1 - DEFINE THE PROBLEM

We are exploring the contents of the imagenet_class_names.txt file, which includes the class labels used in ImageNet-based deep learning models. This notebook will help us understand the label structure and prepare it for use in machine learning pipelines.

2 - IMPORT REQUIRED LIBRARIES

2.1 - Base Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import string
```

2.2 - ML/DL Libraries

In [2]: from sklearn.model_selection import train_test_split

3 - LOAD THE DATA

In [4]: # Convert to DataFrame for analysis

50%

75%

max

4 - EDA (Exploratory Data Analysis)

df = pd.DataFrame(class_names, columns=['class_name'])

```
df['length'] = df['class_name'].apply(len)
        df['first_letter'] = df['class_name'].str[0]
        df.describe(include='all')
Out[4]:
                                               length first_letter
                              class_name
                                    1000 1000.000000
         count
                                                            1000
        unique
                                    1000
                                                 NaN
           top n01440764 tench, Tinca tinca
                                                 NaN
           freq
                                       1
                                                 NaN
                                                            1000
                                            30.675000
                                     NaN
                                                            NaN
          mean
                                            16.886638
                                                            NaN
           std
                                     NaN
           min
                                     NaN
                                            12.000000
                                                            NaN
                                            18.000000
          25%
                                     NaN
                                                            NaN
```

26.000000

37.000000

131.000000

NaN

NaN

NaN

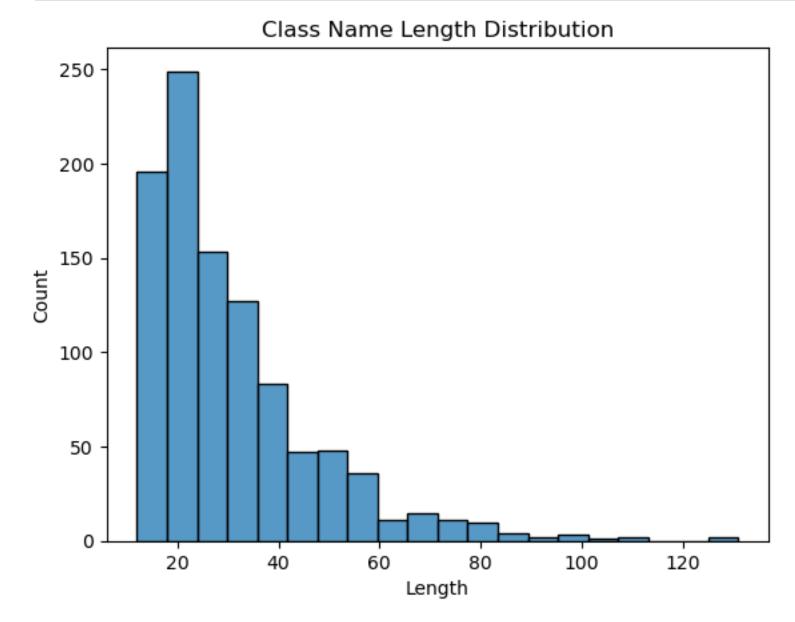
NaN

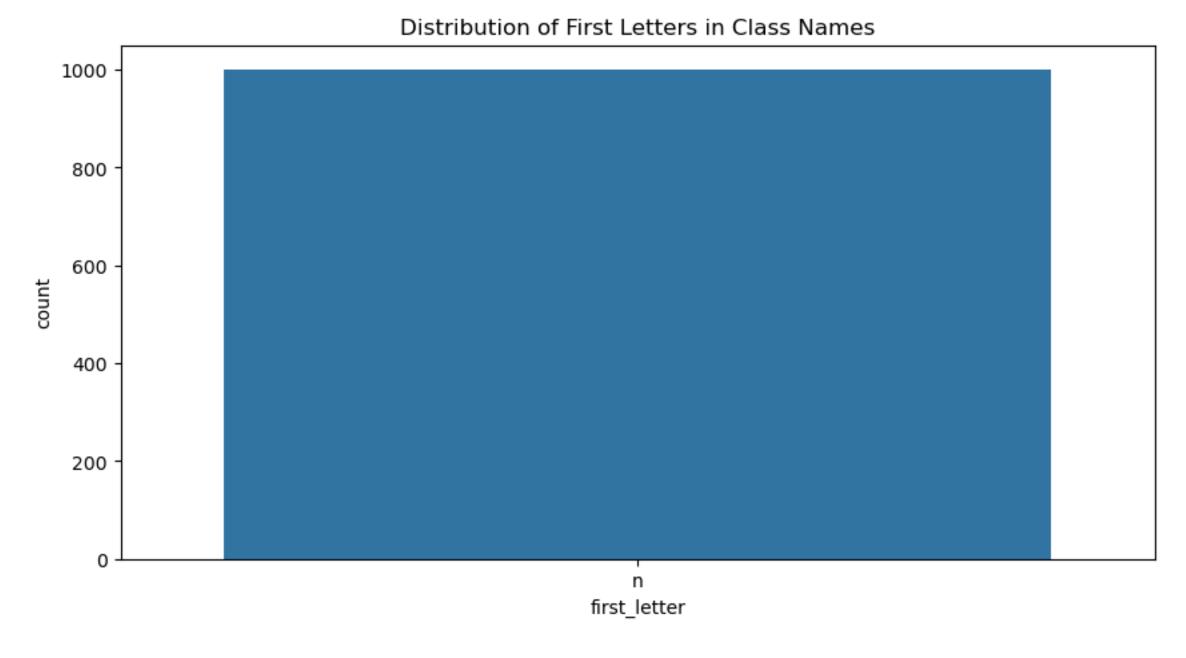
NaN

NaN

5 - VISUALIZE THE DATA

```
In [5]: # Histogram of class name lengths
sns.histplot(df['length'], bins=20)
plt.title('Class Name Length Distribution')
plt.xlabel('Length')
plt.ylabel('Count')
plt.show()
```





6 - PREPROCESS THE DATA

```
In [7]: # Basic standardization
    df['class_name_clean'] = df['class_name'].str.lower().str.replace('-', ' ').str.replace('_', ' ')
    df['class_name_clean'] = df['class_name_clean'].str.replace(r'[^a-z ]+', '', regex=True)
    df.head()
```

Out[7]:		class_name	length	first_letter	class_name_clean
	0	n01440764 tench, Tinca tinca	28	n	n tench tinca tinca
	1	n01443537 goldfish, Carassius auratus	37	n	n goldfish carassius auratus
	2	n01484850 great white shark, white shark, man	93	n	n great white shark white shark man eater man
	3	n01491361 tiger shark, Galeocerdo cuvieri	41	n	n tiger shark galeocerdo cuvieri
	4	n01494475 hammerhead, hammerhead shark	38	n	n hammerhead hammerhead shark

7 - SPLIT THE DATA (Optional example)

```
In [8]: # We can split classes into groups for manual use (e.g., 80% train, 20% holdout)
    train_classes, test_classes = train_test_split(df['class_name_clean'], test_size=0.2, random_state=42)
In [9]: print(f'Train classes: {len(train_classes)}')
    Train classes: 800
```

In [10]: print(f'Test classes: {len(test_classes)}')
 Test classes: 200

In [11]: train_classes[:5]

Out[11]: 29 n axolotl mud puppy ambystoma mexicanum 535 n disk brake disc brake 695 n padlock 557 n flagpole flagstaff 836 n sunglass

Name: class_name_clean, dtype: object