

# Understanding Lego sets popularity

## Background

You recently applied to work as a data analyst intern at the famous Lego Group in Denmark. As part of the job interview process, you received the following take-home assignment:

You are asked to use the provided dataset to understand the popularity of different Lego sets and themes. The idea is to become familiarized with the data to be ready for an interview with a business stakeholder.



## The data

You received access to a database with the following tables. You can also see above a visualization of how the tables are related to each other. ([source](#)):

### inventory\_parts

- "inventory\_id" - id of the inventory the part is in (as in the inventories table)
- "part\_num" - unique id for the part (as in the parts table)
- "color\_id" - id of the color
- "quantity" - the number of copies of the part included in the set
- "is\_spare" - whether or not it is a spare part

### parts

- "part\_num" - unique id for the part (as in the inventory\_parts table)
- "name" - name of the part
- "part\_cat\_id" - part category id (as in part\_categories table)

### part\_categories

- "id" - part category id (as in parts table)
- "name" - name of the category the part belongs to

### colors

- "id" - id of the color (as in inventory\_parts table)
- "name" - color name
- "rgb" - rgb code of the color
- "is\_trans" - whether or not the part is transparent/translucent

### inventories

- "id" - id of the inventory the part is in (as in the inventory\_sets and inventory\_parts tables)
- "version" - version number
- "set\_num" - set number (as in sets table)

### inventory\_sets

- "inventory\_id" - id of the inventory the part is in (as in the inventories table)
- "set\_num" - set number (as in sets table)
- "quantity" - the quantity of sets included

## sets

- "set\_num" - unique set id (as in inventory\_sets and inventories tables)
- "name" - the name of the set
- "year" - the year the set was published
- "theme\_id" - the id of the theme the set belongs to (as in themes table)
- num-parts - the number of parts in the set

## themes

- "id" - the id of the theme (as in the sets table)
- "name" - the name of the theme
- "parent\_id" - the id of the larger theme, if there is one

**\*Acknowledgments:** Rebrickable.com\*

## Challenge

Create a report to summarize your findings. Include:

1. What is the average number of Lego sets released per year?
2. What is the average number of Lego parts per year?
3. Create a visualization for item 2.
4. What are the 5 most popular colors used in Lego parts?
5. [Optional] What proportion of Lego parts are transparent?
6. [Optional] What are the 5 rarest lego bricks?
7. Summarize your findings.

## Checklist before publishing

- Rename your workspace to make it descriptive of your work. N.B. you should leave the notebook name as notebook.ipynb.
- **Remove redundant cells** like the introduction to data science notebooks, so the workbook is focused on your story.
- Check that all the cells run without error.

# Import modules and datasets

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')

%matplotlib inline
```

```

themes = pd.read_csv('themes.csv')
sets = pd.read_csv('sets.csv')
parts = pd.read_csv('parts.csv')
part_categories = pd.read_csv('part_categories.csv')
inventory_sets = pd.read_csv('inventory_sets.csv')
inventory_parts = pd.read_csv('inventory_parts.csv')
inventories = pd.read_csv('inventories.csv')
colors = pd.read_csv('colors.csv')

```

## Rename & Merge Columns

```

In [2]: # renaming individual dataframe columns for merging and differentiation
inventories.rename(columns={'id':'inventory_id'}, inplace=True)
inventory_sets.rename(columns={'quantity':'quantity_inv_sets'}, inplace=True)
sets.rename(columns={'name':'name_sets', 'num_parts':'num_parts_sets'}, inplace=True)
themes.rename(columns={'id':'theme_id', 'name':'name_themes'}, inplace=True)
inventory_parts.rename(columns={'quantity':'inv_parts_quantity'}, inplace=True)
colors.rename(columns={'id':'color_id', 'name':'name_colors'}, inplace=True)
part_categories.rename(columns={'id':'part_cat_id', 'name':'name_part_cat'}, inplace=True)

```

```

In [3]: # merge inventory_sets with sets as df on common columns
df = pd.merge(inventory_sets, sets, on='set_num')

```

```

In [4]: # merge df with inventories via left side
df = df.merge(inventories, on=['inventory_id', 'set_num'], how='left')

```

```

In [5]: df = df.merge(themes, on='theme_id', how='left')

```

```

In [6]: df = df.merge(inventory_parts, on='inventory_id', how='left')

```

```

In [7]: parts.rename(columns={'name':'name_parts'}, inplace=True)
df = df.merge(parts, on='part_num')

```

```

In [8]: df = df.merge(part_categories, on='part_cat_id')

```

```

In [9]: df = df.merge(colors, on='color_id')

```

```

In [10]: # check first 2 rows of df
df.head(2)

```

```

Out[10]:
   inventory_id  set_num  quantity_inv_sets  name_sets  year  theme_id  num_parts_sets
0           311    8593-1                1    Makuta   2003      324             199  https://cdn.rebrickable.com/
1           311    8596-1                1  Takanuva   2003      324             199  https://cdn.rebrickable.com/

```

```

In [11]: # drop irrelevant columns
df.drop(['img_url', 'version', 'is_spare', 'part_material', 'rgb', 'parent_id'], axis=1,

```

```
# reset index
df.reset_index(drop=True, inplace=True)
```

```
In [12]: # inspect for duplicate rows
df.duplicated().sum()
```

```
Out[12]: 0
```

```
In [14]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 725 entries, 0 to 724
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  -
0   inventory_id           725 non-null    int64
1   set_num                725 non-null    object
2   quantity_inv_sets      725 non-null    int64
3   name_sets              725 non-null    object
4   year                   725 non-null    int64
5   theme_id               725 non-null    int64
6   num_parts_sets         725 non-null    int64
7   name_themes            725 non-null    object
8   part_num               725 non-null    object
9   color_id               725 non-null    float64
10  inv_parts_quantity      725 non-null    float64
11  name_parts              725 non-null    object
12  part_cat_id             725 non-null    int64
13  name_part_cat           725 non-null    object
14  name_colors             725 non-null    object
15  is_trans                725 non-null    object
dtypes: float64(2), int64(6), object(8)
memory usage: 90.8+ KB
```

- df has a total of 725 rows (including headers), and 16 columns
- There appears to be no missing values
- most of the 'int64' columns may need to be converted into string for feature engineering

```
In [15]: # confirm null values
df.isnull().any()
```

```
Out[15]: inventory_id           False
set_num                False
quantity_inv_sets      False
name_sets              False
year                   False
theme_id               False
num_parts_sets         False
name_themes            False
part_num               False
color_id               False
inv_parts_quantity      False
name_parts              False
part_cat_id             False
name_part_cat           False
name_colors             False
is_trans                False
dtype: bool
```

```
In [16]: # df.to_excel('mergedlego.xlsx')
```

# Exploratory Data Analysis

```
In [17]: # view statistics of df
df.describe(include='all')
```

```
Out[17]:
```

|        | inventory_id  | set_num | quantity_inv_sets | name_sets | year        | theme_id   | num_parts_sets | name_the  |
|--------|---------------|---------|-------------------|-----------|-------------|------------|----------------|-----------|
| count  | 725.000000    | 725     | 725.000000        | 725       | 725.000000  | 725.000000 | 725.000000     |           |
| unique | NaN           | 167     | NaN               | 166       | NaN         | NaN        | NaN            |           |
| top    | NaN           | 8580-1  | NaN               | Kraata    | NaN         | NaN        | NaN            | Star Wars |
| freq   | NaN           | 76      | NaN               | 76        | NaN         | NaN        | NaN            |           |
| mean   | 29691.263448  | NaN     | 1.135172          | NaN       | 2002.754483 | 374.782069 | 126.787586     |           |
| std    | 37415.223658  | NaN     | 0.346157          | NaN       | 9.255303    | 172.163434 | 192.856806     |           |
| min    | 311.000000    | NaN     | 1.000000          | NaN       | 1969.000000 | 18.000000  | 0.000000       |           |
| 25%    | 9820.000000   | NaN     | 1.000000          | NaN       | 1998.000000 | 236.000000 | 14.000000      |           |
| 50%    | 11554.000000  | NaN     | 1.000000          | NaN       | 2003.000000 | 324.000000 | 59.000000      |           |
| 75%    | 24973.000000  | NaN     | 1.000000          | NaN       | 2009.000000 | 532.000000 | 170.000000     |           |
| max    | 136912.000000 | NaN     | 3.000000          | NaN       | 2022.000000 | 744.000000 | 1412.000000    |           |

- The top sets name is Kraata appearing x76 ('name\_sets' - the name of the set)
- There are 56 unique theme names, with the top themes name; Star Wars appearing x113 ('name\_themes' - the name of the theme)
- There are a total of 167 unique set ids ('set\_num' - unique set id)
- 166 unique set names ('name\_sets' - the name of the set)
- 56 unique theme names ('name\_themes' - the name of the theme)
- 337 unique id for parts ('part\_num' - unique id for the part)
- 336 unique names of parts ('name\_part' - name of the part)
- there are 45 types of colors

```
In [18]: # convert int values to string
df[['inventory_id', 'theme_id', 'color_id', 'part_cat_id']].apply(str)
```

```
Out[18]:
```

|              |   |        |        |          |           |
|--------------|---|--------|--------|----------|-----------|
| inventory_id | 0 | 311\n1 | 311\n2 | 16117\n3 | ...       |
| theme_id     | 0 | 324\n1 | 324\n2 | 324\n3   | 324...    |
| color_id     | 0 | 0.0\n1 | 0.0\n2 | 0.0\n3   | ...       |
| part_cat_id  | 0 | 41\n1  | 41\n2  | 41\n3    | 41\n4 ... |

dtype: object

## Average total of Lego Sets released per Year

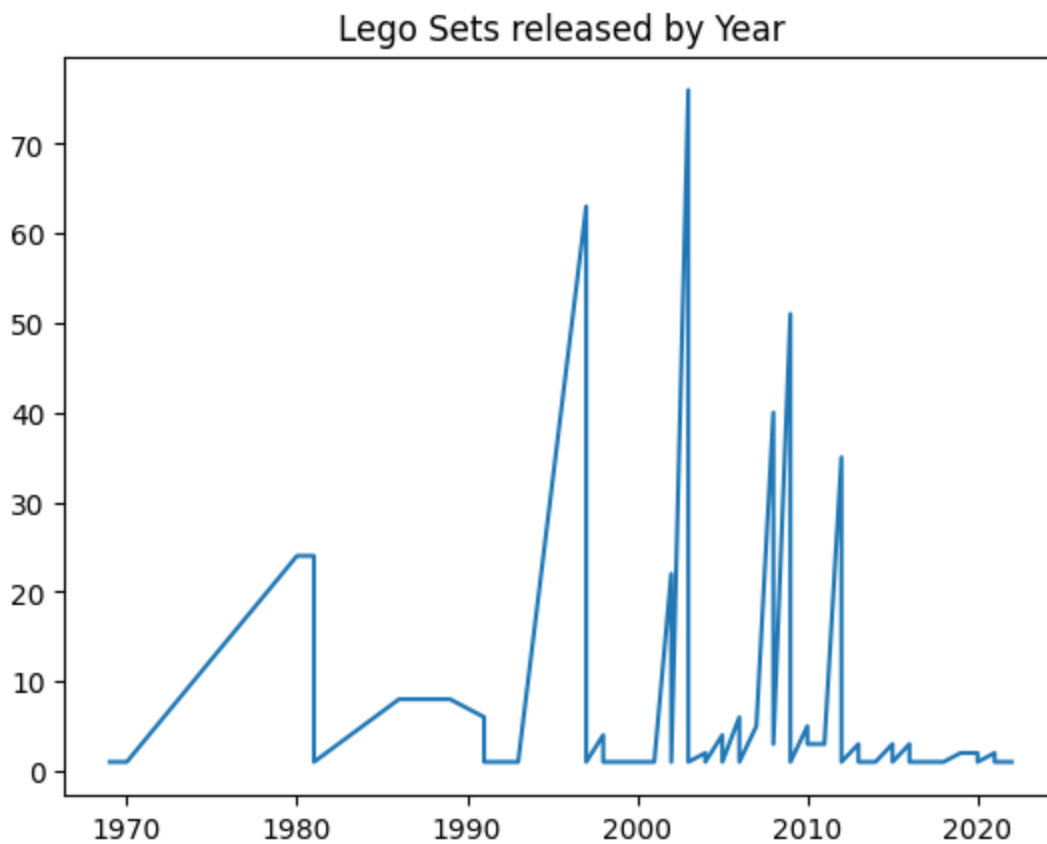
```
In [39]: # create dataframe groupedby 'year' with 'set_num' value counts
year_sets = df.groupby(['year'], as_index=False)['set_num'].value_counts()

# calculate the average
avg_year_sets = round(sum(year_sets['count'])/len(year_sets['year']), 0)
```

```
print('The average total of Lego sets released per year is:', avg_year_sets)
```

The average total of Lego sets released per year is: 4.0

```
In [41]: # plot total of lego sets released per year
plt.plot(year_sets['year'], year_sets['count'])
plt.title('Lego Sets released by Year')
plt.show()
```



```
In [61]: # view highest 5 sets released count
year_sets.sort_values(by='count', ascending=False)[:5]
```

```
Out[61]:
```

|     | year | set_num   | count |
|-----|------|-----------|-------|
| 54  | 2003 | 8580-1    | 76    |
| 16  | 1997 | 9847-1    | 63    |
| 118 | 2009 | 20009-1   | 51    |
| 113 | 2008 | 20004-1   | 40    |
| 128 | 2012 | 5000062-1 | 35    |

```
In [63]: #view smallest 3 sets released counts
year_sets.nsmallest(3, ['count'])
```

```
Out[63]:
```

|   | year | set_num | count |
|---|------|---------|-------|
| 0 | 1969 | 344-1   | 1     |
| 1 | 1969 | 346-2   | 1     |
| 2 | 1970 | 603-3   | 1     |

- 1997 and 2003 recorded the highest amount of Lego sets released
- The trend has remained below 10 sets since 2012

## Average Total of Lego Parts released per Year

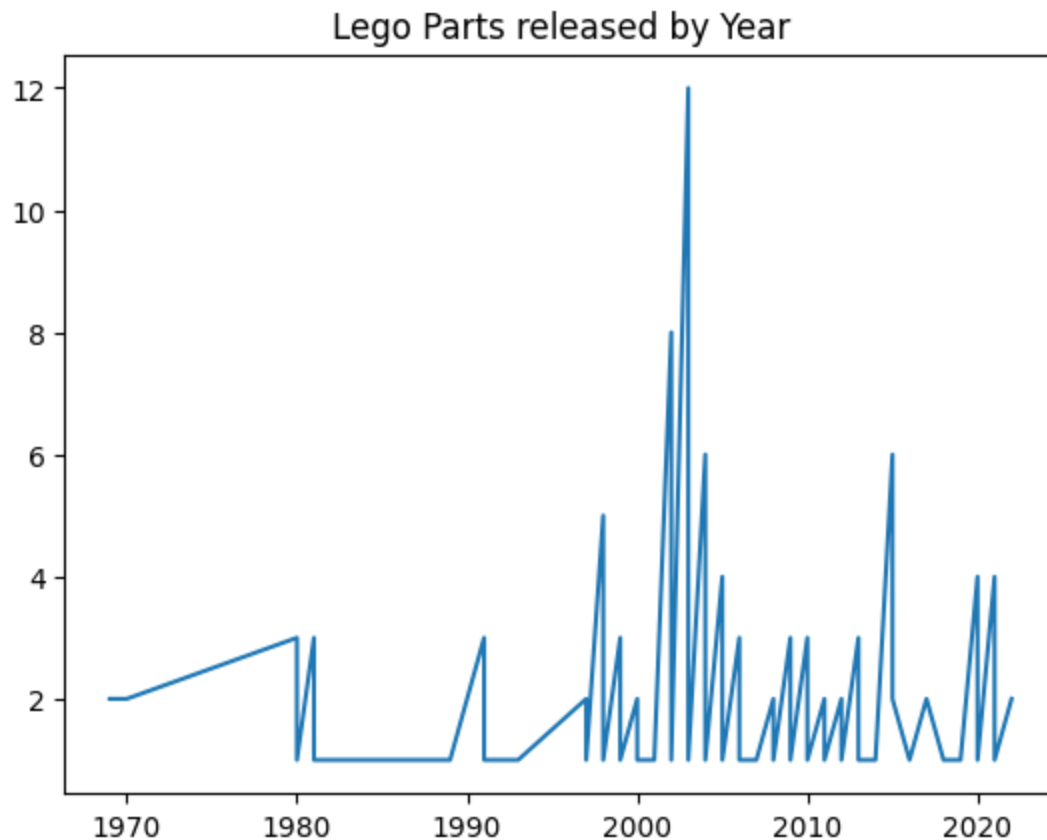
```
In [36]: # create dataframe groupedby 'year' and 'part_num' value counts
year_parts = df.groupby(['year'], as_index=False)['part_num'].value_counts()

# calculate average
avg_year_parts = round(sum(year_parts['count'])/len(year_parts['part_num']), 0)

print('The average total of Lego parts released per year is:', avg_year_parts)
```

The average total of Lego parts released per year is: 2.0

```
In [42]: # plot total Lego parts released per year
plt.plot(year_parts['year'], year_parts['count'])
plt.title('Lego Parts released by Year')
plt.show()
```



```
In [59]: # view top 5 parts released count
year_parts.sort_values(by='count', ascending=False)[:5]
```

```
Out[59]:
```

|     | year | part_num | count |
|-----|------|----------|-------|
| 176 | 2003 | 43559    | 12    |
| 161 | 2002 | 74747    | 8     |
| 169 | 2002 | 70931    | 8     |
| 167 | 2002 | 6035     | 8     |
| 166 | 2002 | 5306c01  | 8     |

```
In [64]: # view smallest 3 parts released count
year_parts.nsmallest(3, ['count'])
```

```
Out[64]:
```

|   | year | part_num | count |
|---|------|----------|-------|
| 4 | 1980 | 3623     | 1     |
| 5 | 1980 | 3633     | 1     |
| 6 | 1980 | 3849     | 1     |

- 2003 recorded the highest amount of Lego parts released, peaking at 12
- Lego parts release count has remained below 7 after the peak, until 2020

## 5 Most Popular Colors used in Lego Parts

```
In [23]: # create groupby dataframe with name_parts and Value_counts of colors
part_color = df.groupby(['name_parts'], as_index=False)['name_colors'].value_counts().so
part_color
```

```
Out[23]:
```

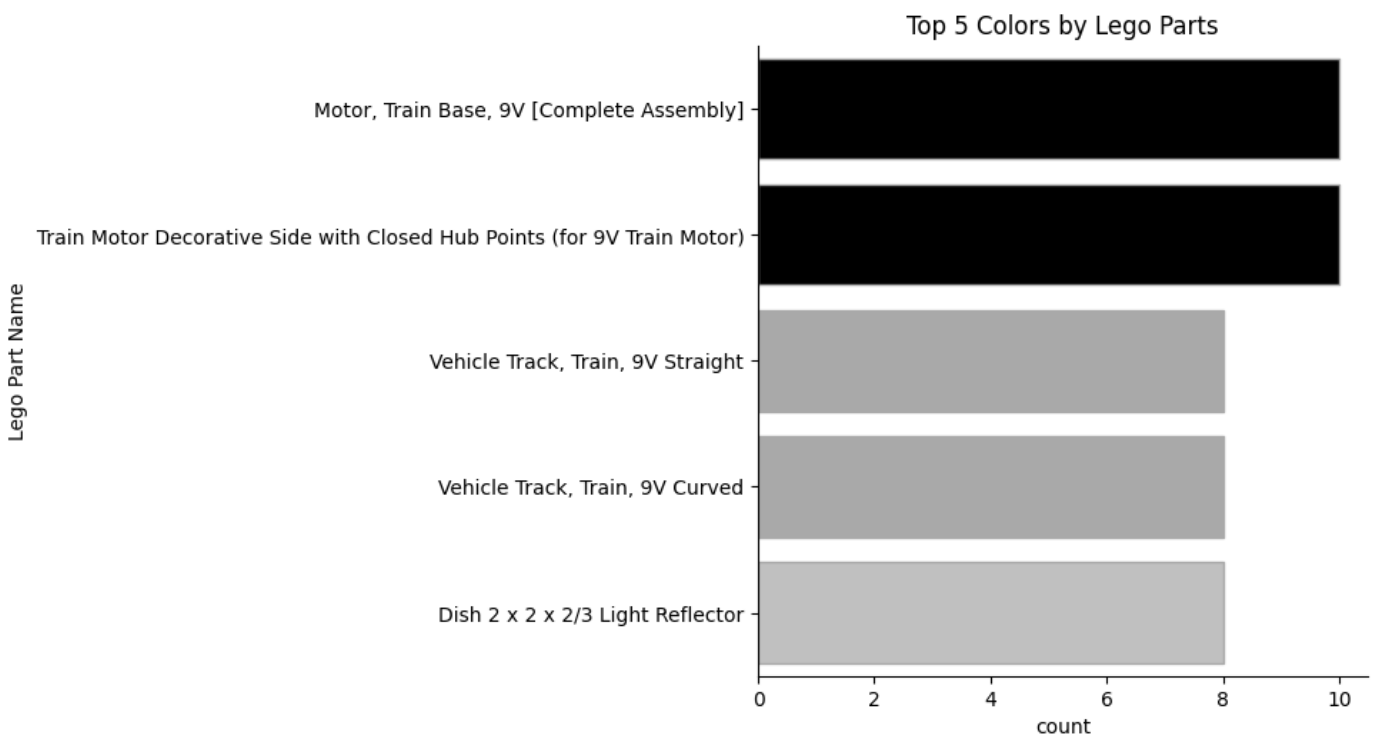
|     | name_parts  | name_colors   | count |
|-----|---|---------------|-------|
| 165 | Motor, Train Base, 9V [Complete Assembly]         | Black         | 10    |
| 377 | Train Motor Decorative Side with Closed Hub Po... | Black         | 10    |
| 384 | Vehicle Track, Train, 9V Straight                 | Dark Gray     | 8     |
| 382 | Vehicle Track, Train, 9V Curved                   | Dark Gray     | 8     |
| 65  | Dish 2 x 2 x 2/3 Light Reflector                  | Chrome Silver | 8     |

```
In [52]: plt.figure(figsize=(12,7))
g = sns.catplot(part_color, y='name_parts', x='count', kind='bar',
                palette=['black', 'black', 'darkgrey', 'darkgrey', 'silver'],
                edgecolor='darkgrey')

plt.title('Top 5 Colors by Lego Parts')
plt.ylabel('Lego Part Name')
plt.show()
```

<Figure size 1200x700 with 0 Axes>





## The most popular Colors Overall

```
In [25]: bricks = df.groupby(['name_colors'], as_index=False)['name_colors'].value_counts()
```

```
In [26]: pop_bricks = bricks.nlargest(5, ['count'])
pop_bricks
```

```
Out[26]:
```

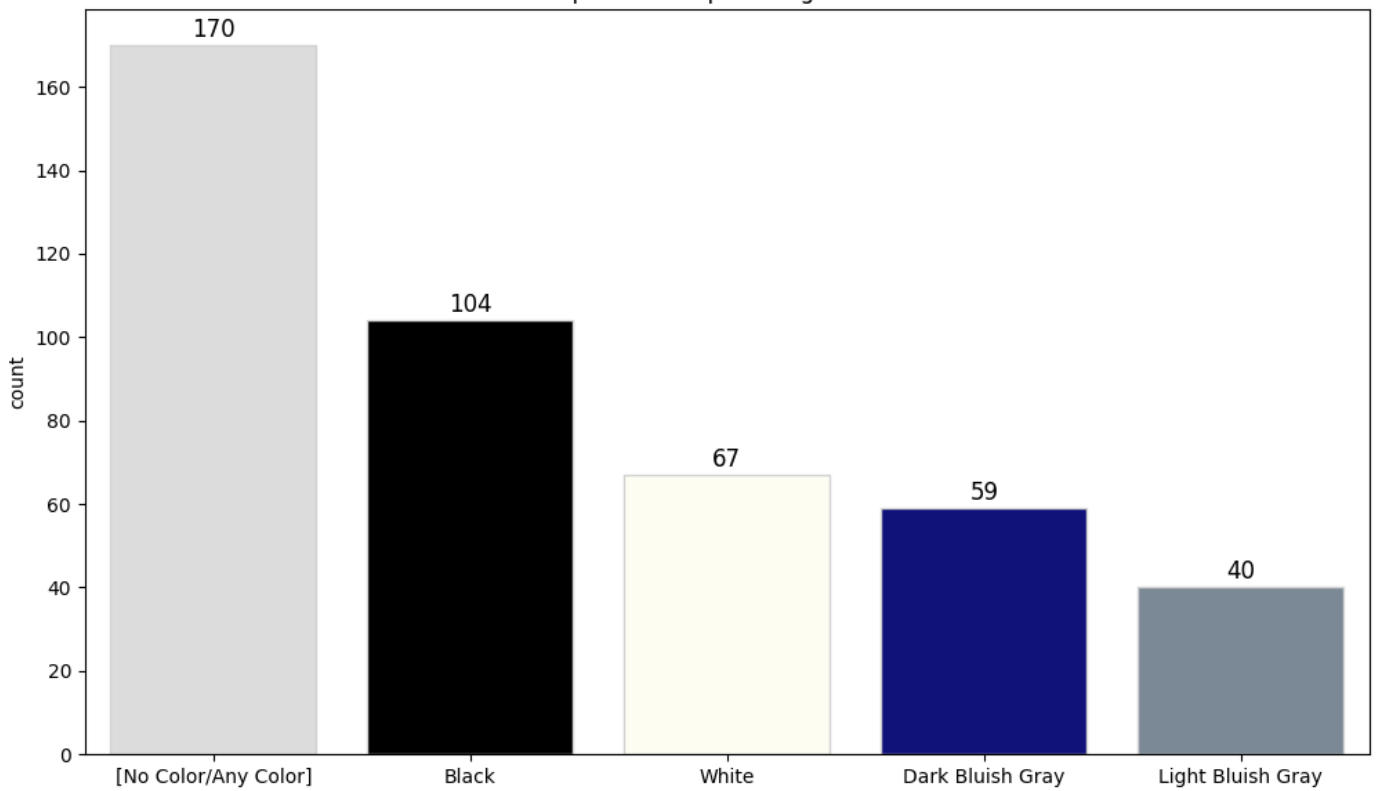
|    | name_colors          | count |
|----|----------------------|-------|
| 43 | [No Color/Any Color] | 170   |
| 0  | Black                | 104   |
| 41 | White                | 67    |
| 6  | Dark Bluish Gray     | 59    |
| 14 | Light Bluish Gray    | 40    |

```
In [27]: # create plot figure
plt.figure(figsize=(12,7))
plots = sns.barplot(pop_bricks, x='name_colors', y='count', palette=['gainsboro', 'black'])

# annotate bars
for bar in plots.patches:
    plots.annotate(format(bar.get_height(), '.0f'),
                    (bar.get_x() + bar.get_width() / 2,
                     bar.get_height()), ha='center', va='center',
                    size=12, xytext=(0, 8),
                    textcoords='offset points')

plt.title('Top 5 most Popular Lego Colors')
plt.xlabel('')
plt.show()
```

Top 5 most Popular Lego Colors



## What proportion of Lego is transparent?

```
In [28]: # value counts of is_trans
df.is_trans.value_counts()
```

```
Out[28]: f    703
         t     22
         Name: is_trans, dtype: int64
```

```
In [29]: # create dataframe with 'is_trans' value counts
trans_lego = df.groupby(['is_trans'], as_index=False)['is_trans'].value_counts()
```

```
In [30]: # add percent column
trans_lego['percent'] = (trans_lego['count']/sum(trans_lego['count'])*100).round(1)
```

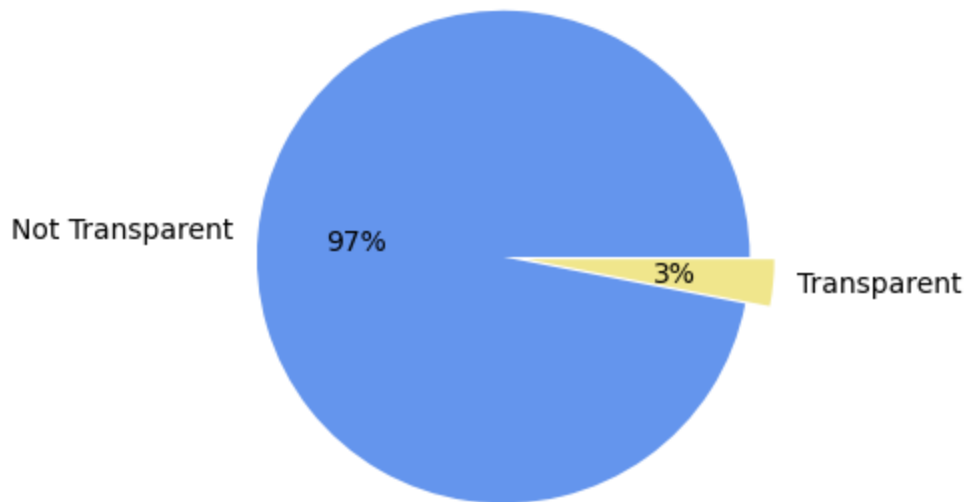
```
In [31]: # rename values
trans_lego['is_trans'] = trans_lego.is_trans.replace(['f', 't'], ['Not Transparent', 'Transparent'])
trans_lego
```

```
Out[31]:
```

|   | is_trans        | count | percent |
|---|-----------------|-------|---------|
| 0 | Not Transparent | 703   | 97.0    |
| 1 | Transparent     | 22    | 3.0     |

```
In [32]: # create pie plot of transparent proportions
plt.figure(figsize=(4,4))
plt.pie(x= trans_lego['count'], labels=trans_lego['is_trans'],
        autopct='%0.0f%%', explode=[0,0.1], colors=['cornflowerblue', 'khaki'])
plt.title('Ratio of Transparent Lego Bricks')
plt.show()
```

## Ratio of Transparent Lego Bricks



- The proportion of transparent Lego is 3%

## What are the 5 Rarest Lego Bricks

### The 5 rarest Lego Bricks by Lego Parts

```
In [33]: # create dataframe of name_parts of Lego bricks count
brick_names = df.groupby(['name_parts'], as_index=False)['name_parts'].value_counts()

# add percent column
brick_names['percent'] = (brick_names['count']/sum(brick_names['count'])*100).round(2)

# select the lowest 5 values
brick_names.sort_values(by='percent', ascending=True)[:5]
```

```
Out[33]:
```

|     | name_parts                                   | count | percent |
|-----|--|-------|---------|
| 0   | Activity Book, EV3 Space Challenge           | 1     | 0.14    |
| 200 | Slope 10° 6 x 8                              | 1     | 0.14    |
| 201 | Slope 18° 4 x 1                              | 1     | 0.14    |
| 202 | Slope 18° 4 x 2                              | 1     | 0.14    |
| 205 | Slope 45° 2 x 1 Triple with Ovoid Bottom Pin | 1     | 0.14    |

### The 5 rarest Lego Bricks by Theme Name

```
In [54]: rare_theme_bricks = df.groupby(['name_themes'], as_index=False)['name_themes'].value_cou
rare_theme_bricks
```

```
Out[54]:
```

|    | name_themes | count |
|----|-------------|-------|
| 27 | Ninja       | 1     |
| 53 | Western     | 1     |

|    |                             |   |
|----|-----------------------------|---|
| 50 | Vehicle                     | 1 |
| 49 | UFO                         | 1 |
| 36 | Role Play Toys and Costumes | 1 |

## The 5 rarest Lego Bricks by Color

```
In [34]: # select the 5 lowest count of Lego brick color
rarest_bricks = bricks.nsmallest(5, ['count'])
rarest_bricks
```

```
Out[34]:
```

|    | name_colors         | count |
|----|---------------------|-------|
| 3  | Brown               | 1     |
| 7  | Dark Brown          | 1     |
| 12 | Glow In Dark Opaque | 1     |
| 18 | Metallic Gold       | 1     |
| 21 | Pearl Blue          | 1     |

## Summary of Analysis

### Average Total of Lego Sets released by Year

- There are a total of 167 unique set\_num and 166 unique name\_sets, which suggests that there is a common name\_set shared between 2 set\_num
- The average total of sets released between 1970 and 2020 is 4 sets per year.
- 2003 marked the highest sets released with a 76 count of set\_num 8580-1, followed by year 1997 with a 63 count of set\_num 9847-1.
- The set\_num released has remained below count of 10 since 2012 until 2020

### Average Total of Lego Parts released by Year

- The average number of Lego parts released per year is 2 parts.
- 2003 recorded the highest amount of Lego parts released of 12 parts for part\_num 43559
- The next highest parts released occurred in 2003, counting 8 parts for part\_num 74747, 70931, 6035 and 5306c01
- The trendline indicates a volatile release of parts throughout 2005 and 2020, consistent to remaining under 7 parts.
- The lowest parts released occurred in 1980

### Most Popular Lego Parts by Color

- The most popular Lego parts by color in 'Black' at 10 counts are for parts; Motor, Train Base, 9V [Complete Assembly], and Train Motor Decorative Side with Closed Hub Points.
- Followed by 'Dark Gray' for both Vehicle Track, Train, 9V Straight and Curved, at 8 counts.
- 'Chrome Silver' for the part Dish 2 x 2 x 2/3 Light Reflector at 8 counts.

### Proportion of Transparent Lego

- The proportion of transparent Lego is 3%, indicating non-transparent are far more popular

### Rarest Lego Bricks

- The rarest Lego bricks by parts are Activity Book, EV3 Space Challenge, and the Slope ranges Slope 10° 6 x 8, Slope 18° 4 x 1, Slope 18° 4 x 2 and Slope 45° 2 x 1 Triple with Ovoid Bottom Pin
- The rarest Lego bricks by theme name are Ninja, Western, Vehicle, UFO and Role Play Toys and Costumes, at 1 count.
- The rarest Lego bricks by color are Brown, Dark Brown, Glow in the Dark Opaque, Metallic Gold and Pearl Blue.

In [ ]: