

Natural Language Processing (NLP) in Stakeholder & Consumer Insight Research to complement interpretivist research approach

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Background



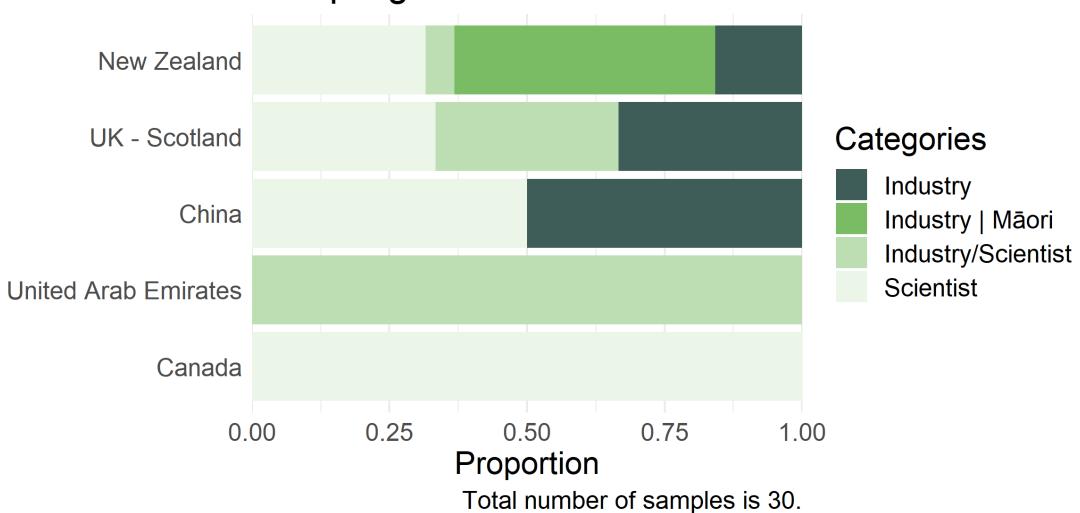
» High level research question:

How do expert stakeholders (local and international, in both science and industry) currently understand and make sense of Controlled Environment Agriculture (CEA) as a future food technology?

» What NLP can help to complement qualitative research?



Sampling distribution in five countries.



A snapshot of the raw data



Speaker 1 01:10	Greetings. Could you introduce your background a little bit?
Speaker 2 02:00	I graduated from XXX in XXX. My current interest My current role at XXX is XXX.
Speaker 1 07:30	Thank you. What is your opinion on topic A?
Speaker 2 08:00	Er, I think overall that I have a mixed feeling about topic A. One side is that Another thing is that I think I know that From my experience, topic A has
Speaker 1 15:10	Really interesting insights you just mentions. I think you are most right on xxx. Could you extend on the second point a bit more?
Speaker 2 16:10	Yeah, for sure. What I mean by xxx is that So the situation will be complicated if we assume that the customer have some perceptions on I think it is hard to predict what future will look like.
Speaker 3 18:55	Totally agreed. Predicting future is very very difficult. So you suggest that xxx. Why would you think that is the case?
Speaker 1 19:30	Em a couple of points there. First, Then, I think the last one maybe around

Some challenges in the pre-processing



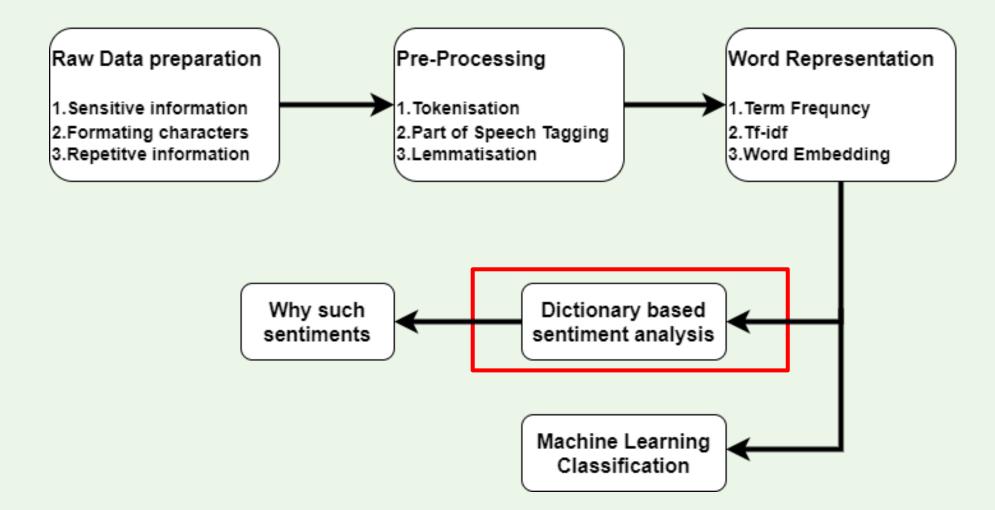
- » Sensitive information that can reveal personal identity
- Time stamps
- Abundance of stop words (words we use filling the gaps)
- Mixture of the question and

 - Not necessary collaborate
 Interview Let's questions to clarify the question.

Workflow

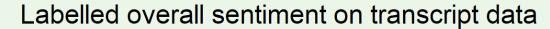


Sentiment labelling

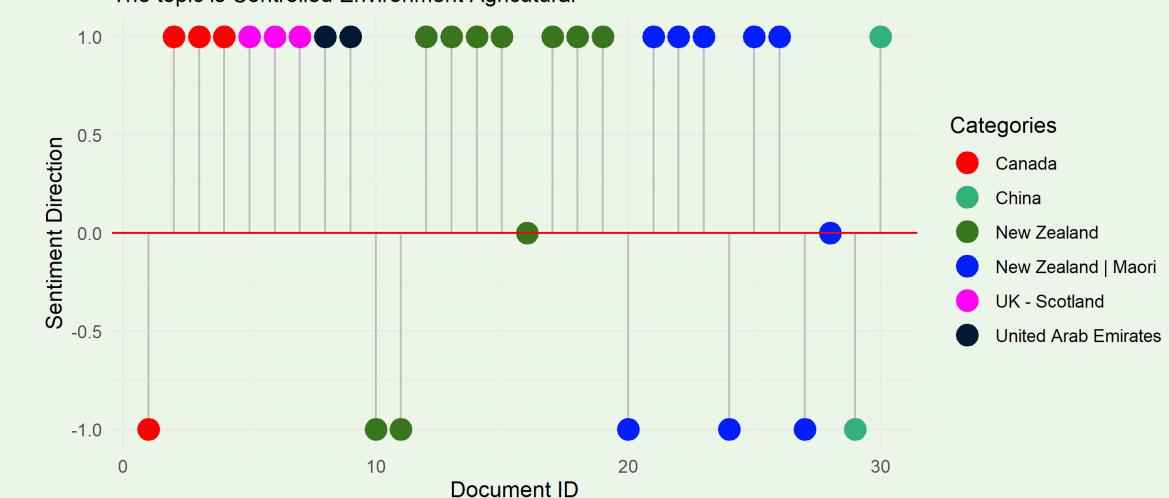


Labelled Data





The topic is Controlled Environment Agricutural



Existing works – term frequency approach



» SentimentAnalysis – 4 built-in dictionaries

» Syuzhet – 4 built-in dictionaries

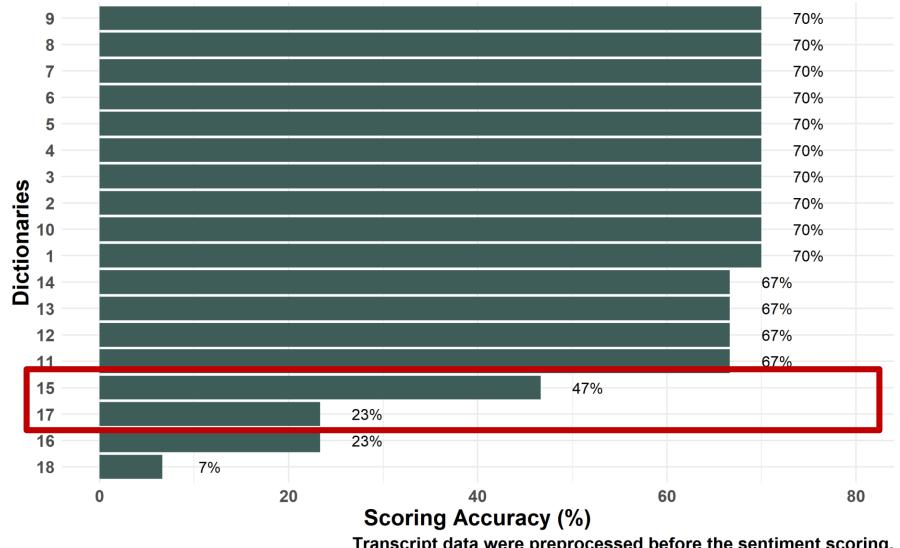
» Sentimentr – 10 built-in dictionaries

» Sentence level scoring → aggregate to document level.

Dictionary scoring



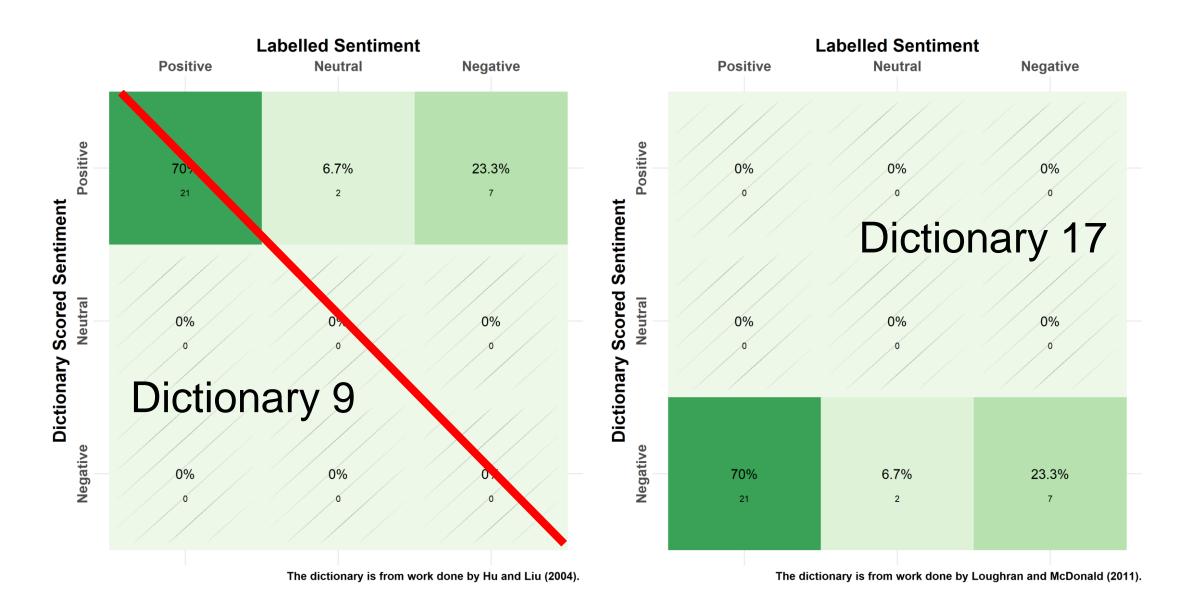
Dictionary-based sentiment analysis on transcript data



Transcript data were preprocessed before the sentiment scoring.

Acceptable vs Poor





Recap dictionary based approach



- » 14 out of 18 built in dictionaries scored ~70% matched the labelled sentiment
- » Pre-built dictionaries are good at scoring positive sentiment
- » Same dictionary with different processing techniques yielded different scores



Word representation – TF-IDF



council end site customer cellfit gate unintended

overall
versedomestic
citruskiwifruit suitable
welcome artificialcommodity
smart

gene mutation hygiene bypass twig**flower** reliant glasshouse biology sugar

Overseas industry

Overseas industry/scientist

Overseas scientist

tray POI I ution
equipment
core vegetable basil
client starter

advantage
suitable affordable
strawberry customer
quinoapotato
glasshouse transportation

greenhouse ordinaryworker smart vegetable perception pollution facility controllable province

New Zealand industry | Māori

methodology utilization teach kumarahua macadamia hectarekaimanuka garden

What's next

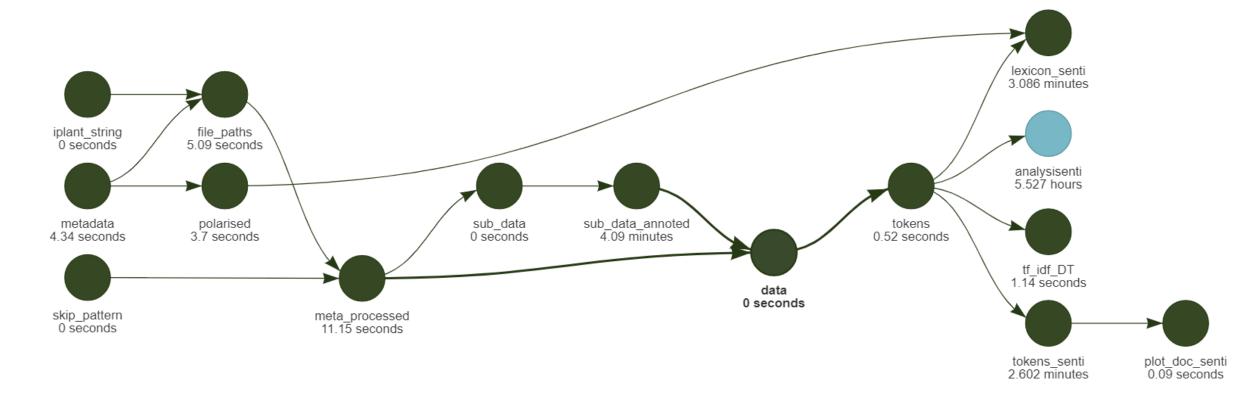


- » Dive into the sentence and word level analysis.
- » Develop context-specific sentiment dictionary.
- » Explore the application of word embedding in sentiment analysis.
- » Develop classification models.

Tools for reproducibility



- renv
- usethis
- targets





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