	Validaciór	n métricas	con HiperParametros	<ul> <li>Laboratorio</li> </ul>
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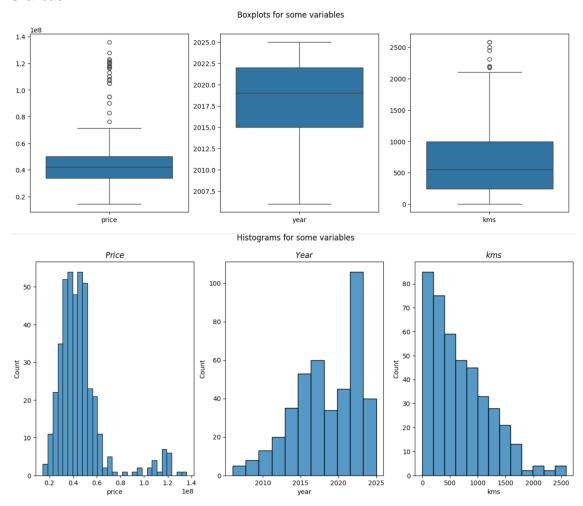
Johan Steven Benavides Guarnizo-88593 Sebastian Morales Devia - 77999

> 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

```
Wultivariate lineal regression

**Multivariate lineal regression

**Image: Second State of the content of
```

RMSE: 12109549.03 MAE: 8817707.26

R2: 0.57

# Light GBM

RMSE: 11126993.07 MAE: 6700310.15

R2: 0.63

#### Random Forest Regressor

```
    Random Forest Regressor

    [96] 1 from sklearn.ensemble import RandomForestRegressor

    [97] 1 model3 = RandomForestRegressor()
    2 model3.fit(X_train, y_train)
    3 y_pred3 = model3.predict(X_test)

    [98] 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("MMSE: %.2f" % rmse)
    6 print("MAE: %.2f" % mae)
    7 print("R2: %.2f" % r2)

    RMSE: 9477736.16
    MAE: 5146986.60
    R2: 0.73
```

RMSE: 9477736.16 MAE: 5146986.60

### Xgboost regressor

```
| 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()) | \( \frac{1}{2} \) | Mean cross-validation score: %.2f" % scores.mean()) | \( \frac{1}{2} \) | Mean cross-validation score: %.2f" % scores.mean()) | \( \frac{1}{2} \) | Mean cross-validation score: %.2f" % scores.mean()) | \( 2 \) | 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()) | \( \frac{1}{2} \) | K-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | K-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | K-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % kf_cv_scores.mean()) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % mse) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % mse) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % mse) | \( \frac{1}{2} \) | \( \frac{1}{2} \) | M-fold CV average score: %.2f" % mse) | \( \frac{1}{2} \) | \(
```

RMSE: 10238249.56 MAE: 5221331.12

R2: 0.69

#### Prueba 1:

#### Light GBM

```
1 # Hyperparameters
               2 params = {
3   'task': 'train',
                      'task': 'train',
'boosting': 'gbdt',
'objective': 'regression',
'num_leaves': 9,
'learning_rate': 0.21,
'metric': {'12','11'},
'header': 'true',
'verbose': 0
             10
             11 }
            12
13 # laoding data
14 lgb_train = lgb.Dataset(X_train, y_train)
             15~{\rm lgb\_eval}~=~{\rm lgb.Dataset}({\rm X\_test,~y\_test,~reference=lgb\_train})
             17 # fitting the model
            18 model2 = lgb.train(params,
19 train_set=lgb_train,
20 valid_sets=lgb_eval)
             21 # Pred
             22 y_pred2 = model2.predict(X_test)
os 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: 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" % rae)
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

RMSE: 11535189.22 MAE: 6249334.76

#### Prueba 2:

# Light GBM

```
[80] 1 # Hyperparameters
        1 # ryperparameters
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 }
       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,
       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: 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 RMSE: 8684526.04
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,
     19
                        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 muest '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
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