

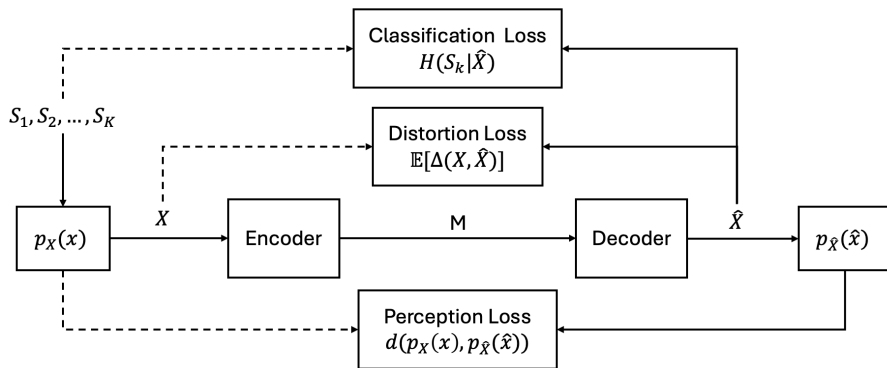
Universal Representations for Classification-enhanced Lossy Compression

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Task-oriented Lossy Compression



Source & Target labels:

► Source: $X \sim p_X(x)$

► Target labels: S_1, \dots, S_K

Lossy Compression:

► Encoder: source $X \mapsto$ message M .

► Decoder: message $M \mapsto$ data \hat{X} , for task-oriented downstream applications.

Rate-Distortion-Perception/Classification Functions

Definition

The rate-distortion-perception function:

$$R(D, P) = \min_{p_{\hat{X}|X}} I(X; \hat{X}) \quad (1a)$$

$$\text{s.t.} \quad \mathbb{E}[\Delta(X, \hat{X})] \leq D, \quad (1b)$$

$$d(p_X, p_{\hat{X}}) \leq P. \quad (1c)$$

Definition

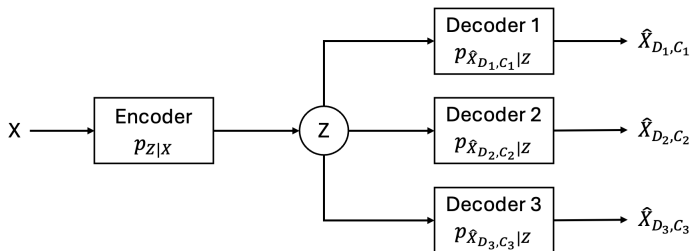
The rate-distortion-classification function:

$$R(D, C) = \min_{p_{\hat{X}|X}} I(X; \hat{X}) \quad (2a)$$

$$\text{s.t.} \quad \mathbb{E}[\Delta(X, \hat{X})] \leq D, \quad (2b)$$

$$H(S|\hat{X}) \leq C. \quad (2c)$$

Universal Rate-Distortion-Classification Function



Definition

Let Z represent X through a transform $p_{Z|X}$. Define:

$$\mathcal{P}_{Z|X}(\Theta) = \{p_{Z|X} : \exists p_{\hat{X}_{D,C}|Z}, (D, C) \in \Theta, \mathbb{E}[\Delta(X, \hat{X}_{D,C})] \leq D, H(S|\hat{X}_{D,C}) \leq C\},$$

The universal rate-distortion-classification function:

$$R(\Theta) = \inf_{p_{Z|X} \in \mathcal{P}_{Z|X}(\Theta)} I(X; Z). \quad (3)$$

Rate Penalty

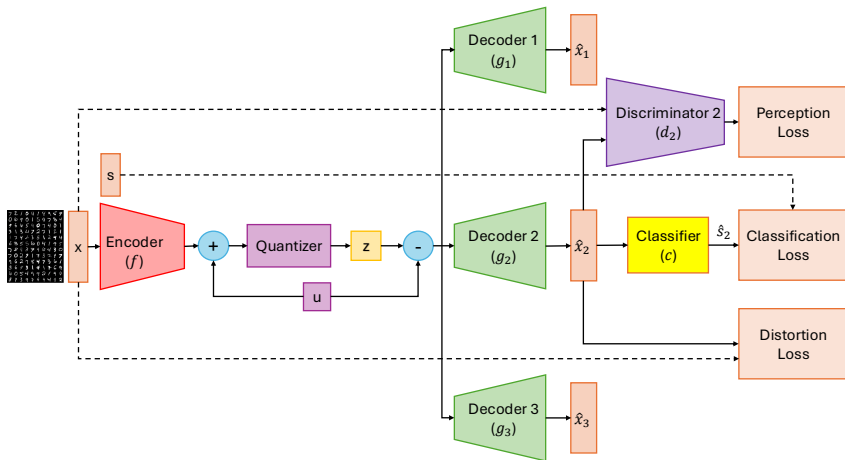
Definition

The rate penalty incurred by meeting all constraints in Θ with **fixed encoder**:

$$A(\Theta) = R(\Theta) - \sup_{(D,C) \in \Theta} R(D,C). \quad (4)$$

- Ideally, $A(\Theta) = 0$, i.e., a **single encoder** suffices for the entire (D,C) tradeoffs.

Universal Representation for Image Lossy Compression



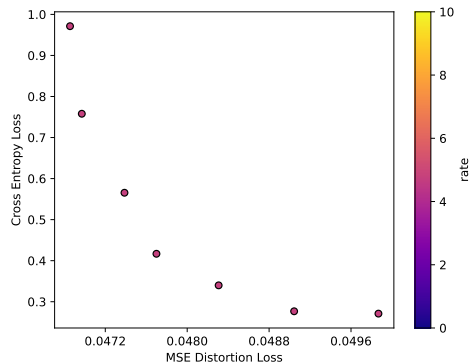
► Dataset: MNIST

► Pre-trained classifier (c)

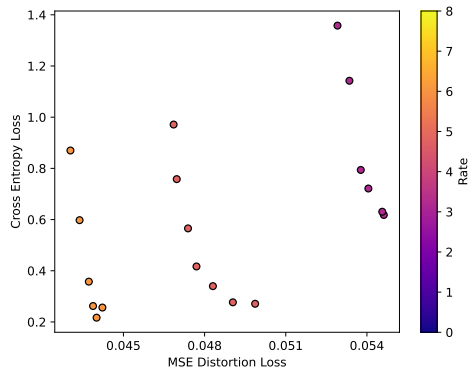
► Train the encoder (f), decoder (g), and GAN-based discriminator (d):

$$\mathcal{L}_{RDC} = \mathbb{E}[\|X - \hat{X}\|^2] + \lambda_c \text{CE}(s, \hat{s}), \quad \mathcal{L}_{RDP} = \mathbb{E}[\|X - \hat{X}\|^2] + \lambda_p W_1(p_X, p_{\hat{X}}).$$

Experimental Results



(a) Classification-distortion-rate function with $R = 4.75$.



(b) Classification-distortion-rate functions with vary rates.

Experimental Results

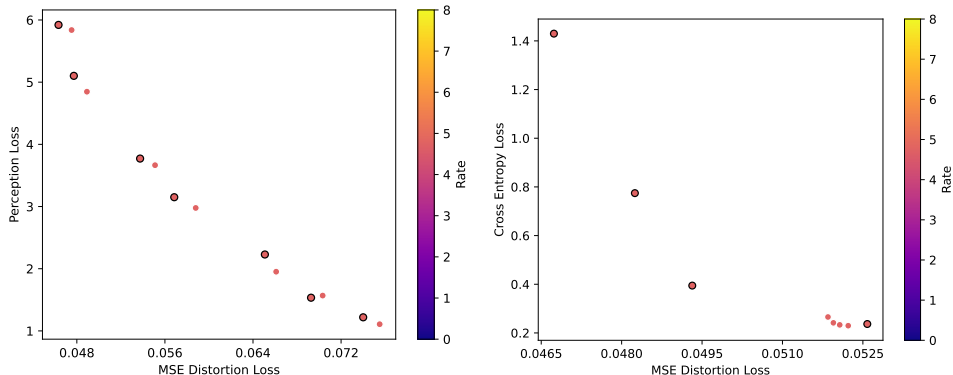


Figure: Perception (classification)-distortion-rate functions with $R = 4.75$. Points with black outlines indicate results from end-to-end trained encoder-decoder models for a particular perception-distortion target. Other points denote losses for universal models, in which decoders are trained separately using a fixed encoder.