Real-World Developments Predict Immigration News in Right-Wing Media: Evidence from Germany

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Appendix

Appendix A: Number of topics

There is no objective answer to the question of how many topics the model should identify. I ran models ranging from 10 to 50 topics and plotted semantic coherence and exclusivity against each other. Semantic coherence is maximized when the most probable words in a given topic frequently co-occur together and "correlates well with human judgment of topic quality" (Roberts et al., 2019, p. 11). However, semantic coherence is generally high in cases of a few topics with many very common words. Examining semantic coherence and exclusivity in tandem counters this problem, since high exclusivity means a topic's most probable terms are unlikely to co-occur with the top words of other topics (Roberts et al., 2019, p. 12). Results show that models between 20 and 30 topics tend to maximize both, see Figure A2. A qualitative evaluation suggested that models with fewer than 30 topics lead to an amalgamation of relevant topics, while more than 30 topics did not lead to additional relevant topics. Thus, I chose to estimate 30 topics. I also examined coherence and exclusivity within the topics of the final model. Figure A3 suggests that the key topics score relatively well in this respect.

Reference

Roberts, M. E., Stewart, B. M., & Tingley, D. (2019). stm: An R Package for Structural Topic Models. *Journal of Statistical Software*, 91(2), 1–40.

https://doi.org/10.18637/jss.v091.i02

Figure A1: Correlations between predictors

Immigration rate	0.74	0.2	0.02
0.74	Foreigner crime rate	0.3	0.08
0.2	0.3	Terror attack	0.24
0.02	0.08	0.24	Terror attack (DE)

Figure A2: Semantic coherence vs. exclusivity for models with varying number of topics

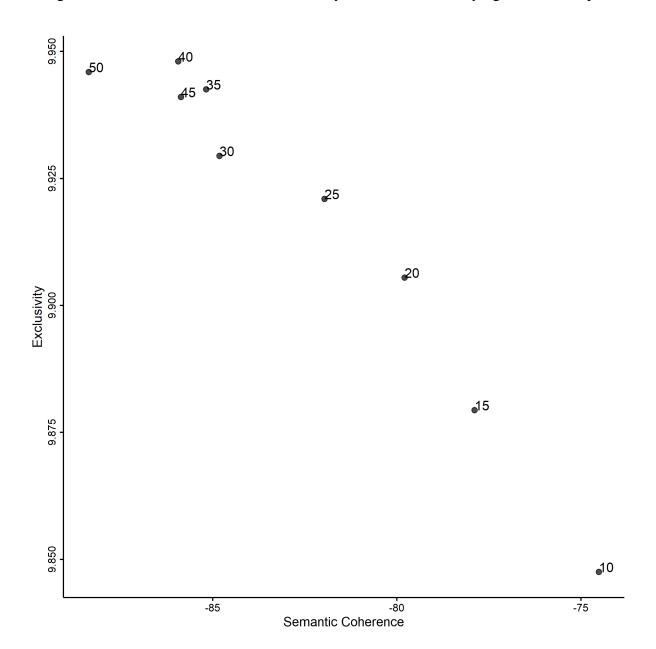


Figure A3: Semantic coherence vs. exclusivity per topic

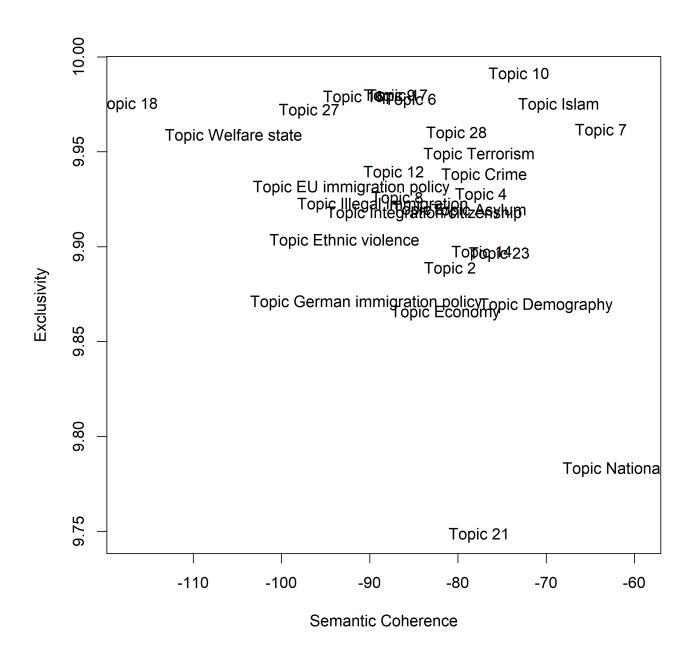


Table A1: Hierarchical linear models predicting immigration news

	Immigration		tion	Crime	te	error	Terror (DE)	
Predictors	Estimates	р	Estimates	p	Estimates	p	Estimates	p
intercept	-13.19 (-14.41 - -11.97)	<0.001	-4.97 (-6.42 – -3.52)	<0.001	-14.43 (-15.74 – -13.13)	<0.001	-15.21 (-16.49 - -13.93)	<0.001
Immigrati on rate	0.13 (0.11 – 0.14)	<0.001						
Time trend	0.01 (0.01 – 0.01)	<0.001	0.00 (0.00 – 0.00)	<0.001	$0.01 \\ (0.01 - \\ 0.01)$	<0.001	0.01 (0.01 – 0.01)	<0.001
Foreigner crime rate			0.15 (0.14 – 0.17)	<0.001				
terror attack					0.04 (0.03 – 0.06)	<0.001		
Terror attack (DE)							0.08 (0.02 – 0.14)	0.008
σ^2	(0.09	0.0	9	0.09		0.09	
τ 00	(0.00 date2	0.0	0 date2	$0.00 \; \mathrm{da}$	te2	$0.00_{ m date2}$	
CCI	(0.02	0.0	1	0.02		0.02	
NOT	-	1053 date2	105	53 date2	1053 d	ate2	1053 date2	2
Observatio	ns :	53818	538	318	53818		53818	

Table A2: Logistic multi-level models predicting immigration news

		Immigra (log)	tion	Crime (log	g) Te	rror (log)		or (DE) log)
Predictors	Odds Ratios	p	Odds Ratios	p	Odds Ratios	p	Odds Ratios	p
Intercept	0.00 (0.00 – 0.00)	<0.001	0.00 (0.00 – 0.00)	<0.001	0.00 (0.00 – 0.00)	<0.001	0.00 (0.00 – 0.00)	<0.001
Immigrati on rate	2.62 (2.23 – 3.07)	<0.001						
Time trend	1.08 (1.08 – 1.08)	<0.001	1.03 (1.03 – 1.03)	<0.001	1.09 (1.09 – 1.09)	<0.001	1.09 (1.09 – 1.09)	<0.001
Foreigner crime rate			3.76 (3.39 – 4.16)	<0.001				
Terror attack					1.38 (1.20 – 1.59)	<0.001		
Terror attack (DE)							1.83 (1.07 – 3.12)	0.027
σ^2		3.29	3	.29	3.29		3.29	
$ au_{00}$		0.15 date2	0	.09 date2	0.19	date2	0.19 _{dat}	e2
ICC		0.04	0	.03	0.05		0.05	
N		1053 date2	1	053 date2	1053	date2	1053 da	te2
Observation	ns	53818	5	3818	5381	8	53818	

Table A3: Hierarchical linear models with random slopes

	(Immigra (random s		Crime andom slo	pe)		rror m slope)	Terror (random	` /
Predictors	Estimates	s p	Estimates	p	Es	timates	p	Estimates	p
Intercept	-12.23 (-13.38 – -11.07)	<0.001	-5.13 (-6.44 – -3.81)	<0.001	(-	14.32 15.61 – 13.04)	<0.001	-15.23 (-16.51 – -13.95)	<0.001
Immigrat ion rate	0.19 (0.16 – 0.21)	<0.001							
Time trend	$0.01 \\ (0.01 - \\ 0.01)$	<0.001	0.00 (0.00 – 0.00)	<0.001	(0.01 0.01 – 0.01)	<0.001	$0.01 \\ (0.01 - \\ 0.01)$	<0.001
Foreigne r crime rate			0.13 (0.12 – 0.15)	<0.001					
Terror attack					(0.04 0.02 – 0.06)	<0.001		
Terror attack (DE)								0.08 (0.04 – 0.12)	<0.001
Random E	ffects								
σ^2	(0.09	0.0)9		0.09		0.09	
$ au_{00}$	(0.00 date2	0.0	00 date2		0.00 date	2	$0.00_{ m date2}$	
τ_{11}		0.02 date2.im	migratio 0.0)1 _{date2.auslkr}	rimpr	0.00 dates	2.week_follo	0.00 date2.wee	ek_following
$ ho_{01}$						0.86 date	2	-1.00 date2	
ICC						0.02		0.02	
N	1	1053 date2	10	53 date2		1053 date	e2	1053 date2	
Observatio	ns 5	53818	53	818		53818		53818	

Table A4: Hierarchical linear models without time trend

		Immigra	ition	Crime	7	Terror	Terro	r (DE)
Predictors	Estimates	p	Estimates	s p	Estimates	p	Estimates	p
Intercept	0.06 (0.06 – 0.07)	<0.001	0.05 (0.04 – 0.05)	<0.001	0.10 (0.09 – 0.10)	<0.001	0.10 (0.10 – 0.11)	<0.001
Immigrati on rate	0.17 (0.15 – 0.19)	<0.001						
Foreigner crime rate			0.18 (0.17 – 0.19)	<0.001				
Terror attack					0.09 (0.07 – 0.11)	<0.001		
Terror attack (DE)							0.13 (0.05 – 0.20)	0.001
σ^2	C	0.09	0.	09	0.09		0.09	
$ au_{00}$	C	0.00 date2	0.	00 date2	0.00	late2	0.00 date	2
ICC	C	0.03	0.	01	0.04		0.05	
N	1	053 date2	10	053 date2	1053	date2	1053 date	e2
Observatio	ns 5	33818	53	8818	53818	3	53818	

Table A5: Hierarchical linear models controlling for 2015/16

		Immigra	tion	Crime	Terror (r (DE)
Predictors	Estimates	p	Estimates	p	Estimates	p	Estimates	p
Intercept	-12.40 (-13.63– -11.16)	<0.001	-5.30 (-6.79 – -3.82)	<0.001	-11.79 (-13.06– -10.53)	<0.001	-12.29 (-13.54– -11.04)	<0.001
Immigrati on rate	0.07 $(0.04 - 0.09)$	<0.001						
Time trend	$0.01 \\ (0.01 - \\ 0.01)$	<0.001	0.00 (0.00 – 0.00)	<0.001	$0.01 \\ (0.01 - \\ 0.01)$	<0.001	$0.01 \\ (0.01 - \\ 0.01)$	<0.001
Refugee crisis	0.05 (0.03 – 0.07)	<0.001	0.01 (-0.00 – 0.03)	0.058	0.09 (0.08 – 0.10)	<0.001	0.09 (0.08 – 0.10)	<0.001
Foreigner crime rate			0.14 (0.12 – 0.16)	<0.001				
Terror attack					0.03 (0.02 – 0.05)	<0.001		
Terror attack (DE)							0.05 (-0.00– 0.11)	0.053
σ^2	0.	09	0.09	9	0.09		0.09	
$ au_{00}$	0.	00 date2	0.00	O date2	$0.00 \; \mathrm{d}$	late2	$0.00_{ m date2}$	
ICC	0.	02	0.0	1	0.02		0.02	
N	10)53 _{date2}	105	3 date2	1053	date2	1053 date2	
Observatio	ns 53	3818	538	18	53818	3	53818	

Table A6: Hierarchical linear models without the years 2015/16

			Immigration (w/o 2015/16)		Terror (w/o 2015/16)		Terror (DE) (w/o 2015/16)	
Predictors	Estimates	s p	Estimates	s p	Estimates	p	Estimates	p
Intercept	-12.33 (-13.41– -11.25)	<0.001	-5.32 (-6.66– -3.98)	<0.001	-11.68 (-12.86– -10.51)	<0.001	-12.24 (-13.40– -11.09)	<0.001
Immigrati on rate	0.20 (0.17 – 0.23)	<0.001						
Time trend	0.01 (0.01 – 0.01)	<0.001	0.00 (0.00 – 0.00)	<0.001	0.01 (0.01 – 0.01)	<0.001	0.01 (0.01 – 0.01)	<0.001
Foreigner crime rate			0.14 (0.12 – 0.16)	<0.001				
Terror attack					0.04 (0.02 – 0.05)	<0.001		
Terror attack (DE)							0.09 (0.02 – 0.16)	0.015
σ^2	(0.08	0.	08	0.08		0.08	
$ au_{00}$	(0.00 date2	0.	00 date2	$0.00 _{ m da}$	te2	$0.00_{ m date2}$	
ICC	(0.01	0.	01	0.01		0.01	
N		953 _{date2}	95	53 date2	953 _{dat}	e2	953 date2	
Observation	ıs 4	48097	48	3097	48097		48097	

Table A7: Full topic model, original German version

Topic	Most probable terms (stemmed)			ed)	Name
1	volk	kultur	kulturell	identitat	National identity
2	recht	demonstration	verein	linksextremist	
					German immigration
3	einwander	einwand	asyl	illegal	policy
4	schul	studi	eltern	bildung	
5	zeitung	journalist	bild	les	
6	itali	spani	italien	lega	
7	merkel	csu	angela	kanzlerin	
8	afd	mitglied	fraktion	gauland	
9	turkisch	turkei	turk	erdogan	
10	österreich	fpö	wien	osterreich	
11	integration	einwand	staatsburgerschaft	zuwander	Integration/citizenship
12	fdp	rot	sachs	hamburg	
13	islamist	anschlag	terrorist	terror	Terrorism
14	brussel	grundgesetz	parlament	kommission	
15	asylbewerb	abschiebung	behord	asyl	Asylum
16	schweiz	konservativ	liberal	schwed	
17	frankreich	franzos	grossbritanni	britisch	
18	griechenland	griechisch	pol	bundeswehr	
19	islam	muslim	religion	moslem	Islam
					EU immigration
20	migration	international	migrant	global	policy
21	stadt	haus	famili	strass	
22	jugend	schul	gewalt	migrationshintergrund	Ethnic violence
23	buch	sarrazin	pegida	leut	
24	illegal	afrikan	migrant	afrika	Illegal immigration
25	milliard	geld	unternehm	kost	Economy
26	bevolker	anteil	demograph	statist	Demography
27	niederland	wild	belgi	belang	
28	usa	us	russland	syri	
29	polizei	tat	opfer	polizist	Crime
30	hartz	iv	arbeitslos	leistung	Welfare state