Identifying Clickbait: Multi-Layer Regular Expression Filter

University of Minnesota Jack Bartels LING 5801 30 June 2019

Executive Summary

While the definition of clickbait is hard to pin down, there is at least a general agreement that it is content designed to make you click on it. As the internet grows through constant uploading of new content, the ability for computers to be able to identify clickbait is becoming increasingly relevant.

Because of this, I created a program in python that uses regular expressions to attempt to identify clickbait titles. The input data is pulled from YouTube's Trending feed, which is a collection of its most popular videos. The titles are then ran through a three-layer filter. The layers filter based on keywords, phrases, and lastly formats. YouTube seemed like a good platform choice due to the variety of cultural influences and the frequent new content, which is ideal for maximizing testing data.

This project is a perfect opportunity to study a few key areas of linguistics. The primary areas pertaining to clickbait are socio-linguistics, lexical variety, and syntax. The intersection of culture and linguistics makes this a particularly interesting and difficult task.

The goal of the program is to identify as many clickbait titles as possible while minimizing false positives. In other words, it is important for this program (if it is to be at all useful) to have a high recall and precision, which are then combined into an overall F1 score.

When testing, I found that formats were by far the most productive filter with a good precision of 80%. The keywords filter had a low recall and lowest precision due to its proneness to overcapture. The phrases filter had the lowest recall but it was still very useful in that it had perfect precision, raising the F1 score. Overall, the program was decently precise, but definitely fell short in terms of recall.

There are a number of reasons why the program may have been unable to identify many of the clickbait titles, but most likely it comes down to two main factors: (1) there is a lot of subjectivity involved in the categorization of clickbait or not; and (2) the program was only exposed to a few hundreds titles of input data, which could be way fewer than is necessary to arrive at a more accurate result.

Overall, it seems that this project would be better suited to using machine learning and large input datasets in order to achieve better results. Indeed many researchers and computational linguists are doing just that.

Introduction

Clickbait; click here to find out what it is! We hear this term on a regular basis, typically when it comes to browsing the internet. Simply put: clickbait is any content that is formatted in a way that is primarily meant to draw the attention of the user. This phenomenon has become more and more common as an increasing number of platforms offer the ability to monetize content through advertising, i.e. clicks. A lot of content creators avoid utilizing clickbait because of the stigma around exploiting their audiences. Others, though, are shameless in their boisterous, cliff-hanging titles and colorful, coming-right-at-you video previews.

While clickbait, on its surface, seems to be a fairly harmless and unavoidable byproduct of the present, view-driven state of the internet, there are a number of reasons the capability to identify it can be really useful. One example is from the perspective of a business or marketing organization that wants to determine how effective clickbait is by comparing viewership of clickbait content versus non-clickbait content. Another example is a user's ability to filter out such content if they so choose. Whether it perpetuates it or stifles it, effective clickbait identification would change the current landscape of the internet.

The Program

As a user of the internet myself, I am frequently bothered by the amount of clickbait content that I have to scroll past. This seemed like the perfect opportunity to explore how feasible it would be to use computational linguistics to capture patterns typical in clickbait titles with decent accuracy. My program is written in Python and it strips titles from YouTube videos. The HTML scraping to strip the titles was achieved by using the lxml and requests libraries. YouTube button titles and URLs were filtered out of the results manually because regular expressions had difficulties isolating the two different kinds of titles solely from HTML tags.

The program has a three-layered filter. If any of the titles to match any of the filter items, the title is flagged as clickbait. The three components of the identifier are keywords, phrases, and what I have generally labeled as 'formats', which are other patterns captured through the use of regular expressions. The filter items can be freely added and removed until testing performance is deemed adequate. Then, the program compares its performance to the human judgement data provided. Typically, a gold standard dataset is generated and then split into a testing and evaluation subset. For this program, I decided to instead opt for generating testing data by stripping titles directly from YouTube's Trending feed and then refining my filters manually,

until satisfied with their performance. This has a principle drawback: it introduces more subjectivity. However, I still decided to adopt this method because of its massive advantage: the ability to test live, dynamic data at any time over the course of the project.

Being exposed to so much more testing data, I believe that the performance of the filters will naturally have a greater ability to generalize effectively beyond testing data. For creating the evaluation dataset, I reserved a day each week in May to not draw testing data from (or around to prevent overlap), but to instead collect into a list for other people to categorize, given only the definition of clickbait provided above, and then aggregated those responses.

The Platform

YouTube is the internet's single largest and most popular user-created video platform. It has content of all varieties and provides, I think, a good representation of the internet as a whole. I chose to build my program using this platform because of the diversity of topics, the frequency of updates to its trending feed that displays popular videos, and the ease of dynamically scraping and stripping its HTML using Python and its libraries. YouTube also provides a challenge when it comes to regular expressions due to the fact that individuals decide what the titles are. The result is that titles have many different capitalization mixes, punctuation formats, and even typos.

There was always the option to pull titles from multiple platforms, but I thought it would serve an additional purpose to focus on only one. That way, it can be tested if the program that I created using YouTube input data is able to generalize well to other platforms. Whether it can or not, it reveals something about how we use language the same or differently depending on which platform we are publishing to. Another interesting aspect of YouTube is that clickbait often has to do with the preview image provided for videos. When given the titles, respondents were not given the accompanying images, evaluating the titles based on the text alone. A possible future project could be to test the effectiveness of the clickbait title identifier, in the context of the image and title presented as a whole.

Linguistics

This project relates to linguistics in a number of ways. Primarily, it deals with sociolinguistics, lexical variations, and syntax. Socio-linguistics deals with how linguistics and socio-cultural factors interact. Many of the titles encountered on YouTube trending have to do with music, food, sports, gaming, events, and other cultural topics. As a result, successfully

identifying the ways in which these cultural factors affect whether or not a title is considered clickbait needs to be examined and included in the filters. Even within the same language, there are numerous lexical variations depending on location, age, education, personality, etc. The program needs to be able to capture these variations otherwise it will suffer in its recall abilities. Additionally, this project is heavily influenced by syntax.

A large part of this class has been focused on ambiguity, regular expressions, and thinking beyond the word level when it comes to computational analysis of language. It is for this exact reason that I decided to make my filter multi-layered. There are some keywords that exist almost exclusively in clickbait titles, likewise with phrases. But targeting specific words and groups of words in this way is very prone to leading to false positives, over-capturing. Because of this, the third layer is focused solely on common formats used in clickbait, which are captured using regular expressions. As a result, the program takes into consideration each of the primary linguistic factors at play.

Approaching the Problem

In order to begin approaching the problem of accurately identifying clickbait, the manner of evaluation needs to be considered. The program will be given an F1 score based on its recall, the ability to identify clickbait titles, and precision, the proportion of identified titles that were indeed clickbait. With this in mind, the goal is to select filter items that successfully capture the majority of clickbait titles without being so broad that they misidentify non-clickbait titles. The importance of accuracy is relevant in the context of the application. For instance, if the identifier is being used to filter out clickbait content that is displayed to the user, the accuracy needs to be very high, especially the precision moreso than recall. The reason is that it would be unfair to content creators who don't use clickbait to get captured by the filter by mistake. For our purposes, an F1 performance above 80% would be sufficient as a proof of concept.

Here are the final filters that were utilized:

Keywords	Phrases	Formats
biggest	here's why	'^Guess .*'
largest	you never knew	'^[0-9]+ (?!in).*'
mind-blowing	won't believe	'^(? I)[A-Z]+[(A-Z 0-9)]+.*'</td
worst	weird trick	·.***.* [,]
amazing	see this	'.*\?!+.* [,]
never	watch this	·.*!\?+.* [,]
click	find out	'.*\?{3,}.* [,]
woah	life changing	'.*\!{3,}.* [,]
WOW	need to see	'.*[A-Z]+ [A-Z]+ [A-Z]+.*'
surprising	will make you	
omg	this is what happens	
crazy	oh my god	
	this is nuts	

The filter items themselves give some linguistic insights. As can be seen, most of the keywords are exclamatory and superlative. When it comes to phrases, they tend to be directed at the user, with a common theme of expressing that they ought to click in order to gain some previously unknown information. Another theme common in the phrase filter items is the imperative. There are many phrases that plainly instruct the user to click the video. The format filter items clearly had three major themes: Numbers, punctuation, and capital letters. Because the format filters are the most productive in terms of recall, that means that those themes are therefore the primary tools used by people when crafting clickbait titles.

Results

The gold standard list was never consulted in order for evaluation to reflect the program's ability to generalize to other data. In order to get a clear picture of the extent to which each filter layer contributes to the overall performance of the identifier, results are split up into four categories: keywords filter only, phrases filter only, formats filter only, and all filters combined.

Each result is broken down into the recall, precision, and F1 score for each test, which are displayed in a percentage form rounded to the nearest tenth of a percent:

Keywords Only:

Recall: 10.6%

Precision: 69.2%

F1 Score: 18.4%

Phrases Only:

Recall: 5.9%

Precision: 100.0%

F1 Score: 11.1%

Formats Only:

Recall: 42.4%

Precision: 80.0%

F1 Score: 55.4%

Combined:

Recall: 52.9%

Precision: 77.6%

F1 Score: 62.9%

What It Reveals

What the results reveal about human use of language when it comes to clickbait is that we tend to utilize specific styles of formatting rather than particular words or phrases in order to create clickbait titles. While the program fell short in identifying all of the clickbait titles, it had decent precision. Most revealing was the fact that the formats filter captured almost all of the clickbait titles identified, with a relatively high precision. The phrases filter had the lowest recall, in other words it was the least productive in helping to find clickbait titles, but it did so with perfect precision. These results show that the primary focuses for filtering should be on phrases, for its precision, and on formats, for its recall. Given that the filters were refined after going through hundreds of input titles and making adjustments and it still had low overall performance when tested on the gold standard means that either my judgement differed greatly from that of the respondents or that there are just so many potential keywords, phrases, and formats for clickbait that a lot more input data is needed—or both. This is where the limitations of subjectivity and lack of machine learning becoming most apparent.

Limitations

Right away, studying clickbait has a large limiting factor: its definition. While I came up with a standard definition to provide to respondents, what actually constitutes clickbait is not so clear cut and varies from person to person. In other words, there is a lot of subjectivity involved which limits the effectiveness of using a single standard filter. As mentioned before, using my own subjectivity in place of a gold standard subset for testing is also an opportunity for more divergence in the agreed upon categorization of clickbait.

On a similar note, the assumption that a title is either clickbait or not could itself be flawed. One could argue that the determination of whether or not something fits the criteria of clickbait is not a black and white issue but rather a sliding scale. For titles that are in a gray area, respondents may be forced to make a rather random decision. In other words, there is ambiguity involved. Such titles could decrease the effectiveness of the identifier by introducing overly general filter items or by not containing any attributes characteristic of typical clickbait.

Another limitation is the need to manually evaluate test data and adjust filters. Not only does this mean that I need to make decisions about each title, I also need to determine from small datasets what tends to be present in clickbait that isn't present in other titles. There is no way for the program to dynamically improve, only manually.

Potential Improvements

Given the time, technological, and knowledge constraints of this project, there are quite a few ways that it could be improved, including even entirely different approaches. Some improvements include sliding scale data, implementing a means of testing frequency and particular filter items individually, and using deep learning.

Sliding scale data rather than boolean decisions about whether a title is clickbait or not. Having that additional data could assist in choosing filter items that better balance recall and precision. It would be useful to be able to compare the frequency of keywords, phrases, and formats in clickbait vs non-clickbait titles. Items that are common in clickbait titles and uncommon in non-clickbait titles could easily be selected as new filter items. With a large enough processed dataset, individual filter items could be evaluated for their recall and precision to determine whether or not they are worth implementing. Finally, the most natural next step for a project like this is to have a program that can analyze large quantities of data and self improve.

Deep learning could do a much better job of creating a clickbait identifier than a human doing it manually, it would only be a matter of how much human processed data it could be fed.

In order to implement these improvements, more resources would be required; namely: time, expertise, input data, and computational power. However, even the current implementation could likely be further improved with more manual adjustment after testing on more data.

Related Projects

In one similar project, Lopez-Sanchez et al. identified one of the primary challenges of making a clickbait identifier as, "the problem of adaptibility. Due to the subjective nature of clickbaits, a single headline may be perceived differently by users of different interests or criteria" (2968). In order to overcome this challenge, they suggest using "Case-Based Reasoning". The basic idea behind this method is to identify and utilize the uniqueness of each user. Any data gathered about the user can be used to predict what they may or may not consider to be clickbait. They achieve this by using frequency analysis and neural networks.

Another related project is by Phillipe Thomas. Thomas used a fairly typical separation of testing and evaluation corpora. Each set was labeled as clickbait or not and then that data was fed into various kinds of neural networks. What is interesting about his work is that he also chose to explore the effect of incorporating the accompanying image. "As the annotation process was supported by the image information, we assume that the teaser images might be helpful to predict clickbait relevance of a given message" (Thomas 2). This prediction is in line with my hypothesis stated earlier in the 'Potential Improvements' section. However, Thomas was unable to get any helpful performance out of this addition, at least using his implementation.

Conclusion

Identifying patterns in human language use certainly seems to be a task better suited to modern techniques that can utilize artificial intelligence and hefty computing power. However, this project still revealed a lot about what clickbait consists of linguistically. Primarily, it showed that clickbait is mostly a matter of formatting, reinforcing the semester-long message that resolving ambiguity and identifying patterns in real world language use requires looking beyond the word and phrase level and instead considering how we can analyze how all of the components of language interact as various scales. Even though I find the results to be disappointing in terms of performance, there does seem to be a lot of room to improve this

program, even in its current implementation. There are a lot of other related projects and research being conducted out there on this and similar topics. Most of that work is being done using neural networks that analyze the frequency of patterns with massive input datasets. Even though there are a lot of challenges when it comes to something as subjective as whether or not a title is trying to exploit your attention, there does seem to be at least a certain extent to which computing can predict—or at least model—human language use in a beneficial way. Being able to identify clickbait could prove to be an extremely valuable ability that could lead to many tools and applications to be used by individuals, companies, and governments. I am excited to see the future progress that is made in this field.

Works Cited

López-Sánchez, Daniel, et al. "Hybridizing Metric Learning and Case-Based Reasoning for Adaptable Clickbait Detection." *Applied Intelligence*, vol. 48, no. 9, 2018, pp. 2967–2982.

Thomas, Philippe. "Clickbait Identification Using Neural Networks." 2017.

Appendix (Code)

```
requests
                     6 ("Yitle Clickhait") muple bise
7 goldStandard = [("Ending the Subscribe to Pewdiepie Meme", False),
8 ("Chiitan: Last Week Tonight with John Oliver (HBO)", False),
9 ("Game of Thrones | Season & Episode 4 | Preview (HBO)", False),
1 ("Game of Thrones Season & Episode 4 Preview Breakdown", False),
2 ("Putting Weird Things In An Air Fryer (TEST)", True),
3 ("Game of Thrones Season & Episode & Review and Breakdown", False),
4 ("Putting Weird Things In An Air Fryer (TEST)", True),
5 ("Game of Thrones Season & Episode & Review and Breakdown", False),
6 ("What To Do When Someone Parks in the Access Aisle", False),
6 ("What To Do When Someone Parks in the Access Aisle", False),
7 ("Game Of Thrones Season & Episode & 'The Long Night' Breakdown!", False),
8 ("Game Of Thrones Season & Episode & 'The Long Night' Breakdown!", False),
9 ("S Giant DIY Foods Challenge & How To Make The Best Avengers Endgame Pancake Art in 24 Hours", True),
9 ("Verything Is a Liability' to Kourtney Kardashian's Law Student Sister Kim", False),
10 ("Verything Is a Liability' to Kourtney Kardashian's Law Student Sister Kim", False),
11 ("Verything Is a Liability' to Kourtney Kardashian's Law Student Sister Kim", False),
12 ("When Your Best Friend Exposes Your Crush To The Entire School *SO EMBARRASSING!*", True),
13 ("12 Details In 'Game of Thrones' Season & Episode & You Might Have Missed", True),
14 ("My First Toxic Relationship", False),
15 ("Game Theory: Hard Mode is a LIE! (Sekiro Easy Mode Controversy)", True),
16 ("Game Theory: Hard Mode is a LIE! (Sekiro Easy Mode Controversy)", True),
17 ("Game Theory: Hard Mode is a LIE! (Sekiro Easy Mode Controversy)", True),
18 ("Game Theory: Hard Mode is a LIE! (Sekiro Easy Mode Controversy)", True),
19 ("Game Theory: Hard Mode is a LIE! (Sekiro Easy Mode Controversy)", True),
```

```
other | CONTAINS STRONG LANGE

of the Spider-Verse", True),

of the Spider-Verse", False),

of the Spider (Official Music Video)", False),

of Netflix", False),

HE HAVING A BABY! (Emotional)", True),

Dude Perfect", False),

of Spider Test", False),

of Netflix", False),

of Doddles - Dood!

es and Emotion True)

of the Spider-Verse", True,

of the Spider-V
87
88
90
91
92
93
94
95
96
97
98
90
101
102
103
104
105
110
111
112
113
114
115
116
117
118
119
120
121
121
122
123
124
125
127
128
129
130
                                                                                                                                            131
132
133
134
135
136
137
138
139
140
141
142
143
144
145
146
147
149
150
151
152
153
154
155
156
157
158
159
160
161
162
164
165
167
168
170
171
172
```

```
Details 15 Game of Photones' Season B Egizode 3 Your Mught Neve Missed*, True),
Miss 7 For Unbocking - 11's All SCHEEM, False),
Lung Welth Shallon Topics to the State New York City Rent*, True),
Welth Shallon Topics to the State New York City Rent*, True),
Welth Shallon Topics to the State State New York City Rent*, True),
A shoot done with high school...ehadt*, False),
Welth Shallon Topics of the State State State Office of Presence (Most York State),
Bearded Geve at Finsibing time gives universe; states departs of a State Stat
176
177
178
179
180
181
182
   183
184
185
186
187
188
189
       190
   191
192
193
194
195
196
197
198
199
200
201
202
203
204
205
206
207
208
210
211
212
213
214
215
217
218
220
221
222
223
224
225
226
227
228
229
231
232
224
225
226
227
228
229
231
232
244
245
256
257
258
259
250
251
252
253
254
255
257
258
259
257
258
259
257
258
259
257
258
259
257
258
259
257
258
259
257
258
259
257
258
259
257
258
259
257
258
257
258
259
257
258
259
257
258
259
```

```
trending_req = requests.get("https://www.youtube.com/feed/trending")
trending_html = trending_req.text
261
262
263
264
265
266
267
270
271
272
273
274
275
276
277
280
281
282
282
282
284
285
               titleListUnf = re.findall(r"(?<=title=\").+?(?=\")", trending html)</pre>
              button titles = ["Queue", "Verified", "Loading icon", "Watch later", "YouTube home",
    "YouTube Video Search", "YouTube Home", "Upload", "Search", "Home",
    "Trending", "History", "Get YouTube Premium", "Get YouTube TV",
    "Music", "Sports", "Gaming", "Movies", "TV Shows", "News", "Live",
    "Spotlight", "360" Video", "Browse channels", "__TITLE_",
    "Previous video", "Play", "Pause", "Next video", "stop"]
               titleList = []
               for title in titleListUnf:
    if title not in buttonT
                                                in buttonTitles and "http" not in title:
                             titleList.append(title)
               return titleList
287
288
289
290
291
292
293
294
295
296
297
298
              keywordFilter(titleList, verbose):
               filteredList = []
               for title in titleList:
                      clickbait = Fals
                     ind((re.search(r"biggest", title, re.IGNORECASE) == None):
    clickbait = True
if not(re.search(r"largest", title, re.IGNORECASE) == None):
    clickbait = True
if not(re.search(r"mind(f, 1941))
                     clickbait =

if not(re
                     if not(re.search(r*morst*, title, re.IGNORECASE) == None):
    clickbait = True
if not(re.search(r*marsis*)
                                 t(re.search(r"mind([-])blowing", title, re.IGNORECASE) == None):
299
300
301
302
                     re.search
clickbait =
if not(re
                     if not(re.search(r"never", title, re.IGNORECASE) == None):
   clickbait = True
if not(re.search(r"clicks", dien.
                                 t(re.search(r"amazing", title, re.IGNORECASE) == None):
303
304
305
                     if not(re.search(r*click*, title, re.IGNORECASE) == None):
    clickbait = True
if not(re.search(r*beab*, title)
306
307
                     if not(re.search(r"wow", title, re.IGNORECASE) == None):
   clickbait = True
if not(re.search(r"surget;);
308
309
                    if not(re.search(r"surprising", title, re.IGNORECASE) == None):
   clickbait = True
   if not(re.search(r"omg", title, re.IGNORECASE) == None):
     clickbait = True
   if not(re.search(r"omg", title, re.IGNORECASE) == None):
     clickbait = True
310
311
312
313
314
315
316
317
318
319
320
321
322
323
324
325
                              clickbait = Tru
                      if verbose:
                              if clickbait:
                             print("Clickbait: ", title)
else:
    print("Not clickbait: ", title)
326
327
328
329
330
                      filteredList.append((title, clickbait))
               return filteredList
               ohraseFilter(titleList, verbose):
331
332
333
334
335
336
337
               filteredList = []
               for title in titleList:
                      clickbait = Fals
338
339
                      if not(re.search(r*here.*s why*, title, re.IGNORECASE) == None):
                      clickbait =
if not(re.search
                                 t(re.search(r"you never knew", title, re.IGNORECASE) == None):
                            not(re.search(r"won.*t believe", title, re.IGNORECASE) == None):
    clickbait = True
not(re.search(r"weird trick", title, re.IGNORECASE) == None):
    clickbait = True
341
342
343
344
```

```
this", title, re.IGNORECASE) == None):
                      (re.search(r
347
348
                   clickbait =
                     t(re.search(r"watch this", title, re.IGNORECASE) == None):
349
350
                   clickbait =
                  not(re.search(r*find out*, title, re.IGNORECASE) == None):
clickbait = True
351
352
                    t(re.search(r"life changing", title, re.IGNORECASE) == None):
353
                  clickbait =
354
                     t(re.search(r"need to see", title, re.IGNORECASE) == None):
355
356
357
358
                  clickbait =
                  not(re.search(r*will make you*, title, re.IGNORECASE) == None):
clickbait = True
                     t(re.search(r"this is what happens", title, re.IGNORECASE) == None):
359
                  clickbait =
360
361
                  not(re.search(r"oh my god", title, re.IGNORECASE) == None):
clickbait = True
362
363
364
365
366
                     (re.search(r"this is nuts", title, re.IGNORECASE) == None):
                   clickbait = Tru
             if verbose:
367
368
                   if clickbait:
                           int("Clickbait: ", title)
369
                          rint("Not clickbait: ", title)
371
372
              filteredList.append((title, clickbait))
373
374
375
376
377
378
379
380
         return filteredList
         formatFilter(titleList, verbose):
         filteredList = []
         for title in titleList:
    clickbait = False
381
382
383
384
              if not(re.search(r"^Guess .*", title, re.IGNORECASE) == None):
385
386
387
                  clickbait =
                    t(re.search(r"^[0-9]+ (?!in ).*", title, re.IGNORECASE) == None):
                  clickbait =
                    t(re.search(r"^(?<!I )[A-Z]+ [(A-Z|0-9)]+ .*", title) == None):
388
389
                  clickbait =
                  not(re.search(r".*\*\*.*", title, re.IGNORECASE) == None):
clickbait = True
390
391
                  not(re.search(r"
  clickbait = True
                                      *\?!+.*", title, re.IGNORECASE) == None):
392
393
394
395
                     t(re.search(r
                                      *!\?+.*", title, re.IGNORECASE) == None):
                  clickbait =
                  not(re.search(r"
  clickbait = Tru
396
                                      .*\?{3,}.*", title, re.IGNORECASE) == None):
397
398
                     t(re.search(r".*\!{3,}.*", title, re.IGNORECASE) == None):
                  clickbait = True
not(re.search(r".*[A-Z]+ [A-Z]+ [A-Z]+.*", title) == None):
clickbait = True
399
400
401
402
403
              if verbose:
404
405
406
407
                   if clickbait:
                          int("Clickbait: ", title)
408
409
410
              filteredList.append((title, clickbait))
411
          return filteredList
412
413
414
415
416
417
418
419
420
421
422
423
               oFilter(titleList, verbose):
         filteredList = []
         for title in titleList:
              clickbait = Fa
              if not(re.search(r"biggest", title, re.IGNORECASE) == None):
                  clickbait =
424
425
                  not(re.search(r*largest*, title, re.IGNORECASE) == None):
clickbait = True
426
                     t(re.search(r"mind([-])blowing", title, re.IGNORECASE) == None):
                   clickbait =
427
428
429
                     (re.search(r"worst", title, re.IGNORECASE) == None):
                  clickbait = True
not(re.search(r*amazing*, title, re.IGNORECASE) == None):
clickbait = True
430
```

```
", title, re.IGNORECASE) == None):
433
434
                  clickbait =
                    t(re.search(r"click", title, re.IGNORECASE) == None):
435
                  clickbait =
436
                                      pah", title, re.IGNORECASE) == None):
                    t(re.search(r
437
                  clickbait =
438
                    t(re.search(r
                                   "wow", title, re.IGNORECASE) == None):
439
440
441
442
443
                  clickbait =
                     (re.search(r
                                    'surprising", title, re.IGNORECASE) == None):
                  clickbait =
                    t(re.search(r"omg", title, re.IGNORECASE) == None):
                  clickbait =
444
                    t(re.search(r"crazy", title, re.IGNORECASE) == None):
445
                  clickbait = T
446
447
448
             if not(re.search(r"here.*s why", title, re.IGNORECASE) == None):
449
                  clickbait = Tr
not(re.search(r
clickbait = Tr
450
                                    'you never knew", title, re.IGNORECASE) == None):
451
452
                    t(re.search(r"w
                                      on.*t believe", title, re.IGNORECASE) == None):
453
                  clickbait =
454
                     (re.search(r"weird trick", title, re.IGNORECASE) == None):
455
456
457
458
                  clickbait =
                    t(re.search(r"see this", title, re.IGNORECASE) == None):
                  clickbait =
                    t(re.search(r"watch this", title, re.IGNORECASE) == None):
459
                  clickbait =
460
461
                    t(re.search(r"find out", title, re.IGNORECASE) == None):
                  clickbait =
462
                     (re.search(r"life changing", title, re.IGNORECASE) == None):
463
464
                  clickbait =
                                     need to see", title, re.IGNORECASE) == None):
                  not(re.search(r'
  clickbait = Tr
465
466
                    t(re.search(r"will make you", title, re.IGNORECASE) == None):
467
468
                  clickbait =
                    t(re.search(r"this is what happens", title, re.IGNORECASE) == None):
469
                  clickbait =
470
                     (re.search(r
                                       my god", title, re.IGNORECASE) == None):
471
472
473
474
                  clickbait =
                     (re.search(r"this is nuts", title, re.IGNORECASE) == None):
                  clickbait = Tr
475
476
477
             if not(re.search(r"^Guess .*", title, re.IGNORECASE) == None):
                  clickbait =
478
                    t(re.search(r
                                     [0-9]+ (?!in ).*", title, re.IGNORECASE) == None):
                  clickbait =
480
481
482
                     (re.search(r
                                     (?<!I )[A-Z]+ [(A-Z]0-9)]+ .*", title) == None):
                  clickbait =
                                      *\*\*.**, title, re.IGNORECASE) == None):
                     (re.search(r
483
                  clickbait =
484
                                     *\?!+.*", title, re.IGNORECASE) == None):
                    t(re.search(r
485
                  clickbait =
486
                    t(re.search(r
                                     *!\?+.*", title, re.IGNORECASE) == None):
487
488
489
                  clickbait =
                     (re.search(r
                                     *\?{3,}.*", title, re.IGNORECASE) == None):
                  clickbait =
490
                    t(re.search(r
                                     *\!{3,}.*", title, re.IGNORECASE) == None):
491
                  clickbait =
492
493
                    t(re.search(r
                  clickbait = T
494
495
496
497
              if verbose:
                   if clickbait:
498
                          int("Clickbait: ", title)
499
500
501
502
              filteredList.append((title, clickbait))
503
504
         return filteredList
505
506 goldTitles = []
507
508
                  goldStandard:
         goldTitles.append(tuple[0])
510
511 goldEvalKeyList = keywordFilter(goldTitles, False)
512 goldEvalPhrList = phraseFilter(goldTitles, False)
513 goldEvalForList = formatFilter(goldTitles, False)
514 goldEvalCmbList = comboFilter(goldTitles, False)
515
515
516
          evaluate(progList, goldList):
         truePos =
         trueNeg =
```

```
| Sample | S
```