

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

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# 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

$\beta_1$ (After war)	Ethnic defendant	Russian defendant
<b>Ethnic judge</b> 	0.034 (0.025) $N = 4327$	<b>0.053***</b> <b>(0.017)</b> $N = 8966$
Russian judge	-0.005 (0.022) $N = 5819$	-0.006 (0.012) $N = 17615$

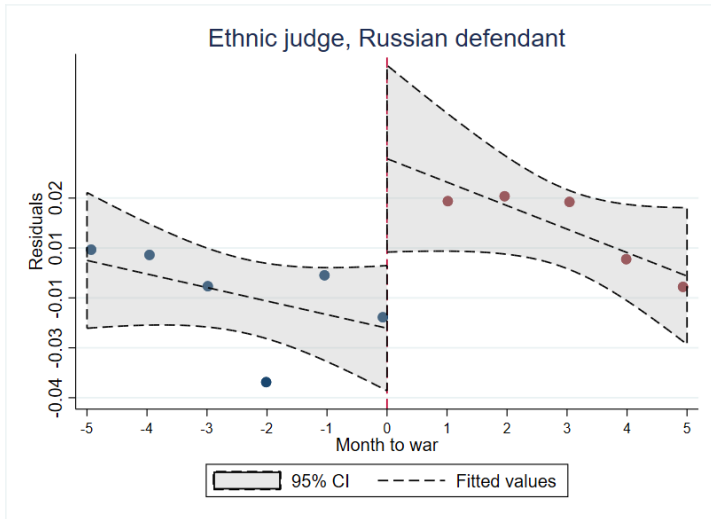
**Outcome: 0/1 detention.** Notes: Includes court FE,  $p(1)$ , uniform kernel, data-driven bandwidth, robust standard errors in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

# 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
$\beta_1$ (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

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## 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](https://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/>

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# 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](#)

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	Ethnic judge, Russian defendant	
	Whole sample	Randomized judges only
$\beta_1$ (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

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	(1)	(2)
$\beta_1$ (After war)	0.033** (0.014)	0.046*** (0.015)
$\beta_1$ 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. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

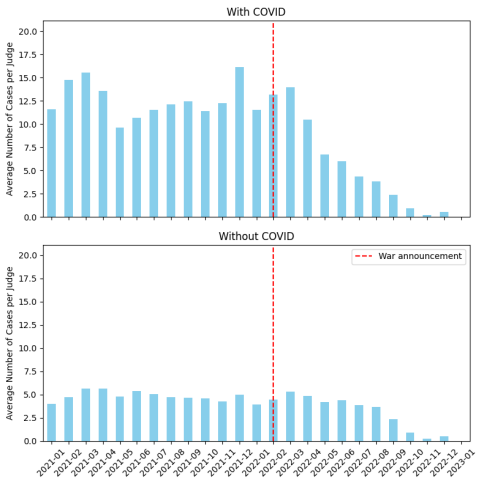
# Balance table (minor offense) [Back to concerns](#)

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

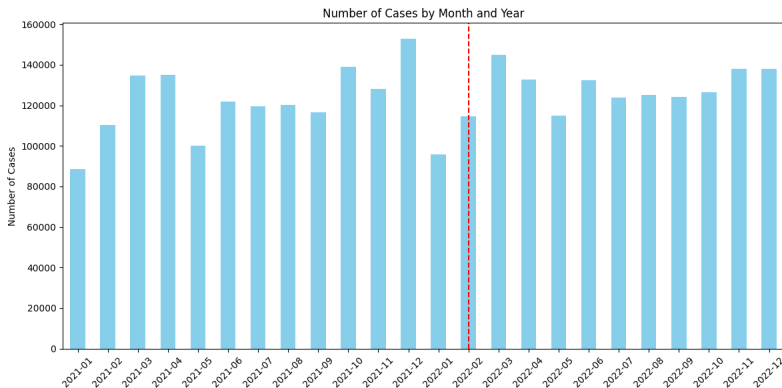
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In general, there are fewer administrative cases after war: This may be explained by alleviating COVID-related restrictions





# Criminal cases are in line with seasonal trends

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third-level administrative divisions.

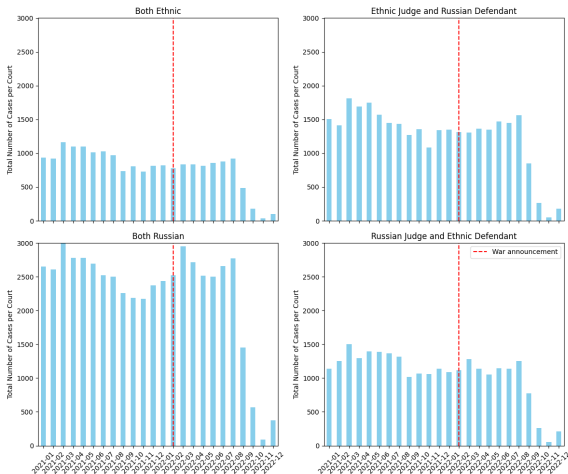
based on the census of 2010



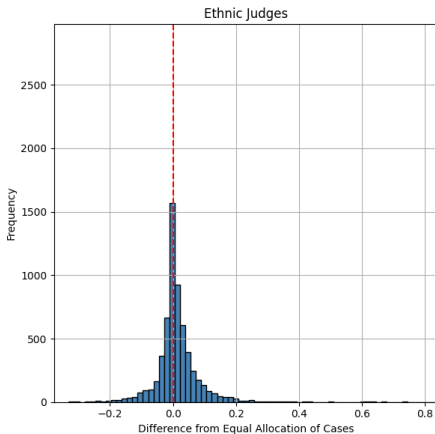
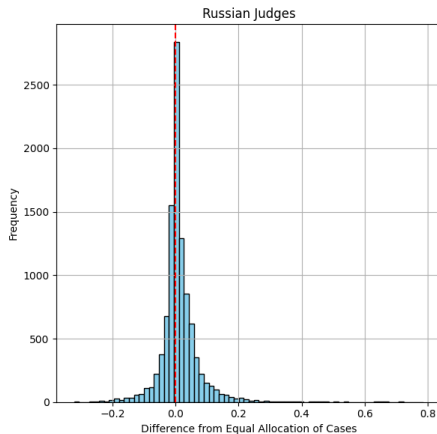
# Distribution of minor offense cases

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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](#)

	(1)	(2)	(3)
Variable	Russian Judge	Ethnic Judge	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)

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## Allocation of cases II

	<b>Ethnic Defendant</b>	<b>Russian Defendant</b>	<b>Undefined</b>
<b>Ethnic Judge</b>	15	20	5
<b>Russian Judge</b>	10	25	3

*Table 1: Level of court, 1 table per court*

	<b>Ethnic Defendant</b>	<b>Russian Defendant</b>	<b>Undefined</b>
<b>This judge</b>	15	20	5
<b>All other judges</b>	10	25	3

*Table 2: Level of judge, 1 table per court*

# Allocation of cases III ○

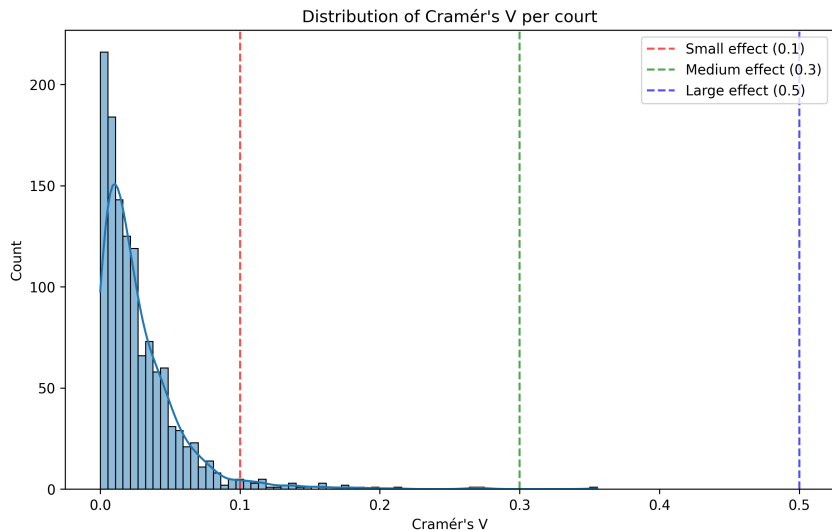


Figure 1: Distribution of Cramér's V across courts



# Allocation of cases IV ○

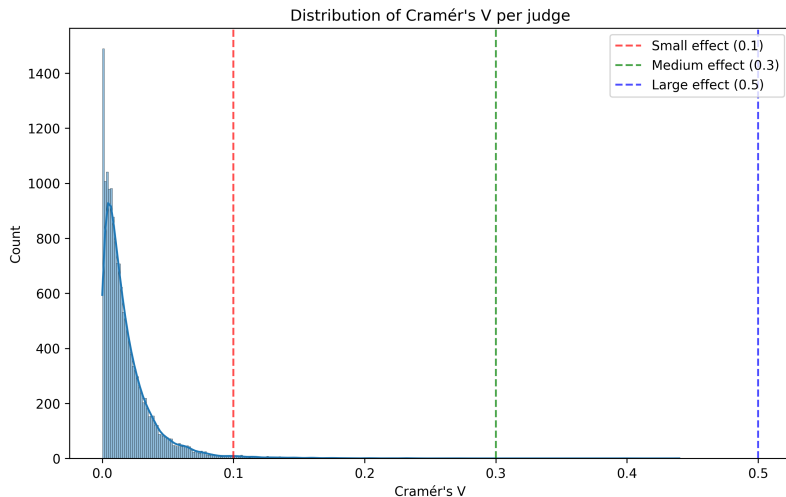
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Figure 2: Distribution of Cramér's V across judges

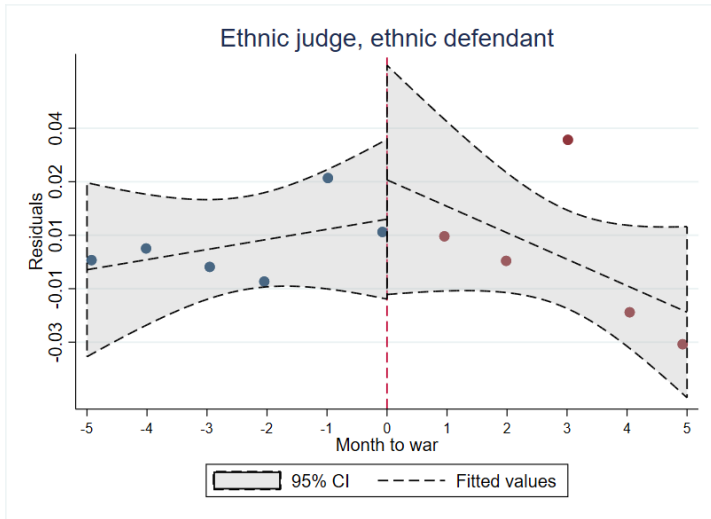
# Robustness checks

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	(1)	(2)	(3)	(4)
	Quadratic polynomial	Quadratic polynomial	Triangular kernel	Pseudo cutoff (6-months prior)
$\beta_1$ (After war)	0.048*	0.053**	0.041**	0.010
	(0.027)	(0.026)	(0.016)	(0.020)
$N$	8309	10541	7539	8568
Bandwidth ( $h$ , months)	[-3,6]	6	[-3,5]	4
Court FE	yes	yes	yes	yes
Polynomial order ( $p$ )	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

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## 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