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ANNOUNCEMENT:

See Gordon M. Hahn’s most recent article “The Caucasus Emirate Jihadists: The Security and Strategic Implications,” in Stephen J. Blank, ed., *Russia’s Homegrown Insurgency: Jihad in the North Caucasus* (Carlisle Barracks, PA: U.S. Army War College Strategic Studies Institute, October 2012), pp. 1-98, www.strategicstudiesinstitute.army.mil/pdf/PUB1116.pdf.

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SPECIAL REPORT

Two-mode Social Network Analysis of the Caucasus Emirate, Subgroups, and Violent Incidences in the Northern Caucasus

Written and Researched by SETH GRAY

[Edited and Supervised by Gordon M. Hahn]

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SPECIAL REPORT: 2-mode Social Network Analysis of the Caucasus Emirate, Subgroups and Violent Incidences in the Northern Caucasus

By Seth Gray

In order to delve into the intricacies of the Northern Caucasian insurgent and terrorist organization, the Caucasus Emirate (CE), a look at several driving dynamics is important. Through a combination of geospatial data and social network analysis, we can analyze the intense localization of violence and question the CE's network structure. This is the first attempt at looking at the CE through a combination of analyses that better illuminates the relationships between the CE network's groups and their alignment with the CE. This first attempt also narrows down on Dagestan as it is well known to be the epicenter of violence in the Northern Caucasus. As other locations have proportionally declined in violence, Dagestan has proportionally increased. There are several factors, which are explored later on that are driving this increase in violence in Dagestan.

A major concern is that historically, violence has been difficult to mitigate in more heterogeneous areas of the Northern Caucasus. The CE has used ethnic, clan, and religious tensions to create a greater disruption, thus fomenting insurgent activities. Dagestan is the least homogenous of the region's republics, offering a greater variety of ethnicity and religion. As time passes, the ability to mitigate violence within the area will lessen and a greater chance of the spread of violence into other regions as well as other countries will increase. More complex mitigating tactics and strategies will be required in order to reduce violence within the area as well as reduce the operational capacity of the Caucasus Emirate.

Geography and Ethnicity

Geography and ethnicity play large roles in how the Caucasus Emirate and its sub-groups operate within each of its spheres of operation. These spheres are represented by the CE's four main vilaiyats: Vilaiyat Nokchikcho (Chechnya), Vilaiyat Gyalgyaiche (Ingushetia and Ossetia), Vilaiyat Dagestan, and the United Vilaiyat of Kabardia, Balkaria, and Adygea. Each of these vilaiyats are comprised of a number of sectors in turn comprised of combat jamaats and special operational groups comprised of a single ethnic group or multiple ethnicities dependent upon the geographic location.

Dagestan consists of sixty indigenous ethnicities and has representatives of 102 different ethnic groups.¹² The largest group, the Avars, along with the Dargins, Lezgins, Laks, Tabasarans, and Chechens belong to Ibero-Caucasian family of languages. The Nakh-speaking people represent the largest collection of ethnicities. There is also the Altai family of languages represented by Kumyks, Nogais and Azeris and the Tat-based language (related to Farsi) represented by Russians, Tats, and mountain Jews. To add to the complexity is Dagestan's village jamaat structure, which differs from the Chechen *teip* system. Jamaats are characterized by a collection of geographically contiguous villages tied together by *tukhumi* or large, tribe-like clan. Historically, these jamaats have governed territories and protected the lands in which these villages are settled. Political structures in Dagestan are for the most part powerless. Strong-man politics, individual actors operating within the greater political context are the main impetus with which negotiations occur and deals are made. They are also the focal points for frustrations as they are commonly targeted instead of the institutions themselves.

¹ Gordon Hahn. *Russia's Islamic Threat*, Yale University Press, 2007, p.23.

² See Appendix 1 for census data for Dagestan.

This gives rise to the potentiality of a high localization of violence as a jamaat or a collection of jamaats are endowed to maintain their area of operations which is commonly a village, an administrative district or in some cases a city e.g. Makhachkala. Kisriev describes these groups as “ethnopolities.” The old definition of jamaat (localized, ethnically homogenous chain of villages) is supplanted by a newer, slightly modified one. The jamaats that are mostly responsible for the violence today comprise members of single and multiple ethnic groups “concerned primarily with local interests.”³ These highly cellular and localized units of society could be equated with that of a porous honeycomb. Although each jamaat is unique, it can be related to other units through language, extended blood and kinship, and religion. Unfortunately, there is no data on the ethnic makeup of specific jamaats that comprise the CE but it cannot be ruled out that these groups exploit ethnic and sociocultural divisions within the populations they operate in.⁴ For example, throughout 1989 there were a series of interethnic and interpolitical disputes which demonstrated that a regional Islamic Congress would be impossible to create. As a result of these quarrels and the dissolution of the Soviet Union, a vacuum in power had emerged resulting in a highlighted fragmentation of society and the movement of traditional jamaats and ethnopolities to fill the void. The members that comprise “ethnopolities” may form alliances with other parties and other jamaats to address local issues.

Another conflict in May of 1997 may exemplify these extremely complicated social structures that form and dissolve so rapidly. In the Dagestan the ‘Kadar zone’ consisting of Kadar, Chabanmakhi, and Karamakhi villages, a dispute occurred between Sufi and Wahhabi Muslims over simple protocol of prayer during a funeral, which resulted in the deaths of multiple

³ Robert W. Schaefer. *The Insurgency in Chechnya and the North Caucasus: From Gazavat to Jihad*, Praeger Security International, 2011,p.73

⁴ See Appendix 1.

civilians and a standoff that lasted until the Interior Military Special Task Force (OMON) blockaded access to the village, thus isolating the tensions and stopping outsiders from joining the fray.⁵ Flash groups are fairly endemic to the Caucasus as they address local issues that require quick resolve. In regards to the conflict in Chabanmakhi, the Wahhabis decided to control the roads passing through their area of operation. A proclamation was issued by Dagestani authorities, declaring force was necessary to quell violence and mobilizing surrounding village jamaats against the ‘Wahhabis’. This decree, a hierarchical approach to an asymmetric and convoluted conflict essentially fell flat as these groups dissolved and then reformed in late December of 1997 to form the “Military Mutual Assistance Treaty” which involved Chechen government forces and the self-styled independent state, the Islamic Jamaat of Dagestan.’

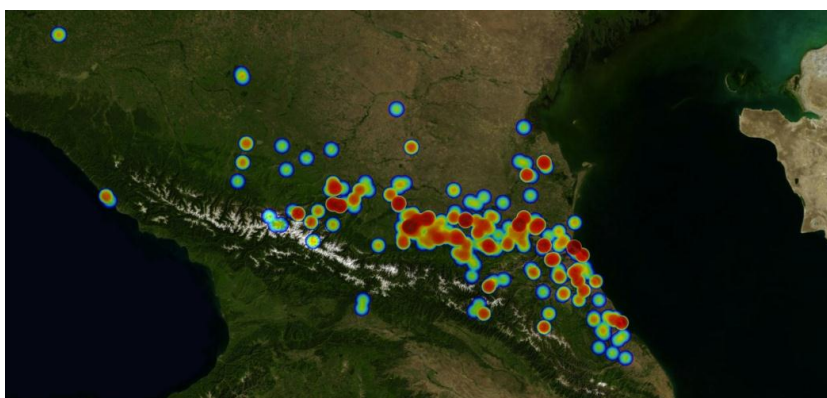


Figure 1: Visualized Hot-Spot map of violence in the Northern Caucasus

⁵ Kisreiv, p.102. Tensions did not subside as sharp politicization in the village continued in the village between Sufis and Wahhabis through 1998 and 1999, leading to the August 1999 invasion of Dagestan led by Khattab, Al Qa'ida's amir in the North Caucasus, and his close associate Shamil Basaev, who served as Deputy prime Minister and then Prime Minister of the Chechen Republic of Ichkeriya from 1997 to July 1998.

Keeping these factors in mind, let us turn to our examination of the CE network and violence, using various geospatial and social network methodologies. Data used for calculating hotspots and other measures was retrieved from Jane's Insurgency and Terrorism Database Event Center.⁶ This data consists of violent incidences occurring from January 1st 2007 to December 31st 2011. The database does not parse data according to specific attributes such as fatality and casualty rates for specific actors such as security forces, non-state combatants and civilians, or for methods of attack. This data was hand coded given the summary of each event generated by the database. Subsequently, this data was uploaded to a Palantir investigation and combined with relational data garnered from Dr. Gordon Hahn's *Islam, Islamism and Politics in Eurasia Report* (IIPER).⁷ Palantir is a proprietary software platform used for information analysis and it allows streamlined data integration processes necessary for data aggregation. Palantir allows the combination of structured data (Excel tables) and unstructured data (doc, PDF, jpeg) upon which a user may combine attribution data and if desired, the export of this data for deeper and more refined analyses.⁸ After aggregation within Palantir, the data was then exported into a file format readable by ArcGIS, a suite of geographic information system (GIS) used for multiple forms of geospatial analysis.

⁶ See <http://home.janes.com/info/jtic/>

⁷ Gordon M. Hahn, *Islam, Islamism, and Politics in Eurasia Report* (IIPER), Center for Strategic and International Studies, <http://csis.org/node/33013/publication>.

⁸ See <http://www.palantir.com/government> for greater software description.

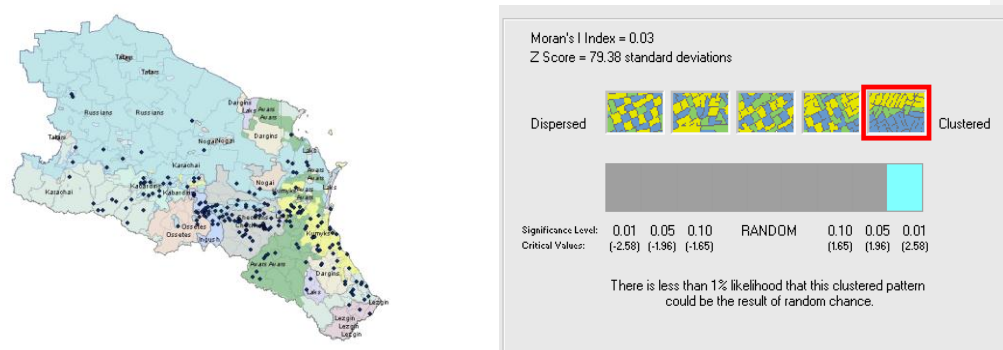


Figure 2: ArcGIS visualization of violent incidences within historical ethnic territories in the Northern Caucasus republics. The localization of violence was determined by running spatial autocorrelation measure Moran's I.

Jane's Database attributes all violence North Caucasus political violence to the CE due to data aggregation purposes. But as discussed earlier, the violence can not be completely attributed to one specific group and must be further defined. In order for all the violence to be attributed to CE, the 28 jamaats in this study would have to claim complete allegiance as well as work with each other in concert. A prime example is the dissonance that occurred between the subgroups of Doka Umarov and Hussein Gakaev in the summer of 2010. After Umarov publicized he would be stepping down as acting Emir and then rescinded the change of leadership to members within the Nokhchicho Villaiyat, Gakaev effectively splintered the group from allegiance with CE and committed subsequent acts of violence independently. Although Gakaev reunited with Umarov months later, this act highlights the ability of these groups to act autonomously, focusing on localized issues such as the governance of territory. Localization also indicates that groups specifically have areas of operation and remain anchored to these spots. Some groups have been documented to move to distances outside of their areas as evidenced by al-Khattab's International Islamic Brigade. However, his brigade was composed mainly of foreign fighters

who did not have ties to a local village. They were mobile in nature.⁹ The spatial autocorrelation measure (Moran's I) found the clustering at an index of 0.03 with a z-score of 1.96. This indicates that there is less than a 1% chance that the acts are occurring at specific locations by random chance.¹⁰ There was no indication that one ethnicity was responsible for more violence than others, but more acts fell into historically controlled Avar and Chechen territories even when normalized for area. Another trend that was found was the gradual eastward movement of the distribution of violence. From 2007 onward, the distribution of violence has moved eastward towards Dagestan, ending with the majority of violence in Dagestan territory in 2011. These ellipses signify a geographical mean center as well as directionality. Factors that may contribute to the eastward movement of the geographical mean center are the suppression of Chechen jamaats by Ramzan Kadyrov. Under Kadyrov, violence has reached unprecedented lows. Another causal factor is that Dagestani jamaats are outperforming Chechen and Ingushetian jamaats.¹¹ These ellipses only show gradual trends in dispersion and do not highlight individual jamaat behavior temporally.

⁹ Kisriev, p.104.

¹⁰ *Moran's I measure:*
$$I = \frac{N}{\sum_i \sum_j w_{ij}} \frac{\sum_i \sum_j w_{ij} (X_i - \bar{X})(X_j - \bar{X})}{\sum_i (X_i - \bar{X})^2}$$

¹¹ Gordon Hahn. "Islam, Islamism and Politics in Eurasia Report," No. 55, 16 April 2012. Available at: <http://csis.org/node/33013/publication>.

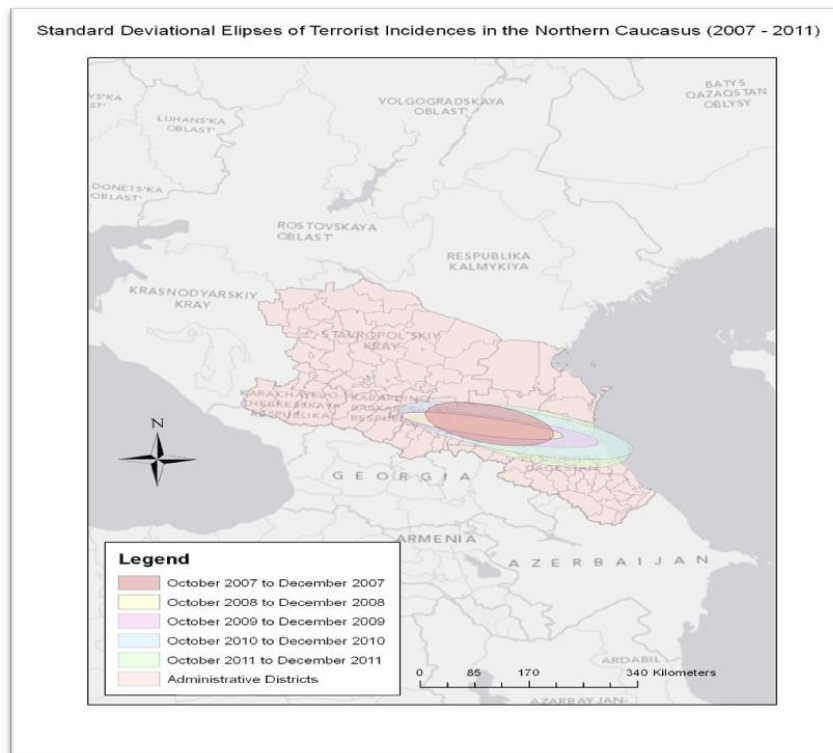


Figure 3: Distributional ellipses signifying the distribution of SIGACT data over 4 month increments. October 2011 through December 2011's ellipse is primarily concentrated in Dagestan territory.

The Caucasus Emirate as a Dark Network

Departing from a geospatial perspective, we turn to social network analysis in order to point out with greater detail, the structure of the CE as well as illuminate future possibilities as to the composition and location(s) of the network. The CE is a dark network. It operates in secrecy. Dark networks are typically illicit organizations that hide themselves from authorities and can range from terrorist and criminal organizations but they can be networks that simply want to

remain unknown.¹² Sean Everton of the Naval Post Graduate School notes the study of dark networks has increased after 9/11 and social network analysis (SNA) has become a tool of choice for governments wishing to analyze criminal, terrorist and insurgent groups.¹³ Dark networks offer a greater challenge to analysts because of a lack of data. Because they operate clandestinely and because much of the information gathered is classified, open-source methods have their limitations. As a result, the limitations of SNA are noted in the small amount of statistically significant analyses performed. However, given the data and in conjunction with other combined qualitative and quantitative analyses, we can shed some light on this network and unravel some of the complexities as to the causes of violence and the actors involved.

First and foremost, let us first discuss what social network analysis is and the analytical tools it will bring to this analysis. SNA has been around since the 1930's stemming from the work of Jacob Moreno.¹⁴ Moreno created sociograms, diagrams used in the analysis of people and their relationships. From "Sociometrics" to current Social Network Analysis, SNA is an analysis of a network of entities: people, groups, events, intuitions, etc. and how these entities interact with each other within a network structure and over time. Through analysis, key nodes (actors) are identified as well as their influence on other nodes. A recent contribution from the social sciences, SNA has been used in research within a multitude of disciplines, from identifying key trading partners and streamlining organizational structures in large corporations to mapping in order to disrupt criminal and terrorist organizations. For this study, three individual SNA platforms were used: Pajek, UCINET, and ORA (Organizational Risk

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¹² Sean Everton, *Disrupting Dark Networks*, Cambridge University Press, November 2012.

¹³ Mark Sageman studied 172 operatives involved in the global Salafi jihad.

¹⁴ Stanley Wasserman and Katherine Faust "Social Network Analysis in the Social and Behavioral Sciences" *Social Network Analysis: Methods and Applications*. Cambridge University Press, pp. 1-27.

Assessment). Each program should be noted as having both strengths and weaknesses and were utilized in tandem to overcome weaknesses they may generate in any analysis if used alone.

Pajek, which means “spider” in Slovenian, was written in 1996. Created by Vladimir Batagelj and Andrej Mrvar at the University of Ljubljana, Pajek was initially designed to analyze organic molecules in Chemistry and protein-receptor interaction networks but quickly became of use to the Social Sciences. Pajek enables an analyst to separate components and subgroups from a large network, highlight links between them, and visualize them using subquadratic algorithms. UCINET was developed by Steve Borgatti, Martin Everett, and Lin Freeman out of the University of Irvine, California.¹⁵ Although it cannot handle datasets within the millions as Pajek can, it offers some measures not present in Pajek and which have been employed in this study. It also utilizes an external program, NetDraw, to visualize datasets. The last SNA platform, ORA, was developed at Carnegie-Melon and hosts similar measures but also allows geospatial measures such as Moran’s I and Gary’s C to run if geospatial attributes are uploaded. After data was exported from the Palantir investigation, there was a serious lack of relational data between individual actors. There was data pertaining to top CE leadership but there wasn’t enough data to correlate with the amount of violent incidences. This indicates either a lack of data as just mentioned, or a greater amount of variance in the causes of violence. Again, we must question how pervasive and how contributive the role of the CE is in the republics and question how we can create a better ontology as to defining violence. So, a 2-mode network analysis was chosen given the amount of data in which two categories are utilized instead of one i.e. agent by agent to agent by event. Instead, a jamaat by event network was used. A two-mode

¹⁵ S.P. Borgatti and M.G. Everett, and L.C. Freeman, *Ucinet for Windows: Software for Social Network Analysis*. Harvard, MA: Analytic Technologies, 2002.

approach analyzes two kinds of entities such as actors and events, which are employed within this thesis. Within an affiliation matrix, rows represent actors and columns represent events. Signifying a relationship between an actor and an event is defined by a numerical value such as 1. So, as a matrix is filled out, a 1 is marked within the row where the actor meets the event, or i meets j . This graph is used when the researcher can suppose that actor(s) participate in the event, which indicates “a potential for some form of social bond between them.”¹⁶ Building on the work of Borgatti and Everett, a two-mode approach and normalization is utilized to account for the adjacency of nodes. For degree centrality (the amount of events and actor is tied to), the following equations are defined as:

$$\text{Actor } x \text{ normalized centrality} = \frac{c_D(x)}{m}$$

$$\text{Event } y \text{ normalized centrality} = \frac{c_D(y)}{n}$$

Figure 5: Normalized degree centralities for actor and event

In other words, the degree centrality for an actor is the number of events tied to that particular actor and the opposite for an event. In this, the actor’s degree is taken and then divided by the number of events and an event’s degree is taken and divided by the number of actors. This process is done for each event and each actor. Output is commonly generated as *indegree* and *outdegree*. In this context, *outdegree* is used because the relationship is directed, meaning the jamaat is the cause of the event. In order to link an event to a jamaat, areas of operation were constructed by using administrative districts controlled by specific jamaat(s). Because areas of operation are not restricted to the boundary of an administrative district, an additional 10km were

¹⁶ Carrington, Scott, Wasserman, ed. *Models and Methods in Social Network Analysis*. Cambridge University Press, 2005.

given to simulate mobility. Events that fall within the established proximities are then linked to the jamaat. Out of 1554 events, 532 were connected to jamaats by proximity and the remaining isolates were removed from the data set for the time being.

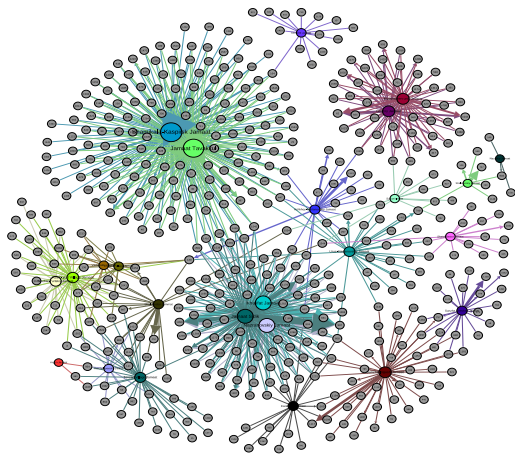
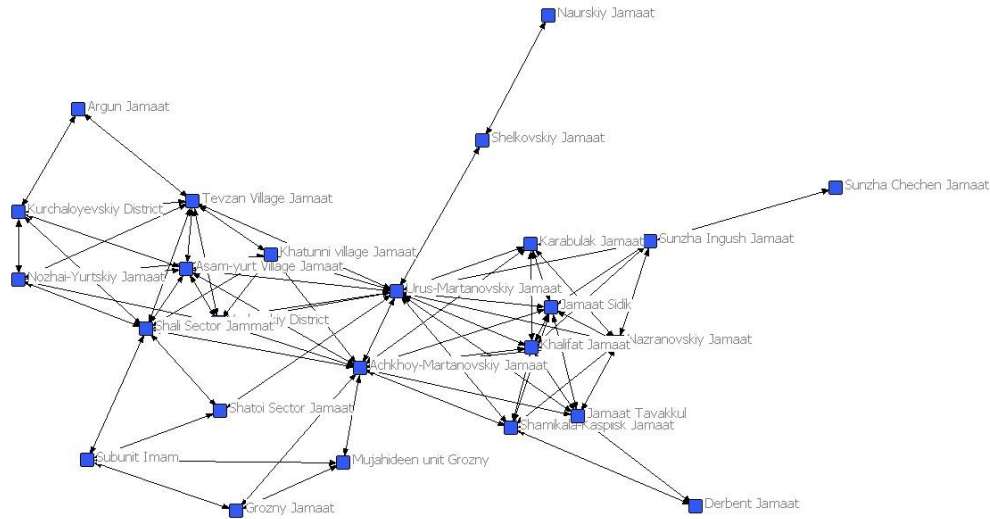


Figure 5. Jamaat by Event network visualization. Jamaats are colored individually and events are left gray.

Figure 5 shows the graph for the North Caucasus in its entirety (552 events connected to 28 jamaats). Isolates were removed due to visualization purposes. As one can visually derive from this graph, the network is fairly sparse, meaning that there is a small percentage of ties out of the entirety of possible connections, e.g. every node is connected to every node. This is fairly normal given the data that has been worked with. We can also see that the network has many sub-groups. This is also normal given the overlap of operational areas. From here the network is conflated from a 2-mode network, into a 1-mode. Currently, there is a serious lack of 2-mode analytical tools and the network is converted in order to run 1-

mode centrality measures et al. After the network is conflated and then binarized, the appropriate centralities can be run:

Figure 6. Conflated and binarized network with a Fruchterman-Reignold layout of the Caucasus Emirate network (see Figure 5 for the previous 2-mode network).



Degree Centrality Measures

Figure 7. UCINET output for Degree Centrality

FREEMAN'S DEGREE CENTRALITY MEASURES					
		OutDegree	InDegree	NrmOutDeg	NrmInDeg
		-----	-----	-----	-----
23	Shamikala-Kaspiisk Jamaat	239	239	1.072	1.072
24	Jamaat Tavakkul	228	228	1.022	1.022
22	Jamaat Sidik	180	180	0.807	0.807

21	Khalifat Jamaat	180	180	0.807	0.807
20	Nazranovskiy Jamaat	180	180	0.807	0.807
14	Mujahideen unit Grozny	86	86	0.386	0.386
13	Grozny Jamaat	86	86	0.386	0.386
18	Sunzha Ingush Jamaat	80	80	0.359	0.359
9	Shali Sector Jammāt	44	44	0.197	0.197
2	Urus-Martanovskiy Jamaat	42	42	0.188	0.188
5	Asam-yurt Village Jamaat	41	41	0.184	0.184
1	Achkhoy-Martanovskiy Jamaat	41	41	0.184	0.184
17	Sunzha Chechen Jamaat	39	39	0.175	0.175
11	Tevzan Village Jamaat	35	35	0.157	0.157
4	Nozhai-Yurtskiy Jamaat	34	34	0.152	0.152
19	Karabulak Jamaat	30	30	0.135	0.135
16	Shelkovskiy Jamaat	23	23	0.103	0.103
7	Kurchaloyevskiy District	23	23	0.103	0.103
10	Shatoi Sector Jamaat	18	18	0.081	0.081
25	Derbent Jamaat	17	17	0.076	0.076
6	Vedenskiy District	17	17	0.076	0.076
12	Khatunni village Jamaat	16	16	0.072	0.072
3	Subunit Imam	11	11	0.049	0.049
15	Naurskiy Jamaat	5	5	0.022	0.022
8	Argun Jamaat	3	3	0.013	0.013

Figure 8. Descriptive Statistics for Degree Centrality

DESCRIPTIVE STATISTICS				
	1	2	3	4
	Outdegree	InDegree	NrmOutDeg	NrmInDeg
-----	---	-----	-----	-----
1 Mean	6	6	25	25
2 Std Dev	3.464	3.464	14.434	14.434
3 Sum	150	150	625	625
4 Variance	12	12	208.333	208.333
5 SSQ	1200	1200	20833.332	20833.332
6 MCSSQ	300	300	5208.333	5208.333
7 Euc Norm	34.641	34.641	144.338	144.338
8 Minimum	1	1	4.167	4.167
9 Maximum	15	15	62.5	62.5
10 N of Obs	25	25	25	25
Network Centralization (Outdegree)		= 39.063%		
Network Centralization (Indegree)		=39.063%		

Given the output of Freeman's Degree Centrality, three of the top five most active jamaats are located in Dagestan. In an information network where ties were determined by information sharing, Shamikala-Kaspisk jamaat and Jamaat Tavaakul would be considered the most influential. One thing that must be accounted for is that areas of operation vary from jamaat to jamaat and the smaller the area; the likelihood of clustering is greater. Another factor that cannot be visualized is the rate at which jamaats share fighters and if a jamaat may claim responsibility for an attack occurring in another area under different jurisdiction. Data garnered from IIPER reports suggests that this is a regular occurrence. However, this data was so sparse that it could not be factored into this analysis of 552 events to be significantly viable. Given that the network centralization for the entire network is 39.063%, we can conclude that there is a moderate amount of concentration or centralization within the entire network i.e. the power of specific jamaats within the entire network varies moderately but this does not suggest that one jamaat may be emphasized over another in terms of importance, power, or authority.

Figure 8. Bonacich output

The Bonacich output differs slightly with the traditional degree centrality measure. The original degree centrality looked at a network as “he who has the most connections wins” or the most connected nodes as being the most powerful. Bonacich degree centrality factors in the centrality of the nodes that a particular node is connected to and avoids the possible misinterpretation of the network. For example, node *A* may have the most connections but node *B* is tied to nodes *C* and *D* who boast high degree centralities of their own. Basically, this measure takes into account the connections of your connections. From Figure 8, we can see that the Chechnya’s Urus-Martanovskiy Jamaat has the highest measure given that it is connected to other Chechen jamaats that have been very active. Although the Dagestan Vilaiyat’s

BONACICH POWER			
		Power	Normal
		-----	-----
2	Urus-Martanovskiy Jamaat	15	2.165
1	Achkhoy-Martanovskiy Jamaat	13	1.876
9	Shali Sector Jammata	10	1.443
5	Asam-yurt Village Jamaat	8	1.155
11	Tevzan Village Jamaat	8	1.155
20	Nazranovskiy Jamaat	8	1.155
21	Khalifat Jamaat	8	1.155
22	Jamaat Sidik	8	1.155
6	Vedenskiy District	7	1.01
12	Khatunni village Jamaat	7	1.01
23	Shamikala-Kaspiisk Jamaat	7	1.01
24	Jamaat Tavakkul	7	1.01
4	Nozhai-Yurtskiy Jamaat	6	0.866
18	Sunzha Ingush Jamaat	6	0.866
19	Karabulak Jamaat	6	0.866
7	Kurchaloyevskiy District	5	0.722
3	Subunit Imam	4	0.577
10	Shatoi Sector Jamaat	3	0.433
13	Grozny Jamaat	3	0.433
14	Mujahideen unit Grozny	3	0.433

8	Argun Jamaat	2	0.289
16	Shelkovskiy Jamaat	2	0.289
25	Derbent Jamaat	2	0.289
15	Naurskiy Jamaat	1	0.144
17	Sunzha Chechen Jamaat	1	0.144

Shamikala-Kaspisk jamaat and Jamaat Tavaakul are both extremely active jamaats, they are geospatially isolated in that particular area of Dagestan according to the data so far.

Betweenness Centrality

Finally, betweenness centrality was run. Betweenness centrality is used to identify nodes that lie on paths to other nodes. For example, if I were to apply for a job, the first person I would talk to would ostensibly be someone who works in human resources (if I wasn't directed to apply online first). The person in HR would lie on a path between myself and the individual hiring me for the position. Therefore, the person in HR has a betweenness of higher value, a power given to them simply because of their position in the network because they can control the flow of information. In regards to a 2-mode network that has been conflated and binarized, the centrality is as follows:

$$Actor\ x = \frac{C_B(x)}{\frac{1}{2} [m^2(s+1)^2 + m(s+1)(2t-s-1) - t(2s-t+3)]}$$

$$s = \left\lfloor \frac{(n-1)}{m} \right\rfloor, t = (n-1) \bmod m$$

Figure 9. Freeman Betweenness Centrality Output

	Freeman Betweenness Centrality	1	2
		Betweenness	nBetweenness
		-----	-----
2	Urus-Martanovskiy Jamaat	116.921	42.363

1	Achkhoy-Martanovskiy Jamaat	61.867	22.415
9	Shali Sector Jammata	30.247	10.959
11	Tevzan Village Jamaat	24.059	8.717
16	Shelkovskiy Jamaat	23	8.333
18	Sunzha Ingush Jamaat	23	8.333
23	Shamikala-Kaspiisk Jamaat	11	3.986
24	Jamaat Tavakkul	11	3.986
5	Asam-yurt Village Jamaat	10.602	3.841
3	Subunit Imam	4.844	1.755
6	Vedenskiy District	4.104	1.487
12	Khatunni village Jamaat	4.104	1.487
10	Shatoi Sector Jamaat	3.9	1.413
22	Jamaat Sidik	3.3	1.196
20	Nazranovskiy Jamaat	3.3	1.196
21	Khalifat Jamaat	3.3	1.196
7	Kurchaloyevskiy District	3.286	1.19
13	Grozny Jamaat	1.733	0.628
14	Mujahideen unit Grozny	1.733	0.628
19	Karabulak Jamaat	1.2	0.435
4	Nozhai-Yurtskiy Jamaat	0.5	0.181
8	Argun Jamaat	0	0
17	Sunzha Chechen Jamaat	0	0
15	Naurskiy Jamaat	0	0
25	Derbent Jamaat	0	0

DESCRIPTIVE STATISTICS FOR EACH MEASURE			
		1	2
		Betweenness	nBetweenness
		-----	-----
1	Mean	13.88	5.029
2	Std Dev	25.069	9.083

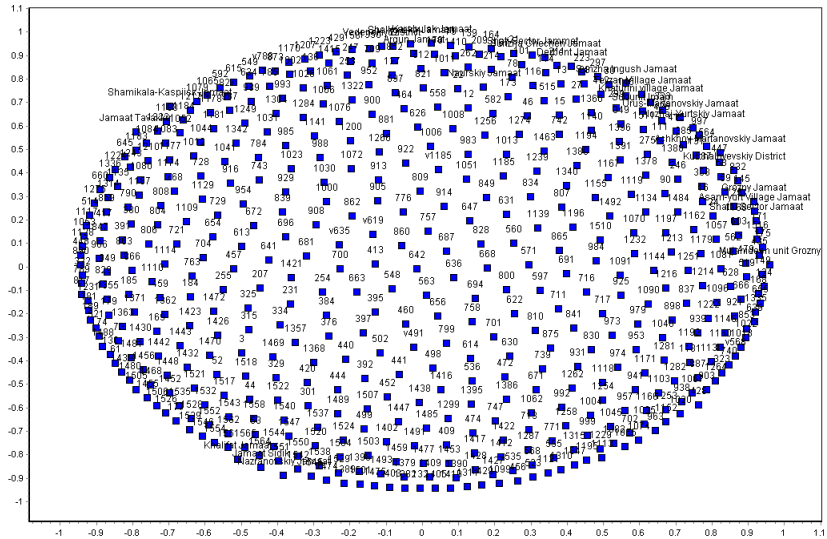
3	Sum	347	125.725
4	Variance	628.448	82.499
5	SSQ	20527.551	2694.753
6	MCSSQ	15711.19	2062.486
7	Euc Norm	143.274	51.911
8	Minimum	0	0
9	Maximum	116.921	42.363
10	N of Obs	25	25
Network Centralization Index			0.3889

With a network centralization index at 0.3889, we can concur that there is a moderate distribution of betweenness value per jamaat within the network. This shows us that the majority of jamaats do not lie on paths between other jamaats, another highlight to the localization of each jamaat. Because there are no short graph theoretic distances between the jamaats, these distances may give the jamaats antagonisms and allows specifically esoteric ways of thinking.¹⁷ In essence, because these jamaats are concerned with local matters, their thinking may not be synchronized with CE leadership even if they all adhere to the same Islamist-Jihadist theo-ideology. Immediacy trumps ideology in this picture and this should be taken into account when analyzing the violence that is attributed to the CE network.

Figure 10. Metric Multi Dimensional Scaling

Figure 10 displays a metric multidimensional scaling visualization of similarities between groups

¹⁷ Carrington p.71.



Metric MDS coordinates (stress = 0.285)

	1	2
1 Achkhoy-Ma	0.049	1.52
2 Urus-Martan	0.049	1.286
3 Subunit Imar	0.854	2.803
4 Nozhai-Yurts	1.45	0.892
5 Asam-yurt Vi	0.992	1.121
6 Vedenskiy Di	0.847	0.697
7 Kurchaloyevs	1.667	1.663
8 Argun Jamaa	1.889	2.143
9 Shali Sector J	0.995	1.68
10 Shatoi Sector	1.284	2.762
11 Tevzan Villag	1.348	1.32
12 Khatunni villi	1.05	0.684
13 Grozny Jama	0.346	3.092
14 Mujahideen	0.564	3.098
15 Naurskiy Jam	-0.45	3.225
16 Shelkovskiy J	-0.593	2.985
17 Sunzha Chec	-1.316	2.528
18 Sunzha Ingus	-1.099	1.894
19 Karabulak Ja	-0.774	1.828
20 Nazranovskiy	-0.729	1.423
21 Khalifat Jama	-0.667	1.191
22 Jamaat Sidik	-0.91	1.193
23 Shamikala-Ka	-0.395	0.604
24 Jamaat Taval	-0.625	0.63
25 Derbent Jam	-0.151	-0.039

and events. Metric multidimensional scaling is used to determine similarity and dissimilarity between data. The points are arranged so that distances between points resemble relations that are similar. So, long distances resemble dissimilarities and short distances resemble similarities. From the data above, Achkoy-Martanovsky and Urus-Martanovsky have exact MDS coordinates resembling high similarity. Since these

two groups operate in heavily overlapped areas of operation, this is logical. The Chechen-Sunzha jamaat differs from the Ingush-Sunzha jamaat slightly.

Conclusions and Considerations

The Caucasus Emirate is a dark network and one of the darkest in nature. Relational data is extremely hard to come by through open-source methods. The few individuals who have escaped the Northern Caucasus with information about the CE have done so at the risk of their lives. Given the analyses that have been undertaken in this paper, the hope that additional information will surface and allow researchers to perform in-depth critiques is great. One thing that can be garnered from this paper is the need for alternative approaches to disrupting and destabilizing the CE. Decentralization of the network may possibly play the largest role in how it should be approached. Leaders at central CE leadership, vilaiyat level, sector level, and jamaat level live only a couple of years at most on average yet this network continues to flourish and violence has increased annually since 2007 (excluding 2010 to 2011).

With the Sochi Olympics rapidly approaching in 2014, not only the Russians but also the global community should be concerned with the amount of violence occurring in the area. If serious advances are not made in regards to quelling violence and stabilizing the republics, the CE leadership will without a doubt have a much easier time in recruiting talent, operating in destabilized areas as well as exporting violence westward and into greater Russia. Although this paper has focused on the localization of violence, it should be noted that this violence is easily exported. A successful mass-casualty attack would augment the global-jihad movement and present a new shift in capabilities. The second and third order effects are difficult to establish but it would be easy to surmise that the CE as well as allied organizations such as Al Qa`ida, the Taliban, the Islamic Movement of Uzbekistan, and Islamic Jihad Union, among others, would be invigorated. Since there were 28 jamaats available for this study, there are potentially at least 28 vectors for the exportation of extremist violence, and this does not include the possible

permutations created through the collaboration of jamaats and specialized Caucasus Emirate's units like the Riyadus-Salkhiin Martyrs Brigade of suicide bombers and their handlers.

Bibliography

- Vladimir Batagelj and Andrej Mrvar. "Pajek: Program for Analysis and Visualization of Large Networks Reference Manual 2.05," Ljubljana, 24 September, 2011.
- Blanch, Lesley. *The Sabres of Paradise*. 1960.
- Borgatti, S.P., Everett, M.G. and Freeman, L.C. 2002. *Ucinet for Windows: Software for Social Network Analysis*. Harvard, MA: Analytic Technologies.
- Borgatti, Everett, Freeman. "UCINET 6 for Windows: Software for Social Network Analysis User's Guide," 2002.
- Carrington, Scott, Wasserman, ed. *Models and Methods in Social Network Analysis*. Cambridge University Press, 2005.
- Carley, Kathleen M. Copyright © 2001-2012. *Organizational Risk Assessment: Software for Social Network Analysis*. Carnegie Mellon University, Institute for Software Research International.
- DeLong-Bas, Natana J. *Wahhabi Islam: From Revival and Reform to Global Jihad*, Oxford University Press, 2004.
- Gammer, Moshe in David Cook Ed. *Jihad and Martyrdom: Critical Concepts in Islamic Studies*, Vol.III, 2010.
- Griffin, Nicholas. *Caucasus: A Journey to the Land Between Christianity and Islam*, 2001.
- Hahn, Gordon. *Russia's Islamic Threat*, Yale University Press, 2007.
- Hahn, Gordon. Islam, Islamism and Politics in Eurasia Reports 1-56. Available at: <http://csis.org/node/33013/publication>.
- Hunter, Shireen T.. *Islam in Russia: The Politics of Identity and Security*, The Center for Strategic and International Studies, 2004.
- Jane's Insurgency and Terrorism Database. Accessible at: <http://home.janes.com/info/jtic/>.
- Kemper, Micheal in David Cook Ed. *Jihad and Martyrdom: Critical Concepts in Islamic Studies*, Vol.III, 2010.
- Kemper, Michael. *The Changing Images of Jihad Leaders, Shamil and Abd al-Qadir in Dagestani and Algerian Historical writing*, Nova Religio: The Journal of Alternative and Emergent Religions Vol.11, no.2, 2007, pp. 28-58.
- Kisriev, Enver. "Islam v. obshchestvenno-politicheskoi zhiznii Dagestana," *Musulmane Izmenaushcheisya Rossii*, pp. 261-277, p.264.
- Kuromiya, Hiroaki and Georges Mamoulia. "Anti-Russian and Anti-Soviet Subversion: The Caucasian-Japanese Nexus, 1904-1945," *Europe-Asia Studies* Vol. 61, No.8, 2009.
- Leitzinger, Antero. *Caucasus and an Unholy Alliance*, 1997.
- Leahy, Kevin Daniel. "From Racketeer to Amir: A Political Portrait of Doku Umarov, Russia's Most Wanted Man," *Caucasian Review of International Affairs*, Vol.4, No.3, 2010. Accessible at http://cria-online.org/12_4.html.
- Lieven, Anatol. *Chechnya: Tombstone of Russian Power*. Yale University Press, 1998.
- Politkovskaya, Anna. *A small corner of hell: Dispatches from Chechnya*. Univerisity of Chicago, 2003.

Rasulov, Yasin. "Jihad in the Northern Caucasus: Supporters and Opponents," [www.kavkazcenter.com](http://www.kavkazcenter.com/russ/islam/jihad_in_ncaucasus/), http://www.kavkazcenter.com/russ/islam/jihad_in_ncaucasus/.

Schaefer, Robert W. *The Insurgency in Chechnya and the North Caucasus: From Gazavat to Jihad*, Praeger Security International, 2011.

Shultz Jr, Richard H., Andrea J. Dew. *Insurgents, Terrorists, and Militias: The Warriors of Contemporary Combat*, Columbia University Press, 2007.

Smith, Sebastian. *Allah's Mountains: Politics and War in the Russian Caucasus*, 1998.

Umarov, Doku. "The official version of Amir Dokka's statement of declaration of the Caucasus Emirate," 22 November 2007, <http://www.kavkazcenter.com/eng/content/2007/11/22/9107.shtml>.

Velichko V.L. *Caucasus: Russian Affairs and Intertribal Problems*, Baku, 2007.

Vatchagaev, Mairbeg. "Karbadino-Balkaria's "Black Hawks": Grassroots vigilantes or FSB Surrogates?", March 11th 2011.

Ware, Robert Bruce and Enver Kisriev. *Dagestan: Russian Hegemony and Islamic Resistance in the North Caucasus*, 2010.

Wasserman, Stanley; Faust, Katherine. "Social Network Analysis in the Social and Behavioral Sciences" *Social Network Analysis: Methods and Applications*. Cambridge University Press, pp. 1-27.

Wilhelmsen, Julie. "Why Separatists Become Islamists," FFI Rapport, 2004.

Appendix A: Distribution of Ethnicities in Dagestan Provinces

Republic	Dagestan (2002.)														
	All	Avars a	Dargin. 2	Kumyk.	Lezgins.	Laks	Tabas .	Aguls	Rutul .	Tsakhur .	Nogay .	Russian	Chechen s .	Azeri	e.g.t. 3
Dagestan	2,576,531	758,438	425,526	365,804	336,698	139,732	110,152	23,314	24,298	8,168	38,168	120,875	87,867	111,656	3,369
	-100%	-29.40%	-16.50%	-14.20%	-13.00%	-5.40%	-4.30%	-0.90%	-0.90%	-0.30%	-1.50%	-4.80%	-3.40%	-4.30%	-0.10%
Makhachkala	462,412	122,765	63,493	64,124	62,810	67,323	10,285	3,161	5,595	1,711	1,45	42,125	1,287	6,475	902
	-100%	-26.50%	-13.70%	-13.90%	-13.60%	-14.60%	-2.20%	-0.70%	-1.20%	-0.40%	-0.30%	-9.10%	-0.30%	-1.40%	-0.20%
Mr. Buinaksk	61,437	28,289	3,791	19,304	475	4,219	136	9	4	6	47	3,71	114	403	91
	-100%	-46.00%	-6.20%	-31.40%	-0.80%	-6.90%	-0.20%	-0.10%	-0.10%	-0.10%	-0.10%	-6.00%	-0.20%	-0.70%	-0.10%
Dagestan, the Lights	26,346	30	2,197	203	5,860	47	9,359	922	86	15	3	505	3	6,917	15
	-100%	-0.10%	-8.30%	-0.80%	-22.20%	-0.20%	-35.50%	-3.50%	-0.30%	-0.10%	-0.10%	-1.90%	-0.10%	-26.30%	-0.10%
Derbent	101,031	442	5,582	552	32,955	436	15,606	2,956	716	78	28	5,073	17	32,064	2,038
	-100%	-0.40%	-5.50%	-0.50%	-32.60%	-0.40%	-15.40%	-2.90%	-0.70%	-0.10%	-0.10%	-5.00%	-0.10%	-31.70%	-2.00%
Mr. Kaspiysk	77,650	10,206	16,040	7,695	13,952	11,737	4,020	880	630	95	43	10,228	64	676	45
	-100%	-13.10%	-20.70%	-9.90%	-18.00%	-15.10%	-5.20%	-1.10%	-0.80%	-0.10%	-0.10%	-13.20%	-0.10%	-0.90%	-0.10%
Khasavyurt	121,817	34,323	5,037	33,104	2,554	4,161	33	8	15	-	165	3,965	36,911	765	90
	-100%	-28.20%	-4.10%	-27.20%	-2.10%	-3.40%	-0.10%	-0.10%	-0.10%	-	-0.10%	-3.30%	-30.30%	-0.60%	-0.10%
Mr. South Sukhums	9,777	4,683	1,812	371	1,046	877	46	74	35	32	58	531	34	52	19
	-100%	-47.90%	-18.50%	-3.80%	-10.70%	-9.00%	-0.50%	-0.80%	-0.40%	-0.30%	-0.60%	-5.40%	-0.30%	-0.50%	-0.20%
Makhachkala	82,846	23,556	12,398	32,492	2,823	3,664	839	496	379	387	3,698	1,519	56	275	6
	-100%	-28.40%	-15.00%	-39.20%	-3.40%	-4.40%	-1.00%	-0.60%	-0.50%	-0.50%	-4.50%	-1.80%	-0.10%	-0.30%	-0.10%
Izberbashskaya mountai ns.	40,987	1,128	26,794	5,889	2,973	834	288	135	14	2	23	2,109	14	395	23
	-100%	-2.80%	-65.40%	-14.40%	-7.30%	-2.00%	-0.70%	-0.30%	-0.10%	-0.10%	-0.10%	-5.10%	-0.10%	-1.00%	-0.10%
Kizilyurt	47,679	33,499	1,108	6,056	1,089	2,569	158	22	105	9	14	2,002	455	94	11
	-100%	-70.30%	-2.30%	-12.70%	-2.30%	-5.40%	-0.30%	-0.10%	-0.20%	-0.10%	-0.10%	-4.20%	-1.00%	-0.20%	-0.10%
Mountai ns . and dministratsiya	51,024	7,956	5,899	2,429	1,986	1,632	649	53	615	79	702	25,111	562	892	36
	-100%	-15.60%	-11.60%	-4.80%	-3.90%	-3.20%	-1.30%	-0.10%	-1.20%	-0.20%	-1.40%	-49.20%	-1.10%	-1.70%	-0.10%
Agul	11,290	7	797	2	5	A	16	10,447	-	4	-	7	-	3	-

District	-100%	-0.10%	-7.10%	-0.10%	-0.10%	-0.10%	-0.10%	-92.50%		-0.10%		-0.10%		-0.10%	
Akushinsky District	52,455	16	50,344	9	3	2,046	A	2	-	-	A	23	-	-	-
	-100%	-0.10%	-96.00%	-0.10%	-0.10%	-3.90%	-0.10%	-0.10%			-0.10%	-0.10%			
Ahvashky District	20,373	20,329	2	A	-	A	-	-	-	-	A	4	28	-	-
	-100%	-99.80%	-0.10%	-0.10%		-0.10%					-0.10%	-0.10%	-0.10%		
Ahtynsky District	31,592	13	A	3	30,956	2	18	A	6	-	-	517	A	8	-
	-100%	-0.10%	-0.10%	-0.10%	-98.00%	-0.10%	-0.10%	-0.10%	-0.10%			-1.60%	-0.10%	-0.10%	
Babayurtovsky District	41,331	7,979	2,563	19,753	68	352	10	12	16	21	7,122	366	2,755	24	-
	-100%	-19.30%	-6.20%	-47.80%	-0.20%	-0.90%	-0.10%	-0.10%	-0.10%	-0.10%	-17.20%	-0.90%	-6.70%	-0.10%	
Botlikh district	50,469	50,361	2	9	6	7	-	-	-	-	A	46	16	3	-
	-100%	-99.80%	-0.10%	-0.10%	-0.10%	-0.10%					-0.10%	-0.10%	-0.10%	-0.10%	
Buinaksk district	65,018	14,883	8,996	40,759	20	167	3	A	-	-	3	145	11	10	A
	-100%	-22.90%	-13.80%	-62.70%	-0.10%	-0.30%	-0.10%	-0.10%			-0.10%	-0.20%	-0.10%	-0.10%	-0.10%
Gergebilsky District	18,366	18,265	33	8	8	13	7	-	-	-	A	16	A	A	-
	-100%	-99.50%	-0.20%	-0.10%	-0.10%	-0.10%	-0.10%				-0.10%	-0.10%	-0.10%	-0.10%	
Gumbetovskiy District	18,177	18,123	6	3	2	4	-	-	-	-	-	7	26	-	-
	-100%	-99.70%	-0.10%	-0.10%	-0.10%	-0.10%						-0.10%	-0.10%		
Gunibsky District	25,106	24,155	784	48	7	27	9	-	-	2	6	41	A	13	-
	-100%	-96.20%	-3.10%	-0.20%	-0.10%	-0.10%	-0.10%			-0.10%	-0.10%	-0.20%	-0.10%	-0.10%	
Dahadavskiy District	38,359	5	38,290	4	3	7	A	-	-	-	2	22	A	4	-
	-100%	-0.10%	-99.80%	-0.10%	-0.10%	-0.10%	-0.10%				-0.10%	-0.10%	-0.10%	-0.10%	
Derbent District	86,494	74	6,828	384	16,278	272	9,229	1,848	365	2	5	651	4	50,247	51
	-100%	-0.10%	-7.90%	-0.40%	-18.80%	-0.30%	-10.70%	-2.10%	-0.40%	-0.10%	-0.10%	-0.80%	-0.10%	-58.10%	-0.10%
Dokuzparrinsky District	14,330	3	3	4	13,399	-	12	233	387	-	-	230	-	22	-
	-100%	-0.10%	-0.10%	-0.10%	-93.50%		-0.10%	-1.60%	-2.70%			-1.60%		-0.20%	
Kazbek district	33,140	29,164	15	32	2	18	-	-	7	3	4	36	3,834	A	-
	-100%	-88.00%	-0.10%	-0.10%	-0.10%	-0.10%			-0.10%	-0.10%	-0.10%	-0.10%	-11.60%	-0.10%	
Kaitag District	26,870	24	24,277	2,39	7	16	41	5	2	-	A	50	A	23	2
	-100%	-0.10%	-90.30%	-8.90%	-0.10%	-0.10%	-0.20%	-0.10%	-0.10%		-0.10%	-0.20%	-0.10%	-0.10%	-0.10%
Karabudakhkent District	60,620	379	20,405	38,839	77	669	13	8	6	-	-	111	9	38	-
	-100%	-0.60%	-33.70%	-64.10%	-0.10%	-1.10%	-0.10%	-0.10%	-0.10%			-0.20%	-0.10%	-0.10%	
Kavakent District	52,739	81	21,614	28,723	251	112	883	846	A	A	3	139	2	57	-
	-100%	-0.20%	-41.00%	-54.50%	-0.50%	-0.20%	-1.70%	-1.60%	-0.10%	-0.10%	-0.10%	-0.30%	-0.10%	-0.10%	
Kizilyurt District	70,440	58,101	431	8,103	99	1,242	9	11	4	-	7	245	1,894	36	-
	-100%	-82.50%	-0.60%	-11.50%	-0.10%	-1.80%	-0.10%	-0.10%	-0.10%		-0.10%	-0.30%	-2.70%	-0.10%	
Kizlyar district	57,748	22,907	10,091	868	2,142	2,052	813	159	862	535	3,754	10,977	516	1,506	2
	-100%	-39.70%	-17.50%	-1.50%	-3.70%	-3.60%	-1.40%	-0.30%	-1.50%	-0.90%	-6.50%	-19.00%	-0.90%	-2.60%	-0.10%
Kulinsky district	10,760	14	55	2	2	10,669	2	-	-	-	-	14	-	-	-
	-100%	-0.10%	-0.50%	-0.10%	-0.10%	-99.20%	-0.10%					-0.10%			
Kumtorkalinsky District	21,053	5,128	1,868	12,576	310	227	163	216	31	123	32	150	69	83	4
	-100%	-24.40%	-8.90%	-59.70%	-1.50%	-1.10%	-0.80%	-1.00%	-0.10%	-0.60%	-0.20%	-0.70%	-0.30%	-0.40%	-0.10%

Kurakh District	15,206	-	4	2	14,992	A	33	21	3	-	-	20	-	124	-
	-100%		-0.10%	-0.10%	-98.60%	-0.10%	-0.20%	-0.10%	-0.10%			-0.10%		-0.80%	
Lak area	12,382	134	275	30	13	11,870	3	16	-	-	A	14	12	4	A
	-100%	-1.10%	-2.20%	-0.20%	-0.10%	-95.90%	-0.10%	-0.10%			-0.10%	-0.10%	-0.10%	-0.10%	-0.10%
Levashin District	64,371	14,737	49,431	33	11	65	4	-	-	-	A	51	6	8	-
	-100%	-22.90%	-76.80%	-0.10%	-0.10%	-0.10%	-0.10%				-0.10%	-0.10%	-0.10%	-0.10%	
Magaramkent sky District	58,694	21	21	16	56,163	22	46	248	370	A	A	982	4	539	22
	-100%	-0.10%	-0.10%	-0.10%	-95.70%	-0.10%	-0.10%	-0.40%	-0.60%	-0.10%	-0.10%	-1.70%	-0.10%	-0.90%	-0.10%
Novolaksy district	22,019	5,229	122	52	24	10,132	5	A	11	-	3	40	6,352	7	-
	-100%	-23.70%	-0.60%	-0.20%	-0.10%	-46.00%	-0.10%	-0.10%	-0.10%		-0.10%	-0.20%	-28.80%	-0.10%	
Nogai District	21,685	61	1,777	233	10	46	-	-	-	-	18,587	256	334	12	-
	-100%	-0.30%	-8.20%	-1.10%	-0.10%	-0.20%					-85.70%	-1.20%	-1.50%	-0.10%	
Rutul District	23,503	557	14	A	2,116	925	11	A	13,975	4,971	2	185	A	603	-
	-100%	-2.40%	-0.10%	-0.10%	-9.00%	-3.90%	-0.10%	-0.10%	-59.50%	-21.20%	-0.10%	-0.80%	-0.10%	-2.60%	
Sergokalinsky District	29,665	25	29,475	19	9	14	10	A	2	-	-	79	-	12	A
	-100%	-0.10%	-99.40%	-0.10%	-0.10%	-0.10%	-0.10%	-0.10%	-0.10%			-0.30%		-0.10%	-0.10%
Suleiman Stalskiy District	54,036	4	9	6	53,824	7	34	8	2	-	-	100	3	8	-
	-100%	-0.10%	-0.10%	-0.10%	-99.60%	-0.10%	-0.10%	-0.10%	-0.10%			-0.20%	-0.10%	-0.10%	
Tabasara District	54,732	7	72	6	60	23	45,224	328	2	-	A	53	2	8,896	-
	-100%	-0.10%	-0.10%	-0.10%	-0.10%	-0.10%	-82.60%	-0.60%	-0.10%		-0.10%	-0.10%	-0.10%	-16.30%	
Tarumov sky District	28,587	8,919	6,165	354	1,066	539	211	136	41	85	2,366	7,174	159	266	2
	-100%	-31.20%	-21.60%	-1.20%	-3.70%	-1.90%	-0.70%	-0.50%	-0.10%	-0.30%	-8.30%	-25.10%	-0.60%	-0.90%	-0.10%
Tlyaratinsky District	22,108	21,903	7	7	6	-	2	A	A	3	-	146	2	9	-
	-100%	-99.10%	-0.10%	-0.10%	-0.10%		-0.10%	-0.10%	-0.10%	-0.10%		-0.70%	-0.10%	-0.10%	
Untsukul sky District	27,460	26,649	179	194	76	154	23	-	A	-	A	125	10	18	-
	-100%	-97.00%	-0.70%	-0.70%	-0.30%	-0.60%	-0.10%		-0.10%		-0.10%	-0.50%	-0.10%	-0.10%	
Khasavurt district	125,454	38,282	6,379	40,068	7,414	348	14	A	5	-	26	298	32,277	42	6
	-100%	-30.50%	-5.10%	-31.90%	-5.90%	-0.30%	-0.10%	-0.10%	-0.10%		-0.10%	-0.20%	-25.70%	-0.10%	-0.10%
Hivsky District	20,747	4	5	2	8,725	16	11,878	47	A	-	-	23	A	17	-
	-100%	-0.10%	-0.10%	-0.10%	-42.10%	-0.10%	-57.30%	-0.20%	-0.10%			-0.10%	-0.10%	-0.10%	
Hunzah District	30,203	29,773	7	14	7	4	2	-	-	-	-	302	8	2	-
	-100%	-98.60%	-0.10%	-0.10%	-0.10%	-0.10%	-0.10%					-1.00%	-0.10%	-0.10%	
Tsumadinsky District	20,632	20,453	2	6	2	-	-	-	3	2	A	116	8	-	A
	-100%	-99.10%	-0.10%	-0.10%	-0.10%				-0.10%	-0.10%	-0.10%	-0.60%	-0.10%		-0.10%
Tsuntin District	17,466	17,183	16	7	10	-	2	-	-	A	A	208	-	A	-
	-100%	-98.40%	-0.10%	-0.10%	-0.10%		-0.10%			-0.10%	-0.10%	-1.20%		-0.10%	
Charodinsky District	11,792	11,601	5	2	-	162	A	-	-	-	A	12	2	-	-
	-100%	-98.40%	-0.10%	-0.10%		-1.40%	-0.10%				-0.10%	-0.10%	-0.10%		
Shamilsky District	26,053	26,008	5	13	2	A	-	-	-	-	2	16	-	A	-
	-100%	-99.80%	-0.10%	-0.10%	-0.10%	-0.10%					-0.10%	-0.10%		-0.10%	

A. with Ando-tsez people and archintsami; proper distribution of the Avars, Ando-Tsez archintsev peoples of Dagestan on the ATE, see the link						
Two. with Kubachins and kaytagtsami ; proper distribution Dargins, Kubachins and kaytagtsev for ATE Dagestan, see the link						

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## **ANNOUNCEMENTS**

### **GORDON M. HAHN'S RECENT PUBLICATIONS AND INTERVIEWS**

Gordon M. Hahn, "The Caucasus Emirate Jihadists: The Security and Strategic Implications," in Stephen J. Blank, ed., *Russia's Homegrown Insurgency: Jihad in the North Caucasus* (Carlisle Barracks, PA: U.S. Army War College Strategic Studies Institute, October 2012), pp. 1-98, [www.strategicstudiesinstitute.army.mil/pdf/PUB1116.pdf](http://www.strategicstudiesinstitute.army.mil/pdf/PUB1116.pdf).

Gordon M. Hahn, "More Russian Election Intrigue," *Russia – Other Points of View*, 24 October 2012, [www.russiaotherpointsofview.com/2012/10/more-russian-election-intrigue.html](http://www.russiaotherpointsofview.com/2012/10/more-russian-election-intrigue.html).

Gordon M. Hahn, "Tandemology 2.0: Putin's Return, Medvedev's Decline, and Tandem Spats," *Russia – Other Points of View*, 7 October 2012, [www.russiaotherpointsofview.com/2012/10/tandemology-20-putins-return-medvedevs-decline-and-tandem-spats.html](http://www.russiaotherpointsofview.com/2012/10/tandemology-20-putins-return-medvedevs-decline-and-tandem-spats.html).

Gordon M. Hahn, "Response to Estonian Defence Minister's Erroneous Article: Problems of Expansionism: Russia, NATO, and Georgia," *Russia – Other Points of View*, 29 September 2012, [www.russiaotherpointsofview.com/2012/09/response-to-estonian-defense-ministers-erroneous-article.html](http://www.russiaotherpointsofview.com/2012/09/response-to-estonian-defense-ministers-erroneous-article.html).

Gordon M. Hahn, "Make It Stop: Another MSM Caricature of Russia" (Critique of David Satter in *The National Review*), *Russia – Other Points of View*, 18 September 2012, [www.russiaotherpointsofview.com/2012/09/make-it-stop-another-msm-caricature-of-russia.html](http://www.russiaotherpointsofview.com/2012/09/make-it-stop-another-msm-caricature-of-russia.html).

Interview excerpts in *Defence Report* article "Russia's homegrown North Caucasus insurgency pushes for Islamist state" by Marion Ghibaudo. See *Defence Report*, 20 September 2012, <http://defencereport.com/russias-homegrown-north-caucasus-insurgency-pushes-for-islamist-state/>.



#### **ABOUT IIPER**

Islam, Islamism and Politics in Eurasia Report (IIPER) is a project of the Russia and Eurasia Program at the Center for Strategic and International Studies. It focuses on all politically-relevant issues involving or bearing on Islam, Islamism, and Jihadism in Russia and Eurasia writ large. All issues of IIPER will soon be permanently archived at <http://csis.org/program/russia-and-eurasia-program>. All back issues temporarily remain archived at: [www.mii.edu/academics/faculty/ghahn/report](http://www.mii.edu/academics/faculty/ghahn/report).

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IIPER welcomes submissions on any aspect of Islamic, Islamist, or Jihadist politics in Eurasia as well as financial contributions to support the project. For related inquiries or to request to be included on IIPER's mailing list, please contact:

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