

## DIRECTED GRAPH DISTRIBUTIONS

## Why

We want to visualize the probabilistic relations between components of outcomes in probabilistic models over large (e.g., product) outcome sets.<sup>1</sup>

## **Definition**

Let  $A = \prod_i A_i$ . For  $x \in A$  and  $S \subset \{1, ..., n\}$ , denote the subvector of x indexed (in order) by S by  $x_S$ .<sup>2</sup>

A distribution  $p:A\to [0,1]$  factors according to a directed graph on  $\{0,1\}$  with parent function pa :  $\{1,\ldots,n\}\to \mathcal{P}(\{1,\ldots,n\})$  if

$$p(x) = \prod_{\mathrm{pa}_i = \varnothing} g_i(x_i) \prod_{\mathrm{pa}_i \neq \varnothing} g_i(x_i, x_{\mathrm{pa}_i}),$$

where  $g_i$  is a distribution for all i which  $pa_i = \emptyset$  and  $g_i(\cdot, \xi)$  is a distribution for all  $\xi \in \prod_{j \in pa_i} A_j$ , i for which  $pa_i \neq \emptyset$ .

**Proposition 1.** p so defined is a distribution, and the  $g_i$  are the marginals and conditionals.<sup>3</sup>

<sup>&</sup>lt;sup>1</sup>Future editions will modify and expand. The title of the sheet may change, since another interpretation for the words "directed graph distribution" is a distribution on directed graphs.

<sup>&</sup>lt;sup>2</sup>Future editions will rework this treatment, perhaps combining it with the sheet Index Matrices, which will possibly be split up.

<sup>&</sup>lt;sup>3</sup>Future editions will be precise and give an account.

## **Examples**

Consider a rooted tree distribution (see Rooted Tree Distributions), or a memory chain (see Memory Chains), or a hidden memory chain (see Hidden Memory Chains).<sup>4</sup>

<sup>&</sup>lt;sup>4</sup>Future editions will expand.

