## Tasks

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### 1 Baseline

ACL : Citation Intent Classification Hyper: HyperPartisan News Detection RCT : Randomized Controlled Trials

#### 1.1 Baseline

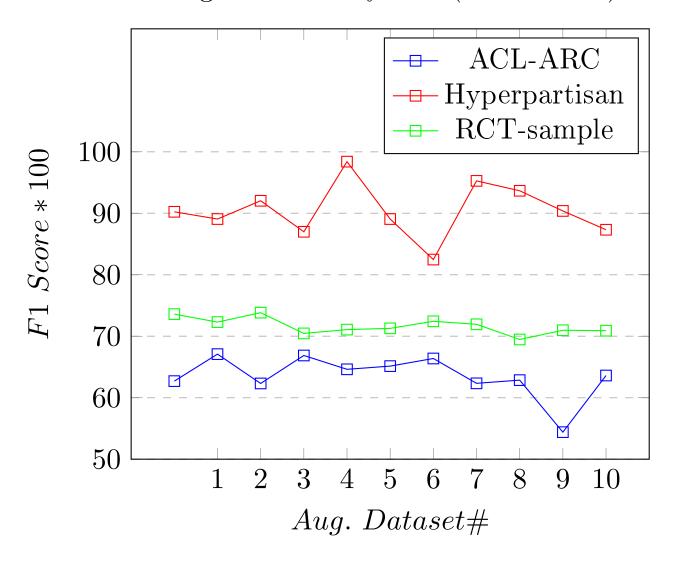
Task	train	dev	test	Classes
ACL	1688	114	139	6
Hyper	516	64	65	2
RCT-sample	500	30212	30135	5
RCT-200k	180040	30212	30135	5

### 2 Tables & Plots

### 2.1 Augmentation by Size (Cumulative)

Dataset #	ACL	Hyper	RCT	Interval
Baseline	62.70	90.24	73.60	-
1	67.09	89.06	72.30	0-3%
2	62.32	92.03	73.83	0-6%
3	66.87	86.99	70.46	0-9%
4	64.62	98.40	71.08	0-12%
5	65.14	89.06	71.28	0-15%
6	66.38	82.47	72.43	0-18%
7	62.34	95.27	71.94	0-21%
8	62.86	93.66	69.47	0-24%
9	54.40	90.38	70.97	0-27%
10	63.61	87.32	70.90	0-30%

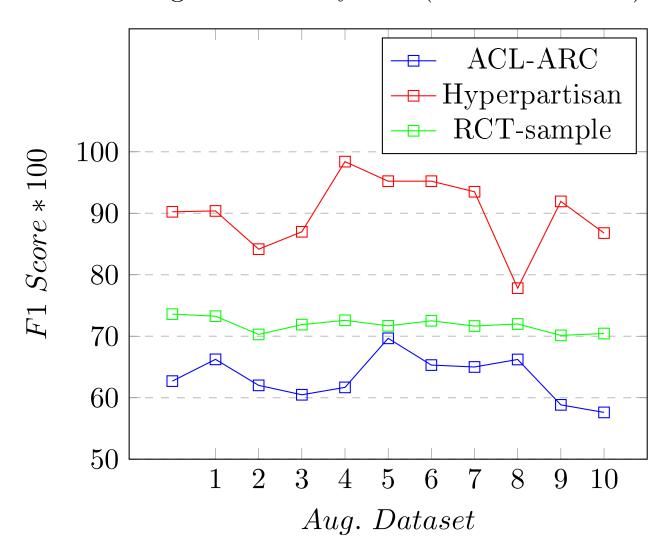
# Augmentation by Size (Cumulative)



### 2.2 Augmentation by Size (Non-cumulative)

Dataset #	ACL	Hyper	RCT	Interval
Baseline	62.70	90.24	73.60	-
1	66.24	90.38	73.26	0-3%
2	62.01	84.16	70.29	3-6%
3	60.48	86.99	71.89	6-9%
4	61.68	98.40	72.59	9-12%
5	69.66	95.22	71.69	12-15%
6	65.31	95.22	72.51	15-18%
7	65.00	93.50	71.66	18-21%
8	66.22	77.82	71.99	21-24%
9	58.84	91.93	70.14	24-27%
10	57.61	86.78	70.44	27-30%

# Augmentation by Size (Non-cumulative)



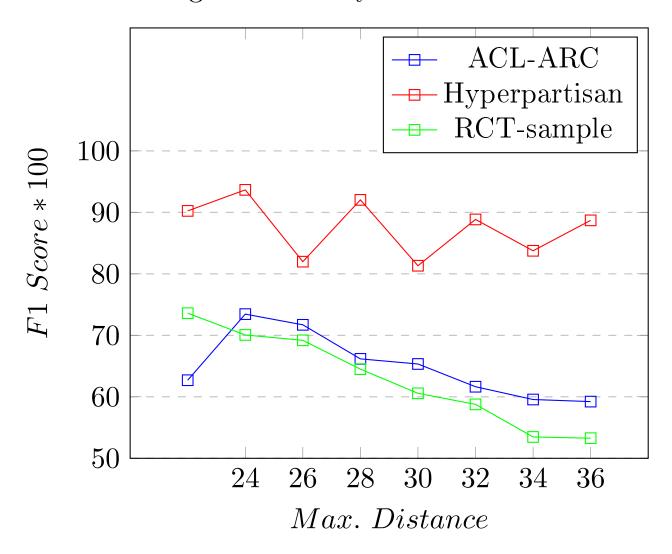
### 2.3 Augmentation by Distance (Data)

Max Distance	ACL	Hyper	RCT
Baseline	1688 (100%)	516 (100%)	500(100%)
24	1746 (103%)	551 (106%)	686(137%)
26	1815 (107%)	567 (109%)	831(166%)
28	1981 (117%)	606 (117%)	1065(213%)
30	2253 (133%)	656 (127%)	1484(296%)
32	2842 (168%)	742 (143%)	2105(421%)
34	3848 (227%)	911 (176%)	2952(590%)
36	5819 (344%)	1127 (218%)	4196(839%)

#### 2.4 Augmentation by Distance (F1 Scores)

Max Distance	ACL	Hyper	RCT
Baseline	62.70	90.24	73.60
24	73.45	93.66	70.06
26	71.71	81.98	69.19
28	66.17	92.03	64.50
30	65.34	81.32	60.57
32	61.64	88.85	58.77
34	59.56	83.75	53.47
36	59.22	88.70	53.27

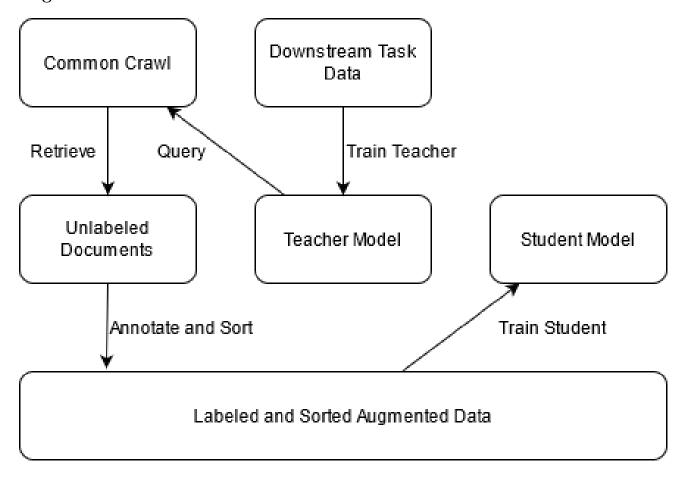
## Augmentation by Maximum Distance



### 3 Baseline models:

• An off-the-shelf RoBERTa model that has been finetuned to perform classification for each of the downstream tasks

#### 4 Augmentation Model



### 5 Algorithm

- 1. Extract failed test examples from the baseline model
- 2. Retrieve passages/sentences from Common Crawl
- 3. Apply augmentation strategy (i)-(iii)
- 4. Augment all the labelled CC data to the training data
- 5. Retrain RoBERTa on the augmented training set

### 6 Augmentation Strategies

- Strategy (i)
  Use baseline model (Teacher) to perform unsupervised labelling on retrieved CC data
- Strategy (ii)
  Using a task specific binary classifier, filter out retrieved CC data that is "out-domain"
  Use baseline model (Teacher) to perform unsupervised labelling on the filtered "in-domain" CC data
- Strategy (iii)
  Using a task specific binary classifier, filter out retrieved CC data that is "out-domain"
  Use ground truth labels of failed test examples and assign labels to the filtered "in-domain" CC data