



Mining data about people from Wikipedia using LLMs

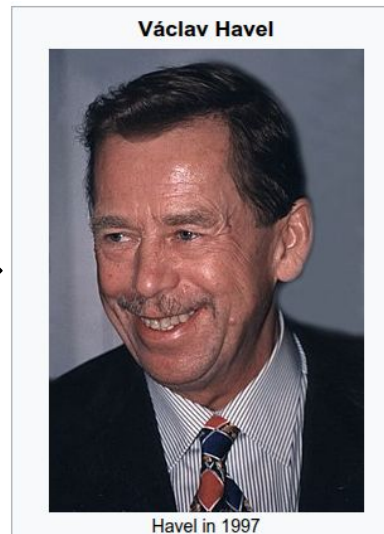
summer semester 2024/2025

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Bc. Maroš Mešter	- regex solution
Bc. Radim Průdek	- LLM-based solution
Bc. Aleš Trna	- data analysis

Abstract

The aim of this project is to utilize **Large Language Models** to automatically extract age-related information from articles about real **people on Wikipedia**. This is done in order to create a database of people's faces annotated with their age. To evaluate this method of annotation, **a ground truth annotation database** together with **a tool for manual annotation** was developed. Finally, the accuracy and coverage of the LLM-based approach was evaluated.

image of a person on Wikipedia →



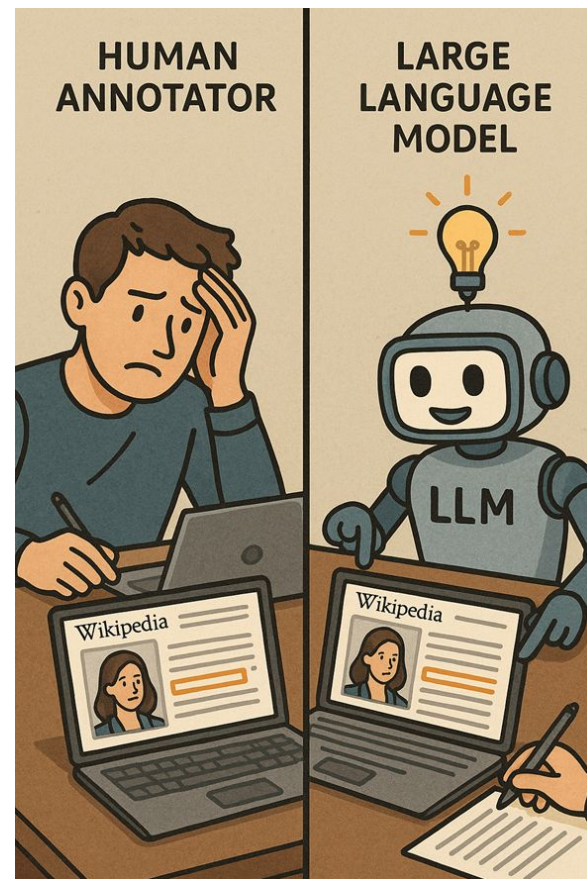
← image caption

Project goals

1. Create a “baseline” automated solution without using AI.
2. Create a ground truth database of an adequate size (can be used as metric).
3. Create manual annotation tool for creating such database.
4. Design a LLM-based automated annotation system.
5. Compare the models and create a performance report.

Background & Motivation

- Databases with reliable annotation are **expensive** and **time consuming**.
- Are Large language models capable of doing such tasks?
- Database of labeled images for training **an age estimating model**.
- Possibility of mining huge amount of free data from a large source (Wikipedia).



Baseline automated solution - **Regex**

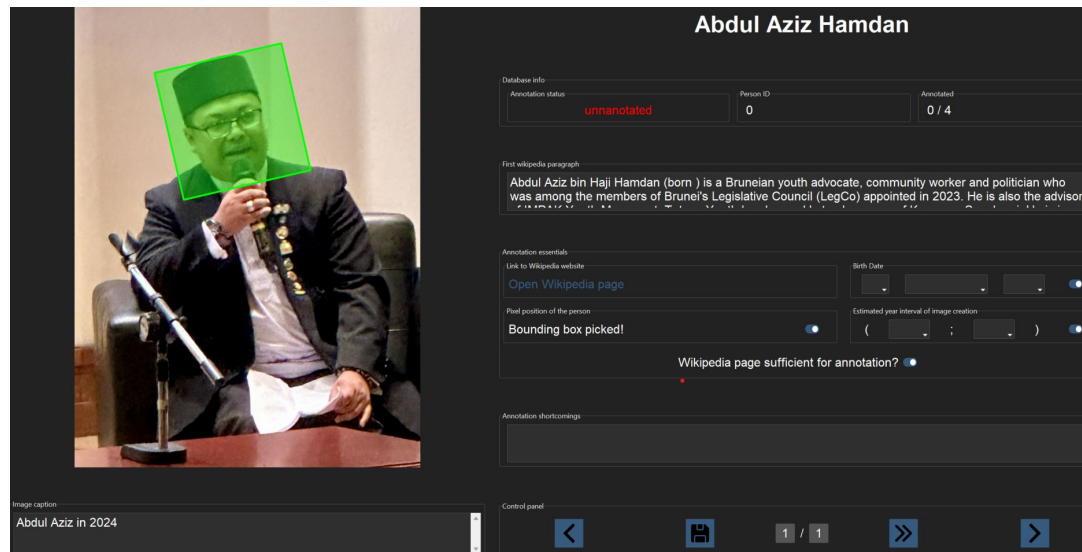
- Python regex library
- extracts birth year from the *category* box on Wikipedia
- tries to extract year from image caption using:
 - `r"\b[12]\d{3}\b"`
 - `r"\b[12]\d{3}['"]?s\b"`

Categories: [Winston Churchill](#) | [1874 births](#) | [1965 deaths](#) |
| [20th-century English historians](#) | [20th-century prime ministe](#)



Image annotation tool

- GUI for manual annotation (Python's ttkbootstrap and tkinter)
- features:
 - selecting bounding box
 - filling annotation data
 - navigating through database
 - clickable link to Wikipedia article
 - noting annotation shortcomings



Annotation tool (GUI) - dark mode.

Image annotation tool

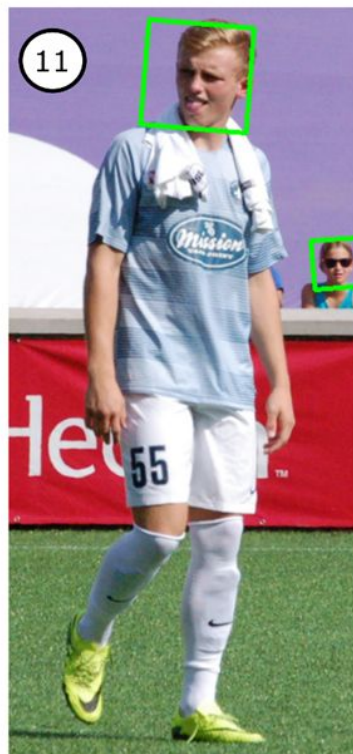


Image caption

Stanley playing for [[Saint Louis FC]] in [[2016 Saint Louis FC season|2016]]

Aedan Stanley

Database info

Annotation status: **unannotated** Person ID: 2 Annotated: 1 / 166

First wikipedia paragraph

Aedan Stanley (born December 13, 1999) is an American professional soccer player who plays as a defender for Indy Eleven in the USL Championship.

Annotation essentials

Link to Wikipedia website: [Open Wikipedia page](#)

Birth Date: 1999 ☒

Pixel position of the person: Click on the image ☒

Estimated year interval of image creation: (;) ☒

Wikipedia page sufficient for annotation? ☒

Annotation shortcomings

Control panel

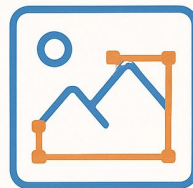
Navigation icons: Previous, Save, 1 / 1, Next, and a right arrow.

Annotation tool (GUI) - light mode.

Ground truth database

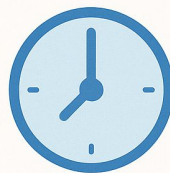
765

IMAGES
ANNOTATED
BY HUMAN



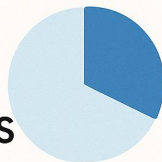
10

MANHOURS
USED FOR
ANNOTATING



71,93%

OF TOTAL IMAGES
ANNOTATABLE



LLM – based automated solution

- Ollama - Local LLM framework.
- Returns structured outputs.
- Tested on different LLMs on RCI server

Model	Mean time per annotation [s]	Mean GPU usage [GB]
Llama3.3-70B	3.4577	43.13
DeepSeek-R1-Distill-Llama-70B	3.5558	43.08
DeepSeek-R1-Distill-Qwen-32B	2.7843	21.21
DeepSeek-R1-Distill-Qwen-14B	1.3514	10.59
DeepSeek-R1-Distill-Llama-8B	0.6689	6.29
DeepSeek-R1-Distill-Qwen-7B	0.8965	5.44

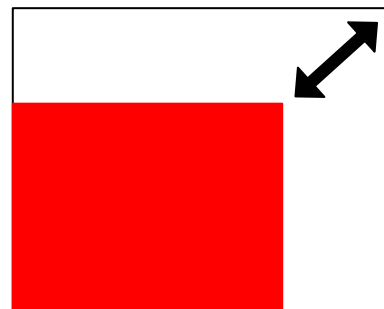
LLM – based automated solution

- **Six different prompts** for estimating the age of the person.
- Example prompt:

```
You are an expert assistant. Please analyze the following input.  
Caption of photo: {caption}  
Person biography: {person_text}  
Identify:  
- Name  
- Birthday (if available)  
- Year photo was taken (based on caption)  
Return the data in this JSON format:  
{{  
  "name": str,  
  "birthday": str or null,  
  "year_of_photo": str or null,  
  "year_of_photo_int": int or null,  
  "can_determine": true or false  
}}
```

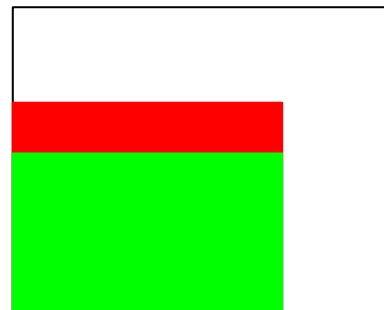
Results

Coverage of each model/prompt



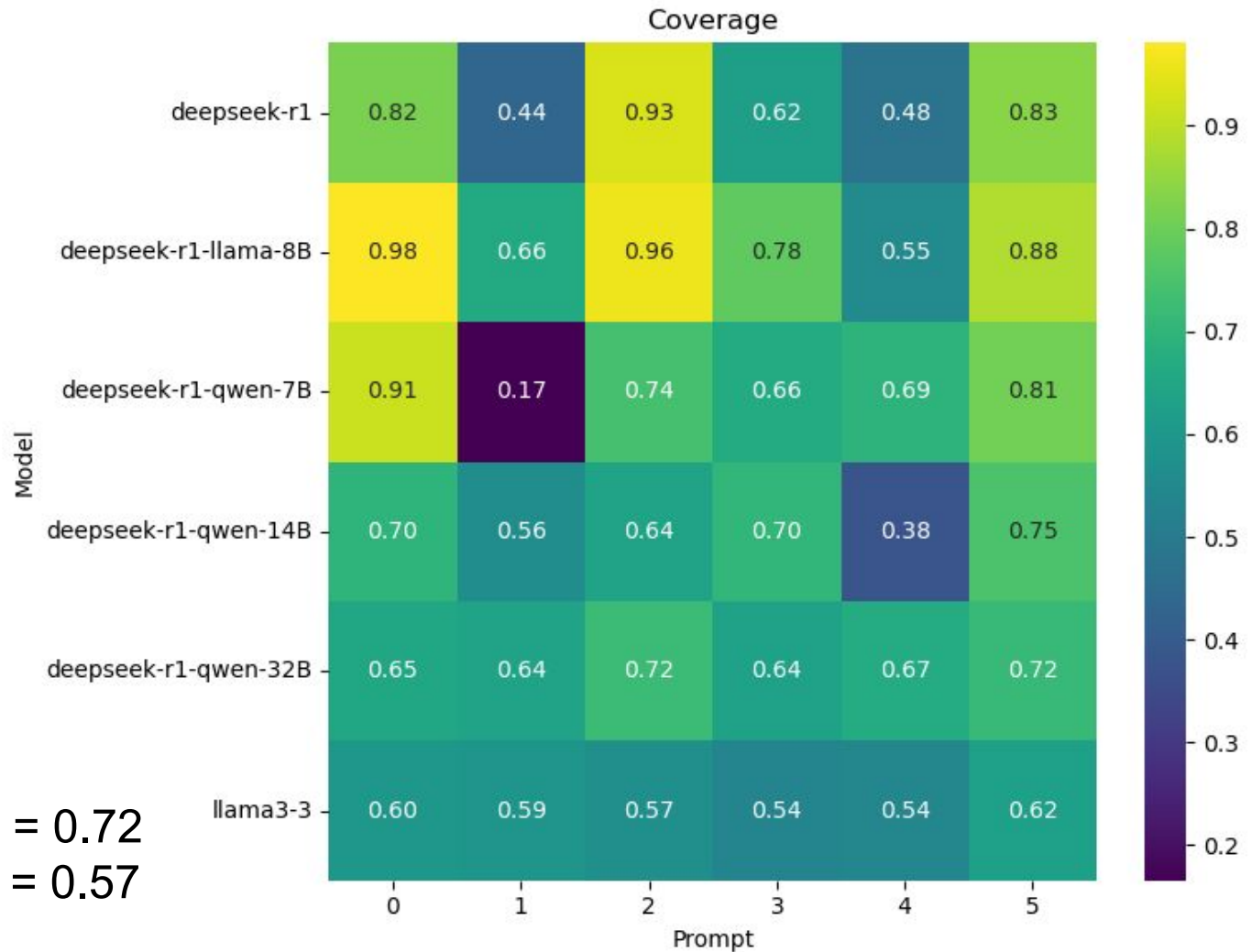
coverage

Accuracy on different quantities tested



accuracy

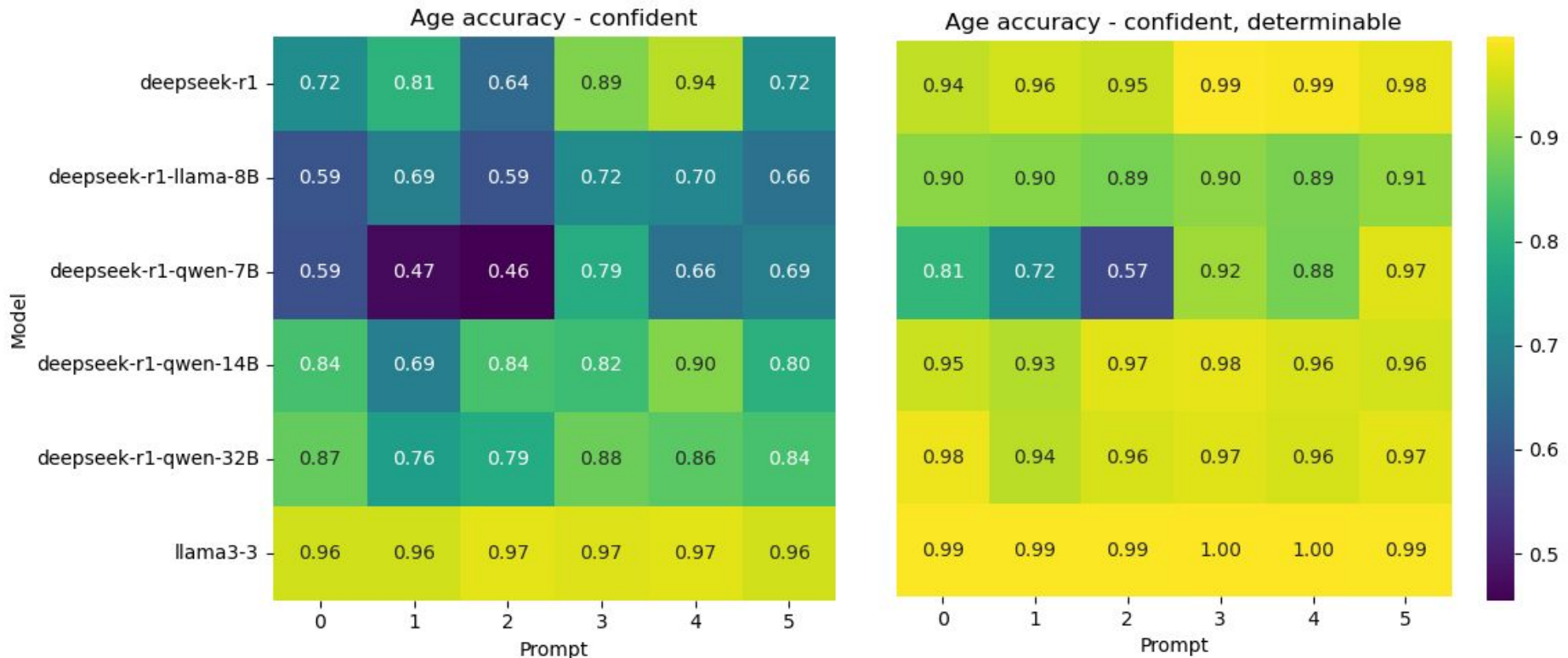
Coverage - ability to estimate age



Human coverage = 0.72

Regex coverage = 0.57

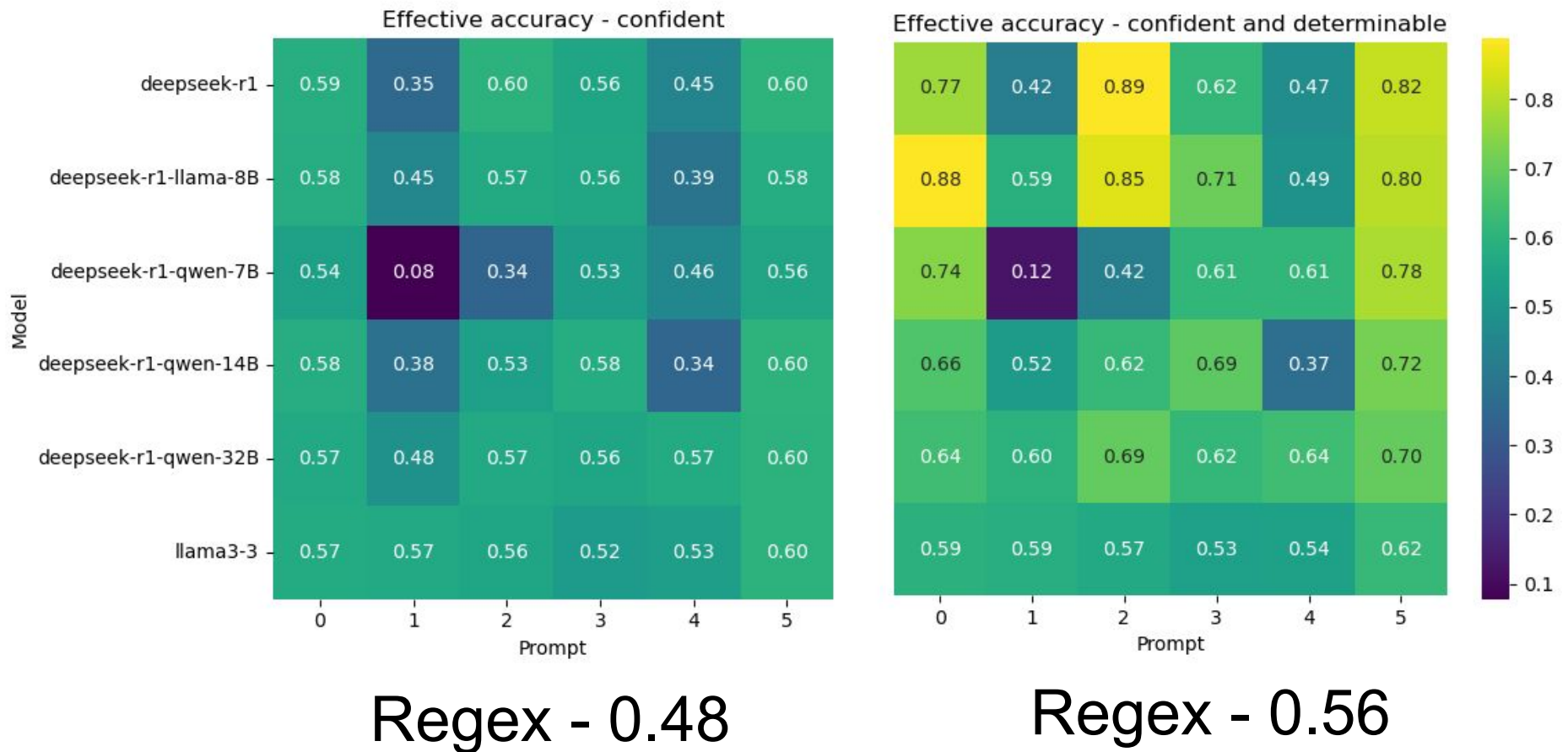
Predicted age
model is confident
age may not be determinable



Regex accuracy = 0.84

Regex accuracy = 0.98

“Effective accuracy” product of accuracy and coverage



Questions?

Resources

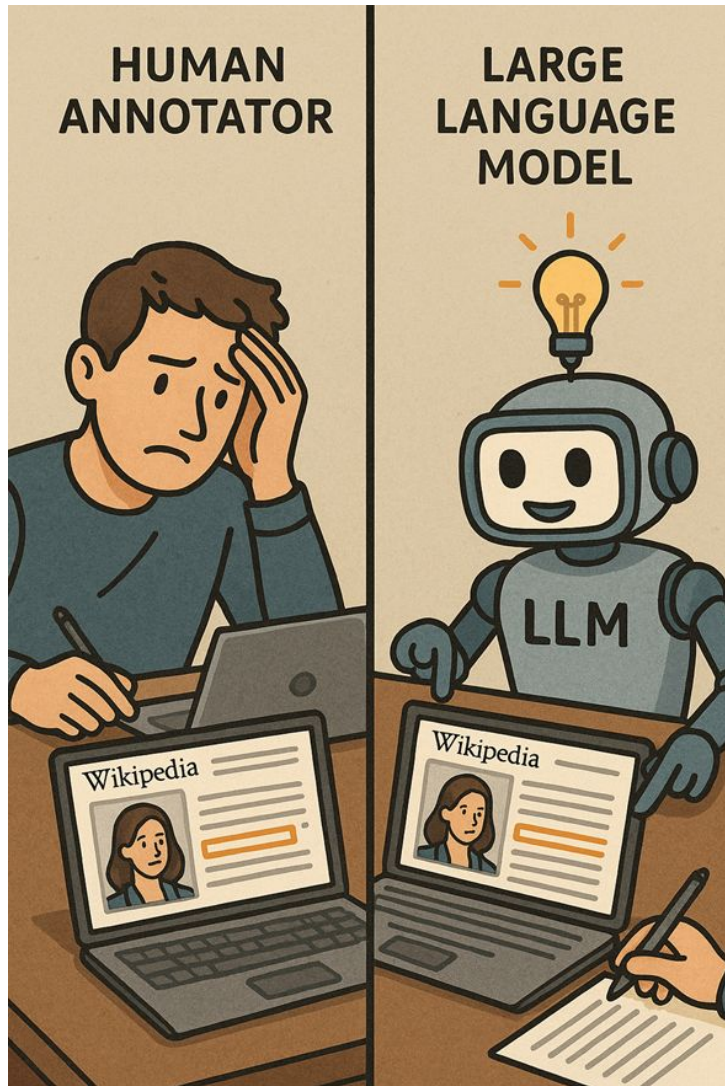
<https://rci.cvut.cz/>

<https://coderpad.io/wp-content/uploads/2022/04/coderpad-regex-the-complete-guide.jpg>

Some illustrative images were generated by OpenAI SORA model

**Thank you for your
attention**

“Create an image which shows human annotator on one side and Large language model doing an annotation task on a person on wikipedia.
Make sure to show the advantages and struggles of both actors.”



HUMAN ANNOTATOR

An annotation