The map posterior factors into a map prior and a measurement likelihood (c.f., Equation (9.11)):

$$(9.21) \quad \log p(m \mid z_{1:t}, x_{1:t}) = \text{const.} + \log p(z_{1:t} \mid x_{1:t}, m) + \log p(m)$$

The log-likelihood $\log p(z_{1:t} \mid x_{1:t}, m)$ decomposes into a sum of individual measurement log-likelihoods:

(9.22)
$$\log p(z_{1:t} \mid x_{1:t}, m) = \sum \log p(z_t \mid x_t, m)$$

Further, the log-prior also decomposes. To see, we note that the prior probability of any map m is given by the following product:

(9.23)
$$p(m) = \prod_{i} p(\mathbf{m})^{\mathbf{m}_{i}} (1 - p(\mathbf{m}))^{1 - \mathbf{m}_{i}}$$

 $= (1 - p(\mathbf{m}))^{N} \prod_{i} p(\mathbf{m})^{\mathbf{m}_{i}} (1 - p(\mathbf{m}))^{-\mathbf{m}_{i}}$
 $= \eta \prod_{i} p(\mathbf{m})^{\mathbf{m}_{i}} (1 - p(\mathbf{m}))^{-\mathbf{m}_{i}}$

Here $p(\mathbf{m})$ is the prior probability of occupancy (e.g., $p(\mathbf{m}) = 0.5$), and N is the number of grid cells in the map. The expression $(1 - p(\mathbf{m}))^N$ is simply a constant, which is replaced by our generic symbol η as usual.

This implies for the log version of the prior:

(9.24)
$$\log p(m) = \text{const.} + \sum_{i} \mathbf{m}_{i} \log p(\mathbf{m}) - \mathbf{m}_{i} \log(1 - p(\mathbf{m}))$$

$$= \text{const.} + \sum_{i} \mathbf{m}_{i} \log \frac{p(\mathbf{m})}{1 - p(\mathbf{m})}$$

$$= \text{const.} + \sum_{i} \mathbf{m}_{i} l_{0}$$

The constant l_0 is adopted from (9.7). The term $N \log(1 - p(\mathbf{m}_i))$ is obviously independent of the map. Hence it suffices to optimize the remaining expression and the data log-likelihood:

(9.25)
$$m^* = \operatorname{argmax}_{m} \sum_{t} \log p(z_t \mid x_t, m) + l_0 \sum_{i} \mathbf{m}_i$$

A hill-climbing algorithm for maximizing this log-probability is provided in Table 9.3. This algorithm starts with the all-free map (line 2). It "flips" the occupancy value of a grid cell when such a flip increases the likelihood of the data (lines 4-6). For this algorithm it is essential that the prior of occupancy