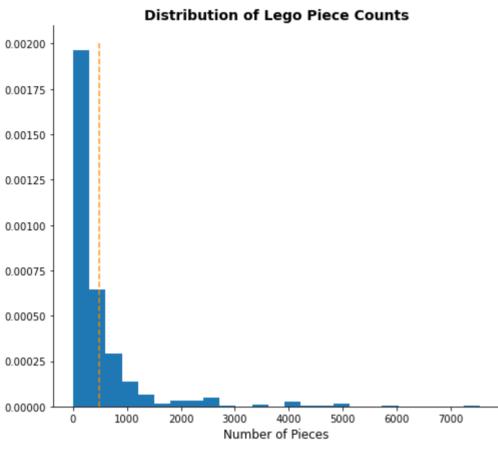
## **Code Demonstration of the Central Limit Theorem**

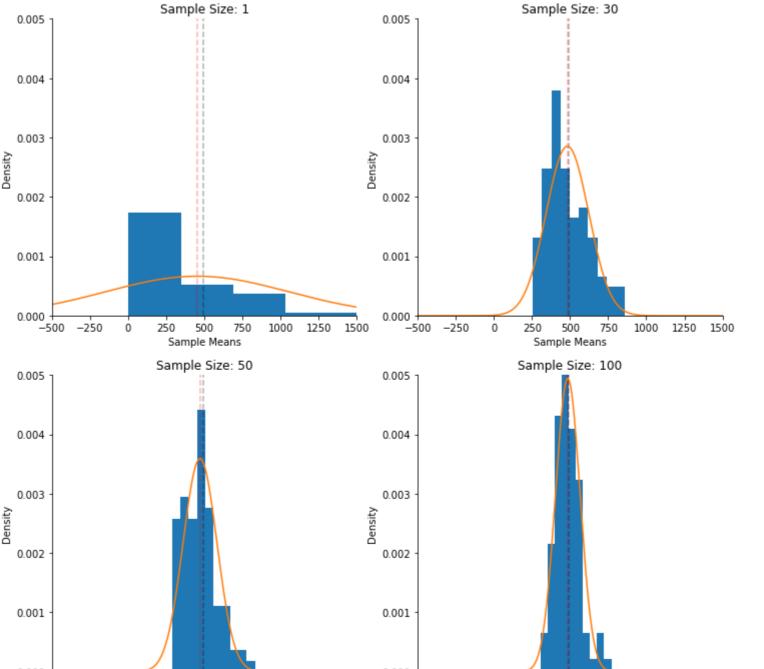
```
import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          from scipy.stats import norm
          # Load data and explore the statistics
          lego = pd.read csv('lego sets.csv')
          lego['piece count'].describe()
Out[13]: count
                  12261.000000
                   493.405921
         std
                    825.364580
                     1.000000
         min
                     97.000000
         25%
         50%
                    216.000000
         75%
                   544.000000
                  7541.000000
         max
         Name: piece count, dtype: float64
In [14]:
          # Create a density histogram for lego piece counts.
          fig, ax = plt.subplots(figsize=(8, 7))
          x = lego['piece count']
          ax.hist(x, bins=25, density=True)
          ax.spines['right'].set visible(False)
          ax.spines['top'].set visible(False)
          # Format the histrogram labels, range, and borders.
          ax.set title("Distribution of Lego Piece Counts", size=14, weight='bold')
          ax.set xlabel("Number of Pieces", fontsize=12)
          x axis = range(-1000, 2000)
          # Plot a vertical line at population mean.
          pop mean = 493.405921
          ax.vlines(x=pop mean, colors='DarkOrange', ymin=0, ymax=.002,
                    linestyles='--', alpha=0.9)
          plt.show()
                            Distribution of Lego Piece Counts
```



```
def get_sample_means(sample_size):
 This function takes 100 random samples from the population of
 Lego Piece Counts and returns a list of the sample means.
 Input parameters:
  sample size: integer specifying sample size
 piece_count = lego['piece_count']
 n = 100
 sample means = []
  # Take 30 random samples from the population and store the sample means.
  for i in range(n):
      sample = np.random.choice(piece count, sample size, replace=True)
      sample means.append(np.mean(sample))
  return sample means
```

```
In [16]:
          def plot_histogram(sample_means_ls):
              This function returns plots the sample means distibution for four
              different sample sizes, specified by the user.
              Input parameters:
              sample_means_ls: list of lists containing
              # Define a grid to plot 4 graphs.
              fig, axes = plt.subplots(2, 2, figsize=(12, 12))
              for idx, ax in enumerate(axes.flatten(), 0):
                  # Find the sample statistics.
                  mean = np.mean(sample_means_ls[idx])
                  std = np.std(sample_means_ls[idx])
                  # Plot a density histogram for each sample.
                  ax.hist(sample_means_ls[idx], density=True)
                  # Format the histrogram labels, range, and borders.
                  ax.set_xlim(-500, 1500)
                  ax.set_ylim(0, 0.005)
                  ax.spines['right'].set_visible(False)
                  ax.spines['top'].set_visible(False)
                  x_axis = range(-1000, 2000)
                  ax.set_ylabel("Density")
                  ax.set_xlabel("Sample Means")
                  # Plot a normal distribution with sample mean and standard deviation.
                  ax.plot(x_axis, norm.pdf(x_axis, mean, std))
                  ax.set_title(f'Sample Size: {sample_size_ls[idx]}')
                  # Plot a verticle line at population mean for comparison
                  pop_mean = np.mean(lego['piece_count'])
                  ax.vlines(x=pop mean, colors='black', ymin=0, ymax=0.005,
                           linestyles='--', alpha=0.3)
                  ax.vlines(x=mean, colors='red', ymin=0, ymax=0.005,
                            linestyles='--', alpha=0.3)
              plt.show()
```

```
# Simulate the central limit theorem for 4 different sample sizes.
sample size ls = [1, 30, 50, 100]
sample means ls = []
for sample size in sample size ls:
   sample_means_ls.append(get_sample_means(sample_size))
# Plot the sampling distributions of the means.
plot_histogram(sample_means_ls)
```



1250

1500

-500

-250

0

250

500

Sample Means

750

1000

-500

-250

0

500

Sample Means

750

1000

1250

1500