

Design and Implementation of hidden Markov Models

Aim:

To Design and Implement the Hidden Markov Models.

Algorithm:

Step 1: we multiply the Initial Probability of state i with the emission probability to observe o from state i at time $t=1$.

Step 2: we find the Maximum value among all the product results and Assign it to the viterbi variable.

Step 3: Generate Initial, Transition and Emission probability Distribution from the Sample Data.

Step 4: Generate, a List of all unknown sequence.

Step 5: Score all unknown sequences and select the best sequence.

Step 6: Termination the following equation depicts the probability of the complete state sequence.

Code:-

```
import os
from hmm.hmm import Discrete HMM
hidden_var_name = ('Sunny', 'Foggy', 'rainy')
observation_var_name = ('no', 'yes')
hidden_var = {}
```

```

hidden_var_name[0] = 0;
hidden_var_name[1] = 1;
hidden_var_name[2] = 2,
}
observation_var = {
    observation_var_name[0] = 0,
    observation_var_name[1] = 1,
}
target = [ ]
obs_seq = [ ]

```

```

with open(os.path.join(os.path.dirname(__file__), 'input.txt')) as f:
    for line in f:

```

```

        hidden, observe = line.strip().split(' ');
        target.append(hidden_var[hidden])
        obs_seq.append(observation_var[observe])

```

```

B = (
    (0.8, 0.2),
    (0.5, 0.5),
    (0.1, 0.9),
)

```

```

hmm = DiscreteHMM(len(hidden_var), len(observation_var), B=B)

```

```

hmm.train(obs_seq, verbose=1)

```

```

hmm.show_model()

```

```

print('checksum:', hmm.check_model())

```

Result:

Thus the Design and Implementation of hidden Markov Model is Successfull and Verified the Output.

train.py - C:\Users\Smart\Downloads\weather_forecast\weather_forecast\train.py (3.9.7)

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```
import os
from hmm.hmm import DiscreteHMM

# Mapping input to variable's id
hidden_var_name = ('sunny', 'foggy', 'rainy')
observation_var_name = ('no', 'yes')
hidden_var = {
    hidden_var_name[0]: 0,
    hidden_var_name[1]: 1,
    hidden_var_name[2]: 2,
}

observation_var = {
    observation_var_name[0]: 0,
    observation_var_name[1]: 1,
}

target = []
obs_seq = []

with open(os.path.join(os.path.dirname(__file__), 'input.txt')) as f:
    for line in f:
        hidden, observe = line.strip().split(',')
        target.append(hidden_var[hidden])
        obs_seq.append(observation_var[observe])

# Setting model
B = (
    (0.8, 0.2),
    (0.5, 0.5),
    (0.1, 0.9),
)

hmm = DiscreteHMM(len(hidden_var), len(observation_var), B=B)

# Training the model best describe the observation
hmm.train(obs_seq, verbose=1)
hmm.show_model()
print('checksum:', hmm.check_model())
```

Ln 1

===== RESTART: C:\Users\user\Desktop\ML\weather_forecast\train.py =====

```
itnum    1 : delta 3.332041
itnum    2 : delta 0.428877
itnum    3 : delta 0.309993
itnum    4 : delta 0.215055
itnum    5 : delta 0.143531
itnum    6 : delta 0.094136
itnum    7 : delta 0.066443
itnum    8 : delta 0.052992
itnum    9 : delta 0.043964
itnum   10 : delta 0.037234
itnum   11 : delta 0.032206
itnum   12 : delta 0.028408
itnum   13 : delta 0.025485
itnum   14 : delta 0.023181
itnum   15 : delta 0.021314
itnum   16 : delta 0.019755
itnum   17 : delta 0.018418
itnum   18 : delta 0.017243
itnum   19 : delta 0.016188
itnum   20 : delta 0.015226
itnum   21 : delta 0.014339
itnum   22 : delta 0.013514
itnum   23 : delta 0.012741
itnum   24 : delta 0.012015
itnum   25 : delta 0.011332
itnum   26 : delta 0.010709
itnum   27 : delta 0.010213
itnum   28 : delta 0.009814
```

-----A: Transition probability-----

```
[[0.5434 0.4386 0.018 ]
 [0.5016 0.0854 0.413 ]
 [0.2865 0.1642 0.5494]]
```

-----B: Emission probability-----

```
[[0.9889 0.0111]
 [0.8347 0.1653]
 [0.1272 0.8728]]
```

-----pi: initial state distribution-----

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

checksum: True

>>>