

# Supplementary Material: No Reference Opinion Unaware Quality Assessment of Authentically Distorted Images

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## A. Steps to obtain the variational approximation loss

Our proposed optimization problem for variational approximation is

$$\min_{\theta} [\text{KL}(p(Y_k, Z_k) \| q_{\theta}(Y_k, Z_k)) - \text{KL}(p(Y_k)p(Z_k) \| q_{\theta}(Y_k, Z_k))]. \quad (1)$$

Following Theorem 3.2 in [1], defining  $q_{\theta}(Y_k, Z_k) = q_{\theta}(Y_k|Z_k)p(Z_k)$ , we can rewrite the above term as

$$\min_{\theta} \left\{ \mathbb{E}_{p(Y_k, Z_k)} \left[ \log \left( \frac{p(Y_k|Z_k)p(Z_k)}{q_{\theta}(Y_k|Z_k)p(Z_k)} \right) \right] - \mathbb{E}_{p(Y_k)p(Z_k)} \left[ \log \left( \frac{p(Y_k)p(Z_k)}{q_{\theta}(Y_k|Z_k)p(Z_k)} \right) \right] \right\}. \quad (2)$$

Removing the terms that do not depend on the parameter  $\theta$ , the optimization problem reduces to

$$\max_{\theta} [\mathbb{E}_{p(Y_k, Z_k)} [\log q_{\theta}(Y_k|Z_k)] - \mathbb{E}_{p(Y_k)p(Z_k)} [\log q_{\theta}(Y_k|Z_k)]]. \quad (3)$$

Similar to [1], we can use samples to obtain the following unbiased estimate of the objective function to be maximized as

$$\mathcal{L}_{\theta_k} = \frac{1}{N} \sum_{n=1}^N \log q_{\theta}(y_k^{(n)} | z_k^{(n)}) - \frac{1}{N^2} \sum_{n=1}^N \sum_{m=1}^N \log q_{\theta}(y_k^{(m)} | z_k^{(n)}). \quad (4)$$

$$\mathcal{L}_{\theta_k} = \frac{1}{N} \sum_{n=1}^N \left[ \log q_{\theta}(y_k^{(n)} | z_k^{(n)}) - \frac{1}{N} \sum_{m=1}^N \log q_{\theta}(y_k^{(m)} | z_k^{(n)}) \right]. \quad (5)$$

## References

- [1] Pengyu Cheng, Weituo Hao, Shuyang Dai, Jiachang Liu, Zhe Gan, and Lawrence Carin. Club: A contrastive log-ratio upper bound of mutual information. In *International conference on machine learning*, pages 1779–1788. PMLR, 2020.