

```
In [1]: import torch
import numpy as np
import os, json
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

from evaluation_based_sampling import evaluate, evaluate_program
from daphne import daphne
```

## Problem 4

```

11 [2]: from load_helper import ast_helper, graph_helper

```

## Importance sampling

- 10k samples in 6.82s implies 879k samples in 10 min

```

In [5]: import parse
import importance_sampling

```

```

In [4]: fname = '4aphne'
         ast = ast_helper(fname)
         ast

Out[4]: [['let',
          ['sprinkler', True],
          ['let',
           ['wet-grass', True],
           ['let',
            ['is-cloudy', ['sample', ['flip', 0.5]]],
            ['let',
             ['is-raining',
              ['if',
               ['=', 'is-cloudy', True],
               ['sample', ['flip', 0.8]],
               ['sample', ['flip', 0.2]]]]],
            ['let',
             ['sprinkler-dist',
              ['if', ['=', 'is-cloudy', True], ['flip', 0.1], ['flip', 0.5]]],
            ['let',
             ['wet-grass-dist',
              ['if',
               ['and', ['=', 'sprinkler', True], ['=', 'is-raining', True]],
               ['flip', 0.99],
               ['if',
                ['and', ['=', 'sprinkler', False], ['=', 'is-raining', False]],
                ['flip', 0.0],
                ['if',
                 ['or', ['=', 'sprinkler', True], ['=', 'is-raining', True]],
                 ['flip', 0.9],
                 None]]],
             ['let',
              ['dontcare0', ['observe', 'sprinkler-dist', 'sprinkler']],
              ['let',
               ['dontcare1', ['observe', 'wet-grass-dist', 'wet-grass']],

```

```
In [26]: %%time
num_samples=879000
samples,sigma = parse.take_samples(num_samples,ast=ast)

CPU times: user 10min 22s, sys: 2.84 s, total: 10min 25s
Wall time: 10min 27s
```

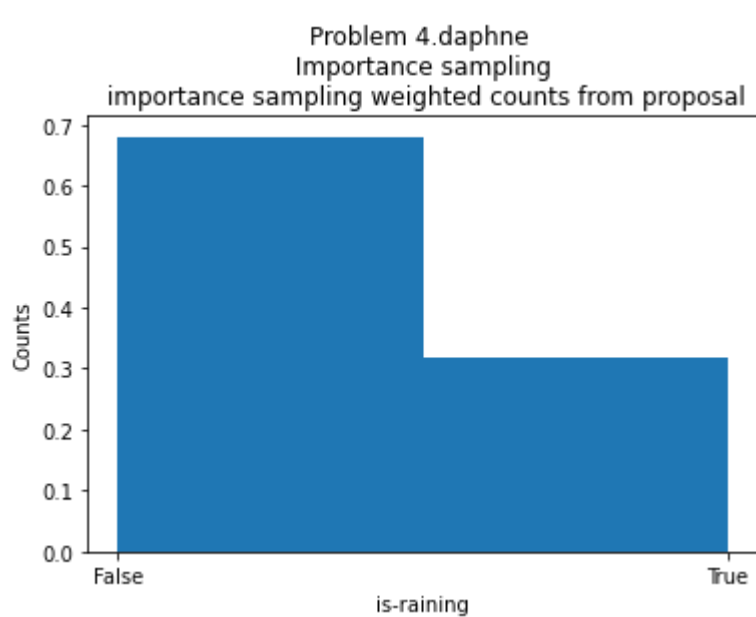
```
In [27]: samples = np.array([sample.item() for sample in samples])
```

```
In [28]: posterior_mean, probs = importance_sampling.weighted_average(samples, sigmas)
         posterior_mean
```

```
Out[28]: 0.31957416455608195
```

```
In [29]: _ = plt.hist(samples.astype(int), weights=probs, bins=2)
plt.xlabel('is-raining')
plt.title('Problem {} \n Importance sampling \n importance sampling weighted counts from proposal'.format(fname))
plt.xticks([0,1],["False","True"])
plt.ylabel('Counts')
```

```
Out[29]: Text(0, 0.5, 'Counts')
```



```
In [30]: """The posterior probability that it is raining, i.e. of "is-raining.": {:.3f}""".format(posterior_mean)
```

```
Out[30]: 'The posterior probability that it is raining, i.e. of "is-raining.": 0.320'
```

MH Gibbs

2k samples in 8.55s implies 140k samples in 10 min

```
In [10]: import graph_based_sampling
import mh_gibbs
from hmc import compute_log_joint_prob
```

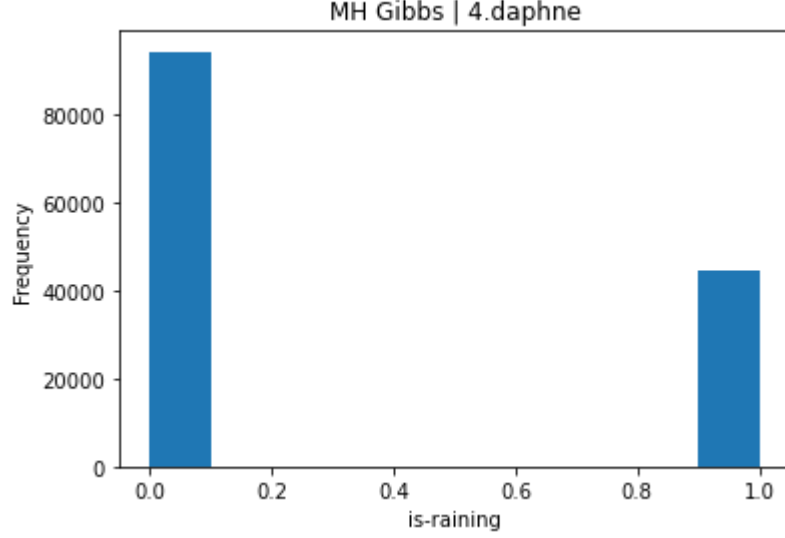
```
fname = '4.daphne'
graph = graph_helper(fname)
graph
```

```
Out[11]: [{],
  {'V': ['observe6', 'sample4', 'sample2', 'sample5', 'sample3'],
   'A': {'sample2': ['observe6', 'observe5'],
        'sample4': ['observe6'],
        'sample3': ['observe6']},
   'P': {'sample2': ['sample*', ['flip', 0.5]],
        'sample3': ['sample*', ['flip', 0.8]],
        'sample4': ['sample*', ['flip', 0.2]],
        'observe5': ['observe*',
                     ['if', ['=', 'sample2', True], ['flip', 0.1], ['flip', 0.5]],
                     True],
        'observe6': ['observe*',
                     ['if',
                      ['and',
                       True,
                       ['=', ['if', ['=', 'sample2', True], 'sample3', 'sample4'], True]],
                      ['flip', 0.99]],
                      ['if',
                       ['and',
                        False,
                        ['=', ['if', ['=', 'sample2', True], 'sample3', 'sample4'], False]],
                       ['flip', 0.0]],
                      ['if',
                       ['or',
                        True,
                        ['=', ['if', ['=', 'sample2', True], 'sample3', 'sample4'], True]],
                        ['flip', 0.9]],
                        None]],
                      True]],
        'Y': {'observe5': True, 'observe6': True}},
   ['if', ['=', 'sample2', True], 'sample3', 'sample4']]
```

```
num_steps=140000
return_list, samples_whole_graph = mh_gibbs.mh_gibbs_wrapper(graph,num_steps)
```

```
In [13]: samples = np.array([sample.item() for sample in return_list])
```

```
In [14]: burn_in = int(0.01*num_steps)
pd.Series(samples[burn_in:]).astype(float).plot.hist()
plt.xlabel('is-raining')
plt.title('MH Gibbs | {}',format(fname))
```



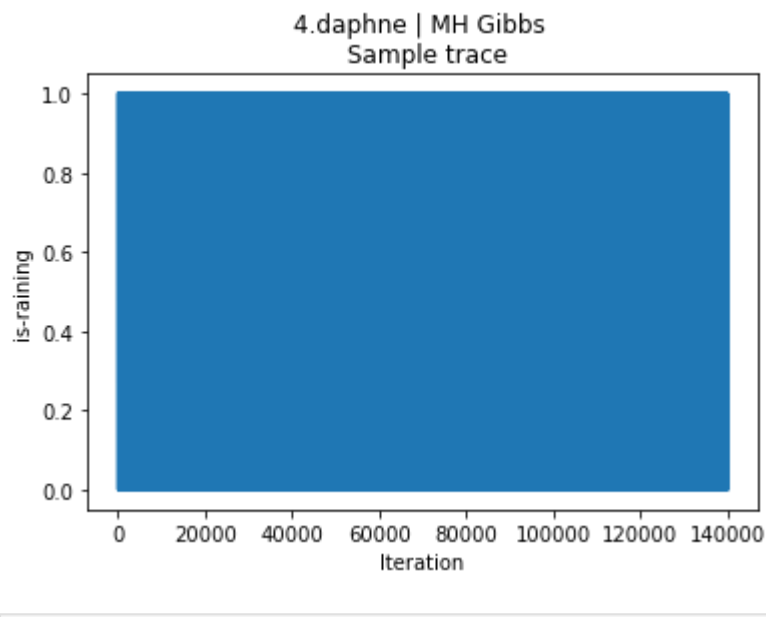
```
In [15]: posterior_mean = samples[burn_in:].mean(0)
```

```
In [16]: """The posterior probability that it is raining, i.e. of "is-raining.": {:.3f}""".format(posterior_mean)
```

```
Out[16]: 'The posterior probability that it is raining, i.e. of "is-raining.": 0.321'
```

```
pd.Series(samples).astype(int).plot()
plt.xlabel('Iteration')
plt.ylabel('is-raining')
plt.title('{} | MH Gibbs \n Sample trace'.format(fname))
```

```
Out[17]: Text(0.5, 1.0, 4.daphne | An Gibbs (N Sample trace)
```



```
Y = G['Y']
Y = {key:evaluate([value])[0] for key,value in Y.items()}
P = G['P']
```

```
jll = np.zeros(size)
for idx in range(size):
    jll[idx] = compute_log_joint_prob(samples_whole_graph[idx], Y, P)
```

```
plt.Series(jll).plot()
plt.xlabel('Iteration (t)')
plt.ylabel('r' '$-\log p(X=x_t, Y=y_t)$')
plt.title('{} | MH Gibbs \n Joint density'.format(fname))
```

```
Out[34]: Text(0.5, 1.0, '4.daphnia | Mr Gibbs (n=60000 density)')
```

