

With this code I will be predicting a sequence from an image using three methods: Using the test image data, train image data and train text data I am trying to calculate three probabilities here:

A. Initial Probability:

Initial Probability for letter[i] is calculated as the number of statement starting from letter[i] over total number of statements in file.

B. Transition Probability:

Transition probability is calculated from the text training file by counting the transitions of letter[i] to letter[j] divided by the total transitions of letter[i]

this gives me $P(l[j]|l[i])$ where j represents the next alphabet in unobserved sequence and i is the current alphabet

C. Emission Probability E:

This E represents the probability of $l(i) = \text{letters}[j]$ given $O(i) = \text{test_letters}[r]$

where $l(i)$ is the i-th element of sequence to be predicted, $O(i)$ is the observed variable which is currently set to the r-th pixel arrangement of test_letters

Using these three probabilities the prediction is done in three ways:

I. Simplified method:

In this method,

1. I will be calculating the probability by taking each observed alphabet O from test sequence
2. Then Each pixel of this observed alphabet O is matched with each of the train alphabets in 'letters'
3. Based on number of pixels matched, I am calculating the emission probability E for each alphabet in letters.
4. The maximum value of this emission probability E gives the most probable alphabet
5. This sequence repeated for each alphabet gives the most probable sequence.

II. HMM using variable elimination:

In this method,

1. let the sequence to be predicted be $l_1, l_2, l_3, \dots, l_n$
2. The observed pixels states be $O_1, O_2, O_3, \dots, O_n$
3. Here the probability for l_1 is found for all letters in LETTERS
4. This is done by multiplying the $\text{initialProb}(l_1)$ by $\text{emission_prob}(l_1|O_1)$
5. Then l_1 is eliminated by marginalising all l_1 terms over all values of l_1
6. l_1 terms include $P(l_1)*P(l_2|O_2)*P(l_2|l_1)$. This can be replaced by $\alpha(l_2)$
7. Similarly eliminating l_2 terms and l_3, l_4, \dots, l_n terms will give us a lookup table with l_1, l_2, \dots, l_n values
8. Selecting the maximum probability values gives the maximum probable sequence

III. HMM using Viterbi Algorithm:

In this method,

1. let the sequence to be predicted be $l_1, l_2, l_3, \dots, l_n$
2. The observed pixels states be $O_1, O_2, O_3, \dots, O_n$
3. Here the probability for l_1 is found for all letters in LETTERS
4. This is done by multiplying the $\text{initialProb}(l_1)$ by $\text{emission_prob}(l_1|O_1)$
5. The max probability from l_1 values is taken as the initial probability for predicting l_2 value
6. probability of l_2 is calculated as,
$$P(l_2) = \max(\text{initial_prob}(l_1)*P(l_2|l_1)*P(l_2|O_2))$$
7. This max value of $P(l_2)$ is used as initial probability to calculate $P(l_3)$ and so on.
8. At each step only the maximum probability value moves forward to the next step.
- 9 At each step the maximum prob alphabet is the predicted alphabet of that test sequence step.
10. Hence at the end we get the most probable sequence.
11. This process takes less time than variable elimination

These are the performances I observed:

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(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-0-0.png
Viterbi Accuracy: 94.1176470588 %
VE Accuracy: 94.1176470588 %
Simple: SUPREME COURT OF THF UNITED STATES
HMM VE: SUPREME COURT OF THF UNITED STATES
HMM Viterbi: SUPREME COURT OF THF UNITED STATES
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-1-0.png
Viterbi Accuracy: 15.7142857143 %
VE Accuracy: 15.7142857143 %
Simple:
HMM VE:
HMM Viterbi:
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-2-0.png
Viterbi Accuracy: 86.2068965517 %
VE Accuracy: 86.2068965517 %
Simple: Nos. 14-556. -rguec -pr11 28, 2015 - Dec1cec June 26, 2015
HMM VE: Nos. 14-556. rguec pr11 28, 2015 - Dec1cec June 26, 2015
HMM Viterbi: Nos. 14-556. rguec pr11 28, 2015 - Dec1cec June 26, 2015
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-3-0.png
Viterbi Accuracy: 95.0819672131 %
VE Accuracy: 95.0819672131 %
Simple: Together with No. 14-562, Tanco et al. v. Haslam, Governor of
HMM VE: Together with No. 14-562, Tanco et al. v. Haslam, Governor of
HMM Viterbi: Together with No. 14-562, Tanco et al. v. Haslam, Governor of
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-4-0.png
Viterbi Accuracy: 42.8571428571 %
VE Accuracy: 42.8571428571 %
Simple: . n -ss , 1.. 1so c t cr 1 to me c r .
HMM VE: . n ss , 1.. 1so c t cr 1 to me c r .
HMM Viterbi: . n ss , .. 1so c t cr 1 to me c r .
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-5-0.png
Viterbi Accuracy: 95.0 %
VE Accuracy: 95.0 %
Simple: Opinion of the Ccourt
HMM VE: Opinion of the Ccourt
HMM Viterbi: Opinion of the Ccourt
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-6-0.png
Viterbi Accuracy: 19.0476190476 %
VE Accuracy: 22.2222222222 %
Simple: m - - - c n r . m 1 g
HMM VE: m - c n r . m 1 g
HMM Viterbi: m - c n . m 1
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-7-0.png
Viterbi Accuracy: 91.6666666667 %
VE Accuracy: 89.5833333333 %
Simple: emcocles a love that may endure even past ceath.
HMM VE: emcocles a love that may endure even past ceath.
HMM Viterbi: embocles a love that may endure even past ceath.
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-8-0.png
Viterbi Accuracy: 15.3846153846 %
VE Accuracy: 15.3846153846 %
Simple: 1
HMM VE: 1
HMM Viterbi:
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-9-0.png
Viterbi Accuracy: 23.8095238095 %

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VE Accuracy: 23.8095238095 %
Simple: - c m 1 .
HMM VE: c m 1 .
HMM Viterbi: c m .
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-10-0.png
Viterbi Accuracy: 92.1875 %
VE Accuracy: 90.625 %
Simple: Their plea is that they do respect it, respect it so deeply that
HMM VE: Their plea is that they do respect it, respect it so deeply that
HMM Viterbi: Their plea is that they co respect it, respect it so ceeply that
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-11-0.png
Viterbi Accuracy: 18.3673469388 %
VE Accuracy: 18.3673469388 %
Simple: - - 1 1 1 111m 1 .
HMM VE: 1 1 1 111m 1 .
HMM Viterbi: 1 1 1 1 111m 1 .
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-12-0.png
Viterbi Accuracy: 82.1428571429 %
VE Accuracy: 80.3571428571 %
Simple: 1heir hope is not to be ccndemned tc 1lv n loneliness,
HMM VE: 1heir hope is not to be ccndemned tc 1lv n loneliness,
HMM Viterbi: 1heir hope is not to be ccndemned tc 1lv n loneliness,
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-13-0.png
Viterbi Accuracy: 94.6428571429 %
VE Accuracy: 94.6428571429 %
Simple: excluded from one of civilization's oldest institutions.
HMM VE: excluded from one of civilization's oldest institutions.
HMM Viterbi: excluded from one of civilization's oldest institutions.
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-14-0.png
Viterbi Accuracy: 78.0 %
VE Accuracy: 80.0 %
Simple: T ey ask for equal c1gn1 y 1n he eyes of he law.
HMM VE: T ey ask for equal c1gn1 y 1n he eyes of he law.
HMM Viterbi: T ey ask for equal c1gn1 y 1n e eyes of he law.
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-15-0.png
Viterbi Accuracy: 97.5 %
VE Accuracy: 95.0 %
Simple: The Constitution grants them that r1ght.
HMM VE: The Constitution grants them that r1ght.
HMM Viterbi: The Constitution grants them that r1ght.
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-16-0.png
Viterbi Accuracy: 18.0555555556 %
VE Accuracy: 18.0555555556 %
Simple: c r 1
HMM VE: c r 1
HMM Viterbi: c r 1
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani\$ python ocr.py courier-train.png
bc.train test-17-0.png
Viterbi Accuracy: 94.1176470588 %
VE Accuracy: 88.2352941176 %

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Simple: 1t 1s so ordered.
HMM VE: 1t 1s so ordered.
HMM Viterbi: 1t is so ordered.
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-18-0.png
Viterbi Accuracy: 17.5438596491 %
VE Accuracy: 17.5438596491 %
Simple: . 1
HMM VE: . 1
HMM Viterbi: .
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$ python ocr.py courier-train.png
bc.train test-19-0.png
Viterbi Accuracy: 26.9230769231 %
VE Accuracy: 28.8461538462 %
Simple: 1N D B . 1 , nc , .. n c.
HMM VE: 1N D B . 1 , nc N, .. n c.
HMM Viterbi: 1N B . 1 , nc , .. n c.
(myenv) VINITAs-MacBook-Air:part2 vinitaboolchandani$

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So the mean performance I got was : **60%**