Descripció Aprèn a gestionar paràmetres amb Python. In [171... import numpy as np import pandas as pd from sklearn.linear model import LinearRegression $\textbf{from} \ \texttt{sklearn.preprocessing} \ \textbf{import} \ \texttt{StandardScaler}$ from sklearn.preprocessing import MinMaxScaler Nivell 1 • Exercici 1 Agafa un conjunt de dades de tema esportiu que t'agradi i normalitza els atributs categòrics en dummy. Estandaritza els atributs numèrics amb StandardScaler. Se importa un dataset con los resultados de los mundiales del League of Legends de 2021. Incluye datos catégoricos (equipos, jugadores campeones usados y resultado) y numéricos (muertes, asistencias, oro ganado, daño, etc). In [172... raw_df = pd.read_csv('lol2021.csv') raw df Out [172... Champion Gold Kill Wards War Creep Player Opponent Position Champion Kills Deaths Assists **Damage** Team Score Earned **Participation Placed** Destroy **Share** UOL 5 0 Boss GS Top Camille 4 3 188 11107 0.17 0.78 8 GS Crazy UOL 9 217 12201 0.20 0.52 10 Top Gwen 2 UOL Ahahacik 2 4 5 0.15 0.78 8 GS Jungle Trundle 156 9048 12 3 GS UOL 5 4 10 194 11234 0.23 0.65 Mojito Jungle Talon UOL Nomanz GS Mid Leblanc 1 3 4 216 9245 0.29 0.56 6 ... 215 BYG UOL 2 15558 0.23 24 Maoan Mid Ryze 4 4 356 0.32 3 9 9554 0.16 0.71 35 216 UOL Argonavt BYG Adc Senna 58 20546 217 BYG Doggo UOL Adc Draven 10 2 4 353 0.38 0.74 15 218 UOL Santas BYG Support Wukong 4 7 200 10354 0.11 0.50 16 5 9 0.47 49 219 BYG Kino UOL Support Rakan 0 41 9304 0.06 220 rows × 20 columns Se normaliza los datos de muertes y torretas contruidas In [173... ss = StandardScaler() Xstd = ss.fit transform(raw df[['Kills', 'Wards Placed']].values) array([[0.50155703, -0.73062203], Out [173... [0.11302694, -0.6030531],[-0.27550316, -0.73062203],[0.89008713, -0.47548418], [-0.66403326, -0.85819095],[0.89008713, -0.15656186], [-0.66403326, -0.79440649],[2.44420752, -0.28413079], [-0.66403326, 1.31048078], [-1.05256335, 1.05534293], [-0.66403326, -0.6030531], [-0.66403326, -0.34791525],[-0.66403326, -0.53926864],[-1.05256335, -0.53926864],[-0.66403326, -0.66683756],[2.05567742, -0.6030531], [-0.66403326, -0.73062203],[0.11302694, -0.41169971], [-1.05256335, 1.62940309], [-0.66403326, 1.88454094], [1.27861723, -0.79440649], [-0.66403326, -0.41169971],[0.89008713, -0.41169971], [0.11302694, -0.41169971], [-1.05256335, -0.22034633], [-0.27550316, -0.0927774],[0.89008713, -0.22034633], [0.11302694, -0.34791525], [-0.66403326, 2.6499545], [-1.05256335, 2.58617003], [0.50155703, -0.47548418], [-0.66403326, -0.28413079],[-1.05256335, 0.09857599], [1.27861723, -0.28413079], [-0.66403326, -0.34791525],[0.50155703, -0.85819095],[-0.27550316, -0.41169971],[1.66714733, -0.41169971], [-1.05256335, 2.33103218], [-0.66403326, 3.03266127], [0.89008713, -0.47548418], [0.11302694, -0.41169971],[-0.66403326, -0.22034633],[-0.27550316, -0.6030531],[-0.66403326, -0.22034633],[-0.27550316, -0.0927774],[0.50155703, -0.28413079],[-0.66403326, -0.02899294], [-0.66403326, 1.94832541], [-1.05256335, 1.75697202], [2.05567742, -0.92197542], [-0.27550316, -0.41169971],[0.50155703, -0.85819095], [-0.66403326, -0.98575988],[-0.27550316, -0.92197542],[-1.05256335, -0.34791525],[-0.66403326, -0.47548418], [-0.27550316, -0.47548418],[-1.05256335, 0.927774], [-1.05256335, 0.80020508], [-0.66403326, -0.41169971],[-0.27550316, -0.15656186],[-0.66403326, -0.02899294],[-1.05256335, -0.28413079],[-1.05256335, 0.22614491], [0.89008713, -0.0927774], [0.50155703, -0.22034633], [1.66714733, 0.09857599], [-1.05256335, 1.50183417], [-0.66403326, 1.88454094], [-1.05256335, -0.28413079],[1.27861723, -0.0927774], [-1.05256335, 0.16236045], [-0.66403326, -0.47548418],[-0.66403326, -0.34791525],[-0.27550316, -0.6030531], [0.11302694, 0.22614491], [1.66714733, -0.34791525], [-0.66403326, 3.03266127], [-0.27550316, 4.62727284], [0.50155703, -0.6030531], [0.11302694, -0.73062203],[1.66714733, -0.53926864], [-0.66403326, -0.79440649],[0.89008713, -0.6030531], [0.11302694, -0.34791525], [1.27861723, -0.47548418],[-0.27550316, -0.41169971], [-1.05256335, 2.07589433], [-1.05256335, 1.24669632], [0.89008713, -0.6030531], [0.50155703, -0.79440649], [0.89008713, -0.85819095], [-1.05256335, -0.66683756],[-0.66403326, -0.6030531],[-0.66403326, -1.04954434],[2.05567742, -0.66683756], [-0.66403326, -0.73062203], [-1.05256335, 0.80020508], [-1.05256335, 0.73642062], [-1.05256335, -0.47548418],[0.50155703, -0.47548418],[-1.05256335, -0.34791525],[0.89008713, -0.6030531],[-1.05256335, -0.41169971],[-0.66403326, -0.28413079],[-1.05256335, -0.53926864],[0.50155703, -0.53926864], [-0.27550316, 1.75697202], [0.11302694, 0.99155847], [0.11302694, -0.0927774], [0.50155703, -0.34791525],[0.50155703, -0.6030531],[0.50155703, -0.0927774],[-0.27550316, -0.66683756],[1.27861723, -0.15656186], [1.66714733, -0.28413079], [2.44420752, -0.28413079], [-1.05256335, 2.71373896], [0.11302694, 2.90509235], [0.11302694, -0.28413079],[0.50155703, -0.53926864],[-0.66403326, -0.53926864],[0.50155703, -0.34791525],[0.11302694, -0.47548418],[-0.66403326, -0.6030531],[-0.27550316, -0.28413079],[2.44420752, -0.28413079], [-0.66403326, 1.50183417], [0.11302694, 1.24669632], [3.99832791, -0.47548418], [0.50155703, -0.02899294],[-0.66403326, -0.15656186],[2.05567742, -0.92197542], [-0.66403326, 0.09857599], [1.27861723, -0.28413079], [-0.27550316, -0.28413079],[0.11302694, -0.53926864], [-1.05256335, 2.71373896], [-1.05256335, 3.1602302], [-1.05256335, -0.41169971],[0.50155703, -0.47548418],[-0.66403326, -0.53926864],[0.11302694, -0.79440649],[-0.66403326, -0.22034633],[-0.27550316, -0.34791525],[-0.27550316, 0.03479153], [1.27861723, -0.41169971], [-0.66403326, 1.11912739], [-1.05256335, 1.94832541], [-1.05256335, -0.22034633],[0.50155703, -0.41169971], [0.50155703, -0.85819095],[2.05567742, -0.47548418], [-0.66403326, -0.41169971],[0.11302694, -0.22034633],[-0.66403326, -0.47548418],[1.27861723, -0.0927774], [-1.05256335, 1.82075648], [-1.05256335, 2.01210987], [0.50155703, -0.66683756],1.27861723, -0.6030531], [-0.66403326, -0.92197542],[0.11302694, -0.53926864],[0.11302694, -0.79440649],[1.27861723, -0.85819095], [-0.27550316, -0.6030531],[-0.27550316, 0.16236045], [-0.66403326, 0.80020508], [-0.27550316, -0.53926864],[2.83273762, -0.73062203], [-0.66403326, -0.34791525],[0.50155703, -0.41169971],[-1.05256335, -0.41169971],[-0.66403326, -0.34791525],[-0.27550316, -0.73062203],[1.66714733, -0.6030531], [-0.66403326, -0.22034633],[-1.05256335, 1.75697202], [-1.05256335, 1.37426524], [1.66714733, -0.02899294], [0.11302694, -0.53926864],[-0.27550316, -1.04954434],[0.50155703, -0.53926864],[2.05567742, -0.28413079], [-0.66403326, -0.34791525],[2.05567742, -0.47548418], [-0.66403326, -0.47548418],[-1.05256335, 1.56561863], [-1.05256335, 0.927774], [2.05567742, -0.85819095], [-0.66403326, -0.73062203],[1.66714733, -0.92197542], [-0.66403326, -0.92197542],[0.11302694, -0.66683756],[-0.66403326, -0.47548418],[-0.27550316, -0.66683756],[-1.05256335, -0.53926864],[-0.66403326, 0.80020508], [-1.05256335, 1.24669632], [0.11302694, -0.66683756], [-0.27550316, -0.0927774],[0.89008713, -0.34791525],[1.66714733, -0.66683756], [1.27861723, -0.15656186], [-1.05256335, -0.22034633],[1.27861723, -0.6030531], [-0.27550316, -0.47548418],[-1.05256335, 2.84130788], [-0.66403326, 2.45860111], [0.89008713, -0.15656186], [0.50155703, -0.41169971],[0.11302694, -0.02899294],[-0.66403326, -0.79440649],[0.89008713, -0.53926864], [0.50155703, 0.28992938], [-0.66403326, 0.99155847], [2.83273762, -0.28413079], [-1.05256335, -0.22034633], [-1.05256335, 1.88454094]]) In [174... print (Xstd.mean()) print (Xstd.std()) 5.652044489000796e-17 1.0 In [175... x = raw_df[['Kills', 'Creep Score', 'Gold Earned']] mms = MinMaxScaler() $Xn = mms.fit_transform(x)$ In [176... Xn.min(axis=0) array([0., 0., 0.]) In [177... Xn.max(axis=0)array([1., 1., 1.]) Out [177... Se muestra un nuevo DF con los datos categórigos tratados con libreria Dummie, generando nuevas columnas para los equipos y los resultados del torneo. In [178... dummies = pd.get_dummies(raw_df, columns=["Team", "Result"]) print("Original DataFrame:") raw df Original DataFrame: Out [178... Champion Kill Wards Creep Gold War Player Opponent Position Champion Kills Deaths Assists **Damage** Team Score **Earned Participation** Placed Destroy **Share** UOL Camille 0 Boss GS Top 4 5 3 188 11107 0.17 0.78 8 9 10 1 GS UOL 3 1 217 12201 0.20 0.52 Crazy Top Gwen 2 UOL Ahahacik GS Jungle 2 4 5 156 9048 0.15 0.78 8 Trundle 3 GS Mojito UOL Jungle Talon 5 4 10 194 11234 0.23 0.65 12 3 0.29 4 UOL Nomanz GS Mid Leblanc 1 4 216 9245 0.56 6 • • • • • • 215 BYG UOL Mid 4 4 2 356 15558 0.23 0.32 24 Maoan Ryze **UOL** Argonavt **BYG** 3 9 0.71 216 Adc Senna 58 9554 0.16 35 217 BYG Doggo UOL Adc Draven 10 2 4 353 20546 0.38 0.74 15 218 UOL Santas **BYG** Support 4 7 200 10354 0.11 0.50 Wukong 0 16 219 BYG Kino 0 5 9 41 9304 0.06 0.47 49 UOL Support Rakan 220 rows × 20 columns In [179... print("DataFrame with Dummies:") dummies DataFrame with Dummies: Out [179... Champion Creep Gold Player Opponent Position Champion Kills Deaths Assists Damage ... Team_DFM Team_GS Team_HLE Score Earned Share 0 Boss GS Camille 4 5 3 188 11107 0.17 ... 0 0 0 Top 1 Crazy UOL Top Gwen 3 1 9 217 12201 0.20 0 0 0 2 Ahahacik GS Jungle Trundle 2 4 5 156 9048 0.15 0 0 3 Mojito UOL Jungle Talon 5 4 10 194 11234 0.23 0 1 4 Leblanc 3 4 216 9245 0.29 ... 0 0 0 Nomanz GS Mid 1 ... 215 Maoan UOL Mid Ryze 4 4 2 356 15558 0.23 ... 0 0 0 0 Argonavt BYG 3 9 58 0 0 216 Adc Senna 1 9554 0.16 ... 0.38 ... 217 UOL 10 2 4 353 20546 0 0 0 Doggo Adc Draven 218 Support 4 200 0 0 0 Santas BYG Wukong 0 10354 0.11 0 0 219 Kino UOL Support Rakan 0 5 9 41 9304 0.06 ... 0 220 rows × 30 columns Nivell 2 • Exercici 2 Continua amb el conjunt de dades de tema esportiu que t'agradi i aplica l'anàlisi de components principals. Nivell 3 • Exercici 3 Continua amb el conjunt de dades de tema esportiu que t'agradi i normalitza les dades tenint en compte els outliers.