



# Analysis of questions

## Post-final presentation

by JaMiMaKa group

Mikołaj Malec, Marcei Korbin, Kacper Grzymkowski, Jakub Fołtyn

# Experimental procedure

- Testing different approaches to topic clustering
  - LDA
  - Sentence embeddings
- Analysis of question complexity
  - LLM prompt engineering
  - DSI measure
  - Bloom's taxonomy
  - Question-words

## Bloom's Taxonomy

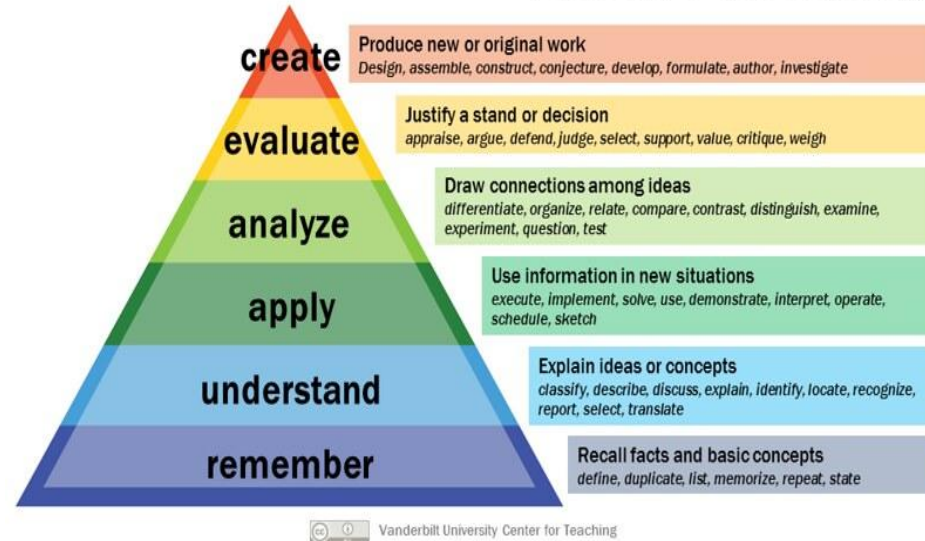


Image source: Vanderbilt University Center for Teaching

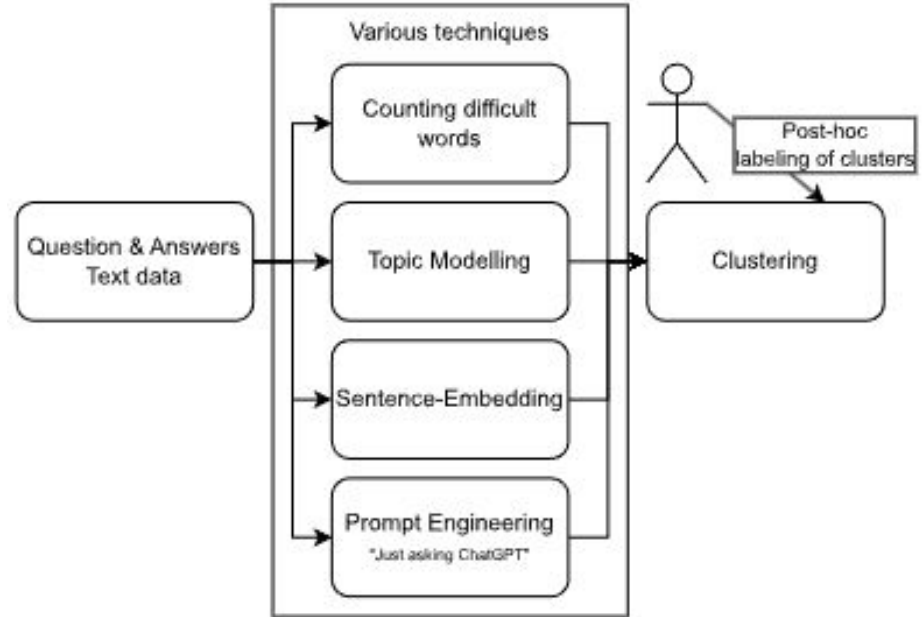


# Dataset

- The Stanford Question Answering Dataset
- Not the best choice, but:
  - Available
  - Decently sized
  - Decently clean
- Time constraints
  - We wanted to get something out of the gate

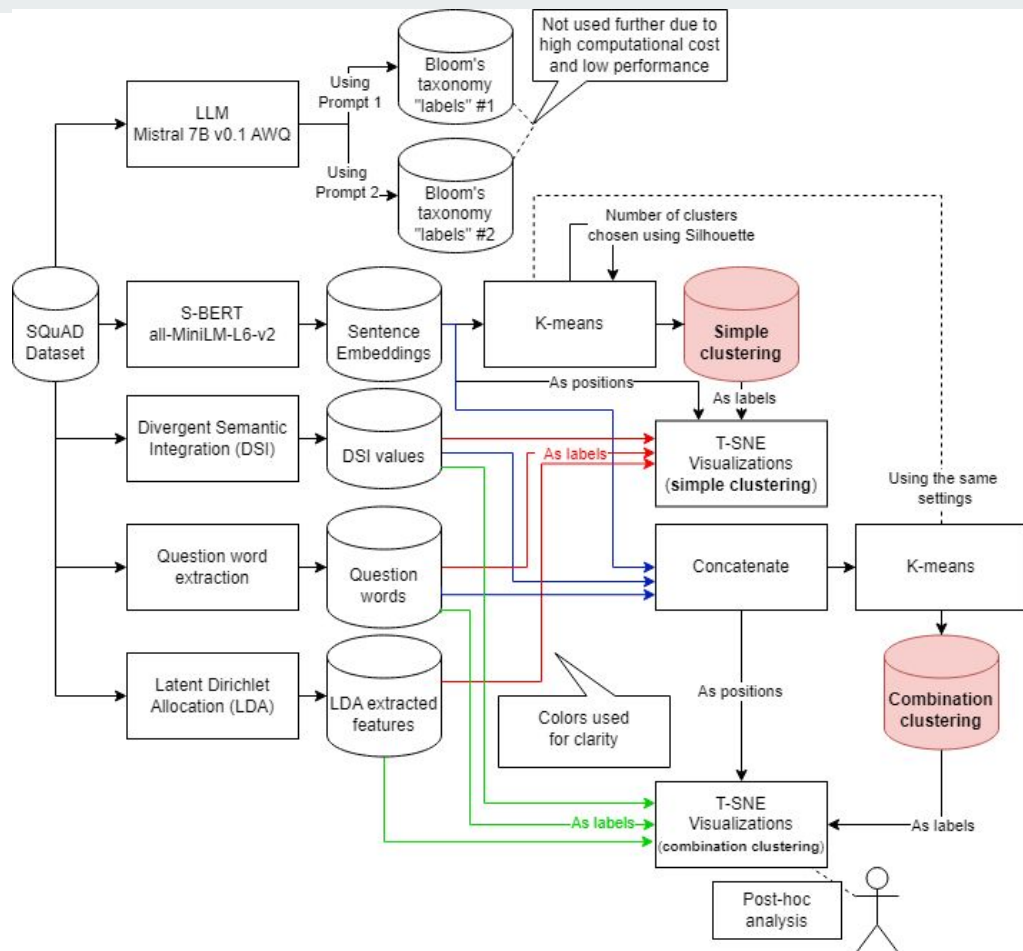
# Overview

- *OK, but what did we actually do?*
- This diagram got a bit more complicated
- While working on the project, it was clear to us...
- But not for anyone else
  - The point of reviews



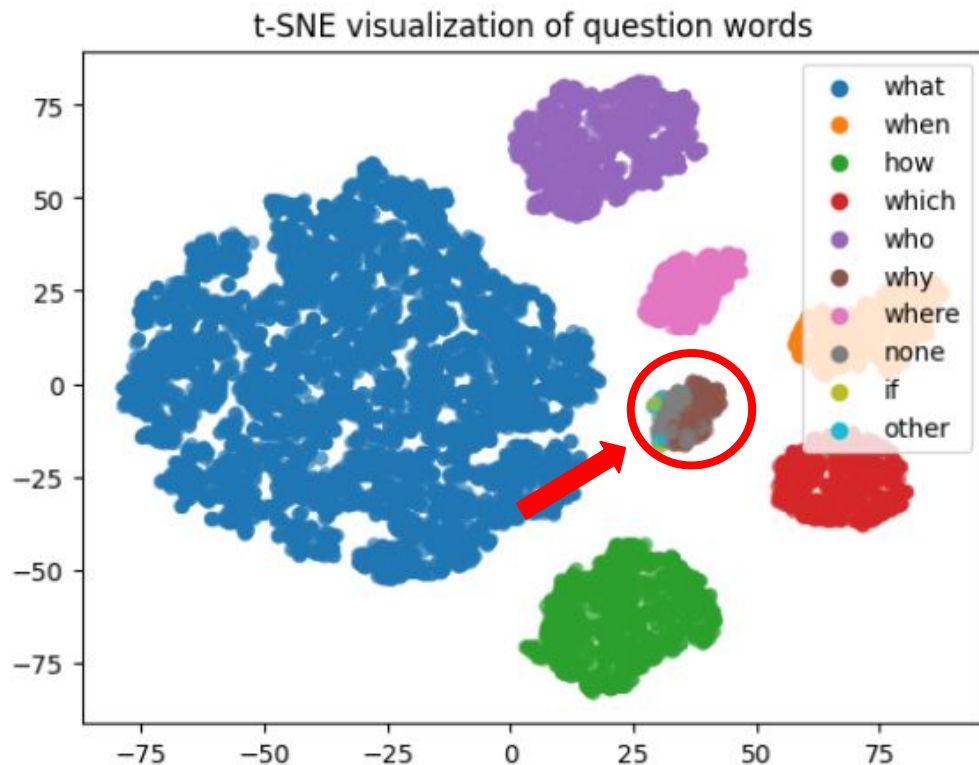
# New diagram

- Fair to say it was needed



## Our contribution

- Explored the data from the perspective of the problem
- We found what worked
- ... and what didn't
- Found *something* interesting



# Lessons learned

- LLMs like Beyoncé





# Actual lessons learned

- LLMs aren't always the best solution to a vague problem
  - Mostly a cost / effectiveness trade-off
  - Still worth exploring as an option
- Sentence embeddings roughly model the topic
- Modelling structure is much more difficult than modelling the topic
  - Especially when trying to model the structure based on the topic...
- More careful selection and preparation of data
- Topic modeling is especially hard on short text data





# Technical lessons learned

- LLMs are large
  - Model quantized to 4 bits
  - Barely fit in memory
- The Python k-modes package doesn't work on large data
- Preprocessing on question data is not as straightforward
  - Certain stop-words are significant for the meaning of the question



## Future works

- Better suited datasets
  - Mix of complex and simple questions
  - Some labels – semi-supervised learning
- Method refinements
  - More statistically sound clustering
  - Analysis using multi-dimensional scaling instead of t-SNE
- More rigorously study structure / topic relationships



# Thank you for your attention

Any questions?