## Supervised: basics

## 1. Preparing the dataset

- Get ready to use the same dataset as for the previous class
- Create a new dataset containing only numerical columns: budget, popularity, revenue, runtime, vote\_average, vote\_count. Clean them by removing rows with NaN
- From the release\_date, extract 2 columns: year, and month. You can use to datetime. Add those columns to the numerical dataset
- Transform the first column, 'adult', into a boolean variable, whose value can be 0
  or 1. You can use pandas get\_dummies or do it manually (or with sklearn
  OneHotEncoder, less convenient)
- Let's say that we want to predict the column popularity, a score given by the
  platform the data was extracted from. Remove this column from the table, and
  keep it in a separate list. The order of elements in the list allows to match them
  with variables.
- Split your dataset into a train and test set. You can do it manually or use sklearn train\_test\_split function. Keep for instance 1/3 as a test set.

## 2. First prediction: linear

- Train a linear regression with sklearn. You can use the LinearRegression class, and method fit.
- Compute the score we have seen, using corresponding functions in sklearn and the <u>predict</u> method of the linearRegression class. Do it first by using the train set, and then using the test set. Compare the difference.
- To get a more intuitive idea of the performance, plot the relation between the target variable and the prediction (e.g., seaborn scatterplot, x=target variable, y=your prediction). With a perfect prediction, you should observe a diagonal line.
- Check the coefficients coef\_ and the intercept intercept\_. Discuss with your peers about their interpretation. What about their magnitude? Sign?

## 3. Decision Tree

- Train a decision tree (DecisionTreeRegressor) with sklearn, using default parameters, using unnormalized data with zero (normalization is useless with tree, and makes interpretation harder)

- Compute the scores, comparing between evaluation training and testing sets, with the linear approach.
- The default parameters make the tree overfit. Play with the parameters max\_depth, min\_samples\_leaf, max\_leaf\_nodes, to try to limit the overfit (you should be able to improve over the linear regression)
- Train a tree small enough to be visualized, and plot it. You can use the built-in tool following the documents <a href="https://scikit-learn.org/stable/modules/tree.html#tree">https://scikit-learn.org/stable/modules/tree.html#tree</a>. The graphviz method usually gives the nicest results.
- Try to interpret the tree. Discuss with your peer.