# Finetune Embeddings - LllamaIndex Example

```
# !pip install llama_index pypdf openai accelerate llama-cpp-python -q
!pip install llama-index langchain pypdf -q

import json

from llama_index import SimpleDirectoryReader
from llama_index.node_parser import SentenceSplitter
from llama_index.schema import MetadataMode
```

#### Download Data

```
!mkdir -p 'data/10k/'
!wget 'https://raw.githubusercontent.com/run-llama/llama_index/main/docs/examples/data/10k/uber_2021.pdf' -0 'da
!wget 'https://raw.githubusercontent.com/run-llama/llama_index/main/docs/examples/data/10k/lyft_2021.pdf' -0 'da
     --2024-01-09 10:47:55-- <a href="https://raw.githubusercontent.com/run-llama_lindex/main/docs/examples/data/10k">https://raw.githubusercontent.com/run-llama_lindex/main/docs/examples/data/10k</a>
     Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 185.199.108.133, 185.199.109.133, 185.199
     Connecting to raw.githubusercontent.com (raw.githubusercontent.com) | 185.199.108.133 | :443... connected.
     HTTP request sent, awaiting response... 200 OK
     Length: 1880483 (1.8M) [application/octet-stream]
     Saving to: 'data/10k/uber_2021.pdf'
     data/10k/uber 2021. 100%[==========] 1.79M --.-KB/s
     2024-01-09 10:47:55 (31.0 MB/s) - 'data/10k/uber 2021.pdf' saved [1880483/1880483]
     --2024-01-09 10:47:55-- https://raw.githubusercontent.com/run-llama/llama_index/main/docs/examples/data/10k
     Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 185.199.108.133, 185.199.109.133, 185.199
     Connecting to raw.githubusercontent.com (raw.githubusercontent.com) | 185.199.108.133 | :443... connected.
     HTTP request sent, awaiting response... 200 OK
     Length: 1440303 (1.4M) [application/octet-stream]
     Saving to: 'data/10k/lyft_2021.pdf'
     data/10k/lyft_2021. 100%[========>] 1.37M --.-KB/s
                                                                            in 0.06s
     2024-01-09 10:47:55 (24.1 MB/s) - 'data/10k/lyft_2021.pdf' saved [1440303/1440303]
TRAIN_FILES = ["./data/10k/lyft_2021.pdf"]
VAL_FILES = ["./data/10k/uber_2021.pdf"]
TRAIN_CORPUS_FPATH = "./data/train_corpus.json"
VAL_CORPUS_FPATH = "./data/val_corpus.json"
```

```
def load_corpus(files, verbose=False):
  if verbose:
    print(f"Loading files {files}")
  reader = SimpleDirectoryReader(input_files=files)
  docs = reader.load_data()
  if verbose:
    print(f"Loaded {len(docs)} docs")
  parser = SentenceSplitter()
  nodes = parser.get_nodes_from_documents(docs, show_progress=verbose)
  if verbose:
    print(f"Parsed {len(nodes)} nodes")
  return nodes
train_nodes = load_corpus(TRAIN_FILES, verbose=True)
val_nodes = load_corpus(VAL_FILES, verbose=True)
     Loading files ['./data/10k/lyft_2021.pdf']
     Loaded 238 docs
     Parsing nodes: 100%
                                                                238/238 [00:01<00:00, 249.04it/s]
     Parsed 344 nodes
     Loading files ['./data/10k/uber_2021.pdf']
     Loaded 307 docs
                                                                307/307 [00:02<00:00, 197.29it/s]
     Parsing nodes: 100%
```

# Generate synthetic queries

Parsed 410 nodes

```
from llama_index.finetuning import (
    generate_qa_embedding_pairs,
    EmbeddingQAFinetuneDataset,
)
from llama_index.llms import OpenAILike
import os
from getpass import getpass
os.environ["TOGETHER API KEY"] = getpass("TOGETHER API KEY")
api_key = os.environ["TOGETHER_API_KEY"]
     TOGETHER API KEY·····
11m = OpenAILike(
    model = "mistralai/Mixtral-8x7B-Instruct-v0.1",
    api_base = "https://api.together.xyz/v1",
    api_key=api_key
train_dataset = generate_qa_embedding_pairs(train_nodes, llm=llm)
val_dataset = generate_qa_embedding_pairs(val_nodes, llm=llm)
train_dataset.save_json("train_dataset.json")
val_dataset.save_json("val_dataset.json")
```

```
100%
                344/344 [10:07<00:00, 1.77s/it]
                410/410 [09:12<00:00, 1.35s/it]
```

# Run Embedding Finetuning

```
from llama_index.finetuning import SentenceTransformersFinetuneEngine
!pip install -U sentence-transformers -q
finetune_engine = SentenceTransformersFinetuneEngine(
    train dataset,
    model_id="BAAI/bge-small-en",
    model_output_path="test_model",
    val_dataset=val_dataset
)
     /usr/local/lib/python3.10/dist-packages/huggingface hub/utils/ token.py:72: UserWarning:
     The secret `HF_TOKEN` does not exist in your Colab secrets.
     To authenticate with the Hugging Face Hub, create a token in your settings tab (https://huggingface.co/setti
     You will be able to reuse this secret in all of your notebooks.
     Please note that authentication is recommended but still optional to access public models or datasets.
       warnings.warn(
      .gitattributes: 100%
                                                                   1.52k/1.52k [00:00<00:00, 51.9kB/s]
      1_Pooling/config.json: 100%
                                                                           190/190 [00:00<00:00, 10.0kB/s]
      README.md: 100%
                                                                    90.8k/90.8k [00:00<00:00, 4.32MB/s]
      config.json: 100%
                                                                  684/684 [00:00<00:00, 27.6kB/s]
      config sentence transformers.json: 100%
                                                                                      124/124 [00:00<00:00, 6.13kB/s]
      model.safetensors: 100%
                                                                        133M/133M [00:01<00:00, 143MB/s]
      pytorch_model.bin: 100%
                                                                        134M/134M [00:01<00:00, 135MB/s]
      sentence_bert_config.json: 100%
                                                                               52.0/52.0 [00:00<00:00, 2.16kB/s]
      special_tokens_map.json: 100%
                                                                              125/125 [00:00<00:00, 6.36kB/s]
      tokenizer.json: 100%
                                                                    711k/711k [00:00<00:00, 12.2MB/s]
                                                                          366/366 [00:00<00:00, 12.7kB/s]
      tokenizer_config.json: 100%
      vocab.txt: 100%
                                                                232k/232k [00:00<00:00, 5.83MB/s]
      modules.json: 100%
                                                                    349/349 [00:00<00:00, 14.7kB/s]
finetune_engine.finetune()
```

Epoch: 100% 2/2 [1:53:16<00:00, 3394.27s/it] Iteration: 100% 79/79 [49:39<00:00, 35.59s/it] Iteration: 100% 79/79 [48:55<00:00, 35.65s/it]

```
embed_model = finetune_engine.get_finetuned_model()
```

#### embed model

HuggingFaceEmbedding(model\_name='test\_model', embed\_batch\_size=10, callback\_manager= <llama\_index.callbacks.base.CallbackManager object at 0x7a00d5db39a0>, tokenizer\_name='test\_model', max\_length=512, pooling=<Pooling.CLS: 'cls'>, normalize=True, query\_instruction=None, text\_instruction=None, cache\_folder=None)

## Evaluate Finetuned Model

```
from llama_index.embeddings import OpenAIEmbedding
from llama_index import ServiceContext, VectorStoreIndex
from llama_index.schema import TextNode
from tqdm.notebook import tqdm
import pandas as pd
def evaluate(dataset, embed_model, top_k=5, verbose=False):
 corpus = dataset.corpus
  queries = dataset.queries
 relevant_docs = dataset.relevant_docs
  service_context= ServiceContext.from_defaults(embed_model=embed_model)
  nodes = [TextNode(id_=id_, text=text) for id_, text in corpus.items()]
  index = VectorStoreIndex(
      nodes,
      service_context=service_context,
      show_progress=True
  retriever = index.as_retriever(similarity_top_k=top_k)
  eval_results = []
  for query_id, query in tqdm(queries.items()):
   retrieved_nodes = retriever.retrieve(query)
    retrieved_ids = [node.node.node_id for node in retrieved_nodes]
    expected_id = relevant_docs[query_id][0]
   is_hit = expected_id in retrieved_ids
   eval_result = {
        "is_hit": is_hit,
        "retrieved": retrieved_ids,
        "expected": expected_id,
        "query": query_id
    eval_results.append(eval_result)
  return eval_results
from sentence_transformers.evaluation import InformationRetrievalEvaluator
from sentence_transformers import SentenceTransformer
from pathlib import Path
def evaluate_st(dataset, model_id, name):
  corpus = dataset.corpus
  queries = dataset.queries
  relevant_docs = dataset.relevant_docs
 evaluator = InformationRetrievalEvaluator(
      queries, corpus, relevant_docs, name=name
  )
 model = SentenceTransformer(model_id)
 output_path = "results/"
 Path(output_path).mkdir(exist_ok=True, parents=True)
  return evaluator(model, output_path=output_path)
```

#### Run Evals

```
1/9/24, 4:00 PM
                                                   Finetune-Embedding-Llama.ipynb - Colaboratory
   # import openai
   # os.environ["OPENAI_API_KEY"] = getpass("OPENAI_API_KEY")
   # openai.api_key = os.environ["OPENAI_API_KEY"]
   # ada = OpenAIEmbedding()
   # ada_val_results = evaluate(val_dataset, ada)
   # df_ada = pd.DataFrame(ada_val_results)
   # hit_rate_ada = df_ada["is_hit"].mean()
   # hit_rate_ada
   bge = "local:BAAI/bge-small-en"
   bge_val_results = evaluate(val_dataset, bge)
   df_bge = pd.DataFrame(bge_val_results)
   hit_rate_bge = df_bge["is_hit"].mean()
   hit_rate_bge
        /usr/local/lib/python3.10/dist-packages/huggingface_hub/utils/_token.py:72: UserWarning:
        The secret `HF_TOKEN` does not exist in your Colab secrets.
        To authenticate with the Hugging Face Hub, create a token in your settings tab (https://huggingface.co/setti
        You will be able to reuse this secret in all of your notebooks.
        Please note that authentication is recommended but still optional to access public models or datasets.
          warnings.warn(
        Generating embeddings: 100%
                                                                           410/410 [07:39<00:00, 1.08s/it]
         100%
                                                       1006/1006 [01:29<00:00, 10.87it/s]
        0.6163021868787276
   evaluate_st(val_dataset, "BAAI/bge-small-en", name="bge")
        0.46355582695619196
   finetuned = "local:test_model"
   val_results_finetuned = evaluate(val_dataset, finetuned)
                                                                           410/410 [07:20<00:00, 1.06s/it]
         Generating embeddings: 100%
         100%
                                                       1006/1006 [01:20<00:00, 11.81it/s]
   df finetuned = pd.DataFrame(val_results_finetuned)
   hit_rate_finetuned = df_finetuned["is_hit"].mean()
   hit_rate_finetuned
        0.668986083499006
   evaluate_st(val_dataset, "test_model", name="finetuned")
        0.5478756097147947
       Summary of Results
```

```
df_bge["model"] = "bge"
df_finetuned["model"] = "fine_tuned"
df_all = pd.concat([df_bge, df_finetuned])
df_all.groupby("model").mean("is_hit")
```



## ✓ Information Retrieval Evaluator

```
df_st_bge = pd.read_csv(
    "results/Information-Retrieval_evaluation_bge_results.csv"
)
df_st_finetuned = pd.read_csv(
    "results/Information-Retrieval_evaluation_finetuned_results.csv"
)
```

```
df_st_bge["model"] = "bge"

df_st_finetuned["model"] = "fine_tuned"

df_st_all = pd.concat([df_st_bge, df_st_finetuned])

df_st_all = df_st_all.set_index("model")

df_st_all
```

	epoch	steps	cos_sim- Accuracy@1	cos_sim- Accuracy@3	cos_sim- Accuracy@5	cos_sim- Accuracy@10	cos_sim- Precision@1	_	cos_sim- Precision@3
model									
bge	-1	-1	0.359841	0.525845	0.581511	0.650099	0.359841	0.359841	0.175282
fine_tuned	-1	-1	0.452286	0.605368	0.668986	0.720676	0.452286	0.452286	0.201789

2 rows × 32 columns