```
In [1]: import re
        import json
        import argparse
        import os
        import sys
        import torch
        import pandas as pd
        import numpy as np
        from tqdm import tqdm
        from nltk.stem import PorterStemmer
        # from transformers import GPT2Tokenizer, GPT2Model
        from transformers import AutoTokenizer, AutoModelForCausalLM
        from huggingface_hub import login
        from swisscom_ai.research_keyphrase.preprocessing.postagging import PosTaggingCoreNLP
        from swisscom_ai.research_keyphrase.model.input_representation import InputTextObj
        from swisscom_ai.research_keyphrase.model.extractor import extract_candidates
       C:\Users\user0\anaconda3\envs\LLM_ENV\Lib\site-packages\tqdm\auto.py:21: TqdmWarning: IProgress n
       ot found. Please update jupyter and ipywidgets. See https://ipywidgets.readthedocs.io/en/stable/u
       ser install.html
        from .autonotebook import tqdm as notebook_tqdm
In [2]: host = 'localhost'
        port = 9000
        pos_tagger = PosTaggingCoreNLP(host, port)
        # Load stopwords
        stopwords = []
        with open('UGIR_stopwords.txt', "r") as f:
            for line in f:
                if line:
                     stopwords.append(line.replace('\n', ''))
        stemmer = PorterStemmer()
        def read_jsonl(path):
            data = []
            with open(path, 'r') as f:
                for line in f:
                    item = json.loads(line.strip())
                    data.append(item)
            return data
In [3]: def get_candidates(core_nlp, text):
            tagged = core_nlp.pos_tag_raw_text(text)
            text_obj = InputTextObj(tagged, 'en')
            candidates = extract_candidates(text_obj)
            return candidates
        def get_phrase_indices(text_tokens, phrase, prefix):
            text_tokens = [t.replace(prefix, '') for t in text_tokens]
            phrase = phrase.replace(' ', '')
            matched_indices = []
            matched_index = []
```

```
target = phrase
    for i in range(len(text_tokens)):
        cur_token = text_tokens[i]
        sub_len = min(len(cur_token), len(phrase))
        if cur_token[:sub_len].lower() == target[:sub_len]:
            matched index.append(i)
            target = target[sub len:]
            if len(target) == 0:
                matched_indices.append([matched_index[0], matched_index[-1] + 1])
                target = phrase
        else:
            matched_index = []
            target = phrase
            if cur_token[:sub_len].lower() == target[:sub_len]:
                matched index.append(i)
                target = target[sub_len:]
                if len(target) == 0:
                    matched_indices.append([matched_index[0], matched_index[-1] + 1])
                    target = phrase
    return matched_indices
def remove_repeated_sub_word(candidates_pos_dict):
    for phrase in candidates_pos_dict.keys():
        split_phrase = re.split(r'\s+ -', phrase)
        split_phrase = list(filter(None, split_phrase))
        if len(split_phrase) > 1:
            for word in split_phrase:
                if word in candidates_pos_dict:
                    single_word_positions = candidates_pos_dict[word]
                    phrase_positions = candidates_pos_dict[phrase]
                    single_word_alone_positions = [pos for pos in single_word_positions if not al
                        pos[0] >= phrase_pos[0] and pos[1] <= phrase_pos[1] for phrase_pos in ph
                    candidates_pos_dict[word] = single_word_alone_positions
    return candidates_pos_dict
def get_all_indices(candidates_pos_dict, window_end):
    all_indices = []
    for phrase in candidates_pos_dict.values():
        for element in phrase:
            start_index = element[0]
            end_index = element[1]
            if(start_index>window_end):
                all_indices.extend(range(start_index, end_index))
    all_indices = sorted(all_indices)
    all_indices = list(dict.fromkeys(all_indices))
   return all_indices
def aggregate_phrase_scores(index_list, tokens_scores):
   total_score = 0.0
   for p_index in index_list:
        part_sum = tokens_scores[p_index[0]:p_index[1]].sum()
        total_score += part_sum
    return total_score
```

```
def get_score_full(candidates, references, maxDepth=15):
   precision = []
   recall = []
   reference_set = set(references)
   referencelen = len(reference_set)
   true positive = 0
   for i in range(maxDepth):
        if len(candidates) > i:
            kp_pred = candidates[i]
            if kp_pred in reference_set:
                true_positive += 1
           precision.append(true_positive / float(i + 1))
            recall.append(true_positive / float(referencelen))
            precision.append(true_positive / float(len(candidates)))
            recall.append(true_positive / float(referencelen))
    return precision, recall
def evaluate_document(candidates, ground_truth):
    results = {}
    precision_scores, recall_scores, f1_scores = {5: [], 10: [], 15: []}, \
                                                 {5: [], 10: [], 15: []}, \
                                                 {5: [], 10: [], 15: []}
   for candidate, gt in zip(candidates, ground_truth):
        p, r = get_score_full(candidate, gt)
        for i in [5, 10, 15]:
           precision = p[i - 1]
           recall = r[i - 1]
           if precision + recall > 0:
                f1_scores[i].append((2 * (precision * recall)) / (precision + recall))
           else:
                f1_scores[i].append(0)
           precision_scores[i].append(precision)
            recall_scores[i].append(recall)
    print("###############\nMetrics")
    for i in precision_scores:
        print("@{}".format(i))
        print("F1:{}".format(np.mean(f1_scores[i])))
        print("P:{}".format(np.mean(precision_scores[i])))
        print("R:{}".format(np.mean(recall_scores[i])))
        top_n_p = 'precision@' + str(i)
        top_n_r = 'recall@' + str(i)
        top_n_f1 = 'f1@' + str(i)
        results[top_n_p] = np.mean(precision_scores[i])
        results[top_n_r] = np.mean(recall_scores[i])
        results[top_n_f1] = np.mean(f1_scores[i])
    print("##############")
   return results
def evaluate_dataset(predicted_top, dataset, score_type, dataset_name):
    experiment_results = []
    gt keyphrase list = []
   predicted_keyphrase_list = []
   for i in range(len(dataset)):
```

```
predicted_keyphrase = predicted_top[i]
    predicted_keyphrase = [phrase.lower() for phrase in predicted_keyphrase]
    predicted_keyphrase_list.append(predicted_keyphrase)

gt_keyphrase = [key.lower() for key in dataset[i]['keyphrases']]
    gt_keyphrase_list.append(gt_keyphrase)

total_score = evaluate_document(predicted_keyphrase_list, gt_keyphrase_list)
    experiment_results.append(total_score)

df = pd.DataFrame(experiment_results)

path = f'experiment_results/{dataset_name}/'
    os.makedirs(path, exist_ok=True)
    df.to_csv(f'{path}score_type_{score_type}.csv', index=False)

top3_f1_5 = df.nlargest(3, 'f1@5').reset_index(drop=True)
    top3_f1_10 = df.nlargest(3, 'f1@10').reset_index(drop=True)
    top3_f1_15 = df.nlargest(3, 'f1@15').reset_index(drop=True)
    return top3_f1_5, top3_f1_10, top3_f1_15
```

```
In [4]: login(token= )
```

The token has not been saved to the git credentials helper. Pass `add_to_git_credential=True` in this function directly or `--add-to-git-credential` if using via `huggingface-cli` if you want to set the git credential as well.

Token is valid (permission: read).

Your token has been saved to C:\Users\user0\.cache\huggingface\token Login successful

```
In [5]: tokenizer = AutoTokenizer.from_pretrained("meta-llama/Meta-Llama-3-8B")
model = AutoModelForCausalLM.from_pretrained("meta-llama/Meta-Llama-3-8B", attn_implementation="opening to be a second or content of the second or content o
```

C:\Users\user0\anaconda3\envs\LLM_ENV\Lib\site-packages\huggingface_hub\file_download.py:1132: Fu tureWarning: `resume_download` is deprecated and will be removed in version 1.0.0. Downloads always resume when possible. If you want to force a new download, use `force_download=True`. warnings.warn(

Special tokens have been added in the vocabulary, make sure the associated word embeddings are fine-tuned or trained.

Loading checkpoint shards: 100% 4/4 [00:12<00:00, 3.08s/it]

C:\Users\user0\anaconda3\envs\LLM_ENV\Lib\site-packages\huggingface_hub\file_download.py:1132: Fu
tureWarning: `resume_download` is deprecated and will be removed in version 1.0.0. Downloads alwa
ys resume when possible. If you want to force a new download, use `force_download=True`.
 warnings.warn(

```
In [6]: device = "cuda" if torch.cuda.is_available() else "cpu"
print(f"device: {device}")
```

device: cuda

DATASET: INSPEC

```
In [7]: dataset_name = "inspec"
   dataset = read_jsonl("KEYWORD_DATA/{}.jsonl".format(dataset_name))
```

```
In [8]: model.to(device)
        model.eval()
        dataset_att_scores_overall = []
        for data in tqdm(dataset):
           with torch.no grad():
               input text = data["text"]
               tokenized_content = tokenizer(input_text, return_tensors='pt')
               outputs = model(**tokenized_content.to(device))
               content_tokens = tokenizer.convert_ids_to_tokens(tokenized_content['input_ids'].squeeze()
               attentions = outputs.attentions
               del outputs
               candidates = get_candidates(pos_tagger, input_text)
               candidates = [phrase for phrase in candidates if phrase.split(' ')[0] not in stopwords]
               candidates_indices = {}
               for phrase in candidates:
                   matched_indices = get_phrase_indices(content_tokens, phrase, prefix)
                   if len(matched_indices) == 0:
                       continue
                   candidates_indices[phrase] = matched_indices
               candidates_indices = remove_repeated_sub_word(candidates_indices)
               # ATTENTION MEASSUREMENT
               attentions = sum(attentions)/len(attentions)
               attentions = attentions.squeeze(0)
               att scores = attentions.mean(0).sum(0)
               att_scores[0] = 0
               phrase_score_dict = {}
               for phrase in candidates_indices.keys():
                   try:
                       phrase_indices = candidates_indices[phrase]
                       if len(phrase_indices) == 0:
                          continue
                   except KeyError:
                       continue
                   final_phrase_score = aggregate_phrase_scores(phrase_indices, att_scores)
                   if len(phrase.split()) == 1:
                       final_phrase_score = final_phrase_score / len(phrase_indices)
                   phrase_score_dict[phrase] = final_phrase_score
           sorted_scores_att_o_s = sorted(phrase_score_dict.items(), key=lambda item: item[1], reverse=
           stemmed_sorted_scores_att_o_s = [(" ".join(stemmer.stem(word) for word in phrase.split()), s
                                      phrase, score in sorted_scores_att_o_s]
           set_stemmed_scores_list_att_o_s = []
           for phrase, score in stemmed_sorted_scores_att_o_s:
               if phrase not in set_stemmed_scores_list_att_o_s:
                   set_stemmed_scores_list_att_o_s.append(phrase)
           pred_stemmed_phrases_att_o_s = set_stemmed_scores_list_att_o_s[:15]
```

```
dataset_att_scores_overall.append(pred_stemmed_phrases_att_o_s)
 att_o_s_top3_f1_5, att_o_s_top3_f1_10, att_o_s_top3_f1_15 = evaluate_dataset(dataset_att_scores_
100% | 500/500 [01:55<00:00, 4.33it/s]
#############################
Metrics
@5
F1:0.3424774238136697
P:0.4784
R:0.29396447932697767
@10
F1:0.3818415536037738
P:0.38055952380952385
R:0.4318405058729329
@15
F1:0.381147979499465
P:0.32708260073260065
R:0.5150753282616488
```

ATTENTIONSEEKER

```
In [9]: model.to(device)
        model.eval()
        dataset att scores overall = []
        for data in tqdm(dataset):
           with torch.no_grad():
               input_text = data["text"]
               tokenized_content = tokenizer(input_text, return_tensors='pt')
               outputs = model(**tokenized_content.to(device))
               content_tokens = tokenizer.convert_ids_to_tokens(tokenized_content['input_ids'].squeeze()
               all_attentions = outputs.attentions
               del outputs
               candidates = get_candidates(pos_tagger, input_text)
               candidates = [phrase for phrase in candidates if phrase.split(' ')[0] not in stopwords]
               candidates_indices = {}
               for phrase in candidates:
                   matched_indices = get_phrase_indices(content_tokens, phrase, prefix)
                   if len(matched_indices) == 0:
                       continue
                   candidates_indices[phrase] = matched_indices
               candidates_indices = remove_repeated_sub_word(candidates_indices)
               all_indices = get_all_indices(candidates_indices,0)
               len_t_tokens = all_attentions[0].squeeze(0)[0].shape[0]
               all_indices_tensor = torch.arange(len_t_tokens)
               mask = torch.isin(all_indices_tensor, torch.tensor(all_indices)).to(device)
               mask_1 = mask*1.0
               # ATTENTION MEASSUREMENT
               attentions = torch.zeros(len_t_tokens).to(device)
               for layer in range(len(all_attentions)):
                   for head in range(32):
```

```
crrn_att_map = all_attentions[layer].squeeze(0)[head].clone()
               crrn_att_map[~mask] = 0
               lh_weight = torch.matmul(crrn_att_map,mask_1)
               attentions += lh_weight.mean(0)*torch.matmul(lh_weight,crrn_att_map)
       att_scores = attentions
       att_scores[0] = 0
       # ATTENTION SEEKER (LHC-SEEKER)
       f_att_scores = torch.zeros_like(att_scores)
       f att scores[mask] = att scores[mask]
       # NEW ATTENTION MEASSUREMENT
       attentions = torch.zeros(len_t_tokens).to(device)
       for layer in range(len(all attentions)):
           for head in range(32):
               crrn_att_map = all_attentions[layer].squeeze(0)[head].clone()
               crrn_att_map[~mask] = 0
               lh_weight = torch.matmul(crrn_att_map,f_att_scores)
               attentions += lh_weight.mean(0)*torch.matmul(lh_weight,crrn_att_map)
       att scores = attentions
       att scores[0] = 0
       phrase_score_dict = {}
       for phrase in candidates_indices.keys():
           try:
               phrase indices = candidates indices[phrase]
               if len(phrase_indices) == 0:
                   continue
           except KeyError:
               continue
           final phrase score = aggregate phrase scores(phrase indices, att scores)
           if len(phrase.split()) == 1:
               final_phrase_score = final_phrase_score / len(phrase_indices)
           phrase_score_dict[phrase] = final_phrase_score
   sorted_scores_att_o_s = sorted(phrase_score_dict.items(), key=lambda item: item[1], reverse=
   stemmed_sorted_scores_att_o_s = [(" ".join(stemmer.stem(word) for word in phrase.split()), s
                               phrase, score in sorted_scores_att_o_s]
   set_stemmed_scores_list_att_o_s = []
   for phrase, score in stemmed_sorted_scores_att_o_s:
       if phrase not in set_stemmed_scores_list_att_o_s:
           set_stemmed_scores_list_att_o_s.append(phrase)
   pred_stemmed_phrases_att_o_s = set_stemmed_scores_list_att_o_s[:15]
   dataset_att_scores_overall.append(pred_stemmed_phrases_att_o_s)
att_o_s_top3_f1_5, att_o_s_top3_f1_10, att_o_s_top3_f1_15 = evaluate_dataset(dataset_att_scores_o
```

DATASET: SEMEVAL 2017

```
In [11]: dataset_name = "semeval2017"
    dataset = read_jsonl("KEYWORD_DATA/{}.jsonl".format(dataset_name))
```

SAMRANK BASE

```
In [12]: |
        model.to(device)
        model.eval()
        dataset_att_scores_overall = []
        for data in tqdm(dataset):
            with torch.no grad():
               input_text = data["text"]
               tokenized_content = tokenizer(input_text, return_tensors='pt')
               outputs = model(**tokenized_content.to(device))
               content_tokens = tokenizer.convert_ids_to_tokens(tokenized_content['input_ids'].squeeze()
               attentions = outputs.attentions
               del outputs
               candidates = get_candidates(pos_tagger, input_text)
               candidates = [phrase for phrase in candidates if phrase.split(' ')[0] not in stopwords]
               candidates_indices = {}
               for phrase in candidates:
                   matched_indices = get_phrase_indices(content_tokens, phrase, prefix)
                   if len(matched_indices) == 0:
                      continue
                   candidates_indices[phrase] = matched_indices
               candidates_indices = remove_repeated_sub_word(candidates_indices)
               # ATTENTION MEASSUREMENT
               attentions = sum(attentions)/len(attentions)
               attentions = attentions.squeeze(0)
               att_scores = attentions.mean(0).sum(0)
               att scores[0] = 0
```

```
phrase_score_dict = {}
         for phrase in candidates_indices.keys():
              try:
                  phrase_indices = candidates_indices[phrase]
                  if len(phrase_indices) == 0:
                      continue
              except KeyError:
                  continue
             final_phrase_score = aggregate_phrase_scores(phrase_indices, att_scores)
             if len(phrase.split()) == 1:
                  final_phrase_score = final_phrase_score / len(phrase_indices)
              phrase_score_dict[phrase] = final_phrase_score
     sorted_scores_att_o_s = sorted(phrase_score_dict.items(), key=lambda item: item[1], reverse=
     stemmed_sorted_scores_att_o_s = [(" ".join(stemmer.stem(word) for word in phrase.split()), setemmed_sorted_scores_att_o_s
                                  phrase, score in sorted scores att o s]
     set_stemmed_scores_list_att_o_s = []
     for phrase, score in stemmed_sorted_scores_att_o_s:
         if phrase not in set_stemmed_scores_list_att_o_s:
              set_stemmed_scores_list_att_o_s.append(phrase)
     pred_stemmed_phrases_att_o_s = set_stemmed_scores_list_att_o_s[:15]
     dataset_att_scores_overall.append(pred_stemmed_phrases_att_o_s)
 att_o_s_top3_f1_5, att_o_s_top3_f1_10, att_o_s_top3_f1_15 = evaluate_dataset(dataset_att_scores_o
100%| 493/493 [02:36<00:00, 3.15it/s]
###########################
Metrics
@5
F1:0.24742282955536773
P:0.5168356997971603
R:0.1700399499230244
@10
F1:0.3350504558633974
P:0.4480730223123732
R:0.2851300385057317
@15
F1:0.3701308557000764
P:0.3989408495493688
R:0.3718292969857578
##############################
```

ATTENTIONSEEKER

```
In [13]: model.to(device)
    model.eval()

dataset_att_scores_overall = []

for data in tqdm(dataset):
    with torch.no_grad():
        input_text = data["text"]

    tokenized_content = tokenizer(input_text, return_tensors='pt')
    outputs = model(**tokenized_content.to(device))
```

```
content_tokens = tokenizer.convert_ids_to_tokens(tokenized_content['input_ids'].squeeze()
all attentions = outputs.attentions
del outputs
candidates = get_candidates(pos_tagger, input_text)
candidates = [phrase for phrase in candidates if phrase.split(' ')[0] not in stopwords]
candidates_indices = {}
for phrase in candidates:
   matched_indices = get_phrase_indices(content_tokens, phrase, prefix)
   if len(matched_indices) == 0:
       continue
   candidates_indices[phrase] = matched_indices
candidates_indices = remove_repeated_sub_word(candidates_indices)
all_indices = get_all_indices(candidates_indices,0)
len t tokens = all attentions[0].squeeze(0)[0].shape[0]
all indices tensor = torch.arange(len t tokens)
mask = torch.isin(all_indices_tensor, torch.tensor(all_indices)).to(device)
mask_1 = mask*1.0
# ATTENTION MEASSUREMENT
attentions = torch.zeros(len_t_tokens).to(device)
for layer in range(len(all_attentions)):
   for head in range(32):
       crrn_att_map = all_attentions[layer].squeeze(0)[head].clone()
       crrn_att_map[~mask] = 0
       lh_weight = torch.matmul(crrn_att_map,mask_1)
       attentions += lh_weight.mean(0)*torch.matmul(lh_weight,crrn_att_map)
att_scores = attentions
att scores[0] = 0
# ATTENTION SEEKER (LHC-SEEKER)
f_att_scores = torch.zeros_like(att_scores)
f_att_scores[mask] = att_scores[mask]
# NEW ATTENTION MEASSUREMENT
attentions = torch.zeros(len_t_tokens).to(device)
for layer in range(len(all attentions)):
   for head in range(32):
       crrn_att_map = all_attentions[layer].squeeze(0)[head].clone()
       crrn_att_map[~mask] = 0
       lh_weight = torch.matmul(crrn_att_map,f_att_scores)
       attentions += lh_weight.mean(0)*torch.matmul(lh_weight,crrn_att_map)
att_scores = attentions
att scores[0] = 0
phrase_score_dict = {}
for phrase in candidates_indices.keys():
   try:
       phrase_indices = candidates_indices[phrase]
       if len(phrase_indices) == 0:
           continue
   except KevError:
       continue
   final_phrase_score = aggregate_phrase_scores(phrase_indices, att_scores)
   if len(phrase.split()) == 1:
       final phrase score = final phrase score / len(phrase indices)
```

```
phrase_score_dict[phrase] = final_phrase_score
     sorted_scores_att_o_s = sorted(phrase_score_dict.items(), key=lambda item: item[1], reverse=
     stemmed_sorted_scores_att_o_s = [(" ".join(stemmer.stem(word) for word in phrase.split()), s
                               phrase, score in sorted_scores_att_o_s]
     set_stemmed_scores_list_att_o_s = []
     for phrase, score in stemmed_sorted_scores_att_o_s:
         if phrase not in set_stemmed_scores_list_att_o_s:
             set_stemmed_scores_list_att_o_s.append(phrase)
     pred_stemmed_phrases_att_o_s = set_stemmed_scores_list_att_o_s[:15]
     dataset_att_scores_overall.append(pred_stemmed_phrases_att_o_s)
 att_o_s_top3_f1_5, att_o_s_top3_f1_10, att_o_s_top3_f1_15 = evaluate_dataset(dataset_att_scores_o
100% | 493/493 [06:20<00:00, 1.30it/s]
Metrics
@5
F1:0.25397632889673977
P:0.5330628803245435
R:0.1741208725412833
@10
F1:0.345326367943574
P:0.46186612576064906
R:0.2942509747499267
F1:0.3849912207748741
P:0.4150328035723573
R:0.38667026905240437
```

In []: