## Appendix B

## Synthesised Textures

## B.1 Textures synthesised with various neighbourhoods

When modelling with MRFs it is not always apparent what neighbourhood size should be used. Certainly there is very little in the literature about how to choose an appropriate neighbourhood size given a certain texture. The most widely accepted technique is to test various neighbourhood sizes until the desired one is found. The test usually involves finding out how well the model performs in the desired application. In this case we wish to know how well the model performs at texture synthesis.

The following Figs. B.1–B.166 demonstrate how well the nonparametric MRF model is able to synthesise various texture types. All textures were synthesised using our multiscale texture synthesis algorithm, as outlined in Fig. 7.4. In each case the training textures were  $128 \times 128$  pixels from which  $256 \times 256$  synthetic versions were produced. The lack of phase discontinuity in the larger synthesised images suggest that the texture characteristics must have been capture by the model.

The first set of training textures, Figs. B.1–B.72, are from the Brodatz album [28]. The figures show how the synthetic texture varies as the neighbourhood size increases from nearest neighbour to a  $9 \times 9$  neighbourhood. For each neighbourhood size two synthesis results are presented, one for the Gibbs sampling method, and one for the ICM sampling method. This is also done for the second set of training textures, Figs. B.73–B.166. This second set is from VisTex [203]. However, in this case the neighbourhood size is only tested from the nearest neighbour up to the  $7 \times 7$  neighbourhood.

When viewing these results, take the time to consider the consequences of choosing a particular neighbourhood size for the classification algorithm. If the neighbourhood size is too small, the model will be too general and the will not produce a reliable likelihood function for the particular texture class. If the neighbourhood size is too large, the model will be too specific and therefore not produce an adequate likelihood for textures of a similar class. When this synthesis experiment was first undertaken, we were looking to see if it was possible to capture all the characteristics of a texture so it could be realistically reproduced over a large area. We found, as the results indicate, that the larger the neighbourhood, the more realistic our reproductions. However these reproductions incurred more repetitive structure as the neighbourhood size increased. This is an indication that as the neighbourhood size increases, the model becomes more "specific" and therefore less adequate for classification. The ideal neighbourhood size for classification will produce a texture that is representative of the training texture, but still stochastic in nature.

Before making a decision on which neighbourhood size to employ for classification, consider the significance of the difference between the Gibbs and ICM sampling results. From the results it is visually obvious that the ICM sampling scheme performs marginally better than the Gibbs sampling scheme. This can be explained, as in Section 7.7, that although the modes of the LCPDF are well estimated, estimating the full distribution of the LCPDF as a sum of Gaussian distributions can produce inherent noise into the estimate. However, it is how well the whole distribution of the LCPDF is modelled that will determine how "ideal" the model is for open-ended classification. Therefore, although it is the Gibbs sampling scheme that performs the worst, it is the Gibbs sampling scheme that will determine how "ideal" the model is.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.1: Brodatz texture D001a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



(h)  $9 \times 9$  neighbourhood

(i)  $9 \times 9$  neighbourhood

Figure B.2: Brodatz texture D002a: (a) original  $128 \times 128$  image; (b-i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.3: Brodatz texture D002b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.4: Brodatz texture D003a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.5: Brodatz texture D003b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.6: Brodatz texture D004a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.7: Brodatz texture D004b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.8: Brodatz texture D004c: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



(a) Training texture



(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.9: Brodatz texture D005a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.10: Brodatz texture D005b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.11: Brodatz texture D005c: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.12: Brodatz texture D006a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5\times 5$  neighbourhood



(g)  $7 \times 7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7\times7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.13: Brodatz texture D007a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.14: Brodatz texture D007b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.15: Brodatz texture D009a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.16: Brodatz texture D009b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.17: Brodatz texture D009c: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



(g)  $7\times7$  neighbourhood

(h)  $9\times9$  neighbourhood

(i)  $9 \times 9$  neighbourhood

Figure B.18: Brodatz texture D010a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.19: Brodatz texture D011a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.20: Brodatz texture D011b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.21: Brodatz texture D012a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.22: Brodatz texture D012b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.23: Brodatz texture D012c: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



(a) Training texture



(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.24: Brodatz texture D013a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.25: Brodatz texture D014a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.26: Brodatz texture D015a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.27: Brodatz texture D015b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

![](_page_30_Figure_1.jpeg)

Figure B.28: Brodatz texture D015c: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

![](_page_31_Figure_1.jpeg)

Figure B.29: Brodatz texture D017a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

![](_page_32_Figure_1.jpeg)

Figure B.30: Brodatz texture D017b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

![](_page_33_Figure_1.jpeg)

Figure B.31: Brodatz texture D017c: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

![](_page_34_Figure_1.jpeg)

Figure B.32: Brodatz texture D017d: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

![](_page_35_Picture_1.jpeg)

![](_page_35_Picture_3.jpeg)

(d)  $5 \times 5$  neighbourhood

![](_page_35_Picture_5.jpeg)

(g)  $7\times7$  neighbourhood

![](_page_35_Picture_7.jpeg)

(b) Nearest neighbour

![](_page_35_Picture_9.jpeg)

(e)  $5 \times 5$  neighbourhood

![](_page_35_Picture_11.jpeg)

(h)  $9 \times 9$  neighbourhood

![](_page_35_Picture_13.jpeg)

(c) Nearest neighbour

![](_page_35_Picture_15.jpeg)

(f)  $7 \times 7$  neighbourhood

![](_page_35_Picture_17.jpeg)

(i)  $9 \times 9$  neighbourhood

Figure B.33: Brodatz texture D018a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.


- (g)  $7\times7$  neighbourhood
- (h)  $9 \times 9$  neighbourhood
- (i)  $9 \times 9$  neighbourhood

Figure B.34: Brodatz texture D019a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.35: Brodatz texture D024a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.36: Brodatz texture D024b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7 \times 7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9\times9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.37: Brodatz texture D026a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.38: Brodatz texture D028a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.39: Brodatz texture D029a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood;



Figure B.40: Brodatz texture D029b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.41: Brodatz texture D031a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



(g)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.42: Brodatz texture D032a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7 \times 7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9\times9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.43: Brodatz texture D033a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood;



- (g)  $7\times7$  neighbourhood
- (h)  $9\times9$  neighbourhood
- (i)  $9 \times 9$  neighbourhood

Figure B.44: Brodatz texture D037a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (i) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.45: Brodatz texture D038a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (i) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.46: Brodatz texture D038b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.47: Brodatz texture D041a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.48: Brodatz texture D051a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.49: Brodatz texture D051b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.50: Brodatz texture D052a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.51: Brodatz texture D054a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (i) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.52: Brodatz texture D055a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.53: Brodatz texture D057a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7 \times 7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9\times9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.54: Brodatz texture D061a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.







(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9\times9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.55: Brodatz texture D062a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.56: Brodatz texture D063a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.57: Brodatz texture D068a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.58: Brodatz texture D068b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.59: Brodatz texture D069a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.60: Brodatz texture D070a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (i) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.61: Brodatz texture D071a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.62: Brodatz texture D071b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.63: Brodatz texture D072a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.64: Brodatz texture D074a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.65: Brodatz texture D077a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.66: Brodatz texture D084a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.67: Brodatz texture D086a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9\times9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.68: Brodatz texture D089a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.





(d)  $5 \times 5$  neighbourhood



(g)  $7 \times 7$  neighbourhood



(b) Nearest neighbour



(e)  $5\times 5$  neighbourhood



(h)  $9\times9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.69: Brodatz texture D090a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.


(a) Training texture



(d)  $5 \times 5$  neighbourhood



(g)  $7\times7$  neighbourhood



(b) Nearest neighbour



(e)  $5 \times 5$  neighbourhood



(h)  $9 \times 9$  neighbourhood



(c) Nearest neighbour



(f)  $7 \times 7$  neighbourhood



(i)  $9 \times 9$  neighbourhood

Figure B.70: Brodatz texture D091a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.71: Brodatz texture D092a: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.



Figure B.72: Brodatz texture D092b: (a) original  $128 \times 128$  image; (b–i) synthesised  $256 \times 256$  images; (b) using Gibbs sampling and nearest neighbour neighbourhood; (c) using ICM sampling and nearest neighbour neighbourhood; (d) using Gibbs sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (e) using ICM sampling and  $5 \times 5$  neighbourhood; (f) using Gibbs sampling and  $7 \times 7$  neighbourhood; (g) using ICM sampling and  $7 \times 7$  neighbourhood; (h) using Gibbs sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood; (i) using ICM sampling and  $9 \times 9$  neighbourhood.

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