

CS461 – RECITATION 07

MACHINE LEARNING PRINCIPLES

Daize Dong
2025-10-27

TODAY'S CONTENT

- Markov Model Assignment
- Quiz 03

MARKOV MODELS

1. States: $S = \{s_1, s_2, \dots, s_N\}$
2. Distributions: π , where $\pi(i) = P(z_1 = i)$
3. Transition Matrix: A , where $A(i, j) = P(z_t = j \mid z_{t-1} = i)$
4. Emission Matrix: B , where $B(i, j) = P(x_i \mid z_j)$

MARKOV EXAMPLE

Imagine a language only containing the following:

- class N, nouns: {wolf, parrot, ...}
- class AND, word {and}
- class V, verbs: {run, fly, ...}

MARKOV EXAMPLE

The state transitions and their probabilities look like:

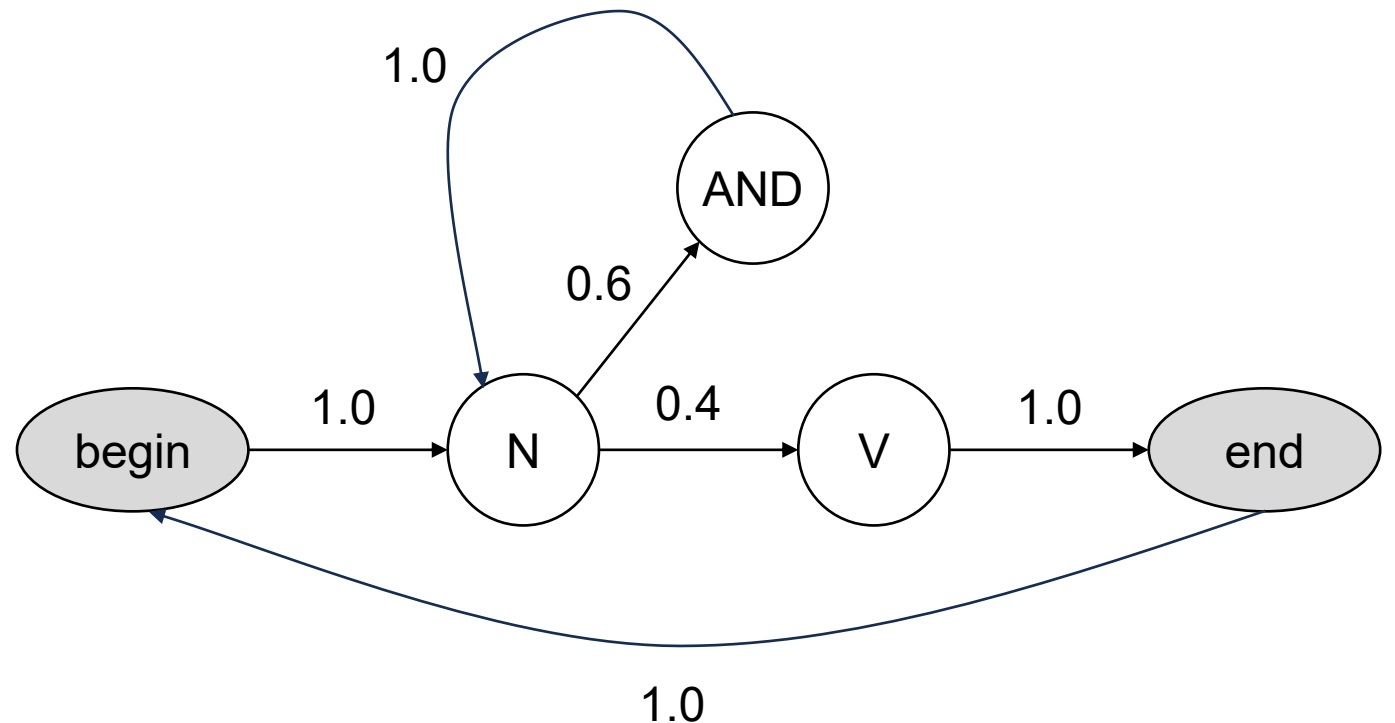
bb: begin

N: {wolf, parrot, ...}

AND: {and}

V: {run, fly, ...}

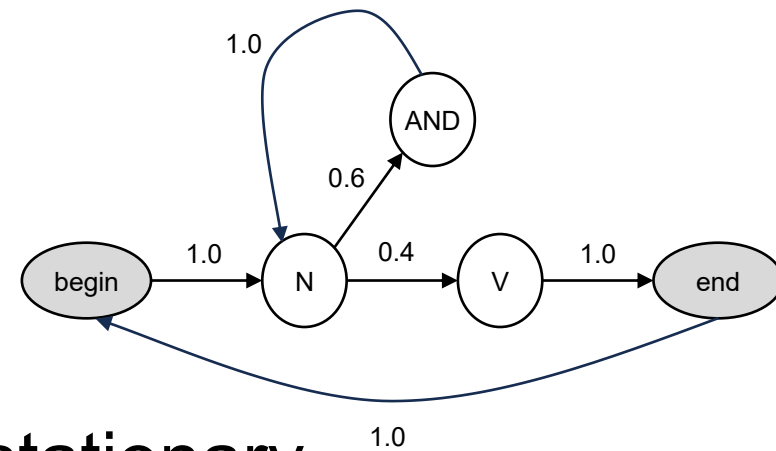
ee: end



MARKOV EXAMPLE

Now solve the following questions:

1. Populate a state transition matrix A .
2. Solve this set of equations to determine the stationary distributions, π .
3. Use the fraction of time spent in the **bb** or **ee** states to deduce the average number of words in a sentence.

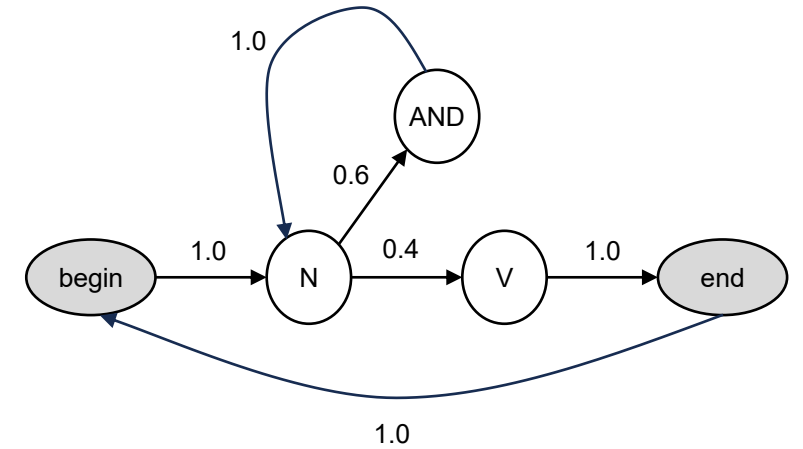


1. TRANSITION MATRIX

End Points

	bb	N	AND	V	ee
bb	0	1.0	0	0	0
N	0	0	0.6	0.4	0
AND	0	1.0	0	0	0
V	0	0	0	0	1.0
ee	1.0	0	0	0	0

Start Points

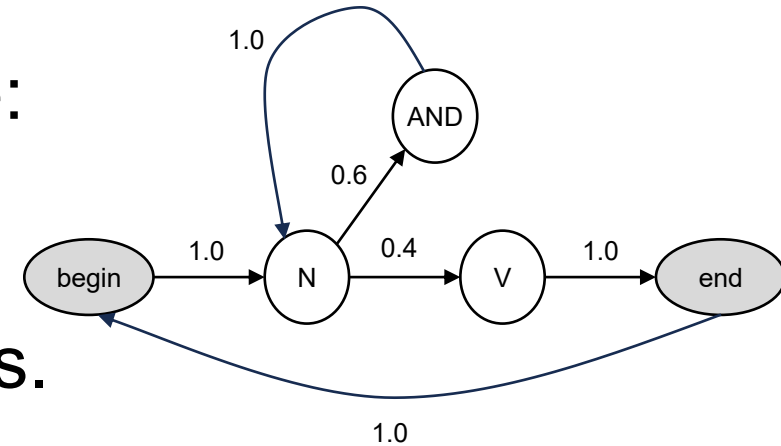


2. STATIONARY DISTRIBUTION

When reached the stationary distribution, we have:

$$\pi = \pi A \text{ and } \sum_{s \in S} \pi(s) = 1$$

Here π can be treated as a vector of all state probs.



$$\pi = \pi A \rightarrow \begin{bmatrix} \pi(bb) \\ \pi(N) \\ \pi(AND) \\ \pi(V) \\ \pi(ee) \end{bmatrix}^T = \begin{bmatrix} \pi(bb) \\ \pi(N) \\ \pi(AND) \\ \pi(V) \\ \pi(ee) \end{bmatrix}^T \begin{bmatrix} 0 & 1.0 & 0 & 0 & 0 \\ 0 & 0 & 0.6 & 0.4 & 0 \\ 0 & 1.0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1.0 \\ 1.0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

2. STATIONARY DISTRIBUTION

$$\pi = \pi A \rightarrow \begin{bmatrix} \pi(bb) \\ \pi(N) \\ \pi(AND) \\ \pi(V) \\ \pi(ee) \end{bmatrix}^T = \begin{bmatrix} \pi(bb) \\ \pi(N) \\ \pi(AND) \\ \pi(V) \\ \pi(ee) \end{bmatrix}^T \begin{bmatrix} 0 & 1.0 & 0 & 0 & 0 \\ 0 & 0 & 0.6 & 0.4 & 0 \\ 0 & 1.0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1.0 \\ 1.0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\rightarrow \begin{cases} \pi(bb) = \pi(ee) \\ \pi(N) = \pi(bb) + \pi(AND) \\ \pi(AND) = 0.6\pi(N) \\ \pi(V) = 0.4\pi(N) \\ \pi(ee) = \pi(V) \end{cases}$$

$$\rightarrow \begin{cases} \pi(bb) = 0.4\pi(N) \\ \pi(N) = \pi(N) \\ \pi(AND) = 0.6\pi(N) \\ \pi(V) = 0.4\pi(N) \\ \pi(ee) = 0.4\pi(N) \end{cases}$$

①

2. STATIONARY DISTRIBUTION

$$\textcircled{1} \begin{cases} \pi(bb) = 0.4\pi(N) \\ \pi(N) = \pi(N) \\ \pi(AND) = 0.6\pi(N) \\ \pi(V) = 0.4\pi(N) \\ \pi(ee) = 0.4\pi(N) \end{cases}$$

$$\begin{aligned} \sum_{s \in S} \pi(s) &= 1 \\ \rightarrow \pi(bb) + \pi(N) + \pi(AND) &+ \pi(V) + \pi(ee) = 1 \end{aligned} \quad \textcircled{2}$$

Combine $\textcircled{1}$ and $\textcircled{2}$, we get:

$$0.4\pi(N) + \pi(N) + 0.6\pi(N) + 0.4\pi(N) + 0.4\pi(N) = 1$$

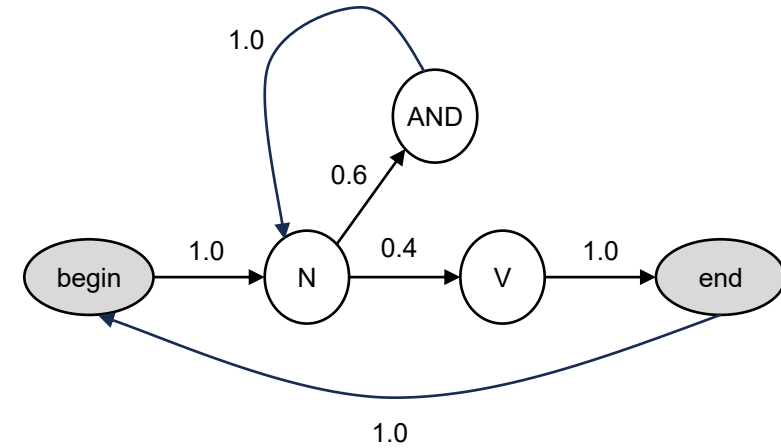
$$2.8\pi(N) = 1$$

$$\pi(N) = \frac{5}{14}$$

2. STATIONARY DISTRIBUTION

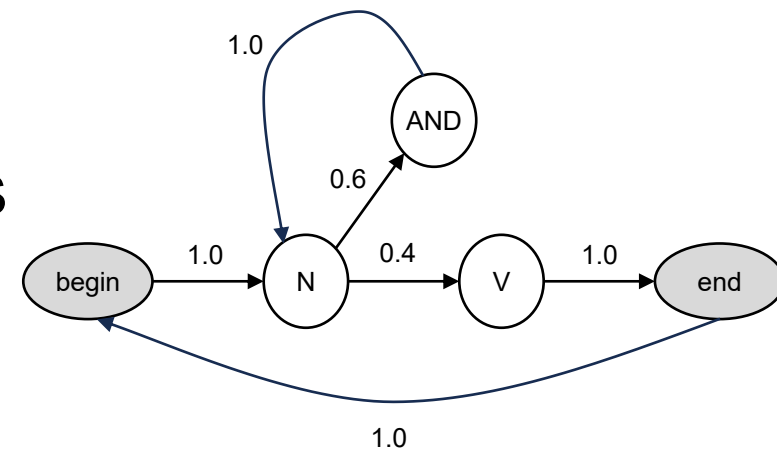
Finally we have:

$$\left\{ \begin{array}{l} \pi(bb) = 0.4\pi(N) = \frac{1}{7} \\ \pi(N) = \frac{5}{14} \\ \pi(AND) = 0.6\pi(N) = \frac{3}{14} \\ \pi(V) = 0.4\pi(N) = \frac{1}{7} \\ \pi(ee) = 0.4\pi(N) = \frac{1}{7} \end{array} \right.$$



3. AVERAGE SENTENCE LENGTH

We just need to find the average number of **bb** OR **ee** states in the sequence (as all sentences must have 1 **bb** and 1 **ee**).



$$\pi(bb) = \pi(ee) = \frac{1}{7}$$
$$|\overline{\pi(bb)}| = \frac{1}{\pi(bb)} = 7$$

Remember **bb** and **ee** don't count as words:

$$L = |\overline{\pi(bb)}| - 2 = 5$$

QUIZ 03

QUIZ 03

A	B	C	D
12	5	0	3

1. What is required to create a good ensemble?

- (a) Some independence between members of the ensemble.
- (b) A data resampling technique to train members of the ensemble.
- (c) Strong individual learners in the ensemble.
- (d) Weak individual learners in the ensemble.

$$\text{Var}\left(\frac{1}{M} \sum_{m=1}^M h_m\right) \approx \rho\sigma^2 + \frac{1-\rho}{M}\sigma^2 \quad , \text{ where } \rho \text{ is the correlation factor}$$

Data resampling is not necessary (random forest uses random features)

QUIZ 03

A	B	C	D
1	12	2	5

2. Which of these does not describe an ensemble technique?

- (a) For each learner, change the weights of the different samples, pushing the learning to form a different decision boundary. AdaBoost
- (b) For each learner, find the principal components of the dataset and drop low variance features.
- (c) For each learner, resample observed samples with replacement, changing the data distribution. Bagging
- (d) For each learner, select a subset of columns, randomly dropping features.

Random Forest

QUIZ 03

A	B	C	D
0	3	0	17

3. Which of these statements about ensemble techniques and the bias-variance trade-off is false?

- (a) Bagging reduces the variance of the training dataset for each ensemble member.
- (b) Boosting can ensemble high-bias models to draw complex decision boundaries within datasets that have high variance.
- (c) Random forests reduce the variance of the training dataset for each ensemble member.
- (d) Adaboost works best with strong learners that already capture the variance of the dataset on their own.

Easy to overfit on strong learners.

$$\text{Var}\left(\frac{1}{M} \sum_{m=1}^M h_m\right) \approx \rho\sigma^2 + \frac{1-\rho}{M}\sigma^2$$

QUIZ 03

A	B	C	D
2	2	13	3

4. Which of these statements about (regular, not hidden) Markov models is false?

- (a) A time-invariant markov model only looks at the-chain of events within a finite number of steps. N-gram
- (b) A 0-gram Markov chain prediction has no dependency upon past tokens.
- (c) Training n-gram models with larger n takes more computation, but not more space or data. scales exponentially
- (d) As the length of n-grams grow, the training data required to fill the transition matrix rapidly grows.

Assume 3-grams with 5 states, then the number of paths is $5^3=125$

QUIZ 03

A	B	C	D
2	0	8	10

5. Which of these statements about a hidden Markov model with 3 hidden states and a vocabulary of 5 observable tokens is true?

- (a) The transition matrix, A, is size 3x5.
- (b) The emission matrix, B, is size 5x5. 3x5
- (c) The parameters of the HMM can be trained with the EM algorithm.
- (d) B and C are both true.

Transition matrix: probabilities transiting to the next state.

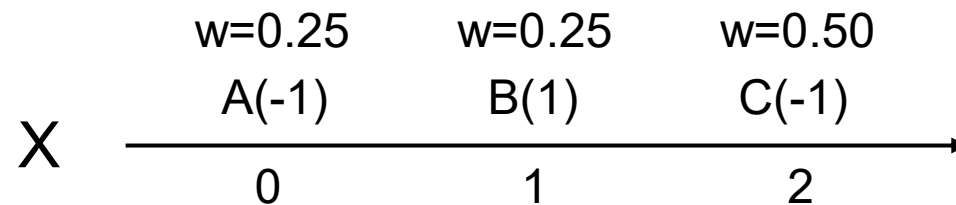
Emission matrix: probabilities of all states during different time steps.

QUIZ 03

6) Consider a set of datapoints with the given x values, classes, and current weights while creating an Adaboost ensemble of decision stumps.

Points	A	B	C
x	0	1	2
class	-1	1	-1
weight	0.25	0.25	0.5

6.1) A decision stump with polarity=1 will classify anything with values $<$ its boundary value as class -1 and any other points as class 1. A stump with polarity=-1 will do the opposite. Using the points in the table, what are all of the possible *weighted errors* for decision stumps at each x value and polarity?



QUIZ 03

Points	A	B	C
x	0	1	2
class	-1	1	-1
weight	0.25	0.25	0.5



For Polarity=1

For Polarity=-1

X=0

A(1)	B(1)	C(1)
e=0.25	e=0	e=0.50

Total error=0.75

A(-1)	B(-1)	C(-1)
e=0	e=0.25	e=0

Total error=0.25

X=1

A(-1)	B(1)	C(1)
e=0	e=0	e=0.50

Total error=0.5

A(1)	B(-1)	C(-1)
e=0.25	e=0.25	e=0

Total error=0.5

X=2

A(-1)	B(-1)	C(1)
e=0	e=0.25	e=0.50

Total error=0.75

A(1)	B(1)	C(-1)
e=0.25	e=0	e=0

Total error=0.25

6.2) Which of those stumps and polarity will be added into the ensemble? Identify the stump by its x value and polarity.

6.3) For simplicity, assign your stump a confidence of $\alpha = 0.7$. It is added into an existing ensemble with two stumps. The first is $x=2$, polarity=-1, and $\alpha = 0.5$. The second is $x=1$, polarity=1, and $\alpha = 0.5$. Use this ensemble of three stumps to classify the three points. Voting is done by summing the confidence-weighted votes and taking the sign. Show the sum and class for each point below.

Point A: _____

Point B: _____

Point C: _____

6.2) Which of those stumps and polarity will be added into the ensemble? Identify the stump by its x value and polarity.

Polarity=-1, X=0 OR 2

Just pick up the model with lowest sum error

6.3) For simplicity, assign your stump a confidence of $\alpha = 0.7$. It is added into an existing ensemble with two stumps. The first is $x=2$, polarity=-1, and $\alpha = 0.5$. The second is $x=1$, polarity=1, and $\alpha = 0.5$. Use this ensemble of three stumps to classify the three points. Voting is done by summing the confidence-weighted votes and taking the sign. Show the sum and class for each point below.

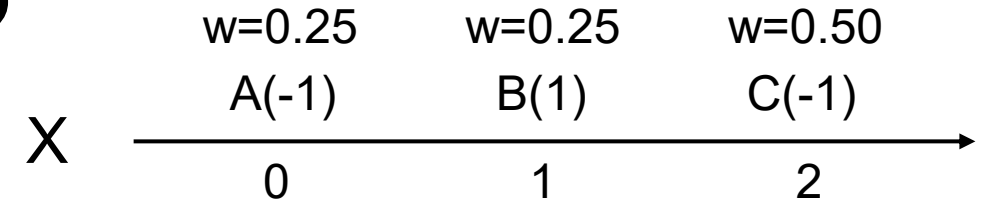
Point A: _____

Point B: _____

Point C: _____

QUIZ 03

Points	A	B	C
x	0	1	2
class	-1	1	-1
weight	0.25	0.25	0.5



(First Choice)

X=0, Polarity=-1, $\alpha=0.7$

A(-1)	B(-1)	C(-1)
-0.7	-0.7	-0.7

X=2, Polarity=-1, $\alpha=0.5$

A(1)	B(1)	C(-1)
0.5	0.5	-0.5

X=1, Polarity=1, $\alpha=0.5$

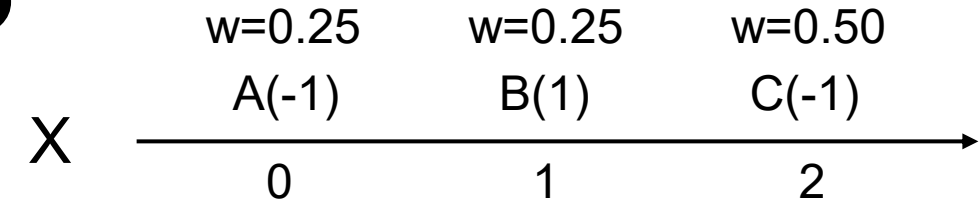
A(-1)	B(1)	C(1)
-0.5	0.5	0.5

SUM

A(-1)	B(1)	C(-1)
-0.7	0.3	-0.7

QUIZ 03

Points	A	B	C
x	0	1	2
class	-1	1	-1
weight	0.25	0.25	0.5



(Another situation)

X=2, Polarity=-1, $\alpha=0.7$

A(1)
0.7

B(1)
0.7

C(-1)
-0.7

X=2, Polarity=-1, $\alpha=0.5$

A(1)
0.5

B(1)
0.5

C(-1)
-0.5

X=1, Polarity=1, $\alpha=0.5$

A(-1)
-0.5

B(1)
0.5

C(1)
0.5

SUM

A(1)
0.7

B(1)
1.7

C(-1)
-0.7

6.2) Which of those stumps and polarity will be added into the ensemble? Identify the stump by its x value and polarity.

Polarity=-1, X=0 OR 2

Just pick up the model with lowest sum error

6.3) For simplicity, assign your stump a confidence of $\alpha = 0.7$. It is added into an existing ensemble with two stumps. The first is $x=2$, polarity=-1, and $\alpha = 0.5$. The second is $x=1$, polarity=1, and $\alpha = 0.5$. Use this ensemble of three stumps to classify the three points. Voting is done by summing the confidence-weighted votes and taking the sign. Show the sum and class for each point below.

Point A: -1, s=-0.7 OR 1, s=0.7

Point B: 1, s=0.3 OR 1, s=1.7

Both are correct.

Point C: -1, s=-0.7 OR -1, s=-0.7

Q&A