## Restricted Boltzmann machine

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The divergence was calculated as follows: in each epoch, each possible pattern was fed to the Boltzmann machine. The dynamics were then run for 100 iterations, and the frequencies at which the different possible patterns occured were counted. This was then used as an approximation for the model distribution  $P_B$  for each epoch, and the Kullback-Leibler divergence was calculated using  $d = \sum_{\mu} P_D(\mu) \log(P_D(\mu)/P_B(i_{\mu}))$ , where  $P_D(\mu) = \frac{1}{14}$  for all data set patterns  $\mu$ , and  $i_{\mu}$  is the index of the data set pattern  $\mu$  in the set of all possible patterns. If  $P_B(\mu) = 0$ , I set  $d = \infty$ . For M = 2, 4, 8, the divergence was infinity for (almost) all epochs, so the plots are not shown. For M = 16, the divergence is shown in figure 1a. The first 10 produced patterns after feeding the first column are shown in figures 1b, 1c, 1d, 1e.

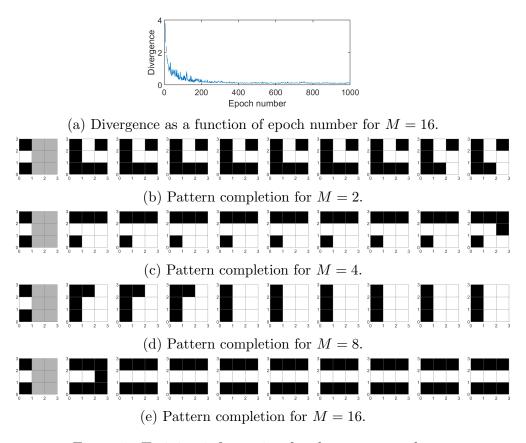


Figure 1: Training information for the two networks.

As expected, the model works well for M=16, and not at all for M=2,4. Interestingly, the model converges to a data set pattern for M=8, despite the divergence being  $\infty$ , but it is not the correct one. This probably means that the model has learned a few of the data set patterns, but not all of them.