Exploiting Image-Text Synergy for Contextual Image Captioning

Sreyasi Nag Chowdhury Rajarshi Bhowmik Hareesh Ravi Gerard de Melo Simon Razniewski Gerhard Weikum



Problem Statement



Design your own Spaghetti

Multimodal documents on the web contain images placed at meaningful locations within the textual narrative.

The image captions are commonly conditioned on the surrounding text. We study the generation of such Contextual Captions, distinct from Arrabiata at our Pasta bar. conventional image captioning.

Dataset

Novel Contextual Captioning dataset:

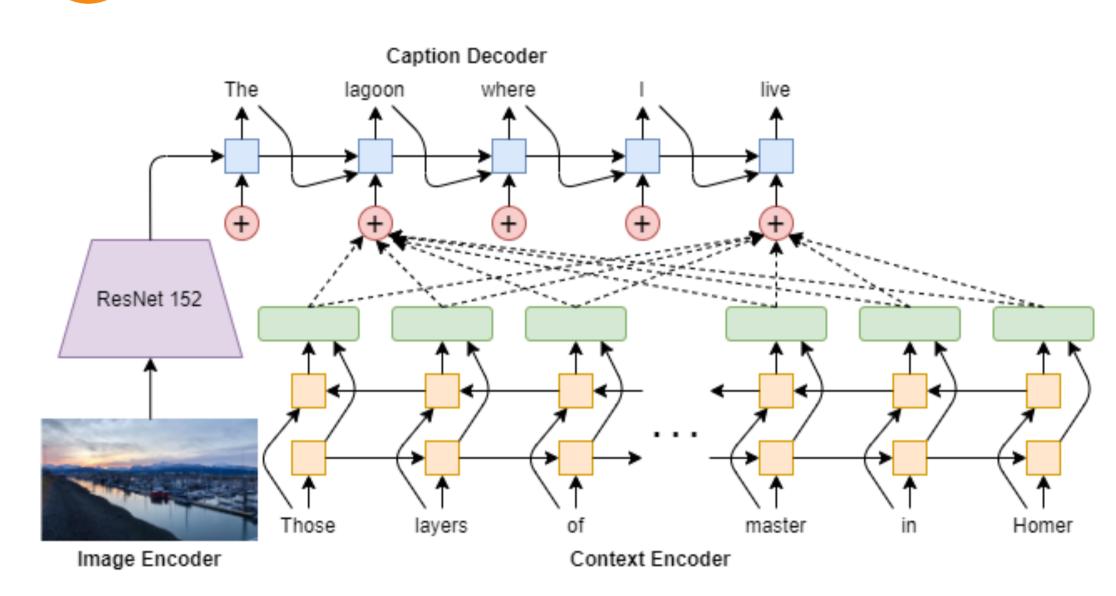
- Data scrapping from subreddit /r/pics
- Domain-agnostic posts
- 250,000 posts spanning one year
- Post: 1 image, caption, 1-10 comments
- Captions contain 10.6 words on avg.
- Concatenated comments serve as image context or associated paragraph
- Paragraphs contain 59.2 words on avg.

Data splits based on Named Entities in image captions:

- 137,732 samples with NE
- 104,653 samples without NE
- Additional splits ensuring overlap between context and caption

SpaCy is leveraged to detect 14 types of NE in image captions.

Model Formulation



Text Encoder: BiLSTM Image Encoder: ResNet152

Decoder: LSTM, attentionweighted sum of encoder states concatenated to current state to incorporate contextual information.

Objective:
$$\mathcal{L}(\theta) = \sum_{t=1}^{N} -\log p(w_t^c \mid w_1^c, \dots, w_{t-1}^c, I, P; \theta)$$

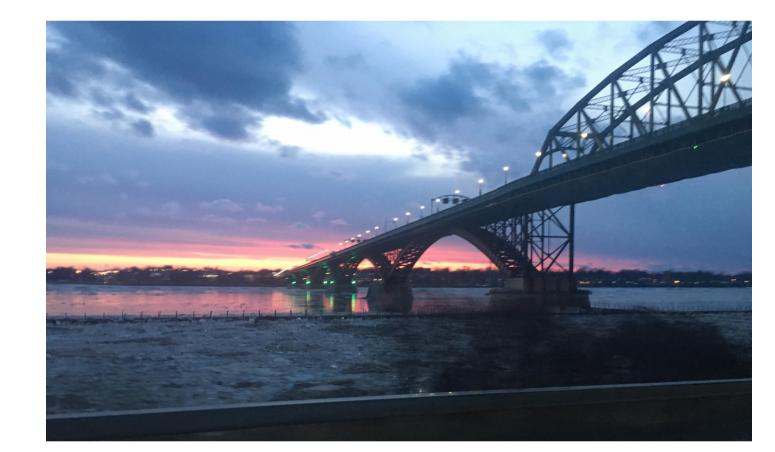
Output of BiLSTM:
$$\tilde{\mathbf{G}}_{\mathbf{t}} = \sum_{i=i}^{M} \alpha_i^t \mathbf{g}_{\mathbf{i}}$$

Attention weights:
$$\alpha_i^t = \frac{\mathbf{v}^{\mathsf{T}}(\mathbf{W_g}\mathbf{g_i} + \mathbf{W_h}\mathbf{h_t} + \mathbf{b})}{\sum_{i'=1}^{M} \mathbf{v}^{\mathsf{T}}(\mathbf{W_g}\mathbf{g_{i'}} + \mathbf{W_h}\mathbf{h_t} + \mathbf{b})}$$

	BLEU-1	ROUGE-L	$CIDE_{\Gamma}$	SPICE	SemSim
Image-only	7.80	7.50	0.38	0.16	0.76
Text-only	6.87	6.54	0.61	0.36	0.72
Contextual	9.30	9.68	0.78	0.50	0.77

Table: Quantitative evaluation of baselines and Contextual Captioning on standard text similarity measures.

- Contextual Captions capture information from both visual and textual modalities.
- They are linguistically rich compared to text-only and image-only captions.



Context: I recently moved to Buffalo, NY... ...every day I am discovering how beautiful this town is. I wanted to share the pallet of colors the sunset had that evening...

Generated Image-only Caption:

- A picture I took of a mossy branch through the shadows of a cloud.

Generated Contextual Captions:

- A beautiful sunset path to heaven.
- A sunset...unknown artist.

References:

- [1] Nag Chowdhury et al. "Illustrate Your Story: Enriching Text...", WSDM 2020
- [2] Nag Chowdhury et al. "SANDI: Story-and-Images Alignment", EACL 2021





