

```
from bertopic import BERTopic
from sentence_transformers import SentenceTransformer
from sklearn.datasets import fetch_20newsgroups
```

```
data = fetch_20newsgroups(subset='train',
                          remove=("headers", "footers", "quotes"))
documents = data.data[:2000] # limit to avoid RAM issues

print("Loaded documents:", len(documents))
```

Loaded documents: 2000

```
# Embedding model
embedding_model = SentenceTransformer("all-MiniLM-L6-v2")
```

```
modules.json: 100% 349/349 [00:00<00:00, 32.2kB/s]
config_sentence_transformers.json: 100% 116/116 [00:00<00:00, 4.57kB/s]
README.md: 10.5k/? [00:00<00:00, 660kB/s]
sentence_bert_config.json: 100% 53.0/53.0 [00:00<00:00, 3.42kB/s]
config.json: 100% 612/612 [00:00<00:00, 61.4kB/s]
model.safetensors: 100% 90.9M/90.9M [00:01<00:00, 113MB/s]
tokenizer_config.json: 100% 350/350 [00:00<00:00, 9.49kB/s]
vocab.txt: 232k/? [00:00<00:00, 2.53MB/s]
tokenizer.json: 466k/? [00:00<00:00, 8.00MB/s]
special_tokens_map.json: 100% 112/112 [00:00<00:00, 2.87kB/s]
config.json: 100% 190/190 [00:00<00:00, 2.54kB/s]
```

```
# BERTopic with simple config (lite mode)
topic_model = BERTopic(
    embedding_model=embedding_model,
    low_memory=True,
    verbose=True
)
```

```
# Fit-transform
topics, probs = topic_model.fit_transform(documents)
```

```
2025-11-23 08:02:39,826 - BERTopic - Embedding - Transforming documents to embeddings.
Batches: 100% 63/63 [02:57<00:00, 1.73it/s]
2025-11-23 08:05:37,735 - BERTopic - Embedding - Completed ✓
2025-11-23 08:05:37,737 - BERTopic - Dimensionality - Fitting the dimensionality reduction algorithm
2025-11-23 08:05:55,876 - BERTopic - Dimensionality - Completed ✓
2025-11-23 08:05:55,878 - BERTopic - Cluster - Start clustering the reduced embeddings
2025-11-23 08:05:55,955 - BERTopic - Cluster - Completed ✓
2025-11-23 08:05:55,971 - BERTopic - Representation - Fine-tuning topics using representation models.
2025-11-23 08:05:56,460 - BERTopic - Representation - Completed ✓
```

```
# Print top topic info
topic_info = topic_model.get_topic_info()
print(topic_info.head())
```

```
   Topic  Count      Name \
0      -1    613  -1_ax_the_max_to
1       0    193   0_he_the_in_to
2       1    154  1_is_of_the_that
3       2    102  2_00_dos_for_good
4       3     94   3_the_to_of_it

      Representation \
0      [ax, the, max, to, and, of, is, for, it, in]
1  [he, the, in, to, and, team, game, was, of, that]
2  [is, of, the, that, to, not, and, it, jesus, in]
3  [00, dos, for, good, 50, excellent, the, offer...
4  [the, to, of, it, be, that, is, clipper, in, and]

      Representative_Docs
```

```

0 [THE WHITE HOUSE\n\n      Office o...
1 [1992-93 Los Angeles Kings notes.\n-----...
2 [My last article included this quote:\n\n "If ...
3 [Discounts! Please take\t$2.00 off each item ...
4 [After reading the debate over the Clipper, I ...

```

```

# Print example topic words
print("\nTopic 0 Keywords:")
print(topic_model.get_topic(0))

```

```

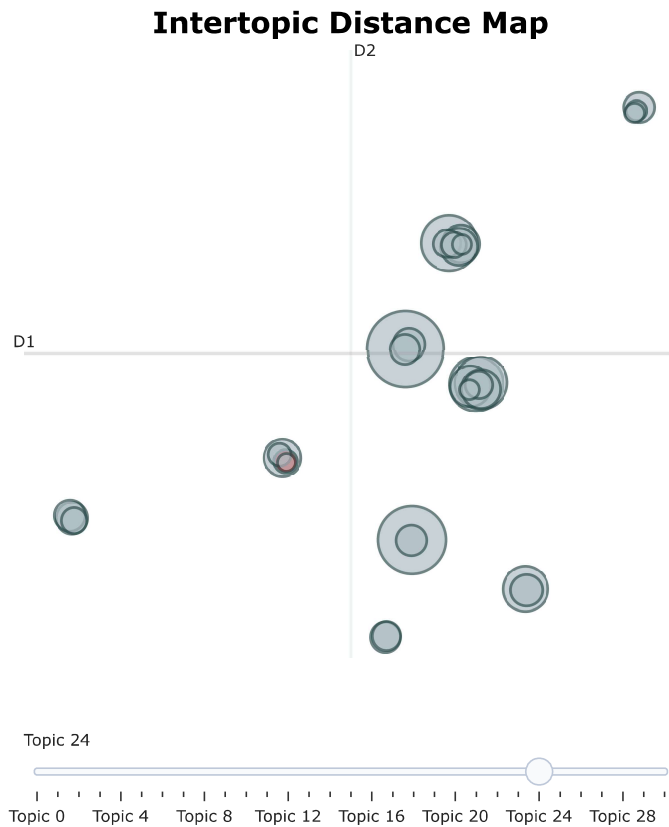
Topic 0 Keywords:
[('he', np.float64(0.026522111785216008)), ('the', np.float64(0.02506393339114429)), ('in', np.float64(0.020598082585780234)), (

```

```

# Basic visualization (works without heavy install)
fig = topic_model.visualize_topics()
fig.show()

```



Start coding or [generate](#) with AI.

