

Characterizing Alternative and Emerging Tobacco Product Transition of Use Behavior on Twitter

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Research note

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Abstract

Objective: The objective of this study was to develop an inductive coding approach specific to characterizing user-generated social media conversations about transition of use of different tobacco and alternative and emerging tobacco products (ATPs).

Results: A total of 40,206 tweets were collected from the Twitter public API stream that were geocoded from 2018-2019. Using data mining approaches, these tweets were then filtered for keywords associated with tobacco and ATP use behavior. This resulted in a subset of 5,718 tweets, with 657 manually annotated and identified as associated with user-generated conversations about tobacco and ATP use behavior. The 657 tweets were coded into 9 parent codes: inquiry, interaction, observation, opinion, promote, reply, share knowledge, use characteristics, and transition of use behavior. The highest number of observations occurred under transition of use (43.38%, n=285), followed by current use (39.27%, n=258), opinions about use (0.07%, n=46), and product promotion (0.06%, n=37). Other codes had less than ten tweets that discussed these themes. Results provide early insights into how social media users discuss topics related to transition of use and their experiences with different and emerging tobacco product use behavior.

Introduction

Social media is now a common source of health-related information [1]. This includes user-generated conversations about a variety of substance use topics, with an emerging field of literature specifically focused on using information from different social media platforms to better understand tobacco and alternative and emerging tobacco (ATP) related knowledge, attitudes, and behaviors [2,3]. As social media platforms become more socially embedded in everyday life, user generated conversations can be assessed [4] to better understand how health behaviors are changing closer to real-time [5]. This approach introduces certain advantages over traditional survey methodology including faster identification of emerging trends and the ability to analyze metadata associated with social media conversations to assess user characteristics, location data, and social network relationships [6]. However, methods to appropriately code social media content for specific health-related topics remain underdeveloped, particularly in the context of identifying and characterizing transitions in behaviors that may change over time.

Twitter is a micro blogging social networking platform that allows users to tweet 280-character messages, which can then be retweeted, favorited, and shared across a large network of online users [7]. Users can form online communities [8] by interacting with other users who share similar beliefs, interests and opinions about topics (by using hashtags, following, retweeting, favoriting, and commenting). This includes user communities who initiate, use and transition between different tobacco and ATP products [9,10]. In fact, Twitter has specifically become a platform for sharing information about electronic cigarettes (e-cigarettes) [11], a nicotine delivery device that has only been introduced commercially in the past decade [12].

Evidencing growing popularity of e-cigarettes and vaping behavior, studies have shown that online searches for electronic cigarettes have increased [13]. However, increased uptake of different types of e-cigarettes and other ATPs (e.g., Juul, heat-not-burn, etc.), particularly among youth and young adults, has not been without controversy [14]. Specifically, ongoing concerns about the long-term health impact of nicotine consumption [15], ATP-related adverse events [16] (such as the 2019 outbreak of e-cigarette or vaping product use-associated lung injury), and mixed evidence about the efficacy of ATPs as cessation devices, continues to generate public health and patient safety concerns about these products [17,18]. These concerns are accentuated when trying to assess the interaction of use behavior between traditional combustible tobacco products (e.g., cigarettes, cigars) and ATPs [19].

Hence, understanding the pathways of transition of tobacco and ATP use - including what products users initiate on, why they switch between products, and unique health harms related to dual-use (i.e., simultaneous use of both combustible and ATPs) – is still a relatively underdeveloped area of study [20]. Hence, the objective of this study was to utilize Twitter as a data source to characterize instances of users' self-reporting behavior in relation to transition of use, with a focus on developing an inductive coding approach specific to identifying key themes of transition of use knowledge, attitudes, and behaviors for use in future social media-based studies.

Methods

We conducted a retrospective observational social media data mining study in two phases: (1) data collection and data mining; and (2) content analysis using an unsupervised machine learning and inductive coding approach. Inductive content analysis to characterize posts relevant to tobacco and ATP behavior and use (i.e., “signal” tweets) were manually annotated by expert coders with training in tobacco and substance use behavior, with results used to generate a codebook of transition and behavioral-related themes that could also be iterated on in future social media-based studies.

Data Collection

To explore conversations related to tobacco and ATP transition of use, data was first collected from the Twitter public streaming API for all tweets that contained geocoded conversations in the US from 2018-2019 as part of a separate study. This initial dataset of geocoded tweets was then filtered for keywords associated with tobacco and ATP use and behavior including vape and vaping. The collected data included textual content of the tweet, user and account information, URLs, and time of post.

Data Mining

To identify themes in our full corpus of tweets, we utilized an unsupervised machine learning approach called the Biterm Topic Model (BTM) that leverages topic modeling and natural language processing (NLP), designed to detect patterns in the data and summarize the entire corpus of tweets into distinct highly correlated categories [21]. It is used to sort short text into highly prevalent themes without the need of predetermined coding or training to classify content into topics and has been previously used for

exploration of key public health topics [22, 23, 24, 25]. For each topic, BTM generates the top 20 words that represent the topic cluster. These topics were then reviewed and selected to identify clusters of user generated Twitter conversations that were pertinent to vaping and transition of use. Since BTM topics are highly correlated we are able to identify “signal” topics based on the BTM output and eliminate irrelevant topics that do not correlate with the study aims. BTM topics were first generated by applying keyword filters of vape and vaping. Outputted topics were included if they were pertinent to vape and vaping only, topics were excluded if they contained irrelevant topics or appeared to correlate with non-user generated conversations (e.g., news tweets, etc.)

We then extracted all the posts from the select vaping topics and then manually coded to ensure they were related to user-generated tobacco and ATP use behavior. Posts were excluded as signals if they were: (1) news related and not organically user-generated content; (2) not written in English; and (3) retweets, the tweets that were retweeted counted as only one tweet. However, all tweets, replies, and tweets containing photos or videos were included to analyze for additional contextual information in addition to content analysis of text contained in the tweet. Transition of use was classified as switching from one tobacco or ATP product to another.

Content Analysis

Tweets and any associated URLs/hyperlinks were aggregated into a table and imported into Atlas.ti qualitative data analysis software for content analysis [26]. A first iterative, inductive analysis of the data was conducted (JSY) to identify thematic areas and classify tweets into codes with code descriptions. Tweets were read for identification of thematic areas in the dataset, then coded based on thematic areas of interest. Codes and coding descriptions were developed and modified iteratively throughout the coding process. A second analysis of the dataset was undertaken to expand the codebook to include subcodes. Subcodes and subcode descriptions were created and modified iteratively during a second round of data coding. Once a coding scheme was developed, the data were coded, extracted, and reviewed to assess the validity of the coding scheme by a second coder (CB). The final coding scheme is presented in **Table 1**.

Results

A total of 40,206 tweets were collected from 2018-2019, with 5,728 (14.25%) comprised of tweets with filtered keywords related to tobacco and ATP use. After manually annotating tweets for relevance to tobacco and ATP use and behavior, we removed all non-signal tweets, leaving 589 signal tweets related to transition of use that were further analyzed. The 589 signal posts were categorized into 10 thematic codes, listed and identified in **Table 1**.

In relation to transition of use, thirteen distinct tobacco/ATP transition pathways involving ATPs were identified; the term “vaping” was used to describe both nicotine vaping and vaping of cannabis-based products. Transitions detected were cannabis to cannabis (0.005%, n=3), cannabis to e-cigarettes (0.006%, n=4), chewing tobacco to e-cigarettes (0.01%, n=6), cigarettes to e-cigarettes (27.16%, n=160),

cigarettes to no product (0.14%, n=8), cigarettes to vape cannabis (0.007%, n=4), e-cigarette to cannabis (0.002, n=1), e-cigarette to cigarette (0.01%, n=7), e-cigarette to e-cigarette (0.2%, n=13), e-cigarette to no product (0.07%, n=43), no product to e-cigarette (0.03%, n=17), no product to vape cannabis (0.03, n=15), and unknown product to e-cigarette (0.007%, n=4).

There were also transitions among different ATP product types as well as cannabis product types, one of which was vaping a cannabis product. Vaping use factors that were observed as influencing transition of use included self-reporting of addiction, adverse symptoms, cost of ATPs, faulty or broken ATPs, preference for flavors, losing or misplacing ATPs, interest in polysubstance use, concern about reducing nicotine levels, stigma, and the alleged therapeutic effects of vaping, especially CBD.

Discussion

This study explored user-generated conversation occurring on Twitter in relation to tobacco and ATP use, with a specific focus on transition of use between these highly addictive products. We observed that this subset of Twitter users actively tweeted about their experience using tobacco and ATPs, representing potentially powerful first-hand information about this behavior influenced by new and emerging nicotine products that may not be readily available using other research and data collection approaches (e.g., cross-sectional surveys, etc.). Overall, the majority of tweets reviewed related to tobacco and ATP use and behavior characteristics including users asking other users about tobacco/ATP products and how to quit, reports of observations and interactions with people in the real-world about tobacco/ATP use and behavior, offering opinions about products and vaping (including vaping as a healthier alternative to tobacco or its alleged therapeutic benefits), replying to and sharing knowledge about tobacco/ATP products with other users, and specific characteristics of use (e.g. addiction, adverse events, costs, flavoring, tricks, etc.)

Specific to Twitter users discussing their transition of use behavior, we observed that users actively discussed both the types of tobacco and ATP products that were used and switched between, as well as provided detailed reasons for this change in product use. A wide variety of tobacco and ATP products were mentioned, including combustible tobacco products (e.g., cigarettes), chewing tobacco, different types of ATPs (Juul, vaping pens, etc.) and cannabis smoking products. Transition was observed between these different products and within these specific product classes (i.e., transitioning from one type of e-cigarette product to another), with some users (n = 32) also specifically self-reporting polytobacco and polysubstance behavior (e.g., smoking cigarettes and also vaping). Users expressed various sentiment about these different products including from the perspective of what products could act as substitutes for others, what products made them feel better, attempts to quit use of one product by switching to another, and issues related to cost and access. Interestingly, some users explicitly stated that cannabis vaping products helped them with cessation of nicotine addiction.

Based on these preliminary results, it appears that for some Twitter users, the platform enables robust conversation and sharing of information related to tobacco and ATP use and behavior and can act as a

digital forum for smokers and vapers to accumulate knowledge, share experiences, and become better acquainted with ATP products.

Limitations

This study was exploratory and meant to generate hypotheses for future research. The study's limitations include the use of a single platform to conduct data collection. Further, while Twitter is a popular social media platform, the demographics of Twitter users may not reflect that of the general population of tobacco and ATP users. The sample of tweets were also limited based on a convenience sample generated on the basis of first collecting for geocoded tweets, and hence, is not representative of Twitter conversations and may be subject to sample bias. Specifically, this data collection methodology was limited to collecting messages from Twitter users that enabled geolocation, a specific limitation to generating a more generalizable dataset as it is estimated that only 1% of all tweets are geocoded [29,30]. Future studies should attempt to use multiple Twitter APIs to generate a more representative Twitter dataset or complement twitter data with other primary and secondary data sources.

Conclusion

The results of our study are exploratory in nature and were derived from a sample of general geolocated tweets over a one-year period, which were then filtered for common tobacco and ATP keywords and then analyzed using unsupervised machine learning. Hence, the results of this study are not generalizable to overall trends in tobacco or ATP behavior, but nevertheless provide important insights into conversations occurring among Twitter users specific to transition of tobacco/ATP use. Themes associated with the transition of use were primarily focused navigating quit attempts or having trouble quitting in the past, those who had relapsed to nicotine addiction, and those who had quit cigarettes but still vaped. Importantly, these results provide early evidence that experiences in transition of use also present opportunities for more targeted and personalized cessation interventions, particularly in the context of increasing knowledge of the known health harms related to continued tobacco use and nicotine addiction and exposure [27, 28].

Future work should conduct further confirmatory studies to assess if themes related to transition of use knowledge, attitudes and behaviors observed in this study hold true in other digital communities and also seek to use more structured research approaches using traditional methods of longitudinal surveys, focus groups, and other approaches to better generalize findings. Future studies could also employ network analysis to determine if online and offline social networks influence perceptions about transition of use and cessation. Further, future studies that leverage social media platforms for self-reported behavioral-related information should look at other platforms now more popular among youth and young adults, such as Instagram, Snapchat, and TikTok, though methods of data collection and analysis may vary.

Declarations

Author Declarations:

Ethics Approval and Consent to Participate: Not applicable/Not required for this study. All information collected from this study was from the public domain and the study did not involve any interaction with users. Any user identifiable information was removed from the study results.

Consent for Publication: Not applicable

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Author Contributions: All authors jointly conceived of the study. JL, CB, and JY conducted data collection and data analyses. All authors jointly wrote the draft and final versions of this manuscript.

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Data Availability Statement: The data that support the findings of this study are available upon request and certain data will be available freely from the website www.ghpolicy.org.

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Tables

Table 1. Emergent coding scheme for Twitter data for tobacco and ATP use behavior

CODE	SUB-CODE	DEFINTION	Paraphrased Example Tweet
INQUIRY (n=12)		Tweet posing a question to other Twitter users	Is anyone a former e-cigarette user? If so, how did you quit? What healthy habit did you replace it with? Feel free to DM me. Is a gas station vape worth purchasing?
INTERACTION ((n=6)		Description of an interaction a Twitter user has with another person related to vaping	The security at the club thought my juuls were chewing gum and when I said it was a vape, he said "well that's embarrassing" I visited my neighbor the other day to try to convince my neighbor to switch from cigarettes to e-cigs, but realized he was smoking cannabis in his backyard instead. It went from a "howdy neighbor" to us smoking together.
OBSERVATION	General (n=1)	Observation a person makes related to vaping	At lunch I heard people talking about the best vape pens and anxiety Person 1: I don't get anxiety Person 2: You are always inebriated; you won't experience anxiety
	Others vaping (n=6)	Observation of another person's vaping behavior	Over the holidays I learned observed that my father began vaping and that's pretty much it My grandma is learning how to vape because she has medical marijuana now, and I am trying my best not to laugh
OPINION OF PRODUCT	Negative (n=3)	An opinion of a vaping-related	Wax vape= lung hurting juice

		product with negative sentiment	
	Positive (n=5)	An opinion of a vaping-related product with positive sentiment	I prefer vapes over juuls please don't come for me
OPINION	Contradiction (n=4)	An opinion about vaping the user finds to be a contradiction in some way	@usernameredacted I haven't been able to refill my vape since cannabis became "legal" it's ridiculous
	General (n=6)	General expression of a vaping-related opinion	Yea e-cigs. The addiction is nicotine which is in the vape juice. Humans have been addicted to nicotine, which is actually beneficial. No new findings here.
	Humor/sarcasm (n=2)	An expression of opinion related to vaping meant to be humorous	What grandfather I did not know you vape (redacted url)
	Transition from no use to e-cigarette initiation (n=7)	Opinion about people who start vaping without doing so as a means to stop smoking	I am highly judging anyone who vapes or juul with nicotine but never began smoking cigarettes. You were never addicted to nicotine. Just trying to be cool.
	Vaping as healthier alternative (n=5)	Opinion about vaping as a healthier alternative to another tobacco-product	Vaping is healthy nicotine
	Vaping helps quit smoking (n=9)	Opinion about vaping as a method of quitting smoking	Reason number (redacted) why smokers should use vape instead ... bonus you won't smell like a disgusting ashtray upon room entrance. (redacted url)
	Vaping (negative) (n=6)	General opinion about vaping behavior that	In my opinion vaping/juuling makes the addiction worse when you're trying to quit smoking cigarettes

		has negative sentiment	
	Vaping as therapeutic (n=2)	Opinion about the therapeutic benefits of vaping	Do not let anyone dispel your depression because you handle it differently. Each person has a different coping mechanism. If you have depression acknowledge it and seek help (there are great antidepressants) do therapy, vape, cannabis, listen to music. Just own it.
	Youth use (n=6)	Opinion about youth vaping	@redactedusername Yes, I agree. It's awful children under 18 use e-cigs but it is better than cigarettes. In an ideal world they wouldn't do neither, but I know when I was a teen I smoked and would have vaped if I could instead.
PROMOTION	Product by user (n=28)	Promotion of a product as a user of that product	I already have a vape but wow this one is so cool. I'm not biased to the colors or anything
	Product by commercial source (n=9)	Promotion of a product by a commercial entity	Request your vape! We have the redacted model voice control mod!
REPLY (n=2)		Reply to another Twitter user's inquiry related to vaping	@redactedusername Suorins give a greater headrush and cheaper to refill. 4, 0.7ml juul pods are the same price as a 30ml suorin bottle of vape liquid. Lol suorins are an overall better investment
SHARE NICOTINE KNOWLEDGE (n=1)		Sharing knowledge of e-cigarette nicotine levels	TIL Juul contains 59mg of nicotine. That's about 3x the amount a usual e-cig or vape. They're not sold in the EU for that exact reason. Wow
TRANSITION	Cannabis to cannabis (n=3)	Transition between cannabis products	Since I can't smoke cannabis, I bought a CBD vape and it's a great substitute
	Cannabis to e-cigarettes (n=4)	Transition cannabis product to nicotine vaping	I stopped smoking weed but I cannot get rid of my vape. I can't wait until Tuesday

Chewing tobacco to e-cigarettes (n=6)	Transition from chewing tobacco to nicotine vaping	I broke down and bought a vape and flavored liquid. Hopefully this is temporary until I quit chewing tobacco. Wish me luck or tell me I'm dumb for exchanging a bad habit for another (redacted url)
Cigarettes to e-cigarettes (n=160)	Transition from conventional cigarettes to nicotine vaping	I feel so much better now that I quit smoking cigs and I hardly use my vape, so I am delighted
Cigarettes to no product (n=8)	Transition from conventional cigarettes to no product	Btw I quit smoking cigs as of Tuesday after 10 consistent years. I am doing it cold turkey no nicotine patches or vape pens. So, if I seem tense keep that in mind. I am having surgery next month and nicotine critically reduces your ability to heal.
Cigarettes to vape cannabis (n=4)	Transition from conventional cigarettes to vaping cannabis product	I take my cbd vape with me everywhere It helped me quit smoking again and keeps the nerves/edge feeling down.
E-cigarette to cannabis (n=1)	Transition from nicotine vaping to a cannabis product	This is funny but addiction to nicotine isn't...It's been 5 weeks and I am so glad I quit. I advise anyone trying to quit to buy cbd vape juice at a vape shop and use it in your mod. It won't block the withdrawals, but it is similar/relaxing.
E-cigarette to cigarette (n=7)	Transition from nicotine vaping to conventional cigarettes	What stage of capitalism is that I am using cigarettes since I can't afford my vape stuff
E-cigarette to e-cigarette (n=13)	Transition between nicotine vaping products	I prefer regular vape after about two to three days of salt nic
E-cigarette to no product (n=43)	Transition from nicotine vaping to no product.	I haven't used my vape or had caffeine in 2 weeks and I just spent the last 30 minutes getting yelled at by

			an elderly tenant without getting a word in. I am so ever the edge right now.
	No product to e-cigarettes (n=17)	Transition from no product to nicotine vaping	You know what's next? Smoking my vape 24/7 when you're not a smoker, so awesome dude
	No product to vape cannabis (n=15)	Transition from no product to vaping cannabis product	According to my husband I am a drug addict now that I got a cbd vape pen
	Unknown product to e-cigarettes (n=4)	Transition from unknown product to nicotine vaping	Man, I know have slowed down... slowly but surely, I will try a vape soon. I got off of cannabis, I really only smoke black after I drink a brew to get a feeling close to the Mary
USE CHARACTERISTICS	Addiction (n=21)	Mention of addiction or descriptions of behavior associated with addiction	Attempting to prove that I don't have an addiction to vape and it's taking too much strength to not buy a new tip for my tank. I have an addiction,
	Adverse symptoms (n=25)	Adverse symptom associated with vaping	Hookah and vape are the worst things for me, vaping makes me dry heave
	E-cigarettes complicated to use (n=1)	Vapes being complicated or hard to use	I almost forgot how to use this 400 vape I bought (redacted url)
	E-cigarettes cost (n=12)	Cost (either high or low) associated with vaping or cigarettes	Why are the juul starter packs so pricey? For \$** I can just buy an actual vape
	Entertainment (n=3)	Entertaining image or video of vaping	Vape warning! (redacted url)
	Faulty or broken e-cigarette (n=9)	Vape being faulty or broken or breaking	P.s I stopped smoking cigarettes, and it stresses me that my vape is broken (redacted url)
	Flavor (n=8)	Flavors used in vape	Flavor of the day is english toffee! delicious

General (n=63)	Uncategorized mentions of vaping	I have run out of vape juice and it is horrendous.
Lose or misplace e-cigarette (n=11)	Misplacing vape, or finding one after having lost or misplaced	I lost my vape pen, damn it lasted so long.
New product (n=12)	Obtaining or using a new product	I am trying to stop using nicotine and I am now a proud vape owner. I prefer Earl Grey tea flavor. (redacted url)
Polytobacco/polysubstance (32)	Concurrent (within 30 days) use of two or more forms of a tobacco product or tobacco product and other substance (e.g., alcohol, marijuana).	It's been a while since I smoked cannabis. I use edibles and vape now. Smoking cigarettes at night. Nicotine in vape is not enough.
Reduce use (n=12)	Reduction in volume of vaping or nicotine used in vaping	No nicotine in my vape. I didn't predict how much my system would be in shock after reducing nicotine use. I moved from vape to juul a couple of days ago and it is very unpleasant
Stigma (n=11)	Negative image or reputation associated with vaping	You can make fun of me because I bought a vape.
Therapeutic effects (n=36)	Therapeutic effects of vaping	My anxiety has been rough lately ***** bought me a CBD vape and it works. It is physically and mentally calming.
Trick (n=2)	Vaping trick, such as blowing clouds	I have been trying to nail this trick for a while and I finally succeed, I was so shocked. #vape#vapetricks

Figures

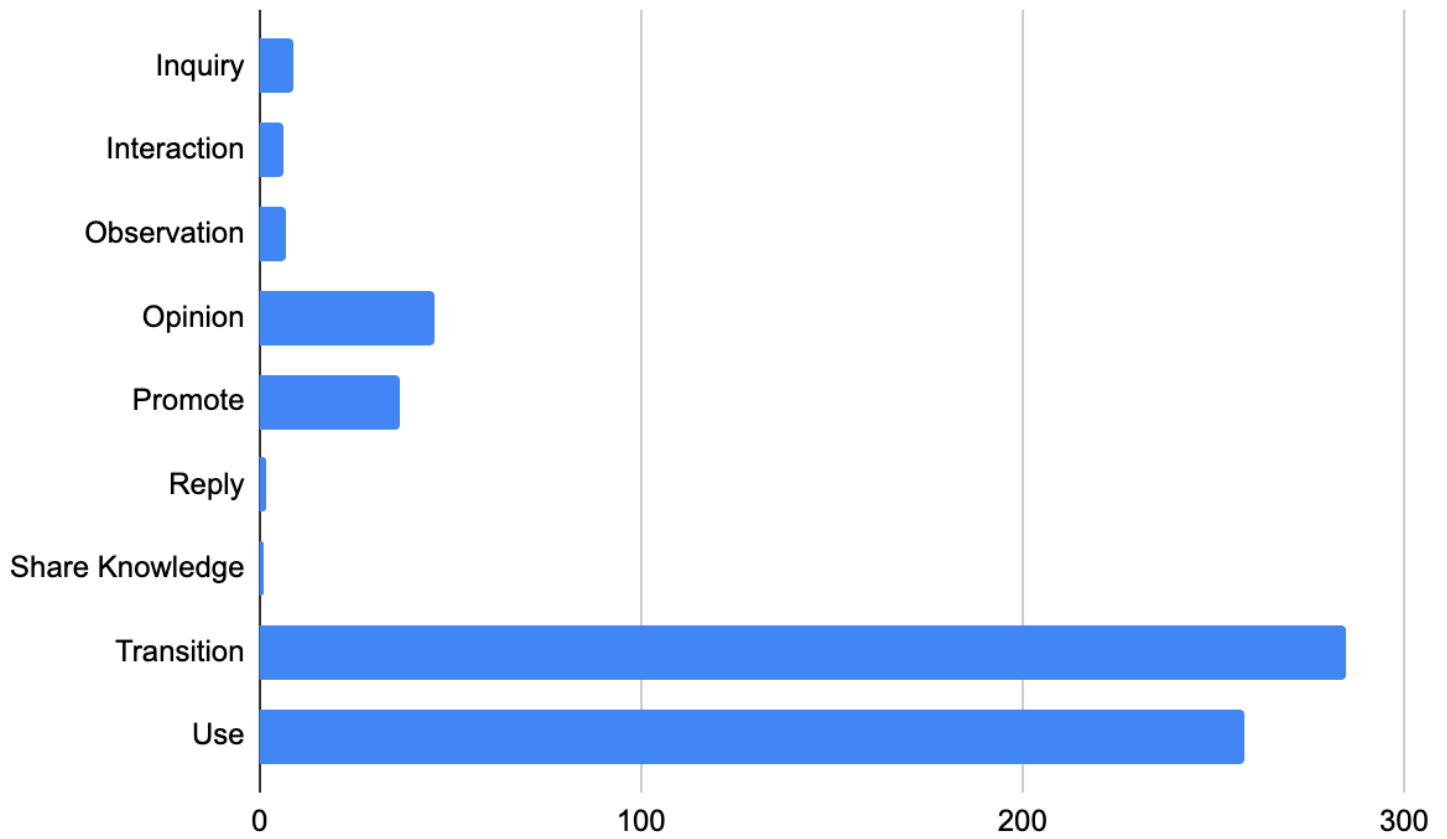


Figure 1

Tweets per tobacco and ATP theme