

This research report shows the methodology and conclusions of finding some prevailing attributes of UFO (undefined flying object) by analyzing the words frequencies from 134k descriptions of UFO eyewitnesses.

What are the prevailing attributes of UFO

Project 1 Final Written Report

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Topic Introduction

In 1977, Peter Sturrock, a professor of space science and astrophysics at Stanford University, did a research and found that 4.6% of members of the American Astronomical Society claimed that they have witnessed inexplicable aerial phenomena. This rate is close to the fact that approximately 5 percent of UFO (undefined flying object) sightings that are never explained (Knuth 2018). The topic of UFOs attracts people from all over the world. Investigations and studies for UFO have been conducted by various governments worldwide nowadays, along with private individuals and organizations (Daniela, G et al. 2021).

Motivation

Few months ago, the US government released a report about UAP (unidentified aerial phenomena). It is based on the review of 144 significant UAP reports involving observations made by military aviators between 2004 and 2021. However, the task force could only give a concrete explanation for one phenomenon, and the rest remain unexplained (Stieb & Danner 2021). That report raised the awareness of people who are concerned with UFO events. Hence, doing a research about identifying some prevailing attributes of UFO is necessary to help people know more about this unfamiliar field. This report will show the research methodology and conclusions in detail.

Dataset

The dataset utilized for this research is based on web scraped data from USA National UFO Reporting Center (Kaggle 2021). It contains 7 unique columns with 134,421 records of UFO sightings from 1946 to 2021 in the USA and was updated in July 2021. This dataset includes information such as time, locations, and descriptions from eyewitnesses about UFO sightings.



Methodology

This research mainly analyses the words frequencies of information contained in the dataset and draws conclusions by categorizing words to identify the potential patterns of UFOs. The research has explored the shape, color, USA state positions, audio, and general positions patterns of UFO. In addition, appearance year, month, time, and duration patterns of UFO have also been investigated. All visualizations were implemented by Python. Three main methods were adopted to explore the main characteristics of UFO, which were 'All Words Frequencies Counting' code (Code 1), 'Target Words Frequencies Counting' code (Code 2) and Microsoft Excel respectively. Firstly, 'All Words Frequencies Counting' code is able to count the frequency for every word in all descriptions. 'Target Words Frequencies Counting' code can count the word frequency for several specific words. These two scripts will not count repeated words from one piece of description. Lastly, Microsoft Excel has powerful searching functions which can identify some patterns that cannot be done with Python.

To find the shape pattern of UFO, 'All Words Frequencies Counting' code was used to obtain the word frequency for all words in a specific column of the dataset containing the shape information of UFO. After getting the frequencies for words describing shapes, the outcome was visualized

(Figure 1). However, Figure 1 only shows 10 shapes of UFO. In order to show the overall shape pattern of UFO, word cloud visualization was utilized (Figure 2). It took a text file as its input which contained different numbers of shape words according to their frequencies and was generated by Code 15. This visualization is able to show the frequencies for all shape words by plotting them with various sizes. From the data, it can be concluded that the top 3 words describing the shapes of UFO are ‘circle’, ‘triangle’ and ‘fireball’, which totally account for 38.85% of all records. Taking into the consideration of people’s stereotype towards UFO, the ratio of circular shapes (circle, fireball, sphere, disk, oval, egg) should be calculated. It turned out to be 57.21% which indicates that people might be affected by their stereotype towards UFO.

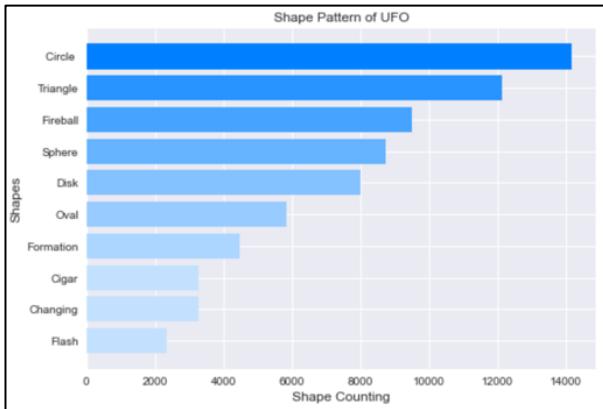


Figure 1
Generated by Code4



Figure 2
Generated by Code7

To find the color pattern of UFO, ‘All Words Frequencies Counting’ code was utilized to count the word frequency for all words from people’s descriptions. Then, color words as well as their words frequencies were recorded and visualized (Figure 3). To make Figure 3 more intuitive, waffle visualization was utilized (Figure 4). From the data, it can be concluded that the top 3 color words are ‘bright’, ‘white’ and ‘orange’, which totally account for 59.44% of all records. In addition, most colors are bright colors, which means that the vast majority of people might witness UFOs at night.

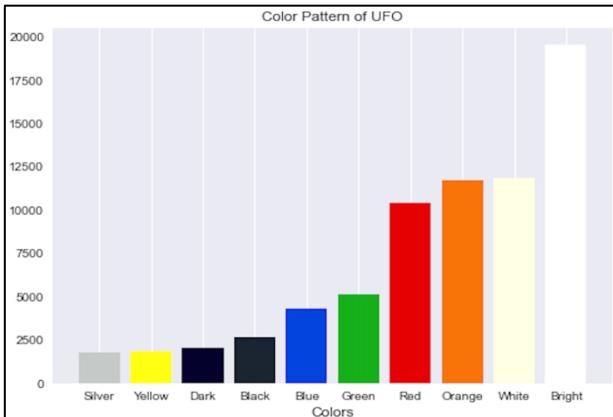


Figure 3
Generated by Code3

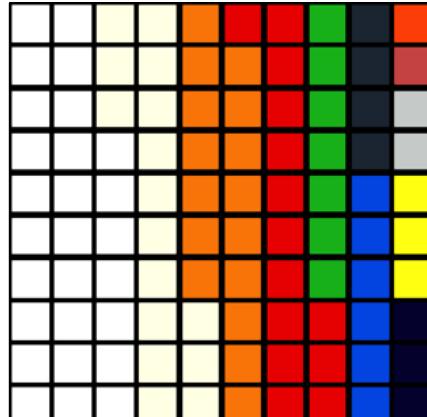


Figure 4
Generated by Code13

The method of identifying the USA state positions of UFO is identical to finding the shape pattern of UFO. Using ‘All Words Frequencies Counting’ code to acquire the word frequency for state names in a specific column of the dataset could produce a list of state names with the corresponding numbers of UFO witnessed in those states. After that, the outcome was visualized

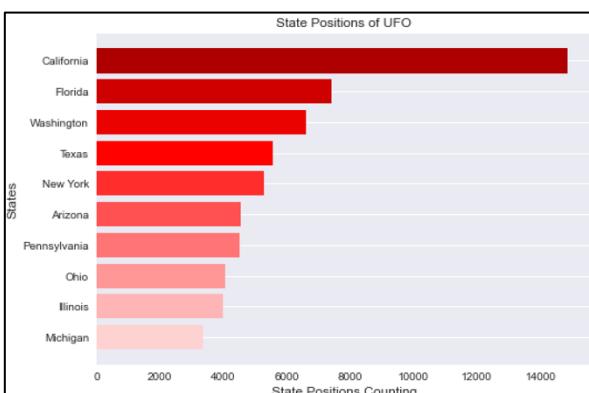


Figure 5
Generated by Code5

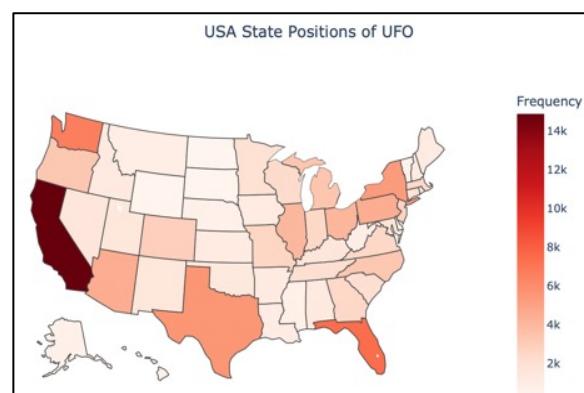


Figure 6
Generated by Code12

(Figure 5). In order to have an overview of the distribution of UFO events, map visualization was adopted (Figure 6). It shows the overall number of UFO events in each state that witnessed from 1946 to 2021. It can be concluded from the data that people are more likely to witness UFOs in California, Florida, and Washington, and the ratio of UFOs to be witnessed in California is 12.44%. In addition, there are 76.61% of UFOs that were witnessed in coastal states.

Finding the general positions of UFO needs both ‘All Words Frequencies Counting’ code and ‘Target Words Frequencies Counting’ code. The first script was used to find all words related to general positions, and the second script was utilized to find the word frequency for a group of similar words. It can be seen from Figure 7 that it is easier for people to see UFOs from home, on the mountain, and on the street. In addition, the percentage of witnessing UFOs at home is about 41.39% higher than on the mountain, and it accounts for approximately 19.74% of all records.

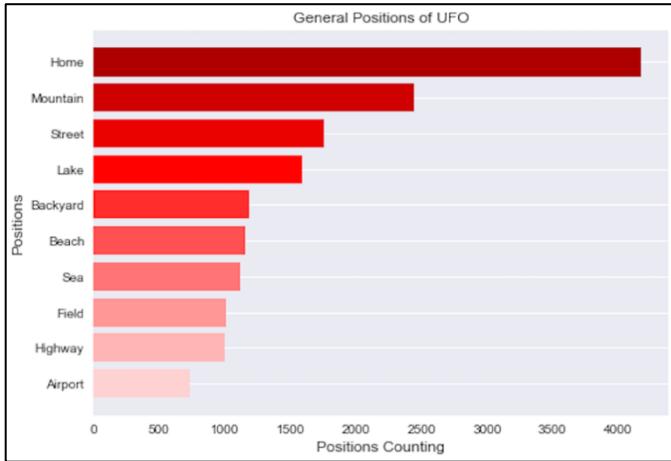


Figure 7
Generated by Code10

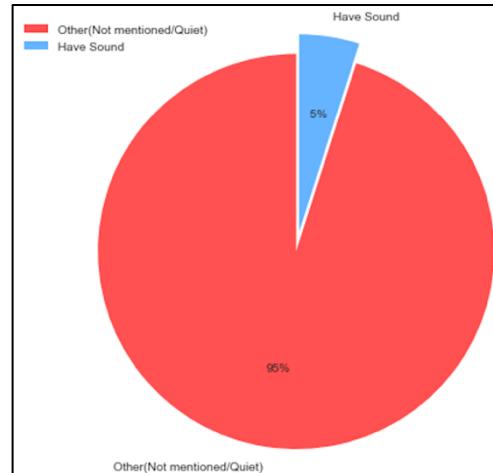


Figure 8
Generated by Code6

Finding the audio pattern of UFO requires ‘Target Words Frequencies Counting’ code which can search and count the number of descriptions containing particular words. For example, setting the target words as ‘noise’, ‘noisy’, ‘hear’ etc. (30~50 words related to hearing sound) can be applied to investigate the audio pattern of UFO. From figure 8, it can be concluded that approximately 5% of eyewitnesses claimed that they heard the sound from UFOs.

One advantage of the dataset is that all records have been sorted according to the time of UFO events. Hence, it is relatively easy to find the number of UFO events in a specific year or month by counting them in Microsoft Excel. It can be seen from Figure 9 that the number of UFO events has been on the rise since 1990 and reached its peak in 2014. After that, the number of sightings dropped sharply until 2018, and has been increasing thereafter. Figure 10 provides an intuitive overview of the number of UFO sightings in each month from 2009 to 2020. It can be concluded that people witnessed more UFOs than usual in July and August from 2012 to 2014 and March to April in 2020.

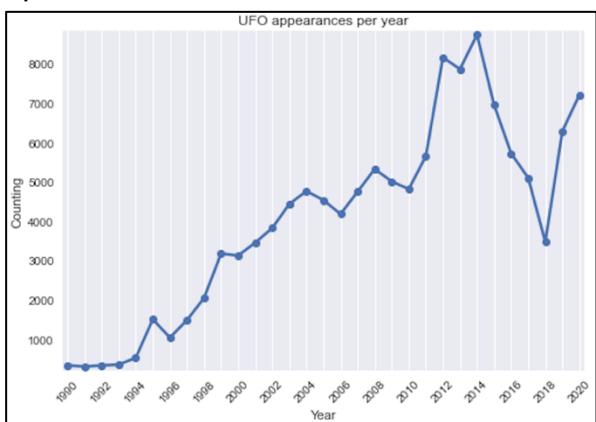


Figure 9
Generated by Code11

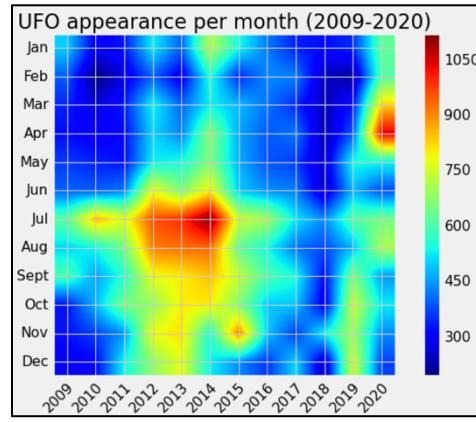


Figure 10
Generated by Code14

Finding the appearance time of UFO needs the ‘Text Filter’ from Microsoft Excel to avoid compatibility issues generated by Python. For example, to find the number of UFO sightings at 20:00 to 21:00, the content in the ‘Text Filter’ could be “contains ‘20:’ ”. It can be seen from Figure

11 that the number of UFO events increases sharply after 4 pm and it reaches a peak at 9 pm. After that, the number declines. This trend is consistent with previous speculation that the vast majority of people might witness UFOs at night.

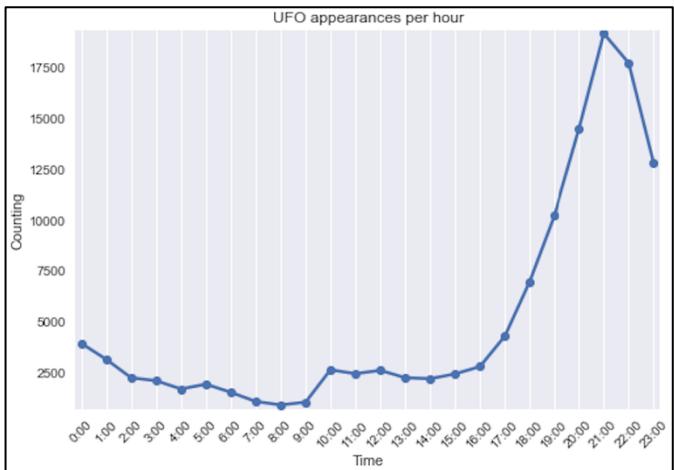


Figure 11
Generated by Code8

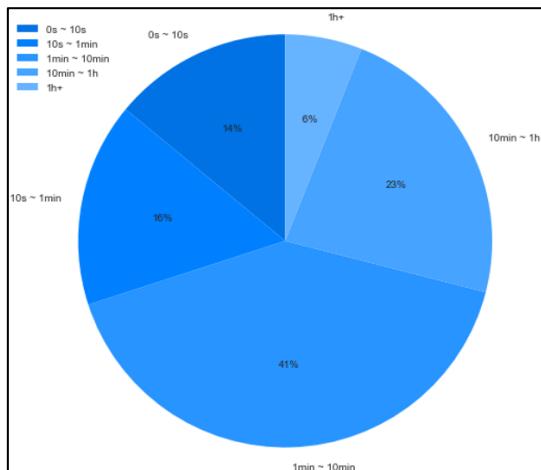


Figure 12
Generated by Code9

Considering the time constraint and actual situations, only data from 2020 were used to investigate the durations of UFO. Identifying this pattern requires Microsoft Excel combined with some manual work. Firstly, meaningless signs and English words were replaced by spaces. Then, empty records were deleted to make sure ‘Text Filter’ was able to work. Next, utilizing ‘Text Filter’ to filter various durations could generate outcomes. During the whole process, filtered data should be removed from the dataset after each round of searching to avoid repetition. It can be concluded from Figure 12 that there are 14% of UFOs will disappear within 10 seconds and 30% of UFOs will vanish within 1 minute. In addition, a vast majority of eyewitnesses (71%) can witness UFOs within 10 minutes.

Conclusion

In conclusion, this report provides people with an overview of the prevailing attributes of UFOs. Numbers of patterns have been identified and concluded from the descriptions of different eyewitnesses. However, the UFO phenomenon is still a mystery and needed to be studied continuously by human beings.

Summary of key findings

- Top 3 shape words describing UFOs are ‘Circle’, ‘Triangle’ and ‘Fireball’.
- Top 3 color words describing UFOs are ‘Bright’, ‘White’ and ‘Orange’.
- UFOs are most likely to be witnessed in ‘California’, ‘Florida’ and ‘Washington’.
- People are more likely to witness UFOs from home, on the mountain, and on the street.
- Approximately 5% of eyewitnesses heard the sound from UFOs.
- The number of UFO sightings has risen since 1990 and reached its peak in 2014.
- People witnessed more UFOs in July, August from 2012 to 2014 and March, April in 2020.
- The number of UFO sightings increases sharply after 4 pm and reaches a peak at 9 pm.
- A vast majority of eyewitnesses (71%) can witness UFOs within 10 minutes.

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Appendix:

```

import pandas as pd
import numpy as np

df=pd.read_csv("./Desktop/Project1/UFO_Dateset.csv")

counts={}
for i in range(134421):
    txt=df.Summary.iloc[i]
    for ch in '#%,.?!"-()':
        txt=txt.replace(ch, " ")
    txt.lower()
    words=txt.split()

    thisRound={}
    for word in words:
        thisRound[word]=1

    for word in words:
        if thisRound[word]==2:
            break;
        else:
            thisRound[word]=2
            if word not in counts:
                counts[word]=1
            else:
                counts[word]+=1

# Ranking
items=list(counts.items())
items.sort(key=lambda x:x[1],reverse=True)
print("\n");
for i in range(5000):
    word,count=items[i]
    print("{0:<15}{1:>10}".format(word,count))
print("\n");

in          42446
the         37263
a           33294
and          28976
lights        26242
of            25035
light         24715
sky           22892
object        19825

```

Code 1 All Words Frequencies Counting Code

```

import numpy as np
import matplotlib.pyplot as plt
x = np.array(["Silver","Yellow","Dark","Black","Blue","Green","Red","Orange","White","Bright"])
y = np.array([1736,1820,2068,2676,4303,5149,10420,11737,11862,19535])

sortIndex = np.argsort(y) # Rank ordering

x_sort = x[sortIndex] # Sorting
y_sort = y[sortIndex]

plt.xticks(np.arange(len(x_sort)), x_sort)
a = plt.bar(np.arange(len(x_sort)),y_sort,color=['#c5c9c7','#fffff14',
'#03012d','#1b2431', '#0343df', '#15b01a','#e50000','#f97306','#fffffe4','#ffffff'])

plt.title('Color Pattern of UFO')
plt.ylabel('Color Counting', fontsize=12)
plt.xlabel('Colors', fontsize=12)
plt.style.use('seaborn')
plt.grid(axis="y",ls='--')
plt.show()

```

Code 3

Bar plot for Color Pattern

```

import numpy as np
import matplotlib.pyplot as plt
y = np.array(["Flash","Changing","Cigar","Formation","Oval",
"Disk","Sphere","Fireball","Triangle","Circle"])
x = np.array([2356,3262,3264,4479,5842,8002,8754,9516,12132,14163])

sortIndex = np.argsort(x) # Rank ordering

x_sort = x[sortIndex] # Sorting
y_sort = y[sortIndex]

plt.yticks(np.arange(len(y_sort)), y_sort)
a = plt.barrh(np.arange(len(y_sort)),x_sort,color=['#0080FF','#2894FF','#46A3FF',
'#66B3FF', '#84C1FF', '#97CBFF', '#ACD6FF', '#C4E1FF', '#D2E9FF', '#ECF5FF'])

plt.plot(kind='barrh')
plt.title('Shape Pattern of UFO')
plt.xlabel('Shape Counting', fontsize=12)
plt.ylabel('Shapes', fontsize=12)
plt.style.use('seaborn')
plt.grid(axis="x",ls='--')
plt.show()

```

Code 4

Bar plot for Shape Pattern

```

import numpy as np
import matplotlib.pyplot as plt
y = np.array(["Michigan","Illinois","Ohio","Pennsylvania","Arizona",
"New York","Texas","Washington","Florida","California"])
x = np.array([3347,3979,4050,4500,4539,5286,5555,6617,7398,14863])

sortIndex = np.argsort(x) # Rank

x_sort = x[sortIndex] # Sorting
y_sort = y[sortIndex]

plt.yticks(np.arange(len(y_sort)), y_sort)
a = plt.barrh(np.arange(len(y_sort)),x_sort,color=['#FFD2D2','#FFB5B5',
'#FF9797', '#FF7575', '#FF5151','#FF2D2D', '#FF0000', '#EA0000', '#CE0000', '#AE0000'])

plt.plot(kind='barrh')
plt.title('State Positions of UFO')
plt.xlabel('State Positions Counting', fontsize=12)
plt.ylabel('States', fontsize=12)
plt.style.use('dark_background')
plt.grid(axis="x",ls='--')
plt.show()

```

Code 5

Bar plot for State Positions of UFO

```

import matplotlib.pyplot as plt
plt.figure(figsize=(7,7))
labels = [u'Other(Not mentioned/Quiet)', u'Have Sound']
sizes = [95.1,4.9]
colors = ['#FF5151', '#66B3FF']

explode = (0.1, 0)

patches, l_text, p_text = plt.pie(sizes, explode=explode, labels=labels, colors=colors,
                                labeldistance=1.1, autopct='%.2f%%', shadow=False,
                                startangle=90, pctdistance=0.6)

for t in l_text:
    t.set_size = 30
for t in p_text:
    t.set_size = 20
plt.axis('equal')
plt.legend(loc='upper left', bbox_to_anchor=(-0.1, 1))
plt.grid()
plt.style.use('seaborn')
plt.show()

```

Code 6

Pie chart for Audio Pattern

Code 2 Target Words Frequencies Counting Code

```

from wordcloud import WordCloud

with open("./Desktop/word.txt", encoding="utf-8") as file:
    # Read text
    text = file.read()

# Some settings
wordcloud = WordCloud(font_path="./Library/Fonts/Times.ttc",
                      background_color="black", width=600,
                      height=400, max_words=100, relative_scaling=0.3).generate(text)
# Generate Word Cloud Visualizations
image = wordcloud.to_image()
# Show this visualization
image.show()

```

Code 7

Word Cloud Visualization for Shape Pattern

```

import matplotlib.pyplot as plt
names = ['1998', '1992', '1994', '1996', '1998', '2000', '2002',
         '2004', '2006', '2008', '2010', '2012', '2014', '2016', '2018', '2020']
x = range(len(names))
y = [332, 303, 331, 351, 527, 1505, 1038, 1483, 2045, 3175, 3124,
      3452, 3826, 4435, 4764, 4534, 4179, 4749, 5323, 5009, 4814,
      5652, 8161, 7865, 8739, 6954, 5728, 5899, 3469, 6286, 7213]

plt.plot(x, y, linewidth=2.5, marker='o')
plt.legend()
plt.xticks(x, names, rotation=45)
plt.margins(0.01)

plt.subplots_adjust(bottom=0.15)
plt.xlabel("Year")
plt.ylabel("Counting")
plt.title("UFO appearances per year")
plt.style.use('seaborn')
plt.grid(axis="y", ls='--')
plt.show()

```

```

import matplotlib.pyplot as plt
names = ['0:00', '1:00', '2:00', '3:00', '4:00', '5:00', '6:00', '7:00', '8:00', '9:00',
         '10:00', '11:00', '12:00', '13:00', '14:00', '15:00', '16:00', '17:00', '18:00',
         '19:00', '20:00', '21:00', '22:00', '23:00']
x = range(len(names))
y = [3919, 3131, 2226, 2083, 1676, 1911, 1513, 1061, 892, 1025, 2618, 2429, 2590,
      2229, 2181, 2415, 2778, 4273, 6962, 10217, 14460, 19152, 17704, 12793]
plt.plot(x, y, linewidth=2.5, marker='o')
plt.legend()
plt.xticks(x, names, rotation=45)
plt.margins(0.01)

plt.subplots_adjust(bottom=0.15)
plt.xlabel("Time")
plt.ylabel("Counting")
plt.title("UFO appearances per hour")
plt.style.use('seaborn')
plt.grid(axis="y", ls='--')
plt.show()

```

Code 8

Line chart for Appearance time of UFO

```

import matplotlib.pyplot as plt
plt.figure(figsize=(8,8))
labels = [u'0s ~ 10s', u'10s ~ 1min', u'1min ~ 10min', u'10min ~ 1h', u'1h+']
sizes = [14, 16, 41, 23, 6]
colors = ['#0072E3', '#0080FF', '#2894FF', '#46A3FF', '#66B3FF']

explode = (0, 0, 0, 0, 0)

patches, l_text, p_text = plt.pie(sizes, explode=explode, labels=labels, colors=colors,
                                  labeldistance=1.1, autopct='%2.0f%%', shadow=False,
                                  startangle=90, pctdistance=0.6)

for t in l_text:
    t.set_size = 30
for t in p_text:
    t.set_size = 20
plt.axis('equal')
plt.legend(loc='upper left', bbox_to_anchor=(-0.1, 1))
plt.grid()
plt.style.use('seaborn')
plt.show()

```

Code 9

Pie chart for Durations of UFO

```

import numpy as np
import matplotlib.pyplot as plt
y = np.array(["Airport", "Highway", "Field", "Sea", "Beach", "Backyard", "Lake",
              "Street", "Mountain", "Home"])
x = np.array([738, 1002, 1014, 1115, 1154, 1187, 1592, 1761, 2448, 4177])

sortIndex = np.argsort(x) # Rank ordering
x_sort = x[sortIndex] # Sorting
y_sort = y[sortIndex]

plt.yticks(np.arange(len(y_sort)), y_sort)
a = plt.barh(np.arange(len(y_sort)), x_sort, color=['#FFD2D2', '#FFB5B5',
          '#FF9797', '#FF7575', '#FF5151', '#FF2D2D', '#FF0000', '#EA0000', '#CE0000', '#AE0000'])

plt.plot(kind='barh')
plt.title('General Positions of UFO')
plt.xlabel('Positions Counting', fontsize=12)
plt.ylabel('Positions', fontsize=12)
plt.style.use('seaborn')
plt.grid(axis="x", ls='--')
plt.show()

```

Code 10

Bar plot for General Positions of UFO

Code 11

Line chart for Appearance year of UFO

```

import plotly.graph_objects as go
import pandas as pd

df=pd.read_excel("./Desktop/Project1/Research outcomereal.xlsx")

fig = go.Figure(data=go.Choropleth(
    locations=df['visualization state'],
    z = df['num4'].astype(float),
    locationmode = 'USA-states',
    colorscale = 'Reds',
    colorbar_title = "Frequency",
))

fig.update_layout(
    title_text = 'USA State Positions of UFO',
    geo_scope='usa',
)

```

Code 12

Map Visualization for State Positions of UFO

```

import matplotlib.pyplot as plt
from pywaffle import Waffle

plt.figure(
    FigureClass=Waffle,
    rows=10,
    columns=10,
    values=[26.92, 16.35, 16.17, 14.36, 7.10, 5.93,
            3.69, 2.85, 2.51, 2.39, 1.06, 0.68],
    colors=( "#ffffff", "#ffffe4", "#f97306", "#e50000",
             "#15b01a", "#0343df", "#1b2431", "#03012d",
             "#fffff14", "#c5c9c7", "#c44240", "#fd3c06"),
)
plt.show()

```

Code 13

Waffle Visualization for Color Pattern

```

import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from pandas import Series, DataFrame
import seaborn as sns
import palettable
from sklearn import datasets

df=pd.read_csv("./Desktop/Project1/month.csv",index_col = ['month'])

df

months = ["Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul", "Aug", "Sept", "Oct", "Nov", "Dec"]
years = ["2009", "2010", "2011", "2012", "2013", "2014", "2015", "2016", "2017", "2018", "2019", "2020"]

data = np.array([[501, 296, 333, 531, 401, 720, 548, 424, 340, 320, 351, 627],
[398, 195, 278, 400, 286, 560, 347, 438, 428, 246, 222, 612],
[344, 267, 338, 540, 401, 531, 483, 417, 350, 240, 329, 828],
[320, 297, 319, 589, 431, 673, 465, 392, 420, 236, 394, 1045],
[364, 339, 324, 532, 546, 663, 473, 388, 367, 256, 549, 541],
[400, 394, 423, 780, 661, 796, 499, 440, 428, 250, 480, 379],
[620, 862, 768, 959, 994, 1122, 746, 703, 522, 441, 611, 647],
[507, 541, 648, 913, 929, 938, 631, 546, 418, 377, 481, 713],
[603, 458, 562, 784, 802, 849, 741, 601, 548, 316, 665, 451],
[329, 482, 649, 686, 884, 805, 667, 496, 495, 286, 746, 528],
[327, 372, 451, 786, 815, 552, 913, 499, 361, 541, 699, 442],
[296, 311, 541, 681, 777, 531, 441, 358, 512, 246, 759, 370]])

fig, ax = plt.subplots()
im = ax.imshow(data)

# We want to show all ticks...
ax.set_xticks(np.arange(len(years)))
ax.set_yticks(np.arange(len(months)))
# ... and label them with the respective list entries
ax.set_xticklabels(years)
ax.set_yticklabels(months)

# Rotate the tick labels and set their alignment.
plt.setp(ax.get_xticklabels(), rotation=45, ha="right",
         rotation_mode="anchor")

ax.set_title("UFO appearance per month (2009-2020)")
fig.tight_layout()
plt.colorbar(im)
plt.style.use('fivethirtyeight')
plt.show()

```

Code 14

Heat Map for Appearance Month of UFO

```

import random
import pandas as pd
import numpy as np

file_handle=open("./Desktop/word"

index=0
words=[0 for x in range(0, 902)]
for i in range(5):
    words[index]="cross"
    index=index+1
for i in range(6):
    words[index]="cone"
    index=index+1
for i in range(13):
    words[index]="teardrop"
    index=index+1
for i in range(13):
    words[index]="egg"
    index=index+1
for i in range(17):
    words[index]="chevron"
    index=index+1
for i in range(21):
    words[index]="diamond"
    index=index+1
for i in range(23):
    words[index]="cylinder"
    index=index+1
for i in range(25):
    words[index]="rectangle"
    index=index+1
for i in range(26):
    words[index]="flash"
    index=index+1
for i in range(35):
    words[index]="changing"
    index=index+1
for i in range(35):
    words[index]="cigar"
    index=index+1

for i in range(26):
    words[index]="flash"
    index=index+1
for i in range(35):
    words[index]="changing"
    index=index+1
for i in range(35):
    words[index]="cigar"
    index=index+1

for i in range(10000):
    random.shuffle(words)

for i in words:
    file_handle.write(i)
    file_handle.write(" ")

file_handle.close()

```

Code 15

Generating Input for Word Cloud Visualization

le changing circle circle s teardrop triangle disk cyl ircle fireball rectangle fo circle disk circle formati k triangle cigar triangle f e teardrop circle triangle disk disk chevron circle ci formation triangle circle rectangle oval sphere circ k diamond chevron formation cle fireball circle cigar c mation circle rectangle cir angle sphere fireball rect ing oval circle cigar circl e triangle disk changing iangle fireball fireball fi hanging sphere triangle fla riangle sphere fireball dia riangle circle cylinder egg disk sphere sphere changing al disk triangle triangle d e oval changing egg cone tr ectangle circle sphere circ triangle formation sphere cigar formation formation s isk fireball triangle recta rdrop disk triangle oval ov ircle rectangle cigar fireb cross triangle diamond tri ireball sphere circle trian riangle oval egg fireball f ngle chevron fireball fireb ircle sphere changing circl riangle cigar triangle fire n changing diamond sphere s cle oval fireball circle di aking disk circle changing triangle cross circle tria egg flash circle triangle c diamond circle sphere disk cylinder oval disk diamond rcle circle triangle disk c hanging oval flash cylinder isk triangle circle triangl ctangle fireball disk cigar ormation sphere circle disk hanging oval sphere circle e triangle oval cigar disk ond triangle triangle circl formation disk sphere tria riangle sphere fireball sph all formation triangle flas oval cigar fireball disk ch riangle changing sphere for mation fireball triangle flash disk circle fireball orformation fireball sphere fi le flash triangle teardrop c tiation triangle triangle circ ball circle cylinder disk t isk triangle rectangle cros ireball sphere circle cylin l fireball fireball triangl eball disk cylinder triangl triangle triangle circle f circle fireball circle dis e triangle rectangle sphere fireball disk oval triangle triangle oval teardrop cir cigar circle sphere circle f

Figure 13

Text File Generated by Code 15

Excel Worksheet

| A | B | C | D | E | F | G | H |
|-----------|----------------|---------------------------------------|-------|-------------------------|------|-------------------------------------|--------|
| 1 Colors | Color Counting | Shape Pattern of UFO | num2 | State Positions Pattern | num3 | Audio Pattern of UFO | |
| 2 silver | | 1736 cross | 435 | NJ | | 2621 Has sound 4.87% | 6553 |
| 3 yellow | | 1820 cone | 537 | CO | | 2905 Not mentiond / no sound 95.13% | 127826 |
| 4 dark | | 2068 teardrop | 1163 | OR | | 3264 | |
| 5 black | | 2676 egg | 1171 | NC | | 3285 | |
| 6 blue | | 4303 chevron | 1537 | MI | | 3347 Michigan | |
| 7 green | | 5149 diamond | 1948 | IL | | 3979 Illinois | |
| 8 red | | 10420 cylinder | 2091 | OH | | 4050 Ohio | |
| 9 orange | | 11737 rectangle | 2284 | PA | | 4500 Pennsylvania | |
| 10 white | | 11862 flash | 2356 | AZ | | 4539 Arizona | |
| 11 bright | | 19535 changing cigar 0 formation oval | 3262 | NY | | 5286 New York | |
| 12 | | | 3264 | TX | | 5555 Texas | |
| 13 | | | 4479 | WA | | 6617 Washington | |
| 14 | | | 5842 | FL | | 7398 Florida | |
| 15 | | disk | 8002 | CA | | 14863 California | |
| 16 | | sphere | 8754 | | | | |
| 17 | | fireball | 9516 | | | | |
| 18 | | triangle | 12132 | | | | |
| 19 | | circle | 14163 | | | | |

| J | K | L | M | N | O | P | Q |
|-----------------------|-------|---|------|------|---|--------|--------|
| 1 visualization state | num4 | | year | num5 | | during | num6 |
| 2 AK | 607 | | 1990 | 332 | | 0:00 | 3919 |
| 3 AL | 1235 | | 1991 | 303 | | 1:00 | 3131 |
| 4 AR | 1136 | | 1992 | 331 | | 2:00 | 2226 |
| 5 AZ | 4539 | | 1993 | 351 | | 3:00 | 2083 |
| 6 CA | 14863 | | 1994 | 527 | | 4:00 | 1676 |
| 7 CO | 2905 | | 1995 | 1505 | | 5:00 | 1911 |
| 8 CT | 1812 | | 1996 | 1038 | | 6:00 | 1513 |
| 9 DE | 361 | | 1997 | 1483 | | 7:00 | 1061 |
| 10 FL | 7398 | | 1998 | 2045 | | 8:00 | 892 |
| 11 GA | 2469 | | 1999 | 3175 | | 9:00 | 1025 |
| 12 HI | 630 | | 2000 | 3124 | | 10:00 | 2618 |
| 13 IA | 1126 | | 2001 | 3452 | | 11:00 | 2429 |
| 14 ID | 1226 | | 2002 | 3826 | | 12:00 | 2590 |
| 15 IL | 3979 | | 2003 | 4435 | | 13:00 | 2229 |
| 16 IN | 2451 | | 2004 | 4764 | | 14:00 | 2181 |
| 17 KS | 1087 | | 2005 | 4534 | | 15:00 | 2415 |
| 18 KY | 1538 | | 2006 | 4179 | | 16:00 | 2778 |
| 19 LA | 1016 | | 2007 | 4749 | | 17:00 | 4273 |
| 20 MA | 2400 | | 2008 | 5323 | | 18:00 | 6962 |
| 21 MD | 1655 | | 2009 | 5009 | | 19:00 | 10217 |
| 22 ME | 1087 | | 2010 | 4814 | | 20:00 | 14460 |
| 23 MI | 3347 | | 2011 | 5652 | | 21:00 | 19152 |
| 24 MN | 1935 | | 2012 | 8161 | | 22:00 | 17704 |
| 25 MO | 2568 | | 2013 | 7865 | | 23:00 | 12793 |
| 26 MS | 729 | | 2014 | 8739 | | | 122238 |
| 27 MT | 937 | | 2015 | 6954 | | | |
| 28 NC | 3285 | | 2016 | 5720 | | | |
| 29 ND | 259 | | 2017 | 5099 | | | |
| 30 NE | 647 | | 2018 | 3469 | | | |
| 31 NH | 1073 | | 2019 | 6286 | | | |
| 32 NJ | 2621 | | 2020 | 7213 | | | |
| 33 NM | 1511 | | | | | | |
| 34 UT | 1389 | | | | | | |
| 35 NV | 1567 | | | | | | |
| 36 NY | 5286 | | | | | | |
| 37 OH | 4050 | | | | | | |
| 38 OK | 1340 | | | | | | |
| 39 OR | 3264 | | | | | | |
| 40 PA | 4500 | | | | | | |
| 41 RI | 556 | | | | | | |
| 42 SC | 2060 | | | | | | |
| 43 SD | 351 | | | | | | |
| 44 TN | 2068 | | | | | | |
| 45 TX | 5555 | | | | | | |
| 46 VA | 2444 | | | | | | |
| 47 VT | 549 | | | | | | |
| 48 WA | 6617 | | | | | | |
| 49 WI | 2261 | | | | | | |
| 50 WV | 828 | | | | | | |
| 51 WY | 379 | | | | | | |

Note: Yellow entries represent coastal states in the USA

| AB | AC | AD | AE | AF | AG |
|-----------|-------|-------------|-------------|-----|-----------|
| cross | 435 | 0.004719745 | 4.719744808 | 5 | cross |
| cone | 537 | 0.005826444 | 5.826443591 | 6 | cone |
| teardrop | 1163 | 0.012618536 | 12.61853612 | 13 | teardrop |
| egg | 1171 | 0.012705336 | 12.70533602 | 13 | egg |
| chevron | 1537 | 0.016676432 | 16.67643166 | 17 | chevron |
| diamond | 1948 | 0.021135777 | 21.13577765 | 21 | diamond |
| cylinder | 2091 | 0.022687325 | 22.68732504 | 23 | cylinder |
| rectangle | 2284 | 0.024781373 | 24.78137274 | 25 | rectangle |
| flash | 2356 | 0.025562572 | 25.56257188 | 26 | flash |
| changing | 3262 | 0.035392661 | 35.9266107 | 35 | changing |
| cigar | 3264 | 0.035414361 | 35.41436104 | 35 | cigar |
| formation | 4479 | 0.048597097 | 48.59709654 | 49 | formation |
| oval | 5842 | 0.06338563 | 63.38563028 | 63 | oval |
| disk | 8002 | 0.086821604 | 86.8216045 | 87 | disk |
| sphere | 8754 | 0.094980796 | 94.98079552 | 95 | sphere |
| other | 9230 | 0.10014539 | 100.1453898 | 100 | other |
| fireball | 9516 | 0.103248486 | 103.2484864 | 103 | fireball |
| triangle | 12132 | 0.131632055 | 131.6320552 | 132 | triangle |
| circle | 14163 | 0.153668381 | 153.668381 | 154 | circle |

Note: Numbers in AF column represent the relative numbers for shape words.

| AQ | AR | AS | AT | AU |
|------------|-------|-------------|-------------|----------|
| red/orange | 492 | 0.006780128 | 0.678012816 | #fd3c06 |
| reddish | 767 | 0.010569834 | 1.056983394 | #c44240 |
| silver | 1736 | 0.023923379 | 2.392337904 | #c5c9c7 |
| yellow | 1820 | 0.025080962 | 2.50809619 | #fffff14 |
| dark | 2068 | 0.028498587 | 2.849858747 | #03012d |
| black | 2676 | 0.036877282 | 3.687728244 | #1b2431 |
| blue | 4303 | 0.05929856 | 5.929855991 | #0343df |
| green | 5149 | 0.070957073 | 7.095707297 | #15b01a |
| red | 10420 | 0.143595397 | 14.35953972 | #e50000 |
| orange | 11737 | 0.161744643 | 16.17446427 | #9f7306 |
| white | 11862 | 0.163467236 | 16.34672363 | #c6fcff |
| bright | 19535 | 0.269206918 | 26.92069179 | #fffff1 |

Note: Numbers in AT column represent the ratio of colors

| AX | AY | AZ |
|-----------|------|-------------|
| 10s] | 789 | 0.136198861 |
| 10s-1min] | 941 | 0.162437424 |
| 1min-10mi | 2373 | 0.409632315 |
| 10min-1h] | 1357 | 0.234248231 |
| 1h+ | 333 | 0.057483169 |
| sum | 5793 | |

Note: Durations of UFO (2020)

| BD | BE | BF |
|---------------------------------------|------|------------|
| place | num | |
| home / house / bedroom / kitchen | 4177 | Home |
| valley / foothills / mountains / hill | 2448 | Mountain |
| crossing / road /street / parkway | 1761 | Street |
| lake | 1592 | Lake |
| backyard / yard / garden | 1187 | Backyards |
| beach | 1154 | Beach |
| ocean / sea / field / farm | 1115 | Sea |
| highway / freeway | 1002 | Highway |
| airport | 738 | Airport |
| coast / gulf / shore | 710 | Park |
| island | 548 | Island |
| woods / forest / grove | 518 | Woods |
| school / college university | 475 | University |
| porch | 390 | Porch |
| station | 286 | Station |
| deck | 268 | Deck |
| desert | 242 | Desert |
| port | 156 | Port |
| creek / stream | 261 | Creek |
| canyon | 118 | Canyon |
| patio | 104 | Patio |
| pool | 96 | Pool |
| camp | 97 | Camp |

| BA | BB |
|--------|-----|
| 2008 1 | 477 |
| 2 | 373 |
| 3 | 343 |
| 4 | 445 |
| 5 | 349 |
| 6 | 494 |
| 7 | 581 |
| 8 | 514 |
| 9 | 391 |
| 10 | 540 |
| 11 | 463 |
| 12 | 352 |

| 2009 1 | 501 |
|--------|-----|
| 2 | 398 |
| 3 | 344 |
| 4 | 320 |
| 5 | 364 |
| 6 | 400 |
| 7 | 620 |
| 8 | 507 |
| 9 | 603 |
| 10 | 329 |
| 11 | 327 |
| 12 | 296 |

| 2010 1 | 296 |
|--------|-----|
| 2 | 195 |
| 3 | 267 |
| 4 | 297 |
| 5 | 339 |
| 6 | 394 |
| 7 | 862 |
| 8 | 541 |
| 9 | 458 |
| 10 | 482 |
| 11 | 372 |
| 12 | 311 |

| 2011 1 | 333 |
|--------|-----|
| 2 | 278 |
| 3 | 338 |
| 4 | 319 |
| 5 | 324 |
| 6 | 423 |
| 7 | 768 |
| 8 | 648 |
| 9 | 562 |
| 10 | 649 |
| 11 | 451 |
| 12 | 541 |

| 2012 1 | 531 |
|--------|-----|
| 2 | 400 |
| 3 | 540 |
| 4 | 509 |
| 5 | 532 |
| 6 | 780 |
| 7 | 959 |
| 8 | 913 |
| 9 | 784 |
| 10 | 686 |
| 11 | 786 |
| 12 | 681 |

| 2013 1 | 401 |
|--------|-----|
| 2 | 286 |
| 3 | 401 |
| 4 | 431 |
| 5 | 546 |
| 6 | 661 |
| 7 | 994 |
| 8 | 929 |
| 9 | 802 |
| 10 | 804 |
| 11 | 815 |
| 12 | 777 |

| 2014 1 | 720 |
|--------|------|
| 2 | 560 |
| 3 | 531 |
| 4 | 673 |
| 5 | 663 |
| 6 | 796 |
| 7 | 1122 |
| 8 | 938 |
| 9 | 849 |
| 10 | 805 |
| 11 | 552 |
| 12 | 531 |

| 2015 1 | 548 |
|--------|-----|
| 2 | 347 |
| 3 | 483 |
| 4 | 465 |
| 5 | 473 |
| 6 | 499 |
| 7 | 746 |
| 8 | 631 |
| 9 | 741 |
| 10 | 667 |
| 11 | 913 |
| 12 | 441 |

| 2016 1 | 424 |
|--------|-----|
| 2 | 438 |
| 3 | 417 |
| 4 | 392 |
| 5 | 388 |
| 6 | 440 |
| 7 | 703 |
| 8 | 546 |
| 9 | 601 |
| 10 | 496 |
| 11 | 499 |
| 12 | 358 |

| 2017 1 | 340 |
|--------|-----|
| 2 | 428 |
| 3 | 350 |
| 4 | 420 |
| 5 | 367 |
| 6 | 428 |
| 7 | 522 |
| 8 | 418 |
| 9 | 548 |
| 10 | 495 |
| 11 | 361 |
| 12 | 512 |

| 2018 1 | 320 |
|--------|-----|
| 2 | 246 |
| 3 | 240 |
| 4 | 236 |
| 5 | 256 |
| 6 | 250 |
| 7 | 441 |
| 8 | 377 |
| 9 | 316 |
| 10 | 286 |
| 11 | 541 |
| 12 | 246 |

| 2019 1 | 351 |
|--------|-----|
| 2 | 222 |
| 3 | 329 |
| 4 | 394 |
| 5 | 549 |
| 6 | 480 |
| 7 | 611 |
| 8 | 481 |
| 9 | 665 |
| 10 | 746 |
| 11 | 699 |
| 12 | 759 |

| 2020 1 | 627 |
|--------|------|
| 2 | 612 |
| 3 | 828 |
| 4 | 1045 |
| 5 | 541 |
| 6 | 379 |
| 7 | 647 |
| 8 | 713 |
| 9 | 451 |
| 10 | 528 |
| 11 | 442 |
| 12 | 370 |

Data for appearance month of UFO