

# *Efficient Estimation of Word Representations in Vector Space* (Mikolov et al., 2013)

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Mikolov et al. (2013) developed the Word2Vec model, which employs the Continuous Bag of Words (CBOW) and Skip-grams (SG) algorithms to represent word vectors. Despite its influential contribution, the paper has several key limitations:

1. *Single vector per word.*

The model assigns a single static vector to each word, which would conflate different senses a word (e.g., *bank* as “financial institution” vs. “riverbank”) and metaphorical uses (*drive a car* vs. *drive me mad*). A potential improvement would be to cluster contextual embeddings by sense *bank*<sub>1</sub>, *bank*<sub>2</sub>, and assign tokens to the nearest cluster based on the specific usage.

2. *Register effects.*

The underlying assumption of the model is that words with similar meanings occur in similar contexts. While it works well with words that often occur in the same register (e.g., *spinach* vs. *bok choy*), it might fail with words that are semantically equivalent but always used in different registers (e.g., *pee* (informal) vs. *urinate* (formal)). These words almost never share overlapping environments, thus might be separated far away by the model.

3. *Limitations of the analogy test.*

The main evaluation method adopted by the paper is the word analogy test. While the analogy test can detect categorical relations, such as A IS THE CAPITAL OF B and *A is the past tense of B*, it cannot well capture the graded or continuous semantic relations such as *warm* vs. *hot* vs. *scorching*. Thus the evaluation does not reflect the real performance of the model. Although the MSR Sentence Completion task partially addresses analogy’s weaknesses, it’s also limited by the multiple-choice format. More comprehensive, continuous evaluations should be adopted.

## References

Mikolov, T., Chen, K., Corrado, G., & Dean, J. (2013). Efficient estimation of word representations in vector space. In *1st International Conference on Learning Representations, ICLR 2013, Scottsdale, Arizona, USA, May 2-4, 2013, Workshop Track Proceedings*. <http://arxiv.org/abs/1301.3781>