# Gboost\_arodriguezsans\_Cad

May 18, 2021

## 1 Cádiz

#### 1.1 Gradient Boosting Trees

## 1.1.1 Gradient Boosting Regressor

```
[1]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     from sklearn.ensemble import GradientBoostingRegressor
     from sklearn.metrics import mean_squared_error
     from sklearn.model_selection import cross_val_score
     from sklearn.model_selection import train_test_split
     from sklearn.model_selection import RepeatedKFold
     from sklearn.model_selection import KFold
     from sklearn.model_selection import GridSearchCV
     from sklearn.model_selection import ParameterGrid
     from sklearn.preprocessing import MinMaxScaler
     from sklearn.inspection import permutation_importance
     from sklearn.metrics import mean_absolute_error, mean_squared_error
     import multiprocessing
     import warnings
     warnings.filterwarnings('once')
```

```
[2]: df_total = pd.read_excel('Total.xls')
# Edit columns names + Lower case column names
df_total.columns = map(str.lower, df_total.columns)
df_total.columns
```

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel\ipkernel.py:287:
DeprecationWarning: `should\_run\_async` will not call `transform\_cell`
automatically in the future. Please pass the result to `transformed\_cell`
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and should run async(code)

```
[2]: Index(['sub_region_2', 'fecha', 'provincia_iso', 'num_casos.x',
             'num_casos_prueba_pcr', 'num_casos_prueba_test_ac',
             'num_casos_prueba_ag', 'num_casos_prueba_elisa',
             'num_casos_prueba_desconocida', 'num_casos.y', 'num_hosp', 'num_uci',
             'num def', 'retail and recreation percent change from baseline',
             'grocery_and_pharmacy_percent_change_from_baseline',
             'parks percent change from baseline',
             'transit_stations_percent_change_from_baseline',
             'workplaces_percent_change_from_baseline',
             'residential_percent_change_from_baseline', 'total'],
            dtype='object')
 [3]: Cad = df_total.loc[df_total['sub_region_2'] == 'Cádiz']
 [4]: # Set index
      Cad = Cad.set_index('fecha')
 [5]: # We select columns of interest (mobility ones)
      Cad = Cad[['num_casos.x']+['num_casos_prueba_pcr']+ list(Cad.loc[:
       →, 'retail_and_recreation_percent_change_from_baseline':'total'])]
 [6]: # We create train and test datasets as in previous scenarios
      X_train, X_test, y_train, y_test = train_test_split( #Cad,
                                                          Cad.drop(columns =
       Cad['num_casos.x'],
                                                          shuffle = False, stratify =⊔
       \rightarrowNone,
                                                          train_size=0.942)
 [7]: type(y_test)
 [7]: pandas.core.series.Series
 [8]: type(X_test)
 [8]: pandas.core.frame.DataFrame
 [9]: | #X test
[10]: # Model generation
      model = GradientBoostingRegressor(n_estimators = 10,
                                        loss
                                                    = 'ls',
                                        max_features = 'auto',
                                        random_state = 123)
      model.fit(X_train, y_train)
```

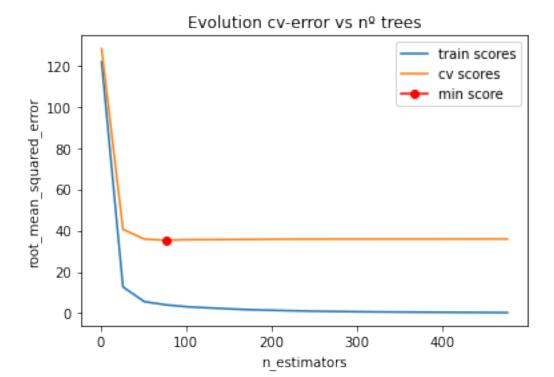
```
# Prediction
predictions = model.predict(X = X_test)
rmse = mean_squared_error(y_test, predictions, squared = False)
print(f" RMSE: {rmse}")
```

RMSE: 82.83093726071736

```
[11]: # Validation with k-cross-validation and neg root mean squared error
      train scores = []
      cv_scores
                = []
      # Values used
      estimator_range = range(1, 500, 25)
      # Train esach model with each values for n estimators and extract its error
      # test and k-cross-validation.
      for n_estimators in estimator_range:
         model = GradientBoostingRegressor(
                      n_estimators = n_estimators,
                      loss
                                = 'ls',
                      max_features = 'auto',
                      random_state = 123)
          # Error train
         model.fit(X_train, y_train)
         predictions = model.predict(X = X_train)
         rmse = mean_squared_error(
                  y_true = y_train,
                 y_pred = predictions,
                  squared = False
         train_scores.append(rmse)
          # Error cv
          scores = cross_val_score(
                     estimator = model,
                     X
                              = X_train,
                              = y_train,
                      scoring = 'neg_root_mean_squared_error',
                      CV
                     n_jobs = multiprocessing.cpu_count() - 1,
          # aggregate scores cross_val_score() and pass to possitive
          cv_scores.append(-1*scores.mean())
      # plot error evolution
      fig, ax = plt.subplots(figsize=(6, 4))
```

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 and should\_run\_async(code)

Optimal n estimators: 76



```
[12]: # Validation k-cross-validation and neg_root_mean_squared_error
results = {}

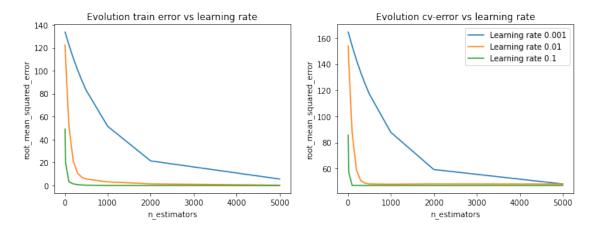
# Values used
learning_rates = [0.001, 0.01, 0.1]
```

```
n estimators = [10, 20, 100, 200, 300, 400, 500, 1000, 2000, 5000]
# model train for each combination of learning_rate + n_estimator
# we get the error for tain and k-cross-validation.
for learning_rate in learning_rates:
   train_scores = []
   cv_scores = []
   for n_estimator in n_estimators:
       model = GradientBoostingRegressor(
                    n_estimators = n_estimator,
                    learning_rate = learning_rate,
                    loss
                                = 'ls',
                   max_features = 'auto',
                   random_state = 123
                 )
        # Error train
       model.fit(X_train, y_train)
       predictions = model.predict(X = X_train)
       rmse = mean_squared_error(
               y_true = y_train,
               y_pred = predictions,
               squared = False
       train_scores.append(rmse)
        # Error CV
       scores = cross_val_score(
                    estimator = model,
                    Х
                             = X_train,
                            = y_train,
                    scoring = 'neg_root_mean_squared_error',
                             = 3,
                    CV
                   n_jobs
                             = multiprocessing.cpu_count() - 1
                 )
        # aggregate scores cross_val_score() and pass to possitive
        cv_scores.append(-1*scores.mean())
   results[learning_rate] = {'train_scores': train_scores, 'cv_scores':u
 →cv scores}
```

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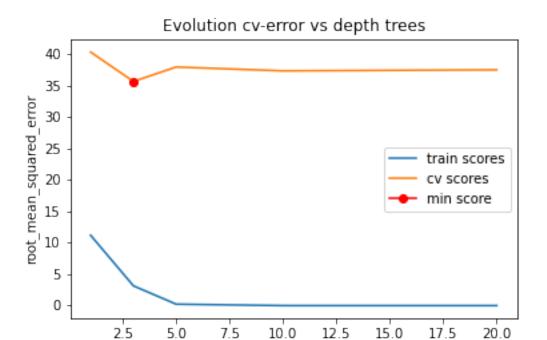


```
[14]: # Validation k-cross-validation and neg_root_mean_squared_error
train_scores = []
cv_scores = []

# Values used
max_depths = [1, 3, 5, 10, 20]
```

```
# Train model for each max_depth
for max_depth in max_depths:
   model = GradientBoostingRegressor(
                n_{estimators} = 100,
                loss
                            = 'ls',
                max_depth = max_depth,
                max_features = 'auto',
                random_state = 123
             )
    # Error train
   model.fit(X_train, y_train)
   predictions = model.predict(X = X_train)
   rmse = mean_squared_error(
            y_true = y_train,
            y_pred = predictions,
            squared = False
   train_scores.append(rmse)
    # Error CV
    scores = cross_val_score(
                estimator = model,
                Х
                        = X_train,
                        = y_train,
                scoring = 'neg_root_mean_squared_error',
                CV
                        = 5,
                         = multiprocessing.cpu_count() - 1
                n_jobs
             )
    # aggregate scores cross_val_score() pass to possitve
    cv_scores.append(-1*scores.mean())
# plots erros evolution
fig, ax = plt.subplots(figsize=(6, 3.84))
ax.plot(max_depths, train_scores, label="train scores")
ax.plot(max_depths, cv_scores, label="cv scores")
ax.plot(max_depths[np.argmin(cv_scores)], min(cv_scores),
       marker='o', color = "red", label="min score")
ax.set_ylabel("root_mean_squared_error")
ax.set_xlabel("max_depth")
ax.set_title("Evolution cv-error vs depth trees")
plt.legend();
print(f"Optimal max_depth: {max_depths[np.argmin(cv_scores)]}")
```

Optimal max\_depth: 3



max depth

```
[15]: # Grid hyperparmeters
      param_grid = {'max_features'
                                   : ['auto', 'sqrt', 'log2'],
                    'max_depth'
                                    : [None, 1, 3, 5, 10, 20],
                                    : [0.5, 1],
                    'subsample'
                    'learning_rate' : [0.001, 0.01, 0.1]
                   }
      # Grid-search with cv
      grid = GridSearchCV(
              estimator = GradientBoostingRegressor(
                              n_estimators
                                                   = 1000,
                              random_state
                                                   = 123,
                              # Early stop #
                              validation_fraction = 0.1,
                              n_iter_no_change
                                                   = 5,
                              tol
                                                   = 0.0001
                          ),
              param_grid = param_grid,
                         = 'neg_root_mean_squared_error',
              scoring
                         = multiprocessing.cpu_count() - 1,
              n_jobs
              cv
                         = RepeatedKFold(n_splits=3, n_repeats=1, random_state=123),
              refit
                         = True,
                         = 0,
              verbose
              return_train_score = True
```

```
grid.fit(X = X_train, y = y_train)
     # Results
     results = pd.DataFrame(grid.cv_results_)
     results.filter(regex = '(param.*|mean t|std t)') \
         .drop(columns = 'params') \
         .sort_values('mean_test_score', ascending = False) \
          .head(4)
     C:\ProgramData\Anaconda3\lib\site-packages\ipykernel\ipkernel.py:287:
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     automatically in the future. Please pass the result to `transformed_cell`
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       and should_run_async(code)
[15]:
        param_learning_rate_param_max_depth param_max_features_param_subsample \
     85
                        0.1
                                                         auto
                                         3
     49
                       0.01
                                         3
                                                                           1
                                                         auto
     66
                       0.01
                                        20
                                                                         0.5
                                                         auto
     36
                       0.01
                                                                         0.5
                                      None
                                                         auto
         mean test score std test score mean train score std train score
                               3.852277
                                                -5.717723
     85
              -14.063480
                                                                  0.545169
              -14.348478
     49
                               3.821078
                                                -6.734509
                                                                 0.428128
     66
              -14.426820
                               2.981705
                                                -6.288895
                                                                  0.659022
                                                                  0.656911
              -14.439484
                               2.995106
                                                -6.303719
[16]: # Best hyperparameters by cv
     print("----")
     print("Best hyperparameters by cv")
     print("----")
     print(grid.best_params_, ":", grid.best_score_, grid.scoring)
     Best hyperparameters by cv
     _____
     {'learning_rate': 0.1, 'max_depth': 3, 'max_features': 'auto', 'subsample': 1} :
     -14.063480148599941 neg_root_mean_squared_error
     C:\ProgramData\Anaconda3\lib\site-packages\ipykernel\ipkernel.py:287:
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     automatically in the future. Please pass the result to `transformed_cell`
     argument and any exception that happen during thetransform in
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       and should_run_async(code)
```

```
[17]: # Error test
      model = grid.best_estimator_
      predictions = model.predict(X = X_test)
      rmse = mean_squared_error(
             y_true = y_test,
             y_pred = predictions,
             squared = False
      print(f"rmse test: {rmse}")
     rmse test: 60.22712986053597
[18]: | importance_predictors = pd.DataFrame({#'predictor': Cad.columns,
                                          'predictor': Cad.drop(columns = 'num casos.
      \rightarrow x').columns,
                                            'importance': model.feature_importances_})
      print("Importance of predictors")
      print("----")
      importance_predictors.sort_values('importance', ascending=False)
     Importance of predictors
     C:\ProgramData\Anaconda3\lib\site-packages\ipykernel\ipkernel.py:287:
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     automatically in the future. Please pass the result to `transformed_cell`
     argument and any exception that happen during thetransform in
     `preprocessing_exc_tuple` in IPython 7.17 and above.
       and should_run_async(code)
                                                 predictor importance
[18]:
      0
                                      num_casos_prueba_pcr
                                                              0.992050
      3
                        parks_percent_change_from_baseline
                                                              0.004198
      6
                 residential_percent_change_from_baseline
                                                              0.001707
      7
                                                              0.000514
                                                     total
      2 grocery_and_pharmacy_percent_change_from_baseline
                                                              0.000475
                   workplaces_percent_change_from_baseline
      5
                                                              0.000448
             transit_stations_percent_change_from_baseline
      4
                                                              0.000399
      1 retail_and_recreation_percent_change_from_base...
                                                            0.000209
[19]: importance = permutation_importance(
                      estimator
                                   = model,
                      X
                                   = X_train,
                                  = y_train,
                      n_repeats
                                   = 5,
                      scoring
                                   = 'neg_root_mean_squared_error',
                                   = multiprocessing.cpu_count() - 1,
                      n_jobs
                      random_state = 123
```

```
# Store results (mean / sd)
      df_importance = pd.DataFrame(
                          {k: importance[k] for k in ['importances_mean', __
      df_importance['feature'] = X_train.columns
      df_importance.sort_values('importances_mean', ascending=False)
     C:\ProgramData\Anaconda3\lib\site-packages\ipykernel\ipkernel.py:287:
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       and should_run_async(code)
[19]:
         importances_mean importances_std \
              179.256190
                                  3.486350
      0
      3
                3.813036
                                  0.217293
      6
                0.765762
                                  0.074297
      7
                0.723144
                                  0.161094
      2
                0.349850
                                  0.055652
                0.302140
                                  0.151518
      4
                0.273565
                                  0.039790
                0.067531
                                  0.022022
      1
                                                   feature
      0
                                      num_casos_prueba_pcr
      3
                       parks_percent_change_from_baseline
      6
                 residential_percent_change_from_baseline
      7
        grocery_and_pharmacy_percent_change_from_baseline
      2
                  workplaces_percent_change_from_baseline
      5
      4
             transit_stations_percent_change_from_baseline
      1 retail_and_recreation_percent_change_from_base...
[20]: # Calculate the mean absolute error (MAE)
      mae = mean_absolute_error(predictions, y_test)
      print('MAE: ' + str(round(mae, 5)))
      # Calculate the root mean squarred error (RMSE)
      rmse = np.sqrt(mean_squared_error(y_test,predictions))
```

MAE: 52.69392 RMSE: 60.22713

print('RMSE: ' + str(round(rmse, 5)))

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 and should\_run\_async(code)

```
[21]: predictions_df = pd.DataFrame(predictions)
    predictions_df.rename(columns={0:'Pred'},inplace=True)
    y_test_df=pd.DataFrame(y_test)
    y_test_df.reset_index(drop=True, inplace=True)
    y_test_df
    predictions_df['yt']=y_test_df['num_casos.x']
    predictions_df
```

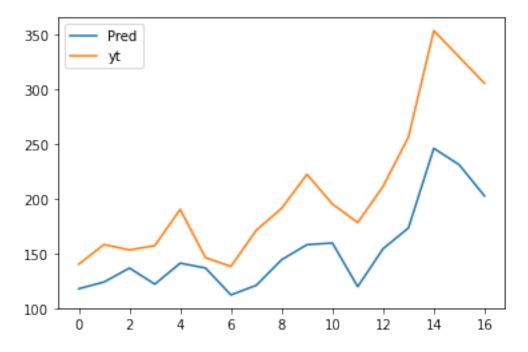
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and should run\_async(code)

```
[21]:
              Pred
                    yt
     0
         117.665262 140
     1 123.799842 158
         136.532269 153
     3 121.832558 157
     4
        141.004841 190
     5
       136.532269 146
        112.022508 138
     6
     7
        120.775663 171
     8
         144.159020 191
         157.834525 222
     10 159.401701 195
     11 119.618124 178
     12 154.155876 211
     13 173.095042 256
     14 245.722347 353
     15 230.786282 329
     16 202.265224 305
```

```
[22]: _ = predictions_df.plot()
```

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and should\_run\_async(code)



# 1.2 XGboost (Supervised)

Brownlee, J., 2021. How to Use XGBoost for Time Series Forecasting. [online] Machine Learning Mastery. Available at: https://machinelearningmastery.com/xgboost-for-time-series-forecasting/[Accessed 17 May 2021].

```
[23]: from numpy import asarray
      from pandas import DataFrame
      from pandas import concat
      from xgboost import XGBRegressor
      from matplotlib import pyplot
      # transform a time series dataset into a supervised learning dataset
      def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
              n_vars = 1 if type(data) is list else data.shape[0]
              df = DataFrame(data)
              cols = list()
              # input sequence (t-n, \ldots t-1)
              for i in range(n_in, 0, -1):
                      cols.append(df.shift(i))
              # forecast sequence (t, t+1, \ldots t+n)
              for i in range(0, n_out):
                      cols.append(df.shift(-i))
              # put it all together
```

```
agg = concat(cols, axis=1)
        # drop rows with NaN values
        if dropnan:
                agg.dropna(inplace=True)
        return agg.values
# split a univariate dataset into train/test sets
def train_test_split(data, n_test):
        return data[:-n_test, :], data[-n_test:, :]
# fit an xqboost model and make a one step prediction
def xgboost_forecast(train, testX):
        # transform list into array
        train = asarray(train)
        # split into input and output columns
       trainX, trainy = train[:, :-1], train[:, -1]
        # fit model
       model = XGBRegressor(objective='reg:squarederror', n_estimators=1000)
       model.fit(trainX, trainy)
        # make a one-step prediction
       yhat = model.predict(asarray([testX]))
       return yhat[0]
# walk-forward validation for univariate data
def walk_forward_validation(data, n_test):
       predictions = list()
        # split dataset
       train, test = train_test_split(data, n_test)
        # seed history with training dataset
       history = [x for x in train]
        # step over each time-step in the test set
        for i in range(len(test)):
                # split test row into input and output columns
                testX, testy = test[i, :-1], test[i, -1]
                # fit model on history and make a prediction
                yhat = xgboost_forecast(history, testX)
                # store forecast in list of predictions
                predictions.append(yhat)
                # add actual observation to history for the next loop
                history.append(test[i])
                # summarize progress
                print('>expected=%.1f, predicted=%.1f' % (testy, yhat))
        # estimate prediction error
        error = mean_squared_error(test[:, -1], predictions, squared = False)
    #error = mean_absolute_error(test[:, -1], predictions)
        return error, test[:, -1], predictions
```

```
# load the dataset
values = Cad['num_casos.x'].values
# transform the time series data into supervised learning
data = series_to_supervised(values, n_in=14)
# evaluate
mae, y, yhat = walk_forward_validation(data,17)
#print('MAE: %.3f' % mae)
print('RMSE: %.3f' % mae)
# plot expected vs preducted
plt.plot(y, label='Expected')
plt.plot(yhat, label='Predicted')
plt.legend()
plt.show()
C:\ProgramData\Anaconda3\lib\site-packages\ipykernel\ipkernel.py:287:
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`preprocessing_exc_tuple` in IPython 7.17 and above.
  and should_run_async(code)
C:\ProgramData\Anaconda3\lib\site-packages\xgboost\data.py:119: UserWarning: Use
subset (sliced data) of np.ndarray is not recommended because it will generate
extra copies and increase memory consumption
  warnings.warn(
>expected=140.0, predicted=103.3
>expected=158.0, predicted=177.8
>expected=153.0, predicted=146.9
>expected=157.0, predicted=177.7
>expected=190.0, predicted=139.5
>expected=146.0, predicted=178.4
>expected=138.0, predicted=139.7
>expected=171.0, predicted=164.8
>expected=191.0, predicted=160.8
>expected=222.0, predicted=166.5
>expected=195.0, predicted=212.6
>expected=178.0, predicted=199.7
>expected=211.0, predicted=200.5
>expected=256.0, predicted=197.5
>expected=353.0, predicted=276.8
>expected=329.0, predicted=295.2
>expected=305.0, predicted=334.1
RMSE: 35.949
```

