

## MITx: 6.008.1x Computational Probability and Inference

<u>Hel</u>j



- Introduction
- Part 1: Probability and Inference
- Part 2: Inference in Graphical Models

Week 5: Introduction to Part 2 on Inference in Graphical Models

Week 5: Efficiency in Computer Programs

Exercises due Oct 20, 2016 at 02:30 IST

Week 5: Graphical Models

Exercises due Oct 20, 2016 at 02:30 IST

Week 5: Homework 4

<u>Homework due Oct 21, 2016 at 02:30 IST</u>

Week 6: Inference in Graphical Models -Marginalization Part 2: Inference in Graphical Models > Week 6: Inference in Graphical Models - Marginalization > Exercise: The Sum-Product Algorithm - Computational Complexity

## Exercise: The Sum-Product Algorithm - Computational Complexity

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Exercise: The Sum-Product Algorithm - Computational Complexity

7/7 points (graded)

Let's take another look at our running five node example:

Exercises due Oct 27, 2016 at 02:30 IST

**B** 

(A)

<u>Week 6: Special Case -</u> <u>Marginalization in Hidden</u> <u>Markov Models</u>

Exercises due Oct 27, 2016 at 02:30 IST

Week 6: Homework 5

Homework due Oct 27, 2016 at 02:30 IST

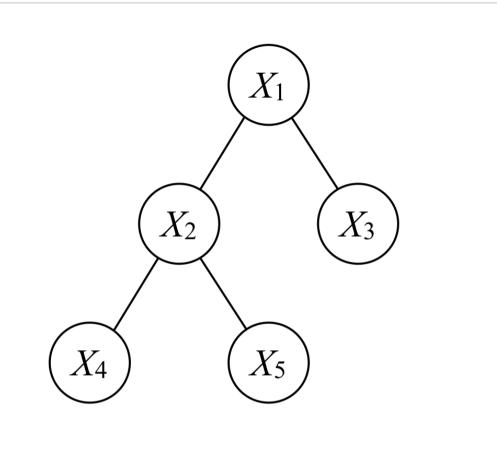
Weeks 6 and 7: Mini-project on Robot Localization

Mini-projects due Nov 03, 2016 at 02:30 IST

Week 7: Inference with Graphical Models - Most Probable Configuration

Exercises due Nov 03, 2016 at 02:30 IST

<u>Week 7: Special Case - MAP</u> <u>Estimation in Hidden Markov</u> Models



When we computed the marginal distribution  $p_{X_1}$  using the sum-product algorithm, note that the ordering in which we pushed summations around matters! We first summed out  $x_5$ . This was a good choice because it only depended on two factors. Consider if instead we tried to sum out  $x_2$  first, which depends on factors  $\phi_2$ ,  $\psi_{12}$ ,  $\psi_{24}$ , and  $\psi_{25}$ :

$$egin{aligned} p_{X_1}(x_1) &\propto \sum_{x_2} \sum_{x_3} \sum_{x_4} \sum_{x_5} igg\{ \phi_1(x_1) \phi_2(x_2) \phi_3(x_3) \phi_4(x_4) \phi_5(x_5) \ & \cdot \psi_{12}(x_1, x_2) \psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\} \ &= \sum_{x_3} \sum_{x_4} \sum_{x_5} igg\{ \phi_1(x_1) \phi_3(x_3) \phi_4(x_4) \phi_5(x_5) \ & \cdot \psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\sum_{x_2} \phi_2(x_2) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{12}(x_1, x_2) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5) igg\}. \ & \underbrace{\psi_{13}(x_1, x_3) \underbrace{\psi_{13}(x_1, x_3) \psi_{13}(x_1, x_3) \psi_{24}(x_2, x_4) \psi_{25}(x_2, x_5)$$

• Suppose that every random variable  $X_i$  takes on exactly k different values. How many operations (additions, multiplications, table lookups) does it take to compute the table  $m_{2\to\{1,4,5\}}$ ? (Note that this table has one entry for every possible value of  $x_1$ ,  $x_4$ , and  $x_5$ .)

Choose the answer with smallest big O bound.

- $\mathcal{O}(k)$
- $\mathcal{O}(k^2)$
- $\mathcal{O}(k^3)$
- ullet  $\mathcal{O}(k^4)$  ullet
- $\mathcal{O}(k^5)$

 ${\mathcal O}(2^k)$ 

• What other summation orders would have been as efficient for computation as the one we chose originally (in computing the marginal for  $X_1$ )?

Select all answers that are as efficient as the one we had chosen in the video/course notes.

(For each of these answers, you can read it as summing out the variable mentioned left-most first and then working rightward.)

- $\quad \square \ \ X_5, X_2, X_3, X_4$
- $\quad \square \ \ X_5, X_3, X_2, X_4$
- $X_4, X_2, X_5, X_3$
- $\quad \ \square \ \ X_4,X_3,X_2,X_5$



Now let's look at the computational complexity (i.e., how many operations it takes) to run the sumproduct algorithm. Let n be the number of nodes in the graph. Assume that every random variable  $X_i$  takes on exactly k different values.

As a reminder, the equation for computing a message  $m_{i \to i}$  is given by

$$m_{i o j}(x_j) = \sum_{x_i} \left[ \phi_i(x_i) \psi_{i,j}(x_i,x_j) \prod_{k\in\mathcal{N}(i) ext{ such that } k
eq j} m_{k o i}(x_i) 
ight]$$

and the equation for computing each node marginal after all messages have been computed is

$$p_{X_i}(x_i) = rac{1}{Z} \underbrace{\phi_i(x_i) \prod_{j \in \mathcal{N}(i)} m_{j 
ightarrow i}(x_i)}_{ ilde{p}_{X_i}(x_i)}.$$

In this part of the exercise, we come up with a worst-case running time of the sum-product algorithm as stated so far.

• When running the sum-product algorithm on a tree, how many operations does it take to compute a message table  $m_{i \to j}$  for a specific i and j (of course i and j are neighbors, and remember that because of how sum-product computes messages, the messages that  $m_{i \to j}$  depends on have already been computed)?

Choose the answer with **smallest** big O bound in terms of k and n (unless one of these doesn't matter).

O (0)	(k)		
O 0	$(k^2)$		
0 O	$(k^3)$		
O 0	$\ell(k^4)$		
O 0	$\ell(nk)$		
• O	$(nk^2)$ $\checkmark$		
0 O	$\ell(nk^3)$		
O 0	$\ell(nk^4)$		
O 0	$\ell(n^2k)$		
0 O	$\ell(n^2k^2)$		
0 0	$\ell(n^2k^3)$		

 $\mathcal{O}(n^2k^4)$ 

• Exactly how many messages do we compute? Please provide your answer in terms of n, the number of nodes in the tree.

In this part, please provide your answer as a mathematical formula (and not as Python code). Use  $^{\land}$  for exponentiation, e.g.,  $x^{\circ}$  denotes  $x^{\circ}$ . Explicitly include multiplication using \*, e.g.  $x^{\circ}$  is xy.

2\*(n-1)  $2 \cdot (n-1)$ 

• Putting together your answers to the previous two parts, how many operations does computing all the messages take?

Choose the answer with **smallest** big O bound in terms of k and n (unless one of these doesn't matter).

 $\circ$   $\mathcal{O}(k)$ 

 $\mathcal{O}(k^2)$ 

 $\mathcal{O}(k^3)$ 

 $\mathcal{O}(k^4)$ 

$\mathcal{O}(nk)$	
$\circ~~\mathcal{O}(nk^2)$	
$\circ~~\mathcal{O}(nk^3)$	
$\mathcal{O}(nk^4)$	
$\mathcal{O}(n^2k)$	
$lacksquare$ $\mathcal{O}(n^2k^2)$ $\checkmark$	
${\mathcal O}(n^2k^3)$	
${\cal O}(n^2k^4)$	
$\mathcal{O}(n^2k^4)$ $\mathcal{O}(n^3k)$	
$\mathcal{O}(n^3k)$	

• After computing all the messages, how many operations does it take to compute the marginal distribution  $p_{X_i}$  for a specific i?

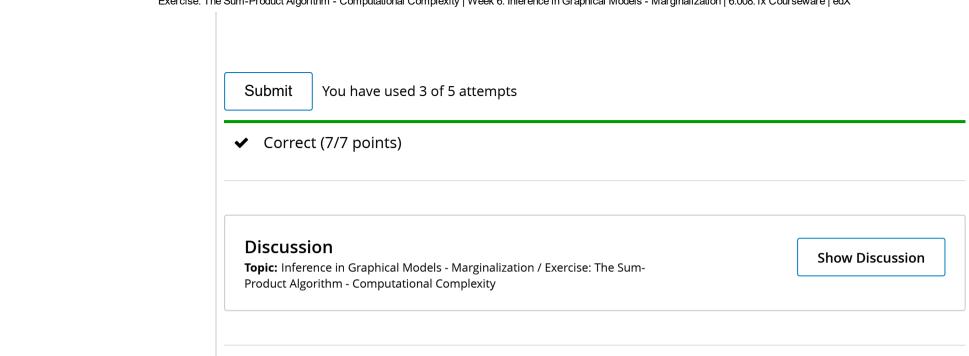
Choose the answer with **smallest** big O bound in terms of k and n (unless one of these doesn't matter).

- $\mathcal{O}(k)$
- $\mathcal{O}(k^2)$
- $\mathcal{O}(k^3)$
- $\mathcal{O}(k^4)$
- ullet  $\mathcal{O}(nk)$
- ${}^{igodot} \, {\cal O}(nk^2)$
- $\mathcal{O}(nk^3)$
- $\mathcal{O}(nk^4)$
- $\mathcal{O}(n^2k)$

$\circ$ $\mathcal{O}(n^2k^2)$
$\circ$ $\mathcal{O}(n^2k^3)$
$\bigcirc \; \mathcal{O}(n^2 k^4)$
How many operations does running the sum-product algorithm take?
Choose the answer with <b>smallest</b> big O bound in terms of $m{k}$ and $m{n}$ (unless one of these doesn't matter).
$\circ$ $\mathcal{O}(k)$
$\circ$ $\mathcal{O}(k^2)$
$\circ$ $\mathcal{O}(k^3)$
${}^{\circ}\;\mathcal{O}(k^4)$
$\circ$ $\mathcal{O}(nk)$
$\circ$ $\mathcal{O}(nk^2)$

${}^{\bigcirc}\;\mathcal{O}(nk^3)$		
$\circ$ $\mathcal{O}(nk^4)$		
$\circ$ $\mathcal{O}(n^2k)$		
$lacksquare \mathcal{O}(n^2k^2)$ $lacksquare$		
$\circ$ $\mathcal{O}(n^2k^3)$		
$\circ$ $\mathcal{O}(n^2k^4)$		
$\circ$ $\mathcal{O}(n^3k)$		
$\odot~\mathcal{O}(n^3k^2)$		
$\mathcal{O}(n^3k^3)$		
$\mathcal{O}(n^3k^4)$		

It turns out that we can improve the running time of the algorithm by some careful bookkeeping!



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