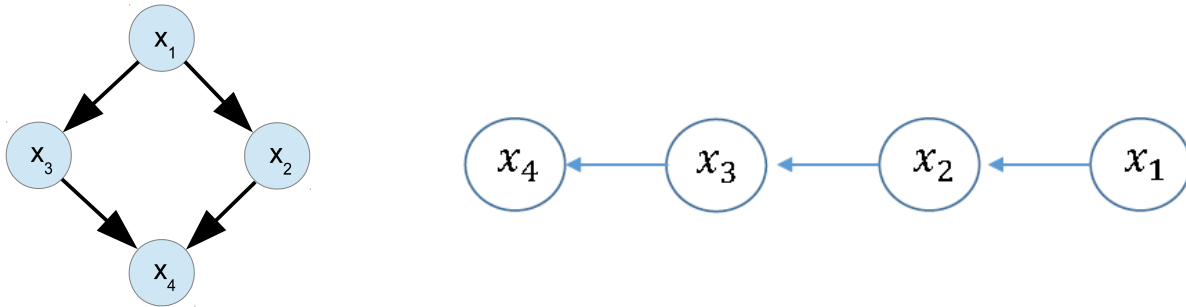


**Exercise 2, Supervised structured learning**

2.1. The joint probability in the variables  $x_1, \dots, x_7$  shall be given as  $p(x_1, x_2, x_3, x_4, x_5, x_6, x_7) = p(x_1)p(x_3)p(x_7)p(x_2|x_1, x_3)p(x_5|x_7, x_2)p(x_4|x_2)p(x_6|x_5, x_4)$ . Draw the directed graphical model for this joint probability!

2.2. Write down the joint probability for the DGM given in the pictures:



2.3. Max marginal inference in chains

A Bus goes from your home to the university where you leave it at the 4th stop. With certain probabilities the bus arrives on time at the stops  $x_i$ . Calculate these probabilities given the directed graphical model (DGM):



Algorithm for marginal computation ("Sum-Product Message Passing")

1. Compute Messages from right to left
- 2.. Read out all marginals

and the following (conditional) probability tables  $p(x_i|x_{i-1})$  ( $x=0$  means late,  $x=1$  on time):

$x_1$	
0	0,29
1	0,71

$x_2 x_1$	$x_1 = 0$	$x_1 = 1$
0	0,39	0,2
1	0,61	0,8

$x_3 x_2$	$x_2 = 0$	$x_2 = 1$
0	0,77	0,51
1	0,23	0,49

$x_4 x_3$	$x_3 = 0$	$x_3 = 1$
0	0,11	0,66
1	0,89	0,34

(equivalently: What is the maximum marginals solution of this probability distribution?)

2.4. Factor graphs

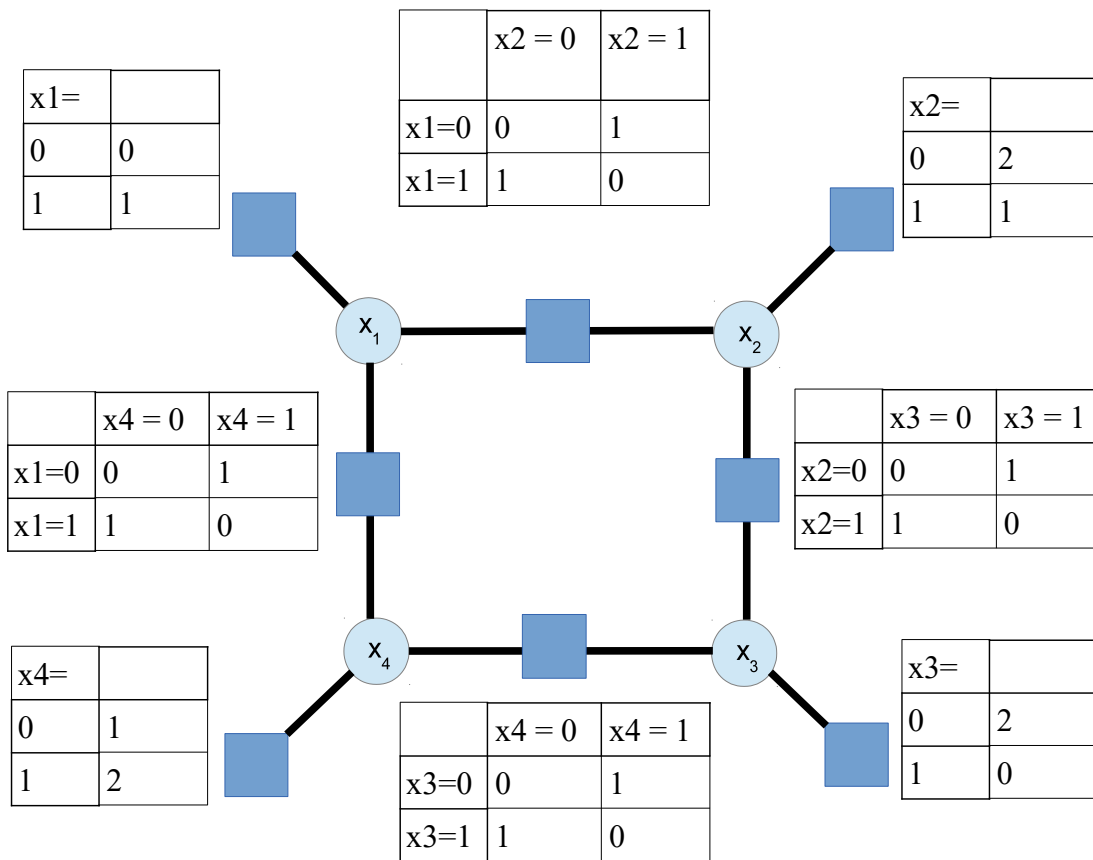
For the chain in 2.2.: Draw the factor graph and write down the formula that corresponds to this factor graph. The conditional probabilities from the directed graphical model are inserted as factors here.

2.5. Let a factor be

$$\psi(x_1, x_2) = \begin{cases} x_1 + x_2 & | 0 \leq x_1, x_2 \leq 2 \\ 0 & | x_1, x_2 < 0; x_1, x_2 > 2 \end{cases}. \text{ There are no more factors.}$$

Which distribution  $p(x_1, x_2)$  follows from that factor?

2.6. Given is the following energy function  
(of a [Gibbs-Boltzmann-Distribution](#)):



Calculate the [MAP](#) solution with [ICM](#).