

# Normalization of Language Embeddings for Cross-Lingual Alignment

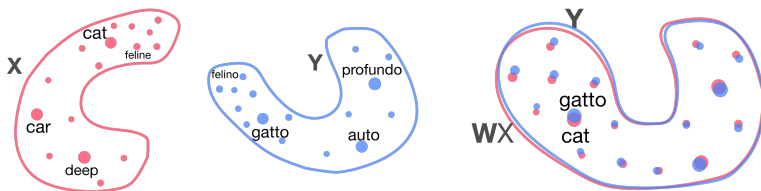
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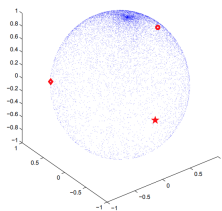
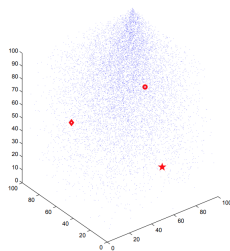
April 15, 2022

# Why Cross-Lingual NLP?

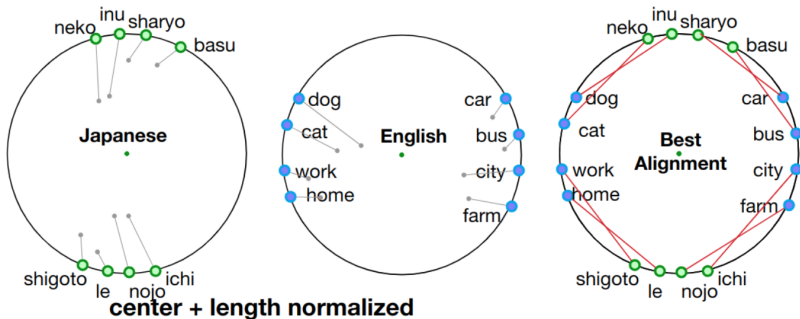


# Preprocessing or Normalization Methods

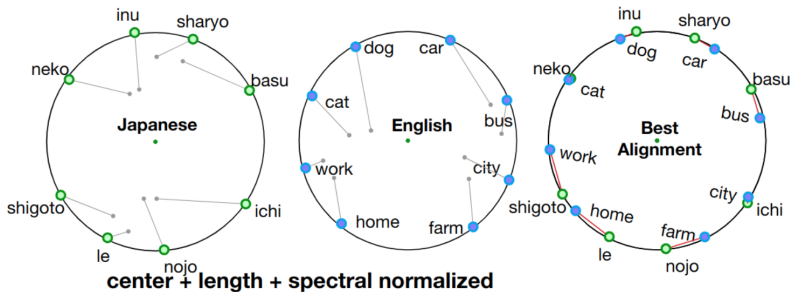
- Mean Centering (C), (*Artetxe et al., 2016*)
- Length Normalization (L), (*Xing et al., 2015*)
- Iterative Normalization (I-C+L), (*Zhang et al., 2019*)



# Spectral Normalization (SN)

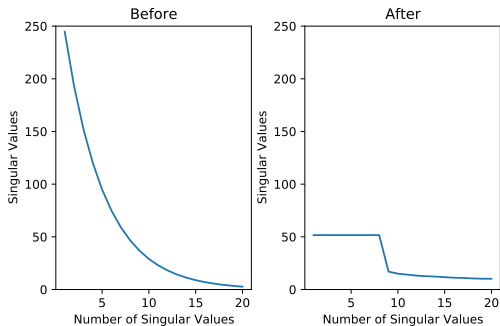


# Spectral Normalization (SN)



# Spectral Normalization (SN)

- Given a matrix,  $A$  of  $n$  words in  $d$  dimensions.
- Compute the Singular Values.
- Even out the top ones and leave the bottom ones below.



# Iterative Spectral Normalization (I-C+SN+L) Algorithm

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## Algorithm 1 I-C+SN+L

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- 1: **for** #Iter steps **do**
  - 2:    $A \leftarrow$  Center  $A$
  - 3:    $A \leftarrow$  SpecNorm( $A$ )
  - 4:    $A \leftarrow$  Unit length normalization of  $A$
  - 5: **return**  $A$
-

# Bilingual Lexicon Induction (BLI) Task

- Retrieve the nearest neighbors in the Target Language.

Language pairs where I-C+SN+L outperforms I-C+L

Models

CCA	PROC	PROC-B	DLV	RCSLS	VECMAP
22/28	24/28	25/28	23/28	14/28	24/28



# Cross-lingual Natural Language Inference (XNLI)

XNLI Performance (Test Set Accuracy)

Model	Dict	EN-DE	EN-FR	EN-TR	EN-RU	Avg
PROC	5k	.607	.534	.568	.585	.574
PROC <sup>I-C+L</sup>	5k	.589	.608	.536	.581	.579
PROC <sup>I-C+SN+L</sup>	5k	.611	.638	.542	.596	<b>.597</b>
PROC-B	3k	.615	.532	.573	.599	.580
PROC-B <sup>I-C+L</sup>	5k	.602	.636	.537	.595	.593
PROC-B <sup>I-C+SN+L</sup>	3k	.624	.638	.548	.601	<b>.603</b>
RCCLS	5k	.390	.363	.387	.399	.385
RCCLS <sup>I-C+L</sup>	5k	.514	.490	.490	.526	.505
RCCLS <sup>I-C+SN+L</sup>	5k	.499	.482	.504	.556	<b>.510</b>

- We introduce a new way to normalize embeddings based on **Spectral Normalization**.
- Our approach:
  - ★ Generalizes previous approaches, and
  - ★ Effectively removes much of the clustering of words.
  - ★ Allows any alignment procedure to find better alignments resulting in improved performance on the BLI and CLDTs.