

# Assignment 1 - Question 12

## Setup: Libraries and Configuration

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

# Plotting settings
plt.style.use('seaborn-v0_8-darkgrid')
sns.set_palette("husl")

# Random seed
seed = 42
np.random.seed(seed)

print(f"NumPy: {np.__version__}")
print(f"Matplotlib: {plt.matplotlib.__version__}")

NumPy: 2.2.6
Matplotlib: 3.10.8
```

## Question 12 (a)

Goal: Simulate  $N = \{10, 100\}$  lifetimes from  $\text{Exponential}(\theta_0)$  with true rate  $\theta_0 = 0.2$ .

### Prepare a Random Uniform Distribution

```
In [2]: def uniform_dist(n_samples, low=0.0, high=1.0, seed=seed):
    rng = np.random.default_rng(seed)
    return rng.uniform(low, high, n_samples)

# Example
n_samples = 10
data = uniform_dist(n_samples, low=-3.0, high=3.0)
assert n_samples == len(data)
print(f"{len(data)} samples from Uniform distribution: {data}")

10 samples from Uniform distribution: [ 1.64373629 -0.36672936  2.15158752  1.18420817 -2.43493591  2.85373411
  1.56683821  1.71638583 -2.2313182  -0.29768437]
```

### The Exponential Update

```
In [3]: # Exponential update
def exp_update(theta, gradient, learning_rate=0.1):
    return theta * np.exp(-learning_rate * gradient)
```

### Configurations

```
In [4]: theta0 = -0.2 # initial weight
gradients = uniform_dist(100, low=-1.0, high=1.0)
```

## The Simulation

```
In [5]: final_thetas = []
for n in range(10, 101): # sample sizes to try
    # Learning
    theta = theta0
    updated_thetas = []
    for gradient in gradients[:n]:
        theta = exp_update(theta, gradient, learning_rate=0.1)
        updated_thetas.append(theta)

    final_thetas.append(theta)
print(f"Final theta after {n} updates: {theta}\n")

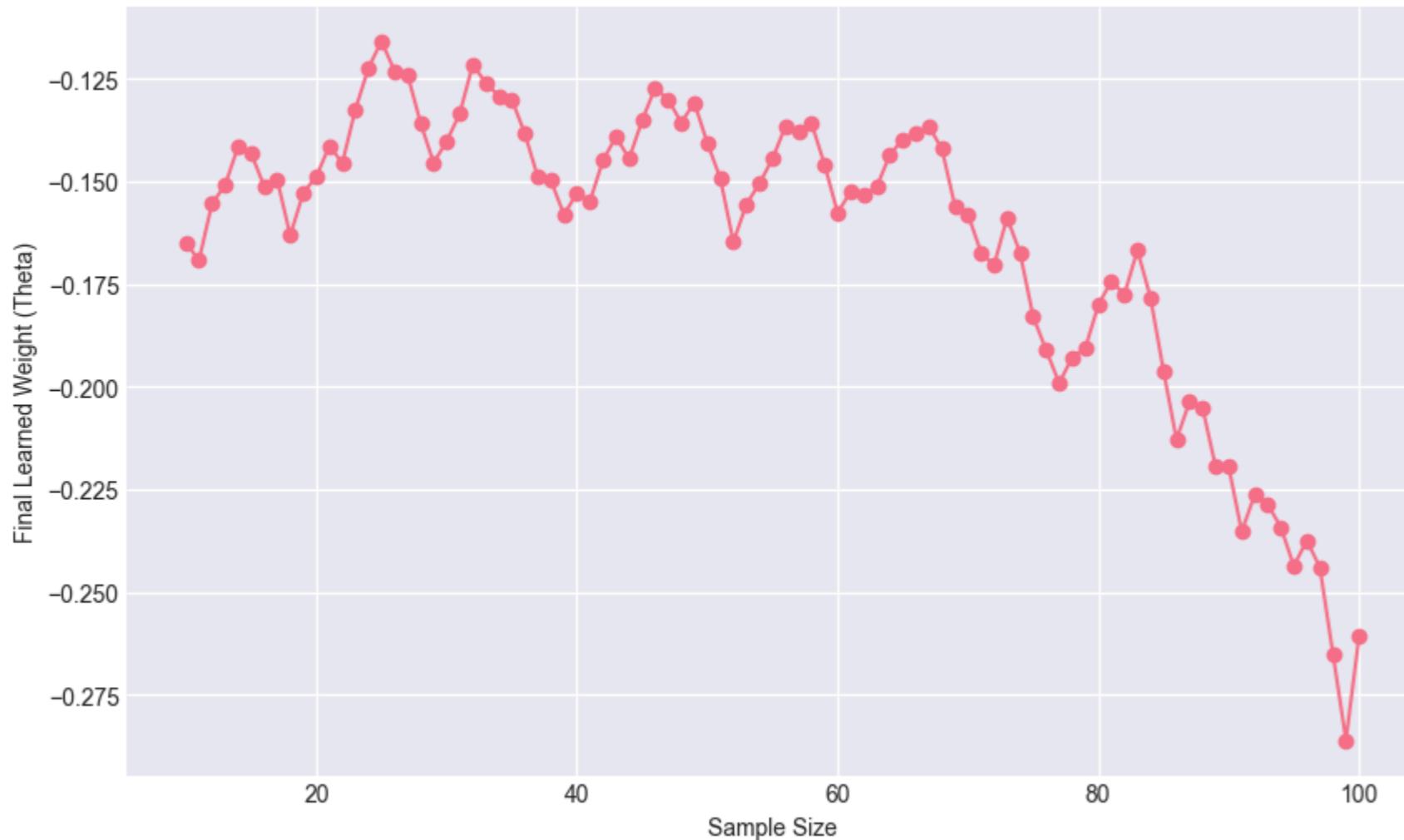
# Plot thetas
plt.figure(figsize=(10, 6))
plt.plot(range(10, 101), final_thetas, marker='o')
plt.title("Final Learned Weights vs Sample Size")
plt.xlabel("Sample Size")
plt.ylabel("Final Learned Weight (Theta)")
plt.grid(True)
plt.show()
```

```
Final theta after 10 updates: -0.16491935941584984
Final theta after 11 updates: -0.16923647864664154
Final theta after 12 updates: -0.15539092227421694
Final theta after 13 updates: -0.15098356608038488
Final theta after 14 updates: -0.14154513964928608
Final theta after 15 updates: -0.1431561273839617
Final theta after 16 updates: -0.15118255652783463
Final theta after 17 updates: -0.14954107926934
Final theta after 18 updates: -0.16317246053792028
Final theta after 19 updates: -0.15282316172092175
Final theta after 20 updates: -0.14885141109010788
Final theta after 21 updates: -0.14136299563890076
Final theta after 22 updates: -0.14553634157816145
Final theta after 23 updates: -0.13246072956320598
Final theta after 24 updates: -0.12244500575564157
Final theta after 25 updates: -0.11581398161234334
Final theta after 26 updates: -0.12310745053312237
Final theta after 27 updates: -0.12392956187972376
Final theta after 28 updates: -0.13576868630631345
Final theta after 29 updates: -0.14548815920648478
Final theta after 30 updates: -0.14025818638240412
Final theta after 31 updates: -0.13355755138286945
Final theta after 32 updates: -0.12163570279620681
Final theta after 33 updates: -0.12594753902813766
Final theta after 34 updates: -0.12925323241952325
Final theta after 35 updates: -0.13004263521363113
Final theta after 36 updates: -0.1383750964506651
Final theta after 37 updates: -0.14900558477958842
Final theta after 38 updates: -0.14973136697981446
Final theta after 39 updates: -0.15813687127645185
Final theta after 40 updates: -0.15285627991875403
Final theta after 41 updates: -0.1547897507315117
```

```
Final theta after 42 updates: -0.14482586794239616
Final theta after 43 updates: -0.13913978699232263
Final theta after 44 updates: -0.1444604490846242
Final theta after 45 updates: -0.1351727785230217
Final theta after 46 updates: -0.1271796850661864
Final theta after 47 updates: -0.13007422560985624
Final theta after 48 updates: -0.13569905907563415
Final theta after 49 updates: -0.13083546378973904
Final theta after 50 updates: -0.1406099908379113
Final theta after 51 updates: -0.1493075682689818
Final theta after 52 updates: -0.16476759090367366
Final theta after 53 updates: -0.1555785992758262
Final theta after 54 updates: -0.15053278397372702
Final theta after 55 updates: -0.14448097217814806
Final theta after 56 updates: -0.13659249610051427
Final theta after 57 updates: -0.13771947923641426
Final theta after 58 updates: -0.13583903491295315
Final theta after 59 updates: -0.14598607177053774
Final theta after 60 updates: -0.1576859195403534
Final theta after 61 updates: -0.15246340671105763
Final theta after 62 updates: -0.15334731325048648
Final theta after 63 updates: -0.15135955249451752
Final theta after 64 updates: -0.1435464070868267
Final theta after 65 updates: -0.139730380647369
Final theta after 66 updates: -0.1382410406835984
Final theta after 67 updates: -0.1366137226830698
Final theta after 68 updates: -0.14207674670576093
Final theta after 69 updates: -0.15605426735929287
Final theta after 70 updates: -0.15804192352535065
Final theta after 71 updates: -0.16732589904772077
Final theta after 72 updates: -0.17041517631146608
Final theta after 73 updates: -0.15878594838168053
```

```
Final theta after 74 updates: -0.16746412705015373
Final theta after 75 updates: -0.1829309232724213
Final theta after 76 updates: -0.19110668545899867
Final theta after 77 updates: -0.19916090844926873
Final theta after 78 updates: -0.19281473010593653
Final theta after 79 updates: -0.19062789797810017
Final theta after 80 updates: -0.18010566526985594
Final theta after 81 updates: -0.17428310197078242
Final theta after 82 updates: -0.17757687745633494
Final theta after 83 updates: -0.16676732053005627
Final theta after 84 updates: -0.17825319201732873
Final theta after 85 updates: -0.19610741641389853
Final theta after 86 updates: -0.21286389692113428
Final theta after 87 updates: -0.2036048508455203
Final theta after 88 updates: -0.20516318024922567
Final theta after 89 updates: -0.21954370139817128
Final theta after 90 updates: -0.21949783143205384
Final theta after 91 updates: -0.23530438523516098
Final theta after 92 updates: -0.22624440598079837
Final theta after 93 updates: -0.22869393980109348
Final theta after 94 updates: -0.23420114894818128
Final theta after 95 updates: -0.24368537320740438
Final theta after 96 updates: -0.2374177910343628
Final theta after 97 updates: -0.24407093455413903
Final theta after 98 updates: -0.26505276372586545
Final theta after 99 updates: -0.28609609020874877
Final theta after 100 updates: -0.2608506969305339
```

Final Learned Weights vs Sample Size



## Question 12 (b)

Goal: Compute and print the MLE  $\hat{\theta}$  from the simulated data.

$$\text{MLE} = n / \sum_{i=1}^N x_i = 1/E[X]$$

```
In [6]: # Compute MLE for each sample size
mle_thetas = []

for n in range(10, 101):
    # Generate samples
    samples = uniform_dist(n, low=0.0, high=1.0, seed=seed)

    # Compute MLE: n / sum(X_i) = 1 / E[X]
    theta_mle = 1.0 / np.mean(samples)
    mle_thetas.append(theta_mle)

    print(f"Sample Size: {n}")
    print(f"True theta: {theta0}")
    print(f"MLE theta: {theta_mle:.4f}")
    print(f"Sample mean: {np.mean(samples):.4f}\n")

# Plot MLE thetas
plt.figure(figsize=(10, 6))
plt.plot(range(10, 101), mle_thetas, marker='o', color='orange')
plt.title("MLE Thetas vs Sample Size")
plt.xlabel("Sample Size")
```

```
plt.ylabel("MLE Theta")
plt.grid(True)
plt.show()
```

Sample Size: 10  
True theta: -0.2  
MLE theta: 1.6766  
Sample mean: 0.5964

Sample Size: 11  
True theta: -0.2  
MLE theta: 1.7364  
Sample mean: 0.5759

Sample Size: 12  
True theta: -0.2  
MLE theta: 1.6525  
Sample mean: 0.6052

Sample Size: 13  
True theta: -0.2  
MLE theta: 1.6444  
Sample mean: 0.6081

Sample Size: 14  
True theta: -0.2  
MLE theta: 1.6039  
Sample mean: 0.6235

Sample Size: 15  
True theta: -0.2  
MLE theta: 1.6354  
Sample mean: 0.6115

Sample Size: 16  
True theta: -0.2  
MLE theta: 1.7023  
Sample mean: 0.5874

Sample Size: 17  
True theta: -0.2  
MLE theta: 1.7079  
Sample mean: 0.5855

Sample Size: 18  
True theta: -0.2  
MLE theta: 1.7968  
Sample mean: 0.5565

Sample Size: 19  
True theta: -0.2  
MLE theta: 1.7519  
Sample mean: 0.5708

Sample Size: 20  
True theta: -0.2  
MLE theta: 1.7426  
Sample mean: 0.5738

Sample Size: 21  
True theta: -0.2  
MLE theta: 1.7164  
Sample mean: 0.5826

Sample Size: 22  
True theta: -0.2  
MLE theta: 1.7475  
Sample mean: 0.5722

Sample Size: 23  
True theta: -0.2  
MLE theta: 1.6961  
Sample mean: 0.5896

Sample Size: 24  
True theta: -0.2  
MLE theta: 1.6605  
Sample mean: 0.6022

Sample Size: 25  
True theta: -0.2  
MLE theta: 1.6413  
Sample mean: 0.6093

Sample Size: 26  
True theta: -0.2  
MLE theta: 1.6854  
Sample mean: 0.5933

Sample Size: 27  
True theta: -0.2  
MLE theta: 1.6989  
Sample mean: 0.5886

Sample Size: 28  
True theta: -0.2  
MLE theta: 1.7569  
Sample mean: 0.5692

Sample Size: 29  
True theta: -0.2  
MLE theta: 1.8022  
Sample mean: 0.5549

Sample Size: 30  
True theta: -0.2  
MLE theta: 1.7885  
Sample mean: 0.5591

Sample Size: 31  
True theta: -0.2  
MLE theta: 1.7695  
Sample mean: 0.5651

Sample Size: 32  
True theta: -0.2  
MLE theta: 1.7310  
Sample mean: 0.5777

Sample Size: 33  
True theta: -0.2  
MLE theta: 1.7542  
Sample mean: 0.5701

Sample Size: 34  
True theta: -0.2  
MLE theta: 1.7724  
Sample mean: 0.5642

Sample Size: 35  
True theta: -0.2  
MLE theta: 1.7810

Sample mean: 0.5615

Sample Size: 36  
True theta: -0.2  
MLE theta: 1.8144  
Sample mean: 0.5512

Sample Size: 37  
True theta: -0.2  
MLE theta: 1.8526  
Sample mean: 0.5398

Sample Size: 38  
True theta: -0.2  
MLE theta: 1.8584  
Sample mean: 0.5381

Sample Size: 39  
True theta: -0.2  
MLE theta: 1.8864  
Sample mean: 0.5301

Sample Size: 40  
True theta: -0.2  
MLE theta: 1.8741  
Sample mean: 0.5336

Sample Size: 41  
True theta: -0.2  
MLE theta: 1.8824  
Sample mean: 0.5312

Sample Size: 42  
True theta: -0.2  
MLE theta: 1.8573  
Sample mean: 0.5384

Sample Size: 43  
True theta: -0.2  
MLE theta: 1.8444  
Sample mean: 0.5422

Sample Size: 44  
True theta: -0.2  
MLE theta: 1.8623  
Sample mean: 0.5370

Sample Size: 45  
True theta: -0.2  
MLE theta: 1.8398  
Sample mean: 0.5435

Sample Size: 46  
True theta: -0.2  
MLE theta: 1.8208  
Sample mean: 0.5492

Sample Size: 47  
True theta: -0.2  
MLE theta: 1.8323  
Sample mean: 0.5458

Sample Size: 48  
True theta: -0.2

MLE theta: 1.8505  
Sample mean: 0.5404

Sample Size: 49  
True theta: -0.2  
MLE theta: 1.8406  
Sample mean: 0.5433

Sample Size: 50  
True theta: -0.2  
MLE theta: 1.8683  
Sample mean: 0.5352

Sample Size: 51  
True theta: -0.2  
MLE theta: 1.8916  
Sample mean: 0.5287

Sample Size: 52  
True theta: -0.2  
MLE theta: 1.9281  
Sample mean: 0.5186

Sample Size: 53  
True theta: -0.2  
MLE theta: 1.9095  
Sample mean: 0.5237

Sample Size: 54  
True theta: -0.2  
MLE theta: 1.9000  
Sample mean: 0.5263

Sample Size: 55  
True theta: -0.2  
MLE theta: 1.8884  
Sample mean: 0.5296

Sample Size: 56  
True theta: -0.2  
MLE theta: 1.8725  
Sample mean: 0.5340

Sample Size: 57  
True theta: -0.2  
MLE theta: 1.8771  
Sample mean: 0.5327

Sample Size: 58  
True theta: -0.2  
MLE theta: 1.8749  
Sample mean: 0.5333

Sample Size: 59  
True theta: -0.2  
MLE theta: 1.8987  
Sample mean: 0.5267

Sample Size: 60  
True theta: -0.2  
MLE theta: 1.9238  
Sample mean: 0.5198

Sample Size: 61

True theta: -0.2  
MLE theta: 1.9148  
Sample mean: 0.5222

Sample Size: 62  
True theta: -0.2  
MLE theta: 1.9178  
Sample mean: 0.5214

Sample Size: 63  
True theta: -0.2  
MLE theta: 1.9153  
Sample mean: 0.5221

Sample Size: 64  
True theta: -0.2  
MLE theta: 1.9015  
Sample mean: 0.5259

Sample Size: 65  
True theta: -0.2  
MLE theta: 1.8954  
Sample mean: 0.5276

Sample Size: 66  
True theta: -0.2  
MLE theta: 1.8940  
Sample mean: 0.5280

Sample Size: 67  
True theta: -0.2  
MLE theta: 1.8923  
Sample mean: 0.5284

Sample Size: 68  
True theta: -0.2  
MLE theta: 1.9042  
Sample mean: 0.5251

Sample Size: 69  
True theta: -0.2  
MLE theta: 1.9306  
Sample mean: 0.5180

Sample Size: 70  
True theta: -0.2  
MLE theta: 1.9349  
Sample mean: 0.5168

Sample Size: 71  
True theta: -0.2  
MLE theta: 1.9510  
Sample mean: 0.5126

Sample Size: 72  
True theta: -0.2  
MLE theta: 1.9565  
Sample mean: 0.5111

Sample Size: 73  
True theta: -0.2  
MLE theta: 1.9387  
Sample mean: 0.5158

Sample Size: 74  
True theta: -0.2  
MLE theta: 1.9531  
Sample mean: 0.5120

Sample Size: 75  
True theta: -0.2  
MLE theta: 1.9765  
Sample mean: 0.5059

Sample Size: 76  
True theta: -0.2  
MLE theta: 1.9881  
Sample mean: 0.5030

Sample Size: 77  
True theta: -0.2  
MLE theta: 1.9989  
Sample mean: 0.5003

Sample Size: 78  
True theta: -0.2  
MLE theta: 1.9907  
Sample mean: 0.5023

Sample Size: 79  
True theta: -0.2  
MLE theta: 1.9879  
Sample mean: 0.5030

Sample Size: 80  
True theta: -0.2  
MLE theta: 1.9741  
Sample mean: 0.5065

Sample Size: 81  
True theta: -0.2  
MLE theta: 1.9666  
Sample mean: 0.5085

Sample Size: 82  
True theta: -0.2  
MLE theta: 1.9714  
Sample mean: 0.5073

Sample Size: 83  
True theta: -0.2  
MLE theta: 1.9572  
Sample mean: 0.5109

Sample Size: 84  
True theta: -0.2  
MLE theta: 1.9730  
Sample mean: 0.5069

Sample Size: 85  
True theta: -0.2  
MLE theta: 1.9954  
Sample mean: 0.5012

Sample Size: 86  
True theta: -0.2  
MLE theta: 2.0146  
Sample mean: 0.4964

Sample Size: 87  
True theta: -0.2  
MLE theta: 2.0041  
Sample mean: 0.4990

Sample Size: 88  
True theta: -0.2  
MLE theta: 2.0058  
Sample mean: 0.4986

Sample Size: 89  
True theta: -0.2  
MLE theta: 2.0212  
Sample mean: 0.4948

Sample Size: 90  
True theta: -0.2  
MLE theta: 2.0209  
Sample mean: 0.4948

Sample Size: 91  
True theta: -0.2  
MLE theta: 2.0364  
Sample mean: 0.4911

Sample Size: 92  
True theta: -0.2  
MLE theta: 2.0272  
Sample mean: 0.4933

Sample Size: 93  
True theta: -0.2  
MLE theta: 2.0293  
Sample mean: 0.4928

Sample Size: 94  
True theta: -0.2  
MLE theta: 2.0342  
Sample mean: 0.4916

Sample Size: 95  
True theta: -0.2  
MLE theta: 2.0425  
Sample mean: 0.4896

Sample Size: 96  
True theta: -0.2  
MLE theta: 2.0364  
Sample mean: 0.4911

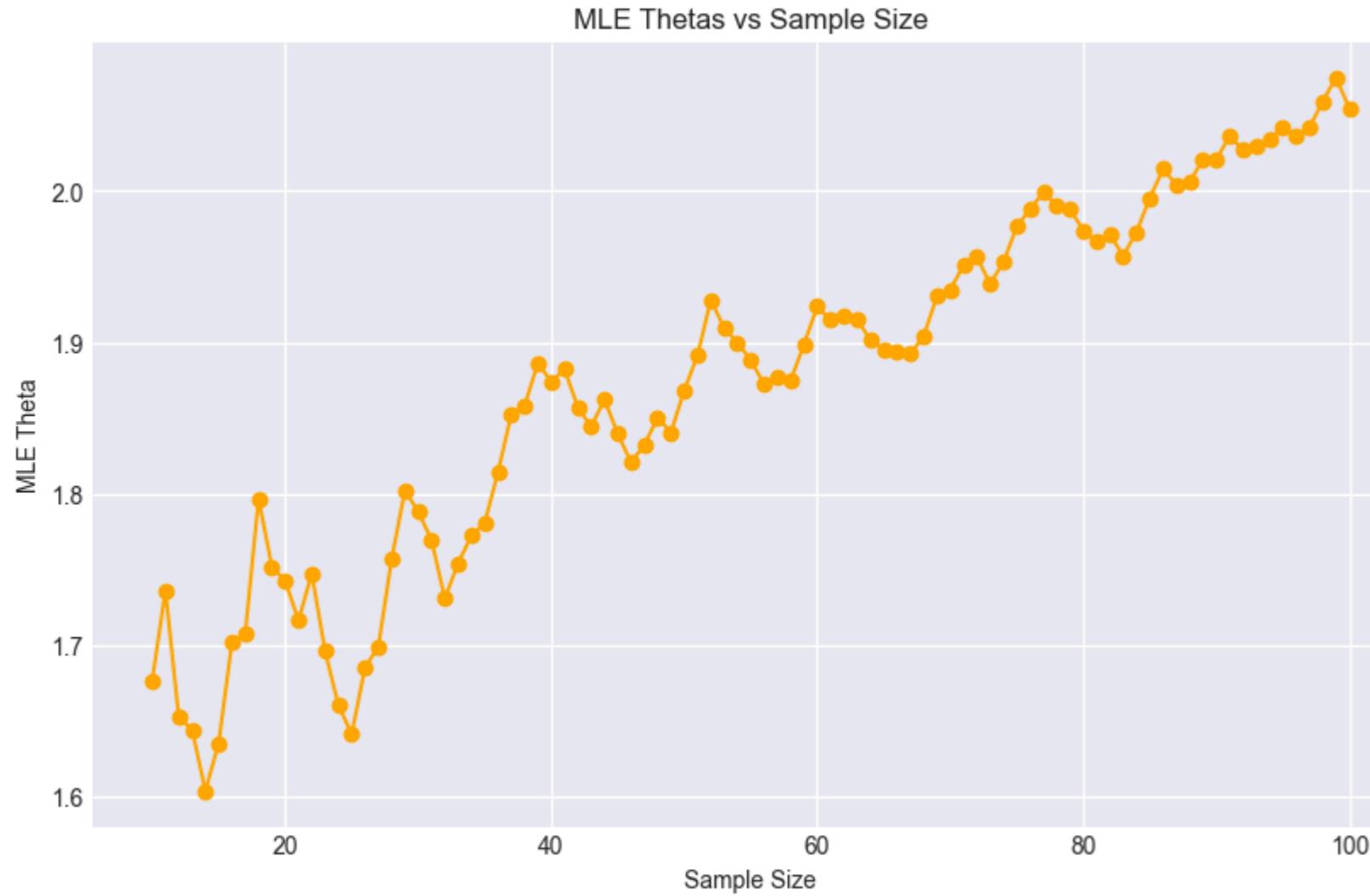
Sample Size: 97  
True theta: -0.2  
MLE theta: 2.0419  
Sample mean: 0.4897

Sample Size: 98  
True theta: -0.2  
MLE theta: 2.0592  
Sample mean: 0.4856

Sample Size: 99  
True theta: -0.2  
MLE theta: 2.0750

```
Sample mean: 0.4819
```

```
Sample Size: 100
True theta: -0.2
MLE theta: 2.0546
Sample mean: 0.4867
```



**Observation:** As the sample size grows, the Maximum Likelihood Estimation (MLE) is enlarged.

## Question 12 (c) - Add a Prior

Goal: Using the given prior, compute the posterior distribution parameters and plot the posterior density. On the same plot, show the likelihood function (up to a constant). and the prior density for comparison.

### Formulate the Prior

```
In [7]: def prior_dist(theta):
    # Exponential Dist
    coeff = theta ** 3
    exponent = -3 * theta
    return coeff * np.exp(exponent)
```

```
# Example
prior_dist(0.2)
```

```
Out[7]: np.float64(0.004390493088752212)
```

## Formulate the Likelihood

```
In [8]: # Define Likelihood function for uniform distribution U(θ, theta)
def likelihood(theta_vals, samples):
    n = len(samples)
    max_sample = np.max(samples)

    likelihood_vals = np.where(
        theta_vals >= max_sample, # Condition: see if weights controlled by theta can explain data (take max sample)
        (1.0 / theta_vals) ** n, # Success case
        0.0 # Else
    )
    return likelihood_vals

# Example
likelihood(np.array([0.5, 1.0, 1.5]), np.array([0.2, 0.4, 0.1]))
```

Out[8]: array([8.0, 1.0, 0.2962963])

## Run Simulations

```
In [9]: # Plot all three on the same figure
def plot_distribution(theta_range, prior, likelihood, posterior, n):
    plt.figure(figsize=(12, 6))
    plt.plot(theta_range, prior, label='Prior', color='blue', linewidth=2)
    plt.plot(theta_range, likelihood, label='Likelihood', color='orange', linewidth=2)
    plt.plot(theta_range, posterior, label='Posterior', color='green', linewidth=2)
    plt.title(f'Prior, Likelihood, and Posterior Densities (N={n})')
    plt.xlabel("θ (theta)")
    plt.ylabel("Probability Density (Normalized)")
    plt.legend(fontsize=10, loc='upper right')
    plt.grid(True, alpha=0.3)
    plt.xlim(0, 2.0)
    plt.show()

selected_samples_plot = [10, 15, 50, 100]

for n in range(10, 101):
    # Generate samples
    samples = uniform_dist(n, low=0.0, high=1.0, seed=seed)

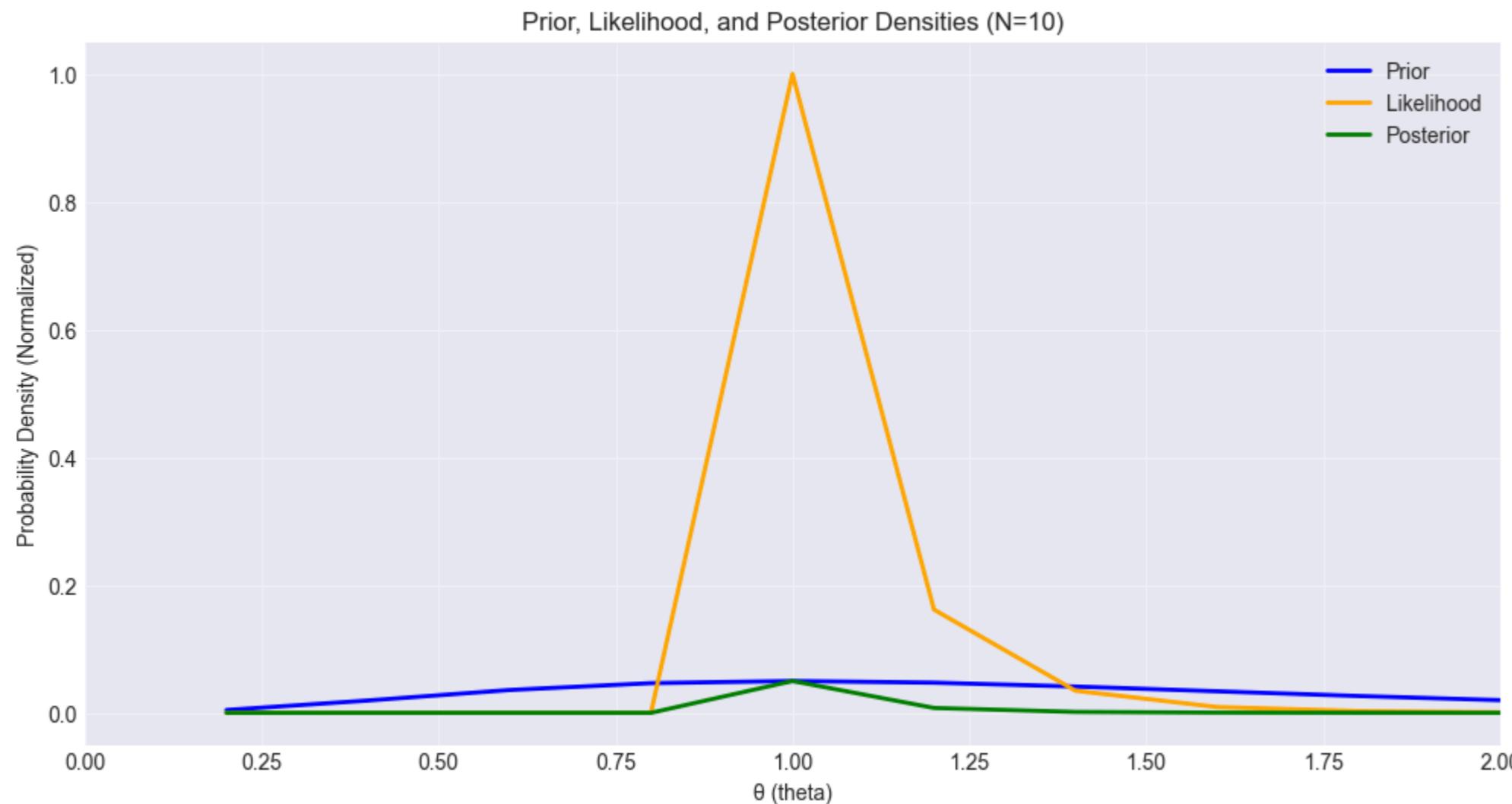
    # For simplicity, MLE for uniform U(θ, theta) is max(samples)
    max_sample = np.max(samples)
    theta_mle_viz = max_sample

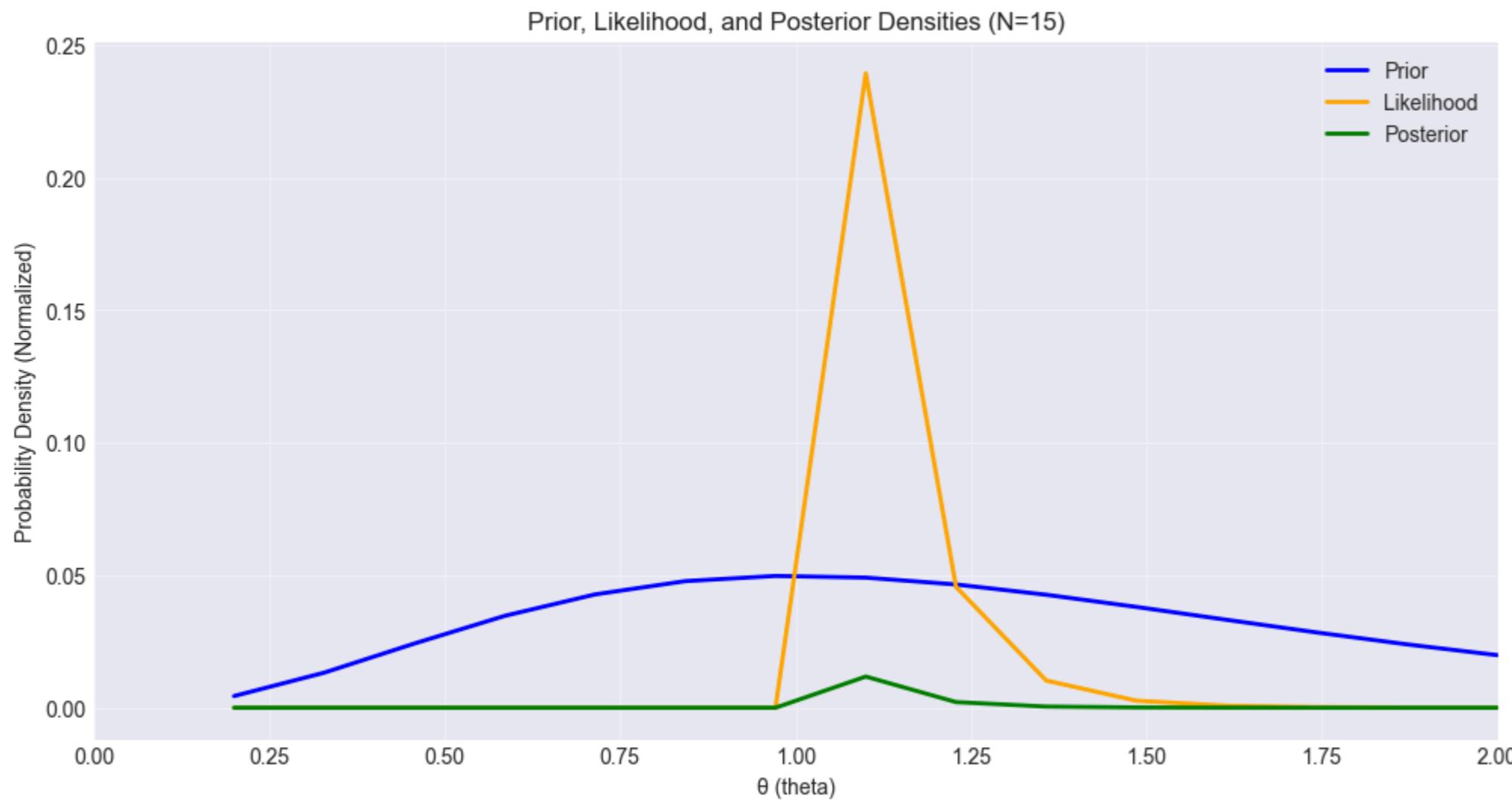
    # Prior
    theta0 = 0.2
    theta_range = np.linspace(theta0, 2.0, n)

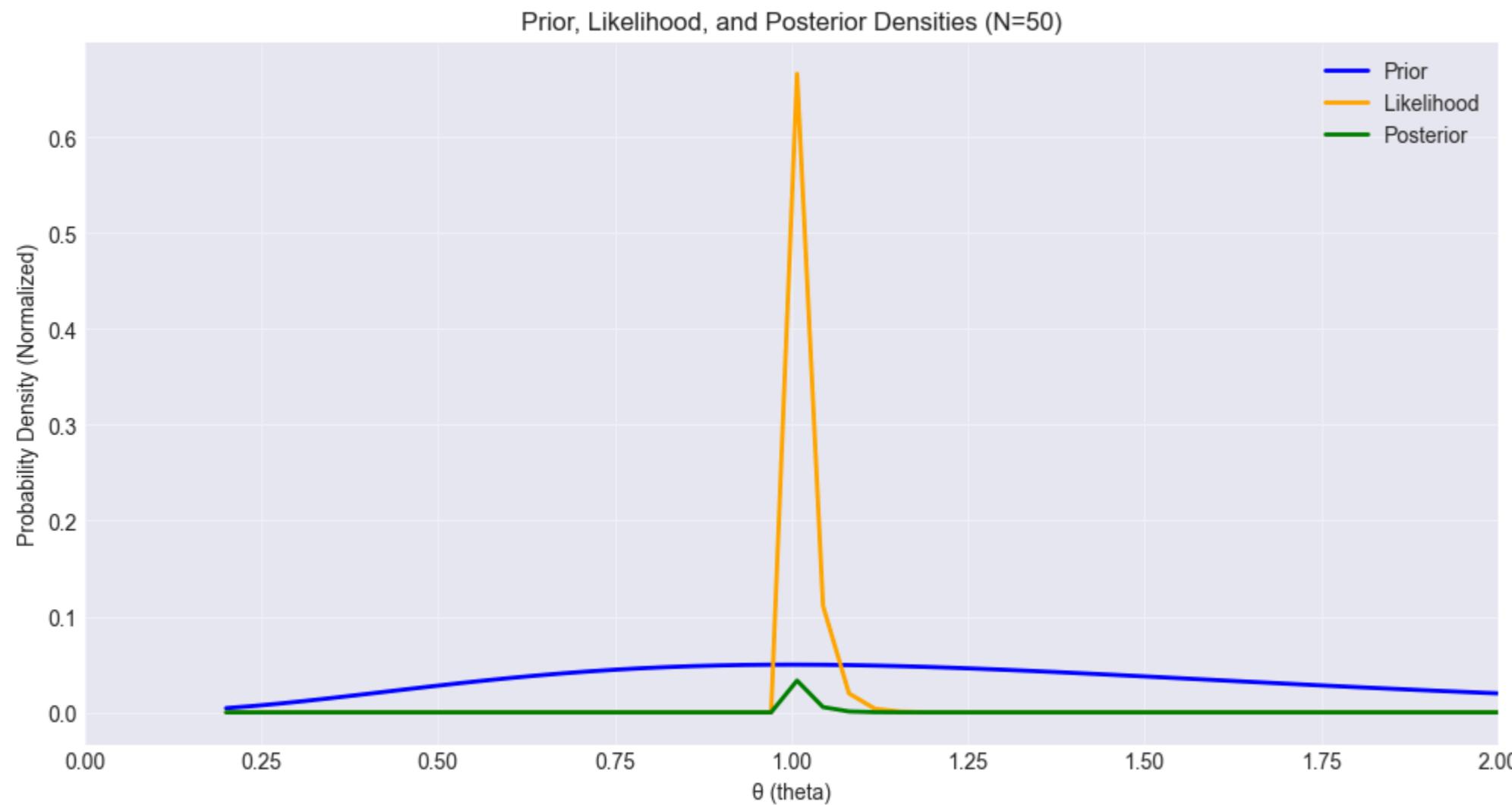
    # Compute prior, likelihood, and posterior
    prior_vals = prior_dist(theta_range)
    likelihood_vals = likelihood(theta_range, samples)

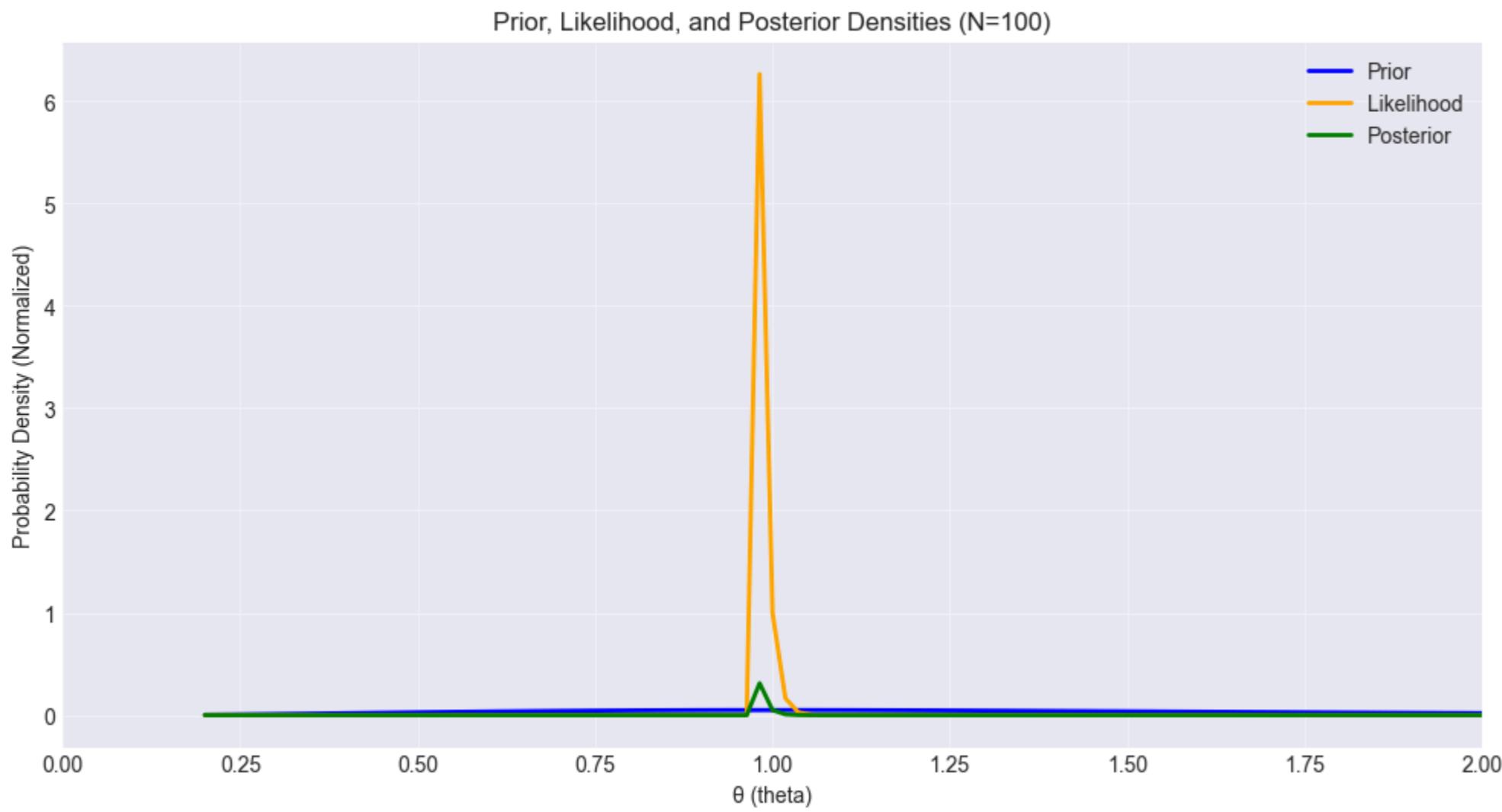
    # Posterior = Likelihood * prior
    posterior = likelihood_vals * prior_vals

    if n in selected_samples_plot:
        plot_distribution(theta_range, prior_vals, likelihood_vals, posterior, n)
```









## Question 12 (d)

We've found a theoretically conjugate **posterior** in question 11 (a).

Goal: Discuss in 3–4 sentences whether the posterior is consistent with the theoretical conjugacy behavior derived.

### Observations

#### Touching Spikes

When the sample size is small enough ( $N = 10$ ), the posterior with the highest probability density could touch the peak of the prior's distribution, but not the likelihood. In fact, the maximum likelihood spikes to a point close to 1, whereas the rest only attains a point below 0.2.

### Expectation

Theoretically, conjugacy is defined when  $posterior = likelihood * prior$ .

### Explanation

From the chosen graph where sample size  $N = 10$ , looking at the highest density points, if  $posterior = prior$ , with a positively large likelihood, the product will never be identical to posterior.

### Conclusion

Posterior is **NOT** conjugate to the theoretical derivation as well.