

CPSC 532W Assignment 6

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Here is a link to the repository:

https://github.com/aliseyfi75/Probabilistic-Programming/tree/master/Assignment_6

1 Code

1.1 evaluator

```
1 def evaluate(exp, env=None):
2
3     if env is None:
4         env = standard_env()
5
6     if type(exp) is list:
7         op, *args = exp
8         if op == 'sample':
9             alpha = evaluate(args[0], env=env)
10            d = evaluate(args[1], env=env)
11            s = d.sample()
12            k = evaluate(args[2], env=env)
13            sigma = {'type' : 'sample', 'alpha' : alpha, 'sample' : s}
14            return k, [s], sigma
15        elif op == 'observe':
16            alpha = evaluate(args[0], env=env)
17            d = evaluate(args[1], env=env)
18            c = evaluate(args[2], env=env)
19            k = evaluate(args[3], env=env)
20            sigma = {'type' : 'observe', 'alpha' : alpha, 'log_prob' : d.log_prob(c)}
21            return k, [c], sigma
22        elif op == 'if':
23            cond, conseq, alt = args
24            if evaluate(cond, env=env):
25                return evaluate(conseq, env=env)
26            else:
27                return evaluate(alt, env=env)
28        elif op == 'fn':
29            params, body = args #fn is: ['fn', ['arg1','arg2','arg3'], body_exp]
30            return Procedure(params, body, env)
31        else: #func eval
32            proc = evaluate(op, env=env)
33            values = [evaluate(e, env=env) for e in args]
34            sigma = {'type' : 'proc'}
35            return proc, values, sigma
36    elif type(exp) is str:
37        if exp[0] == "\": # strings have double, double quotes
38            return exp[1:-1]
39        if exp[0:4] == 'addr':
40            return exp[4:]
41        lowest_env = env.find(exp)
42        return lowest_env[exp]
43    elif type(exp) is float or type(exp) is int or type(exp) is bool:
44        return torch.tensor(exp)
45    else:
46        raise ValueError('Expression type unkown')
```

Listing 1: evaluator.py - evaluate

1.2 SMC

```
1 particles = []
2 weights = []
3 logZs = []
4 output = lambda x: x
5
6 for i in range(n_particles):
7     res = evaluate(exp, env=None)('addr_start', output)
8     logW = 0.
9     particles.append(res)
10    weights.append(logW)
11
12 done = False
13 smc_cnter = 0
14 while not done:
15     new_address = ''
16     print('In SMC step {}, Zs: {}'.format(smc_cnter, logZs))
17     for i in range(n_particles):
18         res = run_until_observe_or_end(particles[i])
19         if 'done' in res[2]:
20             particles[i] = res[0]
21             if i == 0:
22                 done = True
23                 address = ''
24             else:
25                 if not done:
26                     raise RuntimeError('Failed SMC, finished one calculation before the
other')
27             else:
28                 if i == 0:
29                     new_address = res[2]['alpha']
30                 else:
31                     address = res[2]['alpha']
32                     if address != new_address:
33                         raise RuntimeError('Failed SMC, address changed')
34
35             log_prob = res[2]['log_prob']
36             weights[i] = weights[i] + log_prob
37             particles[i] = res
38
39     if not done:
40         logZn, particles = resample_particles(particles, weights)
41         logZs.append(logZn)
42         weights = [0.] * n_particles
43     smc_cnter += 1
44 logZ = sum(logZs)
45 return logZ, particles
```

Listing 2: smc.py - SMC

```
1 def resample_particles(particles, log_weights):
2     paricles_length = len(particles)
3     weights = torch.exp(torch.FloatTensor(log_weights)) # convert to weights
4     normalized_weights = weights + 1e-10 # add a small number to avoid zero weights
5     normalized_weights = normalized_weights / normalized_weights.sum() # normalize weights
6
7     logZ = torch.log(torch.mean(weights)) # calculate logZ
8
9     indices = torch.multinomial(normalized_weights, paricles_length, replacement=True)
10    new_particles = [particles[i] for i in indices]
11
12    return logZ, new_particles
```

Listing 3: smc.py - resample_particles

2 Results

2.1 Task 1

2.1.1 Number of particles: 1

2.1.1.1 Posterior expectation

Posterior expected value of until success is: **22**

2.1.1.2 Histogram

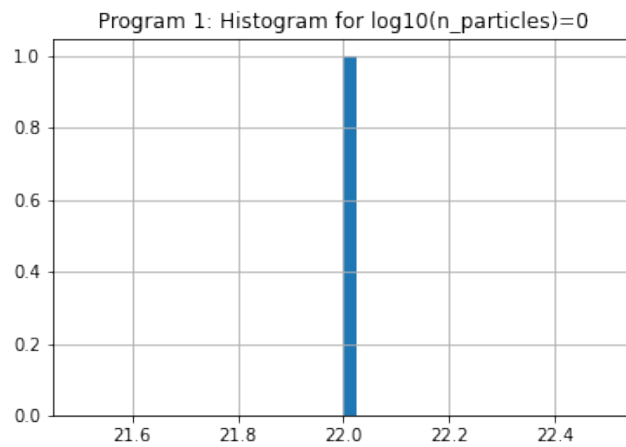


Figure 1: Histogram of posterior distribution of until success

2.1.1.3 marginal evidence estimate

marginal evidence estimate of until success is: **NA**

2.1.2 Number of particles: 10

2.1.2.1 Posterior expectation

Posterior expected value of until success is: **73.7**

2.1.2.2 Histogram

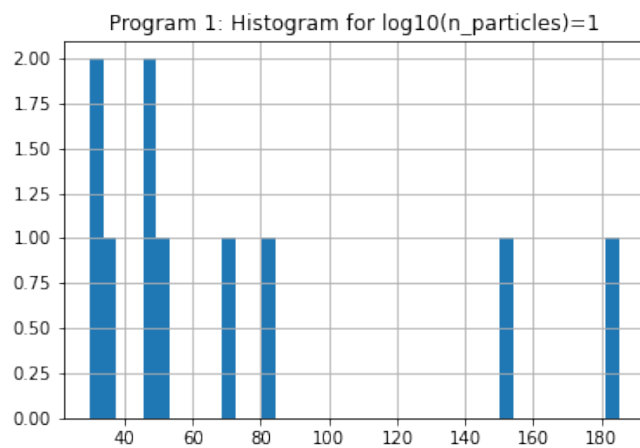


Figure 2: Histogram of posterior distribution of until success

2.1.2.3 marginal evidence estimate

marginal evidence estimate of until success is: **NA**

2.1.3 Number of particles: 10^2

2.1.3.1 Posterior expectation

Posterior expected value of until success is: **97.35**

2.1.3.2 Histogram

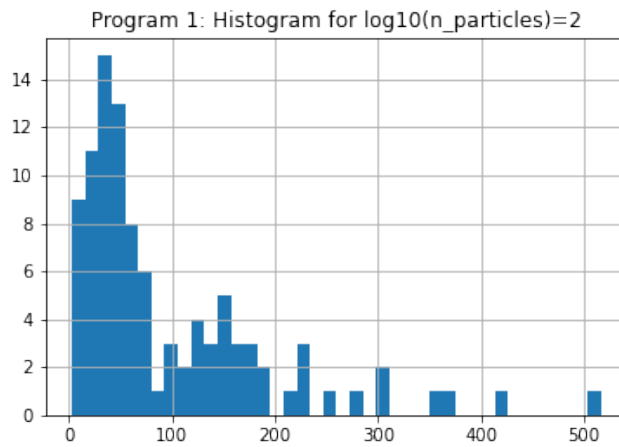


Figure 3: Histogram of posterior distribution of until success

2.1.3.3 marginal evidence estimate

marginal evidence estimate of until success is: **NA**

2.1.4 Number of particles: 10^3

2.1.4.1 Posterior expectation

Posterior expected value of until success is: **97.474**

2.1.4.2 Histogram

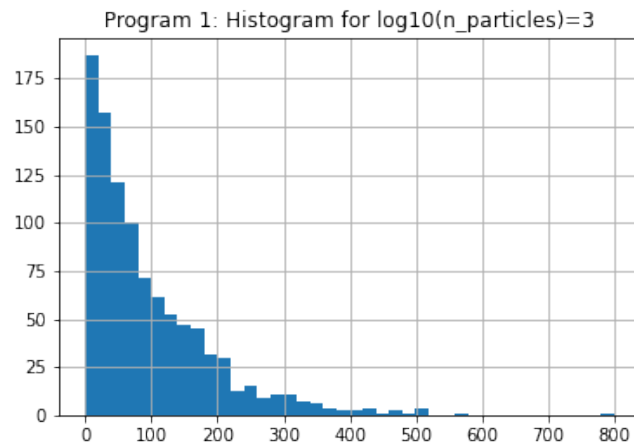


Figure 4: Histogram of posterior distribution of until success

2.1.4.3 marginal evidence estimate

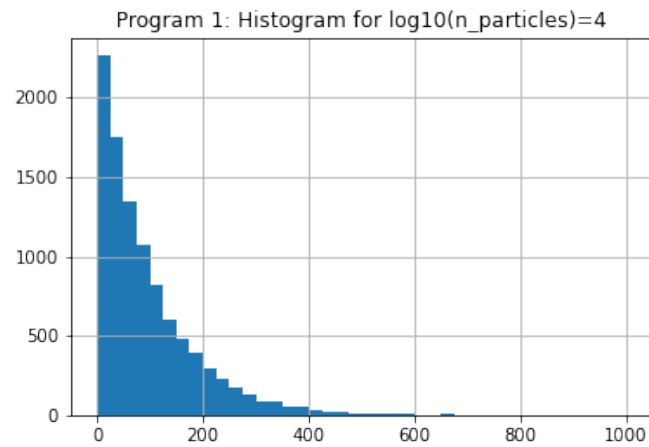
marginal evidence estimate of until success is: **NA**

2.1.5 Number of particles: 10^4

2.1.5.1 Posterior expectation

Posterior expected value of until success is: **96.4183**

2.1.5.2 Histogram



2.1.6 Number of particles: 10^5

2.1.6.1 Posterior expectation

Posterior expected value of until success is: **98.9192**

2.1.6.2 Histogram

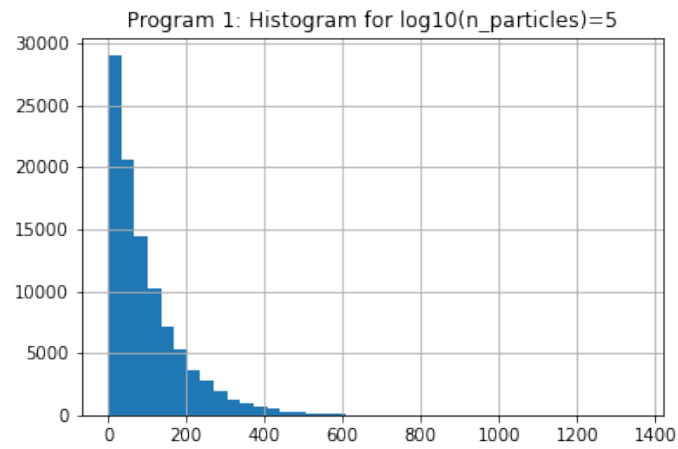


Figure 6: Histogram of posterior distribution of until success

2.1.6.3 marginal evidence estimate

marginal evidence estimate of until success is: **NA**

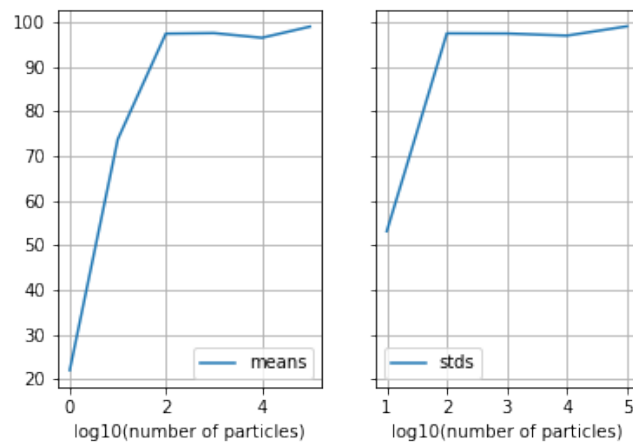


Figure 7: Marginal mean and std of until success for different number of particles

2.2 Task 2

2.2.1 Number of particles: 1

2.2.1.1 Posterior expectation

Posterior expected value of mu is: **0.8076**

2.2.1.2 Histogram

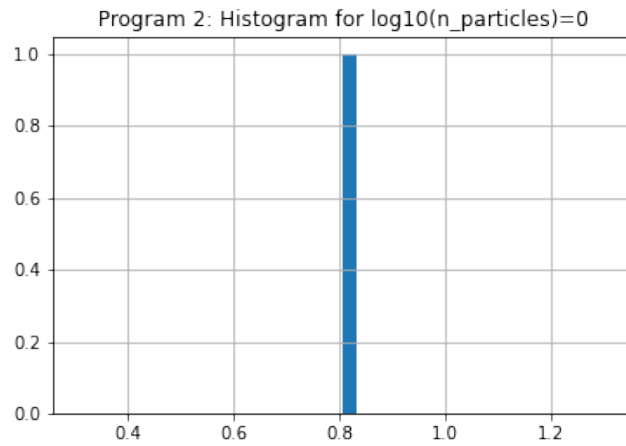


Figure 8: Histogram of posterior distribution of mu

2.2.1.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-35.7038**

marginal evidence estimate of mu is: $3.1191 * 10^{-16}$

2.2.2 Number of particles: 10

2.2.2.1 Posterior expectation

Posterior expected value of mu is: **4.9418**

2.2.2.2 Histogram

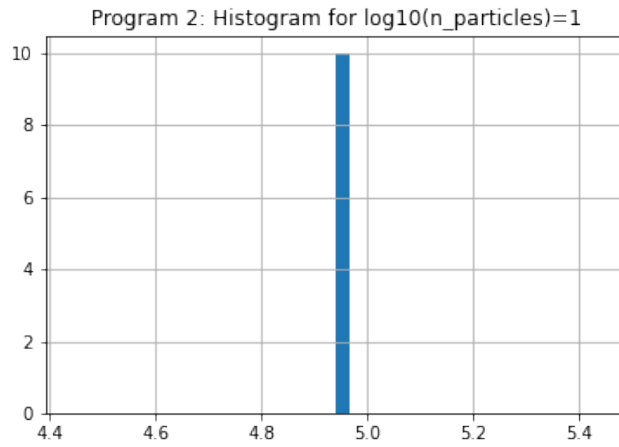


Figure 9: Histogram of posterior distribution of mu

2.2.2.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-15.9816**

marginal evidence estimate of mu is: $1.1462 * 10^{-7}$

2.2.3 Number of particles: 10^2

2.2.3.1 Posterior expectation

Posterior expected value of mu is: **4.3897**

2.2.3.2 Histogram

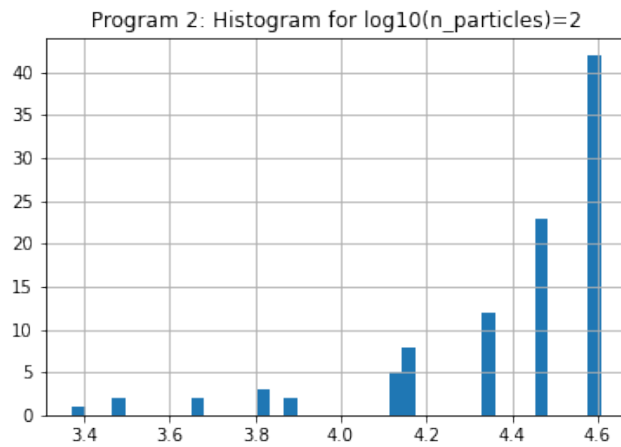


Figure 10: Histogram of posterior distribution of mu

2.2.3.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-9.0384**

marginal evidence estimate of mu is: **0.0001187**

2.2.4 Number of particles: 10^3

2.2.4.1 Posterior expectation

Posterior expected value of mu is: **7.5386**

2.2.4.2 Histogram

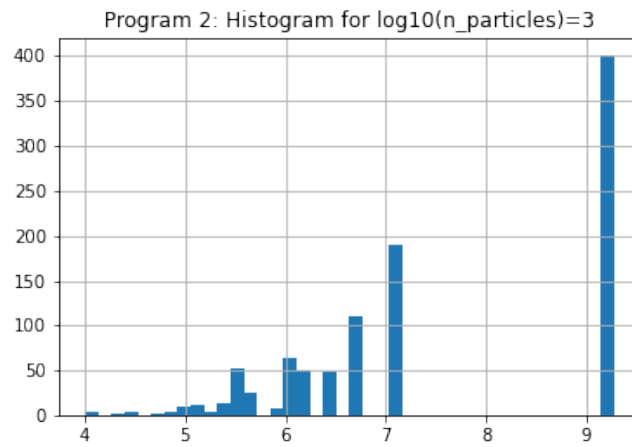


Figure 11: Histogram of posterior distribution of mu

2.2.4.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-7.9729**

marginal evidence estimate of mu is: **0.000344**

2.2.5 Number of particles: 10^4

2.2.5.1 Posterior expectation

Posterior expected value of mu is: **7.2384**

2.2.5.2 Histogram

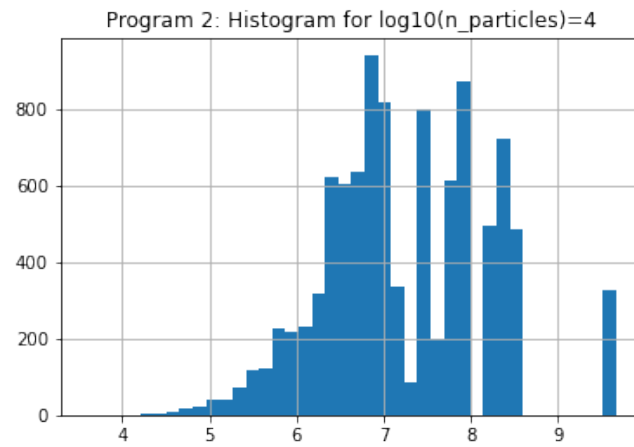


Figure 12: Histogram of posterior distribution of mu

2.2.5.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-8.23083**

marginal evidence estimate of mu is: **0.00027**

2.2.6 Number of particles: 10^5

2.2.6.1 Posterior expectation

Posterior expected value of mu is: **7.2103**

2.2.6.2 Histogram

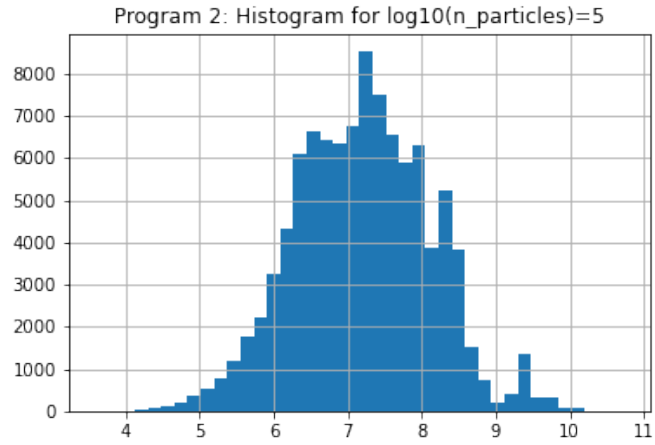


Figure 13: Histogram of posterior distribution of mu

2.2.6.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-8.23384**

marginal evidence estimate of mu is: **0.000266**

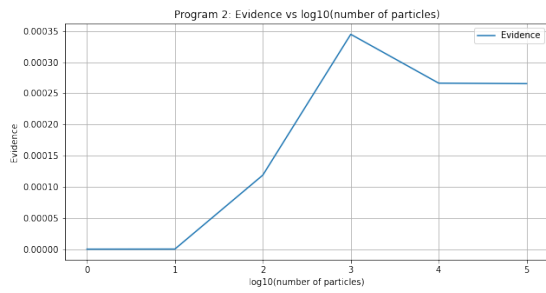


Figure 14: Marginal evidence estimate of mu for different number of particles

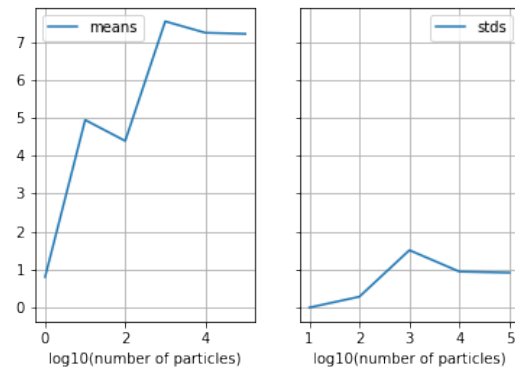


Figure 15: Marginal mean and std of mu for different number of particles

2.3 Task 3

2.3.1 Number of particles: 1

2.3.1.1 Posterior expectation

Posterior expected value of states in each step is: [2., 1., 0., 1., 0., 2., 1., 0., 2., 2., 2., 2., 2., 1., 2., 1., 0.]

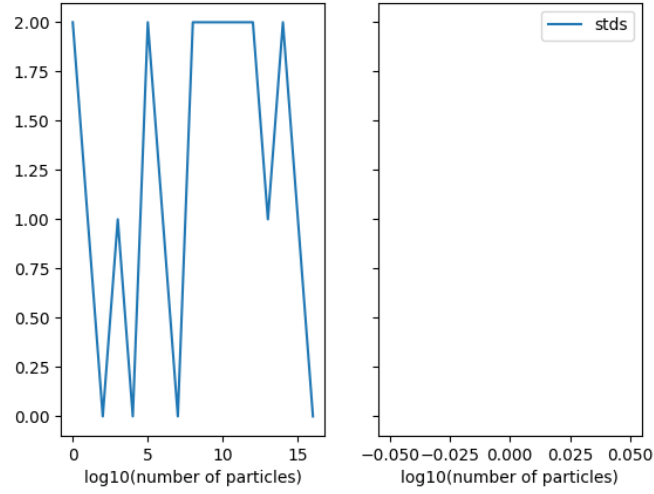


Figure 16: Marginal mean and std of states in each step

2.3.1.2 Histogram

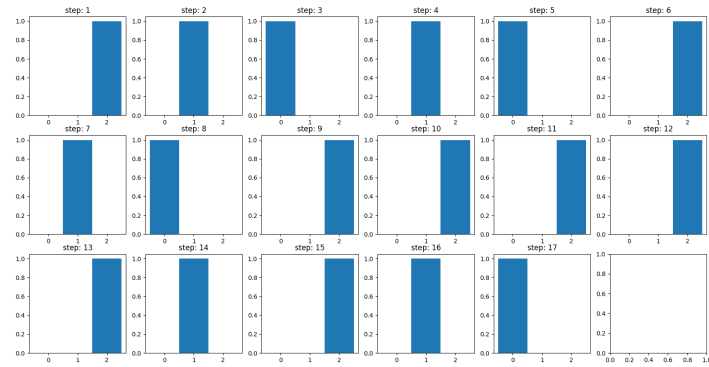


Figure 17: Histogram of posterior distribution of states in each step

2.3.1.3 marginal evidence estimate

marginal log evidence estimate of states in each step is: **-58.723**

marginal evidence estimate of states in each step is: $3.1398 * 10^{-26}$

2.3.2 Number of particles: 10

2.3.2.1 Posterior expectation

Posterior expected value of states in each step is: [1.4, 1.2, 2.0, 2.0, 1.0, 0.2, 1.0, 0.0, 2.0, 2.0, 0.0, 2.0, 2.0, 2.0, 1.0, 0.0, 1.0]

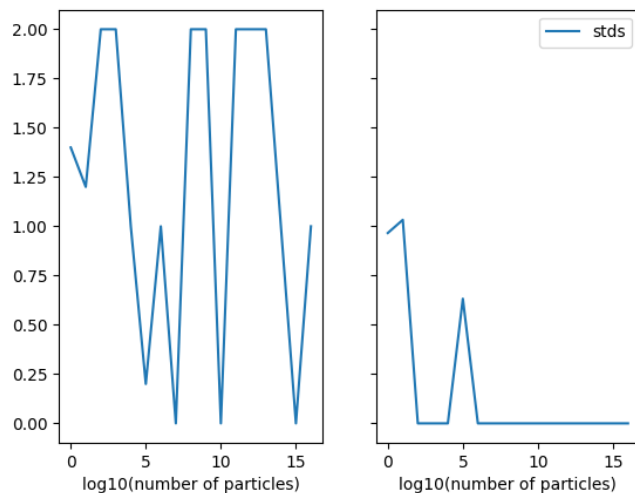


Figure 18: Marginal mean and std of states in each step

2.3.2.2 Histogram

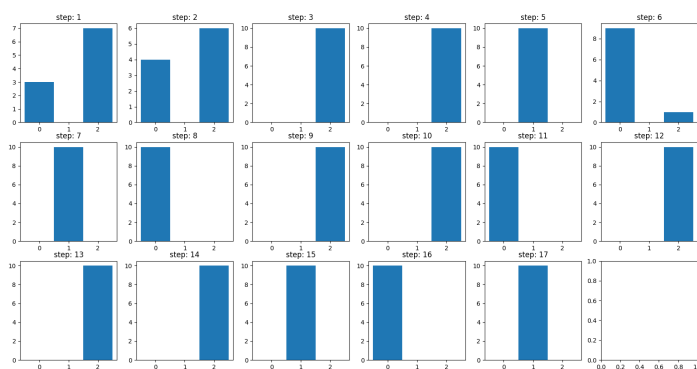


Figure 19: Histogram of posterior distribution of states in each step

2.3.2.3 marginal evidence estimate

marginal log evidence estimate of states in each step is: **-43.748**

marginal evidence estimate of states in each step is: 1.0011^{-19}

2.3.3 Number of particles: 10^2

2.3.3.1 Posterior expectation

Posterior expected value of states in each step is: [1.35, 1.35, 1.66, 1.71, 1.00, 1.66, 1.85, 1.82, 1.17, 0.36, 0.00, 1.42, 1.76, 1.82, 1.28, 1.66, 0.90]

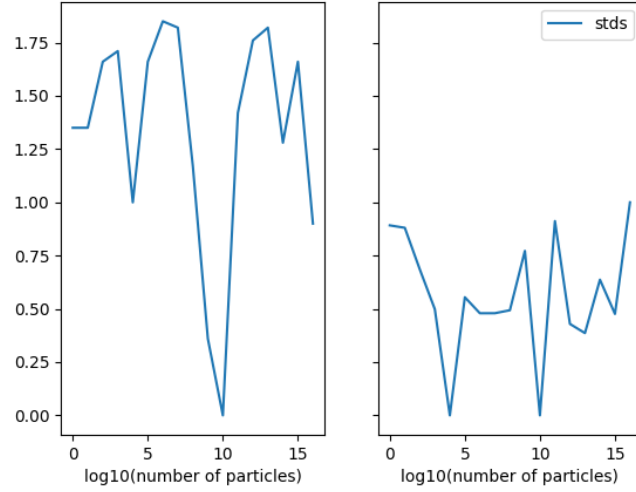


Figure 20: Marginal mean and std of states in each step

2.3.3.2 Histogram

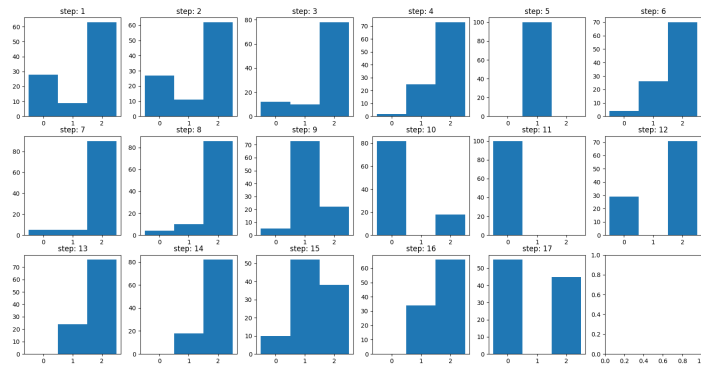


Figure 21: Histogram of posterior distribution of states in each step

2.3.3.3 marginal evidence estimate

marginal log evidence estimate of states in each step is: **-43.931**
marginal evidence estimate of states in each step is: 8.3384×10^{-20}

2.3.4 Number of particles: 10^3

2.3.4.1 Posterior expectation

Posterior expected value of states in each step is: [1.427, 1.617, 1.551, 1.614, 1.021, 1.426, 1.752, 1.749, 1.587, 1.151, 0.170, 1.774, 1.608, 1.723, 1.675, 1.505, 1.178]

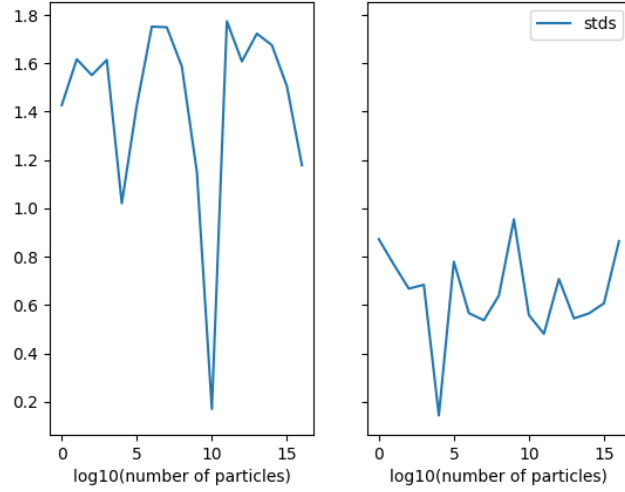


Figure 22: Marginal mean and std of states in each step

2.3.4.2 Histogram

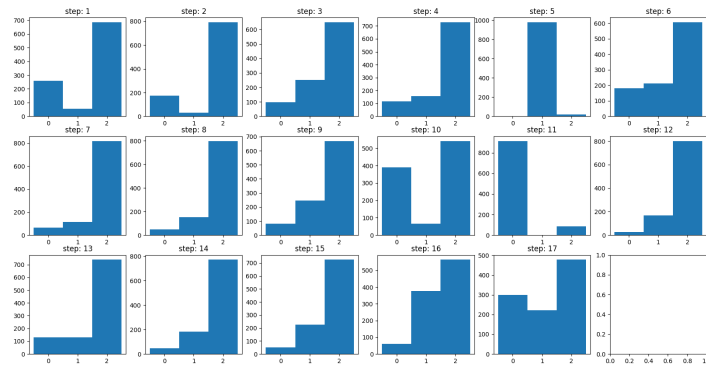


Figure 23: Histogram of posterior distribution of states in each step

2.3.4.3 marginal evidence estimate

marginal log evidence estimate of states in each step is: **-44.359**

marginal evidence estimate of states in each step is: 5.4340×10^{-20}

2.3.5 Number of particles: 10^4

2.3.5.1 Posterior expectation

Posterior expected value of states in each step is: [1.4412, 1.5189, 1.6974, 1.6059, 1.0146, 1.3829, 1.6333, 1.6718, 1.6207, 1.0641, 0.1592, 1.7524, 1.6479, 1.7258, 1.6390, 1.5436, 0.9160]

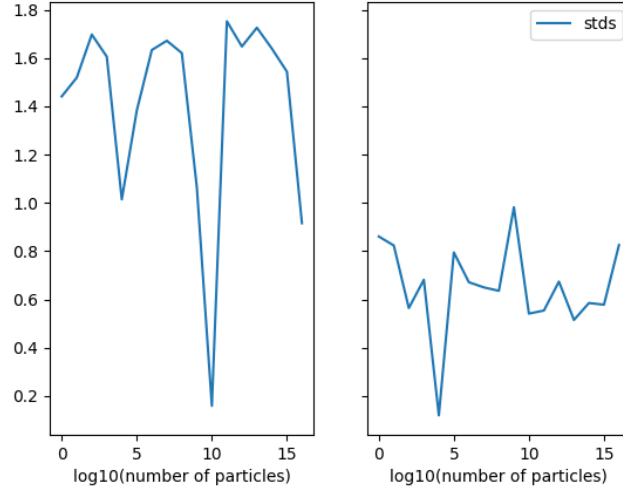


Figure 24: Marginal mean and std of states in each step

2.3.5.2 Histogram

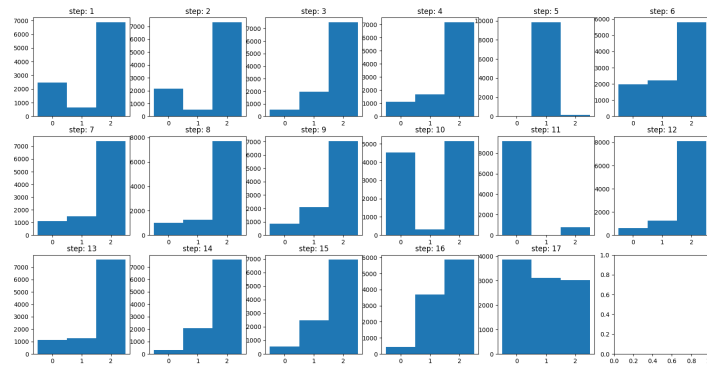


Figure 25: Histogram of posterior distribution of states in each step

2.3.5.3 marginal evidence estimate

marginal log evidence estimate of states in each step is: **-44.404**

marginal evidence estimate of states in each step is: 5.1935×10^{-20}

2.3.6 Number of particles: 10^5

2.3.6.1 Posterior expectation

Posterior expected value of states in each step is: [1.4314, 1.5516, 1.7008, 1.6037, 1.0174, 1.4171, 1.6541, 1.6618, 1.6076, 1.0369, 0.1375, 1.6632, 1.6574, 1.6778, 1.6381, 1.4908, 0.9158]

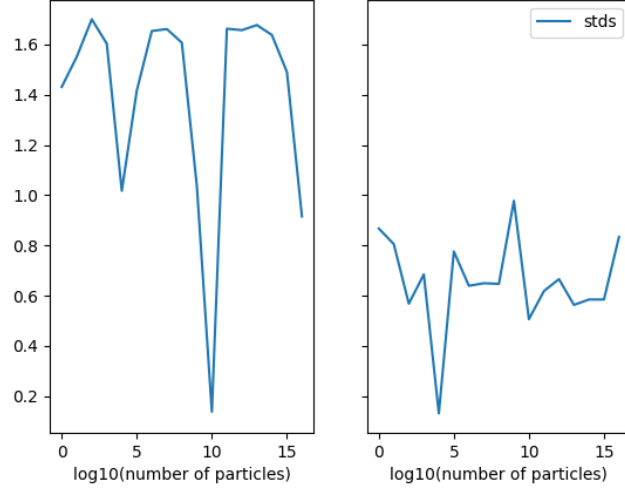


Figure 26: Marginal mean and std of states in each step

2.3.6.2 Histogram

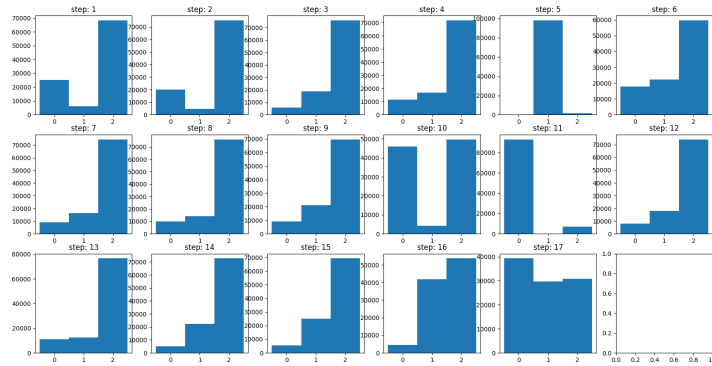


Figure 27: Histogram of posterior distribution of states in each step

2.3.6.3 marginal evidence estimate

marginal log evidence estimate of states in each step is: **-44.407**
marginal evidence estimate of states in each step is: 5.1775×10^{-20}

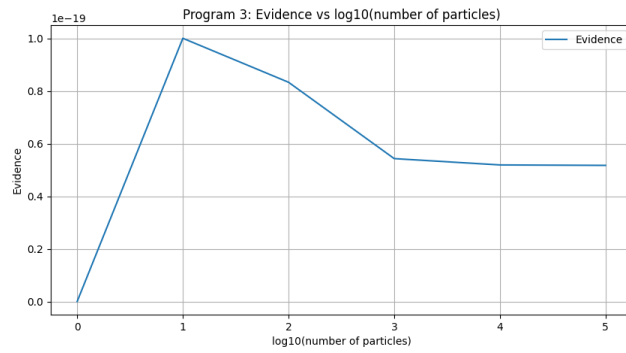


Figure 28: Marginal evidence estimate of μ for different number of particles

2.4 Task 4

2.4.1 Number of particles: 1

2.4.1.1 Posterior expectation

Posterior expected value of mu is: **2.8495**

2.4.1.2 Histogram

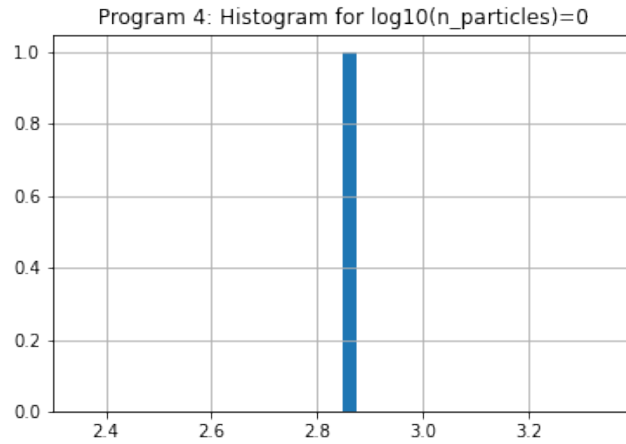


Figure 29: Histogram of posterior distribution of mu

2.4.1.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-35.165577**

marginal evidence estimate of mu is: $5.3429 * 10^{-16}$

2.4.2 Number of particles: 10

2.4.2.1 Posterior expectation

Posterior expected value of mu is: **3.8222**

2.4.2.2 Histogram

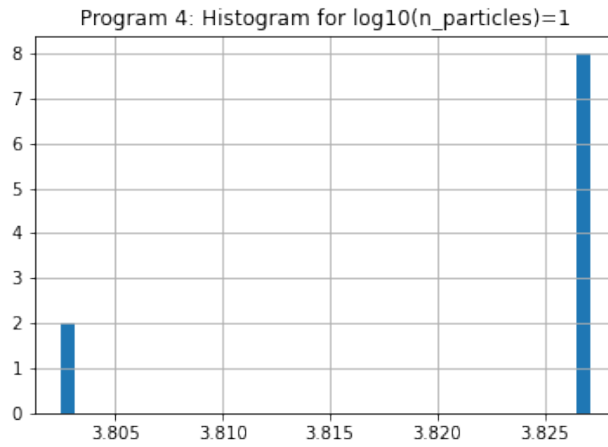


Figure 30: Histogram of posterior distribution of mu

2.4.2.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-15.06296**

marginal evidence estimate of mu is: $2.8724 * 10^{-7}$

2.4.3 Number of particles: 10^2

2.4.3.1 Posterior expectation

Posterior expected value of mu is: **7.3450**

2.4.3.2 Histogram

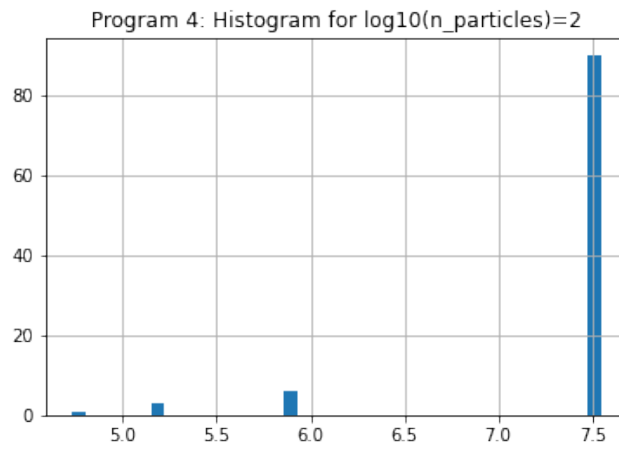


Figure 31: Histogram of posterior distribution of mu

2.4.3.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-7.62482**

marginal evidence estimate of mu is: **0.000488**

2.4.4 Number of particles: 10^3

2.4.4.1 Posterior expectation

Posterior expected value of mu is: **7.1552**

2.4.4.2 Histogram

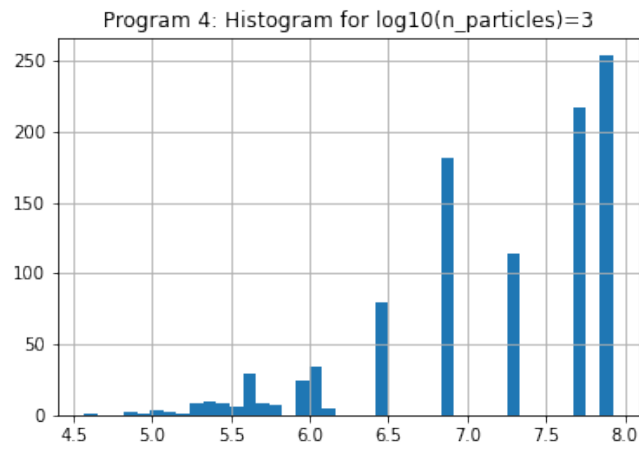


Figure 32: Histogram of posterior distribution of mu

2.4.4.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-7.9910**

marginal evidence estimate of mu is: **0.00034**

2.4.5 Number of particles: 10^4

2.4.5.1 Posterior expectation

Posterior expected value of mu is: **7.1179**

2.4.5.2 Histogram

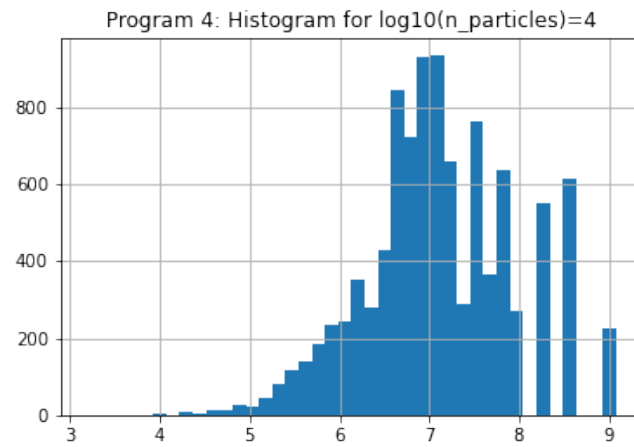


Figure 33: Histogram of posterior distribution of mu

2.4.5.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-8.40138**

marginal evidence estimate of mu is: **0.00023**

2.4.6 Number of particles: 10^5

2.4.6.1 Posterior expectation

Posterior expected value of mu is: **7.2363**

2.4.6.2 Histogram

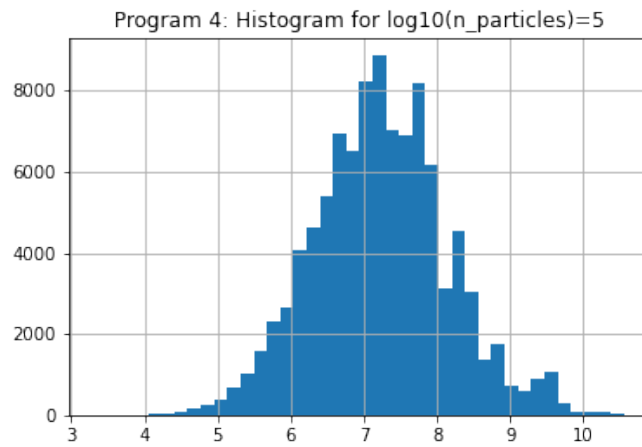


Figure 34: Histogram of posterior distribution of mu

2.4.6.3 marginal evidence estimate

marginal log evidence estimate of mu is: **-8.22765**

marginal evidence estimate of mu is: **0.00027**

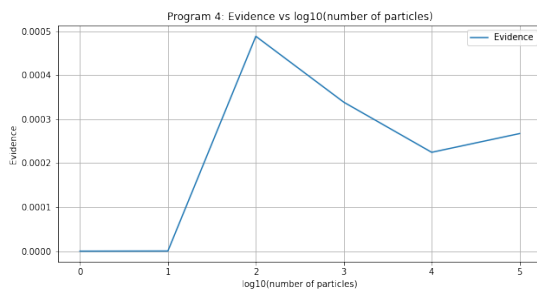


Figure 35: Marginal evidence estimate of mu for different number of particles

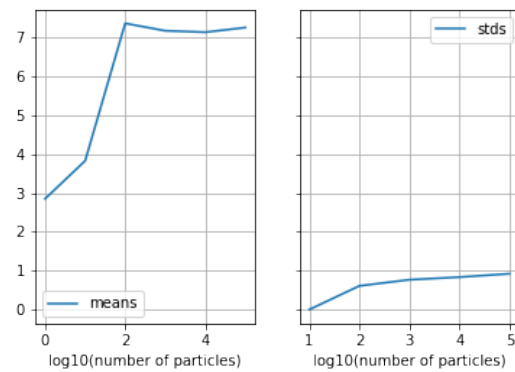


Figure 36: Marginal mean and std of mu for different number of particles