Extended Appendix:

Conflict shapes in flux: Explaining spatial shift in conflict-related violence

1. Constructing the conflict shape

To determine the conflict shape—the area affected by conflict-related violence—we adopt a six-step logic supported by data from the Uppsala Conflict Data Program Georeferenced Event Dataset UCDP GED (Croicu and Sundberg, 2018; Sundberg and Melander, 2013), fieldwork interviews, the literature, and interviews with experts with fieldwork experience in the respective region.

1. We select an *anchor conflict*, that is, the conflict of interest driven by a particular contested issue, or issues, that typically includes multiple conflict actors. The contested issue determines which conflict actors form part of this anchor conflict. Contrary to other studies (see, e.g., Pettersson et al., 2019), we do not specify the types of main contested issues (e.g., over government or territory) as defining criteria since they can mutate over time, often resulting in the involvement of new conflict actors, and because the contested issue is a social construct that different conflict parties can perceive differently (Kriesberg and Gerard, 2018). Instead, we track the contested issue's evolution over time. This conceptualization is consistent with the concept "incompatibility" used by UCDP GED, thus making it possible to draw on this data source. The UCDP defines incompatibility as "stated (in writing or verbally) generally incompatible positions" and distinguishes between incompatibility over government or territory (*Uppsala Conflict Data Program*, 2022)—the concept "contested issue" comprises both of these subtypes.

We examine other, spatially connected conflicts in the region as well as preceding and succeeding conflicts in the same territory to determine whether they relate to the anchor conflict through one or both of the following mechanisms:

- a. <u>Conflict mutation</u> (connecting over time). The original contested issue changes or a new contested issue adds to the conflict. This can occur when actors involved in conflict seem to change but only regroup under a new name (e.g., elements of Iraqi insurgents and military transformed into the Islamic State of Iraq and Syria [ISIS]) or when a new conflict actor emerges (e.g., drug cartels joined Colombia's armed conflict).
- b. <u>Conflict spread</u> (connecting across space). An actor or actors involved in a conflict move(s) to a new location and triggers or escalates a new conflict there that can have the same or a different contested issue (e.g., the leaders of al-Qaeda and the Taliban escaped to the former Federally Administered Tribal Areas [FATA], triggering the area's militarization and radicalization).

If we detect any of these mechanisms, we add the conflicts to the anchor conflict, creating an umbrella conflict.

2. We identify the conflict actors involved. We identify the wider region that comprises countries where relevant conflicts occur, and their neighboring countries, to obtain the list of potentially involved actors from the UCDP GED (Croicu and Sundberg, 2018; Sundberg and Melander, 2013) based on the list of conflicts established in the first step (e.g., we include all countries adjacent to Colombia since violence related to this case often occurs near state borders). Then we filter out the actors who engage violently in the conflicts identified in the previous step. We assess the actors' participation based on their connection to the contested issues. When a conflict mutates or spreads, new actors are likely to join the armed conflict.

- 3. We select all conflict events from the UCDP GED dataset in which the actors identified in the previous step were involved. We follow the Uppsala Conflict Data Program's (UCDP)'s definition of conflict events as "[a]n incident where armed force was used by an organised actor against another organized [sic] actor, or against civilians, resulting in at least 1 direct death at a specific location and a specific date" (Högbladh, 2020: 4). Hence, we exclude governments or non-state actors financing other actors involved in the conflict. An actor needs to have at least a limited level of organizational structure; civilians and community groups are not considered conflict actors. We select those relevant events for which we have relatively precise information about their locations to ensure that we do not distort the conflict shape. We select events with the precision of their location within the first-level administrative unit (one below the state level). We exclude conflict events where organized actors deliberately attack civilians because "civilians" is a generic label and event descriptions often do not provide enough information to be assigned to a specific conflict.
- 4. Using the R package Concaveman (Gombin et al., 2017), based Park and Oh (2013)'s algorithm, we draw a concave hull from the selected events to denote the conflict shape's contours. We set the concavity level for all concave hulls to number two. Following Weidmann (2015), who points out the level of spatial error in data, we add a 50-km buffer to mitigate potential remaining spatial errors in the UCDP GED. A concave hull encloses points more compactly than a convex hull (Asaeedi et al., 2017). It permits internal angles larger than 180 degrees. These internal angles create dents, allowing the researcher to exclude large areas without any data points and hence facilitating more realistic information about the conflict shape's type of environment (see Figure 1). The UCDP family of datasets offers conflict polygons based on conflict dyads. These conflict polygons are convex hulls based on the location of the events allocated to given conflict dyads. Yet the resulting polygon does not allow for dents and consequently includes large regions without conflict events. Since a convex hull does not allow for external angles smaller than 180 degrees, the polygon also includes the red area without any events.
- 5. We identify any outliers (Croicu and Sundberg, 2012) and redraw the conflict shapes if outliers were removed. When events occur far away from the conflict setting, we exclude them from the conflict shape because they would unreasonably stretch the geographical scope. We consider an event an outlier if removing it would lead to a 15 percent contraction of the concave hull area. While Croicu and Sundberg (2012) use a 20 percent threshold for their convex hull-based polygons, we lower the cut-off point because the concave hull we use produces areas smaller than a convex one. The 15 percent rule does not apply if the event involves more than 10 percent of the total number of deaths in the polygon for a given year. In some situations, we consider a group of events to be potential outliers, i.e., when the events occur in the same or nearby location. We do not remove more than 5 percent of the total number of events that are outliers to keep a realistic shape and spatial pattern of each conflict.

To detect outliers for each year separately, we visually detect events—potential outliers—that seem to be located far away from most other events.³ We then calculate the total number of fatalities in the concave hull area. Subsequently, we calculate the concave hull area percentage that contracts after removing the potential outliers. If it contracts less than 15 percent, the event is not considered an outlier.⁴ If it contracts more than 15 percent, we calculate the percentage decrease in the total number of fatalities by removing the event. If the decrease is larger than 5 percent, the event is not considered an outlier. In assessing a group of events that are clustered close to each other, we calculate the percentage decrease in the total number of events. If the reduction is equal to or larger than 5 percent, the events are not considered outliers. Finally, if an event is identified as outlier, we exclude it from calculating the concave hull's new area and from the total number of fatalities. We select the next

closest outlier and repeat the calculations until all potential outliers are assessed, or the number of events removed reaches 5 percent of the total events.

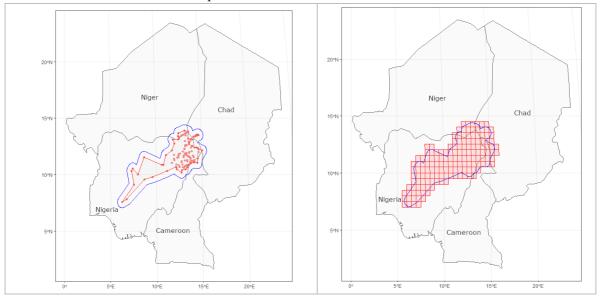


Figure 1: Lake Chad region conflict shape in 2014 overlayed with PRIO-GRIDs. Left: Dots and lines represent conflict events, a concave hull (red) and approximately 50-km buffer (blue), respectively. Right: PRIO-GRIDs (red) covering the conflict shape.

6. We check for what we call the friction mechanism: a conflict intersects substantively in time and space with the anchor conflict yet evolves around a different contested issue. We assume that such a conflict can affect the anchor conflict in the actors involved and in the impact the conflict has on civilians; hence, we add it to the anchor conflict and its relevant conflict events to the conflict shape (yet only in the year[s] in which we observe a substantial overlap). If actors were added via friction mechanism, we check again for outliers. The resulting polygon is the final conflict shape.

Figure 1 shows conflict events used to calculate the concave hull, and the 50km buffer. The red squares are PRIO-GRID, a spatio-temporal grid structure with socioeconomic and environmental information.

2. Analyzing spatial shift of conflict-related violence

Spatial hotspot analysis

We use the Getis-Ord algorithm that analyzes neighboring units to identify hotspots, i.e., spatial clusters of high and low values (Getis and Ord, 1992). Hotspots represent areas with a higher number of conflict events. The Getis-Ord algorithm identifies hotspots by comparing neighboring areas and selecting those with more conflict events. Contrary to other methods such as heatmaps based on kernel density, hotspots identified by the Getis-Ord algorithm are statistically different from their neighboring units rather than based on an arbitrary threshold.

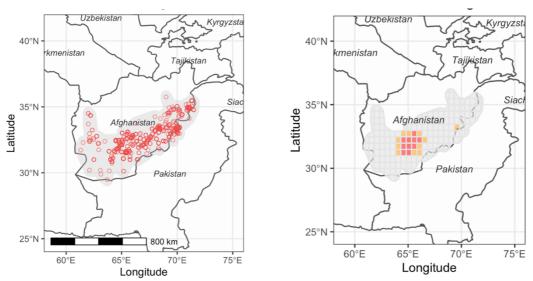


Figure 2: Conflict in the Afghan-Pakistani borderlands in 2006. Left: conflict events (red circles) and conflict shape (gray polygon). Right: Squares used for the calculation of Gi* statistics. Hotspots with a z-score equal to or larger than 1.96 (yellow and red squares).

For the analysis, we build a list of neighboring units, defined as squares that share at least one side, by creating a fishnet with 50x50km squares. We calculate Gi* statistics that return a z-score, a measure of statistical significance, for each square. The statistically significant positive high values of z-scores signal areas of conflict hotspots. We consider squares with a z-score equal to or greater than 1.96 to be a hotspot.

Process tracing

Process tracing helps evaluate whether the proposed relationship between the change in dominant actors and the conflict shape's shifting contraction/expansion is causal or spurious (George and Bennett, 2005: 9). In both probes, we choose two cases as pairs following a most-similar systems design (Gerring, 2017; Przeworski and Teune, 1970):5 the first probe features cases of expansion and the second probe features cases of contraction. The twin objectives of "(1) a representative sample and (2) useful variation on the dimensions of theoretical interest" guide our selection (Seawright and Gerring, 2008: 296). We selected these cases on the independent variable, ensuring in each pair variation in the presence/absence of new dominant actors (explanatory variable) while other characteristics of interest are similar. In both pairs, we conduct a within-case analysis of each case to evaluate the causal mechanism (low-risk/high-opportunity attraction) and a cross-case analysis to compare the two. The second probe includes cases that differ in the conflict duration and contested issues, allowing us to test scope conditions "through progressive pair comparison" (Tarrow, 2010: 251). The second pair thus helps to productively "build contrasts into the research design," and to achieve "a more robust understanding and interpretation" of the first pair of cases (Mukhija, 2010: 423). The differences between the two pairs provide useful "contrast space" to the first pair. Even when comparing different armed multi-actor conflicts, the theoretical dimension—variation in the actor constellation—remains relevant and helps explain variation in the outcome (presence or absence of shift).

Interviews

To establish the causal mechanism that connects the change in dominant actors with spatial shift and to triangulate quantitative data on the conflict shapes, the dominant actors, and the spatial shifts of conflict-related violence, we drew on semi-structured interviews conducted by one of the authors in conflict-affected regions of one case in each probe: Colombia and Syria/Iraq. In Colombia: 59 interviews in Putumayo (2011, 2012), 20 interviews in Catatumbo (2012) and 16 interviews in Arauca (2012, 2016). In Syria/Iraq: 12 interviews in

northern Iraq, close to the Syrian border (2022). For security reasons, we refrain from providing the specific locations. The interviews were conducted in compliance with the authors' home institution's strict ethical and safety guidelines and only after obtaining approval by the institution's Central University Research Ethics Committee (CUREC), as per the reference numbers SSD/CUREC1A/11-240; R48604/RE001; R50663/RE004; CUREC 1A/ODID C1A_22_041. All interviewees provided informed consent. The interview questions evolved around the varying presence of armed actors and their relationship with the regions and the people residing there. The author selected questions from a long catalgoue that were most suitable to the stakeholder group to which the interviewee belonged. Questions were adjusted when necessary: if the author established close trust relationships with the interviewees, she asked them more direct questions, whereas other interviews focused more on contextual information. The interviews generally ranged from 30 minutes to 2 hours. Interviews were conducted in spaces that minimized risks to the interviewees, the interviewer, and any other person involved, e.g. public places, offices, or homes. In all cases, the author only asked questions when interviewees felt comfortable about answering them and when she was confident that the interview would not cause any harm to anyone involved in or affected by the research process. Given the research's sensitive nature, all interviews were anonymous, and all data were treated with absolute confidentiality.

The author used snowball sampling to select interviewees, aiming for a relatively balanced distribution of interviewees across different stakeholder groups to maximize possibilities for triangulation and thus minimize biases in the data. For Colombia, the stakeholder groups comprised ex-combatants, police and military officers, civil society leaders, peasants, refugees, international organization staff, government officials, clerics, and others embedded in or with expert knowledge on conflict actors. For Iraq, the stakeholder groups comprised civil society leaders, youth (above 18 years), international organization staff, government officials, diplomats, and others embedded in or with expert knowledge on conflict actors. In addition to these interviews in challenging conflict settings, we interviewed experts with fieldwork experience in conflict-affected regions to gather additional context information.

3. Description of cases⁷

Armed conflict in Colombia

<u>Anchor conflict:</u> conflict between the Colombian government and left-wing guerrillas based on ideology, conflict actors and events added via mutation mechanism (illegal drug trade component added to the ideological conflict). ⁸

<u>Analysis start date:</u> 1989. While the beginning of the Colombian conflict is a matter of debate among academics, the conflict is possible to trace back at least to the 1950s. The UCDP GED data are available only for the period starting in 1989.

Number of actors included: 11

Region used for the compilation of the list of the potentially involved actors: Colombia, Ecuador, Peru, Brazil, Venezuela, and Panama (in total 34 actors for the period 1989-2016).

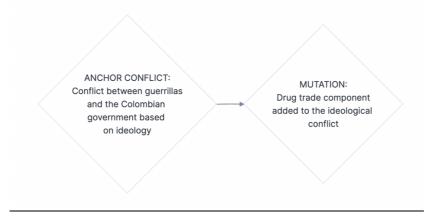


Figure 3: Chart depicting the formation of the armed conflict in Colombia.

Colombia's armed conflict started as a left-wing guerrilla insurgency against Colombia's government in 1964, after bipartisan violence from 1948 to 1958. In the 1990s, the conflict mutated as the illicit drug trade became more central to the conflict. Left-wing guerrillas and paramilitary groups became involved in drug production and trafficking providing them with financial resources. Thus, control over the drug-related resources became a part of the contested issue that resulted in new actors such as drug cartels becoming intertwined with the Colombian conflict (Idler, 2019).

UCDP GED ID	UCDP NAME	FULL/OTHER NAME ⁹	CONFLICT- CONNECTING MECHASNIM
771	AUC	United Self-Defense Forces of Colombia, Autodefensas Unidas de Colombia	Anchor conflict
6506	Bloque Central Bolívar	Central Bolívar Bloc	Anchor conflict
810	Cali Cartel		Mutation
744	ELN	National Liberation Army, Ejército de Liberación Nacional	Anchor conflict
746	EPL	Popular Liberation Army, Ejército Popular de Liberación	Anchor conflict
743	FARC ¹⁰	Revolutionary Armed Forces of Colombia—People's Army, Fuerzas Armadas Revolucionarias de Colombia— Ejército del Pueblo, FARC – EP	Anchor conflict
7013	FARC dissidents		Anchor conflict
17	Government of Colombia ¹¹		Anchor conflict
6579	MAS	Death to Kidnappers, Muerte a Secuestradores	Anchor conflict
774	Medellin Cartel		Mutation
811	PEPES	Persecuted by Pablo Escobar, Perseguidos por Pablo Escobar, Los Pepes	Anchor conflict

Table 1: List of relevant actors for the armed conflict in Colombia

Armed conflict in the Lake Chad region

<u>Anchor conflict:</u> Islamist insurgency led by Boko Haram, conflict actors and events added via spread mechanism (across Nigeria's borders) and friction mechanism (overlap with farmer-pastoralist conflict in central Nigeria). ¹²

Analysis start date: 2009. While Boko Haram was founded in 2002 or 2003, they started using violence in 2009 (Hansen, 2017).

Number of actors included: 13

Region used to compile a list of potentially involved actors: Nigeria, Chad, Cameroon, Niger (in total 58 actors for the period 2009–2016).

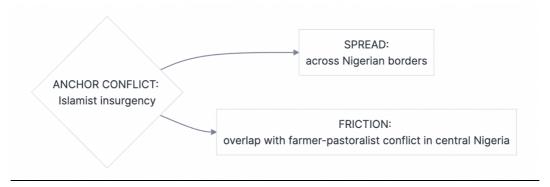


Figure 4: Chart depicting the formation of the armed conflict in the Lake Chad.

Boko Haram started to fight Nigeria's government in 2009 and over time more actors became involved. The military pressure from the Nigerian government and Yan Gora forced the Islamist groups to retreat to the Lake Chad region, and they then engaged in violent activities across Nigeria's borders in Chad, Cameroon, and Niger (Osumah, 2013). This conflict spread led the governments of Cameroon, Chad, and Niger to join Nigeria in the fight against the Islamists (Weeraratne, 2017). Ethno-religious violence in the Jos Plateau and throughout Nigeria is often seen as clashes between farmers and pastoralists. We do not consider the violence in Jos Plateau as a part of the armed conflict because it is not directly connected to the Islamist insurgency in northeast Nigeria. However, the activities of Boko Haram and the Islamic State (IS) spread to central Nigeria and contributed to the blurred distinction between the jihadists from northeast Nigeria and Muslims in central Nigeria. The Islamists' activities enhanced the conflict's narrative as being between Christians and Muslims while omitting the clashes' locally specific roots (Higazi, 2016). Since one of the cleavages between those communities is religion, it is reasonable to assume that increased activity of Boko Haram or Islamic State's West Africa Province (ISWAP) can increase the violent activity there too. Therefore, we include the farmerpastoralist clashes only via the friction mechanism. Since the violence between Christians and Muslims is community-based rather than a conflict between two armed actors, we do not add the UCDP conflict "Christians-Muslims" to the armed conflict, even when there is substantial overlap in time and space. Thus, we treat the resulting events as instances of one-sided violence against civilians.

UCDP			CONFLICT-
GED	UCDP NAME	FULL/OTHER NAME ¹³	CONNECTING
ID			MECHANISM
83	Government of Cameroon		Spread
87	Government of Chad		Spread
75	Government of Niger		Spread
84	Government of Nigeria ¹⁴		Spread
234	\mathbf{IS}^{15}	Islamic State, Daesh	Anchor conflict
1051	Jama'atu Ahlis Sunna	Boko Haram	Anchor conflict
	Lidda'awati wal-Jihad ¹⁶		
4000	Yan Gora		Anchor conflict
4120	Agatu		Friction
4022	Atakar		Friction
589	Birom		Friction
607	Fulani		Friction
608	Hausa		Friction
11040	Sayawa		Friction

Table 2: List of relevant actors for the armed conflict in the Lake Chad region

Actors added through friction mechanism for specific years only:

• 2011: Birom, Fulani, Hausa, Sayawa

• 2014: Birom, Fulani

• 2015: Birom, Fulani, Agatu, Atakar

Armed conflict in the Afghan-Pakistani borderlands

<u>Anchor conflict:</u> Islamist insurgency led by the Taliban Movement of Pakistan (TTP) in FATA, conflict actors and events added via spread mechanisms (conflict in Afghanistan, Islamist violence in wider Pakistan) and friction mechanism (overlap with conflict in Baluchistan). ¹⁷

<u>Analysis start date:</u> 2001. The US-led invasion of Afghanistan in 2001 is seen as one of the casual factors leading to the founding of TTP in 2007.

Number of actors included: 45

Region used to compile a list of potentially involved actors: Afghanistan and Pakistan (in total 67 actors for the period 2001–2016)

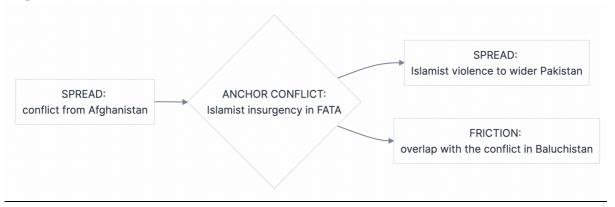


Figure 5: Chart depicting the formation of the armed conflict in the Afghan-Pakistani borderlands.

The TTP's existence and activities are closely connected to the US-led invasion of Afghanistan in 2001, after which high-ranking Taliban and al-Qaeda members fled to the former FATA. We consider the conflict between the Taliban and the US-led coalition connected to the insurgency led by TTP in FATA via the *spread mechanism* for two main reasons. First, the presence of the Taliban and al-Qaeda in FATA resulted in US drone strikes beyond the Afghan border (Aslam, 2011). Second, the presence of Islamists from Afghanistan and foreign fighters from other countries changed the local social fabric. Religious leaders and military commanders replaced the traditional authorities called maliks (Rais, 2019). This change, together with drone strikes that often caused civilian casualties, was conducive to FATA's militarization, the TTP's formation, and the subsequent spread of violence to wider Pakistan. The TTP soon clashed over territorial control with the local tribes and their militias, other armed groups operating in the area, and later also with its splinter groups. Therefore, we include in this armed conflict local militias and other clan-based armed groups that either fought against the TTP or their enemies. Due to their involvement with the TTP, we consider those actors as a part of the anchor conflict. After 2012, the *conflict spread* to wider Pakistan as the TTP carried out attacks in cities such as Karachi and Lahore.

UCDP			CONFLICT-
GED	UCDP NAME	FULL/OTHER NAME ¹⁹	CONNECTING
ID			MECHANISM
769	al-Qaida	Al Qaeda	Spread from
			Afghanistan
398	Ansaar ul-Islam		Anchor conflict
790	Bangesh		Anchor conflict
1046	Fedayeen Islam		Anchor conflict
1100	Forces of Momin Afridi		Anchor conflict
888	Forces of Turkestan		Anchor conflict
	Bhittani		
130	Government of		Spread from
	Afghanistan ²⁰		Afghanistan
142	Government of Pakistan		Spread to Pakistan
3	Government of United	USA	Spread from
	States of America		Afghanistan
359	IMU	Islamic Movement of	Anchor conflict
		Uzbekistan	
234	IS	Islamic State, Daesh	Spread from
			Afghanistan
5928	Jamaat-ul-Ahrar		Anchor conflict
399	Lashkar-e-Islam		Anchor conflict
1177	Lashkar of Akakhel tribe		Anchor conflict
1155	Lashkar of Kukikhel clan		Anchor conflict
2987	Lashkar of Mohmand tribe		Anchor conflict
887	Lashkar of Orakzai tribe		Anchor conflict
796	Lashkar of Salarzai tribe		Anchor conflict
787	Lashkar of Wazir tribe		Anchor conflict
1142	Lashkar of Zakakhel tribe		Anchor conflict
1143	Laskhar of Masozai		Anchor conflict
	Qaumi tribe		

388	LeJ	Lashkar-e-Jhangvi, Army of	Anchor conflict
797	Mangal	Jhangvi	Anchor conflict
303	Taleban ²¹	Taliban	Spread from
000	Taleban	Tunoun	Afghanistan
1178	Tawheed ul-Islam		Anchor conflict
356	TTP	The Taleban Movement of	Anchor conflict
		Pakistan, Tehrik-i-Taliban	Thioner commer
		Pakistan	
3047	TTP-Islahi	TTP - Fazal Saeed Haqqani	Anchor conflict
		faction	Timener commer
5864	TTP-KM	TTP - Khalid Mehsud faction,	Anchor conflict
		TTP - Khan Said Sajna faction	
5865	TTP-SM	TTP - Shehryar Mehsud faction	Anchor conflict
1102	TTP - MR	TTP - Mulla Rafique faction	Anchor conflict
1103	TTP - MT	TTP - Mullah Toofan faction	Anchor conflict
1101	TTP - TA	TTP - Tariq Afridi faction,	Anchor conflict
		Asian Tigers	
791	Turi		Anchor conflict
287	BLA	The Balochistan Liberation	Friction
		Army	
289	BRA	Baloch Republican Army	Friction
375	Forces of Abdul Rahman		Friction
	Khan		
376	Forces of Amanullah		Friction
378	Forces of Arbab Basir		Friction
6356	Government of United		Friction
	Kingdom, Government of		
	United States		
6252	High Council of	Taleban - Faction of Mullah	Friction
	Afghanistan Islamic	Mohammad Rasool Akhund	
	Emirate	and Mullah Mansoor Dadullah	
6318	High Council of	Taleban - Faction of Mullah	Friction
	Afghanistan Islamic	Mohammad Rasool Akhund	
	Emirate, IS	and Mullah Mansoor Dadullah,	
•••		Islamic State	
299	Hizb-i Islami-yi		Friction
	Afghanistan		
292	Jam'iyyat-i Islami-yi		Friction
	Afghanistan		
3192	UBA	United Baloch Army	Friction
304	UIFSA	United Islamic Front for the	Friction
		Salvation of Afghanistan	

Table 3:List of the relevant actors involved in the armed conflict in the Afghan-Pakistani borderlands

Actors added through friction mechanism for specific years only:

- 2001: UIFSA, Government of United Kingdom and Government of United States
- 2003: Hizb-i Islami-yi Afghanistan, Forces of Abdul Rahman Khan, Forces of Amanullah
- 2006: Hizb-i Islami-yi Afghanistan, Forces of Abdul Rahman Khan, Forces of Arbab Basir
- 2007: Hizb-i Islami-yi Afghanistan, Jam'iyyat-i Islami-yi Afghanistan
- 2008: Hizb-i Islami-yi Afghanistan
- 2009: Hizb-i Islami-yi Afghanistan
- 2010: Hizb-i Islami-yi Afghanistan
- 2011: Hizb-i Islami-yi Afghanistan
- 2012: Hizb-i Islami-yi Afghanistan
- 2013: Hizb-i Islami-yi Afghanistan, BLA, UBA
- 2014: Hizb-i Islami-yi Afghanistan
- 2015: Hizb-i Islami-yi Afghanistan, High Council of Afghanistan Islamic Emirate, High Council of Afghanistan Islamic Emirate and IS
- 2016: Hizb-i Islami-yi Afghanistan, High Council of Afghanistan Islamic Emirate, BRA

Armed conflict in Syria/Iraq

<u>Anchor conflict:</u> Islamist insurgency led by IS, conflict actors and events added via spread mechanisms (conflict for Kurdish independence in Iraq, Syria and Turkey; civil war in Syria), mutation (conflict in Iraq after the US-led invasion), and friction mechanisms (conflict between the Iranian government and the Kurds in Iran; conflict between the Turkish government and left-wing groups).

Analysis start date: 2003.

Number of actors included in the analysis: 121

<u>Region used to compile a list of potentially involved actors:</u> Syria, Iraq, Turkey, Iran, Saudi Arabia, Jordan, Lebanon, and Israel (in total 142 actors for the period 2003–2019).

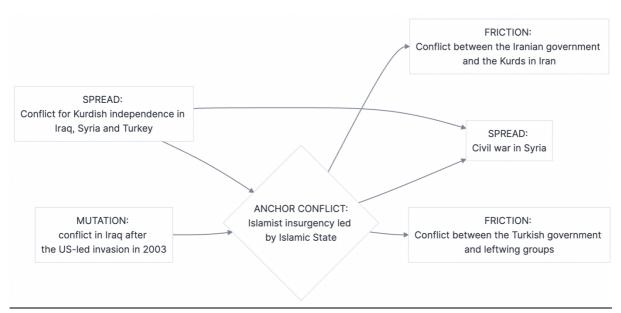


Figure 6: Chart depicting the formation of the armed conflict in Syria/Iraq.

The contemporary Iraqi armed conflict started in 2003 after the US-led invasion. In 2011, Syria's ongoing civil war started after the Syrian government's oppression of the Arab Spring uprising. The same year, the US forces withdrew from Iraq, leaving the Iraqi forces weakened. Al-Qaeda in Iraq (AQI)²² increased offenses on the

Iraqi forces and entered Syria (Brenner, 2019). The newly formed ISIS (Hashim, 2014), AQI's successor, gained positions in large parts of Syria and Iraq, taking advantage of the Sunni grievances bolstered by the Iraqi approach to the Sunni minority. Syrian rebel groups fighting against Syria's government progressed from Syria's peripheries in the north and east toward more central, and then more southern parts, benefiting from local support bases. When they captured Raqqa on March 4, 2013, for example, local inhabitants took down a statue of Assad's father, suggesting their support for the rebels rather than Assad's government (Harding, 2013). As the rebels moved further toward Damascus up to October 2013, they captured territory near Jordan and bombarded Central Damascus. Nonetheless, as the Syrian state forces' bombing of Raqqa shortly after the incident and many other subsequent government-led strikes show, the rebels failed to weaken the Syrian government's position. Syria's government and Syrian rebels are not the only new actors who joined the conflict. Following ISIS's human rights violations committed against Syrian Kurds in 2013, the People's Defense Units (PYD), a Syrian Kurdish armed non-state group, mobilized against the rising ISIS. Evidence of their violent engagements, however, suggests that they were not dominant actors; they did not influence the conflict shape and its hotspots as strongly as ISIS. We consider only those Kurdish armed groups that participated in the conflict in Syria or Iraq as part of this umbrella conflict, and hence the conflict shape. The instability in Iraq after the US-led invasion and the civil war in Syria intensified Kurdish efforts in Iraq and Syria to gain broader autonomy. They fought against other relevant conflict actors in those two conflicts and had their military bases located in or gained recruits from Syria and Iraq. For example, we include the Kurdistan Workers' Party (PKK), a Turkish armed non-state group, as it had several bases in Syria and Iraq and drew recruits from both countries. PKK's battles with the Turkish government took place in Turkey and Syria. In addition, PKK used its military bases in Iraq to launch attacks on the Turkish forces, and the Turkish forces attacked PKK camps in Iraq. We include the Kurdish Democratic Party of Iran (KDPI), an Iranian armed nonstate group only for specific years via the friction mechanism because the group's focus was on regime change in Iran. Although, initially, the KDPI also drew recruits from Iraq, the Iraqi Kurds later formed their own branch.

Note on Syrian rebels

UCDP GED created the aggregated label "Syrian insurgents" because of difficulties identifying which specific groups were involved in individual conflict events. The label covers up to 1,200 groups whose aim was to topple Assad's regime. These include Jabhat al-Nusra, the Free Syrian Army, and many smaller groups—they do not include ISIS or Kurdish groups.²³ Given the composite character of the generic "Syrian insurgents" label, we consider ISIS to continue to be the dominant actor. Even though these groups' total number of engagements is highest (in total 10,551), we can reasonably assume the engagement of each group subsumed under this label to be lower than the total number of ISIS's engagement (412).

Note on Islamic State (IS)

IS is an umbrella identity for loosely connected armed groups across the world. In addition to the brand name, the groups often use the name of the region in which they operate, for example, Islamic State-Khorasan that is active in Afghanistan. We refer to several of these groups in the following way:

- Islamic State of Iraq and Syria (ISIS): armed conflict in Syria and Iraq.
- Islamic State's West Africa Province (ISWAP): armed conflict in the Lake Chad region.
- Islamic State-Khorasan (IS-K): armed conflict in the Afghan-Pakistani borderlands.

Since UCDP GED uses the abbreviation IS for all IS armed groups, we likewise use the abbreviation IS for the groups in the tables that list the conflict actors involved in our case studies and in the network graphs that depict the actors' engagements.

UCDP GED	UCDP NAME ²⁴	FULL/OTHER NAME ²⁵	CONFLICT- CONNECTING
ID			MECHANISM ²⁶
7443	16 th Division, Islamic Front,		Spread (Syrian civil
	Jabhat Fateh al-Sham		war)
783	1920 Revolution Brigades		Spread (Syrian civil
			war)
7520	23 rd Division, Ahrar al-Sham,		Spread (Syrian civil
	Jaysh al Nasr, Jaysh al-Nukhba,		war)
	Nour al-Din al-Zenki, Sham		
	Legion, SNA		
7524	23 rd Division, Jaysh al Nasr,		Spread (Syrian civil
	Jaysh al-Nukhba, Sham Legion,		war)
	SNA, Syrian Liberation Front		
886	AAH	Asaib Ahl al-Haqq	Spread (Syrian civil
			war)
5563	Ahfad al Rasoul Brigade		Spread (Syrian civil
			war)
7672	Ahfad al Rasoul Brigade, Ahrar		Anchor conflict
	al-Sham, al-Farouq Brigades,		
	Ghuraba al-Sham, IS, Jabhat		
	Fateh al-Sham, Jazeera-		
	Euphrates Liberation Front,		
<i>EE(E</i>	Liwa al-Fateh		Anchor conflict
5565	Ahfad al Rasoul Brigade, Ahrar al-Sham, IS, Islamic Kurdish		Anchor conflict
	Front, Jabhat Fateh al-Sham		
5554	Ahrar al-Sham		Spread (Syrian civil
<i>333</i> 4	Alliai ai-Silaili		war)
7676	Ahrar al-Sham, al-Tawhid		Anchor conflict
	Brigade, IS, Islamic Kurdish		
	Front, Jabhat Fateh al-Sham		
5567	Ahrar al-Sham, al-Tawhid		Anchor conflict
	Brigade, IS, Jabhat Fateh al-		
	Sham		
7042	Ahrar al-Sham, FSA, HTS,	FSA: Free Syrian Army,	Spread (Syrian civil
	Jaysh al-Islam	HTS: Hay'at Tahrir al Sham	war)
6702	Ahrar al-Sham, FSA, Jabhat		Spread (Syrian civil
	Fateh al-Sham, Jaysh al-Islam		war)
6572	Ahrar al-Sham, Furqan		Spread (Syrian civil
	Brigades, Jabhat Fateh al-Sham,		war)
	Jaysh al-Islam, Yarmouk Army		
7056	Ahrar al-Sham, HTS, Southern		Spread (Syrian civil
	Front		war)
6389	Ahrar al-Sham, Jabhat Fateh al-		Spread (Syrian civil
	Sham		war)

7498	Ahrar al-Sham, Jabhat Fateh al-Sham, Jaysh al-Islam, Jaysh Asoud al-Sharqiya, Martyr Lieutenant Ahmed Abdou Brigades, Rahman Corps		Spread (Syrian civil war)
6532	Ahrar al-Sham, Jabhat Fateh al- Sham, Southern Front		Spread (Syrian civil war)
7601	Ahrar Ghurayan		Spread (Syrian civil war)
7155	Aknaf Bait al-Maqdis, HTS, Jaysh al-Islam		Spread (Syrian civil war)
7492	Aknaf Bait al-Maqdis, HTS, Jaysh al-Islam, Sons of Yarmouk Movement, Southern Front		Spread (Syrian civil war)
6378	Aknaf Bait al-Maqdis, Jaysh al- Islam, PFLP-GC	PFLP-GC: Popular Front for the Liberation of Palestine- General Command	Spread (Syrian civil war)
5659	al-Mahdi Army		Spread (Syrian civil war)
7426	al-Qa'qa Brigade, Army of Ahl al-Sunni wal Jamaa, Army of al- Ikhlas, Army of Maoata al- Islami, Authenticity and Development Front, Bayareq al- Shaaitat, Islamic Front, Jabhat Fateh al-Sham, Jaish al- Muhajireen wal-Ansar, Liwa al- Qadisiya		Spread (Syrian civil war)
769	al-Qaida		Mutation
3232	al-Tawhid Brigade		Spread (Syrian civil war)
233	Ansar al-Islam		Spread (Syrian civil war)
6393	Authenticity and Development Front		Spread (Syrian civil war)
7455	Authenticity and Development Front, Islamic Front		Spread (Syrian civil war)
7612	Authenticity and Development Front, Islamic Front, Jabhat Fateh al-Sham		Spread (Syrian civil war)
7662	Baz al-Islamiya, Dawn of Freedom Brigades, Islamic Front, Jabhat al-Akrad, Jabhat Fateh al-Sham, Liwa al-Sultan Murad, Majd al-Sham		Spread (Syrian civil war)

7582	Dawn of Freedom Brigades	Spread (Syrian civil war)
7646	Dawn of Freedom Brigades, Euphrates Islamic Liberation Front, Islamic Front, Jabhat al-	Spread (Syrian civil war)
	Akrad, Liwa Thuwar ar-Raqqa	
7663	Dawn of Freedom Brigades, Harkat Hazm, Islamic Front, Jabhat al-Akrad, Jabhat Fateh al-Sham, Mujahideen Army, Nour al-Din al-Zenki, Sham	Spread (Syrian civil war)
	Legion	
7650	Dawn of Freedom Brigades, Jabhat al-Akrad, Northern Storm Brigade	Spread (Syrian civil war)
6410	Eastern Ghouta Unified Military Command	Spread (Syrian civil war)
6390	Eastern Qalamoun Operations Room	Spread (Syrian civil war)
7584	Euphrates Islamic Liberation Front, Islamic Front, Jabhat Fateh al-Sham, Liwa Thuwar ar- Raqqa	Spread (Syrian civil war)
6419	Euphrates Vulcano	Spread (Syrian civil war)
6436	Fatah Halab	Spread (Syrian civil war)
6495	Fatah Halab, Jabhat Fateh al- Sham	Spread (Syrian civil war)
367	Forces of the Caucasus Emirate	Spread (Syrian civil war)
4168	Ghuraba al-Sham, Jabhat Fateh al-Sham	Spread (Syrian civil war)
3822	Government of Australia, Government of United Kingdom, Government of United States of America	Mutation
114	Government of Iran	Friction
116	Government of Iraq	Anchor conflict
121	Government of Israel	Spread (Syrian civil war)
120	Government of Jordan	Spread (Syrian civil war)
119	Government of Lebanon	Spread (Syrian civil war)
57	Government of Russia (Soviet Union)	Spread (Syrian civil war)

118	Government of Syria		Anchor conflict
115	Government of Turkey		Spread
3	Government of United States of America		Mutation
6828	Hamza Division, Liwa al-Sultan Murad		Spread (Syrian civil war)
6015	Harkat Hazm		Spread (Syrian civil war)
6599	Hawar Kilis Operations Room		Spread (Syrian civil war)
366	Hezbollah		Spread (Syrian civil war)
6815	HTS	Hay'at Tahrir al-Sham, Organization for the Liberation of the Levant	Spread (Syrian civil war)
6825	HTS, Rahman Corps		Spread (Syrian civil war)
6866	HTS, Saraya Ahl al-Sham		Spread (Syrian civil war)
7036	HTS, Southern Front		Spread (Syrian civil war)
234	IS	Islamic State	Anchor conflict
5540	IS, Jabhat Fateh al-Sham		Anchor conflict
7669	IS, Syrian insurgents		Anchor conflict
5575	Islamic Front		Spread (Syrian civil war)
6030	Islamic Front, Jabhat al-Akrad, Jabhat Fateh al-Sham, Mujahideen Army		Spread (Syrian civil war)
7437	Islamic Front, Jabhat al-Akrad, Jabhat Fateh al-Sham, Mujahideen Army, SRF		Spread (Syrian civil war)
7649	Islamic Front, Jabhat al-Akrad, Liwa Thuwar ar-Raqqa		Spread (Syrian civil war)
7468	Islamic Front, Jabhat al-Akrad, Mujahideen Army, SRF		Spread (Syrian civil war)
6040	Islamic Front, Jabhat Fateh al- Sham		Spread (Syrian civil war)
7666	Islamic Front, Jabhat Fateh al- Sham, Liwa al-Jihad fi Sabeel Allah, Liwa Thuwar ar-Raqqa		Spread (Syrian civil war)
6069	Islamic Front, Jabhat Fateh al- Sham, Mujahideen Army		Spread (Syrian civil war)
6014	Islamic Front, Mujahideen Army, SRF		Spread (Syrian civil war)
7656	Islamic Front, Northern Storm Brigade		Spread (Syrian civil war)

5541	Jabhat al-Akrad		Spread (Syrian civil war)
1170	Jabhat Fateh al-Sham		Spread (Syrian civil war)
6038	Jabhat Fateh al-Sham, Liwa al- Aqsa		Spread (Syrian civil war)
7615	Jabhat Fateh al-Sham, Liwa al- Qadisiya, Omar al-Mukhtar Battalion		Spread (Syrian civil war)
7461	Jabhat Fateh al-Sham, Rahman Corps		Spread (Syrian civil war)
7674	Jabhat Fateh al-Sham, Salahadin Brigade		Spread (Syrian civil war)
384	Jamaat Jund al-Sahaba		Spread (Syrian civil war)
6380	Jaysh al Fatah Idlib		Spread (Syrian civil war)
6605	Jaysh al-Asha'er		Spread (Syrian civil war)
6377	Jaysh al-Islam		Spread (Syrian civil war)
6600	Jaysh al-Islam, Jaysh Asoud al- Sharqiya, Martyr Lieutenant Ahmed Abdou Brigades, Rahman Corps		Spread (Syrian civil war)
6302	Jaysh al-Jihad (Saraya al- Jihad)		Spread (Syrian civil war)
4357	Jaysh al-Mukhtar		Spread (Syrian civil war)
6367	Jaysh al-Sanadid, Khabour Guards, MFS, PYD	MSF: military wing of Syriac Union Party	Spread (Syrian civil war)
6370	Jaysh al-Sanadid, MFS, PYD		Spread (Syrian civil war)
7150	Jaysh al-Sanadid, PYD		Spread (Syrian civil war)
6626	Jaysh Asoud al-Sharqiya, Martyr Lieutenant Ahmed Abdou Brigades		Spread (Syrian civil war)
4359	JRTN	Army of the Men of the Naqshbandi Order, Naqshbandi Organization	Spread (Syrian civil war)
6037	Liwa al-Aqsa		Spread (Syrian civil war)
6564	Mare' Operations Room		Spread (Syrian civil war)
308	MEK		Mutation

6012	Mujahideen Army		Spread (Syrian civil war)
7128	National Front for Liberation		Spread (Syrian civil war)
7523	National Front for Liberation, SNA		Spread (Syrian civil war)
7525	National Front for Liberation, SNA, Syrian Liberation Front		Spread (Syrian civil war)
7631	Northern Storm Brigade		Spread (Syrian civil war)
6434	Nour al-Din al-Zenki		Spread (Syrian civil war)
7477	Nour al-Din al-Zenki, Sham Legion		Spread (Syrian civil war)
6553	NSA		Spread (Syrian civil war)
1144	Opponents of al-Assad		Spread (Syrian civil war)
323	PKK	Kurdistan Workers' Party	Spread (Kurdish conflict for autonomy)
261	PUK	Patriotic Union of Kurdistan	Spread (Kurdish conflict for autonomy)
4163	PYD	Partiya Yektîya Demokrat, Democratic Union Party	Spread (Kurdish conflict for autonomy)
235	RJF	Reform and Jihad Front	Spread (Syrian civil war)
231	SCIRI	The Supreme Council of the Islamic Revolution in Iraq	Mutation
6288	SDF	Syrian Democratic Forces	Spread (Syrian civil war)
7514	SNA	The Syrian National Army	Spread (Syrian civil war)
6333	Southern Front		Spread (Syrian civil war)
6011	SRF	Syrian Revolutionaries' Front	Spread (Syrian civil war)
1145	Supporters of al-Assad		Spread (Syrian civil war)
7106	Suqour al-Sham Brigades, Syrian Liberation Front		Spread (Syrian civil war)
4456	Syrian insurgents		Spread (Syrian civil war)
6306	Tahrir al- Sham Army		Spread (Syrian civil war)
7371	Tajamu Shuhada al-Sharqiya		Spread (Syrian civil war)

335	MKP	Maoist Communist Party	Friction (left-wing
			groups vs the Turkish
			government)
309	PJAK	Free Life Party of Kurdistan	Friction (Kurdish
			conflict in Iran)
164	KDPI	Kurdish Democratic Party of	Friction (Kurdish
		Iran	conflict in Iran)

Table 4: List of the relevant actors involved in the armed conflict in Syria/Iraq

Actors added through friction mechanism for specific years only:

• 2005: MKP

2016: PJAK, KDPI

2018: PJAK, KDPI

2019: PJAK, KDPI

4. Robustness checks

Alternative data: ACLED data for the armed conflict in the Lake Chad region

Out of our four case studies, only the armed conflict in the Lake Chad region has ACLED data (Raleigh et al., 2010) available for the periods under study. While both data sources provide information on the actors involved, geo-location, and the date of conflict events, ACLED does not impose the threshold of the minimum of 25 battle-related deaths per dyad per year to be included in the dataset. ACLED collects data on more types of conflict events than UCDP GED and consequently contains more data points. Thus, we need to ensure that we have two sets of events complying with the same rules to assess the similarity of the results. To mirror the type of events present in UCDP GED, we include only events that are classified as battles.²⁷

To build the conflict shape, we apply the same procedure we used for the UCDP GED data. The resulting conflict shapes based on ACLED data are similar to those obtained from the UCDP GED data. The number of relevant actors and of conflict events is higher, given the threshold of 25 battle-related death required by the UCDP GED. The additional actors included in the ACLED data are typically communal or ethnic militia involved in less than five battles against Boko Haram or ISWAP. The original dominant actor Boko Haram was replaced by ISWAP while the government of Nigeria remained dominant also in 2016. Contrary to UCDP GED, ACLED data distinguish between police forces and military forces. To compare the results, we consider both military and police forces as representatives of the Nigerian government.

Relevant actors for the Islamist insurgency in Nigeria (2011-2016):

- Agzawaya Communal Militia (Cameroon)
- Attagara Communal Militia (Nigeria)
- Bakaresse Communal Militia (Cameroon)
- Boko Haram Jamaatu Ahli is-Sunnah lid-Dawati wal-Jihad
- BVYG: Borno Vigilance Youths Group
- Civilian JTF: Civilian Joint Task Force
- Djakana Communal Militia (Cameroon)
- Dola Communal Militia (Cameroon)
- Double-Alagarno Communal Militia (Cameroon)
- Faa Communal Militia (Nigeria)

- Gaboua Communal Militia (Cameroon)
- Gangawa Communal Militia (Cameroon)
- Gogone Communal Militia (Niger)
- Islamic State (West Africa)
- Islamic State (West Africa) and/or Boko Haram Jamaatu Ahli is-Sunnah lid-Dawati wal-Jihad
- Jitar Communal Militia (Nigeria)
- Kawuri Communal Militia (Nigeria)
- Kerawa Communal Militia (Cameroon)
- Kolofata Communal Militia (Cameroon)
- Lassa Communal Militia (Nigeria)
- Limani Communal Militia (Cameroon)
- Maiha Communal Militia (Nigeria)
- Maloumri Communal Militia (Cameroon)
- Military Forces of Cameroon (1982-)
- Military Forces of Cameroon (1982-) Rapid Intervention Battalion
- Military Forces of Chad (1990-2021)
- Military Forces of Niger (2011-2021)
- Military Forces of Nigeria (1999-2015)
- Military Forces of Nigeria (2015-)
- Military Forces of Nigeria (2015-) Joint Task Force
- Militia (Ali Kwara)
- Militia (Pro-Government)
- MNJTF: Multinational Joint Task Force
- Mora Communal Militia (Cameroon)
- Mozogo Communal Militia (Cameroon)
- Police Forces of Cameroon (1982-)
- Police Forces of Chad (1990-2021)
- Police Forces of Niger (2011-2021)
- Police Forces of Nigeria (1999-2015)
- Police Forces of Nigeria (2015-)
- Shawa Communal Militia (Nigeria)
- Shuwa Ethnic Militia (Nigeria)
- Tafawa-Balewa Communal Militia (Nigeria)
- Toumour Communal Militia (Niger)
- UN: United Nations
- VGN: Vigilante Group of Nigeria

Actors added through friction mechanism for specific years only:

- 2012: Berom Ethnic Militia (Nigeria)
- 2011, 2012: Christian Militia (Nigeria)
- 2013: Dadiya Ethnic Militia (Nigeria)
- 2014: Eggon Ethnic Militia (Nigeria)
- 2011, 2012, 2014, 2016: Fulani Ethnic Militia (Nigeria)
- 2014: Gbagyi Ethnic Militia (Nigeria)
- 2014: Gbajimba Communal Militia (Nigeria)
- 2012, 2014: Hausa Ethnic Militia (Nigeria)

- 2012: Igbo Ethnic Militia (Nigeria)
- 2014: Irigwe Ethnic Militia (Nigeria)
- 2014: Jukun Ethnic Militia (Nigeria)
- 2014: Kanberi Ethnic Militia (Nigeria)
- 2014: Kotoko Ethnic Militia (Cameroon)
- 2014: Marwa Ethnic Militia (Nigeria)
- 2014: Musgum Ethnic Militia (Cameroon)
- 2011, 2012: Muslim Militia (Nigeria)
- 2011: Muslim Youth Sect Militia (Nigeria)
- 2014: Pai Ethnic Militia (Nigeria)
- 2011: Private Security Forces (Nigeria)
- 2014: Rukuba Ethnic Militia (Nigeria)
- 2014: Shiite Muslim Militia (Nigeria)
- 2014: Tarok Ethnic Militia (Nigeria)
- 2013: Waja Ethnic Militia (Nigeria)

Change in dominant actors

As Table 5 shows, one of the dominant actors changes. The original dominant actor Boko Haram was replaced by ISWAP while the government of Nigeria also remained dominant in 2016. Contrary to UCDP GED, ACLED data distinguish between police forces and military forces. To compare the results, we consider both military and police forces as representatives of the Nigerian government.

ear	ACLED name	Degree centrality
	Boko Haram - Jamaatu Ahli is-Sunnah lid-Dawati wal-Jihad	34
	Police Forces of Nigeria (1999-2015)	22
	Military Forces of Nigeria (1999-2015)	14
2011	Christian Militia (Nigeria)	1
2011	Muslim Militia (Nigeria)	4
	Fulani Ethnic Militia (Nigeria)	:
	Muslim Youth Sect Militia (Nigeria)	
	Private Security Forces (Nigeria)	
	Islamic State (West Africa)	18
	Military Forces of Nigeria (2015-)	17
	Islamic State (West Africa) and/or Boko Haram - Jamaatu	
	Ahli is-Sunnah lid-Dawati wal-Jihad	8
	Military Forces of Cameroon (1982-)	4
	Military Forces of Niger (2011-2021)	1
	Police Forces of Nigeria (2015-)	
	Military Forces of Cameroon (1982-) Rapid Intervention	
	Battalion	
	Boko Haram - Jamaatu Ahli is-Sunnah lid-Dawati wal-Jihad	
	Police Forces of Cameroon (1982-)	
	Civilian JTF: Civilian Joint Task Force	
	Military Forces of Chad (1990-2021)	
	Limani Communal Militia (Cameroon)	
	Kolofata Communal Militia (Cameroon)	
2016	Shawa Communal Militia (Nigeria)	
	Agzawaya Communal Militia (Cameroon)	
	Military Forces of Nigeria (2015-) Joint Task Force	
	UN: United Nations	
	Mozogo Communal Militia (Cameroon)	
	Fulani Ethnic Militia (Nigeria)	
	Mora Communal Militia (Cameroon)	
	Djakana Communal Militia (Cameroon)	
	Bakaresse Communal Militia (Cameroon)	
	Toumour Communal Militia (Niger)	
	Double-Alagarno Communal Militia (Cameroon)	
	Kerawa Communal Militia (Cameroon)	
	Police Forces of Niger (2011-2021)	
	Maloumri Communal Militia (Cameroon)	
	Gangawa Communal Militia (Cameroon)	

Table 5: Degree centrality results based on the ACLED data.

Spatial shift results

Also according to ACLED data, the Lake Chad region conflict shape experienced a shifting contraction. It contracted by 46.7 percent (UCDP: 48.8 percent). It also shifted since the hotspots and conflict shapes overlapped by 25.9 percent (UCDP: 42.7 percent) and 48.3 percent (UCDP: 48.3 percent), respectively. The visual representation of the spatial shift is similar regardless of which conflict event data we use (see Figure 7).

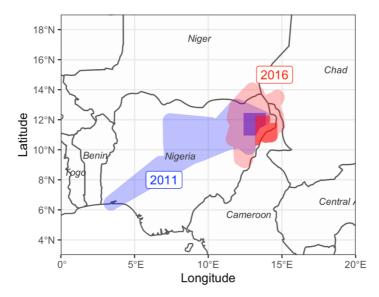


Figure 7: Shifting contraction of the Lake Chad region conflict shape, 2011–2016. Calculations based on ACLED data. Blue shading: conflict shape in 2011. Red shading: conflict shape in 2016. Darker blue and darker red shapes: hotspots.

Alternative operationalization of dominant actors

We use the degree centrality measure (that counts the total number of vertices) for the individual vertices to identify the dominant actors within an armed conflict. Nodes denote conflict actors, and vertices represent conflict events between the actors. Degree centrality thus counts the total number of conflict events in which a given conflict actor was involved. We assume that the most active conflict actors are also the dominant actors. We calculate three further network measures as alternative approaches to identifying the dominant actors (see Table 6).

First, we count the number of unique nodes connected to the given conflict actors. This measure assumes that the most important actors are those who fought many other conflict actors without considering the frequency of those violent encounters. This measure often assigns similar values to actors with a widely different frequency of engagement in conflict events. Consequently, this measure undervalues the importance of actors involved in a high number of conflict events if those events always involve only one other conflict actor. For instance, in 2008, the Taliban fought in 937 events against one conflict actor. The Lashkar-e-Islam militia was engaged in 13 conflict events battling two enemies, meaning that the Taliban is assigned a lower importance score than Lashkar-e-Islam based on the number of nodes. Given our definition of dominant actors, this measure is not suitable for identifying the dominant actors or their change.

Armed conflic	et e				easures							
Name	Year	UCDP ID	Name	Degree	N o	fnodes	nodes Eigenvector			Katz		
				Value	Value	Same DA	Value	Same DA	Value	Same DA		
		17	Government of Colombia	87	3		1		1.35			
		771	AUC	1	1		0.01		1.12			
Armed conflict in	2006	744	ELN	8	1	Yes	0.1	Yes	1.14	Yes		
Colombia -		746	EPL	2	1		0.03		1.14			
period I		743	FARC	78	2		0.99		1.25			
	2011	17	Government of Colombia	44	1	Yes	1	Yes	1.34	Yes		
	2011	743	FARC	44	1	108	1	1 08	1.11	1 03		
	2012		Government of Colombia	27	1	Yes	1	Yes	1.11	Yes		
Armed conflict in	2012	743	FARC	27	1	103	1	1 03	1.11	1 03		
Colombia -		17	Government of Colombia	13	3		1		1.11			
period II	2016		ELN	10	1	Yes	0.98	Yes	1.13	No		
ponou n	2010		EPL	1	1	103	0.1	103	1.13	110		
			FARC	2	1		0.2		1.13			
			Government of Nigeria	55	1		1		1.11			
			Hausa	2	1		0	Yes	1.11			
	2011	589	Birom	13	1	No	0		1.11	No		
	2011	1001	Boko Haram	55	1	110	1		1.11	110		
			Sayawa	2	1		0		1.11			
Armed conflict in			Fulani	13	1		0		1.11			
the Laker Chad			Government of Niger	12	1		0.06		1.16			
region			Government of Nigeria	217	2		1		1.29			
			Government of Cameroon	32	2		0.13		1.29			
	2016		Government of Chad	3	1	No	0.02	Yes	1.16	Yes		
		234		234	5		1		1.61			
			Yan Gora	6	1		0.03		1.16			
			Boko Haram	36	2		0.15		1.26			
		130	Government of Afghanistan	713	2		1		1.22			
	2006	2	Government of United States of America	10								
			Taleban	19	1	No	0	Yes	1.11	Yes		
			Hizb-i Islami-yi Afghanistan	707	1		0.01		1.12			
			al-Qaida	6	1		0.01		1.12			
			Bangesh	19 29	2		0		1.11			
			Government of Pakistan	419			0		1.23			
			Lashkar of Salarzai tribe	7			0		1.15			
Armed conflict in			Mangal	12			0		1.13			
the broader			Ansaar ul-Islam	11	1		0		1.12			
Afghanistan-			Forces of Turkestan Bhittani	1	1		0		1.15			
Pakistan border			Forces of Shah Sahib	4	1		0		1.15			
area		7,72	Government of United States				V		1.15			
	2008	3	of America	31	1	No	0	Yes	1.11	No		
		130	Government of Afghanistan	951	2		1		1.22			
			Turi	39			0		1.25			
			TTP	429			0		1.47			
			Lashkar-e-Islam	13			0		1.24			
			al-Qaida	31			0		1.11			
			Taleban	937	1		1		1.12			
			Hizb-i Islami-yi Afghanistan	14			0.01		1.12			
	لـــــا	277	J. 1.11B.1101111	1 7			0.01		1.12			

Armed conflic	t		Conflict actors	Centrality measures						
Name	Year	UCDP ID	Name	Degree	N o	f nodes	Eigenvector		Katz	
				Value	Value	Same DA	Value	Same DA	Value	Same DA
	2010	114	Government of Iran	7	1		0		1.11	
		116	Government of Iraq	241	1		1		1.11	
		115	Government of Turkey	59	1	No	0	Yes	1.11	No
	2010	309	PJAK	7	1	110	0	103	1.11	140
		234	IS	241	1		1		1.11	
		323	PKK	59	1		0		1.11	
		114	Government of Iran	4	1		0		1.11	
		116	Government of Iraq	309	1		0		1.21	
		115	Government of Turkey	7	1		0		1.11	
		4357	Jaysh al-Mukhtar	3	1		0		1.11	
		366	Hezbollah	2	1		0		1.13	
		118	Government of Syria	10640	3		1		1.54	
		5540	IS, Jabhat Fateh al-Sham	54	2		0		1.32	
		3232	al-Tawhid Brigade	7	1		0		1.21	
			Ahfad al Rasoul Brigade, Ahrar al-							
			Sham, IS, Islamic Kurdish Front,							
		5565	Jabhat Fateh al-Sham	9	1		0		1.21	
			Ahrar al-Sham, al-Tawhid Brigade,							
Armed conflict in			IS, Jabhat Fateh al-Sham	36	1		0		1.21	
Syria and Iraq -		234		412	8		0.01		2.1	
period I			Ahfad al Rasoul Brigade	4	1		0		1.21	
		7582	Dawn of Freedom Brigades	1	1		0		1.21	
	2013		Ahfad al Rasoul Brigade, Ahrar al-			No		Yes		No
			Sham, al-Farouq Brigades, Ghuraba							
			al-Sham, IS, Jabhat Fateh al-Sham,							
			Jazeera-Euphrates Liberation Front,				_			
		7672	Liwa al-Fateh	14	1		0		1.21	
			Ahrar al-Sham, al-Tawhid Brigade,							
		7.7.	IS, Islamic Kurdish Front, Jabhat	22					1.01	
			Fateh al-Sham	22	1		0		1.21	
			PJAK	7	1		0		1.11	
			PKK MEK	3	1		0		1.11	
			Jabhat Fateh al-Sham	5	1 2		0		1.11	
			Syrian insurgents	10551	1		1		1.15	
		4163		160	8		0		2.1	
			Islamic Front	2	1		0		1.21	
			Northern Storm Brigade	10	1		0		1.21	
			Jabhat al-Akrad	4	1		0		1.13	
		3341	Jauman ai-Aktau	4	1		U		1.13	

Armed conflic	t		Conflict actors			Centi	rality measures				
Name	Year	UCDP ID	Name	Degree	N o	f nodes		envector	I	Katz	
				Value	Value	Same DA	Value	Same DA	Value	Same DA	
		116	Government of Iraq	696	1		0		1.72		
		234	IS	2749	35		0.06		7.23		
		118	Government of Syria	10211	3		1		2.06		
		366	Hezbollah	11	3		0		2.07		
		119	Government of Lebanon	8	1		0		1.72		
		3	Government of United States of America	7	1		0		1.11		
			Government of Israel	2	1		0		1.21		
			IS, Jabhat Fateh al-Sham	6	1		0		1.22		
			Jabhat Fateh al-Sham	36	4		0		2.26		
			Harkat Hazm	4	1		0		1.23		
			Jabhat Fateh al-Sham, Liwa al-Aqsa	10	1		0		1.21		
			Euphrates Vulcano	144	1		0		1.72		
			Ahrar al-Sham, Jabhat Fateh al-Sham	3	1		0		1.72		
			Mare Operations Room	1	1		0		1.72		
			16th Division, Islamic Front, Jabhat Fateh al-Sham	7	1				1.72		
		7455	Authenticity and Development Front, Islamic Front	/	1		0		1.72		
		7504	Euphrates Islamic Liberation Front, Islamic Front,	22	1				1.70		
			Jabhat Fateh al-Sham, Liwa Thuwar ar-Raqqa	33	1		0		1.72		
		7601	Ahrar Ghurayan	0	1		0		1.22		
		7612	Authenticity and Development Front, Islamic Front, Jabhat Fateh al-Sham	84	1		0		1.72		
		7012		04	1		0		1.72		
			Dawn of Freedom Brigades, Euphrates Islamic Liberation Front, Islamic Front, Jabhat al-Akrad,								
		7646	Liwa Thuwar ar-Raqqa	37	1		0		1.72		
		7040	Dawn of Freedom Brigades, Jabhat al-Akrad,	37	1		- 0		1.72		
		7650	Northern Storm Brigade	9	1		0		1.72		
		7050	Baz al-Islamiya, Dawn of Freedom Brigades, Islamic		1				1.72		
			Front, Jabhat al-Akrad, Jabhat Fateh al-Sham, Liwa								
		7662	al-Sultan Murad, Majd al-Sham	49	1		0		1.72		
			Dawn of Freedom Brigades, Harkat Hazm, Islamic								
			Front, Jabhat al-Akrad, Jabhat Fateh al-Sham,								
Armed conflict in			Mujahideen Army, Nour al-Din al-Zenki, Sham								
Syria and Iraq -		7663	Legion	14	1		0		1.72		
period II	2014	6393	Authenticity and Development Front	16	1	No	0	Yes	1.72	No	
		115	Government of Turkey	5	1		0		1.11		
		4359	JRTN	6	1		0		1.72		
		4456	Syrian insurgents	9586	1		1		1.21		
			al-Qaida	7	1		0		1.11		
		4163	PYD	313	4		0		2.17		
			Islamic Front	71	1		0		1.72		
		6011		36	3		0		2.07		
			Mujahideen Army	7	1		0		1.72		
		6014	Islamic Front, Mujahideen Army, SRF	246	1		0		1.72		
		6020	Islamic Front, Jabhat al-Akrad, Jabhat Fateh al-	^			_		1 70		
			Sham, Mujahideen Army Islamic Front, Jabhat Fateh al-Sham	136	1		0		1.72		
		0040	Islamic Front, Jabhat Fateh al-Sham, Mujahideen	130	1				1./2		
		6069	Army	31	1		0		1.72		
			Southern Front	1	1		0		1.72		
			Jaysh al-Sanadid, PYD	25	1		0		1.72		
		7150	al-Qa'qa Brigade, Army of Ahl al-Sunni wal Jamaa,		1				1.72		
			Army of al-Ikhlas, Army of Maoata al-Islami,								
			Authenticity and Development Front, Bayareq al-								
			Shaaitat, Islamic Front, Jabhat Fateh al-Sham, Jaish								
		7426	al-Muhajireen wal-Ansar, Liwa al-Qadisiya	99	1		0		1.72		
			Islamic Front, Jabhat al-Akrad, Jabhat Fateh al-								
		7437	Sham, Mujahideen Army, SRF	2	1		0		1.72		
		7460	Islamic Front, Jabhat al-Akrad, Mujahideen Army,	21					1.70		
		7468		21	1		0		1.72		
		7615	Jabhat Fateh al-Sham, Liwa al-Qadisiya, Omar al-	1	1		^		1 72		
		/015	Mukhtar Battalion	1	1		0		1.72		
		7640	Islamic Front, Jabhat al-Akrad, Liwa Thuwar ar- Raqqa	18	1		0		1.72		
			Islamic Front, Northern Storm Brigade	10	1		0		1.72	26	
		, 0.50	Islamic Front, Jabhat Fateh al-Sham, Liwa al-Jihad fi	10	1				1.72	26	
		7666	Sabeel Allah, Liwa Thuwar ar-Raqqa	34	1		0		1.72		
			PKK	5	1		0		1.11		
		343	<u>[= ====</u>			L			4		

Armed conflic	t		Conflict actors	Centrality measures						
Name	Year	· UCDP ID Name		Degree	N o	f nodes		envector	I	Katz
				Value	Value	Same DA		Same DA	Value	Same DA
		114	Government of Iran	18	3		0		1.74	
		116	Government of Iraq	559	1		0.04		1.51	
		115	Government of Turkey	261	2		0.01		1.62	
		234	IS	2367	25		0.26		5.08	
		120	Government of Jordan	4	1		0		1.51	
		366	Hezbollah	3	3		0		1.8	
		119	Government of Lebanon	4	1		0		1.51	
		121	Government of Israel	1	1		0		1.18	
		118	Government of Syria	4356	3		1		1.87	
		6378	Aknaf Bait al-Maqdis, Jaysh al-Islam, PFLP-GC	3	1		0		1.51	
		6436	Fatah Halab	61	2		0		1.75	
		6495	Fatah Halab, Jabhat Fateh al-Sham	28	2		0		1.75	
		6389	Ahrar al-Sham, Jabhat Fateh al-Sham	13	2		0		1.75	
			Ahrar al-Sham, Jabhat Fateh al-Sham, Southern							
		6532	Front	62	1		0		1.51	
		5554	Ahrar al-Sham	17	1		0		1.11	
		6564	Mare, Äô Operations Room	14	1		0		1.51	
		6599	Hawar Kilis Operations Room	150	2		0.01		1.75	
			Jaysh al-Islam, Jaysh Asoud al-Sharqiya, Martyr Lieutenant Ahmed Abdou Brigades,							
Armed conflict in		6600	Rahman Corps	15	1		0		1.11	
Syria and Iraq -	2016		Ahrar al-Sham, FSA, Jabhat Fateh al-Sham,			No		Yes		No
period II		6702	Jaysh al-Islam	5	1		0		1.51	
		7461	Jabhat Fateh al-Sham, Rahman Corps	22	1		0		1.16	
			Ahrar al-Sham, Jabhat Fateh al-Sham, Jaysh al-							
			Islam, Jaysh Asoud al-Sharqiya, Martyr							
		7400	Lieutenant Ahmed Abdou Brigades, Rahman	1.5	1		0		1.51	
			Corps KDPI	15	1		0		1.51	
			PJAK	5	1		0		1.17	
			PKK	191	1		0		1.16	
			Southern Front	4	1		0		1.51	
			Jabhat Fateh al-Sham	29	2		0		1.69	
			Syrian insurgents	3449	1		0.97		1.19	
			Islamic Front	1	1		0		1.51	
		6014	Islamic Front, Mujahideen Army, SRF	4	1		0		1.51	
		6288	· · · · · · · · · · · · · · · · · · ·	685	6		0.05		2.39	
			Jaysh al-Islam	27	2		0		1.62	
		6553	•	8	1		0		1.51	
		6037	Liwa al-Aqsa	17	1		0		1.11	
		6306	Tahrir al- Sham Army	15	1		0		1.11	
			Jaysh Asoud al-Sharqiya, Martyr Lieutenant							
			Ahmed Abdou Brigades	2	1		0		1.51	
		7477	Nour al-Din al-Zenki, Sham Legion	1	1		0		1.51	

Table 6: Comparison of the results of the centrality measures used to identify dominant actors. Dominant actors in blue (beginning of the observed period) and red (end of the observed period). *DA means dominant actors.

Second, we calculate eigenvector centrality. As with the degree centrality, this measure considers the number of connected nodes and the total number of conflict events in which a given conflict actor was involved. In addition, eigenvector centrality assigns a higher score to nodes with well-connected neighbors. In our case, conflict actors fighting other conflict actors with a higher centrality score also receive more points for such connections. In an undirected network of n nodes, the eigenvector centrality x_i of node i is proportional to the sum of the centralities of i's neighbor. Thus, the eigenvector centrality is defined as

$$x_{i} = \kappa^{-1} \sum_{j=1}^{n} A_{ij} x_{j} , \qquad (1)$$

where the adjacency matrix is denoted as A_{ij} and the constant of proportionality κ^{-1} is equal to the largest eigenvalue of the adjacency matrix (Newman, 2018: 159–161). Using eigenvector centrality leads to identifying the same dominant actors as in the case of degree centrality.

Third, we compute Katz centrality, which adds a small value to all nodes regardless of their connections and thus is a convenient measure for less connected networks. We define Katz centrality x_i in a network with n nodes as

$$x_i = \alpha \sum_{j=1}^{n} A_{ij} x_j + \beta$$
 (2)

where α is the eigenvector centrality term summing all centralities of the nodes that point to node i and β is the constant adding extra value to each node in the network (Newman, 2018: 163).

Our results suggest that Katz centrality often assigns the same value to many conflict actors and identifies as important those that are connected to at least two other conflict actors. For instance, in 2008, the measure assigns higher importance to the Mangal tribe, which was involved in 12 conflict events with two other conflict actors, namely Turi and Bangesh tribes, than to the Afghan government, which was involved in 951 conflict events with two other conflict actors. This measure, mainly due to its use of constant β , undervalues less connected actors with high frequency involvement. Katz measure provides similar results as degree centrality and eigenvector centrality if used in less connected networks, such as the armed conflict in Colombia.

In general, network centrality measures identify the most important nodes within the network. They differ based on the applied assumptions of what constitutes an important node within a network. Given how we conceptualize dominant actors, the centrality measures that adequately represent our conceptualization of dominant actors are degree centrality, eigenvector centrality, and to some extent Katz centrality.

Alternative spatial shift explanation

As an alternative explanation of spatial shift, we further consider the change in the proportion of how many actors are fighting the main state actors in relation to the overall number of actors involved in a given conflict. We also consider the change in the proportion conflict events that the state actor was engaged in, in relation to the total number of conflict events assigned to a given conflict. The results do not explain spatial shift, as there is no substantive difference between spatial change (such as contraction or expansion) with or without shift (see Table 7).

Armed	conflict			Actors		Conflict events			
						N of confl.	Total N of	Proportion	
			N of actors inv.	Total N of	Proportion	events with	confl.	of confl.	
Name	State actor	Year	with the gov.	actors	of actors	the gov.	Events	Events	
		2006	3	5	0.6	87	88	1	
		2007	3	4	0.8	77	77	1	
Armed conflict in Colombia	Government of Colon	2008	2	3	0.7	54	54	1	
(2006 - 2011)	dovernment of colon	2009	2	3	0.7	65	65	1	
		2010	2	3	0.7	75	75	1	
		2011	1	2	0.5	44	44	1	
		2012	1	2	0.5	27	27	1	
Armed conflict in Colombia		2013	2	3	0.7	21	21	1	
Armed conflict in Colombia (2012 - 2016)	Government of Color	2014			0.7	35	35	1	
		2015	3	4	0.8	28	28	1	
		2016	3	4	0.8	_		1	
		2011	1	6	0.2	55	70	0.8	
	Government of Nigeria	2012	1	2	0.5	120	120	1	
Armed conflict in the Lake Chad region (2011 - 2016)		2013	1	3	0.3	92	103	0.9	
		2014	1	5	0.2	169	184	0.9	
		2015	2	11	0.2	166	251	0.7	
		2016	2	7	0.3	217	270	0.8	
Armed conflict in the wider	Government of	2006	2	5	0.4	713	732	1	
Afghanistan-Pakistan border	Afghanistan	2007	1	14	0.1	849	956	0.9	
area (2006 - 2008)	, ugnamotan	2008	2	15	0.1	951	1464	0.6	
		2010	1	6	0.2	241	307	0.8	
Armed conflict in Syria and	Government of Iraq	2011	2	12	0.2	141	465	0.3	
Iraq (2010 - 2013)		2012	1	11	0.1	145	8346	0	
		2013		24	0	309	11135	0	
<u> </u>		2010		NA	NA	NA	NA	NA	
Armed conflict in Syria and	Government of	2011			0.2	208		0.4	
Iraq (2010 - 2013)	Syria	2012			-	8073		1	
		2013				10640		1	
Armed conflict in Syria and		2014		-	-	696		0.1	
Iraq (2014 - 2016)	Government of Iraq	2015			_			0.1	
		2016							
Armed conflict in Syria and	Government of	2014				10211		0.8	
Iraq (2014 - 2016)	Syria	2015			-	5994		0.8	
	•	2016	3	36	0.1	4356	6213	0.7	

Table 7: Proportion of the actors involved with the main state actor and proportion of conflict events with the involvement of the main state actor.

Alternative hotspot operationalization

We use the Getis-Ord algorithm to identify hotspot locations (Getis and Ord, 1992). We create squared grids to obtain the list of neighboring units and calculate z-scores for all units. In the article, we rely on the size of the grids, 50x50 km at the equator. Given the potential spatial error in the UCDP GED data estimated by Weidmann (2015), we believe that smaller grids would be inappropriate.²⁹ To test our method's robustness for identifying spatial shift, we calculate Getis-Ord hotspots with larger grids, namely 75x75 km and 100x100 km at the equator (see Table 8). Our results show that even though larger-sized grids tend to produce smaller overlaps, all three grid sizes identify the same type of spatial change. Thus, we are confident that spatial change is not affected by our choice of grid size.

	Hotspot overlap								
Armed conflict	Grid size 50 km	Grid siz	e 75 km	Grid size 100 km					
	%	%	Same SC*	%	Same SC*				
Armed conflict in Colombia (2006 - 2011)	78.7	65.5	Yes	56.8	Yes				
Armed conflict in Colombia (2012 - 2016)	0.0	0.0	Yes	0.0	Yes				
Armed conflict in the Lake Chad region (2011 - 2016)	42.7	51.9	No	23.6	Yes				
Armed conflict in the broader Afghanistan- Pakistan border area (2006 - 2008)	56.1	51.9	Yes	28.6	Yes				
Armed conflict in Syria and Iraq (2010 - 2013)	0.0	0.0	Yes	0.0	Yes				
Armed conflict in Syria and Iraq (2014 - 2016)	74.2	86.4	Yes	68.0	Yes				

*Table 8: Results of the alternative grid specification. *SC means spatial change.*

Alternative conflict shape operationalization

How we form the buffer for the final conflict shape polygons might affect the type of identified spatial change. We use a 50 km buffer to mitigate potential spatial errors in the UCDP GED estimated by Weidmann (2015). We also build conflict shapes with buffers of 75 km and 100 km to test our results' stability. We identify the same spatial shifts regardless of the buffer sizes, the only exception being Colombia in 2012–2016, where the buffer size affected the location of hotspots resulting in the hotspot overlap of 12.2 percent.

We further assess the results of building the conflict shape based on conflict dyads, i.e., pairs of conflict actors, as reflected in georeferenced event datasets, such as UCDP GED (Sundberg and Melander, 2013). Figure 8 shows the difference between our approach and the dyadic one in the conflict in Syria/Iraq. In map 8.1, the conflict shape is based on conceptualizing conflict as a fluid multi-actor phenomenon. In map 8.2, the conflict shape is based on conceptualizing it as a dyad. The conflict in Iraq's dyadic representation as a conflict between Iraq's government and ISIS ignores ISIS battles with other conflict actors across Kurdish territories in Syria, Iraq, Turkey, and Iran. It omits the emergence of a new dominant actor, Syria's government. The conflict shape's shifting expansion into Syria remains an analytical blind spot.

	Conflict overlap								
Armed conflict	Buffer size 50 km	Buffer si	ze 75 km	Buffer size 100 km					
	%	%	Same SC*	%	Same SC*				
Armed conflict in Colombia (2006 - 2011)	88.8	89.5	Yes	90.2	Yes				
Armed conflict in Colombia (2012 - 2016)	91.3	92.0	No	92.8	Yes				
Armed conflict in the Lake Chad region (2011 - 2016)	48.3	53.3	Yes	57.5	Yes				
Armed conflict in the broader Afghanistan- Pakistan border area (2006 - 2008)	96.4	97.0	Yes	97.6	Yes				
Armed conflict in Syria and Iraq (2010 - 2013)	58.9	64.3	Yes	68.3	Yes				
Armed conflict in Syria and Iraq (2014 - 2016)	93.1	93.4	Yes	93.8	Yes				

Table 9: Results of the alternative buffer specification. *SC means spatial change.

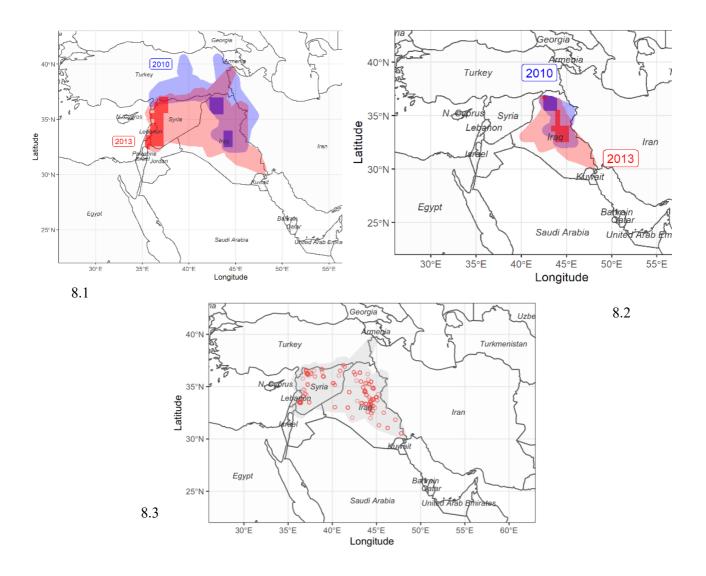


Figure 8: Multi-actor conceptualization of the conflict (8.1) and dyad Government of Iraq and ISIS (8.2), with the conflict shape in light and conflict hotspots in dark color, respectively. Conflict events with the involvement of ISIS in 2013 (8.3). Gray polygon and red circles represent the conflict shape and conflict events, respectively, per the multi-actor conceptualization proposed in this article.

Alternative operationalization of the conflict shape: Wzone dataset

We construct polygons from Kikuta's (2022) Wzone dataset as an alternative to the conflict shapes. The Wzone Dataset relies on UCDP GED data and provides daily polygons for conflict dyads and armed conflicts. Unlike conflict shapes, Wzone polygons apply weights to conflict events based on the number of casualties assigned to a given conflict event. Figure 9 shows Wzone polygons with respect to the conflict events relevant to the armed conflict in the Lake Chad region in 2011 (9.1) and 2016 (9.2). The Wzone polygon for 2011 does not overlap with some of the conflict events. There are no Wzone polygons for the Hausa-Fulani and Hausa-Sayawa dyads. The Wzone polygon for 2016 covers a large area without conflict event.

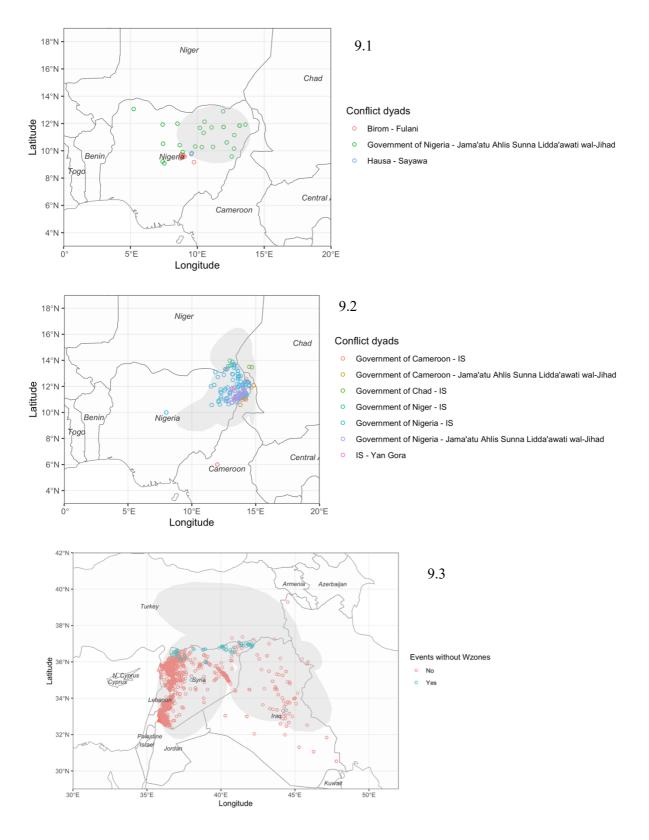


Figure 9: Wzone polygons (gray polygons) and conflict events (circles). Lake Chad region conflict shape in 2011 (9.1) and 2016 (9.2); Syria/Iraq conflict shape in 2013 (9.3).

We replicate the spatial analysis of cases and present the results for all cases. We select the conflict dyads based on the list of actors involved in each case study. Per Kikuta's (2022) recommendation, we create Wzone polygons for individual dyads and aggregate them into a final polygon for the umbrella conflict. We calculate

hotspots based on the Getis-Ord algorithm. WzoneData miss some of the dyads we included in our original analysis, such as Government of Afghanistan-Hizb-i Islami-yi Afghanistan (armed conflict in the Afghan-Pakistani borderlands, 2008) and AUC-FARC (armed conflict in Colombia, 2006). Dyads without Wzone polygons typically represent few conflict events except for the dyad Government of Syria-IS in 2010. However, if spatially clustered, they may affect the hotspot location or the final polygon shape. Figure 9.3 shows how conflict events of dyads without Wzone polygons cluster in the Syria-Turkey border area in 2013.

The replication using the Wzone dataset differs from the results obtained with the methodology to construct conflict shapes. It missed one of the three spatial shifts we demonstrated in the article: it suggests expansion rather than shifting contraction for the conflict shape in the Lake Chad region (2011-2016). We also include non-dominant actors in the analysis to probe our claim's relevance and not bias our empirical analysis against non-dominant actors.

Colombia conflict shape (2006–2011)

Overlap: 95.02 percent (shape: 100 percent, hotspots: 90.04 percent)

Spatial change: no shift

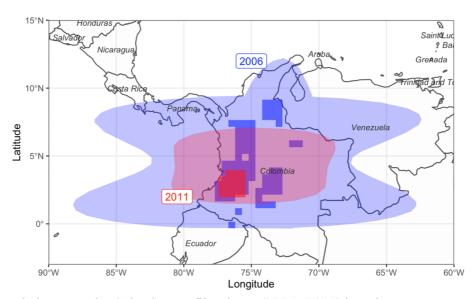


Figure 10: Spatial change in the Colombia conflict shape (2006 - 2011) based on wzoneData (Kikuta, 2022). The hotspots are the darker blue and red shapes.

Colombia conflict shape (2012–2016)

Overlap: 32.55 percent (shape: 65.10 percent, hotspots: 0 percent)

Spatial change: shift

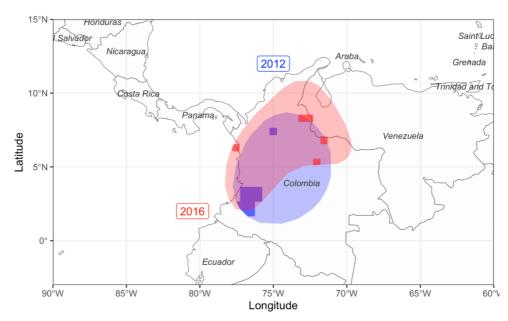


Figure 11: Spatial change in the Colombia conflict shape (2012 - 2016) based on wzoneData (Kikuta, 2022). The hotspots are the darker blue and red shapes.

Lake Chad region conflict shape (2011–2016)

Overlap: 50.56 percent (shape: 65.30 percent, hotspots: 35.81 percent)

Spatial change: no shift

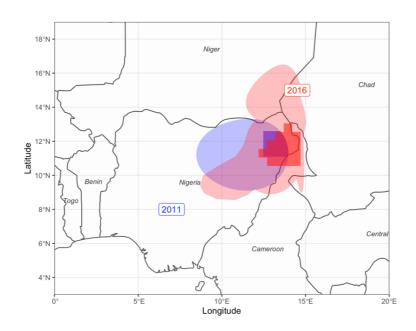


Figure 12: Spatial change in the Lake Chad region conflict shape (2011 - 2016) based on WzoneData (Kikuta, 2022). The hotspots are the darker blue and red shapes.

Afghan-Pakistani conflict shape (2006 – 2008)

For our calculations, we remove the United States-al-Qaeda dyad because not all of the conflict events assigned to this dyad are part of the armed conflict in the Afghan-Pakistani borderlands.

Overlap: 71.48 percent (shape: 100 percent, hotspots: 42.96 percent)

Spatial change: no shift

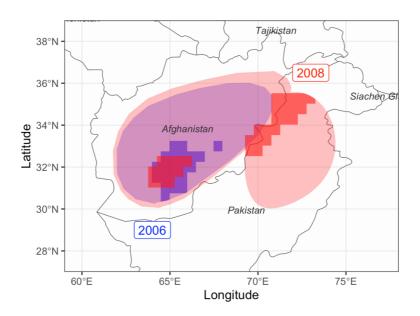


Figure 13: Spatial change in the Afghan-Pakistani conflict shape (2006 – 2008) based on WzoneData (Kikuta, 2022). The hotspots are the darker blue and red shapes.

Syria/Iraq conflict shape (2010–2013)

We remove the United States-al-Qaeda dyad because not all of the conflict events assigned to this dyad are part of the armed conflict in Syria/Iraq.

Overlap: 37.54 percent (shape: 75.09 percent, hotspots: 0 percent)

Spatial change: shift

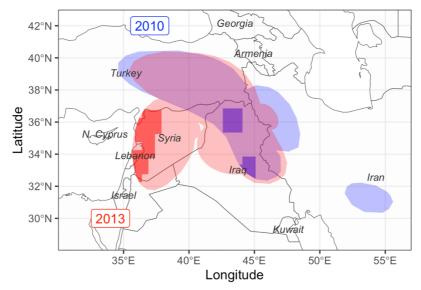


Figure 14: Spatial change in the Syria/Iraq conflict shape (2010 - 2013) based on WzoneData (Kikuta, 2022). The hotspots are the darker blue and red shape

Syria/Iraq conflict shape (2014–2016)

We remove the United States-al Qaeda dyad because not all of the conflict events assigned to this dyad are part of the armed conflict in Syria/Iraq.

Overlap: 54.04 percent (shape: 87.61 percent, hotspots: 20.46 percent)

Spatial change: no shift

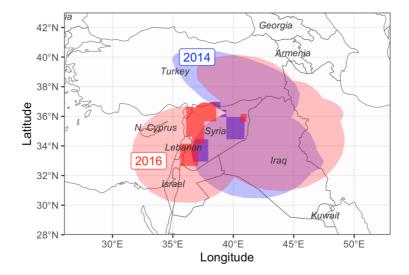


Figure 15: Spatial change in the Syria/Iraq conflict shape (2014 - 2016) based on WzoneData (Kikuta, 2022). The hotspots are the darker blue and red shapes.

Software

We used R software version 4.1.2 "Bird Hippie" for all calculations. List of the key R packages: Tidyverse (Wickham et al., 2019), Concaveman (Gombin et al., 2017), Igraph (Csardi and Nepusz, 2006), Network (Butts, 2008), Rnaturalearth (South, 2017), Sf (Pebesma, 2018), Ggplot2 (Wickham, 2016).

Data availability

Data are available on the journal website as well as on GitHub, together with the relevant R code: https://github.com/Global-Security-Programme/Conflict-shapes-in-flux

References

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¹ The authors and trained research assistants collected the data; the authors double-blind coded a random sample of coded data for each case to ensure the quality and consistency of coding.

² To filter out the events from the UCDP GED we use their variable where_prec and include all events with values equal or smaller than four.

³ The first potential outlier is the most detached event. We select the rest of the possible outliers by moving systematically in a counter-clockwise spiral until all potential outliers are examined.

⁴ The percentage is loosely guided by the methodology for the convex hull of the UCDP polygon, which uses 20 percent.

⁵ We follow George (1979) and consider a case to be 'an instance of a class of events of interest to the investigator,' quoted in Bennett (2004: 20–21).

⁶ For a discussion of 'contrast space,' see Collier and Mahoney (1996: 88).

⁷ Data for the cases were collected by the authors and research assistants.

⁸For more detailed discussions of the Colombian armed conflict see, e.g., (Gray, 2008; McDougall, 2009; Safford and Palacios, 2002; Sanín, 2008).

⁹ For instances when widely used terms or abbreviations are different than the ones in UCDP GED.

¹⁰ Dominant actor in the studied period 2006-2011.

¹¹ Dominant actor in the studied period 2006-2011.

¹² For more detailed discussions on Boko Haram insurgency see, e.g., (Aghedo and Osumah, 2015; Foyou et al., 2018; Iyekekpolo, 2016).

¹³ For instances when widely used terms or abbreviations are different than the ones in UCDP GED.

¹⁴ Dominant actor in the studied period 2011-2016.

¹⁵ Dominant actor in the studied period 2015-2016.

¹⁶ Dominant actor in the studied period 2011-2015.

¹⁷ For more detailed discussions on armed conflict in FATA and its historical context, see, e.g., (Haroon, 2011; Nojumi, 2002; Rehman, 2017; Schoffeld, 2010; Shinwari, 2010).

¹⁸ In 2018, FATA merged with the neighboring province Khyber Pakhtunkhwa.

¹⁹ For instances when widely used terms or abbreviations differ from those used in UCDP GED.

²⁰ Dominant actor in the studied period 2006-2008.

²¹ Dominant actor in the studied period 2006-2008.

²² AQI changed its name to Islamic State in Iraq in 2012 and later to Islamic State in Iraq and Syria (ISIS).

²⁴ Actors often joining forces to fight a common enemy are listed together with their own ID in UCDP GED.

²⁵ For instances when widely used terms or abbreviations differ from those used in UCDP GED.

²⁶ Some of the conflict actors are involved in more than one conflict depicted in Figure 6. If this is the case, the column gives preference to the conflict in which the actor was most involved.

²⁷ ACLED defines battles as 'a violent interaction between two politically organized armed groups at a particular time and location.' (ACLED, 2022: 8).

²⁸ ACLED contains many actors with generic names such as unidentified armed group, unidentified militia, or rioters. Given the lack of information about these actors, we do not include them.

²⁹ See, for example, PRIO GRID (Tollefsen et al., 2012) that uses also grids 50 kmx50 km.