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In [8]: from covid.simulator import Population
         from covid.auxilliary import symptom names
         import numpy as np
         import pandas as pd
         from covid.policy import Policy
In [65]: symptom names = ['covid recovered','covid positive', 'no taste smell',
          'fever', 'headache', 'pneumonia', 'stomach', 'myocarditis', 'blood clots', 'death']
In [9]: | w = np.array([0.2, 0.1, 0.1, 0.1, 0.5, 0.2, 0.5, 1.0, 100])
         assert w.shape[0] == 9, 'Shape of weights does not fit number of symptoms'
In [10]: class Model:
             def init (self, nsymptom, nvacc):
                  # Priors for the beta-bernoulli model
                 self.a = np.ones(shape=[nvacc, nsymptom])/2 # using jeffreys prior
                 self.b = np.ones(shape=[nvacc, nsymptom])/2 # using jeffreys prior
                 self.nvacc = nvacc
                 self.nsymptom = nsymptom
             def update(self, features, actions, outcomes):
                 for index in range(self.nvacc):
                     print(outcomes[np.where(actions == (index-1))])
                     print(actions)
                     print(outcomes)
                     if (np.sum(outcomes[np.where(actions == index - 1)], axis=1).size != 0):
                         self.a[index] += np.sum(outcomes[np.where(actions == index - 1)], axis=1)
                         self.b[index] += np.sum(outcomes[np.where(actions == index - 1)] == 0, axis=1) \setminus 
                              - np.sum(outcomes[np.where(actions == index - 1)], axis=1)
                      else:
                         self.b[index] += np.sum((outcomes==0)[np.where(actions == index - 1)], axis=1)
                  11 11 11
                 for index, outcome in enumerate(outcomes):
                     self.a[int(actions[index])] += outcome[1:]
                     self.b[int(actions[index])] += 1 - outcome[1:]
             def get params(self):
                 return self.a, self.b
             def get prob(self, features, action):
                 return self.a[action] / (self.a[action] + self.b[action])
             def retrain(self, features, actions, outcomes):
                 # Use aggregated database of outcomes etc...
                 self.a = np.ones(shape=[self.nvacc, self.nsymptom])/2 # using jeffreys prior
                 self.b = np.ones(shape=[self.nvacc, self.nsymptom])/2 # using jeffreys prior
                 self.update(features, actions, outcomes)
In [41]: class Naive (Policy):
             def get_utility(self, features, action, outcome):
                 utility = 0
                 for t, o in enumerate(outcome):
                     utility -= np.dot(w, o[1:])*(1+int(action[t] != -1))
                 return utility
             def set model(self, model):
                 self.model = model
             def get action(self, features):
                  """Get a completely random set of actions, but only one for each individual.
                 If there is more than one individual, feature has dimensions t*x matrix, otherwise it is an x-s
         ize array.
                 It assumes a finite set of actions.
                 Returns:
                 A t^*|A| array of actions
                 n obs = features.shape[0]
                 actions = np.zeros(n obs)
                 #u list = np.zeros((n obs, self.n actions))
                 for index, t in enumerate(features):
                     u list = []
                     for a in self.action set:
                         u list.append(self.get expected utility(a, t))
                     actions[index] = np.argmax(np.array(u list))
                 return actions
             def get expected utility(self, action, features):
                 p = self.model.get prob(features, action)
                 return -np.dot(p, w) * (1+int(action != -1))
             def observe(self, features, action, outcomes):
                 self.model.update(features, action, outcomes)
In [52]: ## Baseline simulator parameters
         n_{genes} = 128
         n vaccines = 3
         n treatments = 4
         n_population = 100000
         n_symptoms = 9
         # symptom names for easy reference
         from covid.auxilliary import symptom names
In [53]:
         population = Population(n_genes, n_vaccines, n_treatments)
         X = population.generate(n population)
         n features = X.shape[1]
         < array function internals>:5: VisibleDeprecationWarning: Creating an ndarray from ragged nested s
         equences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) i
         s deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.
In [54]: print("With a for loop")
         vaccine policy = Naive(2, np.array([-1, 0]))
         vaccine_policy.set_model(Model(n_symptoms, 2))
         # The simplest way to work is to go through every individual in the population
         Y = np.zeros((n_population, n_symptoms+1))
         A = np.zeros(n population)
         for t in range(n_population):
             #print("Person nr: ", t)
             a_t = vaccine_policy.get_action(X[t].reshape((1, n_features)))
             A[t] = a t
             # Then you can obtain results for everybody
             y_t = population.vaccinate([t], a_t.reshape((1, 1)))
             Y[t] = y t
             # Feed the results back in your policy. This allows you to fit the
             # statistical model you have.
             vaccine_policy.observe(X[t], a_t, y_t)
         With a for loop
         Initialising policy with 2 actions
         A = \{ [-1 \ 0] \}
In [61]: print(len(symptom names))
In [68]: def print pre statistics(X):
             print(f'Statistic (N={X.shape[0]})')
             for i in range(len(symptom_names)-1):
                 print(f'{symptom_names[i].ljust(15)} {X[:, i].sum()}')
In [75]: print pre statistics(Y)
         print(A)
         Statistic (N=100000)
         covid recovered 0.0
         covid positive 0.0
         no_taste_smell 0.0
                  100000.0
         fever
```

0.0

0.0

In [76]: | a, b = vaccine policy.model.get params()

[5.02542867e-06 5.02542867e-06 9.99994975e-01 5.02542867e-06 5.02542867e-06 5.02542867e-06 5.02542867e-06

[9.84251969e-04 9.84251969e-04 9.99015748e-01 9.84251969e-04 9.84251969e-04 9.84251969e-04 9.84251969e-04

headache pneumonia

stomach

myocarditis 0.0 blood\_clots 0.0

5.02542867e-06]

9.84251969e-04]

[0. 0. 1. ... 0. 0. 0.]

print(a[0]/(a[0] + b[0]))
print(a[1]/(a[1] + b[1]))