

HW3

```
library(tidyverse)
```

Given Quantities

- Transition Matrix

```
transition_m =  
  matrix(c(0.9, 0, # first column - transition from states 1 and 2 to 1  
           .1, 1), # second column - transition from states 1 and 2 to 2  
         nrow = 2, ncol = 2)
```

```
transition_m
```

```
      [,1] [,2]  
[1,]  0.9  0.1  
[2,]  0.0  1.0
```

- Starting Probabilities

```
prob_X1_eq_1 = 0.8  
prob_X1_eq_2 = 0.2
```

- Emission Probabilities

```
emission_m =  
  matrix(c(.99, .01, # first column - transition from states 1 and 2 to 1  
           .96, .04), # second column - transition from states 1 and 2 to 2
```

```

        nrow = 2, ncol = 2)

rownames(emission_m) = c("a", "u")

emission_m

[,1] [,2]
a 0.99 0.96
u 0.01 0.04

```

Forward Algorithm

```

forward_HMM_k <- function(

  ## returns: joint pmf for a given observed signal
  ##           and possible hidden states of a markov chain

  TRANSITION_M, ## obv. a matrix
  EMISSION_M, ## matrix of emitted probabilities
  EMITTED_SIGNALS, ## needs to be a vector
  STARTING_P, ## needs to be a vector
  DESIRED_K
){

  ## initialize
  first_signal = EMITTED_SIGNALS[1] # first signal
  STARTING_P * EMISSION_M[rownames(EMISSION_M) == first_signal] -> working_F

  if(DESIRED_K > 1){
    ## future states 2, 3, etc...
    for( STEPS in 2:DESIRED_K){

      current_F = rep(NA, length(STARTING_P))

      for( STATES in 1:length(STARTING_P)){

        current_F[STATES] <-
          EMISSION_M[
            ## get conditional probs given potential HMM state

```

```

    rownames(EMISSION_M) ==
      ## get probability for the Hidden state we work with here
      EMITTED_SIGNALS[STEPS]][STATES] *

    ## multiply previous Forward F's with transition probabilities
    sum(working_F * TRANSITION_M[,STATES] )

  }

  working_F = current_F
}
}else{
  current_F = working_F
}

return(current_F)
}

### test the algo

forward_HMM_k(
  TRANSITION_M = transition_m,
  EMISSION_M = emission_m,
  EMITTED_SIGNALS = c("a", "u", "a"),
  STARTING_P = c(0.8, 0.2),
  DESIRED_K = 3
) %>% round(., 6)

```

```
[1] 0.006351 0.011098
```

Backward Algo

```

backward_HMM_k <-
function(
  TRANSITION_M,
  EMISSION_M,
  EMITTED_SIGNALS,
  STARTING_P,
  LAST_K

```

```

){

current_B = rep(1, length(STARTING_P))
## past states states 2, 1, etc... {2 and 1 is all we get in this example}

backward_state_sequence = seq(from = length(EMITTED_SIGNALS) - 1, to = LAST_K, by = -1)

for( STEPS in backward_state_sequence){

  working_B = rep(NA, length(STARTING_P))

  for( STATES in 1:length(STARTING_P)){

    working_B[STATES] <-
      sum(
        EMISSION_M[rownames(EMISSION_M) ==
                    EMITTED_SIGNALS[STEPS + 1]] *
        TRANSITION_M[STATES,] *
        current_B
      )
  }

  current_B = working_B

}

return(current_B)
}

backward_HMM_k(
  TRANSITION_M = transition_m,
  EMISSION_M = emission_m,
  EMITTED_SIGNALS = c("a", "u", "a"),
  STARTING_P = c(0.8, 0.2),
  LAST_K = 1
) %>% round(. ,4)

```

```
[1] 0.0127 0.0384
```

```
backward_HMM_k(
  TRANSITION_M = transition_m,
  EMISSION_M = emission_m,
  EMITTED_SIGNALS = c("a", "u", "a"),
  STARTING_P = c(0.8, 0.2),
  LAST_K = 2
) %>% round(. ,4)
```

```
[1] 0.987 0.960
```

Conditional PMF

```
conditional_pmf <-
function(
  DESIRED_STATE,
  DESIRED_K_H,

  EMITTED_SIGNALS_H,
  TRANSITION_M_H,
  EMISSION_M_H,
  STARTING_P_H,

  LAST_K_H

){

  forward_HMM_k(
    TRANSITION_M = TRANSITION_M_H,
    EMISSION_M = EMISSION_M_H,
    EMITTED_SIGNALS = EMITTED_SIGNALS_H,
    STARTING_P = STARTING_P_H,
    DESIRED_K = DESIRED_K_H
  ) -> forward_part

  backward_part <-

  ifelse(
    ## if we want to predict X at the time of last observed
    ## signal, then we will assign B_k all 1
```

```

    ## otherwise, we will use the fucntion to calculate our stuff
    DESIRED_K_H == length(EMITTED_SIGNALS_H),

    rep(1, length(STARTING_P_H)),

    backward_HMM_k(
      TRANSITION_M = TRANSITION_M_H,
      EMISSION_M = EMISSION_M_H,
      EMITTED_SIGNALS = EMITTED_SIGNALS_H,
      STARTING_P = STARTING_P_H,
      LAST_K = LAST_K_H
    )
  )

  ### now we need to recreate

  prod = forward_part * backward_part

  results = prod[DESIRED_STATE]/sum(prod)

  return(results)
}

conditional_pmf(
  DESIRED_STATE = 1,
  DESIRED_K_H = 3,
  TRANSITION_M_H = transition_m,
  EMISSION_M_H = emission_m,
  STARTING_P_H = c(0.8, 0.2),
  EMITTED_SIGNALS_H = c("a", "u", "a") ,

  LAST_K_H = 1
) %>% round(., 3)

```

```
[1] 0.364
```

Viterbi

```
#initial

viterbi_algo <-
function(
  STARTING_P,
  TRANSITION_M,
  EMITTED_SIGNALS,
  EMISSION_M
){

  ## Store two probabilities to chose from for  $V_{\{t\}}(j)$ 
  v_t_j_probs = matrix(rep(NA, length(EMITTED_SIGNALS) * length(STARTING_P) * 2),
    nrow = 4)

  ## Here we will store  $X_{\{t\}}$  states picked on the highest value of  $V_t(j)$  in each step
  x_star = rep(NA, length(EMITTED_SIGNALS))

  ## Store probabilities that were used to pick  $X_{i,t,j}$  at each step
  xai_t_j_probs = matrix(rep(NA, length(EMITTED_SIGNALS) * length(STARTING_P) * 2),
    nrow = 4)

  ## here we will store values of  $X_{i,t}$  function to retrieve most likely values of  $X_{\{t\}}$ 
  xai_star = matrix(rep(NA, length(EMITTED_SIGNALS) * length(STARTING_P)),
    nrow = 2)

  ##### ***** STEP 1: Initialize #####
  initial_V = STARTING_P * EMISSION_M[rownames(EMISSION_M) == EMITTED_SIGNALS[1]]

  x_star[1] = which(initial_V == max(initial_V))

  ## loop over time points
  for(TIMES in 2:length(EMITTED_SIGNALS)){

    working_V = rep(NA, length(STARTING_P))
    working_X = rep(NA, length(STARTING_P))

    ## loop over the number of hidden states
    for(J in 1:length(STARTING_P)){
```

```

    ## Calculate V's
    iter_Vs <-
      EMISSION_M[rownames(EMISSION_M) == EMITTED_SIGNALS[TIMES]][J] *
      TRANSITION_M[J,] * initial_V

    ## store both probabilities
    v_t_j_probs[c(J + (J-1), J + J) ,TIMES] = iter_Vs

    ## pick the highest of the two probabilities and store it as value of V_{t}(j)
    working_V[J] = max(iter_Vs)

    ## Calculate Xai's
    iter_Xs <- TRANSITION_M[,J] * initial_V
    ## store these probabilities
    xai_t_j_probs[c(J + (J-1), J + J) ,TIMES] = iter_Xs

    ## store index of state that corresponds to the highest probability
    xai_star[J, TIMES] <- which(iter_Xs == max(iter_Xs))

  }
  x_star[TIMES] = which(working_V == max(working_V))

  initial_V = working_V
}

recovered_states = rep(NA, length(EMITTED_SIGNALS))

# last recovered state is just the one picked based on the
# highest probability for the last state
recovered_states[length(EMITTED_SIGNALS)] = x_star[length(EMITTED_SIGNALS)]

## backward loop
for(j in seq(from = length(EMITTED_SIGNALS) - 1,
             to = 1,
             by = -1)){

  ## current X_t is based on previous X_{t+1} and previous Xai_{t+1}
  ## X_{t+1} serves as an index for Xai_{t+1}
  ## value retrieved based on this index is assigned to X_t
  prev_x = x_star[j + 1]

```



```

    recovered_states[j] = xai_star[prev_x, (j+1)]
  }

  ## make matrices nicer for printing
  rownames(v_t_j_probs) = c("j = 1, index = 1", "j = 1, index = 2",
                             "j = 2, index = 1", "j = 2, index = 2")
  colnames(v_t_j_probs) = paste0("t = ", seq(from = 1, to = dim(v_t_j_probs)[2], by = 1))

  rownames(xai_t_j_probs) = c("j = 1, index = 1", "j = 1, index = 2",
                              "j = 2, index = 1", "j = 2, index = 2")
  colnames(xai_t_j_probs) = paste0("t = ", seq(from = 1, to = dim(xai_t_j_probs)[2], by

  ## print all results
  print("INTERMEDIATE RESULTS OF ESTIMATION")

  options(scipen = 999)
  print("")
  print("Estiamted Probabilities for chosing value of V_{t}(j) at each time point t")
  print(v_t_j_probs %>% round(., 4))

  print("")
  print("Indexes of higer value of V_{j} at each time point t")
  print(x_star)

  print("")
  print("Estiamted Probabilities for chosing value of Xai_{t}(j) at each time point t")
  print(xai_t_j_probs %>% round(., 4))

  print("")
  print("Indexes of higer value of Xai_{j} at each time point t")
  print(xai_star)

  return(recovered_states)
}

viterbi_algo(
  STARTING_P = c(.8, .2),
  TRANSITION_M = transition_m,
  EMITTED_SIGNALS = c("a", "u", "a", 'u', 'a'),
  EMISSION_M = emission_m
) -> recovered_states_results

```

```

[1] "INTERMEDIATE RESULTS OF ESTIMATION"
[1] ""
[1] "Estiamted Probabilties for chosing value of V_{t}(j) at each time point t"
      t = 1  t = 2  t = 3  t = 4  t = 5
j = 1, index = 1    NA 0.0071 0.0064 0.0001 0.0001
j = 1, index = 2    NA 0.0002 0.0008 0.0000 0.0000
j = 2, index = 1    NA 0.0000 0.0000 0.0000 0.0000
j = 2, index = 2    NA 0.0077 0.0074 0.0003 0.0003
[1] ""
[1] "Indexes of higer value of V_{j} at each time point t"
[1] 1 2 2 2 2
[1] ""
[1] "Estiamted Probabilties for chosing value of Xai_{t}(j) at each time point t"
      t = 1  t = 2  t = 3  t = 4  t = 5
j = 1, index = 1    NA 0.7128 0.0064 0.0057 0.0001
j = 1, index = 2    NA 0.0000 0.0000 0.0000 0.0000
j = 2, index = 1    NA 0.0792 0.0007 0.0006 0.0000
j = 2, index = 2    NA 0.1920 0.0077 0.0074 0.0003
[1] ""
[1] "Indexes of higer value of Xai_{j} at each time point t"
      [,1] [,2] [,3] [,4] [,5]
[1,]    NA     1     1     1     1
[2,]    NA     2     2     2     2

```

Final Answer

Sequence of states that resulted in the highest likelihood of emitted signals (0,1,0,1,0) is:

```
[1] 2 2 2 2 2
```

Just Final Answers

$F_n(j)$ from forward equations:

- $F_1(j)$:

```
[1] 0.792 0.192
```

- $F_2(j)$:

[1] 0.007128 0.010848

- $F_3(j)$:

[1] 0.006351 0.011098

$B_k(j)$ from backward equations for $k = 1, 2$:

- $B_1(i)$:

[1] 0.0127 0.0384

- $B_2(i)$:

[1] 0.987 0.960

- $B_3(i)$:

[1] 1 1

$P(X_3 = 1 | s_3 = (0, 1, 0))$ using forward and backward equations:

[1] 0.364

Likely sequence of states given the sequence of signals $s_5 = (0, 1, 0, 1, 0)$ is:

[1] 2 2 2 2 2