Berliner Hochschule
für Technik

B

Image Captioning in Personal

Photo Management

Studiere Zukunft

Personal Photo Management

Need to manage a lot of photos from different devices

Common sort criteria are when, where, who and what and this is mostly automated thanks to machine learning and metadata

Image captions in personal photo management is on the other hand rarely looked at

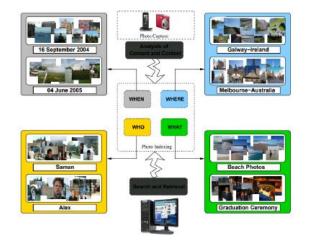


Figure 1.2: W4 entries: when, where, what and who [Cooray 2008]

Image Captions

Large differences between platforms

Different goals from searchability to reacting to other people

Different mechanisms like hashtags or attribution



C Q V · C Gefilit 12.119 Mal google These mountains have the range 😌 #TeamPicel (a) Instagram [@google 2023]



Google's headquarters, the Googleplex

(c) Wikipedia [Wikipedia 2024]



Casey Reges (CoseyCose) 15 511. In Olicogle loading any frem Witzland In worked. The inservents hold lie for its stagnating pretty hard and a liet of big and amail problems are being completely (proved. Also the bolt in unigets offering in getting meansy and no one is _

Nefraseljan

(b) Twitter: Its Just Corpse Dead Body GIF [@killedbygoogle 2023]



Bignage outside Boogle's new Bay Yew campus in Nountain Vew, California, US, on Monday. May 16, 2022.

(d) CNN [Fung 2023]

Figure 1.1: Different captions on platforms

Understanding the needs and limitations of self-hosted systems 3.0

Questionnaire of 20 LibrePhotos Users

Technical Survey, which collected server stats

Needs:

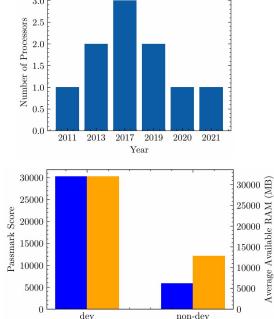
Self-hosted, private, accurate, optional, hardware accelerated, customizable, usage of previous attained knowledge like persons

Wants:

Multiple languages, different styles of captions

Limitations:

Under 8GB of RAM and not use SSE4, AVX and AVX2



Machine Type

Average

im2txt, BLIP and ONNX

im2txt is a CNN LSTM encoder decoder image to text model

Looked into common benchmarks and established a baseline

Evaluate optimization including converting to ONNX

Contrasted this with BLIP, which is based on a transformer based architecture

(always use finetunes and not base models!)

(quantizing and converting models is difficult!)

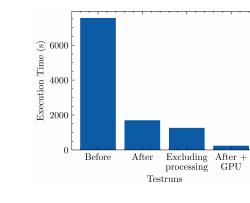


Image Captioning Benchmarks										
Model	Bleu ₁	Bleu ₂	Bleu ₃	Bleu ₄	METEOR	$ROUGE_L$	CIDEr	SPICE		
im2txt	66.4	48.4	34.3	24.5	22.8	49.1	79.8	15.8		
im2txt ONNX	66.4	48.4	34.3	24.5	22.8	49.1	79.8	15.8		
BLIP	65.7	53.4	41.5	31.4	24.6	54.8	102.2	19		
BLIP finetuned	-	-	-	39.7	-	-	133.3	-		

Adding an LLM

Semantic text tasks can be solved by LLMs like adding previous attained knowledge or translating captions

Criteria:

Should be self-hostable, e.g. <8GB of RAM

Needed fitting license

Should perform well enough for the task

Final model: Mistral 7B v0.1 by theBloke (switched by now to an instruct model)

Study LLMs + Transformer based methods

Randomized captions and image order

Users should rate the captions, from 1-5

20 images with 62 responses

Used LimeSurvey as platform

CLIPScore is a reference-free Evaluation Metric



Figure C.1: Rate the following captions for this image: im2txt: a man in a hat is sitting on a bench. blip: a man standing in front of a tree blip+llm: The image shows a man named Niaz standing in front of a tree.

Results and Learnings

CLIPScore can identify bad hallucination as the score decreases and increased in general

Using without prior knowledge bad idea, same with names, but users liked places

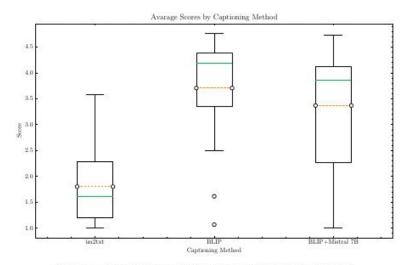


Figure 6.1: Human Ratings comparing im2txt, BLIP and BLIP + Mistral 7B

Future Work and Challenges

Trying out SOTA models like ChatGPT and multimodal models

Benchmark for user preferences with captions is missing

Adding an OCR system to solve TextCaps

Understanding multilanguage support (will look into mistral-nemo-12b)

Evaluating existing fine-tunes and creating new ones

Management of machine learning models

Issue: How to ship models with image captioning service

Shipping the model in docker image is also possible, but does not scale well

Implemented a job to download models from known sources together with settings fields

Checks every time a job gets executed

	er Logs						
0 8	emoving entries						
	lemoving an entry does rigracefully, by inspectio		e job - it's only	sensived tran	the DB. Use only in cases when	yau kriow the	et en johr beslerri
5540.11	Јођ Туре	Progress	Quesed	started	Duration	Started By	Delete
~	Train Faces	134 Asset(n) added	4 days ago	4 days ago	914 milliseconds	admin	Herein
4	Find Similar Faces	149 Asset(s) added	4 days Ago	4 days ago	834 milliseconds	admin	Rences
~	Calculate Clip Embeddings	9 Asset(s) added	4 days ago	4 days ago	2 seconds, 993 milliseconds	admin	forces
~	Scan Photos	332 Assert(s) added	4 days: Ago	4 daya ago	22 seconds, 862 milliseconds	admin	Renau
~	Downized Models	7 Assetisi	4 days	4 days	1 minute, 4 seconds, 668	admin	Rences

Figure 7.8: Worker logs with an entry for download models

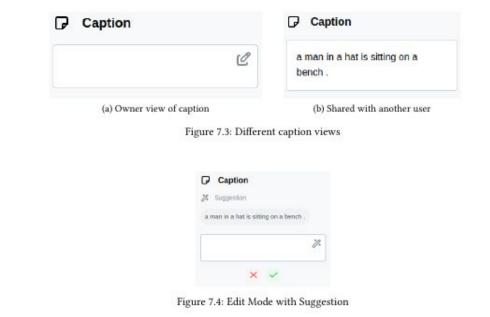
User Experience of generating captions

Issue: User may already have captions and only want to generate new ones sometimes

I added a suggestion box with a suggestion

User has to click to generate a caption on the magic wand icon

Possible different UX when using in a context of sharing images or when improving accessibility





Tags in Caption

Issue: User may prefer caption with hashtags

Implemented automated albums based on tags

This could also be something social, like tagging another user of an instance, where you maybe need an overview



Figure 7.5: Tags implemented with Tiptap

RAM usage in python

Issue: User do not have enough RAM or want more efficient RAM usages after caption creation

Added a warning when selecting large models

Implemented restarting of flask service after a time of non usage to ensure model is unloaded after a time (hard issue!)



Figure 7.2: Warning of High RAM usage

Hardware acceleration

Issue: User want faster image captioning performance

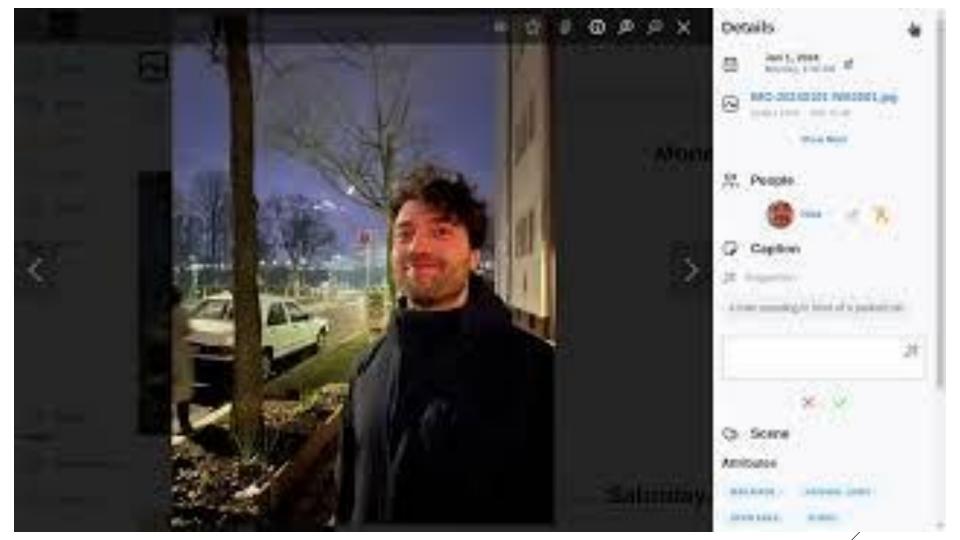
Docker image with Nvidia support used

Issues with supporting users as the GPU does not get recognized sometimes

Issues with unloading models, sometimes the driver reports having no VRAM

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Demo: Integrating it into LibrePhotos

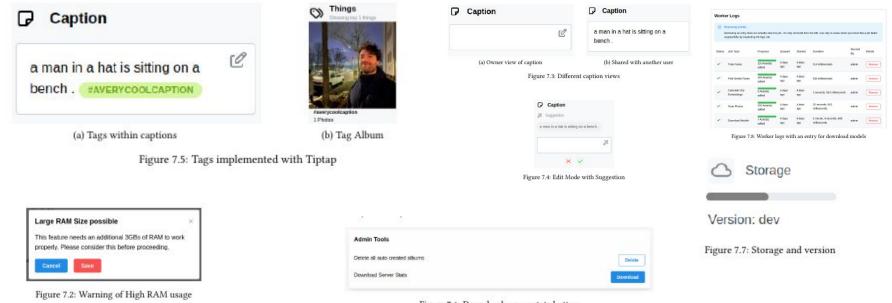


Figure 7.6: Download server stats button

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