

# Additional file 4

## The adaptive community-response (ACR) method for collecting misinformation on social media

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### Outputs of the LMEM

In the present document, we present the outputs of the LMEM for mean text similarity and mean sentiment score, as provided by the STATSMODEL package. We also report the AIC and BIC of the models, and AIC and BIC differences between the nested models.

Output 1: Fixed effects model for the mean text similarity. The AIC and BIC of this model are  $-2364.94$  and  $-2328.96$ , respectively.

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Mixed Linear Model Regression Results
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Model:                MixedLM                Dependent Variable:    value
No. Observations:    1261                Method:                ML
No. Groups:          348                Scale:                0.0051
Min. group size:     1                Log-Likelihood:       1189.4702
Max. group size:     5                Converged:            Yes
Mean group size:     3.6
=====
                                Coef.  Std.Err.  z    P>|z| [0.025 0.975]
-----+-----+-----+-----+-----+-----+-----
Intercept                    0.470    0.006  72.646  0.000   0.458  0.483
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] -0.096    0.006 -14.940  0.000  -0.109 -0.084
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] -0.112    0.006 -18.729  0.000  -0.124 -0.100
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] -0.097    0.006 -16.860  0.000  -0.108 -0.085
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] -0.088    0.008 -10.952  0.000  -0.104 -0.072
Group Var                     0.009    0.014
=====

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Output 2: Random effects model for the mean text similarity. The AIC and BIC of this model are  $-2502.46$  and  $-2394.53$ , respectively. The AIC and BIC differences of the fixed and the random models are  $137.52$  and  $65.57$ , respectively.

Mixed Linear Model Regression Results

Model:		MixedLM	Dependent Variable:		value
No. Observations:		1261	Method:		ML
No. Groups:		348	Scale:		0.0019
Min. group size:		1	Log-Likelihood:		1272.2322
Max. group size:		5	Converged:		Yes
Mean group size:		3.6			

  

	Coef.	Std.Err.	z	P> z	[0.025	0.975]
Intercept	0.470	0.007	66.130	0.000	0.456	0.484
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b1]	-0.099	0.007	-13.326	0.000	-0.113	-0.084
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b2]	-0.115	0.007	-16.538	0.000	-0.128	-0.101
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b3]	-0.099	0.007	-15.059	0.000	-0.111	-0.086
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b4]	-0.088	0.008	-11.486	0.000	-0.103	-0.073
Group Var					0.016	
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b1] Cov					-0.007	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b1] Var					0.011	
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b2] Cov					-0.007	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b1] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b2] Cov					0.010	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b2] Var					0.011	
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b3] Cov					-0.007	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b1] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b3] Cov					0.008	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b2] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b3] Cov					0.010	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b3] Var					0.010	
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b4] Cov					-0.008	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b1] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b4] Cov					0.007	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b2] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b4] Cov					0.007	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b3] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b4] Cov					0.006	
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4']) [T.b4] Var					0.007	

Output 3: Fixed effects model for the mean sentiment score. The AIC and BIC of this model are  $-1568.06$  and  $-1532.08$ , respectively.

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Mixed Linear Model Regression Results
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Model:                MixedLM          Dependent Variable:    value
No. Observations:    1261              Method:                ML
No. Groups:          348                Scale:                 0.0106
Min. group size:     1                  Log-Likelihood:       791.0303
Max. group size:     5                  Converged:             Yes
Mean group size:     3.6
=====
                                Coef. Std.Err.  z    P>|z| [0.025 0.975]
-----+-----+-----+-----+-----+-----+-----+-----
Intercept                0.429   0.008  51.985  0.000  0.412  0.445
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] -0.112  0.009 -12.157  0.000 -0.130 -0.094
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] -0.097  0.009 -11.356  0.000 -0.114 -0.080
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] -0.082  0.008 -10.022  0.000 -0.098 -0.066
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] -0.043  0.011 -3.756  0.000 -0.066 -0.021
Group Var                 0.013   0.015
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Output 4: Random effects model for the mean sentiment score. The AIC and BIC of this model are  $-1598.72$  and  $-1490.79$ , respectively. The AIC and BIC differences of the fixed and the random models are  $30.66$  and  $-41.29$ , respectively.

Mixed Linear Model Regression Results

Model:		MixedLM	Dependent Variable:		value
No. Observations:		1261	Method:		ML
No. Groups:		348	Scale:		0.0062
Min. group size:		1	Log-Likelihood:		820.3623
Max. group size:		5	Converged:		Yes
Mean group size:		3.6			

  

	Coef.	Std.Err.	z	P> z	[0.025	0.975]
Intercept	0.429	0.008	54.693	0.000	0.413	0.444
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1]	-0.111	0.009	-11.730	0.000	-0.129	-0.092
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2]	-0.096	0.009	-10.127	0.000	-0.115	-0.078
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3]	-0.083	0.008	-9.873	0.000	-0.099	-0.067
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4]	-0.042	0.011	-3.758	0.000	-0.064	-0.020
Group Var		0.015				
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] Cov		-0.002				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] Var		0.010				
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] Cov		-0.005				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] Cov		0.011				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] Var		0.014				
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] Cov		-0.003				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] Cov		0.007				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] Cov		0.009				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] Var		0.010				
Group x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] Cov		-0.005				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b1] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] Cov		0.003				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b2] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] Cov		0.006				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b3] x C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] Cov		0.001				
C(t, levels=['s', 'b1', 'b2', 'b3', 'b4'])[T.b4] Var		0.008				

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