A Datasets

Table 21: Mean confidence scores (μ) and standard deviation of confidence scores (σ) across each QT30 subcorpus.

Corpus Name	μ	σ
28May2020	0.76	0.16
4June2020	0.72	0.17
18June2020	0.76	0.16
30July2020	0.75	0.16
2September2020	0.78	0.15
22October2020	0.76	0.16
5November2020	0.77	0.17
19November2020	0.74	0.19
10December2020	0.77	0.16
14January2021	0.74	0.17
28January2021	0.70	0.17
18February2021	0.75	0.16
4March2021	0.76	0.16
18March2021	0.75	0.17
15April2021	0.75	0.15
29April2021	0.70	0.19
20May2021	0.76	0.17
27May2021	0.79	0.15
10June2021	0.75	0.17
24June2021	0.74	0.17
8July2021	0.72	0.17
22July2021	0.44	0.28
5August2021	0.76	0.17
19August2021	0.77	0.17
2September2021	0.78	0.16
16September2021	0.77	0.16
30September2021	0.24	0.08
14October2021	0.75	0.19
28October2021	0.75	0.17
11November2021	0.78	0.16
QT30	0.75	0.17

Table 22: Mean confidence scores (μ) and Standard Deviation of confidence scores (σ) across Moral Maze subcorpora.

Subcorpus	μ	σ
В	0.79	0.14
E	0.76	0.15
M	0.78	0.14
P	0.80	0.15
S	0.74	0.16
G	0.80	0.14
Н	0.74	0.16
W	0.75	0.15

B Results

B.1 In-Domain

Table 23: In-domain results when testing different model architectures on the 4-class problem. F1: Macro-averaged F1, P: precision, R: recall. Highest scores in each column are shown in bold.

Fusio	n Methods		SCS			LCS		US			
Sequence	Multimodal	F1	P	R	F1	P	R	F1	P	R	
			Text	Only							
Early	-	.58	.58	.58	.59	.59	.59	.59	.59	.59	
Late	-	.36	.36	.35	.34	.36	.34	.35	.35	.35	
Audio Only											
Early	-	.43	.48	.44	.41	.41	.42	.20	.31	.26	
Late	-	.28	.29	.29	.29	.31	.29	.29	.29	.29	
			Multi	moda	l						
	Concatenation	.58	.58	.58	.57	.57	.57	.58	.58	.57	
Early	Product	.56	.56	.57	.57	.57	.57	.58	.58	.59	
Larry	CA Text	.57	.56	.58	.46	.46	.48	.57	.56	.57	
	CA Audio	.58	.58	.58	.57	.59	.56	.57	.58	.57	
Late	Concatenation	.36	.36	.35	.37	.36	.36	.35	.36	.35	
	Baselines										
R	andom	.22	.24	.23	.23	.25	.24	.24	.25	.26	
M	Iajority	.14	.09	.25	.14	.09	.25	.14	.09	.25	

Table 24: In-domain results when testing different model architectures on the 3-class problem. F1: Macro-averaged F1, P: precision, R: recall. Highest scores in each column are shown in bold.

Fusio	n Methods		SCS			LCS			US		
Sequence	Multimodal	F1	P	R	F1	P	R	F1	P	R	
			Text	Only							
Early	-	.62	.63	.62	.59	.63	.58	.61	.63	.60	
Late	-	.43	.45	.43	.44	.45	.44	.43	.44	.44	
Audio Only											
Early	-	.21	.16	.33	.54	.54	.55	.22	.16	.33	
Late	-	.33	.33	.34	.34	.33	.34	.34	.33	.35	
			Multi	moda	l						
	Concatenation	.61	.62	.60	.61	.61	.61	.60	.61	.59	
Early	Product	.58	.60	.57	.61	.62	.61	.62	.64	.61	
Larry	CA Text	.53	.52	.54	.53	.52	.55	.52	.51	.53	
	CA Audio	.61	.62	.60	.62	.63	.62	.63	.62	.64	
Late	Concatenation	.43	.44	.43	.45	.47	.45	.44	.45	.44	
Baselines											
Random .28 .33 .33 .30 .34 .36 .24 .25 .26											
M	Iajority	.22	.16	.33	.22	.16	.33	.22	.16	.33	

B.2 Cross-Domain

Table 25: Cross domain, 4-class macro-F1 results trained on SCS. Highest values in each column are shown in bold.

Fusio	n Methods	В	Е	M	P	S	G	Н	W	Mean
Sequence	Multimodal	ь	E	IVI	Г	S	U	п	VV	Mean
			Tex	t-Only	7					
Early	-	.44	.46	.48	.42	.43	.50	.51	.41	.46
Late	-	.31	.28	.32	.26	.30	.31	.33	.29	.30
Audio-Only										
Early	-	.34	.37	.39	.38	.33	.38	.40	.37	.37
Late	-	.20	.22	.24	.22	.22	.28	.25	.22	.23
			Mul	timoda	al					
	Concatenation	.43	.43	.45	.45	.45	.49	.45	.43	.45
Early	Product	.41	.44	.42	.42	.46	.52	.46	.43	.45
Larry	CA Text	.40	.44	.44	.40	.42	.49	.43	.44	.43
	CA Audio	.45	.49	.44	.42	.45	.53	.48	.46	.47
Late	Concatenation	.32	.28	.31	.27	.29	.31	.35	.30	.30
Baselines										
R	andom	.19	.23	.20	.19	.22	.25	.21	.24	.22
M	lajority	.16	.15	.15	.15	.15	.14	.15	.14	.15

Table 26: Cross domain, 4-class macro-F1 results trained on LCS. Highest values in each column are shown in bold.

Fusio	n Methods	В	Е	M	Р	S	G	Н	W	Mean
Sequence	Multimodal	Б	E	IVI	Г	3	G	п	VV	Mean
Text-Only										
Early	-	.41	.46	.46	.44	.45	.52	.45	.46	.46
Late	-	.31	.28	.30	.33	.33	.33	.38	.32	.32
Audio-Onl	y									
Early	-	.33	.37	.38	.36	.32	.36	.40	.36	.36
Late	-	.22	.22	.24	.24	.20	.27	.26	.23	.24
Multimoda	.1									
	Concatenation	.41	.43	.41	.42	.39	.53	.45	.44	.43
Early	Product	.39	.44	.42	.41	.47	.51	.43	.44	.44
Larry	CA Text	.25	.44	.33	.29	.27	.40	.34	.32	.32
	CA Audio	.41	.49	.44	.42	.46	.49	.52	.46	.46
Late	Concatenation	.29	.28	.27	.24	.28	.31	.34	.29	.29
Baselines										
Random		.19	.23	.20	.19	.22	.25	.21	.24	.22
Majority		.16	.15	.15	.15	.15	.14	.15	.14	.15

Table 27: Cross domain, 4-class macro-F1 results trained on US. Highest values in each column are shown in bold.

Fusio	n Methods	В	Е	М	Р	S	G	Н	W	Mean
Sequence	Multimodal	1	Ľ	IVI	Г	S	U	11	VV	Mean
			Tex	t-Only	У					
Early	-	.43	.44	.45	.44	.45	.54	.47	.45	.46
Late	-	.29	.27	.33	.29	.34	.35	.31	.33	.31
Audio-Only										
Early	-	.23	.22	.23	.22	.23	.18	.22	.22	.22
Late	-	.29	.31	.27	.24	.28	.31	.34	.29	.29
			Mul	timod	al					
	Concatenation	.40	.43	.43	.45	.44	.51	.44	.46	.45
Early	Product	.44	.50	.42	.44	.47	.51	.46	.46	.46
Earry	CA Text	.36	.45	.39	.38	.41	.48	.37	.42	.41
	CA Audio	.40	.40	.41	.41	.43	.50	.45	.43	.43
Late	Concatenation	.26	.24	.28	.28	.33	.33	.31	.33	.30
Baselines										
Random		.19	.23	.20	.19	.22	.25	.21	.24	.22
Majority		.16	.15	.15	.15	.15	.14	.15	.14	.15

Table 28: Cross domain, 3-class macro-F1 results trained on SCS. Highest values in each column are shown in bold.

Fusio	n Methods	В	Е	M	P	S	G	Н	W	Mean
Sequence	Multimodal	р .	Ľ	1V1	Г	S	U	11	vv	Mean
			Tex	t-Only	7					
Early	-	.54	.47	.50	.50	.58	.55	.63	.51	.54
Late	-	.45	.38	.35	.36	.36	.37	.43	.34	.38
Audio-Only										
Early	-	.21	.21	.22	.21	.20	.21	.22	.21	.21
Late	-	.33	.32	.32	.33	.32	.34	.34	.32	.33
			Mul	timoda	al					
	Concatenation	.58	.49	.47	.50	.57	.57	.59	.49	.53
Early	Product	.51	.45	.44	.47	.53	.54	.64	.47	.51
Larry	CA Text	.43	.40	.43	.41	.44	.46	.47	.44	.44
	CA Audio	.59	.44	.52	.48	.59	.54	.60	.52	.54
Late	Concatenation	.37	.37	.34	.40	.43	.42	.43	.40	.40
Baselines										
Random		.27	.32	.30	.27	.30	.29	.29	.33	.30
Majority		.21	.21	.22	.21	.20	.21	.22	.20	.21

Table 29: Cross domain, 3-class macro-F1 results trained on LCS. Highest values in each column are shown in bold.

Fusio	n Methods	В	E	M	P	S	G	Н	W	Mean	
Sequence	Multimodal	Б	E	IVI	Г	3	G	п	VV	Mean	
			Tex	t-Only	/						
Early	-	.59	.49	.48	.46	.51	.54	.69	.52	.54	
Late	-	.38	.36	.36	.39	.44	.43	.41	.43	.40	
Audio-Only											
Early	-	.51	.47	.45	.33	.47	.44	.49	.53	.49	
Late	-	.22	.25	.24	.24	.20	.27	.26	.23	.24	
			Mul	timoda	al						
	Concatenation	.51	.47	.45	.48	.56	.53	.58	.49	.51	
Forly	Product	.53	.49	.50	.52	.56	.54	.59	.53	.53	
Early	CA Text	.46	.40	.45	.42	.46	.45	.45	.51	.45	
	CA Audio	.52	.55	.47	.52	.58	.55	.60	.52	.54	
Late	Concatenation	.43	.37	.33	.36	.4	.42	.41	.39	.39	
Baselines											
Random		.27	.32	.30	.27	.30	.29	.29	.33	.30	
Majority		.21	.21	.22	.21	.20	.21	.22	.20	.21	

Table 30: Cross domain, 3-class macro-F1 results trained on US. Highest values in each column are shown in bold.

Fusio	n Methods	В	Е	M	Р	S	G	Н	W	Mean
Sequence	Multimodal	Б	E	IVI	Г	3	G	п	VV	Mean
			Tex	t-Only	7					
Early	-	.57	.48	.47	.50	.62	.55	.60	.54	.54
Late	-	.47	.35	.36	.39	.40	.40	.39	.40	.40
Audio-Only										
Early	-	.21	.21	.22	.21	.20	.21	.22	.21	.21
Late	-	.33	.34	.32	.33	.34	.34	.33	.33	.33
			Mul	timoda	al					
	Concatenation	.59	.46	.47	.50	.55	.55	.60	.50	.53
Forly	Product	.55	.49	.50	.52	.54	.52	.61	.50	.53
Early	CA Text	.42	.39	.42	.41	.42	.46	.49	.40	.43
	CA Audio	.59	.52	.49	.53	.59	.54	.62	.53	.55
Late	Concatenation	.38	.36	.36	.37	.36	.38	.43	.34	.37
Baselines										
Random		.27	.32	.30	.27	.30	.29	.29	.33	.30
Majority		.21	.21	.22	.21	.20	.21	.22	.20	.21