

## Supplementary Materials

$$\mathcal{D}_{KL}(u||\hat{y}) = \sum_{c=0}^{K-1} u^{(x,y,z)}(c) \log \frac{u^{(x,y,z)}(c)}{\hat{y}^{(x,y,z)}(c)} \quad (1)$$

$$\mathcal{L}_{mse}(\hat{y}, y) = \sum_{c=0}^{K-1} \|\hat{y}^{(x,y,z)}(c) - y^{(x,y,z)}(c)\|_2^2 \quad (2)$$

Eq. (1) represents the KL divergence for each voxel, and Eq. (2) represents the mean square error for each voxel. In the code we use a variant of KL divergence, assuming that for a voxel, the  $u$  component is  $u_c$ , in addition  $u_{\frac{1}{K}}, u_1$  stand for  $u_c = \frac{1}{K}$  or 1, respectively. The following proof procedure exists:

$$H(\hat{p}) = - \sum_{c=0}^K \hat{p}_c \log \hat{p}_c, \quad (\text{where } c \text{ is } c^{th}\text{-component}) \quad (3)$$

$$H(u, \hat{p}) = - \sum_{c=0}^K u_c \log \hat{p}_c, \quad (4)$$

$$\begin{aligned} & KL(\hat{p} || u_{\frac{1}{K}}) \\ &= \sum_{c=0}^K \hat{p}_c \log \left( \frac{\hat{p}_c}{u_c} \right) \\ &= \sum_{c=0}^K \hat{p}_c \log \hat{p}_c - \sum_{c=0}^K \hat{p}_c \log u_c \quad \text{Have, } \sum_{c=0}^K \hat{p}_c = 1, \log u_c = -\log K \\ &= -H(\hat{p}) + \log K \stackrel{K}{=} -H(\hat{p}) \end{aligned} \quad (5)$$

Symbol  $\stackrel{K}{=}$  denotes equality up to an additive or multiplicative constant.

$$\begin{aligned} & KL(u_{\frac{1}{K}} || \hat{p}) \\ &= \sum_{c=0}^K u_c \log \left( \frac{u_c}{\hat{p}_c} \right) \\ &= \sum_{c=0}^K u_c \log u_c - \sum_{c=0}^K u_c \log \hat{p}_c \quad \text{Have, } \sum_{c=0}^K u_c = 1, \log u_c = -\log K \\ &= -\log K - \frac{1}{K} \sum_{c=0}^K \log \hat{p}_c \stackrel{K}{=} H(u_{\frac{1}{K}}, \hat{p}). \end{aligned} \quad (6)$$

When  $u_c = 1$ , exist:

$$\begin{aligned}
& KL(u_1 || \hat{p}) \\
&= - \sum_{c=0}^K u_c \log \hat{p}_c = - \sum_{c=0}^K \log \hat{p}_c \\
&= H(u_1, \hat{p}) = K \cdot H(u_{\frac{1}{K}}, \hat{p}) \stackrel{K}{=} K \cdot KL(u_{\frac{1}{K}} || \hat{p}).
\end{aligned} \tag{7}$$

In Torch, if have a tensor  $T \in \mathcal{R}^{B \times K \times W \times H}$ , use `torch.kl( $u_1, T$ )` get Output  $O \in \mathcal{R}^{B \times K \times W \times H}$ , then conduct  $\mathbf{M} = \text{torch.mean}(O, \text{dim} = 1)$ ,  $\mathbf{M} = -\frac{1}{K} \sum_{c=0}^K \log \hat{p}_c$  for each voxel.

So  $\text{torch.mean}(\text{torch.kl}(u_1, T), \text{dim} = 1) = H(u_{\frac{1}{K}}, \hat{p}) \stackrel{K}{=} KL(u_{\frac{1}{K}} || \hat{p})$ .

Then,  $KL(u_{\frac{1}{K}} || \hat{p}) = -\log K - \frac{1}{K} \sum_{c=0}^K \log \hat{p}_c = -\log K + H(u_{\frac{1}{K}}, \hat{p})$ . Taking the derivative of it with respect to the parameters  $\theta$  being optimized:

$$\begin{aligned}
& \frac{\partial KL(u_{\frac{1}{K}} || \hat{p})}{\partial \theta} \\
&= \frac{\partial(-\log K + H(u_{\frac{1}{K}}, \hat{p}))}{\partial \theta} \\
&= \frac{\partial H(u_{\frac{1}{K}}, \hat{p})}{\partial \theta}.
\end{aligned} \tag{8}$$

Eventually we use  $H(u_{\frac{1}{K}}, \hat{p})$  as a substitute for  $KL(u_{\frac{1}{K}} || \hat{p})$ .