

# ECE685D HW5

Zanwen Fu

November 2025

## 1 Solution to 2.1

The joint distribution of a Gaussian–Bernoulli RBM is

$$p(v, h) = \frac{1}{Z} \exp \left( \sum_i \sum_j W_{ij} h_j \frac{v_i}{\sigma_i} - \sum_i \frac{(v_i - b_i)^2}{2\sigma_i^2} + \sum_j \alpha_j h_j \right),$$

derived from the energy

$$E(v, h; \theta) = - \left( \sum_i \sum_j W_{ij} h_j \frac{v_i}{\sigma_i} - \sum_i \frac{(v_i - b_i)^2}{2\sigma_i^2} + \sum_j \alpha_j h_j \right).$$

### (a) Conditional distribution $p(h_j = 1 | v)$

Starting from

$$p(h | v) \propto p(v, h),$$

we keep only the terms depending on  $h$ :

$$p(h | v) \propto \exp \left( \sum_j h_j \left( \alpha_j + \sum_i W_{ij} \frac{v_i}{\sigma_i} \right) \right).$$

Because  $h_j \in \{0, 1\}$  and hidden units factorize,

$$p(h_j = 1 | v) = \frac{\exp(\alpha_j + \sum_i W_{ij} \frac{v_i}{\sigma_i})}{1 + \exp(\alpha_j + \sum_i W_{ij} \frac{v_i}{\sigma_i})}.$$

Thus,

$$p(h_j = 1 | v) = \sigma \left( \alpha_j + \sum_i W_{ij} \frac{v_i}{\sigma_i} \right).$$

### (b) Conditional distribution $p(v_i | h)$

Using Bayes' rule,

$$p(v_i | h) \propto p(v_i, h),$$

and keeping only terms involving  $v_i$ ,

$$p(v_i | h) \propto \exp\left(\sum_j W_{ij} h_j \frac{v_i}{\sigma_i} - \frac{(v_i - b_i)^2}{2\sigma_i^2}\right).$$

Let

$$A_i = \sum_j W_{ij} h_j / \sigma_i.$$

Then the exponent becomes

$$\sum_j W_{ij} h_j \frac{v_i}{\sigma_i} - \frac{(v_i - b_i)^2}{2\sigma_i^2} = -\frac{1}{2\sigma_i^2} [(v_i - b_i)^2 - 2\sigma_i^2 A_i v_i].$$

Completing the square yields

$$p(v_i | h) \propto \exp\left(-\frac{(v_i - \mu_i)^2}{2\sigma_i^2}\right), \quad \mu_i = b_i + \sigma_i \sum_j W_{ij} h_j.$$

Therefore,

$$p(v_i | h) = \mathcal{N}\left(v_i; b_i + \sigma_i \sum_j W_{ij} h_j, \sigma_i^2\right).$$

**Integral form (optional, as allowed in the question):**

$$p(v_i = x | h) = \frac{1}{C(h)} \exp\left(\sum_j W_{ij} h_j \frac{x}{\sigma_i} - \frac{(x - b_i)^2}{2\sigma_i^2}\right),$$

where  $C(h)$  is the normalization constant obtained by integrating over  $x$ .