HW3

November 4, 2022

```
[1]: import sympy
  from sympy import solve
  from sympy.abc import a, b
  import numpy as np
  import scipy
  from scipy.stats import norm, multivariate_normal, invgamma, bernoulli
  import matplotlib.pyplot as plt
  from copy import deepcopy
```

1 Problem 1(a)

1.1 Constant definition and data simulation

```
[2]: # define constants and seed
     loc_gt = 50
     scale_gt = np.sqrt(10)
     N = 100
     mu O = O
     sigma2_0 = 100
     alpha = 2
     beta = 10
     np.random.seed(100)
[3]: y = norm.rvs(loc_gt, scale_gt, N)
[4]: y
[4]: array([44.46675573, 51.08365058, 53.64621936, 49.20172716, 53.1032088,
            51.62610275, 50.69943153, 46.61622588, 49.40076157, 50.80638537,
            48.5515915 , 51.37610778 , 48.15451041 , 52.58309725 , 52.12732998 ,
            49.66982297, 48.31994393, 53.25630067, 48.61449351, 46.46356719,
            55.11966954, 54.8749836, 49.20348823, 47.33598428, 50.58349933,
            52.96331411, 52.31162606, 54.30561852, 48.96834467, 50.17606302,
            50.70328931, 45.43614714, 47.608204 , 52.58185428, 52.3731147 ,
            48.55816922, 53.76191592, 44.65380018, 45.71068959, 46.10269987,
            48.2783322 , 47.88705544, 50.02313068, 48.06171753, 54.1101643 ,
            44.51947043, 46.89050044, 51.13053878, 44.89741675, 54.6508056,
```

```
46.24315849, 48.26154989, 47.02731302, 47.38184798, 50.34425651, 51.60583492, 47.27339772, 53.95117025, 49.74824714, 47.18642201, 47.21150865, 50.05894153, 50.75213073, 50.04284427, 44.82800192, 46.69791843, 51.93859916, 52.3280853, 53.24741073, 45.47101563, 44.17765137, 51.15768843, 48.95082858, 47.82050138, 56.43399404, 48.25848812, 52.3731418, 45.86692732, 51.83593409, 46.5071913, 52.18235571, 52.17213711, 45.04569902, 52.86177945, 52.46285267, 51.35419124, 50.34428346, 50.08944071, 48.16959203, 46.20700227, 44.60530608, 51.16739893, 55.93424623, 48.80812696, 55.79309055, 50.00954196, 49.75959269, 50.01251501, 49.41493401, 42.13493626])
```

1.2 Gradient of log posterior

```
[5]: # let a = \mu, b = \sigma**2
d_logp_d_mu = -(N*a - np.sum(y))/b - (a-mu_0)/sigma2_0
d_logp_d_simga2 = -(N/2+alpha+1)/b + ((N*(a**2) - 2*a*np.sum(y)+np.sum(y**2))/2_\(\text{\top}\) + beta)/(b**2)
```

1.3 Solve μ and σ^2 numerically

```
[6]: MAP = sympy.solve([d_logp_d_mu, d_logp_d_simga2])
[7]: mu_map = MAP[0][a]
    sigma2_map = MAP[0][b]
[8]: MAP[0]
```

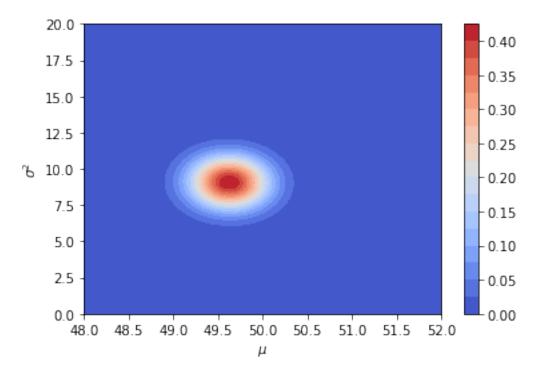
[8]: {a: 49.6256255571400, b: 9.06247322659448}

[11]: cov_mat

```
[12]: increment = 200
X = np.linspace(48, 52, increment)
Y = np.linspace(0, 20, increment)
X, Y = np.meshgrid(X, Y)
```

```
pos = np.dstack((X, Y))
rv = multivariate_normal([mu_map, sigma2_map], cov_mat)
Z = rv.pdf(pos)

plt.contourf(X, Y, Z, 20, cmap='coolwarm')
plt.colorbar();
plt.xlabel('$\mu$')
plt.ylabel('$\sigma^2$')
plt.ylabel('$\sigma^2$')
plt.savefig("./pla.pdf", format="pdf", bbox_inches="tight")
plt.show()
```



2 Problem 1(b)

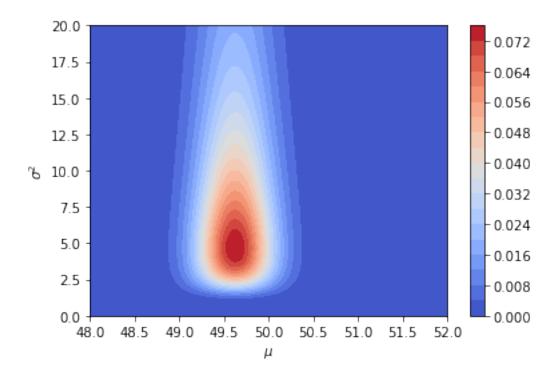
2.1 Initialize m, v^2, b

```
[13]: m = mu_0
v = np.sqrt(sigma2_0)
bb = beta # distinguish with symbol used in 1(a)
aa = 1 # distinguish with symbol used in 1(a)
```

```
[14]: m, v, bb, mu_0, sigma2_0, alpha, beta
```

[14]: (0, 10.0, 10, 0, 100, 2, 10)

```
[15]: total_steps = 1000
      for l in range(total_steps):
          m = (np.sum(y) + bb/aa*(mu_0/sigma2_0)) / (N+bb/(aa*sigma2_0))
          v = np.sqrt((bb/aa) / (N+bb/(aa*sigma2_0)))
          bb = aa*(N*v**2 + np.sum((y-m)**2) + 2*beta) / (2*alpha + N)
[16]: m, v, bb
[16]: (49.624316725225576, 0.3052502279129732, 9.326460345486584)
[17]: q_u = norm(loc=m, scale=v)
      q_sigma2 = invgamma(aa, scale=bb)
[18]: increment = 200
           = np.linspace(48, 52, increment)
          = np.linspace(0, 20, increment)
      XX, YY = np.meshgrid(X, Y)
      pos = np.dstack((XX, YY))
      MU = q_u.pdf(X)
      SIMGA2 = q_sigma2.pdf(Y)
      Z = np.outer(MU, SIMGA2).T
      plt.contourf(X, Y, Z, 20, cmap='coolwarm')
      plt.colorbar();
      plt.xlabel('$\mu$')
      plt.ylabel('$\sigma^2$')
      plt.savefig("./p1b.pdf", format="pdf", bbox_inches="tight")
      plt.show()
```



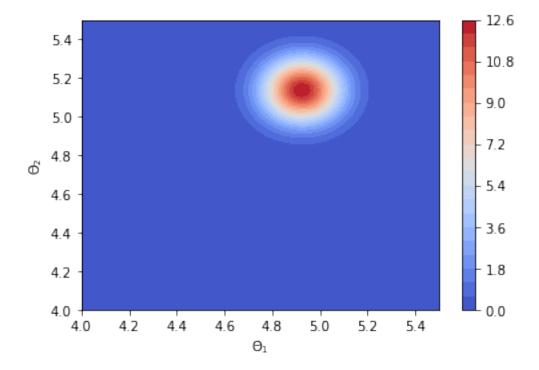
3 Problem 2(a)

3.1 Constant definition and data simulation

```
s_new = s * z
pi = 1 - 1/2 / z * multivariate_normal.pdf(y[n], mean=np.zeros((2, )),
cov=10*np.eye(2))
m_new = m + v*pi*(y[n]-m)/(v+1)
v_new = v - pi*(v**2)/(v+1) + pi*(1-pi)*(v**2*np.sum((y[n]-m)**2)) /
cov=10*np.eye(2))
m_new = m + v*pi*(y[n]-m)/(v+1)
v_new = v - pi*(v**2)/(v+1) + pi*(1-pi)*(v**2*np.sum((y[n]-m)**2)) /
cov=10*np.eye(2))
m_new = m + v*pi*(y[n]-m)/(v+1)
v_new = v - pi*(v**2)/(v+1) + pi*(1-pi)*(v**2*np.sum((y[n]-m)**2)) /
cov=10*np.eye(2))
m_new = m + v*pi*(y[n]-m)/(v+1)
v_new = v - pi*(v**2)/(v+1) + pi*(1-pi)*(v**2*np.sum((y[n]-m)**2)) /
cov=10*np.eye(2))
m_new = m + v*pi*(y[n]-m)/(v+1)
v_new = v - pi*(v**2)/(v+1) + pi*(1-pi)*(v**2*np.sum((y[n]-m)**2)) /
cov=10*np.eye(2))
```

[25]: m_new, v_new

[25]: (array([4.92566649, 5.13535363]), 0.012727630711702492)



4 Problem 2(b)

```
[27]: # initialize prior twems
      v0 = 100
      m0 = np.zeros((2, ))
      s0 = (2*np.pi*v0)**(-2/2)
[28]: # initialize data terms
      v_data = np.ones((N, )) * np.inf
      m_data = np.zeros((N, 2))
      s_data = np.ones((N, ))
[29]: # initialize target variables
      m_new = m0
      v_new = v0
[30]: m_new
[30]: array([0., 0.])
[31]: delta = 1e-4
      iter_num = 0
      while True:
          # save old
          for n in range(N):
              # (a)
              m_new_copy = deepcopy(m_new)
              v_new_copy = deepcopy(v_new)
              v = 1/(1/v_new-1/v_data[n])
              if (v_data[n] < 0 or v_data[n] == np.inf) and iter_num > 1:
                   v data[n] = 1e10
                   v new = v
              m = m_new + v/v_data[n] * (m_new-m_data[n])
              # (b)
              # print(m, m_new, m_data[n], v, v_data[n])
              z = (1-1/2)*multivariate_normal.pdf(y[n], mean=m, cov=(v+1)*np.eye(2))_{\sqcup}
       →+ \
              1/2*multivariate_normal.pdf(y[n], mean=np.zeros((2, )), cov=10*np.
       \rightarroweye(2))
              pi = 1 - 1/2 / z * multivariate_normal.pdf(y[n], mean=np.zeros((2, )),_\[ \]
       \rightarrowcov=10*np.eye(2))
              m_new = m + v*pi*(y[n]-m)/(v+1)
```

```
v_{new} = v - pi*(v**2)/(v+1) + pi*(1-pi)*(v**2*np.sum((y[n]-m)**2)) /_{\square}
       \hookrightarrow (2*(v+1)**2)
              # (c)
              v_{data}[n] = 1/(1/v_{new} - 1/v)
              m_data[n] = m + (v_data[n] + v)/v * (m_new-m)
               # s data[n] = z / (2*np.pi*v data[n]* multivariate normal.
       \rightarrow pdf(m_data[n], mean=m, cov=(v_data[n]+v)*np.eye(2)))
          # check convergence
               # print(np.sum(np.abs(m_new-m_new_copy)))
               if np.sum(np.abs(m_new-m_new_copy)) < delta:</pre>
                   break
          if np.sum(np.abs(m new-m new copy)) < delta:</pre>
                   break
     /var/folders/72/fk532lhj0vl3nqydc2n31h4c0000gn/T/ipykernel_64003/3967665427.py:2
     2: RuntimeWarning: divide by zero encountered in double scalars
       v_{data}[n] = 1/(1/v_{new} - 1/v)
     /var/folders/72/fk532lhj0vl3nqydc2n31h4c0000gn/T/ipykernel_64003/3967665427.py:2
     3: RuntimeWarning: invalid value encountered in multiply
       m_{data}[n] = m + (v_{data}[n] + v)/v * (m_{new-m})
[32]: m_new, v_new
[32]: (array([4.99337862, 5.11811095]), 0.012629772155207391)
[33]: increment = 200
           = np.linspace(4, 5.5, increment)
           = np.linspace(4, 5.5, increment)
      X, Y = np.meshgrid(X, Y)
      pos = np.dstack((X, Y))
      rv = multivariate_normal(m_new, v_new)
           = rv.pdf(pos)
      plt.contourf(X, Y, Z, 20, cmap='coolwarm')
      plt.colorbar();
      plt.xlabel('$\Theta_1$')
      plt.ylabel('$\Theta_2$')
```

plt.savefig("./p2b.pdf", format="pdf", bbox_inches="tight")

plt.show()

