

# KerBS: Kernelized Bayesian Softmax for Text Generation

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## Motivation

- Softmax based generation models assume context embeddings of a word should concentrate around its word embedding as in Figure 1.a.
- However, embeddings of some words actually exhibit properties such as **multi-sense** (Figure 1.b) or **varying variance** (Figure 1.c).
- We need to explicitly model the properties for more accurate word prediction!

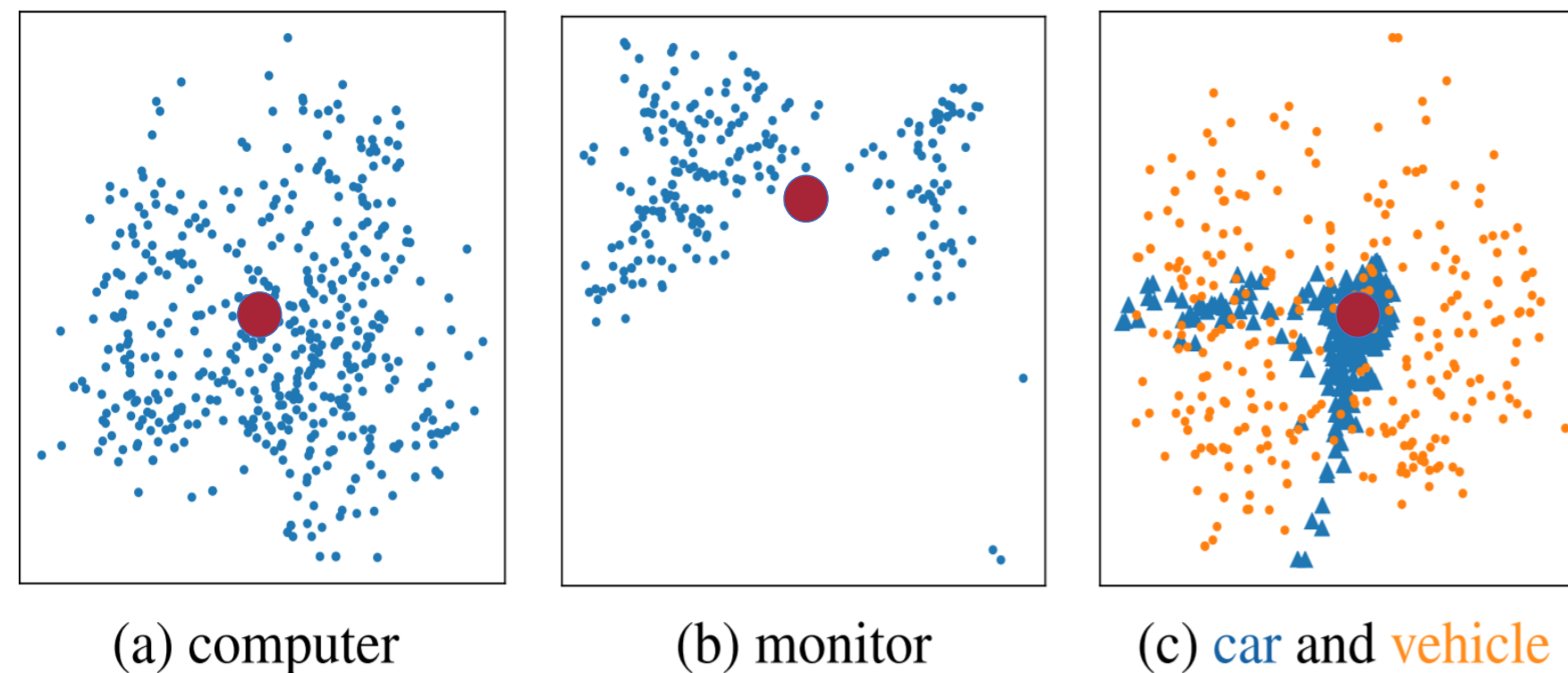


Figure 1: Context embeddings and word embeddings (●) of different word.



## Method

- Multi-sense
  - We replace word embeddings with **sense embeddings**, and sums all sense probabilities of a word to get the probability of the word.

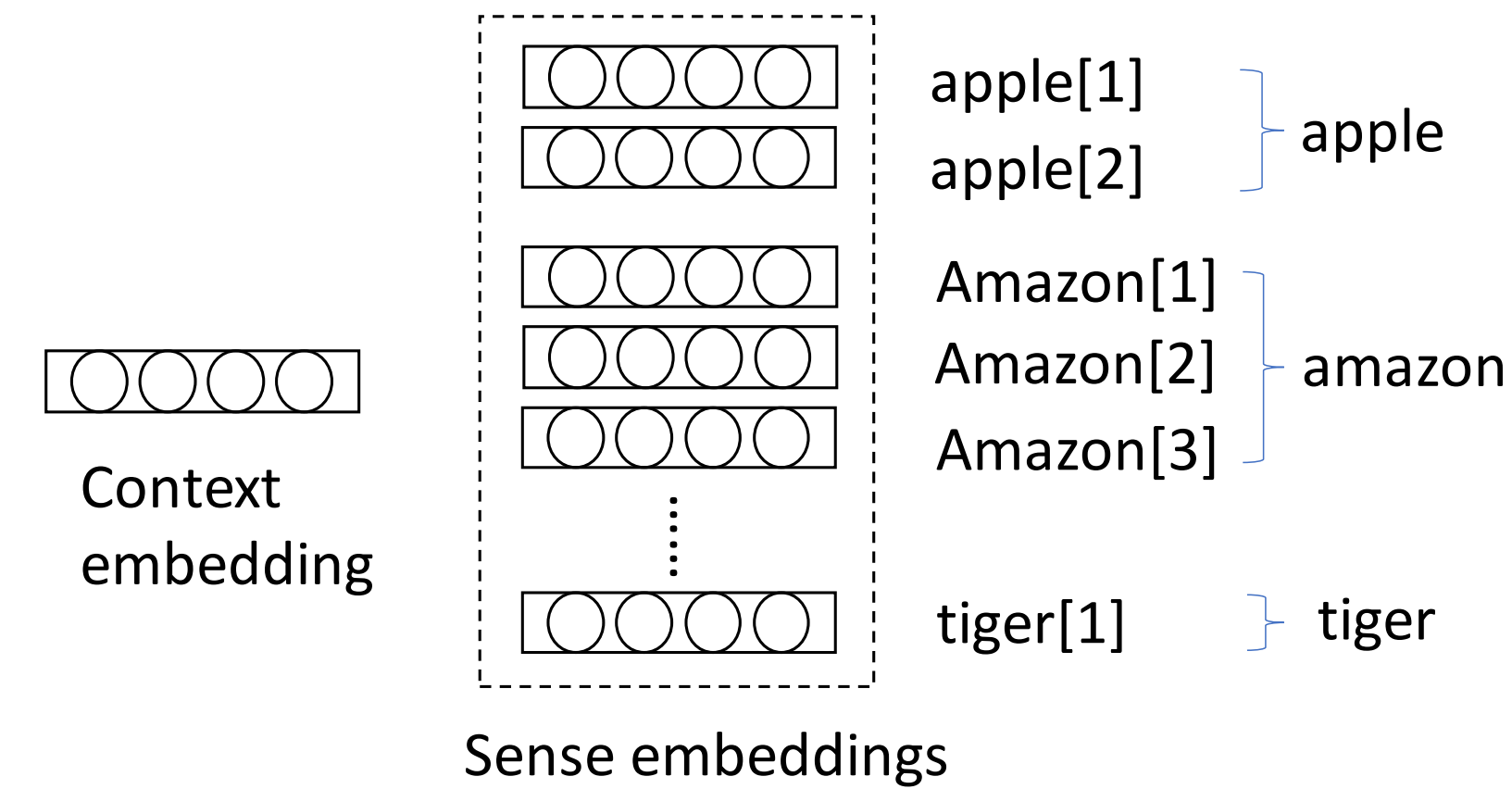


Figure 2: Illustration of sense embedding.

- Kernelized Variances
  - Since Gaussian distribution is instable in high-dimensional space, we designed a **kernel function** to model **different variances** of each word's context embeddings.

$$\mathcal{K}_{\theta}(h, e) = |h||e|(a \exp(-\theta \cos(h, e)) - a)$$

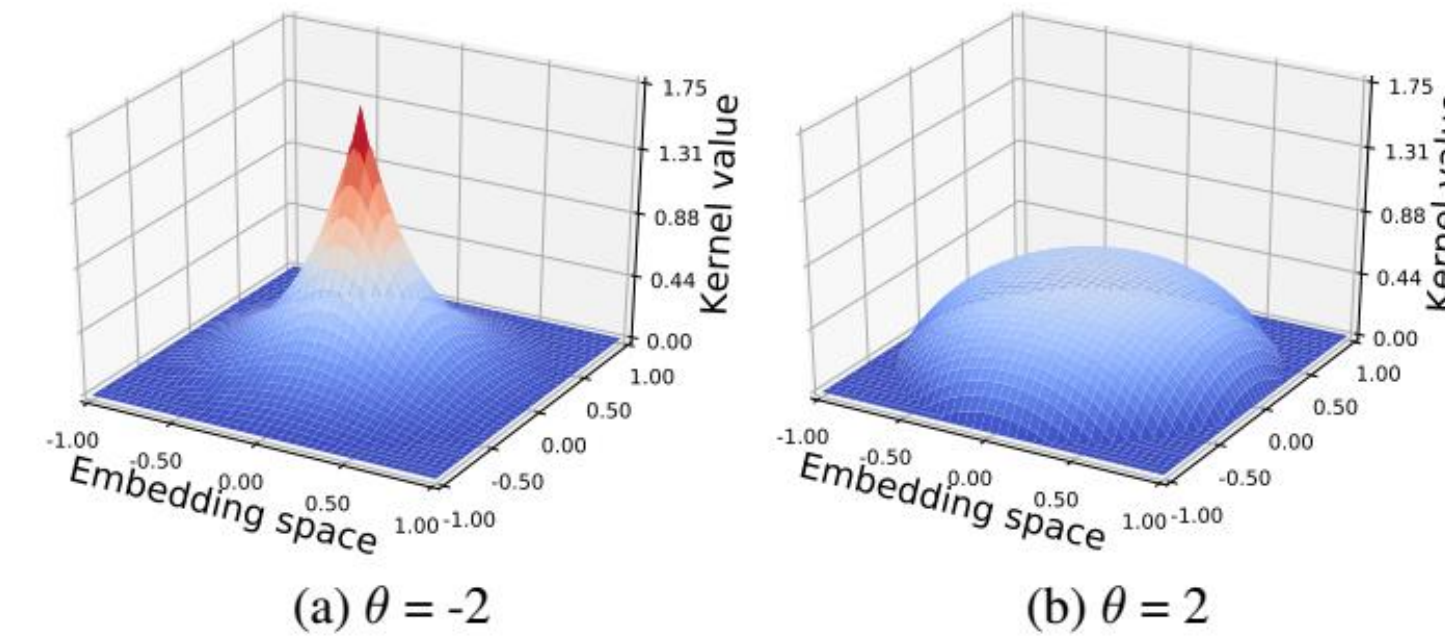


Figure 2: Kernels with different  $\theta$ .

- Dynamic Sense Allocation
  - During training, we record the **usage** of each sense and **log prediction accuracy** of each word.
  - After every M steps, we **reallocate** the least used senses to the poorly predicted words.

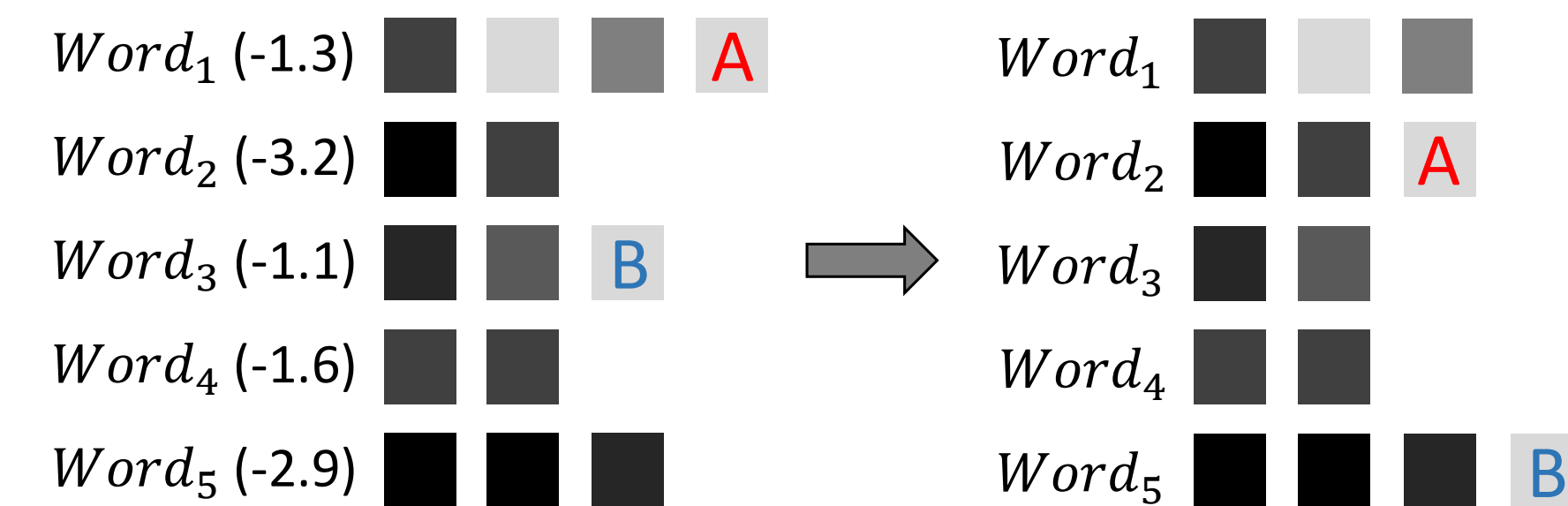


Figure 3: Illustration of sense reallocation. Numbers in the brackets are log prediction accuracies of each word. We use squares to represent senses and darker color means higher usage.

## Experiments and Analyses

- Better performance than baselines.

Tasks	Base models	Metrics	Base	Base+Mos	Bas+KerBS
MT	Transformer	BLEU-4	29.61	28.54	<b>30.90</b>
	Seq2Seq	BLEU-4	25.91	26.45	<b>27.28</b>
LM	GRU	PPL	103.12	102.72	<b>102.17</b>
Dialog	Transformer	BLEU-1	10.61	9.81	<b>10.90</b>
	Seq2Seq	BLEU-1	16.56	13.73	<b>17.85</b>
		Human	1.24	1.04	<b>1.40</b>

- More sense are allocated to words with complex meanings .

Sense	1	2	3	4
word	Redwood heal structural theoretical rotate	particular figure during known size	open order amazing sound base	they work body power change

- $\theta$  reflects the sematic scopes of words.

