Context-aware Captions from Context-agnostic Supervision

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Objective

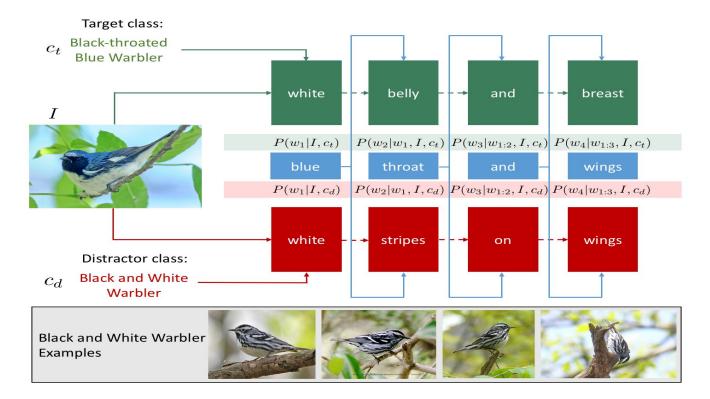
Produce pragmatic, context aware descriptions of images (captions that describe differences between images or visual concepts) using context agnostic data (captions that describe a concept or an image in isolation). We attempt the following two problems.

- Justification:
 - Given an image, a target (ground-truth) class, and a distractor class, describe the target image to explain why it belongs to the target class, and not the distractor class.
- Discriminative image captioning
 - Given two similar images, produce a sentence to identify a target image from the distractor image.

Approach

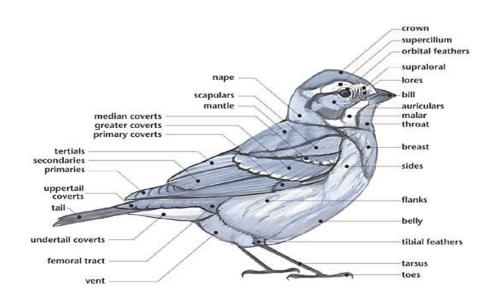
We trained our model using generic context-agnostic data (captions that describe a concept or an image in isolation), in an encoder-decoder paradigm along with attention, and used an inference technique called Emitter-Suppressor Beam Search to produce context aware image captions. Our model develops upon the architecture of Show attend and tell. For justification, apart from the image, the decoder is also conditioned on target-class.

Emitter-Suppressor Beam Search Algorithm



Dataset

We have used the CUB-200-2011 dataset which contains images of birds and their descriptions. The dataset has 200 bird classes (species), each class has 30 images and each image has 10 descriptions. The descriptions are mostly about the morphology of the birds i.e., details about various parts of their body.



Implementation details

Encoder

 We used a pretrained ResNet-34 already available in the PyTorch's torchvision module and discarded the last two layers (pooling and linear layers), since we only need to encode the image, and not classify it.

Decoder

• We used LSTMs with input embedding of size 512 and hidden states of size 1800. For justification the class is embedded into a 512 size vector.

Attention

- We used adaptive pooling over encoder to get a 14*14*512
 vector from the encoder and then applied a linear layer with ReLu
 activation to get the attention weights. Note that we used the soft
 version of the attention.
- We used Adam's optimizer with learning rate of 0.002 which is annealed every 5 epochs. We used dropout with p = 0.5. The batch size used was 64 and the number of epochs were 100. The model was trained on GTX 1060 for 15 hours.

Results

Context Agnostic Captioning

Image	Context Agnostic Caption
	this bird has a white belly and breast with black superciliary and crown



this bird has a white head and breast with grey wings and a yellow beak



this bird has a yellow belly and breast with a black superciliary and gray crown



this bird has a white crown a white breast and grey wings with black edges



this bird has a red crown a short bill and a red breast



this bird has a long neck and a long bill



this bird has a long black bill and a black and white spotted body



this bird has a black crown a white breast and a large wingspan



this bird has a long yellow bill a black crown and red eyes



this bird has a white belly and breast with a brown crown and white wing bars.

Justification Captioning

Image	Target class	Distractor class	Caption
			this bird has a brown crown brown primaries and a brown throat







this bird has a white belly and breast with a brown crown and wing







this bird has a white crown as well as a black bill







this bird has a brown crown brown primaries and a brown belly







this bird has a blue crown green primaries and a yellow belly







this bird has a yellow belly and breast with a black neck and crown







this bird has a red crown red primaries and a red belly







this bird has a pointed yellow bill with a yellow breast



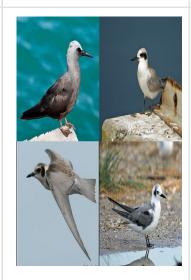




this bird has a yellow crown a short bill and a yellow breast







this bird has a black crown white primaries and a white belly

Discriminative Captioning

Target Image	Distractor Image	Caption
		this bird is brown in color over all of its body except for its wings and tail that have white around them
		this bird has wings that are gray and has a black tail and a black bill
		this bird has a brown crown brown primaries and a brown throat
		this bird has webbed feet with a bright orange wide beak and jet black over the rest of its body





this bird has blackhead and body but there is fins of feathers off of the wing that are black white and red





this bird has a white belly brown breast blue head and white wing bars





a small brown bird with a white throat and red eyerings





a small green bird with a yellow breast and yellow bill





this bird has a white belly and breast with blue wings a light gray eyebrow on the head of a front of black and white striped on the wings and bright pale blue rectrices

Requirements

Kindly use the requirements.txt to set up your machine for replicating this project, some dependencies are :

```
h5py==2.9.0
matplotlib==3.0.3
nltk==3.4.1
numpy==1.16.2
pandas==0.24.2
pillow==5.3.0
python==3.7.3
pytorch==1.0.0
torchfile==0.1.0
torchvision==0.2.1
tqdm==4.31.1
```

You can install these dependencies using pip install -r requirements.txt

Setup

Training

```
python datapreprocess.py \path\to\data\set
\path\to\vocab\
```

```
python train.py
python train justify.py
```

Testing

Download the pretrained models checkpoint_d and checkpoint_j

- Context agnostic captioning: python beamsearch.py c image path
- 2. Justification: python beamsearch.py cj target_image_path target_class_path distractor_class_path
- 3. Discrimination: python beamsearch.py cd target_image_path distractor_image_path

License

This project is licensed under the MIT License - see the LICENSE file for details.

References

- 1. Paper: Context-aware Captions from Context-agnostic Supervision
- 2. Dataset:

Images : CUB-200-2011Captions : Reed et al.

3. A beautiful tutorial on Show, Attend and Tell Implementation