sol_lab2

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Question 1

Creating the probability matricies

```
n = 10
startP = rep(1/n,n)
startP
   P = diag(n) * .5
# 1,2 2,3 3,4... n-1,n
for( i in 1:(n-1)){
 P[i,i+1]=.5
P[n,1] = .5 \#can loop around
head(P)
##
       [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
## [1,] 0.5 0.5 0.0 0.0 0.0 0.0 0.0
                                                  0
## [2,]
       0.0 0.5 0.5 0.0 0.0 0.0 0.0
                                                  0
## [3,]
       0.0 0.0 0.5 0.5 0.0 0.0 0.0
                                             0
## [4,]
       0.0 0.0 0.0
                     0.5
                         0.5 0.0 0.0
                                                  0
## [5,]
       0.0 0.0 0.0 0.0 0.5 0.5 0.0
                                        0
                                                  0
## [6,]
       0.0 0.0 0.0 0.0 0.0 0.5 0.5
emissionP = matrix(data = 0, nrow=n,ncol=n)
for (i in 1:n){
 inds = ((i-2):(i+2))-1
 for ( j in ( inds %% 10 ) ){
   emissionP[i,j+1]=0.2
 }
```

```
head(emissionP)
       [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
## [1,] 0.2 0.2 0.2 0.0 0.0 0.0 0.0 0.0 0.2
       0.2 0.2 0.2 0.2
                          0.0
                             0.0
                                  0.0
       0.2 0.2 0.2
                              0.0
                                  0.0
## [3,]
                     0.2
                          0.2
## [4,]
       0.0 0.2 0.2
                     0.2
                          0.2
                             0.2
                                  0.0
       0.0 0.0 0.2 0.2 0.2 0.2 0.2 0.0
## [5,]
## [6,] 0.0 0.0 0.0 0.2 0.2 0.2 0.2
                                                 0.0
```

Starting probabilities, Transission probabilities and Emission probabilities, are all shown above.

```
symbols = LETTERS[1:n]
states = rep("State_",n)
for ( i in 1:n){
    states[i] = paste(states[i],symbols[i],sep="")
}
mod = initHMM(States=symbols, Symbols = symbols, startProbs=startP, transProbs=P, emissionProbs = emiss
```

Question 2

```
set.seed(123)
simulated_path = simHMM(mod,100)
print(simulated_path)
## $states
    [1] "D" "E" "F" "G" "H" "H" "I" "I" "I" "J" "J" "A" "A" "A" "B" "B" "C" "C"
   [55] "G" "G" "H" "I" "J" "J" "A" "B" "C" "D" "D" "E" "E" "E" "F" "G" "H"
   [73] "T" "J" "A" "B" "C" "D" "D" "D" "D" "E" "F" "G" "G" "G" "H" "T" "T" "J"
   [91] "A" "A" "B" "B" "C" "C" "D" "D" "D" "E"
##
## $observation
    [1] "E" "C" "E" "F" "J" "F" "H" "J" "A" "J" "I" "B" "I" "C" "J" "B" "A" "C"
   [19] "A" "E" "D" "C" "C" "C" "C" "F" "G" "G" "H" "D" "I" "F" "J" "J" "B" "J"
   [37] "C" "J" "C" "A" "D" "A" "B" "B" "C" "D" "D" "B" "D" "B" "F" "E" "C" "G"
   [55] "H" "G" "E" "G" "A" "H" "B" "B" "J" "D" "C" "E" "D" "C" "E" "G" "I" "H"
   [73] "H" "A" "J" "J" "D" "E" "F" "E" "F" "C" "D" "G" "E" "E" "G" "A" "H" "I"
   [91] "B" "B" "C" "J" "C" "A" "E" "B" "D" "E"
```

Question 3

```
obs = simulated_path$observation

f = forward(mod,obs)

filtered = prop.table(t(exp(f)),margin=1)

b = backward(mod, obs)

smoothed = prop.table(t(exp(f + b)),margin=1)

most_probable_path = viterbi(mod,obs)
```

```
guesses = matrix(nrow=100,ncol=3)
colnames(guesses) = c("filtered", "smoothed", "viterbi")

for (i in 1:100){
    filtered_guess = symbols[ which.max(filtered[i,])]
    smoothed_guess = symbols[ which.max(smoothed[i,])]
    guesses[i,1:2] = c(filtered_guess, smoothed_guess)
}
guesses[,3] = most_probable_path

acc = colMeans(guesses == simulated_path$states)
acc

## filtered smoothed viterbi
## 0.53     0.64     0.36
```

Question 5

```
simulated_path = simHMM(mod,100)
obs = simulated_path$observation

f = forward(mod,obs)

filtered = prop.table(t(exp(f)),margin=1)

b = backward(mod, obs)

smoothed = prop.table(t(exp(f + b)),margin=1)
```

```
most_probable_path = viterbi(mod,obs)
guesses = matrix(nrow=100,ncol=3)
colnames(guesses) = c("filtered", "smoothed", "viterbi")

for (i in 1:100){
    filtered_guess = symbols[ which.max(filtered[i,])]
    smoothed_guess = symbols[ which.max(smoothed[i,])]
    guesses[i,1:2] = c(filtered_guess, smoothed_guess)
}
guesses[,3] = most_probable_path
acc = colMeans(guesses == simulated_path$states)
acc

## filtered smoothed viterbi
## 0.59 0.71 0.48
```

```
acc_mat = matrix(ncol=4, nrow=0)
colnames(acc_mat) = c("NumSteps", "filtered_acc", "smoothed_acc", "viterbi_acc")
for ( temp in 1:50){
 n = 100
 simulated_path = simHMM(mod,n)
 obs = simulated_path$observation
 f = forward(mod,obs)
 filtered = prop.table(t(exp(f)),margin=1)
  b = backward(mod, obs)
  smoothed = prop.table(t(exp(f + b)) ,margin=1)
 most_probable_path = viterbi(mod,obs)
  guesses = matrix(nrow=n,ncol=3)
  colnames(guesses) = c("filtered", "smoothed", "viterbi")
  for (i in 1:n){
   filtered_guess = symbols[ which.max(filtered[i,])]
   smoothed_guess = symbols[ which.max(smoothed[i,])]
```

```
guesses[i,1:2] = c(filtered_guess, smoothed_guess)
}
guesses[,3] = most_probable_path

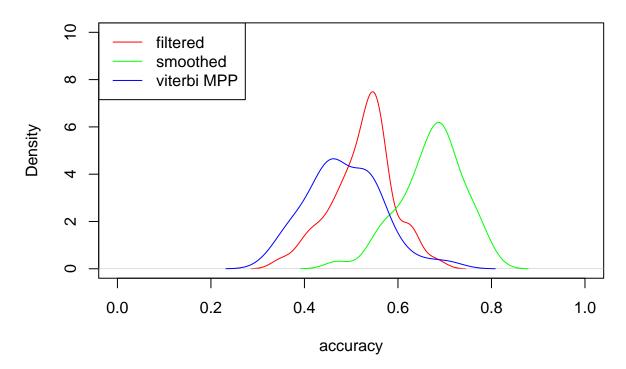
acc = colMeans(guesses == simulated_path$states)
#print(acc)
acc_mat = rbind(acc_mat, c(n,acc))

}

plot(density(acc_mat[,2]), col ="red",xlab="accuracy", xlim=c(0,1), ylim=c(0,10))
lines(density(acc_mat[,3]), col ="green")
lines(density(acc_mat[,4]), col ="blue")

legend("topleft",legend = c("filtered", "smoothed", "viterbi MPP"), col=c("red", "green", "blue"), lty
```

density.default(x = acc_mat[, 2])



The smoothed distribution takes future data into account and thus has more information availabe than the the filtered. This is an "advantage".

The most probable path has the be a valid path which is a restriction not placed on the smoothed distribution.

```
library()
mat_ents = matrix(nrow=300, ncol=150)

for ( i in 1:150){
    simulated_path = simHMM(mod,300)
    obs = simulated_path$observation

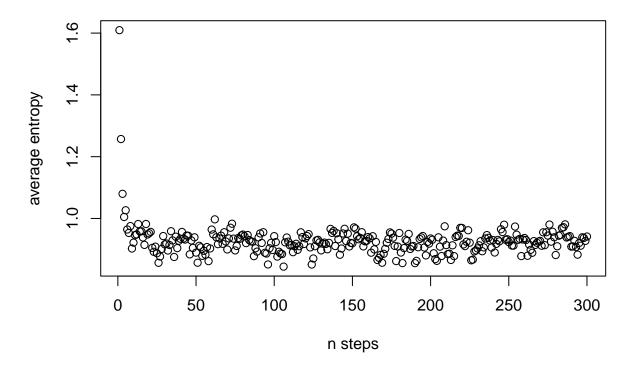
    f = forward(mod,obs)

    filtered = prop.table(t(exp(f)),margin=1)
    entropy_vec = apply(filtered,1, entropy.empirical))
    mat_ents[,i] = entropy_vec

}

ent_means = apply(mat_ents,1,mean)
plot(x=1:300, y = ent_means, main="averag entropy after n steps", xlab="n steps", ylab="average entropy")
```

averag entropy after n steps



There is a sharp decline after the first (\sim 3) steps. After that the values seem the be relatively stable

```
set.seed(123)
simulated_path = simHMM(mod,100)
obs = simulated_path$observation

f = forward(mod,obs)

filtered = prop.table(t(exp(f)),margin=1)

prob_101 = round(filtered[100,] %*% P, digits = 3)
rownames(prob_101) = "Inferred Probabilty"
colnames(prob_101) = symbols
prob_101

## A B C D E F G H I J
```

Inferred Probabilty 0 0 0.049 0.223 0.375 0.277 0.076 0 0 0 $\,$