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Identifying Sentences with Recipe information with Natural Language Processing

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1. BACKGROUND

Natural language processing (NLP) is a branch of artificial intelligence that helps computers understand, interpret and manipulate human language. NLP software is used to translate text, identify spam in your email inbox, and identify relevant information online when using a search engine.

For a frame of reference, when reading text from another language, we may recognize letters, numbers, or punctuation, yet the underlying meaning is a complete mystery to us. With the help of NLP, computers can answer questions much more accurately, summarize entire documents, and quantify the connotation of an article.

The goal of this project is to determine whether a sentence contains a recipe step. We identify four features from the text: It's semantics, digits, chemicals, and wikification possibilities (in-text entities that are found on Wikipedia pages). These features are then used to train the Naïve Bayes Classifier. Afterwards the classifier was tested on documents within the same field.

3. RESULTS

Naïve Bayes Classifier Results

	Precision	Recall	F1-Score	Support
Positive	0.75	0.11	0.19	27
Negative	0.53	0.96	0.68	28
Average	0.64	0.55	0.44	55

- Positives are examples of sentences that contain recipe information
- Negatives are examples of sentences that do not contain recipe information
- The Support Column informs us of how many sentences were used for this project

Identifying Ingredients Results

	Precision	Recall	Total
Positives	1	0.64286	14

- All ingredients were correctly identified
- There exists redundant repetition in the output that was not accounted for.
- F1-Score is 0.7826

20ml

20ml

20_ml

20_ml

20_ml

20_ml

of

The most desirable Quadgram with no ovelap

Less desirable Quadgrams with overlap

2. METHODS

dissolved in

dissolved in

dissolved in

dissolved in

dissolved in

dissolved in

was

was

was

was

was

was

- Break the text into sentences and then extract every word in the sentence.
- Tag the words with its Part of Speech identifier
- Begin extracting features:
 - 1. Chemical names and formulas via PubChemPy API
 - 2. Digits (in terms of measurements such as 1 ml, 1 mg, and 1 kg)
 - 3. Wikification possibilities
- Manually Annotate the text and identify the recipe contain sentences for training/testing purposes.
- Train the Naïve Bayes Classifier.
- Test the Naïve Bayes Classifier.
- Create additional algorithms to create Quadgrams.
- Identify recipes via positioning.

TRAIN CLASSIFIER SEPARATE EXTRACT ANNOTATE SENTENCES & TAG **FEATURES** DOCUMENTS AND TEST NOUNS, ADJECTIVES, NUMBERS, ESTING DAT RESULTS SETS FEATURES INCLUDE: NLTK.SENT_TOKENIZE() NLTK.WORD_TOKENIZE() MANUALLY LABEL SENTENCES AS PART OF SPEECH TAGS NAIVE BAYES CLASSIFIER "RECIPE CONTAINING SENTENCES" NLTK.POS TAG() MEASUREMENTS Typical Annotations within the Papers followed by 20ml deionized water dissolved in was

deionized water

deionized water

deionized water

deionized water

deionized water

deionized water

followed by

followed by

followed by

followed by

followed by

followed by

4. SUMMARY/INTERPRETATION

- Detecting positives worked 75% of the time.
- Collectively there are too many false positives being detected, henceforth the low recall rate of 11%
- There is no correlation between wikification possibilities and the recipe containing sentences.
- 75% of the sentences have wikification possibilities
- The average wikification possibilities are 3 per sentence
- Our first attempts at identifying ingredient only were fairly successful due to the high F1-Score

5. FUTURE WORK

For the future of this project we will look closer for the recipe components of the manufactured materials. This way, we may continue improving the previously developed algorithms to identify digits, chemicals, and semantics of the text. We will also use deep learning for labeling the role of an ingredient for recipe assembly. We should also continue to annotate more documents to see if the accuracy of this work improves. Typically, we see other projects of this nature use thousands of examples to train classifiers.

Alongside this we may also exclude the wikification feature due to it's limited utility in identifying the recipe steps. We may also expand the measurement category to include Moles, molarity, and acidity (PH).

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