

1. Imports, tokenization, and model definition

This appendix contains the inference script used to load the trained RNN/LSTM model and predict topic and ideology labels for a single news article.

How to Run Inference block

```
[ ]: !python inference.py \
      --checkpoint "outputs/best_lstm.pt" \
      --text_file "assignment3-1_train_unzipped/assignment3-1_train/1.txt"
```

The following code implements the inference pipeline for the trained RNN/LSTM model. The script takes the path to a single text file as input and outputs two integers:

```
inference.py

[13]: import argparse
      import re
      import torch
      import torch.nn as nn

      def simple_tokenize(text: str):
          return re.findall(r"[A-Za-z']+", text.lower())

      Model (must match train.py)

[14]: class SeqClassifier(nn.Module):
      def __init__(self, vocab_size, embed_dim, hidden_dim, num_layers, dropout, model_type, bidirectional, pad_idx=0):
          super().__init__()
          self.embedding = nn.Embedding(vocab_size, embed_dim, padding_idx=pad_idx)

          rnn_cls = {"rnn": nn.RNN, "lstm": nn.LSTM, "gru": nn.GRU}[model_type.lower()]
          self.rnn = rnn_cls(
              embed_dim,
              hidden_dim,
              num_layers=num_layers,
              batch_first=True,
              dropout=dropout if num_layers > 1 else 0.0,
              bidirectional=bidirectional,
          )

          out_dim = hidden_dim * (2 if bidirectional else 1)
          self.drop = nn.Dropout(dropout)
          self.topic_head = nn.Linear(out_dim, 5)
          self.ideology_head = nn.Linear(out_dim, 4)
          self.model_type = model_type.lower()

      def forward(self, x):
          emb = self.embedding(x)
          _, h = self.rnn(emb)

          if self.model_type == "lstm":
              h_n, _ = h
          else:
              h_n = h

          last = self.drop(h_n[-1])
          return self.topic_head(last), self.ideology_head(last)
```

2. Encoding and main inference routine

Encode helper

```
15]: def encode(text, vocab, max_len):
    tokens = simple_tokenize(text)
    ids = [vocab.get(t, vocab["<UNK>"]) for t in tokens[:max_len]]
    if len(ids) < max_len:
        ids += [vocab["<PAD>"]] * (max_len - len(ids))
    return torch.tensor(ids, dtype=torch.long).unsqueeze(0)

17]: !python inference.py --checkpoint "outputs/best_lstm.pt" --text_file "assignment3-1_train_unzipped/assignment3-1_train/1.txt"

5 0
```

```
18]: !python inference.py \
    --checkpoint "outputs/best_lstm.pt" \
    --text_file "assignment3-1_train_unzipped/assignment3-1_train/1.txt"

[ ]: def main():
    ap = argparse.ArgumentParser()
    ap.add_argument("--checkpoint", required=True)
    ap.add_argument("--text_file", required=True)
    args = ap.parse_args()

    device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
    ckpt = torch.load(args.checkpoint, map_location=device)

    vocab = ckpt["vocab"]
    max_len = ckpt["max_len"]

    model = SeqClassifier(
        vocab_size=len(vocab),
        embed_dim=ckpt["embed_dim"],
        hidden_dim=ckpt["hidden_dim"],
        num_layers=ckpt["num_layers"],
        dropout=ckpt["dropout"],
        model_type=ckpt["model_type"],
        bidirectional=ckpt["bidirectional"],
        pad_idx=vocab["<PAD>"],
    ).to(device)

    model.load_state_dict(ckpt["model_state"])
    model.eval()

    with open(args.text_file, "r", encoding="utf-8", errors="ignore") as f:
        text = f.read()

    x = encode(text, vocab, max_len).to(device)

    with torch.no_grad():
        topic_logits, ideo_logits = model(x)

    topic_pred = int(torch.argmax(topic_logits, dim=1).item()) + 1 # 1..5

    idx_to_ideology = {0: -1, 1: 0, 2: 1, 3: 99}
    ideo_pred = idx_to_ideology[int(torch.argmax(ideo_logits, dim=1).item())]

    # Print exactly two integers
    print(topic_pred, ideo_pred)

if __name__ == "__main__":
    main()
```

Output Description:

The inference script prints two predicted labels:

- Topic label $\in \{1: \text{Politics}, 2: \text{Entertainment}, 3: \text{Sports}, 4: \text{Technology}, 5: \text{Economics}\}$
- Ideology label $\in \{-1: \text{Left}, 0: \text{Neutral}, 1: \text{Right}, 99: \text{Not Political}\}$