

Are Pretrained Multilingual Models Equally Fair Across Languages?

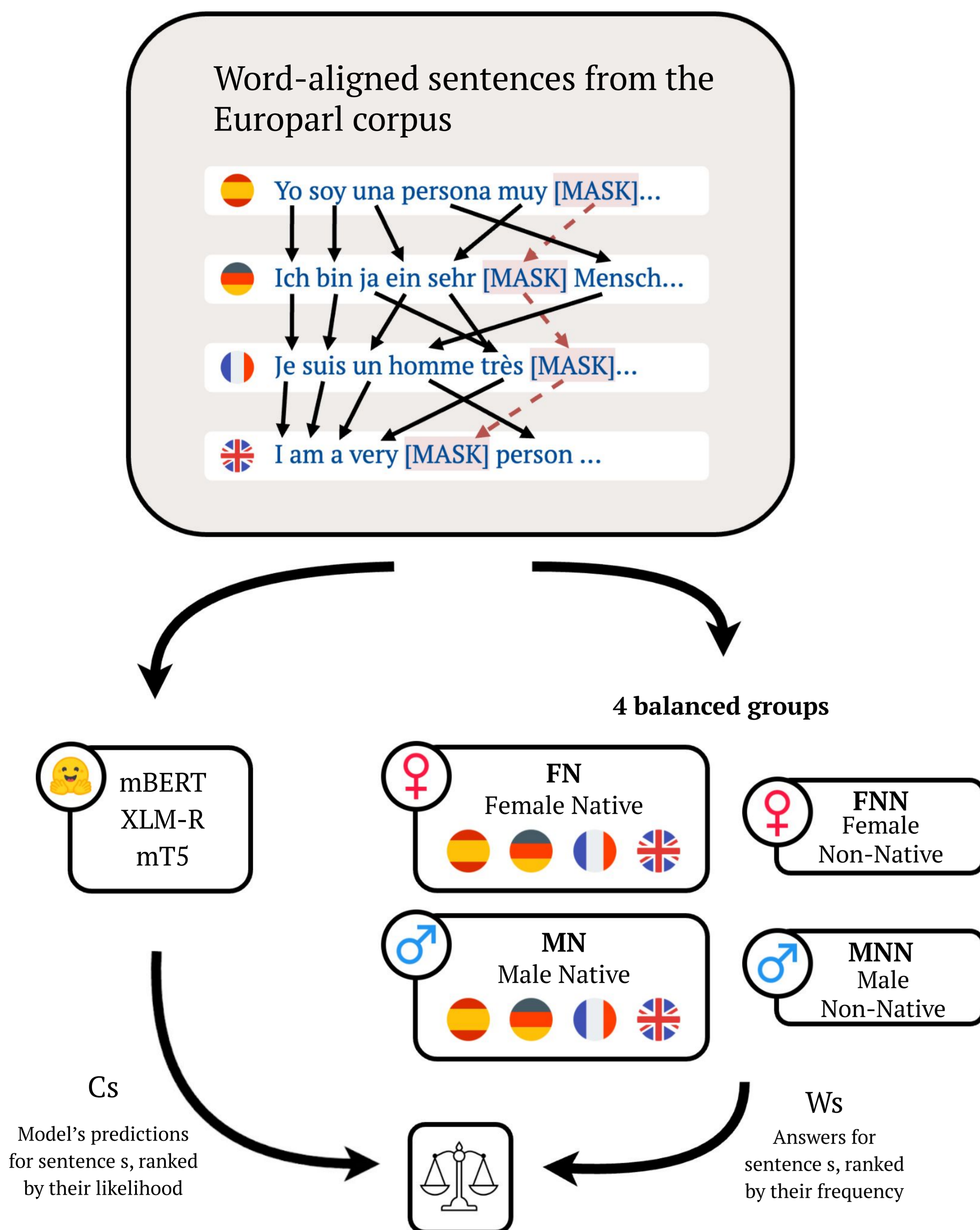
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Main contribution

MozArt: A new multilingual dataset annotated with balanced demographics. github.com/coastalcp/mozart

Method



Metrics

Fairness metric: group disparity
(group-level performance differences)

$$\sigma_{gd} = \sqrt{\frac{\sum_{j=1}^G (P@k_j - \overline{P@k})^2}{G}}$$

$P@k = \mathbb{1}[c_i \in W_s]$
 $G = \text{number of groups}$

Mean Reciprocal Rank

$$MRR = \frac{1}{|S|} \sum_{s=1}^{|S|} \frac{1}{Rank_i^{C_s^n}}$$

$|S| = 100 \text{ sentences}$

Rank Correlations

Spearman's rho, Kendall's tau (see paper for results)

MozArt details

	EN	ES	DE	FR
WordPiece (avg. #tokens)	19.7	22.0	23.6	23.1
SentencePiece (avg. #tokens)	22.3	22.9	24.9	25.3
#Sentences	100	100	100	100
#Annotations	600	600	600	600
#Annotators	60	60	60	60
Demographics	id_u, id_s, gender, age, nationality, first language, fluent languages, current country of residence, country of birth, time taken			

Experimental Results

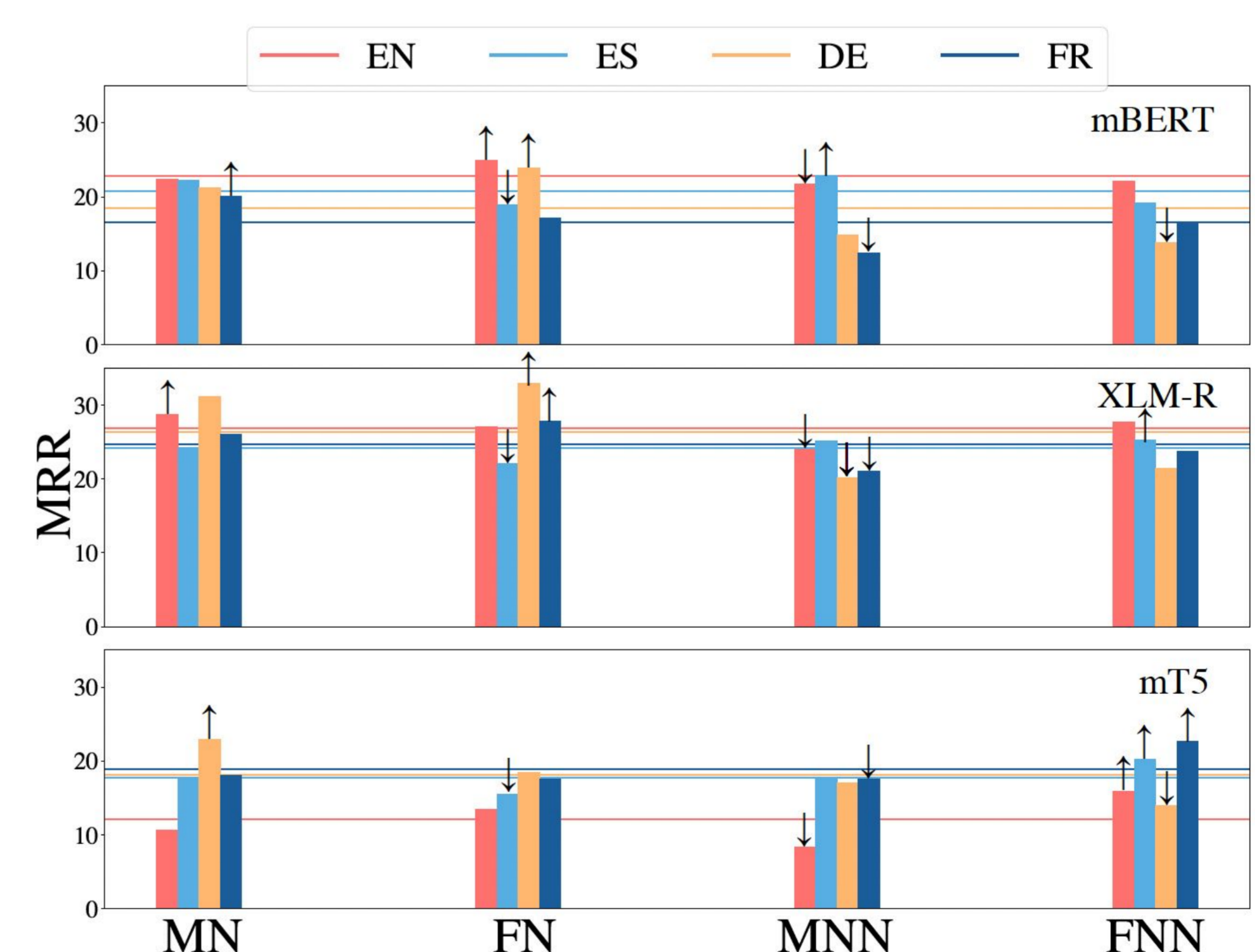
Precision + std. dev.

mBERT					
P@1	EN	ES	DE	FR	
MN	13.3	12.7	11.3	10.7	12.0 (1.0)
FN	13.3	12.0	15.3	8.0	12.2 (2.7)
MNN	12.7	12.4	11.4	3.6	10.0 (3.8)
FNN	13.3	10.0	5.6	6.9	9.0 (3.0)
	13.2 (0.3)	11.8 (1.1)	10.8 (3.5)	7.3 (2.5)	P@1(σ_{gd})

XLM-R					
P@1	EN	ES	DE	FR	
MN	16.7	13.3	20.7	16.7	16.9 (2.6)
FN	16.0	15.3	24.0	17.3	18.2 (3.5)
MNN	15.3	13.5	15.0	11.4	13.8 (1.5)
FNN	20.0	14.7	13.1	12.7	15.1 (3.0)
	17.0 (1.8)	14.2 (0.8)	18.2 (4.4)	14.5 (2.6)	P@1(σ_{gd})

mT5					
P@1	EN	ES	DE	FR	
MN	2.0	4.7	8.7	5.3	5.2 (2.4)
FN	4.0	3.3	6.7	3.3	4.3 (1.4)
MNN	2.0	4.7	6.4	4.3	4.4 (1.6)
FNN	3.3	6.7	1.9	6.2	4.5 (2.0)
	2.8 (0.9)	4.8 (1.2)	5.8 (2.5)	4.8 (1.1)	P@1(σ_{gd})

Mean Reciprocal Rank



Conclusion

MozArt! A new multilingual dataset of parallel cloze examples with demographic information from annotators. We show that **multilingual PLM are not equally fair** across languages; neither across groups of users.