```
document_cnt = []
for i in range(len(flattened_train_cycle_passage)):
    document_cnt.append(len(flattened_train_cycle_passage[i]))

print('cycle_passage가 제일 많은 개수', max(document_cnt))
print('cycle_passage가 제일 적은 개수', min(document_cnt))
print('total cycle_passage 개수', sum(document_cnt))
print('B코 cycle_passage 개수', sum(document_cnt)/len(document_cnt))
median_value = statistics.median(document_cnt)
print('cycle_passage 개수 중간 값', median_value)
```

- 1. 노드 64개로 만드는 과정 DPR 부터 진행하면 됨
- 2. 노드 32개는 오늘 밤에 colab에 돌리기

그러면 인제 진행해야 할 것은 세이프디...

한 번 정리해보자면 다음과 같다.

- 1. object detection을 통해 건설현장에서 위험요인의 위치 인식
- 2.
- 3. 인식된 이미지(bounding box)와 위험요인(text)를 simVLM모델에 넣는다.
- 4. 넣어서 나오는 output은 재해예방대책
- 5. classification 부분에서 재해예방대책과

```
import faiss
from transformers import DPRContextEncoder, DPRContextEncoderTokenizer, DPRQuestionEncoder, DPRQuestionEncoderTokenizer
# Initialize a GPU device
# res = faiss.StandardGpuResources()
ngpus = faiss.get_num_gpus()
print("number of GPUs:", ngpus)
\tt def\ calculate\_dpr\_similarity\_with\_faiss\_gpu(model\_name\_or\_path,\ qa\_context,\ cycle\_passage):
    # Load the DPR models and tokenizers
    context_model = DPRContextEncoder.from_pretrained(model_name_or_path)
    context_tokenizer = DPRContextEncoderTokenizer.from_pretrained(model_name_or_path)
    question model = DPROuestionEncoder.from pretrained(model name or path)
    \verb| question_tokenizer = DPRQuestionEncoderTokenizer.from_pretrained(model_name_or_path)|
    top_documents = []
    for query_list, doc_list_for_query in zip(qa_context, cycle_passage):
        if not doc_list_for_query:
            top_documents.append([])
            continue
        query = query_list if query_list else ""
        doc\_texts = [doc[0] for doc in doc\_list\_for\_query if doc]
        \ensuremath{\text{\#}} Tokenize and encode the query and documents using the DPR models
        query_inputs = context_tokenizer(query, return_tensors="pt", padding=True, truncation=True)
```

```
doc_inputs = context_tokenizer(doc_texts, return_tensors="pt", padding=True, truncation=True)
    with torch.no_grad():
        query_embeddings = question_model(**query_inputs).pooler_output
        passage_embeddings = context_model(**doc_inputs).pooler_output
   # Initialize a GPU device
    # res = faiss.StandardGpuResources()
    # Initialize your FAISS index on the GPU
   dimension = 768  # The dimension of DPR embeddings
cpu_index = faiss.index_factory(dimension, "Flat", faiss.METRIC_INNER_PRODUCT)
    # index = faiss.IndexFlatL2(dimension)
   gpu_index = faiss.index_cpu_to_all_gpus( # build the index
        cpu_index
   # Convert PyTorch tensors to NumPy arrays
    {\tt query\_embeddings\_np = query\_embeddings.cpu().contiguous().numpy()}
   passage_embeddings_np = passage_embeddings.cpu().contiguous().numpy()
    # Add data to the GPU-based index
    gpu_index.add(passage_embeddings_np)
    # Calculate top_n based on the length of doc_texts
   if len(doc\_texts) < 50:
        top_n = len(doc_texts)
    else:
        top_n = 50
   distances, top doc indices = qpu index.search(query embeddings np, top n)
   top_docs = top_doc_indices[0].tolist() # Convert to a list of top document indices
    # print(top_docs)
    # assert -1 == 0
    top_documents.append(top_docs)
    print('hi')
return top_documents
```

노드 64개를 대상으로하는 DPR은 너무 오래걸리는 상황

→ 일단 이것은 BM25 + reranking만 진행하자

노드 32개로 하는것은 일단 새로 진행해야하므로 DPR 진행 중

#### 중간에 바람쐬러 나가서 생각한것 정리

- ☑ GSC에서 노드 개수 32개가 only qanode라고 적혀있지만 only qanode가 아님
- ▼ DPR algorithm을 노드 개수 64개를 대상으로 진행하고 있는거 중단하기
- ☐ original gsc 돌리기
- ☐ edge masking 생각하기
  - 기존에는 graph-pruning을 생각했었음. → 이것은 cycle count 실험을 완료 후 진행
- ▼ <del>노드개수 64개에 대해서 BM25 적용하기</del>
  - ✓ reranking과정도 진행하기에 앞서 비율 상태 확인 + dev 데이터도 진행하기
- □ reranking 진행하기 → 진행중
- □ 32개에 대해서 DPR 알고리즘은 colab에서 진행 중, 진행 완료하고 전처리 끝까지(statement.jsonl) 진행하기

```
import json

# Load the original data from the existing input_file
original_data = []
with open('./data/csqa/statement/dev.statement.jsonl', 'r', encoding='utf-8') as f:
    for line in f.readlines():
        data = json.loads(line)
        original_data.append(data)
```

```
# Create a new list to hold the modified data
new_data = []
# Now, for each example, modify the 'statements' to match 'rerank_text'
# rerank_only_text : 1221
for data, rerank_labels in zip(original_data, dev_rerank_only_text): # rerank_labels = 1번문제
    for statement, rerank_text_list in zip(data["statements"], rerank_labels): # rerank_text_list = 1번 문제 1번 정답
       if not rerank_text_list:
           statement["statement"] = ""
           continue
        statement["statement"] = "<SEP>".join(rerank_text_list)
        # print(statement['statement'])
    # Append the modified data to the new list
    new_data.append(data)
# Write the modified data to a new JSONL file
output_file_path = './data/csqa/statement/dev_modified.statement.jsonl'
with open(output_file_path, 'w', encoding='utf-8') as output_file:
    for data in new_data:
        json.dump(data, output_file, ensure_ascii=False)
        output_file.write('\n')
print(f"Modified data saved to {output_file_path}")
```

- 1. 목요일까지 cycle count에 대해서 실험을 진행한다. 실험은 0,1,2, 3개의 ssed에 대해서 실험을 진행한다.
- 2. 실험 순서는 다음과 같다. 각 방법은 노드 개수 32, 64개를 대상으로 진행한다.
  - a. original gsc
  - b. original gsc + original cycle count
  - c. original gsc + new cycle count(not one-hot vector)
  - d. original gsc + new cycle count(one-hot vector)
  - e. new context score

#### New context score 제작 코드

```
def load_bert_xlnet_roberta_input_tensors(statement_jsonl_path, model_type, model_name, max_seq_length, pretrained=None):
    cache_path = statement_jsonl_path + '.' + model_name +'.sentvecs.loaded_cache'
    class InputExample(object):
        def __init__(self, example_id, question, contexts, endings, label=None, statements=None):
            self.example_id = example_id
            self.question = question
            self.contexts = contexts
            self.endings = endings
            self.label = label
            self.statements = statements # Add statements attribute
    class InputFeatures(object):
        def __init__(self, example_id, choices_features, label):
    self.example_id = example_id
            self.choices_features = [
                    'input_ids': input_ids,
                    'input_mask': input_mask,
                    'segment_ids': segment_ids,
                    'output_mask': output_mask,
                for _, input_ids, input_mask, segment_ids, output_mask in choices_features
            self.label = label
    def read_examples(input_file,input_file1='./data/csqa/statement/train_modified.statement.jsonl'):
        with open(input_file, "r", encoding="utf-8") as f:
            examples = []
            for line in f.readlines():
                json_dic = json.loads(line)
                label = ord(json_dic["answerKey"]) - ord("A") if 'answerKey' in json_dic else 0
```

```
contexts = json_dic["question"]["stem"]
            if "para" in json_dic:
                contexts = json_dic["para"] + " " + contexts
            if "fact1" in json_dic:
                \verb|contexts = json_dic["fact1"] + " " + contexts|\\
            examples.append(
                InputExample(
                    example id=json_dic["id"],
                    contexts=[contexts] * len(json_dic["question"]["choices"]), # 이건 말그대로 문제(question) # contexts=[statement["statement"] for statement in json_dic["statements"]],
                    # contexts=[ending["text"] + "." for ending in json_dic["question"]["choices"]],
                    question="",
                     # question=json_dic["question"]["stem"],
                    endings=[ending["text"] for ending in json_dic["question"]["choices"]],
                     # endings=["" for ending in json_dic["question"]["choices"]],
                ))
    # print('input_file',examples[1].contexts)
    # print('input_file',examples[1].endings)
    # Now, add examples from input_file1
    with \ open('./data/csqa/statement/train\_modified.statement.jsonl', \ "r", \ encoding="utf-8") \ as \ f:
        examples1 = []
        for line in f.readlines():
            json_dic = json.loads(line)
            label = ord(json_dic["answerKey"]) - ord("A") if 'answerKey' in json_dic else 0
            contexts = [statement["statement"] \ for \ statement \ in \ json\_dic['statements']]
            # print('contexts',contexts)
            # assert -1 == 0
            # if "para" in json_dic:
                 contexts = json_dic["para"] + " " + contexts
            # if "fact1" in json_dic:
                  contexts = json_dic["fact1"] + " " + contexts
            examples1.append(
                InputExample(
                    example_id=json_dic["id"],
                     contexts=contexts, # 이건 말그대로 문제(question)
                     question=""
                     # endings=[ending["text"] for ending in json_dic["question"]["choices"]],
                    endinas=""
                    # endings=["" for ending in json_dic["question"]["choices"]],
                    label=label
                ))
    # print('input_file1',examples1[1].contexts)
    # print(examples1[1].endings)
    # assert -1 == 0
    return examples, examples1
# class InputExample(object):
      def __init__(self, example_id, question, contexts, endings, label=None):
          self.example_id = example_id
          self.question = question
          self.contexts = contexts
          self.endings = endings
          self.label = label
# class InputFeatures(object):
      def __init__(self, example_id, choices_features, label):
    self.example_id = example_id
          self.choices_features = [
             {
                  'input_ids': input_ids,
'input_mask': input_mask,
                  'segment_ids': segment_ids,
                  'output_mask': output_mask,
              for _, input_ids, input_mask, segment_ids, output_mask in choices_features
          self.label = label
# def read_examples(input_file):
      with open(input_file, "r", encoding="utf-8") as f:
          examples = []
          for line in f.readlines():
              json_dic = json.loads(line)
              label = ord(json_dic["answerKey"]) - ord("A") if 'answerKey' in json_dic else 0
```

```
contexts = json_dic["question"]["stem"]
              if "para" in json_dic:
                  contexts = json_dic["para"] + " " + contexts
              if "fact1" in json_dic:
                  contexts = json_dic["fact1"] + " " + contexts
              examples.append(
                  InputExample(
                      example id=json_dic["id"],
                      contexts=[contexts] * len(json_dic["question"]["choices"]), # 이건 말그대로 문제(question) # contexts=[statement["statement"] for statement in json_dic["statements"]],
                      # contexts=[ending["text"] + "." for ending in json_dic["question"]["choices"]],
                      question="",
                      # question=ison dic["question"]["stem"].
                      endings=[ending["text"] for ending in json_dic["question"]["choices"]],
                      # endings=["" for ending in json_dic["question"]["choices"]],
                  ))
              # print('examples',examples[0].endings)
                ''examples ['The sanctions against the school were a punishing blow, and they seemed to what the efforts the school
              'The sanctions against the school were a punishing blow, and they seemed to what the efforts the school had made to d
              'The sanctions against the school were a punishing blow, and they seemed to what the efforts the school had made to d
              'The sanctions against the school were a punishing blow, and they seemed to what the efforts the school had made to d
              'The sanctions against the school were a punishing blow, and they seemed to what the efforts the school had made to d
              # assert -1 == 0
      return examples
def convert examples to features(examples, examples1, label list, max seg length,
                                  tokenizer,
                                  cls_token_at_end=False,
                                  cls_token='[CLS]',
                                  cls_token_segment_id=1,
                                  sep_token='[SEP]',
                                  sequence_a_segment_id=0,
                                  sequence_b_segment_id=1,
                                  sequence_c_segment_id=2,
                                  sep_token_extra=False,
                                  pad_token_segment_id=0,
                                  pad_on_left=False,
                                  pad_token=0,
                                  {\tt mask\_padding\_with\_zero=True):}
    """ Loads a data file into a list of `InputBatch`s
        `cls_token_at_end` define the location of the CLS token:
            - False (Default, BERT/XLM pattern): [CLS] + A + [SEP] + B + [SEP]
            - True (XLNet/GPT pattern): A + [SEP] + B + [SEP] + [CLS]
        `cls_token_segment_id` define the segment id associated to the CLS token (0 for BERT, 2 for XLNet)
   label_map = {label: i for i, label in enumerate(label_list)}
    tokenc_len = []
    features = []
    for ex_index, (example, example1) in enumerate(zip(examples, examples1)):
       # print('ex_index',ex_index)
        # print(example)
        # print('example',example.contexts)
        # print('examples1',example1.contexts)
        # print('example1', example.endings)
        # print('example.question',example.question)
        # assert -1 == 0
        choices_features = []
        for ending_idx, (context, ending,cycle_passage) in enumerate(zip(example.contexts, example.endings,example1.contexts)):
            tokens a = tokenizer.tokenize(context)
            tokens_b = tokenizer.tokenize(example.question + " " + ending)
            # print('tokens_a', tokens_a)
            # print('tokens_b',tokens_b)
            # assert -1 == 0
            special tokens count = 4 if sep token extra else 3
            if not cycle_passage:
                continue
            else:
                special_tokens_count += cycle_passage.count('<SEP>')
                # cycle_passage = cycle_passage.replace()
                cycle_passage = cycle_passage.replace('<SEP>',' ')
                tokens_c = tokenizer.tokenize(example1.question + " " + cycle_passage)
                # print('tokens_c',tokens_c)
                # print(len(tokens_c)) # 412
            tokenc_len.append(len(tokens_c))
            # assert -1 == 0
            # special_tokens_count = 4 if sep_token_extra else 3
```

```
# special_tokens_count = 5 if sep_token_extra else 3
# print('special_tokens_count',special_tokens_count) # 4
# assert -1 == 0
_truncate_seq_pair(tokens_a, tokens_b,tokens_c, max_seq_length - special_tokens_count)
# print('after_tokens_a',tokens_a)
# print('after_tokens_b',tokens_b)
# print('after_tokens_c', tokens_c)
# print(len(tokens c))
# print('special_token_extra',sep_token_extra) # True
# assert -1 == 0
    # The convention in BERT is:
    # (a) For sequence pairs:
    \mbox{\#} tokens: [CLS] is this jack \mbox{\#\#son} \mbox{\#\#ville} ? [SEP] no it is not . [SEP]
    # type_ids: 0 0 0 0 0 0 0 0 1 1 1 1 1 1
    # (b) For single sequences:
    # tokens: [CLS] the dog is hairy . [SEP]
    # type_ids:
                  0 0 0
                               0 0
    # Where "type_ids" are used to indicate whether this is the first
    \# sequence or the second sequence. The embedding vectors for `type=0` and
    \ensuremath{\mbox{\# `type=1`}} were learned during pre-training and are added to the wordpiece
    # embedding vector (and position vector). This is not *strictly* necessary
    # since the [SEP] token unambiguously separates the sequences, but it makes
    # it easier for the model to learn the concept of sequences.
   \# For classification tasks, the first vector (corresponding to [CLS]) is \# used as as the "sentence vector". Note that this only makes sense because
    # the entire model is fine-tuned.
tokens = tokens_a + [sep_token]
if sep_token_extra:
    # roberta uses an extra separator b/w pairs of sentences
    tokens += [sep_token]
    # print('tokens',tokens)
   # assert -1 == 0
segment_ids = [sequence_a_segment_id] * len(tokens)
# print('segment_ids',segment_ids)
# print(len(segment_ids))
# assert -1 == 0
if tokens_b and tokens_c: # roberta 경우에는 tokens_b까지 type_ids를 0으로봄 그래야 qa_context가 되기 때문임
   tokens += tokens_b + [sep_token]
    segment_ids += [sequence_b_segment_id] * (len(tokens_b) + 1)
   # print('segment_ids',segment_ids)
    # print(len(seament ids))
    # print('tokensb',tokens)
    # print('cls_token_at_end',cls_token_at_end) # false
   tokens += tokens_c + [sep_token]
    segment_ids += [sequence_c_segment_id] * (len(tokens_c)+1)
    # print('3segment_ids',segment_ids)
    # print(len(segment_ids))
    # print('3tokensb', tokens)
    # print('3cls_token_at_end',cls_token_at_end) # false
    # assert -1 == 0
# Roberta 인ㄱ 경우 cls_token_segment_id = 0
if cls token at end:
   tokens = tokens + [cls_token]
    segment_ids = segment_ids + [cls_token_segment_id]
else:
   tokens = [cls_token] + tokens
   # print('##############"")
   # print('tokens',tokens)
    # assert -1 == 0
   segment_ids = [cls_token_segment_id] + segment_ids
    # print('3segment_ids',segment_ids)
    # print(len(segment_ids))
    # assert -1 == 0
input_ids = tokenizer.convert_tokens_to_ids(tokens)
# print('input_ids',input_ids) # 길이 : 29
# print(len(input_ids))
# assert -1 == 0
# print('mask_padding_with_zero',mask_padding_with_zero) # True
# print('Cls_token',cls_token)
# print('sep_token',sep_token)
# print('max_seq_length', max_seq_length) # 88
```

```
# assert -1 == 0
            # The mask has 1 for real tokens and 0 for padding tokens. Only real
            # tokens are attended to.
           input_mask = [1 if mask_padding_with_zero else 0] * len(input_ids)
            # print('input_mask',input_mask)
            # print(len(input_mask))
            # assert -1 == 0
            special_token_id = tokenizer.convert_tokens_to_ids([cls_token, sep_token])
            # print('special_token_id',special_token_id)
           # print(len(special_token_id))
           output mask = [1 if id in special token id else 0 for id in input ids] # 1 for mask
            # print('output_mask',output_mask)
            # assert -1 == 0
            # Zero-pad up to the sequence length.
            padding_length = max_seq_length - len(input_ids)
            # print(max_seq_length) # 88
            # print('padding_length',padding_length)
            # print('pad_on_left',pad_on_left) # false
            # print('pad_token',pad_token)
            # assert -1 == 0
            if pad_on_left:
               input_ids = ([pad_token] * padding_length) + input_ids
               segment_ids = ([pad_token_segment_id] * padding_length) + segment_ids
            else:
               input_ids = input_ids + ([pad_token] * padding_length)
                # print('input_ids',input_ids)
                # print(len(input_ids))
                # assert -1 == 0
               input_mask = input_mask + ([0 if mask_padding_with_zero else 1] * padding_length)
               output_mask = output_mask + ([1] * padding_length)
segment_ids = segment_ids + ([pad_token_segment_id] * padding_length)
                # print('input_mask',input_mask)
                # print(len(input_mask))
               # print('output_mask',output_mask)
                # print(len(output_mask))
               # print('segment_ids',segment_ids)
                # print(len(segment_ids))
               # assert -1 == 0
            assert len(input_ids) == max_seq_length
            assert len(output_mask) == max_seq_length
           assert len(input_mask) == max_seq_length
            assert len(segment_ids) == max_seq_length
           choices_features.append((tokens, input_ids, input_mask, segment_ids, output_mask))
        label = label_map[example.label]
        features.append(InputFeatures(example id=example.example id, choices features=choices features, label=label))
    print(max(tokenc_len))
    return features
# def _truncate_seq_pair(tokens_a, tokens_b,tokens_c, max_length):
      """Truncates a sequence pair in place to the maximum length."""
     \# This is a simple heuristic which will always truncate the longer sequence
     # one token at a time. This makes more sense than truncating an equal percent
     \ensuremath{\text{\#}} of tokens from each, since if one sequence is very short then each token
     \# that's truncated likely contains more information than a longer sequence.
     # print('max_length', max_length) # 84
     # assert -1 == 0
     while True:
         total_length = len(tokens_a) + len(tokens_b) + len(tokens_c)
         if total_length <= max_length:
             break
         if len(tokens_a) > len(tokens_b):
             tokens_a.pop()
         else:
             tokens b.pop()
def _truncate_seq_pair(tokens_a, tokens_b, tokens_c, max_length):
    """Truncates a sequence pair in place to the maximum length."""
    # Calculate the total length of the three sequences
    total_length = len(tokens_a) + len(tokens_b) + len(tokens_c)
    # Check if truncation is necessary
   if total length <= max_length:
       return # No truncation needed
    # Calculate how many tokens to remove from each sequence
    tokens to remove = total length - max length
```

```
# Remove tokens from the sequences
    while tokens_to_remove > 0:
        # Remove tokens from tokens_c if there are any left to remove
        if len(tokens_c) > 0:
            tokens_c.pop()
        # Remove tokens from tokens_b if there are any left to remove
        elif len(tokens b) > 0:
           tokens_b.pop()
        # Remove tokens from tokens a if there are any left to remove
        elif len(tokens a) > 0:
           tokens_a.pop()
        tokens to remove -= 1
def select_field(features, field):
    return [[choice[field] for choice in feature.choices_features] for feature in features]
def convert_features_to_tensors(features):
    all_input_ids = torch.tensor(select_field(features, 'input_ids'), dtype=torch.long)
    all_input_mask = torch.tensor(select_field(features, 'input_mask'), dtype=torch.long)
    all_segment_ids = torch.tensor(select_field(features, 'segment_ids'), dtype=torch.long)
    all_output_mask = torch.tensor(select_field(features, 'output_mask'), dtype=torch.bool)
    all_label = torch.tensor([f.label for f in features], dtype=torch.long)
    # print('pretrained',pretrained) # None
    # assert -1==0
   if pretrained:
        if os.path.exists(cache_path):
            with open(cache_path, 'rb') as f:
    all_sent_vecs, all_logits = pickle.load(f)
            model, old_args = torch.load(pretrained)
            _all_input_ids = all_input_ids.view(-1, all_input_ids.size(-1)).cuda()
            _all_input_mask = all_input_mask.view(-1, all_input_mask.size(-1)).cuda()
            _all_segment_ids = all_segment_ids.view(-1, all_segment_ids.size(-1)).cuda()
            _all_output_mask = all_output_mask.view(-1, all_output_mask.size(-1)).cuda()
            model.cuda()
            model.eval()
            bs = 256
            n = _all_input_ids.size(0)
            all_sent_vecs = []
            all_logits = []
            for a in tqdm(range(0, n, bs)):
                b = min(n, a + bs)
                lm_inputs = _all_input_ids[a:b], _all_input_mask[a:b], _all_segment_ids[a:b], _all_output_mask[a:b]
                with torch.no_grad():
                    sent vecs, all hidden states = model.encoder(*lm inputs)
                    logits = model.decoder(sent_vecs)
                all sent vecs.append(sent vecs.to(all input ids.device))
                all_logits.append(logits.to(all_input_ids.device))
            all\_sent\_vecs = torch.cat(all\_sent\_vecs, \ dim=0).view(all\_input\_ids.size(0), \ all\_input\_ids.size(1), \ -1)
            all_logits = torch.cat(all_logits, dim=0).view(all_input_ids.size(0), all_input_ids.size(1), -1)
            with open(cache_path, 'wb') as f:
                pickle.dump([all_sent_vecs, all_logits], f)
        return all_input_ids, all_input_mask, all_segment_ids, all_output_mask, all_sent_vecs, all_logits, all_label
    return all_input_ids, all_input_mask, all_segment_ids, all_output_mask, all_label
   tokenizer_class = {'bert': BertTokenizer, 'xlnet': XLNetTokenizer, 'roberta': RobertaTokenizer, 'albert': AlbertTokenizer}.get
   tokenizer_class = {'bert': BertTokenizer, 'xlnet': XLNetTokenizer, 'roberta': RobertaTokenizer}.get(model_type)
tokenizer = tokenizer class.from pretrained(model name)
examples, examples1 = read_examples(statement_jsonl_path, './data/csqa/statement/train_modified.statement.jsonl')
# format of 'add ga prefix' or 'fairseg'
# for example in examples:
   # example.contexts = ['0: ' + c for c in example.contexts]
    # example.endings = ['A: ' + e for e in example.endings]
   # # example.contexts = ['question is ' + c for c in example.contexts]
# example.endings = ['answer is ' + e + '.' for e in example.endings]
features = convert_examples_to_features(examples, examples1, list(range(len(examples[0].endings))), max_seq_length, tokenizer,
                                         cls_token_at_end=bool(model_type in ['xlnet']), # xlnet has a cls token at the end
                                         cls_token=tokenizer.cls_token,
                                         sep_token=tokenizer.sep_token,
                                         sep_token_extra=bool(model_type in ['roberta', 'albert']),
                                         cls_token_segment_id=2 if model_type in ['xlnet'] else 0,
                                         pad_on_left=bool(model_type in ['xlnet']), # pad on the left for xlnet
                                         pad_token_segment_id=4 if model_type in ['xlnet'] else 0,
                                         sequence_b_segment_id=0 if model_type in ['roberta', 'albert'] else 1,sequence_c_segment_id
example_ids = [f.example_id for f in features]
*data_tensors, all_label = convert_features_to_tensors(features)
return (example_ids, all_label, *data_tensors)
```

```
def load_input_tensors(input_jsonl_path, model_type, model_name, max_seq_length, pretrained):
    if model_type in ('lstm',):
        raise NotImplementedError
    elif model_type in ('gpt',):
        return\ load\_gpt\_input\_tensors(input\_jsonl\_path,\ max\_seq\_length)
    elif model_type in ('bert', 'xlnet', 'roberta', 'albert'):
        return\ load\_bert\_xlnet\_roberta\_input\_tensors(input\_jsonl\_path,\ model\_type,\ model\_name,\ max\_seq\_length,\ pretrained)
def load_info(statement_path: str):
    n = sum(1 for _ in open(statement_path, "r"))
    num_choice = None
    with open(statement_path, "r", encoding="utf-8") as fin:
        ids = []
        labels = []
        for line in fin:
            input_json = json.loads(line)
            labels.append(ord(input_json.get("answerKey", "A")) - ord("A"))
            ids.append(input_json['id'])
            if num_choice is None:
                num_choice = len(input_json["question"]["choices"])
        labels = torch.tensor(labels, dtype=torch.long)
    return ids, labels, num_choice
def load_statement_dict(statement_path):
    all dict = {}
    with open(statement_path, 'r', encoding='utf-8') as fin:
        for line in fin:
            instance_dict = json.loads(line)
            qid = instance_dict['id']
            all_dict[qid] = {
                 'question': instance_dict['question']['stem'],
                'answers': [dic['text'] for dic in instance_dict['question']['choices']]
    return all_dict
```

gsc.py에서 dataset 호출하는 부분이다.

여기서 args.train\_cycle\_statements, args.dev\_cycle\_statements를 주어야함

modeling.py에서

LM\_QAGSC\_DataLoader 부분을보면

```
self.train_qids, self.train_labels, *self.train_encoder_data = load_input_tensors(train_statement_path, model_type, model_name, max_sec_self.dev_qids, self.dev_labels, *self.dev_encoder_data = load_input_tensors(dev_statement_path, model_type, model_name, max_sec_self.dev_encoder_data = load_input_tensors(dev_statement_path, model_type, model_name, model_type, model_name, model_type, model_name, model_type, model_name, model_type, model_name, model_type, model_
```

load\_input\_tensors의 첫번째 인자가 train\_statement\_path, dev\_statement\_path

추가로 train\_cycle\_statements, dev\_cycle\_statements 넣기

그러면 sh파일에서도 수정해야함

```
export CUDA_VISIBLE_DEVICES=1,2
dt='date '+%Y%m%d_%H%M%S'`
dataset="csqa"
```

```
model='roberta-large'
 shift
 shift
args=$@
elr="1e-5"
dlr="1e-2"
weight_decay="1e-3"
n_epochs=30
bs=128
# mbs=16
# ebs=32
mbs=4
k=2 #num of gnn layers
echo "**** hyperparameters ****"
echo "dataset: $dataset"
echo "enc_name: $model"
echo "batch_size: $bs"
echo "learning_rate: elr $elr dlr $dlr"
echo "edge_encoder_dim $enc_dim gsc_layer $k" echo "***************
save_dir_pref='saved_models'
logs_dir_pref='logs/gsc'
mkdir -p $save dir pref
mkdir -p $logs_dir_pref
###### Training ######
 for seed in 0; do
     echo "Training with seed: $seed"
    python3 -u gsc1.py --dataset $dataset \
               --encoder $model -k $k --enc_dim $enc_dim \
              -elr -dr -dr -bs -bs -bs -bs -ebs {ebs} --weight_decay {weight_decay} --seed -elc -cay --seed --weight_decay --seed --seed --weight_decay --weight_decay --seed --weight_decay --weight_decay --seed --weight_decay --w
             --n_epochs $n_epochs --max_epochs_before_stop 10
             --train_adj data/${dataset}/graph/train.graph.adj.pk \
             --dev_adj data/{\frac{dataset}{graph/dev.graph.adj.pk}}
             --test_adj data/{\frac{\phi}{\phi}}
             --train_statements data/\frac{dataset}{statement/train.statement.jsonl}
             --dev\_statements \quad data/\$\{dataset\}/statement/dev.statement.jsonl~`
             --test_statements data/${dataset}/statement/test.statement.jsonl \
             --max_seq_len 88
             --num relation 38
             --inhouse True \
             --unfreeze_epoch 4 \
             --log_interval 20 \
             --save model \
               --save_dir {\sve_dir_pref}/{\dataset}/enc-{\model}_k{k}_encdim{enc_dim}_bs{bs}_sed{\seed}_${dt} \ \args \ \
```

sh파일에서 받는 인자가 gsc.py의 dataset에 들어가고, dataset에 맞는 클래스(LM\_QAGSC\_dataloader)은 modeling.py에 있고 이 값은 load\_input\_tensor 함수에 들어감

## 이것은 Load\_input\_tensor

```
def load_input_tensors(input_jsonl_path, model_type, model_name, max_seq_length, pretrained):
    if model_type in ('lstm',):
        raise NotImplementedError
    elif model_type in ('gpt',):
        return load_gpt_input_tensors(input_jsonl_path, max_seq_length)
    elif model_type in ('bert', 'xlnet', 'roberta', 'albert'):
        return load_bert_xlnet_roberta_input_tensors(input_jsonl_path, model_type, model_name, max_seq_length, pretrained)
```

우리는 load\_bert\_xlnet\_roberta\_input\_tensors에 input\_cycle\_jsonl\_path를 넣는다.

아예 context score에 추가하는 cycle passage를 위해

run\_idgsc\_\_csqa\_cycle.sh, gsc1\_cycle.py, modeling\_model\_cycle.py, data\_utils\_cycle.py 에서 진행하자. 테스트 데이터에 대해서 context score에 추가할 cycle\_passage를 만들경우는 in-house가 아닌 경우에 필요함

### 노드 개수가 늘어날수록 DPR이 너무 오래걸려 64개부터 진행 ㄴㄴ. 그 이유는 이렇게 둘러대면 되지 않을까?

1. 노드 개수가 늘어날수록 Other node가 증가한다. 증가하면 물론 보조적인 도움을 줘서 영향을 줄 수도 있지만 그렇지 않을수도 있다. 그래서 BM25만으로 먼저 많이 중복되는 단어를 거르는것만으로도 효과가 있을것이다. 그 후에 reranking을 진행하는게 time-saving이다.