

A2-write-up

Requirement 3

A screenshot of loss change

The following is an example with the hyperparameters:

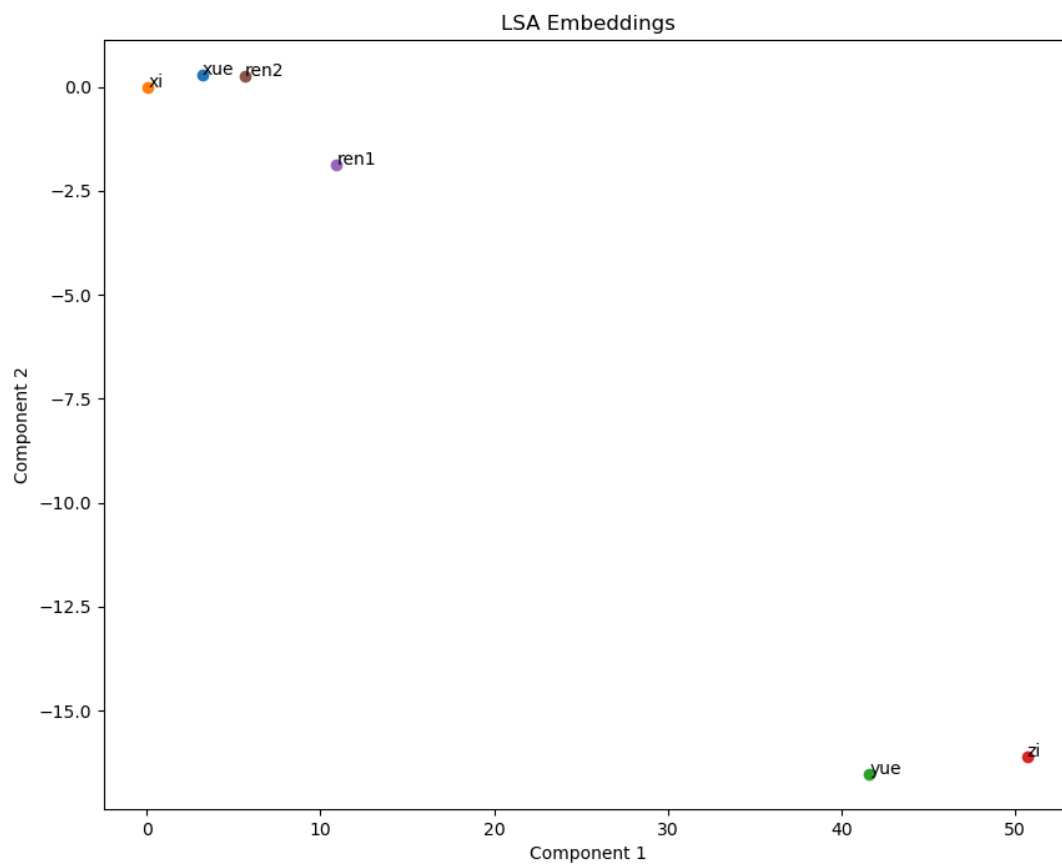
emb_sizes=50, k_values=5, window_sizes=1, epoch=10

```
1 |-----embeddings_50_5_1-----
2 39746
3 Epoch 1, Batch 1, Loss: 66.5421
4 Epoch 1, Batch 1001, Loss: 49.9358
5 Epoch 1, Batch 2001, Loss: 48.0158
6 Epoch 2, Batch 1, Loss: 38.3472
7 Epoch 2, Batch 1001, Loss: 40.8440
8 Epoch 2, Batch 2001, Loss: 39.1583
9 Epoch 3, Batch 1, Loss: 29.7991
10 Epoch 3, Batch 1001, Loss: 34.1294
11 Epoch 3, Batch 2001, Loss: 33.0272
12 Epoch 4, Batch 1, Loss: 16.7178
13 Epoch 4, Batch 1001, Loss: 30.2180
14 Epoch 4, Batch 2001, Loss: 29.4422
15 Epoch 5, Batch 1, Loss: 16.6874
16 Epoch 5, Batch 1001, Loss: 27.9282
17 Epoch 5, Batch 2001, Loss: 27.3069
18 Epoch 6, Batch 1, Loss: 16.6677
19 Epoch 6, Batch 1001, Loss: 26.5302
20 Epoch 6, Batch 2001, Loss: 25.9674
21 Epoch 7, Batch 1, Loss: 17.4516
22 Epoch 7, Batch 1001, Loss: 25.6120
23 Epoch 7, Batch 2001, Loss: 25.0424
24 Epoch 8, Batch 1, Loss: 18.1206
25 Epoch 8, Batch 1001, Loss: 25.0201
26 Epoch 8, Batch 2001, Loss: 24.3878
27 Epoch 9, Batch 1, Loss: 17.8904
28 Epoch 9, Batch 1001, Loss: 24.6181
29 Epoch 9, Batch 2001, Loss: 23.8985
30 Epoch 10, Batch 1, Loss: 17.3013
31 Epoch 10, Batch 1001, Loss: 24.3075
32 Epoch 10, Batch 2001, Loss: 23.4977
```

Requirement 5:

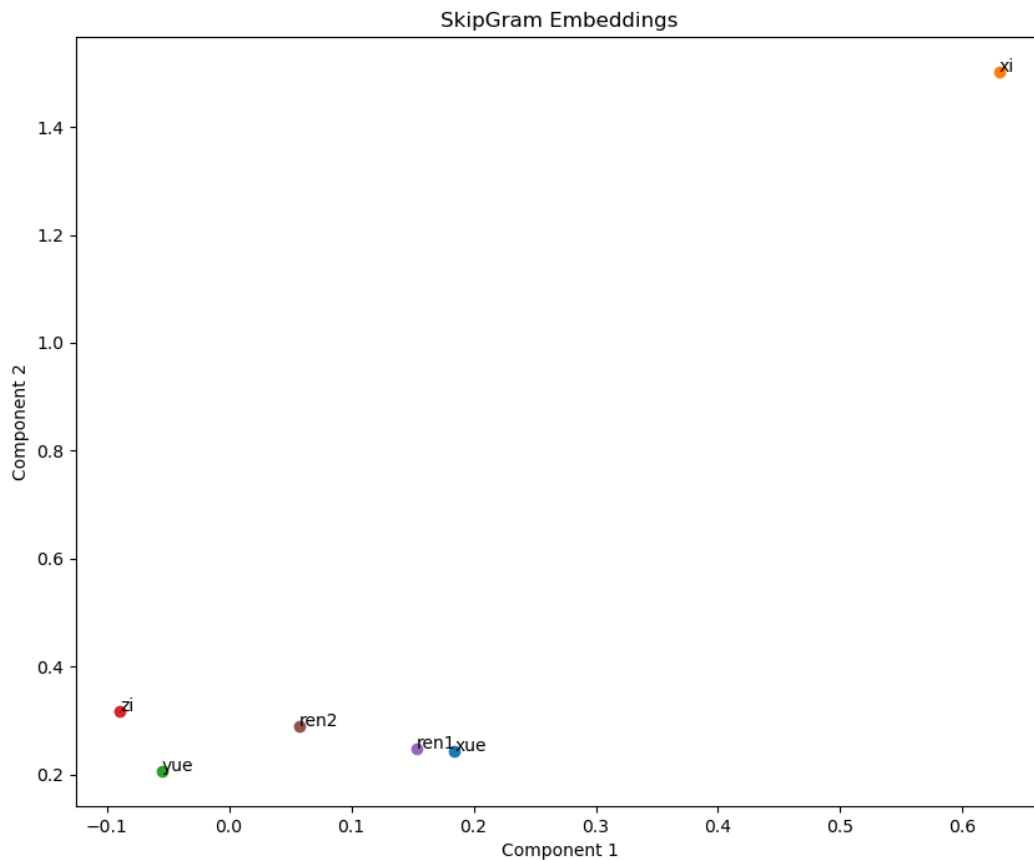
Compare your favorite embedding plot with the one we obtained from the LSA Lab.

LSA



Skip-Gram

emb_sizes=50, k_values=5, window_sizes=1



Briefly describe the difference in your write-up :

The obvious difference is that In the context of applying the Skip-Gram model, the character "习" has longer distance with other characters compare to LSA model. Even "学" is far away from "学" with Skip-Gram model.

The main reason is that Skip-Gram embeddings capture semantic similarities. Since "习" is used uniformly across diverse contexts, it lacks strong semantic associations with particular words. Additionally, the meaning of "习" might be broader than just "学". Therefore, its embedding could reflect this breadth, causing it to be distant from specific words, even those that are semantically related.

Compare to LSA, "学" and "习" are very close. This is because LSA focuses on the co-occurrence of words across the entire collection of documents, rather than within specific contexts or windows. Since "学" and "习" frequently co-occur in the same or similar documents, LSA would capture this global co-occurrence pattern and represent them as being close in the embedding space.