

NQA Seminar: ProPara

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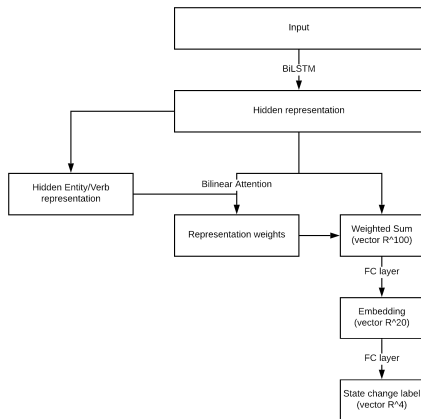
Overview

- 1 ProPara Dataset
- 2 Models
- 3 Semi-supervised Learning
- 4 Results

ProPara Dataset

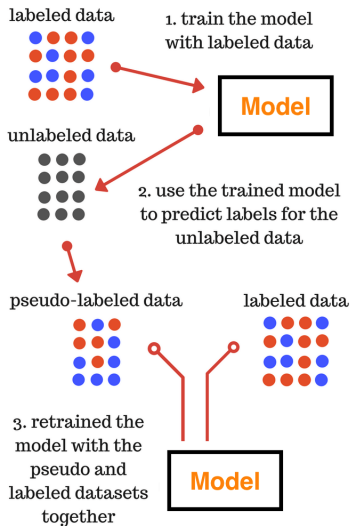
- Procedural text, try to track objects during the paragraph.
- 2k training, 300 testing samples.

Paragraph (seq. of steps):		Participants:					
		water	light	CO2	mixture	sugar	Time ↓
	state0	soil	sun	?	-	-	
<i>Roots absorb water from soil</i>	state1	roots	sun	?	-	-	
<i>The water flows to the leaf.</i>	state2	leaf	sun	?	-	-	
<i>Light from the sun and CO2 enter the leaf.</i>	state3	leaf	leaf	leaf	-	-	
<i>The light, water, and CO2 combine into a mixture.</i>	state4	-	-	-	leaf	-	
<i>Mixture forms sugar.</i>	state5	-	-	-	-	leaf	

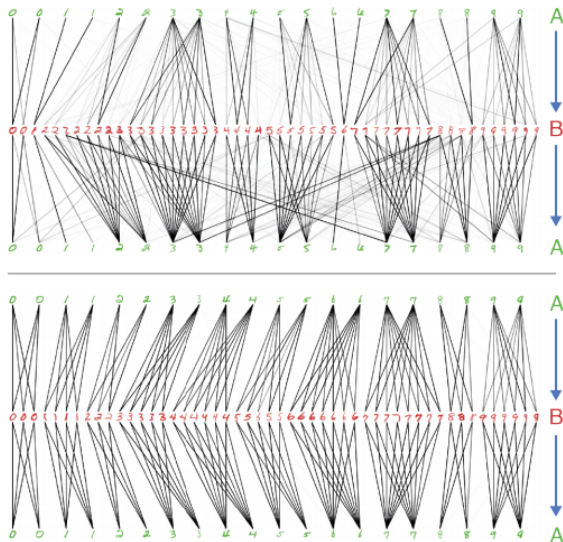


ProLocal + BERT

Pseudo-label



Learning by Association



Learning by Association

- Given labeled samples A , unlabeled samples B
- Calculate embeddings using NN, E_A, E_B
- Similarity between samples given by $E_{A_i} \cdot E_{B_j} =: M_{ij}$
- Define walker randomly walking between labeled and unlabeled samples by this similarity
- Probability of walking from A_i to $B_j = P_{ij}^{AB} = \frac{\exp(M_{ij})}{\sum_k \exp(M_{ik})}$, and similarly P^{BA}
- Probability of round-trip walking is $P^{ABA} := P^{AB} P^{BA}$
- Probability of “correct” walks is $\frac{1}{|A|} \sum_{i \sim j} P_{ij}^{ABA}$

Results - Learning by Association

F1 scores on test set

	without unlabeled	with unlabeled
256 labeled	0.353 (0.035)	0.400 (0.044)
512 labeled	0.423 (0.042)	0.436 (0.024)
all labeled	0.522 (0.018)	0.535 (0.015)

- Apply SSL on ProLocal + BERT model
- Use other similar datasets as unlabeled samples