	Validación métrica:	s con HiperParametros	 Laboratorio 3
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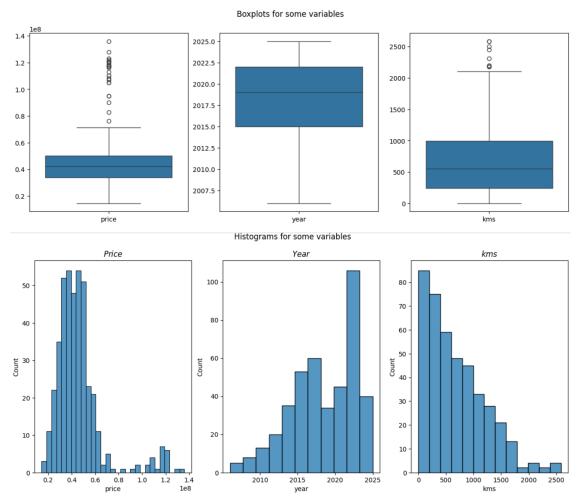
Luis Felipe Sanchez Sanchez - 90613 Harold Shneider Martinez Tapiero - 77999

> Universidad ECCI Facultad de Ingeniería

Elias Buitrago Bolivar

Resultados de los diferentes modelos con los datos subidos sin seleccionar ni filtrar ninguna info:

Graficas:



Multivariate lineal regression

```
✓ Multivariate lineal regression

✓ Multivariate lineal regression

✓ I # Define model and prediction
2 ols = LinearRegression()
3 model1 = ols.fit(X_train, y_train)
4 y_pred1 = model1.predict(X_test)

✓ I # accuracy check
2 rmse = MSE(y_test, y_pred1, squared=False)
3 mae = MAE(y_test, y_pred1)
4 r2 = r2_score(y_test, y_pred1)
5 print("RMSE: %.2f" % rmse)
6 print("MAE: %.2f" % mae)
7 print("R2: %.2f" % mae)
7 print("R2: %.2f" % r2)

✓ RMSE: 12109549.03
MAE: 8817707.26
R2: 0.57
```

RMSE: 12109549.03 MAE: 8817707.26

R2: 0.57

Light GBM

RMSE: 11126993.07 MAE: 6700310.15

R2: 0.63

Random Forest Regressor

RMSE: 9477736.16 MAE: 5146986.60

Xgboost regressor

```
12. [183] 1 #K-fold cross validation
2 scores = cross_val_score(model4, X_train, y_train, cv=10)
3 print("Mean cross-validation score: %.2f" % scores.mean())

→ Mean cross-validation score: 0.68

12. ② 1 kfold = KFold(n_splits=10, shuffle=True)
2 kf_cv_scores = cross_val_score(model4, X_train, y_train, cv=kfold)
3 print("K-fold CV average score: %.2f" % kf_cv_scores.mean())

→ K-fold CV average score: %.2f" % kf_cv_scores.mean())

→ K-fold CV average score: %.2f" % kf_cv_scores.mean())

1 # Pred
2 y_pred4 = model4.predict(X_test)

1 # accuracy_check
2 rmse = MSE(y_test, y_pred4, squared=False)
3 mae = MAE(y_test, y_pred4)
4 r2 = r2_score(y_test, y_pred4)
5 print("MSE: %.2f" % rmse)
6 print("MSE: %.2f" % mae)
7 print("RSE: %.2f" % rae)

→ RMSE: 10238249.56
ME: 5221331.12
R2: 0.69
```

RMSE: 10238249.56 MAE: 5221331.12

R2: 0.69

Prueba 1:

Light GBM

```
1 # Hyperparameters
                1 # Hyperparameters
2 params = {
3    'task': 'train',
4    'boosting': 'gbdt',
5    'objective': 'regression',
6    'num_leaves': 9,
7    'learning_rate': 0.21,
8    'metric': {'12','11'},
9    'header': 'true',
10    'verbose': 0
11 }
              11 }
              13 # laoding data
              14 lgb_train = lgb.Dataset(X_train, y_train)
15 lgb_eval = lgb.Dataset(X_test, y_test, reference=lgb_train)
             16
17 # fitting the model
              18 model2 = lgb.train(params,
              19
                                              train_set=lgb_train,
                                                 valid_sets=lgb_eval)
              21 # Pred
              22 y_pred2 = model2.predict(X_test)
1 # accuracy check
2 rmse = MSE(y_test, y_pred2, squared=False)
                3 mae = MAE(y_test, y_pred2)
               4 ro = role(y_test, y_pred2)
4 ro = role(y_test, y_pred2)
5 print("RMSE: %.2f" % rmse)
6 print("MAE: %.2f" % mae)
7 print("R2: %.2f" % r2)
     RMSE: 12625286.83
MAE: 6797070.60
R2: 0.70
```

RMSE: 12625286.83 MAE: 6797070.60

Random Forest Regressor

```
[ ] 1 from sklearn.ensemble import RandomForestRegressor

[ ] 1 model3 = RandomForestRegressor()
2 model3.fit(X_train, y_train)
3 y_pred3 = model3.predict(X_test)

[ ] 1 # accuracy check
2 rmse = MSE(y_test, y_pred3, squared=False)
3 mae = MAE(y_test, y_pred3)
4 r2 = r2_score(y_test, y_pred3)
5 print("RMSE: %.2f" % rmse)
6 print("MAE: %.2f" % rmse)
7 print("R2: %.2f" % r2)

RMSE: 8993766.98
MAE: 5157716.40
R2: 0.85
```

RMSE: 8993766.98 MAE: 5157716.40

R2: 0.85

Xgboost regressor

```
[ ] 1 # Pred
2 y_pred4 = model4.predict(X_test)

[ ] 1 # accuracy check
2 rmse = MSE(y_test, y_pred4, squared=False)
3 mae = MAE(y_test, y_pred4)
4 r2 = r2_score(y_test, y_pred4)
5 print("RMSE: %.2f" % rmse)
6 print("MAE: %.2f" % mae)
7 print("R2: %.2f" % r2)

→ RMSE: 11535189.22
MAE: 6249334.76
R2: 0.75
```

RMSE: 11535189.22 MAE: 6249334.76

Prueba 2:

Light GBM

```
[80] 1 # Hyperparameters
         1 # nyperparameters
2 params = {
3   'task': 'train',
4   'boosting': 'gbdt',
5   'objective': 'regression',
6   'num_leaves': 5,
7   'learning_rate': 0.52,
8   'metric': {'12','11'},
9   'header': 'true',
10   'verbose': 0
        10
        11 }
        12
        13 # laoding data
        14 lgb_train = lgb.Dataset(X_train, y_train)
        15 lgb_eval = lgb.Dataset(X_test, y_test, reference=lgb_train)
        17\ \text{\# fitting the model}
        18 model2 = lgb.train(params,
                                 train_set=lgb_train,
valid_sets=lgb_eval)
        19
        20
        21 # Pred
        22 y_pred2 = model2.predict(X_test)
 1 # accuracy check
         2 rmse = MSE(y_test, y_pred2, squared=False)
         3 mae = MAE(y_test, y_pred2)
        4 r2 = r2_score(y_test, y_pred2)
5 print("RMSE: %.2f" % rmse)
         6 print("MAE: %.2f" % mae)
         7 print("R2: %.2f" % r2)
 RMSE: 11861025.88
MAE: 6673291.02
        R2: 0.74
```

RMSE: 11861025.88 MAE: 6673291.02

R2: 0.74

Random Forest Regressor

RMSE: 8900388.53 MAE: 5350869.41

R2: 0.83

Xgboost regressor

```
    1 # accuracy check
    2 rmse = MSE(y_test, y_pred4, squared=False)
    3 mae = MAE(y_test, y_pred4)
    4 r2 = r2_score(y_test, y_pred4)
    5 print("RMSE: %.2f" % rmse)
    6 print("MAE: %.2f" % mae)
    7 print("R2: %.2f" % r2)

    RMSE: 12371077.69
    MAE: 7126827.66
    R2: 0.67
```

RMSE: 12371077.69 MAE: 7126827.66

R2: 0.67

Prueba 3:

Light GBM

RMSE: 11597556.08 MAE: 6994310.87

R2: 0.75

Random Forest Regressor

```
1 param_grid = {
        'n_estimators': [ 150, 200, 350], # Número de árboles en el
'max_depth': [None, 10, 20], # Profundidad máxima de los ár
        'min_samples_split': [2, 9, 10], # Número mínimo de muestra
        'min_samples_leaf': [2, 3, 4], # Número mínimo de muestras 'max_features': ['auto', 'sqrt', 'log2'] # Número máximo de
 8 rf = RandomForestRegressor(random_state=42)
 9 random_search = RandomizedSearchCV(estimator=rf, param_distribut
                                            scoring=scoring, refit='rmse
13 random_search.fit(X_train, y_train)
14 best_model = random_search.best_estimator_
15 y_pred = best_model.predict(X_test)
18 rmse = np.sqrt(mean_squared_error(y_test, y_pred))
19 mae = mean_absolute_error(y_test, y_pred)
20 r2 = r2_score(y_test, y_pred)
22 print("RMSE: %.2f" % rmse)
23 print("MAE: %.2f" % mae)
24 print("R2: %.2f" % r2)
Fitting 5 folds for each of 100 candidates, totalling 500 fits
MAE: 5170397.83
R2: 0.84
```

RMSE: 8684526.04 MAE: 5170397.83

R2: 0.84

Xgboost regressor

```
[50] 1 # Pred
2 y_pred4 = model4.predict(X_test)

1 # accuracy check
2 rmse = MSE(y_test, y_pred4, squared=False)
3 mae = MAE(y_test, y_pred4)
4 r2 = r2_score(y_test, y_pred4)
5 print("RMSE: %.2f" % rmse)
6 print("MAE: %.2f" % mae)
7 print("R2: %.2f" % r2)

RMSE: 11715237.49
MAE: 6850894.90
R2: 0.71
```

RMSE: 11715237.49 MAE: 6850894.90

R2: 0.71

Prueba 4:

Light GBM

```
13 # laoding data
    14 lgb_train = lgb.Dataset(X_train, y_train)
    15 lgb_eval = lgb.Dataset(X_test, y_test, reference=lgb_train)
    17 # fitting the model
    18 model2 = lgb.train(params,
                     train_set=lgb_train,
    20
                     valid_sets=lgb_eval)
    21 # Pred
    22 y_pred2 = model2.predict(X_test)
1 # accuracy check
     2 rmse = MSE(y_test, y_pred2, squared=False)
    3 mae = MAE(y_test, y_pred2)
    4 r2 = r2_score(y_test, y_pred2)
     5 print("RMSE: %.2f" % rmse)
     6 print("MAE: %.2f" % mae)
    7 print("R2: %.2f" % r2)
₹ RMSE: 11228274.90
    MAE: 6582336.57
    R2: 0.76
```

RMSE: 11228274.90 MAE: 6582336.57

Random Forest Regressor

```
1 param_grid = {
         'n_estimators': [ 200, 400], # Número de árboles en el bo
'max_depth': [None, 10, 20, 30], # Profundidad máxima de
'min_samples_split': [2, 9, 10], # Número mínimo de muest
         'min_samples_leaf': [2, 3, 4, 5], # Número mínimo de mues
'max_features': ['auto', 'sqrt', 'log2'] # Número máximo
 8 rf = RandomForestRegressor(random_state=42)
 9 random_search = RandomizedSearchCV(estimator=rf, param_distrib
                                                    scoring=scoring, refit='rms
13 random_search.fit(X_train, y_train)
14 best_model = random_search.best_estimator_
15 y_pred = best_model.predict(X_test)
18 rmse = np.sqrt(mean_squared_error(y_test, y_pred))
19 mae = mean_absolute_error(y_test, y_pred)
20 r2 = r2_score(y_test, y_pred)
22 print("RMSE: %.2f" % rmse)
23 print("MAE: %.2f" % mae)
24 print("R2: %.2f" % r2)
Fitting 5 folds for each of 100 candidates, totalling 500 fits
RMSE: 8955562.17
MAE: 5367564.00
R2: 0.83
```

RMSE: 8955562.17 MAE: 5367564.00

R2: 0.83

Xgboost regressor

```
[63] 1 # Pred
2 y_pred4 = model4.predict(X_test)

1 # accuracy check
2 rmse = MSE(y_test, y_pred4, squared=False)
3 mae = MAE(y_test, y_pred4)
4 r2 = r2_score(y_test, y_pred4)
5 print("RMSE: %.2f" % rmse)
6 print("MAE: %.2f" % mae)
7 print("R2: %.2f" % r2)

RMSE: 10775624.04
MAE: 6505791.87
R2: 0.75
```

RMSE: 10775624.04 MAE: 6505791.87