Prostate data modeling

July 3, 2019

```
import modules
In [1]: import numpy as np
        import pandas as pd
   reading data and variable setting
In [2]: dat = pd.read_csv('prostate.csv',encoding='UTF8').drop(columns = 'Unnamed: 0')
In [3]: train = dat[dat['train']].drop(columns = 'train')
In [4]: y = train['lpsa']
        X = train.drop(columns = 'lpsa')
        X = X.values
In [5]: N,p = X.shape
   Make the function wich return XX'
In [6]: def product(a):
            n = len(a)
            out = np.zeros([n,n])
            for i in range(n):
                for j in range(n):
                     out[i,j] = a[i]*a[j]
            return(out)
   Make expit function which is inverse function of logit
In [7]: def expit(x):
            if x < 100:
                return(np.exp(x)/(1+np.exp(x)))
            else:
                return(1)
```

 $Bernoulli(w_i)$

Start Variational inference which return w_i which is the parameter of the $q_4^*(\gamma_i) \sim$

```
In [8]: def rhotest(rho):
            sigmab = 1
            A = 10**(-7)
           B = 10**(-7)
            tau = 1
            #rho = 0.8
            w = np.repeat(0.5,p)
            lamb= np.log(rho/(1-rho))
            for iteration in range(1000):
                test= False
                W = np.diag(w)
                omega = product(w) + W.dot(np.eye(p)-W)
                sigma = np.linalg.inv(tau*np.multiply(X.T.dot(X),omega)+ (1/sigmab) * np.eye(p)
                mu = tau*sigma.dot(W.dot(X.T.dot(y)))
                s = B + 0.5*(np.linalg.norm(y)**2 -2*y.T.dot(X).dot(W).dot(mu)
                             + np.trace(np.multiply(X.T.dot(X),omega).dot(product(mu) +sigma))
                tau = (A+N/2)/s
                eta =np.zeros(len(w))
                for j in range(p):
                    eta[j] = lamb - 0.5*tau *(mu[j]**2 + sigma[j,j])*np.linalg.norm(X[:,j])**2
                    + tau*X[:,j].T.dot(y*mu[j]-np.delete(X,j,1).dot(np.diag(np.delete(w,j))).d
                    w[j] = expit(eta[j])
                t=t+1
            return(w)
In [9]: rhopost = []
        rholst = [0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9]
        for i in rholst:
            rhopost.append(rhotest(i))
  Plotting the change of w_i
In [10]: df = pd.DataFrame(rhopost)
         df.columns = dat.columns[:-2]
         df.index = rholst
         df.round(3)
Out [10]:
                     lweight age
              lcavol
                                     lbph
                                             svi
                                                    1cp gleason pgg45
                          0.0 0.0 0.000 0.039 0.000
         0.1
                 0.0
                                                             0.0
                                                                    0.0
         0.2
                 0.0
                          0.0 0.0 0.000 0.075 0.000
                                                             0.0
                                                                    0.0
         0.3
                          0.0 0.0 0.000 0.107 0.000
                                                             0.0
                                                                    0.0
                 0.0
         0.4
                 0.0
                          0.0 0.0 0.000 0.134 0.000
                                                             0.0
                                                                    0.0
                          0.0 0.0 0.000 0.161 0.000
         0.5
                 0.0
                                                             0.0
                                                                    0.0
                          0.0 0.0 0.000 0.334 0.000
                                                                    0.0
         0.6
                 0.0
                                                             0.0
         0.7
                 0.0
                          0.0 0.0 0.000 0.464 0.442
                                                             0.0
                                                                    0.0
         0.8
                          0.0 0.0 0.328 0.601 0.642
                                                                    0.0
                 0.0
                                                             0.0
                          0.0 0.0 0.669 0.768 0.822
         0.9
                 0.0
                                                             0.0
                                                                    0.0
```