

Re-TACRED: Addressing Shortcomings of the TACRED Dataset

George Stoica, Emmanouil Antonios Platanios, Barnabás Póczos
gis@cs.cmu.edu, emplata@microsoft.com, bapoczos@cs.cmu.edu

Relation Extraction

Relation Extraction (RE): Determine Relationships between **subjects** and **objects** in text

"Farland joined the FBI in 1942."
per:employee_of

TACRED [1]

- One of the *largest* and *most-popular* crowd-sourced RE datasets
- Collected from 2009-2014 TAC KBP evaluations
- .54 Fleiss' Kappa from 761 randomly sampled annotations
- 106,264 instances spread among 42 relations

TACREV [2]

- 5K most challenging examples yield >50% error and reduce performance by 8.1% average F1-score
- However,
 1. Selection bias constrained error estimation
 2. Analysis limited by largely uncorrected dataset

Re-TACRED

- **Comprehensive:** We verify *entire* TACRED dataset
- **Improved Annotation:** We deploy an improved crowd-sourced annotation strategy
- **High Quality:** We achieve .77 Fleiss' Kappa over *entire* dataset
- **Performance:** We improve model F1-score performance on average by 13%

Improved Annotation Strategy

Relation Refinements

Addressing Ambiguous Relations

"Holly showed off her jewelry."
per:identity

Merging Very Similar Relations

"... Badr is the armed wing of the ISCI."
org:member_of or org:parents?

Relaxing Challenging Criteria

"ORGANIZATION from CITY."
org:city_of_headquarters?

Enforcing Mutual-Exclusivity

"He is a native of Pittsburgh, PA."
per:city_of_birth or per:city_of_residence?

Previous work provides workers with *all* type-compatible relations between sentence entities

• However, what if assigned entity types are incorrect?

"Thomas More Law Center" → PERSON or ORGANIZATION?

We address this issue in two ways:

1. Add a "wrong_type" relation to each relation set
2. Extend label sets to include relations from frequently confused types (PERSON:CITY, STATE/PROVINCE, COUNTRY, LOCATION) → (PERSON:LOCMULTI)

Edge-Case Handling

All workers must be:

- **Experienced:** Have previously completed at least 500 tasks
- **Reliable:** Have at least 95% task approval rate
- **Specialized:** Have passed custom qualification exams for each labeling task
- **Careful:** Have correctly answered at least 80% of observed gold-sentences

Results

Model	Dataset	All Labels			Non-Refined Labels						Refined Labels			
		Metric			Train Split	Test Split	Metrics			Dataset	Category F1			
		F1	Precision	Recall			F1	Precision	Recall		Ambiguous	Similar	Challenging	Exclusive
PA-LSTM*	TACRED	66.2	68.1	64.5	TACRED _{train}	TACRED _{test}	72.3	71.3	73.3	TACRED	46.7	21.2	55.9	51.9
	Re-TACRED	77.9	78.3	77.6	TACRED _{train}	Re-TACRED _{test}	73.3	76.7	70.2	Re-TACRED	87.6	48.8	68.8	53.4
	Change %	+11.7	+10.2	+13.1	Re-TACRED _{train}	Re-TACRED _{test}	75.9	75.8	76.1	Change %	+30.9	+27.6	+12.9	+1.5
C-GCN*	TACRED	66.3	68.5	64.4	TACRED _{train}	TACRED _{test}	72.6	71.1	74.3	TACRED	14.6	22.7	56.7	51.5
	Re-TACRED	79.1	79.7	78.5	TACRED _{train}	Re-TACRED _{test}	73.2	76.0	70.6	Re-TACRED	88.1	51.9	73.7	54.2
	Change %	+12.8	+11.2	+14.1	Re-TACRED _{train}	Re-TACRED _{test}	77.3	78.2	76.5	Change %	+73.5	+29.2	+17.0	+2.7
SpanBERT*	TACRED	69.7	70.1	69.2	TACRED _{train}	TACRED _{test}	75.0	74.7	75.3	TACRED	44.1	51.9	66.8	55.9
	Re-TACRED	84.2	84.6	83.9	TACRED _{train}	Re-TACRED _{test}	76.8	81.2	72.8	Re-TACRED	91.7	65.1	74.0	69.8
	Change %	+14.5	+14.5	+14.7	Re-TACRED _{train}	Re-TACRED _{test}	84.1	85.0	83.1	Change %	+47.6	+13.2	+7.2	+13.9