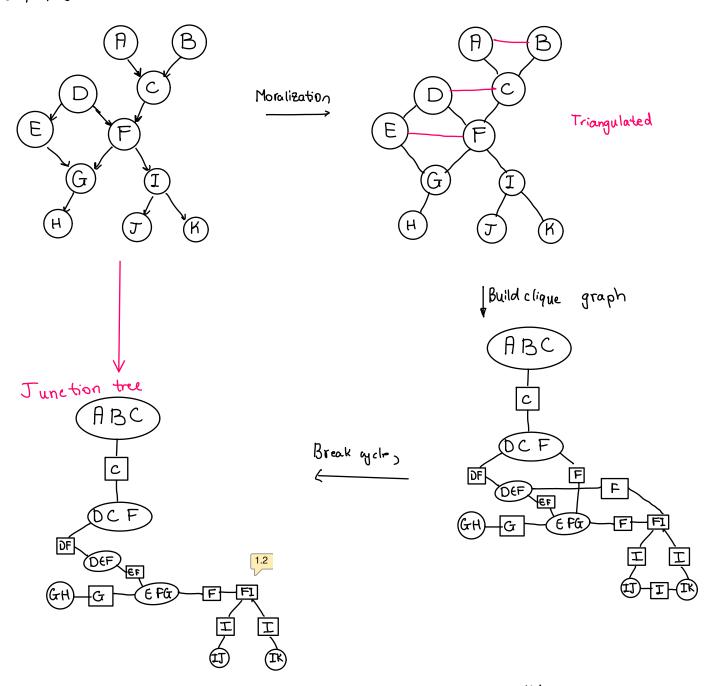
Tuesday, April 16, 2024 15:05

Problem 17 1.1

(B) U=(A,...,K)



(b) P(U) → product of clique potentials divided by product of the separator potentials

 $P(\mathcal{U}) = \frac{\Psi(\mathsf{A.B.C}) \cdot \Psi(\mathsf{D.C.F}) \cdot \Psi(\mathsf{D.E.F}) \cdot \Psi(\mathsf{E.F.G}) \cdot \Psi(\mathsf{G.H}) \cdot \Psi(\mathsf{F.D.\Psi(I.J}) \cdot \Psi(\mathsf{I.L.K})}{\Psi(\mathsf{C}) \cdot \Psi(\mathsf{D.F}) \cdot \Psi(\mathsf{E.F}) \cdot \Psi(\mathsf{G.F}) \cdot \Psi(\mathsf{G.H}) \cdot \Psi(\mathsf{I.D.F})}$ 

PEOPSIEM 18

a) FOR WERD

$$\mu_{\alpha}(x_n) = \sum_{x_{n-1}} \psi_{n-1,n}(x_{n-1},x_n) \mu_{\alpha}(x_{n-1})$$

$$\mu_{b}(x_{n}) = \sum_{x_{n+1}} \psi_{n,n+1}(x_{n},x_{n+1}) \mu_{b}(x_{n+1})$$

μα (xa= k) ... μα (xu= k)

FOR A FIXED & WE WEED K SUH STEPS FOR  $\mu_d(x;=k)$ . FOR Y POSSIBLE & WE WEED  $K^2$  SUH. STEPS.

THE FOLLOWING  $\mu_M(x_{i+1}=k)$  weeds K sum. Steps for a fixed k if all measures  $\mu_M(x_i=k)$  were stocked. For  $\theta$  possible k, we need  $\xi^2$  sum. Steps.

THEREFORE, SINCE WE NEED TO COMPUTE MY (K=k)... My (Ky=k)... My (Ky=k)

3. K2 SUMMATION STEPS

3.1

MB(XH=R) FOR YR WE WILL NEED K2 SUMMATION STEPS.

AND TO SUM PRODUCTS  $\mu_{\alpha}(x_{\mu}=k)\mu_{\beta}(x_{\mu}=k)$  we will need ADDITIONAL K STEPS.

P(xu=1) = 1 4 4a(xu=1) 4p(xu=1)

2 - NEEDS (8K2+K2 + K) = KK2+K STEPS

Ma (Xn=1) & Mp (Xn=1) ARE ASSUMED TO BE STORED & NOT COMPUTED AGAIN.

THEREFORE WE NEED 4K2+K STEPS TO COMPLIE

COMPLEXITY IN "BIG O" NOTATION: O(NK)

c) TO COMPUTE & HORGIPHL DISTRIBUTIONS, WE WELD TO COMPUTE & POSSIBLE HESPOOLS.

FOR  $\forall k$  we will need  $K^2$  STERS.

FOR FIXED &  $\mu_{cl}(x_3=k)$  & nowing  $\forall \mu_{cl}(x_2=k)$  are any computed use used k steps. For  $\forall$  k, we will used  $k^2$  steps

... IT IS APPLIED FOR  $\mu_A(x_a=k)\dots\mu_A(x_5=k)$ . Therefore we werd  $(5-1)\,k^2=4\,k^2$  steps to compute 4 forward missours.

FOR BACKWARD HENDERS WE START FROM  $\mu_B(X_u=R)$  & CONTINUO UNTIL  $\mu_B(X_1=R)$  FOR  $\forall R$  ARE COMPUTED, WHICH ALSO REQUES IN  $\mu_K^2$  SUMMATION STEPS.

TO COMPUTE ONE HARCHIAL PROBABILITY P(X;= E) (FOR FIXED E) WE WELD ADDITIONAL & SUMMATION STEPS (FOR I COMPUTATION)

WE KOWE 5 NODES & RE POWHER NODE SHOPES, THURS TO COMPUTE Y HORGINAL DISTRIBUTIONS WE NEET

COMPRESIDE IN "PIP O" NOTHTION:  $O(N_{K_3})$ 

## Project 7

#### Contents

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## Problem 19

5.1

### Message passing on a chain

## 0 0.1333333 0.5333333 ## 1 0.1111111 0.2222222

## , , Psi23

We will now use R to perform the message passing shown in Figure 1. Suppose that we have the following (conditional) probability distributions:

```
# Conditional probabilities

x1 <- 1/3

x2_1 <- c(4/5, 2/3)

x3_2 <- c(5/7, 1/3)

x4_3 <- c(3/5, 2/5)

x5_4 <- c(1/2, 7/9)
```

Note that these equations fully determine each (conditional) probability distribution, since  $X_i \in \{0, 1\}$  for  $i \in \{1, ..., 5\}$ .

#### (a) Store clique potentials in an R object

```
# Potential matrix
pot <- array(dim = c(2, 2, 4), dimnames = list(c("0", "1"), c("0", "1"),
c("Psi12", "Psi23", "Psi34", "Psi45")))

# Filling it up
pot[, , "Psi12"] <- cbind(1-x2_1, x2_1) * c(1-x1,x1)
pot[, , "Psi23"] <- cbind(1-x3_2, x3_2)
pot[, , "Psi34"] <- cbind(1-x4_3, x4_3)
pot[, , "Psi45"] <- cbind(1-x5_4, x5_4)</pre>
pot

## , , Psi12
##
## 0 1
```

```
##
## 0 0.2857143 0.7142857
## 1 0.6666667 0.3333333
##
## , , Psi34
##
## 0 1
## 0 0.4 0.6
## 1 0.6 0.4
##
## , , Psi45
##
## 0 1
## 0 0.5000000 0.5000000
## 1 0.2222222 0.7777778
```

#### (b) Computing forward messages

```
# Forward message
mu_a <- array(dim = c(5, 2), dimnames = list(c("X1", "X2", "X3", "X4", "X5"), c("0", "1")))
# Initialization
mu_a[1, ] <- 1
# Computation
for (i in 2:5){
   mu_a[i, ] <- mu_a[i-1, ] %*% pot[, , i-1]
}
mu_a</pre>
```

```
## X1 1.0000000 1.0000000
## X2 0.2444444 0.7555556
## X3 0.5735450 0.4264550
## X4 0.4852910 0.5147090
## X5 0.3570253 0.6429747
```

#### (c) Computing backward message

```
# Backward message
mu_b <- array(dim = c(5, 2), dimnames = list(c("X1", "X2", "X3", "X4", "X5"), c("0", "1")))
# Initialization
mu_b[5, ] <- 1
# Computation
for (i in 4:1){
    mu_b[i, ] <- mu_b[i+1, ] %*% pot[, , i]
}
mu_b</pre>
```

```
##
## X1 0.2429159 0.7341564
## X2 0.9312169 1.0687831
## X3 1.0555556 0.9444444
## X4 0.722222 1.2777778
## X5 1.0000000 1.0000000
(d)) Compute the marginal probability distribution for each no
# Marginal prob.
marg \leftarrow array(dim = c(5, 2), dimnames = list(c("X1", "X2", "X3", "X4", "X5"), c("0", "1")))
# Calculate
for (i in 1:5){
  marg[i, ] <- mu_a[i, ] * t(mu_b)[, i]</pre>
marg
              0
## X1 0.2429159 0.7341564
## X2 0.2276308 0.8075250
## X3 0.6054086 0.4027631
## X4 0.3504879 0.6576837
## X5 0.3570253 0.6429747
Normalizing constant Z
Z <- rowSums(marg)</pre>
Z
##
                                                   Х5
          Х1
                    X2
                               ХЗ
                                         Х4
## 0.9770723 1.0351558 1.0081717 1.0081717 1.0000000
Normalizing
marg <- marg/Z</pre>
print(rowSums(marg))
## X1 X2 X3 X4 X5
## 1 1 1 1 1
print(marg)
##
## X1 0.2486161 0.7513839
## X2 0.2199000 0.7801000
## X3 0.6005015 0.3994985
## X4 0.3476471 0.6523529
## X5 0.3570253 0.6429747
library(rmarkdown)
```

render("Gibbs.Rmd", pdf\_document(TRUE), "Indilewitsch\_Toidze\_Houhamdi\_Pudziuvelyte\_Project6.pdf")

# Index of comments

- 1.1 2/2
- 1.2 FI is probably not meant as a separator-node here.
- 2.1 2/3
- 3.1 some symbols here are rather difficult to discern in handwriting
- 4.1 -1: Complexity without storing messages is naively  $O(N * NK^2)$ . Using stored messages it is  $O(NK^2)$ .
- 5.1 4.5/5
- 6.1 -0.5: should be transposed here.  $mu\_b[i,\,] <- mu\_b[i+1,\,] \ensuremath{\,\%^{\star}\%}\ t(pot[,\,,\,i])$