Phases in Disaster Cycle	Task the benchmark is made for	Type (ML)
Provention Foregoting and Prediction	Drought forecasting [18]	multiclass (ordinal) classification
Prevention - Forecasting and Prediction	Earth surface forecasting; extreme summer	video prediction
	prediction; seasonal cycle prediction[22]	
Preparedness - Early warning	Wildfire spread prediction [16]	image segmentation
Preparedness - Monitoring and Detection	Building detection, road network extraction	multi-class segmentation
	and then flood detection [14]	+ binary classification
	Ground deformation [8]	binary classification
	Real-time wildfire smoke detection [11]	binary classification
	Disaster detection [23]	multiclass multilabel classification
	Disaster detection, damage detection, incident detection in scene-centric images [25]	multiclass classification
		multiclass multilabel
	Disaster detection [26]	classification
	Video Incident classification [24]	multiclass classification
	Flood detection [6]	binary classification
	Flood extent detection [12]	image segmentation
	Disaster type detection [3]	multiclass classification
Response - Damage assessment		multiclass (ordinal)
	Damage severity assessment [3]	classification
	Change detection and damage	multiclass (ordinal)
	assessment of buildings [13]	classification
	Damage assessment [10]	semantic segmentation
		image segmentation,
	Damage assessment of buildings [27]	multiclass (ordinal)
		classification
		image segmentation +
	Damaged Building Detection [9]	multiclass (ordinal)
		classification
	Post Flood Scene Understanding	image classification, semantic
	(Flood detection, distinguishing different	segmentation task, and
	water bodies and flood)[21]	visual question answering
		Semantic segmentation;
	Damage assessment [17]	multilabel classification
	Damage severity assessment [20]	multiclass (ordinal) classification
	Damage assessment [19]	multiclass classification
	Categorization of humanitarian tasks [4]	multiclass classification
	Informativeness classification;	binary classification;
Response - Post-disaster	Categorization of humanitarian tasks;	multiclass classification;
Coordination and Response	Damage severity assessment [2]	multiclass (ordinal) classification
	Informativeness classification [3]	binary classification
	Categorization of humanitarian tasks [3]	multiclass classification
	Informativeness detection; categorization of	binary classification;
	humanitarian tasks [5]	multiclass classification
Recovery	Sentiment analysis [15]	multiclass multilabel classificatio
Miscellaneous (Preparedness -		
Monitoring and Detection; Response -	Road passability assessment, flood detection	
Post-disaster Coordination	and flood classification [7]	multilabel classification
and Response)		
Miscellaneous (Preparedness -		
· -	Disaster type detection;	
Monitoring and Detection;	Informativeness classification;	multitade loaming
Response - Post-disaster Coordination	Categorization of humanitarian tasks;	multitask learning
and Response; Response -	Damage severity assessment [1]	
Damage assessment)		

Table 1: Task and type of benchmark datasets for Natural disaster

Note: This table has a brief summary of all the papers of reviewed in this site. We encourage you to read the original

paper(referenced in the table) if you're interested in that benchmark dataset.

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