

Detecting personal attributes through analyzing online forums

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Introduction

Personal attributes refers to specific ways of thinking/feeling/behavior that individuals have, formed from the interaction between a person’s innate tendencies and their need to adapt to their environment. An example of such attributes include various types of problems individuals face in their lives such as particular relationship dynamics as well as reasons for partaking in various social activities.

In the case of understanding problems that individuals face, it might be great value to categorize individual experiences such that individuals can feel that they are not alone in their experiences and can learn how others have attempted to mitigate similar problems (Bandura, 1977).

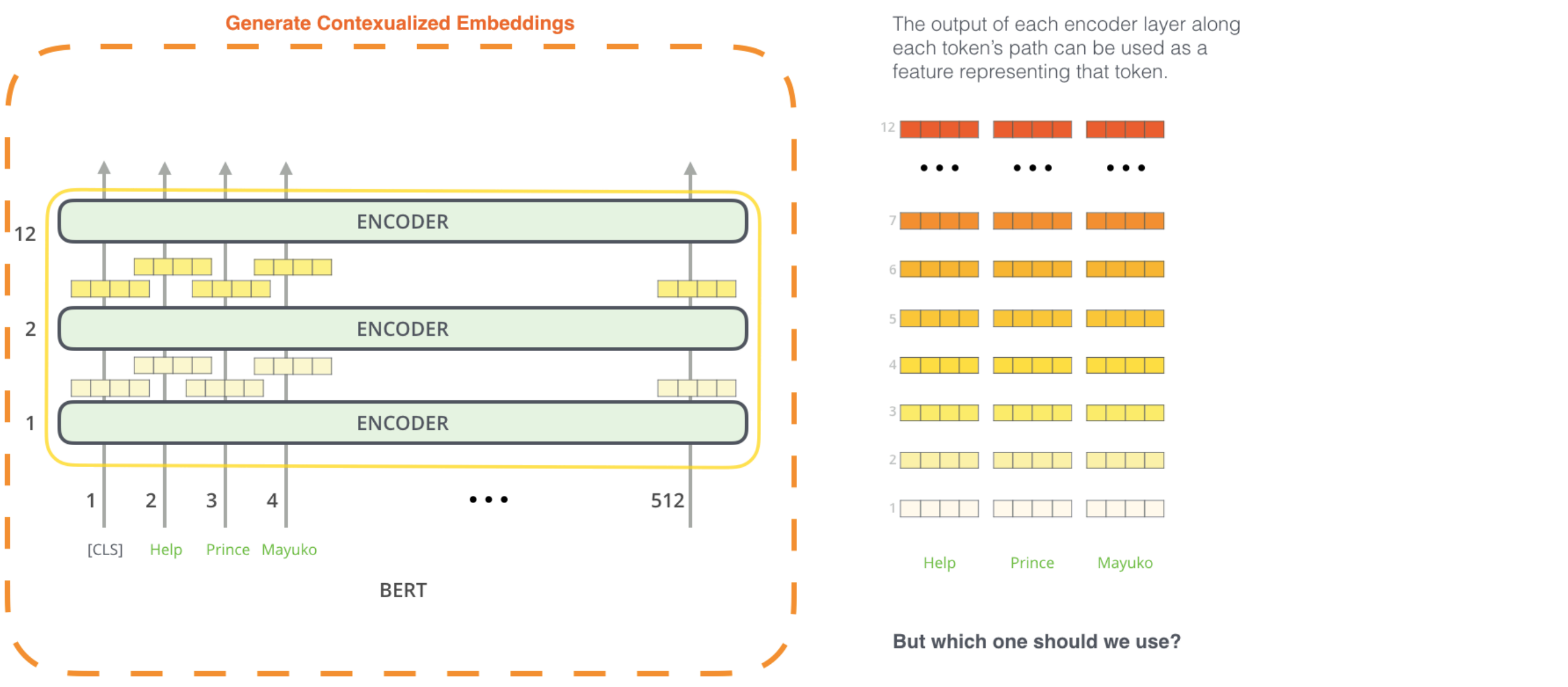
Human judges have long been able to classify many of such attributes (McAdams, Hoffman, Mansfield & Day, 1996; Blagov & Singer, 2004) but automated textual analysis (NLP) techniques have only been successful at a few (Bosson, Swann & Pennebaker, 2000). This is possibly due to the extensive use of the word-count approach (Pennebaker, Boyd, Jordan & Blackburn, 2015), which has largely been replaced by advanced language models in many NLP tasks.

These language models have achieved SoTA performance in many common NLP tasks (Question Answering, Translation etc), demonstrating that they have achieved a great extent of semantic and syntactic ‘understanding’. This suggests that these language models might be capable of interpreting and categorizing personal attributes from writing in a similar manner as an experienced counselor/psychotherapist.

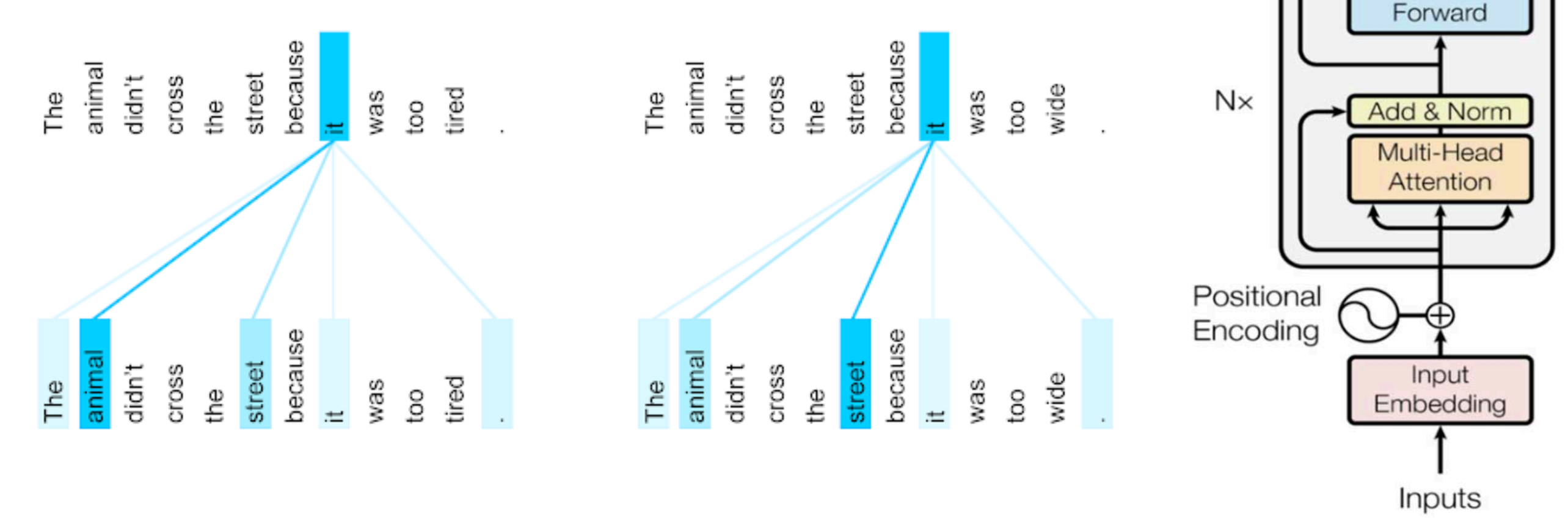
In this research, we will seek to apply recent advances in language models, particularly the BERT model to understand personal attributes of Reddit authors who have sought for advice for their personal situations.

Methods

1. Mining >8000 posts from r/Advice of Reddit
2. Generate 768-length vector embedding of each post using Google BERT

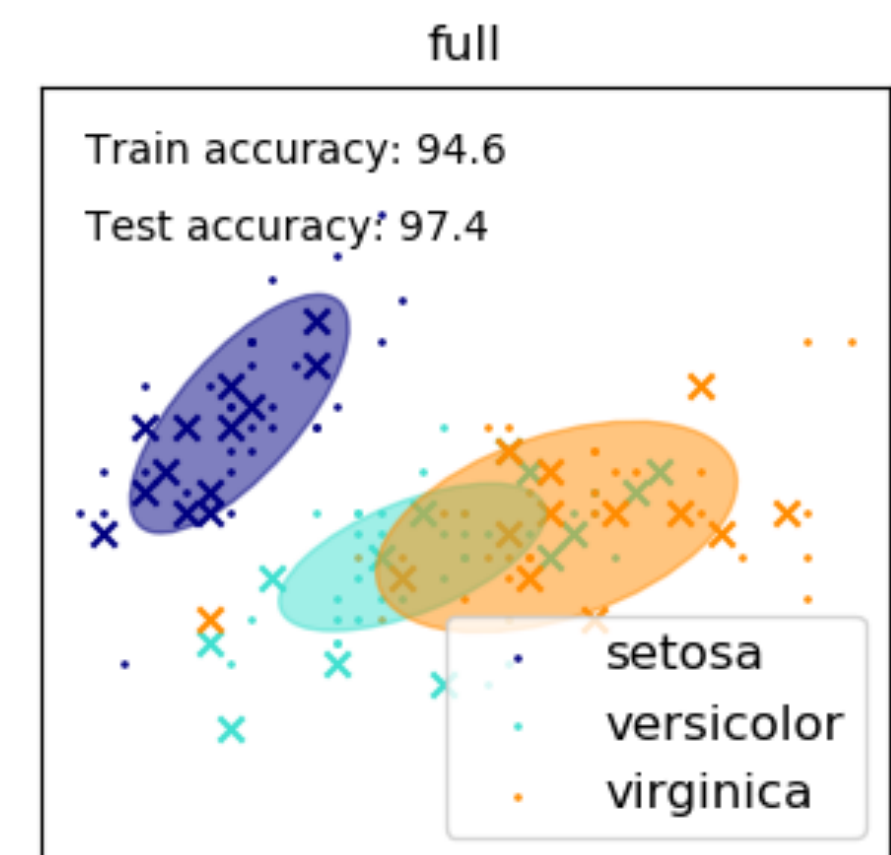


BERT Architecture by Alammar, Jay (2018). The Illustrated Transformer [Blog post]. Retrieved from <https://jalammr.github.io/illustrated-transformer/>



Detail of encoder: Transformer aka ‘Attention’ by Google AI Blog

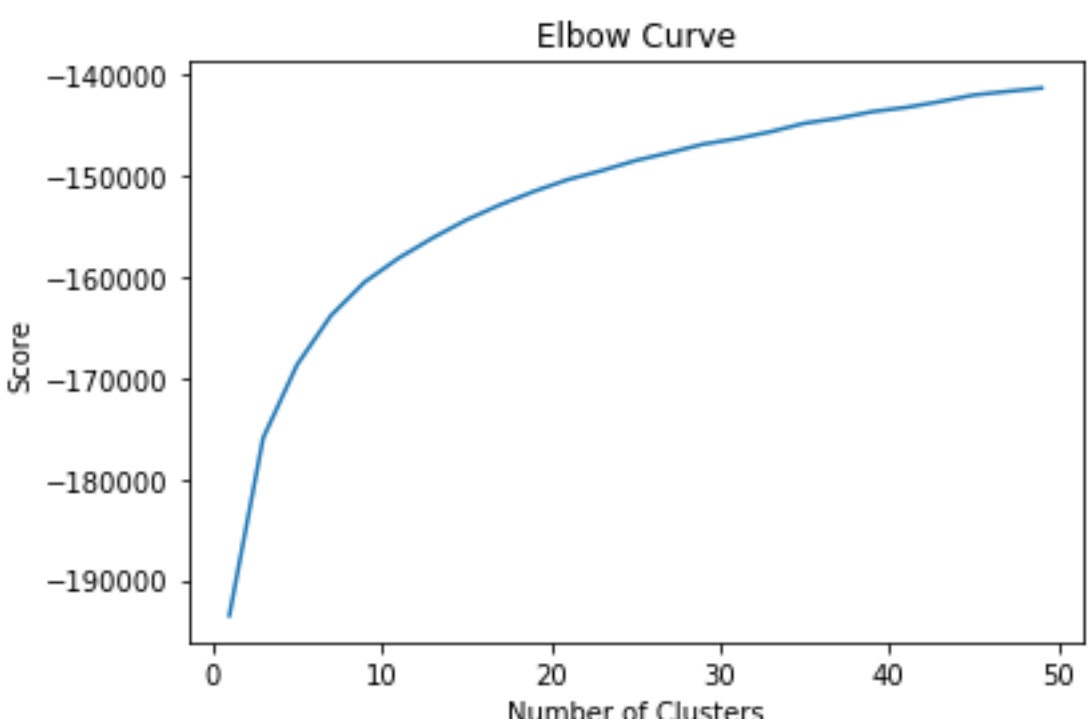
3. Clustering vector embedding using Gaussian Mixture Models



How do Gaussian Mixtures Models work? by Scikit Learn

Relative to KMeans

- Variance in each dimension can be different: ‘ellipsis’ rather than ‘sphere’
- This can detect particular personality attributes that depends on covariance between vector elements
- More even cluster sizes

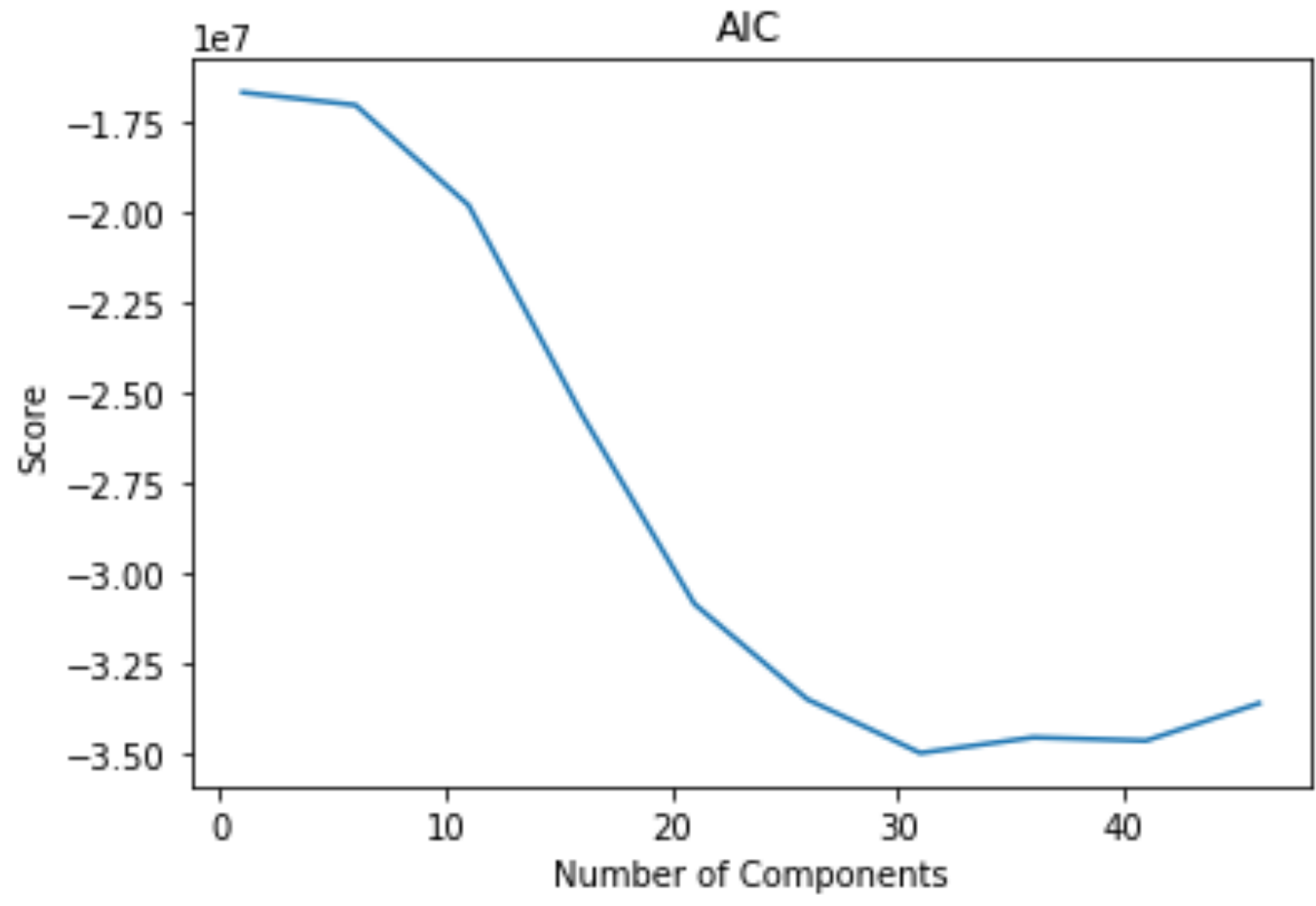


Sum of distance to centroids for KMeans models with varying number of clusters (higher is better). Note lack of distinct elbow-shaped bent

Results

Titles of posts with vectors most similar to “How do I get Vaccinated as a Minor?”

1. Hi, I'm 13, unvaccinated, and damn it seems like I have a lot of stuff to catch up on. What do I do?
2. Lost Health Insurance at 21. Advice?
3. How do I tell my parents I don't believe in god?
4. Should I get vaccinated?
5. My religion is taking away my friends



Akaike Information Criteria for Gaussian Mixture Models with variable number of components (lower is better). Note: plateau at 30 components.

Labels for each component of the Gaussian Mixture Model (Human-labelled)

- component 0: toxic friendships
- component 1: some form of abuse/harassment by family/friends (author are usually early adolescents)
- component 2: best/girl/boy friend issues with some form harassment
- component 3: job related but more short term and minor
- component 4: girl/boyfriend --> breaking up
- component 5: biological related stuff such as illness as well as use of various substances but less severe
- component 6: academics/work related but more long term, not knowing how to make a decision
- component 7: House/neighbour related
- component 8: mental health but more severe
- component 9: disappointment with best/girl/boy friend
- component 10: supporting best/girl/boy friend
- component 11: sexual preferences and behaviors
- component 12: dilemma between what the person wants and what significant others including family wants
- component 13: issues with trying to get into a relationship
- component 14: befriending others
- component 15: embarrassment related
- component 16: tensions at work/school
- component 17: self-improvement
- component 18: harassment by people they don't know
- component 19: mental health but very severe such as PTSD, depression and suicide
- component 20: adolescents unsure about future
- component 21: tensions with boy/girlfriend but not yet breaking up
- component 22: gaming, videos, internet
- component 23: family tensions
- component 24: abuse in the family (author are usually adults)
- component 25: teenage issues with a focus on relationship
- component 26: Transitioning from high school --> university or university to job
- component 27: teenage issues with a focus on biological aspects
- component 28: Issues relating to money
- component 29: motivation (or the lack thereof) for high school/university

Discussion

Preliminary experiments suggest that language models can to a large extent categorize personal attributes, at least in terms of personal problems. However, the occurrence of misclassification suggest the need for more investigation to be conducted in optimizing their perform. These are some of the changes we hope to experiment with in the short term.

Alternative embedding methods

We hope to experiment with alternative language models including OpenAI’s GPT-2, which is architecturally similar to BERT but pre-trained on a different corpus. In addition, we will also try traditional word count models that emphasize different semantic aspects such as SPLICE (Farnadi et al., 2016), which focuses on author self-evaluation and MRC psycholinguistic database (Coltheart, 1981) which provides semantic, phonological and syntactic information of words

Alternative clustering methods

We hope to experiment with density-based clustering such as DBSCAN/ OPTICS as well as tree-based clustering such as Birch (in place of Gaussian Mixture Models).

Aggregate multiple posts by the same author

This might be able to remove the influence of writing style and other personal attribute-irrelevant ‘noise’ by establishing a baseline from which the post in question can be removed of.

Use of supervised or semi-supervised learning

Our current approach is wholly unsupervised. This has the advantage of discovering features that are not intuitive to humans. However, a supervised/semisupervised approach might be more accurate because the existence of a clear training objective can lead to more optimized model selection. This can be done by mining for posts from specific subreddits such as r/family or r/ breakups. Alternatively, we can use existing coding schemes for humans as well as their accompanying examples and see if language models can interpret them adequately (Demorest, Crits-Christoph, Hatch & Luborsky, 1999; Hy & Loevinger, 1996; McClelland & Koestner, 1992; McAdams, 1992).

