

Word translation without parallel data

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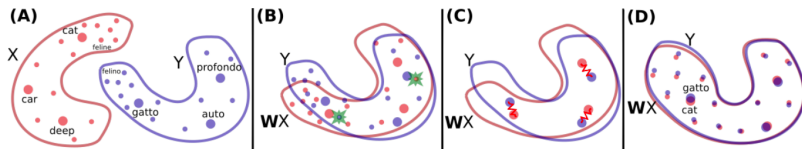
План

- ▶ машинный перевод: необходимы данные
- ▶ два уровня: word & sentence
 - ▶ adversarial training
 - ▶ метрика Cross-Domain Similarity Local Scaling (CSLS)
 - ▶ unsupervised validation criterion
- ▶ результаты

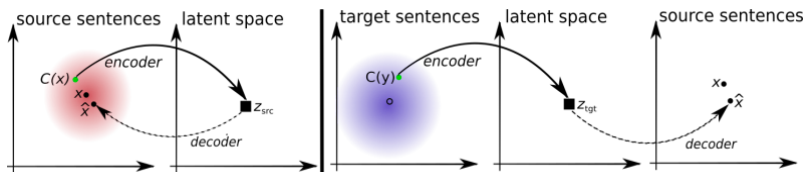
Word level

$$W^* = \operatorname{argmin}_{W \in M_d(\mathbb{R})} \|WX - Y\|_F$$

translation t of any word s : $t = \operatorname{argmax}_t \cos(Wx_s, y_t)$



Sentence level



Adversarial Training. Word Level

source: $X = x_1, \dots, x_n$, **target:** $Y = y_1, \dots, y_m$

Discriminator objective θ_D

probability $P_{\theta_D}(\text{source} = 1|z)$ that z is a vector from source language

$$\begin{aligned}\mathcal{L}_D(\theta_D|W) = & -\frac{1}{n} \sum_{i=1}^n \log P_{\theta_D}(\text{source} = 1|Wx_i) - \\ & -\frac{1}{m} \sum_{i=1}^m \log P_{\theta_D}(\text{source} = 0|y_i)\end{aligned}$$

Mapping objective W

$$\begin{aligned}\mathcal{L}_W(W|\theta_D) = & -\frac{1}{n} \sum_{i=1}^n \log P_{\theta_D}(\text{source} = 0|Wx_i) - \\ & -\frac{1}{m} \sum_{i=1}^m \log P_{\theta_D}(\text{source} = 1|y_i)\end{aligned}$$

Cross-Domain Similarity Local Scaling (CSLS)

- ▶ K-NN is asymmetric
- ▶ $r_T(W_{X_S}) = \frac{1}{K} \sum_{y_t \in \mathcal{N}_T(W_{X_S})} \cos(W_{X_S}, y_t)$
- ▶ $CSLS(W_{X_S}, y_t) = 2 \cos(W_{X_S}, y_t) - r_t(W_{X_S}) - r_s(y_t)$

Sentence level

$$\mathcal{Z}^S = (x_1^s, \dots, x_{|\mathcal{W}_S|}^s)$$

$$\mathcal{Z}^N = (x_1^t, \dots, x_{|\mathcal{W}_T|}^t)$$

input sentence: $\mathbf{x} = (x_1, \dots, x_m)$, **language** l

encoder: $e_{\theta_{enc}, \mathcal{Z}}(\mathbf{x}, l) = e(\mathbf{x}, l) \implies \mathbf{z} = (z_1, \dots, z_m)$

decoder: $d_{\theta_{dec}, \mathcal{Z}}(\mathbf{x}, l) = d(\mathbf{x}, l) \implies \mathbf{y} = (y_1, \dots, y_k)$

encoder and decoder - Bi-LSTM

Objective functions

Denoising Auto-Encoding

$$\mathcal{L}_{auto}(\theta_{enc}, \theta_{dec}, \mathcal{Z}, l) = \mathbb{E}_{x \sim \mathcal{D}_l, \hat{x} \sim d(e(C(x), l), l)} [\Delta(\hat{x}, x)]$$

Cross Domain Training

$$\mathcal{L}_{cd}(\theta_{enc}, \theta_{dec}, \mathcal{Z}, l_1, l_2) = \mathbb{E}_{x \sim \mathcal{D}_{l_1}, \hat{x} \sim d(e(C(M(x)), l_2), l_1)} [\Delta(\hat{x}, x)]$$

Adversarial Training

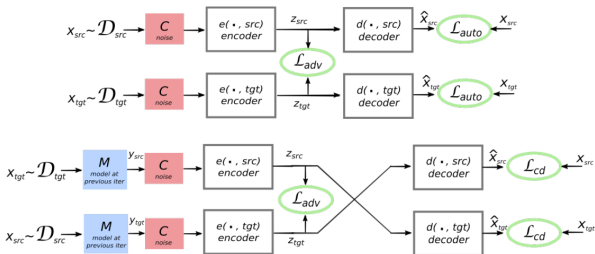
$$\mathcal{L}_{adv}(\theta_{enc}, \mathcal{Z} | \theta_D) = -\mathbb{E}_{(x_i, l_i)} [\log p_D(l_j | e(x_i, l_i))]$$

$$\mathcal{L}_D(\theta_D | \theta, \mathcal{Z}) = -\mathbb{E}_{(x_i, l_i)} [\log p_D(l_i | e(x_i, l_i))]$$

Final Objective

Final Objective Function

$$\begin{aligned}\mathcal{L}(\theta_{enc}, \theta_{dec}, \mathcal{Z}) = & \lambda_{adv} \mathcal{L}_{adv}(\theta_{enc}, \mathcal{Z} | \theta_D) + \\ & + \lambda_{cd} [\mathcal{L}_{cd}(\theta_{enc}, \theta_{dec}, \mathcal{Z}, src, tgt) + \mathcal{L}_{cd}(\theta_{enc}, \theta_{dec}, \mathcal{Z}, tgt, src)] + \\ & + \lambda_{auto} [\mathcal{L}_{auto}(\theta_{enc}, \theta_{dec}, \mathcal{Z}, src) + \mathcal{L}_{auto}(\theta_{enc}, \theta_{dec}, \mathcal{Z}, tgt)]\end{aligned}$$

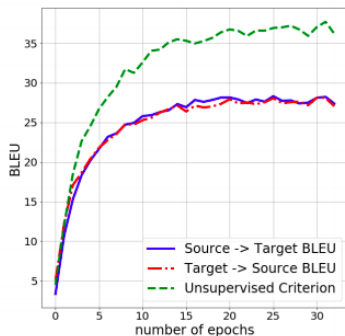
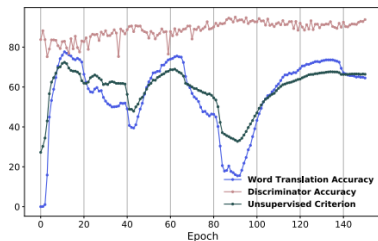


Validation criterion

$$M_{src \rightarrow tgt}(x) = d(e(x, src), tgt)$$

$$MS(e, d, \mathcal{D}_{src}, \mathcal{D}_{tgt}) = \frac{1}{2} \mathbb{E}_{x \sim \mathcal{D}_{src}} [BLEU(x, M_{src \rightarrow tgt} \circ M_{tgt \rightarrow src}(x))] + \\ + \frac{1}{2} \mathbb{E}_{x \sim \mathcal{D}_{tgt}} [BLEU(x, M_{tgt \rightarrow src} \circ M_{src \rightarrow tgt}(x))]$$

Validation criterion



Experiments

| | en-es | es-en | en-fr | fr-en | en-de | de-en | en-ru | ru-en | en-zh | zh-en | en-eo | eo-en |
|--|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|
| <i>Methods with cross-lingual supervision and fastText embeddings</i> | | | | | | | | | | | | |
| Procrustes - NN | 77.4 | 77.3 | 74.9 | 76.1 | 68.4 | 67.7 | 47.0 | 58.2 | 40.6 | 30.2 | 22.1 | 20.4 |
| Procrustes - ISF | 81.1 | 82.6 | 81.1 | 81.3 | 71.1 | 71.5 | 49.5 | 63.8 | 35.7 | 37.5 | 29.0 | 27.9 |
| Procrustes - CSLS | 81.4 | 82.9 | 81.1 | 82.4 | 73.5 | 72.4 | 51.7 | 63.7 | 42.7 | 36.7 | 29.3 | 25.3 |
| <i>Methods without cross-lingual supervision and fastText embeddings</i> | | | | | | | | | | | | |
| Adv - NN | 69.8 | 71.3 | 70.4 | 61.9 | 63.1 | 59.6 | 29.1 | 41.5 | 18.5 | 22.3 | 13.5 | 12.1 |
| Adv - CSLS | 75.7 | 79.7 | 77.8 | 71.2 | 70.1 | 66.4 | 37.2 | 48.1 | 23.4 | 28.3 | 18.6 | 16.6 |
| Adv - Refine - NN | 79.1 | 78.1 | 78.1 | 78.2 | 71.3 | 69.6 | 37.3 | 54.3 | 30.9 | 21.9 | 20.7 | 20.6 |
| Adv - Refine - CSLS | 81.7 | 83.3 | 82.3 | 82.1 | 74.0 | 72.2 | 44.0 | 59.1 | 32.5 | 31.4 | 28.2 | 25.6 |

| | English to Italian | | | Italian to English | | |
|--|--------------------|-------------|-------------|--------------------|-------------|-------------|
| | P@1 | P@5 | P@10 | P@1 | P@5 | P@10 |
| <i>Methods with cross-lingual supervision (WaCky)</i> | | | | | | |
| Mikolov et al. (2013b) [†] | 33.8 | 48.3 | 53.9 | 24.9 | 41.0 | 47.4 |
| Dinu et al. (2015) [†] | 38.5 | 56.4 | 63.9 | 24.6 | 45.4 | 54.1 |
| CCA [†] | 36.1 | 52.7 | 58.1 | 31.0 | 49.9 | 57.0 |
| Artetxe et al. (2017) | 39.7 | 54.7 | 60.5 | 33.8 | 52.4 | 59.1 |
| Smith et al. (2017) [†] | 43.1 | 60.7 | 66.4 | 38.0 | 58.5 | 63.6 |
| Procrustes - CSLS | 44.9 | 61.8 | 66.6 | 38.5 | 57.2 | 63.0 |
| <i>Methods without cross-lingual supervision (WaCky)</i> | | | | | | |
| Adv - Refine - CSLS | 45.1 | 60.7 | 65.1 | 38.3 | 57.8 | 62.8 |
| <i>Methods with cross-lingual supervision (Wiki)</i> | | | | | | |
| Procrustes - CSLS | 63.7 | 78.6 | 81.1 | 56.3 | 76.2 | 80.6 |
| <i>Methods without cross-lingual supervision (Wiki)</i> | | | | | | |
| Adv - Refine - CSLS | 66.2 | 80.4 | 83.4 | 58.7 | 76.5 | 80.9 |

References

[1]. <https://arxiv.org/pdf/1710.04087.pdf>

[2]. <https://arxiv.org/abs/1711.00043>