Ejercicio1_Clustering_012 Carla Mayela De la Garza Fernandez - Alan Zamarron Medrano **Ejercicio Regresion Lineal** In [1]: import pandas as pd import numpy as np import matplotlib.pyplot as plt import statsmodels.formula.api as smf %matplotlib inline from sklearn.linear_model import LinearRegression In [2]: data = pd.read csv("RegresionLineal.csv") Out[2]: Peso Altura **0** 68.78 162 **1** 74.11 212 2 71.73 220 **3** 69.88 206 **4** 67.25 152 **5** 68.78 183 6 68.34 167 **7** 67.01 175 **8** 63.45 156 **9** 71.19 186 **10** 67.19 183 **11** 65.80 163 **12** 64.30 163 **13** 67.97 172 **14** 72.18 194 **15** 65.27 168 **16** 66.09 161 **17** 67.51 164 **18** 70.10 188 **19** 68.25 187 **20** 67.89 162 **21** 68.14 192 **22** 69.08 184 **23** 72.80 206 **24** 67.42 175 **25** 68.49 154 **26** 68.61 187 **27** 74.03 212 **28** 71.52 195 **29** 69.18 205 data.plot(kind="scatter", x="Altura", y="Peso") Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x174b4cc2880> 74 72 70 68 66 64 150 160 170 180 190 200 210 220 Altura In [4]: | lm = smf.ols(formula="Peso~Altura", data=data).fit() lm.params Out[4]: Intercept 49.071634 0.108611 Altura dtype: float64 In [9]: | altura_pred = lm.predict(pd.DataFrame(data["Altura"])) altura_pred Out[9]: 0 66.666580 72.097119 1 2 72.966006 3 71.445455 65.580473 5 68.947407 6 67.209634 7 68.078521 8 66.014916 9 69.273239 68.947407 10 11 66.775191 66.775191 67.752688 13 70.142125 14 15 67.318245 16 66.557970 66.883802 17 69.490461 18 69.381850 19 66.666580 20 21 69.924904 22 69.056018 71.445455 23 24 68.078521 25 65.797694 26 69.381850 72.097119 27 28 70.250736 29 71.336844 dtype: float64 In [10]: data.plot(kind="scatter", x="Altura", y="Peso") plt.plot(pd.DataFrame(data["Altura"]),altura pred,c="green",linewidth=2) Out[10]: [<matplotlib.lines.Line2D at 0x174b4cac6d0>] 72 70 66 64 150 160 190 Altura In []: **Ejercicio Algoritmo Apriori** 1era parte: ejercicio resuelto sin libreria mixtend 2da parte: ejercicio resuelto con libreria mixtend In [1]: import numpy as np import matplotlib.pyplot as plt import pandas as pd In [2]: data = pd.read csv("Asosiacion.csv", header=None) Out[2]: 0 3 1 1 B E NaN NaN 2 C D E NaN 3 A C D NaN 4 A C E NaN A=0In [3]: B=0C=0D=0E=0print("Nivel k=1") for i in range (0, 5): for j in range (0, 4): if (data.values[i,j] == 'A'): if (data.values[i,j] == 'B'): B=B+1if (data.values[i,j] == 'C'): C=C+1if (data.values[i,j] == 'D'): D=D+1if (data.values[i,j] == 'E'): E=E+1SA=A/5SB=B/5SC=C/5SD=D/5SE=E/5print("Soporte A="+str(SA)) print("Soporte B="+str(SB)) print("Soporte C="+str(SC)) print("Soporte D="+str(SD)) print("Soporte E="+str(SE)) print("Se eliminan B y D, debido a que tienen soporte menor del umbral de .5") Nivel k=1 Soporte A=0.6 Soporte B=0.4Soporte C=0.8 Soporte D=0.4Soporte E=0.8Se eliminan B y D, debido a que tienen soporte menor del umbral de .5 In [4]: Uno = ['A','C','E'],['E'],['C','E'],['A','C'],['A','C','E'] import csv with open('Nivel2.csv', 'w', newline='') as file: writer = csv.writer(file) writer.writerow(['A','C','E','NaN']) writer.writerow(['E','NaN','NaN','NaN']) writer.writerow(['C','E','NaN','NaN']) writer.writerow(['A','C','NaN','NaN']) writer.writerow(['A','C','E','NaN']) In [5]: dos = pd.read_csv("Nivel2.csv", header=None) Out[5]: 2 3 E NaN 1 E NaN NaN NaN **2** C E NaN NaN **3** A C NaN NaN С E NaN **4** A In [6]: print("Nivel k=2") AC=0AE=0CE=0x=0y=1 for i in range (0, 5): y=1 j=0**for** j **in** range(0, 4): if (dos.values[i,j]=='A' and dos.values[i,y]=='C'): AC=AC+1if (dos.values[i,j]=='A' and dos.values[i,y]=='E'): y=y-1 AE = AE + 1if (dos.values[i,j]=='C' and dos.values[i,y]=='E'): CE=CE+1y=y+1 SAC=AC/5 SAE=AE/5 SCE=CE/5 print("SoporteAC= "+ str(SAC)) print("SoporteAE= "+ str(SAE)) print("SoporteCE= "+ str(SCE)) print ("Se eliminan AE, debido a que tiene soporte menor del umbral de .5") Nivel k=2SoporteAC= 0.6 SoporteAE= 0.4 SoporteCE= 0.6 Se eliminan AE, debido a que tiene soporte menor del umbral de .5 In []: Otra forma de solucionar el ejercicio utilizando mixtend In [11]: pip install mlxtend Requirement already satisfied: mlxtend in c:\users\carla\anaconda3\lib\site-packages (0.17.3) Requirement already satisfied: matplotlib>=3.0.0 in c:\users\carla\anaconda3\lib\site-packages (from Requirement already satisfied: scikit-learn>=0.20.3 in c:\users\carla\anaconda3\lib\site-packages (fr om mlxtend) (0.23.1) Requirement already satisfied: setuptools in c:\users\carla\anaconda3\lib\site-packages (from mlxten d) (49.2.0.post20200714) Requirement already satisfied: scipy>=1.2.1 in c:\users\carla\anaconda3\lib\site-packages (from mlxte nd) (1.5.0)Requirement already satisfied: pandas>=0.24.2 in c:\users\carla\anaconda3\lib\site-packages (from mlx α) tend) (1.0.5)Requirement already satisfied: numpy>=1.16.2 in c:\users\carla\anaconda3\lib\site-packages (from mlxt end) (1.18.5)Requirement already satisfied: joblib>=0.13.2 in c:\users\carla\anaconda3\lib\site-packages (from mlx tend) (0.16.0)Requirement already satisfied: python-dateutil>=2.1 in c:\users\carla\anaconda3\lib\site-packages (fr om matplotlib>=3.0.0->mlxtend) (2.8.1) Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\carla\anaconda3\lib\site-packages (from matplotlib >= 3.0.0 -> mlxtend) (1.2.0) Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in c:\users\carla\anaconda3\l ib\site-packages (from matplotlib>=3.0.0->mlxtend) (2.4.7) Requirement already satisfied: cycler>=0.10 in c:\users\carla\anaconda3\lib\site-packages (from matpl otlib>=3.0.0-mlxtend) (0.10.0) Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\carla\anaconda3\lib\site-packages (fr om scikit-learn>=0.20.3->mlxtend) (2.1.0) Requirement already satisfied: pytz>=2017.2 in c:\users\carla\anaconda3\lib\site-packages (from panda $s \ge 0.24.2 - mlxtend)$ (2020.1) Requirement already satisfied: six>=1.5 in c:\users\carla\anaconda3\lib\site-packages (from python-da teutil >= 2.1 - matplotlib >= 3.0.0 - mlxtend) (1.15.0) Note: you may need to restart the kernel to use updated packages. In [13]: import pandas as pd from mlxtend.frequent patterns import apriori file = pd.read_excel('ejercicio1_parte2.xlsx') file.head() Out[13]: ABCDE **0** 1 1 1 0 1 **1** 0 1 0 0 1 2 0 0 1 1 1 **3** 1 0 1 1 0 4 1 0 1 0 1 In [18]: frequent_itemsets = apriori(file, min_support=0.5, use_colnames=**True**) frequent_itemsets Out[18]: support itemsets 0.6 (A) 1 8.0 (C) 8.0 (E) 3 0.6 (A, C) 0.6 (C, E) In []: