assignment3 07/02/2025, 17:25

```
In [18]: import numpy as np
         states = ["Rain", "no Rain"]
         observations = ["Umbrella", "no Umbrella"]
         #transition matrix P(X_t \mid X_{t-1})
         transition_matrix = np.array ([
              [0.7, 0.3],
              [0.3, 0.7]
         ])
         observations_matrix = np.array([
              [0.9, 0.2],
              [0.1, 0.8]
         ])
         #initial state, assumed equal probability of Rain and no Rain
         initial_state = np.array([0.5, 0.5])
         print("Hidden Markov Model (HMM) Representation\n")
         print(f"states: {states}")
         print(f"observations: {observations}")
         print("\nTransition Matrix P(X t | X t-1)")
         print(transition matrix)
         print("\n0bservations Matrix P(E_t | X_t)")
         print(observations_matrix)
         print("\nInitial State")
         print(initial_state)
        Hidden Markov Model (HMM) Representation
        states: ['Rain', 'no Rain']
        observations: ['Umbrella', 'no Umbrella']
        Transition Matrix P(X t \mid X t-1)
        [[0.7 \ 0.3]
         [0.3 0.7]]
        Observations Matrix P(E_t | X_t)
        [[0.9 \ 0.2]
         [0.1 0.8]]
        Initial State
        [0.5 0.5]
In [19]: observation_sequence = [0, 0, 1, 0, 0] # {Umbrella, Umbrella, No Umbrell
In [20]: def forward_algorithm(observation_sequence, transition_matrix, observation
```

assignment3 07/02/2025, 17:25

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Computes the forward probabilities f_t using the Forward Algorithm.
     Returns a list of all normalized forward messages.
     num states = len(initial state)
     num observations = len(observation sequence)
     # store forward messages in the hashset
     forward_messages = []
     # initialize with prior belief and first observation
     f_t = initial_state * observation_matrix[observation_sequence[0]]
     f_t /= np.sum(f_t) # Normalize
     forward messages.append(f t.copy())
     # recursive filtering for t >= 1
     for t in range(1, num_observations):
         f t = observation matrix[observation sequence[t]] * (transition m
         f t /= np.sum(f t) # Normalize
         forward_messages.append(f_t.copy())
     return forward_messages
 # run forward algorithm
 forward_messages = forward_algorithm(observation_sequence, transition_mat
 # display results
 for t, f_t in enumerate(forward_messages, start=1):
     print(f"P(X_{t} | e_1:\{t\}): \{f_t\}")
 # extract and print final probability of rain at day 2 and 5
 p_rain_day_2 = forward_messages[1][0] # P(Rain | e1:2)
 p_rain_day_5 = forward_messages[4][0] # P(Rain | e1:5)
 print(f"\nProbability of Rain at Day 2: {p_rain_day_2:.3f} (Expected: 0.8
 print(f"Probability of Rain at Day 5: {p_rain_day_5:.3f}")
P(X_1 | e_1:1): [0.81818182 0.18181818]
P(X_2 | e_1:2): [0.88335704 0.11664296]
P(X 3 | e 1:3): [0.19066794 0.80933206]
P(X_4 | e_1:4): [0.730794 0.269206]
P(X_5 \mid e_1:5): [0.86733889 0.13266111]
Probability of Rain at Day 2: 0.883 (Expected: 0.883)
Probability of Rain at Day 5: 0.867
```

In []: