

# problem\_6

April 9, 2020

## 1 Problem 6, MCMC Sampling

### 1.1 Problem a)

```
[132]: import numpy as np
np.random.seed(0)

[24]: def gen_sample():
    r = np.random.rand(1)
    if r > 0.5:
        return np.random.normal(5, 1)
    else:
        return np.random.normal(-5, 1)

def gen_samples(n):
    samples = np.array([gen_sample() for i in range(n)])
    return samples

[27]: num_x_samples = 100
x_samples = gen_samples(num_x_samples)
```

### 1.2 Problem b)

```
[34]: def cal_acceptance_rate(x_samples, mu1, mu2, mu1_h, mu2_h, sigma):
    alpha = cal_log_posterior(x_samples, mu1_h, mu2_h) -
    ↪cal_log_posterior(x_samples, mu1, mu2)
    alpha = np.exp(alpha)

    # We do not need to calculate the transition probability,
    # since it is 1 always in this case.
    transition_prob = 1
    alpha = alpha * 1

    return alpha

def cal_log_posterior(x_samples, mu1, mu2):
    # log p(mu1, mu2 | x)
```

```

log_prob_mus = -1 * (mu1 ** 2 + mu2 ** 2) / 200

# p(x | mu1, mu2)
prob_x_1 = 0.5 * np.exp(-0.5 * (x_samples - mu1) ** 2)
prob_x_2 = 0.5 * np.exp(-0.5 * (x_samples - mu2) ** 2)
prob_x = prob_x_1 + prob_x_2

return np.sum(np.log(prob_x)) + log_prob_mus

def metropolis_hastings_sampling(x_samples, num_samples, sigma):
    # Initialize mu1 and mu2 be 0 and 0
    mu1, mu2 = 0, 0
    mu1s, mu2s = np.zeros(num_samples), np.zeros(num_samples)
    accept_count = 0

    # Run Metropolis Hastings Sampling
    for n in range(num_samples):
        mu1_h = np.random.normal(mu1, sigma)
        mu2_h = np.random.normal(mu2, sigma)
        alpha = max(0, min(1, cal_acceptance_rate(x_samples, mu1, mu2, mu1_h,
→mu2_h, sigma)))
        if np.random.rand(1) < alpha:
            mu1 = mu1_h
            mu2 = mu2_h
            accept_count += 1
        mu1s[n] = mu1
        mu2s[n] = mu2

    acceptance_rate = accept_count / num_samples
    return mu1s, mu2s, acceptance_rate

```

```

[74]: import matplotlib.pyplot as plt

num_samples = 11000
mean_mu1s = {}
mean_mu2s = {}
acc_rates = {}

for sigma in [0.5, 5]:
    print('sigma = {}'.format(sigma))

    mean_mu1s[sigma] = []
    mean_mu2s[sigma] = []
    acc_rates[sigma] = []

    plt.figure(figsize=(30, 10))
    for i in range(6):

```

```

# Run Metropolis Hastings Sampling
mu1s, mu2s, acceptance_rate = metropolis_hastings_sampling(
    x_samples,
    num_samples,
    sigma
)

# Discard the first 10000 samples
mu1s = mu1s[10000:]
mu2s = mu2s[10000:]

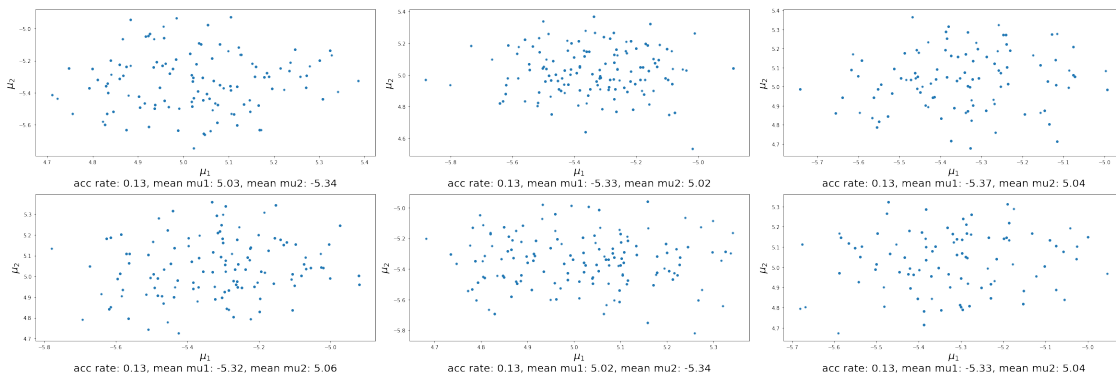
# Plot the next 1000 samples of mu1 and mu2
plt.subplot(231 + i)
plt.scatter(mu1s, mu2s, s=10)
plt.xlabel(r'$\mu_1$' + '\nacc rate: %.2lf, mean mu1: %.2lf, mean mu2: %'
→%.2lf' % (acceptance_rate, np.mean(mu1s), np.mean(mu2s)), fontsize=20)
plt.ylabel(r'$\mu_2$', fontsize=20)
plt.tight_layout()

# Record the data
mean_mu1s[sigma].append(np.mean(mu1s))
mean_mu2s[sigma].append(np.mean(mu2s))
acc_rates[sigma].append(acceptance_rate)

plt.show()

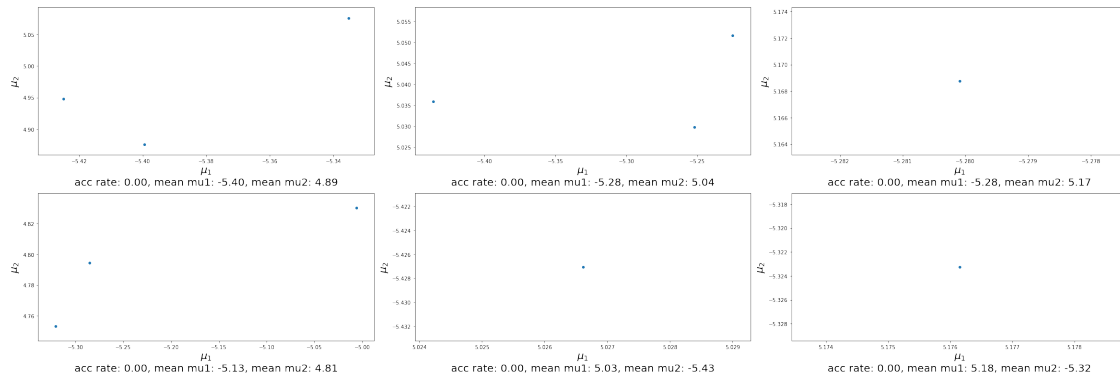
```

sigma = 0.5



sigma = 5

/Users/haekyu/anaconda3/envs/massif/lib/python3.7/site-packages/ipykernel\_launcher.py:3: RuntimeWarning: overflow encountered in exp  
This is separate from the ipykernel package so we can avoid doing imports until



```
[77]: import pandas as pd

sigma = 0.5
df_1 = pd.DataFrame()
df_1['mean_mu1'] = mean_mu1s[sigma]
df_1['mean_mu2'] = mean_mu2s[sigma]
df_1['acceptance rate'] = acc_rates[sigma]

df_1
```

```
[77]:   mean_mu1  mean_mu2  acceptance rate
0   5.025064 -5.340499         0.131000
1  -5.332442  5.021185         0.131091
2  -5.365892  5.035718         0.130727
3  -5.316841  5.056698         0.127909
4   5.019107 -5.335424         0.134545
5  -5.333710  5.039708         0.127909
```

```
[79]: sigma = 5
df_2 = pd.DataFrame()
df_2['mean_mu1'] = mean_mu1s[sigma]
df_2['mean_mu2'] = mean_mu2s[sigma]
df_2['acceptance rate'] = acc_rates[sigma]

df_2
```

```
[79]:   mean_mu1  mean_mu2  acceptance rate
0  -5.401915  4.894947         0.002273
1  -5.277091  5.037582         0.001455
2  -5.280091  5.168751         0.002182
3  -5.128976  4.807459         0.003273
4   5.026610 -5.427057         0.001909
5   5.176153 -5.323255         0.002182
```

### 1.3 Problem c)

```
[156]: def get_z_samples(x_samples, mu1, mu2):
        z = np.zeros(len(x_samples), dtype=np.int8)

        prob = np.exp(-0.5 * (x_samples - mu1) ** 2) / (np.exp(-0.5 * (x_samples - mu1) ** 2) + np.exp(-0.5 * (x_samples - mu2) ** 2))
        us = np.random.rand(len(z))

        # Mark that it represents data from mu1
        z[us < prob] = 1

        # Mark that it represents data from mu2
        z[us >= prob] = 2

        return z

def gibbs_sampling(x_samples, num_samples):
    # Initialize Gibbs sampling
    mu1, mu2 = 0, 0
    mu1s, mu2s = np.zeros(num_samples), np.zeros(num_samples)

    # Run Gibbs Sampling
    for n in range(num_samples):
        z = get_z_samples(x_samples, mu1, mu2).copy()

        mu1 = np.sum(z == 1)
        mu2 = np.sum(z == 2)

        mu1_h = 100 * np.sum((z == 1) * x_samples) / (100 * mu1 + 1)
        mu2_h = 100 * np.sum((z == 2) * x_samples) / (100 * mu2 + 1)

        mu1 = np.random.normal(mu1_h, np.sqrt(200 / (100 * mu1 + 1)))
        mu2 = np.random.normal(mu2_h, np.sqrt(200 / (100 * mu2 + 1)))

        mu1s[n] = mu1
        mu2s[n] = mu2

    return mu1s, mu2s

[162]: num_samples = 11000
        gibbs_mean_mu1s = []
        gibbs_mean_mu2s = []

        plt.figure(figsize=(30, 10))
        for i in range(6):
            # Run Gibbs Sampling
            gibbs_mu1s, gibbs_mu2s = gibbs_sampling(x_samples, num_samples)
```

```

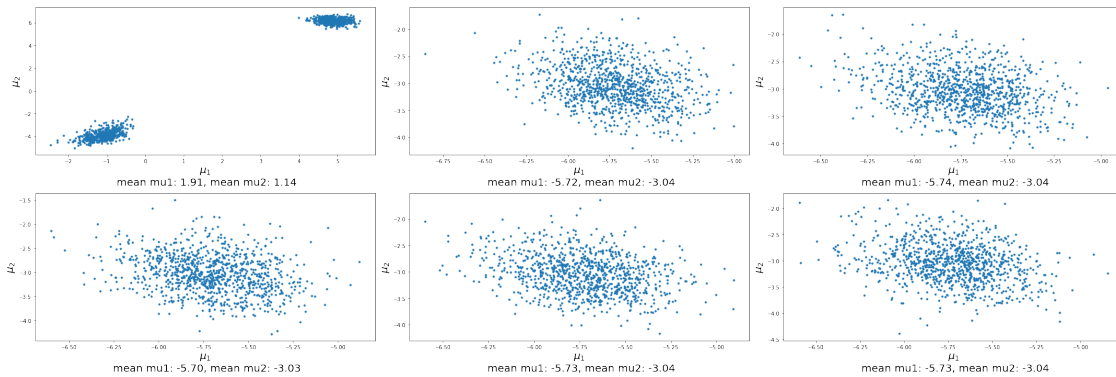
# Discard the first 10000 samples
gibbs_mu1s = gibbs_mu1s[10000:]
gibbs_mu2s = gibbs_mu2s[10000:]

# Plot the next 1000 samples of mu1 and mu2
plt.subplot(231 + i)
plt.scatter(gibbs_mu1s, gibbs_mu2s, s=10)
plt.xlabel(r'$\mu_1$' + '\n mean mu1: %.2lf, mean mu2: %.2lf' % (np.
↪mean(gibbs_mu1s), np.mean(gibbs_mu2s)), fontsize=20)
plt.ylabel(r'$\mu_2$', fontsize=20)
plt.tight_layout()

# Record the data
gibbs_mean_mu1s.append(np.mean(gibbs_mu1s))
gibbs_mean_mu2s.append(np.mean(gibbs_mu2s))

plt.show()

```



```

[163]: df = pd.DataFrame()
df['mean_mu1'] = gibbs_mean_mu1s
df['mean_mu2'] = gibbs_mean_mu2s

df

```

```

[163]:   mean_mu1  mean_mu2
0    1.912066  1.142613
1   -5.724099 -3.042464
2   -5.738514 -3.035543
3   -5.704386 -3.032615
4   -5.725242 -3.040966
5   -5.728205 -3.044541

```

```

[]:

```