

# Violence Motivated Migration: A cross-national analysis of the role of nonwartime violence in migration decision-making

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# Section 1

## Motivating Questions

# How does violence affect migration?

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# How does violence affect migration?

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- Yet, violence is a multi-faceted social phenomenon (Collins 2008) and has **features that exist along a spectrum that spans peace and war-times**
- Nonwartime violence is overlooked, so the full relationship eludes us

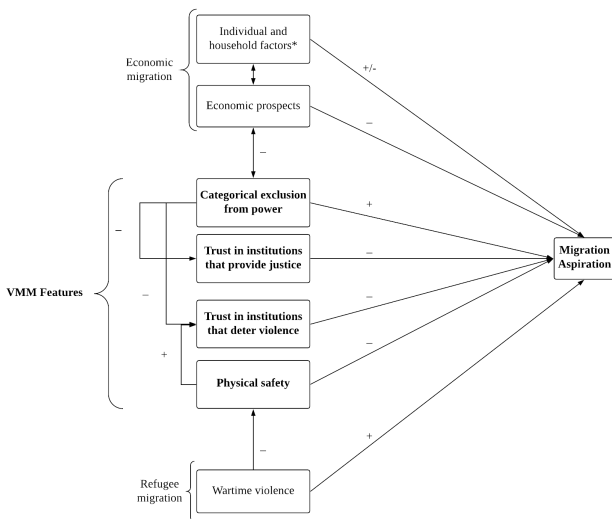
## Section 2

# A New Conceptualization: Violence Motivated Migration

# Aims, Assumptions, and Scope Conditions

- Conceptualization designed to incorporate violence into migration theory
- Types of migration: internal and international, out of one neighborhood into another
- Types of violence: systematic, wartime and nonwartime (restriction: *not* interpersonal)
  - ▶ e.g. hate crimes





## Section 3

# Proof of Concept

# Analytical approach

- First question: Do the specified features of violence correlate with migration aspirations?
  - ▶ If not, basic fundamental premise of conceptualization flawed
- Second question: Are the results driven by war or armed conflict?
  - ▶ If so, VMM is not more useful than current siloed research on refugees or economic migrants
  - ▶ If not, VMM is a valid conceptualization that should be further pursued

# Data

- Gallup World Poll
- International Labour Organisation
- Uppsala Conflict Data Program

# Question 1: Migration Aspirations

- Gallup World Poll (GWP)
- "How likely are you to move away or stay?"
- Over 844,000 valid observations over 160 countries, 2005-2019
- 14.2% say they are likely to move away

# VMM Subjective Perceptions of Violence

- Categorical Exclusion from Power
  - ▶ GWP: Is your community a good place for minorities?
- Perceived risk of violence
  - ▶ GWP: Safe walking alone at night in your community?
- Institutional trust in the mechanisms that deter violence
  - ▶ GWP: Trust local police in your community?
- Institutional trust in the mechanisms that pursue justice
  - ▶ GWP: Trust courts and judicial system?

# Controls

- Subjective economic prospects (GWP)
- Objective economic prospects (unemployment rate + change in annual unemployment rate)
- Individual-level factors
  - ▶ Age
  - ▶ Gender
  - ▶ Marital status
  - ▶ Survey year

## Question 2: Wartime and Armed Conflict

- Uppsala Conflict Geocoded Event Dataset (GED)
- Attached to GWP observations at the country-year level
- War = 1000+ battle deaths in survey year
- Armed conflict = 25+ battle deaths in survey year



# Progressively complex multilevel models

Multilevel models with migration aspirations as DV

Individuals nested within countries (89 obs.) within country-years (510 obs.)

- Model 1 = individual + economic
- Model 2 = Model 1 + safety + police + courts + minorities
- Model 3 = Model 2 + Armed Conflict
- Model 4 = Model 2 + War

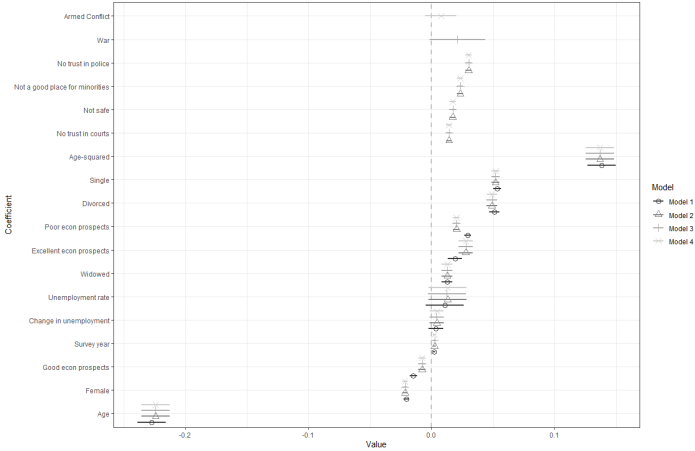
# Section 4

## Findings

## Summary: Broad support for VMM framework

- Perception of community as bad place for minorities associated with a higher likelihood of migration aspirations ( $\beta = .023$ )
  - ▶ **Supports VMM 'Categorical Exclusion' feature**
- Those who do not feel safe are more likely to say they are likely to move away ( $\beta = .017$ )
  - ▶ **Supports VMM 'Safety' feature**
- Those who do not trust the police ( $\beta = .030$ ) and those who do not trust the courts ( $\beta = .014$ ) are more likely to have migration aspirations.
  - ▶ **Supports VMM 'Institutional Trust' features**

# Does war drive the results?



# Conclusion: A VMM Research Agenda

- Subjective perceptions of violence an important but overlooked aspect of migration aspirations
- Proof of concept: Enough evidence to continue pursuing line of inquiry
  - ▶ Case studies to unpack mechanisms
    - ★ e.g. Latin American countries - lots of migration, lots of violence, most countries technically not "at war" for long stretches of time
  - ▶ Examine the role of police brutality as a push factor in U.S. context
    - ★ Draw on Tolnay and Beck's (1992) work in understanding the role of lynching in the Great Migration
  - ▶ Police brutality vs. Structural failures
  - ▶ Deeper dive into Categorical Exclusion

Thank You!

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# Appendix: Distribution of risk and trust

**Table 2. Distribution of risk and institutional trust in the Gallup World Poll sample**

## Trust in Local Police

		<b>Yes</b>	<b>No</b>	<b>Don't know</b>
<b>Safety (Risk)</b>	<b>Yes</b>	(.452)	(.116)	(.024)
	<b>No</b>	(.159)	(.206)	(.017)
	<b>Don't know</b>	(.014)	(.006)	(.006)

## Trust in Courts

		<b>Yes</b>	<b>No</b>	<b>Don't know</b>
<b>Safety (Risk)</b>	<b>Yes</b>	(.330)	(.220)	(.044)
	<b>No</b>	(.142)	(.212)	(.027)
	<b>Don't know</b>	(.009)	(.010)	(.006)

# Appendix: VMM coefficients Part I

**Table 4. Results from logistic regression models, dependent variable = migration aspirations**

	<i>Base Model</i>	<i>VMM Model</i>	<i>VMM Model with MAR score</i>
	Coeff (Std. Error)	Coeff (Std. Error)	Coeff (Std. Error)
Categorical Exclusion from Power			.168*** (.008)
Safety (ref = Yes)			
Don't Know		-.079* (.033)	-.122** (.038)
No		.216*** (.009)	.198*** (.010)
Trust in Police (ref = Yes)			
Don't Know		.090*** (.022)	.035 (.026)
No		.297*** (.009)	.281*** (.010)
Trust in Courts			
Don't Know		-.171*** (.018)	-.192*** (.020)
No		.073*** (.009)	.058*** (.009)



# Appendix: Controls (VMM Models Part II)

Economic Conditions (ref = Excellent)	<i>Base Model</i>	<i>VMM Model</i>	<i>VMM Model with MAR score</i>
Good	-.173*** (.025)	-.318*** (.022)	-.301*** (.024)
Fair	.063** (.023)	-.205*** (.021)	-.199*** (.023)
Poor	.361*** (.024)	-.015 (.021)	-.0002 (.023)
Year	.009*** (.002)	.025*** (.002)	.022*** (.002)
Age	.005*** (.001)	-.004*** (.001)	-.003* (.001)
Age squared	-.0006*** (.00002)	-.0004*** (.001)	-.0004*** (.00002)
Male	.130*** (.010)	.153*** (.008)	.152*** (.009)
Marital Status (ref = married)			
Single	.498*** (.013)	.358*** (.010)	.372*** (.011)
Separated/Divorced	.301*** (.025)	.314*** (.017)	.302*** (.019)
Widowed	.009 (.030)	.109*** (.023)	.065* (.026)
Intercept	-55.95*** (2.81)	-50.93*** (3.07)	-46.54*** (3.36)
N =	574,658	483,975	388,444
AIC	490669	408090	329023

$\alpha < .001$  \*\*\*\*  $\alpha < 0.001$  \*\*\*  $\alpha < 0.01$  \*\*  $\alpha < 0.05$  \*  $\alpha < 0.1$

# Appendix: VMM Coefficients with Wartime Variables Part I

**Table 5. Results from logistic regression models including wartime measures**  
**dependent variable = migration aspirations**

	<i>VMM Model with MAR score (from Table 4)</i>	<i>Wartime</i>	<i>Wartime (t-1)</i>	<i>Armed Conflict</i>	<i>Armed Conflict (t-1)</i>
	Coeff. (Std. Error)	Coeff. (Std. Error)	Coeff. (Std. Error)	Coeff. (Std. Error)	Coeff. (Std. Error)
Wartime		.237*** (.015)			
Wartime in previous year (t-1)			.275*** (.015)		
Armed conflict				.191*** (.010)	
Armed conflict in previous year (t-1)					.142*** (.010)
Categorical Exclusion from Power	.168*** (.008)	.174*** (.008)	.173*** (.008)	.163*** (.008)	.154*** (.008)
Safety (ref = Yes)					
Don't Know	-.122** (.038)	-.126*** (.038)	-.128*** (.038)	-.116** (.038)	-.117** (.038)
No	.198*** (.010)	.193*** (.010)	.192*** (.010)	.197*** (.010)	.196*** (.010)
Trust in Police (ref = Yes)					
Don't Know	.035 (.026)	.040 (.026)	.039 (.026)	.044^ (.026)	.039 (.026)
No	.281*** (.010)	.281*** (.010)	.281*** (.010)	.281*** (.010)	.281*** (.010)
Trust in Courts					
Don't Know	-.192*** (.020)	-.190*** (.020)	-.190*** (.020)	-.182*** (.020)	-.187*** (.020)
No	.058*** (.009)	.057*** (.010)	.058*** (.010)	.066*** (.010)	.064*** (.010)

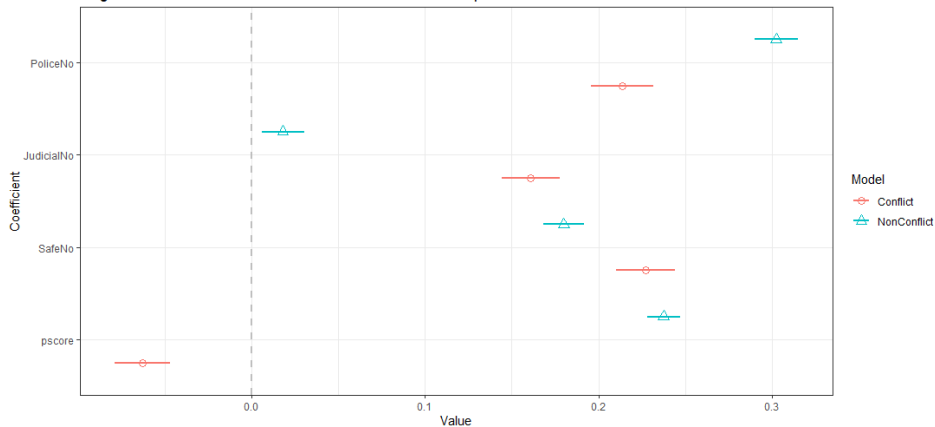
# Appendix: Controls in VMM Wartime models (VMM wartime models Part II)

Economic Conditions (ref = Excellent)	<i>VMM Model with MAR score (from Table 4)</i>	<i>Wartime</i>	<i>Wartime (t-1)</i>	<i>Armed Conflict</i>	<i>Armed Conflict (t-1)</i>
Good	-.301*** (.024)	-.306*** (.024)	-.304*** (.024)	-.290*** (.024)	-.296*** (.023)
Fair	-.199*** (.023)	-.199*** (.023)	-.197*** (.023)	-.179*** (.023)	-.189*** (.023)
Poor	-.002 (.023)	-.001 (.024)	-.0003 (.023)	.022 (.023)	.011 (.023)
Year	.022*** (.002)	.024*** (.002)	.024*** (.002)	.023*** (.002)	.024*** (.002)
Age	-.003* (.001)	-.002* (.001)	-.002* (.001)	-.002^ (.001)	-.002^ (.001)
Age squared	-.0004*** (.00002)	-.0004*** (.00002)	-.0004*** (.00002)	-.0004*** (.00001)	-.0004*** (.00002)
Male	.152*** (.009)	.146*** (.009)	.144*** (.009)	.142*** (.009)	.145*** (.009)
Marital Status (ref = married)					
Single	.372*** (.011)	.382*** (.011)	.383*** (.011)	.392*** (.012)	.387*** (.012)
Separated/Divorced	.302*** (.019)	.310*** (.019)	.312*** (.019)	.326*** (.019)	.319*** (.019)
Widowed	.065* (.026)	.064* (.026)	.063* (.026)	.060* (.026)	.062* (.026)
Intercept	-46.54*** (3.36)	-49.84*** (3.37)	-49.09*** (3.37)	-48.18*** (3.36)	-49.22*** (3.37)
N =	388,444	388,443	388,443	388,443	388,443
AIC	329023	328790	328721	328631	328812

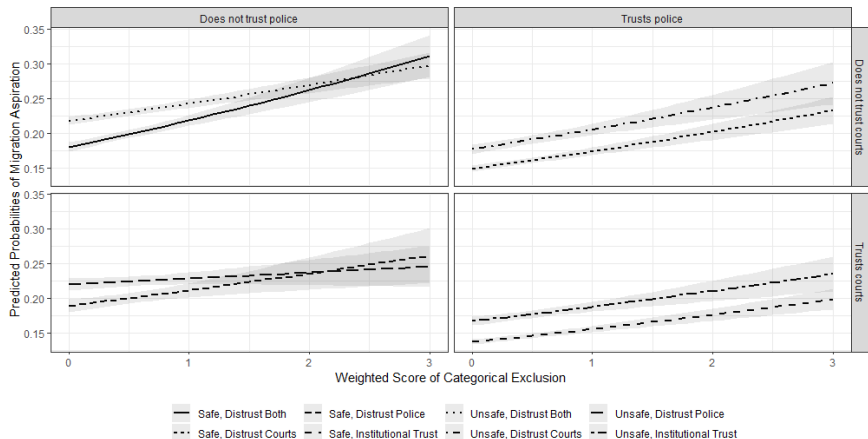
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# Appendix: Does armed conflict drive the results?

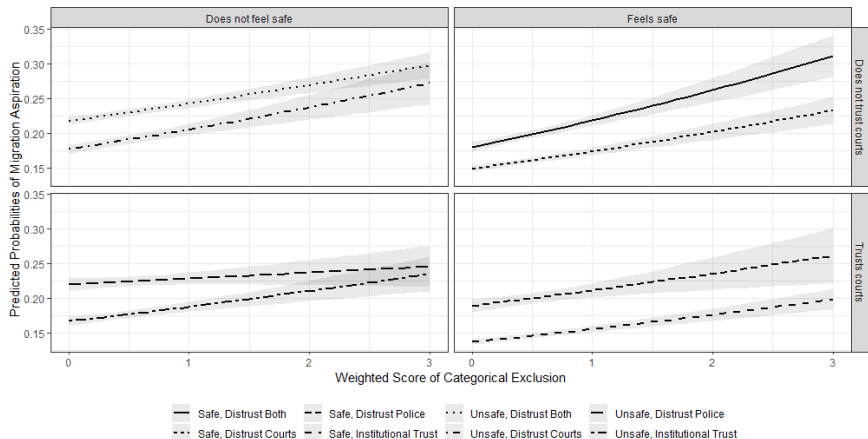
Regression Coefficients for Conflict and NonConflict Samples



# Appendix: Four-way interactions; effects of safety



# Appendix: Four-way interactions; effects of police



# Appendix: Four-way interactions; effects of courts

