SKKU Lab Recommendation Service: FindMyLab

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Backend Part – Tagging

Tag Collection



Abstract of Paper

The next generation of artificial intelligence (AI) is required to be capable of proper communication to enable eloquent interaction with human beings. Thus, AI models require powerful language understanding and generation capabilities.

GPT Prompt Engineering

Artificial intelligence

Al communication

Language understanding

task-oriented dialogue systems

dialogue research

language generation

Backend Part – tagging

Clustering

Synonym

Database manage

Database management

Database management system

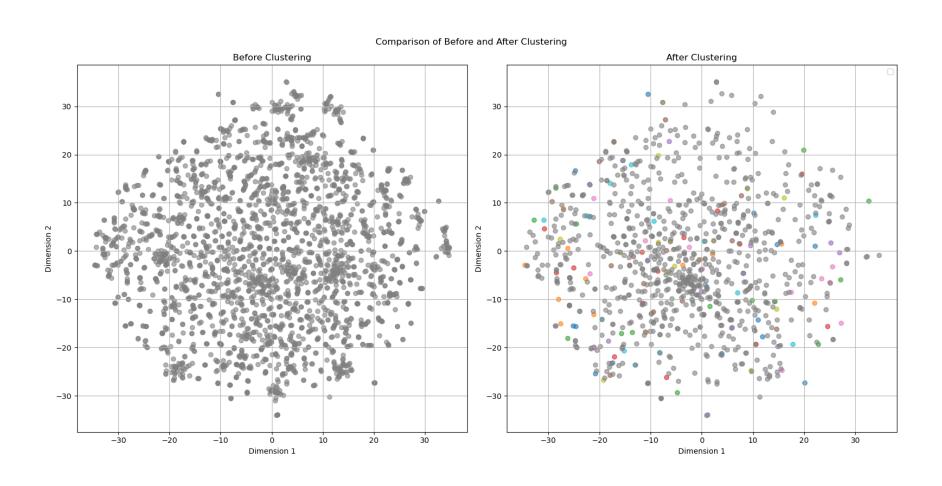
Similar Word

Medical technology

Healthcare technology

Backend Part – tagging

Clustering



Backend Part – tagging

Clustering Result

Fuzzy logic system

Fuzzy logic

Fuzzy logic

Language understanding

Language learning

Language learning

Medical technology

Healthcare technology

Medical technology

Reinforcement learning

Deep reinforcement learning

Model based reingorcement learning

Reinforcement learning

System Log

```
app listening on port 3000

DB is Connected

Get request

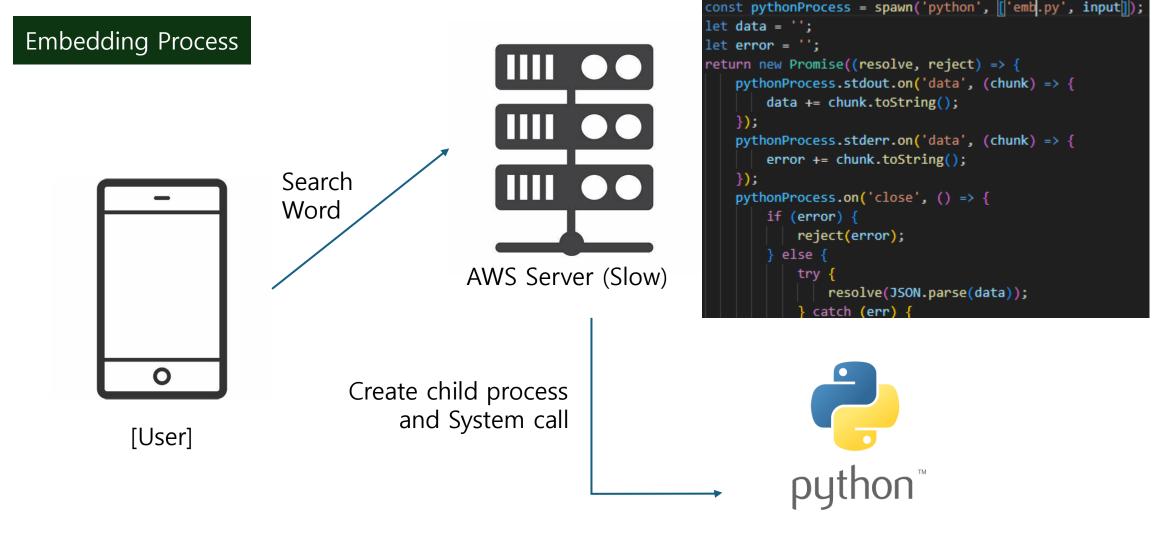
Embedding Time: 5.966s

Query Search Time: 61.59ms

Tag Search Time: 99.441ms

End
```

[Server system log]



[Embedding Program]

Embedding Process Search Word AWS Server (Slow) EC2 Database **Embedding** Word 0 [User] Desktop Server (Fast)

Result

Issue of time

> 5secs

```
app listening on port 3000

DB is Connected

Get request

Embedding Time: 52.795ms

Query Search Time: 9.91ms

Tag Search Time: 92.14ms

End
```

[Server system log]

data-preprocessing

CSV → json

NAME	Abstract	Keywords
고영중		
고영중		
고영중		
•••		



```
{
    "query" : key word (expected search query: using extracted keyword from abstract ),
    "pos": [source abstract]
}
```

data-preprocessing

512 tokenize

```
{ "query" : key word A, "pos": [source abstract A-1]}
{ "query" : key word A, "pos": [next source abstract A-2]}

{ "query" : key word B, "pos": [source abstract B-1]}
{ "query" : key word B, "pos": [next source abstract B-2] }
```

```
{"query": "Dialogue-Specific Pre-Training", "pos": ["The next generation of artificial intelligence {"query": "Dialogue-Specific Pre-Training", "pos": ["ures of dialogues. In addition, we introduce {"query": "Encoder-Decoder Transformer", "pos": ["The next generation of artificial intelligence ("query": "Encoder-Decoder Transformer", "pos": ["ures of dialogues. In addition, we introduce an {"query": "Auxiliary Task", "pos": ["The next generation of artificial intelligence (AI) is require {"query": "Auxiliary Task", "pos": ["ures of dialogues. In addition, we introduce an efficient metalogues."]
```

total: 18536

data-preprocessing

only pos json →hard negative mining

```
!python -m FlagEmbedding.baai_general_embedding.finetune.hn_mine \
--model_name_or_path BAAI/bge-m3 \
--input_file /data/projects/jaehee/jupyter/processed_dataset_pos_only.jsonl \
--output_file /data/projects/jaehee/jupyter/processed_dataset_hdn.jsonl \
--range_for_sampling 15-200 \
--negative_number 10 \
--use_gpu_for_searching
```

data-preprocessing

result of hard negative mining

```
{ "query" : key word A, "pos": [source abstract A-1], "neg":[mining result 1], ..., [mining result N] }
{ "query" : key word A, "pos": [next source abstract A-2], "neg":[mining result 1], ..., [mining result N] }

{ "query" : key word B, "pos": [source abstract B-1], "neg":[mining result 1], ..., [mining result N] }

{ "query" : key word B, "pos": [next source abstract B-2], "neg":[mining result 1], ..., [mining result N] }
```

Model Change

! issue of out of memory → changed to small model

SBERT-base-nli-v2 →

BAAI/bge-small-en-v1.5

Al Part – search algorithm

Model Change

BAAI/bge-small-en-v1.5

Model Name	Dimension	Sequence Length	Average (56)	Retrieval (15)	Clustering (11)	Pair Classification (3)
BAAI/bge- large-en- v1.5	1024	512	64.23	54.29	46.08	87.12
BAAI/bge- base-en- v1.5	768	512	63.55	53.25	45.77	86.55
BAAI/bge- small-en- v1.5	384	512	62.17	51.68	43.82	84.92

Model config

```
!torchrun --nproc_per_node 1 \
-m FlagEmbedding.baai_general_embedding.finetune.run \
--output_dir ./checkpoint_v1_bs32 \
--model_name_or_path BAAI/bge-small-en-v1.5 \
--train_data ./processed_dataset_hdn.jsonl \
--learning_rate 1e-5 \
--fp16 \
--num_train_epochs 14 \
--per_device_train_batch_size 32 \
--dataloader drop last True \
--normlized True \
--temperature 0.02 \
--query_max_len 64 \
--passage_max_len 512 \
--train_group_size 2 \
--negatives_cross_device \
--logging_steps 10 \
--save_steps 1000 \
--query_instruction_for_retrieval ""
```

batch size: 32

epoch: 14

Ex)

search keywords

```
['Compiler toolchain provenance',
  'BERT-based system',
  'Toolchain classification',
  'Binary code similarity detection',
  'Machine learning',
  'Fine-tuning process',
  'Binary analysis',
  'Signature-based tool',
  'Emerging compilation toolchains (Rust Go Nim)',
  'C and C++ compilers (GCC clang)']
```

Ex)



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ToolPhet: Inference of Compiler Provenance From Stripped Binaries With Emerging Compilation Toolchains

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The keywords are highly similar to those used in Professor Koo Hyungjun's actual research papers

RESULT

<before>

```
Top 5 유사한 교수님:
1. 구형준: 0.7517 ] 0.01
2. 이지형: 0.7372
3. 우홍욱: 0.7202
4. 이은석: 0.7200
5. 우사이먼성일: 0.7151
```

<after>

```
가장 키워드와 유사한 교수님: 구형준
유사도: 0.91418282868880352
Top 5 유사한 교수님:
1. 구형준: 0.9142
2. 우사이먼성일: 0.8734
3. 이은석: 0.8705
4. 김재광: 0.8604
5. 이지형: 0.8475
```

Compare

```
from FlagEmbedding import FlagModel
sentences_1 = ["Dialogue-Specific Pre-Training"]
sentences_2 = ["The next generation of artificial intelligence (AI) is required to be capable of proper communication to enable e
model = FlagModel('BAAI/bge-small-en', use_fp16=True)
fine_tuned_model = FlagModel('/data/projects/jaehee/jupyter/checkpoint_v1_bs32/checkpoint_3000', use_fp16=True)
# 기존 모델로 각각 임베딩
embeddings_1 = model.encode(sentences 1)
embeddings_2 = model.encode(sentences_2)
# 기존 모델로부터 나온 임베딩으로 유사도 계산
similarity_from_base_model = embeddings_1 @ embeddings_2.T
# 파인 튜닝 모델로 각각 임베딩
embeddings_1 = fine_tuned_model.encode(sentences_1)
embeddings_2 = fine_tuned_model.encode(sentences_2)
# 파인 튜닝 모델로부터 나온 임베딩으로 유사도 계산
similarity_from_fine_tuned_model = embeddings_1 @ embeddings_2.T
print('기존 모델:', similarity_from_base_model)
print('파인 튜닝된 모델:', similarity_from_fine_tuned_model)
```

기존 모델: [[0.923 0.813]]

파인 튜닝된 모델: [[0.825 0.578]]

Frontend Part

Image data processing : base 64

{"고영중": "/9j/4Q/

eRXhpZgAASUkqAAgAAAAQAAABAwABAAAArA4AAAEBAwABAAAAQRUAAAIBAwADAAAAZgAA AAYBAwABAAAAAgAAAABABAgAFAAAA1AAAABABAgAJAAAA2QAAABIBAwABAAAAAQAAABUBA wABAAAAAwAAABOBBQABAAAA4gAAABSBBQABAAAA6gAAACgBAwABAAAAAgAAADEBAgAFAA AA8gAAADIBAgAUAAAAEQEAADSBAgAJAAAAJQEAAJICAgArAAAALgEAAGmHBAABAAAAXAE AABAEAAAIAAgACABTT05ZAElMQ0UtN00zAAAJPQAQJwAAAAk9ABAAAABBZG9iZSBQaG90 b3Nob3AgMjMuMCAoV2luZG93cykAMjAyMjowNzowNyAwMDowMjowOQBtb25nZGFuaQBDb3B5cmlnaHQoQykgTkVTVCBzdC4gYWxsIHJpZ2h0cyByZXNlcnZlZC4AAAAAKACaggUAAQ AAAEIDAACdggUAAQAAAEoDAAAiiAMAAQAAAAEAAAAniAMAAQAAAFAAAAAwiAMAAQAAAAI

alt=""
onError={(e) => {
 e.target.src = defaultImg;
}}
/>

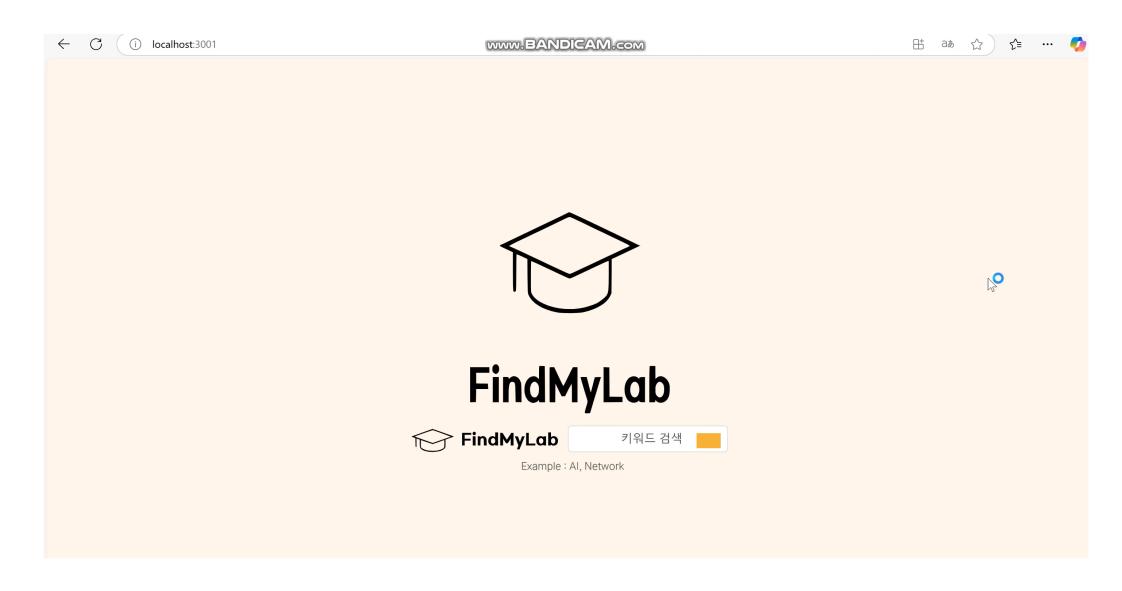
src={`data:image/jpeg;base64,\${img}`}

style={{ width: "auto", height: "150px", border: "0.1px solid #4F4E4E" }}

Base 64 encoded image data

Decode and output the encoded image data

Demo Video



Thank you.