

## Why

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## Definition

Let X and Y be sets. A hidden memory chain (hidden markov chain, hidden memory model, hidden markov model<sup>2</sup>, HMM) with hiddens (or latents) Y and observations Y of length n is a joint distribution  $p: X^n \times Y^n \to [0,1]$  satisfying

$$p(x,y) = f(x)g(y_1, x_1) \prod_{i=1}^{n} h(x_i, x_{i-1})g(y_i, x_i),$$

where  $f: X \to [0, 1]$  is a distribution, and g and h are functions satisfying  $g(\cdot, \xi)$  and  $h(\cdot, \xi)$  are distributions on Y and X, respectively, for all  $\xi \in X$ .

**Proposition 1.** p so defined is a distribution. The function f is the distribution  $p_1$ . For all i = 1, ..., n,  $p_{n+1|i} \equiv gg$ . For all i = 2, ..., n,  $p_{i|i-1} \equiv h$ .<sup>3</sup>

Clearly,  $p_{1,...,n}$  is a memory chain (see Memory Chains). For this reason, we continue to refer to h as the *conditional distribution*. We continute to refer to f as the *initial distribution*. We refer to h as the *observation distribution*.

The word "hidden" refers to the situation in which we observe outcomes y, and we hypothesize that they were "generated by" unobserved outcomes x.

<sup>&</sup>lt;sup>1</sup>Future editions will include.

<sup>&</sup>lt;sup>2</sup>This term is universal. We avoid it because of the Bourbaki project's policy on naming. The skeptical reader will note (as in Memory Chains) that our term and this term have the same initials.

<sup>&</sup>lt;sup>3</sup>Future editions will define everything needed in this proposition in the proposition statement, as ooposed to saying "p so defined".

