

NATURAL LANGUAGE PROCESSING IN URGENCY DETECTION

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What is Urgency Detection?

- Humanitarian disasters: climate change, refugee crisis.
- Technology assisting: aid agencies and first responders
- Detecting *relevant* or *informative* tweets
- Sparseness of data, varying characteristics of disasters



- *Expresses an actionable need that needs to be resolved in a short time frame.*
- Information Retrieval (IR) problem
- Goal: flag messages that *express urgency* panic-ridden emotion
- Examples

Urgent and Non-Urgent example from a real-world dataset

Dataset

- Urgent sentences
- Non-Urgent sentences

Macedonia

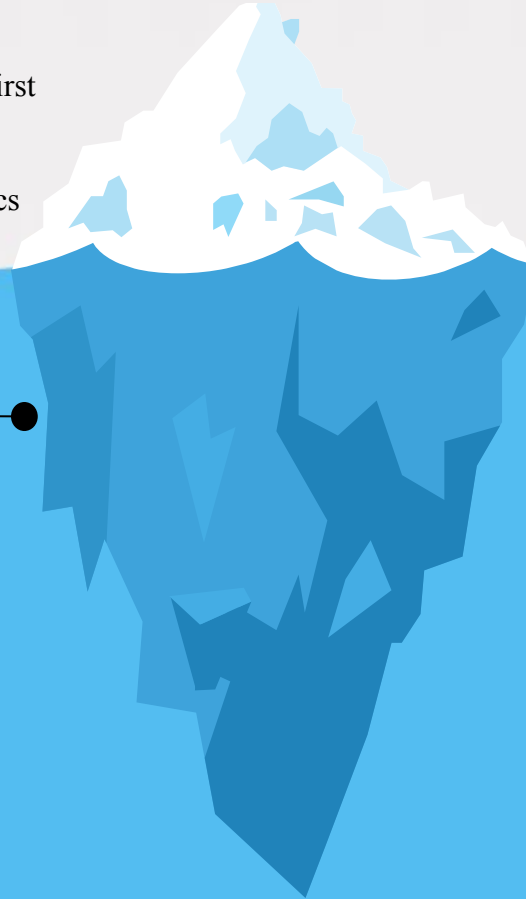
- Four locals are trapped in the marketplace need help.
- Avalanche in Macedonia caused four deaths

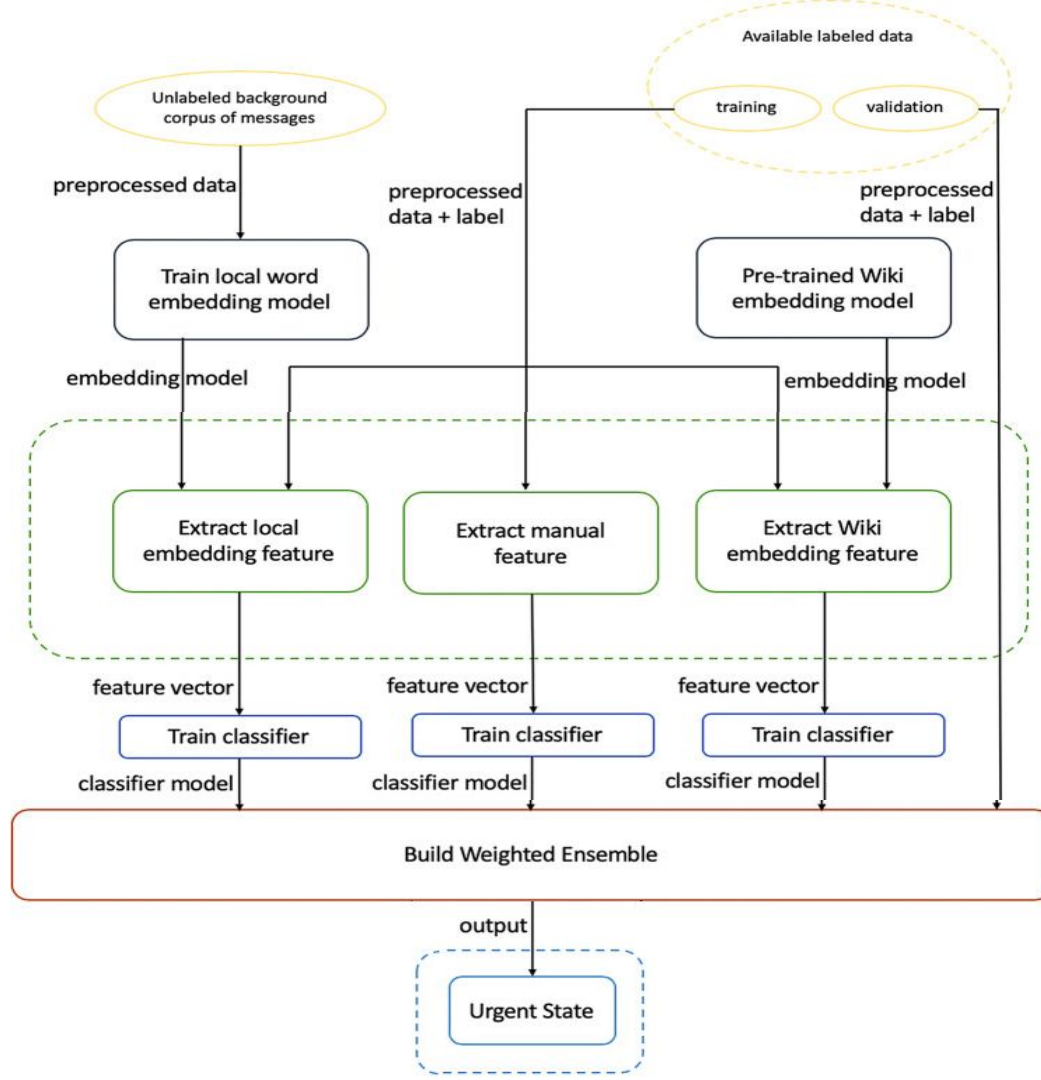
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Role of Machine Learning —●





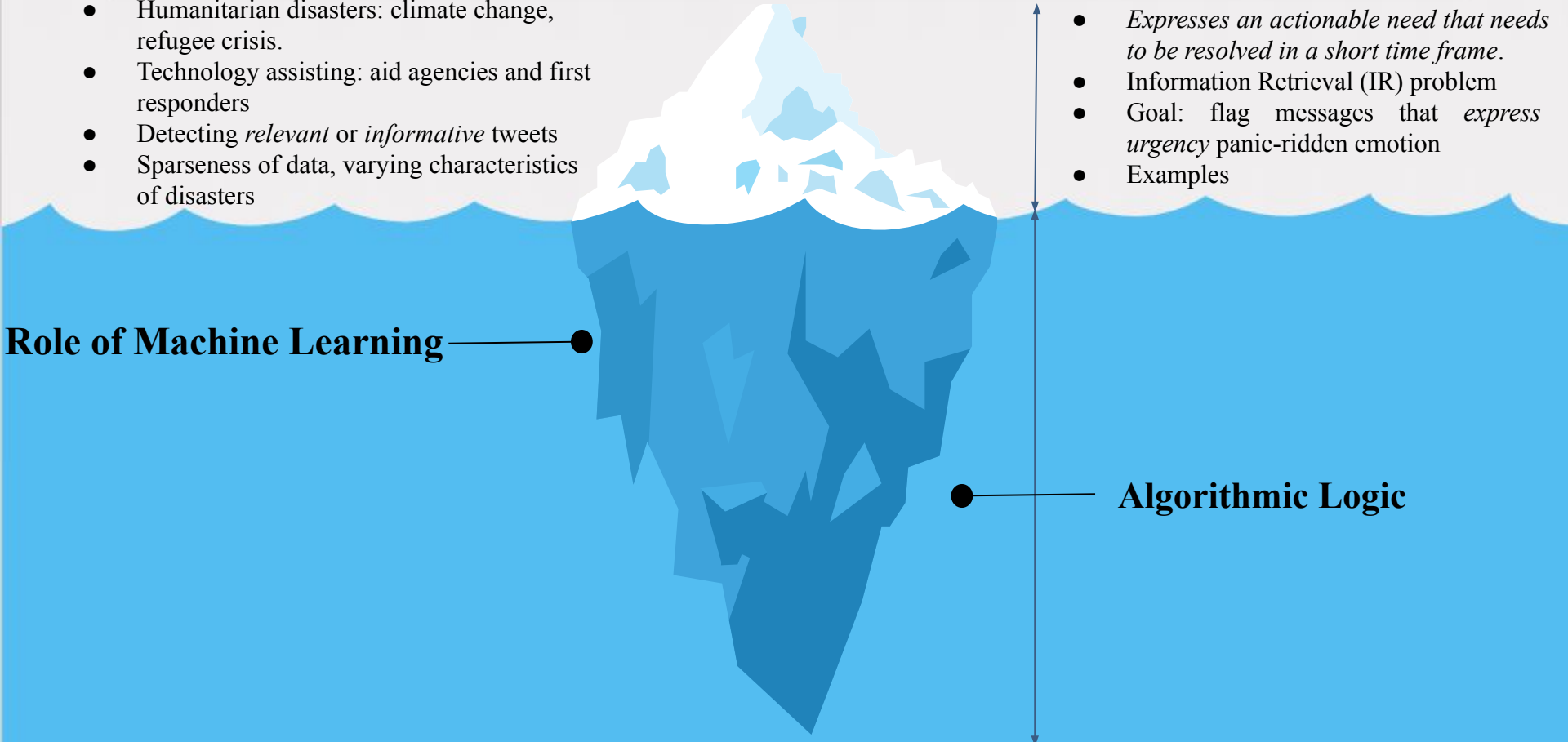
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Role of Machine Learning —●

● — **Algorithmic Logic**





Algorithmic Logic: Transfer Learning for Urgency Detection

INPUT :

- Labeled dataset in target domain: D_t
- Labeled dataset in source domain: D_{sl}
- Unlabeled corpus in source domain: D_{su}
- Pre-trained Wikipedia Embedding Model: W_w
- Up-sampling parameter: u

OUTPUT :

- Classifier for Urgency Detection: C

METHOD :

- 1) Train word embedding W_s on text in $D_{su} \cup D_{sl}$;
- 2) Up-sample D_t by factor u and 'mix' with D_{sl} to get expanded training set, $D_{train} : D_{tu} \cup D_{sl}$
- 3) Extract manual feature set F_m , source embedding feature set F_s (using W_s), and Wiki feature set F_w (using W_w) from each message in D_{train} ;
- 4) Train linear regression models C_s , C_m and C_w on F_s , F_m and F_w resp. to get classifier;
- 5) Return final classifier model $C : \text{avg score}(C_s, C_m, C_w)$;

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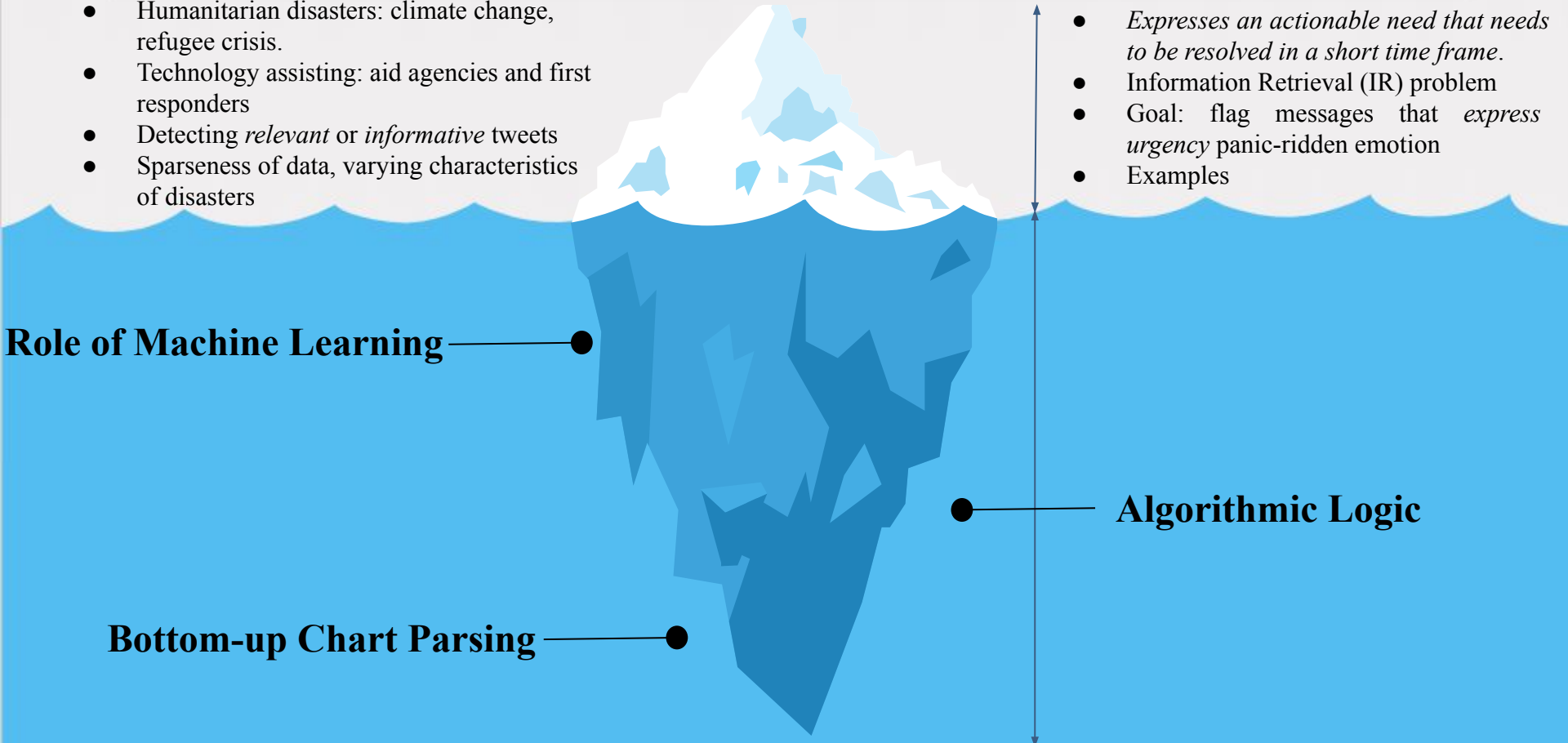
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Role of Machine Learning —●

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Bottom-up Chart Parsing —●





Bottom-up Chart Parsing

Grammar

1. $S \rightarrow NP \ VP$
2. $NP \rightarrow Det \ N$
3. $VP \rightarrow VG \ NP$
4. $VG \rightarrow V$

Lexicons

1. Esha: NP
2. plays: V
3. the: Det
4. piano: N

Sentence: ₁ Esha ₂ plays ₃ the ₄ piano ₅

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(11) (0, 4, $S \rightarrow NP \ VP$ *)			
--	(10) (1, 4, $VP \rightarrow VG \ NP$ *)		
--	(5) (1, 2, $VP \rightarrow VG \ * \ NP$)	(9) (2, 4, $NP \rightarrow Det \ N$ *)	
(2) (0, 1, $S \rightarrow NP \ * \ VP$)	(4) (1, 2, $VG \rightarrow V$ *)	(7) (2, 3, $NP \rightarrow Det \ * \ N$)	--
(1) "Esha" (0, 1, $NP \rightarrow \text{"Esha"} \ *$)	(3) "plays" (1, 2, $V \rightarrow \text{"plays"} \ *$)	(6) "the" (2, 3, $Det \rightarrow \text{"the"} \ *$)	(8) "piano" (3, 4, $N \rightarrow \text{"piano"} \ *$)

REFERENCES

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