

Product \rightarrow Summation

$$P_\Phi(x) \propto \prod_k \phi_k(D_k)$$

$$\text{Argmax}_k \prod_k \phi_k(D_k)$$

$$\text{Argmax}_k \sum_k \theta_k(D_k)$$

a ¹	b ¹	8
a ¹	b ²	1
a ²	b ¹	0.5
a ²	b ²	2

$\downarrow \log_2$

a ¹	b ¹	3
a ¹	b ²	0
a ²	b ¹	-1
a ²	b ²	1

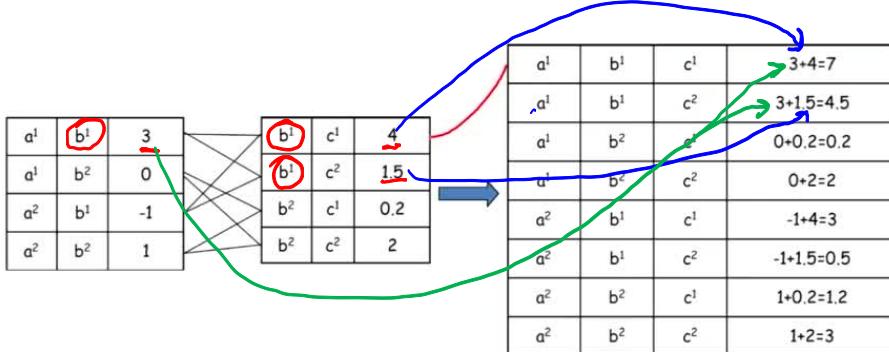


$$\max_D \max_C \max_B \max_A (\theta_1(A, B) + \theta_2(B, C) + \theta_3(C, D) + \theta_4(D, E))$$

$$\max_D \max_C \max_B (\theta_2(B, C) + \theta_3(C, D) + \theta_4(D, E) + \max_A \theta_1(A, B))$$

$\lambda_1(B)$

Factor Summation



Factor Maximization

max-marginalization

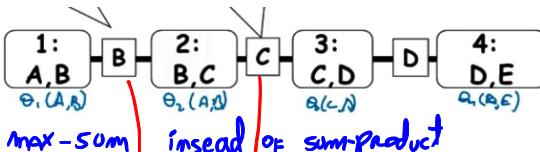
a^1	b^1	c^1	$\rightarrow 0$
a^1	b^1	c^2	4.5
a^1	b^2	c^1	0.2
a^2	b^1	c^1	3
a^2	b^1	c^2	0.5
a^2	b^2	c^1	1.2
a^2	b^2	c^2	3

a^1	c^1	7
a^1	c^2	4.5
a^2	c^1	3
a^2	c^2	3

$$\lambda_4(e) = \max_{a, b, c, d} \theta(a, b, c, d, e)$$

Daphne Koller

Best value that I can get
if we mandate $E=e$
 $= \text{MAX MARGINAL}$



$$\lambda_{1 \rightarrow 2}(B) = \max_A \theta_1, \quad \lambda_{2 \rightarrow 3}(C) = \max_B (\theta_2 + \lambda_{1 \rightarrow 2})$$

Map Assignment

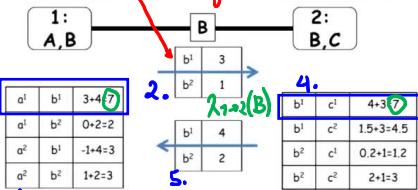
$\theta_1(A, B)$			$\theta_2(B, C)$		
a^1	b^1	3	b^1	c^1	4
a^1	b^2	0	b^1	c^2	1.5
a^2	b^1	-1	b^2	c^1	0.2
a^2	b^2	1	b^2	c^2	2

a^1	b^1	c^1	$3+4 = 7$
a^1	b^1	c^2	$3+1.5 = 4.5$
a^1	b^2	c^1	$0+0.2 = 0.2$
a^1	b^2	c^2	$0+2 = 2$
a^2	b^1	c^1	$-1+4 = 3$
a^2	b^1	c^2	$-1+1.5 = 0.5$
a^2	b^2	c^1	$1+0.2 = 1.2$
a^2	b^2	c^2	$1+2 = 3$

7. 0₁ Simple Example 0₂

a ¹	b ¹	3
a ¹	b ²	0
a ²	b ¹	-1
a ²	b ²	1

b ¹	c ¹	4
b ¹	c ²	1.5
b ²	c ¹	0.2
b ²	c ²	2



Daphne Koller

$$\beta_i(C_i) = \theta_i(C_i) + \sum_k \lambda_{k \rightarrow i} \quad W_i = \{x_1, \dots, x_n\} - C_i$$

All incoming messages

$$\beta_i(C_i) = \max_{W_i} \theta_i(C_i, W_i) \rightarrow \text{max assignment to variable not assigned to the clique}$$

Belief of variable clique

max-marginal

-Calibration: cliques agree on shared variables

$$\max_{C_i \rightarrow S_{i,j}} \beta_i(C_i) = \max_{C_j \rightarrow S_{i,j}} \beta_j(C_j)$$

Assignment

Assignment

a ¹	b ¹	3+4=7
a ¹	b ²	0+2=2
a ²	b ¹	-1+4=3
a ²	b ²	1+2=3

$$b^1=7$$

$$b^2=3$$

b ¹	c ¹	4+3=7
b ¹	c ²	1.5+3=4.5
b ²	c ¹	0.2+1=1.2
b ²	c ²	2+1=3

$$\beta_1 = \theta_1 + \lambda_{2 \rightarrow 1}$$

$$\beta_2 = \theta_2 + \lambda_{1 \rightarrow 2}$$

-The same clique tree algorithm used for sum-product can be used for max-sum

-As in sum-product, convergence is achieved after a single up-down pass

-Result is a max-marginal at each clique C:

-For each Assignment c to C, what is the score of the best completion to c