## thompson

#### December 9, 2021

```
[2]: from covid.simulator import Population
     from covid.auxilliary import symptom_names
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     from covid.policy import Policy
     from scipy.stats import beta, bernoulli, uniform
[3]: np.random.laplace(0,0.1)
[3]: -0.19614104708356772
[4]: class ThompsonSampling:
         def __init__(self, nvacc):
             # Priors for the beta-bernoulli model
             self.a = np.ones(nvacc) # uniform prior
             self.b = np.ones(nvacc) # uniform prior
             self.nvacc = nvacc
         def update(self, action, outcome):
             self.a[action] += outcome
             self.b[action] += 1 - outcome
         def update_with_laplace(self, action, outcome):
             outcome = outcome + np.random.laplace(0, 0.1)
             self.a[action] += outcome
             self.b[action] += 1 - outcome
         def get_params(self):
             # Returns the parameters of all the beta distrobutions.
             return self.a, self.b
```

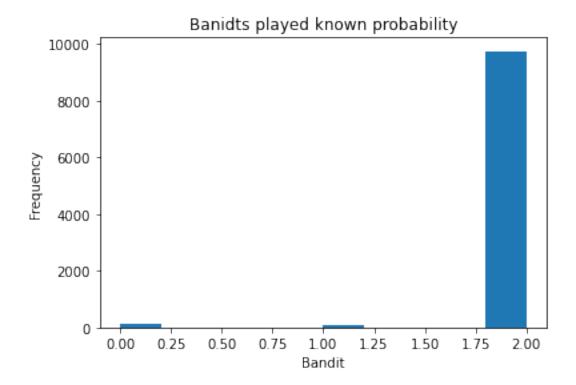
def get\_prob(self):

return beta.rvs(self.a, self.b)

```
[5]: class Naive(Policy):
         def set_model(self, model):
             self.model = model
         def get_action(self):
             probs = self.model.get_prob()
             #print(probs)
             ret_val = np.argmin(probs)
             #print(ret_val)
             #print(ret val)
             return ret_val
         def observe(self, action, outcome):
             self.model.update(action, outcome)
         def observe_with_laplace(self, action, outcome):
             self.model.update_with_laplace(action, outcome)
[6]: class NaiveForExample(Policy):
         def set_model(self, model):
             self.model = model
         def get_action(self):
             probs = self.model.get_prob()
             #print(probs)
             ret_val = np.argmax(probs)
             #print(ret_val)
             #print(ret_val)
             return ret_val
         def observe(self, action, outcome):
             self.model.update(action, outcome)
         def observe_with_laplace(self, action, outcome):
             self.model.update_with_laplace(action, outcome)
[7]: thetas = np.array([0.8, 0.75, 0.85])
[8]: n_genes = 128
     n_vaccines = 3
     n_{treatment} = 4
     #population = Population(n_qenes, n_vaccines, n_treatment)
     N = 10000
     #X = population.generate(N)
```

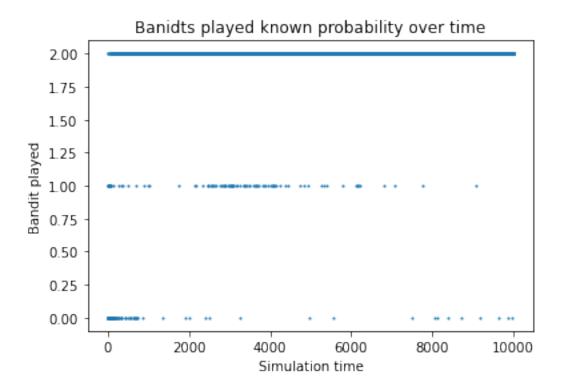
```
[9]: policy = NaiveForExample(n_actions=3, action_set=[0, 1, 2])
      model = ThompsonSampling(nvacc=3)
      policy.set_model(model)
      action_arr = np.zeros(N)
      symptom_arr = np.zeros(N)
      for i in range(N):
          action = policy.get_action()
          #print(thetas[action])
          reward = bernoulli.rvs(thetas[action])
          #print(reward)
          policy.observe(action, reward)
          action_arr[i] = action
          symptom_arr[i] = reward
          #print(action, reward)
          #policy.observe(action, reward)
          #print(action)
      alphas, betas = policy.model.get_params()
      for i, j in zip(alphas, betas):
          print(i,j)
      print(alphas[0]/(alphas[0] + betas[0]))
      print(alphas[1]/(alphas[1] + betas[1]))
      print(alphas[2]/(alphas[2] + betas[2]))
     Initialising policy with 3 actions
     A = \{ [0, 1, 2] \}
     114.0 34.0
     76.0 28.0
     8338.0 1416.0
     0.7702702702702703
     0.7307692307692307
     0.8548287881894607
[10]: #print(action_arr)
      #plt.scatter(range(N), action_arr)
      def plot_action_known_theta(actions):
          plt.hist(actions)
          plt.xlabel('Bandit')
          plt.ylabel('Frequency')
          plt.title('Banidts played known probability')
          #plt.legend(loc='upper right')
          plt.savefig('figures/banidts_played_known_prob_hist.png')
          plt.show()
```

### [11]: plot\_action\_known\_theta(action\_arr)



```
def plot_bandits_played_time(actions, N):
    plt.scatter(range(N), actions, s=1)
    plt.ylabel('Bandit played')
    plt.xlabel('Simulation time')
    plt.title('Banidts played known probability over time')
    #plt.legend(loc='upper right')
    plt.savefig('figures/banidts_played_known_prob_over_time.png')
    plt.show()
```

[13]: plot\_bandits\_played\_time(action\_arr, N)



```
n_vaccines = 3 # DO NOT CHANGE, breaks the simulator.
      n_{treatments} = 4
      n_population = 10_000
      n_{symptoms} = 10
      # symptom names for easy reference
      from covid.auxilliary import symptom_names
      np.random.seed(1)
      population = Population(n_genes, n_vaccines, n_treatments)
      X = population.generate(n_population)
      n_features = X.shape[1]
[15]: \#new\_col = X[:,[5,7,8]].sum(axis=1)
      #new col0 = new col > 0
      #new_col01 = new_col0.astype(int)
      #new_col01 = np.reshape(new_col01, (n_population, -1))
      \#X_new = np.hstack((X,new_col01))
[16]: action_space = np.array([-1,0,1,2])
      n_actions = action_space.shape[0]
```

[14]: n\_genes = 128

```
[17]: policy = Naive(n_actions, action_space)
      model = ThompsonSampling(n_actions)
      policy.set_model(model)
     Initialising policy with 4 actions
     A = \{ [-1 \ 0 \ 1 \ 2] \}
[18]: action_arr = np.zeros(n_population)
      symptom_arr = np.zeros((n_population, n_symptoms+1))
      symptom_arr.shape
[18]: (10000, 11)
[19]: print("With a for loop")
      # The simplest way to work is to go through every individual in the population
      for t in range(n_population):
          a_t = policy.get_action()
          #print(type(a_t))
          \#a \ t = np.int(3)
          # Then you can obtain results for everybody
          y_t = population.vaccinate([t], a_t.reshape((1, 1)))
          #print(y_t.shape)
          new_col = y_t[:,[5,7,8]].sum(axis=1)
          #print(y t)
          new_col0 = new_col > 0
          new_col01 = new_col0.astype(int)
          #new_col01 = np.reshape(new_col01, (n_population, -1))
          new_col01 = np.reshape(new_col01,(1,-1))
          y_t_new = np.hstack((y_t,new_col01))
          # Feed the results back in your policy. This allows you to fit the
          # statistical model you have.
          #print( y t new[:,10])
          policy.observe(a_t, y_t_new[:,10])
          \#policy.observe(a_t, y_t_new[:,1])
          action_arr[t] = a_t
          symptom_arr[t] = y_t_new
```

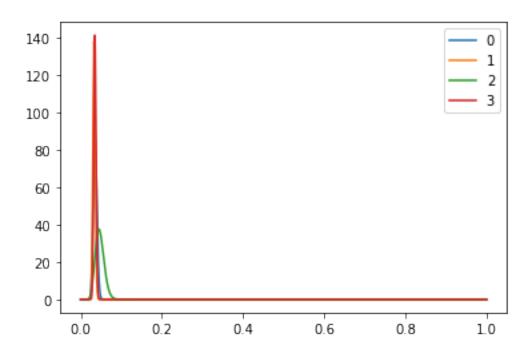
With a for loop

```
[20]: print(action_arr)
```

```
[3. 3. 3. ... 1. 0. 1.]
```

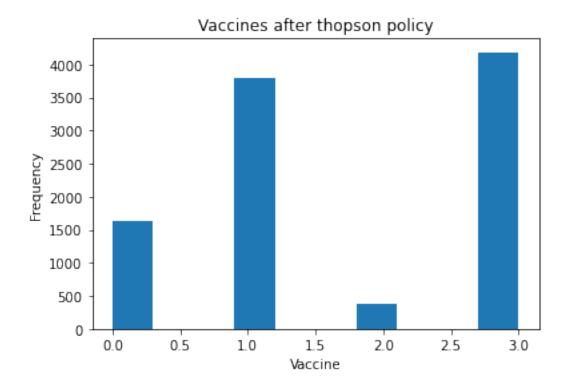
```
[21]: alphas, betas = policy.model.get_params()
      for i, j in zip(alphas, betas):
          print(i,j)
      print(alphas[0]/(alphas[0] + betas[0]))
      print(alphas[1]/(alphas[1] + betas[1]))
      print(alphas[2]/(alphas[2] + betas[2]))
      print(alphas[3]/(alphas[3] + betas[3]))
     58.0 1579.0
     126.0 3676.0
     18.0 364.0
     146.0 4041.0
     0.0354306658521686
     0.03314045239347712
     0.04712041884816754
     0.03486983520420349
[22]: utility_normal = -sum(alphas)
      utility_normal
[22]: -348.0
[23]: xs = np.linspace(0, 1, 10000)
      plt.plot(xs, beta.pdf(xs, alphas[0], betas[0]))
      plt.plot(xs, beta.pdf(xs, alphas[1], betas[1]))
      plt.plot(xs, beta.pdf(xs, alphas[2], betas[2]))
      plt.plot(xs, beta.pdf(xs, alphas[3], betas[3]))
     plt.legend(['0', '1', '2', '3'])
      #
```

[23]: <matplotlib.legend.Legend at 0x7f91780fde80>



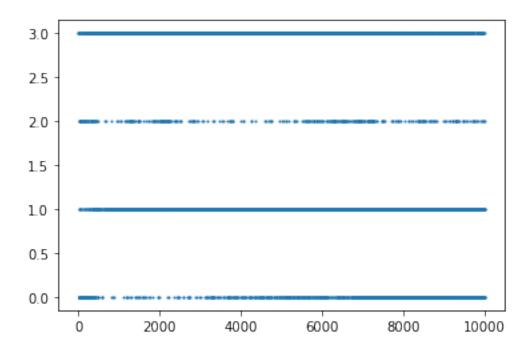
```
[24]: def plot_thompson_pol(actions):
    #plt.hist(action_arr)
    plt.hist(actions)#, label=['0','1','2','3'])
    plt.xlabel('Vaccine')
    plt.ylabel('Frequency')
    plt.title('Vaccines after thopson policy')
    #plt.legend(loc='upper right')
    plt.savefig('figures/vaccines_thompson_policy.png')
    plt.show()
```

[25]: plot\_thompson\_pol(action\_arr)



[26]: plt.scatter(range(n\_population), action\_arr, s=1)

[26]: <matplotlib.collections.PathCollection at 0x7f917852a9d0>



```
def plot_fair_pol(actions, data):
    mm = np.where(data[:,11]==1) #male
    ff = np.where(data[:,11]==0) #female

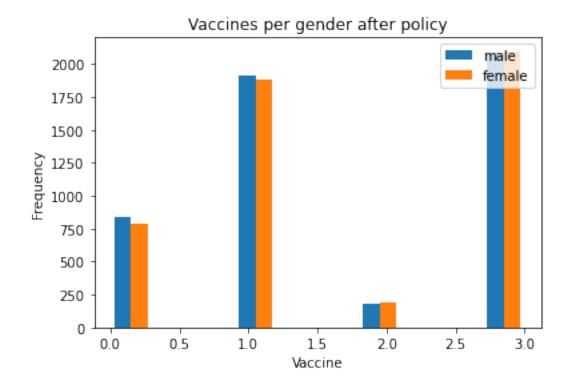
    m_a = np.take(actions, mm)
    m_a = m_a.flatten()

    f_a = np.take(actions, ff)
    f_a = f_a.flatten()

    plt.hist([m_a, f_a],label=['male','female'])
    plt.xlabel('Vaccine')
    plt.ylabel('Frequency')
    plt.title('Vaccines per gender after policy')
    plt.legend(loc='upper right')
    plt.savefig('figures/fair_policy_plot.png')
    plt.show()
```

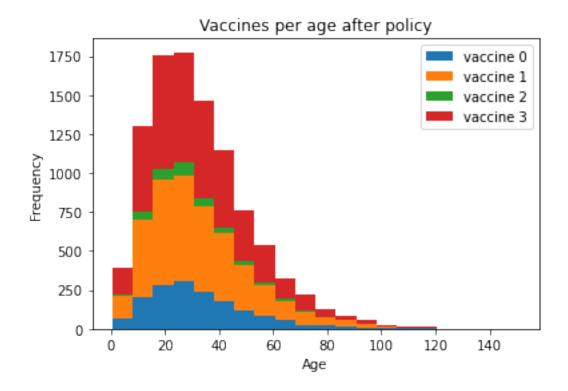
(10000,)

# [28]: plot\_fair\_pol(action\_arr, X)



```
[29]: def plot_vacc_age(data, actions):
          vacc_0 = np.where(actions==0) #no vaccine
          vacc_1 = np.where(actions==1) #vaccine 1
          vacc_2 = np.where(actions==2) #vaccine 2
          vacc_3 = np.where(actions==3) #vaccine 3
          v0_vacc = np.take(data[:,10], vacc_0)
          v0_vacc = v0_vacc.flatten()
          v1_vacc = np.take(data[:,10], vacc_1)
          v1_vacc = v1_vacc.flatten()
          v2_vacc = np.take(data[:,10], vacc_2)
          v2_vacc = v2_vacc.flatten()
          v3_vacc = np.take(data[:,10], vacc_3)
          v3_vacc = v3_vacc.flatten()
          #f_a = np.take(actions, ff)
          #f_a = f_a.flatten()
          plt.hist([v0_vacc, v1_vacc, v2_vacc, v3_vacc],label=['vaccine 0','vaccine_
       →1','vaccine 2','vaccine 3'], bins=20, stacked=True)
          plt.xlabel('Age')
          plt.ylabel('Frequency')
          plt.title('Vaccines per age after policy')
          plt.legend(loc='upper right')
          plt.savefig('figures/fair_age_plot.png')
          plt.show()
```

```
[30]: plot_vacc_age(X, action_arr)
```



```
[31]: def plot_vacc_inc(data, actions):
    vacc_0 = np.where(actions==0) #no vaccine
    vacc_1 = np.where(actions==1) #vaccine 1
    vacc_2 = np.where(actions==2) #vaccine 2
    vacc_3 = np.where(actions==3) #vaccine 3

    v0_vacc = np.take(data[:,12], vacc_0)
    v0_vacc = v0_vacc.flatten()

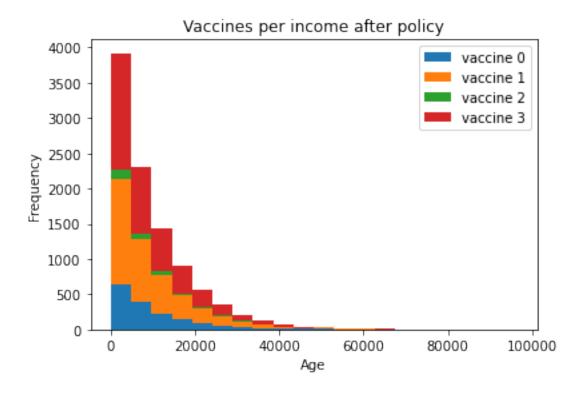
    v1_vacc = np.take(data[:,12], vacc_1)
    v1_vacc = v1_vacc.flatten()

    v2_vacc = np.take(data[:,12], vacc_2)
    v2_vacc = v2_vacc.flatten()

    v3_vacc = np.take(data[:,12], vacc_3)
    v3_vacc = v3_vacc.flatten()

#f_a = np.take(actions, ff)
#f_a = f_a.flatten()
```

### [32]: plot\_vacc\_inc(X, action\_arr)

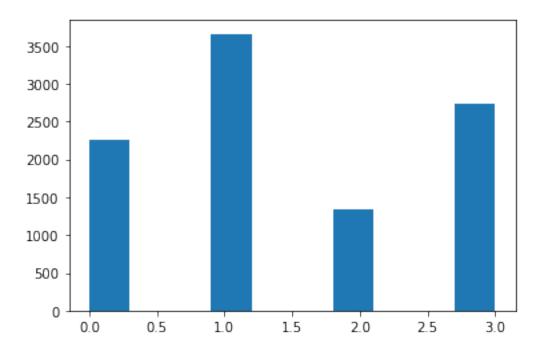


## 1 Thompson with laplace

```
[33]: n_{genes} = 128
      n_vaccines = 3 # DO NOT CHANGE, breaks the simulator.
      n treatments = 4
      n_population = 10_000
      n_{symptoms} = 10
      # symptom names for easy reference
      from covid.auxilliary import symptom_names
      np.random.seed(1)
      population = Population(n_genes, n_vaccines, n_treatments)
      X = population.generate(n_population)
      n_features = X.shape[1]
[34]: action\_space = np.array([-1,0,1,2])
      n_actions = action_space.shape[0]
      action_arr_lap = np.zeros(n_population)
      symptom_arr_lap = np.zeros((n_population, n_symptoms+1))
[35]: policy_1 = Naive(n_actions, action_space)
      model_l = ThompsonSampling(n_actions)
      policy_l.set_model(model)
     Initialising policy with 4 actions
     A = \{ [-1 \ 0 \ 1 \ 2] \}
 []:
[36]: print("With a for loop")
      # The simplest way to work is to go through every individual in the population
      for t in range(n population):
          a_t = policy_l.get_action()
          # Then you can obtain results for everybody
          y_t = population.vaccinate([t], a_t.reshape((1, 1)))
          new_col = y_t[:,[5,7,8]].sum(axis=1)
          new_col0 = new_col > 0
          new_col01 = new_col0.astype(int)
          new_col01 = np.reshape(new_col01,(1,-1))
          y_t_new = np.hstack((y_t,new_col01))
```

```
# Feed the results back in your policy. This allows you to fit the
          # statistical model you have.
          policy_l.observe_with_laplace(a_t, y_t_new[:,10])
          #policy.observe(a_t, y_t_new[:,1])
          action_arr_lap[t] = a_t
          symptom_arr_lap[t] = y_t_new
     With a for loop
[37]: alphas_1, betas_1 = policy_1.model.get_params()
      for i, j in zip(alphas_l, betas_l):
          print(i,j)
      print(alphas_1[0]/(alphas_1[0] + betas_1[0]))
      print(alphas_1[1]/(alphas_1[1] + betas_1[1]))
      print(alphas_1[2]/(alphas_1[2] + betas_1[2]))
     142.49802833539425 3757.5019716646075
     261.14442632401193 7199.8555736759945
     64.81713641433676 1660.1828635856637
     243.64451720493534 6678.355482795076
     0.03653795598343441
     0.035001263412948895
     0.03757515154454304
[38]: plt.hist(action_arr_lap)
                       0., 0., 3659.,
                                             0., 0., 1343.,
                                                                  0.,
                                                                         0.,
```

```
[38]: (array([2263.,
             2735.]),
      array([0., 0.3, 0.6, 0.9, 1.2, 1.5, 1.8, 2.1, 2.4, 2.7, 3.]),
       <BarContainer object of 10 artists>)
```



[]:

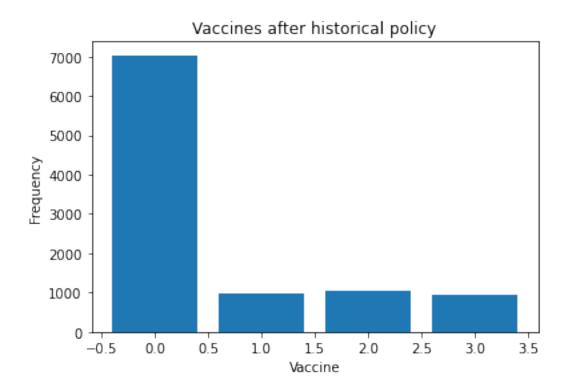
## 2 Utility

```
[39]: utility_laplace = -sum(alphas_l)
[40]: utility_laplace
[40]: -712.1041082786783
[41]: #utility_normal = -sum(alphas)
[42]: utility_normal
[42]: -348.0
[43]: alphas
[43]: array([142.49802834, 261.14442632, 64.81713641, 243.6445172])
```

## 3 Experiment design

#### 3.1 historical

```
[44]: n_{genes} = 128
      n_vaccines = 3 # DO NOT CHANGE, breaks the simulator.
      n_{treatments} = 4
      n_population = 10_000
      n_symptoms = 10
      # symptom names for easy reference
      from covid.auxilliary import symptom_names
      np.random.seed(1)
      population = Population(n_genes, n_vaccines, n_treatments)
      X = population.generate(n_population)
      n_features = X.shape[1]
[45]: new_col = X[:,[5,7,8]].sum(axis=1)
      new_col0 = new_col > 0
      new col01 = new col0.astype(int)
      new_col01 = np.reshape(new_col01,(n_population,-1))
      X_new = np.hstack((X,new_col01))
      hist_pol_utility = -X_new[:,150].sum(axis=0)
[46]: hist_pol_utility
[46]: -243.0
[47]: def plot_hist_policy(data):
          v1 = data[:,147].sum(axis=0)
          v2 = data[:,148].sum(axis=0)
          v3 = data[:,149].sum(axis=0)
          v0 = (data[:,[147,148,149]].sum(axis=1)==0).sum()
          plt.bar([0,1,2,3], [v0,v1,v2,v3])
          plt.xlabel('Vaccine')
          plt.ylabel('Frequency')
          plt.title('Vaccines after historical policy')
          plt.savefig('figures/vaccices_historical_policy.png')
          plt.show()
[48]: plot_hist_policy(X)
```



### [49]: from covid.policy import RandomPolicy

```
[50]: """
      vaccine\_policy = RandomPolicy(n\_vaccines, action\_space) \# make sure to add -1_{\sqcup}
       ⇔for 'no vaccine'
      Y_rand_pol = np.zeros((n_population, n_symptoms))
      A_rand_pol = np.zeros(n_population)
      print("With a for loop")
      # The simplest way to work is to go through every individual in the population
      for t in range(n_population):
          #a_t = vaccine_policy.get_action(X[t])
          a_t = vaccine_policy.get_action(X[t])
          # Then you can obtain results for everybody
          y_t = population.vaccinate([t], a_t)
          \#y_t = population.vaccinate([t], a_t.reshape((1, 1)))
          # 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)
```

```
vaccine_policy.observe(X[t], a_t, y_t)

#print(a_t.shape)
Y_rand_pol[t] = y_t
#A_rand_pol[t] = a_t
"""
```

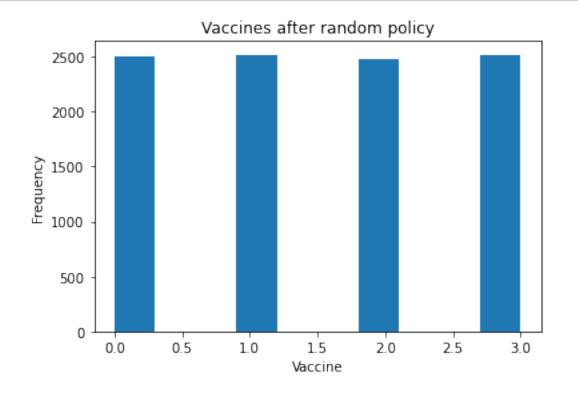
```
[50]: '\nvaccine_policy = RandomPolicy(n_vaccines, action_space) # make sure to add -1
     for \'no vaccine\'\n\nY_rand_pol =
     np.zeros((n_population,n_symptoms))\nA_rand_pol =
     np.zeros(n population)\n\nprint("With a for loop")\n# The simplest way to work
     is to go through every individual in the population\nfor t in
     range(n population):\n
                               #a_t = vaccine_policy.get_action(X[t])\n
                                                                           at =
     vaccine_policy.get_action(X[t])\n
                                          print(a_t)\n
                                                          # Then you can obtain
     results for everybody\n
                                y t = population.vaccinate([t], a t)\n
                                                                          #y t =
     population.vaccinate([t], a_t.reshape((1, 1)))\n
                                                       \n # Feed the results
     back in your policy. This allows you to fit the \n
                                                          # statistical model you
                #vaccine_policy.observe(X[t], a_t, y_t)\n
     vaccine_policy.observe(X[t], a_t, y_t)\n \n
                                                       #print(a_t.shape)\n
     Y_{rand_pol[t]} = y_t n #A_{rand_pol[t]} = a_t n'
```

#### 3.2 random

```
print("Vaccinate'em all")
# Here you can get an action for everybody in the population
A = vaccine_policy.get_action(X)
A = np.array([int(i) for i in A])
A = A.astype(int)
A = A.reshape((n_population, 1))+1
print(A)
# Then you can obtain results for everybody
Y = population.vaccinate(list(range(n_population)), A)
# Feed the results back in your policy.
vaccine_policy.observe(X, A, Y)
```

```
Initialising policy with 3 actions
A = { [-1 0 1 2] }
Vaccinate'em all
[[0]
  [2]
  [1]
```

```
[1]
      [2]
      [3]]
[52]: print(A)
      [[0]]
      [2]
      [1]
      [1]
      [2]
      [3]]
[53]: def plot_random_pol(actions):
          plt.hist(actions)# label=['0','1','2','3'])
          plt.xlabel('Vaccine')
          plt.ylabel('Frequency')
          plt.title('Vaccines after random policy')
          #plt.legend(loc='upper right')
          plt.savefig('figures/vaccines_random_policy.png')
          plt.show()
[54]: plot_random_pol(A)
```



```
[55]: print(Y)

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

[56]: new_col = Y[:,[5,7,8]].sum(axis=1)
       new_col0 = new_col > 0
       new_col01 = new_col0.astype(int)
       new_col01 = np.reshape(new_col01,(n_population,-1))
       Y_new = np.hstack((Y,new_col01))
      rand_pol_utility = -Y_new[:,10].sum(axis=0)

[57]: rand_pol_utility
```

## 4 Bootstrap historical and simulated

#### 4.1

```
[58]: utility_simalated = np.zeros(100)

for i in range(100):
    n_genes = 128
    n_vaccines = 3 # DO NOT CHANGE, breaks the simulator.
    n_treatments = 4
    n_population = 10_000
    n_symptoms = 10

# symptom names for easy reference
from covid.auxilliary import symptom_names

np.random.seed(i)

population = Population(n_genes, n_vaccines, n_treatments)
X = population.generate(n_population)
    n_features = X.shape[1]

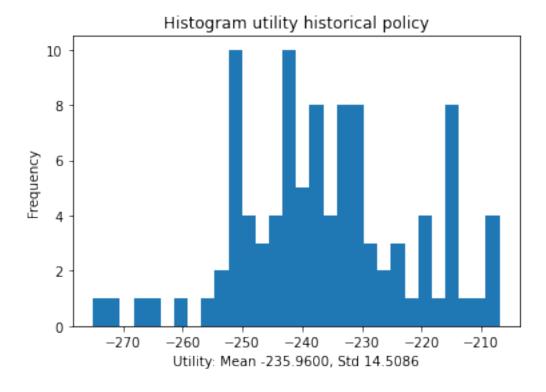
new_col = X[:,[5,7,8]].sum(axis=1)
```

```
new_col0 = new_col > 0
new_col01 = new_col0.astype(int)
new_col01 = np.reshape(new_col01,(n_population,-1))
X_new = np.hstack((X,new_col01))

hist_pol_utility = -X_new[:,150].sum(axis=0)

utility_simalated[i] = hist_pol_utility
```

```
[71]: plt.hist(utility_simalated, bins=30)
   mean = np.mean(utility_simalated)
   std = np.std(utility_simalated)
   plt.ylabel('Frequency')
   xlabel_str = f"Utility: Mean {mean:.4f}, Std {std:.4f}"
   plt.xlabel(xlabel_str)
   plt.title('Histogram utility historical policy')
   plt.savefig('figures/histogram_utility_historical_policy.png')
```



```
[63]: utility_random = np.zeros(100)
```

```
for i in range(100):
    \#n\_genes = 128
    #n_vaccines = 3 # DO NOT CHANGE, breaks the simulator.
    \#n\_treatments = 4
    \#n_population = 10_000
    \#n_symptoms = 10
    # symptom names for easy reference
    #from covid.auxilliary import symptom_names
    np.random.seed(i)
    population = Population(n_genes, n_vaccines, n_treatments)
    X = population.generate(n_population)
    n_features = X.shape[1]
    vaccine_policy = RandomPolicy(n_vaccines, action_space)
    print("Vaccinate'em all")
    # Here you can get an action for everybody in the population
    A = vaccine_policy.get_action(X)
    A = np.array([int(i) for i in A])
    A = A.astype(int)
    A = A.reshape((n population, 1))+1
    #print(A)
    # Then you can obtain results for everybody
    Y = population.vaccinate(list(range(n_population)), A)
    # Feed the results back in your policy.
    vaccine_policy.observe(X, A, Y)
    new_col = Y[:,[5,7,8]].sum(axis=1)
    new_col0 = new_col > 0
    new_col01 = new_col0.astype(int)
    new_col01 = np.reshape(new_col01,(n_population,-1))
    Y_new = np.hstack((Y,new_col01))
    rand_pol_utility = -Y_new[:,10].sum(axis=0)
    utility_random[i] = rand_pol_utility
```

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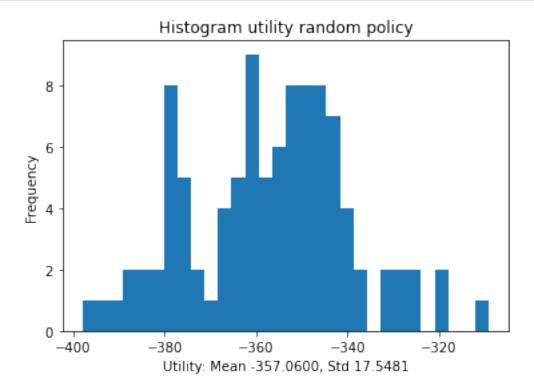
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[64]: print("======="")
print("========="")
print("========="")
```

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```
[70]: plt.hist(utility_random,bins=30)
mean = np.mean(utility_random)
std = np.std(utility_random)
plt.ylabel('Frequency')
xlabel_str = f"Utility: Mean {mean:.4f}, Std {std:.4f}"
plt.xlabel(xlabel_str)
plt.title('Histogram utility random policy')
plt.savefig('figures/histogram_utility_random_policy.png')
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