## FinalStoch

February 19, 2022

# 1 Q1

```
[276]: import numpy as np
       import json
       from tqdm import tqdm
       train_data = np.load('train_data.npy')
       class HiddenMarkovModel():
           def __init__(self, initial_probabilities, transition_probs, emission_probs):
               self.transition_probs = transition_probs
               self.emission_probs = emission_probs
               self.initial_probabilities = initial_probabilities
               self._log_transition_probs = self._compute_log_probs(transition_probs)
               self. log emission probs = self. compute log probs(emission probs)
               if not self._valid_hmm(): raise ValueError("not a valid markov model")
           def baum_welch(self, seq, iterations=100):
               for itr in tqdm(xrange(iterations)):
                   initial_update = dict([(s, 0.0) for s in self.transition_probs])
                   transition_update1 = dict(
                       [(s, dict([(s, 0) for s in self.transition_probs])) for s in_
        →self.transition_probs])
                   transition_update2 = dict([(s, 0) for s in self.transition_probs])
                   emission_update1 = dict([(s, dict([(o, 0) for o in self.
        →emission_probs[s]])) for s in self.emission_probs])
                   emission_update2 = dict([(s, dict([(o, 0) for o in self.
       →emission_probs[s]])) for s in self.emission_probs])
                   for s in seq:
                       init, tr1, tr2, em1, em2 = self._baum_welch_iteration(s)
                       for s in self.transition_probs:
                           initial_update[s] += init[s]
                       for s in self.transition_probs:
                           for x in self.transition_probs[s]:
                               transition_update1[s][x] += tr1[s][x]
```

```
for s in self.transition_probs:
                   transition_update2[s] += tr2[s]
               for s in self.emission_probs:
                   for o in self.emission_probs[s]:
                       emission_update1[s][o] += em1[s][o]
                       emission_update2[s][o] += em2[s][o]
           self.initial_probabilities = dict([(s, (1 / len(seq)) *_
→initial_update[s]) for s in self.transition_probs])
           for i in self.transition_probs:
               for j in self.transition_probs[i]:
                   self.transition_probs[i][j] = (transition_update1[i][j] /__
→transition_update2[i])
           for i in self.emission_probs:
               for o in self.emission probs[i]:
                   self.emission_probs[i][o] = (emission_update1[i][o] /_u
→emission_update2[i][o])
           self._log_transition_probs = self._compute_log_probs(self.
→transition_probs)
           self._log_emission_probs = self._compute_log_probs(self.
→emission_probs)
   def _baum_welch_iteration(self, seq, cons=100):
       forward_probabilities = [dict([(s, 0.0) for s in self.
→transition_probs]) for _ in xrange(len(seq))]
       eps = 0.000000000000000000001
       forward_probabilities[0] = dict(
           [(s, np.log(self.initial_probabilities[s] + eps) + self.
→ log_emission_probs[s][seq[0]]) for s in
            self.transition probs])
       for i in xrange(1, len(seq)):
           for j in self.transition_probs:
               a = self._log_emission_probs[j][seq[i]]
               b = self._log_sum_exp([forward_probabilities[i - 1][k] + self.
→ log_transition_probs[k][j] for k in
                                      self.transition_probs[j]])
               forward_probabilities[i][j] = a + b
       backward_probabilities = [dict([(s, 0.0) for s in self.
→transition_probs]) for _ in xrange(len(seq))]
```

```
for j in self.transition_probs:
           backward_probabilities[len(seq) - 1][j] = 0
       for i in range(len(seq) - 2, -1, -1):
           for j in self.transition_probs:
               backward_probabilities[i][j] = self._log_sum_exp([
                   backward_probabilities[i + 1][k] + self.
→_log_transition_probs[j][k] + self._log_emission_probs[k][
                       seq[i + 1]]
                   for k in self.transition_probs[j]
               ])
       gamma = [dict([(s, 0) for s in self.transition_probs]) for t in_
→xrange(len(seq))]
       for t in xrange(len(seq)):
           norm = sum(
               [np.exp(forward_probabilities[t][k] +__
→backward_probabilities[t][k] + 1200) for k in
                self.transition_probs])
           for j in self.transition_probs:
               gamma[t][j] = np.exp(forward_probabilities[t][j] +

→backward_probabilities[t][j] + 1200) / norm
       zeta = [dict([(s, dict([(s, 0) for s in self.transition_probs])) for su
→in self.transition_probs]) for i in
               xrange(len(seq))]
       for t in xrange(len(seq) - 1):
           norm = sum([sum([
               np.exp(forward_probabilities[t][i] + self.
→ log_transition_probs[i][j]
                      + backward_probabilities[t + 1][j] + self.
\rightarrow_log_emission_probs[j][seq[t + 1]] + 1200)
               for j in self.transition_probs[i]]) for i in self.
→transition_probs])
           for i in self.transition_probs:
               for j in self.transition_probs:
                   if j in self._log_transition_probs[i]:
                       zeta[t][i][j] = np.exp(forward_probabilities[t][i] + L
⇒self._log_transition_probs[i][j]
                                               + backward_probabilities[t +_
→1][j] + self._log_emission_probs[j][
                                                   seq[t + 1]] + 1200) / norm
                   else:
                       zeta[t][i][j] = 0
```

```
init = dict([(s, gamma[0][s]) for s in self.transition_probs])
       trans_prob1 = {}
       trans_prob2 = {}
       for i in self.transition_probs:
           trans_prob2[i] = 0
           for t in range(len(seq) - 1):
               trans_prob2[i] += gamma[t][i]
       for i in self.transition probs:
           trans_prob1[i] = {}
           for j in self.transition_probs[i]:
               trans_prob1[i][j] = 0
               for t in range(len(seq) - 1):
                   trans_prob1[i][j] += zeta[t][i][j]
       em_prob1 = \{\}
       for i in self.emission_probs:
           em_prob1[i] = {}
           for o in self.emission_probs[i]:
               em_prob1[i][o] = sum([(seq[t] == o) * gamma[t][i] for t in_u
→xrange(len(seq))])
       em_prob2 = \{\}
       for i in self.emission_probs:
           em_prob2[i] = {}
           for o in self.emission_probs[i]:
               em_prob2[i][o] = sum([gamma[t][i] for t in xrange(len(seq))])
       return init, trans_prob1, trans_prob2, em_prob1, em_prob2
   def _log_sum_exp(self, arr):
       a = np.max(arr)
       # print(arr)
       return a + np.log(np.sum([np.exp(x - a) for x in arr]))
   def _compute_log_probs(self, prob_matrix):
       log_prob_matrix = {}
       for i in prob_matrix:
           log_prob_matrix[i] = {}
           for j in prob_matrix[i]:
               log_prob_matrix[i][j] = np.log(prob_matrix[i][j] + 0.
→00000000001)
       return log_prob_matrix
   def _valid_hmm(self):
       if (self.transition_probs != None) and (self.emission_probs != None):
           valid = True
```

```
# print("1", set(self.transition_probs.keys()), set(self.
\hookrightarrow emission_probs.keys()))
           valid &= (set(self.transition_probs.keys()) == set(self.
→emission probs.keys()))
           return valid
       else:
           return True
   def likelihood(self, seq, log=True):
       alpha = [dict([(s, 0) for s in self.transition_probs]) for _ in__
alpha[0] = dict([(i, np.log(self.initial_probabilities[i]) + self.
→ log_emission_probs[i][seq[0]]) for i in
                        self.emission_probs])
       for i in xrange(1, len(seq)):
           for j in self.transition_probs:
               alpha[i][j] = self._log_emission_probs[j][seq[i]] + self.
→_log_sum_exp([
                   alpha[i - 1][k] + self._log_transition_probs[k][j] for k in_
⇔self.transition_probs[j]])
       if log:
           return self._log_sum_exp(list(alpha[-1].values()))
       else:
           return sum([np.exp(p) for p in alpha[-1].values()]) # return the
\rightarrow sum of the last row
   def backward_likelihood(self, seq, log=True):
       backward_probabilities = [dict([(s, 0.0) for s in self.
→transition_probs]) for _ in xrange(len(seq))]
       for j in self.transition_probs:
           backward_probabilities[len(seq) - 1][j] = 0
       for i in range(len(seq) - 2, -1, -1):
           for j in self.transition_probs:
               backward_probabilities[i][j] = self._log_sum_exp([
                   backward_probabilities[i + 1][k] + self.
→ log_transition_probs[j][k] + self._log_emission_probs[k][
                       seq[i + 1]]
                   for k in self.transition_probs[j]
               ])
       if log:
```

```
return self._log_sum_exp([(backward_probabilities[0][k] + self.
 → log_emission_probs[k][seq[0]] + np.log(
                self.initial_probabilities[k])) for k in self.transition_probs])
        else:
            # return sum([np.exp(p) for p in alpha[-1].values()]) # return the
 \rightarrowsum of the last row
            return
    def viterbi(self, seq):
        T = [dict([(s, 0) \text{ for s in self.transition_probs}]) \text{ for } in_{\sqcup}]
 →xrange(len(seq))]
        T[0] = dict([(i, np.log(self.initial_probabilities[i]) + self.
 → log_emission_probs[i][seq[0]]) for i in
                      self.emission_probs])
        T_bp = [dict([(s, 0) \text{ for s in self.transition_probs}]) \text{ for } in_{\sqcup}
 →xrange(len(seq))]
        T_bp[0] = dict([(i, i) for i in self.emission_probs])
        for i in xrange(1, len(seq)):
            for j in self.transition_probs:
                T_bp[i][j] = max(self.transition_probs[j],
                                  key=lambda k: T[i - 1][k] + self.
 →_log_transition_probs[k][j] +
                                                 self.
→_log_emission_probs[j][seq[i]])
                prev_state = T_bp[i][j]
                T[i][j] = T[i - 1][prev_state] + self.
 →_log_transition_probs[prev_state][j] + self._log_emission_probs[j][seq[i]]
        # construct the most likely sequence
        state_sequence = [None for i in xrange(len(seq))]
        state_sequence[-1] = max(self.transition_probs,
                                  key=lambda k: T[-1][k]
        for i in xrange(len(seq) - 1):
            i = len(seq) - 2 - i # go in reverse order
            state_sequence[i] = T_bp[i + 1][state_sequence[i + 1]]
        return state_sequence
transition = {
    0: {1: 0.5, 5: 0.5},
    1: {0: 0.5, 2: 0.5},
    2: {1: 0.5, 3: 0.5},
```

```
3: \{2: (1 / 3), 4: (1 / 3), 8: (1 / 3)\},
         4: {3: 0.5, 9: 0.5},
         5: \{0: (1 / 3), 6: (1 / 3), 10: (1 / 3)\},
         6: {5: 0.25, 7: 0.25, 11: 0.25, 12: 0.25},
         7: \{6: (1/3), 8: (1/3), 12: (1/3)\},\
         8: {3: 0.2, 7: 0.2, 9: 0.2, 12: 0.2, 13: 0.2},
         9: \{8: (1/3), 4: (1/3), 14: (1/3)\},
         10: {5: 0.5, 15: 0.5},
         11: \{6: (1/3), 12: (1/3), 16: (1/3)\},
         12: {6: 0.125, 7: 0.125, 8: 0.125, 11: 0.125, 13: 0.125, 16: 0.125, 17: 0.
      \rightarrow125, 18: 0.125},
         13: {8: 0.25, 12: 0.25, 14: 0.25, 18: 0.25},
         14: \{9: (1 / 3), 13: (1 / 3), 19: (1 / 3)\},
         15: {10: 0.5, 20: 0.5},
         16: {11: (0.25), 12: (0.25), 17: (0.25), 21: 0.25},
         17: {12: (0.25), 16: (0.25), 18: (0.25), 22: 0.25},
         18: {12: 0.2, 13: 0.2, 17: 0.2, 19: 0.2, 23: 0.2},
         19: \{14: (1 / 3), 18: (1 / 3), 24: (1 / 3)\},
         20: {15: 0.5, 21: 0.5},
         21: {16: (1 / 3), 20: (1 / 3), 22: (1 / 3)},
         22: {17: (1 / 3), 21: (1 / 3), 23: (1 / 3)},
         23: \{18: (1 / 3), 22: (1 / 3), 24: (1 / 3)\},
         24: {19: 0.5, 23: 0.5}
     }
     emission = {}
     for i in range(25):
         emission[i] = \{0: (1 / 11), 1: (1 / 11), 2: (1 / 11), 3: (1 / 11), 4: (1 / 11)\}
     \rightarrow11), 5: (1 / 11), 6: (1 / 11), 7: (1 / 11),
             8: (1 / 11), 9: (1 / 11), 10: (1 / 11)}
     initial = {0: 0.04, 1: 0.04, 2: 0.04, 3: 0.04, 4: 0.04, 5: 0.04, 6: 0.04, 7: 0.
      \rightarrow04, 8: 0.04, 9: 0.04, 10: 0.04, 11: 0.04,
                12: 0.04, 13: 0.04, 14: 0.04, 15: 0.04, 16: 0.04, 17: 0.04, 18: 0.
     →04, 19: 0.04, 20: 0.04, 21: 0.04, 22: 0.04,
                23: 0.04, 24: 0.04}
[]: model = HiddenMarkovModel(initial, transition, emission)
     model.baum_welch(train_data)
     output = json.dumps(model.transition_probs)
     f = open("transition_probs.json", "w")
     f.write(output)
     f.close()
     output = json.dumps(model.emission probs)
```

f = open("emission\_probs.json", "w")

```
f.write(output)
       f.close()
       output = json.dumps(model.initial_probabilities)
       f = open("initial_probs.json", "w")
       f.write(output)
       f.close()
[278]: from pathlib import Path
       initial_converged = json.loads(Path('initial_probs.json').read_text())
       transition converged = json.loads(Path('transition probs.json').read_text())
       emission_converged = json.loads(Path('emission_probs.json').read_text())
       initial_converged = {int(k): v for k, v in initial_converged.items()}
       transition converged = {int(k): {int(x): z for x, z in v.items()} for k, v in_
       →transition_converged.items()}
       emission_converged = {int(k): {int(x): z for x, z in v.items()} for k, v inu
       →emission_converged.items()}
       model =
        → HiddenMarkovModel(initial_converged, transition_converged, emission_converged)
```

## 2 Q2

```
[275]: test_data= np.load('test_data.npy')
      with open('forward_backward_output.txt', 'w') as f:
          for test in tqdm(test_data):
              f prob = model.likelihood(test, log=True)
              b_prob = model.backward_likelihood(test, log=True)
              print(f_prob, "\t", np.exp(f_prob))
              print(b_prob, "\t", np.exp(b_prob))
              print("----")
              f.write(str(np.exp(f\_prob))+"\t"+str(np.exp(b\_prob))+"\n")
                    | 2/100 [00:00<00:13, 7.21it/s]
       2%|
      -87.40822221594169
                             1.0941866674560593e-38
      -87.40822221594168
                             1.0941866674560748e-38
      ______
      -96.61173593224723
                             1.1016818575067096e-42
      -96.61173593224727
                             1.1016818575066626e-42
        4%|
             | 4/100 [00:00<00:12, 7.44it/s]
```

-92.86014462227607	4.691930838968532e-41
-92.8601446222761	4.691930838968398e-41
-94.30465596383002 -94.30465596383004	
6%    6/1	100 [00:00<00:12, 7.41it/s]
-82.78489959778312	1.1142301856444717e-36
-82.78489959778302	1.1142301856445827e-36
-92.82644902727122	4.852722004156363e-41
-92.82644902727121	4.852722004156433e-41
8%    8/3	100 [00:01<00:12, 7.33it/s]
-89.58799416423282	1.2371674629415378e-39
-89.58799416423292	1.2371674629414148e-39
	1.4776657300430085e-34 1.4776657300429664e-34
10%    10,	/100 [00:01<00:12, 7.11it/s]
-85.43701721839028	7.855518790205181e-38
-85.43701721839041	7.855518790204177e-38
	- 1.5328619896912804e-48 1.532861989691215e-48
12%    12/	100 [00:01<00:12, 7.17it/s]
-86.32498227308703	3.232485224350067e-38
-86.32498227308707	3.2324852243499294e-38
-89.46806690436863	1.3948009576418284e-39
-89.46806690436867	1.3948009576417688e-39
14%    14/	100 [00:01<00:11, 7.32it/s]
-76.12267947149063	8.71646434179358e-34
-76.12267947149067	8.716464341793208e-34
-92.233509597237	8.780024653149411e-41
-92.23350959723705	8.780024653149036e-41
16%    16/	100 [00:02<00:11, 7.38it/s]

-79.720734 -79.720734		2.3862983304 2.3862983304	
-88.967777 -88.967777		2.3003037080 2.3003037080	
18%	18/100	[00:02<00:11,	7.39it/s]
-87.676163 -87.676163		8.370013896 8.370013896	
-93.331999 -93.331999		2.9270319758 2.9270319758	
20%	20/100	[00:02<00:10,	7.44it/s]
-74.678111 -74.678111		3.695798363° 3.695798363°	
-94.411261 -94.411261		9.947411496 9.947411496	
22%	22/100	[00:03<00:10,	7.21it/s]
-96.244204 -96.244204		1.591010154 1.591010154	
-89.208352 -89.208352		1.808443561 1.808443561	
24%	24/100	[00:03<00:10,	7.25it/s]
-95.206281 -95.206281		4.4919760893 4.4919760893	
-95.863686 -95.863686		2.3277156928 2.3277156928	
26%	26/100	[00:03<00:10,	7.38it/s]
-88.718219 -88.718219		2.9523419619 2.9523419619	
-91.414677 -91.414677		1.9911781840 1.9911781840	
28%	28/100	[00:03<00:09,	7.26it/s]

-84.46502179894641 -84.46502179894637	2.0763813474967497e-37 2.0763813474968382e-37
-90.39670154992757 -90.39670154992758	5.510757934273053e-40 5.510757934272975e-40
30%    30/100	[00:04<00:10, 6.82it/s]
-83.43356570651397 -83.43356570651393	5.8245557722310415e-37 5.82455577223129e-37
-85.24263616957687 -85.24263616957694	9.54099105462171e-38 9.54099105462103e-38
32%    32/100	[00:04<00:10, 6.55it/s]
-108.3600088687375 -108.36000886873757	8.706549817337239e-48 8.70654981733662e-48
-80.69702990879753 -80.69702990879752	8.989286136980333e-36 8.989286136980461e-36
34%    34/100	[00:04<00:09, 6.82it/s]
-88.93375316634295 -88.9337531663429	2.3799166365846443e-39 2.3799166365847797e-39
-79.7246428064165 -79.72464280641643	2.3769911100295996e-35 2.3769911100297685e-35
36%    36/100	[00:05<00:09, 6.52it/s]
-97.77431326612859 -97.77431326612852	3.4447307085015524e-43 3.444730708501797e-43
-82.83795802259169 -82.83795802259172	1.056651900881171e-36 1.0566519008811409e-36
39%    39/100	[00:05<00:07, 8.66it/s]
-96.23146604232804	1.6114072465802652e-42
-96.23146604232801	1.611407246580311e-42
-85.37467418816064	8.360843721370662e-38
-85.37467418816067	8.360843721370424e-38
-89.48353211462116	1.3733960101993952e-39

-89.48353211462118 	1.373396010199356e-39
41%    41/10	0 [00:05<00:06, 9.76it/s]
-97.0700156740002 -97.07001567400025	6.966711632751486e-43 6.966711632751189e-43
-100.72028238787712 -100.72028238787705	1.81024411049534e-44 1.8102441104954685e-44
-84.15934908173891 -84.15934908173895	2.8187664468113575e-37 2.8187664468112372e-37
45%    45/100	[00:05<00:04, 11.33it/s]
-88.55051706951673 -88.55051706951677	3.491395654405544e-39 3.491395654405395e-39
-83.0923545087351 -83.09235450873508	8.19311307143034e-37 8.193113071430458e-37
-84.56911402210281 -84.56911402210285	1.8711148293847844e-37 1.8711148293847046e-37
47%    47/100	0 [00:05<00:04, 11.91it/s]
47%   47/100 -104.77868156501671 -104.77868156501674	3.1274960885960125e-46 3.1274960885959234e-46
-104.77868156501671	3.1274960885960125e-46
-104.77868156501671 -104.77868156501674 	3.1274960885960125e-46 3.1274960885959234e-46 9.056516524497316e-38
-104.77868156501671 -104.77868156501674 	3.1274960885960125e-46 3.1274960885959234e-46 9.056516524497316e-38 9.05651652449693e-38 1.1455882873647317e-34
-104.77868156501671 -104.77868156501674 	3.1274960885960125e-46 3.1274960885959234e-46 9.056516524497316e-38 9.05651652449693e-38 1.1455882873647317e-34 1.145588287364504e-34
-104.77868156501671 -104.77868156501674 	3.1274960885960125e-46 3.1274960885959234e-46  9.056516524497316e-38 9.05651652449693e-38  1.1455882873647317e-34 1.145588287364504e-34  0 [00:06<00:04, 12.16it/s] 4.842165982706708e-39
-104.77868156501671 -104.77868156501674	3.1274960885960125e-46 3.1274960885959234e-46  9.056516524497316e-38 9.05651652449693e-38  1.1455882873647317e-34 1.145588287364504e-34  0 [00:06<00:04, 12.16it/s] 4.842165982706708e-39 4.842165982706777e-39  1.1283249071996753e-39

-87.0192308458694 -87.01923084586947	1.6144634750761637e-38 1.614463475076049e-38
-94.67955922563597 -94.67955922563598	7.606587742950879e-42 7.60658774295077e-42
-80.08134438715506 -80.08134438715507	1.6638494568055258e-35 1.6638494568055023e-35
57%    57/100	[00:06<00:03, 12.76it/s]
-78.74913040170539 -78.74913040170547	6.305030791610312e-35 6.305030791609774e-35
-91.9892161841703 -91.98921618417036	1.1209623255062784e-40 1.1209623255062148e-40
-84.86323534339658 -84.86323534339657	1.3943287231048913e-37 1.3943287231049111e-37
59%    59/100	[00:06<00:03, 12.90it/s]
-95.44014252875895 -95.44014252875903	3.5552713056826696e-42 3.555271305682417e-42
-77.67738555392103	1.841365853404063e-34
-77.67738555392107	1.8413658534039845e-34
-99.46139090397452	6.374802994627711e-44
-99.46139090397452	6.374802994627711e-44
63%    63/100	[00:07<00:02, 13.02it/s]
-79.3590738144842	3.4260378246379254e-35
-79.35907381448418	3.4260378246380227e-35
-89.26822913313974	1.7033373840019986e-39
-89.26822913313983	1.7033373840018534e-39
-92.34380214104321	7.863145142123464e-41
-92.34380214104314	7.863145142124023e-41
67%    67/100	[00:07<00:02, 13.44it/s]
-97.55762225940336	4.278218734086457e-43
-97.55762225940337	4.278218734086397e-43
-74.08024314532335	6.719843889124379e-33

-74.08024314532337	6.719843889124283e-33
-89.46985072590093	1.3923150994831403e-39
-89.46985072590093	1.3923150994831403e-39
-93.53214853494003	2.3960948776385237e-41
-93.53214853494005	2.3960948776384895e-41
69%    69/100	[00:07<00:02, 13.45it/s]
-89.75822792080102	1.0435107047283274e-39
-89.75822792080093	1.0435107047284163e-39
-94.11244207472551	1.3411758618634945e-41
-94.1124420747255	1.3411758618635137e-41
-92.7607900983177	5.182039550720493e-41
-92.76079009831767	5.182039550720641e-41
73%    73/100	[00:07<00:02, 12.92it/s]
-85.71546437227586	5.946299905501121e-38
-85.71546437227586	5.946299905501121e-38
-101.76610049832081	6.361274093218553e-45
-101.76610049832074	6.361274093219005e-45
-85.80169650808843	5.455024036966462e-38
-85.80169650808841	5.455024036966539e-38
75%    75/100	[00:08<00:02, 12.14it/s]
-95.83219188936275	2.402192870878932e-42
-95.83219188936279	2.4021928708788293e-42
-92.20531347753138	9.031110467468759e-41
-92.20531347753135	9.031110467469016e-41
-100.95550384875033	1.4308093210985211e-44
-100.95550384875031	1.4308093210985413e-44
79%    79/100	[00:08<00:01, 12.33it/s]
-95.51918532740343	3.2850720290462225e-42
-95.51918532740349	3.285072029046036e-42
-83.52554612554279	5.3127113406937134e-37
-83.52554612554289	5.312711340693185e-37

	1.7549459603667815e-35 1.7549459603668564e-35
83%    83/100	[00:08<00:01, 13.00it/s]
-79.89658996072365	2.001482776400959e-35
-79.89658996072366	2.0014827764009307e-35
-100.26602924093567	2.851128414417456e-44
-100.26602924093565	2.851128414417496e-44
-87.77547967364865	7.578679326328344e-39
-87.77547967364868	7.578679326328129e-39
-89.77416246064237	1.0270146196328767e-39
-89.77416246064236	1.0270146196328914e-39
87%    87/100	[00:09<00:00, 13.68it/s]
-86.78206685282748	2.0465729876691823e-38
-86.78206685282753	2.046572987669095e-38
-82.54134969095658	1.4215014282104077e-36
-82.54134969095655	1.4215014282104481e-36
-89.94704926650442	8.639584170620975e-40
-89.94704926650445	8.639584170620728e-40
-86.57053316188643	2.5286874455360856e-38
-86.57053316188644	2.5286874455360496e-38
89%    89/100	[00:09<00:00, 13.58it/s]
-77.56779224395252	2.0546405500782094e-34
-77.56779224395248	2.054640550078297e-34
-99.93085977925003	3.9863830359225533e-44
-99.93085977925004	3.9863830359224966e-44
-105.45007169439032	1.5981434045489174e-46
-105.45007169439035	1.5981434045488719e-46
93%    93/100 [	[00:09<00:00, 13.30it/s]
-94.75320538551817	7.066522700501191e-42
-94.75320538551826	7.066522700500588e-42

```
-95.06009364773068
                             5.199072571359702e-42
     -95.06009364773072
                             5.1990725713594805e-42
      -90.88138295281098
                             3.394039371099483e-40
     -90.88138295281095
                             3.3940393710995794e-40
               | 95/100 [00:09<00:00, 13.49it/s]
     -91.24412018883183
                             2.3614681685546735e-40
      -91.24412018883181
                             2.361468168554707e-40
      _____
      -90.69811940737132
                             4.076685628753131e-40
     -90.69811940737128
                             4.076685628753305e-40
      _____
      -96.87895765951693
                             8.433411191589496e-43
     -96.87895765951696
                             8.433411191589257e-43
      99%1
               | 99/100 [00:09<00:00, 12.81it/s]
     -95.96284038104089
                             2.10798712865435e-42
      -95.96284038104089
                             2.10798712865435e-42
      -90.09279498686587
                             7.467861897169648e-40
      -90.0927949868658
                             7.46786189717018e-40
      _____
     -100.38579699711615
                             2.529311407187092e-44
     -100.3857969971162
                             2.529311407186948e-44
      _____
     100%|
               | 100/100 [00:10<00:00, 9.93it/s]
      -89.59896282555579
                             1.223671543435187e-39
      -89.59896282555573
                             1.2236715434352567e-39
[277]: test data = np.load('test data.npy')
      with open('viterbi_output.txt', 'w') as f:
          for test in tqdm(test_data):
              s = model.viterbi(test)
              print(s)
              print("======="")
              f.write(str(s)+"\n")
       8%1
                   | 8/100 [00:00<00:02, 37.02it/s]
```

[16, 11, 12, 16, 12, 18, 17, 18, 17, 22, 23, 22, 23, 22, 21, 16, 21, 20, 21, 22, 23, 22, 21, 22, 23, 22, 17, 12, 8, 9, 4, 3, 2, 3, 4, 3, 4, 3, 8, 7, 8, 13, 12,

11, 12, 8, 7, 8, 3, 2]

\_\_\_\_\_

[16, 21, 16, 12, 17, 12, 11, 16, 21, 22, 17, 12, 16, 11, 12, 6, 11, 16, 11, 12, 11, 6, 5, 10, 5, 6, 11, 12, 16, 21, 20, 15, 20, 21, 16, 21, 20, 15, 20, 21, 16, 21, 16, 12, 6, 5, 6, 5, 6, 5, 6, 12]

\_\_\_\_\_

[21, 16, 12, 11, 6, 5, 6, 11, 6, 11, 6, 7, 8, 7, 8, 12, 17, 12, 11, 12, 7, 8, 9, 4, 3, 4, 9, 8, 12, 16, 21, 20, 21, 20, 15, 20, 21, 16, 12, 17, 12, 18, 17, 22, 17, 12, 13, 18, 17, 18]

\_\_\_\_\_

[17, 12, 11, 12, 11, 12, 16, 21, 16, 21, 16, 11, 12, 17, 22, 23, 22, 17, 12, 11, 12, 6, 7, 8, 13, 14, 19, 14, 19, 14, 13, 12, 16, 21, 16, 11, 12, 6, 11, 12, 8, 7, 8, 13, 12, 18, 19, 18, 19, 14]

\_\_\_\_\_

[21, 16, 11, 6, 7, 6, 12, 18, 17, 18, 17, 12, 18, 12, 6, 11, 16, 21, 16, 12, 11, 6, 5, 6, 5, 10, 5, 0, 5, 10, 5, 10, 5, 10, 5, 0, 1, 2, 1, 0, 1, 2, 1, 0, 1, 0, 1, 0, 1, 0]

\_\_\_\_\_

[20, 21, 16, 21, 22, 23, 22, 23, 22, 17, 18, 19, 18, 19, 18, 17, 18, 12, 18, 17, 22, 23, 22, 17, 12, 18, 19, 24, 23, 22, 23, 22, 23, 22, 21, 16, 11, 6, 12, 18, 17, 12, 13, 18, 17, 18, 17, 12, 7, 6]

\_\_\_\_\_

[5, 10, 5, 0, 5, 6, 11, 16, 12, 16, 21, 22, 23, 22, 21, 22, 17, 12, 16, 21, 22, 23, 22, 23, 22, 17, 18, 17, 22, 17, 18, 17, 22, 23, 24, 19, 18, 12, 18, 19, 18, 19, 18, 17, 18, 19, 24, 19, 18, 19]

\_\_\_\_\_

[18, 12, 11, 12, 18, 12, 17, 12, 18, 17, 22, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 22, 23, 22, 17, 22, 23, 22, 17, 12, 17, 22, 23, 22, 21, 20, 15, 20, 15, 20, 15]

\_\_\_\_\_

[12, 11, 6, 5, 10, 5, 6, 5, 6, 7, 6, 5, 10, 5, 0, 5, 10, 5, 10, 5, 6, 7, 8, 13, 14, 13, 14, 13, 8, 13, 12, 18, 17, 22, 17, 22, 23, 22, 17, 22, 17, 12, 13, 18, 19, 18, 17, 18, 13, 14]

\_\_\_\_\_

16% | 16/100 [00:00<00:02, 37.65it/s]

[19, 14, 13, 8, 3, 2, 1, 0, 1, 0, 1, 2, 1, 0, 1, 0, 5, 0, 1, 0, 1, 0, 1, 2, 3, 2, 1, 2, 3, 8, 9, 4, 9, 8, 12, 13, 8, 9, 4, 9, 14, 13, 14, 13, 14, 19, 24, 23, 22, 23]

\_\_\_\_\_

[5, 0, 1, 2, 3, 4, 9, 8, 12, 17, 18, 19, 18, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 22, 17, 22, 23, 22, 23, 22, 23, 22, 17, 12, 6, 5, 6, 7, 8, 7, 6, 12, 8, 7, 6, 7, 6, 11, 12]

[15, 20, 15, 20, 15, 20, 15, 20, 21, 22, 23, 22, 23, 22, 17, 18, 17, 22, 17, 22, 23, 22, 17, 18, 12, 16, 12, 13, 18, 19, 14, 19, 18, 12, 17, 22, 23, 24, 19, 24, 23, 22, 17, 18, 17, 12, 16, 12, 16, 21]

[4, 3, 2, 1, 0, 1, 2, 1, 0, 1, 0, 1, 2, 1, 2, 1, 2, 3, 4, 3, 4, 3, 8, 13, 12, 18, 17, 22, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 18, 12, 16, 11, 6, 5, 10, 5, 6]

================

[15, 20, 15, 20, 21, 20, 21, 20, 15, 20, 21, 16, 11, 6, 5, 10, 5, 6, 11, 12, 11, 16, 21, 20, 21, 16, 11, 12, 18, 12, 11, 16, 11, 12, 18, 17, 22, 23, 22, 21, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15]

\_\_\_\_\_

[6, 11, 12, 18, 17, 18, 19, 14, 19, 24, 23, 22, 23, 22, 17, 18, 19, 24, 23, 22, 23, 22, 17, 18, 17, 18, 17, 12, 11, 16, 11, 12, 18, 17, 22, 23, 22, 23, 22, 17, 18, 12, 11, 6, 5, 0, 1, 2, 3, 4]

\_\_\_\_\_

[22, 17, 12, 6, 11, 16, 21, 22, 17, 22, 23, 22, 23, 22, 17, 18, 17, 22, 23, 22, 23, 22, 23, 22, 17, 12, 8, 3, 2, 1, 2, 3, 8, 13, 12, 17, 12, 11, 12, 18, 12, 16, 21, 22, 17, 22, 23, 22, 23, 22]

\_\_\_\_\_

[16, 11, 12, 11, 12, 11, 16, 12, 16, 11, 12, 17, 22, 21, 22, 23, 22, 17, 18, 17, 22, 23, 22, 23, 22, 23, 22, 17, 22, 21, 16, 21, 20, 21, 22, 23, 22, 23, 22, 23, 24, 19, 14, 13, 12, 6, 5, 6, 5, 10]

-----

24%| | 24/100 [00:00<00:02, 37.39it/s]

[2, 1, 0, 5, 10, 5, 6, 7, 12, 16, 21, 22, 23, 22, 23, 22, 17, 22, 23, 22, 23, 22, 17, 18, 19, 18, 12, 8, 7, 8, 7, 6, 12, 7, 8, 12, 17, 12, 16, 11, 12, 11, 12, 17, 18, 19, 24, 23, 22, 17]

\_\_\_\_\_

[15, 20, 15, 20, 15, 20, 21, 16, 21, 16, 12, 16, 12, 18, 12, 16, 12, 18, 19, 18, 17, 22, 23, 22, 23, 24, 19, 18, 19, 18, 17, 22, 21, 16, 21, 16, 21, 20, 21, 16, 12, 16, 21, 22, 23, 22, 21, 22, 23, 22]

\_\_\_\_\_

[23, 24, 19, 24, 23, 22, 17, 22, 17, 22, 17, 16, 12, 6, 5, 6, 12, 18, 19, 14, 9, 8, 12, 6, 5, 6, 11, 12, 16, 17, 12, 18, 12, 17, 18, 19, 14, 19, 18, 13, 14, 19, 14, 13, 8, 13, 12, 18, 17, 22]

\_\_\_\_\_

[13, 8, 9, 14, 19, 24, 23, 22, 23, 22, 17, 12, 13, 14, 9, 8, 9, 8, 3, 2, 3, 8, 9, 4, 9, 8, 13, 14, 19, 18, 17, 12, 17, 22, 23, 24, 19, 14, 19, 14, 19, 18, 17, 22, 23, 22, 23, 22, 23, 22]

[18, 17, 18, 17, 22, 23, 22, 23, 22, 21, 16, 12, 11, 16, 11, 12, 16, 11, 12, 11, 16, 11, 12, 16, 21, 20, 15, 20, 15, 20, 15, 20, 21, 20, 21, 20, 15, 20, 21, 16, 11, 6, 12, 16, 21, 20, 15, 20]

\_\_\_\_\_

[16, 12, 16, 11, 12, 16, 12, 17, 22, 23, 22, 17, 18, 17, 22, 23, 22, 23, 22, 23, 24, 19, 18, 19, 18, 17, 12, 8, 3, 4, 3, 2, 1, 0, 1, 0, 5, 6, 7, 6, 11, 16, 12, 18, 12, 16, 21, 20, 15, 20]

[10, 5, 10, 5, 6, 11, 6, 12, 11, 6, 5, 10, 5, 10, 5, 6, 11, 16, 21, 20, 21, 16, 12, 17, 16, 12, 17, 18, 12, 6, 5, 0, 1, 2, 1, 2, 1, 0, 1, 2, 3, 4, 3, 8, 3, 4,

### 3, 8, 3, 2]

[13, 18, 17, 22, 23, 22, 23, 22, 23, 22, 17, 16, 12, 18, 19, 24, 19, 18, 17, 12, 8, 7, 12, 8, 7, 12, 11, 12, 16, 12, 16, 17, 18, 17, 22, 23, 22, 23, 22, 21, 16, 12, 7, 8, 12, 8, 7, 6, 11, 12]

\_\_\_\_\_

[9, 14, 19, 14, 19, 14, 19, 14, 19, 18, 17, 18, 17, 18, 17, 22, 21, 16, 12, 16, 11, 6, 5, 6, 12, 11, 16, 11, 12, 11, 12, 6, 7, 6, 11, 12, 17, 18, 19, 18, 12, 11, 16, 21, 20, 15, 20, 15, 20, 15]

33% | 33/100 [00:00<00:01, 39.04it/s]

[11, 12, 11, 12, 16, 21, 20, 15, 20, 15, 20, 15, 20, 15, 20, 21, 16, 21, 16, 12, 18, 12, 16, 12, 17, 12, 16, 21, 22, 17, 18, 19, 18, 12, 11, 16, 12, 16, 11, 6, 12, 18, 12, 16, 12, 18, 17, 12, 16, 21]

\_\_\_\_\_

[1, 2, 3, 8, 13, 12, 17, 18, 19, 18, 17, 22, 21, 16, 12, 16, 21, 16, 21, 20, 21, 22, 17, 18, 17, 22, 23, 22, 23, 22, 23, 22, 21, 16, 12, 13, 18, 17, 22, 23, 22, 21, 20, 15, 20, 15, 20, 15, 20, 21]

\_\_\_\_\_

[22, 23, 24, 19, 18, 17, 22, 17, 22, 17, 12, 18, 17, 22, 23, 22, 17, 18, 17, 12, 18, 17, 22, 21, 20, 15, 20, 21, 22, 17, 18, 17, 12, 18, 17, 22, 23, 22, 23, 22, 23, 22, 17, 18, 17, 22, 17, 12, 16, 11]

[6, 5, 6, 7, 8, 9, 4, 3, 4, 3, 4, 9, 4, 3, 4, 3, 8, 7, 6, 5, 6, 12, 16, 11, 12, 8, 3, 2, 1, 2, 1, 0, 5, 0, 5, 10, 15, 20, 21, 20, 21, 16, 11, 6, 7, 6, 5, 6]

[16, 12, 11, 12, 17, 22, 17, 12, 11, 16, 11, 16, 17, 16, 21, 20, 21, 16, 17, 12, 17, 12, 11, 16, 21, 16, 11, 12, 6, 11, 12, 18, 17, 18, 17, 18, 19, 18, 19, 24, 19, 14, 19, 18, 17, 22, 17, 22]

\_\_\_\_\_

[18, 17, 18, 17, 22, 23, 22, 17, 22, 17, 22, 17, 12, 17, 22, 23, 22, 23, 22, 21, 20, 15, 20, 15, 20, 21, 22, 17, 22, 23, 22, 17, 22, 21, 20, 21, 16, 11, 6, 12, 18, 19, 14, 19, 14, 19, 18, 12, 16, 21]

[15, 20, 21, 20, 21, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 12, 17, 22, 17, 12, 17, 16, 21, 20, 21, 16, 12, 7, 8, 12, 16, 21, 22, 23, 22, 23, 22, 17, 18, 17, 12, 16, 12, 17, 18, 19, 18, 12, 16, 21]

\_\_\_\_\_

[13, 12, 18, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 22, 21, 20, 15, 20, 15, 20, 15, 20, 21, 20, 15, 20, 15, 20, 21, 16, 11, 16, 12, 11, 6, 5, 10, 5, 10, 5, 6, 11, 6, 11, 6, 5, 0]

\_\_\_\_\_

[15, 20, 21, 16, 12, 17, 22, 17, 18, 17, 22, 17, 12, 11, 16, 11, 12, 11, 16, 12, 17, 22, 17, 12, 11, 12, 17, 12, 7, 8, 13, 14, 13, 12, 6, 11, 12, 11, 12, 11, 6, 7, 8, 3, 8, 12, 16, 17, 22, 21]

[19, 18, 17, 12, 16, 21, 22, 23, 22, 23, 22, 21, 20, 15, 20, 15, 20, 15, 20, 21, 22, 17, 22, 23, 24, 19, 18, 17, 12, 18, 17, 22, 17, 22, 21, 20, 15, 20, 21, 20, 15, 20, 21, 22, 17, 18, 19, 14, 19, 14]

\_\_\_\_\_

42% | 42/100 [00:01<00:01, 39.03it/s]

[14, 13, 12, 17, 12, 17, 22, 23, 22, 17, 12, 16, 12, 6, 11, 12, 11, 12, 11, 6, 12, 16, 21, 22, 23, 22, 23, 22, 21, 16, 11, 12, 18, 19, 18, 12, 11, 12, 17, 22, 21, 16, 12, 6, 5, 6, 12, 17, 18, 19]

[21, 16, 11, 16, 21, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 21, 22, 17, 12, 16, 21, 20, 15, 20, 21, 16, 12, 18, 12, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 18, 12, 18]

\_\_\_\_\_

[6, 5, 6, 7, 6, 7, 6, 12, 11, 12, 18, 17, 12, 8, 9, 8, 12, 7, 8, 7, 8, 7, 8, 12, 16, 21, 20, 15, 20, 15, 20, 21, 20, 21, 16, 21, 22, 23, 24, 19, 14, 19, 14, 19, 18, 17, 12, 18, 12, 16]

[12, 11, 12, 17, 12, 16, 11, 12, 8, 7, 8, 9, 8, 13, 14, 13, 8, 12, 16, 12, 17, 12, 6, 11, 6, 5, 6, 5, 6, 12, 17, 22, 23, 22, 23, 22, 23, 22, 17, 16, 12, 16, 21, 20, 15, 20, 21, 16, 21, 16]

\_\_\_\_\_

[0, 5, 0, 1, 2, 3, 4, 9, 14, 19, 24, 23, 24, 19, 18, 19, 24, 23, 22, 21, 16, 11, 12, 16, 21, 22, 23, 22, 23, 22, 23, 22, 17, 12, 8, 9, 4, 3, 2, 1, 0, 5, 10, 5, 10, 5, 6, 11, 16, 21]

\_\_\_\_\_

\_\_\_\_\_

\_\_\_\_\_

[4, 9, 4, 9, 4, 9, 4, 3, 2, 1, 2, 1, 0, 1, 0, 1, 0, 1, 2, 1, 0, 1, 0, 5, 6, 11, 16, 17, 18, 19, 24, 23, 24, 19, 18, 17, 22, 17, 12, 17, 18, 17, 22, 23, 22, 21, 20, 15, 20, 15]

[18, 19, 18, 17, 12, 11, 12, 7, 8, 3, 2, 3, 8, 9, 8, 13, 12, 13, 14, 19, 14, 19, 14, 13, 14, 19, 18, 12, 11, 16, 21, 22, 23, 22, 23, 22, 17, 18, 17, 22, 23, 22, 23, 22, 23, 22, 17, 22]

[21, 20, 15, 20, 21, 20, 15, 20, 15, 20, 15, 20, 21, 16, 12, 16, 21, 22, 17, 22, 23, 22, 23, 22, 17, 12, 8, 3, 2, 1, 0, 1, 0, 1, 0, 5, 0, 5, 10, 5, 6, 11, 6, 7, 6, 12, 8, 7, 6, 11]

\_\_\_\_\_

[1, 0, 1, 2, 1, 2, 3, 2, 3, 4, 3, 4, 3, 2, 3, 8, 13, 12, 8, 13, 14, 19, 18, 17, 12, 17, 12, 18, 19, 18, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 21, 16, 11, 12, 17, 12, 16, 21, 20, 15]

-----

52% | 52/100 [00:01<00:01, 40.42it/s]

[17, 22, 17, 12, 18, 17, 12, 6, 5, 0, 1, 0, 5, 0, 1, 0, 5, 6, 5, 6, 5, 10, 5, 6, 12, 11, 16, 21, 22, 17, 18, 12, 13, 14, 13, 14, 19, 14, 13, 8, 3, 2, 1, 2, 1, 0, 5, 10, 5, 10]

-----

- [5, 10, 5, 6, 11, 12, 18, 17, 22, 23, 22, 17, 22, 23, 22, 23, 22, 17, 22, 23, 22, 23, 22, 17, 12, 18, 12, 8, 7, 6, 5, 0, 1, 2, 3, 4, 9, 8, 12, 16, 11, 12, 8, 13, 12, 8, 3, 2, 1, 0]
- \_\_\_\_\_
- 19, 14, 19, 14, 19, 14, 19]

### \_\_\_\_\_

[22, 23, 22, 23, 22, 23, 24, 19, 14, 13, 8, 12, 17, 22, 17, 22, 23, 22, 17, 18, 17, 22, 21, 20, 15, 20, 21, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 16, 12, 18, 12, 7, 8, 7, 12, 18, 19, 24]

### \_\_\_\_\_

[15, 20, 21, 20, 21, 16, 21, 16, 12, 16, 12, 6, 7, 8, 9, 8, 3, 2, 1, 2, 3, 4, 9, 8, 9, 8, 7, 8, 9, 4, 3, 8, 12, 7, 8, 13, 12, 16, 11, 16, 11, 12, 16, 21, 16, 12, 18, 12, 7, 8]

### \_\_\_\_\_

- [5, 10, 5, 0, 1, 0, 1, 2, 3, 4, 9, 8, 12, 17, 18, 17, 18, 17, 18, 17, 18, 12, 18, 19, 14, 13, 14, 13, 18, 17, 12, 11, 12, 6, 7, 8, 9, 4, 9, 4, 9, 14, 19, 18, 19, 24, 19, 18, 17, 12]
- \_\_\_\_\_

\_\_\_\_\_

- [7, 6, 11, 6, 5, 10, 5, 0, 5, 10, 5, 6, 5, 10, 5, 10, 5, 10, 5, 10, 5, 10, 5, 6, 12, 16, 17, 18, 19, 18, 19, 24, 19, 14, 13, 12, 11, 12, 11, 16, 21, 16, 11, 12, 17, 22, 17, 12, 11, 12, 11, 12]
- [22, 17, 12, 18, 19, 18, 17, 18, 17, 18, 17, 12, 18, 17, 18, 17, 12, 11, 6, 7, 8, 3, 8, 12, 18, 12, 17, 22, 17, 22, 23, 22, 23, 22, 17, 18, 17, 18, 17, 12, 18, 19, 14, 19, 18, 19, 18, 17, 22, 17]
- [22, 23, 22, 17, 22, 23]

### 62% | 62/100 [00:01<00:00, 40.80it/s]

[16, 21, 20, 21, 16, 11, 16, 12, 16, 11, 12, 16, 17, 12, 16, 21, 16, 12, 11, 12, 18, 12, 18, 17, 22, 17, 22, 17, 18, 17, 22, 23, 22, 23, 24, 19, 24, 23, 22, 23, 24, 19, 18, 19, 18, 12, 11, 12, 16, 12]

- [9, 4, 9, 14, 19, 18, 17, 22, 23, 22, 23, 22, 17, 12, 16, 17, 22, 17, 22, 17, 22, 17, 12, 7, 6, 7, 8, 12, 8, 7, 6, 11, 16, 12, 11, 12, 17, 22, 23, 22, 17, 22, 21, 20, 15, 20, 15, 20, 21, 16]

23, 22, 17, 18, 19, 24, 23, 22, 23]

\_\_\_\_\_

[11, 6, 7, 6, 5, 6, 5, 10, 5, 0, 5, 10, 5, 10, 5, 0, 1, 2, 1, 0, 1, 2, 3, 2, 1, 0, 5, 10, 5, 6, 11, 16, 11, 6, 11, 16, 11, 16, 11, 16, 21, 22, 17, 12, 18, 19, 14, 19, 24, 23]

\_\_\_\_\_

[14, 13, 18, 19, 18, 12, 6, 7, 6, 7, 6, 5, 10, 5, 0, 1, 0, 1, 0, 5, 0, 5, 0, 1, 0, 1, 2, 1, 0, 5, 0, 1, 2, 3, 2, 1, 2, 1, 0, 5, 10, 5, 0, 5, 10, 15, 20, 15]

\_\_\_\_\_

[11, 12, 17, 22, 17, 18, 17, 22, 23, 22, 23, 22, 23, 22, 23, 22, 17, 12, 11, 12, 6, 11, 16, 12, 16, 17, 16, 12, 17, 18, 12, 13, 8, 3, 2, 1, 2, 1, 0, 1, 2, 3, 2, 1, 0, 5, 0, 5, 6, 12]

\_\_\_\_\_

[2, 3, 2, 1, 0, 5, 10, 5, 0, 1, 0, 5, 6, 12, 16, 21, 16, 12, 16, 21, 22, 17, 18, 17, 18, 19, 14, 19, 18, 17, 12, 11, 16, 17, 22, 23, 22, 23, 22, 17, 12, 18, 19, 14, 13, 12, 16, 21, 20, 15]

\_\_\_\_\_

[12, 16, 11, 6, 11, 12, 17, 22, 23, 22, 23, 22, 23, 22, 17, 18, 17, 18, 17, 18, 17, 12, 11, 16, 12, 17, 18, 12, 7, 6, 11, 12, 18, 17, 22, 17, 12, 6, 5, 0, 1, 2, 1, 2, 1, 2, 1, 0, 1, 0]

\_\_\_\_\_

[15, 20, 21, 16, 11, 16, 11, 6, 11, 16, 11, 12, 18, 17, 22, 17, 12, 16, 21, 16, 21, 20, 21, 16, 11, 12, 16, 17, 22, 17, 22, 17, 22, 23, 22, 17, 22, 23, 22, 21, 16, 17, 18, 17, 12, 18, 17, 22, 23, 22]

\_\_\_\_\_

72% | 72/100 [00:01<00:00, 40.71it/s]

[6, 5, 10, 5, 10, 5, 6, 7, 8, 9, 8, 7, 6, 5, 0, 5, 10, 5, 6, 12, 16, 11, 12, 6, 11, 6, 5, 10, 5, 10, 5, 0, 5, 0, 1, 0, 1, 2, 1, 2, 3, 8, 7, 8, 3, 8, 12, 6, 7, 6]

[20, 21, 16, 11, 6, 7, 12, 18, 19, 18, 19, 14, 13, 18, 17, 12, 17, 22, 23, 22, 21, 20, 15, 20, 21, 20, 21, 20, 21, 16, 12, 11, 12, 18, 12, 17, 22, 17, 18, 17, 12, 13, 14, 13, 12, 17, 22, 17, 12, 18]

[10, 5, 6, 5, 10, 5, 6, 5, 6, 11, 12, 16, 11, 12, 16, 11, 6, 5, 10, 5, 10, 5, 6, 11, 6, 11, 16, 12, 16, 17, 12, 11, 12, 17, 22, 17, 12, 11, 12, 16, 12, 18, 17, 22, 17, 22, 21, 16, 12, 18]

[15, 20, 15, 20, 15, 20, 21, 16, 21, 22, 17, 18, 13, 14, 13, 14, 19, 18, 12, 11, 12, 11, 16, 11, 12, 13, 18, 19, 18, 17, 16, 12, 8, 9, 4, 9, 4, 9, 8, 9, 4, 9, 8, 9, 4, 3, 2, 1, 0, 1]

\_\_\_\_\_

[18, 19, 18, 17, 22, 17, 22, 23, 22, 17, 18, 17, 22, 17, 22, 23, 22, 23, 22, 23, 22, 17, 18, 12, 11, 16, 21, 20, 21, 20, 21, 20, 15, 20, 15, 20, 15, 20, 21, 20, 21, 22, 17, 18, 17, 18, 19, 14, 13, 14]

[20, 15, 20, 15, 20, 21, 22, 23, 22, 17, 18, 12, 17, 22, 17, 12, 16, 21, 20, 21, 16, 11, 16, 12, 17, 22, 17, 22, 23, 22, 17, 22, 17, 18, 17, 12, 16, 21, 16, 12, 16, 21, 20, 15, 20, 15, 20, 15, 20, 15]

[16, 12, 16, 11, 16, 12, 18, 17, 12, 13, 8, 3, 2, 1, 0, 5, 6, 7, 12, 7, 6, 7, 8, 9, 4, 9, 8, 12, 16, 21, 16, 21, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 21]

\_\_\_\_\_

[7, 8, 13, 14, 19, 18, 17, 18, 17, 22, 23, 22, 17, 18, 17, 18, 17, 22, 23, 22, 17, 18, 12, 17, 18, 19, 14, 19, 14, 9, 8, 13, 8, 12, 16, 11, 6, 5, 0, 5, 10, 5, 10, 5, 6, 11, 6, 7, 8, 9]

\_\_\_\_\_

[5, 10, 5, 10, 5, 0, 1, 0, 1, 2, 3, 2, 1, 2, 1, 2, 3, 8, 9, 4, 3, 8, 7, 6, 7, 6, 11, 12, 16, 12, 6, 7, 8, 3, 8, 12, 17, 18, 19, 14, 19, 14, 19, 14, 19, 18, 19, 18, 17, 18]

\_\_\_\_\_

[15, 20, 15, 20, 15, 20, 15, 10, 5, 10, 5, 6, 11, 6, 5, 6, 5, 6, 5, 10, 5, 6, 5, 6, 5, 10, 5, 6, 11, 16, 12, 7, 6, 7, 6, 11, 16, 21, 16, 12, 16, 12, 8, 7, 8, 13, 12, 7, 8, 12]

===============

81% | 81/100 [00:02<00:00, 39.59it/s]

[13, 14, 19, 18, 17, 18, 19, 14, 19, 18, 12, 18, 17, 22, 21, 22, 23, 22, 17, 18, 17, 18, 17, 22, 21, 16, 12, 11, 6, 5, 10, 5, 10, 5, 10, 5, 10, 5, 6, 12, 11, 16, 21, 20, 15, 20, 15, 20, 15, 20]

\_\_\_\_\_

[16, 11, 12, 18, 19, 14, 13, 12, 7, 8, 3, 2, 3, 2, 1, 2, 3, 4, 3, 2, 3, 8, 12, 7, 8, 12, 13, 18, 17, 22, 17, 16, 12, 18, 19, 18, 17, 22, 17, 16, 12, 11, 16, 21, 16, 12, 13, 8, 7, 8]

-----

[13, 14, 19, 24, 23, 24, 19, 18, 17, 12, 11, 12, 11, 6, 7, 8, 7, 6, 12, 11, 12, 16, 21, 16, 21, 16, 11, 12, 11, 16, 12, 13, 8, 7, 8, 3, 4, 9, 14, 19, 24, 23, 22, 23, 22, 17, 22, 23, 24, 19]

\_\_\_\_\_

[2, 3, 8, 9, 8, 3, 4, 3, 2, 1, 0, 1, 0, 1, 2, 1, 0, 1, 0, 1, 2, 3, 2, 3, 8, 12, 17, 12, 11, 16, 17, 18, 12, 11, 12, 18, 19, 18, 12, 17, 22, 17, 12, 18, 17, 22, 23, 22, 17, 12]

\_\_\_\_\_

[19, 24, 23, 22, 23, 22, 17, 18, 17, 18, 12, 8, 7, 6, 7, 6, 7, 6, 11, 12, 18, 12, 17, 22, 17, 12, 18, 17, 18, 19, 14, 19, 14, 19, 24, 23, 22, 23, 22, 17, 18, 19, 18, 17, 12, 7, 8, 9, 8, 3]

\_\_\_\_\_

[4, 9, 8, 3, 4, 3, 4, 3, 4, 9, 8, 9, 4, 3, 2, 1, 2, 1, 0, 1, 2, 1, 2, 1, 0, 1, 2, 1, 0, 5, 6, 11, 6, 11, 12, 16, 21, 16, 11, 12, 16, 21, 20, 21, 20, 15, 20, 21, 20, 15]

[12, 11, 6, 7, 6, 7, 8, 12, 16, 21, 16, 11, 12, 7, 8, 12, 11, 12, 7, 8, 13, 14, 19, 18, 12, 11, 16, 21, 16, 21, 20, 21, 20, 15, 20, 21, 22, 23, 22, 17, 18, 17,

12, 18, 12, 11, 12, 18, 19, 24]

\_\_\_\_\_

[2, 3, 8, 13, 18, 17, 22, 17, 12, 7, 8, 12, 16, 12, 18, 17, 12, 18, 17, 22, 23, 22, 21, 16, 12, 18, 12, 16, 21, 22, 23, 22, 21, 20, 15, 20, 15, 20, 21, 16, 17, 22, 23, 22, 21, 20, 21, 20, 15, 20]

\_\_\_\_\_

[15, 20, 15, 20, 21, 16, 11, 16, 11, 6, 7, 6, 12, 11, 16, 21, 20, 21, 16, 11, 12, 18, 17, 22, 23, 22, 17, 12, 18, 17, 12, 11, 12, 16, 12, 16, 21, 16, 12, 16, 12, 11, 12, 18, 19, 24, 23, 22, 23, 22]

[22, 17, 18, 17, 12, 16, 11, 16, 21, 16, 21, 20, 21, 22, 23, 22, 21, 20, 15, 20, 15, 20, 21, 20, 21, 22, 23, 22, 17, 18, 12, 16, 17, 22, 23, 22, 23, 22, 23, 24, 19, 18, 19, 14, 13, 14, 13, 12, 16, 21]

\_\_\_\_\_

89% | 89/100 [00:02<00:00, 39.21it/s]

[22, 17, 18, 19, 18, 19, 18, 17, 18, 19, 18, 17, 18, 19, 24, 23, 22, 17, 18, 12, 16, 11, 12, 11, 6, 5, 6, 5, 10, 5, 10, 5, 6, 11, 12, 18, 17, 22, 23, 22, 23, 22, 23, 22, 21, 16, 12, 11, 12, 16]

[2, 1, 0, 1, 2, 1, 0, 5, 6, 5, 10, 5, 10, 5, 6, 5, 6, 5, 10, 5, 0, 5, 0, 1, 0, 1, 2, 1, 0, 5, 6, 5, 0, 1, 2, 1, 0, 5, 6, 12, 8, 9, 8, 12, 17, 18, 17, 12, 16, 21]

\_\_\_\_\_

[5, 10, 5, 10, 5, 0, 1, 2, 3, 2, 3, 2, 1, 2, 3, 4, 3, 8, 3, 4, 3, 4, 9, 8, 12, 16, 11, 12, 18, 12, 16, 11, 12, 13, 8, 7, 6, 12, 11, 16, 21, 20, 21, 16, 12, 7, 8, 7, 6, 5]

\_\_\_\_\_

[15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15, 20, 21, 22, 23, 22, 23, 22, 23, 22, 23, 22, 21, 20, 21, 16, 21, 20, 15, 20, 15, 20, 15, 20]

\_\_\_\_\_

[10, 5, 6, 5, 6, 5, 6, 5, 6, 11, 16, 21, 16, 12, 16, 21, 16, 12, 6, 11, 6, 5, 6, 12, 13, 8, 9, 4, 3, 2, 3, 8, 7, 8, 7, 8, 7, 12, 16, 21, 16, 21, 20, 21, 22, 23, 22, 17, 12, 16]

[5, 10, 5, 6, 12, 18, 12, 18, 12, 7, 8, 12, 13, 14, 13, 12, 16, 21, 20, 21, 20, 21, 16, 12, 13, 14, 19, 24, 19, 24, 23, 22, 17, 12, 6, 5, 10, 5, 10, 5, 6, 11, 12, 11, 16, 21, 16, 21, 16, 12]

\_\_\_\_\_\_

[5, 10, 5, 10, 5, 10, 5, 10, 5, 10, 5, 0, 1, 0, 1, 0, 1, 0, 1, 2, 1, 0, 5, 6, 5, 6, 12, 11, 12, 18, 12, 17, 18, 12, 16, 17, 12, 8, 3, 8, 9, 8, 12, 8, 13, 12, 8, 7, 8, 9]

\_\_\_\_\_

[6, 7, 6, 11, 6, 5, 6, 12, 17, 18, 17, 18, 19, 18, 12, 18, 12, 8, 7, 6, 5, 10, 5, 6, 5, 10, 5, 6, 11, 16, 21, 20, 21, 20, 21, 16, 11, 6, 5, 6, 5, 6, 5, 10, 5, 6, 11, 12, 8, 7]

-----

100% | 100/100 [00:02<00:00, 39.40it/s]

[14, 19, 24, 23, 22, 23, 24, 19, 18, 17, 18, 17, 22, 17, 12, 16, 12, 18, 17, 22, 21, 16, 12, 11, 16, 12, 7, 8, 12, 6, 11, 12, 8, 13, 14, 19, 24, 23, 24, 19, 18, 17, 12, 11, 6, 7, 8, 7, 6, 11]

\_\_\_\_\_

[7, 6, 7, 8, 13, 12, 6, 7, 8, 12, 17, 18, 12, 17, 22, 23, 22, 17, 22, 23, 22, 17, 12, 13, 14, 13, 8, 12, 18, 17, 18, 12, 11, 12, 17, 18, 17, 18, 17, 18, 12, 18, 12, 11, 12, 11, 12, 7, 8, 12]

\_\_\_\_\_

[12, 18, 17, 18, 19, 24, 23, 22, 23, 24, 19, 14, 9, 4, 9, 8, 13, 18, 19, 24, 23, 18, 17, 22, 23, 22, 17, 18, 17, 12, 16, 21, 20, 15, 20, 15, 20, 21, 20, 21, 20, 15, 20, 15, 20, 15, 20, 15, 20, 15]

\_\_\_\_\_

[23, 22, 17, 22, 17, 12, 16, 21, 16, 21, 16, 12, 16, 21, 16, 21, 16, 21, 16, 12, 17, 22, 21, 20, 21, 16, 21, 16, 11, 16, 12, 11, 6, 5, 6, 7, 8, 13, 12, 18, 19, 24, 23, 22, 17, 22, 23, 24, 19, 18, 12, 6]

\_\_\_\_\_

[5, 10, 5, 0, 5, 10, 5, 6, 7, 8, 7, 8, 3, 2, 1, 0, 1, 0, 1, 2, 1, 2, 1, 2, 3, 2, 1, 0, 1, 0, 5, 0, 1, 2, 3, 4, 9, 8, 7, 8, 13, 14, 19, 24, 23, 22, 17, 12, 16, 21]

\_\_\_\_\_

[19, 14, 19, 14, 19, 14, 13, 8, 12, 17, 22, 23, 22, 23, 22, 21, 20, 15, 20, 21, 16, 12, 11, 16, 12, 7, 8, 9, 8, 12, 16, 11, 6, 7, 6, 5, 0, 1, 0, 5, 10, 5, 6, 5, 10, 5, 0, 5, 6, 11]

\_\_\_\_\_

[23, 22, 21, 16, 12, 17, 18, 12, 16, 11, 6, 11, 16, 21, 16, 12, 8, 7, 12, 18, 12, 17, 22, 17, 22, 17, 12, 11, 16, 21, 20, 15, 20, 21, 22, 23, 22, 23, 22, 23, 22, 17, 22, 17, 16, 12, 7, 8, 12, 6]

[5, 6, 5, 6, 11, 6, 11, 16, 12, 17, 12, 16, 17, 22, 23, 22, 23, 22, 21, 20, 15, 20, 15, 20, 21, 22, 17, 22, 17, 12, 8, 9, 4, 9, 4, 3, 4, 3, 2, 1, 0, 1, 2, 1, 0, 5, 10, 5, 6, 5]

# 3 Question 4

```
[168]: def convert_dictionary_to_matrix(dictionary):
           rows=len(dictionary)
           matrix=np.zeros((rows,rows),dtype=np.float64)
           for i in range(rows):
               for j in range(rows):
                   value = dictionary[i].get(j)
                   matrix[i,j] = value if value is not None else 0
           return matrix
[169]: def get_stationary_distribution(Q):
           evals, evecs = np.linalg.eig(Q.T)
           evec1 = evecs[:,np.isclose(evals, 1)]
           evec1 = evec1[:,0]
           stationary = evec1 / evec1.sum()
           stationary = stationary.real
           stationary = stationary / stationary.sum()
           return stationary
[170]: def remove_edge(dictionary, left, right):
           dictionary[left].pop(right)
           dictionary[right].pop(left)
[171]: def normalize_matrix(matrix):
           input: two dimensional matrix
           returns a matrix which its rows sum to 1
           return matrix / np.sum(matrix,1)[:,None]
[216]: import copy
       transition_matrix = convert_dictionary_to_matrix(model.transition_probs)
       transition2 =copy.deepcopy(model.transition probs)
       remove_edge(transition2,8,12)
       remove_edge(transition2,12,13)
       remove_edge(transition2,18,19)
       remove_edge(transition2,16,17)
       remove_edge(transition2,6,12)
       transition_matrix2 = convert_dictionary_to_matrix(transition2)
       transition_matrix2 = normalize_matrix(transition_matrix2)
       stationary distribution=get stationary distribution(transition matrix)
       stationary_distribution
[216]: array([0.02112507, 0.0285067, 0.02265319, 0.01969918, 0.01347683,
```

0.03876496, 0.04224324, 0.02300952, 0.03733263, 0.01653681,

```
0.0491273 , 0.05711506, 0.08066256, 0.05072891, 0.01322225])
[180]: from random import uniform
       import random
       import pandas as pd
       def w_choice(seq):
           choose an item from list of items whose probability is given
           example: input [0.3,0.4,0.3]
           output: it may choose from [0,1,2] with given probabilty
           total_prob = sum([item for item in seq])
           chosen = random.uniform(0,total prob)
           cumulative = 0
           for i in range(len(seq)):
               cumulative += seq[i]
               if cumulative > chosen:
                   return i
[181]: state count = len(transition matrix)
       a_array = np.zeros((state_count, state_count))
       for i in range(state count):
           for j in range(state_count):
               r_nominator = stationary_distribution[j]*transition_matrix2[j,i]
               r_denominator = stationary_distribution[i]*transition_matrix2[i,j]
               a = 0
               if r_denominator == 0:
                   a = 1
               else:
                   a = min(r_nominator / r_denominator,1)
               a_{array}[i,j] = a
[182]: new_transition_matrix=a_array*transition_matrix2
       stationary2=_

    get_stationary_distribution(normalize matrix(new_transition_matrix))
       stationary2
[182]: array([0.02180875, 0.02994216, 0.0234925, 0.02023764, 0.01302789,
              0.04241142, 0.03536303, 0.02535294, 0.03216247, 0.01639951,
              0.01965901, 0.0513758, 0.10217224, 0.02017778, 0.02124298,
              0.03131307, 0.07050133, 0.08362803, 0.05162414, 0.01804102,
              0.05382912, 0.06143372, 0.08667532, 0.05355925, 0.01456888
```

0.01811561, 0.04662704, 0.11368048, 0.02230126, 0.01947599, 0.02869246, 0.07049459, 0.07837297, 0.05633585, 0.03169953,

```
[241]: ### check if they are close
       (stationary2-stationary_distribution) < 0.01</pre>
[241]: array([ True,
                      True,
                             True,
                                     True,
                                            True,
                                                   True,
                                                          True,
                                                                 True,
                                                                         True,
               True,
                      True,
                             True,
                                    True,
                                           True,
                                                   True,
                                                          True,
                                                                 True,
                                                                         True,
               True,
                      True,
                             True,
                                    True,
                                           True,
                                                   True,
                                                          True])
[100]: ### sampling
       iterations = 1000
       sample_items = 100
       x_array = np.zeros((iterations, sample_items))
       for e in range(iterations):
           x_array[e,0] = int(random.uniform(0,state_count))
           for i in range(sample_items-1):
               x = int(x_array[e,i])
               y = w_choice(transition_matrix2[x])
               if random.uniform(0,1) <= a_array[x,y]:</pre>
                   x_array[e,i+1] = y
               else:
                   x_{array}[e,i+1] = x
[113]: x_array=x_array.reshape(-1)
       imperical_stationary=pd.Series(x_array).value_counts().sort_index()/len(x_array)
       np.array(imperical_stationary)-stationary_distribution < 0.01</pre>
[113]: array([ True,
                                    True,
                      True,
                             True,
                                            True,
                                                   True,
                                                          True,
                                                                 True,
                                                                         True,
               True,
                      True,
                             True,
                                    True,
                                            True,
                                                   True,
                                                          True,
                                                                 True,
                                                                        True,
                                    True,
               True,
                      True,
                             True,
                                            True,
                                                   True,
                                                          True])
[114]: stationary2=__

—get_stationary_distribution(normalize_matrix(new_transition_matrix))
       stationary2
[114]: array([0.02180875, 0.02994216, 0.0234925, 0.02023764, 0.01302789,
              0.04241142, 0.03536303, 0.02535294, 0.03216247, 0.01639951,
              0.01965901, 0.0513758, 0.10217224, 0.02017778, 0.02124298,
              0.03131307, 0.07050133, 0.08362803, 0.05162414, 0.01804102,
              0.05382912, 0.06143372, 0.08667532, 0.05355925, 0.01456888])
      (stationary2-stationary_distribution) < 0.01
[115]:
[115]: array([ True,
                      True,
                             True,
                                    True,
                                           True, True,
                                                          True,
                                                                 True,
                                                                         True,
               True.
                      True.
                             True, True, True, True,
                                                          True,
                                                                 True.
                      True,
                            True, True, True, True,
                                                         Truel)
               True.
[49]:
```

### **3.0.1** Question **5**

[157]: min\_matrix=np.where(transition\_matrix<transition\_matrix.

→T, transition\_matrix, transition\_matrix.T)

```
proposal_stationary=get_stationary_distribution(normalize_matrix(min_matrix))
       proposal_stationary
[157]: array([0.03623193, 0.05314085, 0.04338589, 0.039074 , 0.04035088,
              0.05016627, 0.0429694, 0.02380878, 0.04097378, 0.04377247,
              0.0249732, 0.03266443, 0.05314085, 0.02670919, 0.03966445,
              0.03096373, 0.04686786, 0.04565732, 0.03981049, 0.04184281,
              0.04970957, 0.04558968, 0.05314085, 0.03842504, 0.01696628])
           Yes, It is logical,
      4 Extra Section for testing purposes
[191]: import numpy as np
       transition = np.array([[0.1,0.2,0.7],
                                               [0.3, 0.6, 0.1],
                                               [0.3, 0.4, 0.3]
[192]: np.linalg.matrix_power(transition, 10000)
[192]: array([[0.25], 0.4375, 0.3125],
              [0.25, 0.4375, 0.3125],
              [0.25 , 0.4375, 0.3125]])
[193]: stationary_dist=get_stationary_distribution(transition)
       stationary_dist
[193]: array([0.25 , 0.4375, 0.3125])
[205]: def metropolis hastings(sample items, transition, stationary dist):
           sample = np.zeros(sample_items,dtype=np.int32)
          for i in range(sample_items-1):
                   x = sample[i]
                   y = w_choice(transition[x])
                   r_nominator = stationary_dist[y]*transition[y,x]
                   r_denominator = stationary_dist[x]*transition[x,y]
                   if r_denominator == 0:
                       a = 1
```

```
else:
    a = min(r_nominator / r_denominator,1)

if random.uniform(0,1) <= a:
    sample[i+1] = y

else:
    sample[i+1] = x

return sample
print(metropolis_hastings(10,transition,stationary_dist))</pre>
```

### [0 0 0 1 1 2 0 0 2 0]

[213]: 0.0 0.260744 1.0 0.427493 2.0 0.311763 dtype: float64

[]: