

Children often ask their caregivers to find episodes of their favorite TV shows based only on a very short (and loosely relevant!) description of it ("the one where Arthur has a wiggly tooth") but video services like Netflix and Amazon don't currently provide such content-based search. Given summaries of each episode, can we use sequence embeddings to solve this retrieval problem?

Before beginning this homework, install the following libraries:

```
conda install -c huggingface transformers
pip install -U sentence-transformers
conda install -c conda-forge ipywidgets
```

First, let's read in our data for the TV show "Wild Kratts" (from [Wikipedia](#)), which has the following (tab-separated) form:

Episode	Title	Summary
1	Mom of a Croc	<p>At the Nile River, zoologists Chris and Martin Kratt (voiced by their real-world selves) are on a mission to show one of their fellow Wild Kratts team members—brilliant young inventor Aviva Corcovado (Athena Karkanis)—that there's more to crocodiles than just violence and snapping jaws. After shrinking themselves down to a few inches tall by using Aviva's Miniaturizer invention, the Kratt Brothers disguise themselves as crocodile eggs and sneak into a mother crocodile's new nest. In the Wild Kratts team's turtle-shaped aircraft and headquarters—the Tortuga, one of Aviva's greatest inventions—the Wild Kratts tech team, consisting of Aviva, communications expert and mechanic Koki (Heather Bambrick), and skilled pilot Jimmy Z (Jonathan Malen) monitor Chris and Martin and watch as the mother crocodile faithfully guards her nest against predators for months without even eating anything. Eventually, as the crocodile eggs hatch and the crocodile mom uses her mouth to carry several of her newly hatched babies to the river, Aviva changes her mind about crocodiles and decides that these reptiles are in fact caring and dedicated mothers. But when the mother crocodile leaves the river to go get more hatchlings from her nest, predators threaten the first batch of baby crocodiles. The Kratt Brothers must use the incredible Creature Power Suits—two of Aviva's inventions—to gain the abilities of crocodiles and protect the vulnerable crocodile hatchlings.</p>
2	Whale of a Squid	<p>The Kratt Brothers use Aviva's amphipod-inspired submersible, the Amphisub, to dive into the deep waters of the Southern Ocean. There, they witness a never-before-seen wildlife moment: a battle between a sperm whale and a giant squid. However, the water pressure at the extreme depths where the battle is taking place badly damages and partially crushes the Amphisub, forcing Aviva to use her new ExtendoArm invention to pull the submersible back to the Tortuga. To allow Chris and Martin to return to the site of the whale-versus-squid battle, Aviva programs two new Creature Power Suits—Sperm Whale Power for Chris, and Squid Power for Martin. The Kratt Brothers use their new Creature Powers to dive back into the deep sea, where the sperm whale and the giant squid are still locked in combat. Suddenly, the sperm whale becomes entangled in a discarded fishing net and begins sinking toward an area full of underwater volcanoes. To make matters worse, a colossal squid attacks the sperm whale's calf. Chris and Martin must put their Creature Powers of both sperm whale and squid to good use to rescue the mother sperm whale and her calf.</p>

```
In [1]: def read_data(filename):
        data=[]
        with open(filename, encoding="utf-8") as file:
            for line in file:
                cols=line.rstrip().split("\t")
                episode=cols[0]
                title=cols[1]
                summary=cols[2]
                data.append((episode, title, summary))
        return data
```

```
In [2]: data=read_data("../data/wild_kratts_episodes.txt")
```

```
In [3]: def get_document_reps_for_data(data, sequence_embedding_function, model):  
  
    # This function applies the sequence_embedding_function argument (a function itself  
    # element in the input data list, and returns a copy of that list with an embedding  
  
    data_with_reps=[]  
  
    for episode, title, summary in data:  
        data_with_reps.append((episode, title, summary, sequence_embedding_function(mod  
  
    return data_with_reps
```

```
In [4]: def cosine_similarity(a, b):  
    return np.dot(a, b)/(np.linalg.norm(a)*np.linalg.norm(b))
```

First, we may be tempted to use the [CLS] token for BERT to represent an entire input string (as is often done in *supervised* document classification models). How well does this work as an out-of-the-box document representation not optimized for our particular task?

```
In [5]: from transformers import BertModel, BertTokenizer  
import numpy as np  
from sentence_transformers import SentenceTransformer
```

```
In [6]: tokenizer = BertTokenizer.from_pretrained('bert-base-uncased')  
model = BertModel.from_pretrained('bert-base-uncased')
```

Some weights of the model checkpoint at bert-base-uncased were not used when initializing BertModel: ['cls.predictions.bias', 'cls.predictions.transform.dense.bias', 'cls.predictions.transform.dense.weight', 'cls.predictions.transform.LayerNorm.bias', 'cls.seq_relationship.bias', 'cls.seq_relationship.weight', 'cls.predictions.transform.LayerNorm.weight', 'cls.predictions.decoder.weight']

- This IS expected if you are initializing BertModel from the checkpoint of a model trained on another task or with another architecture (e.g. initializing a BertForSequenceClassification model from a BertForPreTraining model).

- This IS NOT expected if you are initializing BertModel from the checkpoint of a model that you expect to be exactly identical (initializing a BertForSequenceClassification model from a BertForSequenceClassification model).

Q1: Fill out the `get_cls_token_for_doc` function to return the [CLS] embedding for the input string. The output should be a single 768-dimensional numpy vector (see `4.embeddings/BERT.ipynb` for converting between a pytorch tensor and a numpy object).

```
In [68]: def get_cls_token_for_doc(model, string):  
    inputs = tokenizer(string, return_tensors="pt")  
    # your code goes here  
  
    #getting the tokens  
    tokenizer.convert_ids_to_tokens(inputs["input_ids"][0])
```

```

outputs = model(**inputs)

#last_hidden_states = outputs.last_hidden_state

return outputs.last_hidden_state[0].detach().numpy()[2]

```

```

In [69]: bert_cls_data=get_document_reps_for_data(data, get_cls_token_for_doc, model)

```

Q2: Use these representations to find the episode that is most similar to the description "The one where they bounce back in time" by having the highest cosine similarity between representations. A sample function shell `run_query` is provided below, along with the only arguments you need, but feel free to adapt it as you see fit.

```

In [71]: query="The one where they bounce back in time"

```

```

In [82]: def run_query(query, data_with_reps, sequence_embedding_function, model):
# your code goes here
vals = []

for eps in data_with_reps:
    comp_rep = sequence_embedding_function(model, query)
    cos_sim = cosine_similarity(eps[3], comp_rep)
    vals.append((cos_sim, query, eps[0]))

for c, q, s in reversed(sorted(vals)):
    print("%.3f\t%s\t%s" % (c, q, s))

```

```

In [83]: run_query(query, bert_cls_data, get_cls_token_for_doc, model)

```

```

0.396 The one where they bounce back in time 62
0.377 The one where they bounce back in time 111
0.373 The one where they bounce back in time 75
0.369 The one where they bounce back in time 92
0.366 The one where they bounce back in time 44
0.365 The one where they bounce back in time 81
0.350 The one where they bounce back in time 65
0.348 The one where they bounce back in time 32
0.347 The one where they bounce back in time 1
0.347 The one where they bounce back in time 118
0.345 The one where they bounce back in time 102
0.342 The one where they bounce back in time 150
0.333 The one where they bounce back in time 5
0.332 The one where they bounce back in time 96
0.330 The one where they bounce back in time 30
0.328 The one where they bounce back in time 36
0.323 The one where they bounce back in time 46
0.318 The one where they bounce back in time 91
0.318 The one where they bounce back in time 88
0.315 The one where they bounce back in time 121
0.312 The one where they bounce back in time 132
0.307 The one where they bounce back in time 55
0.304 The one where they bounce back in time 87
0.302 The one where they bounce back in time 149
0.302 The one where they bounce back in time 119

```

0.300	The one where they bounce back in time	74
0.299	The one where they bounce back in time	139
0.299	The one where they bounce back in time	4
0.298	The one where they bounce back in time	130
0.297	The one where they bounce back in time	11
0.296	The one where they bounce back in time	127
0.295	The one where they bounce back in time	59
0.294	The one where they bounce back in time	86
0.294	The one where they bounce back in time	66
0.294	The one where they bounce back in time	131
0.294	The one where they bounce back in time	58
0.290	The one where they bounce back in time	73
0.290	The one where they bounce back in time	8
0.287	The one where they bounce back in time	15
0.287	The one where they bounce back in time	112
0.287	The one where they bounce back in time	77
0.286	The one where they bounce back in time	138
0.286	The one where they bounce back in time	21
0.286	The one where they bounce back in time	9
0.285	The one where they bounce back in time	10
0.284	The one where they bounce back in time	24
0.284	The one where they bounce back in time	12
0.283	The one where they bounce back in time	128
0.283	The one where they bounce back in time	89
0.282	The one where they bounce back in time	100
0.280	The one where they bounce back in time	33
0.279	The one where they bounce back in time	70
0.278	The one where they bounce back in time	84
0.278	The one where they bounce back in time	25
0.277	The one where they bounce back in time	141
0.277	The one where they bounce back in time	122
0.276	The one where they bounce back in time	137
0.275	The one where they bounce back in time	72
0.273	The one where they bounce back in time	95
0.273	The one where they bounce back in time	90
0.270	The one where they bounce back in time	117
0.269	The one where they bounce back in time	41
0.269	The one where they bounce back in time	106
0.268	The one where they bounce back in time	79
0.268	The one where they bounce back in time	98
0.268	The one where they bounce back in time	152
0.267	The one where they bounce back in time	82
0.266	The one where they bounce back in time	69
0.266	The one where they bounce back in time	145
0.264	The one where they bounce back in time	146
0.264	The one where they bounce back in time	31
0.263	The one where they bounce back in time	71
0.263	The one where they bounce back in time	103
0.259	The one where they bounce back in time	68
0.259	The one where they bounce back in time	29
0.258	The one where they bounce back in time	148
0.257	The one where they bounce back in time	27
0.256	The one where they bounce back in time	123
0.256	The one where they bounce back in time	67
0.254	The one where they bounce back in time	126
0.254	The one where they bounce back in time	124
0.252	The one where they bounce back in time	26
0.252	The one where they bounce back in time	80
0.251	The one where they bounce back in time	54
0.251	The one where they bounce back in time	136
0.250	The one where they bounce back in time	147

0.249	The one where they bounce back in time	115
0.249	The one where they bounce back in time	125
0.248	The one where they bounce back in time	52
0.247	The one where they bounce back in time	28
0.247	The one where they bounce back in time	34
0.247	The one where they bounce back in time	108
0.246	The one where they bounce back in time	83
0.245	The one where they bounce back in time	35
0.243	The one where they bounce back in time	53
0.242	The one where they bounce back in time	47
0.240	The one where they bounce back in time	50
0.240	The one where they bounce back in time	116
0.240	The one where they bounce back in time	113
0.240	The one where they bounce back in time	42
0.239	The one where they bounce back in time	143
0.239	The one where they bounce back in time	99
0.239	The one where they bounce back in time	40
0.238	The one where they bounce back in time	57
0.235	The one where they bounce back in time	120
0.235	The one where they bounce back in time	78
0.233	The one where they bounce back in time	17
0.232	The one where they bounce back in time	37
0.231	The one where they bounce back in time	43
0.227	The one where they bounce back in time	51
0.224	The one where they bounce back in time	105
0.224	The one where they bounce back in time	85
0.223	The one where they bounce back in time	151
0.222	The one where they bounce back in time	104
0.221	The one where they bounce back in time	56
0.221	The one where they bounce back in time	19
0.219	The one where they bounce back in time	60
0.218	The one where they bounce back in time	76
0.217	The one where they bounce back in time	20
0.216	The one where they bounce back in time	140
0.216	The one where they bounce back in time	107
0.215	The one where they bounce back in time	16
0.212	The one where they bounce back in time	7
0.210	The one where they bounce back in time	3
0.210	The one where they bounce back in time	14
0.209	The one where they bounce back in time	110
0.208	The one where they bounce back in time	129
0.207	The one where they bounce back in time	94
0.206	The one where they bounce back in time	114
0.206	The one where they bounce back in time	63
0.205	The one where they bounce back in time	93
0.203	The one where they bounce back in time	101
0.196	The one where they bounce back in time	23
0.193	The one where they bounce back in time	45
0.192	The one where they bounce back in time	18
0.191	The one where they bounce back in time	142
0.190	The one where they bounce back in time	49
0.189	The one where they bounce back in time	48
0.187	The one where they bounce back in time	109
0.186	The one where they bounce back in time	39
0.182	The one where they bounce back in time	61
0.181	The one where they bounce back in time	64
0.179	The one where they bounce back in time	97
0.178	The one where they bounce back in time	133
0.171	The one where they bounce back in time	6
0.170	The one where they bounce back in time	134
0.167	The one where they bounce back in time	13

```
0.162 The one where they bounce back in time 2
0.143 The one where they bounce back in time 22
0.130 The one where they bounce back in time 135
0.124 The one where they bounce back in time 144
0.105 The one where they bounce back in time 38
```

Now let's try a sentence embedding model that was optimized for generating sentence representations: Sentence-BERT (Reimers and Gurevych 2019). Example usage (in the context of the Huggingface transformers library) can be found [here](#).

```
In [84]: sentence_model = SentenceTransformer('sentence-transformers/all-distilroberta-v1')
```

Q3: Fill out the `get_sentence_embedding` function below to return the sentence embedding for the input string, and use it again to find the episode that is most similar to the description "The one where they bounce back in time" by having the highest cosine similarity between representations. Which method for generating sentence embeddings appears better for this task?

```
In [105... def get_sentence_embedding(model, string):
                # your code goes here
                embeddings = sentence_model.encode(string)
                return embeddings
```

```
In [106... get_sentence_embedding(model, query)
```

```
Out[106... array([ 2.21291883e-03,  1.46007640e-02,  2.03430578e-02,  1.27137015e-02,
        1.35774920e-02,  3.19657773e-02, -7.94165395e-03,  5.24583161e-02,
        2.90296637e-02, -1.41363693e-02,  4.94237766e-02, -2.31884580e-04,
       -4.03048992e-02, -6.15077540e-02,  9.52694193e-03, -7.15107704e-03,
       -6.67671785e-02, -4.34239171e-02, -1.10399239e-02, -2.21695676e-02,
       -1.34791841e-03, -3.53775956e-02, -7.93617591e-02,  8.38638619e-02,
       -3.64556462e-02,  1.69570819e-02, -8.85298103e-02,  1.07683558e-02,
       -8.18700530e-03,  3.06761004e-02,  1.76261924e-02, -1.28581850e-02,
        2.35316176e-02,  6.67219236e-02,  5.61144166e-02,  3.27455346e-03,
       -4.85332571e-02, -9.01158492e-04,  6.27200492e-03, -5.37139736e-03,
        1.40093248e-02,  1.00574549e-03,  2.96798851e-02,  1.23534417e-02,
       -1.32677779e-02, -6.48900867e-03,  1.86410528e-02,  4.17390689e-02,
       -1.89139359e-02, -3.09142452e-02,  3.27500366e-02,  2.01369170e-02,
        3.56953824e-03, -3.77495885e-02,  1.06609203e-02,  5.49859665e-02,
       -1.41724269e-03, -1.52485780e-02, -1.37812691e-02,  1.48049816e-02,
        2.49714293e-02, -9.09428224e-02, -2.07985006e-03,  2.63621733e-02,
        3.11674410e-03,  2.14692149e-02, -1.30433282e-02, -7.28110597e-02,
        1.96592454e-02, -1.46719180e-02,  1.44420527e-02,  2.45864019e-02,
       -4.54400778e-02,  2.89853720e-04,  7.60145532e-03,  3.64723196e-03,
       -1.69990733e-02, -6.93664700e-02, -1.38491075e-02,  2.56028473e-02,
```

-3.50530520e-02, 7.26699308e-02, -4.05667834e-02, 1.06763688e-03,
4.28503491e-02, -6.79032430e-02, -1.14855748e-02, -1.70884803e-02,
9.46743507e-03, 2.67948955e-02, -9.88203473e-03, 4.40861844e-02,
5.08677177e-02, 3.89164500e-02, -3.79183963e-02, -9.36778076e-03,
-2.02575196e-02, 1.25979796e-01, -1.92892291e-02, 5.68098156e-03,
-4.03489619e-02, 5.32163270e-02, -5.44315018e-03, 5.55830076e-02,
-7.20151141e-03, 1.84596481e-03, -3.76855652e-03, 2.94402204e-02,
2.28764433e-02, 3.08500882e-02, -4.25108112e-02, 2.08726246e-02,
-3.65397800e-03, 5.86143807e-02, -1.33511070e-02, -1.16881272e-02,
2.20061373e-02, -2.04156861e-02, -3.02557778e-02, 3.20698917e-02,
2.08282471e-02, -1.23807220e-02, 4.24493626e-02, 3.32105085e-02,
7.04018101e-02, 9.06893332e-03, -4.06467430e-02, -1.51359066e-01,
-3.29957567e-02, -8.76825154e-02, 1.20234238e-02, -4.67043854e-02,
4.51379009e-02, 7.87090417e-03, -1.95024759e-02, 3.00189126e-02,
4.17987891e-02, 1.60717387e-02, 4.48036101e-03, 3.45836878e-02,
3.99757698e-02, 2.72923447e-02, -1.91768585e-03, 1.79524515e-02,
3.38673145e-02, -3.37349288e-02, 1.51583888e-02, 2.06432026e-02,
-2.84546819e-02, -5.17852344e-02, 8.12441483e-02, -1.81690678e-02,
2.23133969e-03, 4.92140017e-02, -5.06111374e-03, 1.92196853e-02,
-4.71705347e-02, 9.11280606e-03, 3.81658785e-02, 4.57479917e-02,
1.16539588e-02, -1.53999710e-02, -1.48039078e-02, 1.64523441e-02,
-3.35796946e-03, -4.39754650e-02, -3.30474451e-02, -8.44774942e-04,
6.27329051e-02, -2.32955087e-02, -2.20720116e-02, 4.58791927e-02,
5.94791099e-02, -5.52829504e-02, 3.45610492e-02, 6.18197396e-02,
-5.13229556e-02, 8.08027610e-02, 2.34802328e-02, 3.61699909e-02,
-4.68614744e-03, -2.40609725e-03, 7.38980100e-02, -5.79677196e-03,
-1.63228661e-02, 1.58600044e-02, -3.31827104e-02, -2.26014368e-02,
-4.26743664e-02, -2.03661229e-02, 4.15895320e-02, -3.00418194e-02,
-2.63435598e-02, -1.28771281e-02, -2.58608758e-02, -6.30370453e-02,
9.14851651e-02, -1.95028745e-02, 1.56084625e-02, -1.98379196e-02,
1.71796791e-02, 1.22316694e-02, 1.09745778e-01, -1.12986630e-02,
4.90177125e-02, -1.59105342e-02, -1.33282430e-02, 5.45077436e-02,
-1.35276113e-02, -1.49595123e-02, 1.20830687e-03, -4.54155961e-03,
-1.98911913e-02, -2.28482541e-02, -2.25817561e-02, 3.55753466e-03,
-2.80248076e-02, 8.04722607e-02, -1.69929937e-02, -1.18243219e-02,
7.42609054e-02, -3.13200988e-02, -2.71249246e-02, -3.02734300e-02,
-5.73949069e-02, 1.61248166e-02, -1.76638830e-02, 4.97527719e-02,
-1.86391233e-03, 3.01872119e-02, -5.42424023e-02, 5.21208532e-02,
3.47083695e-02, 8.34543444e-03, 2.89557818e-02, -4.73210029e-03,
1.00078307e-01, -2.02751048e-02, -6.53888285e-03, -1.58774480e-02,
-2.55581457e-02, -4.82113939e-03, 2.49504037e-02, -5.64011298e-02,
1.60764754e-02, -1.76009759e-02, -1.66858104e-03, 3.18602063e-02,
2.87729315e-02, 1.61234401e-02, 4.71981131e-02, 3.60711776e-02,
8.81029107e-03, -4.05638739e-02, 3.06993891e-02, -2.26861779e-02,
-2.86178105e-02, 1.46398861e-02, 9.10236873e-03, 6.47793263e-02,
-9.16156732e-03, 3.49191763e-02, -3.98512296e-02, 2.17506830e-02,
-3.80389877e-02, 2.51834132e-02, 2.37333756e-02, 1.13705471e-02,
-4.66702394e-02, -2.15412546e-02, -4.15026024e-03, -1.34915877e-02,
3.03957798e-02, 1.65170282e-02, -4.68902588e-02, 4.30448763e-02,
4.75014932e-03, -2.10781526e-02, -9.00529325e-03, -7.51913711e-03,
-3.50393308e-03, -3.08090951e-02, -1.25842728e-02, 1.97249874e-02,
-3.02418899e-02, 2.49120388e-02, 1.95122194e-02, -1.48757761e-02,
-8.45433958e-03, 2.07474772e-02, 4.29975204e-02, 3.06963157e-02,
-2.70412583e-02, 2.15281337e-03, -1.35459034e-02, -5.44551685e-02,
-2.00507045e-02, 7.40589714e-03, 4.53561923e-04, -3.56789082e-02,
-4.48385105e-02, -4.19514365e-02, 5.66280410e-02, 3.45417224e-02,
-3.31271850e-02, -5.91354333e-02, -2.94692256e-02, 1.45531455e-02,
-1.29651614e-02, -2.08167769e-02, -4.14324440e-02, -7.38277275e-04,
-8.37282650e-03, 5.14540374e-02, -7.97048733e-02, 5.63222857e-04,
1.38438931e-02, -8.54356389e-04, -1.02657499e-02, 7.20910877e-02,
2.64527444e-02, 4.63393927e-02, -5.44629944e-03, -2.24378761e-02,

-6.05147444e-02, -8.12144857e-03, 3.84929497e-03, 3.33278328e-02,
1.48666631e-02, 2.95839664e-02, -1.04039289e-01, -2.54411660e-02,
-3.08515970e-02, 2.64406819e-02, -2.93540582e-02, 6.08844534e-02,
1.52755696e-02, -2.47605890e-02, 3.36184545e-04, 3.19134742e-02,
-4.39078249e-02, 1.57307629e-02, 1.04721906e-02, -6.71743508e-03,
-3.31027294e-03, -6.52832910e-02, 2.69277804e-02, 1.78861199e-03,
4.97643016e-02, 4.67903614e-02, -3.02304495e-02, -2.36002225e-02,
-2.11849064e-03, -3.95075791e-03, -6.50295382e-03, -1.50645478e-02,
-2.40596390e-04, 1.52773233e-02, 3.91860958e-03, -2.34858394e-02,
1.75554713e-03, -1.51093747e-03, 2.17265170e-02, -3.52287851e-02,
1.72481798e-02, -5.95338494e-02, -4.09956351e-02, -6.63001230e-03,
7.63502419e-02, 1.01583241e-03, -7.55239604e-03, 1.41998539e-02,
-1.46027403e-02, 3.80894728e-02, 2.47641969e-02, -2.05197316e-02,
6.24312870e-02, -8.97597745e-02, 1.36640323e-02, -1.64970178e-02,
5.14223166e-02, -3.51534002e-02, -1.45202391e-02, 2.31130440e-02,
-2.37241816e-02, 1.81850139e-02, 5.46971336e-03, -4.16800603e-02,
9.78655517e-02, -1.81357097e-02, 2.49483678e-02, -2.88970657e-02,
-3.74293490e-03, 6.91632256e-02, 9.33621917e-03, 1.82613228e-02,
3.82086150e-02, 2.88215522e-02, 6.31359965e-02, -5.21778911e-02,
5.33160567e-02, -2.54744161e-02, 2.70235687e-02, 1.54799288e-02,
-4.76785526e-02, 5.17638624e-02, -1.59045160e-02, -1.64057184e-02,
3.79515179e-02, 7.65309529e-03, 9.12372023e-03, 6.43153489e-02,
1.27697969e-02, 5.90448491e-02, 1.05696190e-02, 4.63568280e-03,
-1.77777316e-02, -8.73347744e-03, 2.29637343e-02, 7.95334801e-02,
4.97696549e-02, 2.25280114e-02, 3.95001546e-02, 3.02420463e-03,
1.80288237e-02, 2.56689684e-03, -1.29529685e-02, -1.27854105e-03,
-7.42591918e-03, -2.24011764e-02, -2.48855017e-02, -8.55687484e-02,
-5.85518638e-03, -3.30785438e-02, 1.11911877e-03, -5.49122021e-02,
4.05883193e-02, 2.52610967e-02, 1.92099754e-02, -4.84749489e-02,
-1.82878878e-02, -3.54616344e-02, 5.12124151e-02, 4.21108939e-02,
-1.46670556e-02, -1.53476857e-02, 6.14334792e-02, 3.49272713e-02,
3.27575356e-02, 1.67912350e-03, -7.55056888e-02, -1.45413158e-02,
-1.32042579e-02, -5.17149568e-02, -2.84248814e-02, -2.16986239e-02,
-3.08782863e-03, 3.47218998e-02, 1.38619170e-03, 3.78282666e-02,
2.83214240e-03, -1.75256226e-02, -1.62327848e-02, 9.29554552e-03,
1.39869740e-02, -4.91586886e-03, -1.27024734e-02, 5.08511774e-02,
2.84793340e-02, 2.01693531e-02, -2.07730308e-02, -1.24611938e-02,
-4.24916372e-02, -4.03589010e-02, -3.96966795e-03, -1.12628322e-02,
-9.84559581e-02, -3.65020894e-02, 3.17138359e-02, 1.71383545e-02,
-1.38486652e-02, -1.92760099e-02, 1.92616116e-02, -1.44866873e-02,
-1.22793848e-02, 3.09865773e-02, -2.00249758e-02, 4.56785709e-02,
2.67754328e-02, -1.33322980e-02, 3.19032334e-02, 2.64700856e-02,
-1.30730476e-02, -3.52753839e-03, 2.17156510e-33, 1.33138550e-02,
-6.32691607e-02, -5.39007001e-02, 8.55311304e-02, -3.79473157e-02,
6.71645254e-03, -4.49737310e-02, 7.68038929e-02, -1.82718318e-02,
-1.45877227e-02, -4.57998272e-03, -4.13810369e-03, -4.84352894e-02,
-1.85312815e-02, 2.96622496e-02, 4.62089293e-02, -5.45271393e-03,
-4.27619554e-02, -2.26776628e-03, -7.45747704e-03, 3.03296335e-02,
7.30959419e-03, -3.76255140e-02, 2.93690450e-02, 1.10472729e-02,
3.24803330e-02, 2.54544634e-02, -8.85757431e-03, 1.32278968e-02,
-3.85051593e-02, -7.65438098e-03, 8.48209951e-03, 4.03504185e-02,
8.13582074e-03, 8.61408375e-03, -5.66714481e-02, -2.90653631e-02,
1.69131458e-02, -5.38208173e-04, 2.04201452e-02, -5.79386542e-04,
-2.68766545e-02, 3.19870259e-03, -2.54053008e-02, 3.18425372e-02,
-3.37387510e-02, -3.89852971e-02, -1.87378339e-02, -3.57849523e-02,
8.93797167e-03, -1.23559637e-02, 5.30179963e-02, -9.12249554e-03,
6.24284893e-03, -4.70225215e-02, 3.70354876e-02, -6.30526096e-02,
-3.80523615e-02, -1.43564194e-02, -8.69926662e-05, -1.60269008e-03,
4.75796163e-02, 3.00323078e-03, 5.53539507e-02, -3.72805223e-02,
-2.27514841e-02, 5.51960664e-03, -1.16593288e-02, 1.18770134e-02,
-5.34966253e-02, -3.12041733e-02, -2.53740009e-02, -1.18508544e-02,


```

3.17209847e-02, 2.41814051e-02, 1.74945872e-02, -5.58640435e-02,
-4.69565811e-03, -2.11711954e-02, 1.17525766e-02, -3.19689922e-02,
-1.41299926e-02, 1.79970302e-02, 5.25570028e-02, 3.70890764e-03,
3.91017720e-02, 3.32126804e-02, -3.53420153e-02, -6.92135915e-02,
4.45570238e-02, 3.84343080e-02, 4.53064851e-02, 3.14874351e-02,
-4.38765965e-05, 1.99763384e-02, -9.83823277e-03, -5.66110872e-02,
-5.35560818e-03, 2.28485130e-02, 2.36424543e-02, 2.74402928e-02,
-3.58455218e-02, -5.33464327e-02, 1.65300667e-02, 4.65191714e-03,
-7.90528953e-03, -9.17639732e-02, 6.47867518e-03, -6.85673431e-02,
-3.51829119e-02, -1.01479264e-02, 7.02062175e-02, 3.16270487e-03,
1.72238797e-02, -4.37075458e-02, -4.09999024e-03, 1.74359996e-02,
-2.13136040e-02, -8.76731984e-03, 3.55058610e-02, 5.91446571e-02,
-1.74465440e-02, 3.10921557e-02, -3.56930234e-02, 7.54021853e-02,
2.39817072e-02, -1.43353995e-02, -4.20640446e-02, 3.74372415e-02,
5.21604940e-02, -3.17335650e-02, 2.33216751e-02, 1.02511495e-01,
1.40924603e-02, 8.06194991e-02, 2.54230406e-02, 3.54625061e-02,
-3.01836245e-02, 1.40863471e-02, -6.63911998e-02, -6.05957769e-02,
-4.75138240e-03, 1.26640638e-03, 7.68455677e-03, -1.69133488e-02,
-6.21553808e-02, 6.56201541e-02, 5.58134727e-02, -7.45311156e-02,
1.21466238e-02, 1.41489552e-02, -6.20018654e-02, 9.46553797e-02,
2.27337517e-02, -2.48843320e-02, 1.13459947e-02, 8.44585337e-03,
-9.08971578e-02, -4.53343056e-02, 2.83331480e-02, 6.48867488e-02,
1.71370171e-02, 2.89198346e-02, -3.13544832e-02, -2.93983202e-02,
2.54119895e-02, -4.29376401e-02, -3.70732130e-04, -4.01982293e-02,
-2.23859493e-02, -1.43441539e-02, 3.32331322e-02, -1.36592612e-01,
2.10027490e-02, 3.45640369e-02, -2.24993401e-03, 2.95645036e-02,
-2.94183362e-02, 7.56334662e-02, 4.42030206e-02, -6.86599836e-02,
-3.15298028e-02, -5.26868254e-02, -3.67746279e-02, 2.23003961e-02,
-5.77544421e-02, 2.40107290e-02, 4.00874950e-02, -2.44222209e-03,
3.21263596e-02, -5.11016436e-02, -4.71268734e-03, 4.81284633e-02,
-2.35034097e-02, -1.31728360e-02, -1.04371291e-02, -1.75945032e-02,
1.85968529e-03, 2.49489211e-02, -3.04331002e-03, 2.60488000e-02,
2.57156733e-02, -1.70342624e-02, -6.66107088e-02, 1.63565725e-02,
3.75571032e-03, 2.32964684e-03, 1.29806595e-02, -2.93646497e-03,
8.54911841e-03, 9.20998398e-03, -1.05059324e-02, -1.95238553e-02,
-3.26132141e-02, 5.25490846e-03, -2.80559703e-04, 1.46071287e-02,
-4.10063751e-03, 1.46662667e-02, 3.24111283e-02, 1.15096429e-03,
2.27696523e-02, -5.82251325e-02, 2.30602436e-02, 9.00164992e-03,
-9.11707000e-04, 4.68654893e-02, 3.94302234e-03, -3.11784148e-02,
-3.03201471e-02, -6.22763997e-03, -1.36993630e-02, -7.15379640e-02,
-2.83833966e-02, -4.58457768e-02, 4.19994956e-03, 9.53249075e-03,
-4.10476513e-03, -2.92242263e-02, 3.37043102e-03, -5.71409427e-03,
-4.49105576e-02, 5.04756812e-03, -2.82579064e-02, -1.36668226e-02,
1.31686712e-02, -2.64742263e-02, 1.83658360e-03, -7.73750171e-02,
6.81264549e-02, 6.01057969e-02, -2.66325986e-03, 1.95708573e-02,
6.41155392e-02, -1.23819280e-02, 1.39256716e-02, 1.46918604e-02,
-2.29183193e-02, 7.19881104e-03, -1.75621156e-02, -1.85609125e-02,
-4.86978143e-02, -2.84883601e-04, -3.20848599e-02, 8.21653008e-03,
-8.27786922e-02, 4.78361845e-02, 6.14723284e-03, 2.46668793e-02,
-1.25666236e-04, -4.68023643e-02, 4.28869911e-02, 2.85773873e-02],
dtype=float32)

```

In [94]:

```

def get_document_reps_for_data(data, sequence_embedding_function, model):

    # This function applies the sequence_embedding_function argument (a function itself
# element in the input data list, and returns a copy of that list with an embedding

    data_with_reps=[]

    for episode, title, summary in data:

```

```

        data_with_reps.append((episode, title, summary, sequence_embedding_function(mod

return data_with_reps

```

```

In [107... sentence_transformer_data=get_document_reps_for_data_new(data, get_sentence_embedding,

```

```

In [97]: def get_document_reps_for_data_new(data, sequence_embedding_function, model):

    # This function applies the sequence_embedding_function argument (a function itself
    # element in the input data list, and returns a copy of that list with an embedding

    data_with_reps=[]

    for episode, title, summary in data:
        data_with_reps.append((episode, title, summary, sequence_embedding_function(sen

    return data_with_reps

```

```

In [109... query="The one where they bounce back in time"
run_query(query, sentence_transformer_data, get_sentence_embedding, sentence_model)

```

```

0.349 The one where they bounce back in time 91
0.341 The one where they bounce back in time 76
0.335 The one where they bounce back in time 143
0.334 The one where they bounce back in time 82
0.330 The one where they bounce back in time 47
0.325 The one where they bounce back in time 17
0.321 The one where they bounce back in time 83
0.313 The one where they bounce back in time 121
0.306 The one where they bounce back in time 123
0.304 The one where they bounce back in time 29
0.303 The one where they bounce back in time 84
0.293 The one where they bounce back in time 15
0.292 The one where they bounce back in time 130
0.290 The one where they bounce back in time 43
0.289 The one where they bounce back in time 98
0.285 The one where they bounce back in time 27
0.283 The one where they bounce back in time 13
0.281 The one where they bounce back in time 92
0.278 The one where they bounce back in time 59
0.277 The one where they bounce back in time 6
0.277 The one where they bounce back in time 99
0.277 The one where they bounce back in time 18
0.275 The one where they bounce back in time 104
0.274 The one where they bounce back in time 54
0.274 The one where they bounce back in time 74
0.273 The one where they bounce back in time 52
0.272 The one where they bounce back in time 120
0.270 The one where they bounce back in time 8
0.268 The one where they bounce back in time 22
0.267 The one where they bounce back in time 131
0.266 The one where they bounce back in time 61
0.265 The one where they bounce back in time 23
0.265 The one where they bounce back in time 68
0.264 The one where they bounce back in time 62
0.261 The one where they bounce back in time 89

```

0.261	The one where they bounce back in time	117
0.260	The one where they bounce back in time	81
0.260	The one where they bounce back in time	10
0.260	The one where they bounce back in time	133
0.258	The one where they bounce back in time	5
0.257	The one where they bounce back in time	57
0.257	The one where they bounce back in time	151
0.255	The one where they bounce back in time	67
0.254	The one where they bounce back in time	9
0.254	The one where they bounce back in time	140
0.248	The one where they bounce back in time	66
0.247	The one where they bounce back in time	55
0.247	The one where they bounce back in time	48
0.247	The one where they bounce back in time	73
0.244	The one where they bounce back in time	19
0.243	The one where they bounce back in time	58
0.243	The one where they bounce back in time	87
0.242	The one where they bounce back in time	124
0.242	The one where they bounce back in time	146
0.240	The one where they bounce back in time	103
0.240	The one where they bounce back in time	16
0.240	The one where they bounce back in time	32
0.240	The one where they bounce back in time	41
0.239	The one where they bounce back in time	37
0.237	The one where they bounce back in time	64
0.237	The one where they bounce back in time	80
0.237	The one where they bounce back in time	12
0.237	The one where they bounce back in time	79
0.236	The one where they bounce back in time	126
0.234	The one where they bounce back in time	105
0.234	The one where they bounce back in time	71
0.233	The one where they bounce back in time	21
0.232	The one where they bounce back in time	2
0.231	The one where they bounce back in time	38
0.229	The one where they bounce back in time	39
0.229	The one where they bounce back in time	56
0.224	The one where they bounce back in time	102
0.221	The one where they bounce back in time	112
0.220	The one where they bounce back in time	88
0.220	The one where they bounce back in time	90
0.219	The one where they bounce back in time	49
0.218	The one where they bounce back in time	97
0.217	The one where they bounce back in time	119
0.217	The one where they bounce back in time	69
0.217	The one where they bounce back in time	25
0.215	The one where they bounce back in time	144
0.214	The one where they bounce back in time	1
0.213	The one where they bounce back in time	110
0.210	The one where they bounce back in time	106
0.207	The one where they bounce back in time	31
0.207	The one where they bounce back in time	11
0.206	The one where they bounce back in time	65
0.206	The one where they bounce back in time	96
0.205	The one where they bounce back in time	116
0.204	The one where they bounce back in time	20
0.204	The one where they bounce back in time	50
0.204	The one where they bounce back in time	46
0.203	The one where they bounce back in time	107
0.202	The one where they bounce back in time	33
0.201	The one where they bounce back in time	109
0.197	The one where they bounce back in time	34

0.197	The one where they bounce back in time	86
0.197	The one where they bounce back in time	36
0.194	The one where they bounce back in time	100
0.192	The one where they bounce back in time	142
0.191	The one where they bounce back in time	134
0.188	The one where they bounce back in time	108
0.188	The one where they bounce back in time	115
0.187	The one where they bounce back in time	35
0.187	The one where they bounce back in time	114
0.186	The one where they bounce back in time	70
0.185	The one where they bounce back in time	135
0.185	The one where they bounce back in time	45
0.184	The one where they bounce back in time	141
0.181	The one where they bounce back in time	3
0.180	The one where they bounce back in time	127
0.179	The one where they bounce back in time	113
0.179	The one where they bounce back in time	101
0.177	The one where they bounce back in time	53
0.174	The one where they bounce back in time	148
0.172	The one where they bounce back in time	118
0.171	The one where they bounce back in time	28
0.171	The one where they bounce back in time	94
0.170	The one where they bounce back in time	44
0.170	The one where they bounce back in time	24
0.169	The one where they bounce back in time	26
0.168	The one where they bounce back in time	30
0.168	The one where they bounce back in time	145
0.167	The one where they bounce back in time	147
0.167	The one where they bounce back in time	95
0.157	The one where they bounce back in time	125
0.153	The one where they bounce back in time	7
0.152	The one where they bounce back in time	14
0.150	The one where they bounce back in time	129
0.149	The one where they bounce back in time	132
0.148	The one where they bounce back in time	78
0.143	The one where they bounce back in time	77
0.141	The one where they bounce back in time	4
0.140	The one where they bounce back in time	60
0.138	The one where they bounce back in time	111
0.137	The one where they bounce back in time	152
0.136	The one where they bounce back in time	85
0.131	The one where they bounce back in time	42
0.131	The one where they bounce back in time	139
0.128	The one where they bounce back in time	122
0.127	The one where they bounce back in time	93
0.120	The one where they bounce back in time	72
0.115	The one where they bounce back in time	138
0.112	The one where they bounce back in time	137
0.107	The one where they bounce back in time	128
0.103	The one where they bounce back in time	149
0.095	The one where they bounce back in time	150
0.089	The one where they bounce back in time	51
0.079	The one where they bounce back in time	63
0.077	The one where they bounce back in time	40
0.050	The one where they bounce back in time	136
0.035	The one where they bounce back in time	75

In []: