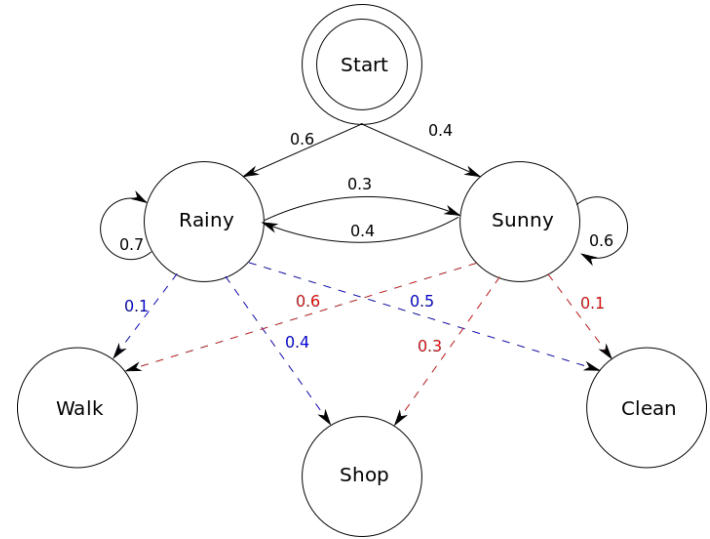


Viterbi Decoder

CS6230: CAD For VLSI

Hidden Markov Model

- **Markov Property:** Future state depends only on the current state and not the the events that occurred before it.
- **Hidden States:** Actual states of the system are not directly observable.
- **Observations:** Can only observe outputs generated by the hidden states



Viterbi Algorithm

- A dynamic programming algorithm that finds the **most likely sequence of hidden states** for a given sequence of observations made.
Commonly used with Hidden Markov Models.
- **Viterbi path:** Resulting sequence of hidden states.
- Applications:
 - Decoding convolutional codes in digital communication
 - Parts of speech tagging

Mathematical Formulation

States: $q_0, q_1, q_2, \dots, q_N$ (q_0 = start state) N - no of states

Observations: k_1, k_2, \dots, k_M M - no of observations

Transition probabilities: a_{ij} (probability of transitioning from state q_i to state q_j)

Emission probabilities: $b_j(o_t)$ (probability of observing o_t given the current state q_j)

Viterbi Path Probability: $v_t(j)$ (probability of the most likely path ending in state q_j at time t)

Input sequence of observations: $o_1, o_2, o_3, \dots, o_T$ (o_t can be any of k_1, k_2, \dots, k_M)

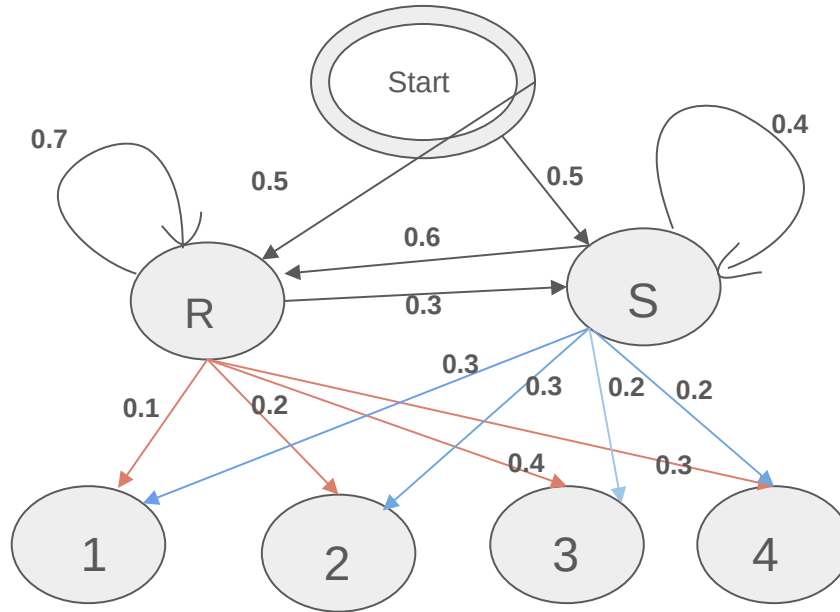
$$(t > 1) \quad V_t(j) = \max_{(0 \leq i < N+1)} (V_{t-1}(i) * a_{ij} * b_j(o_t))$$

When $t = 1$, $V_1(j) = a_{q_0, j} * b_j(o_1)$

Example

$N = 2$ (Rainy, Sunny)

$M = 4$ (Walk, Shop, Study, Clean)



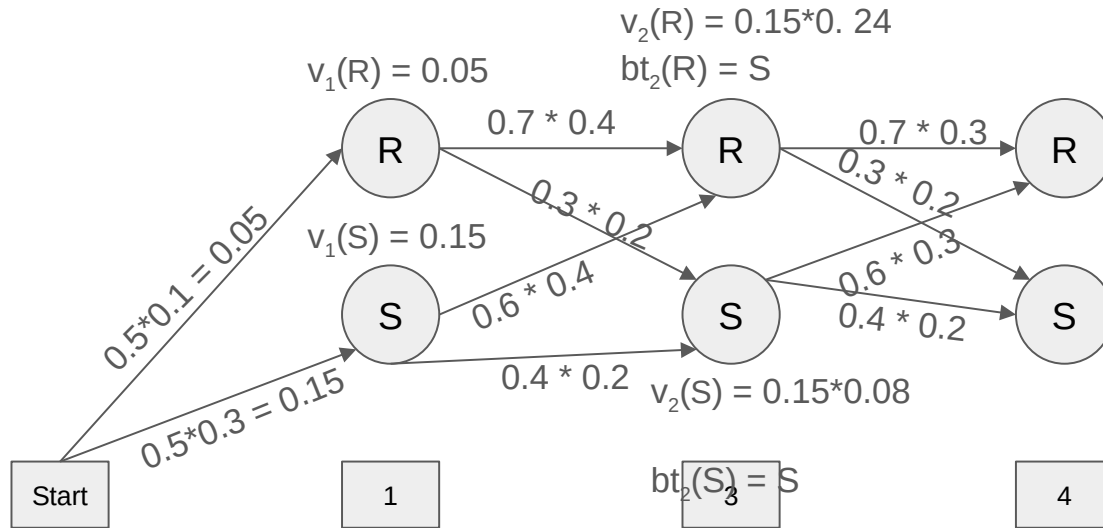
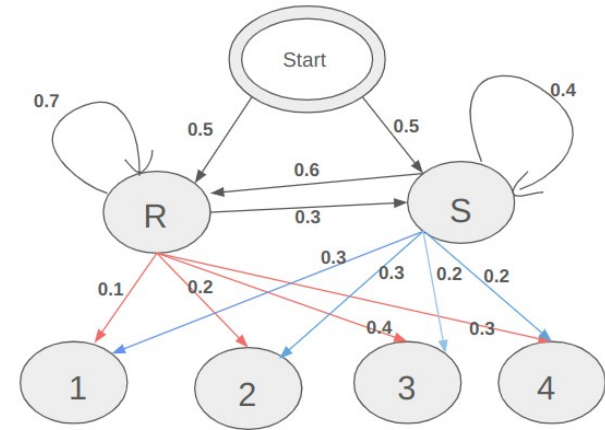
Observed for 3 consecutive days: 1,3,4 (W,S,C)

Most probable sequence of weather?

Observed for 3 consecutive days: 1,3,4 (W,S,C)

Most probable sequence of weather?

Ans: S,R,R (2,1,1) with prob = $0.15 * 0.24 * 0.21$



$$v_3(R) = 0.15 * 0.24 * 0.21$$

$$bt_3(R) = R$$

$$v_3(S) = 0.15 * 0.24 * 0.06$$

$$bt_3(S) = R$$

Pseudocode

T - no of observations in input sequence

bt[t][j]: backtrace array which records prev state that led to the max probability ending at state j at time t

prob[]: current max path probability ending at each state

temp[]: temporary array to store updated values of probabilities at each time step

Initialization(t = 1): $\text{prob}[j] = a[0][j] * b[j][o_1]$ for all $j : 1 \text{ to } N$
 $\text{bt}[1][j] = q_0$

Iterative step:

for each time step t from 2 to T do

for each state j from 1 to N do

 {

 max_val = 0

 max_state = -1

for i in 1 to N:

 { val = $\text{prob}[i] * a[i][j] * b[j][o_t]$

if val > max_val then:

 max_val = val;

 max_state = i;

 }

 temp[j] = max_val, bt[t][j] = max_state

 }

prob = temp; //copy temp array back to prob

Contd.

Termination:

path[]: final viterbi path of hidden states

vprob: final probability of mostly likely sequence

vprob =0;

for i in 1 to N:

{

if prob[i] > vprob:

 vprob = prob[i] ;

 path[T] = i ;

}

for t in T-1 -> 1:

{

 path[t] = bt[t+1][path[t+1]];

}