

## HW3

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### Q1. Text Classification using CNNs

1.

$$Y_{n,f} = b_f + \sum_c X_{n,c} *_{filt} W_{f,c}^{conv}$$

2.

The size of  $Y_{n,f}$  is  $(1, 1, H - H' + 1)$ .

3.

The size of the output of the pooling layer is  $(N, F, 1)$

4.

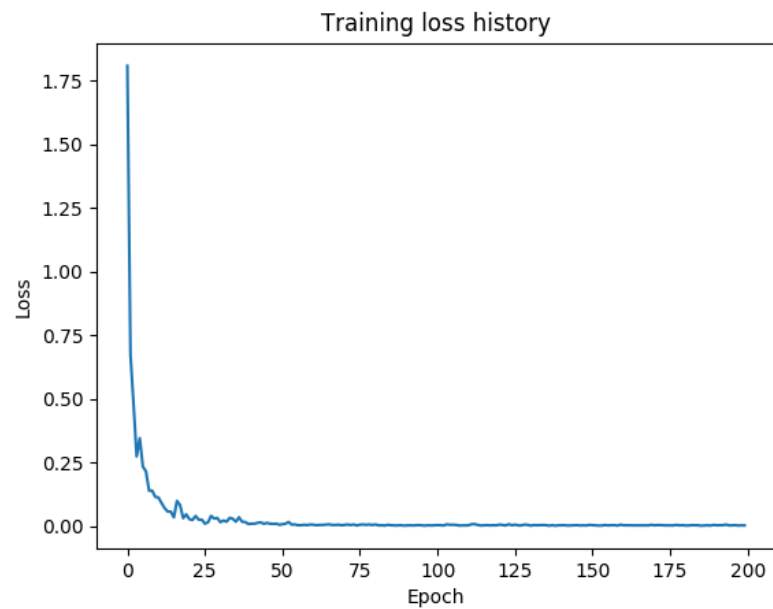
Use GloVe pre-trained embedding with “update\_embedding” option opened during training, the accuracies under different settings are:

	Global average- pooling	Global max- pooling
<b>Kernel size: 5</b>	93.92%	95.87%
<b>Kernel size: 7</b>	91.82%	95.28%

## Q2. Siamese Networks for Learning Embeddings

2.

1) Learning curve:



2) Results on training set:



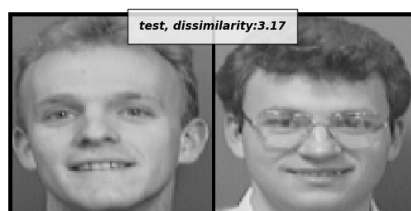
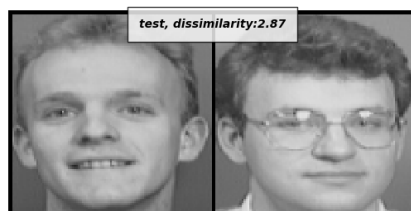


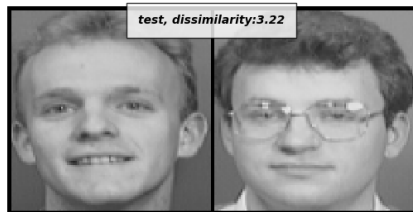
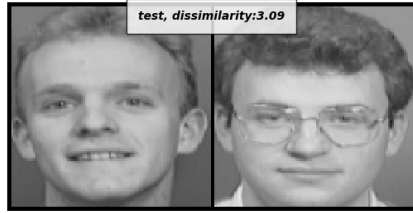
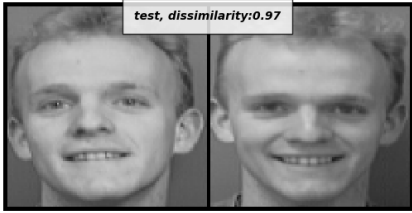




### 3) Results on Testset







### Q3. Conditional Variational Autoencoders

1.

$$\begin{aligned}
 \log p_\theta(x|y) &= \int q_\phi(z|x, y) \log p_\theta(x|y) dz \\
 &= \int q_\phi(z|x, y) \log \frac{p_\theta(x, z|y)}{p_\theta(z|x, y)} dz \\
 &= \int q_\phi(z|x, y) \log \frac{p_\theta(x|y, z)p_\theta(z|y)}{p_\theta(z|x, y)} dz \\
 &= \int q_\phi(z|x, y) (-\log p_\theta(z|x, y) + \log p_\theta(x|y, z) + \log p_\theta(z|y)) dz \\
 &= \int q_\phi(z|x, y) (-\log p_\theta(z|x, y) + \log q_\phi(z|x, y) + \log p_\theta(x|y, z) - \\
 &\quad \log q_\phi(z|x, y) + \log p_\theta(z|y)) dz \\
 &= KL(q_\phi(z|x, y) \| p_\theta(z|x, y)) - KL(q_\phi(z|x, y) \| p_\theta(z|y)) + \\
 &\quad \int q_\phi(z|x, y) \log p_\theta(x|y, z) dz \\
 &= KL(q_\phi(z|x, y) \| p_\theta(z|x, y)) - KL(q_\phi(z|x, y) \| p_\theta(z|y)) + \\
 &\quad \mathbb{E}_{q_\phi(z|x, y)} [\log p_\theta(x|y, z)] dz \\
 &\geq -KL(q_\phi(z|x, y) \| p_\theta(z|y)) + \mathbb{E}_{q_\phi(z|x, y)} [\log p_\theta(x|y, z)] dz
 \end{aligned}$$

2.

Let  $J$  be the dimensionality of  $z$ . Let  $\mu$  and  $\sigma$  denote the variational mean and s.d. evaluated at datapoint  $i$ , and let  $\mu_j$  and  $\sigma_j$  simply denote the  $j$ -th element of these vectors. Then:

$$\begin{aligned}
 KL(q_\phi(z|x, y) \| p_\theta(z|y)) &= \int q_\phi(z|x, y) (\log p_\theta(z|x, y) - \log q_\phi(z|x, y)) dz \\
 &= \int q_\phi(z|x, y) \log p_\theta(z|x, y) - \int q_\phi(z|x, y) \log q_\phi(z|x, y) \\
 &= \int \mathcal{N}(z; \mu, \sigma^2) \log \mathcal{N}(z; 0, I) dz - \int \mathcal{N}(z; \mu, \sigma^2) \log \mathcal{N}(z; \mu, \sigma^2) dz \\
 &= -\frac{J}{2} \log(2\pi) - \frac{1}{2} \sum_{j=1}^J (\mu_j^2 + \sigma_j^2) + \frac{J}{2} \log(2\pi) + \frac{1}{2} \sum_{j=1}^J (1 + \log \sigma_j^2) \\
 &= \frac{1}{2} \sum_{j=1}^J (1 + \log \sigma_j^2 - \mu_j^2 - \sigma_j^2)
 \end{aligned}$$

3. Generated images by CVAE.





#### Q4. Generative Adversarial Networks

Generated images by DCGAN:

