ECE/CS 498 DSU/DSG Spring 2020
In-Class Activity 4

NetID: _____

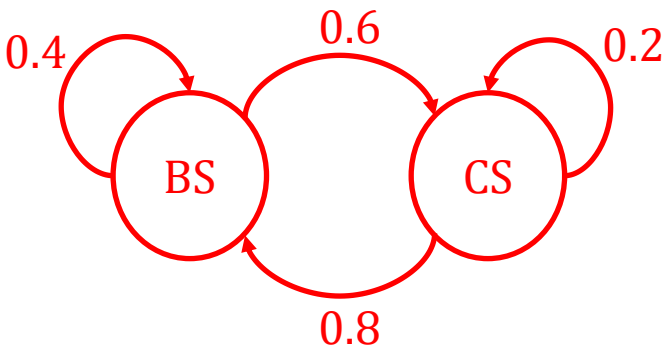
The purpose of the in-class activity is for you to:

(i) Understand how to model a time series prediction problem as an HMM.
(ii) Go through the forward-backward algorithm for predicting the most likely hidden state given the time series observations.

Problem 1

The security state of a computer can be either in a **benign state (BS)** or in a **compromised state (CS)**. The computer is constantly being attacked by hackers. The probability that an attack is successful, and the computer moves from benign to compromised is **0.6**. The probability that the computer, given that is in a compromised state, detects the attacker and transitions from the compromised state to the benign state is **0.8**. In all other situations, the state remains unchanged. **The transition probabilities are independent of the past states given the current state of the system.** At any point in time, it is believed, that the computer is in the benign state with probability **0.9**.

a) Draw the states and state transition probabilities that describe this system:



There is no way of directly observing the state of the computer. On the other hand, there are **system events** like **port scanning (PS)** and **web browsing (WB)** that can be observed. The probability of observing an event depends only on the state of the computer. The probability that a benign user does a port scan is **0.4** and does web browsing is **0.6**. An attacker will perform a port scan with a probability of **0.7** and web browsing with a probability of **0.3**.

During an observation period, the following sequence of events was observed: [WB, PS, WB] corresponding to t=1, 2 and 3 respectively. Answer the following questions.

b) Is it possible to identify the exact state of the computer at time instant two (t=2)? If not, state a condition when you can fully determine (with 100% probability) the system's state after observing an event.
No, observing the event PS at t=2 does not give us a definitive answer on what the current state of the system is. This is because both states BS and CS can produce the event PS with some probability. Only when an event is exclusive to a system state can we fully determine the system's state.

c) Is it possible to identify the most likely state of the computer at t=2?
Yes. Since we are given both the state transition probabilities and the observation probabilities given the state, we can find the most probable security state at t=2.

d) Mathematically express the property of the transition probability of states mentioned above. What is the property known as?
Markov Property: $P(S_t|S_{t-1}, S_{t-2} \dots S_1) = P(S_t|S_{t-1})$

e) Which of the following models can be used to answer the question in part (c)? Explain your answer.
Linear Regression:
No. Linear regression cannot model the time series evolution of states.

Markov Models:
No. While the state of the Markov Model is known, we don't know the precise state of the system at the time of observation.

Hidden Markov Models:
Yes, HMMs, using the observations and the underlying Markov Model can estimate the state of the system for the given observations.

Now that is clear that we need to use HMM to solve answer the question in part (c), let us set up the model.

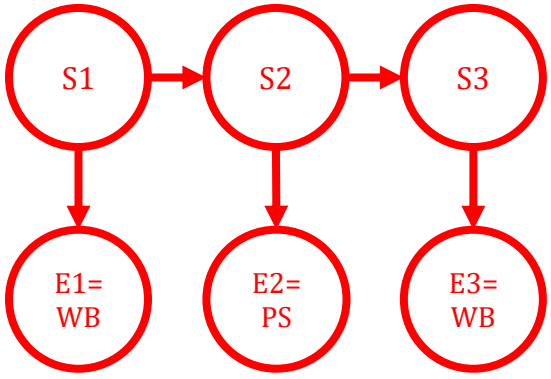
f) Write down the state transition probability matrix **A**, observation matrix **B** and the initial distribution of hidden states $\boldsymbol{\pi}$.

$$\boldsymbol{A} = \begin{array}{c|cc} & BS & CS \\ \hline BS & \mathbf{0.4} & \mathbf{0.6} \\ CS & \mathbf{0.8} & \mathbf{0.2} \end{array}$$

$$\boldsymbol{B} = \begin{array}{c|cc} & PS & WB \\ \hline BS & \mathbf{0.4} & \mathbf{0.6} \\ CS & \mathbf{0.7} & \mathbf{0.3} \end{array}$$

$$\boldsymbol{\pi} = \begin{array}{cc} BS & CS \\ \hline \mathbf{0.9} & \mathbf{0.1} \end{array}$$

g) Draw the HMM model. Denote the hidden states as S_t for $t \in \{1,2,3\}$.



To predict the most likely state for S_2 , we need to compute

$$S_2^* = \underset{\sigma_j \in \{BS,CS\}}{\text{argmax}}\ \gamma_2(j)$$

where

$$\gamma_2(j) = P\big(S_2 = \sigma_j \big| E_1,\ E_2,\ E_3\big)$$

Recall from the lecture slides, that to calculate γ_2, we need to perform the forward algorithm which gives us α_2, and the backward algorithm that produces β_2.

h) Compute α_2 recursively using the forward algorithm.

<WB> (t=1)			
States	$\boldsymbol{\alpha_1}$		Normalize α_1
BS	$\alpha_1(BS)$	$P(S_1 = BS) \times P(E_1 = WB \mid S_1 = BS)$ $= 0.9 \times 0.6 = 0.54$	$\frac{0.54}{0.57} = 0.947$
CS	$\alpha_1(CS)$	$P(S_1 = CS) \times P(E_1 = WB \mid S_1 = CS)$ $= 0.1 \times 0.3 = 0.03$	$\frac{0.03}{0.57} = 0.053$

<WB, PS> (t=2)			
States	$\boldsymbol{\alpha_2}$		Normalize α_2
BS	$\alpha_2(BS)$	$[\alpha_1(BS) \times P(S_2 = BS \mid S_1 = BS)$ $+ \alpha_1(CS) \times P(S_2 = BS \mid S_1 = CS)] \times P(E_2 = PS \mid S_2 = BS)$ $= [0.947 \times 0.4 + 0.053 \times 0.8] \times 0.4 = 0.168$	$\frac{0.168}{0.573} = 0.293$
CS	$\alpha_2(CS)$	$[\alpha_1(BS) \times P(S_2 = CS \mid S_1 = BS)$ $+ \alpha_1(CS) \times P(S_2 = CS \mid S_1 = CS)] \times P(E_2 = PS \mid S_2 = CS)$ $= [0.947 \times 0.6 + 0.053 \times 0.2] \times 0.7 = 0.405$	$\frac{0.405}{0.573} = 0.707$

i) Compute β_2 recursively using the backward algorithm.
Note: We initialize $\beta_3(BS) = 1,\ \beta_3(CS) = 1$

<WB, PS> (t=2) (WB observed at t=3)		
States	$\boldsymbol{\beta_2}$	
BS	$\beta_2(BS)$	$P(S_3 = BS \mid S_2 = BS) \times P(E_3 = WB \mid S_3 = BS) \times \beta_3(BS)$ $+P(S_3 = CS \mid S_2 = BS) \times P(E_3 = WB \mid S_3 = CS) \times \beta_3(CS)$ $= 0.4 \times 0.6 \times 1 + 0.6 \times 0.3 \times 1 = 0.42$
CS	$\beta_2(CS)$	$P(S_3 = BS \mid S_2 = CS) \times P(E_3 = WB \mid S_3 = BS) \times \beta_3(BS)$ $+ P(S_3 = CS \mid S_2 = CS) \times P(E_3 = WB \mid S_2 = CS) \times \beta_3(CS)$ $= 0.8 \times 0.6 \times 1 + 0.2 \times 0.3 \times 1 = 0.54$

j) Compute γ_2 and find S_2^*

<WB, PS> (t=2)			
States	$\boldsymbol{\gamma_2}$		Normalize γ_2
BS	$\gamma_2(BS)$	$\beta_2(BS) \times \alpha_2(BS) = 0.42 \times 0.293 = 0.123$	$\frac{0.123}{0.505} = 0.244$
CS	$\gamma_2(CS)$	$\beta_2(CS) \times \alpha_2(CS) = 0.54 \times 0.707 = 0.382$	$\frac{0.382}{0.505} = 0.756$

By taking the argmax of γ_2, we can see that the most likely security state of the computer at t=2 is $S_2 = CS$.

k) What if the observation matrix B was modified to be the following: $B = \begin{bmatrix} 0.9 & 0.1 \\ 0.9 & 0.1 \end{bmatrix}$.
What is the most likely state? (Hint: The observation matrix does not give us any additional information of which hidden state is more likely given an observation)

Since the new observation matrix B does not give us additional information on the hidden state at any point in time, we can look at the prior probabilities π and the transition matrix A and infer the most likely state to be CS.