### Event Causality Identification with Causal News Corpus - Shared Task 3, CASE 2023

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#### TASK

## **D1** Event Causality Identification Shared Task involved two subtasks related to Classification and Span Detection.



#### DATASET

### <sup>02</sup> Subtask 1 worked directly on the Causal News Corpus V2 (CNC-V2) / RECESS (Tan et al., 2023).

Stat.	Label	Train	Dev	Test	Total
#	Causal	1624	185	173	1982
Sent-	Non-causal	1451	155	179	1785
ences	Total	3075	340	352	3767
Avg.	Causal	33.44	34.41	35.93	33.75
#	Non-causal	26.69	26.85	28.67	26.90
words	Total	30.25	30.96	32.24	30.50

Subtask 1 Data Summary Statistics.

- Data Source: Causal News Corpus V2 *a.k.a.* RECESS (Tan et al., 2023)
- A sentence is *Causal* if "one argument provides the reason, explanation or justification for the situation described by the other"(Webber et al., 2019) and contains at least a pair of events.

#### DATASET

## 03

### Subtask 2 worked directly on the Causal News Corpus V2 (CNC-V2) / RECESS (Tan et al., 2023).

Statistic	Train	Dev	Test	Total
# Sentences	1624	185	173	1982
# Relations	2257	249	248	2754
Avg. rels/sent	1.39	1.35	1.43	1.39
Avg. # words	33.44	34.41	35.93	33.75
Cause	11.56	12.20	12.96	11.74
Effect	10.71	10.18	11.54	10.74
Signal	1.45	1.53	1.46	1.46
Avg # Sig./rel	0.70	0.64	0.79	0.70
Prop. of rels w/ Sig.	0.68	0.63	0.76	0.69

- A **Cause** is a reason, explanation or justification that led to an **Effect**.
- **Signals** are words that help to identify the structure of the discourse.

Subtask 2 Data Summary Statistics.

### **EVALUATION**

# We provided multiple evaluation metrics, but model performance was eventually ranked by F1.

- The following evaluation metrics were provided:
  - Subtask 1: Accuracy, Binary Precision (P), Binary Recall (R), Binary F1 and Matthews Correlation Coefficient
  - Subtask 2: Macro P, R and F1 based on word labels
- Leader board was ranked by F1 for both tasks
- For Subtask 2, to handle predictions for examples with multiple causal relations:
  - If more predictions (p) are provided than true relations (n), we only consider the first n relations.
  - If fewer predictions (p) are provided than true relations (n), we assume the missing n-p relations have all "Other" tokens.
  - Once n=p, we calculate every combination of pairs of prediction and true relations and retain the combination that gives us the highest score.

### **COMPETITION**

# We used the Codalab website to host our competition.



### COMPETITION

06 Timeline



*Timeline of competition.* 

### COMPETITION

# <sup>07</sup> There were 23 registered participants throughout the competition.



Number of teams per stage of competition.

### RESULTS

### **08** The best F1 score for Subtask 1 was 84.66%.

Rank	Team Name	Codalab Username	R	Р	F1	Acc	MCC
1	-	DeepBlueAI	86.13	83.24	84.66	84.66	69.37
2	InterosML (Patel, 2023)	rpate112	87.28	81.62	84.36	84.09	68.37
3	BoschAI (Schrader et al., 2023)	timos	87.86	80.00	83.75	83.24	66.83
4	CSECU-DSG (Hossain et al., 2023)	csecudsg	85.55	80.00	82.68	82.39	64.95
5	-	elhammohammadi	89.60	76.35	82.45	81.25	63.52
6	BERT Baseline	tanfiona	89.02	75.86	81.91	80.68	62.37
7	Anonymous	sgopala4	86.13	78.01	81.87	81.25	62.88
8	MLModeler5 (Bhatia et al., 2023)	nitanshjain	87.28	65.37	74.75	71.02	44.83
9	VISU	kunwarv4	52.60	85.85	65.23	72.44	48.19
10	-	pakapro	47.40	44.09	45.68	44.60	-10.72

Subtask 1 Leaderboard.

#### RESULTS

### **10** The best F1 score for Subtask 2 was 72.79%.

Ra-	Teem Name	Codalab	Overall		
nk	Team Name	Username	R	Р	F1
1	BoschAI (Schrader et al., 2023)	timos	63.98	84.42	72.79
2	1Cademy Baseline	tanfiona	59.18	60.25	59.71
3	CSECU-DSG (Hossain et al., 2023)	csecudsg	36.12	40.00	37.96
4	-	pakapro	0.00	0.00	0.00

Subtask 2 Leaderboard.

## **12 Conclusion & Future Work**

- Two subtasks:
  - 1) Causal Event Classification, and
  - 2) Cause-Effect-Signal Span Detection.
- Each subtask attracted predictions from models that beat our baselines.



## Thank you.

- Link to repository: <a href="https://github.com/tanfiona/CausalNewsCorpus">https://github.com/tanfiona/CausalNewsCorpus</a>
- Please share your feedback with us: Fiona Anting Tan (tan.f@u.nus.edu)