

Problem 1:

1. Since each color will be associated with a weight  $W_a \in \mathbb{R}$  for each  $a \in \{1, \dots, K\}$ , when  $W_a = a$ ,

the joint probability  $P(x) = \frac{1}{Z} \prod_{i \in V} e^{x_i} \prod_{(i,j) \in E} 1_{x_i \neq x_j}$

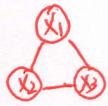
$$\text{with } Z = \sum_{x_v \in \{1, \dots, K\}^{|V|}} \prod_{i \in V} e^{x_i} \prod_{(i,j) \in E} 1_{x_i \neq x_j};$$

since max assignment  $x^* = \arg \max p(x)$ , for the cycle  $G$  with even number of vertices ( $|V| = 2t$ ,  $t \in \mathbb{N}$ ), in order to maximize  $p(x)$ , it's conceivable to assign the color  $(K-1)$  and  $(K)$  repetitively to the adjacent vertices. More specifically, the MAP assignment will be for  $x_i$  ( $i \in \{1, 2, \dots, 2t\}$ ),

- { when  $i = 2l-1$  ( $l \in \{1, 2, \dots, t\}$ ),  $x_i = K-1$ ;
- { and when  $i = 2l$  ( $l \in \{1, 2, \dots, t\}$ ),  $x_i = K$ ;
- or { when  $i = 2l-1$  ( $l \in \{1, 2, \dots, t\}$ ),  $x_i = K$ ;
- { and when  $i = 2l$  ( $l \in \{1, 2, \dots, t\}$ ),  $x_i = K-1$ ;

As a specific example, when  $|V| = 6$ , we can find  $x^*$  as  $= (K-1, K, K-1, K, K-1, K)$  or  $(K, K-1, K, K-1, K, K-1)$ .

2.



for MAP, we need to

$$\max_{q \in Q} \sum_{i \in V} \sum_{x_i} q_i(x_i) \log \phi_i(x_i) + \sum_{(i,j) \in E} \sum_{x_i, x_j} q_{ij}(x_i, x_j) \log \psi_{ij}(x_i, x_j)$$

$$\text{in our example, } \phi_i(x_i) = e^{x_i}, \psi_{ij}(x_i, x_j) = 1_{x_i \neq x_j};$$

$$= \max_{q \in Q} \sum_{i \in V} \sum_{x_i} q_i(x_i) \cdot x_i + \sum_{(i,j) \in E} \sum_{x_i, x_j} q_{ij}(x_i, x_j) \log \psi_{ij}(x_i, x_j)$$

for both IP and LP, the common constraints include:

Set 1: when  $x_i = x_j$ ,  $1_{x_i \neq x_j} = 0$ , and its log will be negative infinity, thus in those cases,

$$q_{ij}(x_i, x_j) = 0;$$

$$q_{12}(x_1=1, x_2=1) = 0, q_{12}(x_1=2, x_2=2) = 0, q_{12}(x_1=3, x_2=3) = 0$$

$$q_{13}(x_1=1, x_3=1) = 0, q_{13}(x_1=2, x_3=2) = 0, q_{13}(x_1=3, x_3=3) = 0$$

$$q_{23}(x_2=1, x_3=1) = 0, q_{23}(x_2=2, x_3=2) = 0, q_{23}(x_2=3, x_3=3) = 0$$

$$\text{Set 2: } q_1(x_1=1) + q_1(x_1=2) + q_1(x_1=3) = 1$$

$$q_2(x_2=1) + q_2(x_2=2) + q_2(x_2=3) = 1$$

$$q_3(x_3=1) + q_3(x_3=2) + q_3(x_3=3) = 1$$

$$\text{Set 3: } \sum_{x_i} q_{ij}(x_i, x_j) = q_i(x_i) \text{ for all } (i, j) \in E, x_i;$$

$$\text{such as: } q_{12}(x_1=1, x_2=1) + q_{12}(x_1=1, x_2=2) + q_{12}(x_1=1, x_2=3) = q_1(x_1=1)$$

$$q_{12}(x_1=2, x_2=1) + q_{12}(x_1=2, x_2=2) + q_{12}(x_1=2, x_2=3) = q_1(x_1=2)$$

Then in addition, for IP:

set 4: all  $q_i(x_i) \in \{0,1\}$ , and  $q_{ij}(x_i, x_j) \in \{0,1\}$

for LP:

set 4: all  $q_i(x_i) \in [0,1]$ , and  $q_{ij}(x_i, x_j) \in [0,1]$

$$\text{Thus: } \max_{q \in Q} \sum_{i \in V} \sum_{x_i} q_i(x_i) \cdot x_i + \sum_{(i,j) \in E} \sum_{x_i, x_j} q_{ij}(x_i, x_j) \cdot \log 1_{x_i \neq x_j}$$

$$= \max_{q \in Q} [2 \cdot q_1(x_1=2) + 1 \cdot q_1(x_1=1) + 3 \cdot q_1(x_1=3) + 1 \cdot q_2(x_2=1) + 2 \cdot q_2(x_2=2) + 3 \cdot q_2(x_2=3) \\ + 1 \cdot q_3(x_3=1) + 2 \cdot q_3(x_3=2) + 3 \cdot q_3(x_3=3)]$$

For IP: since  $x_1, x_2, x_3$  are symmetric, if we assign  $q_1(x_1=3)=1$ , thus giving its largest possible sum of  $1 \cdot q_1(x_1=1) + 2 \cdot q_1(x_1=2) + 3 \cdot q_1(x_1=3)$ .

Then for the constraints we can derive that either  $q_2(x_2=2)=1$ , and  $q_3(x_3=1)=1$ ,  
or  $q_2(x_2=1)=1$ , and  $q_3(x_3=2)=1$ ,

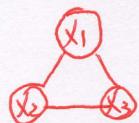
Thus in all cases, the maximal value will be  $3 \cdot 1 + 2 \cdot 1 + 1 \cdot 1 = 6$ ;

For LP: if we subject the  $\max_{q \in Q}$  function to all the constraints, and use linear programming solver (such as [www.comnun.com/cmnn03/cmnn03004](http://www.comnun.com/cmnn03/cmnn03004)).

The solutions are  $q_1(x_1=2) = q_1(x_1=3) = q_2(x_2=2) = q_2(x_2=3) = q_3(x_3=2) = q_3(x_3=3) = 1/2$ ;

Thus the maximal value will be  $(2 \cdot \frac{1}{2} + 3 \cdot \frac{1}{2}) \cdot 3 = 7.5$ ;

Problem 2. Part 1: Here I use a simple example to illustrate the procedures of the MATLAB code; 3



$$\phi_i(x_i) = e^{x_i}, \quad \psi_{ij}(x_i, x_j) = 1_{x_i \neq x_j}; \quad x_i \in \{1, 2, 3\}$$

Step 1: calculate the messages:

$$\text{first assign } M_{1 \rightarrow 2}^0(x_2) = M_{1 \rightarrow 3}^0(x_3) = M_{2 \rightarrow 3}^0(x_3) = M_{2 \rightarrow 1}^0(x_1) = M_{3 \rightarrow 1}^0(x_1) = M_{3 \rightarrow 2}^0(x_2) =$$

$$\text{Then } M_{i \rightarrow j}^t(x_j) = \sum_{x_i} [\phi_i(x_i) \psi_{ij}(x_i, x_j) \prod_{k \in M(t) \setminus j} M_{k \rightarrow i}^{t-1}(x_k)]$$

$$\begin{aligned} \text{for example, } M_{1 \rightarrow 2}^1(x_2=1) &= \sum_{x_1} [\phi_1(x_1) \psi_{12}(x_1, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1)] \\ &= \phi_1(x_1=1) \psi_{12}(x_1=1, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1=1) + \phi_1(x_1=2) \cdot \psi_{12}(x_1=2, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1=2) \\ &\quad + \phi_1(x_1=3) \psi_{12}(x_1=3, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1=3) \\ &= e^2 \cdot 1 + e^3 \cdot 1 \end{aligned}$$

$$\text{similarly, } M_{1 \rightarrow 2}^1(x_2=2) = e^1 \cdot 1 + e^3 \cdot 1$$

$$M_{1 \rightarrow 2}^1(x_2=3) = e^1 \cdot 1 + e^2 \cdot 1$$

In order to avoid overflow, we sum  $\sum_{x_3}^t M_{i \rightarrow j}(x_3)$ , and use  $1 / (\sum_{x_3}^t M_{i \rightarrow j}(x_3))$  as the  $\eta_{i \rightarrow j}^t$ , in above case, after normalization:

$$M_{1 \rightarrow 2}^1(x_2=1) = (e^2 + e^3) / (2e + 2e^2 + 2e^3)$$

$$M_{1 \rightarrow 2}^1(x_2=2) = (e + e^3) / (2e + 2e^2 + 2e^3)$$

$$M_{1 \rightarrow 2}^1(x_2=3) = (e + e^2) / (2e + 2e^2 + 2e^3)$$

We will repeat this for all other five messages, normalize each, and finish this iteration;

We will repeat this whole iteration its times, and hopefully it will converge;

For this specific example, after converging (11 iterations), the values are like:

$$M_{1 \rightarrow 2}(x_2=1) = 0.4293, \quad M_{1 \rightarrow 2}(x_2=2) = 0.3443, \quad M_{1 \rightarrow 2}(x_2=3) = 0.2254$$

$$M_{1 \rightarrow 3}(x_3=1) = 0.4291, \quad M_{1 \rightarrow 3}(x_3=2) = 0.3445, \quad M_{1 \rightarrow 3}(x_3=3) = 0.2265$$

$$M_{2 \rightarrow 1}(x_1=1) = 0.4291, \quad M_{2 \rightarrow 1}(x_1=2) = 0.3445, \quad M_{2 \rightarrow 1}(x_1=3) = 0.2265$$

$$M_{2 \rightarrow 3}(x_3=1) = 0.4293, \quad M_{2 \rightarrow 3}(x_3=2) = 0.3443, \quad M_{2 \rightarrow 3}(x_3=3) = 0.2255$$

$$M_{3 \rightarrow 1}(x_1=1) = 0.4293, \quad M_{3 \rightarrow 1}(x_1=2) = 0.3443, \quad M_{3 \rightarrow 1}(x_1=3) = 0.2254$$

$$M_{3 \rightarrow 2}(x_2=1) = 0.4294, \quad M_{3 \rightarrow 2}(x_2=2) = 0.3449, \quad M_{3 \rightarrow 2}(x_2=3) = 0.2247$$

Step 2: calculate the beliefs:

$$\text{for } b_i(x_i) = \phi_i(x_i) \prod_{k \in N(i)} M_{k \rightarrow i}(x_k)$$

$$\text{for example, } b_1(x_1=1) = \phi_1(x_1=1) \cdot M_{2 \rightarrow 1}(x_1=1) \cdot M_{3 \rightarrow 1}(x_1=1) \\ = e^1 \cdot 0.4291 \cdot 0.4293$$

$$b_1(x_1=2) = \phi_1(x_1=2) \cdot M_{2 \rightarrow 1}(x_1=2) \cdot M_{3 \rightarrow 1}(x_1=2) \\ = e^2 \cdot 0.3445 \cdot 0.3453$$

$$b_1(x_1=3) = \phi_1(x_1=3) \cdot M_{2 \rightarrow 1}(x_1=3) \cdot M_{3 \rightarrow 1}(x_1=3) \\ = e^3 \cdot 0.2265 \cdot 0.2254$$

since  $\sum_{x_i} b_i(x_i) = 1$ , we normalize by  $\gamma_i = 1 / (b_1(x_1=1) + b_1(x_1=2) + b_1(x_1=3))$   
and we get  $b_1(x_1=1) = 0.21$ ,  $b_1(x_1=2) = 0.36$ ,  $b_1(x_1=3) = 0.43$ ;

we repeat this for  $b_2(x_2)$  and  $b_3(x_3)$ ;

$$\text{and } b_2(x_2=1) = 0.21, b_2(x_2=2) = 0.36, b_2(x_2=3) = 0.43; \\ b_3(x_3=1) = 0.21, b_3(x_3=2) = 0.36, b_3(x_3=3) = 0.43;$$

$$\text{for } b_{i3}(x_i, x_3) = \phi_i(x_i) \phi_j(x_3) \psi_{ij}(x_i, x_3) \cdot \prod_{k \in N(i) \setminus j} M_{k \rightarrow i}(x_k) \cdot \prod_{k \in N(j) \setminus i} M_{k \rightarrow j}(x_k)$$

$$\text{for example, } b_{12}(x_1=1, x_2=1) = \phi_1(x_1=1) \cdot \phi_2(x_2=1) \psi_{12}(x_1=1, x_2=1) \cdot M_{3 \rightarrow 1}(x_1=1) \cdot M_{3 \rightarrow 2}(x_2=1) = 0$$

$$b_{12}(x_1=1, x_2=2) = e^1 \cdot e^2 \cdot M_{3 \rightarrow 1}(x_1=1) \cdot M_{3 \rightarrow 2}(x_2=2)$$

$$b_{12}(x_1=1, x_2=3) = e^1 \cdot e^3 \cdot M_{3 \rightarrow 1}(x_1=1) \cdot M_{3 \rightarrow 2}(x_2=3)$$

$$b_{12}(x_1=2, x_2=1) = e^2 \cdot e \cdot M_{3 \rightarrow 1}(x_1=2) \cdot M_{3 \rightarrow 2}(x_2=1)$$

$$b_{12}(x_1=2, x_2=2) = 0$$

$$b_{12}(x_1=2, x_2=3) = e^2 \cdot e^3 \cdot M_{3 \rightarrow 1}(x_1=2) \cdot M_{3 \rightarrow 2}(x_2=3)$$

$$b_{12}(x_1=3, x_2=1) = e^3 \cdot e \cdot M_{3 \rightarrow 1}(x_1=3) \cdot M_{3 \rightarrow 2}(x_2=1)$$

$$b_{12}(x_1=3, x_2=2) = e^3 \cdot e^2 \cdot M_{3 \rightarrow 1}(x_1=3) \cdot M_{3 \rightarrow 2}(x_2=2)$$

$$b_{12}(x_1=3, x_2=3) = 0$$

since  $\sum_{x_1 x_2} b_{i3}(x_i, x_3) = 1$ , we normalize  $b_{12} = 1 / \sum_{x_1 x_2} b_{12}(x_1, x_2)$ ;

$$\text{and we get: } b_{12}(x_1=1, x_2=1) = 0, b_{12}(x_1=1, x_2=2) = 0.0723, b_{12}(x_1=1, x_2=3) = 0.1330$$

$$b_{12}(x_1=2, x_2=1) = 0.0722, b_{12}(x_1=2, x_2=2) = 0, b_{12}(x_1=2, x_2=3) = 0.2908$$

$$b_{12}(x_1=3, x_2=1) = 0.1335, b_{12}(x_1=3, x_2=2) = 0.2923, b_{12}(x_1=3, x_2=3) = 0$$

We repeat this for  $b_{13}(X_1, X_3)$  and  $b_{23}(X_2, X_3)$ ,

and we get :  $b_{13}(X_1=1, X_3=1) = 0$ ,  $b_{13}(X_1=1, X_3=2) = 0.0750$ ,  $b_{13}(X_1=1, X_3=3) = 0.1331$

$b_{13}(X_1=2, X_3=1) = 0.0749$ ,  $b_{13}(X_1=2, X_3=2) = 0$ ,  $b_{13}(X_1=2, X_3=3) = 0.2906$  ;

$b_{13}(X_1=3, X_3=1) = 0.1338$ ,  $b_{13}(X_1=3, X_3=2) = 0.2926$ ,  $b_{13}(X_1=3, X_3=3) = 0$  ;

$b_{23}(X_2=1, X_3=1) = 0$ ,  $b_{23}(X_2=1, X_3=2) = 0.0749$ ,  $b_{23}(X_2=1, X_3=3) = 0.1338$

$b_{23}(X_2=2, X_3=1) = 0.0750$ ,  $b_{23}(X_2=2, X_3=2) = 0$ ,  $b_{23}(X_2=2, X_3=3) = 0.2926$  ;

$b_{23}(X_2=3, X_3=1) = 0.1331$ ,  $b_{23}(X_2=3, X_3=2) = 0.2905$ ,  $b_{23}(X_2=3, X_3=3) = 0$  ;

All match the results produced by the MATLAB code .

Step 3: calculate the partition function  $Z$ ;

$$\text{since } p(x) = \frac{1}{Z} \prod_{i \in V} \phi_i(x_i) \cdot \prod_{(i,j) \in E} \mathbb{1}_{x_i \neq x_j}$$

$$\begin{aligned} D(q||p) &= -H(q) + \log Z - \sum_x q(x) \left[ \sum_{i \in V} \log \phi_i(x_i) + \sum_{(i,j) \in E} \log \mathbb{1}_{x_i \neq x_j} \right] \\ &= -H(q) + \log Z - \sum_{i \in V} \sum_{x_i} q_i(x_i) \log \phi_i(x_i) - \sum_{(i,j) \in E} \sum_{x_i x_j} q_{ij}(x_i, x_j) \cdot \log \mathbb{1}_{x_i \neq x_j} \end{aligned}$$

Thus when local maximization is achieved:

$$\log Z = H(q) + \sum_{i \in V} \sum_{x_i} b_i(x_i) \log \phi_i(x_i) + \sum_{(i,j) \in E} \sum_{x_i x_j} b_{ij}(x_i, x_j) \cdot \log \mathbb{1}_{x_i \neq x_j}$$

$$H(q) = -\sum_{i \in V} \sum_{x_i} b_i(x_i) \cdot \log b_i(x_i) - \sum_{(i,j) \in E} \sum_{x_i x_j} b_{ij}(x_i, x_j) \cdot \log \frac{b_{ij}(x_i, x_j)}{b_i(x_i) \cdot b_j(x_j)}$$

In above case, the  $Z$  is calculated as 4493.2;

Notes for the mathtable:

For a graph  $G: (V, E)$ , it has  $N$  nodes, each node can take  $K$  value;

Step 1: create the messages. The final matrix is named as  $C$ , and it's  $N^2 \times K$  dimension, as

$k=1 \quad k=2 \quad \dots \quad k=K$

$$m_{i \rightarrow j} \rightsquigarrow \begin{matrix} 11 \\ 12 \\ \vdots \\ 1N \\ \vdots \\ N1 \\ \vdots \\ NN \end{matrix}$$

;

Step 2: create the singleton, named as  $B$ , and it's  $N \times K$  dimension, as

$k=1 \quad k=2 \quad \dots \quad k=K$

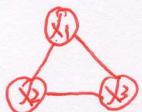
$$b_i(x_i) \rightsquigarrow \begin{matrix} 1 \\ 2 \\ \vdots \\ N \end{matrix}$$

and pairwise, named as  $E$ , and it's  $M(N-1)/2 \times K^2$  dimension, as

$k, k=1, 1 \quad k, k=1, 2 \quad \dots \quad k, k=K, K$

$$b_{i2}(x_i, x_2) \rightsquigarrow \begin{matrix} 12 \\ 13 \\ \vdots \\ 1N \\ 23 \\ \vdots \\ 2N \\ \vdots \\ (M-1)N \end{matrix}$$

Problem 2. Part 2: using the same example.



$$\phi_i(x_i) = e^{x_i}, \quad \psi_{i3}(x_i, x_3) = 1_{x_i \neq x_3}; \quad x_i \in \{1, 2, 3\}$$

Step 1: calculate the messages:

$$\text{initially } M_{1 \rightarrow 2}^0(x_2) = M_{1 \rightarrow 3}^0(x_3) = M_{2 \rightarrow 3}^0(x_3) = M_{2 \rightarrow 1}^0(x_1) = M_{3 \rightarrow 1}^0(x_1) = M_{3 \rightarrow 2}^0(x_2) = 1;$$

$$\text{Then } M_{i \rightarrow j}^t(x_j) = \max_{x_i} [\phi_i(x_i) \psi_{ij}(x_i, x_j) \prod_{k \in N(i) \setminus j} M_{k \rightarrow i}^{t-1}(x_k)]$$

$$\text{for example, } M_{1 \rightarrow 2}^1(x_2=1) = \max_{x_1} [\phi_1(x_1) \psi_{12}(x_1, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1)]$$

$$= \max_{x_1} [\phi_1(x_1=1) \cdot \psi_{12}(x_1=1, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1=1), \phi_1(x_1=2) \cdot \psi_{12}(x_1=2, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1=2), \\ \phi_1(x_1=3) \psi_{12}(x_1=3, x_2=1) \cdot M_{3 \rightarrow 1}^0(x_1=3)]$$

$$= \max_{x_1} (0, e^2 \cdot 1, e^3 \cdot 1) = e^3, \text{ record } x_1=3;$$

$$\text{similarly, } M_{1 \rightarrow 2}^1(x_2=2) = \max_{x_1} (e \cdot 1, 0, e^3 \cdot 1) = e^3, \text{ record } x_1=3;$$

$$M_{1 \rightarrow 2}^1(x_2=3) = \max_{x_1} (e \cdot 1, e^2 \cdot 1, 0) = e^2, \text{ record } x_1=2;$$

In order to avoid overflow, we sum  $\sum_{x_3} M_{i \rightarrow j}^t(x_3)$ , and use  $1 / (\sum_{x_3} M_{i \rightarrow j}^t(x_3))$

as the  $\eta_{i \rightarrow j}^t$ , in above case, after normalization:

$$M_{1 \rightarrow 2}^1(x_2=1) = e^3 / (2e^3 + e^2); \text{ when } x_1=3$$

$$M_{1 \rightarrow 2}^1(x_2=2) = e^3 / (2e^3 + e^2); \text{ when } x_1=3$$

$$M_{1 \rightarrow 2}^1(x_2=3) = e^2 / (2e^3 + e^2); \text{ when } x_1=2.$$

We will use this to calculate all other 5 messages, normalize each, and finish this iteration

(note: also need to record the maximum assignment, such as above  $x_1=3$  for  $M_{1 \rightarrow 2}^1(x_2=1)$ ).

We will then repeat this whole iteration 5 times, and each time update the assignment matrix as well.

For this specific example, after 11 iterations, the values are like:

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$$M_{1 \rightarrow 2}(x_2=1) = 0.4223, M_{1 \rightarrow 2}(x_2=2) = 0.4223, M_{1 \rightarrow 2}(x_2=3) = 0.1554;$$
$$x_1=3 \quad x_1=3 \quad x_1=2$$

$$M_{1 \rightarrow 3}(x_3=1) = 0.3333, M_{1 \rightarrow 3}(x_3=2) = 0.3333, M_{1 \rightarrow 3}(x_3=3) = 0.3333;$$
$$x_1=2 \quad x_1=3 \quad x_1=2$$

$$M_{2 \rightarrow 1}(x_1=1) = 0.3333, M_{2 \rightarrow 1}(x_1=2) = 0.3333, M_{2 \rightarrow 1}(x_1=3) = 0.3333;$$
$$x_2=2 \quad x_2=3 \quad x_2=2$$

$$M_{2 \rightarrow 3}(x_3=1) = 0.3333, M_{2 \rightarrow 3}(x_3=2) = 0.3333, M_{2 \rightarrow 3}(x_3=3) = 0.3333;$$
$$x_2=2 \quad x_2=3 \quad x_2=2$$

$$M_{3 \rightarrow 1}(x_1=1) = 0.4223, M_{3 \rightarrow 1}(x_1=2) = 0.4223, M_{3 \rightarrow 1}(x_1=3) = 0.1554;$$
$$x_3=3 \quad x_3=3 \quad x_3=2$$

$$M_{3 \rightarrow 2}(x_2=1) = 0.4223, M_{3 \rightarrow 2}(x_2=2) = 0.4223, M_{3 \rightarrow 2}(x_2=3) = 0.1554;$$
$$x_3=3 \quad x_3=3 \quad x_3=2$$

Note for Matlab code maxprod.m :

Step 1: Using Max-product to calculate the message matrix, which is named as C, and it's

$N^2 \times K$  dimension, as:

$k=1 \quad k=2 \quad \dots \quad k=K$

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M1 = [11; 12; ...; 1N; 21; ...; N1; ...; NN];
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