A Corpus of Biology Analogy Questions as a Challenge for Explainable AI

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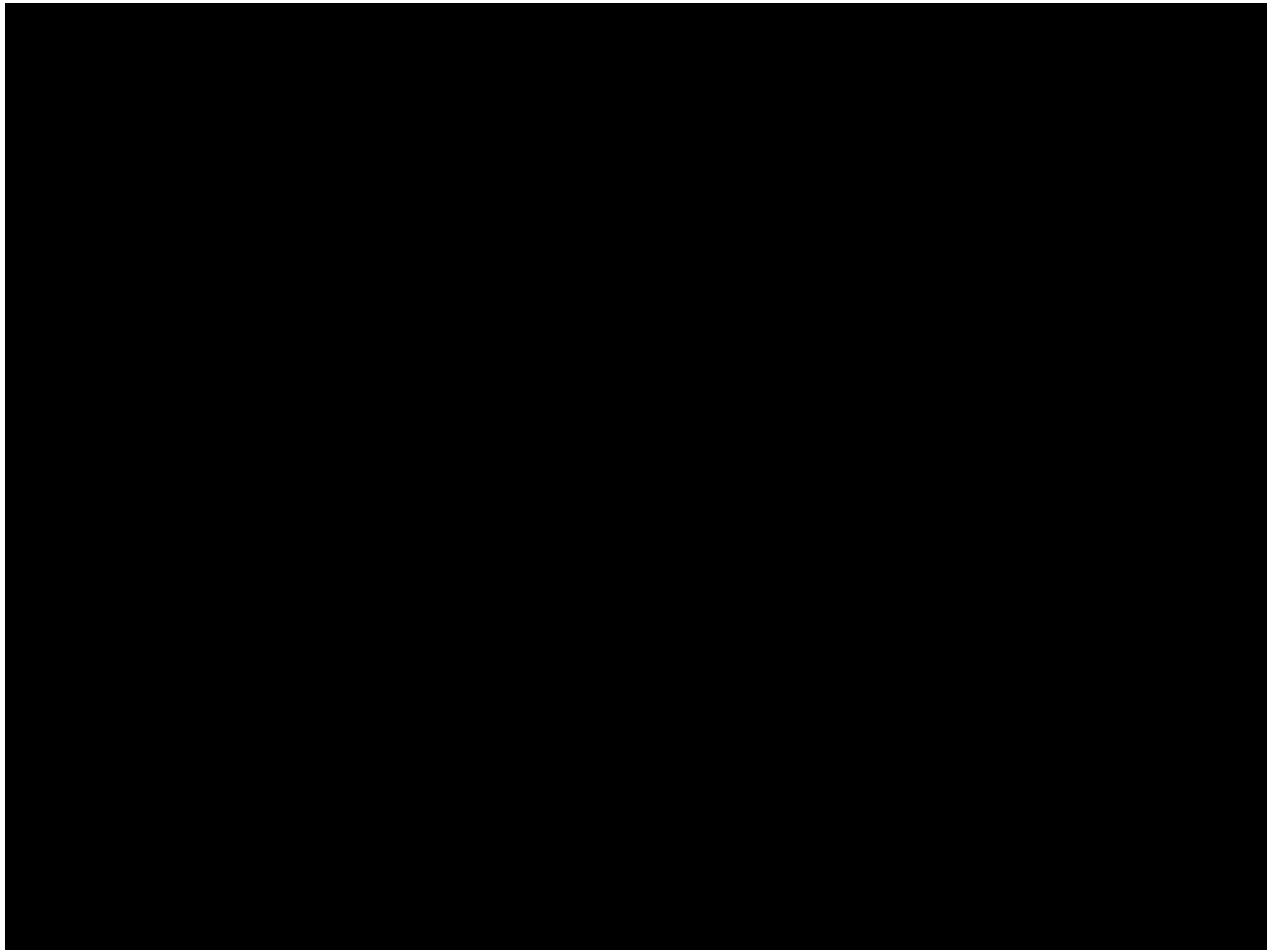
A is to B as C is to what?

Man: King as Woman: Queen

Motivation

Analogy as a learning tool for students in discovering relations esp. in science

Intelligent Biology Textbook (Inquire)



Related work on analogy datasets

Google Analogy Dataset[1]

The Bigger Analogy Test Set (BATS)[2]

1. Mikolov, T., Chen, K., Corrado, G., & Dean, J. (2013). Efficient estimation of word representations in vector space. In Proceedings of International Conference on Learning Representations (ICLR).
2. Gladkova, A., Drozd, A., & Matsuoka, S. (2016). Analogy-based detection of morphological and semantic relations with word embeddings: what works and what doesn't. In Proceedings of the NAACL-HLT SRW (pp. 47–54). San Diego, California, June 12-17, 2016: ACL. Retrieved from https://www.aclweb.org/anthology/N/N16/N16-2002.pdf

Related Work on answering analogy questions

Word embedding[1]

SemEval 2012 Task 2[2]

- Measuring Degrees of Relational Similarity

Pair-pattern matrix[3]

Deep learning

- BERT[4]

1. Jeffrey Pennington, Richard Socher, and Christopher D. Manning. Glove: Global vectors for word representation. In EMNLP, 2014
2. David A. Jurgens, Saif M. Mohammad, Peter D. Turney, and Keith J. Holyoak (2012), SemEval-2012 Task 2: Measuring Degrees of Relational Similarity, First Joint Conference on Lexical and Computational Semantics (*SEM), Montreal, Canada, June 2012, pp. 356–364.
3. Peter D. Turney and Patrick Pantel. From frequency to meaning: Vector space models of semantics.J. Artif. Int. Res., 37(1):141–188, January 2010.
4. Jacob Devlin, Ming-Wei Chang, Kenton Lee, and Kristina Toutanova. Bert: Pre-training of deep bidirectional transformers for language understanding, 2018

Dataset

Biology Analogy Questions Corpus:

Analogy questions generated from Biology Knowledge Base

Text data:

LIFE Biology textbook

OpenStax Biology textbooks

Generation of biology analogy corpus using knowledge base

Generation of Biology Analogy Questions

KB_Bio_101[1]: hand-curated knowledge base

Extraction of analogy from knowledge base

- Crawl of knowledge base of different semantic relationships in **KB_Bio_101**
- Relations include: subclass-of, has-part, has-region, possesses, etc
- Allows interpretability of **multi-hop** analogies
  - E.g. **A** SUBCLASS-OF **B** SUBCLASS-OF **C**
- 70,000+ analogy questions

1. Vinay K. Chaudhri, Daniel Elenius, Sue Hinojoza, Michael A. Wessel. KB_Bio_101: Content and Challenges. In *In the Proceedings of International Conference on Formal Ontologies in Information Systems, 2014.*

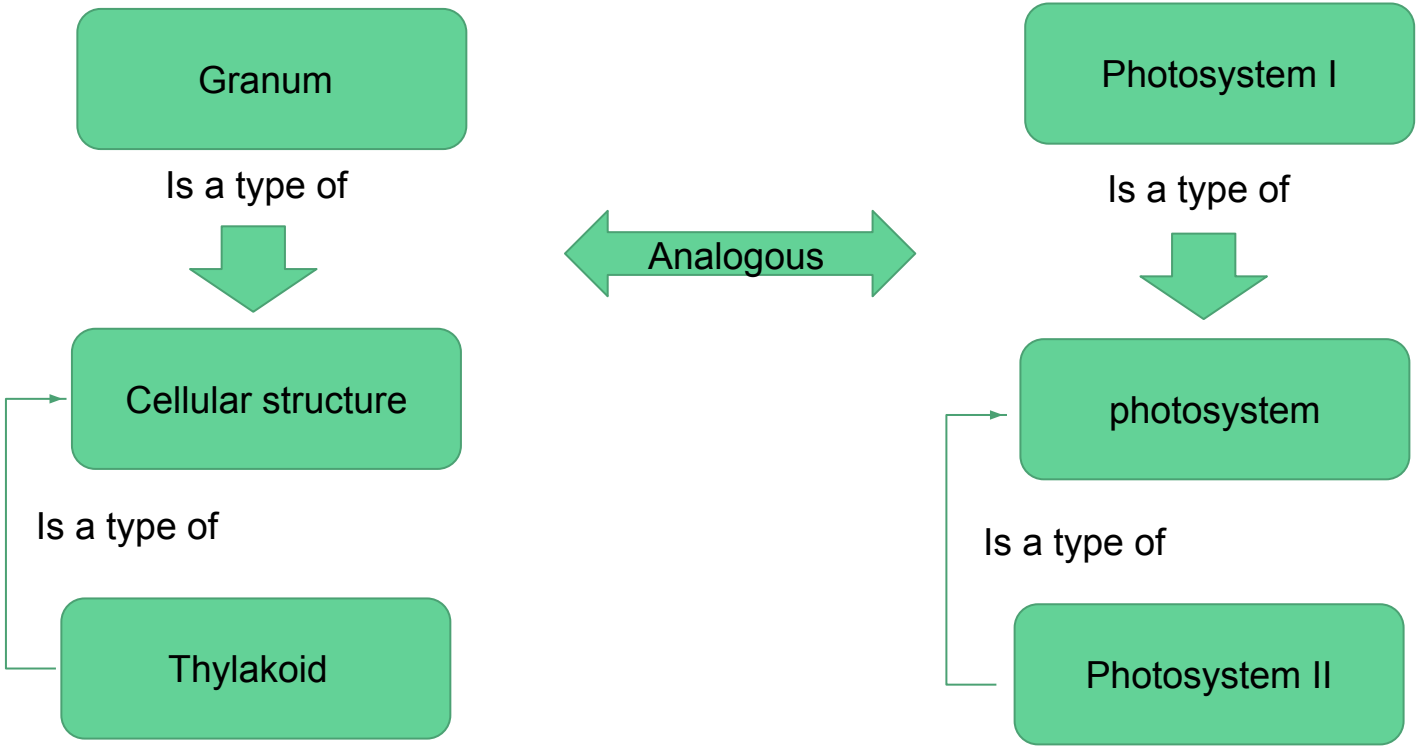
Generation of Biology Analogy Questions

Analogy examples

Type of Analogy	Example Question	Answer
subclass-of	Phospholipid is to a lipid as margarine is to what?	fat
has-part	Chloroplast is to a granum as mitochondrion is to what?	ribosome
has-region	Phospholipid is to a fatty acid tail as polar amino acid is to what?	polar side chain
possesses	ATP synthase is to a peptide linkage as oligosaccharide is to what?	glycosidic linkage
element	Granum is to a thylakoid as photo system I is to what?	light-harvesting complex
is-inside	Aquaporin is to phospholipid bilyaer as stroma is to what?	chloroplast
has-function	Chloroplast is to photosynthesis as lysosome is to what?	autophagy

Generation of Biology Analogy Questions

Interpretability



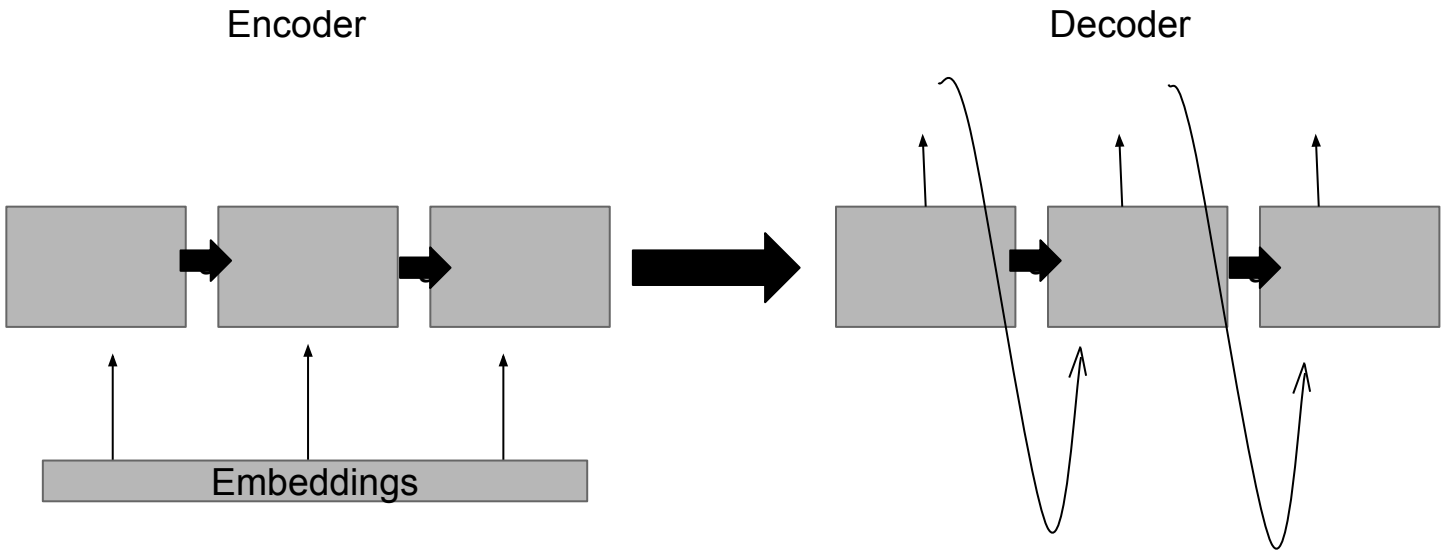
Automated answering of analogy questions

Automated answering of analogy questions

	GLoVe embeddings	FastText embeddings	ELMo embeddings	BERT/BioBERT embeddings
Word embeddings	x	x	x	
Seq2Seq	x		x	
Seq2Vec	x		x	x

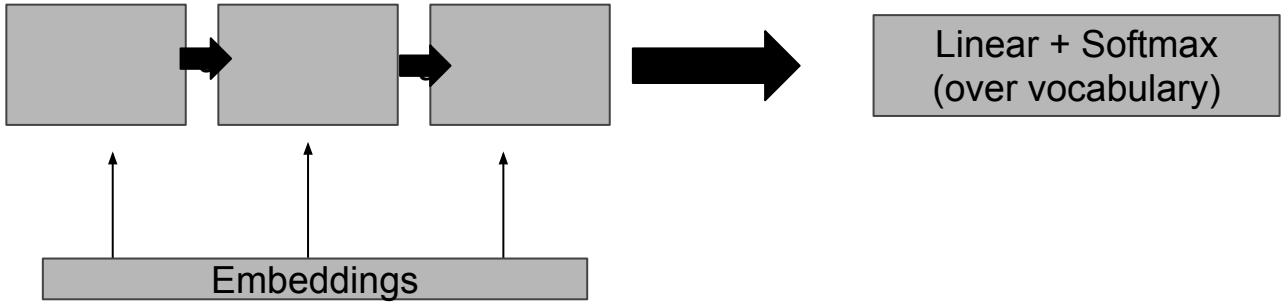
The word embeddings are trained on the LIFE Biology and OpenStax textbooks.

Seq2Seq



Each block is a LSTM cell. The encoder is bi-directional LSTM.

Seq2Vec



Each block is a LSTM cell. The encoder is bi-directional LSTM.

Results

Seq2Seq

	Vanilla Seq2Seq	ELMo
Any Match Acc	**0.580**	0.579
Corpus BLEU	49.873	**54.08**

Table 2: Best Seq2Seq models results

Seq2Vec

	ELMo	BERT	BioBERT
Any Match Acc	**0.580**	.332	.381
Corpus BLEU	**56.35**	34.19	38.63

Table 3: Best Seq2Vec models results

Results - Examples (Seq2Vec)

Correct examples:

Nonpolar covalent bond | chemical bond | ultraviolet ray : **light**

- Prediction: light

carbon atom | atom | cytochrome A3 : **cytochrome**

- Prediction: cytochrome

Incorrect example:

thymine | pyrimidine | water molecule : **polar molecule** (Wrong relationship)

Prediction: **oxygen atom**

Conclusion & Future Work

Utility of knowledge base
- Interpretability, generation of questions

Automated interpretability of analogies
- Relation extraction as a means for explanation

Open problem for more research!

Comments, Questions, Suggestions?

Appendix: More info on dataset

Data split on corpus:

Train/val/test: 70/10/20

Specific Future directions for analogy evaluation:

- Currently treats multi-hop relations that have same relations for all hops as one relation in dataset evaluation (causes some analogies to be “far-fetched”)
- Quantitative Evaluation on analogy by domain expert
- Evaluation on unseen concepts for analogies

Appendix: Word embedding evaluation

Word embeddings evaluation

	Correct / Total	Top 10 Accuracy	Cosine Similarity
`ELMo` **pretrained**	507/2154	**.235**	.54
`GloVe` **trained on Wikipedia**	159/1510	.105	.42
`GloVe` **trained on Biology textbook**	135/1203	.112	.424
`fastText` **trained on Wikipedia**	330/1931	.171	.548
`fastText` **trained on Biology textbook**	181/1990	.0909	**.658**

Table 2: Performance on analogy between concepts with names that are single words

Appendix: Specific results

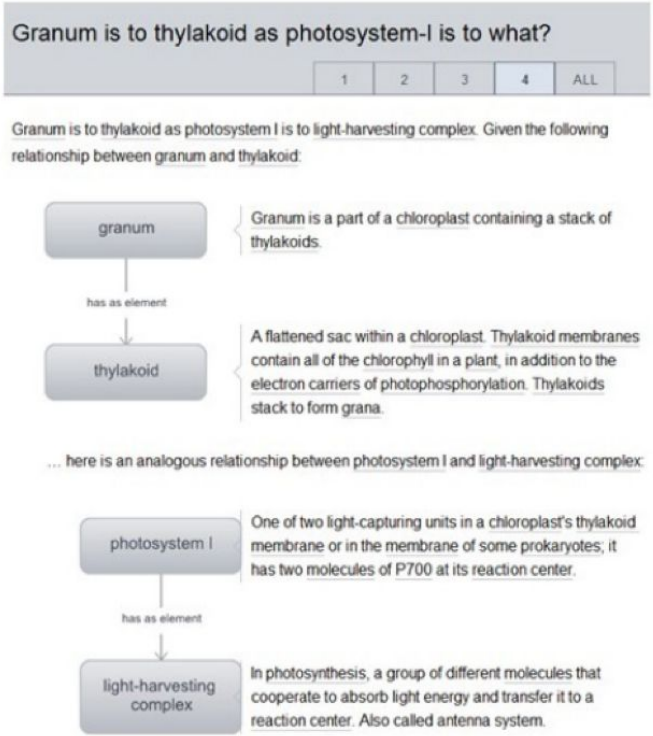
	ELMo	BERT	BioBERT
Top 1 Acc	**0.507**	.277	.320
Top 2 Acc	**0.789**	.454	.516
Top 3 Acc	**0.930**	.572	.647
Top 4 Acc	**0.980**	.649	.722
Any Match Acc	**0.580**	.332	.381
Corpus BLEU	**56.35**	34.19	38.63

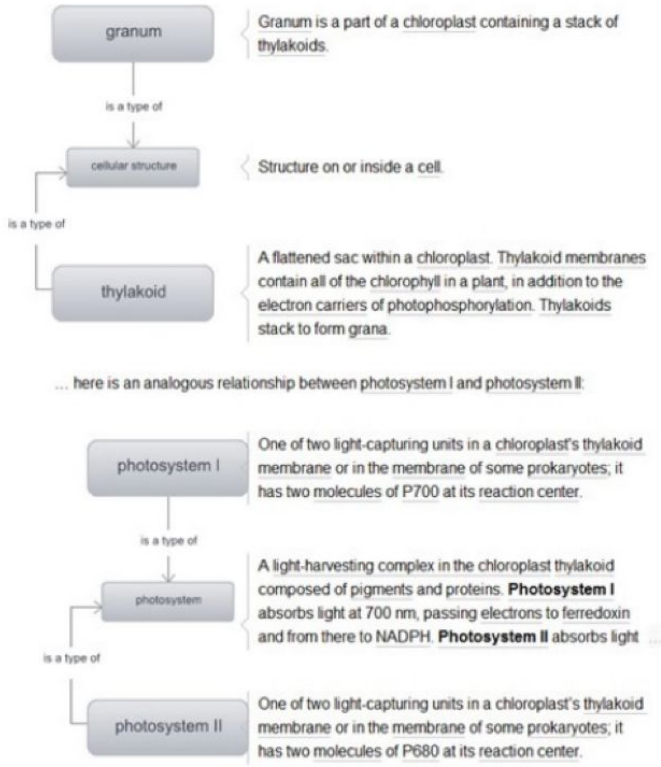
Table 3: Best Seq2Vec models results

	Vanilla Seq2Seq	ELMo
Top 1 Acc	**0.502**	**0.502**
Top 2 Acc	**0.788**	0.784
Top 3 Acc	**0.930**	0.926
Top 4 Acc	**0.980**	0.968
Any Match Acc	**0.580**	0.579
Corpus BLEU	49.873	**54.08**

Table 2: Best Seq2Seq models results

Appendix: Specific interpretability results





Appendix: Additional examples

More correct examples

diacylglycerol | amphipathic molecule | nucleotide | molecule

Prediction: organic molecule

AMP | phosphate group | AMP | carbon skeleton

Prediction: carbon skeleton

“Incorrect” examples from data:

glucose | polar covalent bond | oxygen molecule | nonpolar covalent bond
Prediction: double bond

nonpolar covalent bond | covalent bond | plant cell-wall | cell wall
Prediction: cellular structure

Appendix: Full demo