

# HMWK2

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Probabilistic Graphical Models

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**Exercise 1.1**

The implied factorization for any  $p \in \mathcal{L}(G)$  is straightforward from the definition :

$$p(t, Z, x, y) = p(t|Z)p(Z|x, y)p(x)p(y)$$

**Exercise 1.2**

(a) Assuming that  $Z$  is binary variable, we can without loss of generality suppose  $Z(\Omega) = \{0, 1\}$ . Let  $\pi = \mathbb{P}(Z = 0)$ ,  $\forall x, y \in X(\Omega) \times Y(\Omega)$

$$\begin{aligned} p(x, y) &= p(x, y|Z = 0)\pi + p(x, y|Z = 1)(1 - \pi) \\ &= p(x|Z = 0)p(y|Z = 0)\pi + p(x|Z = 1)p(y|Z = 1)(1 - \pi) \end{aligned} \quad (X \perp\!\!\!\perp Y \mid Z)$$

Furthermore,

$$\begin{aligned} p(x, y) &= p(x)p(y) \\ &= (p(x|Z = 0)\pi + p(x|Z = 1)(1 - \pi)) (p(y|Z = 0)\pi + p(y|Z = 1)(1 - \pi)) \end{aligned} \quad (X \perp\!\!\!\perp Y)$$

Hence, if we note  $p_0 = p(\cdot|Z = 0)$  and  $p_1 = p(\cdot|Z = 1)$  :

$$\begin{aligned} p_0(x)p_0(y)\pi + p_1(x)p_1(y)(1 - \pi) &= (p_0(x)\pi + p_1(x)(1 - \pi)) (p_0(y)\pi + p_1(y)(1 - \pi)) \\ \Rightarrow p_0(x)p_0(y)\pi(1 - \pi) + p_1(x)p_1(y)\pi(1 - \pi) - (p_0(x)p_1(y) + p_1(x)p_0(y))\pi(1 - \pi) &= 0 \\ \Rightarrow p_0(x)(p_0(y) - p_1(y)) + p_1(x)(p_1(y) - p_0(y)) &= 0 \\ \Rightarrow (p_0(x) - p_1(x))(p_0(y) - p_1(y)) &= 0 \end{aligned}$$

Thus,  $p(x|Z = 0) = p(x|Z = 1)$  or  $p(y|Z = 0) = p(y|Z = 1)$

<b>Exercise 2.1</b>
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<b>Exercise 2.2</b>
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Exercise 3.a
Exercise 3.b (derivation)
Exercise 3.c (estimator covariance)
Exercise 3.d

**K-Means**

**EM isotropic**

**EM General**