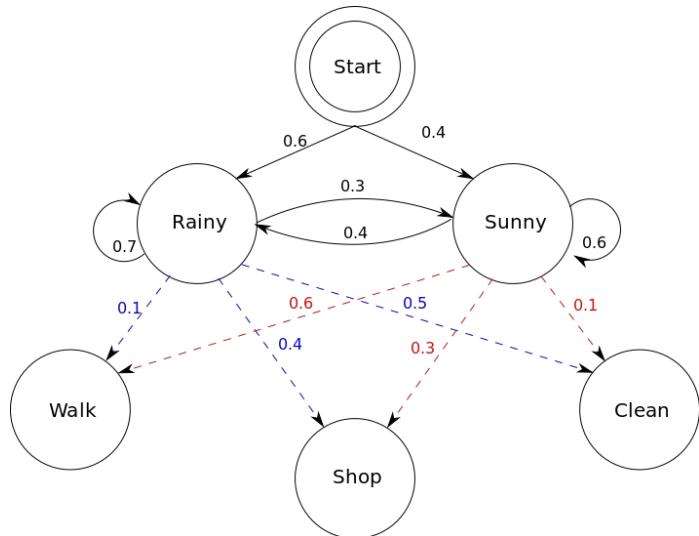


# Viterbi Decoder

CS6230: CAD For VLSI

# Hidden Markov Model

- **Markov Property:** Future state depends only on the current state and not 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  $k_1, k_2, \dots, k_M$ )

(  $t > 1$  )

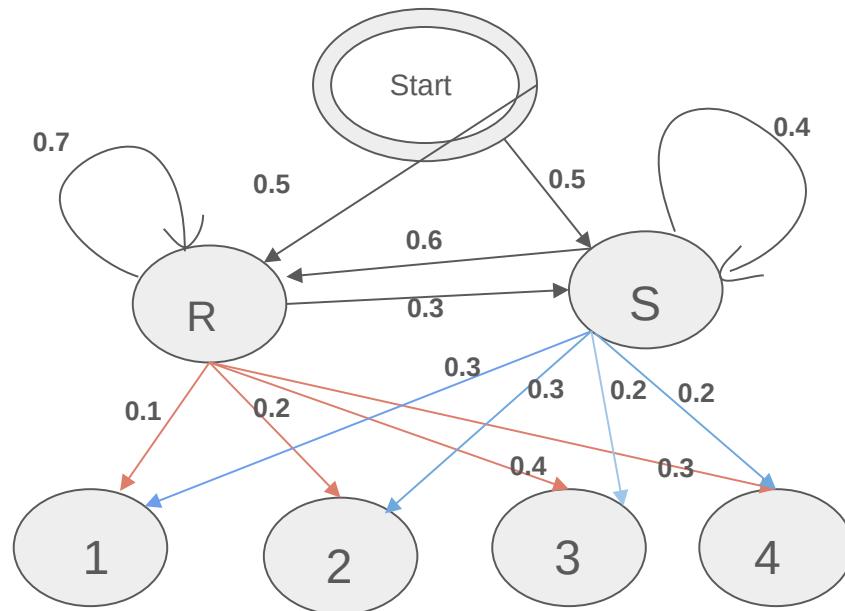
$$V_t(j) = \max_{(0 < i < N+1)} ( V_{t-1}(i) * a_{ij} * b_j(o_t) )$$

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

## Example

$N = 2$  (Rainy, Sunny)

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



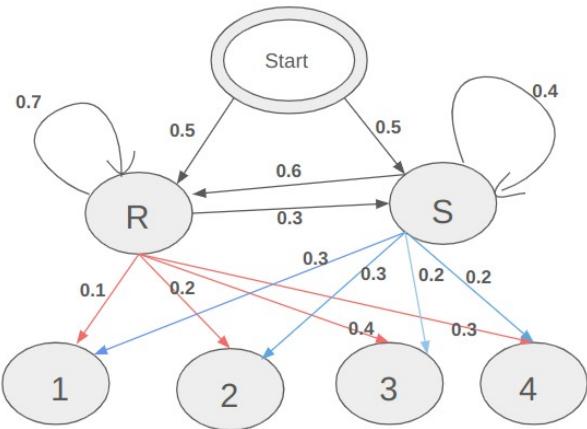
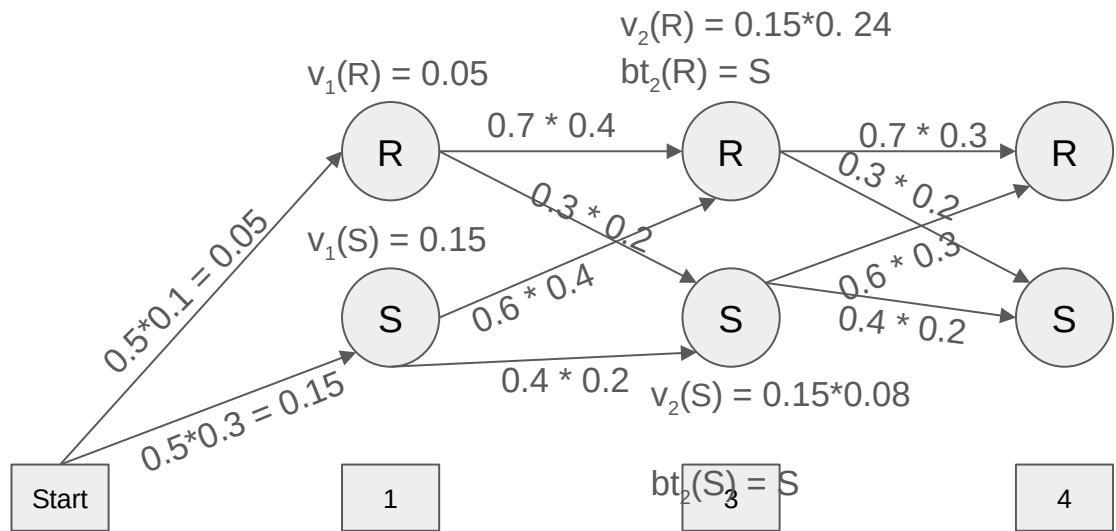
Observed for 3 consecutive days: 1,3,4 (W,S,C)

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  $j : 1$  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 = prob[i] \* a[i][j] \* b[j][o<sub>t</sub>]

**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]];  
}
```