HW1

April 6, 2025

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[1]: import numpy as np
     import matplotlib.pyplot as plt
     import math
[2]: SEED = 42
[3]: # Initialize different RNG algorithms
     bg_mt19937=
                    np.random.MT19937(SEED)
     bg_pcg64=
                    np.random.PCG64(SEED)
                    np.random.PCG64DXSM(SEED)
     bg_pcg64dxsm=
     bg_philox=
                    np.random.Philox(SEED)
     bg_sfc64=
                    np.random.SFC64(SEED)
[4]: # Initialize different RNGs, one for each algorithm
     rng_mt19937=
                      np.random.Generator(bg_mt19937)
     rng_pcg64=
                      np.random.Generator(bg_pcg64)
     rng_pcg64dxsm=
                      np.random.Generator(bg_pcg64dxsm)
                      np.random.Generator(bg philox)
     rng philox=
     rng_sfc64=
                      np.random.Generator(bg_sfc64)
     all_rngs = [rng_mt19937, rng_pcg64, rng_pcg64dxsm, rng_philox, rng_sfc64]
[5]: # Gaussians' parameters
     gaussian_parameters = [
                             (-2, 2),
                             (4, 1),
                             (10, 3),
                             (15, 2)
                         ]
     # List of probability of drawing from each distribution
     choice_weights = np.array([0.15, 0.25, 0.35, 0.25])
[6]: \#NB: I am avoiding vectorized operations on purpose, since is not clear (to me)
      →how this could affect the rng state evolution
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DATASET_SIZE = int(1e6)

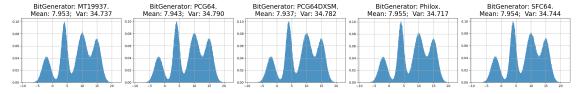
samples_from_rng = {}

for rng in all_rngs:
    samples = []

for i in range(DATASET_SIZE):
    choice = rng.choice(a=gaussian_parameters, replace=True,__
p=choice_weights)
    samples.append(rng.normal(loc=choice[0], scale=np.sqrt(choice[1])))

samples_from_rng[rng.bit_generator.__class__.__name__] = [samples]
```

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[7]: #Plotting from different rng algorithms
     n = len(samples_from_rng)
     rows = 1
     cols = n
     fig, axes = plt.subplots(rows, cols, figsize=(5 * cols, 4 * rows), sharex=True)
     axes = axes.flatten()
     for ax, (name, samples) in zip(axes, samples_from_rng.items()):
         mean = np.mean(samples)
         var = np.var(samples)
         ax.hist(samples, bins=500, density=True, alpha=0.8)
         ax.set_title(f"BitGenerator: {name}.\nMean: {mean:.3f}; Var: {var:.3f}",__
      →fontsize=20)
         ax.grid(True)
     for i in range(len(samples_from_rng), len(axes)):
         fig.delaxes(axes[i])
     plt.tight_layout()
     plt.show()
```



1 Exercise 2

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[8]: # Initialize different RNGs, one for each algorithm
      rng_mt19937=
                       np.random.Generator(bg_mt19937)
      rng_pcg64=
                      np.random.Generator(bg_pcg64)
                      np.random.Generator(bg_pcg64dxsm)
      rng_pcg64dxsm=
                      np.random.Generator(bg_philox)
      rng_philox=
      rng_sfc64=
                       np.random.Generator(bg_sfc64)
      all_rngs = [rng_mt19937, rng_pcg64, rng_pcg64dxsm, rng_philox, rng_sfc64]
 [9]: # Exponential's parameter
      exponential_mean = 1.0
      # Uniform's parameters
      uniform low = 0.0
      uniform_high = 5.0
[10]: \#DATASET\_SIZE = int(1e6)
      samples_from_rng = {}
      for rng in all_rngs:
          samples = []
          for i in range(DATASET_SIZE):
              sample = {}
              sample['exponential'] = rng.exponential(scale=exponential_mean)
              sample['uniform'] = rng.uniform(low=uniform_low, high=uniform_high)
              samples.append(sample)
          samples_from_rng[rng.bit_generator.__class__.__name__] = samples
[11]: #Plotting from different rng algorithms
      n = len(samples_from_rng)
      rows = 1
      cols = n
      fig, axes = plt.subplots(rows, cols, figsize=(5 * cols, 4 * rows), sharex=True)
      axes = axes.flatten()
```

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for ax, (name, samples) in zip(axes, samples_from_rng.items()):
    # Calculate deltas just for plotting
    deltas = [sample['exponential'] - sample['uniform'] for sample in samples]
    mean = np.mean(deltas)
    var = np.var(deltas)
    # Probability of X > Y
    positive_rate = np.sum(np.array(deltas) > 0) / len(deltas)
    ax.hist(deltas, bins=500, density=True, alpha=0.8)
    ax.set_title(f"BitGenerator: {name}.\nMean: {mean:.3f}; Var: {var:.3f};
 \neg nP(X > Y) = \{positive\_rate:.3f\}", fontsize=20)
    ax.grid(True)
    # Add a vertical red line at x = 0
    ax.axvline(x=0, color='red', linestyle='-', linewidth=1.5)
for i in range(len(samples_from_rng), len(axes)):
    fig.delaxes(axes[i])
plt.tight_layout()
plt.show()
```

