

Bournemouth

An intersectional analysis of stranger, acquaintance, and domestic violence victimisation in England and Wales using MAIHDA

Dr Ferhat Tura & Dr Jane Healy Bournemouth University Dr Clare R. Evans, University of Oregon Dr George Leckie, University of Bristol Contact:<u>ftura@bournemouth.ac.uk</u>



Background & Evidence Gap

- Violence victimisation peaked in 1995 (4.7%) and declined to 0.8% in 2022 (ONS, 2022).
- Certain marginalised groups bear disproportionate victimisation risks.
- Victims face emotional, physical, and social impacts.
- Additionally...
- Limited quantitative intersectional studies on violence victimisation.
- Prior focus on individual-level risk factors with interaction terms and intimate partner violence.
- Lack of comprehensive analysis of diverse violence types stranger, acquaintance, and domestic.





Theoretical framework

- Intersectionality/Framework examining the interplay of identities and systems of oppression (Crenshaw, 1989, 1991).
- Incorporates gender, ethnicity, age, SEC, and disability.
- Challenges additive approaches, highlighting systemic inequalities.





Research Process

- Objectives
- Map intersectional disparities in victimisation risks by intersectional strata.
- Analyse violence victimisation by type (stranger, acquaintance, domestic).
- Inform targeted interventions to address systemic inequalities in violence victimisation

Data Source

- Crime Survey for England and Wales (CSEW):
 - Years: 2015/16 to 2019/20
 - Sample: 165,661 respondents aged 16+ in private households.
 - Captures self-reported victimisation experiences in 12-month period.

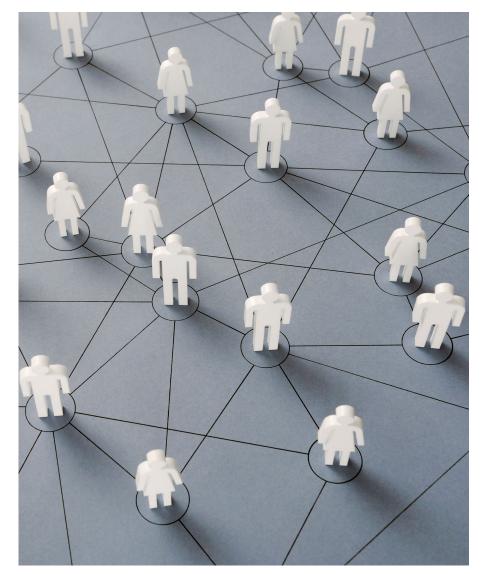


- MAIHDA(Multilevel Analysis of Individual Heterogeneity and Discriminatory Accuracy):
 - From a multilevel modelling perspective, the individual outcome is regressed on individual exposures in the usual way. The key novelty is that intersectional MAIHDA treats individuals (level1) as nested within intersectional strata (level-2) formed by combinations of the key individual sociodemographic characterises of research focus. The regression intercept and slope coefficients are then modelled as varying across these strata (Evans et al., 2024; University of Bristol website).
- Strata defined by sex, ethnicity, age, SEC, and disability.



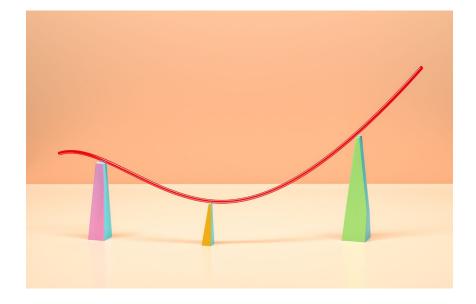
Social Strata

- Five Variables
 - Sex (Male/Female)
 - Ethnicity (White, Asian, Black, Mixed/Chinese/Other)
 - Age (1624, 25-44, 45+)
 - Socioeconomic Class (Managerial, Intermediate, Routine, Never worked/Long term unemployed)
 - Long-standing disability (No/Yes)
- 191 (2*4*3*4*2 = 192) unique strata analysed.





Outcome Variables



- Binary measures for three violence types:
 - Stranger Violence
 - Acquaintance Violence
 - Domestic Violence



Sample Characteristics

Strata variables (re	fence category: ref)	ID	Frequenc	Percenta
		coding	У	ge
Gender	Man (ref)	1	76,235	46.0
	Woman	2	89,426	54.0
Ethnicity	White (ref)	1	149,453	90.2
	Asian or Asian British	2	8,261	5.0
	Black or Black British	3	4,311	2.6
	Mixed or Chinese/Other	4	3,636	2.2
Age	16-24 (ref)	1	7,417	4.5
	25-44	2	51,765	31.2
	45 plus	3	106,479	64.3
SEC	Higher managerial (ref)	1	61,719	37.3
	Intermediate occupations	2	39,837	24.0
	Routine and manual	3	58,315	35.2
	occupations Never worked and long-term	4	E 700	3.5
	unemployed	4	5,790	3.5
Disability	No long-standing illness (ref)	1	127,425	76.9
	Long-standing illness	2	38,236	23.1
Outcome v				
Stranger violence	No	-	164,540	99.3
	Yes	-	1,121	0.7
Acquaintance violence	No	-	164,712	99.4
	Yes	-	949	0.6
Domestic violence	No	-	165,131	99.7
	Yes	-	530	0.3



Model Results

	Stranger null		Stranger full		Acquaintance null		Acquaintance full		Domestic null		Domestic full	
	OR	95% CI	OR	95% CI	OR	95% CI	OR	95% CI	OR	95% CI	OR	95% CI
Main effects: Regression coefficients												
(Intercept)	0.01 ***	0.01 – 0.01	0.04 ***	0.03-0.05	0.01 ***	0.01 – 0.01	0.02 ***	0.01 – 0.02	0.00 ***	0.00-0.00	0.00 ***	0.00-0.01
Female (ref: Male)			0.37 ***	0.32-0.43			0.71 ***	0.59 – 0.85			2.27 ***	1.78 – 2.90
Asian (ref: White)			0.71 *	0.52 - 0.97			0.47 ***	0.31 – 0.71			0.45 **	0.26 - 0.79
Black			1.09	0.76 – 1.57			1.03	0.69 – 1.54			0.8	0.45 - 1.42
Mixed, Chinese, Other			1.25	0.89 – 1.76			0.82	0.52 – 1.28			0.71	0.38 – 1.32
Age, 25-44 (ref: Age, 16-24)			0.41 ***	0.33 – 0.51			0.48 ***	0.37 – 0.62			0.55 ***	0.40 - 0.76
Age, 45 plus			0.17 ***	0.14-0.21			0.19 ***	0.15 – 0.25			0.16 ***	0.11 – 0.22
SEC, Intermediate (ref: Managerial)			0.95	0.78 – 1.14			1.03	0.80 – 1.34			0.95	0.69 - 1.30
SEC, Routine			0.79 **	0.66 - 0.94			1.23	0.98 – 1.55			1.1	0.82 – 1.47
SEC, Never worked			0.63 *	0.44 - 0.92			1.12	0.78 – 1.61			1.03	0.66 - 1.62
Disability, Long standing (ref: No long standing)			1.64 ***	1.39 – 1.95			2.16 ***	1.77 – 2.64			3.04 ***	2.40 - 3.86
Random effects: Variances												
Stratum-level	0.8		0.02		0.69		0.05		1.13		0.05	
Individual-level~	3.29		3.29		3.29		3.29		3.29		3.29	
Summary statistics												
VPC	20%		1%		17%		2%		26%		2%	
PCV			98%				93%				96%	
AIC	12954		12832		11445		11346		6832		6739	
AUC	0.713		0.699		0.677		0.667		0.753		0.738	
LRT	$\chi^{2}(1) = 485.21 * * *$		χ ² (9) = 139.96***		χ ² (1) = 249.85***		χ ² (9) = 102.34***		$\chi^{2}(1) = 319.72^{***}$		χ ² (9) = 102.74***	
N stratum	191		191		191		191		191		191	
N individuals	165,661		165,661		165,661		165,661		165,661		165,661	



Key Findings

Stranger Violence

- High-risk groups: Young, disabled men
- VPC: 20%

 variance
 explained by
 strata in the null
 model but
 reduced to 1%
 when main
 effects added

Acquaintance Violence

- High-risk groups: Young, disabled men
- VPC: 17% variance explained by strata in the null model but reduced to 2% when main effects added

Domestic Violence

- High-risk groups: Young, disabled women
- VPC: 26%
 variance
 explained by
 strata in the null
 model but
 reduced to 2%
 when main
 effects added

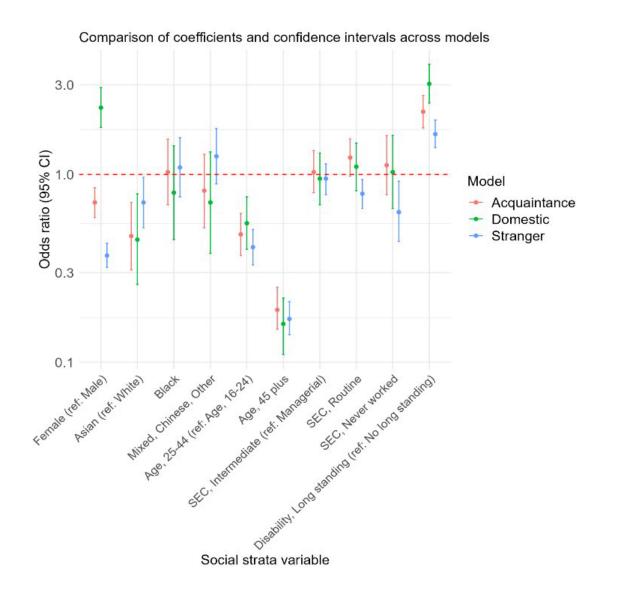


Interaction Effects

- Interaction effects accounted for 2%7% unexplained variance.
- Highlights need for tailored policy interventions.



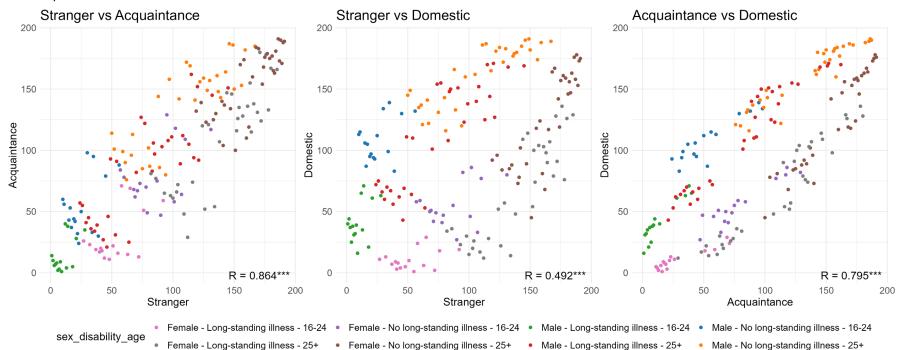
Visual Insights





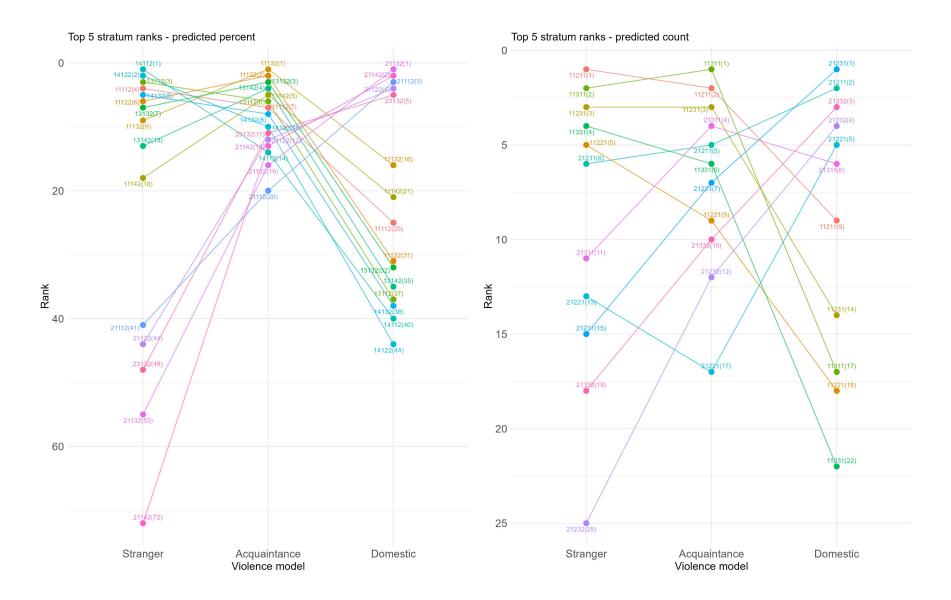
Visual Insights

Scatterplot matrix of stratum ranks across models





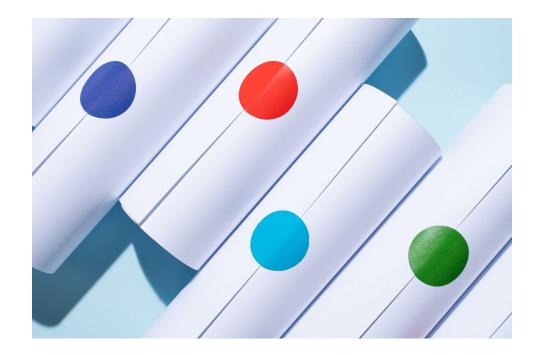
Visual Insights





Policy implications

- Target interventions for:
 - Young, disabled men: Stranger/acquaintance violence.
 - Young, disabled women: Domestic violence.
- Prioritise training for law enforcement and social services.





- Binary gender variable; lacks insights into non-binary individuals.
- Disability types not differentiated.
- Underrepresentation of care home populations.





Future Research Directions

- Include gender minorities and specific disability types.
- Explore underreporting patterns by demographic strata.
- Assess longitudinal intervention impacts.







- Demonstrates systemic and intersectional inequalities.
- Highlights young, disabled individuals as highly vulnerable groups.
- Calls for intersectionality-driven policymaking and resource allocation.





ANY QUESTIONS



- Crenshaw, K. 1989). Demarginalizing the Intersection of Race and Sex: A Black Feminist Critique of Antidiscrimination Doctrine, Feminist Theory and Antiracist Politics. University of Chicago Legal Forum, 1(8): 139-67.
- Crenshaw, K. (1991). Mapping the Margins: Intersectionality, Identity Politics, and Violence Against Women of Color. Stanford Law Review, 43(6): 1241-1299.
- Evans, C.R., Leckie, G., Subramanian, S.V., Bell, A. and Merlo, J., 2024. A tutorial for conducting intersectional multilevel analysis of individual heterogeneity and discriminatory accuracy (MAIHDA). SSM-population health
- Office for National Statistics (2022). The Nature of Violent Crime in England and Wales: Year Ending March 2022. Available at:

https://www.ons.gov.uk/peoplepopulationandcommunity/crimeandjustice/articles/thenatureofviolentcrime inenglandandwales/yearendingmarch2022

- Tura, F., Healy, J., Evans, C.R. and Leckie, G., 2024. An Intersectional Analysis of Stranger, Acquaintance, and Domestic Violence Victimisation in England and Wales using MAIHDA. CrimRxiv. Available at: https://www.crimrxiv.com/pub/ndr3vykw/release/1
- University of Bristol website: Intersectional MAIHDA | Centre for Multilevel Modelling | University of
 Bristol