

# 1 The ISIC model

The issue-sensitive image captioning model (ISIC model) by Nie, Cohn-Gordon, and Potts, 2020 is defined as follows:

- $\mathcal{I}$  is a set of images;  $W$  is the lexicon (a set of words); sequences of words (of arbitrary positive length) are denoted as  $\vec{w}$
- the **literal speaker**  $S_0$  is given by a pretrained language model:

$$S_0(\vec{w} \mid i) \quad [\text{pretrained}]$$

- the **pragmatic listener**  $L_1$  is defined via Bayes rule (where prior probabilities  $P(i)$  are assumed to be flat)

$$L_1(i \mid \vec{w}) \propto P(i) S_0(\vec{w} \mid i)$$

- the **issue-sensitive speaker** is defined in terms of three utility components:

$$S_1(\vec{w} \mid i, C) \propto \exp [\alpha((1 - \beta)U_1(i, \vec{w}, C) + \beta U_2(i, \vec{w}, C)) - \text{Cost}(i, \vec{w})] , \text{ where}$$

$$U_1(i, \vec{w}, C) = \log L(C(i) \mid \vec{w})$$

$$U_2(i, \vec{w}, C) = \mathcal{H}(L_1(\cdot \mid \vec{w}, C(i)))$$

$$\text{Cost}(i, \vec{w}) = -\log S_0(\vec{w} \mid i)$$

## 2 Minimal model setup

To explore the behavior and predictions of this model outside of neural language models, we can look at a minimalized, discrete setup with a ground-truth semantics.

- $\mathcal{I} = F_1 \times \cdots \times F_n$  is a set of objects identified uniquely via a list of  $n$  features  $F_1, \dots, F_n$ . All  $F_j$  are non-empty, finite sets which do not contain  $\emptyset$  and are mutually disjoint.  $k_j$  is the number of feature values in feature set  $F_j$ .
- Instead of words and sequences thereof, we look at a set of **messages**  $\mathcal{M} = F'_1 \times \cdots \times F'_n$ , where  $F'_j = F_j \cup \{\emptyset\}$ .
- The **meaning function**  $\mathcal{B}: \mathcal{I} \times \mathcal{M} \rightarrow [0; 1]$  is defined as:

$$\mathcal{B}(i, m) = \begin{cases} 1 & \text{if } \forall j((1 \leq j \leq n) \wedge (m_j \neq \emptyset)) \rightarrow m_j = i_j \\ \epsilon & \text{otherwise.} \end{cases}$$

- An **issue** is a partition on  $\mathcal{I}$  derived from a feature. The partition  $C^j$  derived from feature  $j$  is:

$$C^j = \left\{ \left\{ i \in \mathcal{I} \mid i_j = f \right\} \mid f \in F_j \right\}$$

- The **literal speaker** is defined as:

$$S_0(m \mid i) \propto \mathcal{B}(i, m)$$

- The **parameters** of this setup are  $n$ , the list of  $k_j$ ,  $\alpha$  and  $\beta$  and  $\epsilon$ . The most basic case is  $n = 3$  and  $k_j = 2$  for all  $j$ .  $\epsilon$  is supposed to be small, e.g.,  $1 \times 10^{-4}$ . For extreme (noise-free) predictions, we can set  $\epsilon = 0$  and  $\alpha \rightarrow \infty$ .

## References

Nie, Allen, Reuben Cohn-Gordon, and Christopher Potts (Nov. 2020). “Pragmatic Issue-Sensitive Image Captioning”. In: *Findings of the Association for Computational Linguistics: EMNLP 2020*. Online: Association for Computational Linguistics, pp. 1924–1938.