absolute\_error

# **Some Hints**

- scoring parameter for cross-validation : scoring='neg mean absolute error'
- start with a very simple approach first: clean the data but do not do any feature engineering such as scaling and log normalization. You can try more complicated things if you have time.
- In order to save time in the beginning, do data preparation on all data (i.e. do not split data before data preparation step).
- if an attribute is too complicated or requires too much cleaning work (especially text attributes), then just remove it! You can deal with it later if you have time ...
- try different manual feature selection strategies.