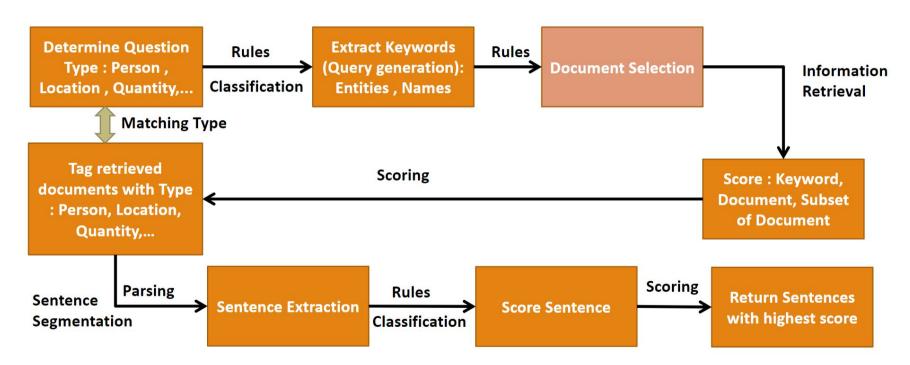
Question and Answer System

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Methodology



Pre-processing step

- Transform the corpus into a document-term matrix
 - Each document is a news article
 - 35898 entries in total
- Store the matrix into Elasticsearch database
 - Index by date + article number of the day
 - Localhost:9200 to access it

Question Classification

- Automatic classifier approach
 - How similar the input question is to all four sample questions
 - Use the maximum cosine-similarity score to determine the type
 - Sample questions are:
 - Question 1 = "which companies went bankrupt in month x of year y?"
 - Question 2 = "what affects gdp?"
 - Question 3 = "what percentage % of drop or increase change in gdp is associated with results from this property consumption consumer spending government investment imports exports foreign trade?"
 - Question 4 = "who is the ceo of company x?"

Query Formation and Document Retrieval

- Use "Query string" type for Elasticsearch
- Queries are usually keywords
 - For example, for company's ceo name, the query will be "CEO and XXX(company)"
- Use elasticsearch to retrieve top candidates of the document

Answer generation

- Select candidate sentences based on keyword-filtering
- Use NER to tag nouns
 - For CEO, looking for Person
 - For Bankruptcy, looking for Organization and GPE (geopolitical entity)
- For GDP reasons, tf-idf for terms are calculated as NER will not work
- For the impact of the specific reason for GDP, regex is used
 - "\d+(?:\.\d+)?(?:%| percent(?:(?:age|ile)? points)?)"
 - Min value in the list of percentages is used because some extreme large percentages are not for the specific reason
 - For example, "40% of China's GDP came from Shanghai" will be extracted
 - Taking average will be inaccurate as these extreme values will shift the true value to a bigger one

Interface

- Use the function "answer_the_question"
- If downloaded as ".py" file, then can use command line to enter the function

Business Insights

- Rule-based approaches are easier to execute
- Factual questions are easier to answer, and yield accurate results
 - Even if we don't have prior knowledge about the topic
- Closed-domain allows the search to be quicker
- If datasets are downloaded locally, no web connection is necessary for answering the questions

Further steps

- Based on this corpus
 - Other factual questions can be developed for analysis
 - Market sentiment analysis can also be analyzed from the text
- If possible, enlarge the corpus
 - To get a broader range of questions answered
- Get a training dataset of questions
 - To formally train the dataset in order to classify the input string more accurately
- A better interface
 - Use Python GUI packages to make it more user-friendly to input the question