

# BayesHW1

September 18, 2022

## 0.1 Question 1

```
[1]: from IPython.display import IFrame
     IFrame("BayesHW1Q1.pdf", width=600, height=300)
```

```
[1]: <IPython.lib.display.IFrame at 0x7f2e15477280>
```

## 0.2 Question 2

```
[2]: from sympy import symbols, Eq, solve
     import numpy as np
     import math
     import seaborn as sns
     import matplotlib.pyplot as plt
     from scipy.stats import beta

     a, b = symbols('a b')

     n = [2, 10, 50, 100]
     m = [0.25, 0.5, 0.75, 0.9]

     A = [[]]*(len(n)*len(m))
     B = [[]]*(len(n)*len(m))

     ct = 0
     for i in m:
         for ii in n:
             eq1 = Eq((i-1)*a + i*b, 0)
             eq2 = Eq(a + b - ii, 0)

             C = solve((eq1, eq2), (a, b))

             A[ct] = C.get(a)
             B[ct] = C.get(b)

             ct = ct + 1
```

```

[3]: def plot_prior_post(X,N,A,B):

    cols = ['m={}'.format(col) for col in m ]
    rows = ['n={}'.format(row) for row in n]

    fig, axes = plt.subplots(4,4,sharex=True,sharey=True)
    ct = 0
    row = 0

    for i in range(len(A)):

        x = np.linspace(0, 1, 1000)
        fx = beta.pdf(x, float(A[i]), float(B[i]))
        prior = {'x': x, 'fx': fx}

        updated_a = A[i] + X
        updated_b = B[i] + N - X

        fxx = beta.pdf(x, float(updated_a), float(updated_b))
        posterior = {'x': x, 'fxx': fxx}

        #Plot prior and posterior
        if ct < 4:
            #ax[row, ct].plot(prior['x'], prior['fx'])
            axes[row,ct].plot(posterior['x'], posterior['fxx'])

            ct = ct + 1

        else:
            ct = 0
            row = row + 1
            #ax[row, ct].plot(prior['x'], prior['fx'])
            axes[row,ct].plot(posterior['x'], posterior['fxx'])
            ct = ct + 1

    fig.tight_layout()
    fig.subplots_adjust(left=.15, top=1.5)

    for ax, col in zip(axes[0], cols):
        ax.set_title(col)

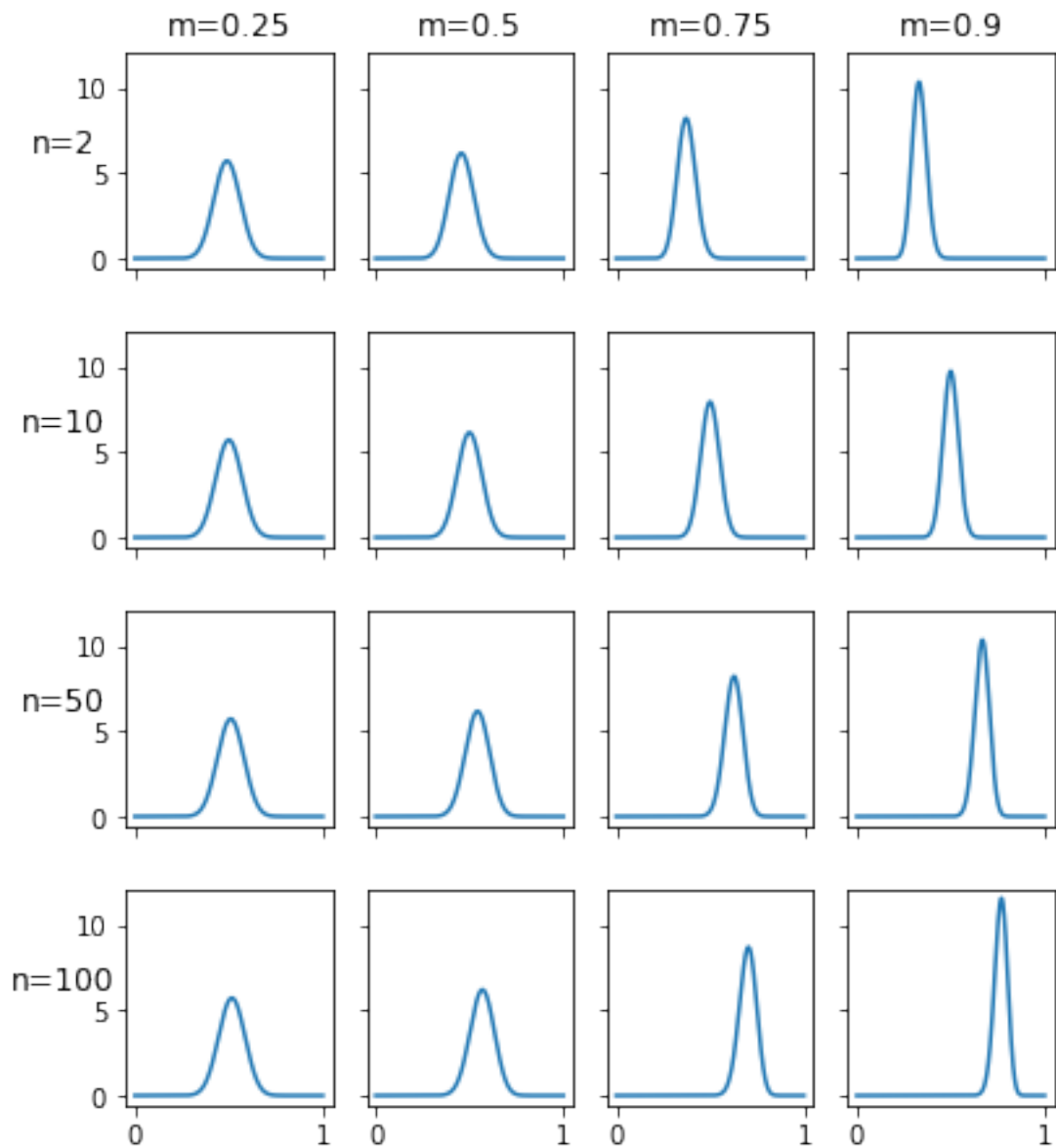
    for axes, r in zip(axes[:,0], rows):
        axes.set_ylabel(r, rotation=0, size='large')

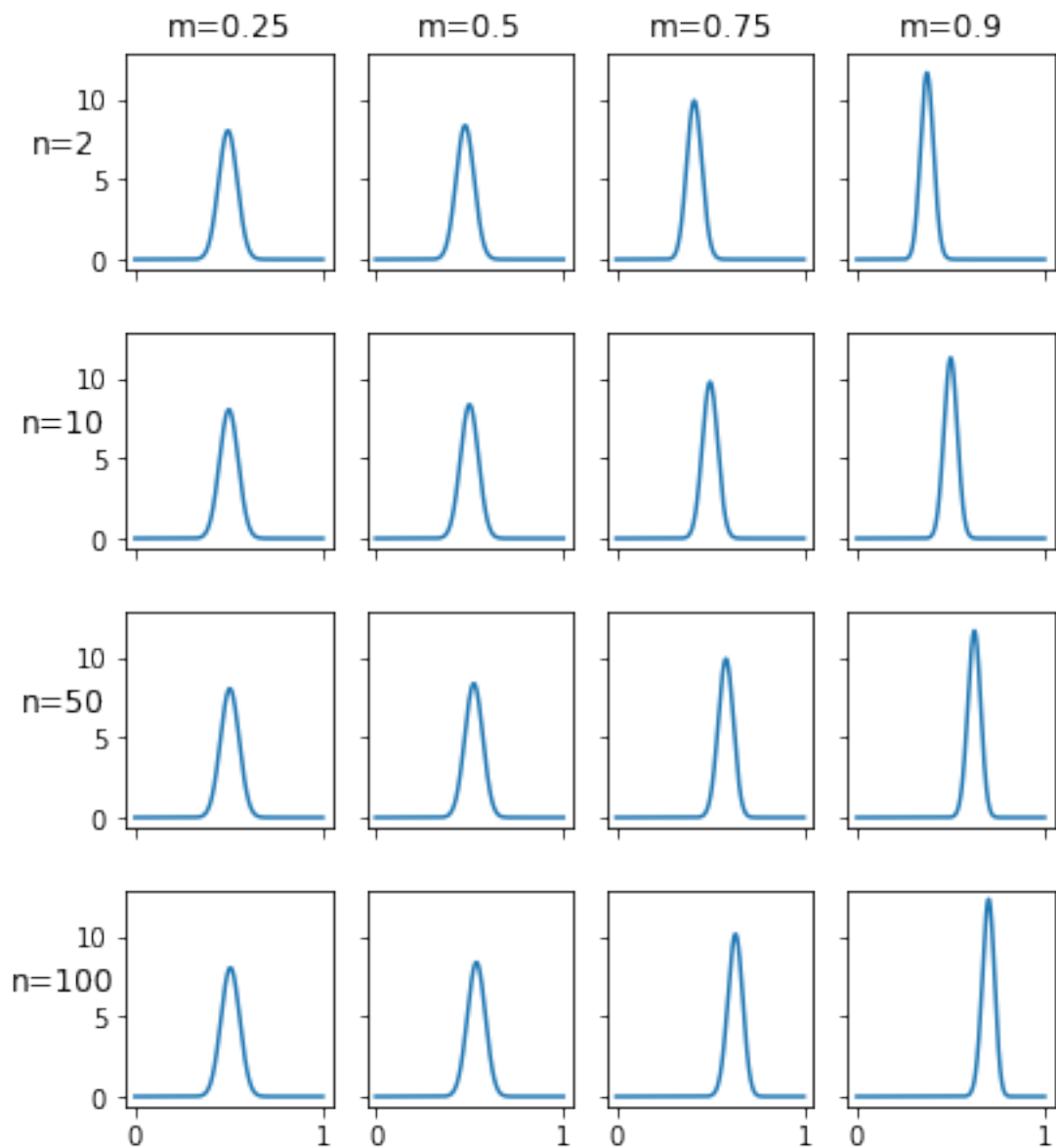
```

```

A = [round(float(item), 2) for item in A]
B = [round(float(item), 2) for item in B]
plot_prior_post(25,50,A,B)
plot_prior_post(50,100,A,B)

```





For the second data set with more observations the posterior converges to the mean faster. Also, if there are more data points in the prior it has a larger influence on the posterior.

### 0.3 Question 3

```
[4]: from IPython.display import IFrame
      IFrame("BayesHW1Q1.pdf", width=600, height=300)
```

```
[4]: <IPython.lib.display.IFrame at 0x7f2de2c368b0>
```

## 0.4 Question 4

```
[5]: import pandas as pd

data = pd.read_csv('ArtHistBooks.csv')
df = data.drop('Purchase', axis=1)

def old_books(df,a_0,b_0):
    df[df > 1] = 1

    a_post_0 = [[]]*(df.shape[1])
    b_post_0 = [[]]*(df.shape[1])
    successes = df.sum()

    N = len(data)

    ct = 0

    for X in successes:
        a_post_0[ct] = a_0 + X
        b_post_0[ct] = b_0 + N - X

        ct = ct + 1

    theta_hat_0 = [[]]*(df.shape[1])

    for i in range(len(a_post_0)):
        theta_hat_0[i] = a_post_0[i] / (a_post_0[i] + b_post_0[i])

    book_types = ['A','H','C']
    column_names = ['Book Type','a_post','b_post','theta_post']
    vals = [book_types,a_post_0,b_post_0,theta_hat_0]
    return pd.DataFrame(dict(zip(column_names, vals)))

old_books(df,1,1)
```

```
-----
FileNotFoundError                                Traceback (most recent call last)
<ipython-input-5-c2eb5d5f032d> in <module>
      1 import pandas as pd
      2
----> 3 data = pd.read_csv('ArtHistBooks.csv')
      4 df = data.drop('Purchase', axis=1)
      5
```

```

~/.local/lib/python3.8/site-packages/pandas/util/_decorators.py in
↳ wrapper(*args, **kwargs)
    309         stacklevel=stacklevel,
    310     )
--> 311     return func(*args, **kwargs)
    312
    313     return wrapper

~/.local/lib/python3.8/site-packages/pandas/io/parsers/readers.py in
↳ read_csv(filepath_or_buffer, sep, delimiter, header, names, index_col,
↳ usecols, squeeze, prefix, mangle_dupe_cols, dtype, engine, converters,
↳ true_values, false_values, skipinitialspace, skiprows, skipfooter, nrows,
↳ na_values, keep_default_na, na_filter, verbose, skip_blank_lines, parse_dates,
↳ infer_datetime_format, keep_date_col, date_parser, dayfirst, cache_dates,
↳ iterator, chunksize, compression, thousands, decimal, lineterminator,
↳ quotechar, quoting, doublequote, escapechar, comment, encoding,
↳ encoding_errors, dialect, error_bad_lines, warn_bad_lines, on_bad_lines,
↳ delim_whitespace, low_memory, memory_map, float_precision, storage_options)
    678     kwds.update(kwds_defaults)
    679
--> 680     return _read(filepath_or_buffer, kwds)
    681
    682

~/.local/lib/python3.8/site-packages/pandas/io/parsers/readers.py in
↳ _read(filepath_or_buffer, kwds)
    573
    574     # Create the parser.
--> 575     parser = TextFileReader(filepath_or_buffer, **kwds)
    576
    577     if chunksize or iterator:

~/.local/lib/python3.8/site-packages/pandas/io/parsers/readers.py in
↳ __init__(self, f, engine, **kwds)
    932
    933     self.handles: IOHandles | None = None
--> 934     self._engine = self._make_engine(f, self.engine)
    935
    936     def close(self):

~/.local/lib/python3.8/site-packages/pandas/io/parsers/readers.py in
↳ _make_engine(self, f, engine)
    1216         # "Union[str, PathLike[str], ReadCsvBuffer[bytes],
↳ ReadCsvBuffer[str]]"
    1217         # , "str", "bool", "Any", "Any", "Any", "Any", "Any"
-> 1218         self.handles = get_handle( # type: ignore[call-overload]

    1219             f,
    1220             mode,

```

```

~/local/lib/python3.8/site-packages/pandas/io/common.py in
↳get_handle(path_or_buf, mode, encoding, compression, memory_map, is_text,
↳errors, storage_options)
    784         if ioargs.encoding and "b" not in ioargs.mode:
    785             # Encoding
--> 786             handle = open(
    787                 handle,
    788                 ioargs.mode,

FileNotFoundError: [Errno 2] No such file or directory: 'ArtHistBooks.csv'

```

```
[ ]: old_books(df,1,100)
```

```

[ ]: def new_book(data,a_0,b_0):

    cases = {0:{'A':1,'H':0,'C':0,'P':1},1:{'A':1,'H':1,'C':0,'P':1},2:{'A':
↳1,'H':1,'C':1,'P':1},
           3:{'A':0,'H':1,'C':0,'P':1},4:{'A':0,'H':1,'C':1,'P':1},5:{'A':
↳0,'H':0,'C':1,'P':1},
           6:{'A':1,'H':0,'C':1,'P':1},7:{'A':0,'H':0,'C':0,'P':1}}

    successes = [0]*len(cases)

    total_cases = [0]*len(cases)

    for i in range(0,8):
        for index, row in data.iterrows():
            case = cases.get(i)
            A = case.get('A') == row[0]
            H = case.get('H') == row[1]
            C = case.get('C') == row[2]
            P = case.get('P') == row[3]
            if A == True and H == True and C == True and P == True:
                successes[i] = successes[i] + 1
            if A == True and H == True and C == True:
                total_cases[i] = total_cases[i] + 1

    ct = 0

    a_post_0 = [[]]*(len(cases))
    b_post_0 = [[]]*(len(cases))

    for i in range(0,len(successes)):
        a_post_0[ct] = a_0 + successes[i]
        b_post_0[ct] = b_0 + total_cases[i] - successes[i]

```

```

        ct = ct + 1

    theta_hat_0 = [[]]*(len(cases))

    for i in range(len(a_post_0)):
        theta_hat_0[i] = a_post_0[i] / (a_post_0[i] + b_post_0[i])

    cases_list = list(cases.values())
    column_names = ['Case', 'a_post', 'b_post', 'theta_post']
    vals = [cases_list, a_post_0, b_post_0, theta_hat_0]
    return pd.DataFrame(dict(zip(column_names, vals)))

data[data > 1] = 1
new_book(data, 1, 1)

```

```
[ ]: new_book(data, 1, 100)
```

Increasing beta prior decreased the likelihood for all situations as expected

## 0.5 Question 5

```

[ ]: chd_data = pd.read_csv('CHDdata.csv')
    chd_data = chd_data.drop(['famhist'], axis=1)

    for column in chd_data:
        if column != 'chd':
            chd_data[column] = (chd_data[column] - chd_data[column].mean()) / ↳
            ↳chd_data[column].std()

    chd_data_1 = chd_data[chd_data['chd']==1].drop('chd', axis=1).
    ↳reset_index(drop=True)

    chd_data_0 = chd_data[chd_data['chd']==0].drop('chd', axis=1).
    ↳reset_index(drop=True)

    mu_0 = np.transpose(np.zeros(8))
    v = 1
    a_prior = 9
    dim = (8,8)
    r_prior = np.identity(8)

    N_1 = len(chd_data_1)
    N_0 = len(chd_data_0)

    x_bar1 = np.matrix(chd_data_1.mean().values)

```



```

x_bar1 = x_bar1.T
x_bar0 = np.matrix(chd_data_0.mean().values)
x_bar0 = x_bar0.T

mu_post_1 = ((v*mu_0) + (N_1*chd_data_1.mean().values)) / (v + N_1)
dof_1 = a_prior + N_1 - 8 + 1

mu_post_0 = ((v*mu_0) + (N_0*chd_data_0.mean().values)) / (v + N_0)
dof_0 = a_prior + N_0 - 8 + 1

S_1 = np.zeros(dim)
for index, row in chd_data_1.iterrows():
    x_j = np.matrix(chd_data_1.iloc[index].values)
    j_bar = x_j.T - x_bar1
    S_temp = np.dot(j_bar, np.transpose(j_bar))
    S_1 = S_1 + S_temp

S_0 = np.zeros(dim)
for index, row in chd_data_0.iterrows():
    x_j = np.matrix(chd_data_0.iloc[index].values)
    j_bar = x_j.T - x_bar0
    S_temp = np.dot(j_bar, np.transpose(j_bar))
    S_0 = S_0 + S_temp

r_post_1 = r_prior + S_1 + ((v * N_1) / (v + N_1)) * np.outer((mu_0.T - x_bar1.
    ↪T), (mu_0.T - x_bar1.T))
r_post_0 = r_prior + S_0 + ((v * N_0) / (v + N_0)) * np.outer((mu_0.T - x_bar0.
    ↪T), (mu_0.T - x_bar0.T))

percision_1 = (v+N_1)*dof*np.linalg.pinv(r_post_1)
percision_0 = (v+N_0)*dof*np.linalg.pinv(r_post_0)

D = np.matrix(abs(mu_post_0 - mu_post_1))
D = D.T

const = (1/N_0) + (1/N_1)
top = ((N_0 - 1)*np.linalg.pinv(r_post_0)) + ((N_1 - 1) * np.linalg.
    ↪pinv(r_post_1))
r_pooled = (const * top) / (N_0 + N_1 - 2)

T_2 = np.transpose(D) * np.linalg.pinv(r_pooled) * D
print(T_2)

```

Since  $T^2$  is large we will reject the null that the means are the same

[ ]: