Week-6 Questions

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- 1. Which of the following is the correct formula for the KL divergence between the encoder distribution $q(z|x) = \mathcal{N}(\mu, \sigma^2)$ and the prior $p(z) = \mathcal{N}(0, 1)$ in a VAE?
 - (A) $\frac{1}{2}(\sigma^2 + \mu^2 1)$
 - (B) $\frac{1}{2}(\mu^2 + \sigma^2 \log \sigma^2 1)$
 - (C) $\mu^2 + \sigma$
 - (D) $\log(\sigma^2) + \mu$

Answer: (B)

- 2. Increasing the β value in β -VAE:
 - (A) Decreases the importance of KL divergence
 - (B) Encourages latent variables to deviate from the prior
 - (C) Encourages disentanglement by penalizing latent capacity
 - (D) Has no effect on the latent space

Answer: (C)

- 3. Assuming a latent space of dimension 2 in a VAE trained on MNIST, which of the following is true?
 - (A) KL divergence will be zero
 - (B) Decoder may struggle to reconstruct due to low capacity
 - (C) Training loss will vanish
 - (D) The posterior becomes non-Gaussian

Answer: (B)

- 4. The KL divergence term in a VAE becomes very small (close to zero). This indicates:
 - (A) Encoder overfits
 - (B) Decoder collapsed

- (C) β too large
- (D) Latent Posterior matches the prior

Answer: (D)

```
(A)
std = torch.exp(logvar / 2)
eps = torch.randn_like(std)
z = mu + eps * std

(B)
z = torch.randn(mu.size()) * logvar + mu

(C)
z = mu + torch.randn_like(mu) + logvar

(D)
z = mu * torch.randn_like(logvar)
```

Answer: (A)

5. Which of the following decoder blocks is appropriate for a VAE trained on MNIST using binary cross-entropy loss?

```
(A) nn.Sequential(
    nn.Linear(20, 400),
    nn.ReLU(),
    nn.Linear(400, 784),
    nn.Tanh()
)
```

```
(D) nn.Sequential(
nn.Linear(20, 784)
```

Answer: (B)

6. Which of the following code snippets correctly flattens an encoder output $z \in \mathbb{R}^{[B,D,H,W]}$ to compute distances to codebook vectors?

Answer: (B)

7. Which code samples 16 latent vectors from a trained codebook of size 512 with embedding dim 64?

```
(A) indices = torch.randint(0, 512, (16,))

z_q = model.codebook(indices)

(B) z_q = torch.randn(16, 64)

(C) z_q = model.encoder(torch.randn(16, 3, 32, 32))

(D) z_q = torch.ones(16, 64)
```

Answer: (A)

- 8. Theoretically it is not possible to use a deep neural network as an encoder and reduce the dimension of latent space to one. Is the this statement true??
 - (A) Yes
 - (B) No

Answer: (A)

9. VAE is

- (A) An optimization problem where the optimal encoder and decoder are found by finding the 'arg min' of the reconstruction error between data and encoded decoded data
- (B) An optimization problem where the optimal encoder and decoder are found by finding the 'arg min' of the reconstruction error between data and decoded- encoded data
- (C) An optimization problem where the optimal encoder and decoder are found by finding the 'arg max' of the reconstruction error between data and encoded decoded data
- (D) An optimization problem where the optimal encoder and decoder are found by finding the 'arg max' of the reconstruction error between data and decoded encoded data

Answer: (B)

10. The basic idea in a VQ VAE is

- (A) Use a discrete latent space with a fixed finite size dictionary and replace the output of encoder with the element with greatest distance in the dictionary
- (B) Use a discrete latent space with a fixed finite size dictionary and replace the output of encoder with the element with least distance in the dictionary
- (C) Use a discrete latent space with a learnable finite size dictionary and replace the output of encoder with the element with greatest distance in the dictionary
- (D) Use a discrete latent space with a learnable finite size dictionary and replace the output of encoder with the element with least distance in the dictionary

Answer: (D)