

Report for Programming Assignment-4

CS7015 - Deep Learning

April 26, 2019

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1 Question 1

Figures 1 to 4 shows the plots of the learned representations (plotted in a 2-D space using t-SNE) for three different values of number of latent variables (n) - 50, 100, 200, 500 respectively. We use $k = 1$ in all four cases.

The different classes are:

- Class 0: T-shirt/top
- Class 1: Trouser/pants
- Class 2: Pullover shirt
- Class 3: Dress
- Class 4: Coat
- Class 5: Sandal
- Class 6: Shirt
- Class 7: Sneaker
- Class 8: Bag
- Class 9: Ankle boot

As the number of latent variables is increased, the variance within the cluster corresponding to each class decreases and the clusters become more condensed (this can clearly be observed in the transition from $n = 50$ to $n = 100$). Thus,

more information about the classes is captured, resulting in lower intra-cluster variance and higher inter-cluster variance. There is an overlap between classes with similar-looking objects (for example, "shirt" and "t-shirt/top", "dress" and "coat", "sandal" and "sneaker"). While some of these overlaps are resolved as the number of latent variables is increased (for example, "sandal" and "sneaker" get separated as n is increased), some classes are too close together for proper differentiation even with large n ("shirt" and "t-shirt/top", "dress" and "coat" still overlap in the case of $n = 500$).

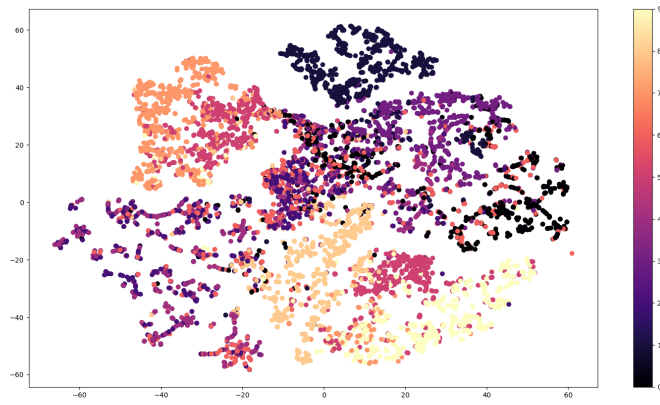


Figure 1: $n = 50$

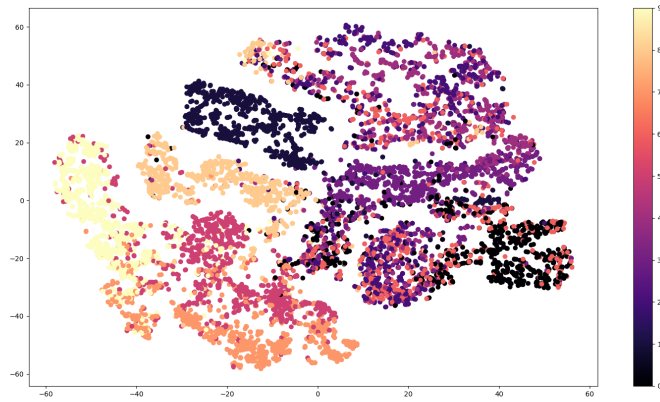


Figure 2: $n = 100$

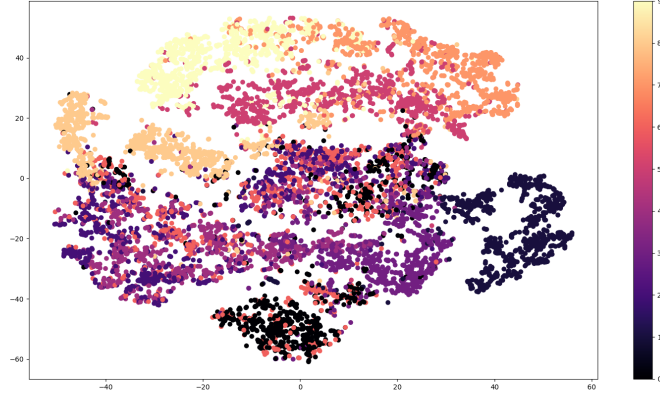


Figure 3: $n = 200$

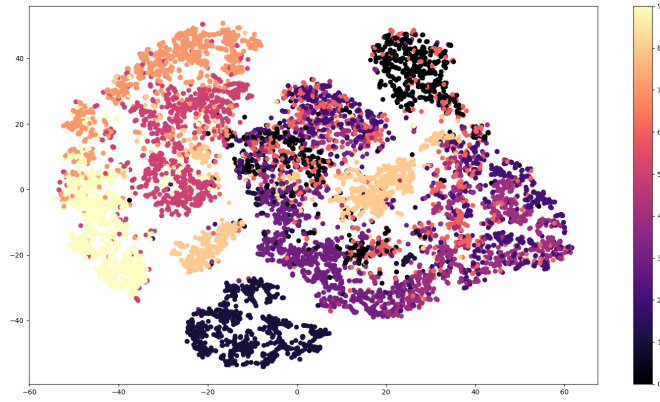


Figure 4: $n = 500$

2 Question 2

- Figure 5 shows the average reconstruction loss (Sum of square error between V_d and \tilde{V}) vs iterations for various values of k .
- We plot the samples generated by the Gibbs chain (after intervals of 100 iterations) for different values of k . Figure 6 shows the input samples and Figures 7 to 11 shows the samples from the Gibbs chains.

We use $n = 512$ in all five cases.

We find that as k increases, \tilde{V} (which was initialized to V_d) undergoes an increasing number of transformations, and is hence more different compared to V_d . Hence the reconstruction loss increases with k . We can also qualitatively see that the reconstructed images are better for smaller values of k .

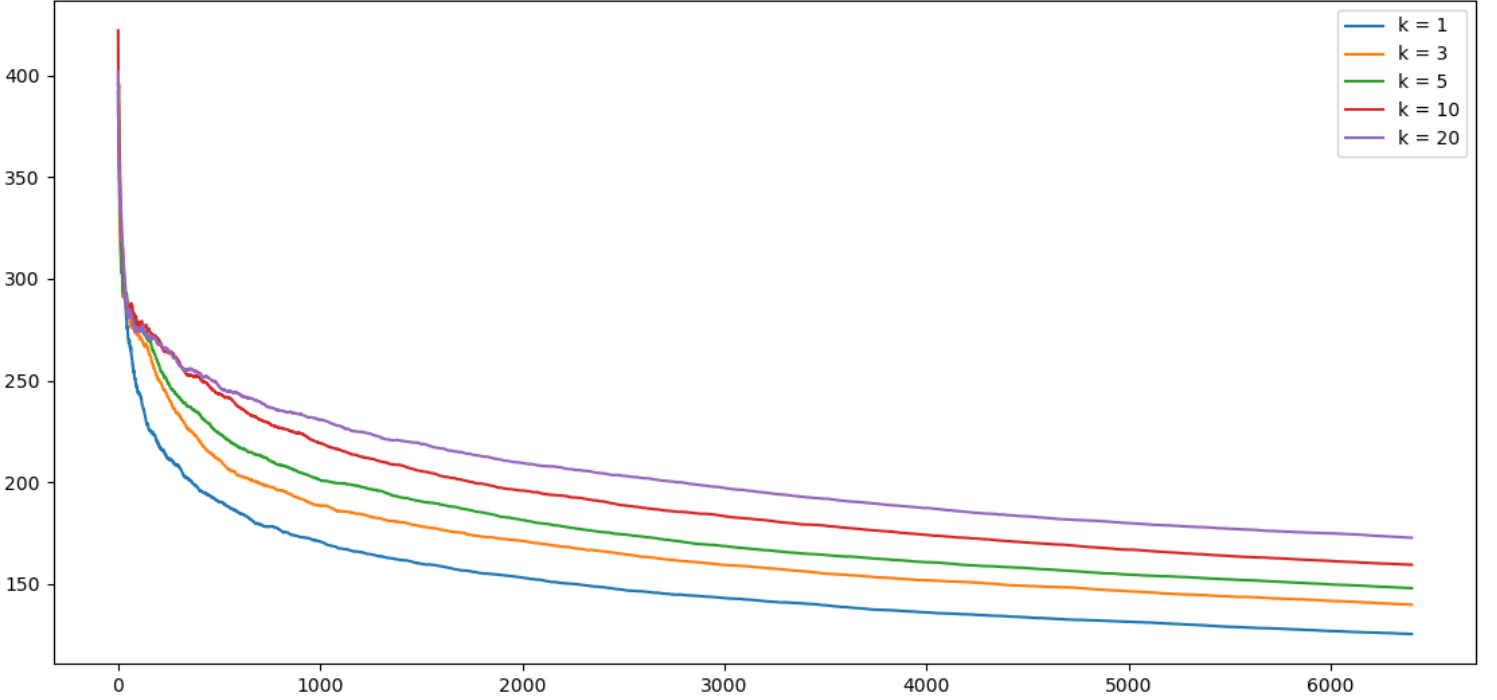


Figure 5: Average Reconstruction vs iterations

3 Question 3

Figure 6 shows the input samples used after intervals of 100 iterations. Figures 7 shows the corresponding samples from the Gibbs chain for $k = 1$, $n = 512$. We find that the quality of reconstructed images get better with the number of iterations. Qualitatively, they appear more and more similar to the corresponding sample input at that iteration.

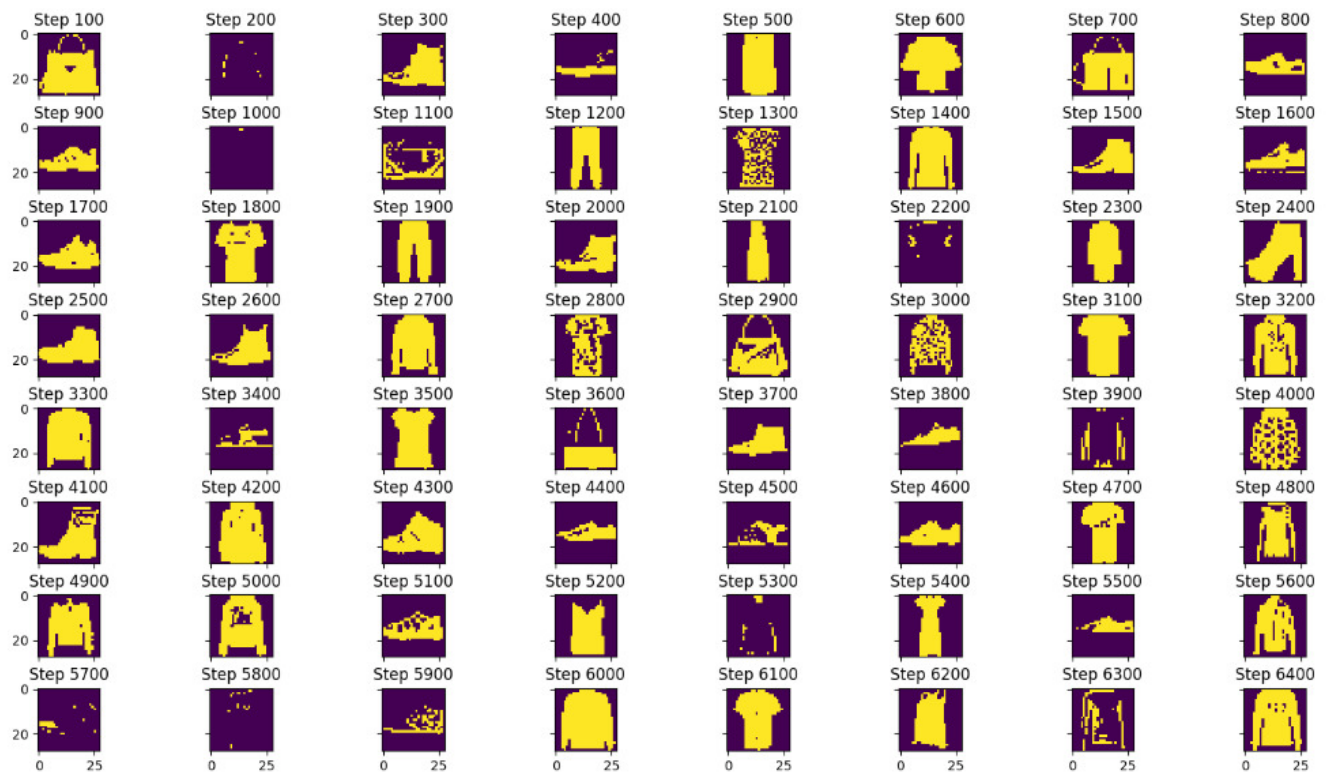


Figure 6: Input samples

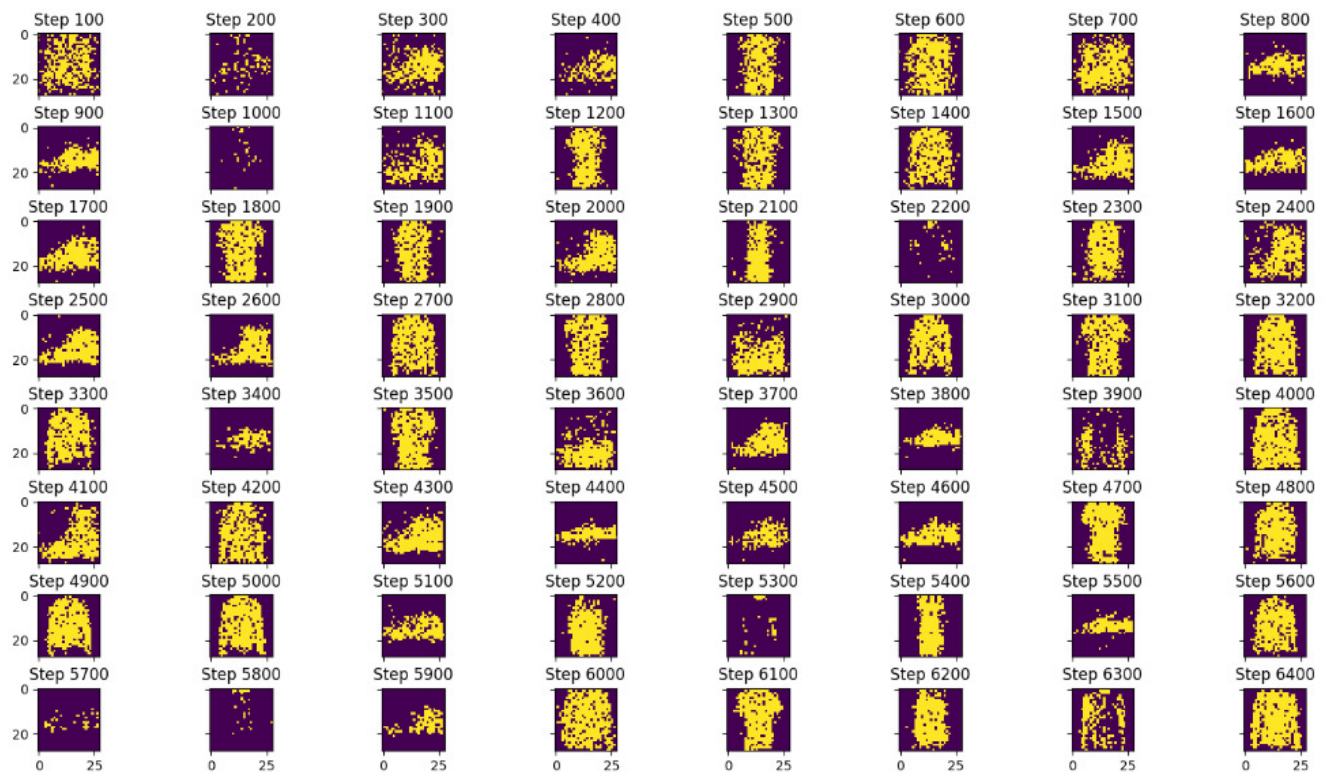


Figure 7: $k = 1$



Figure 8: $k = 3$

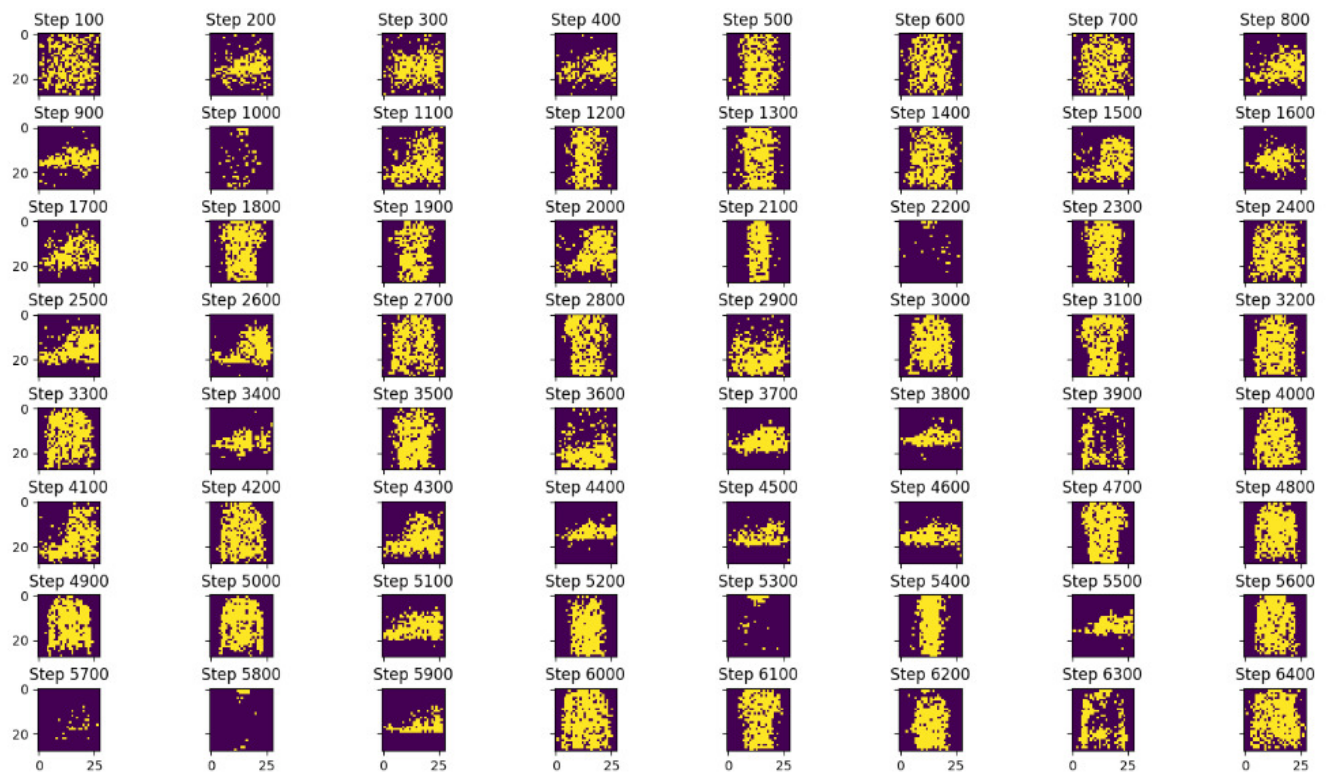


Figure 9: $k = 5$

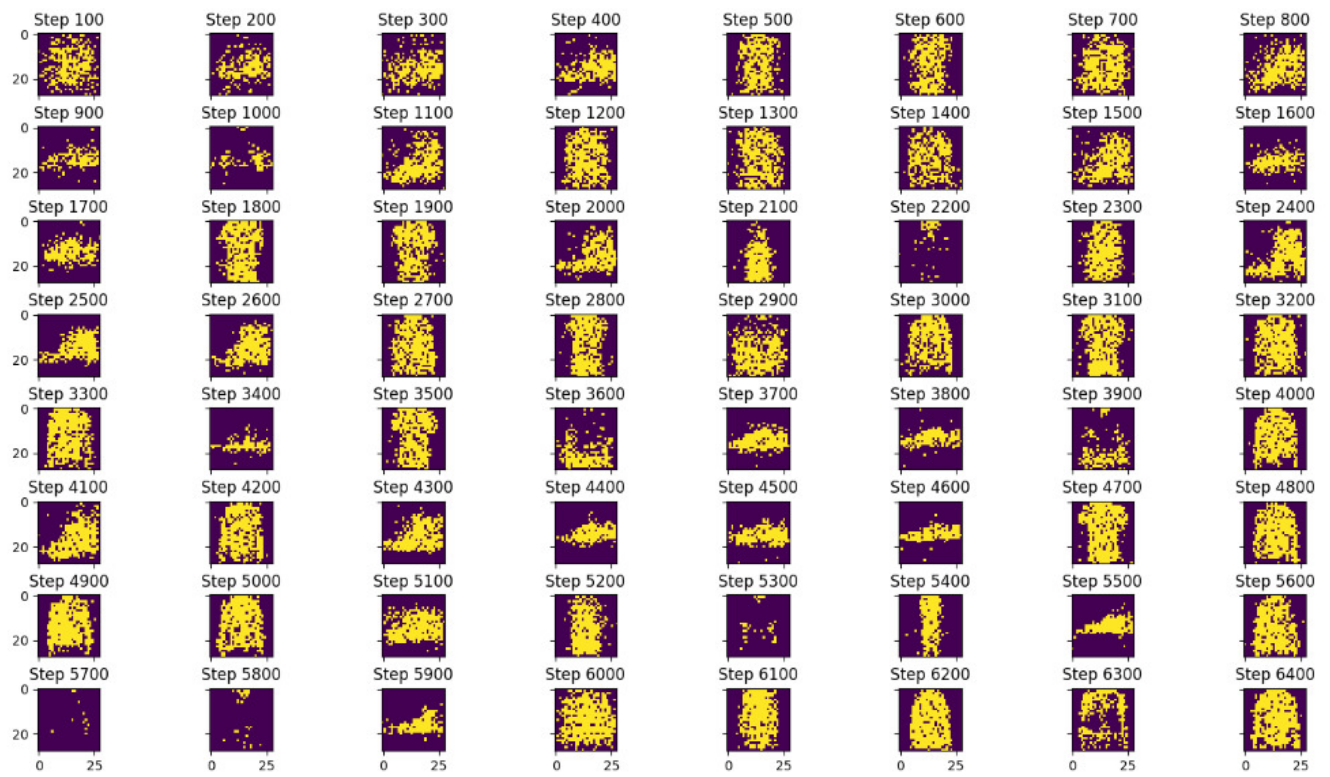


Figure 10: $k = 10$

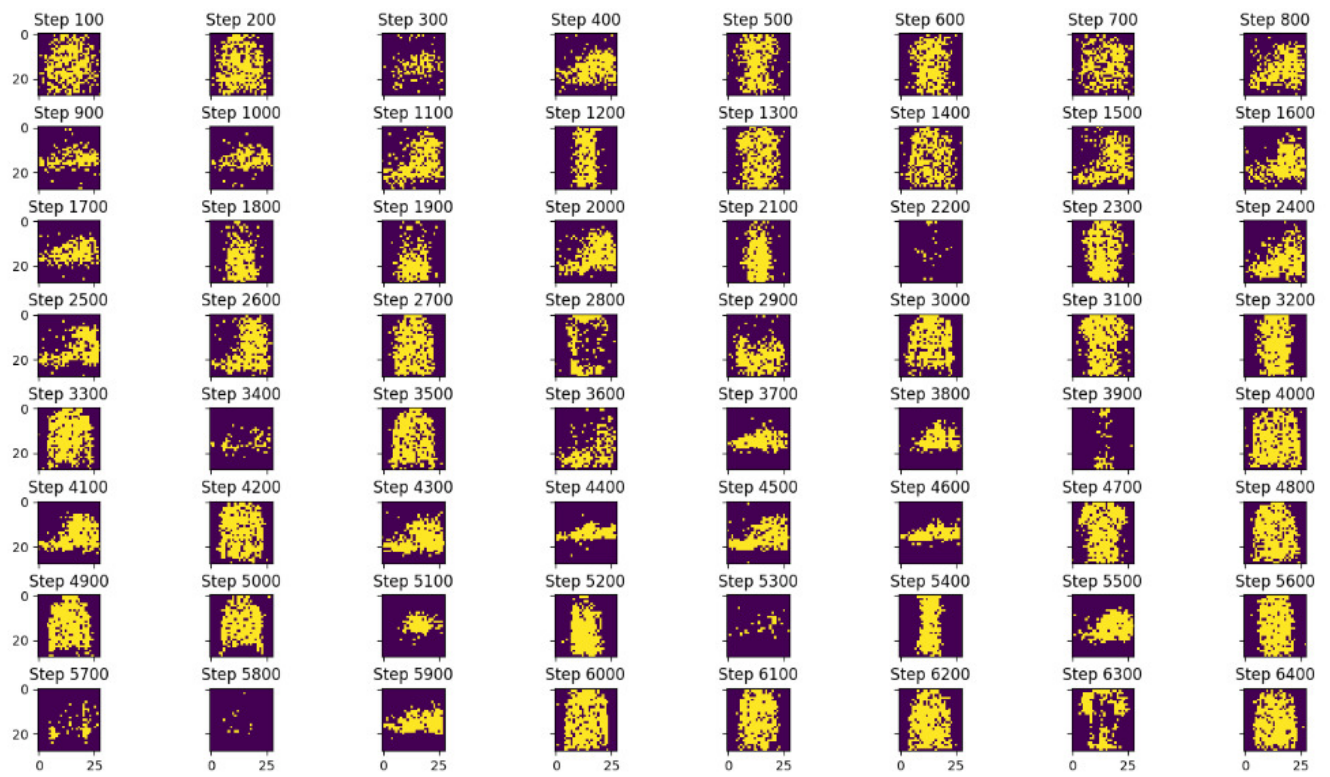


Figure 11: $k = 20$