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# Do Language Models Learn about Legal Entity Types during Pretraining?

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# Do language model acquire any legal knowledge?

- Language models learn the probabilistic structure of a language
- One hypothesis is that LMs acquire diverse **linguistic and semantic knowledge** during pretraining (Petroni et al, 2019, Jiang et al., 2020)
- **Domain oriented pretraining** and task specific pretraining enhance a model performance and adaptability (Gururangan et al., 2020)
- The **lack of resources** and updated annotated datasets is a notable bottleneck of the application of NLP in law

## Research questions

What type of signals do LMs learn during pretraining?

Are those signals reliable enough to be used as a source of incidental supervision for downstream tasks in Legal NLP?

Legal entity typing is a foundation for:

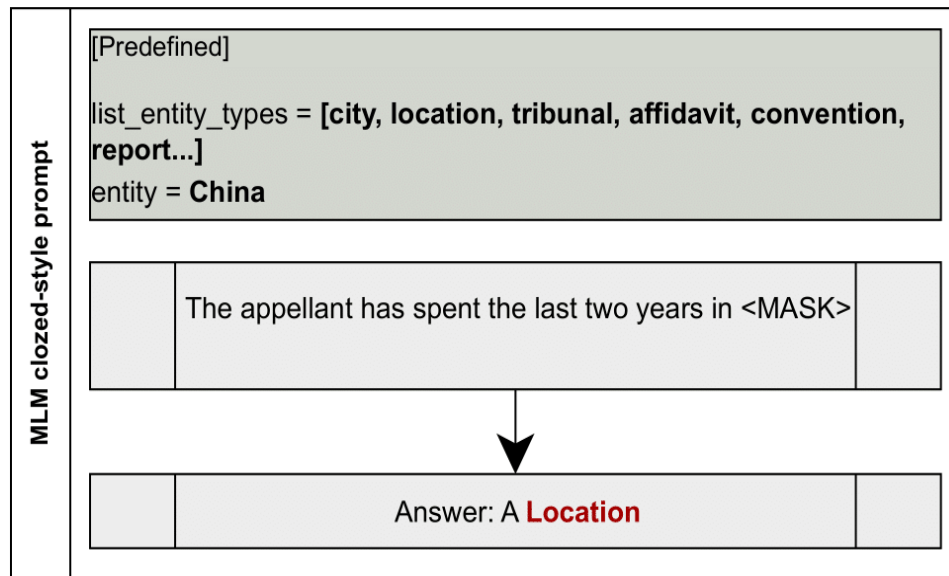
- (i) incidental supervision and text labelling
- (ii) downstream tasks
- (iii) text comprehension

# Probing Legal Entity Types

## A surrogate for legal knowledge

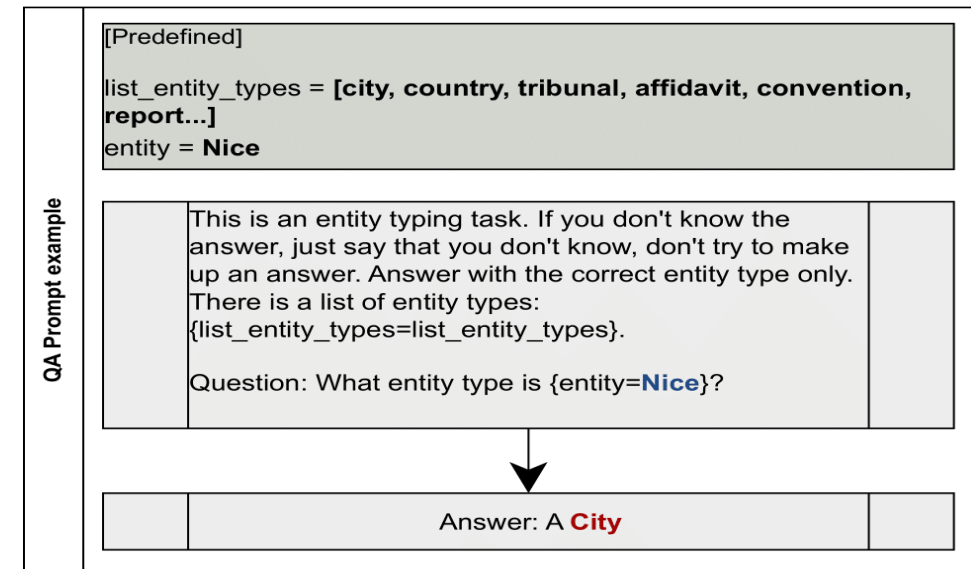
- Task: Entity Typing (predict the entity type), in a zero-shot setting, as a Multiple-Choice QA task
- Dataset: Asylex, a dataset of refugee status determination decisions
- Models: Encoder-only vs Decoder-only

### 1. MLM – Cloze prompts- with BERT-based models



RoBERTa, DeBERTav3, CaseHOLD, Pile of Law, LexLM

### 2. Llama2: QA style prompts



Llama2-7B

# Experiments results

How proficient are Language Models at acquiring knowledge about domain-specific entities like legal entities during pretraining?

Averaged F1 scores

	Gen	CH	PoL	LexLM	Llama2
Generic	11.59	8.51	20.42	11.97	63.26
Gen Legal	17.98	20.89	15.40	12.48	29.52
Refugee Law	10.01	5.93	4.68	4.92	13.03

**Generic:** location, date, adjective location  
**Gen Legal:** org, legal citation, claimant info, procedure, doc evidence, precedent case  
**Refugee Law:** credibility, determination, explanation, legal ground, NGO reports

*Gen groups the results of RoBERTa and DeBERTa-v3*  
*CH refers to CaseHOLD*

# Experiments results: error analysis

	Error Type	Prompt example	Prediction	Gold	%
MLM	Random Prediction	under <mask> of the Republic of China, they cannot take on a second citizenship	lawsuit	law	70.71
	Contextually Accurate	the applicant has not returned to <mask> since 2008	employment	location	12.43
	Closely Related	my colleague relied on this <mask> in her conclusion	ngo report	doc_evidence	16.86
Llama2	Random Prediction	What is <i>Subsection 648</i> ?	country	law	22.22
	Closely Related	What is <i>vietnamese</i> ?	country	nationality (norp)	18.52
	False Negative	What is <i>female claimant</i> ?	female claimant	gender (claimant_info)	33.33
	Prompt Error	What is <i>removal order</i> ?	It is a type of judicial decision.	procedure	25.93

We compare:

- Language models and corpora
- Single-token vs Multi-token → Not significant
- Prompt templates → Optimized templates
- Entity Types → Generic, Law, Refugee Law
- Syntactic vs. Semantic Signals

Vocabulary  
Overlap  
Matrix

Gen	100.0	34.8	46.2	41.2
CH	34.8	100.0	55.7	44.5
PoL	46.2	55.7	100.0	55.1
LexLM	41.2	44.5	55.1	100.0
	Gen	CH	PoL	LexLM

Thank you!

Link to the paper

