In [1]: #j'importe les libraries necessaires import pandas as pd import numpy as np import matplotlib.pylab as plt #pour les analyses des données et representation graphic from datetime import datetime #gestion des dates et des temps import seaborn as sns #pour les analyses des données et representation graphic from scipy import stats as st #pour les analyses statistics Je telecharge les données pour faire les verifications habituelles: recherches des NAN ou valeurs manquants ou null II y a 3 fichiers: customers, products et transactions In [2]: cust = pd.read_csv("customers.csv") cust.head() Out[2]: client_id sex birth c_4410 f 1967 c_7839 f 1975 c_1699 f 1984 c_5961 f 1962 c_5320 m 1943 In [3]: cust.describe(include = "all") Out[3]: birth client_id sex 8623 8623 8623.000000 count 2 unique 8623 NaN NaN top c_325 1 4491 NaN freq 1978.280877 NaN mean NaN NaN 16.919535 std NaN 1929.000000 min NaN NaN 25% NaN NaN 1966.000000 50% NaN 1979.000000 NaN **75%** NaN 1992.000000 NaN NaN 2004.000000 max NaN cust.isna().any() client id False Out[4]: False sex birth False dtype: bool In [5]: cust.isnull().any() client id False Out[5]: False sex False birth dtype: bool In [6]: prod = pd.read csv("products.csv") prod.head() Out[6]: id_prod price categ 0_1421 19.99 0 0_1368 5.13 0 0_731 17.99 0 1_587 4.99 1 0_1507 3.99 0 prod.describe(include = "all") In [7]: #describe permet une analyse de statitstique basic. include "all" permet de inclure toutes les colonnes, #des variables quantitatifs comme qualitatifs en meme temps. Sinon on peut voir que les colonnes quantitatifs. Out[7]: id_prod price categ 3287 3287.000000 3287.000000 count NaN unique 3287 NaN 2_125 NaN NaN top freq 1 NaN NaN 21.856641 0.370246 NaN mean 29.847908 NaN 0.615387 std 0.000000 NaN -1.000000 min 25% NaN 6.990000 0.000000 50% NaN 13.060000 0.000000 **75**% NaN 22.990000 1.000000 300.000000 2.000000 max NaN il ya des prix negatifs In [8]: # je visualize les valeurs < 0</pre> prod_neg = prod[prod["price"] < 0]</pre> prod_neg.head() Out[8]: id_prod price categ 0 731 T_0 -1.0 In [9]: prod neg.shape #shape permet de voir la taille du df Out[9]: (1, 3) In [10]: prod.isna().any() #detection des Nan id_prod False Out[10]: price False categ False dtype: bool prod.isnull().any() #detection des valeurs 0 id prod False Out[11]: price False False categ dtype: bool In [12]: | transac = pd.read_csv("transactions.csv") transac.head() Out[12]: id_prod date session_id client_id 2021-04-10 18:37:28.723910 s_18746 c_4450 0_1483 2_226 2022-02-03 01:55:53.276402 s_159142 c_277 c_4270 2021-09-23 15:13:46.938559 s_94290 2021-10-17 03:27:18.783634 s_105936 c_4597 0_2186 s_63642 c_1242 transac.describe(include = "all") Out[13]: id_prod date session_id client_id **count** 337016 337016 337016 337016 3266 336855 169195 8602 unique 1_369 test_2021-03-01 02:30:02.237413 s_0 c_1609 1081 13 200 12855 freq il y a un format de date test, donc inhabituel In [14]: | transac.isna().any() id_prod False Out[14]: date False session id False client_id False dtype: bool In [15]: transac.isnull().any() id prod False date False session id False client id False dtype: bool In [16]: #je visualise les dates "test" transac test = transac[transac['date'].str.contains("test")] transac_test.head(40) Out[16]: id_prod date session_id client_id T_0 test_2021-03-01 02:30:02.237420 1431 s_0 ct_1 2365 T_0 test_2021-03-01 02:30:02.237446 s_0 ct_1 2895 T_0 test_2021-03-01 02:30:02.237414 s_0 ct_1 5955 T_0 test_2021-03-01 02:30:02.237441 ct_0 s_0 7283 T_0 test_2021-03-01 02:30:02.237434 s_0 ct_1 13745 T_0 test_2021-03-01 02:30:02.237443 s_0 ct_0 20470 T_0 test_2021-03-01 02:30:02.237442 s_0 ct_0 22347 T_0 test_2021-03-01 02:30:02.237412 ct_1 s_0 26359 T_0 test_2021-03-01 02:30:02.237439 s_0 ct_1 26407 T_0 test_2021-03-01 02:30:02.237426 ct_0 s_0 26814 T_0 test_2021-03-01 02:30:02.237447 s_0 ct_1 27161 T_0 test_2021-03-01 02:30:02.237434 ct_0 s_0 32675 T_0 test_2021-03-01 02:30:02.237438 s_0 ct_0 T_0 test_2021-03-01 02:30:02.237443 34387 ct_0 s_0 36700 T_0 test_2021-03-01 02:30:02.237437 s_0 ct_0 38779 T_0 test_2021-03-01 02:30:02.237450 s_0 ct_0 39224 T_0 test_2021-03-01 02:30:02.237429 s_0 ct_0 T_0 test_2021-03-01 02:30:02.237435 40476 ct_0 s_0 47170 T_0 test_2021-03-01 02:30:02.237430 s_0 ct_1 48425 T_0 test_2021-03-01 02:30:02.237443 ct_1 s_0 54813 T_0 test_2021-03-01 02:30:02.237412 s_0 ct_1 56373 T_0 test_2021-03-01 02:30:02.237446 ct_0 s_0 57261 T_0 test_2021-03-01 02:30:02.237439 s_0 ct_1 57682 T_0 test_2021-03-01 02:30:02.237419 ct_0 s_0 57755 T_0 test_2021-03-01 02:30:02.237413 ct_1 s_0 58802 T_0 test_2021-03-01 02:30:02.237429 s_0 ct_0 59043 T_0 test_2021-03-01 02:30:02.237413 ct_0 59422 T_0 test_2021-03-01 02:30:02.237445 s_0 ct_0 ct_0 60170 T_0 test_2021-03-01 02:30:02.237446 s_0 60485 T_0 test_2021-03-01 02:30:02.237438 s_0 ct_1 62365 T_0 test_2021-03-01 02:30:02.237412 s_0 ct_1 70341 T_0 test_2021-03-01 02:30:02.237424 s_0 ct_1 71097 T_0 test_2021-03-01 02:30:02.237446 s_0 ct_0 72068 T_0 test_2021-03-01 02:30:02.237427 s_0 ct_0 72102 T_0 test_2021-03-01 02:30:02.237428 s_0 ct_0 72405 T_0 test_2021-03-01 02:30:02.237424 s_0 ct_1 72850 T_0 test_2021-03-01 02:30:02.237433 s_0 ct_1 75176 T_0 test_2021-03-01 02:30:02.237443 s_0 ct_1 76735 T_0 test_2021-03-01 02:30:02.237442 s_0 ct_0 77758 T_0 test_2021-03-01 02:30:02.237429 s_0 ct_1 transac test.id prod.unique() #unique permet de visualiser les differents objets dans une colonne array(['T_0'], dtype=object) Out[17]: In [18]: transac_test.session_id.unique() array(['s_0'], dtype=object) Out[18]: In [19]: transac_test.client_id.unique() array(['ct_1', 'ct_0'], dtype=object) Out[19]: Resume: - Dans les 3 df pas de valeurs manquants, ou NaN, ni 0. - Dans la df products il y une ligne de prix negatif qui correspond au id_prod T_0 et categ = 0. il faut exclure les prix < 0 - Dans le df transactions il y a un format de date inhabituel contenant le prefix 'test'. Il y a 200 lignes. Il est associé avec id_prod = T_0 et session_id = s_0 et client_id = ct_0/ct_1. Le id_prod T_ 0 est associé uvec un prix = -1 dans la table products. J'exclus donc. In [20]: #j'exclus le prix = -1 dans products prod1 = prod[prod["price"] != -1] prod1.describe(include = "all") Out[20]: id_prod price categ 3286 3286.000000 3286.000000 count 3286 NaN NaN unique 2_125 NaN NaN top 1 NaN NaN freq 0.370359 21.863597 NaN mean 29.849786 NaN 0.615446 std 0.620000 0.000000 min NaN 25% NaN 6.990000 0.000000 13.075000 0.000000 50% NaN 22.990000 1.000000 **75%** NaN max NaN 300.000000 2.000000 #j"exclus les lignes avec les dates test... transT = transac[transac["id_prod"] != "T_0"] transT.describe(include = "all") Out[21]: id_prod date session_id client_id **count** 336816 336816 336816 336816 3265 336816 169194 8600 unique 1_369 2021-03-01 04:48:32.880371 s_118668 c_1609 top 1081 1 12855 freq 14 In [22]: #je verfie si il reste des lignes avec une date = 'test..' transT[transT['date'].str.contains("test")] Out[22]: id_prod date session_id client_id Exactement 200 lignes correspondent aux lignes date = 'test...' sont exclus dans la table transT In [23]: #Je procede au merge des tables products et transactions sur id_prod tp = pd.merge(transT, prod1, on = 'id_prod', how = 'left') tp.head() Out[23]: date session_id client_id price categ id_prod s_18746 c_4450 0_1483 2021-04-10 18:37:28.723910 4.99 0.0 s_159142 c_277 65.75 2.0 1_374 2021-09-23 15:13:46.938559 s_94290 c_4270 10.71 1.0 0_2186 2021-10-17 03:27:18.783634 s_105936 c_4597 4.20 0.0 0_1351 2021-07-17 20:34:25.800563 s_63642 c_1242 8.99 0.0 In [24]: | tp.describe(include = "all") Out [24]: date session_id client_id price id_prod categ **count** 336816 336816 336816 336713.000000 336713.000000 336816 336816 unique 3265 169194 8600 NaN NaN NaN NaN 1081 12855 NaN NaN freq 0.430156 NaN NaN NaN 17.215189 mean NaN 17.855445 0.591082 NaN NaN NaN NaN std 0.000000 min NaN NaN NaN NaN 0.620000 25% NaN NaN 8.610000 0.000000 NaN NaN **50%** NaN NaN NaN NaN 13.900000 0.000000 **75%** NaN NaN NaN NaN 18.990000 1.000000 NaN NaN 2.000000 max NaN NaN 300.000000 Absence de lignes test ou de prix = -1In [25]: | tp.isnull().any() id_prod False Out [25]: date False session_id False client_id False price True categ True dtype: bool In [26]: | tp.isna().any() False id_prod Out[26]: date False session_id False client_id False price True categ True dtype: bool In [27]: #Il y a des NaN dans price et categ. Je les visualize. tp_isna = tp[tp.price.isna() == True] tp_isna.head() Out[27]: date session_id client_id price categ id_prod 0_2245 2021-06-17 03:03:12.668129 6231 s_49705 c_1533 NaN NaN 10797 0_2245 2021-06-16 05:53:01.627491 s_49323 c_7954 NaN NaN 14045 0_2245 2021-11-24 17:35:59.911427 s_124474 c_5120 NaN NaN 17480 0_2245 2022-02-28 18:08:49.875709 s_172304 c_4964 NaN NaN c_580 **21071** 0_2245 2021-03-01 00:09:29.301897 s_3 NaN NaN tp_isna.describe(include = "all") date session_id client_id price categ Out[28]: id_prod count 1 103 103 100 NaN NaN unique 0_2245 2022-01-15 06:19:44.035000 s_167890 c_1025 NaN NaN 103 1 1 2 NaN NaN freq NaN NaN mean NaN std NaN NaN NaN NaN NaN min NaN 25% NaN NaN NaN NaN NaN NaN 50% NaN NaN NaN NaN NaN NaN NaN **75**% NaN max NaN In [29]: tp_isna.id_prod.unique() array(['0_2245'], dtype=object) Out[29]: In [30]: tp_isnull = tp[tp.price.isnull() == True] tp_isnull.head() date session_id client_id price categ Out[30]: id_prod 0_2245 c_1533 2021-06-17 03:03:12.668129 s_49705 NaN NaN 2021-06-16 05:53:01.627491 10797 0_2245 s_49323 c_7954 NaN NaN 0_2245 2021-11-24 17:35:59.911427 14045 s_124474 c_5120 NaN NaN 17480 s_172304 c_4964 NaN NaN **21071** 0_2245 2021-03-01 00:09:29.301897 s_3 c_580 NaN NaN In [31]: tp_isnull.describe(include = "all") Out[31]: date session_id client_id price categ id_prod 103 0.0 0.0 103 103 103 count 103 103 unique 100 NaN NaN s_167890 0_2245 2022-01-15 06:19:44.035000 c_1025 NaN NaN 103 1 1 2 NaN NaN freq NaN NaN mean NaN NaN NaN NaN NaN NaN std NaN NaN NaN NaN min NaN NaN NaN NaN NaN NaN **25%** NaN NaN NaN NaN NaN NaN **50**% NaN NaN NaN NaN NaN NaN **75%** NaN NaN NaN NaN NaN NaN NaN NaN max NaN NaN NaN NaN In [32]: tp_isnull.id_prod.unique() array(['0_2245'], dtype=object) Out[32]: Resumé: - les prix et categ manquants sont associés exclusivement avec le id_prod = 0_2245 - le id_prod = 0_2245 appartient a la categorie 0 a cause du prefix 0_, j'attribue donc 0 pour la categorie 0 actegorie j'attribue le prix moyenne de la categorie 0 comme prix pour le id_prod = 0_2245 In []: #Je determine le prix moyen pour la categorie 0 afin de l'attribuer au id prod = 0 2245 tp_cat0 = tp[tp["categ"] == 0.0] tp_cat0.head() date session_id client_id price categ Out[33]: id_prod 0_1483 2021-04-10 18:37:28.723910 s_18746 c_4450 4.99 0.0 0_2186 2021-10-17 03:27:18.783634 s_105936 c_4597 4.20 0.0 2021-07-17 20:34:25.800563 0.0 0_1351 s_63642 c_1242 8.99 0_1085 2021-09-15 05:47:48.215162 s_90139 c_2526 3.99 0.0 c_5799 s_82100 8.03 0.0 In [34]: | price_mean_cat0 = tp_cat0["price"].mean() price_mean_cat0 10.646828235274288 Out[34]: In [35]: #J'ajoute le id prod = 0 2245 a la table initiale products' en utilisant la method 'append'. prod_02245 = prod1.append(pd.DataFrame([["0_2245", 10.65, 0.0]],columns=['id_prod', 'price', 'categ'])) prod 02245.describe(include = "all") Out[35]: id_prod price categ 3287 3287.000000 3287.000000 count 3287 NaN NaN unique 2_125 NaN NaN top freq 1 NaN NaN 21.860186 0.370246 NaN mean 29.845884 0.615387 NaN std 0.620000 0.000000 NaN min 0.000000 25% 6.990000 NaN 50% NaN 13.060000 0.000000 **75**% 22.990000 1.000000 NaN 300.000000 2.000000 NaN max In [36]: #Je verifie si la ligne etait bien ajouté prod_02245_T = prod_02245[prod_02245["id_prod"] == "0_2245"] prod_02245_T Out[36]: id_prod price categ **0** 0_2245 10.65 0.0 Comme indiqué dans la ligne top id_prod etait bien ajouté au df products et la table products est passée de 3286 ligne a 3287 In [37]: #Je procede au merge des tables products ou prodT(contenant le id prod 0 2245) et transactions sur id prod tp2 = pd.merge(transT, prod_02245, on = 'id_prod', how = 'left') tp2.head() Out[37]: id_prod date session_id client_id price categ 0_1483 2021-04-10 18:37:28.723910 s_18746 c_4450 0.0 4.99 2_226 2022-02-03 01:55:53.276402 s_159142 c_277 65.75 2.0 2021-09-23 15:13:46.938559 s_94290 c_4270 10.71 1.0 1_374 c_4597 0.0 0_2186 2021-10-17 03:27:18.783634 s_105936 4.20 0_1351 2021-07-17 20:34:25.800563 s_63642 c_1242 8.99 0.0 tp2.describe(include = 'all') In [38]: Out[38]: date session_id client_id categ id_prod price **count** 336816 336816 336816.000000 336816.000000 336816 336816 8600 3265 336816 169194 NaN unique NaN 1_369 2021-03-01 04:48:32.880371 s_118668 c_1609 NaN NaN top 1081 1 12855 NaN freq 14 NaN 17.213181 0.430024 NaN NaN NaN NaN mean 0.591039 std NaN NaN 17.853084 NaN NaN 0.620000 0.000000 min NaN NaN NaN NaN 8.610000 25% NaN NaN NaN NaN 0.000000 **50%** NaN 13.900000 0.000000 NaN NaN NaN 1.000000 **75%** NaN NaN 18.990000 NaN NaN 300.000000 2.000000 NaN NaN NaN NaN max In [39]: tp2.isnull().any() id prod False Out[39]: date False session_id False client id False price False categ False dtype: bool In [40]: | tp2.isna().any() id_prod False Out [40]: False date session_id False client id False price False categ False dtype: bool pas de données manquantes dans tp2. Je procede au merge avec le df customers Pour rapelle le df transactions contenait une date test associé avec le client_id ct_0/ct_1 Avant le merge je verifie si customers contient ces client_id. si oui je vais les supprimers avant le merge. In [41]: #ctp evol10 = ctp[(ctp.date > startdate) & (ctp.date < enddate)]</pre> cust_ct01 = cust[(cust.client_id == "ct_0") | (cust.client_id == "ct_1")] cust_ct01.head() client_id sex birth Out[41]: 2735 f 2001 ct_0 8494 ct_1 m 2001 In [42]: cust_ct01.shape Out[42]: (2, 3) la table customers contient bien deux lignes de client_id ct0 et c1. il faut les eclure car ces client_id sont exclus dans les autres tables In [43]: custT = cust[(cust["client_id"] != "ct_0") & (cust["client_id"] != "ct_1")] custT.shape (8621, 3) Out[43]: In [44]: cust.shape (8623, 3) Out[44]: In [45]: #je verifie si ct0/1 etaient bien exclus custT_ct01 = custT[(custT.client_id == "ct_0") | (custT.client_id == "ct_1")] custT_ct01 Out[45]: client_id sex birth Les client_id ct0/1 n'apparaissent plus dans la table customers In [46]: #je procede au merge de la table tp avec la table custT(sans ct0/1) ctp = pd.merge(tp2, custT, on='client_id', how='left') ctp.head() Out[46]: id_prod date session_id client_id price categ sex birth 0_1483 2021-04-10 18:37:28.723910 c_4450 4.99 0.0 f 1977 s_18746 f 2000 2_226 2022-02-03 01:55:53.276402 s_159142 c_277 65.75 2.0 1_374 2021-09-23 15:13:46.938559 s_94290 c_4270 10.71 1.0 f 1979 m 1963 0_2186 2021-10-17 03:27:18.783634 s_105936 c_4597 4.20 0.0 s_63642 c_1242 8.99 0.0 f 1980 0_1351 2021-07-17 20:34:25.800563 ctp.describe(include = "all") date session_id client_id Out[47]: price birth id_prod categ sex count 336816 336816 336816 336816 336816.000000 336816.000000 336816 336816.000000 3265 336816 169194 8600 2 unique NaN NaN NaN c_1609 1_369 2021-03-01 04:48:32.880371 s_118668 NaN NaN m NaN top 1081 1 14 12855 NaN NaN 169241 NaN freq NaN 17.213181 0.430024 1977.823396 mean NaN NaN NaN NaN std NaN NaN NaN NaN 17.853084 0.591039 NaN 13.523923 NaN 0.620000 0.000000 1929.000000 min NaN NaN NaN NaN 25% NaN NaN 8.610000 0.000000 1971.000000 NaN NaN NaN 50% 0.000000 NaN NaN NaN NaN 13.900000 NaN 1980.000000 **75%** NaN NaN NaN NaN 18.990000 1.000000 NaN 1987.000000 NaN NaN NaN NaN 300.000000 2.000000 NaN 2004.000000 max In [48]: ctp.isnull().any() id_prod False Out [48]: date False session id False client id False price False categ False False sex birth False dtype: bool In [49]: ctp.isna().any() False id_prod Out [49]: date False session_id client_id False price False categ False False sex birth False dtype: bool la nouvelle table ne contient plus des valeurs manquantes. In [50]: #exportation des données du bilans alimentaires hors sousnutrition ctp.to_csv('ctp_partie2.csv', index=False) In []: