



A young boy with a tree made of trees: Domain Adaptation of an LSTM-based Image Caption Generation Model

A project by Dominik Künkele and Maria Irena Szawerna
for the LT2318 HT22 Artificial Intelligence: Cognitive
Systems course.

Introduction (Maria)

1. Usefulness of image captioning software.
2. Do we need to re-train models for every domain?
3. CNN+LSTM image captioning models.
4. Can we adapt a pretrained model to a new domain?

Materials and Methods (Dominik)

1. The image captioning model from Xu et al. 2015, implemented by Nikolai for one of the tutorials in the course.
2. Changes:
 1. Relation filter(s)
 2. Unknown filter
 3. Excluding UNK tokens from caption generation.
3. Flickr8k (original) and imageCLEF (new domain)
4. Human judgement questionnaire

Results (Dominik)

number of samples	highest BLEU score	fine-tuned epoch
100	0.065	13
200	0.091	9
500	0.146	19
1000	0.165	18
2000	0.201	13
5000	0.222	9

Table 1: BLEU scores for different sample sizes (unknown filter 0.1)

unknown filter	highest BLEU score	fine-tuned epoch
0.1	0.065	13
0.15	0.104	18
0.2	0.176	12
1	0.162	19

Table 2: BLEU scores for different unknown filters (number of samples: 100)

Results (Maria)

Image	Best model (% responses)
15551	100_0.1 (50%)
20144	original (50%)
20343	100_0.1 (33.3%) or 100_1.0 (33.3%)
12761	original (33.3%) or 100_1.0 (33.3%)
22381	100_0.1 (83.3%)
10821	100_0.1 (100%)
18448	original (100%)
17173	original (66.7%)
13152	original (66.7%)
20272	original (100%)
11310	100_0.1 (66.7%)
14183	100_1.0 (50%)
10622	original (50%)
13123	original (100%)
19181	100_1.0 (83.3%)
16920	100_0.1 (100%)
11308	original (50%)
14016	original (83.3%)
21029	original (100%)
18355	original (33.3%) or 100_0.15 (33.3%) or 100_0.2 (33.3%)

Table 3: Best captions per image in the unfiltered imageCLEF test set.

Image	Best model (% responses)
40416	original (83.3%)
39158	original (50%)
25053	original (83.3%)
30620	100_1.0 (66.7%)
32397	100_0.2 (66.7%)
38937	original (50%)
39005	100_0.1 (33.3%) or 100_0.15 (33.3%) or 100_0.2 (33.3%)
40120	original (66.7%)
23588	100_1.0 (83.3%)
39472	100_0.1 (66.7%)
30138	original (50%)
40202	100_0.1 (83.3%)
35895	original (66.7%)
32663	100_0.2 (83.3%)
30705	original (83.3%)
38081	100_0.15 (66.7%)
31571	original (100%)
35858	original (66.7%)
37836	100_0.1 (33.3%) or 100_0.15 (33.3%)
39239	100_0.15, 100_0.2, and 100_1.0 (50%) ¹

Table 4: Best captions per image in the filtered imageCLEF test set.

Image	Best model (% responses)
2116444946[...]	original (100%)
2316097768[...]	original (100%)
2439384468[...]	original (66.7%)
2112921744[...]	100_0.2 (50%)
2392460773[...]	original (100%)
2434006663[...]	original (100%)
2308256827[...]	original (100%)
2111360187[...]	original (66.7%)
2271671533[...]	original (66.7%)
2328616978[...]	original (100%)
2456907314[...]	original (66.7%)
2229179070[...]	100_0.1 (100%)
2279980395[...]	100_0.1 (88.3%)
2393971707[...]	original (83.3%)
211277478[...]	100_0.1 (50%)
2337919839[...]	original (83.3%)
2447035752[...]	original (50%)
23445819[...]	100_0.2 (66.7%)
2448210587[...]	original (33.3%)
2445654384[...]	100_0.1 (50%) or 100_0.2 (50%)

Table 5: Best captions per image in the Flickr8k test set.

Overall: captions were deemed to be low-quality, *rubbish*, to quote one of the participants.

Discussion (Dominik)

1. The fine-tuning was much more disruptive than expected.
2. Differences in caption structure (syntax) and vocabulary are major.
 1. Less data - grammatical but thematically unfitting captions.
 2. More data - ungrammatical, theoretically thematically fitting captions (UNK tokens).
3. Small improvements based on human judgements, but the captions are still bad.

Conclusions (Maria)

1. We have addressed our questions and found out what issues impede domain adaptation.
2. Ideas for future research:
 1. Testing the influence of other hyperparameters.
 2. Testing the same thing on a different model architecture.
 3. Fine-tuning on a more similar dataset.
 4. Exploring ways of mitigating the discovered issues.

Thank you for your
attention!
