

Gaining The Competitive Edge

Analyzing the Importance and Potential Drivers of Success for Political Marketing Strategies Regarding the 2021 Federal Elections in Germany

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Management Summary

Empirical evidence suggests that a substantial amount of voters is susceptible to persuasion: 23% of polled German voters were uncertain of their voting decision two weeks before the election, a similar share of 10 to 20% of undecided voters can also be found in other European countries like Italy and France (Spiegel 2021, Alam & Riva 2019, La Repubblica 2006, Mannheimer 2003). Hence, political marketing should be considered a suitable tool to gain voter support and gain a competitive edge against other parties – particularly in close elections.

In addition to the content stances, the measurements of sentiments, emotions, and text similarity can be used to validly determine party positioning (Silva & Proksch 2021, Burst et al. 2021, Jolly et al. 2019, Laver et al. 2002). Based on the work of Giavazzi et al. (2020), which identified a shift in text similarity between German political parties following terror attacks, this research conducted a topic-specific analysis to identify similar shifts in central election topics. The topic-specific stance analysis detected several shifts in language by the parties. The topic of Corona was discussed with more right-wing language by all parties compared to their overall stance position. In contrast, the topic Environment was discussed with more left-wing language by all parties compared to their overall stance position. For this purpose, this research collected two large data sets from Twitter and the 2017- and 2021-party programs of the six major political parties.

As a unique contribution to political marketing research, this paper conducted a comprehensive text analysis of the party activities on Twitter regarding the 2021 Federal Elections in Germany. Additionally, this research derived practical, data-driven recommendations for potential drivers of success for political marketing regarding elections. These recommendations address the public candidate image, emotive strategies, and practical techniques to manipulate voter behavior.

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1. Introduction

On September 26th, 2021, voters in Germany elected a new federal government for the upcoming legislature period. As the first incumbent chancellor in German history, Angela Merkel (CDU/CSU) did not run for office again after 16 years as head of government. The campaigning for her successor was a tight and suspenseful race with substantial fluctuation in voter supports: three parties in the Conservatives (CDU/CSU), Greens (Grüne), and Social Democrats (SPD) all held supremacy in the polls at some point. On election day, the SPD prevailed by 1.6% – a difference of only approximately 700,000 votes out of 61 million eligible voters (Politico 2021).

This paper argues that political marketing should be considered a capable tool to generate a competitive edge over opposing parties in tight campaign races like this one for several reasons. Firstly, political marketing can generate a competitive edge because most politicians utilize social media to amplify messages and as platform for public display (Silva & Proksch 2021). Particularly Twitter as a text-centric platform has gained popularity among German politicians: roughly 70% of the current parliament members utilize the platform to connect with the voting population and colleagues (Pollytix 2021). As the digital form of election posters, it has been argued that Twitter may soon replace its analog predecessors (Tawadrous 2021, Jaursch 2021, Jaursch 2020). Secondly, political marketing can generate a competitive edge because many voters are susceptible to persuasion. Two weeks before the elections, 23% of polled German voters were uncertain of their voting decision (Spiegel 2021). A similar pattern of 10 to 20% of undecided voters can also be empirically found in other European countries like Italy and France (Alam & Riva 2019, La Repubblica 2006, Mannheimer 2003). Lastly, political marketing can generate a competitive edge because it can persuade people in their voting decision. After becoming president in 2016, Donald Trump famously stated: “I think I would not be here if I did not have social media” (Wired 2016). His digital media director specified that Twitter was crucial for gaining votes, and Facebook was fundamentally important for fundraising (Fujiwara et al., 2020). Indeed, recent research substantiates evidence for the effect of Twitter and Facebook on the voter turnout in favor of candidates dependent on the party (Fujiwara et al. 2020, Liberini et al. 2018). Hence, this paper argues that social media's effect on the elections results from the perceptions of party positioning and the election-specific circumstances.

In addition to the content stances, the measurements of sentiments, emotions, and text similarity can be used to validly determine party positioning (Silva & Proksch 2021, Burst et al. 2021, Jolly et al. 2019, Laver et al. 2002). Based on the work of Giavazzi et al. (2020), which identified a shift in text similarity between German political parties following terror attacks, this research conducted a topic-specific analysis to identify similar shifts in central election topics. For this purpose, this research collected two large data sets from Twitter and the 2017- and 2021-party programs of the six major political parties. The first data set represents the posts of verified 2021 candidate accounts from the six major parties within the campaign period (Dec. 2020 – Sep. 2021). The second data set represents the posts from user-generated discussions based on election-relevant hashtag queries within the last four weeks before the elections. As initial positioning of parties, the 2019 Chapel Hill Expert Survey was deployed for the left-to-right positioning of the six major political parties and assigned to the texts of the respective 2017 party programs (Jolly et al. 2019).

In line with current research, the two incumbent government parties (CDU/CSU & SPD) have utilized more positive sentiment than the opposition parties (Crabtree et al. 2020). Furthermore, the sentimental and emotive strategy of the fringe parties displays the expected results: whereas the Left exhibits the highest scores for negative sentiment and the emotion sadness, the Alternative (AfD) showcases the lowest score for positive sentiment and the highest scores for the emotions anger, disgust, and fear in their political marketing activities. The text-similarity of the six major political parties has continuously and considerably increased between 2017 and 2021 party programs and campaign period tweets - thus resulting in the most similar relative positioning for all parties for the campaign period tweets. The topic-specific stance analysis detected several shifts in language by the parties. The topic of Corona was discussed with more right-wing language by all parties compared to their overall stance position. This shift was arguably led by the Liberal Democrats (FDP) and the Alternative (AfD) based on tweet volume. In contrast, the topic Environment was discussed with more left-wing language by all parties compared to their overall stance position. This shift was arguably led by the Greens based on tweet volume.

2. Literature Review

In order to lay proper foundation for this research, a delineation of political marketing (chapter 2.1), as well as the current empirical evidence regarding the impact of political marketing (chapter 2.2) and user-generated discussions (chapter 2.3) in elections are displayed in the following. Hence, this literature review serves as the academic framework for this research's conceptual model.

2.1 Delineation of Political Marketing

Political marketing can be delineated as the usage of marketing concepts and practices in political campaigns by numerous individuals or organizations to appeal to their stakeholders. It is operationalized by influencing public opinion, promoting certain ideologies, winning elections, or passing legislation corresponding to the needs and wants of a selected group of people in a particular society (Hughes & Dann 2006, Newman 1999a, Newman & Perloff 2004). These stakeholders do not necessarily possess aligned interests, as political marketing includes tasks like communication with party members, potential sources of funding, media channels, and its arguable core in regards to elections: the electorate (Lock & Harris 1996). Building on the delineation of its characteristics, success in political marketing reflects the materialization or realization of the desired behavior of the focal groups. Subsequently, success explicitly includes shaping the public opinion, gaining new funding partners or winning votes in elections. Regarding the topic of this research, marketing for election campaigns, political marketing mainly targets voter behavior, as it is the measuring stick of success.

From an *academically-oriented perspective*, voter behavior can be construed in a model by Newman (1999c) that operates with two assumptions. Firstly, the model asserts that voter behavior results from a set of individual cognitive beliefs drawn from a vast variety of communicational and media sources (e.g., word-of-mouth marketing) and personal affiliations to particular groups or social environments. Secondly, the model assumes that voters are consumers of a service they receive from politicians. Axiomatically, the consumers choose the candidate with the highest perceived value. The model proposes five cognitive domains that affect this perceived value: political issues, social imagery, candidate personality, situational contingency, and epistemic value.

While political issues refer to the policies the candidate indicates to introduce once elected, **social imagery and candidate personality** reflect the usage of stereotypes to induce specific associations (e.g., with archetypes) to establish the desired image in the voter's mind. **Situational contingency** utilizes the presentation of hypothetical events to illustrate why the opposing candidate is not fit for the office that he is in contention for (Newman 1999c, Newman & Perloff 2004). Situational contingency is relevant for the 2021 German Federal Elections because the lead candidate Armin Laschet (CDU), was confronted with situational contingency. His adversaries stressed the perceived danger of a pending government led by him in public events and social media. The hashtags *#laschetverhindern* (meaning: prevent Laschet) and *#laschetdarfnichtkanzlerwerden* (meaning: Laschet must not become chancellor) are popularly used examples for this (Tagesspiegel 2021). **Epistemic value** is a concept that is supposed to induce curiosity in choosing a particular candidate. In election campaigns, smaller parties usually utilize epistemic value to compensate for the discrepancy in credibility compared to bigger parties resulting from the lower levels of awareness of and knowledge about smaller parties (Newman 1999c, Newman & Perloff 2004).

From a *practice-oriented perspective*, political marketing as a persuasion tool to influence voter behavior is nowadays based on data-driven insights drawn from recent polls and market research results. Stakeholder and SWOT analysis are deployed to effectively reach the identified target groups of potential voters (Magnecon 2021). According to political consultant Johannes Hilje, specializing in political marketing in Germany, there are five critical factors to the success of political marketing strategies (Hilje 2021):

1. The contents should be attractive, mobilizing, and platform-optimized.
2. Building a loyal supporter group (community building) is crucial in creating and preparing content.
3. The communication and content should be tailored to small, distinct target groups to improve their effectiveness (**microtargeting**).
4. Influencers associated with the party can facilitate a bigger stage and hence organically introduce the message to a larger audience.
5. Parties should react quickly to content interactions by users to spark ongoing, potentially viral conversations on the platforms.

In social media, the Alternative (AfD) is the frontrunner in terms of followers and subscribers on the platforms of Facebook and YouTube, whereas the Greens lead on Twitter and Instagram. Additionally, the Alternative (AfD) has the highest percentual user interaction regarding comments, likes, and shares on all mentioned platforms (Hilje 2021).

2.2 Impact of Political Marketing in Elections

10 to 20% of polled voters are uncertain about their voting decision until one week before the election date, suggesting the opportunity for a large scale of voters to be targeted successfully by political marketing efforts (Alam & Riva 2019, La Repubblica 2006, Mannheimer 2003).

Activities are most likely to affect the voting decision when conducted in a personal manner like with direct messages on social media platforms (+4.1%) (Jacobsen 2015, Dale & Strauss 2009, Green et al. 2012, Bond et al. 2012). Political advertising has no significant effect on the overall turnout of voters in elections across all parties because the mobilizing and demobilizing effects of ads verge on neutralizing each other. However, it increases the probability for parties that more voters of the party's respective target group decide to exercise their voting right, thereby shaping the partisan composition of the electorate in its favor (Spenkuch & Toniatti 2018).

Regarding the 2021 Federal Elections in Germany, there were several interesting peculiarities in the political advertising activities: all six parties spent considerably more money on political advertising, as the respective budgets ranged from 6.5 to 20 million EUR. Even though its share is growing, the deployment of online advertising is still comparatively small. In the last 90 days before the elections, the parties spent up to 575,000 EUR on Google and Facebook ads. Except for the Greens, no party initiated a voluntary disclosure of their activities (Zandt 2021). The Alternative (AfD) regularly blurred the lines between negative campaigning and disinformation to discredit their political adversaries. This allegation includes the evidence-free narrative of systematic voter fraud and sexist, migrant, and racist disinformation (Jaursch 2021). The Liberal Democrats (FDP) currently face allegations for using ethically questionable microtargeting techniques on Facebook, like presenting mutually exclusive stances on climate changes depending on individual preferences (RND 2021). Partially acknowledged by Facebook, the Facebook Ad Library is missing thousands of ads by political parties and candidates (TargetLeaks 2021).

2.3 Impact of User-Generated Discussions in Elections

User-Generated Discussions can be defined as user-made content to initiate or participate in conversations on online platforms like social media (Techopedia 2021). Parties can exploit user-generated content to improve their party image, add authenticity, or humanize the candidate in the public eye (Chia 2017). User-Generated Discussions are particularly relevant to this research because they are indicators for the public perceptions of parties and candidates. Additionally, they arguably indirectly affect voter behavior, as the recent example of YouTuber Rezo and his viral video “*The Destruction of the CDU*” (19.3 million views) displays (Rezo 2021, Allgaier 2020). One week before the European Elections in 2019, the popular German YouTube creator Rezo (1.56 million subscribers) posted a controversially discussed video that gained nationwide attention. In this video, he accused the government of climate inaction and openly invoked his followers to specifically not vote for the current government parties and the right-wing Alternative (AfD). Because of the expertise and credibility associated with social media creators like Rezo by their followers, it can be insinuated that incidences like this directly affect their followers' opinions and subsequently indirectly affect their voting behavior (Allgaier 2020).

The impact of user-generated discussions is also suitably illustrated by studies about the phenomenon labeled as “hate speech”. This term refers to aggressive or generally degrading statements towards people based on a group affiliation, usually issued on online platforms or social media. Measured by the number of accounts, a small group of accounts seems to be responsible for disseminating the majority of hate speech posts on Facebook: an estimated 5% of the hate speech creating accounts generate over 50% of the likes for degrading comments. These accounts can be linked to the groups of supporters for the right-wing party Alternative for Germany (AfD) and the more radical Identitarian Movement. This dissemination of hate speech is often coordinated in terms of content and timing (Kreißel et al. 2018). Both papers conclude that this coordination aims to place political content in the middle of society and create the impression that the communicated attitudes toward this content correspond with the attitudes of the average user. Subsequently, a distorted representation of opinion is obtained, which also affects the users' actual opinions (Geschke et al. 2019, Kreißel et al., 2018). Applying these research results to election campaigns could mean that this distorted representation of opinions could cause a change in party allegiances and voting choices.

3. German Political System

Based on the 2017 federal election results and the weekly polls in 2021, six parties were projected to satisfy the necessary conditions of being included in the new parliament in July 2021, when the work for this research started (Bundestagswahl 2021b). This projection was confirmed by the results of the 2021 federal elections (Politico 2021). Subsequently, these six parties will be referred to as the "major" parties in the following and introduced briefly, hierarchically ordered based on the second vote proportions earned in the 2021 federal elections (Bundeswahlleiter 2021d, Bundeswahlleiter 2021f). The party introductions are based on the dossiers of the impartial state institution German Federal Agency for Civic Education to exclude potential biases. Besides these established players, a total of 47 other, smaller parties were also eligible for the federal elections in 2021, usually summarized as "Others" in polls and election results (Bundeswahlleiter 2021a). Parties that represent minority populations are explicitly excluded from the necessary regulations to be part of the parliament. In 2021, this rule solely applies to the Southern Schleswig Voters' Association (SSW), representing the Danish minority in the state Schleswig-Holstein (SSW 2021).

Social Democrats (SPD) – 25.7% (2021) / 20.5% (2017)

The Social Democrats (SPD) embody the oldest German party. Giving up its adherence to Marxism with the Godesberg Program in 1959, the SPD began to open up as a party of the people in the late 1950s, becoming the typical representation of its traditional clientele of industrial workers, as well as a large proportion of middle-class voters. Its program can be characterized by the values of freedom, justice, and solidarity (Decker 2021c, Krell & Woyke 2015).

Christian Democrats (CDU/CSU) – 24.1% (2021) / 32.9% (2017)

The Christian Democrats (CDU/CSU) represent a conservatively oriented party of the Christian-bourgeois camp in Germany. Typically associated with anti-socialist orientation, its program is characterized by conservative, liberal and Christian-social positions. As the only major twin party in Germany, the union consists of technically two different parties, as the CDU acts as its representative in 15 of 16 German states. At the same time, the CSU only exists in Bavaria instead of the CDU (Decker 2021c, Bösch 2018, Zolleis & Schmid 2013, Bösch 2002).

The Greens (Die Grünen) – 14.8% (2021) / 8.9% (2017)

The Greens were the most successful newly founded party in the history of the Federal Republic until the emergence of the Alternative (AfD). They were founded in protest against environmental destruction, the use of nuclear energy, and nuclear armament. Since the 1980s, they have established themselves permanently as the fourth force in the party system alongside the CDU/CSU, SPD, and FDP (Decker 2020c, Probst 2015, Probst 2013, Anan 2013).

Liberal Democrats (FDP) – 11.5% (2021) / 10.7% (2017)

The Liberal Democrats (FDP) were the only minor party formed at the beginning of the Federal Republic to permanently secure its parliamentary existence aside from the big parties of CDU/CSU and SPD. Occupying a pivotal role in forming coalitions, they have been the favorite partner in coalitions by both CDU/CSU and SPD. With their liberal economic positions and restrictive stance on refugee and European policy, they have an above-average reach among higher-income voters (Decker 2021a, Treibel 2018, Walter 2010).

Alternative for Germany (AfD) – 10.3% (2021) / 12.6% (2017)

Founded in response to measures to combat the European currency crisis, the Alternative has most notably fought for restrictive immigration policy, a conservative social policy, and an anti-establishment orientation. Now typically identified as a right-wing populist party, the Alternative has been successful in all elections since 2014. The Alternative became the first right-wing party in post-war Germany to earn representatives in each of the 16 state parliaments and the Bundestag (Decker 2020b, Lewandowsky 2018, Decker 2016).

The Left (Die Linke) – 4.9% (2021) / 9.2% (2017)

The Left Party assembles the remnants of the East German Socialist Union Party (*SED*), previously organized in its immediate successor organizations Party of Democratic Socialism (*PDS*) and later also the Labor and Social Justice Party (*WASG*) since the German reunification in 1990. It thus has its roots both in the pro-labor-union environment and the protest against the social policies of the 2000s (Decker 2021b, Neu 2018, Pfahl-Traugher 2013).

4. Conceptual Model

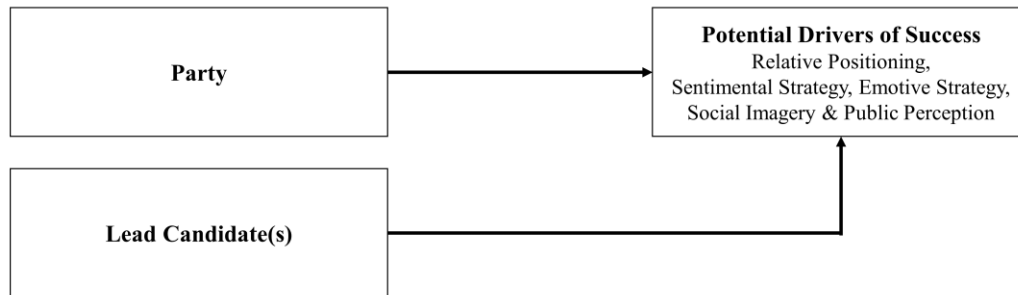


Figure 1: Conceptual Model (own representation)

As figure 1 displays, this research proposes that the success of political marketing is dependent on the actions and perceptions of parties and lead candidates. This proposition of potential drivers of success particularly includes the relative positioning for parties and social imagery for lead candidates. It is examined by identifying differences among parties and candidates as distinct categorical entities through various text analysis methods. Firstly, the conceptual framework (chapter 4.1) delineates what precisely constitutes success for political marketing in this research. Secondly, the relevance and validity of the chosen measurements (chapter 4.2) to identify potential drivers of success are demonstrated. The methodology explains the specific procedure for each measurement in detail (chapter 6). The choice of the central election topics for the topic-specific stance analysis is derived from journalistic election coverages (chapter 4.3). The specific hypotheses and expectations (chapter 4.4) are presented and reasoned.

4.1 Conceptual Framework

Table 1

Data basis for the Conceptual Framework (own representation, source: Politico 2021)

Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Election Result I September 24 th 2017	20.5%	33.0%	8.9%	10.7%	12.6%	9.2%
Campaign Start Polls December 8 th 2020	16.0%	36.0%	19.0%	6.0%	9.0%	7.0%
Election Result II September 26 th 2021	25.7%	24.1%	14.8%	11.5%	10.3%	4.9%
Legislature Period Delta	+5.2%	-8.9%	+5.9%	+0.8%	-2.3%	-4.3%
Campaign Period Delta	+9.7%	-11.9%	-4.2%	+5.5%	+1.3%	-2.1%
Election Winners New government	yes	no	yes	yes	no	no

As previously established, success in political marketing is constituted when the activities lead to the wanted appeal to its stakeholders (Hughes & Dann 2006, Newman 1999a, Newman & Perloff 2004, Lock & Harris 1996). Measured in voter behavior, success in political marketing is subsequently constituted when votes are won compared to the last election (1) or compared to the campaign start (2). Additionally, it can be considered a success if the party takes part in the new government coalition (3) because it allows the party to implement beneficial policies for its stakeholders. Applying this to the current German political landscape, only the Social Democrats (SPD) and Liberal Democrats (FDP) managed to be successful in all three dimensions (see table 1). Hence, these two parties will serve as representatives of success in political marketing for the 2021 federal elections in this research. There is no legal regulation in the German constitution that determines election campaigns' specific beginning or ending. The prevailing opinion is that the determination of the election date can be viewed as the starting point for the election campaigns of parties and candidates (Bundeswahlleiter 2020d). On December 8th, 2020, President Frank-Walter Steinmeier announced that the election would take place on September 26th, 2021 (Bundeswahlleiter 2020e). Hence, the campaign period is limited between December 8th, 2020, and September 26th, 2021.

4.2 Measurements

In addition to the content stances, the measurements of sentiments, emotions, and text similarity can be used to validly determine party positioning (Silva & Proksch 2021, Burst et al. 2021, Jolly et al. 2019, Laver et al. 2002). Based on the work of Giavazzi et al. (2020), which identified a shift in text similarity between German political parties following terror attacks, this research conducted a topic-specific analysis to identify similar shifts in central election topics. The following demonstrates the relevance and validity of sentiment and emotion analysis, stance analysis, and text similarity to identify potential drivers of success.

Sentiments and Emotions. Sentiments and emotions characterize two of the focal points in the history of modern political science. From a static point of view, emotions tie-up behavior and attitudes towards political actors as the affective “glue” to assure reliance on longstanding complexions (Sears 1993). From a dynamic point of view, emotions as affective responses play a crucial role in subsuming judgments of political actors and current circumstances (Marcus 2000, Lodge & Taber 1998, Hastie & Park 1986). These capabilities have been used many times in political marketing strategies, subsequently creating research areas like “affective polarization”, particularly on social media platforms like Facebook and Twitter (Park et al. 2021). Hence, a sentimental strategy expressed in sentiments and an emotive strategy expressed in emotions can be viewed as potential drivers of success in political marketing campaigns. Sentiments and emotions can be measured with sentiment and emotion lexicons (Mohammad & Turney 2013, Jockers 2020). This research utilizes the Syuzhet, Afinn, Bing and NRC lexicons (see chapter 6 for more details).

Stances. Political stances refer to publicly stated opinions by parties and their representatives and thereby their relative contentwise positioning compared to other parties (Chou et al., 2021). In Germany, the general stances of parties are regularly expressed in party programs (also called manifestos) approaching state or federal elections, in which parties explain their opinions and define their positionings in regards to specific topics (DW 2021). These stances change relatively little over time, as most partisans display longstanding loyalty in party allegiances, and repositioning is perceived as costly due to the often-occurring alienation of their voters in the process (Chou et al., 2021). However, the emergence of new parties has raised the

necessity for contentwise differentiation and the attractiveness for repositionings multiple times in German history, as the empirical evidence suggests regarding the Greens and the Left (Decker 2021b, Schumacher et al. 2013, Decker 2016, Decker 2013). The right-radical Alternative for Germany (AfD) has most recently raised the question for the established parties whether and how to recapture voters with repositionings. In line with this, recent empirical evidence suggests that AfD voters can be persuaded when established parties accommodate their preferences (Chou et al., 2021). Hence, a stance strategy expressed in ideological positioning and text similarity can be viewed as potential drivers of success in political marketing campaigns.

Stances can be measured based on expert evaluations of their ideological position and based on their text similarity to attain a relative positioning (Burst et al. 2021, Giavazzi et al. 2020, Jolly et al. 2019, Laver et al. 2002). Both approaches are utilized in this research (see chapter 6 for more details).

4.3 Central Election Topics

Three topics were selected for the topic-specific stance analysis. Each of them was chosen due to its arguably high relevance for the elections and the justifiable expectation of differences in stances across parties.

Corona. This topic was selected because its regulations altered the living conditions throughout the entire campaign period. This omnipresence especially applies during the time of the second "hard" lockdown with severe social distancing measures (Dec. 2020 – Apr. 2021) and later on in the discussion about the vaccination policies (Wiwo 2021b). Political actors often displayed different stances: e.g., portraying themselves as the defeaters of the pandemic or as defenders of constitutional freedoms and rights (Junginger 2021).

Environment. This topic was selected because the floods in July 2021 in western Germany and their aftermath caused a controversial societal debate about climate change in public discussions and political parties. The event caused 180 deaths and over 4.5 billion euros in damages (bpb 2021c). Fiercely debated on social media as one of the most controversial moments in the campaign period is the laughter of Conservative lead candidate Laschet (CDU) during a speech of German president Steinmeier expressing his condolences to the victims of the

catastrophe at the site of the disaster (Lütz 2021, Handelsblatt 2021). The range of stances is extensive: the Greens declared climate change as the essential topic in their campaign, and the Alternative (AfD) barely acknowledges its existence and importance with their program stances (BBW 2021).

Digitization. This topic was selected because it is considered as one of the most pivotal challenges of the 21st century. With the Covid-19 pandemic as a catalyst, the accelerated digitization is omnipresent in society today: home office, hybrid working, digital university, virtual conferences and many more peculiarities have drastically gained importance during the last two years (Brings & Weber 2021). Political parties have repeatedly stressed the importance of the digitization, even though their stances differ considerably: the Social Democrats (SPD) guarantees the provision of fast Internet for everyone, and the Greens even want a legal entitlement. The Left wants broadband and mobile networks to be in public hands, while the Alternative (AfD) wants to promote regional structures more strongly in the expansion process (Tagesschau 2021). The topic is particularly interesting for young voters (Gensing & Reisin 2021).

4.4 Hypotheses

Sentiments and Emotions

H1: The sentiments and emotions utilized in the tweets of official party and candidate accounts differ across parties.

H2: The scores for negative sentiment and emotions connoted in the tweets of official party and candidate accounts are highest for the political fringe parties (The Left & AfD).

The rise in political polarization refers to a contemporary phenomenon that describes the development of increasingly unfavorable sentiment towards opposite political parties or ideologies. There is substantial empirical evidence that social media as a “filter bubble” or “echo chamber” has been a significant driver for this phenomenon. Social media makes it exceedingly simpler to connect with like-minded people, find confirmation for the own selective attention in a process that is often referred to as self-sorting or enclave extremism (Boxell 2017, Sunstein

2007). On the one hand, social media makes information considerably more accessible to voters. However, on the other hand, it also augments information asymmetries by amplifying the views and influence of minority and extremist voters (Matejka & Tabellini 2020). Whereas incumbent government parties use the most positive sentiment in their activities, the rhetoric of radical or populist parties elicits anxiety, anger, resentment, and nostalgia (Betz 2020, Crabtree et al. 2020, Salmela & von Scheve 2018). Firstly, this gives reason to expect that sentiments and emotions differ depending on the political party and candidate. Secondly, this gives reason to expect that the two fringe parties in the German political system (The Left & AfD) utilize negative sentiment and emotions the most in their political marketing activities.

Stances

H3: For central election topics, the stances of parties change towards the stances of the parties that are the public opinion leader for the topic in the public discussion.

Following exogenous events that are perceived as threats from other cultures (e.g., terror attacks), a shift in text-similarity on Twitter among German political parties can be substantiated. This shift is characterized by the assimilation of party language towards the right-wing party Alternative (AfD) by the other five major political parties (Giavazzi et al. 2020). This finding enables two critical inferences. Firstly, following a sudden, exogenous event that affects the entire society (e.g., terror attack), all parties seem to assimilate their positions and language towards the same relative positioning. Secondly, this assimilation arguably strengthens the party's public support that initially represented the eventual consensus.

Applying this to the 2021 German Federal Election campaigns could mean that there have been similar shifts in text-similarity of party language regarding relevant exogenous events this year. As the reasoning for the central election topics displays, the Covid-19 related lockdown at the beginning of 2021 and the floods in western Germany in July 2021 can be viewed as such relevant exogenous events. Subsequently, being a public opinion leader for a central election topic could be a significant success driver in the political marketing for election campaigns (bpb 2021c).

5. Data

In this chapter, a summarizing overview of the utilized data sets of this research shall be presented. To gain an easier understanding, the explanations for the raw data collection (chapter 5.1.1), data cleaning (chapter 5.1.2), as well as data preparation and variable operationalization (chapter 5.1.3), are displayed separately for Twitter and Party Program Data because they differ from each other considerably. For the goal to reproduce this research analysis or reuse the utilized data sets, encoding readers must be considerate of German special characters (e.g., ä) to attain an undistorted display of the data.

5.1 Data Collection and Transformation

5.1.1 Raw Data Collection

Twitter Data

The Twitter data was collected using the Twitter Search API and an R download script. In this download script, several data partitions were scraped. Firstly, within the four weeks before the election date on September 26th, 2021, tweet data based on relevant hashtag queries was collected, which was subsequently used to analyze the user-generated discussions. The relevant hashtags were drawn from the trending hashtags lists on the real-time social media dashboard specifically created for the 2021 federal election by the Berlin newspaper Tagesspiegel in cooperation with Democracy Reporting International and Stiftung Mercator (Tagesspiegel 2021). The lists of hashtags can be viewed in Appendix A. Secondly, the political marketing data was collected from the timelines from verified Twitter accounts of parties and candidates (bpb 2021a). Every scraped data set was exported as a CSV file in UTF-8 encoding.

Party Program Data

The texts of the party programs for 2017 and 2021 were collected directly from the official websites of the major political parties as PDF files (AfD 2017, AfD 2021, CDU/CSU 2017, CDU/CSU 2021, FDP 2017, FDP 2021a, Gruene 2017, Gruene 2021, Linke 2017, Linke 2021, SPD 2021). Except for the 2017 SPD program, which was not available for download anymore. Instead, the text was obtained from the ManifestoR database (Burst et al. 2021, SPD 2017). Lastly, the downloaded PDF files were converted into TXT files.

5.1.2 Data Cleaning

Twitter Data

Firstly, the data was adjusted by dropping irrelevant columns for this analysis. This step reduced the number of variables in the political marketing data sets from 90 to 37 and the discussion data sets from 74 to 32. With a matching number of columns, all data sets were merged into one political marketing (timeline queries) and one discussion (hashtag queries) data set. Thirdly, a time adjustment was made by excluding tweets outside the campaign term between December 8th, 2020, and September 26th, 2021. Fourthly, the respective text variables were cleaned by removing unwanted special characters, smilies, numbers, HTML codes, URLs, and excess white spaces using regular expressions and the dplyr manipulation language. Lastly, NA values were handled by replacing them with 'None' for character variables and '0' for numeric variables. The list of kept variables for both data sets can be viewed in the summary statistics in Appendix A.

Party Program Data

To make the text data amenable to natural processing tools, punctuations, numbers, symbols, URLs, and German and English stopwords were removed except German special characters (e.g., ä, ö and ü).

5.1.3 Data Preparation & Variable Operationalization

Twitter Data

Firstly, an id variable for parties was created to identify each tweet's party allegiance quickly. The resulting categorical variable has six distinct character values representing the respective party: AfD, CDU, FDP, Gruene, Linke, and SPD. In the political marketing data set, this was operationalized by using the origins of the respective initial data stems that resulted from the necessary separation in the collection process. The discussions data set was operationalized by using a structure detect code based on keywords for the respective party (party name, lead candidate name, and nicknames for party supporters). Additionally, in the discussions data set, a second categorical variable with ten distinct values was created for each of the lead candidates of the parties: Baerbock, Bartsch, Chrupalla, Dobrindt, Habeck, Laschet, Lindner, Scholz, Weidel, and Wissler. Secondly, dummy variables were created for each categorical value of the party id and candidate id variable, resulting in 16 variables that display the value 1 if the condition is

satisfied (e.g., tweet of SPD candidate) and the value 0 if the condition is not satisfied. Thirdly, an election id variable was created that displays the value winner if the party is part of the new government coalition (SPD, Greens & FDP). The value loser was assigned if the party is part of the opposition of the new parliament (CDU/CSU, AfD & Left). Fourthly, the same logic was applied to creating the campaign id variable, which displays the value gained, if the party was able to gain percentual voter support during the campaign term (SPD, FDP & AfD). The value lost was assigned if the party lost percentual voter support between the campaign beginning and the actual election (CDU, Greens & Left).

Party Program Data

The preparation of party programs and tweets for the stance analysis followed five steps. Firstly, the relevant texts were saved in a data frame and converted into a corpus. In this research, the relevant texts are the 2017 and 2021 party programs and the collected tweets from political marketing. The party programs were loaded in as character values (e.g., PP_FDP_2021). The Twitter texts were aggregated for the unique entity (either entire party or screen name) by collapsing the rows and subsuming the contents of the text column into one object with the delimiter of a space. The collapsing is done so that each entity has one text (aggregation of tweets) that is scored as a whole. Secondly, the reference scores for the texts with known orientation were added to the corpus. In this research, the reference scores of the Manifesto Corpus and 2019 Chapel Hill Expert Survey were utilized (Burst et al. 2021, Jolly et al. 2019). Thirdly, the texts within the corpus were tokenized. Fourthly, the text tokens were converted into a document-feature matrix (DFM). Fifthly, these document-feature matrices were utilized in the analysis functions of wordscores and similarity between authors. The theoretical foundation for this procedure is explained in the methodology (chapter 6).

5.2 Data Overview

The main takeaways (figure 2, tables 2 and 3) derived from the overview of the collected data are presented in the following. The full summary statistics and additional descriptive visualizations can be found in Appendix A. As figure 2 displays, apart from two short-term sharp declines in volume around Christmas 2020 and early April 2021, it can be observed that all six major parties steadily increased their tweet volume approaching the 2021 federal elections - especially the

Greens intensified their efforts within the last four weeks. As table 2 displays, the Social Democrats (SPD) exhibit the highest average favorite and followers count, whereas the Greens have the highest average retweet and friends count. It is also noteworthy that only the fringe parties (AfD & The Left) have a tweet without hashtag share that is considerably below 50 percent.

Table 2 – Political Marketing Data (65 variables x 218,256 observations)

Summary statistics (own representation)







Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Obs.	39435	40133	47116	41364	19361	30847
						
Reach (average per account)						
Favorites	144.91	60.75	91.82	61.07	133.40	62.76
Retweets	85.89	83.26	138.33	74.47	107.26	88.03
Followers	62224.65	46932.89	44464.20	29245.91	23505.91	35611.61
Friends	1407.92	1305.86	2048.05	1768.77	945.58	1192.26
Dissemination (average per account)						
Text Width	149.30	151.39	153.11	152.85	167.88	173.52
Hashtag Count	0.82	0.93	1.07	0.92	1.86	1.34
Tweets with no hashtag	58.39%	53.46%	49.16%	55.60%	41.53%	40.61%

Table 3 – User Generated Discussions Data (70 variables x 297,775 observations)

Summary statistics (own representation)

No Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
118438	19127	67417	19509	10966	54907	7411
Reach & Dissemination (average per account)						
Favorites	2.85	2.38	2.87	3.26	2.51	2.61
Retweets	99.97	127.04	95.58	162.23	97.05	32.96
Text Width	148.96	156.19	151.68	147.90	147.52	154.53

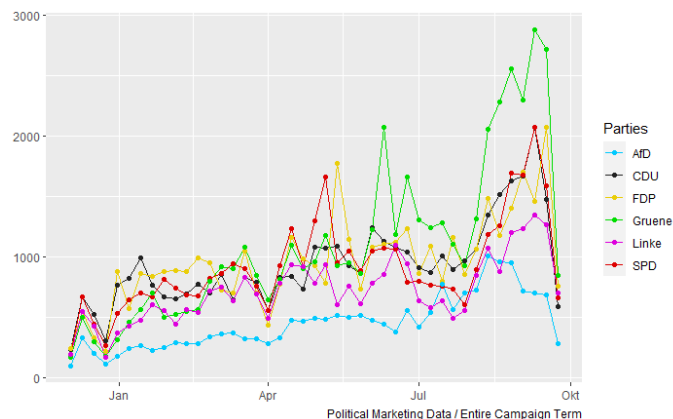


Figure 2: Total volume of tweets by the political marketing accounts (own rep.).

6. Methodology

Sentiment & Emotion Analysis

The sentiment and emotion analysis was conducted with the Syuzhet Package. This package enables sentiment extraction based on four different sentiment lexicons (Syuzhet, Afinn, Bing, and NRC) (Jockers 2020). Additionally, the package enables the extraction of emotions with the NRC Emotion Lexicon. Usable in over 40 languages, including German, this lexicon annotates words (also called unigrams) with their associations with two sentiments (positive, negative) and eight emotions (anger, anticipation, disgust, fear, joy, sadness, surprise, and trust). The lexicon was developed manually by two computational linguists and crowdsourcing (Mohammad & Turney 2013, Jockers 2020). The computation of sentiments and emotions is simple: the words of the character input vector (in this case: a singular tweet) are aligned with the words in the respective lexicon. If a word in the input vector matches a word in the lexicon, the annotation of the word is added to the respective tweet. If a sentiment or emotion applies multiple times in a singular tweet, the algebraic sum is added to the respective tweet (Naldi 2019). As formulas 1 and 2 display, two additional indices were created to add a simplified basis for interpreting the analysis results.

Sentiment Index. measures the avg. sentiment score of the deployed lexicons per tweet.

$$SENI = \frac{Syuzhet + Bing + Afinn + NRC}{4}$$

Intensity Index. measures the sum of deployed emotions per tweet.

$$INTI = \frac{\sum Emotion\ Scores}{8}$$

Stance Analysis

In this research, **stances** refer to the relative positioning of parties based on wordscores. Wordscores are a commonly utilized content analysis method for inferring policy positions for virgin documents based on the scores for words from documents with known scores. Hence, wordscores display a method that can be deployed to attain a relative positioning of parties based on their used language (Lowe 2017). The method utilized in this research generally follows three steps.

Firstly, reference texts (e.g., 2017 party programs) are obtained with a priori known positions. These a priori positions can be derived from expert evaluations. This research utilized two expert evaluations for comparative purposes: the Manifesto Corpus and the 2019 Chapel Hill Expert Survey (Burst et al. 2021, Jolly et al. 2019).

The *Manifesto Corpus* is a multilingual, annotated collection of electoral programs, supported by the two leading German scientific organizations Wissenschaftszentrum für Berlin (WZB) and Deutsche Forschungsgemeinschaft (DFG) (Burst et al. 2021, WZB 2021, DFG 2021). The Manifesto Corpus was chosen as an expert evaluation because its CMP coding (left-to-right positioning score) directly applies to specific party program texts.

The *2019 Chapel Hill Expert Survey* contains evaluations and trend files (1999-2019) of 277 parties on political ideology. The 2019 Chapel Hill Expert Survey was chosen because it is the most comprehensive expert survey on political parties in Europe (Jolly et al. 2019). In this research, the Manifesto Corpus was utilized to determine the ideological position of parties for each federal election since 1990. However, because its CMP coding operationalizes left-wing positionings with negative scores, it cannot be utilized in the wordscore algorithm (Quanteda 2021). Hence, the 2019 Chapel Hill Expert Survey was deployed as reference scores for the 2017 party programs instead because it operationalizes with a scale of 0 to 10. In this scale, the value 0 represents holistically left, and the value 1 represents holistically right (Jolly et al., 2019).

Secondly, wordscores are generated from the reference texts. Thirdly, the to-be-scored “virgin” texts are scored using the previously attained wordscores (Laver et al. 2002, Burst et al. 2021, Jolly et al. 2019). The usage of corpora and document-feature matrices is essential for deploying stance analysis. A **corpus** represents the storage library for the UTF-8 encoded text derived from the original documents, along with meta-data at corpus- and document-level. Hence, a corpus is the data source for a **document-feature matrix (DFM)**, in which the documents represent the rows and the features (broad term analysis variables, e.g., syntactic dependencies) are displayed by the columns. Subsequently, each row displays one unit of analysis, and each column represents one analysis variable. For stance analysis, the text in document-feature matrices must be prepared by tokenization, lower-casing texts, and removing punctuation and stopwords. **Tokenization** refers to the extraction of singular, unique words as ‘tokens’ and consequently represents one of the essential requirements for computing text similarities and similarity between authors (Quanteda 2021a).

Lexical Diversity & Similarity Between Authors (SBA)

Lexical Diversity can be measured based on the same tokenized document-feature matrix (DFM) as the general stance analysis wordscores. The measure calculates the lexical richness based on the number of unique types of tokens and the length of a document. It is a valuable indicator to measure linguistic skills or the complexity of ideas (Quanteda 2021b, Quanteda 2021c).

The measurement of Similarity Between Authors requires grouping the document-feature matrix by a specific entity. This research applies the Twitter handle (also known as screen name) as the grouping entity. This approach enables us to calculate the similarity between authors within each party. The SBA is measured in a distance matrix. This distance matrix can be visualized with dendrograms (Quanteda 2021d).

Topic-Specific Stance Analysis

The general stance analysis generates wordscores for the 2021 party programs and campaign tweets based on the reference texts (2017 party programs) and their reference scores. The topic-specific stance analysis uses the campaign tweets and their score as the reference material for the tweets solely about the specific topics.

7. Results

In the following, the results of this research are presented. Every table and graph can also be found with additional information in Appendix B.

7.1 Sentiment & Emotion Analysis

Table 4 – Sentiment Index (SENI) and Intensity Index (INTI)

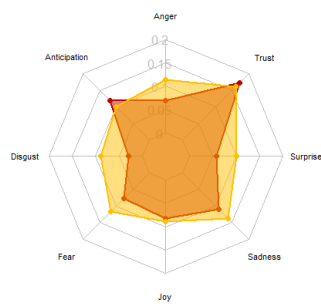
Average index score for each party. Highest (green) and lowest scores (red) are highlighted (own rep.).

Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
SENI	-0.6251	-0.6292	-0.6772	-0.6981	-0.7348	-0.8205
Strategy						
SENI	-0.7552	-0.7963	-0.7209	-0.7894	-0.7681	-0.9238
Image						
INTI	0.8718	0.8766	0.8803	0.8609	0.8767	0.9161
Strategy						
INTI	1.1318	0.9679	0.8249	1.0771	0.9766	0.8699
Image						

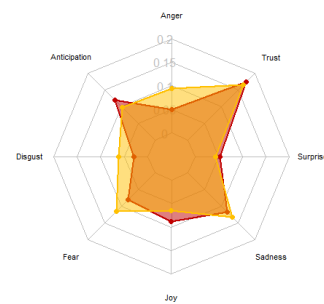
This research provides empirical evidence that sentiments and emotions expressed in tweets differ depending on parties and lead candidates (hypothesis 1). In line with current research, table 4 shows that the two incumbent government parties (CDU/CSU & SPD) have utilized more positive sentiment than the opposition parties (Crabtree et al. 2020). Compared to the public image in the user-generated discussions, this picture changes. Whereas the Social Democrats (SPD) have the second most positively connoted public image behind the Greens, only the Left has more negatively connoted tweets than the Conservatives (CDU/CSU). The sentimental and emotive strategy of the fringe parties displays the expected results (hypothesis 2): the Left exhibits the highest scores for negative sentiment and the emotion sadness. The Alternative (AfD) showcases the lowest score for positive sentiment and the highest scores for the emotions anger, disgust, and fear in their political marketing activities (see fig. 3 and table 4).

The comparison between the emotive strategies and the induced emotions in user-generated discussions is also intriguing. All parties induce more emotions in the user-generated discussions than their emotive strategy in the political marketing utilized: the Greens are the notable exception here, as they induce fewer emotions in the user-generated discussions than their

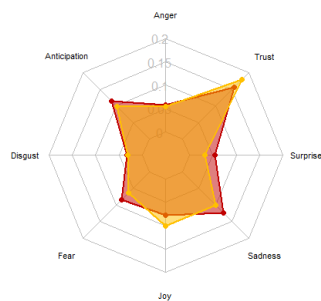
emotive strategy displays. Interestingly, Social Democrats (SPD) and Liberal Democrats (FDP) utilized the two most minor amounts of emotions in their strategies. However, they induced the two highest amounts in the user-generated discussions. The epistemic value expressed in the utilization of the emotion surprise played a crucial role for the Social Democrats (SPD). They display the highest score for this emotion both in their marketing activities and user-generated discussions. Additionally, as it pertains to the Left and the Greens, their emotive strategy almost matches their public image in terms of emotions.



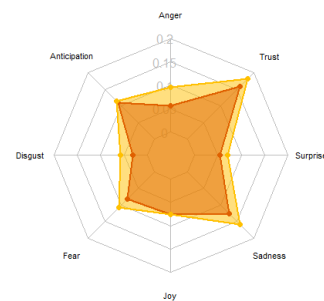
Social Democrats (SPD)



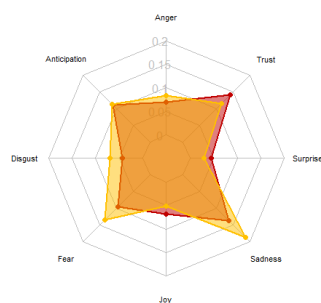
Conservatives (CDU/CSU)



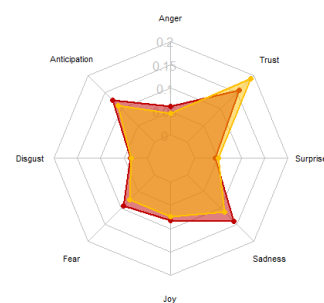
Greens



Liberal Democrats (FDP)

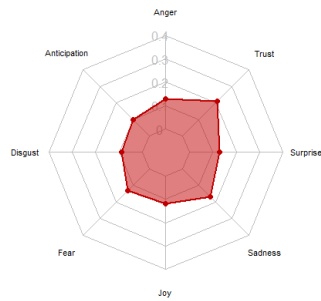


Alternative (AfD)



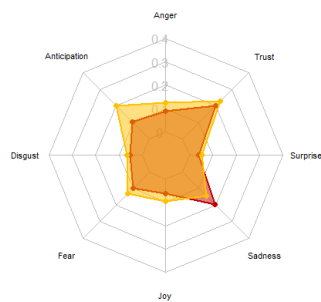
The Left

Figure 3: Average emotion score for each party. Emotive Strategies (red) of the political marketing in comparison with the public perception in the user-generated discussions (golden) (own rep.).



Social Democrats (SPD)

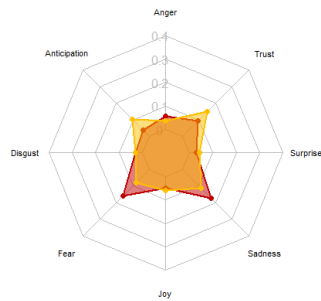
Olaf Scholz (red)



Greens

Annalena Baerbock (red)

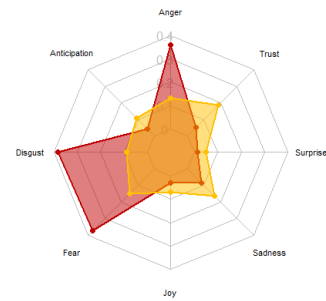
Robert Habeck (golden)



Alternative (AfD)

Tino Chrupalla (red)

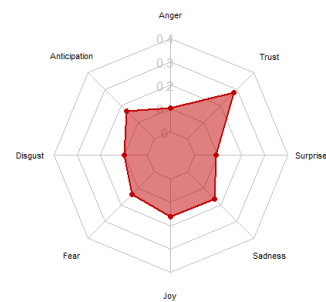
Alice Weidel (golden)



Conservatives (CDU/CSU)

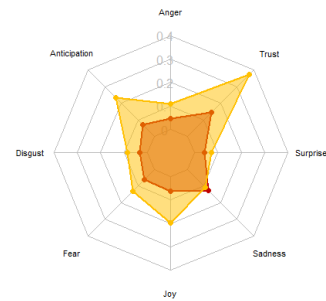
Alexander Dobrindt (red)

Armin Laschet (golden)



Liberal Democrats (FDP)

Christian Lindner (red)



The Left

Dietmar Bartsch (red)

Janine Wissler (golden)

Figure 4: Public Party and Candidate Images expressed in emotions (own rep.).

Comparing figure 3 and 4, the Alternative (AfD) and its two lead candidates induce considerably fewer emotions as the party utilized in their marketing activities. The tweets about lead candidate Alice Weidel (AfD) are most positively connoted out of all lead candidates in the Sentiment

Index, subsequently being perceived similarly positive as Christian Lindner (FDP), who displays the highest amount of positive sentiment. These results indicate that the lead candidates of incumbent opposition parties (Greens, Left, FDP & AfD) were rewarded with a positively connoted image of their lead candidates for their criticism of current politics by the incumbent government (CDU/CSU & SPD). Additionally, even though Laschet (CDU) was the target of situational contingency (e.g., #laschetdarfnichtkanzlerwerden), Dobrindt (CSU) was perceived considerably worse in the public eye, displaying the worst overall public candidate image in terms of sentiments and emotions.

7.2 General Stance Analysis

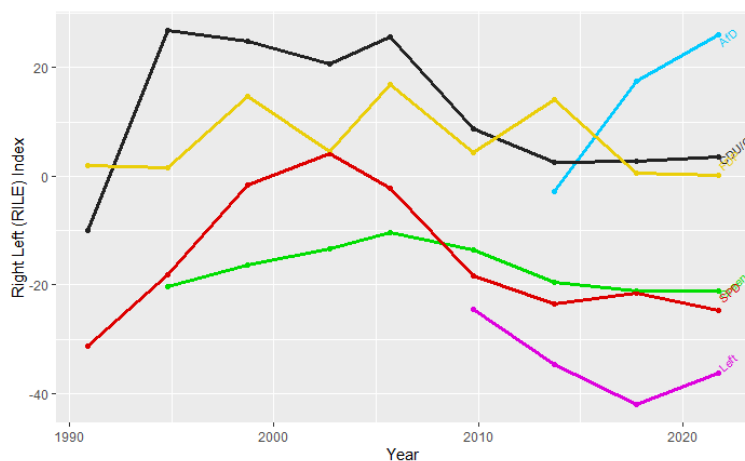


Figure 5: Relative Positioning of Major German Parties based on their overall ideological position (1990-2021). Measured with the ManifestoR RILE Index based on party programs for each federal election (own representation based on Burst et al. 2021)

This research utilizes two sources for the relative positioning of parties. Firstly, the Manifesto Corpus is deployed to attain the relative positioning based on the overall ideological position. Secondly, the 2019 Chapel Hill Expert Survey is exerted to attain the relative positioning based on their wordscores. As figure 5 displays, the ManifestoR RILE Index indicates that the democratic middle of Conservatives (CDU/CSU), Liberal Democrats (FDP), Greens, and Social Democrats (SPD) have roughly maintained their ideological positions from 2017 with their respective 2021 party programs. The Alternative (AfD) has opted for a right-wing shift and a more radical ideological position. In contrast, their left-wing counterpart, the Left, chose to abate some of their 2017 positions for a more moderate ideological position in 2021.

Table 5 – Reference Scores

Initial scores for ideological positioning of the reference texts (2017 party programs). Scores are derived from the average overall ideological score of the 2019 Chapel Hill Expert Survey (own rep. based on Jolly et al. 2019). The used Chapel Hill scale ranges from the value 0 representing holistically left-wing to the value 10 representing holistically right-wing positioning.

SPD	CDU/CSU	Greens	FDP	AfD	Left
3.619	6.524	3.238	6.429	9.238	1.429

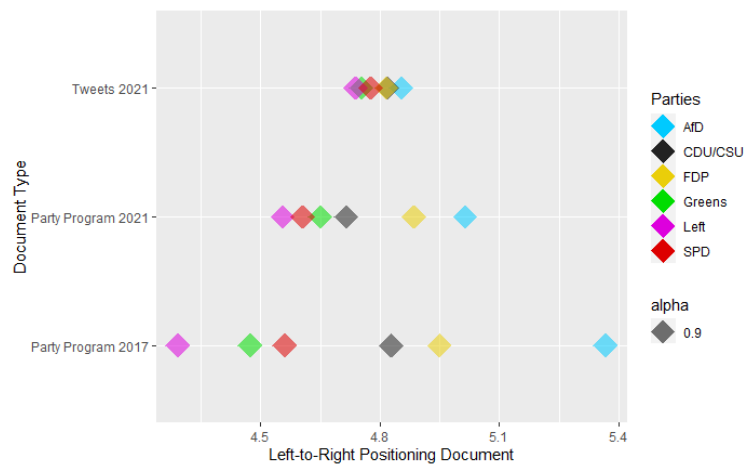


Figure 6: Relative Positioning of Major Political Parties based on their general wordscores (own rep.).

As figure 6 displays, the relative positionings offer several interesting insights. Firstly, the left-to-right range in relative positioning has decreased between the 2017 and 2021 party programs by 42.69%, from 1.0757 to 0.4592. Secondly, the left-to-right range decreased even further from the 2021 party programs compared to the campaign tweets by 24.93% from 0.4592 to 0.1145. These results indicate that between 2017 and 2021, the used wordings in party programs and tweets have assimilated across all parties, resulting in a very similar relative wordscore positioning for all parties. Thirdly, the relative positioning of all documents indicates that the political language in Germany is slightly skewed to the left: the median wordscore (4.797) for all documents is allocated on the left of the political middle equivalent value of 5. Fourthly, as established in the general stance analysis (fig. 5), the Alternative (AfD) has been evaluated to have conducted a right-wing ideological shift between the 2017 and 2021 federal elections (Burst et al. 2021). Possibly controversially, this research also indicates that they performed this right-wing shift with a simultaneous left-wing shift in text similarity (fig. 6).

Table 6 – Lexical Diversity

Calculated for party programs (PP) and campaign tweets (T). Highest (green) and lowest scores (red) are highlighted (own rep.).

Party	SPD	CDU	Greens	FDP	AfD	Left
PP 2017	0.4622	0.4392	0.3959	0.4728	0.6152	0.5036
PP 2021	0.4905	0.3969	0.4087	0.4960	0.6035	0.4686
T 2021	0.1279	0.1200	0.1047	0.1181	0.1619	0.1222

Table 7 – Similarity Between Authors (SBA)

Differences in text similarity translate into distance between authors. The respective party dendrograms can be found in Appendix B. Highest (green) and lowest scores (red) are highlighted (own rep.).

Party	SPD	CDU	Greens	FDP	AfD	Left
Max. Dist.	767.30	518.83	714.76	574.00	454.81	578.57
Median	196.60	202.20	233.52	222.94	145.30	257.82
Dist.						
Mean	219.40	209.23	270.75	242.75	176.66	281.75
Dist.						
IQR Dist.	137.10	166.89	137.10	124.05	167.02	180.06
Min. Dist.	13.00	13.42	34.42	23.77	15.03	41.61

The results for lexical diversity and similarity between authors are surprising. As table 6 displays, the lexical diversity of all parties has decreased between 2017 and 2021 party programs, and again in comparison to the campaign period tweets. Particularly the Alternative (AfD) proves to be a difficult conundrum to interpret. As table 7 displays, the Alternative (AfD) showcases remarkable coherence among their candidates with the highest Similarity Between Authors (SBA). This result is surprising, as the party also scores the highest lexical diversity, indicating the most voluble language of all parties. This counterintuitive pairing of the highest SBA and lexical diversity for the Alternative (AfD) could indicate considerably higher coordination in the contents and dissemination of tweets compared to the other parties to achieve this result. However, a deviation might have been caused by the smaller sample size for the Alternative (19,361 observations) compared to the sample sizes of the other five parties (30,847-47,116 observations, respectively).

7.3 Topic-Specific Stance Analysis

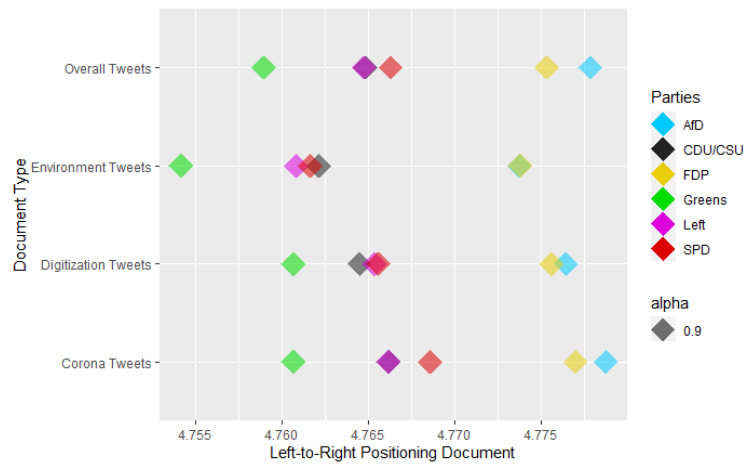


Figure 7: Topic-specific tweet stances for the topics Corona, Environment and Digitization in comparison to the respective overall stances for all tweets (own rep.).

Similar to the findings of Giavazzi et al. 2020, this research provides empirical evidence that there were several intriguing shifts in topic-specific stances in regards to the topics of the Covid-19 pandemic (topic 1), environment (topic 2), and digitization (topic 3) by the six major political parties (hypothesis 3). The results indicate that two major exogenous events played a crucial role in the political marketing activities within the campaign term, which enables a rough distinction of the campaign activities into three phases: the Covid-19 lockdown and the floods in western Germany.

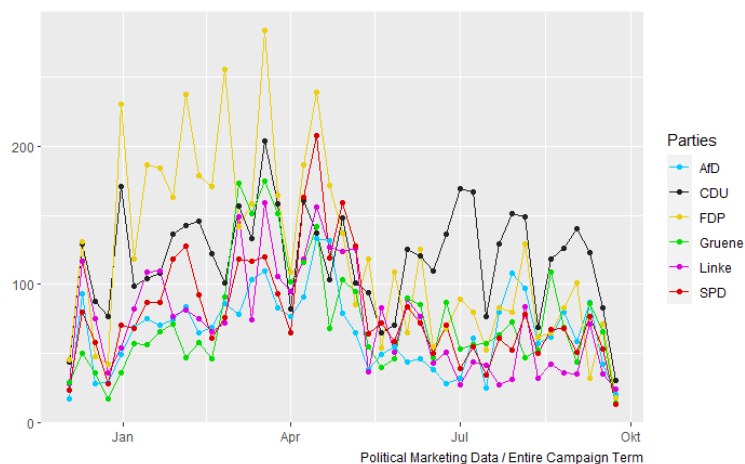


Figure 8: Volume of Corona-related tweets during the entire campaign period (own rep.).

The *first campaign phase* (Dec. 2020 – Apr. 2021) was predominantly determined by the alternate living conditions resulting from the regulations of the “hard“ Covid-19 lockdown (Wiwo 2021b). As figure 8 displays, particularly the Liberal Democrats (FDP) and Alternative (AfD) exercised an intensified volume of covid-19-specific tweets compared to other parties' efforts in this time frame. This finding is also reflected in the corona-specific stances, indicating a shift in text-similarity (figure 7). Arguably led by the most right-wing stances of FDP and AfD, the corona-specific tweets of all parties are positioned more right-wing than their respective overall tweet stances. With the lockdown expiring in April 2021, the volume of corona-related tweets considerably decreased, just like the public opinion leadership changed, which was now taken by the Conservatives (CDU/CSU) measured in tweet volume up until the elections.

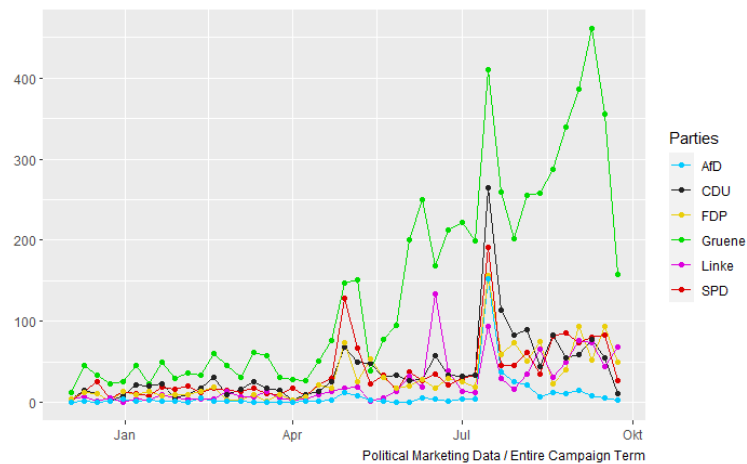


Figure 9: Volume of environment-related tweets during the entire campaign period (own rep.).

The *second campaign phase* (May 2021 – Aug. 2021) marks the replacement of Corona as the predominant campaign topic, as the tweet volume for the environment- and climate-change-related tweets drastically increase. Easily identifiable in figure 9: a dramatic, short-term spike in tweet volume at the time of the floods in West Germany and Central Europe in late July 2021 (bpb 2021c). As expected, the Greens pushed this topic forward the most in the public discussion, as their environment-related tweet volume in the second and third campaign phase by far exceeds the efforts of all other parties. This finding can also be measured in stances (figure 7). Arguably led by the most left-wing stances of the Greens, the environment-specific tweets of all parties are positioned more left-wing than their respective overall tweet stances.

The *third campaign phase* (Sep. 2021) characterizes the global maximum of digitization-related tweets volume (figure 10), yet mainly driven by the Conservatives (CDU/CSU). The corresponding stances offer an ambivalent scenery (figure 7): the digitization-specific stances of Conservatives (CDU/CSU), Social Democrats (SPD), and the Alternative (AfD) display a more left-wing relative positioning compared to their respective overall stance positions. In contrast, the topic-specific stances of Liberal Democrats (FDP), the Left, and the Greens are all positioned more right-wing than their respective overall stance positions.

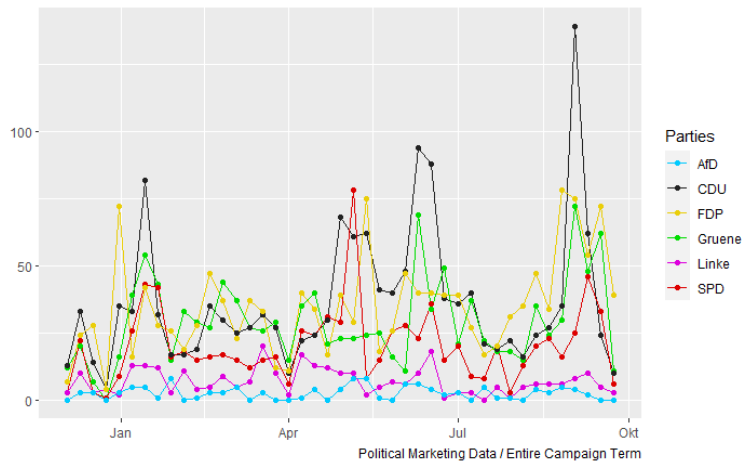


Figure 10: Volume of digitization-related tweets during the entire campaign period (own rep.)

8. Discussion

In this chapter, the results of this research shall be summarized, as well as contextualized with possible interpretations and relevant research findings. To round off this work, the academic and practical relevance of this research is highlighted in the marketing implications (chapter 8.2) and suggestions for future research (chapter 8.3).

8.1 Summary

Empirical evidence suggests that a substantial amount of voters is susceptible to persuasion: 23% of polled German voters were uncertain of their voting decision two weeks before the election, a similar share of 10 to 20% of undecided voters can also be established in other European countries like Italy and France (Spiegel 2021, Alam & Riva 2019, La Repubblica 2006, Mannheimer 2003). Hence, political marketing should be considered a suitable tool to gain voter support and gain a competitive edge against other parties – particularly in close elections. In characterizing the potential drivers of success of Social Democrats (SPD) and Liberal Democrats (FDP), it should be noted that the two parties followed different approaches, likely as a result of their difference in incumbent status (government vs. opposition). The Social Democrats (SPD) generated the highest interaction in terms of favorite and follower counts per candidate profile with the shortest and most positively connoted tweet texts. The epistemic value expressed in the utilization of the emotion surprise played a crucial role for the Social Democrats (SPD). They display the highest score for this emotion both in their marketing activities and user-generated discussions. Embodying the motif of freedom, The Liberal Democrats (FDP) positioned themselves as the arguable public opinion leader in the topics of the Covid-19 pandemic with their stances against lockdown measures - thereby criticizing the activities of the incumbent government (FDP 2021b, FDP 2021c, Welt 2021).

The results of this research allow several fascinating interpretations. Firstly, the increasing text similarity across parties could result from a political power struggle in which all parties attempted to gain voter support in the “middle“ of society. This conclusion is arguable because the fringe parties entered into a competition with the more centralistic parties for the same voter groups by displaying similar stances and similar language. Subsequently, it would represent a development in which parties would be less discernable by contentwise stances and wordings.

However, this interpretation requires further empirical evidence. Secondly, the simultaneous shifts of intensified right-wing ideology and left-wing language could indicate that the right-wing Alternative (AfD) attempted to mask their more radical views with more moderate language, arguably hiding their actual positions. Thirdly, the counterintuitive pairing of the highest SBA and lexical diversity for the Alternative (AfD) could indicate considerably higher coordination in the contents and dissemination of tweets compared to the other parties to achieve this result. These particular indications would be in line with the presented contemporary research. Researchers were able to link the majority of hate-speech creating accounts on social media to the Alternative (AfD) and Identarian Movement. Additionally, they were able to identify a high degree of coordination for those two groups on various social media platforms (Geschke et al., 2019, Kreißel et al., 2018).

8.2 Marketing Implications

In general, using situational contingency was a popular tool to decrease the eligibility of particular candidates and parties in the 2021 campaigns. Arguably successful, as both major implementations have been proposed as drivers for substantial voter fluctuation. On the one hand, there is the supposedly terrifying “slide to the left” expressed in a very left-oriented government coalition (SPD, Greens & Left) stirred up by the Conservatives (CDU/CSU) days before the election. On the other hand, there are individual campaigns against supposedly unbearable candidates Laschet (CDU/CSU), Baerbock (Greens), and Scholz (SPD) (Kahrs 2021). Closely before the elections, polls regarding the perceptions of the lead candidates underlined the potential importance of social imagery in voting decisions. While the majority of polled voters attested Social Democrat Olaf Scholz (67%) the aptitude for the position of chancellor, his rivals Armin Laschet (29%) and Annalena Baerbock (23%) were perceived considerably less qualified for the position. The same picture arises in the individual chancellor preferences, where eventual chancellor Olaf Scholz (48%) was perceived more favorable than his adversaries Armin Laschet (24%) and Annalena Baerbock (14%) combined (Forschungsgruppe Wahlen 2021).

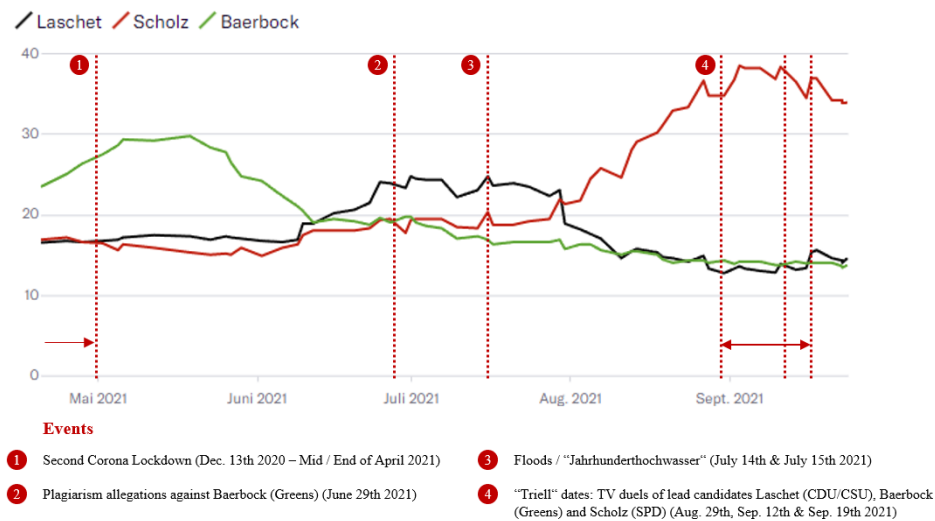


Figure 11: Polled preferences for the chancellor position in a comparison between lead candidates (Sources: Eckstein et. al. 2021, Wahlen 2021, bpb 2021c)

However, it should be noted that these perceptions changed considerably over time. As figure 11 displays, all three candidates were perceived as the most preferential at some time (Eckstein et al., 2021, Wahlen 2021, bpb 2021c). The eventual chancellor Olaf Scholz (SPD) gained his lead in preference in August 2021 after his adversaries suffered from personal controversies. On June 29th, 2021, plagiarism allegations against Annalena Baerbock (Greens) were made regarding her recently published book (Dake 2021). A couple of weeks prior to this, the lead candidate of the Greens was criticized for imprecise and false claims in her published curriculum vitae (Gensing 2021). As previously presented, the public image of Armin Laschet (CDU/CSU) suffered from the laughter at the catastrophe site of the floods during a speech of the German president Steinmeier, which was widely perceived as misplaced in the public discussion (Lütz 2021, Handelsblatt 2021). Interestingly, similarly negative publicity of Olaf Scholz's involvement in the CumEx tax scandal did not cause a substantial decrease in popularity (Ott 2021, Finanzwende 2021).

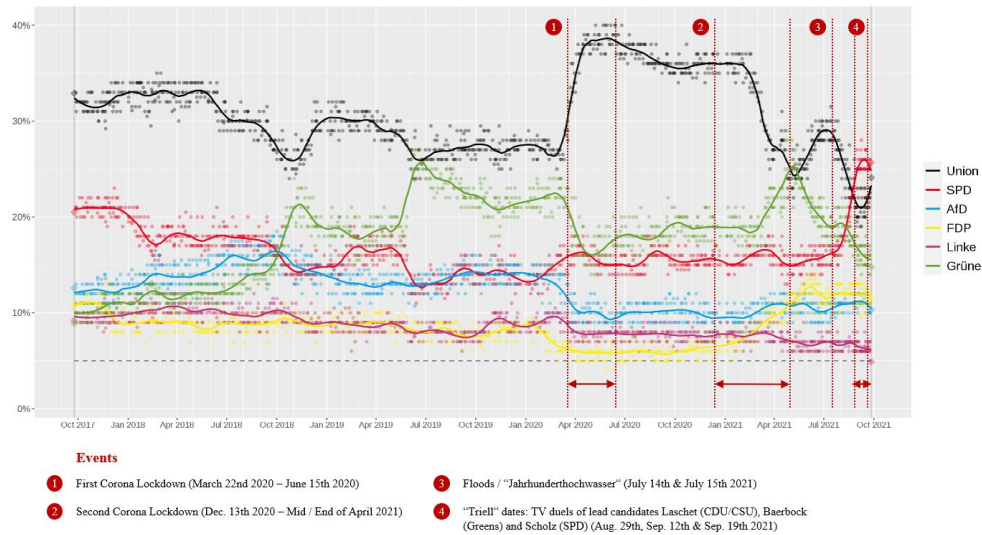


Figure 12: Regression of Poll Data for the Major Political Parties in the last legislature period (2017-2021)

(Sources: Politico 2021, Bundeswahlleiter 2021f, Eckstein et. al. 2021, Wahlrecht 2021, Wiwo 2021a, Wiwo 2021b, bpb 2021c)

In summary, multiple practical implications can be derived from these results. Firstly, as the Social Democrats (SPD) and Liberal Democrats (FDP) display, effective political marketing on Twitter is conductible with short texts and low lexical diversity, so the message is easily understandable. Secondly, the question of whether these messages should be in the majority positively or negatively connoted is primarily determined by the incumbent status of parties. Whereas incumbent government parties tend to use positive sentiment to appetize their policies to the public, incumbent opposition parties tend to use negative sentiment to be rewarded for governmental criticism and seen as a suitable political alternative. Thirdly, the public image of the lead candidate seems to be one of the central drivers for voter behavior. As figures 11 and 12 display, the increase in popularity of Social Democrats (SPD) and their chancellor candidate Olaf Scholz happened in similar proportions and timing in August and September 2021. Fourthly, situational contingency and epistemic value seem to be suitable tools to enhance the candidate image. Situational contingency accomplishes the depreciated public perception of opposing candidates while also stressing the differentiating value of the own candidate. Epistemic value offers the opportunity to persuade voters with uncertain voting preferences, as the employed emotion surprise is equipped to gain the voter's attention.

8.4 Limitations and Future Research

While interpreting the results of the sentiment and emotion analysis, the equating of particular sentiments or emotions with value judgments should be done carefully: negative sentiment does not necessarily result in an unfavorable perception (see the incumbent status of government vs. opposition). The same issue is also relevant for stance analysis. Left-wing and right-wing stances are hardly assignable to an unbiased, evaluative scale. While the method of wordscores is a powerful tool, its inherent inefficiencies in measurements have also been covered in contemporary research. Firstly, with being scaled in one dimension (e.g., left to right), the method assumes that parties only represent one side in an issue. However, this assumption often does not hold in reality, especially regarding divisively discussed issues within the party itself (e.g., refugee policies). If applicable, this limitation could be remedied by adding a second dimension (e.g., negative to positive) (Mikhaylov 2009). Secondly, if entire party programs are analyzed with wordscores, the inherent assumption is made that each mentioned topic or policy has the same importance for the parties' stakeholders. However, this is arguably not true: some policies are more dispensable than others, as the results of coalition negotiations display. Thirdly, the overall absolute position of wordscore positions should be interpreted carefully, as the absolute difference between parties is often only decided by small decimal figures. Hence, this research only analyses the relative changes in stances and not their absolute amounts. This assessment is consistent with contemporary, relevant research: even though the wordscore method generates similar estimates for party positionings in the left-to-right dimension in different countries, the degree of congruence differs considerably across countries, party types, and party programs that they are based on (Bräuninger et al. 2013).

Based on the results of this research, a responsive relationship between tweet volume of parties and exogenous events relevant for election topics can be asserted - as illustrated with the Covid-19 lockdown and July floods (*Jahrhunderthochwasser*) (bpb 2021c). Additionally, 40 to 60% of the posts by the six major political parties were issued without including hashtags. This finding raises the question of whether the usage of hashtags is necessary for successful political marketing – or if the algorithm already connects users with their searched contents with the usage of intelligent methods like latent semantic scaling (Watanabe 2020). Both findings could be exciting outlines for future research.

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Appendices

Appendix A. Data

Appendix B. Results

GitHub Repository

<https://github.com/philippklaeger/Gaining-The-Competitive-Edge>

Data Sets

Before Analysis

Political Marketing

<https://drive.google.com/file/d/1BtQ1bhOUISmhqDY-fRcXS0X4WJOIXxtL/view?usp=sharing>

User-Generated Discussions

https://drive.google.com/file/d/1f_QpmKQt2yiO-InzM41nmzIUZBVcgWSw/view?usp=sharing

After Analysis

Marketing

<https://drive.google.com/file/d/1wRvswelFP5wBGvSuMUZUYEEXA4HfXy-j/view?usp=sharing>

Users







<https://drive.google.com/file/d/19eIIgpT3X3k-IXmUdI1dFAkNhccNjYDW/view?usp=sharing>

Appendix A. Data

Data Set 1 – Political Marketing

Table A1 – Summary Statistics for the Political Marketing Data

65 variables x 218,256 observations

ID Variables	Time Variables	Text Variables	Location Variables	Analysis Variables	Dummy Variables	Count & Type Variables
user_id, reply_to_status_id, reply_to_user_id, mentions_user_id, retweet_status_id, retweet_user_id, retweet_name, mentions_screen_name, name	created_at, retweet_created_at, account_created_at	text, hashtags, retweet_text, description	location, retweet_location	party_id, election_id, campaign_id, M_syuzhet, M_bing, M_afinn, M_nrc, M_Sentiment_Index, anger, anticipation, disgust, fear, joy, sadness, surprise, trust, positive, negative, Polarity_Index, Emotion_Index, Intensity_Index	is_quote, is_retweet, AfD, CDU, FDP, Gruene, Linke, SPD	favorite_count, retweet_count, followers_count, friends_count, retweet_favorite_count, retweet_retweet_count, retweet_followers_count, retweet_friends_count, urls_url, media_url, media_type, photo, ext_media_url, verified, retweet_verified
Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Observations	39435	40133	47116	41364	19361	30847
Tweet WordCloud						
Most Used Words	1. spd (2987) 2. geht (1841) 3. cdu (1761) 4. menschen (1648) 5. mal (1505) 6. schon (1423) 7. danke (1393) 8. innen (1373) 9. dann (1354) 10. laschet (1250)	1. deutschland (2166) 2. cdu (1974) 3. corona (1899) 4. geht (1743) 5. menschen (1699) 6. spd (1557) 7. wollen (1423) 8. danke (1351) 9. gibt (1285) 10. mio (1251)	1. klimaschutz (3617) 2. menschen (2728) 3. geht (2208) 4. cdu (2201) 5. btw (1994) 6. laschet (1985) 7. bundesregierung (1923) 8. wollen (1915) 9. bundestag (1899) 10. innen (1856)	1. fdp (3225) 2. vielzutun (1924) 3. corona (1852) (1688) 5. geht (1654) 6. menschen (1638) 7. dann (1551) 8. immer (1514) 9. schon (1488) 10. gibt (1461)	1. afd (7674) 2. deutschland (1619) 3. bundestag (1564) 4. corona (1301) 5. btw (925) 6. deutschland- (842) 7. merkel (830) 8. kinderkongress (793) 9. schon (758) 10. cdu (735)	1. menschen (2137) 2. linke (1838) 3. bundesregierung (1493) 4. cdu (1461) 5. bundestag (1439) 6. btw (1361) 7. geht (1254) 8. corona (1206) 9. deutschland (1173) 10. gibt (1127)
Reach (average per account)						
Favorites	144.91	60.75	91.82	61.07	133.40	62.76
Retweets	85.89	83.26	138.33	74.47	107.26	88.03
Followers	62224.65	46932.89	44464.20	29245.91	23505.91	35611.61
Friends	1407.92	1305.86	2048.05	1768.77	945.58	1192.26
Dissemination (average per account)						
Text Width	149.30	151.39	153.11	152.85	167.88	173.52
Hashtag Count	0.82	0.93	1.07	0.92	1.86	1.34
Tweets with no hashtag	58.39%	53.46%	49.16%	55.60%	41.53%	40.61%
Most Used Hashtags (for all accounts)	1. #spd (835) 2. #triell (575) 3. #laschet (519) 4. #scholzpacktdasan (452) 5. #corona (424) 6. #cdu (407) 7. #bundestag (362) 8. #btw21 (322) 9. #klimaschutz (251) 10. #sauerland (227)	1. #corona (820) 2. #wegenmorgen (647) 3. #btw21 (523) 4. #cdu (521) 5. #corona (456) 6. #afghanistan (431) 7. #bundestag (430) 8. #spd (380) 9. #scholz (337) 10. #bundeswehr (317)	1. #klimaschutz (1339) 2. #btw21 (1076) 3. #laschet (1027) 4. #triell (863) 5. #afghanistan (861) 6. #allesistdrin (663) 7. #bundestag (648) 8. #scholz (583) 9. #cdu (514) 10. #corona (448)	1. #vielzutun (1206) 2. #fdp (1039) 3. #corona (814) 4. #bpt21 (586) 5. #bundestag (583) 6. #btw21 (541) 7. #bundesregierung (372) 8. #afghanistan (307) 9. #eu (305) 10. #lockdown (286)	1. #afd (5139) 2. #bundestag (1064) 3. #corona (779) 4. #deutschland- (627) 5. #deutschland (545) 6. #abernormal (498) 7. #merkel (486) 8. #berlin (446) 9. #btw21 (438) 10. #lockdown (432)	1. #btw21 (664) 2. #bundestag (660) 3. #afghanistan (632) 4. #corona (625) 5. #cdu (611) 6. #machtdasland- (513) 7. #bundesregierung (478) 8. #dielinke (451) 9. #linke (414) 10. #mietendeckel (406)

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Data Set 2 – User-Generated Discussions

Table A2 – Summary Statistics for the User-Generated Discussions Data

70 variables x 297,775 observations

ID Variables	Time Variables	Text Variables	Location Variables	Analysis Variables	Dummy Variables	Count & Type Variables
query, user_id,	created_at,	text, hashtags,	location,	party_id,	is_quote, is_retweet,	favorite_count,
reply_to_status_id,	retweet_created_at,	retweet_text	retweet_location	candidate_id,	AfD, CDU, FDP,	retweet_count,
reply_to_user_id,	account_created_at	description		election_id,	Gruene, Linke, SPD,	retweet_favorite_count,
mentions_user_id,				campaign_id,	Weidel, Chrupalla,	retweet_retweet_count,
retweet_status_id,				U_syuzhet, U_bing,	Laschet, Dobrindt,	retweet_followers_count,
retweet_user_id,				U_afinn, U_nrc,	Lindner, Baerbock,	retweet_friends_count,
retweet_name,				U_Sentiment_Index,	Habeck, Wissler,	urls_url, media_url,
mentions_screen_name,				anger, anticipation,	Bartsch, Scholz	media_type, photo,
name				disgust, fear, joy,		ext_media_url, verified,
				sadness, surprise,		retweet_verified
				trust, positive,		
				negative,		
				Polarity_Index,		
				Emotion_Index,		
				Intensity_Index		
No Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
118438	19127	67417	19509	10966	54907	7411
Reach & Dissemination (per Account)						
- Avg. Favorite Count	2.85	2.38	2.87	3.26	2.51	2.61
- Avg. Retweet Count	99.97	127.04	95.58	162.23	97.05	32.96
- Avg. Text Width	148.96	156.19	151.68	147.90	147.52	154.53

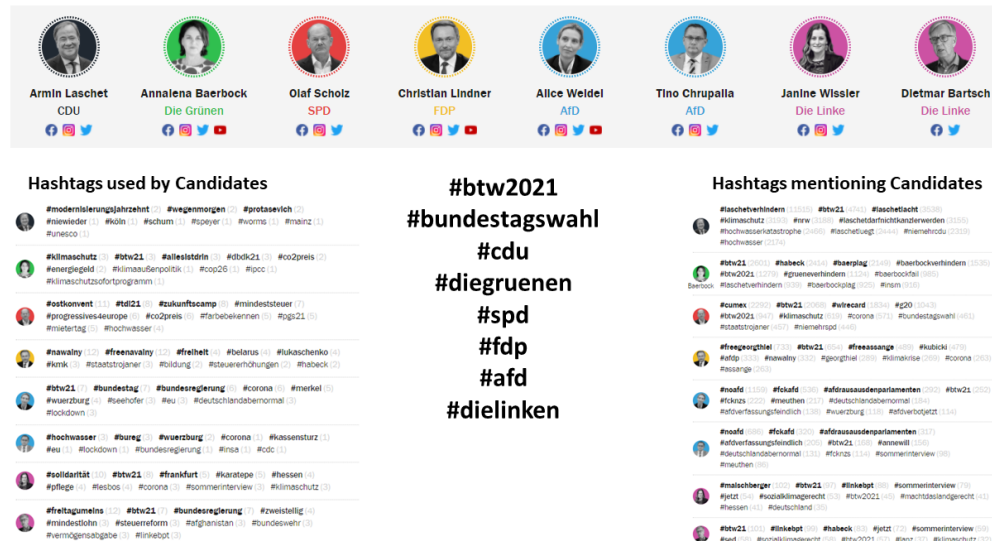


Figures A15 – WordCloud of Tweet Texts (User-Generated Discussions)

User-Generated Discussions – Search Queries (Hashtags)

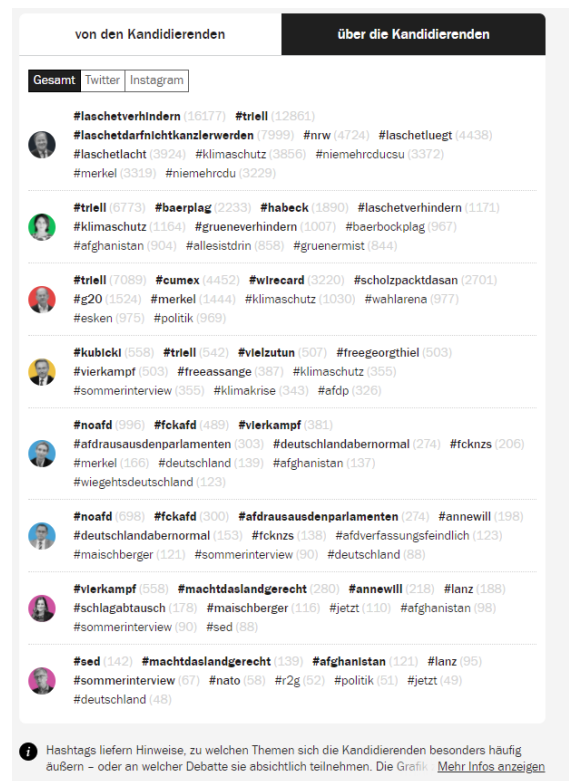
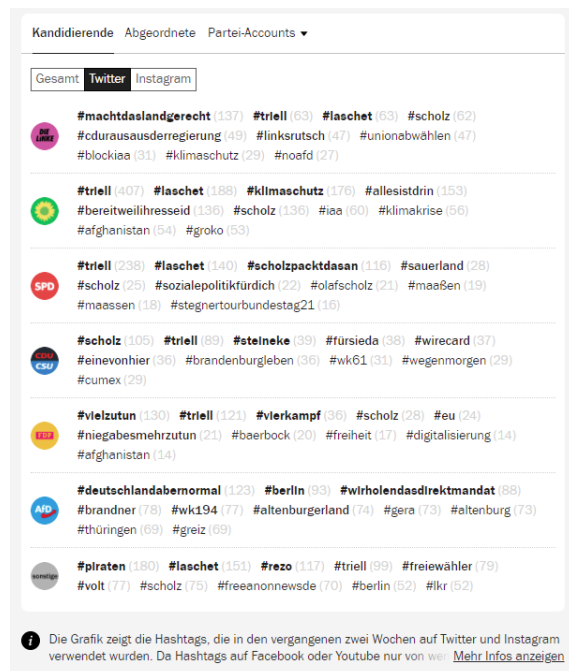
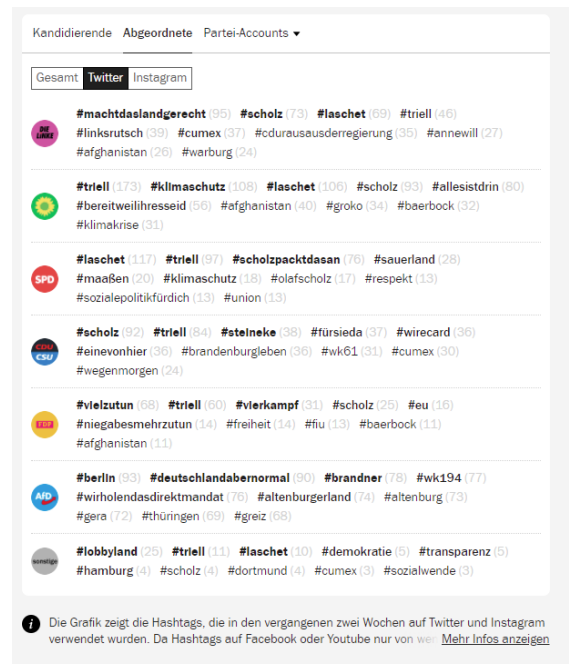
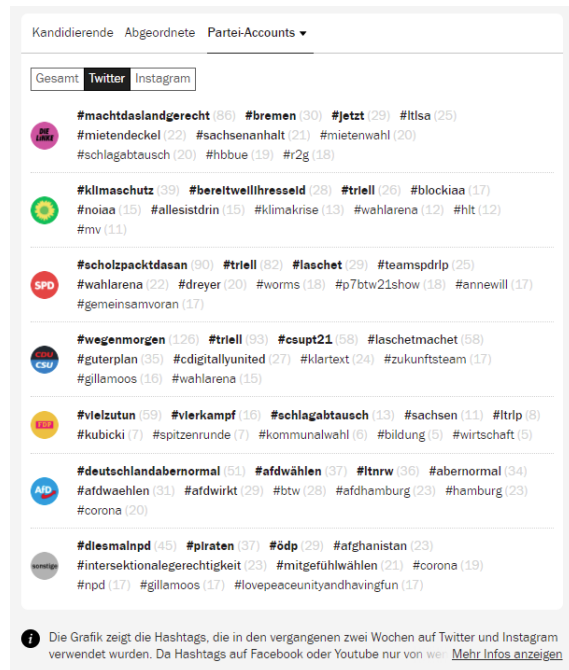
The list of search queries continuously increased weekly with the progress of this research and additional conversation topics in the following weeks. Hence, there are multiple iterations of query lists. In the following, the used queries are presented.

Seeds For Collecting Twitter Data



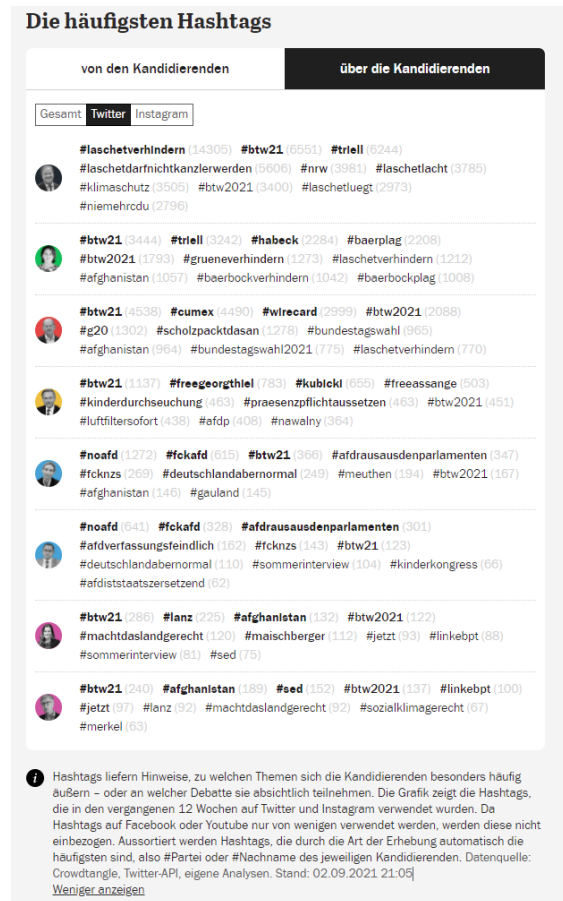
Seeds – First Iteration (first week of data collection)

Source: Tagesspiegel 2021



Seeds – Second Iteration (second and third week of data collection)

Source: Tagesspiegel 2021



Seeds – Third Iteration (fourth week of data collection)

Source: Tagesspiegel 2021

The relevant hashtags were drawn from the trending hashtags lists on the real-time social media dashboard specifically created for the 2021 federal election by the Berlin newspaper *Tagesspiegel* in cooperation with Democracy Reporting International and Stiftung Mercator (Tagesspiegel 2021).

Appendix B. Results

Sentiment & Emotion Analysis

Table B1 – Sentiment and Emotions of Parties (Political Marketing)

	SPD	CDU/CSU	Greens	FDP	AfD	Left
Sentiment Index	-0.6255	-0.6307	0.6777	-0.6983	-0.7348	-0.8203
Syuzhet	-0.3353	-0.3302	-0.3587	-0.3679	-0.3953	0.4433
Bing	-0.3988	-0.4076	-0.4448	-0.4684	-0.4910	-0.5378
Afinn	-1.2823	-1.3101	-1.3785	-1.3994	-1.4699	-1.6494
NRC	-0.4856	-0.4748	-0.5287	-0.5575	-0.5830	-0.6508
Positive	0.4491	0.4237	0.4046	0.4134	0.3462	0.3835
Negative	0.2636	0.2694	0.2992	0.3128	0.3189	0.3268
Intensity Index	0.8718	0.8766	0.8803	0.8609	0.8767	0.9160
Anger	0.0615	0.0611	0.0728	0.0696	0.0866	0.0761
Anticipation	0.1472	0.1490	0.1433	0.1374	0.1358	0.1557
Disgust	0.0370	0.0368	0.0417	0.0392	0.0544	0.0459
Fear	0.0923	0.0941	0.1023	0.0984	0.1160	0.1120
Joy	0.1104	0.1106	0.0968	0.0957	0.0836	0.1037
Sadness	0.1368	0.1446	0.1547	0.1568	0.1701	0.1739
Surprise	0.0721	0.0663	0.0675	0.0684	0.0547	0.0564
Trust	0.2145	0.2141	0.2011	0.1954	0.1754	0.1924

Table B2 – Sentiment and Emotions of Parties (User-Generated Discussions)

	SPD	CDU/CSU	Greens	FDP	AfD	Left
Sentiment Index	-0.7552	-0.7963	-0.7209	-0.7894	-0.7681	-0.9238
Syuzhet	-0.4054	-0.4199	-0.4066	-0.4381	-0.4135	-0.5015
Bing	-0.5258	-0.5518	-0.4737	-0.5863	-0.5122	-0.6345
Afinn	-1.4952	-1.5821	-1.4565	-1.5297	-1.5651	-1.8459
NRC	-0.5945	-0.6313	-0.5467	-0.6037	-0.5818	-0.7135
Positive	0.4254	0.4194	0.4604	0.4360	0.3201	0.4713
Negative	0.3547	0.3768	0.2802	0.4704	0.3869	0.2888
Intensity Index	1.1318	0.9679	0.8250	1.0771	0.9766	0.8699
Anger	0.1430	0.1187	0.0691	0.1206	0.1036	0.0583
Anticipation	0.1263	0.1221	0.1214	0.1434	0.1402	0.1374
Disgust	0.1109	0.0781	0.0389	0.0726	0.0877	0.0429
Fear	0.1445	0.1433	0.0784	0.1326	0.1684	0.0935
Joy	0.1109	0.0815	0.1264	0.0954	0.0645	0.0930
Sadness	0.1712	0.1647	0.1264	0.1973	0.2327	0.1410
Surprise	0.1254	0.0541	0.0409	0.0883	0.0371	0.0650
Trust	0.1996	0.2054	0.2235	0.2270	0.1424	0.2389

Table B3 – Sentiment and Emotions of Candidates (User-Generated Discussions)

	SPD	CDU/CSU	Greens	FDP	AfD	Left
	<i>Scholz</i>	<i>Dobrindt</i> <i>Laschet</i>	<i>Baerbock</i> <i>Habeck</i>	<i>Lindner</i>	<i>Chrupalla</i> <i>Weidel</i>	<i>Bartsch</i> <i>Wissler</i>
Sentiment Index	-0.6734	-1.3565 -0.6816	-0.6332 -0.6640	-0.7092	-0.8306 -0.5625	-0.9696 -1.0216
Syuzhet	-0.3600	-0.7085 -0.3508	-0.3937 -0.3508	-0.3769	-0.5101 -0.2812	-0.5263 -0.5674
Bing	-0.4857	-0.9238 -0.4801	-0.4056 -0.4107	-0.5165	-0.5101 -0.2812	-0.6854 -0.7121
Afinn	-1.3391	-2.8156 -1.3192	-1.2858 -1.3130	-1.4160	-1.8393 -1.1103	-1.8318 -2.0532
NRC	-0.5089	-0.9780 -0.5757	-0.4480 -0.5814	-0.5274	-0.5948 -0.5670	-0.8349 -0.7534
Positive	0.4032	0.1202 0.4071	0.4248 0.5089	0.5321	0.2339 0.2750	0.3146 0.5230
Negative	0.3120	1.9599 0.3890	0.2540 0.4950	0.4783	0.4157 0.2401	0.2430 0.5548
Intensity Index	1.0682	1.3367 0.9423	0.8385 1.0452	1.2034	0.6312 0.6088	0.5857 1.3064
Anger	0.1278	0.3607 0.1307	0.0894 0.1248	0.1010	0.0576 0.0357	0.0467 0.1083
Anticipation	0.0979	0.0421 0.1068	0.1015 0.1995	0.1666	0.0373 0.1013	0.0717 0.2328
Disgust	0.0885	0.3828 0.0900	0.0541 0.0647	0.0977	0.0308 0.0286	0.0343 0.0868
Fear	0.1312	0.3707 0.1509	0.0959 0.1326	0.1345	0.1588 0.0804	0.0592 0.1287
Joy	0.1195	0.0301 0.0684	0.0629 0.0960	0.1630	0.0497 0.0606	0.0623 0.1987
Sadness	0.1655	0.0842 0.1609	0.1953 0.1444	0.1649	0.1736 0.1106	0.1246 0.1065
Surprise	0.1292	0.0140 0.0502	0.0385 0.0533	0.0941	0.0299 0.0445	0.0436 0.0736
Trust	0.2085	0.0521 0.1843	0.2010 0.2297	0.2815	0.0935 0.1471	0.1433 0.3710

Note: Parties with two lead candidates are displayed in alphabetical order of their surnames.

General Stance Analysis

Table B4 – RILE Indices

	SPD	CDU/CSU	Greens	FDP	AfD	Left
RILE Index	-21.453	2.758	-21.057	0.576	17.429	-41.912
Program						
Election I						
RILE Index	-24.673	3.495	-21.037	0.265	26.050	-36.168
Program						
Election II						
RILE	-3.22	+0.737	-0.020	-0.311	+8.621	+5.744
Difference						
Between						
Programs						

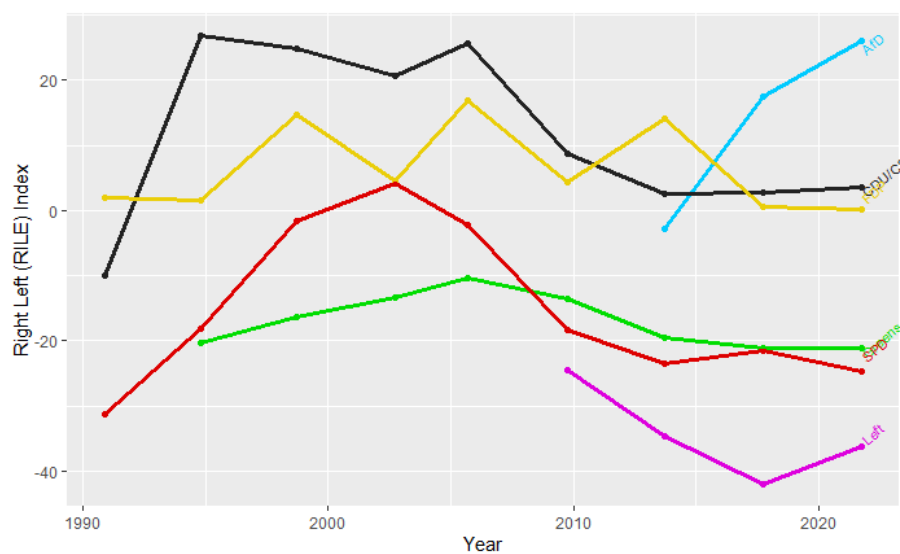


Figure B1 – RILE Index (1990-2021)

Table B5 – Reference Scores (2019 Chapel Hill Expert Survey)

	SPD	CDU/CSU	Greens	FDP	AfD	Left
Overall	3.619	6.524	3.238	6.429	9.238	1.429
Ideological Score						
left vs. right						
Economic Issues Score	3.714	6.143	3.810	7.905	7.000	1.286
left vs. right						
Social & Cultural Values Score	3.381	6.571	1.095	3.429	9.524	2.810
libertarian vs. traditional						
Immigration Policy Score	4.150	6.850	1.700	5.950	9.900	2.700
liberal vs. restrictive						
Environmental Score	4.700	6.125	1.550	7.526	8.450	4.250
sustainability vs. econ. growth						
State Spendings Score	3.053	6.395	3.158	8.895	7.286	1.000
public services vs. tax reduction						
Identity Score	3.421	6.211	1.105	3.895	9.684	2.895
cosmopolitanism vs. nationalism						
Trade Orientation Score	4.222	3.778	3.882	1.611	6.688	6.222
liberalization vs. protectionism						
Civil Score	4.050	7.150	1.850	3.450	9.500	2.421
civil liberties vs. law & order						

Reference scores for the Quanteda WordScore algorithm. The values for each party were taken from the Chapel Hill Expert Survey 2019 (n:21), an expert evaluation of stances based on party programs in 2019, used as reference scores for the 2017 party programs in the prediction of wordscores (left to right; scale 0-10).

Table B6 – General Stance Analysis

Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Reference	3.619	6.524	3.238	6.429	9.238	1.429
Score						
CH Overall						
Ideo. Score						
Party Program (2017)	4.560	4.828	4.474	4.950	5.368	4.292
Quanteda score						
PP 2017 SE	0.0040	0.0062	0.0031	0.0046	0.0094	0.0038
Party Program (2021)	4.605	4.715	4.650	4.885	5.015	4.556
Quanteda score						
PP 2021 SE	0.0056	0.0038	0.0033	0.0047	0.0068	0.0040
Tweets (2021)	4.777	4.818	4.753	4.817	4.854	4.739
Quanteda score						
T 2021 SE	0.0010	0.0009	0.0009	0.0009	0.0014	0.0011
Difference	+0.045	+0.113	+0.176	-0.065	-0.353	+0.264
Program- Program (2017- 2021)						
Difference	+0.172	+0.103	+0.103	-0.068	-0.161	+0.183
Tweets-Program (2021)						

Relative Left-Right WordScore positions of party programs and tweets for each of the major political parties (2017+2021). Based on the assignment of the Chapel Hill Expert Survey (2019) scores as reference values for the 2017 federal election party programs, the relative position of the party programs and tweets in 2021 was predicted by the Quanteda WordScore algorithm.

Table B7 – Lexical Diversity

Party	SPD	CDU	Greens	FDP	AfD	Left
PP 2017	0.4622	0.4392	0.3959 *	0.4728	0.6152 **	0.5036
PP 2021	0.4905	0.3969 *	0.4087	0.4960	0.6035 **	0.4686
T 2021	0.1279	0.1200	0.1047 *	0.1181	0.1619 **	0.1222

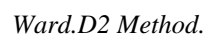
*Lexical Diversity of the party programs in 2017 and 2021, as well as the tweets in the Political Marketing data. Grouped by party. Maximum (**) and minimum (*) values.*

Table B8 – Similarity Between Authors (SBA)

Party	SPD	CDU	Greens	FDP	AfD	Left
Median	196.60	202.20	233.52	222.94	145.30 *	257.82 **
Distance						
Mean	219.40	209.23	270.75	242.75	176.66 *	281.75 **
Distance						
IQR	137.10	166.89	137.10	124.05 *	167.02	180.06 **
Distance						

*Similarity Between Authors (SBA) in the Political Marketing data. Grouped by party. Maximum (**) and minimum (*) values.*

Clustering is an adequate visualization of Similarity Between Authors (SBA). The clustering algorithm *ward.D2* utilizes the Ward's minimum variance method by squaring the dissimilarities before clustering (UA Alberta 2021). This method was chosen because of two reasons. Firstly, because it allows us to identify latent groups within each party that are dissimilar to each other measured in SBA. Secondly, because by squaring the dissimilarities before clustering, this method makes small dissimilarities proportionally easier to identify – an issue that is likely to arise with large sample sizes for clustering like the number of candidates for each party in this research.



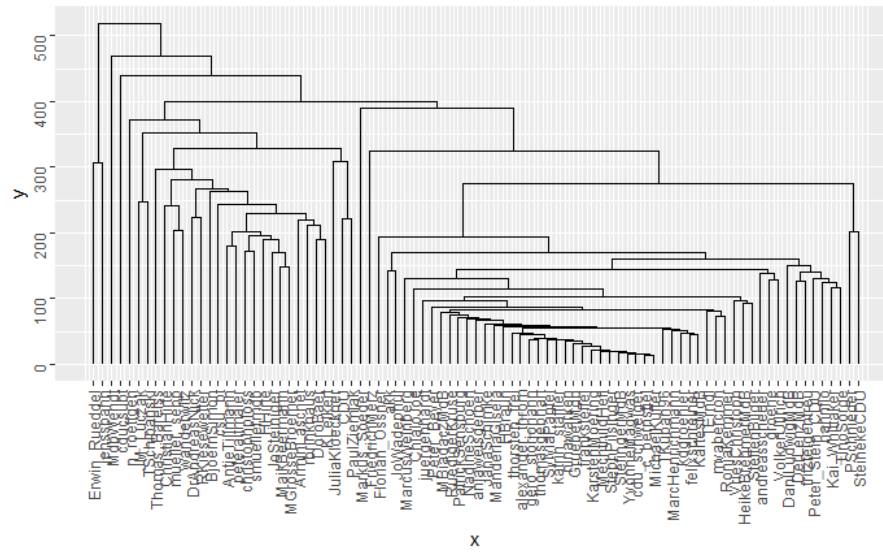


Figure B5 – CDU/CSU Dendrogram I (Similarity Between Authors Clustering)

Complete Linkage Method.

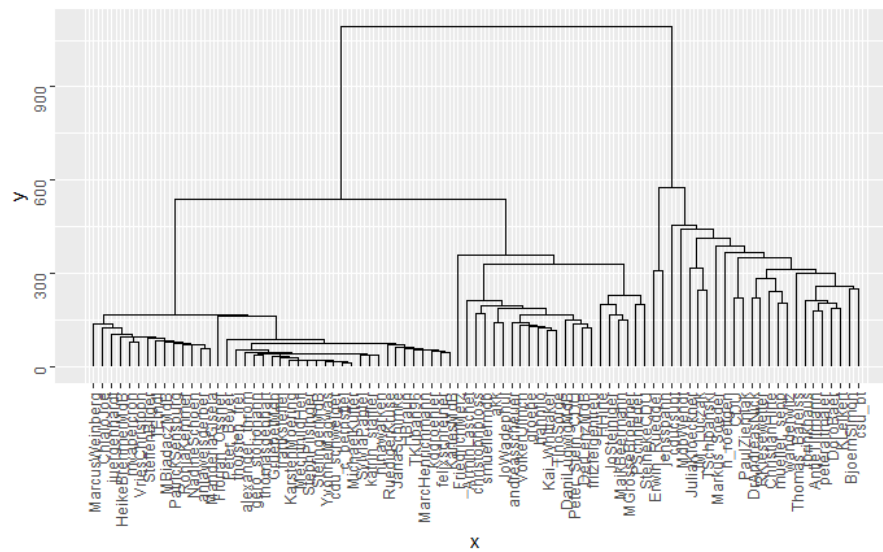
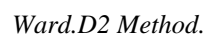
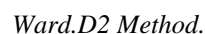


Figure B6 – CDU/CSU Dendrogram II (Similarity Between Authors Clustering)

Ward.D2 Method.





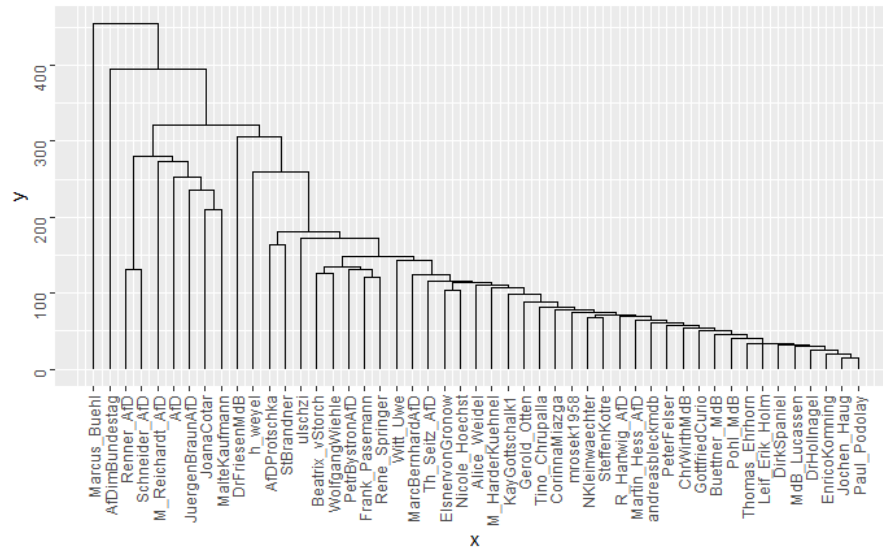


Figure B11 – Afd Dendrogram I (Similarity Between Authors Clustering)

Complete Linkage Method.

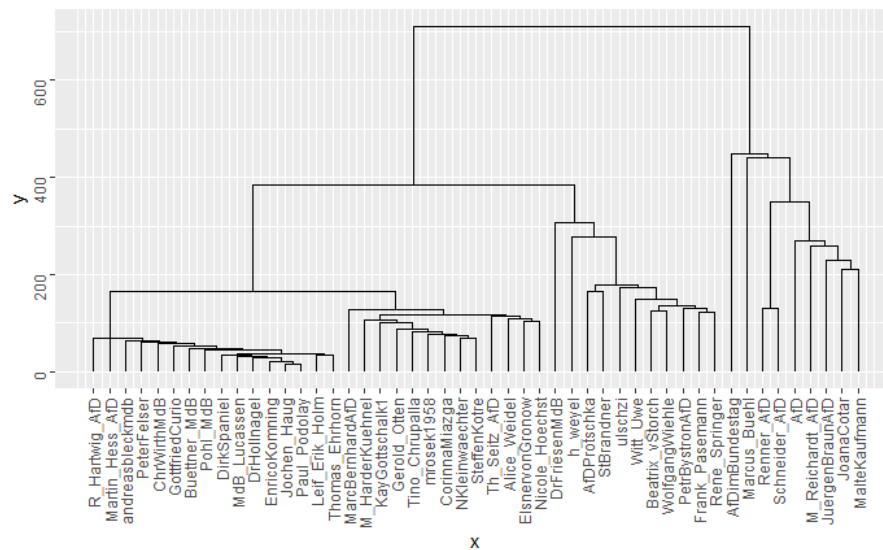


Figure B12 – Afd Dendrogram II (Similarity Between Authors Clustering)

Ward.D2 Method.

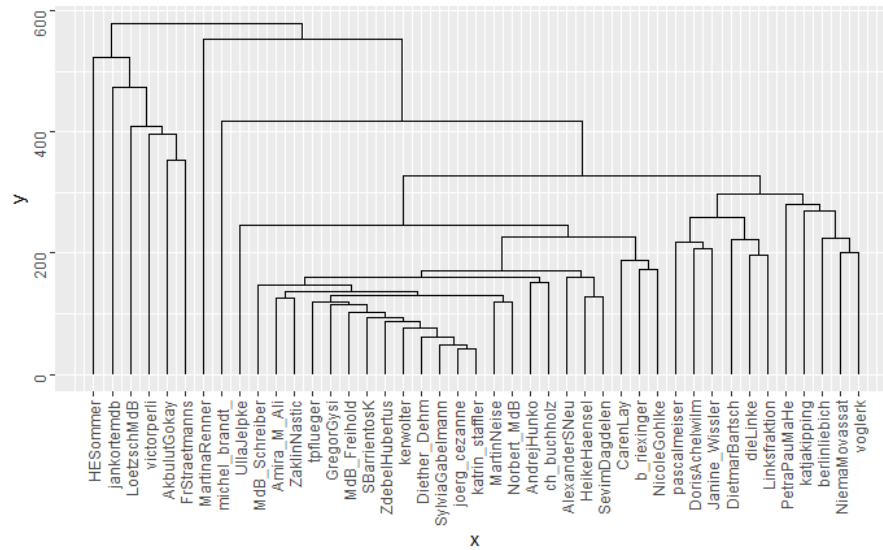


Figure B13 – The Left Dendrogram I (Similarity Between Authors Clustering)

Complete Linkage Method.

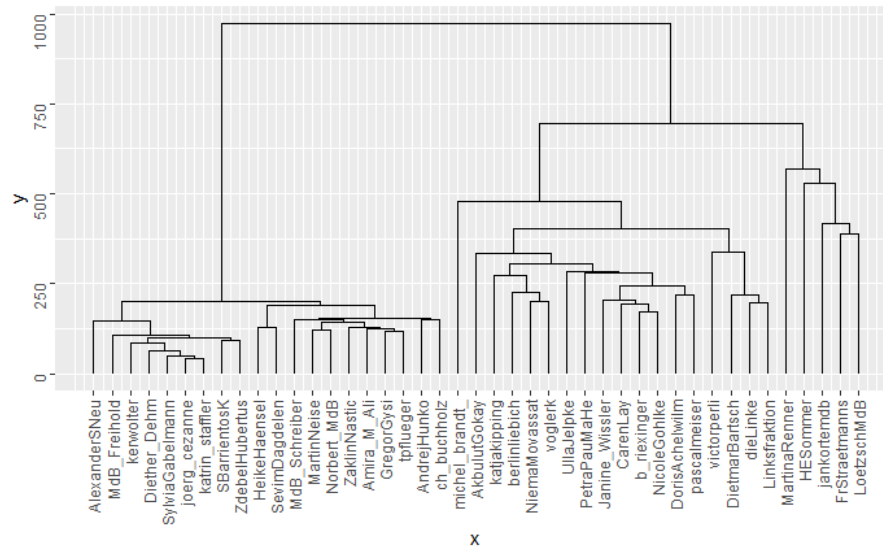
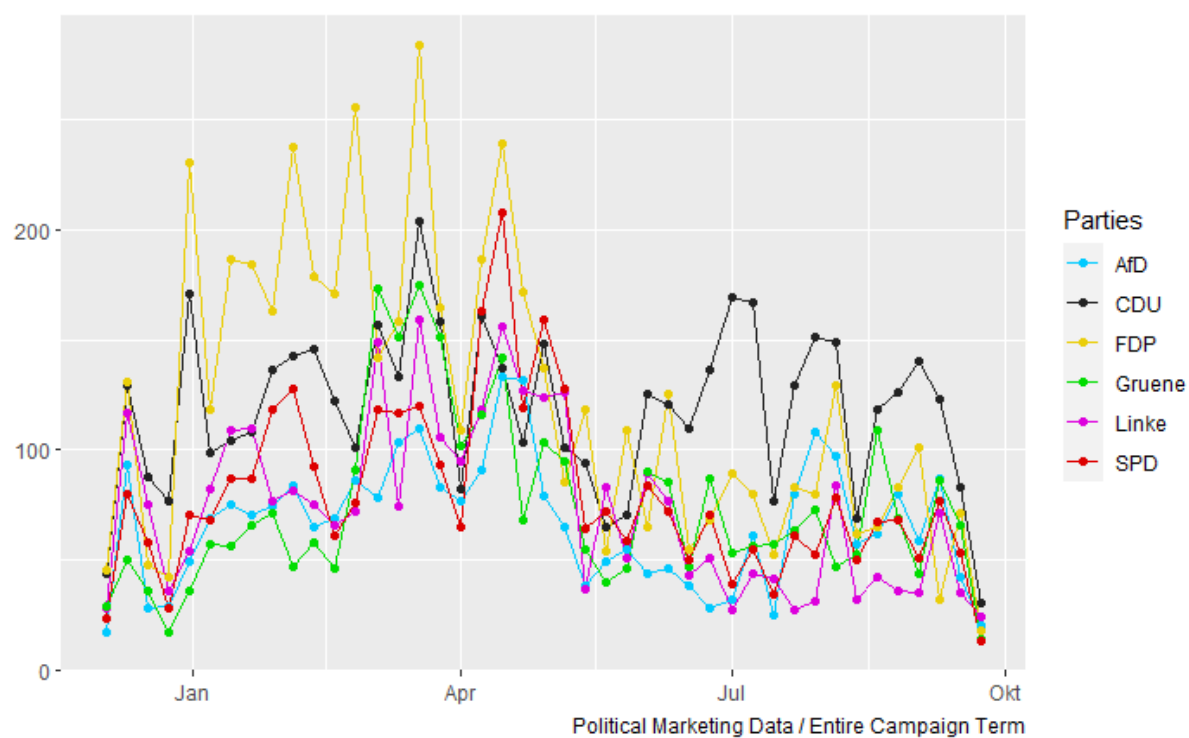
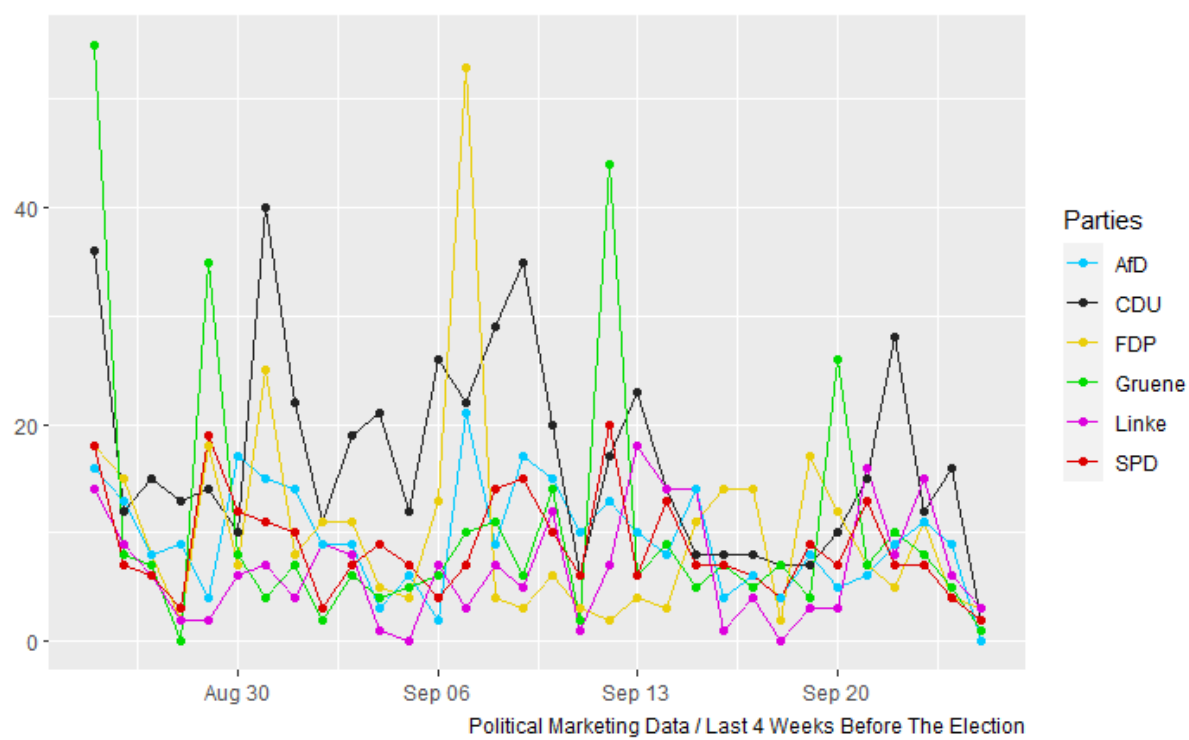


Figure B14 – The Left Dendrogram II (Similarity Between Authors Clustering)

Ward.D2 Method.

Topic-Specific Stance Analysis**Tables B9-B11 – Topic-Specific Stances**

Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Tweets (2021)	4.7663	4.7648	4.7590	4.7753	4.7778	4.7648
Quanteda score						
Corona Tweets (2021)	4.7686	4.7662	4.7607	4.7770	4.7787	4.7662
Quanteda score						
Difference	+0.0023	+0.0014	+0.0017	+0.0017	+0.0009	+0.0012
Tweets-Corona-Tweets (2021)						
SE Tweets	2.5377e-05	2.4547e-05	2.7749e-05	3.4189e-05	4.6831e-05	2.6589e-05
SE Cor. Tweets	7.3095e-05	6.3244e-05	1.0885e-04	8.5752e-05	1.1300e-04	8.0612e-05
Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Tweets (2021)	4.7663	4.7648	4.7590	4.7753	4.7778	4.7648
Quanteda score						
Environment Tweets (2021)	4.7617	4.7622	4.7542	4.7738	4.7737	4.7617
Quanteda score						
Difference	-0.0046	-0.0026	-0.0048	-0.0015	-0.0041	-0.0031
Tweets- Env. - Tweets (2021)						
SE Tweets	2.5378e-05	2.4548e-05	2.7750e-05	3.4187e-05	4.6833e-05	2.6590e-05
SE Env. Tweets	1.3489e-04	1.2767e-04	8.6502e-05	1.9730e-04	3.2660e-04	1.7234e-04
Party	SPD	CDU/CSU	Greens	FDP	AfD	Left
Tweets (2021)	4.7663	4.7648	4.7590	4.7753	4.7778	4.7648
Quanteda score						
Digitization Tweets (2021)	4.7656	4.7645	4.7607	4.7756	4.7764	4.7654
Quanteda score						
Difference	-0.0007	-0.0003	+0.0017	+0.0003	-0.0014	-0.0006
Tweets- Dig.- Tweets (2021)						
SE Tweets	2.5378e-05	2.4549e-05	2.7750e-05	3.4186e-05	4.6833e-05	2.6590e-05
SE Dig. Tweets	1.5616e-04	1.1466e-04	1.4796e-04	1.6474e-04	5.5794e-04	2.6575e-04

Topic 1 - Corona / Covid-19 Pandemic**Figure B16 – Corona-specific tweets (entire campaign term)****Figure B17 – Corona-specific tweets (last four weeks)**

Topic 2 - Environment / Climate Change

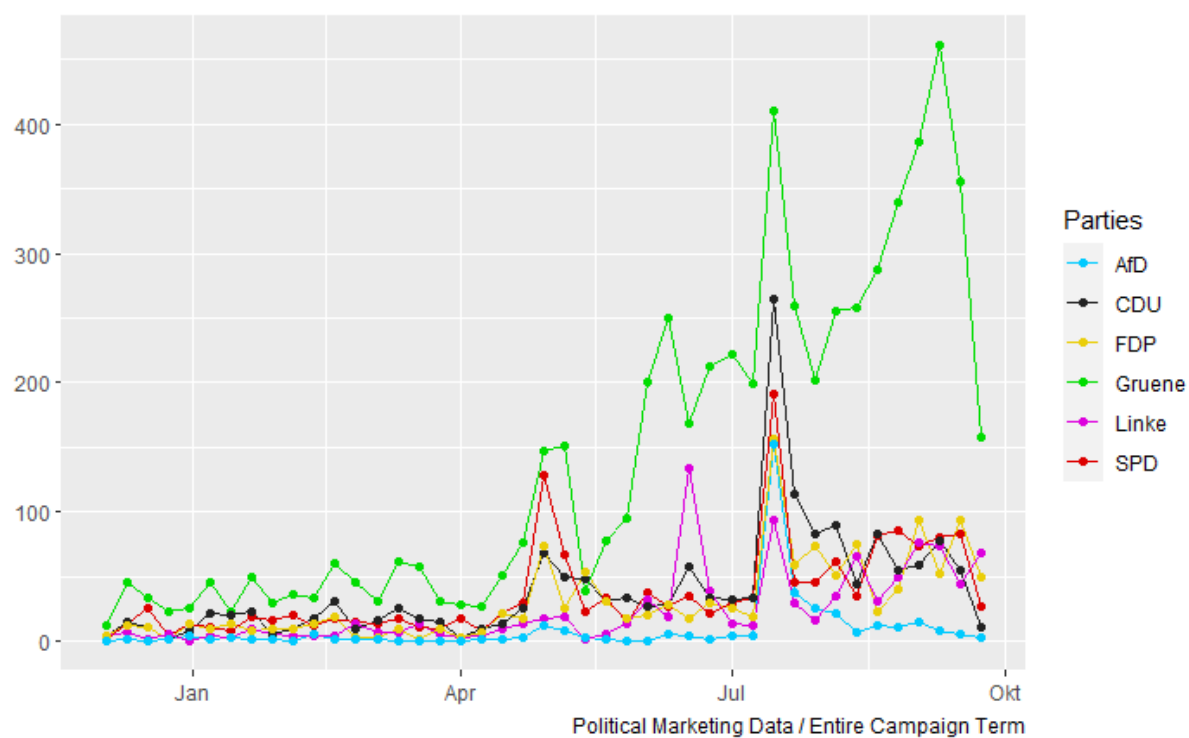


Figure B18 – Environment-specific tweets (entire campaign term)

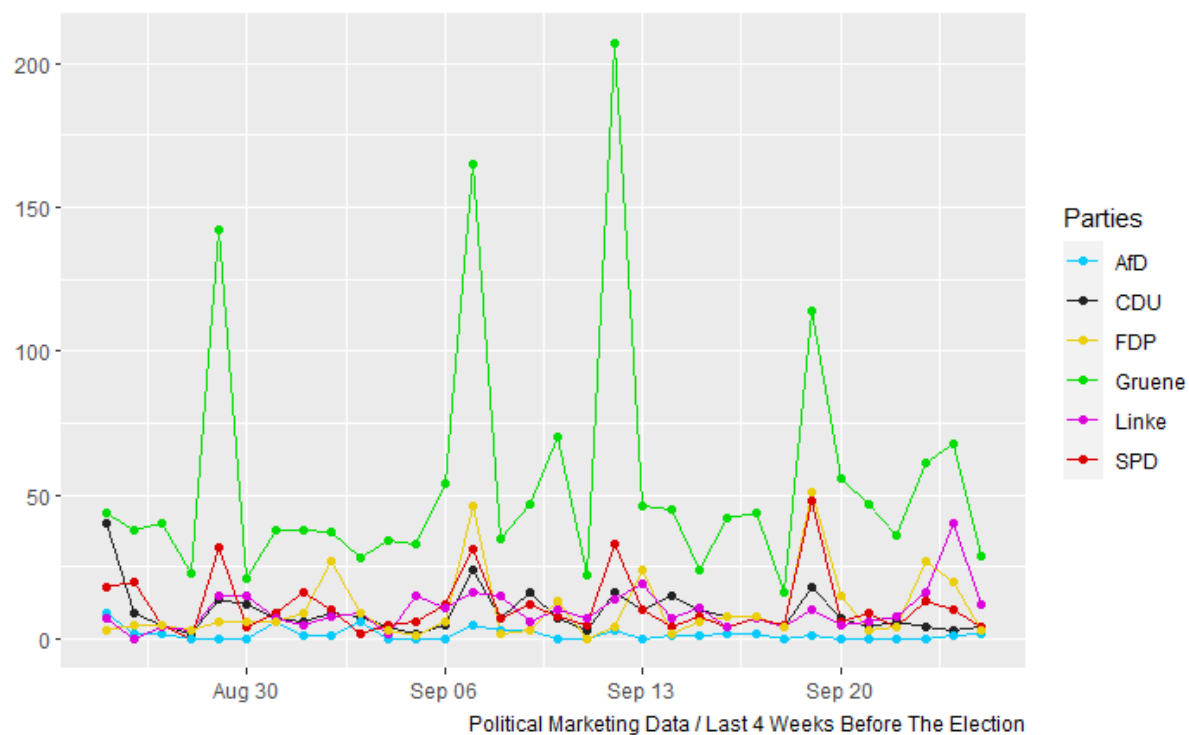


Figure B19 – Environment -specific tweets (last four weeks)

Topic 3 - Digitization

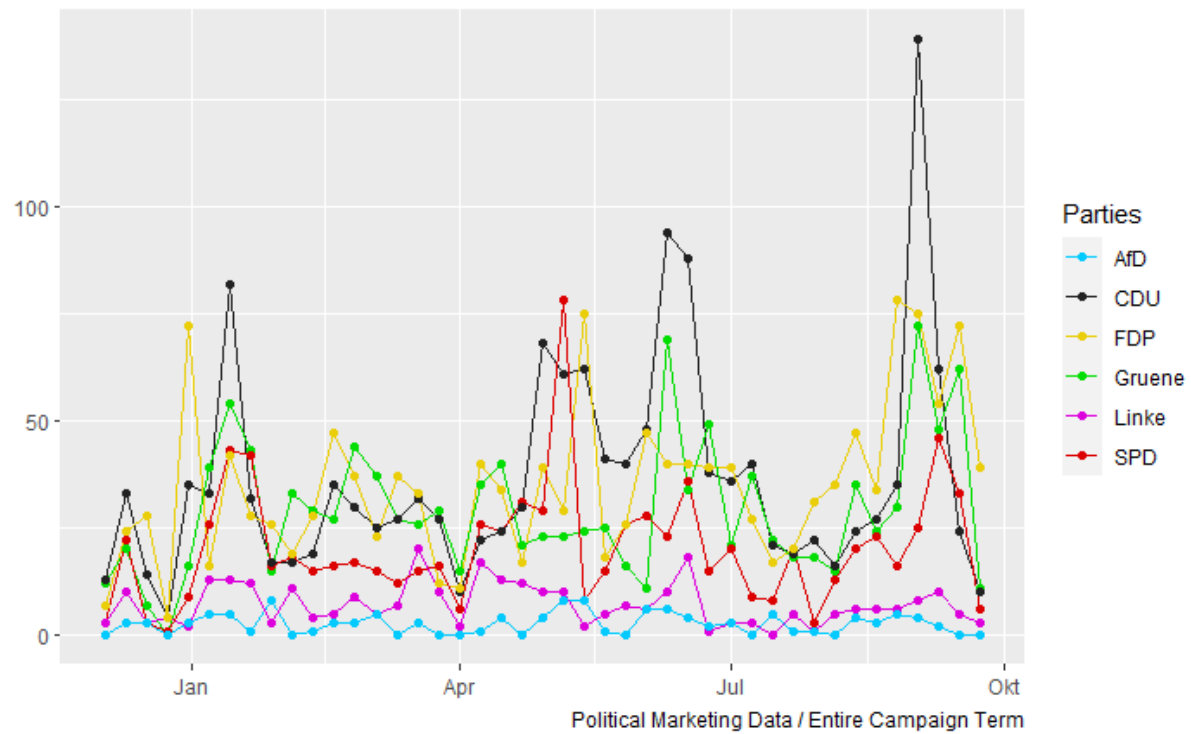


Figure B20 – Digitization-specific tweets (entire campaign term)

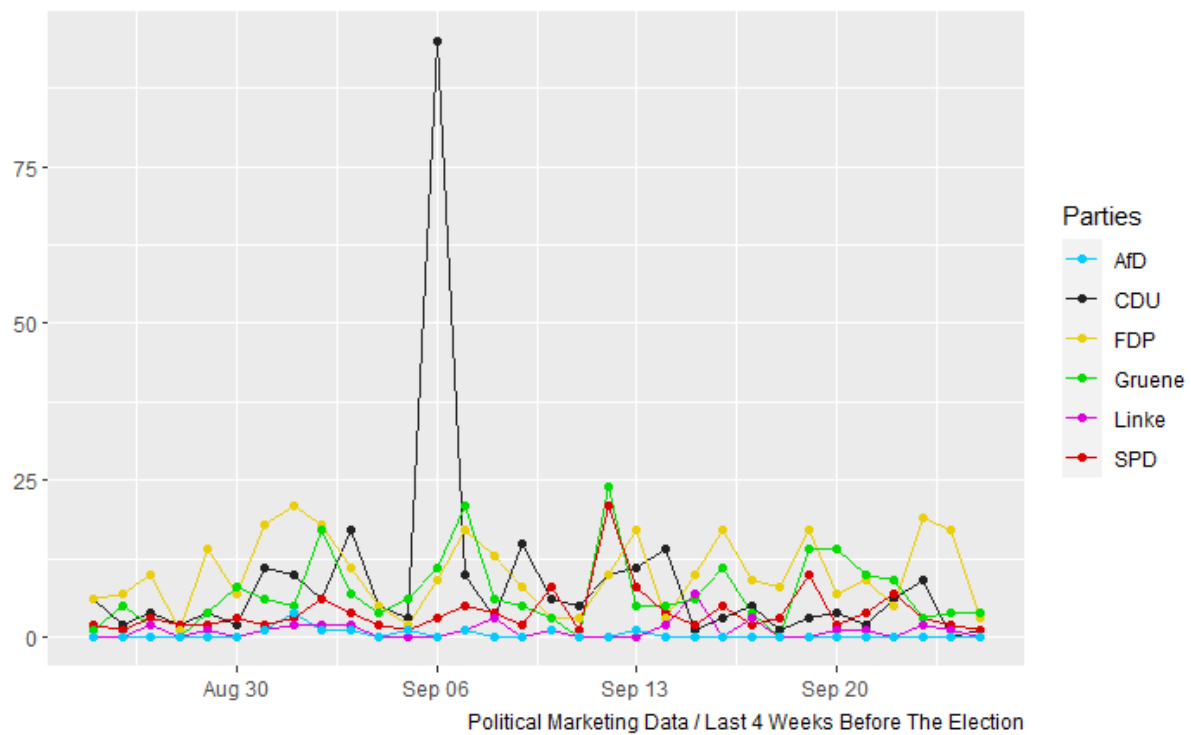


Figure B21 – Digitization-specific tweets (last four weeks)