DS5100

2020 Tokyo Olympics Reilly Meinert, Max Ryoo, Said Mrad, Sydney Masterson



Project Scenario

Argentina - 1 2 3

Armenia - 2 2 4

Australia 17 7 22 46

Gold

Silver

Bronze

Total \$

- Brief Introduction About the data set
 - Olympics
 - https://olympics.com/en/olympic-games/tokyo-2020/medals
 - GDP
 - https://www.worldometers.info/gdp/gdp-by-country/
 - Kaggle Integration
 - https://www.kaggle.com/arjunprasadsarkhel/2021-olympics-in-tokyo?select=Teams.xlsx

Team ‡

Dataset Introduction

- Pandas Dataframe to csv
 - https://github.com/hyunsukr/DS5100-Final/tree/main/data
- Data was webscrapped and engineered to produce a final dataframe with the information below.

Name	Gold	Silver	Bronze	Total	Country	GDP	GDP abbreviated	GDP growth	Population	GDP per capita	NOC	Discipline	Continents
USA United States of America	39	41	33	113	United States	\$19,485,394,000,000.00	\$19.485 trillion	2.27%	325,084,756	\$59,939.00	United States of America	47	North America
CHN People's Republic of China	38	32	18	88	China	\$12,237,700,479,375.00	\$12.238 trillion	6.90%	1,421,021,791	\$8,612.00	People's Republic of China	33	Asia
JPN Japan	27	14	17	58	Japan	\$4,872,415,104,315.00	\$4.872 trillion	1.71%	127,502,725	\$38,214.00	Japan	48	Asia
GBR Great Britain	22	21	22	65	United Kingdom	\$2,637,866,340,434.00	\$2.638 trillion	1.79%	66,727,461	\$39,532.00	Great Britain	28	Europe
ROC ROC	20	28	23	71	Russia	\$1,578,417,211,937.00	\$1.578 trillion	1.55%	145,530,082	\$10,846.00	ROC	34	Asia
AUS Australia	17	7	22	46	Australia	\$1,323,421,072,479.00	\$1.323 trillion	1.96%	24,584,620	\$53,831.00	Australia	35	Australia
NED Netherlands	10	12	14	36	Netherlands	\$830,572,618,850.00	\$831 billion	3.16%	17,021,347	\$48,796.00	Netherlands	27	Europe
FRA France	10	12	11	33	France	\$2,582,501,307,216.00	\$2.583 trillion	1.82%	64,842,509	\$39,827.00	France	33	Europe
	USA United States of America	USA United States of America 39 CHN People's Republic of China 38 JPN Japan 27 GBR Great Britain 22 ROC ROC 20 AUS Australia 17 NED Netherlands 10	USA United States of America 39 41 CHN People's Republic of China 38 32 JPN Japan 27 14 GBR Great Britain 22 21 ROC ROC 20 28 AUS Australia 17 7 NED Netherlands 10 12	USA United States of America 39 41 33 CHN People's Republic of China 38 32 18 JPN Japan 27 14 17 GBR Great Britain 22 21 22 ROC ROC 20 28 23 AUS Australia 17 7 22 NED Netherlands 10 12 14	USA United States of America 39 41 33 113 CHN People's Republic of China 38 32 18 88 JPN Japan 27 14 17 58 GBR Great Britain 22 21 22 65 ROC ROC 20 28 23 71 AUS Australia 17 7 22 46 NED Netherlands 10 12 14 36	USA United States of America 39 41 33 113 United States CHN People's Republic of China 38 32 18 88 China JPN Japan 27 14 17 58 Japan GBR Great Britain 22 21 22 65 United Kingdom ROC ROC 20 28 23 71 Russia AUS Australia 17 7 22 46 Australia NED Netherlands 10 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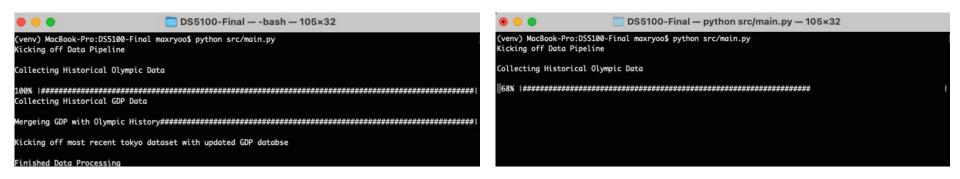
Dataset Web Scraping

- Class called Web_Scrapper
- Three data sets were webscrapped
 - Olympics
 - GDP
 - Only gives 2020 (recent GDP)
 - GDP Historical Data

```
class Web_Scrapper():
   def __init__(self, baselink="https://olympics.com/tokyo-2020/olympic-games/en/results/all-sports/", history = {}):
       self.baselink = baselink
      with open('src/resources/history.json') as json_file:
          history = json.load(json file)
       self.history = history
   def scrape_gdp_history(self, years):
       ## Not Tested Yet
       time series = pd.DataFrame()
       pbar.start()
       for i in pbar(range(0,len(years))):
          if int(years[i]) > 1949:
              df = self.scrape_gdp_economy(years[i])
              time series = time series.append(df)
       return time series
   def scrape_gdp_economy(self, year):
       URL = 'https://countryeconomy.com/gdp?year=' + year
       r = requests.get(URL) #http requests tot ehs specified url and save it in R
       soup = BeautifulSoup(r.content, 'html5lib')
       tables = soup.find_all('table', {'id':'tbA'})
       tables percap = soup.find all('table', {'id':'tbPC'})
       tempList = []
       for table in tables:
           for child in table.children:
              for td in child:
                   for tr in td:
                       tempList.append(tr.get text())
       second_tempList = []
      for table in tables_percap:
           for child in table.children:
              for td in child:
                   for tr in td:
                       second tempList.append(tr.get text())
       tempList = tempList[5:len(tempList)-1]
       second tempList = second tempList[6:len(second tempList) - 1]
```

Data Processing - Interaction

- Interaction with user.
- Progress bar to show % data pull
- Give feedback to user how much the data pull is complete.



Data Processing - Data Engineering

```
"United States of America" : "United States" ,
"People's Republic of China" : "China",
"Japan" : "Japan",
"Great Britain" : "United Kingdom",
"ROC" : "Russia",
"Australia" : "Australia",
"Netherlands": "Netherlands",
```

- Joined datasets based on country name
 - Some country names were different
 - Had to map countries names through a json (dictionary) through data cleaning
- Added geographical location for the data
 - Continents each country is located
 - Utilized a third party package
 - pycountry-convert

```
pycountry-convert 0.7.2

pip install pycountry-convert
```

```
class Cleaner():
   def __init__(self):
       # Opening JSON Mapping file
       with open('src/resources/mapping.json') as json_file:
           mapping = json.load(json file)
       with open('src/resources/mapping_continents.json') as json_file:
           cont_map = json.load(json_file)
       self.continent_maps = cont_map
       self.country_maps = mapping
   def join_gdp(self, gdp, olympic, join_cols=['Country']):
       temp_olympic = olympic.copy()
       temp_olympic["Country"] = temp_olympic["Name"].str[4:].map(self.country_maps)
       joined = pd.merge(temp_olympic, gdp, how='left', on=join_cols)
       return joined
   def join aggregate teams(self, teams, olympic):
       teams = teams.groupby("NOC")["Discipline"].count().reset_index()
       temp = olympic.copy()
       temp['tempName'] = temp["Name"].str[4:]
       joined = pd.merge(temp, teams, how='inner', left_on='tempName', right_on='NOC')
       joined = joined.drop(columns=["tempName"])
       return joined
```

Testing

def test scarpe summary():

webscrapper = Web_Scrapper()

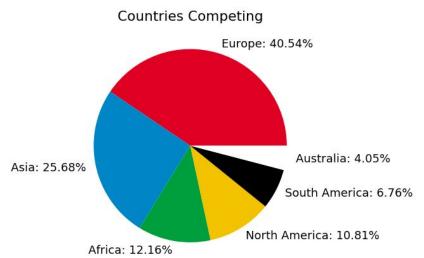
- Pytest to test the code / coverage
- Data engineering functions and data quality testing
- All methods relating to data
- Code coverage of 100% except main

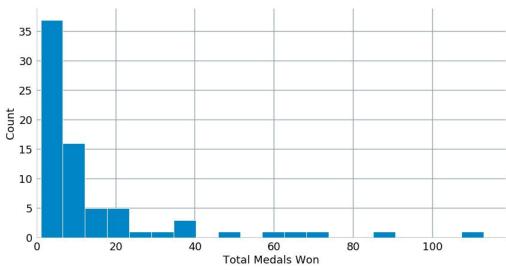
```
platform darwin -- Python 3.7.1, pytest-6.2.5, py-1.11.0, pluggy-1.0.0 -- /Users/maxryoo/Documents/MSDS/DS5100/DS5100-Final/venv/bin/python
                                                                                                                                          cachedir: .pvtest_cache
                                                                                                                                         rootdir: /Users/maxryoo/Documents/MSDS/DS5100/DS5100-Final
                                                                                                                                         plugins: mock-3.6.1, cov-3.0.0
                                                                                                                                         collected 11 items
                                                                                                                                          tests/utils/test_cleaner.py::test_join_gdp_single PASSED
                                                                                                                                         tests/utils/test_cleaner.py::test_get_continents_map PASSED
                                                                                                                                                                                                                 18%
                                                                                                                                         tests/utils/test_cleaner.py::test_get_continents_map_exceptions PASSED
                                                                                                                                                                                                                [ 27%]
                                                                                                                                          tests/utils/test_cleaner.py::test_convert_continent PASSED
                                                                                                                                                                                                                 36%]
                                                                                                                                         tests/utils/test_cleaner.pv::test_convert_continent_not_available PASSED [ 45%]
                                                                                                                                         tests/utils/test_cleaner.py::test_join_gdp PASSED
                                                                                                                                         tests/utils/test_cleaner.py::test_join_aggregate_teams PASSED
                                                                                                                                                                                                                 [ 72%]
                                                                                                                                          tests/utils/test_webscrapper.py::test_scrape_gdp PASSED
                                                                                                                                          tests/utils/test_webscrapper.py::test_scarpe_summary PASSED
                                                                                                                                                                                                                 81%]
                                                                                                                                          tests/utils/test_webscrapper.py::test_scrape_gdp_history PASSED
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                                                                                                                                          tests/utils/test_webscrapper.py::test_scrape_history PASSED
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                                                                                                                                           /Users/maxryoo/Documents/MSDS/DS5100/DS5100-Final/venv/lib/python3.7/site-packages/pkg_resources/_vendor/pyparsing.py:943: DeprecationWarnin
                                                                                                                                           : Using or importing the ABCs from 'collections' instead of from 'collections.abc' is deprecated, and in 3.8 it will stop working
                                                                                                                                             collections.MutableMapping.register(ParseResults)
                                                                                                                                          ests/utils/test_cleaner.py::test_get_continents_map
                                                                                                                                           /Users/maxryoo/Documents/MSDS/DS5100/DS5100-Final/veny/lib/python3.7/site-packages/pkg_resources/_vendor/pyparsing.py:3226: DeprecationWarni
                                                                                                                                          ng: Using or importing the ABCs from 'collections' instead of from 'collections.abc' is deprecated, and in 3.8 it will stop working
                                                                                                                                             elif isinstance( exprs, collections.Iterable ):
                                                                                                                                           Docs: https://docs.pytest.org/en/stable/warnings.html
                                                                                                                                           ------ coverage: platform darwin, python 3.7.1-final-0 ------
                                                                                                                                                                   Stmts Miss Cover
                                                                                                                                         src/__init__.pv
                                                                                                                                                                            0 100%
                                                                                                                                         src/main.py
                                                                                                                                                                            34 0%
                                                                                                                                          src/utils/__init__.py
                                                                                                                                                                            0 100%
scrapped_olympic = webscrapper.scrape_summary('https://olympics.com/en/olympic-games/tokyo-2020/medals')
```

T100%7

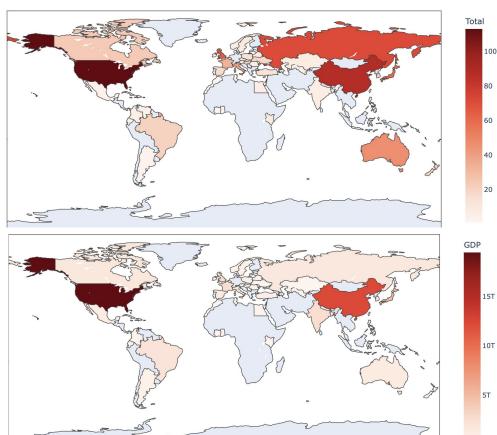
```
computed_total = scrapped_olympic["Gold"].astype(int) + scrapped_olympic["Silver"].astype(int) + scrapped_olympic["Bronze"].astype(int)
sanity_check_medal_count = sum(scrapped_olympic["Total"].astype(int) > 500)
assert set(scrapped_olympic) == set(['Name', 'Gold', 'Silver', 'Bronze', 'Total'])
assert scrapped olympic.equals(scrapped olympic.dropna())
assert computed_total.equals(scrapped_olympic["Total"].astype(int))
assert sanity_check_medal_count == 0
```

Exploratory Data Analysis





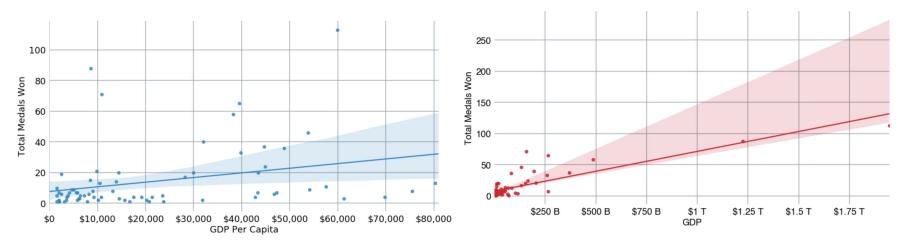
World Statistics



Conclusions:

- US has highest GDP and number of medals won
- China second for both
- Relationship not as strong for rest of world
 - Russia, Australia, & Great Britain have high medal counts but lower GDPs compared to US and China

Medal Count versus GDP



Medals Won versus GDP Per Capita: .2975

Medals Won versus GDP: 0.8362

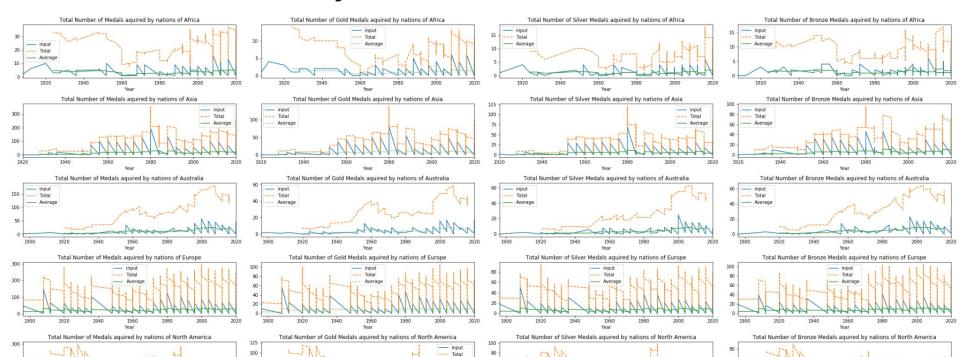
Conclusion: The relationship between GDP and medals won is much stronger than the relationship between GDP per capita and medals won.

Model Building

- Multiple Linear Regression
 - R^2 = 0.8479990568528668
 - Mean Squared Error = 109.1391895919251
 - Root Mean Squared Error = 10.446970354697342
- Possible Next Steps
 - Multicollinearity
 - Linear Regression assumption checking
 - GDP and Population had a beta of 0, which may raise eyebrows

	Feature	Coefficient
7	Continents_Europe	6.200415
8	Continents_North America	4.484664
5	Continents_Asia	3.548857
6	Continents_Australia	2.443772
4	Discipline	0.784625
0	GDP	0.000000
2	Population	0.000000
9	Continents_South America	-6.659379
1	GDP growth	-0.138194
3	GDP per capita	-0.000042

Time Series Analysis



Conclusion / Next Steps

- Code in github is available with virtual environments
 - Making the github repo a package repo will make it so that we can deploy the package.
 - Setup.py
- Dive deeper in the Multiple Linear Regression model such as multicollinearity etc.