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July 1, 2025

[62]: !pip install scikit-learn transformers

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Requirement already satisfied: scikit-learn in /usr/local/lib/python3.11/dist-
packages (1.6.1)
Requirement already satisfied: transformers in /usr/local/lib/python3.11/dist-
packages (4.52.4)
Requirement already satisfied: numpy>=1.19.5 in /usr/local/lib/python3.11/dist-
packages (from scikit-learn) (2.0.2)
Requirement already satisfied: scipy>=1.6.0 in /usr/local/lib/python3.11/dist-
packages (from scikit-learn) (1.15.3)
Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-
packages (from scikit-learn) (1.5.1)
Requirement already satisfied: threadpoolctl>=3.1.0 in
/usr/local/lib/python3.11/dist-packages (from scikit-learn) (3.6.0)
Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-
packages (from transformers) (3.18.0)
Requirement already satisfied: huggingface-hub<1.0,>=0.30.0 in
/usr/local/lib/python3.11/dist-packages (from transformers) (0.33.0)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.11/dist-packages (from transformers) (24.2)
Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.11/dist-
packages (from transformers) (6.0.2)
Requirement already satisfied: regex!=2019.12.17 in
/usr/local/lib/python3.11/dist-packages (from transformers) (2024.11.6)
Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-
packages (from transformers) (2.32.3)
Requirement already satisfied: tokenizers<0.22,>=0.21 in
/usr/local/lib/python3.11/dist-packages (from transformers) (0.21.2)
Requirement already satisfied: safetensors>=0.4.3 in
/usr/local/lib/python3.11/dist-packages (from transformers) (0.5.3)
Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.11/dist-
packages (from transformers) (4.67.1)
Requirement already satisfied: fsspec>=2023.5.0 in
/usr/local/lib/python3.11/dist-packages (from huggingface-
hub<1.0,>=0.30.0->transformers) (2025.3.2)
Requirement already satisfied: typing-extensions>=3.7.4.3 in
/usr/local/lib/python3.11/dist-packages (from huggingface-
hub<1.0,>=0.30.0->transformers) (4.14.0)
```

```
Requirement already satisfied: hf-xet<2.0.0,>=1.1.2 in
     /usr/local/lib/python3.11/dist-packages (from huggingface-
     hub<1.0,>=0.30.0->transformers) (1.1.5)
     Requirement already satisfied: charset-normalizer<4,>=2 in
     /usr/local/lib/python3.11/dist-packages (from requests->transformers) (3.4.2)
     Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-
     packages (from requests->transformers) (3.10)
     Requirement already satisfied: urllib3<3,>=1.21.1 in
     /usr/local/lib/python3.11/dist-packages (from requests->transformers) (2.4.0)
     Requirement already satisfied: certifi>=2017.4.17 in
     /usr/local/lib/python3.11/dist-packages (from requests->transformers)
     (2025.6.15)
[63]: from typing import List
      from dataclasses import dataclass, field
      from sklearn.feature_extraction.text import TfidfVectorizer
      from sklearn.metrics.pairwise import cosine_similarity
      from transformers import pipeline
[64]: | @dataclass(eq=True, frozen=True)
      class TreeNode:
          name: str
          children: List['TreeNode'] = field(default_factory=list, compare=False,__
       ⇔hash=False)
          documents: List[str] = field(default factory=list, compare=False,,,
       ⇔hash=False)
          def add_child(self, child: 'TreeNode') -> None:
              self.children.append(child)
          def add_document(self, doc: str) -> None:
              self.documents.append(doc)
[65]: class DocumentEncoder:
          def __init__(self):
              self.vectorizer = TfidfVectorizer()
          def fit_transform(self, docs: List[str]):
              return self.vectorizer.fit_transform(docs).toarray()
          def transform(self, doc: str):
              return self.vectorizer.transform([doc]).toarray()
[66]: class HierarchyTree:
          def __init__(self, root_name: str):
              self.root = TreeNode(name=root_name)
```

```
def add_topic(self, path: List[str]) -> TreeNode:
      node = self.root
      for name in path:
          match = next((child for child in node.children if child.name ==__
⇔name), None)
          if not match:
              match = TreeNode(name=name)
              node.add child(match)
          node = match
      return node
  def classify_and_assign(self, doc: str, topic_paths: List[List[str]]) ->__
→None:
      for path in topic_paths:
          node = self.add_topic(path)
          node.add_document(doc)
  def get_all_documents_under(self, node: TreeNode) -> List[str]:
      docs = list(node.documents)
      for child in node.children:
          docs.extend(self.get_all_documents_under(child))
      return docs
```

```
[67]: class SearchModule:
          def __init__(self, tree: HierarchyTree):
              self.tree = tree
          def search(self, query: str, top_k: int = 3) -> List[TreeNode]:
              candidates = []
              node\_doc\_map = \{\}
              def collect_nodes(node: TreeNode):
                  if node.documents:
                      text = ' '.join(node.documents)
                      node_doc_map[id(node)] = (node, text)
                      candidates.append(node)
                  for child in node.children:
                      collect_nodes(child)
              collect_nodes(self.tree.root)
              corpus = [val[1] for val in node_doc_map.values()]
              if not corpus:
                  return []
              encoder = DocumentEncoder()
              corpus_vectors = encoder.fit_transform(corpus)
```

```
query_vector = encoder.transform(query)

sims = cosine_similarity(query_vector, corpus_vectors).flatten()
top_indices = sims.argsort()[::-1][:top_k]

return [node_doc_map[list(node_doc_map.keys())[i]][0] for i in_u

otop_indices]
```

```
[69]: # Sample documents
      documents = [
          "Neural networks and deep learning techniques",
          "Transformers in NLP applications",
          "Classical machine learning models like SVM",
          "Database indexing and SQL queries",
          "NoSQL systems and distributed databases"
      ]
      # Build the tree and classify documents
      tree = HierarchyTree("AI")
      tree.classify_and_assign(documents[0], [["ML", "Deep Learning"]])
      tree.classify_and_assign(documents[1], [["ML", "NLP"]])
      tree.classify_and_assign(documents[2], [["ML", "Classical ML"]])
      tree.classify_and_assign(documents[3], [["Databases", "SQL"]])
      tree.classify_and_assign(documents[4], [["Databases", "NoSQL"]])
      # Create the model
      rag = RAGModel(tree)
```

Device set to use cpu

```
[70]: #Ask a test question
question = "What are deep learning techniques?"
print("\nBot:", rag.query(question))
```

Truncation was not explicitly activated but `max_length` is provided a specific value, please use `truncation=True` to explicitly truncate examples to max length. Defaulting to 'longest_first' truncation strategy. If you encode pairs of sequences (GLUE-style) with the tokenizer you can select this strategy more precisely by providing a specific strategy to `truncation`.

Setting `pad_token_id` to `eos_token_id`:50256 for open-end generation.

Both `max_new_tokens` (=256) and `max_length`(=150) seem to have been set.

`max_new_tokens` will take precedence. Please refer to the documentation for more information.

(https://huggingface.co/docs/transformers/main/en/main_classes/text_generation)

Bot: Deep learning techniques are the tools that allow you to learn and learn from a large number of topics.

Deep learning is the technology that allows you to use neural networks or machine learning techniques to generate, process, and store information in real-time.

Deep learning methods are the tools that allow you to learn and learn from large number of topics.

Some deep learning techniques are:

SVDT (Simultaneous Neural Networks)

SVDT is a new approach to deep learning that allows you to make deep learning decisions using the inputs and outputs of a deep learning model. This model can be used to automatically classify, model, and predict a variety of data sources. This technique uses deep learning data to teach and learn about the social, cultural, and other dimensions of a person's behavior. It does not apply to non-social networks like Facebook, LinkedIn, or Google+.

SVDT is a new approach to deep learning that allows you to make deep learning decisions using the inputs and outputs of a deep learning model. This model can be used to automatically classify, model, and predict a variety of data sources. SVDT is a new approach to deep learning that allows you to make deep learning decisions using the inputs and outputs of a deep learning

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