

# Metaphors in Pre-Trained Language Models: Probing and Generalization Across Datasets and Languages

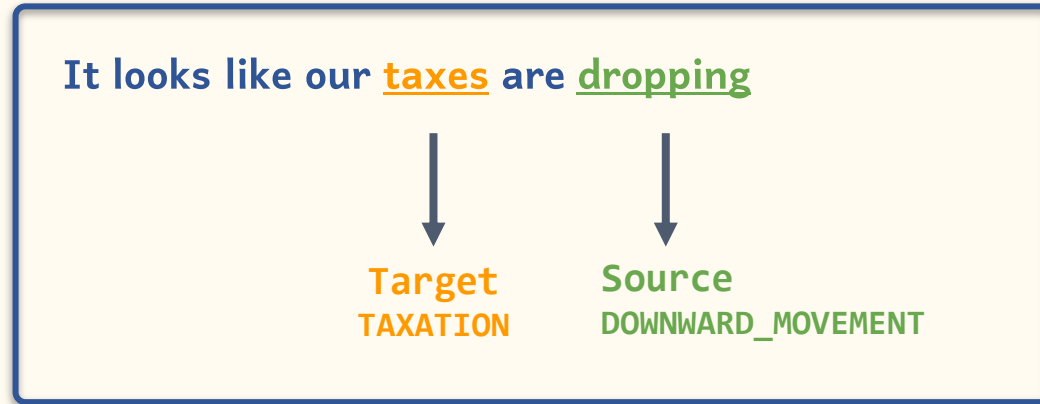
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# Metaphor definition

- Conceptual Metaphor Theory [1]



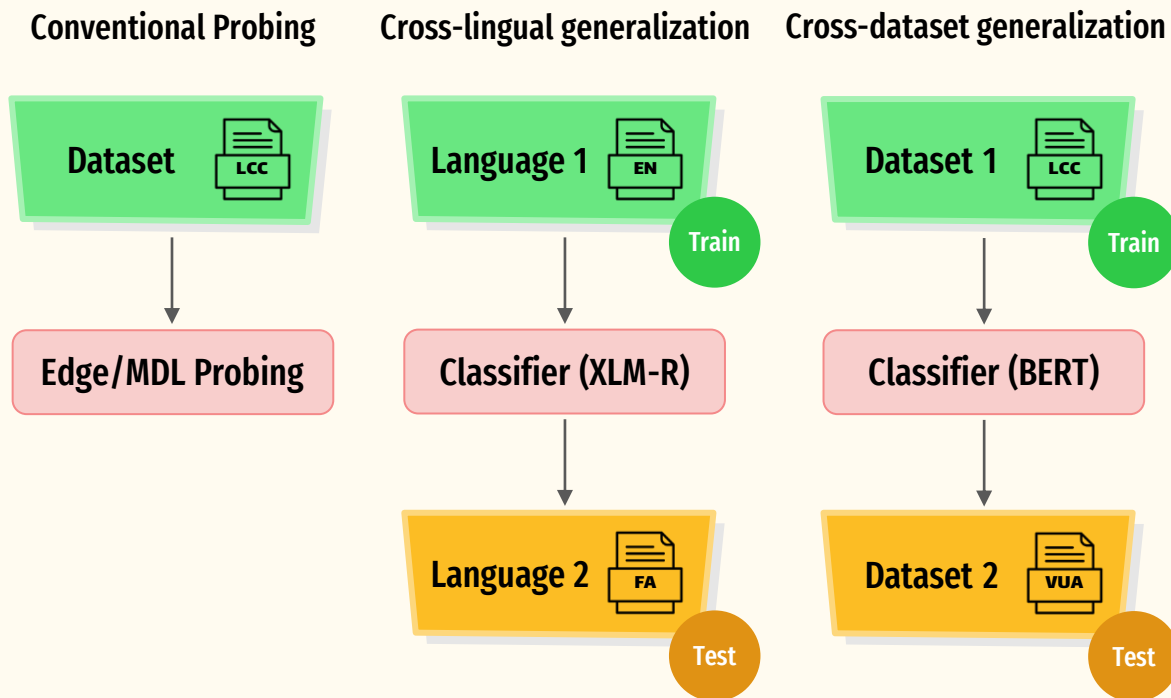
# Motivation

- Metaphors are essential in human communication and constructing human-like computational systems.
- The creativity and generalization to new problems depend on the metaphors.

*Do pre-trained language models represent metaphors?*



# Our methodology



# Datasets

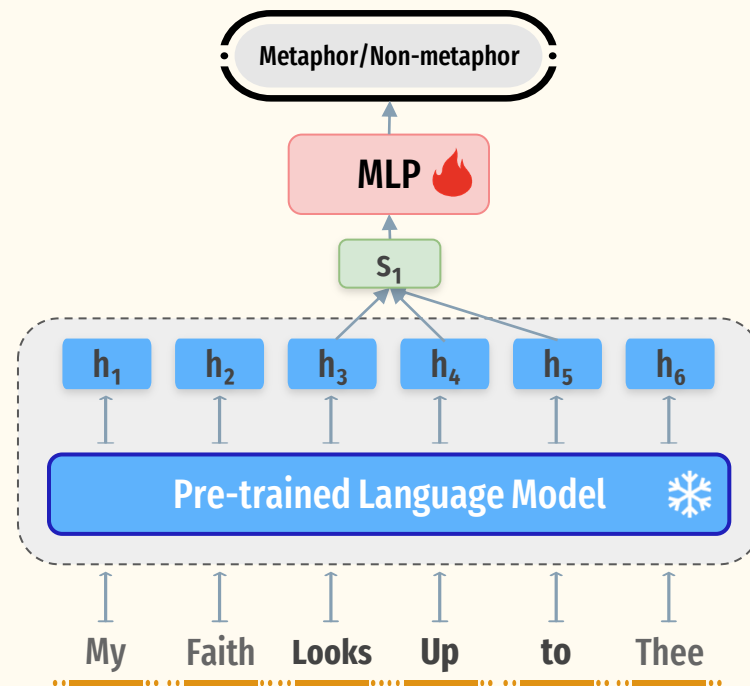
VUA Verbs	He [ <b>finds</b> ] <sub>1</sub> it hard to communicate with people , not least his separated parents . → 1 He finds it hard to [ <b>communicate</b> ] <sub>1</sub> with people , not least his separated parents . → 0
VUA POS	They picked up power from a [ <b>spider</b> ] <sub>1</sub> 's web of unsightly overhead wires . → 1 They picked up power from a spider 's web of unsightly overhead [ <b>wires</b> ] <sub>1</sub> . → 0
TroFi	“ Locals [ <b>absorbed</b> ] <sub>1</sub> a lot of losses , ” said Mr. Sandor of Drexel → nonliteral Vitamins could be passed right out of the body without being [ <b>absorbed</b> ] <sub>1</sub> → literal
LCC	Lawful gun ownership is not a [ <b>disease</b> ] <sub>1</sub> . → 3.0 But the Supreme Court says it's not a way to [ <b>hurt</b> ] <sub>1</sub> the Second Amendment → 2.0 Is he angry that gun rights [ <b>progress</b> ] <sub>1</sub> has been done without him? → 1.0 I mean the 2nd amendment [ <b>suggests</b> ] <sub>1</sub> a level playing field for all of us. → 0.0

- ★ Positive and Negative examples are balanced
- ★ LCC dataset has four languages (English, Spanish, Russian, Farsi)
- ★ LCC scores less than 0.5 are considered literal and more than 1.5 non-literal



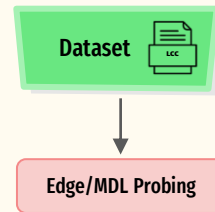
# Probing

Evaluating linguistic knowledge in  
neural representations



# Conventional probing

## Conventional Probing

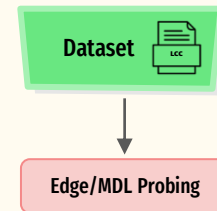


- To answer:
  - Do PLMs represent metaphorical information within their representations?
  - If so, how is it distributed throughout their layers?
- Methods
  - Edge probing [1]
  - MDL probing [2]



[1] Ian Tenney, Patrick Xia, Berlin Chen, Alex Wang, Adam Poliak, R. Thomas McCoy, Najoung Kim, Benjamin Van Durme, Samuel R. Bowman, Dipanjan Das, and Ellie Pavlick. 2019b. What do you learn from context? Probing for sentence structure in contextualized word representations. In International Conference on Learning Representations.  
[2] Elena Voita and Ivan Titov. 2020. Information-theoretic probing with minimum description length. In Proceedings of the 2020 Conference on Empirical Methods in Natural Language Processing (EMNLP), pages 183–196, Online. Association for Computational Linguistics.





# Conventional probing - results

Dataset	Baseline		BERT		RoBERTa		ELECTRA	
	Acc.	Comp.	Acc.	Comp.	Acc.	Comp.	Acc.	Comp.
LCC (en)	74.86	1.05 <sub>2</sub>	88.25	1.85 <sub>6</sub>	88.06	1.96 <sub>5</sub>	<b>89.30</b>	<b>2.05<sub>5</sub></b>
TroFi	67.34	1.01 <sub>4</sub>	<b>68.58</b>	1.07 <sub>4</sub>	68.46	<b>1.09<sub>6</sub></b>	68.07	1.08 <sub>3</sub>
VUA POS	65.92	1.03 <sub>0</sub>	80.32	1.43 <sub>5</sub>	81.72	1.48 <sub>6</sub>	<b>83.03</b>	<b>1.51<sub>4</sub></b>
VUA Verbs	65.97	1.04 <sub>9</sub>	78.29	1.28 <sub>9</sub>	78.88	<b>1.34<sub>5</sub></b>	<b>79.96</b>	1.31 <sub>4</sub>

## Conclusions

- PLMs do represent metaphors
- RoBERTa and ELECTRA >>> BERT
  - Better pre-training objectives 👍
  - Extensive pre-training data 👍





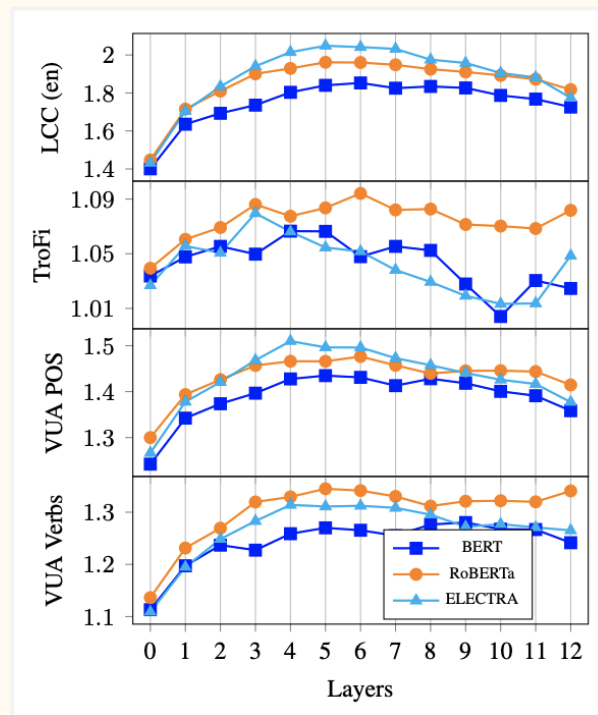


Edge/MDL Probing

# Layer-wise analysis

## Conclusions

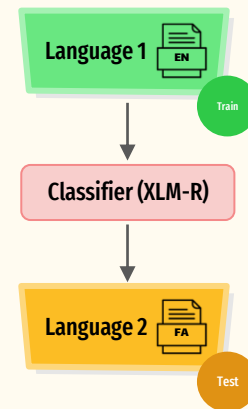
- Middle layers > Deeper layers
  - Not highly contextualized
  - Forecasting source domain ✗ target domain



# Cross-lingual generalization

- Are metaphors transferable across languages?
- Using edge probing
  - train a classifier on **language S**
  - predict metaphoricity in **language T**.

Cross-lingual generalization



# Cross-lingual generalization - results

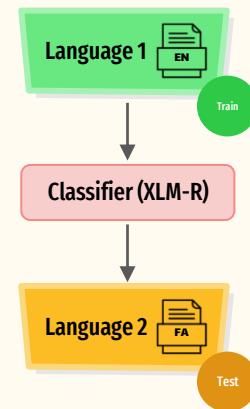
		Train Language	
		English	Spanish
Test Language	English	85.14 (65.37)	79.31 (52.71)
	Spanish	79.40 (53.17)	84.59 (66.09)

★ XLM-R (randomly initialized XLM-R)

## ● Conclusions

- XLM-R >>> Random
  - Transferability of metaphorical information between languages 🧠
  - Capacity of XLM-R 👍

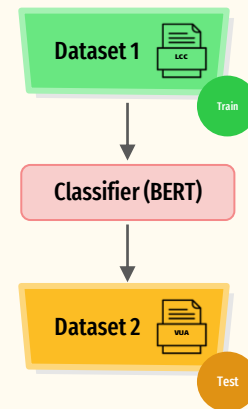
Cross-lingual generalization



# Cross-dataset generalization

- Are metaphors transferable across datasets?
- Using edge probing
  - train a classifier on **dataset S**  
to predict metaphoricity in **dataset T**.

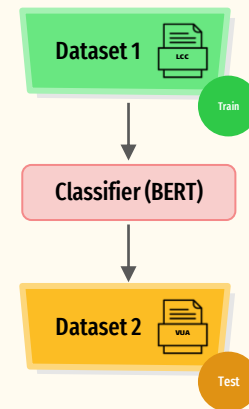
## Cross-dataset generalization



# Cross-dataset generalization - results

		Train Dataset	
		LCC(en)	VUA Verbs
Test Dataset	LCC(en)	84.26 (54.93)	70.37 (50.14)
	VUA POS	62.23 (51.47)	71.6 (53.47)
	VUA Verbs	60.20 (50.88)	75.21 (60.03)

★ BERT (randomly initialized BERT)



## Conclusions

- PLM > Random
  - Generalizable metaphorical information ✓
- Consistent data annotation → better results



# Conclusion

- Contextual representations in PLMs do encode metaphorical knowledge
- Metaphorical knowledge is encoded better in the middle layers of PLMs
- Metaphorical knowledge is transferable between languages and datasets ~ Consistency of the annotation



# THANK YOU!

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**Paper:** <https://arxiv.org/abs/2203.14139>

**Code:** [github.com/EhsanAghazadeh/Metaphors in PLMs](https://github.com/EhsanAghazadeh/Metaphors_in_PLMs)



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