pwd 'C:\\Users\\Alfredo\\Desktop\\Machine Learning' 1s Directorio de C:\Users\Alfredo\Desktop\Machine Learning 19/04/2021 22:06 19/04/2021 22:06 19/04/2021 20:49 5.037 agrupacion por columna.ipynb 16/04/2021 20:34 185.160 cufflinks graficos .ipynb 14/04/2021 21:00 4.592 data frame en una tabla SQL.ipynb 28/07/2020 18:58 137.766 dataframe df.ipynb 7.495 DataFrames modificacion de los nombres de las filas .ipynb 17/06/2020 18:23 7.468 dictionaries.ipynb 07/04/2021 19:14 14.378 ejercicio 1 matplotlib.ipynb 17/06/2020 18:23 1.911 filter.ipynb 1.745 funciones.ipynb 07/04/2021 21:03 33.836 Graficos de columna para tipo categoria seaborn.ipynb 09/04/2021 19:31 171.852 Graficos de Regresion.ipynb 08/04/2021 16:08 63.630 heat maps.ipynb 19/04/2021 22:06 31.598 linear regression.ipynb 07/04/2021 18:40 03/07/2020 12:24 6.613 merge con dataframe.ipynb 2.469 numpy con listas.ipynb 08/04/2021 16:08 201.001 pairgrid & facegrid.ipynb 07/04/2021 19:24 02/07/2020 19:34 11.249 seleccion de datos pandas.ipynb 08/01/2021 18:55 16.098 Untitled.ipynb 1.385 Untitled1.ipynb 16/04/2021 20:15 08/04/2021 13:05 14/04/2021 22:01 3.656.309 Untitled3.ipynb 15/04/2021 18:00 555 Untitled4.ipynb 09/02/2021 15:58 1.695 Untitled5.ipynb 15/04/2021 18:32 15/04/2021 23:06 72.266 Untitled7.ipynb 16/04/2021 20:01 1.398 Untitled9.ipynb 19/04/2021 21:13 726.209 USA_Housing.csv 02/07/2020 19:35 2.766 valores nulos.ipynb 5.505.736 bytes 3 dirs 354.973.999.104 bytes libres casas = pd.read_csv('C://Users//Alfredo//Desktop//Machine Learning//USA Housing.csv') casas Avg. Area Number of Avg. Area Number of Avg. Area Avg. Area House Area Price **Address** Income **Bedrooms Population** Rooms 208 Michael Ferry Apt. 23086.800503 1.059034e+06 79545.458574 5.682861 7.009188 4.09 674\nLaurabury, NE 3701.. 188 Johnson Views Suite 079\nLake 79248.642455 6.002900 40173.072174 1.505891e+06 36882.159400 1.058988e+06 9127 Elizabeth Stravenue\nDanieltown, WI 06482... 5.865890 8.512727 5.13 61287.067179 63345.240046 34310.242831 1.260617e+06 USS Barnett\nFPO AP 44820 7.188236 5.586729 3.26 59982.197226 5.040555 7.839388 4.23 26354.109472 6.309435e+05 USNS Raymond\nFPO AE 09386 22837.361035 1.060194e+06 USNS Williams\nFPO AP 30153-7653 **4995** 60567.944140 3.46 7.830362 6.137356 PSC 9258, Box 8489\nAPO AA 42991-25616.115489 1.482618e+06 **4996** 78491.275435 6.999135 6.576763 4.02 33266.145490 1.030730e+06 4213 Hacy Surdon 076\nJoshualand, VA 01... **4997** 63390.686886 7.250591 4.805081 2.13 42625.620156 1.198657e+06 USS Wallace\nFPO AE 73316 **4998** 68001.331235 5.534388 7.130144 5.44 37778 George Ridges Apt. 509\nEast **4999** 65510.581804 5.992305 6.792336 4.07 46501.283803 1.298950e+06 Holly, NV 2... 5000 rows × 7 columns casas info <bound method DataFrame.info of</pre> Avg. Area Income Avg. Area House Age Avg. Area Number of Rooms \ 79545.458574 79248.642455 6.002900 6.730821 5.865890 63345.240046 7.188236 5.586729 4995 4996 78491.275435 6.999135 6.576763 63390.686886 4.805081 4998 68001.331235 5.534388 65510.581804 4999 Avg. Area Number of Bedrooms Area Population 4.09 23086.800503 1.059034e+06 36882.159400 1.058988e+06 34310.242831 1.260617e+06 26354.109472 6.309435e+05 4995 22837.361035 1.060194e+06 25616.115489 1.482618e+06 33266.145490 1.030730e+06 4998 4999 46501.283803 1.298950e+06 Address 9127 Elizabeth Stravenue\nDanieltown, WI 06482... USS Barnett\nFPO AP 44820 USNS Raymond\nFPO AE 09386 USNS Williams\nFPO AP 30153-7653 PSC 9258, Box 8489\nAPO AA 42991-3352 4996 4997 USS Wallace\nFPO AE 73316 4999 37778 George Ridges Apt. 509\nEast Holly, NV 2... [5000 rows x 7 columns]> casas describe Avg. Area Income Avg. Area House Age Avg. Area Number of Rooms \ 79545.458574 5.682861 79248.642455 6.002900 63345.240046 5.586729 5.040555 59982.197226 7.839388 60567.944140 6.137356 4995 78491.275435 4996 63390.686886 4.805081 68001.331235 7.130144 4998 5.534388 65510.581804 5.992305 4999 Avg. Area Number of Bedrooms Area Population 23086.800503 1.059034e+06 36882.159400 1.058988e+06 34310.242831 1.260617e+06 26354.109472 6.309435e+05 22837.361035 1.060194e+06 4995 4996 25616.115489 1.482618e+06 33266.145490 1.030730e+06 42625.620156 1.198657e+06 4998 46501.283803 1.298950e+06 4999 Address 9127 Elizabeth Stravenue\nDanieltown, WI 06482... USS Barnett\nFPO AP 44820 USNS Raymond\nFPO AE 09386 4995 USNS Williams\nFPO AP 30153-7653 PSC 9258, Box 8489\nAPO AA 42991-3352 4996 4215 Tracy Garden Suite 076\nJoshualand, VA 01... 4997 4998 USS Wallace\nFPO AE 73316 37778 George Ridges Apt. 509\nEast Holly, NV 2... [5000 rows x 7 columns]> casas columns Index(['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of Rooms', 'Avg. Area Number of Bedrooms', 'Area Population', 'Price', 'Address'], election import train_test_split rn.linear_model import LinearRegression casas head() Avg. Area Number of Avg. Area Avg. Area House Avg. Area Number of Area **Price Address** Rooms Bedrooms **Population** Income 208 Michael Ferry Apt. 674\nLaurabury, **0** 79545.458574 5.682861 7.009188 4.09 23086.800503 1.059034e+06 188 Johnson Views Suite 079\nLake **1** 79248.642455 6.002900 6.730821 3.09 40173.072174 1.505891e+06 Kathleen, CA... 9127 Elizabeth Stravenue\nDanieltown, **2** 61287.067179 5.865890 8.512727 5.13 36882.159400 1.058988e+06 WI 06482... **3** 63345.240046 7.188236 5.586729 3.26 34310.242831 1.260617e+06 USS Barnett\nFPO AP 44820 **4** 59982.197226 5.040555 7.839388 26354.109472 6.309435e+05 USNS Raymond\nFPO AE 09386 X= casas[['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of Rooms', X_train, X_test, Y_train, Y_test =train_test_split(X, Y, test_size=0.3,random_state =42) X_train Avg. Area Income Avg. Area House Age Avg. Area Number of Rooms Avg. Area Number of Bedrooms Area Population **1840** 55245.337288 3.965745 8.961106 43557.943435 **2115** 62305.638407 3.19 47066.617420 6.490769 7.647362 5.407515 8.243178 3.11 26706.911029 **4437** 77345.472379 **1146** 65846.171039 6.385374 6.804131 3.18 28214.363551 **2486** 69350.793357 6.910415 8.288048 4.29 36779.058567 **4426** 76223.561256 6.371627 5.342217 2.42 30165.337445 **466** 56685.014442 3.38 43322.166854 6.958045 7.502115 3.14 **3092** 66195.337714 6.507971 6.611861 37288.923574 **3772** 58694.515017 7.394768 9.269453 4.32 49960.977236 7.880521 6.04 **860** 61162.580254 5.896316 36033.701431 lrm = LinearRegression() lrm fit(X_train, Y_train) LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False) from **sklearn** import metrics predicciones = lrm.predict(X_test) predicciones #los datos de test se comparan con los valores de Y (datos de pruebas) 1457119.79297226, 1483428.95309313, 1047510.59737201]) plt scatter(Y_test, predicciones) <matplotlib.collections.PathCollection at 0x2116d11a0c8>

series de metricas para evaluar el modelo de forma numerica

metrics.mean_absolute_error(Y_test, predicciones)
81135.5660933688

MAE (Mean absoulute error) - Media del valor absoluto de los errores

MSE (Media de los erroes al cuadrado)
metrics.mean_squared_error(Y_test, predicciones)

10068422551.400928