

CONTEXT-AWARE STEREOTYPE DETECTION:

CONVERSATIONAL THREAD ANALYSIS ON BERT-BASED MODELS

Pol Pastells, Wolfgang S. Schmeisser-Nieto, Simona Frenda, Mariona Taulé



WHY Human annotators need context to label a text as containing stereotypes.

WHAT Do discursive contexts also play a role to interpret a message in NLP?

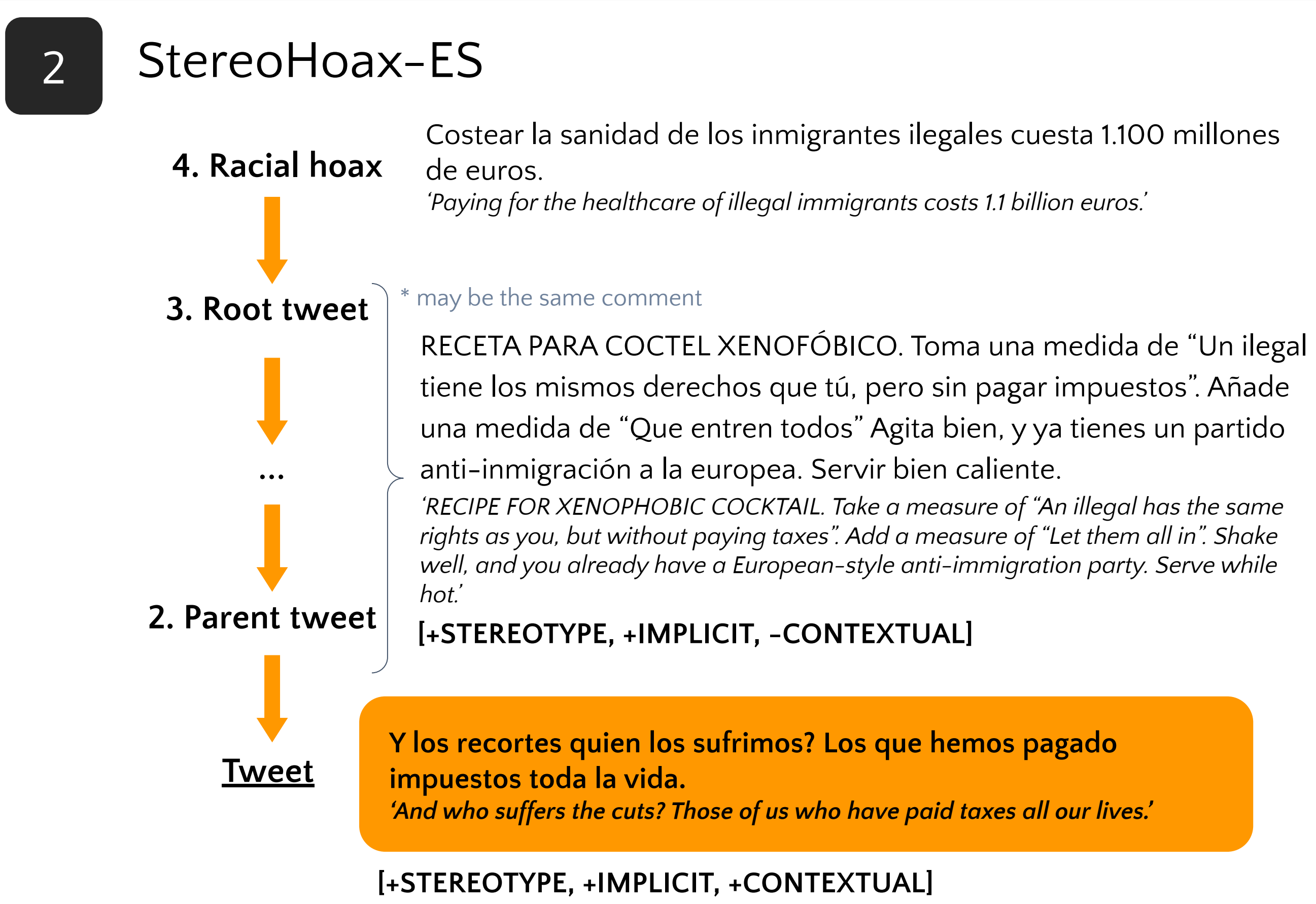
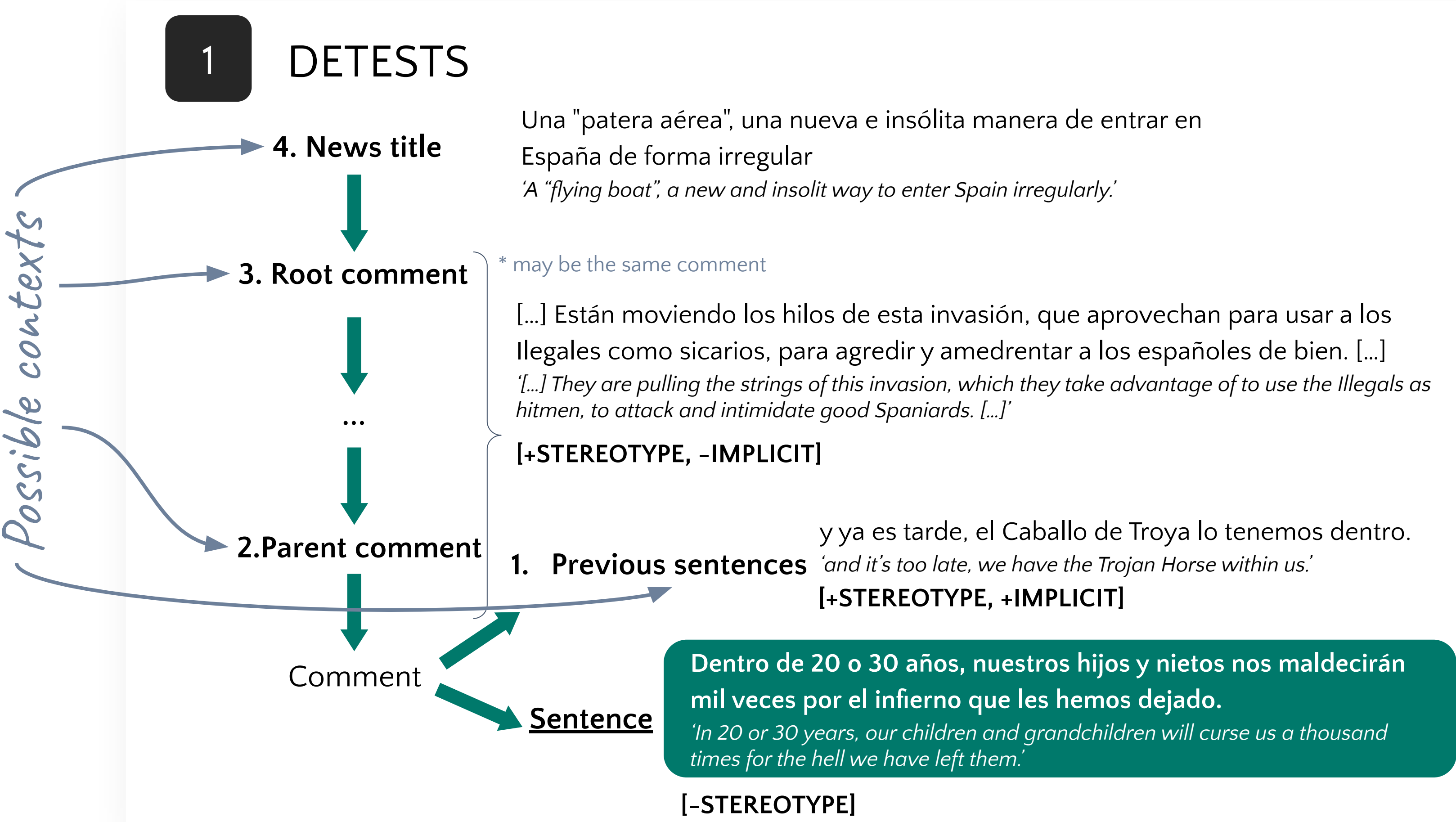
HOW Fine-tune classification models to detect stereotypes related to immigrants in Spanish. We add different contexts after [SEP] token of the models:

[CLS] + TEXT TO CLASSIFY + [SEP] + CONTEXT

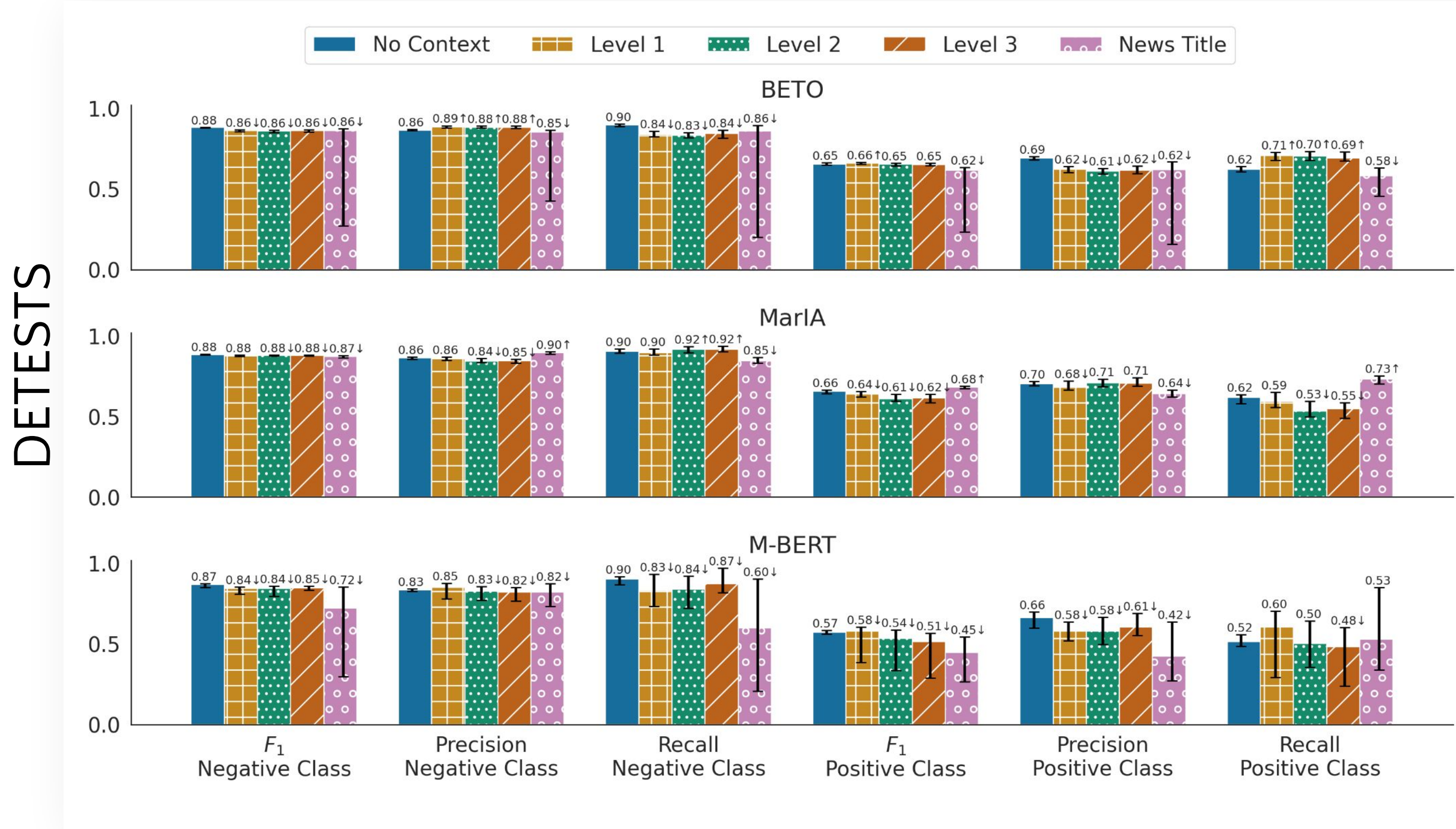
	Labels	N° texts DETESTS	N° texts StereoHoax-ES
Stereotype	Contextual	-	590
	Explicit	303	1,260
	Implicit	1,056	344
	Total	1,359	1,604
No Stereotype		4,270	3,745
Total		5,629	5,349

MODELS BETO (dccuchile/bert-base-spanish-wwm-cased)
MarIA (PlanTL-GOB-ES/roberta-base-bne)
M-BERT (google-bert/bert-base-multilingual-cased)

Datasets

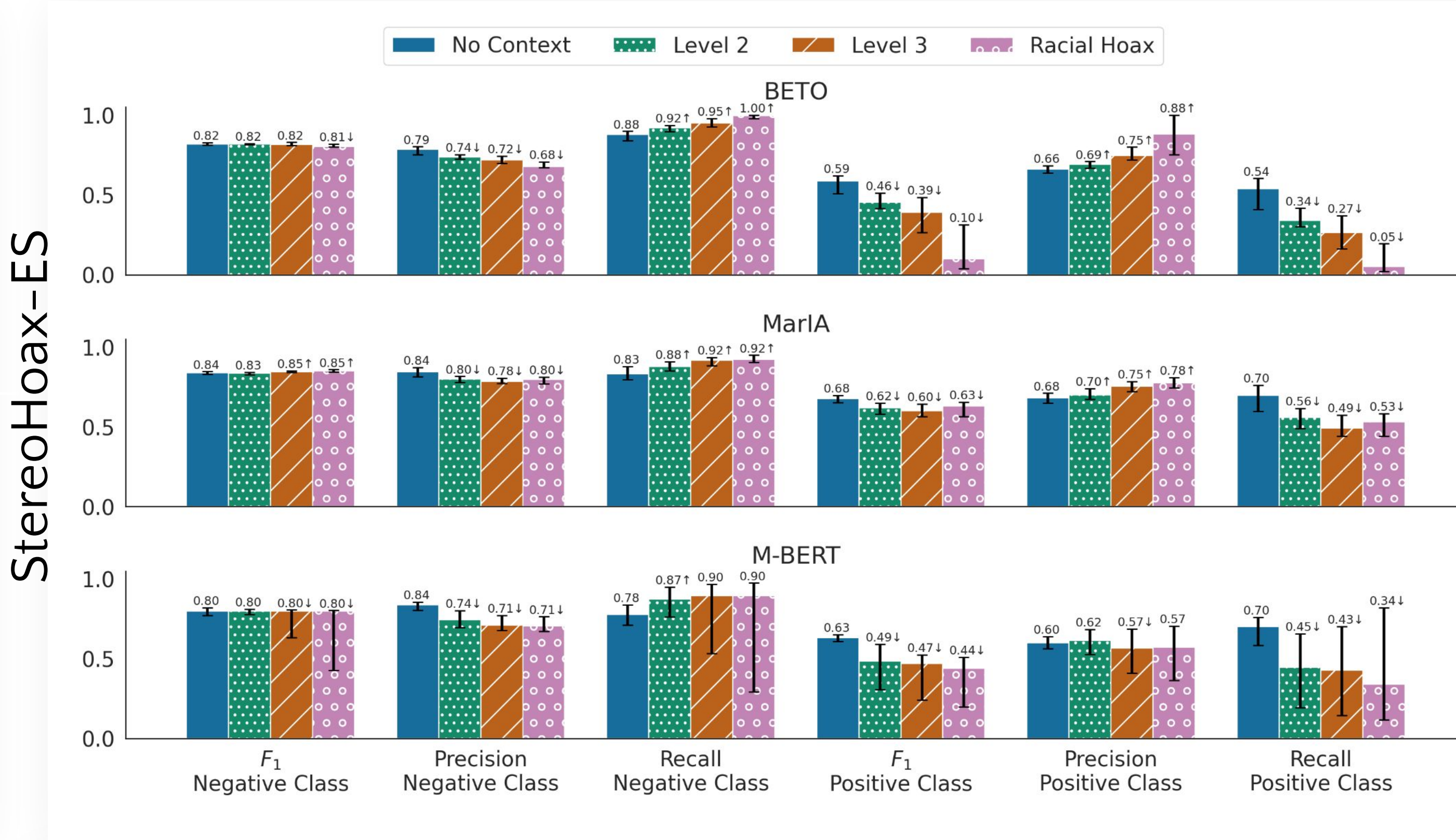


Results



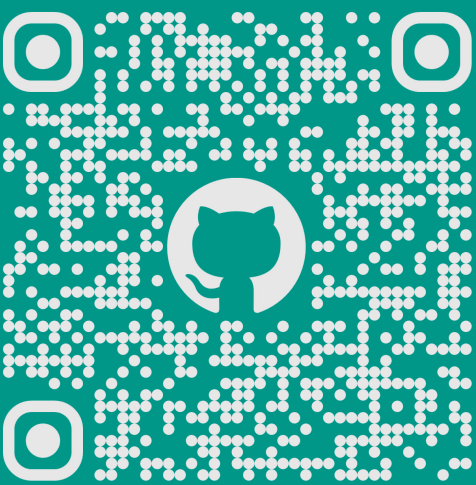
Model	Category	No Context > 65% seeds	Level 1 Changes	Level 2 Changes	Level 3 Changes	Level 4 Changes
BETO	FP	163	40 (25% ↓)	50 (31% ↓)	37 (23% ↓)	4 (2% ↓)
	TN	1173	0	1 (0% ↑)	2 (0% ↑)	3 (0% ↑)
	FN	112	0	1 (1% ↓)	3 (3% ↓)	3 (3% ↓)
	TP	292	16 (5% ↑)	19 (7% ↑)	14 (5% ↑)	2 (1% ↑)
MarIA	FP	169	11 (7% ↓)	7 (4% ↓)	7 (4% ↓)	41 (24% ↓)
	TN	1187	12 (1% ↑)	14 (1% ↑)	16 (1% ↑)	1 (0% ↑)
	FN	98	8 (8% ↓)	21 (21% ↓)	28 (29% ↓)	0
	TP	277	6 (2% ↑)	3 (1% ↑)	3 (1% ↑)	35 (13% ↑)
M-BERT	FP	198	6 (3% ↓)	1 (1% ↓)	0	3 (2% ↓)
	TN	1167	0	2 (0% ↑)	2 (0% ↑)	0
	FN	86	1 (1% ↓)	1 (1% ↓)	11 (13% ↓)	0
	TP	223	3 (1% ↑)	0	1 (1% ↑)	1 (0% ↑)

- ❖ BETO does **worse** with contexts (level 1, 2 and 3), with more False Positives.
- ❖ MarIA **worsens** with context level 2 and 3 with more False Negatives. Level 4 has a positive bias.



Model	Category	No Context > 65% seeds	Level 2 Changes	Level 3 Changes	Level 4 Changes
BETO	FP	142	12 (8% ↓)	0	0
	TN	609	28 (5% ↑)	34 (6% ↑)	55 (9% ↑)
	FN	58	29 (50% ↓)	51 (88% ↓)	113 (195% ↓)
	TP	148	2 (1% ↑)	1 (1% ↑)	0
MarIA	FP	83	1 (1% ↓)	1 (1% ↓)	0
	TN	566	11 (2% ↑)	18 (3% ↑)	23 (4% ↑)
	FN	68	22 (32% ↓)	22 (32% ↓)	27 (40% ↓)
	TP	226	3 (1% ↑)	0	0
M-BERT	FP	80	0	0	0
	TN	504	24 (5% ↑)	34 (7% ↑)	39 (8% ↑)
	FN	112	30 (27% ↓)	22 (20% ↓)	28 (25% ↓)
	TP	227	0	0	0

- ❖ All 3 models do **worse** with contexts (level 2, 3 and 4), with more False Negatives.



- ❖ No general improvement using contexts after the [SEP] token on BERT-based models.
- ❖ Results were highly dependent on the dataset used.