In [1]: **import** pandas **as** pd populations = pd.read_csv("population.csv") In [6]: populations Country (or dependency) Population (2020) Yearly Change Net Change Density (P/Km²) Land Area (Km²) Migrants (net) Fert. Rate Med. Age Urban Pop % World Share Out[6]: 0 0.39% 5540090 153 9388211 -348399.0 1.7 38 61% 18.47% China 1440297825 13586631 464 2973190 -532687.0 28 35% 1 India 1382345085 0.99% 2.2 17.70% 2 **United States** 331341050 0.59% 1937734 36 9147420 954806.0 1.8 38 83% 4.25% 3 274021604 1.07% 2898047 151 1811570 -98955.0 30 56% 3.51% Indonesia 2.3 4 Pakistan 221612785 2.00% 4327022 287 770880 -233379.0 3.6 23 35% 2.83% 3 50 230 Montserrat 4993 0.06% 100 NaN N.A. 10% 0.00% N.A. 231 3497 103 0 12170 66% 0.00% Falkland Islands 3.05% NaN N.A. N.A. 6 232 1628 0.68% 11 260 N.A. 46% 0.00% Niue NaN N.A. 233 1360 1.27% 17 136 10 0% 0.00% Tokelau NaN N.A. N.A. 234 Holy See 801 0.25% 2 2003 0 N.A. N.A. N.A. 0.00% NaN 235 rows × 11 columns Describing data In [7]: populations.dtypes Country (or dependency) object Out[7]: Population (2020) int64 Yearly Change object Net Change int64 Density (P/Km²) int64 Land Area (Km²) int64 Migrants (net) float64 Fert. Rate object Med. Age object Urban Pop % object World Share object dtype: object In [10]: populations.index RangeIndex(start=0, stop=235, step=1) Out[10]: populations.describe() In [11]: Out[11]: Population (2020) Net Change Density (P/Km²) Land Area (Km²) Migrants (net) count 2.350000e+02 2.350000e+02 235.000000 2.350000e+02 201.000000 3.322744e+07 3.460878e+05 475.770213 5.535918e+05 6.283582 mean std 1.353034e+08 1.128260e+06 2331.285935 1.687796e+06 123291.887548 min 8.010000e+02 -3.838400e+05 0.000000 0.000000e+00 -653249.000000 -10047.000000 25% 3.994905e+05 4.240000e+02 37.000000 2.545000e+03 5.460109e+06 3.917000e+04 **50**% 95.000000 7.724000e+04 -852.000000 4.038200e+05 **75**% 2.067170e+07 2.496600e+05 239.500000 9741.000000 1.440298e+09 1.358663e+07 26337.000000 1.637687e+07 954806.000000 max In [12]: populations.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 235 entries, 0 to 234 Data columns (total 11 columns): Non-Null Count Dtype Column -----0 Country (or dependency) 235 non-null object 235 non-null 1 Population (2020) int64 2 Yearly Change 235 non-null object 3 Net Change 235 non-null int64 Density (P/Km²) 235 non-null 4 int64 235 non-null Land Area (Km²) int64 float64 Migrants (net) 201 non-null Fert. Rate 235 non-null object 8 Med. Age 235 non-null object Urban Pop % 235 non-null object World Share 235 non-null object dtypes: float64(1), int64(4), object(6) memory usage: 20.3+ KB In [11]: populations.mean() C:\Users\Blessing N\AppData\Local\Temp\ipykernel_15588\94686229.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is depre cated; in a future version this will raise TypeError. Select only valid columns before calling the reduction. populations.mean() Population (2020) 3.322744e+07 Out[11]: Net Change 3.460878e+05 Density (P/Km²) 4.757702e+02 Land Area (Km²) 5.535918e+05 Migrants (net) 6.283582e+00 dtype: float64 In [13]: populations.sum() Country (or dependency) ChinaIndiaUnited StatesIndonesiaPakistanBrazil... Out[13]: Population (2020) 7808449406 Yearly Change 0.39%0.99%0.59%1.07%2.00%0.72%2.58%1.01%0.04%1... Net Change 81330639 Density (P/Km²) 111806 Land Area (Km²) 130094083 Migrants (net) 1263.0 Fert. Rate 1.72.21.82.33.61.75.42.11.82.11.44.32.63.32.16... Med. Age 3828383023331828402948192625321732324640404247... Urban Pop % $61\%35\%83\%56\%35\%88\%52\%39\%74\%84\%92\%21\%47\%43\%38\%4\dots$ World Share 18.47%17.70%4.25%3.51%2.83%2.73%2.64%2.11%1.87... dtype: object len(populations) In [14]: Out[14]: populations.plot() <AxesSubplot:> Out[15]: Population (2020) 1.4 Net Change 1.2 Density (P/Km2) Land Area (Km2) 1.0 Migrants (net) 0.8 0.6 0.4 0.2 0.0 100 150 50 200 In [16]: populations 0 1440297825 -348399.0 18.47% China 0.39% 5540090 153 9388211 1.7 38 61% -532687.0 1 1382345085 13586631 464 2973190 28 35% 17.70% India 0.99% 2.2 2 **United States** 331341050 0.59% 1937734 36 9147420 954806.0 1.8 38 83% 4.25% 3 1.07% 2898047 151 1811570 30 56% 3.51% Indonesia 274021604 -98955.0 2.3 4 Pakistan 221612785 2.00% 4327022 287 770880 -233379.0 3.6 23 35% 2.83% 230 Montserrat 4993 0.06% 3 50 100 NaN N.A. N.A. 10% 0.00% 231 Falkland Islands 3497 3.05% 103 0 12170 NaN N.A. N.A. 66% 0.00% 232 6 0.00% Niue 1628 0.68% 11 260 NaN N.A. N.A. 46% 17 0.00% 233 Tokelau 1360 1.27% 136 10 0% NaN N.A. N.A. 2 234 Holy See 801 0.25% 2003 0 0.00% NaN N.A. N.A. N.A. 235 rows × 11 columns $\#populations['Urban Pop \%'] = populations['Urban Pop \%'].object.replace('[\\%\,\.]','').astype(int)$ In [17]: populations.dropna(inplace=True) populations In [19]: Out[19]: Country (or dependency) Population (2020) Yearly Change Net Change Density (P/Km²) Land Area (Km²) Migrants (net) Fert. Rate Med. Age Urban Pop % World Share 18.47% 0 1440297825 5540090 9388211 -348399.0 38 61% China 0.39% 153 1.7 1 India 1382345085 0.99% 13586631 464 2973190 -532687.0 2.2 28 35% 17.70% 2 **United States** 331341050 0.59% 1937734 36 9147420 954806.0 83% 4.25% 1.8 38 3 Indonesia 274021604 1.07% 2898047 151 1811570 -98955.0 2.3 30 56% 3.51% 4 Pakistan 221612785 2.00% 4327022 287 770880 -233379.0 23 35% 2.83% 3.6 196 106845 452 593 41 44% 0.00% Aruba 0.43% 180 201.0 1.9 197 105901 1.15% 1201 147 720 -800.0 3.6 22 24% 0.00% Tonga 198 U.S. Virgin Islands 104398 -0.15% -153 298 350 -451.0 2 43 96% 0.00% 199 98453 0.62% 608 214 460 -200.0 2.5 34 56% 0.00% Seychelles 0.00% 200 98069 0.84% 811 223 440 0.0 2 34 26% Antigua and Barbuda 201 rows × 11 columns In [20]: populations.plot() <AxesSubplot:> Out[20]: Population (2020) 1.4 Net Change 1.2 Density (P/Km²) Land Area (Km2) 1.0 Migrants (net) 0.8 0.6 0.4 0.2 0.0 100 125 150 175 In [21]: populations.sample(frac=0.5) Country (or dependency) Population (2020) Yearly Change Net Change Density (P/Km²) Land Area (Km²) Migrants (net) Fert. Rate Med. Age Urban Pop % World Share Out[21]: 185 Mayotte 273905 2.50% 6665 375 3.7 20 46% 0.00% 198 U.S. Virgin Islands 104398 -0.15% 298 350 43 96% 0.00% -153 -451.0 2 62 Chile 19144605 0.87% 164163 26 743532 111708.0 1.7 35 85% 0.25% 33 44019263 2.42% 1036022 25 1765048 -50000.0 20 35% Sudan 4.4 0.56% 117 Slovakia 5460109 0.05% 2629 114 48088 1485.0 1.5 41 54% 0.07% 87 69 Senegal 16816539 2.75% 447563 192530 -20000.0 4.7 19 49% 0.21% 195 St. Vincent & Grenadines 111002 0.32% 351 284 390 -200.0 1.9 33 53% 0.00% 36 Afghanistan 39074280 2.33% 886592 60 652860 -62920.0 4.6 18 25% 0.50% 37799407 0.89% 331107 4 9093510 242032.0 81% 0.48% 38 Canada 41 1.5 31 63 Kazakhstan 18815231 1.21% 225280 7 2699700 -18000.0 2.8 58% 0.24% 100 rows × 11 columns populations.sample(frac=1) In [22]: Country (or dependency) Population (2020) Yearly Change Net Change Density (P/Km²) Land Area (Km²) Migrants (net) Fert. Rate Med. Age Urban Pop % World Share Out[22]: 136 0.04% Armenia 2964219 0.19% 5512 104 28470 -4998.0 1.8 35 63% 28 Colombia 50976248 1.08% 543448 46 1109500 204796.0 1.8 31 80% 0.65% 133 3475842 0.35% 11996 20 175020 -3000.0 2 36 96% 0.04% Uruguay 93 Hungary 9655983 -0.25% -24328 107 90530 6000.0 1.5 43 72% 0.12% 84 1.01% 225 48320 -30000.0 28 85% 0.14% Dominican Republic 10866667 108952 2.4 ... 1 India 1382345085 0.99% 13586631 464 2973190 -532687.0 2.2 28 35% 17.70% 102 Sierra Leone 8004158 2.10% 163768 111 72180 -4200.0 4.3 19 43% 0.10% 285972 0.00% 183 New Caledonia 0.97% 2748 16 18280 502.0 2 34 72% 44 **150** Latvia 1882408 -1.08% -20545 30 62200 -14837.0 1.7 69% 0.02% 173 441750 0.27% 1171 1380 320 900.0 1.5 43 93% 0.01% Malta 201 rows × 11 columns