

Lista de artículos y BDE

	Autor	BDE
1	[Abtahi et al., 2018]	Abtahi Dataset
2	[Abu Shaqra et al., 2019]	AVANEmo
3	[Bagher Zadeh et al., 2018]	CMU-MOSEI
4	[Barros et al., 2018]	OMG Emotion
5	[Ben et al., 2022]	MMEW
6	[Bhattacharya et al., 2020]	EWalk
7	[Bian et al., 2019]	OL-SFED
8	[Boccignone et al., 2017]	AMHUSE
9	[Conneau et al., 2017]	EMOEEG
10	[Davison et al., 2018]	SAMM
11	[Doyran et al., 2021]	MUMBAI
12	[Dudzic et al., 2021]	Mementos
13	[Firdaus et al., 2020]	SEMD
14	[Gorbova et al., 2019]	ME dataset
15	[Grobova at al., 2022]	HSDB
16	[Guimard et al., 2022]	PEM360
17	[Gupta et al., 2018]	DAiSEE
18	[Happy et al., 2017]	ISED
19	[Husák et al., 2017]	MEVIEW
20	[Jo et al., 2020]	ROSBAG-AD
21	[Kollias and Zafeiriou, 2021]	Aff-Wild2
22	[Kossaifi et al., 2017]	AFEW-VA
23	[Li et al., 2017a]	DIVRV
24	[Li et al., 2017b]	CHEAVD
25	[Li et al., 2020a]	EEG-BMMED
26	[Li et al., 2020b]	MEMOS
27	[Li et al., 2021]	DEFE
28	[Li et al., 2022a]	CAS(ME)3
29	[Li et al., 2022b]	PPB-Emo
30	[Li et al., 2022c]	DEFE+
31	[Li et al., 2022d]	4DME
32	[Liang et al., 2021]	FG-Emotions
33	[Liliana et al., 2018]	IMED
34	[Liu et al., 2022]	MAFW
35	[Livingstone and Russo, 2018]	RAVDESS
36	[Luo et al., 2020]	BoLD
37	[Ma et al., 2019]	ElderReact
38	[Miolla et al., 2021]	PEDFE
39	[Miranda-Correa et al., 2021]	AMIGOS
40	[Mo et al., 2021]	MFED
41	[Nguyen et al., 2019]	SIE-Intensity
42	[Niu et al., 2018]	DaLC
43	[Nonis et al., 2021]	EVFEV
44	[Perepelkina et al., 2018]	RAMAS
45	[Poria et al., 2018]	MELD
46	[Qu et al., 2018]	CAS(ME)2

47	[Quach et al., 2022]	GEVC-GroupVid
48	[Quiroz et al., 2022]	DatasetROS
49	[Saganowski et al., 2022]	Emognition
50	[Sapiński et al., 2019]	POD
51	[Seuss et al., 2019]	ASD
52	[Shen et al., 2020]	MEMoR
53	[Shukla et al., 2017]	Shukla Dataset
54	[Shukla et al., 2022]	ADdataset
55	[Soleymani and Pantic, 2017]	MAHNOB HCI
56	[Song et al., 2019]	MPED
57	[Subramanian et al., 2018]	ASCERTAIN
58	[Suhaimi et al., 2018]	VR Dataset
59	[Tabbaa et al., 2022]	VRE:ED
60	[Tripathi et al., 2019]	EmoWare
61	[Vidal et al., 2020]	MSP-Face
62	[Wang et al., 2022a]	EmVidCap
63	[Wang et al., 2022b]	SWUME
64	[Wei et al., 2017]	BNU-LSVED 2.0
65	[Xue et al., 2021]	CEAP-360VR
66	[Yang et al., 2018]	DAG
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68	[Zhang et al., 2021a]	Werewolf-XL
69	[Zhang et al., 2021b]	CPED
70	[Zlatintsi et al., 2017]	COGNIMUSE

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