

Silent Protest on the Bench: The Impact of War on Ethnic Sentencing Disparities in Russian Courts*

*This project is supported by the Charles University Grant Agency no. 69224

EWMES, December 2025

Vlada Kosenkova, CERGE-EI

Arsenii Shcherbov, CERGE-EI

Motivation

- Political events may affect the emotions of people in positions of authority and lead to more subjective outcomes
 - E.g.: judges become more punitive (Eren and Mocan, 2018; Shayo and Zussman, 2011)
 - Propaganda and media persuasion can lead to increase in animosity between ethnic groups (Adena et al., 2015; DellaVigna et al., 2014; Yanagizawa-Drott, 2014)
 - 2022 Russia - Ukraine war
 - Rise of propaganda and censorship (state media budget ↑ 200%)
 - Emphasis on 'traditional Russian values' → distancing of ethnic minorities from ethnic Russians (EEAS Report; [Holod.Media](#); [Ideal.Realii](#))
- This paper studies effects of the war announcement on ethnic bias of judges in Russian courts

This Paper

- We use the month of war announcement as a cutoff to examine sentencing bias in different ethnic pairs of judges and defendants

Data:

- Cases on minor offenses from all courts for 2021-2022
- Implied ethnicity of judges and defendants based on their names

Results:

- Ethnic judges became more punitive to Russian defendants: 5.3pp more likely to sentence to detention for minor offense
 - Effect can be caused by the backlash of minorities to propaganda as it increases in localities with higher predisposition to be affected by it
 - Namely, with lower social trust, lower ethnic fractionalization, and less democratic

Contribution to Related Literature

- **Decision of judges** are affected by emotional stress, characteristics of defendants, exposure to tragic events, media exposure

(Eren and Mocan, 2018; Mustard, 2001; Choi et al., 2022; Shayo and Zussman, 2011; Philippe and Ouss, 2018)

→ *We examine the role of war announcement and intensified propaganda on judges in Russia*

- Propaganda and media persuasion can increase **interethnic hostility** (DellaVigna et al., 2014; Adena et al., 2015; Petrova and Yanagizawa-Drott, 2016)

→ *We study sentencing disparities in ethnic pairs of judges and defendants*

- We aim to extend existing research on **Russian judicial system**, which is scarce and mainly focused on criminal cases

(Volkov, 2016; Zhuchkova and Kazun, 2023; Knorre et al., 2024)

Background

Judicial system:

- 2018 reform: courts should have switched to **automatic allocation** of cases by late 2019
- Overall, about 16,000 judges (around 70% are ethnic Russian) and 1,600 courts

Ethnic minorities: →

- Around 80% of Russian population are ethnic Russians, others belong to different ethnic groups: Ukrainians, Tatars, Bashkirs, etc.
- Names are a good predictor (*Azat Zainullin vs Pavel Ivanov*)
- We divide population into Russian – ethnic Russians, and ethnic minority – non-ethnic Russian (Brunarska and Soral, 2022)

Data

- All reported minor offense cases (Justice database) and court hearings data (to check allocation of cases)
- ~ 5,000 minor offense cases per month

Data analysis:

- ML model to imply ethnicity based on personal names (Bessudnov et al., 2023)
- Regular expressions and NLP libraries to extract sentencing decisions from texts [More details](#)

Empirical Strategy: RD in Time

$$Y_i = \beta_0 + \beta_1 \mathbb{1}(m \geq \bar{c}) + \beta_2 f(m) + \beta_3 \mathbb{1}(m \geq \bar{c}) \times f(m) + \varepsilon_i$$

- i is case, m is year-month of sentencing decision, \bar{c} = February 2022
- focus on minor offenses (Article 21.1 Part 1 of Administrative Code)

More details

- main outcome – probability to receive **detention** as punishment instead of fine

Threats to Identification

Possible concerns:

- Changes in allocation of cases to judges around the cutoff →
 - 2018 reform introduced automated distribution system
 - Checking results only for judges with randomly allocated cases
- Changes in severity of crimes
 - Focusing on easily comparable cases (minor offenses) (Shayo and Zussman, 2011; Kricheli-Katz and Weinshall, 2023; Zhan and Qiao, 2024)
 - Checking the length of decisions texts
- Changes in demographic characteristics of defendants →
 - Estimate local effect, only for months before mobilization and mass migration (Anastasiadou et al., 2024)
- Changes in supply across different types of crime →

Results: ethnic judges are more punitive to Russian defendants

β_1 (After war)	Ethnic defendant	Russian defendant
Ethnic judge ●	0.034 (0.025) $N = 4327$	0.053*** (0.017) $N = 8966$
Russian judge	-0.005 (0.022) $N = 5819$	-0.006 (0.012) $N = 17615$

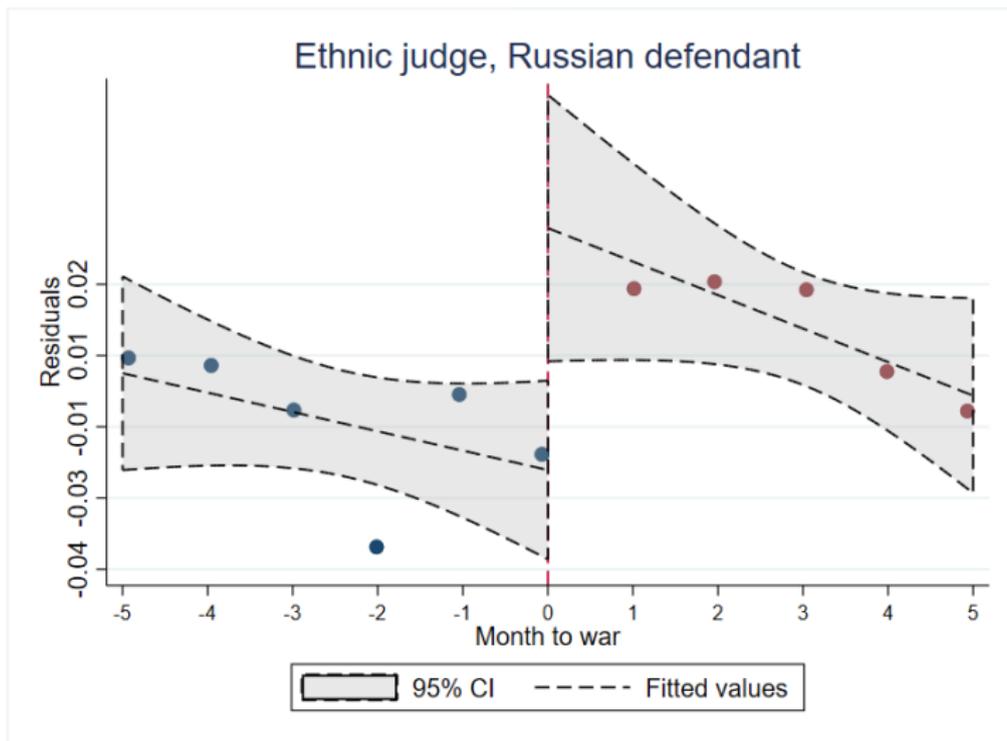
Outcome: 0/1 detention. Notes: Includes court FE, $p(1)$, uniform kernel, data-driven bandwidth, robust standard errors in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Results Plot

Robustness

Random Alloc.

Ethnic Defendant



Potential Drivers I

- To study potential drivers of the effect, we use additional data on city-level (Census 2010 and Enikolopov et al., 2020)
 - **Social trust:** answer positively to the question 'Do you think most people can be trusted or you can't be too cautious with them?'
 - **Ethnic fractionalization:** probability that two randomly picked individuals in a city belong to the same ethnic group
 - **Democratic values:** incidence of protests after 2011 rigged elections in the city

Potential Drivers II

- Results are more pronounced in the localities with lower social trust, low ethnic diversity and less democratic regions
 - Likely to be affected by propaganda to a greater effect (Adena et al., 2015)
 - Minorities can feel more isolated or marginalized in these contexts

	Ethnic judge, Russian defendant			
	All	Low trust	Low ethnic diversity	No protest 2011
β_1 (After war)	0.053*** (0.017)	0.071* (0.037)	0.074*** (0.026)	0.066** (0.033)
N	8966	1732	4783	2671

Outcome: 0/1 detention. Notes: Includes court FE, $p(1)$, uniform kernel, data-driven bandwidth, robust standard errors in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

- Results are not driven by specific ethnic group (more detailed analysis) or Moscow/Saint Petersburg

Conclusion

- We examined the sentencing disparities for different ethnic pairs of judges and defendants before and after war
- We used **data** on sentencing decisions for minor offense cases happened in 2021-2022
- **Results:** ethnic judges become more punitive towards Russian defendants
 - Effect is stronger in localities with lower ethnic diversity, lower social trust, and lower potential to protest
 - Might be explained by the reaction to increased propaganda-induced hostility from the ethnic majority

Further Steps

- Extending dataset to other types of crimes (currently DUI, theft)
 - Extracting characteristics of the case and defendant (expand on Zhuchkova and Kazun, 2023)
 - Sentiment analysis of the sentencing decisions (i.e. Choi et al., 2022; Gennaro and Ash, 2022)
- Look deeper at the underlying mechanisms
 - Measure exposure to propaganda through VK and/or Telegram channels activity in localities
 - Information about killed-in-action soldiers from the localities prior to the sentencing decisions

Appendix

References I

- Ozkan Eren and Naci Mocan. Emotional judges and unlucky juveniles. *American Economic Journal: Applied Economics*, 10(3): 171–205, 2018.
- Moses Shayo and Asaf Zussman. Judicial ingroup bias in the shadow of terrorism. *The Quarterly journal of economics*, 126(3): 1447–1484, 2011.
- Maja Adena, Ruben Enikolopov, Maria Petrova, Veronica Santarosa, and Ekaterina Zhuravskaya. Radio and the rise of the nazis in prewar germany. *The Quarterly Journal of Economics*, 130(4):1885–1939, 2015.
- Stefano DellaVigna, Ruben Enikolopov, Vera Mironova, Maria Petrova, and Ekaterina Zhuravskaya. Cross-border media and nationalism: Evidence from serbian radio in croatia. *American Economic Journal: Applied Economics*, 6(3):103–132, 2014.
- David Yanagizawa-Drott. Propaganda and conflict: Evidence from the rwandan genocide. *The Quarterly Journal of Economics*, 129(4):1947–1994, 2014.
- EAAS Report. 2022 Report on EEAS activities to counter FIMI, 2022. URL https://www.eeas.europa.eu/eeas/2022-report-eeas-activities-counter-fimi_en.
- David B Mustard. Racial, ethnic, and gender disparities in sentencing: Evidence from the us federal courts. *The Journal of Law and Economics*, 44(1):285–314, 2001.
- Donghyun Danny Choi, J Andrew Harris, and Fiona Shen-Bayh. Ethnic bias in judicial decision making: Evidence from criminal appeals in kenya. *American Political Science Review*, 116(3):1067–1080, 2022.
- Arnaud Philippe and Aurélie Ouss. “no hatred or malice, fear or affection”: Media and sentencing. *Journal of Political Economy*, 126(5):2134–2178, 2018.
- Maria Petrova and David Yanagizawa-Drott. Media persuasion, ethnic hatred, and mass violence. *Economic aspects of genocides, other mass atrocities, and their prevention*, 274, 2016.

References II

- Vadim Volkov. Legal and extralegal origins of sentencing disparities: Evidence from russia's criminal courts. *Journal of Empirical Legal Studies*, 13(4):637–665, 2016.
- Svetlana Zhuchkova and Anton Kazun. Exploring gender bias in homicide sentencing: An empirical study of russian court decisions using text mining. *Homicide Studies*, page 10887679231217159, 2023.
- Alex Knorre, Vladimir Kudryavtsev, Ekaterina Khodzhaeva, Kseniia Runova, and Kirill Titaev. The lull before the storm? criminal justice, crime and incarceration in russia (2000–2020). *Europe-Asia Studies*, 76(8):1229–1252, 2024.
- Zuzanna Brunarska and Wiktor Soral. Does origin matter? ethnic group position and attitudes toward immigrants: The case of russia. *Nationalities Papers*, 50(2):219–236, 2022.
- Alexey Bessudnov, Denis Tarasov, Viacheslav Panasovets, Veronica Kostenko, Ivan Smirnov, and Vladimir Uspenskiy. Predicting perceived ethnicity with data on personal names in russia. *Journal of Computational Social Science*, 6(2):589–608, 2023.
- Tamar Kricheli-Katz and Keren Weinshall. Judging fast or slow: The effects of reduced caseloads on gender-and ethnic-based disparities in case outcomes. *Journal of Empirical Legal Studies*, 20(4):961–1004, 2023.
- Chaoqun Zhan and Shitong Qiao. Workload, legal doctrine, and judicial review in an authoritarian regime: A study of expropriation judgments in china. *International Review of Law and Economics*, page 106232, 2024.
- Athina Anastasiadou, Artem Volgin, and Douglas R Leasure. War and mobility. *Demographic Research*, 50:205–220, 2024.
- Ruben Enikolopov, Alexey Makarin, and Maria Petrova. Social media and protest participation: Evidence from russia. *Econometrica*, 88(4):1479–1514, 2020.
- Gloria Gennaro and Elliott Ash. Emotion and reason in political language. *The Economic Journal*, 132(643):1037–1059, 2022.

Extraction of data from text

Regular expressions: used to extract most names, articles of law, and sentencing decisions

Natasha NLP library: used to extract names of defendants from court hearings data and cross-check regular expressions [More details](#)

Identification of ethnicity

Ethnicity is identified based on first name and surname of a person, using a predictive model by Bessudnov et al. (2023) [More details](#)

Ethnic Model

- We use Bessudnov et al. (2023) machine learning classifier to predict ethnicity based on personal names
- It is based on tokenization (division of words into part of different size), accuracy – 0.85
- Training set: data from popular social media website VK (www.vk.com), including names, gender, location, and languages a person speak
- Testing set: historical dataset with the names and ethnicity of the victims of political repressions from the Memorial society:
<https://lists.memo.ru/>

[Back to data analysis](#)

Natasha Library

- **Natasha** is an open-source python library for Russian natural language processing. It combines several packages, including Named-Entity Recognition models (Slovnet) and rule-based facts extractor (Yargy)
- To identify names, Natasha uses rule-based approach and accuracy is 0.95 when it is used on Russian names, and 0.89 on all names (including non-Russian names written in Russian)
- It used large corpus of Russian news (more than 700,000) from Lenta.ru agency to create rules and dictionaries
- More information is available on [Github](#)

[Back to data analysis](#)

Ethnic judge, Russian defendant		
	Whole sample	Randomized judges only
β_1 (After war)	0.053*** (0.018)	0.054*** (0.018)
N	8966	8561
Judges	322	308

Outcome: 0/1 detention. Notes: Includes court FE, $p(1)$, uniform kernel, data-driven bandwidth, robust standard errors in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Ethnic Judge, All Cases

[Back to results](#)

	(1)	(2)
β_1 (After war)	0.033** (0.014)	0.046*** (0.015)
β_1 x Ethnic defendant		-0.035** (0.014)
Ethnic defendant		0.022** (0.009)
N	14974	14974

Outcome: 0/1 detention. Notes: Includes court FE, p(1), uniform kernel, data-driven bandwidth, robust standard errors in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

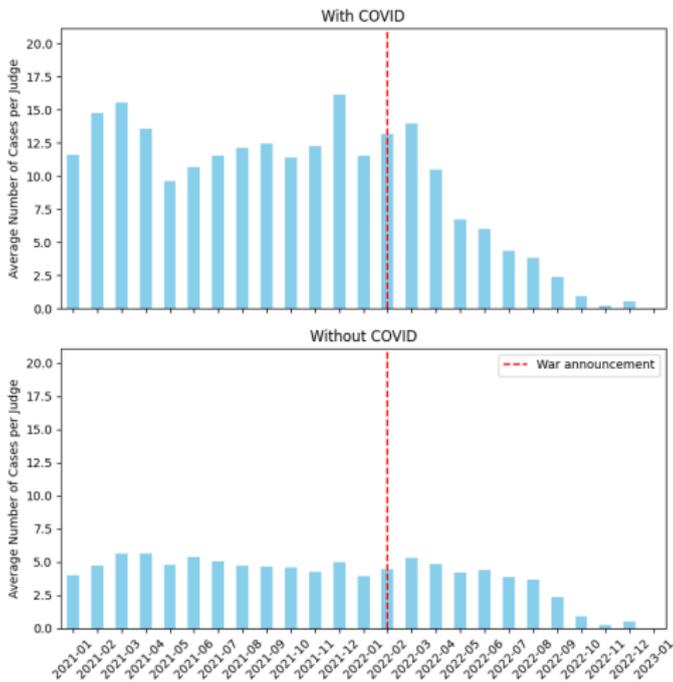
Balance table (minor offense) [Back to concerns](#)

Variable	(1) Before war	(2) After war	(3) Difference
Ethnic judge	0.379 (0.485)	0.373 (0.484)	-0.004 [0.004]
Ethnic defendant	0.316 (0.469)	0.321 (0.465)	-0.015 [0.010]
Text length	8,269.188 (3,007.621)	8,574.368 (3,061.639)	98.262*** [44.212]
Text length: ethnic defendant	2,734.281 (4,286.916)	2,730.283 (4,372.853)	13.990 [38.378]
Text length: Russian defendant	5,579.169 (4,602.214)	5,749.955 (4,653.245)	52.136 [40.671]
Text length: ethnic judges	3,194.249 (4,537.438)	3,177.256 (4,517.316)	-38.242 [37.768]
Text length: Russian judges	5,6578.456 (4,600.405)	5,789.206 (4,761.128)	-85.073 [95.102]
Text length: ethnic judge, Russian defendant	2,612.071 (3,916.479)	2,785.620 (3,882.002)	183.296** [90.956]
<i>N</i>	4,196	4,708	8,904

Distribution of all administrative cases

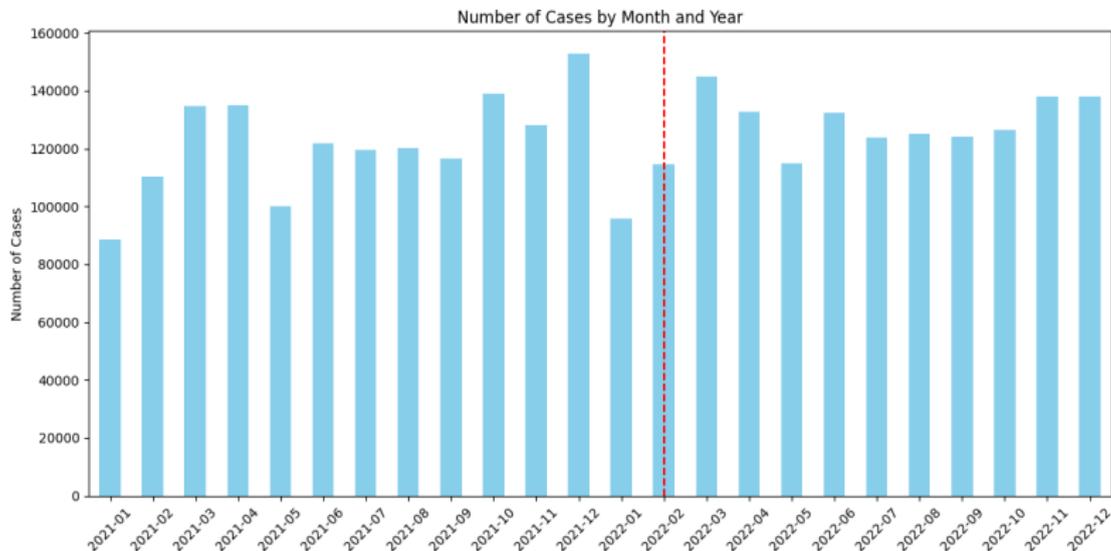
[Back to concerns](#)

In general, there are fewer administrative cases after war: This may be explained by alleviating COVID-related restrictions



Criminal cases are in line with seasonal trends

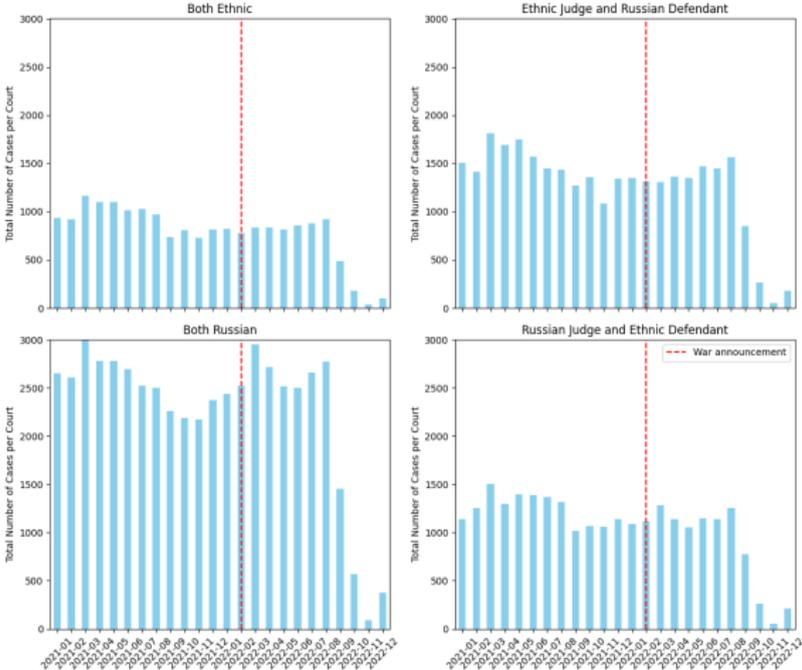
[Back to concerns](#)



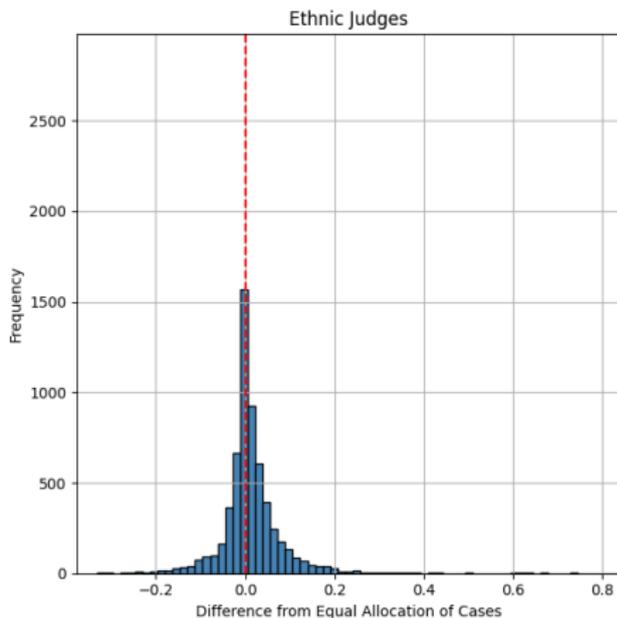
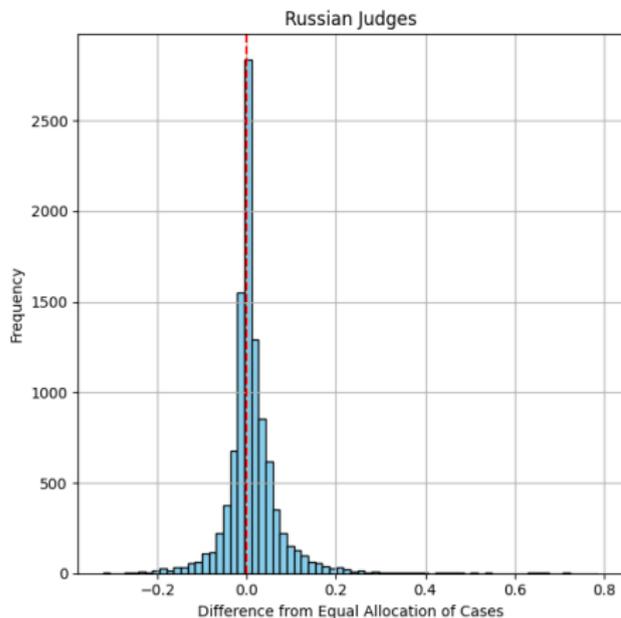
Distribution of minor offense cases

[Back to concerns](#)

Distribution of the cases among ethnic pairs for a specific article hasn't changed much before and after war



Caseload of judged is similar in both ethnic groups



Pre-war differences [Back to concerns](#)

Variable	(1) Russian Judge	(2) Ethnic Judge	(3) Difference
Share of ethnic def.	0.324 (0.468)	0.386 (0.487)	0.001 (0.004)
Prob. of detention	0.546 (0.498)	0.543 (0.498)	-0.006* (0.004)
Amount of fine	685.755 (225.244)	702.478 (235.211)	9.141*** (2.654)
Length of arrest	3.680 (3.088)	3.641 (3.171)	0.048 (0.035)
Prob. to issue high fine	0.463 (0.499)	0.467 (0.499)	0.003 (0.006)
<i>N</i>	44,575	28,678	73,253

Allocation of cases I

- Allocation of cases should be random and based on the automatization procedure
- We use Cramér's V to measure how strongly ethnicity of judge is associated with the ethnicity of defendant on the level of the court; and how different the judge is from other judges in the same court
 - Based on Pearson's chi-squared statistic
 - Varies from 0 (no association) to 1 (strong association)

[Back to concerns](#)

Allocation of cases II

	Ethnic Defendant	Russian Defendant	Undefined
Ethnic Judge	15	20	5
Russian Judge	10	25	3

Table 1: Level of court, 1 table per court

	Ethnic Defendant	Russian Defendant	Undefined
This judge	15	20	5
All other judges	10	25	3

Table 2: Level of judge, 1 table per court

Allocation of cases III ○

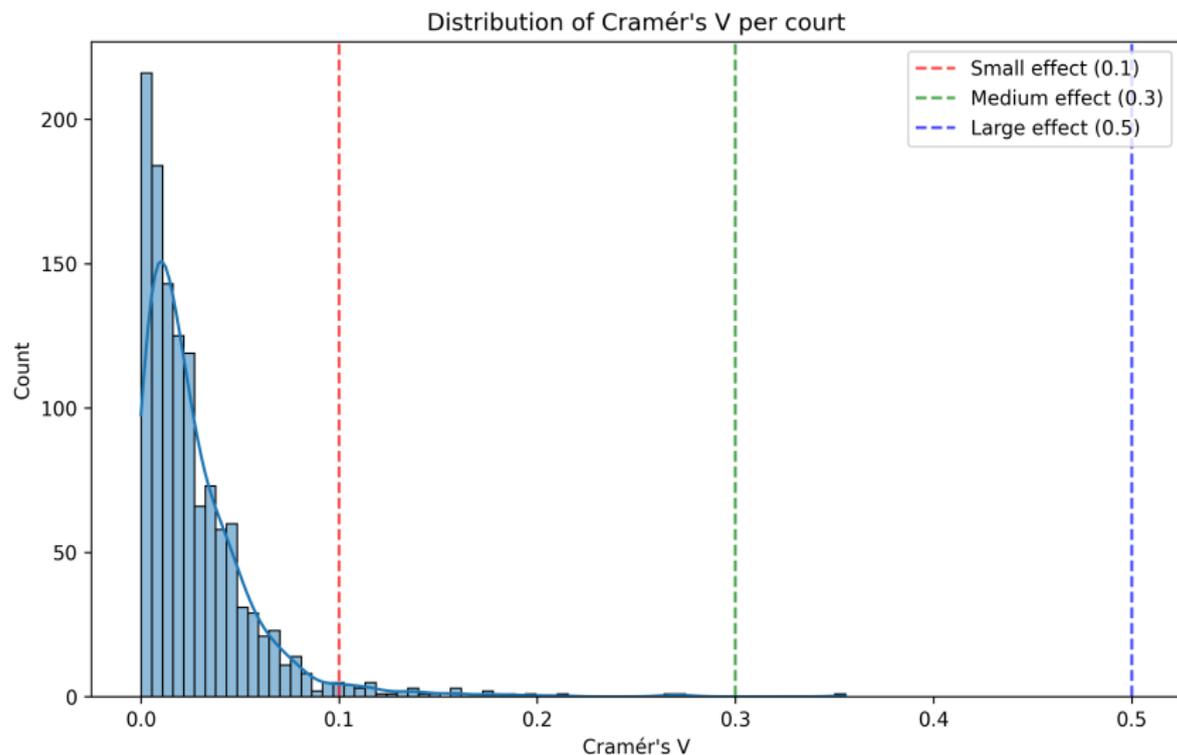


Figure 1: Distribution of Cramér's V across courts

Allocation of cases IV ○

[Back to concerns](#)

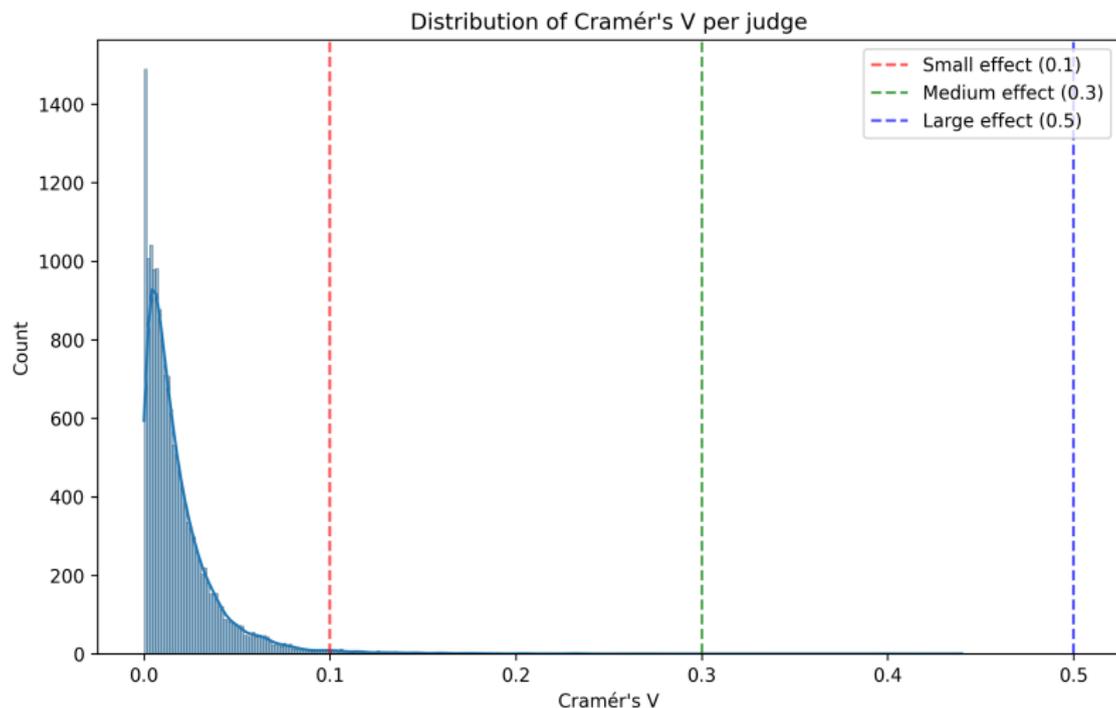


Figure 2: Distribution of Cramér's V across judges

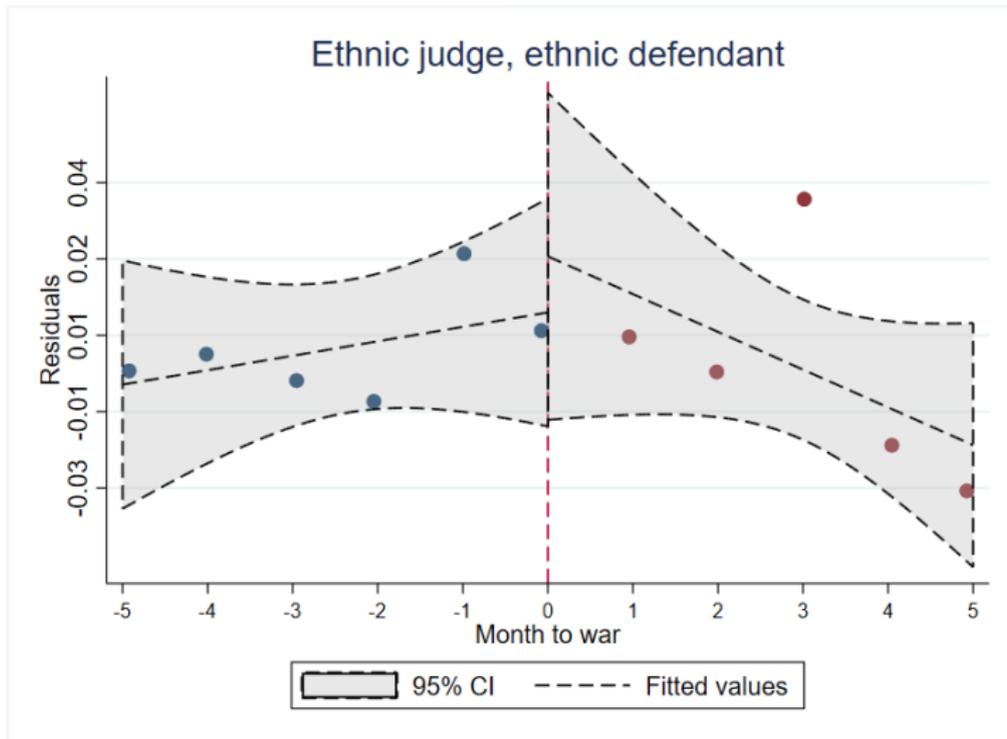
Robustness checks [Back to results](#)

	(1)	(2)	(3)	(4)
	Quadratic polynomial	Quadratic polynomial	Triangular kernel	Pseudo cutoff (6-months prior)
β_1 (After war)	0.048*	0.053**	0.041**	0.010
	(0.027)	(0.026)	(0.016)	(0.020)
<i>N</i>	8309	10541	7539	8568
Bandwidth (<i>h</i> , months)	[-3,6]	6	[-3,5]	4
Court FE	yes	yes	yes	yes
Polynomial order (<i>p</i>)	2	2	1	1
Kernel	uniform	uniform	triangular	uniform

Notes: Robust standard errors in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

Results Plot II

[Back to Results](#)



Minor offenses: Article 20.1 Part 1 [Back to Strategy](#)

Minor offense, that is, a violation of public order expressing clear disrespect to society, accompanied by obscene language in public places, offensive harassment of citizens, as well as destruction or damage to others' property

- Administrative fine from 500 to 1,000 RUB (5-10 USD)
- Administrative arrest (detention) from 1 to 15 days
- Examples: use of obscene language in public places, minor destruction or damage to others' property