14.770: Media Lecture 20-21b

Ben Olken

Outline

- Media's impact on policy.
 - Affecting politicians' effort
 - And in turn affecting outcomes
- Media and protest.

- Setup is a voting model with politician moral hazard and reputations
- Citizens are of two types:
 - Vulnerable (fraction $\gamma < \frac{1}{2}$) care about effort.
 - Non-vulnerable care about politician ideology.
- Politicians can put in effort e ∈ [0, E] to help vulnerable. Effort unobservable and costs politician E.
- Politicians are of three types with positive probability:
 - Altruistic (always performs E)
 - Selfish (always performs 0)
 - $\bullet\,$ Opportunistic. Opportunistic values re-election only with value $\Omega.$

Timing

• Period 1:

- Politician chooses effort
- Media reports on politician effort.
- Let q(e, m) be the fraction of vulnerable citizens who receive a signal that the politician exerted positive effort.
- Key conditions are that $q_{em}(e,m) > 0$ (media and effort are complements) and $q_{ee}(e,m) < 0$
- Period 2:
 - Vote to re-elect politician or re-draw politician from initial distribution.
 - Politician chooses effort
 - Game ends

Voting

- Voting:
 - Vulnerable vote optimally to maximize politician effort in period 2.
 - Vulnerable voters vote for re-election if they observe effort in first period, and not otherwise.
 - Why? Types make this voting optimal, since observing effort increases posterior probability of a politician being altruistic. The model therefore combines effort and competence.
 - What if there were only opportunistic types?
 - Non-vulnerable vote for ideological reasons.
 - Ideological vote share for incumbent is $v = b + \varepsilon$
 - ε is distributed as [-b + a, b a]. So v is distributed uniformly on [a, 2b a]. b is expected support for incumbent and a measures accuracy.
 - So incumbent wins re-election if

$$\gamma q(e, m) + (1 - \gamma) v > \frac{1}{2}$$

Solution

• Probability of re-election is

$$P = \begin{cases} 1 & \text{if} \quad \gamma q \left(e, m\right) + \left(1 - \gamma\right) a > \frac{1}{2} \\ \frac{\left(2b - a\right) + \frac{\gamma}{1 - \gamma} q\left(e, m\right) - \frac{1}{2\left(1 - \gamma\right)}}{2\left(b - a\right)} & \text{if} \quad \text{otherwise} \\ 0 & \text{if} \quad \gamma q \left(e, m\right) + \left(1 - \gamma\right) \left(2b - a\right) < \frac{1}{2} \end{cases}$$

• FOC for optimal effort, at interior, is

$$rac{\gamma}{2\left(b-a
ight)\left(1-\gamma
ight)}\Omega q_{e}=1$$

• Effort increases if:

- Media *m* increases
- Incumbent advantage *b* decreases
- Election becomes less noisy (i.e. a increases)
- $\bullet\,$ Share of population that cares about effort γ increases



Image is in the public domain.

"President Bush's job approval rating took a hit in the wake of Hurricane Katrina, dropping to a historic low of 41%, a new Zogby America poll reveals. The public rates the performance of all levels of government in the aftermath of Hurricane Katrina negatively, with 36% giving the President passing marks on his handling of the crisis" -BBC, 9/8/05

Some model notes

- How is this model different from a conventional moral hazard model?
 - Types (why?) Voters can't write contracts (except in prospective voting models). So need to motivate voters' decisions.
 - Two types of noise (why?) What would change if $\gamma = 1$?
 - $q = \frac{1}{2}$ in equilibrium just enough effort to win. Media effect less clear.)
 - Repeated game version?
- Media
 - Is media a complement to effort or a substitute?

Empirics from the US

Snyder and Stromberg (2008)

- Setting: US congressional districts
- Empirical strategy:
 - Examine overlap between newspaper markets and congressional districts
 - Idea: those districts where overlap is less clear get less media coverage, and so congressmen put in less effort
- Driving empirical idea:

$$q_{md} = \alpha ReaderShare_{md}$$

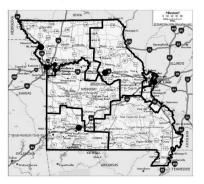
where q_{md} is quantity of articles about congressman d in media source m, and ReaderShare is the share of m's readers in district d
Define

$$Congruence_{cd} = \sum_{m=1}^{M} MarketShare_{mc}ReaderShare_{md}$$

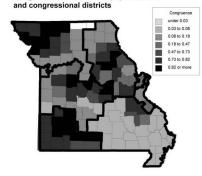
where *c* is a county.

• Idea: voter information increasing in congruence

Example



Congressional districts



Congruence between newspaper markets

FIG. 2.-Congruence in Missouri

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Graphical results

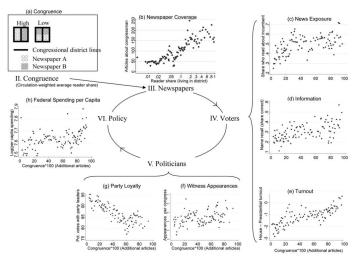


FIG. 1.-Structure of empirical investigation

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- Identifying assumption: congruence is not related to interest in politics.Validity? Endogenous demand?
- Step 1: Does ReaderShare increase political coverage
 - q_{mdt} is number of articles in *m* in district *d* in year *t*. $q_{mt} = \sum_{d} q_{mdt}$.
 - Data from 161 newspapers.
 - Regress

$$\frac{q_{mdt}}{q_{mt}} = ReaderShare_{mdt} + X + \varepsilon$$

where X are controls like party leaders, seniority, freshmen, majority party, scandals, etc

• Validity? Placebo using general political knowledge

Empirics

Coverage

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			J.S. House Mem Articles about Co		
		(1)	(2)	(3)	(4)
	ReaderShare	177.25	164.14		
© The University of Chicago Press.	Congruence	(17.95)***	(17.06)***	171.10 (19.42)***	170.64 (6.18)***
All rights reserved. This	Party leader		154.62 (50.53)***	(19.42)*** 191.93 (72.45)***	(0.13) 122.70 (10.65)***
content is excluded from our	Scandal		70.21 (18.24)***	82.15 (27.37)***	45.17 (10.76)***
Creative Commons license.	Higher_office (ran or appointed)		90.25	98.21	82.61
For more information see	Out_of_state		(11.22)*** -34.75	$(13.02)^{***}$ -10.45	(8.25)*** -19.99
https:// ocw.mit.edu/help/	Close_race		(9.38)*** 36.02 (16.87)**	(12.26) 53.63 (20.56)**	(4.19)*** 33.00 (11.01)***
faq-fair-use/	Freshman		5.32 (3.63)	8.07 (5.08)	9.66 (4.09)**
	Retired		18.38 (7.42)**	29.43 (9.26)***	19.94 (5.88)***
	% urban		-18.40 (12.39)	.19 (13.37)	-34.36 (5.40)***
	Median income		24.67 (37.71)	14.57 (45.38)	-24.79 (17.78)
	Observations R^2	4,206 .18	4,206 .27	2,308 .26	3,421 .28

TABLE 2

* Significant at 10 percent.

** Significant at 5 percent.

- Step 2: Does Congruence increase voter information
 - Regress

 $info_{ict} = \gamma Congruence_{ct} + x_{ict} + \alpha_t + \alpha_r + \varepsilon$

where *info* are measures about how much an individual knows about their congressman, x are individual controls (party, education, income, age, gender, race) representative controls (tenure, majority, etc) and election controls, α_t are state & year FE and α_r are congressman * 3 term FE

- Where is identification coming from? thoughts?
- They also run a specification with county fixed effects, which is identified off of changes in district every ten years.

Empirics

Voter Information

	VOTER K	TABLE NOWLEDGE OF HO	4 USE REPRESENTATIV	/E.
	Ва	SELINE	WITHIN- RACE	REDISTRICTING
	(1)	(2)	(3)	(4)
Controls	No	Yes	Yes	Yes
Fixed effects	Year	State × year	District × year	State × year, count
		Dependent Va	iable: ReadAboutIn	cumbent
Congruence	.29	.42	.40	.30
	(.08)***	(.09)***	(.12)***	(.09)***
Observations	8,985	8,985	8,985	8,985
R^2	.12	.22	.24	.18
		Depender	t Variable: NameRe	ecall
Congruence	.28	.35	.42	.27
-	(.07)***	(.07)***	(.07)***	(.06)***
Observations	14,139	14,139	14,139	14,139
R^2	.16	.27	.30	.24
		Dependent V	ariable: NameReco	gnition
Congruence	.04	.08	.10	.07
	(.05)	(.05)	(.06)*	(.06)
Observations R ²	9,624	9,624 39	9,624 42	9,624
ĸ	.27			.31
	-	-	e: FeelingThermon	
Congruence	.21	.20	.19	.29
	(.05)***	(.06)***	(.07)***	(.09)***
Observations B ^a	12,459 18	12,459 .25	12,459 .28	12,459 .19
n	.10		.20 ble: IdeologicalRati	
-			0	8
Congruence	.22 (.09)***	.25 (.09)***	.30 (.10)***	.20 (.09)**
Observations	(.09)***	(.09)***	7.441	(.09)** 7.441
B ²	18	.25	.27	19
n	.10		able: LikesOrDislike	
C	.28	.26	.21	.30
Congruence	.28 (.08)***	.20 (.09)***	.21 (.09)**	.30 (.07)***
Observations	10,775	10.775	10.775	10.775
R ²	.17	29	32	94

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* Significant at 10 percent.

** Significant at 5 percent. *** Significant at 1 percent. NOTE .- Results are from OLS regressions. Robust standard errors, clustered by county, are in parentheses.



- Step 3: Does Congruence increase politician effort
 - Aggregate Congruence to district level

$$Congruence_d = \sum_{m=1}^{M} MarketShare_{md} ReaderShare_{md}$$

$$effort_d = Congruence_d + X_d + \varepsilon$$

where X is district population controls and (sometimes) district FE.

• Note that here district FE are not as persuasive, since district boundaries change

TABLE 13 Dependent Variable: Number of Witness Appearances before Congressional Hearings									
	(1)	(2)	(3)	(4)	(5)	(6)	(7)		
Congruence	.41 (.17)**	.42 (.16)***	.41 (.15)***	.43 (.21)**	.41 (.18)**	.44 (.23)*	.38 (.22)*		
District controls	X	X	Х	X	Х	Х	Х		
Race and rep- resentative controls			х	х	х	х	х		
Fixed effects	State, year	State, year	State, year	District, year	Rep., year	State, year	State, year		
Estimation procedure	Poisson	NB	NB	NB	NB	Poisson	NB		
Appearance before committee	All	All	All	All	All	Appr., W&M	Appr., W&M		
Observations	4,890	4,890	4,890	4,890	4,890	4,890	4,890		

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Empirics

Effort

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		Commit	TEE ASSIGN	MENTS		
			Depende	NT VARIABLE		
		utive Comr Assignment			icy Commit Assignment	
	(1)	(2)	(3)	(4)	(5)	(6)
Congruence	.41 (.07)***	.15 (.09)	.05 (.08)	18 (.06)***	07 (.08)	21 (.07)***
Controls	No	Yes	Yes	No	Yes	Yes
Fixed effects	State	State	Year,	State	State	Year,
	× year	× year	district	× year	× year	district
Observations	4,508	4,508	4,508	4,771	4,771	4,771
R^2	.18	.37	.56	.12	.24	.54

 TABLE 14

 Committee Assignments

NOTE.-Results are from OLS regressions. The unit of observation is House representative by congressional session. Standard errors clustered by House representative are in parentheses.

* Significant at 10 percent.

** Significant at 5 percent.

*** Significant at 1 percent.

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Effort

Empirics

	(1)	(2)	(3)	(4)	(5)
Congruence	-5.38 (2.06)***	-4.75 (2.03)**	-4.65 (2.21)**	-6.75 (2.63)**	-3.27 (1.40)**
District controls	x	X	X	X	X
Race and represen- tative controls		Х	Х	Х	х
Fixed effects	State,	State,	State ×	District,	Rep.,
	year	year	year	state ×	state ×
Observations R^2	4,534 .19	4,534 .32	4,534 .38	year 4,534 .68	year 4,534 .91

 TABLE 15

 Dependent Variable: Percentage of Roll Call Votes with Party Leadership

NOTE.-Results are from OLS regressions. The unit of observation is House representative by congressional session. Standard errors, clustered by congressional district, are in parentheses.

* Significant at 10 percent.

** Significant at 5 percent.

*** Significant at 1 percent.

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	(1)	(2)	(3)	(4)	(5)
(Democrats win) ×					
Congruence	.18	.16	.18	.16	.15
0	(.05) * * *	(.05) * * *	$(.06)^{***}$	(.05)***	(.06)***
Democrats win	75	75	81	75	80
	$(.03)^{***}$	(.03) * * *	(.08)***	(.03)***	(.09)***
Congruence	04	09	09	06	04
0	(.04)	(.04)**	(.05)*	(.04)	(.04)
Controls	No	Main	Main,	No	Main,
			urban		urban
			interacted		interacted
Fixed effects	No	No	No	Year,	Year,
				district	district
Observations	3,959	3,959	3,959	3,959	3,959
R^2	.87	.90	.90	.96	.96

 TABLE 16

 Dependent Variable: Nominate Scores First Dimension

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- Step 4: Does *Congruence* increase political outcomes for the district
 - Aggregate Congruence to county level
 - Regress congressional spending on *Congruence* with same county level controls and county FE as before controls and county FE as before

	BASE	LINE	WITHIN-RACE		WITHIN N Cour		REDISTRICTING	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Congruence	.092 (.030)***	.137 (.027)***	.064 (.030)**	.094 (.030)***	.106 (.039)***	.096 (.038)**	.051 (.021)**	.035 (.020)*
Controls	No	Yes	No	Yes	No	Yes	No	Yes
Fixed effects	State × year	State × year	District × year	District × year	Year, neighbor	Year, neighbor	State × year, county	State × year county
Observations R ²	33,085 .259	33,085 .393	28,787 .441	28,787 .516	16,698 .638	16,698 .677	33,085 .8	33,085 .890

TABLE 17 DISTRIBUTION OF FEDERAL FUNDS ACROSS COUNTIES, 1984–2004 Dependent Variable: Log Spending per Capita

NOTE - Results are from OLS regressions. The unit of observation is county by congressional session. Standard errors, clustered by county, are in parentheses.

* Significant at 10 percent.

** Significant at 5 percent.

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*** Significant at 1 percent.

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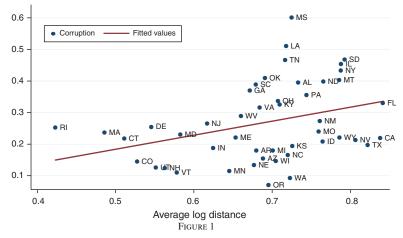
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Another identification idea

Campante and Do (2014): "Isolated Capital Cities, Accountability, and Corruption: Evidence from US States,"

- Campante and Do have another identification idea: isolated capital cities
- Idea is that newspaper market serves major economic hubs.
 - In some states, e.g. Massachusetts, this is the same as the political hub (Boston)
 - In some states, e.g. New York, this is not the same as the political hub (New York City vs. Albany)
- Examine whether there is more corruption in states with isolated capitals
- Views?
- Instrument using deviation of population from centroid of state
 - Idea is that capitals were intended to be in geographic center of state, which may or may not be the same of population center

Results



Note: Corruption __Federal convictions of public officials for corruption-related crime (average 1976–2002); independent variable: *AvgLogDistance_{not}* (average 1920–1970).

0	k	е	n

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
AvgLogDistance _{not}	1.0477*** [0.215]	1.1666*** [0.247]	1.0307*** [0.322]	0.7932*** [0.276]				
$AvgLogDistance_{adj}$					0.8245*** [0.168]	0.8383*** [0.190]	0.8023*** [0.200]	0.5734** [0.223]
Basic control variables		Х	Х	Х		Х	Х	Х
Control I			Х	Х			Х	Х
Control II				Х				х
Observations R^2	48 0.257	48 0.465	48 0.532	48 0.609	48 0.232	48 0.406	48 0.525	48 0.598

TABLE 2-CORRUPTION AND ISOLATION OF THE CAPITAL CITY: AVG LOG DISTANCE

Notes: Robust standard errors in brackets. OLS regressions. Dependent variable: Corruption = Federal convictions for corruption-related crime relative to population, average 1976–2002. Independent variables as of 1970 (AvgLogDistance average 1920–1970). All $AvgLogDistance_mas$ specifications include log *Area* and log *Maximum Distance*. Basic control variables: log *income*, log *population*, *percent college*. Control I: *Share of government employment*, *percent urban*, *census region* dummies. Control II: *racial dissimilarity*, *regulation index*, Share of value added in mining.

*** Significant at the 1 percent level.

** Significant at the 5 percent level.

*Significant at the 10 percent level.

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	1st	1st						
	stage (1)	stage (2)	2SLS (3)	2SLS (4)	2SLS (5)	2SLS (6)	2SLS (7)	2SLS (8)
Panel A. Population:	Centroid							
$AvgLogDistance_{not}$	0.8708*** [0.250]		1.8280*** [0.583]	1.7360*** [0.546]	1.5857*** [0.567]			
$AvgLogDistance_{adj}$		1.0851*** [0.287]				1.4880*** [0.489]	1.3880*** [0.441]	1.2725*** [0.458]
Basic Control	Х	Х	Х	Х	Х	Х	Х	Х
Control I	Х	Х		Х	Х		Х	Х
Control II	Х	Х			Х			Х
Observations	48	48	48	48	48	48	48	48
R^2	0.851	0.677	0.387	0.463	0.538	0.398	0.481	0.551
F-statistic	12.15	14.27	_	_	_		_	_
AR p-value	-	-	0.002	0.002	0.003	0.002	0.002	0.003
Panel B. Land suitab	ility: Centroid							
$AvgLogDistance_{not}$	1.2427** [0.456]		1.1403 [0.976]	1.7231** [0.858]	1.4375** [0.681]			
$AvgLogDistance_{adj}$		1.4166** [0.530]				0.8999 [0.776]	1.4495** [0.734]	1.2610** [0.618]
Basic control	х	х	х	х	х	х	х	Х
Control I	Х	Х		Х	Х		Х	Х
Control II	Х	Х			Х			Х
Observations	48	48	48	48	48	48	48	48
R^2 (centered)	0.828	0.607	0.465	0.465	0.562	0.456	0.469	0.553
F-statistic	7.42	7.15		_	_	_	_	_
AR p-value		_	0.333	0.033	0.014	0.333	0.033	0.015

TABLE 5-CORRUPTION AND ISOLATION OF THE CAPITAL CITY: ADDRESSING CAUSALITY

Newspaper Coverage

Number of search hits	State elections (1)	State budget (2)	State government (3)	Governor's name (4) 1,377.846*** [239.350]	
ReaderConcentr	884.057*** [304.295]	983.524*** [254.500]	1,164.911** [555.114]		
Observations R^2	431 0.783	436 0.770	436 0.789	436 0.716	

TABLE 6—NEWSPAPER COVERAGE OF STATE POLITICS AND THE CONCENTRATION OF CIRCULATION AROUND THE CAPITAL

Notes: Robust standard errors in brackets, clustered by state. OLS regressions. Dependent variable: Number of search hits for each term in NewsLibrary.com (January 1, 2008 to December 31, 2009). Control variables: log of daily circulation, Number of search hits for "Monday," state fixed effects.

***Significant at the 1 percent level.

**Significant at the 5 percent level.

*Significant at the 10 percent level.

	Knowledge (1)	Knowledge (2)	Interest (3)	Interest (4)	Gen. interest (5)	Gen. interest (6)
log distance to capital	-0.0623*** [0.0205]	-0.0836*** [0.0252]	-0.0326 [0.0227]	-0.0649** [0.0288]	-0.0001 [0.0218]	-0.0120 [0.0275]
State fixed effects County controls Individual controls	X X	X X X	X X	X X X	X X	X X X
Observations Mean of dependent variable	780 0.662	780 0.662	652 0.403	648 0.403	780 0.590	776 0.590
Pseudo R ²	0.033	0.172	0.021	0.160	0.014	0.207

TABLE 8—DISTANCE TO THE CAPITAL AND INDIVIDUAL ENGAGEMENT WITH STATE POLITICS

Notes: Robust standard errors in brackets, clustered by county. Probit regressions, marginal effects reported. Dependent variables: *Knowledge* = dummy for knowing name of at least one candidate in gubernatorial elections; *Interest* = dummy for caring "a great deal," "quite a bit," or "some" about newspaper articles regarding gubernatorial elections (conditional on reading newspapers); *General interest* = dummy for reporting interest in government and public affairs "most of the time" or "some of the time." County controls: *population, percent urban, population density, percent non-White, median age, median income, and median schooling* (from 1990 Census); Individual controls: dummies for *age, occupation, sex, income, and political party identification, and number of children and general level of information* (from ANES). All columns include state fixed effects.

***Significant at the 1 percent level.

**Significant at the 5 percent level.

*Significant at the 10 percent level.

	OLS	OLS	OLS	2SLS	2SLS	2SLS
	PG exp.	Oth. exp.	PG prov.	PG exp.	Oth. exp.	PG prov.
	(1)	(2)	(3)	(4)	(5)	(6)
AvgLogDistance _{not}	-0.478***	0.319***	-2.690*	-0.552**	0.330**	-0.405
	[0.137]	[0.102]	[1.533]	[0.217]	[0.149]	[2.517]
Observations	48	48	48	48	48	48
AR <i>p</i> -value				0.021	0.041	0.874
R^2	0.451	0.593	0.877	0.448	0.593	0.871

TABLE 12—PUBLIC GOODS AND ISOLATION OF THE CAPITAL CITY

Notes: Robust standard errors in brackets. Dependent variables: *PG exp.* (public good expenditures) = share of state expenditures on education, public welfare, health, and hospitals in 2008; *Oth. exp.* (other expenditures) = share of state expenditures on government administration, interest on debt, and "other" in 2008; *PG prov.* (public good provision) = first principal component of "Smart State" Index (2005), percent of population with health insurance (2008–2009), and log of hospital beds per capita (2009). Independent variables: *AvgLogDistance_{mat}* are age 1920–2000. Control variables: log *area* and log *maximum distance*, log *income*, log *population, percent celege, share of government employment, racial dissimilarity, percent urban, regional* dummies (all specifications). IV: *centroid AvgLogDistance_{mat}* of population. AR *p*-value: *p*-value from Anderson-Rubin (minimum distance) test.

***Significant at the 1 percent level.

**Significant at the 5 percent level.

*Significant at the 10 percent level.

Media Crowd-out and Policy

Eisensee and Stromberg (2007)

- Setting: US response to foreign disasters
- Empirical idea:
 - Disasters that strike when the news is focused on other things get less media coverage, and therefore less political response
- Two versions of this:
 - The Olympics
 - "Daily news pressure" which is the average amount of time news spends on the top 3 stories (average of the 40 days after disaster)
 - Include yea, month, country, and disaster type FE and controls for intensity of disaster

	Dependent variable: News				Dependent variable: Relief			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
News Pressure	-0.0162 (0.0041)***	-0.0163 (0.0041)***	-0.0177 (0.0057)***	-0.0142 (0.0037)***	-0.0117 (0.0045)***	-0.0119 (0.0045)***	-0.0094 (0.0058)	-0.0078 (0.0040)**
Olympics	-0.1078 (0.0470)**	-0.1079 (0.0470)**	-0.0871 (-0.0628)	-0.111 (0.0413)***	-0.1231 (0.0521)**	-0.1232 (0.0521)**	-0.1071 (0.0763)	-0.1098 (0.0479)**
World Series	-0.1133 (-0.1065)				-0.1324 (0.1031)			
log Killed			0.0605 (0.0040)***				0.0582 (0.0044)***	
log Affected			0.0123 (0.0024)***			0.0376	(0.0024)***	
Imputed log Killed				0.0491 (0.0034)***				0.0442
Imputed log Affected				0.0151 (0.0020)***				0.0394 (0.0020)***
Observations	5,212	5,212	2,926	5,212	5,212	5,212	2,926	5,212
R-squared	0.1799	0.1797	0.3624	0.2875	0.1991	0.1989	0.4115	0.3726

TABLE IV EFFECT OF THE PRESSURE FOR NEWS TIME ON DISASTER NEWS AND RELIEF

Linear probability OLS regressions. All regressions include year, month, country, and disaster type fixed effects. Regressions with imputed values (4 and 8) also include fixed effects for the interaction of missing values and disaster type. Robust standard errors in parentheses: * significant at 10 percent; ** significant at 5 percent; *** significant at 1 percent.

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- We've focused on the "news" coverage of politics
- But what about explicitly politcal advertising
- Surprisingly there is much less here
- Llareguy, Marshall, and Snyder (2014) examine this in the context of Mexico
 - Exploit a reform where radio and TV ad shares are allocated based on previous election results
 - Look at spillovers due to the fact that media markets are not coincident with electoral boundaries
- Seems like a good topic for more work, particularly in the US

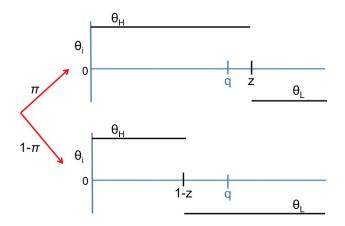
- Newer area of research has emphasized the role of media as a coordination device
 - Suppose you hate the government and want to protest.
 - There is safety in numbers. Government can easily quash 10 person protest; much harder to surpress 1 million people
 - Then successful protests involve coordination everyone may want to protest, but will only do so if they believe everyone else will also protest with them
- Media are important as a coordination device
 - Traditional media: broadcasting plans for rallies etc.
 - Aside: this is why step 1 of a coup is to seize the TV stations
 - Social media: interactivity allowing people to agree among themselves

- Barbera and Jackson write down a formal version of coordination game I sketched earlier to study the impact of better information on successful protests
- Setup:
 - Players of mass 1 indexed by *i*. Each person has type θ_H or θ_L .
 - Collective action (protest, revolution, whatever) is successful iff at least fraction *q* ∈ [0, 1] participates.
 - Coordination comes from payoffs: Each individual receives payoffs $\frac{\theta}{C}$ if they participate and protest succeeds and payoff -C if they participate and protest fails.
 - Note: each individual is small relative to success of protest . So these are the individual utility gains/losses from participation in the protest.

Information

- Two states of the world: With probability π , in "High" state with fraction $z > q \ge \frac{1}{2}$ are H types; with probability 1π , in "Low" state with fraction 1 z < q are H types.
- Note: coordination games usually involve multiple equiliria. In this case, equilibrium where nobody participates is always an equilibrium. They focus on the other equilibrium, i.e., equilibrium with most protest, which sometimes exists.
- Information questions come from the inference question: I know my type, and maybe I get some other signals. What's my inference about this?

Setup



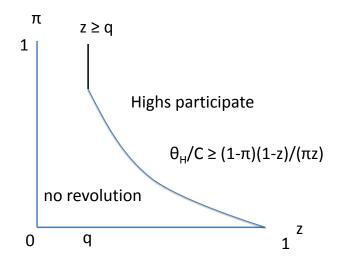
• Suppose no other signal. I am an *H* type. By Bayes rule, probability of "High" state is

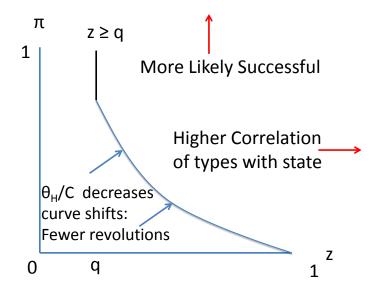
$$\frac{\pi z}{\pi z + (1 - \pi)(1 - z)}$$

• Should I revolt if I'm a high type? I will do so if

$$\begin{aligned} \theta_{H} Pr(\textit{HighState}) &\geq C(1 - Pr(\textit{HighState})) \\ \frac{\theta_{H}}{C} &\geq \frac{(1 - Pr(\textit{HighState}))}{Pr(\textit{HighState})} \\ \frac{\theta_{H}}{C} &\geq \frac{(1 - \pi)(1 - z)}{\pi z} \end{aligned}$$

• Note that this implies that probability of revolt is increasing in both π and z, through Bayes rule





- Suppose each agent can see 1 other randomly chosen agent's type. What happens?
- Changes updating through Bayes rule about probability of High state.
- If an *H* agent sees another *H* agent, then probability of High state (through Bayes rule) is

$$\frac{\pi z^2}{\pi z^2 + (1 - \pi)(1 - z)^2}$$

• If an H agents sees a low agent, then probability of High state is

$$\frac{\pi z(1-z)}{\pi z(1-z) + (1-\pi)z(1-z) = \pi}$$

- So now two cases to consider:
 - High type will show up if they see a low type if

$$\pi\theta_H \ge (1-\pi)C$$

Note this is more demanding condition than before, since you are less optimistic about high state than if you'd never seen a signal.

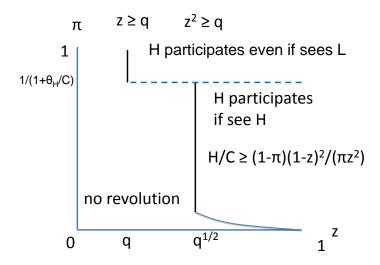
• High type will show up if they only see a low type if both previous condition (with new Bayes rule)

and there are enough other H types who see other H for revolution to work, i.e.

$$z^2 \ge q$$

Olken

Equilibrium



Effect of information

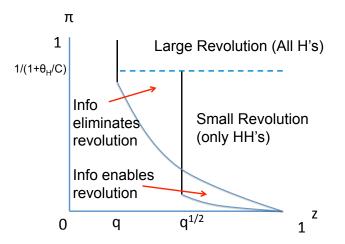


Figure 6: Sometimes information aids the revolution and other times it blocks it

types are ex ante worse off and the L types are better off. This happens if $z^2 < q < z$ while (2) holds.

Olken

Media Lecture 2€Ë€Fà

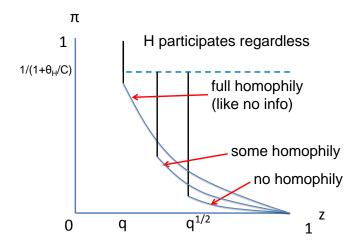
Effect of information

- Key point: information doesn't necessarily always facilitate revolution
- Why?
- There is now a region where H types would have shown up before, but now, will only show up if they see another H type. So you get a "smaller" revolution than before (only the HH participate. Only an equilibrium if $z^2 > q$.
- On the other hand, when π is low, there's a region where the information value of the signal helps a lot relative to what you would had from Bayes rule with no signals

Social media

- To extent to social media, now suppose that there is some correlation in matching.
- That is, suppose a fraction h ∈ [0, 1] of matches that would have been cross-type are always same type. So h = 0 is random and h = 1 is perfect homophily.
- This changes the Bayes rule since you now have to account for *h* in your updating.
- Equilibrium is now somewhere in between the two models.

Effect of homophily



Enikolopov et al 2016: Social Media and Protest Participation: Evidence from Russia

- Does this matter? Lots of interest in role of social networks in facilitating protests, esp. vis-a-vis Arab spring
- Enikolopov et al look at this in context of Russia, looking at VK (Russian social network)
- Empirical idea: VK was launched by Pavel Durov in 2006, and started by inviting his classmates to participate. Network is largest in these cities.
- They show that there are more protests in 2011 in cities with more classmates of Durov.
- They control for average number of students from various cities studying at same university in other cohorts.

First stage

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Table 1 Determinants of VK penetration in 2011 (first stage regression)

Table 1. Determinar	its of VK per	netration in 2	tori (first sta	age regressi	on).		
			Log (numb	per of VK users), Aug 2011		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Log (SPbSU students), same 5-year cohort as VK founder	0.4847***	0.1581***	0.1416***	0.1322***	0.1393***	0.1371***	0.1360***
Log (SPbSU students), one cohort younger than VK founder	[0.1443] 0.5741***	[0.0425] -0.0292	[0.0466] -0.0259	[0.0489] -0.0452	[0.0482] -0.0433	[0.0463] -0.0464	[0.0488] -0.0457
Log (3PD30 students), one conort younger than VK lounder	[0.1064]	[0.0552]	[0.0463]	[0.0461]	[0.0468]	[0.0472]	[0.0474]
Log (SPbSU students), one cohort older than VK founder	0.3101	0.0250	0.0058	0.0161	0.0175	0.0137	0.0142
Log (of boo statema), one consist state than the logitation	[0.1866]	10.05231	[0.0472]	[0.0468]	10.04671	[0.0445]	[0.0454]
Regional center	[]	0.2952***	0.3932***	0.3015*	0.2563*	0.3008*	0.3026*
		[0.0899]	[0.1268]	[0.1583]	[0.1526]	[0.1539]	[0.1523]
Distance to Saint Petersburg, km			0.0002	0.0001	0.0001	0.0002	0.0000
			[0.0001]	[0.0001]	[0.0001]	[0.0001]	[0.0001]
Distance to Moscow, km			-0.0002	-0.0002	-0.0002	-0.0003	-0.0001
			[0.0001]	[0.0001]	[0.0001]	[0.0002]	[0.0001]
Rayon center (county seat)			0.0045	-0.0142	-0.0134	-0.0056	-0.0155
			[0.0916]	[0.0873]	[0.0869]	[0.0906]	[0.0843]
Log (average wage), city-level, 2011			0.1688	0.2108	0.1977	0.1756	0.1386
Presence of a university in a city, 2011			[0.1573]	[0.1637] -0.0224	[0.1686] -0.0087	[0.1691] -0.0348	[0.1571] -0.0056
Presence of a university in a city, 2011				[0.1496]	[0.1468]	[0.1478]	[0.1441]
Internet penetration, region-level, 2011				-0.1190	-0.1572	-0.0677	-0.0875
internet penetration, region level, 2011				[0.2304]	[0.2144]	[0.2272]	[0.2254]
Log (number of Odnoklassniki users), 2014				0.1475*	0.1391*	0.1322	0.1706**
				[0.0798]	[0.0806]	[0.0801]	[0.0793]
Ethnic fractionalization, 2010				0.4041*	0.4872**	0.5660***	0.4599**
				[0.2149]	[0.2073]	[0.2016]	[0.2197]
Observations	625	625	625	625	625	625	625
R-squared	0.4031	0.8263	0.8486	0.8517	0.8546	0.8550	0.8540
Population controls		Yes***	Yes***	Yes***	Yes**	Yes***	Yes***
Age cohort controls			Yes**	Yes**	Yes***	Yes**	Yes**
Education controls			Yes***	Yes***	Yes***	Yes***	Yes***
Electoral controls, 1995					Yes		
Electoral controls, 1999						Yes*	
Electoral controls, 2003	0.555	0.045**	0.0501	0.0571	0.040**	0.0548	Yes
p-value for equility of coefficients for three cohorts	0.555	0.045**	0.059* 0.021**	0.057*	0.048** 0.015**	0.051*	0.047**
p-value for equility of coefficients of Durov's and younger cohort p-value for equility of coefficients of Durov's and older cohort	0.679	0.019** 0.054*	0.021** 0.049**	0.017**	0.015**	0.016** 0.069*	0.014** 0.072*
p-value for equility of coefficients of Durov's and older conort	U.458	0.054*	0.049**	0.088"	0.072*	0.009*	0.072*

**** p-0.01, ** p-0.05, **p-0.1. Robust standard errors in brackets are adjusted by clusters within regions. Unit of observation is city. Logarithm of any variable is calculated with 1 added niske. When **es' is added to indicate inclusion of a group of controls, a significance level is reported immediately after for this group of controls. Fieldle controls for population (5th polynomial) are included in all specifications. Age cohort controls include the number of people aged 20-24, 25-29, 30-34, 35-39, 40-44, 45-49, 50 and older years, in each city according to 2010 Russian Census. Education controls include the share of population with higher education overall according to 2002 Russian Census and segnately in each of the age contrast according to 2010 Russian Census. Education control include the share of population with higher education overall according to 2002 Russian Census and segnately in each of the age contrast according to 2010 Russian Census. The agencount for both the levels and the change in educaton. Electratic controls include the share of a corresponding user.

Log (SPbSU students), same 5-year cohort as VK founder 0.252** 0.259** 0.263** 0.274** 0.062**** 0.062**** 0.062**** 0.062**** 0.062**** 0.062**** 0.062**** 0.062***** 0.062**** 0.061***** 0.060***** 0.065***** 0.065***** 0.065***** 0.065****** 0.065***********************************	(7) (8) 064*** 0.065** 0.020] [0.021] 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.000] [0.000] 0.000] [0.000] 0.000] [0.000]
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Log (SPbSU students), one ochort younger than VK founder 0.152 0.150 0.137 0.100 0.012 0.011 0.021 0.011 0.021 0.011 0.020 0.022 -0.088 0.318 0.228 0.013 0.013 0.013 0.013 0.013 0.013 0.013 0.013 0.013 0.013 0.013 0.020 0.000	0.009 0.012 0.020] [0.020] 0.018 -0.015 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.009 -0.014 0.009 -0.014 0.009 -0.014 0.009 -0.014 0.000] [0.009 0.000] [0.000] 0.000] -0.000
Log (SPbSU students), one ochort younger than VK founder 0.152 0.150 0.137 0.100 0.012 0.011 0 Log (SPbSU students), one ochort older than VK founder -0.075 -0.072 -0.082 -0.088 0.0133 10.020 [0.020]	0.009 0.012 0.020] [0.020] 0.018 -0.015 0.020] [0.020] 0.020] [0.020] 0.020] [0.020] 0.009 -0.014 0.009 -0.014 0.009 -0.014 0.009 -0.014 0.000] [0.009 0.000] [0.000] 0.000] -0.000
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Log (SPhSU students), one ochort older than VK founder -0.072 -0.082 -0.068 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.017 -0.016 -0.028 0.018 0.0292 10.0201 [0.020] [0.020] [0.020] [0.020] [0.037] [0.0487] [0.487] [0.487] [0.090] [0.000] [0.018 -0.015 0.020] [0.020 0.009 -0.014 0.096] [0.098 0.000 0.000 0.000] [0.000 0.000] [0.000
Construction Construction<	0.020] [0.020 0.009 -0.014 0.096] [0.098 0.000 0.000 0.000] [0.000 0.000] -0.000
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ID:176[[0.190] [0.193] [0.184] [0.034] [0.037] [0.799] Presence of a university in a city, 2011 0.870° 0.880° 0.880° 0.989° 0.195° 0.195° 0.195° 0.195° 0.195° 0.195° 0.195° 0.195° 0.195° 0.069° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.028° 1.098° 1.038° 1.028° 1.038° 1.038° 1.028° 1.038° 1.038° 1.028° 1.038° 1.038° 1.038° 1.038° 1.038° 1.038° 1.038° 1.038° 1.038°	0.010] [0.011]
Presence of a university in a city, 2011 D.870 ⁺⁺ D.876 ⁺⁺ D.889 ⁺⁺ D.196 ⁺⁺ D.132 D.0480 D.1088 D.118 ⁺⁺ D.148 ⁺⁺ D.148 ⁺⁺ D.138 ⁺⁺ D.138 ⁺⁺⁺ D.138 ⁺⁺⁺ D.138 ⁺⁺⁺ D.148 ⁺⁺⁺⁺ D.148 ⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺⁺	.007 -0.014
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Internet penetration, region-level, 2011 0.18 0.181 0.175 0.149 -0.013 0.005 -1 Log (number of Odnoklassniki users), 2014 10.443 10.2401 10.2801 10.2877 10.0451 [0.045] [0.140] 10.0451 [0.140] 10.0451 [0.140] 10.0451 [0.140] 10.0451 [0.140] 10.161 [0.1919] 10.071 10.0191 [0.017] [0.047] [0.149] [0.171] [0.173] [0.048] [0.560] -0.516 -0.468 -0.506 -0.089 -0.081 -	.195** 0.200*
ID2431 [D2443] [D2457] [D2567] [D2567] [D2567] [D2562] [D257] [D2453]	0.097] [0.097]
Log (number of Odnoklassniki users), 2014 0.104 0.081 0.157 0.133 0.022* 0.024* 0 [D109] [D.120] [D.119] [D.017] [D.019]	0.003 -0.007
[0.109] [0.120] [0.123] [0.119] [0 Ethnic fractionalization, 2010 - 0.580 - 0.516 - 0.488 - 0.506 - 0.089 - 0.081 - 1 [0.221] [0.335] [0.337] [0.343] [0.059] [0.061] [0 Observations 625 625 625 625 625 625	0.054] [0.048
Ethnic fractionalization, 2010 -0.516 -0.468 -0.506 -0.089 -0.081 - [0.321] [0.335] [0.337] [0.343] [0.069] [0.061] [0 Observations 625 625 625 625 625 625	.041* 0.034*
[0.321] [0.335] [0.337] [0.343] [0.059] [0.061] [0 Observations 625 625 625 625 625 625	0.021] [0.019
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	0.062] [0.062]
	625 625
	0.781 0.781
	'es*** Yes***
	'es*** Yes***
	Yes* Yes*
Electoral controls, 1995 Yes** Yes**	
	Yes*
Electoral controls, 2003 Yes* p-value for equity of coefficients for three cohorts 0.271 0.271 0.250 0.247 0.078* 0.071* 0	Yes*** .058* 0.069*
	.058 0.069
p-value for equility of coefficients of Durov's and younger conort 0.528 0.489 0.430 0.430 0.487 0.089* 0.073* 0 p-value for equility of coefficients of Durov's and older cohort 0.115 0.111 0.099* 0.102 0.031** 0.032** 0.	

Table 2. Student cohorts and protest participation in 2011. Reduced form estimation

^{am} p-0.01, ^{am} p-0.05, ^{am} p-0.1. Robust standard errors in brackets are adjusted by dusters within regions. Unit of observation is a givil Logarithm of any vaniable is calculated with 1 added inside. When ^{am} vers is added to indicate inclusion of a group of controls, is supplicated to ever the reported immediately after for this group of controls. Floated is controls for the group of controls. Floated is controls for the set of the set o

Table 3. VK penetration and protest participation in 2011.

Panel A. Number of protesters

• • • • • • • • • • • • • • • • • • • •	Log (number of protesters), Dec 2011								
	IV	IV	IV	IV	OLS	OLS	OLS	OLS	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Log (number of VK users), Aug 2011	1.912**	1.863**	1.920**	2.015**	0.228***	0.216***	0.216***	0.227***	
	[0.900]	[0.862]	[0.886]	[0.906]	[0.072]	[0.072]	[0.074]	[0.076]	
Log (SPbSU students), one cohort younger than VK founder	0.238*	0.231*	0.227*	0.252*	0.224**	0.224**	0.211*	0.236**	
	[0.124]	[0.125]	[0.125]	[0.131]	[0.107]	[0.109]	[0.108]	[0.108]	
Log (SPbSU students), one cohort older than VK founder	-0.106	-0.105	-0.108	-0.097	0.013	0.019	0.011	0.027	
	[0.143]	[0.143]	[0.136]	[0.144]	[0.092]	[0.091]	[0.089]	[0.092]	
Population controls	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	
Age cohort controls	Yes	Yes	Yes	Yes	Yes*	Yes*	Yes**	Yes**	
Education controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Other controls	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	
Electoral controls, 1995		Yes				Yes			
Electoral controls, 1999			Yes				Yes		
Electoral controls, 2003				Yes*				Yes**	
Observations	625	625	625	625	625	625	625	625	
Effective F-statistics (Olea Montiel and Pflueger 2013)	276.8	274	274	274					

Panel B. Probability of protests

			Inc	idence of protes	ts, dummy, Der	2011		
	IV	IV	IV	IV .	OLS	OLS	OLS	OLS
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Log (number of VK users), Aug 2011	0.466***	0.446***	0.464***	0.481***	0.039***	0.037***	0.037***	0.039***
	[0.180]	[0.169]	[0.174]	[0.181]	[0.013]	[0.013]	[0.013]	[0.014]
Log (SPbSU students), one cohort younger than VK founder	0.033	0.030	0.031	0.034	0.029	0.029	0.027	0.031
	[0.025]	[0.026]	[0.026]	[0.027]	[0.020]	[0.021]	[0.021]	[0.020]
Log (SPbSU students), one cohort older than VK founder	-0.024	-0.023	-0.025	-0.021	0.006	0.007	0.005	0.009
	[0.029]	[0.029]	[0.028]	[0.030]	[0.017]	[0.017]	[0.017]	[0.018]
Population controls	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***
Age cohort controls	Yes	Yes	Yes	Yes	Yes**	Yes**	Yes**	Yes**
Education controls	Yes	Yes	Yes*	Yes	Yes	Yes	Yes	Yes
Other controls	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***
Electoral controls, 1995		Yes				Yes**		
Electoral controls, 1999			Yes				Yes	
Electoral controls, 2003				Yes				Yes**
Observations	625	625	625	625	625	625	625	625
Effective F-stat (Montiel Olea and Pflueger 2013)	276.8	274	274	274				

¹¹⁰ p-030,²¹ p-0305, ²¹ p-01. Robust standard errors in brackets are adjusted by clusters within regions. Unit of observations is a city Logarithm of any variable is calculated with 1 added india. When ¹¹⁰ errors is added to indicate inclusion of a group of controls, a singlificance level is reported immediately after for this group of controls. Fixelito controls for population (Sth polynomia) are included in all specifications. Age cohort controls include the number of people aged 20-24, 25-29, 30-34, 35-59, 40-44, 54-24, 50 and older years, in each dty according to 2017 Russian Commun. Education controls include the share of population with higher education or overall according to 2022 Russian Census and abgrated by in a dty of the age cohorts in uning party (Cur Home is Russia in 1995, Unity in 1999, United Russia in 2003), vote against all, and electoral lumout for a corresponding year. Other controls include during for regional county center, distances to Maccow and St Peterbury, log everage warge), haire of people with higher education or 2021, internet perentation in 2011, log (Cohortsanik iu 2011, log Cohortsanik iu 2011).

18	IDIE 4. VK PE	enetration	and pre-vr	Protests.					
Panel A. Participation in earlier protests									
	Log (I	number of pro	testers), 198	7-1992	Log (pro	-democracy p	protesters), 19	87-1992	
Log (number of VK users), Aug 2011	0.534	0.427	0.284	0.493	0.144	-0.011	0.017	0.141	
	[1.883]	[1.943]	[1.839]	[1.927]	[1,495]	[1.510]	[1,491]	[1.573]	
P-value for equality of coefficients with that in Table 4	0.492	0.488	0.413	0.463	0.295	0.277	0.265	0.288	
	Log (parti	icipants in lab	or protests),	1997-2002	Log (p	articipants in :	social protests	s), 2005	
Log (number of VK users), Aug 2011	-0.562	-0.537	-1.380	-0.497	-0.313	-0.292	-0.075	-0.042	
	[1.877]	[1.716]	[1.831]	[1.962]	[1.632]	[1.497]	[1.569]	[1.600]	
P-value for equality of coefficients with that in Table 4	0.216	0.193	0.094*	0.220	0.273	0.256	0.314	0.304	
Panel B. Incidence of earlier protests									
	Inc	idence of pro	tests, 1987-1	992	Incidence of pro-democracy protests, 1987-1992				
Log (number of VK users), Aug 2011	0.009	0.007	-0.015	0.024	-0.011	-0.020	-0.023	0.004	
	[0.281]	[0.282]	[0.267]	[0.281]	[0,195]	[0,195]	[0,191]	[0.198]	
P-value for equality of coefficient with that in Table 5	0.194	0.202	0.155	0.197	0.090*	0.092*	0.078*	0.091*	
	Incide	ence of labor	protests, 199	7-2002	Incidence of social protests, 2005				
Log (number of VK users), Aug 2011	-0.070	-0.060	-0.172	-0.036	-0.057	-0.055	-0.022	-0.019	
	[0.243]	[0.219]	[0.238]	[0.256]	[0.239]	[0.221]	[0.230]	[0.235]	
P-value for equality of coefficient with that in Table 5	0.056*	0.047**	0.021**	0.065*	0.105	0.099*	0.123	0.117	
Population, Age cohorts, Education, and Other controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Electoral controls, 1995		Yes				Yes			
Electoral controls, 1999			Yes				Yes		
Electoral controls, 2003				Yes				Yes	
Observations	625	625	625	625	625	625	625	625	
00001101010	323	525	525	020	323	525	525	525	

Table 4. VK Penetration and pre-VK Protests.

Table 6. VK penetration and Voting Outcomes.

	Vo	Voting share for United Russia, 2007				Voting share for United Russia, 2011				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)		
.og (number of VK users), Aug 2011	0.035	0.019	0.045	0.003	0.230*	0.179*	0.230*	0.182		
	[0.050]	[0.041]	[0.046]	[0.037]	[0.128]	[0.099]	[0.118]	[0.104		
.og (SPbSU students), one cohort younger than VK founder	-0.007	-0.004	-0.006	-0.007	-0.002	0.002	-0.001	0.000		
	[0.009]	[0.008]	[0.008]	[0.007]	[0.017]	[0.014]	[0.016]	[0.013		
.og (SPbSU students), one cohort older than VK founder	0.002	0.001	-0.000	-0.003	0.004	0.006	0.001	-0.002		
	[0.008]	[0.007]	[0.008]	[0.006]	[0.017]	[0.013]	[0.015]	[0.013		
Population controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes		
Age cohort controls	Yes***	Yes***	Yes***	Yes**	Yes	Yes	Yes	Yes		
Education controls	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes**		
Other controls	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes***	Yes**		
Electoral controls, 1995		Yes***				Yes***				
Electoral controls, 1999			Yes***				Yes***			
Electoral controls, 2003				Yes***				Yes**		
Observations	625	625	625	625	625	625	625	625		
ffective F-statistics (Olea Montiel and Pflueger 2013)	276.8	274	274	274	276.8	274	274	274		
		Voting share for Medvedev, 2008					Voting Share for Putin, 2012			
.og (number of VK users). Aug 2011	0.125*	0.115*	0 137**	0.098*	0.127*	0 111*	0.127*	0.096		
.og (number of VK users), Aug 2011	0.125	0.0621	0.137	0.098	[0.073]	[0.065]	0.127	10.058		
.og (SPbSU students), one cohort younger than VK founder	-0.005	-0.003	-0.005	-0.004	0.002	0.003	0.003	0.002		
og (SPDSO sludenis), one conon younger than VK tounder	[0.011]	-0.003	[0.010]	10.0081	[0.011]	[0.010]	[0.003	10.002		
						0.007	0.005	0.003		
			0.002							
.og (SPDSU students), one conort older than VK tounder	0.001	-0.000	-0.003	-0.003	0.008					
	0.001 [0.009]	-0.000 [0.008]	[0.009]	[0.007]	[0.011]	[0.010]	[0.010]	[0.009		
.og (SPbSU students), one cohort older than VK founder	0.001 [0.009] Yes	-0.000 [0.008] Yes	[0.009] Yes*	[0.007] Yes**	[0.011] Yes	[0.010] Yes	[0.010] Yes*	[0.009 Yes*		
opulation controls ge cohort controls	0.001 [0.009] Yes Yes**	-0.000 [0.008] Yes Yes*	[0.009] Yes* Yes**	[0.007] Yes** Yes	[0.011]	[0.010] Yes Yes	[0.010] Yes* Yes	[0.009 Yes* Yes		
ropulation controls ge cohort controls ducation controls	0.001 [0.009] Yes Yes** Yes	-0.000 [0.008] Yes	[0.009] Yes*	[0.007] Yes**	[0.011] Yes Yes	[0.010] Yes	[0.010] Yes*	[0.009 Yes* Yes Yes**		
opulation controls ge cohort controls ducation controls ther controls	0.001 [0.009] Yes Yes**	-0.000 [0.008] Yes Yes* Yes	[0.009] Yes* Yes** Yes	[0.007] Yes** Yes Yes	[0.011] Yes Yes Yes***	[0.010] Yes Yes Yes***	[0.010] Yes* Yes Yes****	[0.009 Yes* Yes Yes**		
opulation controls ge cohort controls ducation controls ther controls lectoral controls, 1995	0.001 [0.009] Yes Yes** Yes	-0.000 [0.008] Yes Yes* Yes Yes	[0.009] Yes* Yes** Yes	[0.007] Yes** Yes Yes	[0.011] Yes Yes Yes***	[0.010] Yes Yes*** Yes***	[0.010] Yes* Yes Yes*** Yes***	[0.009 Yes* Yes Yes**		
Socializa controls lige cohort controls ducation controls Tuber controls Tectoral controls, 1995 Tectoral controls, 1999	0.001 [0.009] Yes Yes** Yes	-0.000 [0.008] Yes Yes* Yes Yes	[0.009] Yes* Yes** Yes Yes	[0.007] Yes** Yes Yes Yes***	[0.011] Yes Yes Yes***	[0.010] Yes Yes*** Yes***	[0.010] Yes* Yes Yes****	[0.009 Yes*		
	0.001 [0.009] Yes Yes** Yes	-0.000 [0.008] Yes Yes* Yes Yes	[0.009] Yes* Yes** Yes Yes	[0.007] Yes** Yes Yes	[0.011] Yes Yes Yes***	[0.010] Yes Yes*** Yes***	[0.010] Yes* Yes Yes*** Yes***	[0.009 Yes* Yes Yes** Yes**		

¹¹¹ p-0105, ¹² p-01, ¹² p-0105, ¹² p-01. Robust standard errors in brackets are adjusted by clustern within regions. Unit of observation is a civil, constraintim of any variable is calculated with 1 added inside. When Yare's added to indicide inclusion of a group of controls, significance level is exported immediately after for this group of controls. Fixelike controls for pould include the number of people aged 20-24, 25-29, 9-34, 55-39, 40-44, 45-49, 50 and odder years, in each civil according to 2010 Russian Coreaus. Education controls include the neuropear of people aged 20-24, 25-29, 9-34, 55-39, 40-44, 45-49, 50 and odder years, in each civil according to 2010 Russian Coreaus. Education controls include the neuropear in education. Electral controls include the neuropear in education in education. Electral controls include the neuropear in education in education

Summing up...

- Media can have important roles in policy
 - Through accountability channel
 - And as a coordination device.

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