

# APIAnalysis

November 10, 2024

## 1 Assignment 6: API Analysis

This notebook is for analyzing the data you have gathered from your API. See Moodle for the requirements. This notebook must assume that your API Gather notebook has saved data gathered into an appropriate format to be read into this notebook for analysis. In almost all cases, this notebook should use pandas. If you think you have data that cannot be analyzed in that way, please get in touch with the instructor well before the deadline to discuss if another approach is acceptable.

You are responsible for creating all markdown and code cells as needed in this assignment so that your purpose, analysis, and interpretation of results are clear.

### 1.1 Import modules for data manipulation and analysis

```
[12]: import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns
```

### 1.2 Read saved csv file (st\_species.csv) into a pandas DataFrame

```
[13]: file_path = "/home/mkmueler/cs703f24/Assignment6/st_species.csv"  
  
df = pd.read_csv(file_path)  
df.head()  
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 759 entries, 0 to 758  
Data columns (total 16 columns):  
 #   Column           Non-Null Count  Dtype     
 ---  --  
 0   uid              759 non-null    object    
 1   name             759 non-null    object    
 2   homeworld        286 non-null    object    
 3   quadrant          288 non-null    object    
 4   extinctSpecies   759 non-null    bool      
 5   warpCapableSpecies 759 non-null    bool      
 6   extraGalacticSpecies 759 non-null    bool
```

```

7  humanoidSpecies          759 non-null    bool
8  reptilianSpecies         759 non-null    bool
9  nonCorporealSpecies     759 non-null    bool
10 shapeshiftingSpecies    759 non-null    bool
11 spaceborneSpecies        759 non-null    bool
12 telepathicSpecies        759 non-null    bool
13 transDimensionalSpecies  759 non-null    bool
14 unnamedSpecies           759 non-null    bool
15 alternateReality          759 non-null    bool
dtypes: bool(12), object(4)
memory usage: 32.7+ KB

```

### 1.3 Analysis of the different classes of species

- Make a new dataframe containing only the classes of species
- Typecast the boolean values to integers
- Sum up the boolean columns

```
[14]: cols = ["extinctSpecies", "warpCapableSpecies", "extraGalacticSpecies", ↴
    ↴ "humanoidSpecies", "reptilianSpecies", "nonCorporealSpecies", ↴
    ↴ "shapeshiftingSpecies", "spaceborneSpecies", "telepathicSpecies", ↴
    ↴ "transDimensionalSpecies", "unnamedSpecies"]
df_sp_classes = df[cols]
#typecast boolean as int
df_sp_classes = df_sp_classes.astype(int)
print(df_sp_classes.head(11))
# summing up the boolean columns
df_sp_classes_sum = df_sp_classes.sum()
print(df_sp_classes_sum)
```

	extinctSpecies	warpCapableSpecies	extraGalacticSpecies	humanoidSpecies	\
0	0	0	0	1	
1	0	0	0	1	
2	0	0	0	1	
3	0	0	0	1	
4	0	0	0	1	
5	0	0	0	0	
6	0	1	0	1	
7	0	0	0	1	
8	0	0	0	1	
9	0	0	0	0	
10	0	0	0	1	

	reptilianSpecies	nonCorporealSpecies	shapeshiftingSpecies	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	

```

4          0          0          0
5          0          0          0
6          0          0          0
7          0          0          0
8          0          0          0
9          0          0          0
10         0          0          0

    spaceborneSpecies  telepathicSpecies  transDimensionalSpecies  \
0                  0                  0                  0
1                  0                  0                  0
2                  0                  0                  0
3                  0                  0                  0
4                  0                  0                  0
5                  0                  1                  0
6                  0                  0                  0
7                  0                  0                  0
8                  0                  1                  0
9                  0                  0                  0
10                 0                  0                  0

    unnamedSpecies
0                  1
1                  0
2                  0
3                  1
4                  1
5                  0
6                  0
7                  1
8                  0
9                  1
10                 0

extinctSpecies      3
warpCapableSpecies 100
extraGalacticSpecies 9
humanoidSpecies     490
reptilianSpecies    8
nonCorporealSpecies 31
shapeshiftingSpecies 23
spaceborneSpecies   15
telepathicSpecies   38
transDimensionalSpecies 0
unnamedSpecies      147
dtype: int64

```

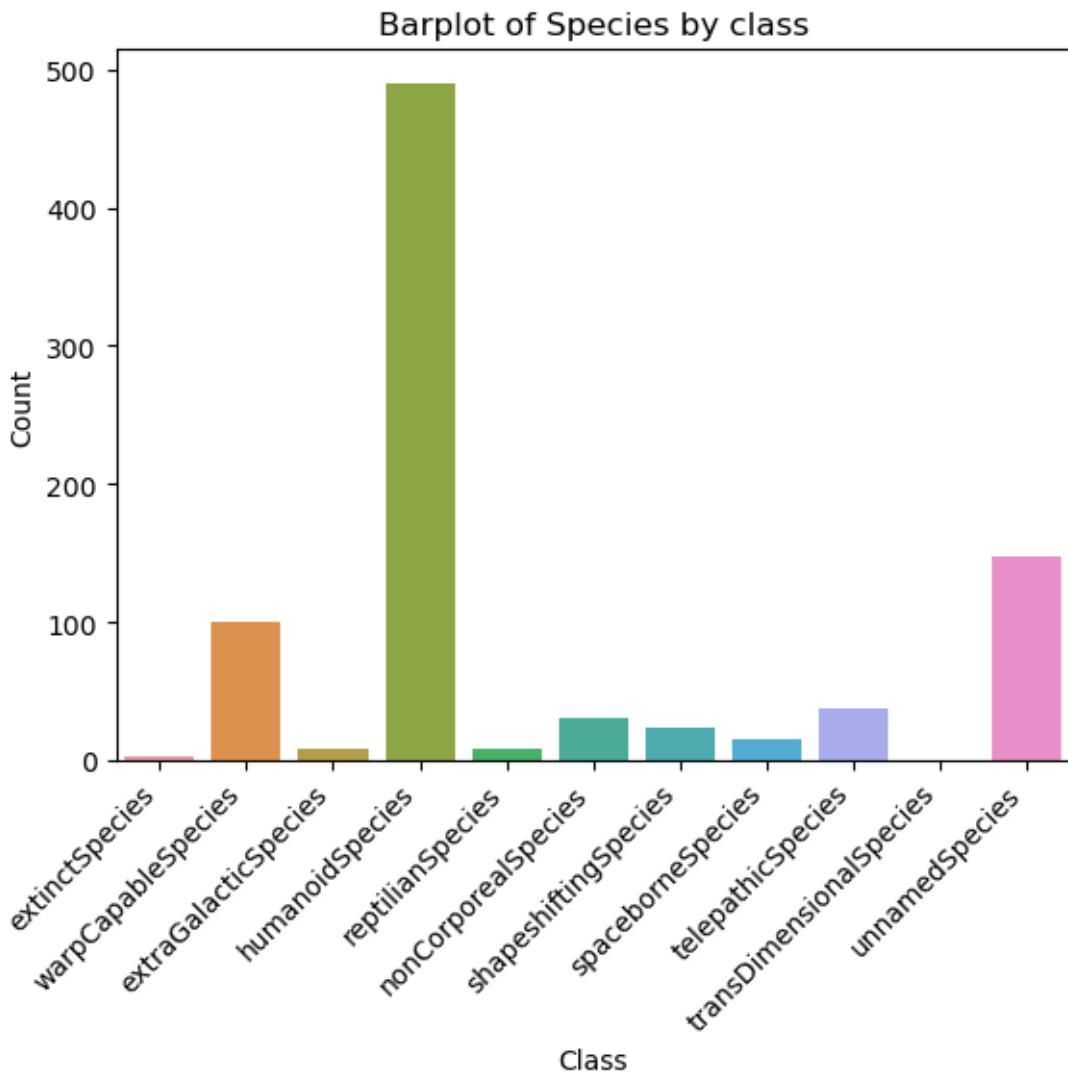
```
[15]: #make new dataframe with index as first column
df_sum = df_sp_classes_sum.reset_index()
# name the new columns
df_sum.columns = ['Class', 'Count']
print(df_sum)
```

	Class	Count
0	extinctSpecies	3
1	warpCapableSpecies	100
2	extraGalacticSpecies	9
3	humanoidSpecies	490
4	reptilianSpecies	8
5	nonCorporealSpecies	31
6	shapeshiftingSpecies	23
7	spaceborneSpecies	15
8	telepathicSpecies	38
9	transDimensionalSpecies	0
10	unnamedSpecies	147

### 1.3.1 Visualize the different classes of species:

- Make a barplot to visualize frequency of each class of species

```
[16]: sns.barplot(data=df_sum, x="Class", y="Count")
plt.xticks(rotation=45, ha = "right")
plt.title("Barplot of Species by class")
plt.show()
```



Filter warpCapableSpecies:

- Count the proportion of humanoid species in the warpCapableSpecies class
- Visualize the result in a pie chart

```
[17]: warp_capable_species = df[df['warpCapableSpecies']]
humanoid_warp = warp_capable_species['humanoidSpecies'].sum()
total_warp_capable = len(warp_capable_species)
proportion_humanoid_warp = humanoid_warp / total_warp_capable
proportion_non_humanoid_warp = (total_warp_capable - humanoid_warp) / total_warp_capable
print(proportion_humanoid_warp)
print(proportion_non_humanoid_warp)
```

```

labels=["Humanoid", "Non-humanoid"]
plt.pie([proportion_humanoid_warp, proportion_non_humanoid_warp],  

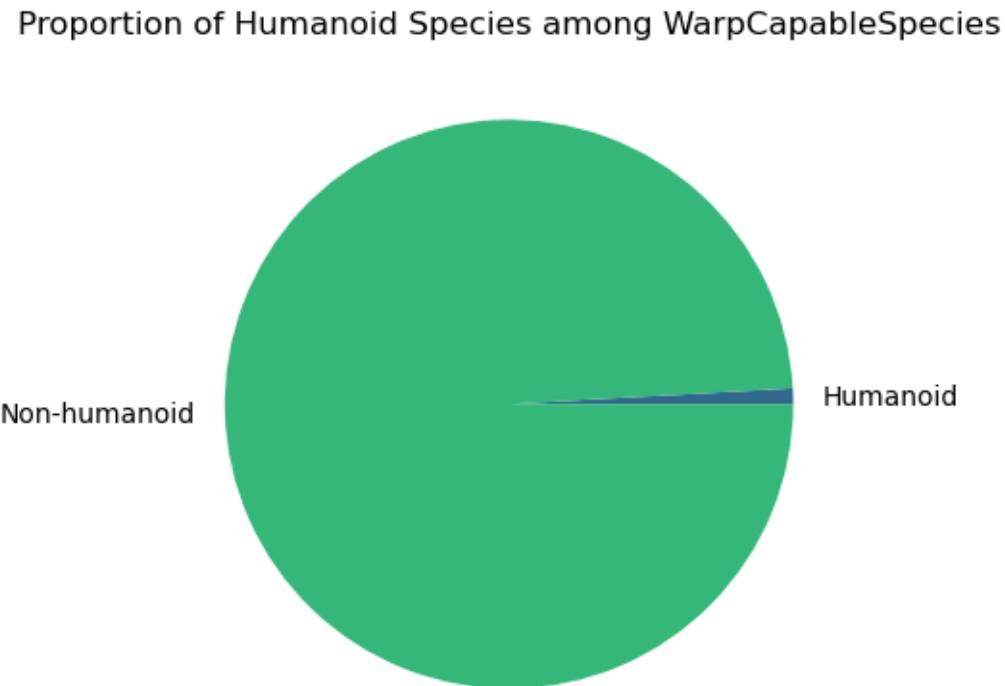
        labels=labels, colors=sns.color_palette('viridis', len(labels)))
plt.title('Proportion of Humanoid Species among WarpCapableSpecies')
plt.show()

```

0.89

99.11

[17]: Text(0.5, 1.0, 'Proportion of Humanoid Species among WarpCapableSpecies')



### Interpretation of Barplot and Pie chart

- humanoid species are the most common species (490 out of 759)
- out of 759 species in the galaxy only about 100 are capable of employing warp drive technology
- since there was no way of telling which classes of species are making up the warpCapableSpecies, filtering by warpCapable and counting the number of humnanoid vs. non-humanoid species provides a clearer picture:

The proportion of humanoids in the class of species capable of warp drive is less than 1%, although humanoids make up c. 65% of the galactic population. In other words, over 99% of the species that

employ warp drive technology are non-humanoid.

#### 1.4 Analyze population of species:

- Count number of different species living in each quadrant

```
[18]: # Group by quadrant and count species
species_by_quadrant = df['quadrant'].value_counts().reset_index()
species_by_quadrant.columns = ['Quadrant', 'Count']
print(species_by_quadrant)
```

	Quadrant	Count
0	Delta Quadrant	162
1	Alpha Quadrant	87
2	Gamma Quadrant	20
3	Beta Quadrant	19

#### Interpretation of the population data:

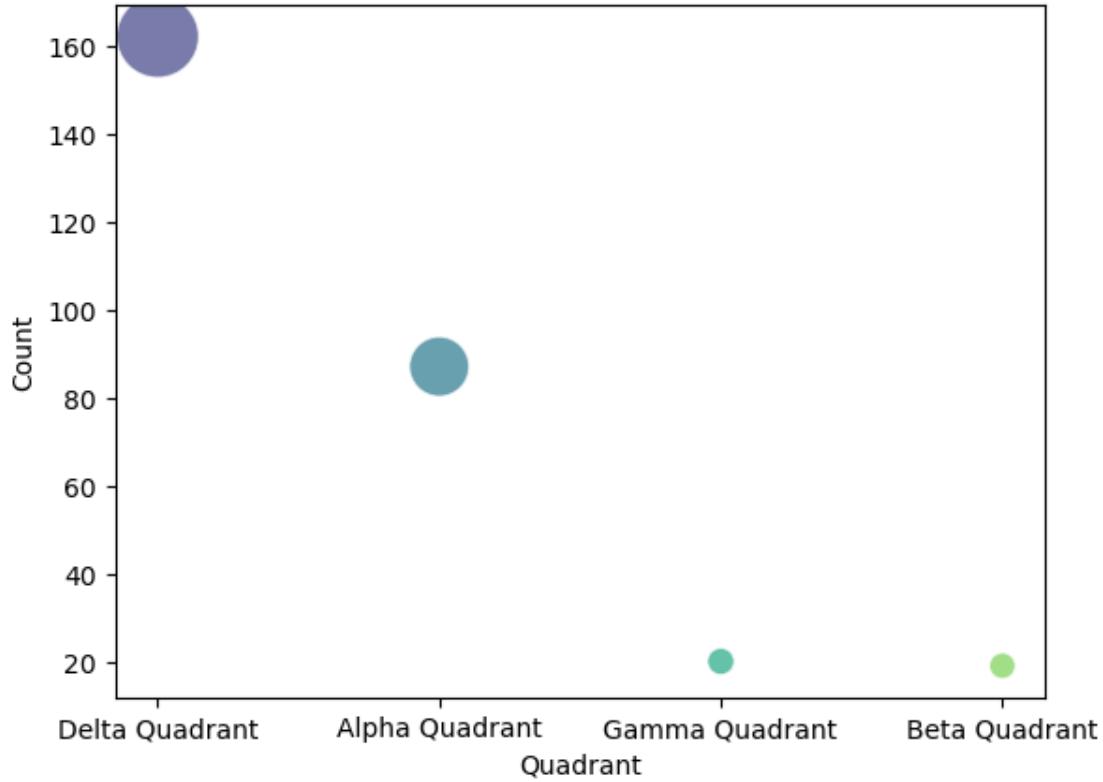
- Compared to the Alpha Quadrant (home to Earth) almost twice as many different species live in the Delta Quadrant

##### 1.4.1 Visualize the population in each quadrant

- bubble plot
- pie chart

```
[19]: sns.scatterplot(data=species_by_quadrant, x='Quadrant', y='Count',
                     size='Count', sizes=(100, 1000), hue='Quadrant', palette='viridis',
                     legend=False, alpha=0.7)
plt.show()
```

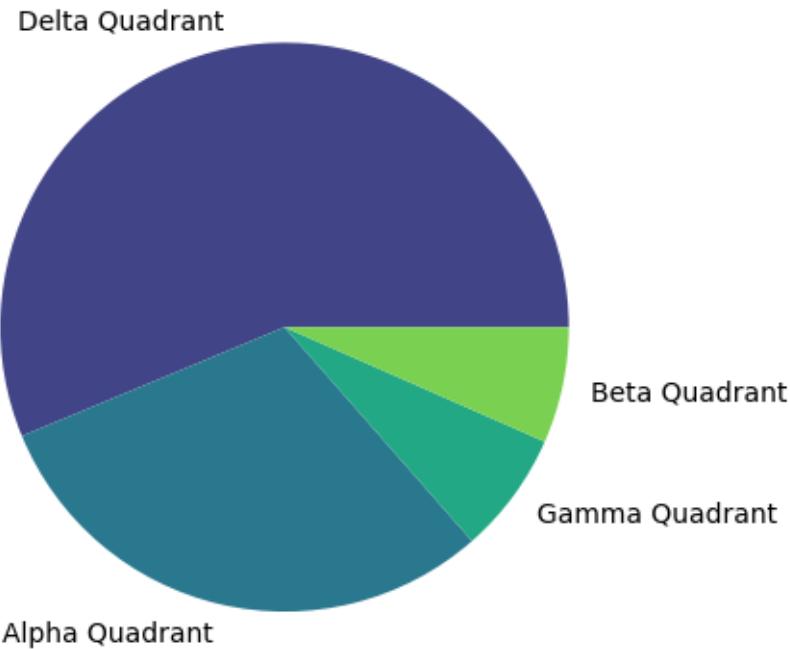
```
[19]: <Axes: xlabel='Quadrant', ylabel='Count'>
```



```
[20]: # visualize population with a pie chart
plt.pie(species_by_quadrant['Count'], labels=species_by_quadrant['Quadrant'],
         colors=sns.color_palette('viridis', len(species_by_quadrant)))
plt.title('Population of Species in Each Quadrant')
plt.show()
```

[20]: Text(0.5, 1.0, 'Population of Species in Each Quadrant')

Population of Species in Each Quadrant



#### 1.4.2 Analyze the proportion of humanoid species in the Delta Quadrant

- Filter by Delta Quadrant
- Count the number of humanoid species among the total population in the Delta Quadrant
- Visualize the result in a pie chart

```
[21]: #filter Delta Quadrant
delta_quadrant_species = df[df["quadrant"] == "Delta Quadrant"]

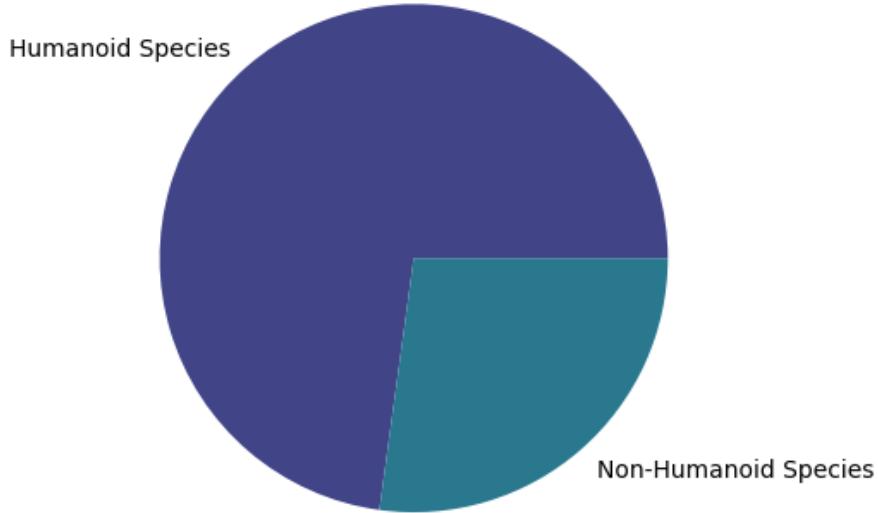
total_delta_quadrant_species = len(delta_quadrant_species)
humanoid_delta = delta_quadrant_species["humanoidSpecies"].sum()
non_humanoid_delta = total_delta_quadrant_species - humanoid_delta
print(total_delta_quadrant_species)
print(humanoid_delta)
proportion_humanoid_delta = humanoid_delta / total_delta_quadrant_species
proportion_non_humanoid_delta = (total_delta_quadrant_species - humanoid_delta) ↴
    / total_delta_quadrant_species
print(proportion_humanoid_delta)
print(proportion_non_humanoid_delta)
```

```
0.7283950617283951  
0.2716049382716049
```

```
[22]: labels = ["Humanoid Species", "Non-Humanoid Species"]  
sizes = [humanoid_delta, non_humanoid_delta]  
plt.pie(sizes, labels=labels, colors=sns.color_palette('viridis',  
len(species_by_quadrant)))  
plt.title("Proportion of Humanoid vs. Non-humanoid Population in the Delta  
Quadrant")  
plt.show()
```

```
[22]: Text(0.5, 1.0, 'Proportion of Humanoid vs. Non-humanoid Population in the Delta  
Quadrant')
```

Proportion of Humanoid vs. Non-humanoid Population in the Delta Quadrant



### Interpretation of Proportions and Pie chart

- humanoid species are the most common species in the Delta Quadrant
  - c. 73% of the population in the Delta Quadrant are humanoid
  - c. 27% of the population in the Delta Quadrant are non-humanoid
- (either way, we are not alone :))