

# Probabilistic Graphical Models

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1. Consider a joint distribution over a black and white  $3 \times 3$  image with pixels  $x_1, \dots, x_9$  as shown below:

$x_1$	$x_4$	$x_7$
$x_2$	$x_5$	$x_8$
$x_3$	$x_6$	$x_9$

Note that  $\mathbf{x}$  can take on one of  $2^9$  values in  $\{0, 1\}^9$ . Without further assumption, we would need to estimate  $2^9$  parameters to estimate  $p(\mathbf{x})$ .

- a) Recall that by repeated application of the product rule of probability, a joint distribution  $p(\mathbf{x})$  can be written as:

$$p(\mathbf{x}) = p(x_1) \prod_{i=2}^n p(x_i | x_1, \dots, x_{i-1}) \quad (1)$$

Write the expression for the joint distribution  $p(\mathbf{x})$ ,  $\mathbf{x} = [X_1, X_2, \dots, X_9]^T$  using the factorization in expression (1).

- b) Is the factorization unique?
- c) Assume that each pixel, conditioned on the values of the four pixels immediately to the left/right/above/below, is independent of all the other pixels in the image. This *conditional independence* assumption implies, for example

$$p(x_5 | x_1, x_2, x_3, x_4, x_6, x_7, x_8, x_9) = p(x_5 | x_2, x_4, x_6, x_8).$$

Assume that for a pixel on the edge of the image, for example,  $x_3$ ,

$$p(x_3 | x_1, x_2, x_4, x_5, x_6, x_7, x_8, x_9) = p(x_3 | x_2, x_6).$$

Simplify your expression from (a) by incorporating the conditional independence assumption.

- d) Draw a directed graphical model (i.e, a DAG) that matches the decomposition and conditional independence structure found above.
- e) How many parameters do you need to estimate to estimate  $p(\boldsymbol{x})$  with the conditional independence assumption?