

rebrick_04

October 17, 2020

0.1 Example of how to use the rebrick python wrapper for rebrickable api

```
[92]: import rebrick
import json

# init rebrick module for general reading
rebrick.init("1c3972e38c0479079cfafb4c80b6e11d")

# get set info
response = rebrick.lego.get_set(6608)
print(json.loads(response.read()))
```

```
{'set_num': '6608-1', 'name': 'Tractor', 'year': 1982, 'theme_id': 73,
'num_parts': 21, 'set_img_url':
'https://cdn.rebrickable.com/media/sets/6608-1/9875.jpg', 'set_url':
'https://rebrickable.com/sets/6608-1/tractor/', 'last_modified_dt':
'2015-11-02T10:43:45.980214Z'}
```

```
[93]: # init rebrick module including user reading
rebrick.init("your_API_KEY_here", "your_USER_TOKEN_here")

# if you don't know the user token you can use your login credentials
rebrick.init("1c3972e38c0479079cfafb4c80b6e11d", "rpmoo", "mcH.3~.EsV;hs*qd")

# get user partlists
response = rebrick.users.get_partlists()
print(json.loads(response.read()))
```

```
{'count': 0, 'next': None, 'previous': None, 'results': []}
```

1 1. Load the latest data sets from rebrickable website

First scrape www.Rebrickable.com for the links we want

```
[94]: #https://pythonspot.com/extract-links-from-webpage-beautifulsoup/
import os
import requests
from bs4 import BeautifulSoup
```

```
import re
import shutil
import gzip
```

```
[95]: # https://www.dataquest.io/blog/web-scraping-tutorial-python/
def getLinks(url): # scrape the url
    html_page = requests.get(url) #urllib2.urlopen(url)
    soup = BeautifulSoup(html_page.content, 'lxml')

    html = list(soup.children)[4] # gets the html tag
    nav = list(html.children)[5] # nav is the 'navigable string'

    # get all the links on the page
    links = []
    for link in nav.findAll('a', attrs={'href': re.compile("https://")}):
        links.append(link.get('href'))
    #print(len(links), 'links')

    # now get only the link we want to download
    downloads_list = []
    for item in links:
        if item[0:44] == 'https://cdn.rebrickable.com/media/downloads/' and
        ↪item[-3:] != 'zip':
            print(item)
            downloads_list.append(item)
    print("We have", len(downloads_list), "links to download.")
    return downloads_list
```

```
[96]: URL='https://rebrickable.com/downloads/'
linklist = getLinks(URL)
```

```
https://cdn.rebrickable.com/media/downloads/themes.csv.gz?1602930200.2377865
https://cdn.rebrickable.com/media/downloads/colors.csv.gz?1602930200.3457866
https://cdn.rebrickable.com/media/downloads/part_categories.csv.gz?1602930200.45
77868
https://cdn.rebrickable.com/media/downloads/parts.csv.gz?1602930200.8817878
https://cdn.rebrickable.com/media/downloads/part_relationships.csv.gz?1602930205
.2857969
https://cdn.rebrickable.com/media/downloads/elements.csv.gz?1602930201.1537883
https://cdn.rebrickable.com/media/downloads/sets.csv.gz?1602930201.6497893
https://cdn.rebrickable.com/media/downloads/minifigs.csv.gz?1602930201.8897898
https://cdn.rebrickable.com/media/downloads/inventories.csv.gz?1602930201.401789
https://cdn.rebrickable.com/media/downloads/inventory_parts.csv.gz?1602930204.62
17954
https://cdn.rebrickable.com/media/downloads/inventory_sets.csv.gz?1602930204.821
796
https://cdn.rebrickable.com/media/downloads/inventory_minifigs.csv.gz?1602930205
```

.0737965

We have 12 links to download.

Next download these links

```
[97]: # Test the file_url construction
r = re.compile('[?]*') # https://stackoverflow.com/a/2175096/8971265
file_url = r.split("https://cdn.rebrickable.com/media/downloads/themes.csv.gz?
↳1592816880.2813647")
file_url
```

```
[97]: ['https://cdn.rebrickable.com/media/downloads/themes.csv.gz',
      '1592816880.2813647']
```

```
[98]: # Test the file name construction
r = re.compile('https://cdn.rebrickable.com/media/downloads/')
file_name = r.split(file_url[0])[1]
file_name
```

```
[98]: 'themes.csv.gz'
```

```
[99]: def download_file(url, _path):

    """function to download a file at a given url and to a given path"""

    # get the file url w/o the slug
    r = re.compile('[?]*') # https://stackoverflow.com/a/2175096/8971265
    file_url = r.split(url)
    print("file_url:", file_url[0])

    # get the file name
    r = re.compile('https://cdn.rebrickable.com/media/downloads/')
    file_name = r.split(file_url[0])[1]
    print("file_name:", file_name)

    # define headers dictionary containing the user agent (https://
↳stackoverflow.com/a/10606260/8971265)
    # https://www.scrapehero.com/
↳how-to-fake-and-rotate-user-agents-using-python-3/
    # https://developers.whatismybrowser.com/useragents/explore/
↳operating_system_name/windows/
    headers = {'User-Agent': "Mozilla/5.0 (Windows NT 10.0; Win64; x64)␣
↳AppleWebKit/537.36 (KHTML, like Gecko) Chrome/80.0.3987.149 Safari/537.36"}

    # download the file (https://stackoverflow.com/a/39217788/8971265)
    print("Downloading:", file_name)
    with requests.get(url, stream=True, headers=headers) as r:
        with open(os.path.join(_path, file_name), 'wb') as f:
```

```

        shutil.copyfileobj(r.raw, f)

# unzip the file (https://stackoverflow.com/a/44712152/8971265)
print("Unzipping:", file_name, "\n")
with gzip.open(os.path.join(_path, file_name), 'rb') as f_src:
    with open(os.path.join(_path, file_name[:-3]), 'wb') as f_dst:
        shutil.copyfileobj(f_src, f_dst)

return

```

```

[100]: def fetch_files(_linklist, _path):

        """a function that creates a folder and download files to
        that folder from a list of links """

        if not os.path.isdir(_path): # if it is not there create it
            os.makedirs(_path)

        for link in linklist:

            # use the download_file function to download from the given link
            download_file(link, _path)

```

```

[101]: # pass the folder name we want to create and save downloads in
data_path = os.path.join("data")

# download the files to that folder
fetch_files(linklist, data_path)

```

```

file_url: https://cdn.rebrickable.com/media/downloads/themes.csv.gz
file_name: themes.csv.gz
Downloading: themes.csv.gz
Unzipping: themes.csv.gz

```

```

file_url: https://cdn.rebrickable.com/media/downloads/colors.csv.gz
file_name: colors.csv.gz
Downloading: colors.csv.gz
Unzipping: colors.csv.gz

```

```

file_url: https://cdn.rebrickable.com/media/downloads/part_categories.csv.gz
file_name: part_categories.csv.gz
Downloading: part_categories.csv.gz
Unzipping: part_categories.csv.gz

```

```

file_url: https://cdn.rebrickable.com/media/downloads/parts.csv.gz
file_name: parts.csv.gz
Downloading: parts.csv.gz

```

Unzipping: parts.csv.gz

file_url: https://cdn.rebrickable.com/media/downloads/part_relationships.csv.gz

file_name: part_relationships.csv.gz

Downloading: part_relationships.csv.gz

Unzipping: part_relationships.csv.gz

file_url: <https://cdn.rebrickable.com/media/downloads/elements.csv.gz>

file_name: elements.csv.gz

Downloading: elements.csv.gz

Unzipping: elements.csv.gz

file_url: <https://cdn.rebrickable.com/media/downloads/sets.csv.gz>

file_name: sets.csv.gz

Downloading: sets.csv.gz

Unzipping: sets.csv.gz

file_url: <https://cdn.rebrickable.com/media/downloads/minifigs.csv.gz>

file_name: minifigs.csv.gz

Downloading: minifigs.csv.gz

Unzipping: minifigs.csv.gz

file_url: <https://cdn.rebrickable.com/media/downloads/inventories.csv.gz>

file_name: inventories.csv.gz

Downloading: inventories.csv.gz

Unzipping: inventories.csv.gz

file_url: https://cdn.rebrickable.com/media/downloads/inventory_parts.csv.gz

file_name: inventory_parts.csv.gz

Downloading: inventory_parts.csv.gz

Unzipping: inventory_parts.csv.gz

file_url: https://cdn.rebrickable.com/media/downloads/inventory_sets.csv.gz

file_name: inventory_sets.csv.gz

Downloading: inventory_sets.csv.gz

Unzipping: inventory_sets.csv.gz

file_url: https://cdn.rebrickable.com/media/downloads/inventory_minifigs.csv.gz

file_name: inventory_minifigs.csv.gz

Downloading: inventory_minifigs.csv.gz

Unzipping: inventory_minifigs.csv.gz

1.1 2. Load the data to a data dictionary

Function to load data.

```
[102]: import pandas as pd
def load_csv_data(file_name, _path):
    file_path = os.path.join(_path, file_name)
    return pd.read_csv(file_path)
```

Make a list of files in the directory.

```
[103]: files = os.listdir(data_path)
```

Make a list of csv files only.

```
[104]: files_csv = [f for f in files if f[-4:] == '.csv'] # find all files ending with
↳ '.csv'
display(files_csv)
```

```
['colors.csv',
 'df_sets_new.csv',
 'df_sets_used.csv',
 'elements.csv',
 'inventories.csv',
 'inventory_minifigs.csv',
 'inventory_parts.csv',
 'inventory_sets.csv',
 'minifigs.csv',
 'parts.csv',
 'part_categories.csv',
 'part_relationships.csv',
 'sets.csv',
 'themes.csv']
```

Make a data dictionary with the csv file names as keys.

```
[105]: lego={} #https://stackoverflow.com/a/56217834/8971265
for i in range(len(files_csv)):
    lego[files_csv[i][:4]] = load_csv_data(files_csv[i], data_path)

# print out one dataframe from the data dictionary to check
lego['themes'].head()
```

```
[105]:
```

| | id | name | parent_id |
|---|----|----------------|-----------|
| 0 | 1 | Technic | NaN |
| 1 | 2 | Arctic Technic | 1.0 |
| 2 | 3 | Competition | 1.0 |
| 3 | 4 | Expert Builder | 1.0 |
| 4 | 5 | Model | 1.0 |

1.2 3. Explore the data

(Taking some initial inspiration from this Kaggle notebook: [Lego - let's play](#))

Colors

```
[106]: df_colors = lego['colors'].reset_index(drop=True)
df_colors.head()
```

```
[106]:
```

| | id | name | rgb | is_trans |
|---|----|----------------|--------|----------|
| 0 | -1 | [Unknown] | 0033B2 | f |
| 1 | 0 | Black | 05131D | f |
| 2 | 1 | Blue | 0055BF | f |
| 3 | 2 | Green | 237841 | f |
| 4 | 3 | Dark Turquoise | 008F9B | f |

```
[147]: # import plotly's 'graph objects' library
import plotly.graph_objs as go

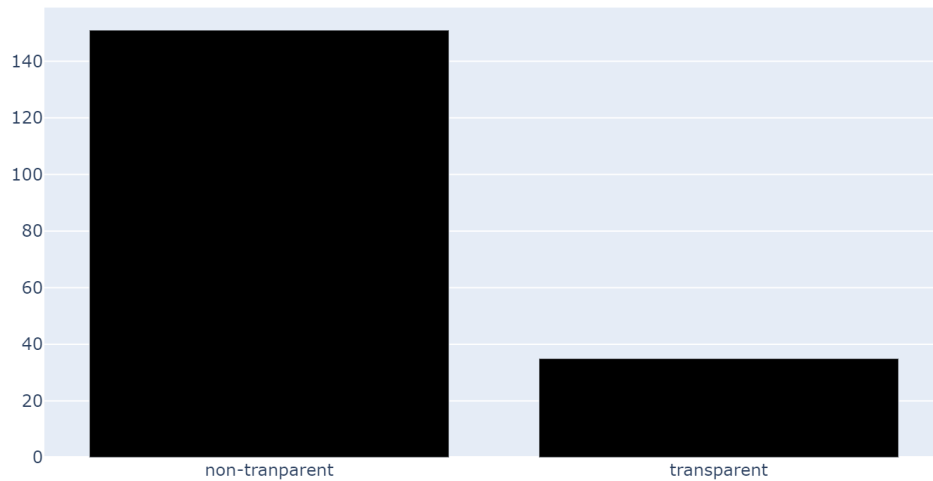
# group dataframe by count of 'is_trans' column
group = df_colors.groupby('is_trans')[['is_trans']].count()
group.head()
```

```
[147]:
```

| | is_trans |
|----------|----------|
| is_trans | |
| f | 151 |
| t | 35 |

```
[154]: # plot it
plot_name="Count of transparent and non-transparent unique Lego colours"
data = [go.Bar(x=['non-transparent', 'transparent'],
               y=group['is_trans'],
               marker=dict(color='black'))
]
layout = go.Layout(title=plot_name)
fig = go.Figure(data=data, layout=layout)
fig.write_image("1.png", width=800, scale=2)
#fig.show()
```

Count of transparent and non-transparent unique Lego colours



Sets

```
[156]: # dataframe of every lego set in the reirckable database
df_sets = lego['sets']
df_sets.head()
```

```
[156]:
```

| | set_num | name | year | theme_id | num_parts |
|---|---------|----------------------------|------|----------|-----------|
| 0 | 001-1 | Gears | 1965 | 1 | 43 |
| 1 | 0011-2 | Town Mini-Figures | 1978 | 84 | 12 |
| 2 | 0011-3 | Castle 2 for 1 Bonus Offer | 1987 | 199 | 0 |
| 3 | 0012-1 | Space Mini-Figures | 1979 | 143 | 12 |
| 4 | 0013-1 | Space Mini-Figures | 1979 | 143 | 12 |

```
[157]: # the years from that dataframe
df_sets.groupby(['year']).groups.keys()
```

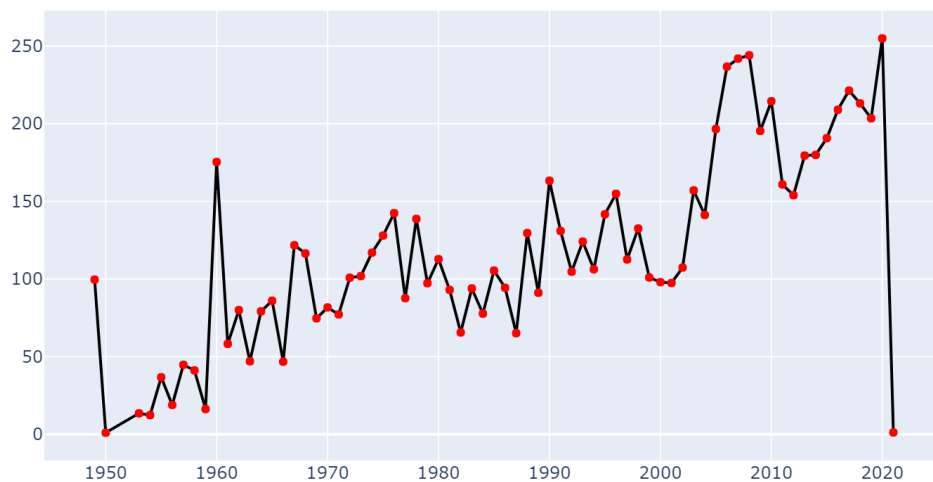
```
[157]: dict_keys([1949, 1950, 1953, 1954, 1955, 1956, 1957, 1958, 1959, 1960, 1961,
1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969, 1970, 1971, 1972, 1973, 1974,
1975, 1976, 1977, 1978, 1979, 1980, 1981, 1982, 1983, 1984, 1985, 1986, 1987,
1988, 1989, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997, 1998, 1999, 2000,
2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013,
2014, 2015, 2016, 2017, 2018, 2019, 2020, 2021])
```

```
[158]: column_to_plot = 'num_parts'
# group by the 'column_to_plot' and display
group = df_sets.groupby('year')[[column_to_plot]].mean()
display(group.head())
```


| | num_parts |
|------|-----------|
| year | |
| 1949 | 99.600000 |
| 1950 | 1.000000 |
| 1953 | 13.500000 |
| 1954 | 12.357143 |
| 1955 | 36.714286 |

```
[160]: # plot it
data = [go.Scatter(x=group.index,
                    y=group[column_to_plot],
                    mode='lines+markers',
                    marker=dict(color='red'),
                    line=dict(color='black'))
]
layout = go.Layout(title='Mean number of parts in Lego sets by year')
fig = go.Figure(data=data, layout=layout)
fig.write_image("2.png", width=800, scale=2)
```

Mean number of parts in Lego sets by year



Themes

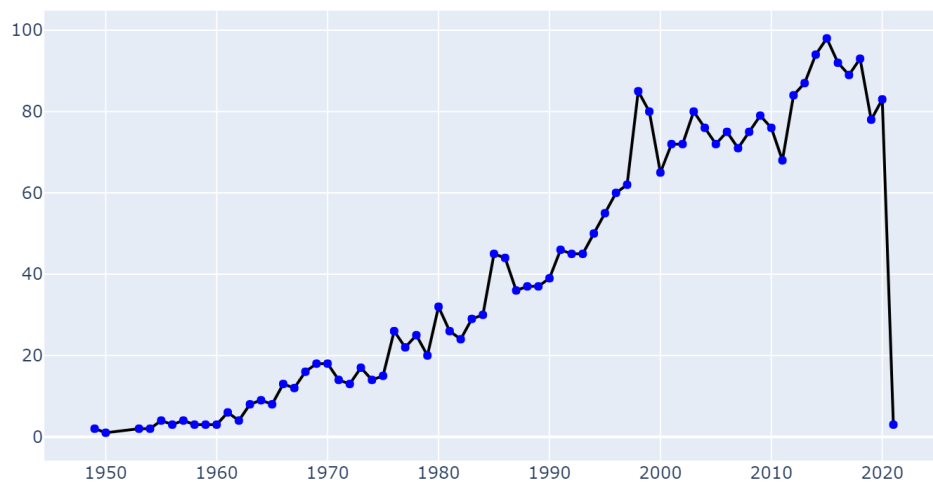
Group by number of unique theme_id [ref](#)

```
[161]: group = df_sets.groupby('year')[['theme_id']].nunique()
display(group.head())
```

| | theme_id |
|------|----------|
| year | |
| 1949 | 2 |
| 1950 | 1 |
| 1953 | 2 |
| 1954 | 2 |
| 1955 | 4 |

```
[162]: # plot it
data = [go.Scatter(x=group.index,
                    y=group['theme_id'],
                    mode='lines+markers',
                    marker=dict(color='blue'),
                    line=dict(color='black'))
]
layout = go.Layout(title='Number of unique Lego themes by year')
fig = go.Figure(data=data, layout=layout)
#fig.show()
fig.write_image("3.png", width=800, scale=2)
```

Number of unique Lego themes by year



Back to the Colors

```
[164]: df_colors.head()
```

```
[164]:   id      name      rgb is_trans
      0  -1    [Unknown] 0033B2      f
      1   0      Black 05131D      f
      2   1      Blue 0055BF      f
      3   2      Green 237841      f
      4   3  Dark Turquoise 008F9B      f
```

What's the difference between Inventories and Sets?

```
[165]: df_inventories = lego['inventories']
      df_inventories.head()
```

```
[165]:   id  version set_num
      0    1         1  7922-1
      1    3         1  3931-1
      2    4         1  6942-1
      3   15         1  5158-1
      4   16         1   903-1
```

```
[166]: df_inventories.shape
```

```
[166]: (26759, 3)
```

The inventories table has 'versions' of unique sets.

Look at the sets table again.

```
[167]: df_sets.head()
```

```
[167]:   set_num      name  year  theme_id  num_parts
      0  001-1      Gears  1965         1         43
      1  0011-2  Town Mini-Figures  1978         84         12
      2  0011-3  Castle 2 for 1 Bonus Offer  1987        199         0
      3  0012-1  Space Mini-Figures  1979        143         12
      4  0013-1  Space Mini-Figures  1979        143         12
```

```
[168]: # join the sets and inventories tables on 'set_num'
      df_inv_sets = pd.merge(df_inventories, df_sets, on='set_num', how='outer').
      ↪sort_values(by='set_num')
      df_inv_sets
```

```
[168]:   id  version      set_num      name \
      1925   2836         1      00-6      Special Offer
      12129  24696         1      001-1      Gears
      3423   5087         1     0011-2  Town Mini-Figures
      1492   2216         1     0011-3  Castle 2 for 1 Bonus Offer
      955   1414         1     0012-1  Space Mini-Figures
      ...    ...    ...    ...    ...
```

| | | | | |
|-------|-------|---|--------------|-------------------------------------|
| 1303 | 1936 | 1 | tominifigs-1 | Town Minifig Packs 2-Pack |
| 11031 | 16524 | 1 | trucapam-1 | Captain America Mosaic |
| 9884 | 14717 | 1 | tsuper-1 | Technic Super Set |
| 3278 | 4868 | 1 | vwkit-1 | Volkswagen Kit |
| 507 | 758 | 1 | wwgp1-1 | Wild West Limited Edition Gift Pack |

| | year | theme_id | num_parts |
|-------|--------|----------|-----------|
| 1925 | 1985.0 | 67.0 | 0.0 |
| 12129 | 1965.0 | 1.0 | 43.0 |
| 3423 | 1978.0 | 84.0 | 12.0 |
| 1492 | 1987.0 | 199.0 | 0.0 |
| 955 | 1979.0 | 143.0 | 12.0 |
| ... | ... | ... | ... |
| 1303 | 2000.0 | 50.0 | 0.0 |
| 11031 | 2016.0 | 696.0 | 72.0 |
| 9884 | 1991.0 | 12.0 | 0.0 |
| 3278 | 1959.0 | 366.0 | 22.0 |
| 507 | 1996.0 | 476.0 | 0.0 |

[26759 rows x 7 columns]

```
[169]: # take a look at the NaNs
df_inv_sets[df_inv_sets.isnull().any(axis=1)]
```

| | id | version | set_num | name | year | theme_id | num_parts |
|-------|-------|---------|------------|------|------|----------|-----------|
| 15621 | 48649 | 1 | fig-000001 | NaN | NaN | NaN | NaN |
| 15622 | 48650 | 1 | fig-000002 | NaN | NaN | NaN | NaN |
| 15624 | 48681 | 1 | fig-000003 | NaN | NaN | NaN | NaN |
| 15626 | 48758 | 1 | fig-000004 | NaN | NaN | NaN | NaN |
| 15636 | 49548 | 1 | fig-000005 | NaN | NaN | NaN | NaN |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 26725 | 76081 | 1 | fig-010711 | NaN | NaN | NaN | NaN |
| 26726 | 76082 | 1 | fig-010712 | NaN | NaN | NaN | NaN |
| 26728 | 76167 | 1 | fig-010713 | NaN | NaN | NaN | NaN |
| 26729 | 76168 | 1 | fig-010714 | NaN | NaN | NaN | NaN |
| 26739 | 76650 | 1 | fig-010715 | NaN | NaN | NaN | NaN |

[10470 rows x 7 columns]

The difference between the number of unique inventory ids and set_num's is that fact that figures don't have a set number.

```
[170]: # drop NaNs
df_inv_sets.dropna(axis=0, inplace=True)
df_inv_sets
```

```
[170]:
```

| | id | version | set_num | name \ |
|-------|-------|---------|--------------|-------------------------------------|
| 1925 | 2836 | 1 | 00-6 | Special Offer |
| 12129 | 24696 | 1 | 001-1 | Gears |
| 3423 | 5087 | 1 | 0011-2 | Town Mini-Figures |
| 1492 | 2216 | 1 | 0011-3 | Castle 2 for 1 Bonus Offer |
| 955 | 1414 | 1 | 0012-1 | Space Mini-Figures |
| ... | ... | ... | ... | ... |
| 1303 | 1936 | 1 | tominifigs-1 | Town Minifig Packs 2-Pack |
| 11031 | 16524 | 1 | trucapam-1 | Captain America Mosaic |
| 9884 | 14717 | 1 | tsuper-1 | Technic Super Set |
| 3278 | 4868 | 1 | vwkit-1 | Volkswagen Kit |
| 507 | 758 | 1 | wgwp1-1 | Wild West Limited Edition Gift Pack |

| | year | theme_id | num_parts |
|-------|--------|----------|-----------|
| 1925 | 1985.0 | 67.0 | 0.0 |
| 12129 | 1965.0 | 1.0 | 43.0 |
| 3423 | 1978.0 | 84.0 | 12.0 |
| 1492 | 1987.0 | 199.0 | 0.0 |
| 955 | 1979.0 | 143.0 | 12.0 |
| ... | ... | ... | ... |
| 1303 | 2000.0 | 50.0 | 0.0 |
| 11031 | 2016.0 | 696.0 | 72.0 |
| 9884 | 1991.0 | 12.0 | 0.0 |
| 3278 | 1959.0 | 366.0 | 22.0 |
| 507 | 1996.0 | 476.0 | 0.0 |

[16289 rows x 7 columns]

```
[171]: # join the result with 'inventory_parts' table to get a table of all part for
↳ all sets
df_inventory_parts = lego['inventory_parts']
df_all = pd.merge(df_inv_sets, df_inventory_parts, how='outer', left_on='id',
↳ right_on='inventory_id')
df_all
```

```
[171]:
```

| | id | version | set_num | name | year | theme_id | num_parts \ |
|--------|---------|---------|---------|---------------|--------|----------|-------------|
| 0 | 2836.0 | 1.0 | 00-6 | Special Offer | 1985.0 | 67.0 | 0.0 |
| 1 | 24696.0 | 1.0 | 001-1 | Gears | 1965.0 | 1.0 | 43.0 |
| 2 | 24696.0 | 1.0 | 001-1 | Gears | 1965.0 | 1.0 | 43.0 |
| 3 | 24696.0 | 1.0 | 001-1 | Gears | 1965.0 | 1.0 | 43.0 |
| 4 | 24696.0 | 1.0 | 001-1 | Gears | 1965.0 | 1.0 | 43.0 |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 853171 | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 853172 | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 853173 | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 853174 | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 853175 | NaN | NaN | NaN | NaN | NaN | NaN | NaN |

| | inventory_id | part_num | color_id | quantity | is_spare |
|--------|--------------|--------------|----------|----------|----------|
| 0 | NaN | NaN | NaN | NaN | NaN |
| 1 | 24696.0 | 132a | 7.0 | 4.0 | f |
| 2 | 24696.0 | 3020 | 15.0 | 4.0 | f |
| 3 | 24696.0 | 3062c | 15.0 | 1.0 | f |
| 4 | 24696.0 | 3404bc01 | 15.0 | 4.0 | f |
| ... | ... | ... | ... | ... | ... |
| 853171 | 76650.0 | 2446 | 484.0 | 1.0 | f |
| 853172 | 76650.0 | 2447 | 40.0 | 1.0 | f |
| 853173 | 76650.0 | 3626cpr3026 | 14.0 | 1.0 | f |
| 853174 | 76650.0 | 970c00 | 0.0 | 1.0 | f |
| 853175 | 76650.0 | 973pr5038c01 | 288.0 | 1.0 | f |

[853176 rows x 12 columns]

```
[172]: # drop NaNs
df_all.dropna(axis=0, inplace=True)
```

```
[173]: # join with colors table to bring in the piece-wise color codes
df_all = pd.merge(df_all, df_colors, left_on='color_id', right_on='id',
↳how='outer')
df_all.head()
```

```
[173]:      id_x  version set_num      name_x  year  theme_id  num_parts \
0  24696.0      1.0   001-1      Gears  1965.0      1.0      43.0
1  24696.0      1.0   001-1      Gears  1965.0      1.0      43.0
2  13351.0      1.0  0016-1  Castle Mini Figures  1978.0     186.0      15.0
3  24702.0      1.0   003-1  Master Mechanic Set  1966.0     366.0     403.0
4  24702.0      1.0   003-1  Master Mechanic Set  1966.0     366.0     403.0
```

| | inventory_id | part_num | color_id | quantity | is_spare | id_y | name_y | \ |
|---|--------------|----------|----------|----------|----------|------|------------|---|
| 0 | 24696.0 | 132a | 7.0 | 4.0 | f | 7 | Light Gray | |
| 1 | 24696.0 | 36 | 7.0 | 4.0 | f | 7 | Light Gray | |
| 2 | 13351.0 | 3847a | 7.0 | 3.0 | f | 7 | Light Gray | |
| 3 | 24702.0 | 132a | 7.0 | 8.0 | f | 7 | Light Gray | |
| 4 | 24702.0 | 36 | 7.0 | 4.0 | f | 7 | Light Gray | |

| | rgb | is_trans |
|---|--------|----------|
| 0 | 9BA19D | f |
| 1 | 9BA19D | f |
| 2 | 9BA19D | f |
| 3 | 9BA19D | f |
| 4 | 9BA19D | f |

```
[174]: # drop NaNs and create a new table with the columns we are interested in
df_all.dropna(axis=0, inplace=True)
```

```
df_all_colors =
↳df_all[['set_num', 'name_x', 'year', 'theme_id', 'color_id', 'quantity', 'name_y', 'rgb', 'is_trans
df_all_colors
```

```
[174]:
```

| | set_num | | name_x | year | theme_id | \ |
|--------|----------------|-----------------------------------|---------------------|--------|----------|---|
| 0 | 001-1 | | Gears | 1965.0 | 1.0 | |
| 1 | 001-1 | | Gears | 1965.0 | 1.0 | |
| 2 | 0016-1 | | Castle Mini Figures | 1978.0 | 186.0 | |
| 3 | 003-1 | | Master Mechanic Set | 1966.0 | 366.0 | |
| 4 | 003-1 | | Master Mechanic Set | 1966.0 | 366.0 | |
| ... | ... | | ... | ... | ... | |
| 803683 | LEGO-Modulex-1 | Unused Modulex parts sold by LEGO | 2019.0 | 408.0 | | |
| 803684 | LEGO-Modulex-1 | Unused Modulex parts sold by LEGO | 2019.0 | 408.0 | | |
| 803685 | LEGO-Modulex-1 | Unused Modulex parts sold by LEGO | 2019.0 | 408.0 | | |
| 803686 | LEGO-Modulex-1 | Unused Modulex parts sold by LEGO | 2019.0 | 408.0 | | |
| 803687 | LEGO-Modulex-1 | Unused Modulex parts sold by LEGO | 2019.0 | 408.0 | | |

| | color_id | quantity | name_y | rgb | is_trans |
|--------|----------|----------|----------------------|--------|----------|
| 0 | 7.0 | 4.0 | Light Gray | 9BA19D | f |
| 1 | 7.0 | 4.0 | Light Gray | 9BA19D | f |
| 2 | 7.0 | 3.0 | Light Gray | 9BA19D | f |
| 3 | 7.0 | 8.0 | Light Gray | 9BA19D | f |
| 4 | 7.0 | 4.0 | Light Gray | 9BA19D | f |
| ... | ... | ... | ... | ... | ... |
| 803683 | 1037.0 | 1.0 | Modulex Violet | BD7D85 | f |
| 803684 | 1037.0 | 1.0 | Modulex Violet | BD7D85 | f |
| 803685 | 1024.0 | 1.0 | Modulex Pink Red | F45C40 | f |
| 803686 | 1035.0 | 1.0 | Modulex Medium Blue | 61AFFF | f |
| 803687 | 1027.0 | 1.0 | Modulex Light Yellow | FFE371 | f |

[803688 rows x 9 columns]

```
[175]: # group by year and aggregate the by number of unique colour codes
group = df_all_colors.groupby('year')[['rgb']].nunique()
display(group.head())
```

| year | rgb |
|--------|-----|
| 1949.0 | 10 |
| 1950.0 | 3 |
| 1953.0 | 5 |
| 1954.0 | 7 |
| 1955.0 | 7 |

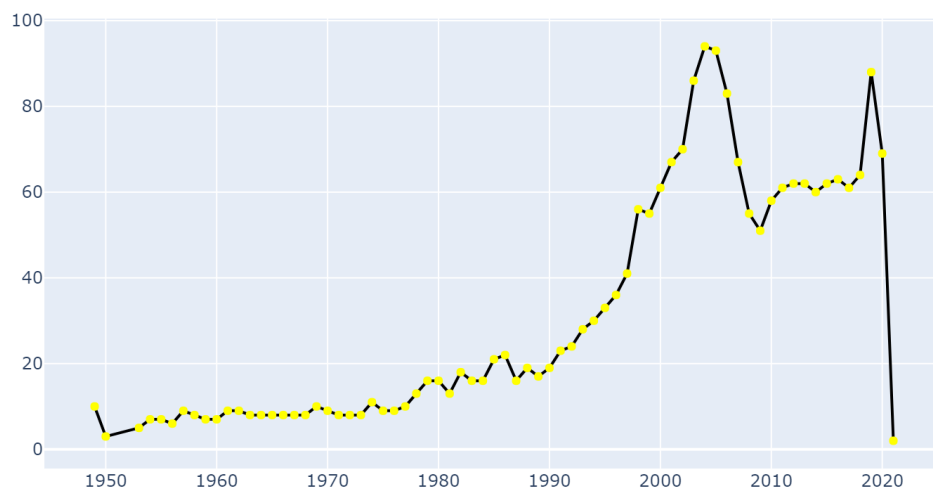
```
[179]: # plot it
data = [go.Scatter(x=group.index,
```

```

        y=group['rgb'],
        mode='lines+markers',
        marker=dict(color='yellow'),
        line=dict(color='black')
    )
]
layout = go.Layout(title='Number of unique Lego colours by year')
fig = go.Figure(data=data, layout=layout)
fig.write_image("4.png", width=800, scale=2)

```

Number of unique Lego colours by year



```

[180]: # get the list of unique colours for each year
colors = list(df_all_colors['rgb'].unique())
years = group.index.astype(int)
years

```

```

[180]: Int64Index([1949, 1950, 1953, 1954, 1955, 1956, 1957, 1958, 1959, 1960, 1961,
                1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969, 1970, 1971, 1972,
                1973, 1974, 1975, 1976, 1977, 1978, 1979, 1980, 1981, 1982, 1983,
                1984, 1985, 1986, 1987, 1988, 1989, 1990, 1991, 1992, 1993, 1994,
                1995, 1996, 1997, 1998, 1999, 2000, 2001, 2002, 2003, 2004, 2005,
                2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016,
                2017, 2018, 2019, 2020, 2021],
                dtype='int64')

```

```

[181]: # get the unique color lists by year
colors_list_by_year = []

```



```

for year in years:
    temp = df_all_colors[df_all_colors['year']==year]
    colors_for_year = list(temp['rgb'].unique())
    colors_list_by_year.append(colors_for_year)

```

```
[182]: colors_list_by_year[0:9]
```

```

[182]: [['FFFFFF',
        'C91A09',
        'F2CD37',
        '0055BF',
        '237841',
        '4B9F4A',
        'CA1F08',
        '1E601E',
        'F3C305',
        '039CBD'],
        ['C91A09', '0055BF', '237841'],
        ['FFFFFF', 'C91A09', 'F2CD37', '0055BF', '237841'],
        ['9BA19D', 'FFFFFF', 'C91A09', 'F2CD37', '0055BF', 'FCFCFC', '237841'],
        ['FFFFFF', 'C91A09', 'F2CD37', '0055BF', 'FCFCFC', '237841', '05131D'],
        ['FFFFFF', 'C91A09', 'F2CD37', '0055BF', 'FCFCFC', '05131D'],
        ['9BA19D',
        'FFFFFF',
        'C91A09',
        'F2CD37',
        '0055BF',
        '05131D',
        'FCFCFC',
        '237841',
        'A5A9B4'],
        ['FFFFFF',
        'C91A09',
        'F2CD37',
        '0055BF',
        '05131D',
        'FCFCFC',
        '237841',
        'A5A9B4'],
        ['9BA19D', 'FFFFFF', 'C91A09', 'F2CD37', '0055BF', 'FCFCFC', '05131D']]

```

```

[183]: # define a function to count features
def count_a_feature(feature, year, _df):

    """function to count values of a given feature
    and its proportion of the total count for a
    dataframe passed, and return a dataframe

```

```
with the result for a given year"""
```

```
temp = _df[_df['year']==year] # filter for the year
col1 = temp[feature].value_counts()
total = temp[feature].value_counts().sum()
col2 = round(100*temp[feature].value_counts() / total, 1)
data = {'count': col1, 'pc': col2}
temp = pd.DataFrame(data).sort_values(by='count', ascending=False)
#display(temp)

return temp, feature
```

```
[184]: # test it
count_a_feature('rgb', 1950, df_all_colors)
```

```
[184]: (      count    pc
0055BF      10  45.5
237841       9  40.9
C91A09       3  13.6,
'rgb')
```

Create a dataframe of all unique colours (index) by year (columns) with the proportion of those colours for each year as the values.

```
[185]: import numpy as np
# get the list of unique colours for each year
colors = list(df_all_colors['rgb'].unique())
# get the unique years
year = list(df_all_colors['year'].unique())
# set up a new dataframe
df_stack = pd.DataFrame(index=colors, columns=years)
# get the portion of all colors for each colour in each year
for year in years:
    df_feature, feature = count_a_feature('rgb', year, df_all_colors)
    for color in colors:
        try:
            df_stack.loc[color, year] = df_feature.loc[color, 'pc']
        except:
            df_stack.loc[color, year] = np.nan
df_stack = df_stack[::-1] # [::-1] reverses order
# df_stack
```

```
[191]: # function to plot it
def stacked_plot(plot_name, _df):

    data = []
    font_size=12
```

```

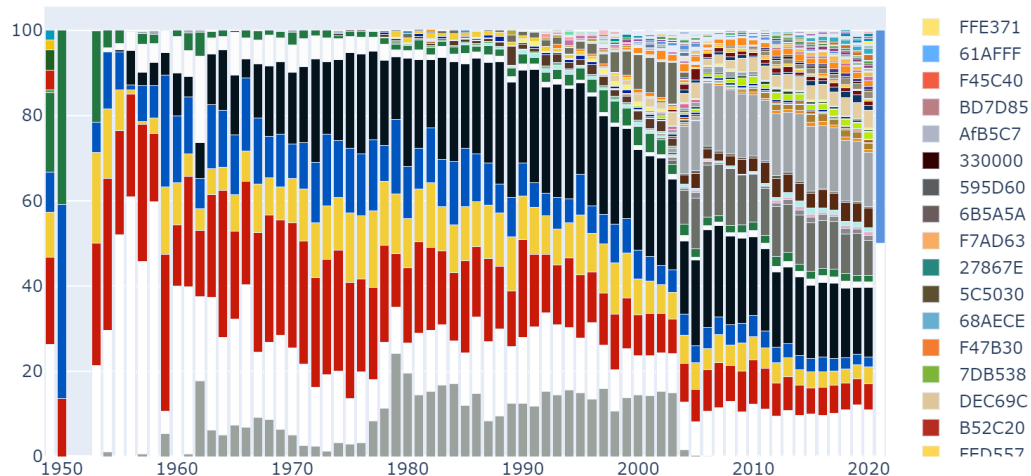
for color in colors:
    trace = go.Bar(x=_df.columns,
                    y=_df.loc[str(color)],
                    name=str(color),
                    marker=dict(color='#'+str(color)))

    )
    data.append(trace)
layout = go.Layout(title=plot_name,
                    barmode='stack',
                    font=dict(size=font_size))
fig = go.Figure(data=data, layout=layout)
#fig.show()
fig.write_image(f"{plot_name}.png", width=800, scale=2)

```

```
[194]: stacked_plot("Proportion_of_Lego_piece_colours_for_all_sets_by_year", df_stack)
```

Proportion_of_Lego_piece_colours_for_all_sets_by_year



Now plot by count.

```

[195]: import numpy as np
# get the list of unique colours for each year
colors = list(df_all_colors['rgb'].unique())
# get the unique years
year = list(df_all_colors['year'].unique())
# set up a new dataframe
df_stack_count = pd.DataFrame(index=colors, columns=years)
# get the portion of all colors for each colours in each year

```

```

for year in years:
    df_feature, feature = count_a_feature('rgb', year, df_all_colors)
    for color in colors:
        try:
            df_stack_count.loc[color, year] = df_feature.loc[color, 'count']
        except:
            df_stack_count.loc[color, year] = np.nan
df_stack_count = df_stack_count[::-1] # reverse order
#df_stack_count

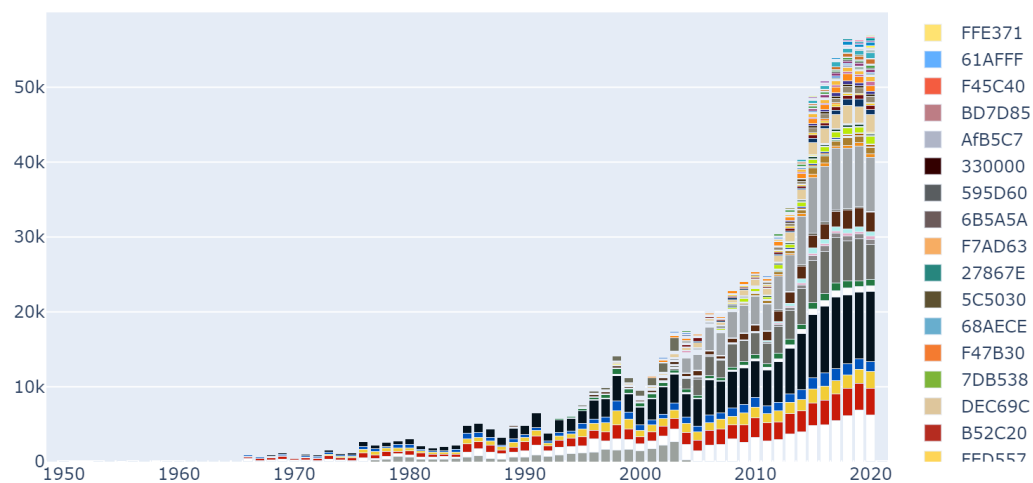
```

```

[196]: stacked_plot("Count_of_Lego_pieces_by_colour_in_all_sets_by_year",
    ↪df_stack_count)
# could filter this by theme
# could add title to legend as 'rgb'

```

Count_of_Lego_pieces_by_colour_in_all_sets_by_year



1.3 4. Get market price data

Use the [Brick Link API](#) to get guide price data of all sets

```

[ ]: from bricklink_api.auth import oauth
from bricklink_api.catalog_item import get_price_guide, Type, NewOrUsed

# authorisation details from a seller account https://www.bricklink.com/v2/api/
    ↪register_consumer.page
consumer_key = "9667E5445FA043538C4C34B2AFE74CFB"
consumer_secret = "D9AAF83E1F574360896C81A3072B48C6"
token_value = "8AFE17DBFFAB46629A6214E5EAD57F37"

```

```

token_secret = "967AB439A70E4D45831686F1E200817C"
auth = oauth(consumer_key, consumer_secret, token_value, token_secret)

# Example, get price guide for a used 42100-1 (Lego Technic Liebherr R 9800)
json_obj = get_price_guide(Type.SET, "42100-1", new_or_used='U', auth=auth)
#json_obj

```

```

[ ]: def get_set_price_data(_set, N_or_U):

    """function to get the price data of a given
    set number as new (N) or used (U)"""

    try:
        json_obj = get_price_guide(Type.SET, _set, new_or_used=N_or_U,
        ↪auth=auth)
        average = json_obj['data']['avg_price']
        minimum = json_obj['data']['min_price']
        maximum = json_obj['data']['max_price']
        quantity = json_obj['data']['total_quantity']
        currency = json_obj['data']['currency_code']
    except:
        average, minimum, maximum, quantity, currency = 0., 0., 0., 0., 'none'

    return average, minimum, maximum, quantity, currency

```

Get all new price data for all sets

```

[ ]: #https://stackoverflow.com/a/20627316/8971265
pd.options.mode.chained_assignment = None # default='warn'

import time

# new columns
df_sets['price_mean_N'] = ''
df_sets['price_min_N'] = ''
df_sets['price_max_N'] = ''
df_sets['set_qty_N'] = ''
df_sets['currency_N'] = ''

i = 0 # row counter
for _set in df_sets['set_num']:

    print('Set:', _set)

    # get the set price data
    result = get_set_price_data(_set, 'N')
    df_sets.loc[[i], ['price_mean_N']] = result[0]

```

```

df_sets.loc[[i], ['price_min_N']] = result[1]
df_sets.loc[[i], ['price_max_N']] = result[2]
df_sets.loc[[i], ['set_qty_N']] = result[3]
df_sets.loc[[i], ['currency_N']] = result[4]

# print the row number and Bricklink result
print('Row:', i, result)

#bump the counter
i += 1

# sleep 24 hours every 4000 calls (to avoid exceeding the API limitations)
if i % 4000 == 0:

    print("Saving...")
    print(f"Set, i={i}, day={time.localtime().tm_mday}")
    # wait 24 hours
    print("Sleeping...")
    time.sleep(60*60*24)

df_sets.to_csv("data/df_sets_new.csv")
del df_sets

```

Load from csv

```
[197]: df_sets = pd.read_csv("data/df_sets_new.csv", index_col=0)
df_sets.tail()
```

```
[197]:
```

| | set_num | name | year | theme_id | \ |
|-------|-------------|-------------------------------------|------|----------|---|
| 15492 | wgwp1-1 | Wild West Limited Edition Gift Pack | 1996 | 476 | |
| 15493 | XMASTREE-1 | Christmas Tree | 2019 | 410 | |
| 15494 | XWING-1 | Mini X-Wing Fighter | 2019 | 158 | |
| 15495 | XWING-2 | X-Wing Trench Run | 2019 | 158 | |
| 15496 | YODACHRON-1 | Yoda Chronicles Promotional Set | 2013 | 158 | |

| | num_parts | price_mean_N | price_min_N | price_max_N | set_qty_N | currency_N |
|-------|-----------|--------------|-------------|-------------|-----------|------------|
| 15492 | 0 | 0.0000 | 0.0000 | 0.0000 | 0.0 | USD |
| 15493 | 26 | 0.0000 | 0.0000 | 0.0000 | 0.0 | none |
| 15494 | 60 | 0.0000 | 0.0000 | 0.0000 | 0.0 | none |
| 15495 | 52 | 0.0000 | 0.0000 | 0.0000 | 0.0 | none |
| 15496 | 413 | 265.2236 | 265.2236 | 265.2236 | 2.0 | USD |

```
[198]: # Typical example, The Chicken Coop
df_sets[df_sets['set_num']=='21140-1']
```

```
[198]:
```

| | set_num | name | year | theme_id | num_parts | price_mean_N | \ |
|------|---------|------------------|------|----------|-----------|--------------|---|
| 1902 | 21140-1 | The Chicken Coop | 2018 | 577 | 198 | 22.8214 | |

| | price_min_N | price_max_N | set_qty_N | currency_N |
|------|-------------|-------------|-----------|------------|
| 1902 | 14.99 | 35.5129 | 328.0 | USD |

```
[199]: # remove the zero-priced sets
df_prices_N = df_sets[df_sets['price_mean_N'] > 0.0]
df_prices_N
```

```
[199]:
```

| | set_num | | name | year | \ |
|-------|-------------|---|---------------------------------|------|---|
| 30 | 044-1 | | Basic Building Set | 1968 | |
| 51 | 074-1 | | Pre-School Set | 1976 | |
| 67 | 088-1 | | Super Set | 1969 | |
| 68 | 10000-1 | | Guarded Inn | 2001 | |
| 70 | 10001-1 | | Metroliner | 2001 | |
| ... | ... | | ... | ... | |
| 15481 | Watford-1 | LEGO Store Grand Opening Exclusive Set, Watfor... | | 2013 | |
| 15482 | Wauwatosa-1 | LEGO Store Grand Opening Exclusive Set, Mayfai... | | 2012 | |
| 15489 | Wiesbaden-1 | LEGO Store Grand Opening Exclusive Set, Wiesba... | | 2010 | |
| 15490 | WILLIAM-1 | | Will.i.am | 2016 | |
| 15496 | YODACHRON-1 | | Yoda Chronicles Promotional Set | 2013 | |

| | theme_id | num_parts | price_mean_N | price_min_N | price_max_N | set_qty_N | \ |
|-------|----------|-----------|--------------|-------------|-------------|-----------|---|
| 30 | 366 | 225 | 165.6266 | 165.6266 | 165.6266 | 1.0 | |
| 51 | 505 | 20 | 60.0000 | 60.0000 | 60.0000 | 1.0 | |
| 67 | 469 | 615 | 2000.0000 | 2000.0000 | 2000.0000 | 1.0 | |
| 68 | 199 | 256 | 262.5159 | 199.0000 | 352.2405 | 14.0 | |
| 70 | 233 | 787 | 1477.5102 | 899.9900 | 3640.4192 | 5.0 | |
| ... | ... | ... | ... | ... | ... | ... | |
| 15481 | 408 | 15 | 236.0687 | 236.0687 | 236.0687 | 1.0 | |
| 15482 | 408 | 15 | 191.4508 | 150.0000 | 234.8883 | 5.0 | |
| 15489 | 408 | 146 | 211.2815 | 211.2815 | 211.2815 | 1.0 | |
| 15490 | 535 | 3 | 1770.4561 | 1770.4561 | 1770.4561 | 1.0 | |
| 15496 | 158 | 413 | 265.2236 | 265.2236 | 265.2236 | 2.0 | |

| | currency_N |
|-------|------------|
| 30 | USD |
| 51 | USD |
| 67 | USD |
| 68 | USD |
| 70 | USD |
| ... | ... |
| 15481 | USD |
| 15482 | USD |
| 15489 | USD |
| 15490 | USD |
| 15496 | USD |

[9527 rows x 10 columns]

```
[201]: # plot scatter of price and num_parts
data = [go.Scatter(x=df_prices_N['price_mean_N'],
                    y=df_prices_N['num_parts'],
                    mode='markers'
                )
]
layout=go.Layout(title='Number of parts versus price',
                  xaxis=dict(title='mean price'),
                  yaxis=dict(title='num_parts'))
fig = go.Figure(data=data, layout=layout)
#fig.show()
fig.write_image("5.png", width=800, scale=2)
```



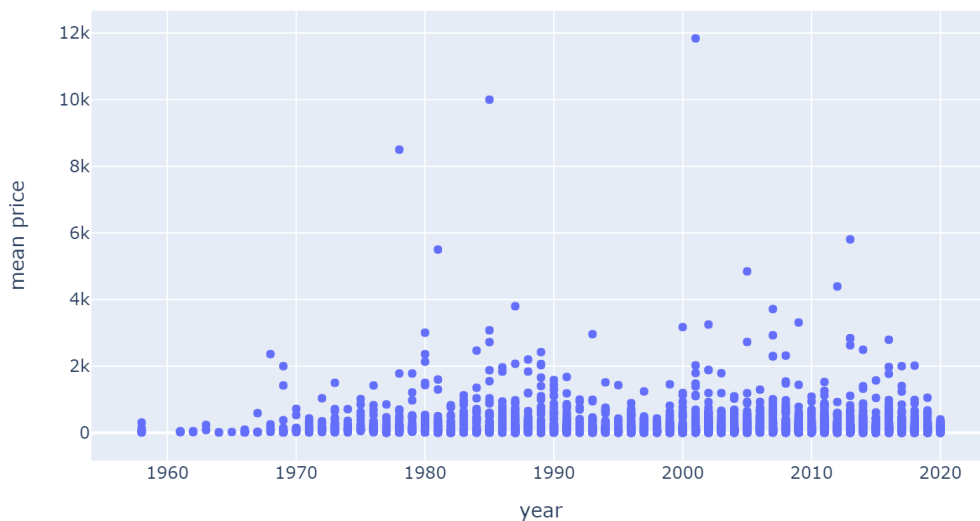
```
[202]: corr_matrix = df_prices_N.corr()
corr_matrix['price_mean_N'].sort_values(ascending=False)
```

```
[202]: price_mean_N    1.000000
price_min_N    0.972301
price_max_N    0.896906
num_parts      0.304013
set_qty_N     -0.077959
theme_id      -0.084471
year          -0.141753
Name: price_mean_N, dtype: float64
```


This result tells us that the minimum and maximum prices are strongly correlated to the mean price, both having Pearson correlation number close to 1. But that information is not useful. The number of parts in a set is correlated to the mean price having a score of 0.3. The positive score means that the higher the number of parts, the higher the price. Interestingly, we can see that the year has a slight negative correlation to price; the lower the year the higher the price.

```
[203]: # plot scatter of price and num_parts
data = [go.Scatter(y=df_prices_N['price_mean_N'],
                  x=df_prices_N['year'],
                  mode='markers',
                  hovertext=df_prices_N['name'])
        ]
layout=go.Layout(title='Price versus year',
                 yaxis=dict(title='mean price'),
                 xaxis=dict(title='year'))
fig = go.Figure(data=data, layout=layout)
#fig.show()
fig.write_image("6.png", width=800, scale=2)
```

Price versus year



```
[145]: # (perhaps identify and plot by theme separately)
# list of unique themes names
df_themes.name.unique()
```

```
[145]: array(['Arctic Technic', 'Competition', 'Expert Builder', 'Model',
            'Airport', 'Construction', 'Farm', 'Fire', 'Harbor', 'Off-Road',
```

'Race', 'Riding Cycle', 'Robot', 'Traffic', 'RoboRiders',
'Speed Slammers', 'Star Wars', 'Supplemental', 'Throwbot Slizer',
'Universal Building Set', 'Basic Model', 'Castle', 'Train',
'Creature', 'Food & Drink', 'Building', 'Cargo', 'Basic Set',
'Recreation', 'Mecha', 'Arctic', 'City', 'Coast Guard', 'Hospital',
'Police', 'Trains', 'Classic Town', 'Station', 'Post Office',
'Divers', 'Extreme Team', 'Launch Command', 'Outback', 'Paradisa',
'Res-Q', 'Space Port', 'Town Jr.', 'Gas Station', 'Town Plan',
'World City', 'Drome Racers', 'Ferrari', 'Lamborghini',
'Power Racers', 'Radio Control', 'Speed Racer', 'Tiny Turbos',
'Track System', 'Williams F1', 'World Racers', 'Xalax',
'Alien Conquest', 'Blacktron I', 'Blacktron II', 'Classic Space',
'Exploriens', 'Futuron', 'Ice Planet 2002', 'Insectoids',
'Life On Mars', 'M:Tron', 'Mars Mission', 'RoboForce',
'Space Police I', 'Space Police II', 'Space Police III', 'Spyrius',
'UFO', 'Unitron', 'Galaxy Squad', 'Pirates I', 'Imperial Armada',
'Imperial Soldiers', 'Islanders', 'Pirates II', 'Pirates III',
'Mini', 'Bricktober', 'Ultimate Collector Series', 'Black Falcons',
'Black Knights', 'Classic Castle', 'Crusaders', 'Dark Forest',
'Fantasy Era', 'Forestmen', 'Fright Knights', 'Kingdoms',
'Knights Kingdom I', 'Knights Kingdom II', 'Lion Knights',
'My Own Creation', 'Royal Knights', 'Wolfpack', 'Advent',
'Belville', 'Classic Basic', 'Clikits', 'Creator', 'Pirates',
'Friends', 'Christmas', 'Easter', 'Halloween', 'Thanksgiving',
'Valentine', '12V', '4.5V', '9V', 'My Own Train', 'RC Train',
'Skylines', 'Technic', 'NXT', 'RCX', 'EV3', 'Jurassic Park III',
'Mosaic', 'Jack Stone', 'Spider-Man', 'Desert', 'Dino Island',
'Jungle', 'Orient Expedition', 'Ultra Agents', 'Mission Deep Sea',
'Mission Deep Freeze', 'Aquanauts', 'Aquaraiders I',
'Aquaraiders II', 'Aquasharks', 'Hydronauts', 'Stingrays',
'Fairy-Tale', 'Golden Land', 'Playhouse', 'Agori', 'Barraki',
'Battle Vehicles', 'Bohrok', 'Bohrok Va', 'Bohrok-Kal',
'Glatorian', 'Glatorian Legends', 'Matoran of Light',
'Matoran of Mahri Nui', 'Matoran of Mata Nui',
'Matoran of Metru Nui', 'Matoran of Voya Nui', 'Mistika',
'Phantoka', 'Piraka', 'Playsets', 'Rahaga', 'Rahi', 'Rahkshi',
'Stars', 'Titans', 'Toa', 'Toa Hagah', 'Toa Hordika', 'Toa Inika',
'Toa Mahri', 'Toa Metru', 'Toa Nuva', 'Tohunga', 'Turaga', 'Vahki',
'Visorak', 'Warriors', 'Protectors', 'Skull Spiders', 'Toa Okoto',
'HO 1:87 Vehicles', 'Jumbo Bricks', 'Vehicle', 'Wooden Box Set',
'Heroes', 'Vehicles', 'Villains', 'Monthly Mini Model Build',
'Pick A Model', 'Space', 'Western', 'Bionicle', 'Holiday',
'Airjitzu', 'Adventurers', 'Aquazone', 'Fabuland', 'Primo',
'Scala', 'Town', 'Basketball', 'Gravity Games', 'Hockey', 'Soccer',
'Basic', 'Ferries', 'Gears', 'Cowboys', 'Indians', 'Game',
'Key Chain', 'Cars', 'Boat', 'Building Set with People', 'Classic',
'Dinosaurs', 'Duplo and Explore', 'Learning', 'Mindstorms', 'WeDo',

```
'Samsonite', 'Service Packs', 'Soft Bricks', 'Control Lab', 'eLAB',
'Series 1 Minifigures', 'Series 2 Minifigures',
'Series 3 Minifigures', 'Series 4 Minifigures',
'Series 5 Minifigures', 'Series 6 Minifigures',
'Series 7 Minifigures', 'Series 8 Minifigures',
'Series 9 Minifigures', 'Series 10 Minifigures', 'Team GB',
'Series 11 Minifigures', 'Series 12 Minifigures',
'The LEGO Movie Series', 'The Simpsons', 'Series 13 Minifigures',
'Series 14 Minifigures', 'The Simpsons Series 2',
'Series 15 Minifigures', 'Disney Series 1',
'Series 16 Minifigures', 'DFB Minifigures', 'The Hobbit',
'The Lord of the Rings', 'Speedorz', 'Constraction',
'Legend Beasts', 'Series 1', 'Series 2', 'Series 3', 'Series 4',
'Series 5', 'Series 6', 'Series 7', 'Series 8', 'Series 9',
'DC Comics Super Heroes', 'Disney Princess', 'Marvel Super Heroes',
'Ninjago', 'The LEGO Batman Movie', 'Series 17 Minifigures',
'Ninjago The Movie', 'Cities of Wonders',
'Jurassic World: Fallen Kingdom', 'Series 18 Minifigures',
'Bob the Builder', 'Disney Planes', 'Dora the Explorer',
'Miles From Tomorrowland', 'Spiderman', 'Thomas & Friends',
'Toy Story', 'Winnie the Pooh', 'My Town',
'Jake and the Never Land Pirates', 'Sofia the First',
'Disney's Mickey Mouse', 'Doc McStuffins', 'Legoville',
'DC Comics', 'Powered Up',
'Harry Potter and Fantastic Beasts Series 1', 'Action Wheelers',
'Dino', 'Dolls', 'Little Forest Friends', 'Little Robots',
'Princess Castle', 'Rattles', 'Toolo', 'Zooters',
'Fantastic Beasts', 'The LEGO Movie II', 'Creator 3-in-1',
'Creator Expert', 'Early Creator', 'The LEGO Movie Series II',
'Disney Series 2', 'Mars Exploration', 'Series 19 Minifigures',
'DC Super Heroes', 'Frozen II',
'Harry Potter and Fantastic Beasts Series 2',
'Series 20 Minifigures', 'Batman', 'UCS', 'Justice League',
'Superman', 'Avengers', 'Guardians of the Galaxy', 'Iron Man',
'X-Men', 'Harry Potter'], dtype=object)
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```
[ ]: #-----example of price guide-----
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[91]: # get price guide
json_obj = get_price_guide(Type.SET, "21138-1", new_or_used='U', auth=auth)
json_obj
```

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[91]: {'meta': {'description': 'OK', 'message': 'OK', 'code': 200},
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               'currency_code': 'USD',
               'min_price': '19.0000',
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  'shipping_available': True,
  'qunatity': 1},
{'quantity': 1,
  'unit_price': '24.0000',
  'shipping_available': False,
  'qunatity': 1},
{'quantity': 1,
  'unit_price': '19.0000',
  'shipping_available': False,
  'qunatity': 1}]]}
```

[]: