

## HW3

```
[1]: import pandas as pd
import numpy as np
from scipy.optimize import minimize
from scipy.stats import norm

[42]: data = pd.read_excel("/Users/lee/Downloads/CilibertoTamerEconometrica.xlsx")
```

## Q1

## Q1

```
[53]: n = len(data)
data['mk'] = 1
X = np.array(data[['mk', 'marketsize', 'marketdistance']])
X

[53]: array([[1.      , 1.81654286, 0.69199997],
 [1.      , 1.00100732, 0.685      ],
 [1.      , 1.23256075, 0.50300002],
 ...,
 [1.      , 1.45095992, 1.72099996],
 [1.      , 1.23586535, 0.54699999],
 [1.      , 0.8648805 , 0.90600002]])

[30]: N = np.array(data[['airlineAA', 'airlineDL', 'airlineUA', 'airlineAL', 'airlineLCC', 'airlineWN']].sum(axis=1))
N

[30]: array([1, 1, 0, ..., 4, 2, 1])

[39]: def nl(para):
    beta = np.array(para[:3])
    delta = np.array(para[3])

    nll = 0
    for i in range(n):
        if N[i] == 0:
            nll += -np.log(norm.cdf(np.dot(-X[i], beta)))
        elif N[i] == 6:
            nll += -np.log(1 - norm.cdf(np.dot(-X[i], beta) + delta*np.log(6)))
        else:
            nll += -np.log(norm.cdf(np.dot(-X[i], beta) + delta*np.log(N[i] + 1)) - norm.cdf(np.dot(-X[i], beta) + de
    return nll

[40]: para = [1.3, 1.3, 1.3, 1]

[41]: ans = minimize(nl, para, options={'disp': True})

Warning: Desired error not necessarily achieved due to precision loss.
Current function value: 4599.864692
Iterations: 15
Function evaluations: 472
Gradient evaluations: 76

[43]: para = ans.x
beta, delta = para[:3], para[3]
cvmt = ans.hess_inv
se = [np.sqrt(cvmt[i,i]) for i in range(len(cvmt))]

[46]: print('Beta:',beta,'\nBeta\'s standard error:',se[:3])
print('Delta:',delta,'\nDelta\'s standard error:',se[3])

Beta: [0.98325407 0.06796466 0.47342029]
Beta's standard error: [0.016073392954031825, 0.017222813891410622, 0.012128213291040213]
Delta: 1.9211390416649874
Delta's standard error: 0.03196461970500831
```

Q2

Q2

```
[59]: data = np.array(data)
data2 = []
for i in range(n):
    imk = []
    for j in range(6):
        imk.append(np.append(X[i], (data[i][j+15], data[i][j+21])))
    data2.append(np.vstack(imk))

[60]: data2[0]

[60]: array([[1.          , 1.81654286, 0.69199997, 0.18000001, 0.81831986],
        [1.          , 1.81654286, 0.69199997, 0.55909091, 0.          ],
        [1.          , 1.81654286, 0.69199997, 0.24636364, 0.81831986],
        [1.          , 1.81654286, 0.69199997, 0.48030305, 0.02098957],
        [1.          , 1.81654286, 0.69199997, 0.10666667, 0.          ],
        [1.          , 1.81654286, 0.69199997, 0.          , 0.00691858]])

[64]: def of(para):
    beta_alpha = np.array(para[:5])
    delta = np.array(para[5])
    phi = np.array(para[6])

    E_N = np.zeros(n)
    np.random.seed(7323014)
    for i in range(n):
        N_list = np.zeros(T)
        for j in range(T):
            u_i0 = np.random.randn(1)
            u_ik = np.random.randn(6)
            for k in range(6,0,-1):
                profit = np.dot(data2[i], beta_alpha) - delta*np.log(k) + phi*u_i0 + np.sqrt(1-phi**2)*u_ik
                profit_num_firm = sum(profit > 0)
                if profit_num_firm >= k:
                    N_list[j] = k
                    break
            E_N[i] = np.mean(N_list)
    Error = N - E_N
    mu = np.dot(Error, X)
    return np.dot(mu, mu)

[68]: para = [0, 0, 0, 0, 0, 0.5, 0.5]

[69]: theta_list = []
T = 30
S = 30
seed_list = np.random.randint(0,high=10000,size=S)
for seed in range(S):
    ith_seed = seed_list[seed]
    def of(para):
        beta_alpha = np.array(para[:5])
        delta = np.array(para[5])
        phi = np.array(para[6])

        E_N = np.zeros(n)
        np.random.seed(ith_seed)
        for i in range(n):
            N_list = np.zeros(T)
            for j in range(T):
                u_i0 = np.random.randn(1)
                u_ik = np.random.randn(6)
                for k in range(6,0,-1):
                    profit = np.dot(data2[i], beta_alpha) - delta*np.log(k) + phi*u_i0 + np.sqrt(1-phi**2)*u_ik
                    profit_num_firm = sum(profit > 0)
                    if profit_num_firm >= k:
                        N_list[j] = k
                        break
                E_N[i] = np.mean(N_list)
            Error = N - E_N
            mu = np.dot(Error, X)
            return np.dot(mu, mu)
    ans = minimize(of,
                    para,
                    method='Nelder-Mead',
                    options={'disp': True})
    theta_list.append(ans.x)
```

```
beta: [0.00047961 0.00116044 0.00014594]
beta's standard error: [1.35488485e-04 5.41428688e-05 1.26644489e-04]
alpha: [0.00022884 0.00045789]
alpha's standard error: [0.00021857 0.00022327]
delta: 0.24409692574675956
delta's standard error: 0.0030761758032818412
phi: 0.4932415561359996
phi's standard error: 0.004369318858032602
```