

Implementation_2

June 9, 2025

0.1 Challenge: A deep dive into Dialogue Act Recognition - PART 2/2

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```
[17]: %load_ext autoreload
      %autoreload 2

      import sys
      import os

      # Assuming your notebook is in a subfolder, and you want to go up 1 or more
      ↪ levels to project root
      project_root = os.path.dirname(os.getcwd())

      # Optional: add root to Python path for module imports
      if project_root not in sys.path:
          sys.path.append(project_root)

      import torch
      import json
      import torch.utils.data
      from tqdm import tqdm
      import torch.nn as nn
      import numpy as np
      from torch.utils.data import Dataset, DataLoader
      from precompute_embeddings import run_embedding_extraction
      from collections import Counter
      import re
      import csv
      import time
      from models import ContextNet
      from utils import compute_avg_training_time
      from visualisations import plot_curve, generate_and_plot_confusion_matrix, \
      plot_top_symmetric_misclassifications, plot_tsne, plot_tsne_zoomed
      from train_structures import run_training_loop, evaluate

      # Set random seed for reproducibility
      seed = 0
```

```

np.random.seed(seed)
torch.manual_seed(seed)

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")

```

The autoreload extension is already loaded. To reload it, use:

```
%reload_ext autoreload
```

```

[5]: parent_dir = os.path.dirname(os.getcwd())
path_to_dataset = os.path.join(parent_dir, 'SGD Dataset') # Assuming we're in a
↳ subfolder of the root
path_to_train = os.path.join(path_to_dataset, "TrainSet.json")
path_to_test = os.path.join(path_to_dataset, "TestSet.json")

```

0.1.1 Implementation 2: Paper - “A Context-based Approach for Dialogue Act Recognition using Simple Recurrent Neural Networks”

The first implementation used embeddings of words, sequentially fed into an RNN, to build context-aware representations, based on a single utterance.

This implementation is inspired by the approach proposed in Bothe et al. (2018), *A Context-Based Approach for Dialogue Act Recognition Using Simple Recurrent Neural Networks*, presented at the Eleventh International Conference on Language Resources and Evaluation (LREC 2018).

It has two major upgrades compared to the first implementation: - First: it makes use of utterance-level embeddings, acquired from a model pretrained on a large corpus (~ 80 million amazon product reviews). - Second: the final model stacks these utterance-level embeddings across dialogue turns to model dialogue context.

As a result, the model benefits from semantically richer representations while also having the context of multiple turns in the dialogue.

Preprocessing The preprocessing is the same as implementation 1, we just want to get the act_labels and their corresponding number labels. The rest of the preprocessing is done in the dataset class itself.

```

[6]: EOD = "<eod>" # Define End of dialogue token as a global variable

```

```

[7]: def extract_turn_data(file_path):
    all_data = []
    with open(file_path, 'r') as f:
        data = json.load(f)
        for dialogue in data:
            for turn in dialogue['turns']:
                all_data.append({
                    'speaker': turn['speaker'],
                    'utterance': turn['utterance'],
                    'act': turn["dialogue_act"]
                })

```

```

        all_data[-1]['utterance'] += f" {EOD}" # Add an End Of Dialogue
        ↪marker at the end of last utterance in turn
    return all_data

```

```

extracted_train = extract_turn_data(path_to_train)
extracted_test = extract_turn_data(path_to_test)

```

```

[8]: # A set makes it easy to find unique elements (removes duplicates)
acts = list(set([data['act'] for data in extracted_train] + [data['act'] for
        ↪data in extracted_test]))

acts_labels = {act : i for i, act in enumerate(acts)}
inv_acts_labels = {v: k for k, v in acts_labels.items()}

```

We're only interested in the raw utterances of the test data, since it's just for visualisation purposes.

Code for precomputing an embedding vector for each utterance This code will create a `train_embeddings_[average|last].pt` `test_embeddings_[average|last].pt` depending on whether you want the mLSTM with pretrained weights to use the embedding from the last hidden state `h_t` or from the average of all hidden states. These embeddings can then more easily be acquired during the training of our model.

The paper: **A context-based Approach for DA Recognition using Simple RNN** reads: “*This model consists of a single multiplicative long-short-term memory (mLSTM) network (Krause et al., 2016) layer with 4,096 hidden units. The mLSTM is composed of an LSTM and a multiplicative RNN and considers each possible input in a recurrent transition function. It is trained as a character language model on 80 million Amazon product reviews (Radford et al., 2017).*”

The original pretrained mLSTM can be found [here](#) and can be used as is, as a feature extractor. However, since that codebase uses TensorFlow, the one used in this notebook is the one cited as a [pytorch fork](#) on that repository.

Running the following code cell will result in the creation of `test_embeddings_average.pt` and `train_embeddings_average.pt` files that contain a list of dialogues, where each dialogue contains an embedding of every utterance that occurs in the dialogue. For the actual implementation, the reader is referred to the code in `precompute_embeddings.py`.

```

[9]: run_embedding_extraction(model_path='mlstm_ns.pt', path_to_train=path_to_train,
        ↪path_to_test=path_to_test)

```

Embeddings already exist. Skipping extraction.

Dataset In the dataset some extra preprocessing is done, we want to collect the previous `n` utterances as context, so we make every sample a list of `n` utterances, with their corresponding labels. Instead of using raw utterances however, we use the embedding version acquired from the pretrained mLSTM, these were precomputed in the previous step.

```

[10]: class DialogueWindowDataset(Dataset):
        def __init__(self, embeddings_path, file_path, window_size=2):

```

```

    """
    Args:
        embeddings_path (str): Path to the precomputed embeddings file.
        file_path (str): Path to the JSON file containing dialogue data.
        window_size (int): Size of the context window. (e.g. 2 is 2
↪previous turns + current turn)
    """
    self.embeddings = torch.load(embeddings_path)
    self.window_size = window_size

    with open(file_path, 'r') as f:
        file_contents = json.load(f)
        dialogues = [dialogue['turns'] for dialogue in file_contents]
        self.data = []
        for dialog_idx, dialog in enumerate(tqdm(dialogues)):
            embeddings = self.embeddings[dialog_idx]
            speakers = [0 if turn['speaker'] == 'USER' else 1 for turn in
↪dialog]

            labels = [acts_labels[turn['dialogue_act']] for turn in dialog]

            n = len(embeddings)
            for t in range(n):
                window_emb = []
                window_speaker = []
                for i in range(t - self.window_size, t + 1):
                    if i >= 0:
                        emb = embeddings[i]
                        spk = torch.nn.functional.one_hot(
                            torch.tensor(speakers[i]), num_classes=2
                        ).float()
                    else:
                        emb = torch.zeros(4096)
                        spk = torch.zeros(2)
                window_emb.append(emb)
                window_speaker.append(spk)

            window_emb = torch.stack(window_emb) # (window_size + 1,
↪embed_dim)
            window_speaker = torch.stack(window_speaker) #
↪(window_size + 1, 2)
            combined = torch.cat([window_emb, window_speaker], dim=1)
↪# (window_size + 1, embed_dim + 2)
            label = labels[t]

            self.data.append((combined, label))

    def __len__(self):

```

```

        return len(self.data)

    def __getitem__(self, idx):
        return self.data[idx]

```

The result is a dataset where each batch contains n embeddings per sample (except at the start of a dialogue where there is no “previous utterance” to speak of).

```

[11]: # Define the dataset and dataloader
# In accordance with the paper, the average embedding is used for the mLSTM_
      ↪ model, as it contains more information
train_dataset = DialogueWindowDataset(embeddings_path='train_embeddings_average.
      ↪ pt', file_path=path_to_train)
train_loader = DataLoader(train_dataset, batch_size=128, shuffle=True)
test_dataset = DialogueWindowDataset(embeddings_path='test_embeddings_average.
      ↪ pt', file_path=path_to_test)
test_loader = DataLoader(test_dataset, batch_size=128, shuffle=False)

# Example batch
for batch in train_loader:
    combined, labels = batch
    print(combined.shape) # (batch_size, 3, embed_dim + 2)
    print(labels.shape)   # (batch_size,)
    break

```

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```

    self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:12<00:00, 436.69it/s]
100%|      | 1331/1331 [00:03<00:00, 409.82it/s]

torch.Size([128, 3, 4098])
torch.Size([128])

```

Architecture The final model’s input size is based on the embedding size returned by our pre-trained mLSTM model, which is 4096. The hidden size is kept the same as our previous model.

```
[12]: # Hyperparameters
input_size = 4096 + 2 # Embedding dimension + speaker one-hot
hidden_size = 32
num_classes = len(acts_labels) # Number of dialogue acts
pooling_type = 'average' # 'last_hidden_state' or 'average'

# Initialize model, optimizer, and criterion
model = ContextNet(input_size, hidden_size, num_classes, pooling_type).
    ↪to(device)
```

Training The training, evaluation and visualisation section remains similar to the first implementation, with small adaptations to account for the different input (speaker info is already added in the dataset creation).

```
[13]: def prepare_batch_v2(batch):
        combined, labels = batch
        return [combined], labels

def forward_v2(model, combined):
    return model(combined)
```

```
[14]: optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
criterion = nn.CrossEntropyLoss()
num_epochs = 50
```

```
[ ]: losses, accuracies = run_training_loop(
    model=model,
    train_loader=train_loader,
    optimizer=optimizer,
    criterion=criterion,
    num_epochs=num_epochs,
    prepare_batch=prepare_batch_v2,
    forward=forward_v2,
    save_dir="train_logs"
)
```

```
100%|      | 646/646 [00:05<00:00, 119.86it/s]
```

```
Epoch 1/50, Train Loss: 1.640, Train Accuracy: 57.985
```

```
100%|      | 646/646 [00:05<00:00, 126.50it/s]
```

```
Epoch 2/50, Train Loss: 0.589, Train Accuracy: 86.195
```

```
100%|      | 646/646 [00:05<00:00, 122.87it/s]
```

```
Epoch 3/50, Train Loss: 0.362, Train Accuracy: 92.334
```

```
100%|      | 646/646 [00:05<00:00, 126.33it/s]
```

```
Epoch 4/50, Train Loss: 0.248, Train Accuracy: 94.950
```

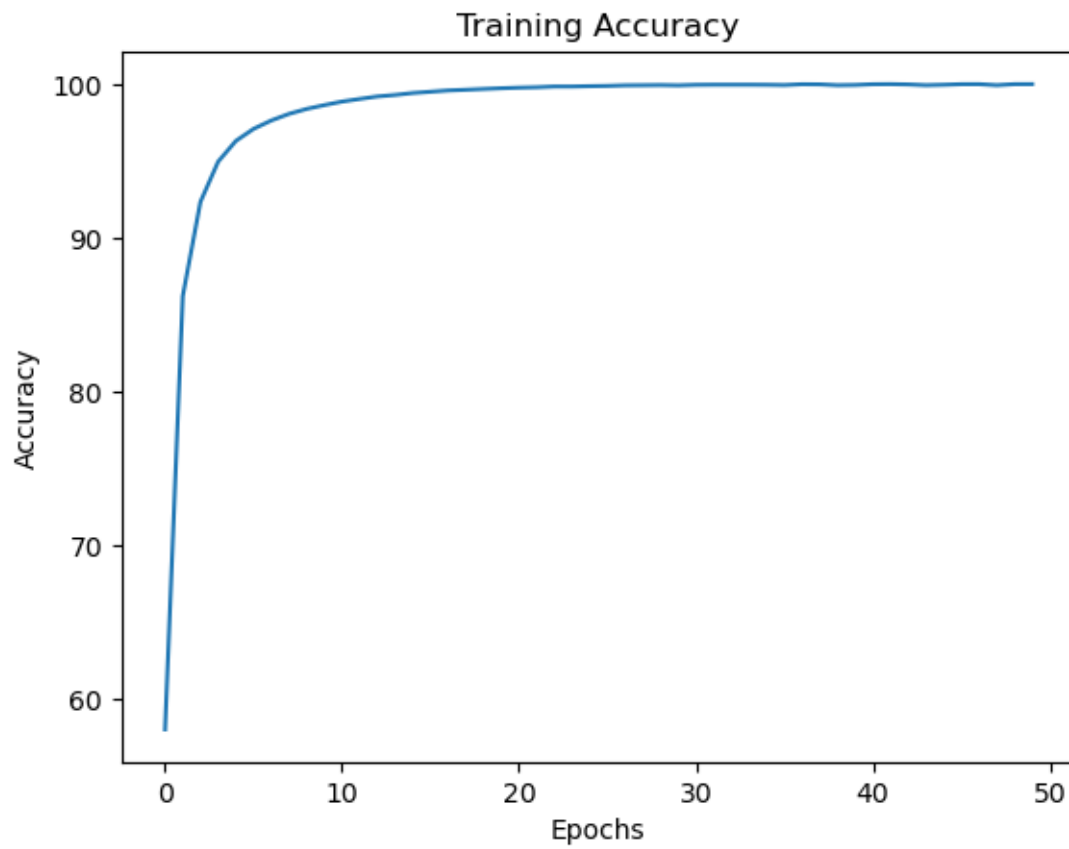
100%| | 646/646 [00:05<00:00, 125.00it/s]
Epoch 5/50, Train Loss: 0.182, Train Accuracy: 96.296
100%| | 646/646 [00:05<00:00, 118.53it/s]
Epoch 6/50, Train Loss: 0.141, Train Accuracy: 97.082
100%| | 646/646 [00:05<00:00, 122.37it/s]
Epoch 7/50, Train Loss: 0.112, Train Accuracy: 97.627
100%| | 646/646 [00:05<00:00, 123.75it/s]
Epoch 8/50, Train Loss: 0.092, Train Accuracy: 98.057
100%| | 646/646 [00:05<00:00, 121.70it/s]
Epoch 9/50, Train Loss: 0.076, Train Accuracy: 98.380
100%| | 646/646 [00:05<00:00, 115.44it/s]
Epoch 10/50, Train Loss: 0.065, Train Accuracy: 98.629
100%| | 646/646 [00:05<00:00, 121.43it/s]
Epoch 11/50, Train Loss: 0.055, Train Accuracy: 98.859
100%| | 646/646 [00:05<00:00, 115.18it/s]
Epoch 12/50, Train Loss: 0.047, Train Accuracy: 99.030
100%| | 646/646 [00:05<00:00, 119.76it/s]
Epoch 13/50, Train Loss: 0.040, Train Accuracy: 99.192
100%| | 646/646 [00:05<00:00, 120.36it/s]
Epoch 14/50, Train Loss: 0.035, Train Accuracy: 99.289
100%| | 646/646 [00:05<00:00, 120.05it/s]
Epoch 15/50, Train Loss: 0.031, Train Accuracy: 99.416
100%| | 646/646 [00:05<00:00, 123.49it/s]
Epoch 16/50, Train Loss: 0.027, Train Accuracy: 99.499
100%| | 646/646 [00:05<00:00, 117.47it/s]
Epoch 17/50, Train Loss: 0.024, Train Accuracy: 99.583
100%| | 646/646 [00:05<00:00, 120.68it/s]
Epoch 18/50, Train Loss: 0.021, Train Accuracy: 99.628
100%| | 646/646 [00:05<00:00, 113.98it/s]
Epoch 19/50, Train Loss: 0.018, Train Accuracy: 99.679
100%| | 646/646 [00:05<00:00, 119.60it/s]
Epoch 20/50, Train Loss: 0.016, Train Accuracy: 99.728

100%| | 646/646 [00:05<00:00, 117.50it/s]
Epoch 21/50, Train Loss: 0.014, Train Accuracy: 99.765
100%| | 646/646 [00:05<00:00, 116.06it/s]
Epoch 22/50, Train Loss: 0.013, Train Accuracy: 99.792
100%| | 646/646 [00:05<00:00, 122.35it/s]
Epoch 23/50, Train Loss: 0.011, Train Accuracy: 99.841
100%| | 646/646 [00:05<00:00, 112.67it/s]
Epoch 24/50, Train Loss: 0.010, Train Accuracy: 99.841
100%| | 646/646 [00:05<00:00, 121.27it/s]
Epoch 25/50, Train Loss: 0.009, Train Accuracy: 99.864
100%| | 646/646 [00:05<00:00, 127.50it/s]
Epoch 26/50, Train Loss: 0.007, Train Accuracy: 99.881
100%| | 646/646 [00:05<00:00, 129.03it/s]
Epoch 27/50, Train Loss: 0.007, Train Accuracy: 99.916
100%| | 646/646 [00:05<00:00, 128.62it/s]
Epoch 28/50, Train Loss: 0.006, Train Accuracy: 99.924
100%| | 646/646 [00:05<00:00, 128.63it/s]
Epoch 29/50, Train Loss: 0.005, Train Accuracy: 99.935
100%| | 646/646 [00:04<00:00, 129.50it/s]
Epoch 30/50, Train Loss: 0.005, Train Accuracy: 99.915
100%| | 646/646 [00:05<00:00, 127.06it/s]
Epoch 31/50, Train Loss: 0.004, Train Accuracy: 99.949
100%| | 646/646 [00:05<00:00, 128.08it/s]
Epoch 32/50, Train Loss: 0.004, Train Accuracy: 99.958
100%| | 646/646 [00:05<00:00, 127.89it/s]
Epoch 33/50, Train Loss: 0.004, Train Accuracy: 99.958
100%| | 646/646 [00:05<00:00, 125.89it/s]
Epoch 34/50, Train Loss: 0.003, Train Accuracy: 99.958
100%| | 646/646 [00:05<00:00, 123.95it/s]
Epoch 35/50, Train Loss: 0.004, Train Accuracy: 99.952
100%| | 646/646 [00:04<00:00, 130.66it/s]
Epoch 36/50, Train Loss: 0.004, Train Accuracy: 99.935

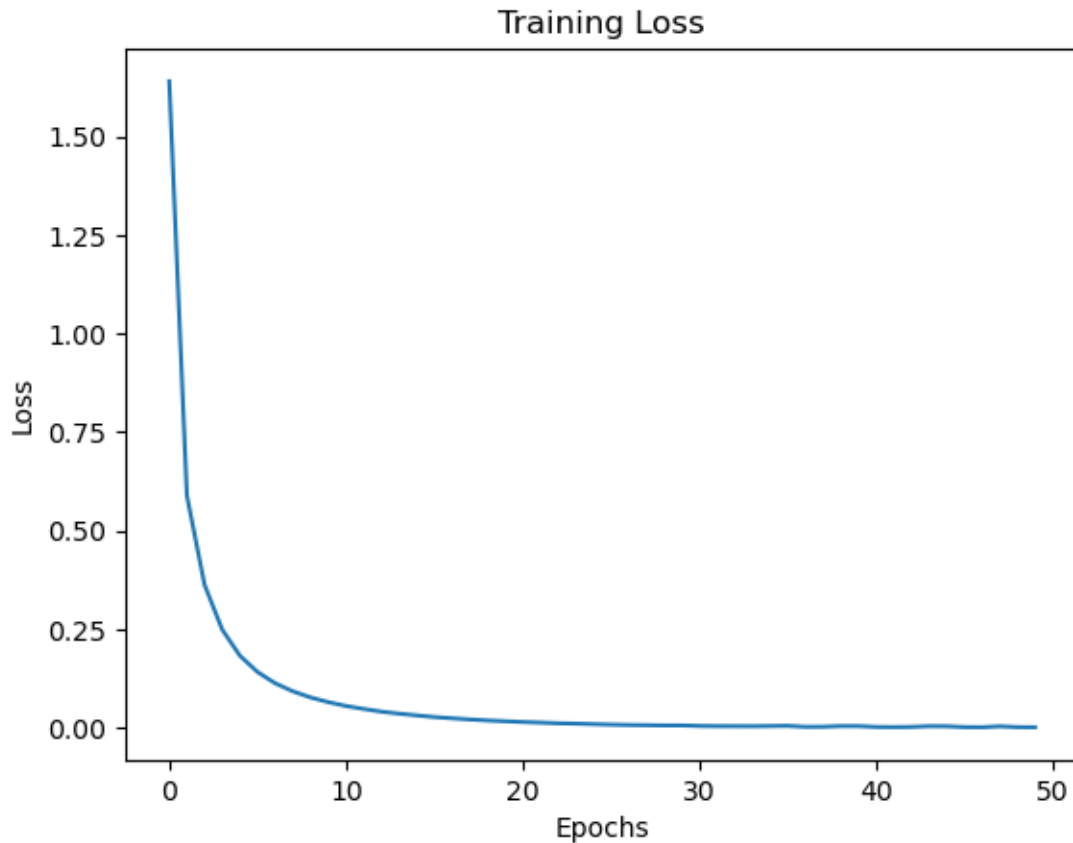
100%| | 646/646 [00:04<00:00, 129.27it/s]
Epoch 37/50, Train Loss: 0.002, Train Accuracy: 99.988
100%| | 646/646 [00:04<00:00, 129.29it/s]
Epoch 38/50, Train Loss: 0.002, Train Accuracy: 99.979
100%| | 646/646 [00:05<00:00, 128.92it/s]
Epoch 39/50, Train Loss: 0.004, Train Accuracy: 99.925
100%| | 646/646 [00:05<00:00, 128.67it/s]
Epoch 40/50, Train Loss: 0.004, Train Accuracy: 99.941
100%| | 646/646 [00:04<00:00, 129.48it/s]
Epoch 41/50, Train Loss: 0.002, Train Accuracy: 99.989
100%| | 646/646 [00:05<00:00, 129.11it/s]
Epoch 42/50, Train Loss: 0.001, Train Accuracy: 99.996
100%| | 646/646 [00:05<00:00, 128.19it/s]
Epoch 43/50, Train Loss: 0.002, Train Accuracy: 99.975
100%| | 646/646 [00:05<00:00, 127.87it/s]
Epoch 44/50, Train Loss: 0.003, Train Accuracy: 99.926
100%| | 646/646 [00:05<00:00, 128.12it/s]
Epoch 45/50, Train Loss: 0.003, Train Accuracy: 99.953
100%| | 646/646 [00:05<00:00, 122.35it/s]
Epoch 46/50, Train Loss: 0.001, Train Accuracy: 99.992
100%| | 646/646 [00:05<00:00, 121.52it/s]
Epoch 47/50, Train Loss: 0.001, Train Accuracy: 99.994
100%| | 646/646 [00:05<00:00, 128.57it/s]
Epoch 48/50, Train Loss: 0.003, Train Accuracy: 99.926
100%| | 646/646 [00:05<00:00, 128.91it/s]
Epoch 49/50, Train Loss: 0.001, Train Accuracy: 99.994
100%| | 646/646 [00:05<00:00, 128.65it/s]
Epoch 50/50, Train Loss: 0.001, Train Accuracy: 99.996

Training complete. Model saved to 'train_logs\model.pth', stats saved to
'train_logs\train_stats.csv'

```
[54]: plot_curve(accuracies, "Epochs", "Accuracy", "Training Accuracy")
```



```
[55]: plot_curve(losses, "Epochs", "Loss", "Training Loss")
```



Evaluation

```
[72]: # Evaluate on the test set
metrics = evaluate(model, test_loader, prepare_batch_fn=prepare_batch_v2,
↳forward_fn=forward_v2, inv_act_labels=inv_acts_labels)
```

100%| | 132/132 [00:05<00:00, 23.21it/s]

c:\Users\Ward\anaconda3\envs\multibench\Lib\site-packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains classes not in y_true
 warnings.warn("y_pred contains classes not in y_true")

Evaluation Results:
 Top-1 Accuracy: 94.49%
 Top-5 Accuracy: 99.49%
 Balanced Accuracy: 85.61%

Classification Report:

	precision	recall	f1-score
support			

1113	SYSTEM_CONFIRM	0.98	0.95	0.96
1331	SYSTEM_GOODBYE	0.98	0.98	0.98
719	SYSTEM_INFORM	0.99	0.95	0.97
310	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.99	0.96	0.98
17	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.79	0.88	0.83
79	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.99	0.92	0.95
499	SYSTEM_NOTIFY_SUCCESS	0.97	0.99	0.98
806	SYSTEM_OFFER	0.91	0.97	0.94
383	SYSTEM_OFFER_INTENT	0.92	0.98	0.95
608	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.97	0.96	0.97
14	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.93	0.96
1961	SYSTEM_REQUEST	1.00	0.99	0.99
585	SYSTEM_REQ_MORE	0.96	0.98	0.97
572	USER_AFFIRM	0.98	0.97	0.97
139	USER_AFFIRM_INTENT	0.90	0.93	0.91
103	USER_AFFIRM_INTENT USER_INFORM	0.89	0.85	0.87
1961	USER_INFORM	0.99	0.98	0.98
658	USER_INFORM_INTENT	0.96	0.93	0.94
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.81	0.90	0.85
12	USER_INFORM_INTENT USER_THANK_YOU	0.44	0.33	0.38
772	USER_INFORM USER_INFORM_INTENT	0.94	0.95	0.95
0	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.84	0.84	0.84

14	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.60	0.21	0.32
216	USER_INFORM USER_NEGATE	0.95	0.90	0.93
184	USER_INFORM USER_REQUEST_ALTS	0.84	0.86	0.85
7	USER_NEGATE	0.75	0.43	0.55
69	USER_NEGATE_INTENT	0.60	0.72	0.66
72	USER_NEGATE_INTENT USER_GOODBYE	0.71	0.76	0.73
9	USER_NEGATE USER_GOODBYE	0.60	0.67	0.63
587	USER_NEGATE USER_THANK_YOU	0.99	0.99	0.99
719	USER_REQUEST	0.97	0.94	0.95
152	USER_REQUEST_ALTS	0.91	0.93	0.92
340	USER_REQUEST USER_AFFIRM	0.95	0.96	0.95
515	USER_SELECT	0.85	0.89	0.87
279	USER_SELECT USER_GOODBYE	0.79	0.68	0.73
377	USER_THANK_YOU	0.80	0.77	0.79
384	USER_THANK_YOU USER_GOODBYE	0.71	0.83	0.77
16850	accuracy			0.94
16850	macro avg	0.83	0.81	0.81
16850	weighted avg	0.95	0.94	0.95

```
[2]: compute_avg_training_time('train_logs/train_stats.csv')
```

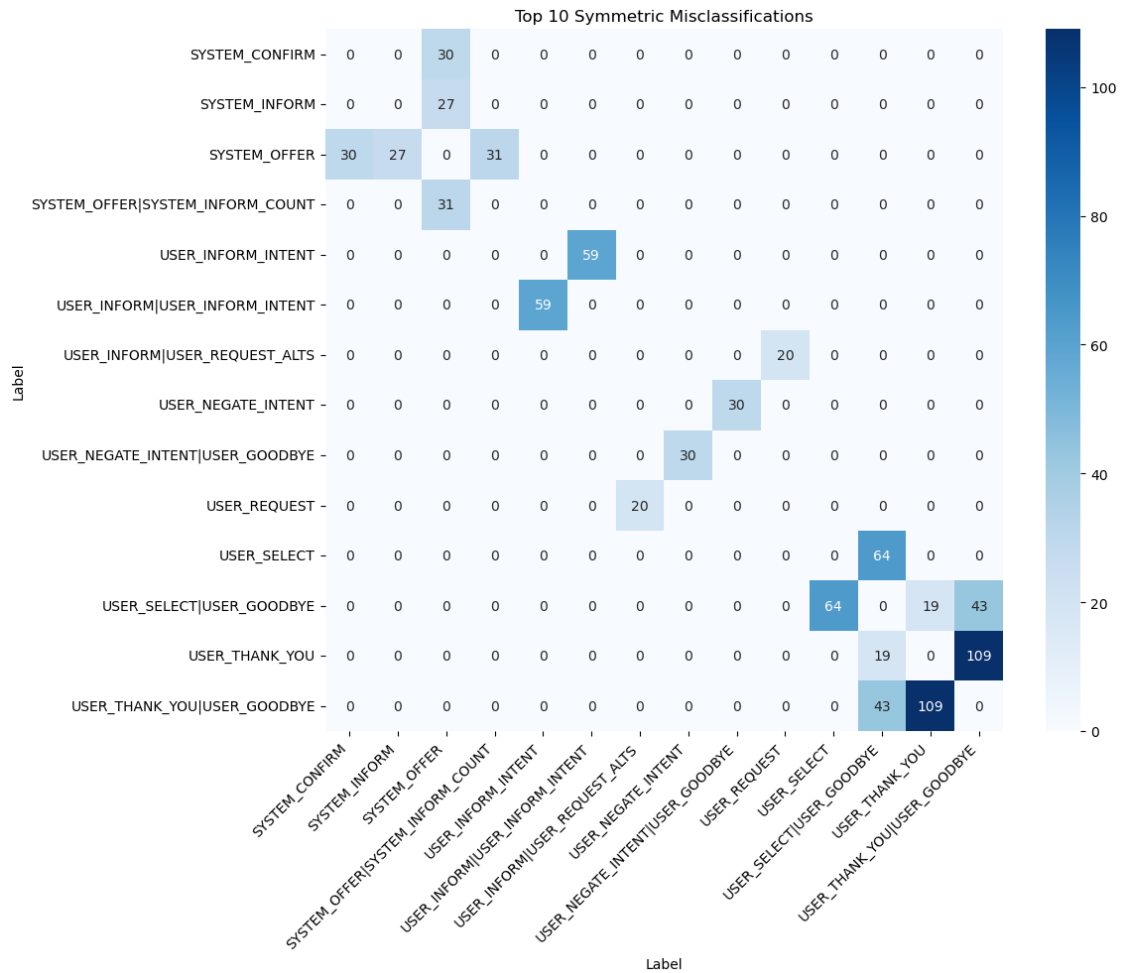
Average training time is: 5.2202076244354245 seconds

Visualisations

```
[61]: all_targets_named = metrics['y_true']
      all_predictions_named = metrics['y_pred']

      # Call the function to plot the top 10 misclassifications
```

```
plot_top_symmetric_misclassifications(all_targets_named, all_predictions_named,
↪top_n=10)
```

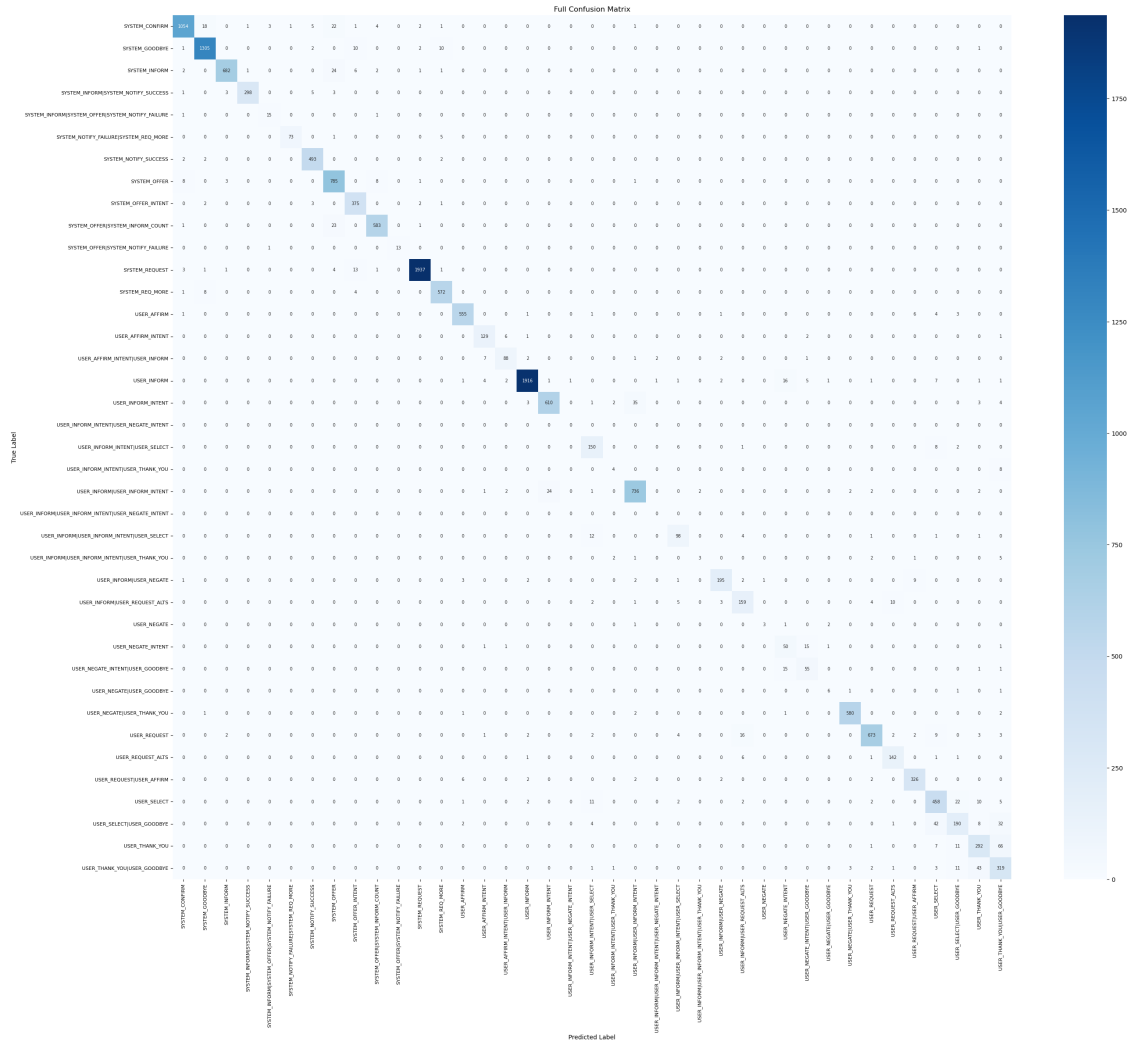


```
[58]: generate_and_plot_confusion_matrix(
    all_targets_named,
    all_predictions_named,
    figsize=(28, 24),           # Customize plot size
    annot_kws={"size": 7}      # Customize annotation font size
)
```

Found 39 unique classes.

Generating confusion matrix plot...

Saved matrix plots to full_confusion_matrix.pdf and .png

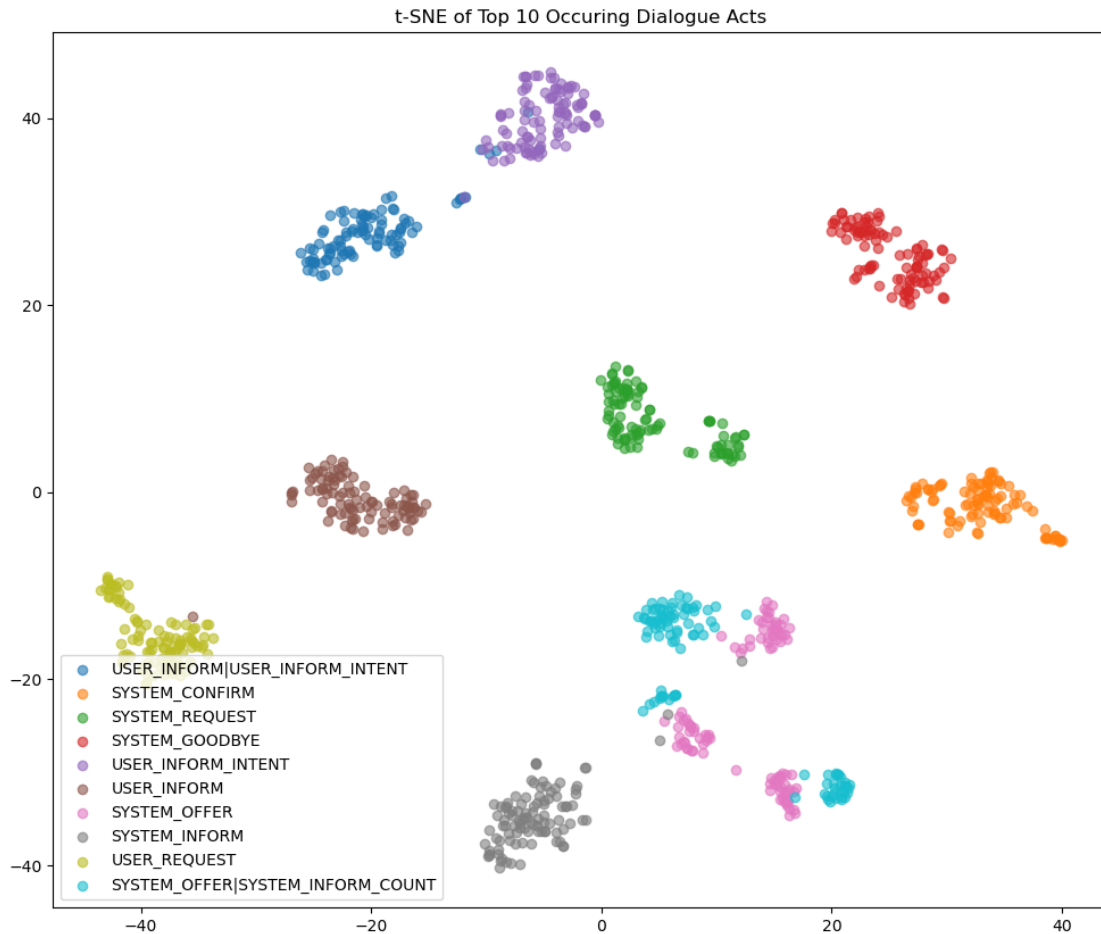


```
[73]: utterances = metrics["utterance_vectors"]
      labels = metrics["y_true"]

      # Filter the utterances and labels to only include the top labels
      label_counts = Counter(metrics["y_true"])
      top_labels = [label for label, _ in label_counts.most_common(10)]

      plot_tsne(utterances, labels, top_labels, max_per_label=100, title="t-SNE of
      ↳Top 10 Occuring Dialogue Acts")
```

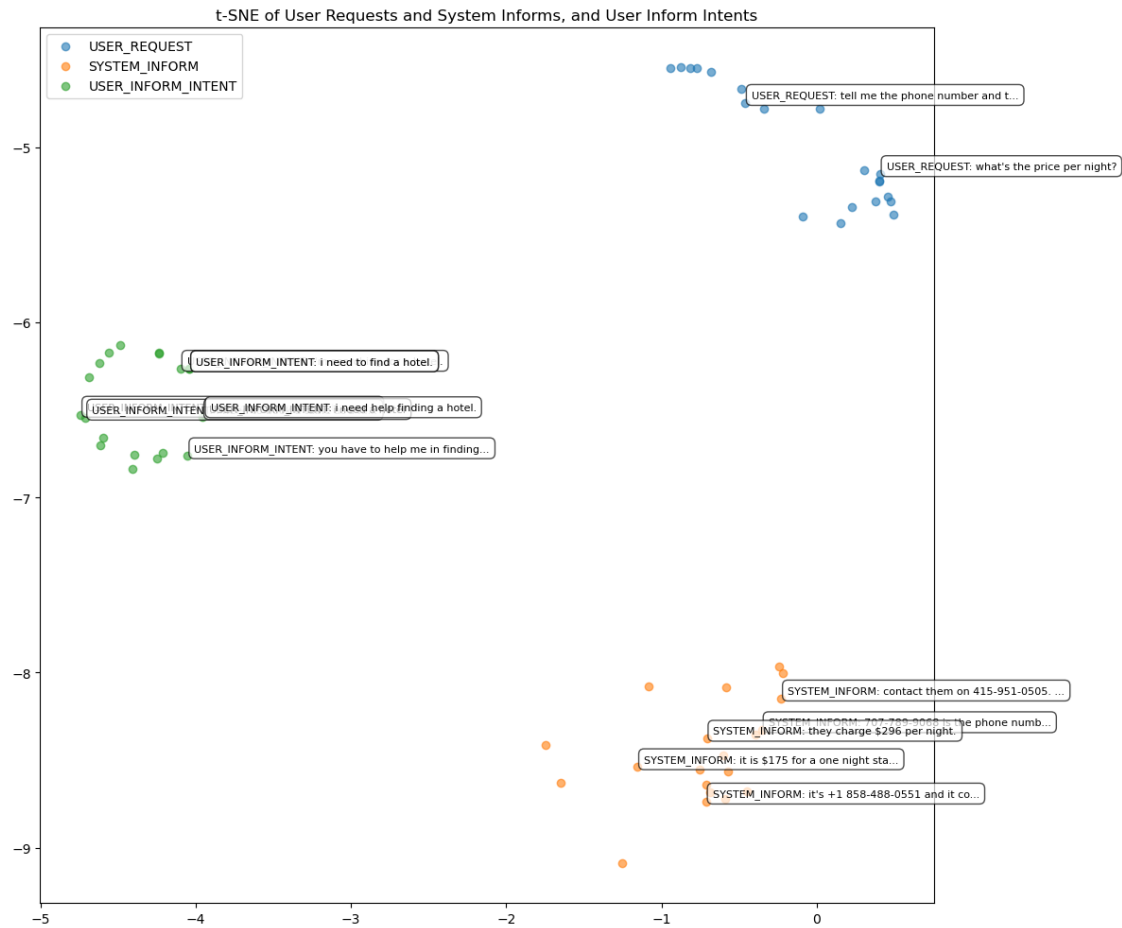
Saved plots to t-SNE of Top 10 Occuring Dialogue Acts.png



```
[75]: utterances_test = [data['utterance'].lower() for data in extracted_test]
```

```
[83]: # Plot zoomed version for just a specific subset of acts
plot_tsne_zoomed(
    vectors=utterances,
    labels=labels,
    texts=utterances_test,
    selected_labels=['USER_REQUEST', 'SYSTEM_INFORM', 'USER_INFORM_INTENT'], #_
    ↪Focus on these classes
    max_per_label=20, # Limit points per class
    annotate_sample=0.3, # Annotate 30% of points
    zoom_scale=1.0,
    title="t-SNE of User Requests and System Informs, and User Inform Intents",
)
```

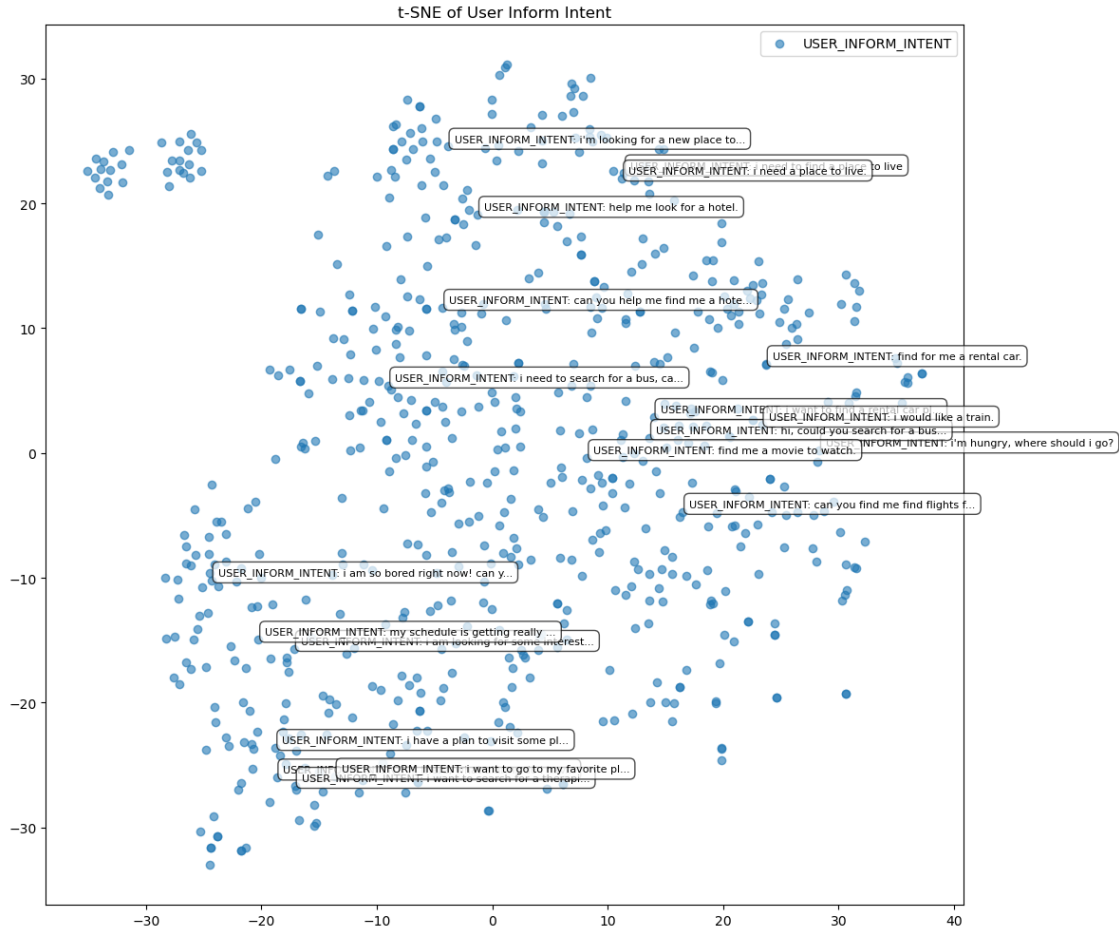
Saved plots to t-SNE of User Requests and System Informs, and User Inform Intents.png



The model is so good at clustering that the clusters being this far apart makes it harder to differentiate, let's take a look at one specific dialogue act:

```
[99]: # Plot zoomed version for just a specific subset of acts
plot_tsne_zoomed(
    vectors=utterances,
    labels=labels,
    texts=utterances_test,
    selected_labels=['USER_INFORM_INTENT'], # Focus on these classes
    max_per_label=1000,                    # Limit points per class
    annotate_sample=0.03,                  # Annotate 3% of points
    zoom_scale=1.0,
    title="t-SNE of User Inform Intent",
)
```

Saved plots to t-SNE of User Inform Intent.png



Ablation study: context window size We will now run an experiment to analyse the impact of window size in the dialogue recognition task. The window size decides how many previous utterances are used as context for the current one. Sizes [0, 10] will be checked, and the resulting accuracies and training times will be saved in a .csv file.

```
[20]: # Define where to save your results
results_file = "ablation_window_size_results.csv"

# Open CSV file for writing metrics
with open(results_file, mode='w', newline='') as f:
    writer = csv.writer(f)
    writer.writerow([
        "window_size", "accuracy", "top5_accuracy", "balanced_accuracy",
        "dataset_creation_time_sec", "train_time_sec"
    ])

    for window_size in range(0, 11): # From 0 to 10 inclusive
        print(f"\n=== Running training for window size: {window_size} ===")
```

```

# Measure training time
start_time_dataset = time.time()

# Load datasets
train_dataset = DialogueWindowDataset(
    embeddings_path='train_embeddings_average.pt',
    file_path=path_to_train,
    window_size=window_size
)
train_loader = DataLoader(train_dataset, batch_size=128, shuffle=True)

test_dataset = DialogueWindowDataset(
    embeddings_path='test_embeddings_average.pt',
    file_path=path_to_test,
    window_size=window_size
)
test_loader = DataLoader(test_dataset, batch_size=128, shuffle=False)

end_time_dataset = time.time()
dataset_creation_time = end_time_dataset - start_time_dataset

# Initialize model, optimizer, criterion
model = ContextNet(input_size, hidden_size, num_classes, pooling_type).
    ↪to(device)
optimizer = torch.optim.Adam(model.parameters(), lr=1e-3)
criterion = torch.nn.CrossEntropyLoss()

train_start_time = time.time()

# Train the model
losses, accuracies = run_training_loop(
    model=model,
    train_loader=train_loader,
    optimizer=optimizer,
    criterion=criterion,
    num_epochs=num_epochs,
    prepare_batch=prepare_batch_v2,
    forward=forward_v2,
    save_dir=f"train_logs/window_{window_size}"
)

train_end_time = time.time()
train_time = train_end_time - train_start_time

# Evaluate on the test set
metrics = evaluate(

```

```

        model=model,
        test_loader=test_loader,
        prepare_batch_fn=prepare_batch_v2,
        forward_fn=forward_v2,
        inv_act_labels=inv_acts_labels
    )

    # Write metrics to CSV
    writer.writerow([
        window_size,
        metrics["accuracy"],
        metrics["top5_accuracy"],
        metrics["balanced_accuracy"],
        round(dataset_creation_time, 2),
        round(train_time, 2)
    ])

print(f"\nAblation study completed! Results saved to '{results_file}'")

```

=== Running training for window size: 0 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```

    self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:04<00:00, 1081.68it/s]
100%|      | 1331/1331 [00:00<00:00, 1398.08it/s]
100%|      | 646/646 [00:02<00:00, 231.94it/s]

Epoch 1/50, Train Loss: 1.130, Train Accuracy: 74.588
100%|      | 646/646 [00:02<00:00, 281.60it/s]

Epoch 2/50, Train Loss: 0.401, Train Accuracy: 89.584
100%|      | 646/646 [00:02<00:00, 262.92it/s]

Epoch 3/50, Train Loss: 0.293, Train Accuracy: 91.936

```

100%| | 646/646 [00:02<00:00, 274.78it/s]
Epoch 4/50, Train Loss: 0.245, Train Accuracy: 92.974
100%| | 646/646 [00:02<00:00, 276.54it/s]
Epoch 5/50, Train Loss: 0.217, Train Accuracy: 93.730
100%| | 646/646 [00:02<00:00, 277.85it/s]
Epoch 6/50, Train Loss: 0.197, Train Accuracy: 94.238
100%| | 646/646 [00:02<00:00, 282.34it/s]
Epoch 7/50, Train Loss: 0.182, Train Accuracy: 94.611
100%| | 646/646 [00:02<00:00, 273.84it/s]
Epoch 8/50, Train Loss: 0.170, Train Accuracy: 94.976
100%| | 646/646 [00:02<00:00, 282.47it/s]
Epoch 9/50, Train Loss: 0.160, Train Accuracy: 95.222
100%| | 646/646 [00:02<00:00, 278.09it/s]
Epoch 10/50, Train Loss: 0.151, Train Accuracy: 95.406
100%| | 646/646 [00:02<00:00, 280.75it/s]
Epoch 11/50, Train Loss: 0.144, Train Accuracy: 95.601
100%| | 646/646 [00:02<00:00, 276.42it/s]
Epoch 12/50, Train Loss: 0.137, Train Accuracy: 95.869
100%| | 646/646 [00:02<00:00, 274.78it/s]
Epoch 13/50, Train Loss: 0.131, Train Accuracy: 95.966
100%| | 646/646 [00:02<00:00, 261.86it/s]
Epoch 14/50, Train Loss: 0.125, Train Accuracy: 96.186
100%| | 646/646 [00:02<00:00, 270.29it/s]
Epoch 15/50, Train Loss: 0.121, Train Accuracy: 96.328
100%| | 646/646 [00:02<00:00, 269.95it/s]
Epoch 16/50, Train Loss: 0.116, Train Accuracy: 96.449
100%| | 646/646 [00:02<00:00, 259.86it/s]
Epoch 17/50, Train Loss: 0.112, Train Accuracy: 96.585
100%| | 646/646 [00:02<00:00, 276.42it/s]
Epoch 18/50, Train Loss: 0.108, Train Accuracy: 96.698
100%| | 646/646 [00:02<00:00, 281.97it/s]
Epoch 19/50, Train Loss: 0.103, Train Accuracy: 96.882

100%| | 646/646 [00:02<00:00, 281.73it/s]
Epoch 20/50, Train Loss: 0.100, Train Accuracy: 96.962
100%| | 646/646 [00:02<00:00, 281.11it/s]
Epoch 21/50, Train Loss: 0.097, Train Accuracy: 97.033
100%| | 646/646 [00:02<00:00, 274.19it/s]
Epoch 22/50, Train Loss: 0.094, Train Accuracy: 97.117
100%| | 646/646 [00:02<00:00, 275.39it/s]
Epoch 23/50, Train Loss: 0.091, Train Accuracy: 97.231
100%| | 646/646 [00:02<00:00, 272.63it/s]
Epoch 24/50, Train Loss: 0.088, Train Accuracy: 97.259
100%| | 646/646 [00:02<00:00, 276.53it/s]
Epoch 25/50, Train Loss: 0.086, Train Accuracy: 97.346
100%| | 646/646 [00:02<00:00, 264.52it/s]
Epoch 26/50, Train Loss: 0.084, Train Accuracy: 97.432
100%| | 646/646 [00:02<00:00, 275.09it/s]
Epoch 27/50, Train Loss: 0.081, Train Accuracy: 97.582
100%| | 646/646 [00:02<00:00, 278.28it/s]
Epoch 28/50, Train Loss: 0.079, Train Accuracy: 97.582
100%| | 646/646 [00:02<00:00, 277.36it/s]
Epoch 29/50, Train Loss: 0.078, Train Accuracy: 97.588
100%| | 646/646 [00:02<00:00, 254.49it/s]
Epoch 30/50, Train Loss: 0.076, Train Accuracy: 97.664
100%| | 646/646 [00:02<00:00, 244.41it/s]
Epoch 31/50, Train Loss: 0.074, Train Accuracy: 97.686
100%| | 646/646 [00:02<00:00, 261.30it/s]
Epoch 32/50, Train Loss: 0.073, Train Accuracy: 97.728
100%| | 646/646 [00:02<00:00, 267.97it/s]
Epoch 33/50, Train Loss: 0.071, Train Accuracy: 97.783
100%| | 646/646 [00:02<00:00, 272.74it/s]
Epoch 34/50, Train Loss: 0.069, Train Accuracy: 97.787
100%| | 646/646 [00:02<00:00, 273.39it/s]
Epoch 35/50, Train Loss: 0.068, Train Accuracy: 97.833

100%| | 646/646 [00:02<00:00, 274.46it/s]
Epoch 36/50, Train Loss: 0.067, Train Accuracy: 97.845
100%| | 646/646 [00:02<00:00, 276.91it/s]
Epoch 37/50, Train Loss: 0.066, Train Accuracy: 97.944
100%| | 646/646 [00:02<00:00, 268.63it/s]
Epoch 38/50, Train Loss: 0.065, Train Accuracy: 97.926
100%| | 646/646 [00:02<00:00, 275.06it/s]
Epoch 39/50, Train Loss: 0.064, Train Accuracy: 97.950
100%| | 646/646 [00:02<00:00, 271.14it/s]
Epoch 40/50, Train Loss: 0.063, Train Accuracy: 97.969
100%| | 646/646 [00:02<00:00, 282.12it/s]
Epoch 41/50, Train Loss: 0.062, Train Accuracy: 98.007
100%| | 646/646 [00:02<00:00, 251.97it/s]
Epoch 42/50, Train Loss: 0.061, Train Accuracy: 98.013
100%| | 646/646 [00:02<00:00, 276.74it/s]
Epoch 43/50, Train Loss: 0.061, Train Accuracy: 98.066
100%| | 646/646 [00:02<00:00, 280.50it/s]
Epoch 44/50, Train Loss: 0.060, Train Accuracy: 98.029
100%| | 646/646 [00:02<00:00, 276.55it/s]
Epoch 45/50, Train Loss: 0.059, Train Accuracy: 98.081
100%| | 646/646 [00:02<00:00, 277.53it/s]
Epoch 46/50, Train Loss: 0.058, Train Accuracy: 98.076
100%| | 646/646 [00:02<00:00, 275.83it/s]
Epoch 47/50, Train Loss: 0.058, Train Accuracy: 98.114
100%| | 646/646 [00:02<00:00, 274.25it/s]
Epoch 48/50, Train Loss: 0.057, Train Accuracy: 98.134
100%| | 646/646 [00:02<00:00, 278.12it/s]
Epoch 49/50, Train Loss: 0.057, Train Accuracy: 98.077
100%| | 646/646 [00:02<00:00, 275.38it/s]
Epoch 50/50, Train Loss: 0.056, Train Accuracy: 98.122

Training complete. Model saved to 'train_logs/window_0\model.pth', stats saved to 'train_logs/window_0\train_stats.csv'

```

100%|      | 132/132 [00:00<00:00, 234.04it/s]
c:\Users\Ward\anaconda3\envs\multibench\Lib\site-
packages\sklearn\metrics\_classification.py:2480: UserWarning: y_pred contains
classes not in y_true
  warnings.warn("y_pred contains classes not in y_true")

```

Evaluation Results:
 Top-1 Accuracy: 87.69%
 Top-5 Accuracy: 98.53%
 Balanced Accuracy: 74.69%

Classification Report:		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.98	0.96	0.97
1113				
	SYSTEM_GOODBYE	0.99	0.98	0.98
1331				
	SYSTEM_INFORM	0.82	0.92	0.87
719				
	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.83	0.83	0.83
310				
	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.79	0.88	0.83
17				
	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.93	0.99	0.96
79				
	SYSTEM_NOTIFY_SUCCESS	0.92	0.92	0.92
499				
	SYSTEM_OFFER	0.88	0.85	0.86
806				
	SYSTEM_OFFER_INTENT	0.87	0.97	0.92
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.98	0.96	0.97
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.36	0.53
14				
	SYSTEM_REQUEST	0.99	0.96	0.98
1961				
	SYSTEM_REQ_MORE	0.97	0.97	0.97
585				
	USER_AFFIRM	0.83	0.80	0.81
572				
	USER_AFFIRM_INTENT	0.70	0.82	0.75
139				
	USER_AFFIRM_INTENT USER_INFORM	0.64	0.71	0.67
103				

1961	USER_INFORM	0.82	0.89	0.85
658	USER_INFORM_INTENT	0.93	0.87	0.90
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.68	0.83	0.75
12	USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
772	USER_INFORM USER_INFORM_INTENT	0.82	0.56	0.67
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.66	0.78	0.72
14	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
216	USER_INFORM USER_NEGATE	0.81	0.86	0.84
184	USER_INFORM USER_REQUEST_ALTS	0.66	0.72	0.69
7	USER_NEGATE	1.00	0.14	0.25
69	USER_NEGATE_INTENT	0.56	0.75	0.64
72	USER_NEGATE_INTENT USER_GOODBYE	0.63	0.72	0.67
9	USER_NEGATE USER_GOODBYE	0.00	0.00	0.00
587	USER_NEGATE USER_THANK_YOU	0.93	0.89	0.91
719	USER_REQUEST	0.93	0.89	0.91
152	USER_REQUEST_ALTS	0.94	0.97	0.95
340	USER_REQUEST USER_AFFIRM	0.86	0.86	0.86
515	USER_SELECT	0.70	0.77	0.73
279	USER_SELECT USER_GOODBYE	0.76	0.66	0.71
377	USER_THANK_YOU	0.79	0.77	0.78
384	USER_THANK_YOU USER_GOODBYE	0.68	0.82	0.74
16850	accuracy			0.88
	macro avg	0.74	0.73	0.72

16850

weighted avg 0.88 0.88 0.88

16850

=== Running training for window size: 1 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
```

```
100%|      | 5403/5403 [00:10<00:00, 525.01it/s]
```

```
100%|      | 1331/1331 [00:02<00:00, 629.46it/s]
```

```
100%|      | 646/646 [00:03<00:00, 164.08it/s]
```

Epoch 1/50, Train Loss: 1.349, Train Accuracy: 68.132

```
100%|      | 646/646 [00:03<00:00, 186.59it/s]
```

Epoch 2/50, Train Loss: 0.409, Train Accuracy: 91.294

```
100%|      | 646/646 [00:03<00:00, 183.57it/s]
```

Epoch 3/50, Train Loss: 0.232, Train Accuracy: 95.252

```
100%|      | 646/646 [00:03<00:00, 172.96it/s]
```

Epoch 4/50, Train Loss: 0.158, Train Accuracy: 96.585

```
100%|      | 646/646 [00:03<00:00, 187.11it/s]
```

Epoch 5/50, Train Loss: 0.118, Train Accuracy: 97.365

```
100%|      | 646/646 [00:03<00:00, 184.28it/s]
```

Epoch 6/50, Train Loss: 0.094, Train Accuracy: 97.850

```
100%|      | 646/646 [00:03<00:00, 178.96it/s]
```

Epoch 7/50, Train Loss: 0.077, Train Accuracy: 98.219

```
100%|      | 646/646 [00:03<00:00, 181.87it/s]
```

Epoch 8/50, Train Loss: 0.065, Train Accuracy: 98.468

100%| | 646/646 [00:03<00:00, 183.68it/s]
Epoch 9/50, Train Loss: 0.055, Train Accuracy: 98.701
100%| | 646/646 [00:03<00:00, 183.02it/s]
Epoch 10/50, Train Loss: 0.048, Train Accuracy: 98.862
100%| | 646/646 [00:03<00:00, 184.12it/s]
Epoch 11/50, Train Loss: 0.041, Train Accuracy: 99.039
100%| | 646/646 [00:03<00:00, 184.04it/s]
Epoch 12/50, Train Loss: 0.036, Train Accuracy: 99.180
100%| | 646/646 [00:03<00:00, 184.35it/s]
Epoch 13/50, Train Loss: 0.032, Train Accuracy: 99.242
100%| | 646/646 [00:03<00:00, 182.21it/s]
Epoch 14/50, Train Loss: 0.028, Train Accuracy: 99.379
100%| | 646/646 [00:03<00:00, 182.69it/s]
Epoch 15/50, Train Loss: 0.025, Train Accuracy: 99.421
100%| | 646/646 [00:03<00:00, 185.83it/s]
Epoch 16/50, Train Loss: 0.022, Train Accuracy: 99.477
100%| | 646/646 [00:03<00:00, 176.86it/s]
Epoch 17/50, Train Loss: 0.020, Train Accuracy: 99.536
100%| | 646/646 [00:03<00:00, 182.86it/s]
Epoch 18/50, Train Loss: 0.018, Train Accuracy: 99.571
100%| | 646/646 [00:03<00:00, 183.08it/s]
Epoch 19/50, Train Loss: 0.016, Train Accuracy: 99.621
100%| | 646/646 [00:03<00:00, 184.31it/s]
Epoch 20/50, Train Loss: 0.014, Train Accuracy: 99.649
100%| | 646/646 [00:03<00:00, 168.14it/s]
Epoch 21/50, Train Loss: 0.013, Train Accuracy: 99.692
100%| | 646/646 [00:03<00:00, 177.93it/s]
Epoch 22/50, Train Loss: 0.011, Train Accuracy: 99.731
100%| | 646/646 [00:03<00:00, 178.04it/s]
Epoch 23/50, Train Loss: 0.010, Train Accuracy: 99.766
100%| | 646/646 [00:03<00:00, 173.30it/s]
Epoch 24/50, Train Loss: 0.010, Train Accuracy: 99.760

100%| | 646/646 [00:03<00:00, 173.12it/s]
Epoch 25/50, Train Loss: 0.009, Train Accuracy: 99.811
100%| | 646/646 [00:03<00:00, 173.53it/s]
Epoch 26/50, Train Loss: 0.008, Train Accuracy: 99.818
100%| | 646/646 [00:03<00:00, 172.56it/s]
Epoch 27/50, Train Loss: 0.007, Train Accuracy: 99.838
100%| | 646/646 [00:03<00:00, 177.06it/s]
Epoch 28/50, Train Loss: 0.007, Train Accuracy: 99.847
100%| | 646/646 [00:03<00:00, 168.71it/s]
Epoch 29/50, Train Loss: 0.006, Train Accuracy: 99.851
100%| | 646/646 [00:03<00:00, 170.63it/s]
Epoch 30/50, Train Loss: 0.006, Train Accuracy: 99.873
100%| | 646/646 [00:03<00:00, 181.72it/s]
Epoch 31/50, Train Loss: 0.005, Train Accuracy: 99.872
100%| | 646/646 [00:03<00:00, 170.54it/s]
Epoch 32/50, Train Loss: 0.005, Train Accuracy: 99.876
100%| | 646/646 [00:03<00:00, 188.50it/s]
Epoch 33/50, Train Loss: 0.005, Train Accuracy: 99.875
100%| | 646/646 [00:03<00:00, 181.89it/s]
Epoch 34/50, Train Loss: 0.004, Train Accuracy: 99.904
100%| | 646/646 [00:03<00:00, 172.82it/s]
Epoch 35/50, Train Loss: 0.004, Train Accuracy: 99.915
100%| | 646/646 [00:03<00:00, 182.43it/s]
Epoch 36/50, Train Loss: 0.004, Train Accuracy: 99.907
100%| | 646/646 [00:03<00:00, 165.41it/s]
Epoch 37/50, Train Loss: 0.004, Train Accuracy: 99.903
100%| | 646/646 [00:03<00:00, 186.43it/s]
Epoch 38/50, Train Loss: 0.003, Train Accuracy: 99.921
100%| | 646/646 [00:03<00:00, 185.23it/s]
Epoch 39/50, Train Loss: 0.003, Train Accuracy: 99.926
100%| | 646/646 [00:03<00:00, 178.81it/s]
Epoch 40/50, Train Loss: 0.003, Train Accuracy: 99.903

```

100%|      | 646/646 [00:03<00:00, 181.31it/s]
Epoch 41/50, Train Loss: 0.003, Train Accuracy: 99.909
100%|      | 646/646 [00:03<00:00, 178.68it/s]
Epoch 42/50, Train Loss: 0.004, Train Accuracy: 99.913
100%|      | 646/646 [00:03<00:00, 175.67it/s]
Epoch 43/50, Train Loss: 0.003, Train Accuracy: 99.931
100%|      | 646/646 [00:03<00:00, 175.90it/s]
Epoch 44/50, Train Loss: 0.004, Train Accuracy: 99.901
100%|      | 646/646 [00:03<00:00, 179.22it/s]
Epoch 45/50, Train Loss: 0.003, Train Accuracy: 99.931
100%|      | 646/646 [00:03<00:00, 182.30it/s]
Epoch 46/50, Train Loss: 0.002, Train Accuracy: 99.944
100%|      | 646/646 [00:03<00:00, 184.43it/s]
Epoch 47/50, Train Loss: 0.003, Train Accuracy: 99.924
100%|      | 646/646 [00:03<00:00, 189.21it/s]
Epoch 48/50, Train Loss: 0.003, Train Accuracy: 99.918
100%|      | 646/646 [00:03<00:00, 182.67it/s]
Epoch 49/50, Train Loss: 0.002, Train Accuracy: 99.955
100%|      | 646/646 [00:03<00:00, 180.95it/s]
Epoch 50/50, Train Loss: 0.002, Train Accuracy: 99.949

```

```

Training complete. Model saved to 'train_logs/window_1\model.pth', stats saved
to 'train_logs/window_1\train_stats.csv'

```

```

100%|      | 132/132 [00:00<00:00, 212.12it/s]
c:\Users\Ward\anaconda3\envs\multibench\Lib\site-
packages\sklearn\metrics\_classification.py:2480: UserWarning: y_pred contains
classes not in y_true
  warnings.warn("y_pred contains classes not in y_true")

```

```

Evaluation Results:
Top-1 Accuracy: 94.64%
Top-5 Accuracy: 99.57%
Balanced Accuracy: 86.63%

```

```

Classification Report:

```

	precision	recall	f1-score
support			

1113	SYSTEM_CONFIRM	0.99	0.94	0.96
1331	SYSTEM_GOODBYE	0.98	0.99	0.99
719	SYSTEM_INFORM	0.95	0.95	0.95
310	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.96	0.89	0.92
17	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.88	0.82	0.85
79	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	1.00	0.95	0.97
499	SYSTEM_NOTIFY_SUCCESS	0.97	0.98	0.97
806	SYSTEM_OFFER	0.91	0.97	0.94
383	SYSTEM_OFFER_INTENT	0.92	0.97	0.94
608	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.96	0.97	0.97
14	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.93	0.93	0.93
1961	SYSTEM_REQUEST	0.99	0.98	0.98
585	SYSTEM_REQ_MORE	0.97	0.99	0.98
572	USER_AFFIRM	0.98	0.96	0.97
139	USER_AFFIRM_INTENT	0.90	0.94	0.92
103	USER_AFFIRM_INTENT USER_INFORM	0.96	0.90	0.93
1961	USER_INFORM	1.00	0.97	0.98
658	USER_INFORM_INTENT	0.96	0.93	0.94
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.83	0.89	0.86
12	USER_INFORM_INTENT USER_THANK_YOU	0.50	0.33	0.40
772	USER_INFORM USER_INFORM_INTENT	0.95	0.96	0.95
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.88	0.85	0.87
	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.50	0.14	0.22

14				
	USER_INFORM USER_NEGATE	0.97	0.94	0.96
216				
	USER_INFORM USER_REQUEST_ALTS	0.85	0.92	0.89
184				
	USER_NEGATE	0.75	0.86	0.80
7				
	USER_NEGATE_INTENT	0.68	0.78	0.72
69				
	USER_NEGATE_INTENT USER_GOODBYE	0.76	0.69	0.72
72				
	USER_NEGATE USER_GOODBYE	0.62	0.56	0.59
9				
	USER_NEGATE USER_THANK_YOU	0.96	0.99	0.98
587				
	USER_REQUEST	0.98	0.95	0.96
719				
	USER_REQUEST_ALTS	0.95	0.93	0.94
152				
	USER_REQUEST USER_AFFIRM	0.96	0.96	0.96
340				
	USER_SELECT	0.84	0.91	0.87
515				
	USER_SELECT USER_GOODBYE	0.78	0.65	0.71
279				
	USER_THANK_YOU	0.81	0.86	0.83
377				
	USER_THANK_YOU USER_GOODBYE	0.73	0.81	0.77
384				
	accuracy			0.95
16850				
	macro avg	0.86	0.84	0.85
16850				
	weighted avg	0.95	0.95	0.95
16850				

=== Running training for window size: 2 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning:
You are using `torch.load` with `weights_only=False` (the current default
value), which uses the default pickle module implicitly. It is possible to
construct malicious pickle data which will execute arbitrary code during
unpickling (See
<https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for
more details). In a future release, the default value for `weights_only` will be
flipped to `True`. This limits the functions that could be executed during

unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via ``torch.serialization.add_safe_globals``. We recommend you start setting ``weights_only=True`` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:12<00:00, 427.38it/s]
100%|      | 1331/1331 [00:01<00:00, 708.23it/s]
100%|      | 646/646 [00:05<00:00, 121.61it/s]

Epoch 1/50, Train Loss: 1.709, Train Accuracy: 54.532
100%|      | 646/646 [00:04<00:00, 137.06it/s]

Epoch 2/50, Train Loss: 0.610, Train Accuracy: 84.959
100%|      | 646/646 [00:04<00:00, 136.14it/s]

Epoch 3/50, Train Loss: 0.385, Train Accuracy: 90.795
100%|      | 646/646 [00:04<00:00, 138.87it/s]

Epoch 4/50, Train Loss: 0.272, Train Accuracy: 94.206
100%|      | 646/646 [00:04<00:00, 136.13it/s]

Epoch 5/50, Train Loss: 0.202, Train Accuracy: 95.978
100%|      | 646/646 [00:04<00:00, 135.38it/s]

Epoch 6/50, Train Loss: 0.155, Train Accuracy: 96.918
100%|      | 646/646 [00:04<00:00, 136.34it/s]

Epoch 7/50, Train Loss: 0.122, Train Accuracy: 97.506
100%|      | 646/646 [00:04<00:00, 140.50it/s]

Epoch 8/50, Train Loss: 0.099, Train Accuracy: 97.974
100%|      | 646/646 [00:04<00:00, 132.89it/s]

Epoch 9/50, Train Loss: 0.082, Train Accuracy: 98.317
100%|      | 646/646 [00:04<00:00, 138.09it/s]

Epoch 10/50, Train Loss: 0.069, Train Accuracy: 98.617
100%|      | 646/646 [00:04<00:00, 139.55it/s]

Epoch 11/50, Train Loss: 0.058, Train Accuracy: 98.859
100%|      | 646/646 [00:04<00:00, 136.47it/s]

Epoch 12/50, Train Loss: 0.050, Train Accuracy: 99.029
100%|      | 646/646 [00:04<00:00, 137.84it/s]

Epoch 13/50, Train Loss: 0.043, Train Accuracy: 99.173
```


100%| | 646/646 [00:04<00:00, 135.98it/s]
Epoch 14/50, Train Loss: 0.037, Train Accuracy: 99.309
100%| | 646/646 [00:04<00:00, 135.26it/s]
Epoch 15/50, Train Loss: 0.033, Train Accuracy: 99.338
100%| | 646/646 [00:04<00:00, 133.36it/s]
Epoch 16/50, Train Loss: 0.028, Train Accuracy: 99.447
100%| | 646/646 [00:04<00:00, 137.06it/s]
Epoch 17/50, Train Loss: 0.025, Train Accuracy: 99.548
100%| | 646/646 [00:04<00:00, 137.79it/s]
Epoch 18/50, Train Loss: 0.022, Train Accuracy: 99.602
100%| | 646/646 [00:04<00:00, 135.28it/s]
Epoch 19/50, Train Loss: 0.019, Train Accuracy: 99.663
100%| | 646/646 [00:05<00:00, 128.56it/s]
Epoch 20/50, Train Loss: 0.017, Train Accuracy: 99.712
100%| | 646/646 [00:05<00:00, 114.29it/s]
Epoch 21/50, Train Loss: 0.015, Train Accuracy: 99.734
100%| | 646/646 [00:05<00:00, 118.62it/s]
Epoch 22/50, Train Loss: 0.013, Train Accuracy: 99.787
100%| | 646/646 [00:05<00:00, 118.15it/s]
Epoch 23/50, Train Loss: 0.012, Train Accuracy: 99.821
100%| | 646/646 [00:05<00:00, 124.66it/s]
Epoch 24/50, Train Loss: 0.010, Train Accuracy: 99.847
100%| | 646/646 [00:05<00:00, 127.71it/s]
Epoch 25/50, Train Loss: 0.009, Train Accuracy: 99.858
100%| | 646/646 [00:05<00:00, 127.17it/s]
Epoch 26/50, Train Loss: 0.007, Train Accuracy: 99.909
100%| | 646/646 [00:05<00:00, 123.37it/s]
Epoch 27/50, Train Loss: 0.007, Train Accuracy: 99.900
100%| | 646/646 [00:05<00:00, 125.19it/s]
Epoch 28/50, Train Loss: 0.007, Train Accuracy: 99.912
100%| | 646/646 [00:04<00:00, 133.33it/s]
Epoch 29/50, Train Loss: 0.005, Train Accuracy: 99.935

100%| | 646/646 [00:04<00:00, 132.33it/s]
Epoch 30/50, Train Loss: 0.006, Train Accuracy: 99.912
100%| | 646/646 [00:04<00:00, 129.95it/s]
Epoch 31/50, Train Loss: 0.005, Train Accuracy: 99.938
100%| | 646/646 [00:04<00:00, 132.64it/s]
Epoch 32/50, Train Loss: 0.004, Train Accuracy: 99.942
100%| | 646/646 [00:05<00:00, 129.00it/s]
Epoch 33/50, Train Loss: 0.004, Train Accuracy: 99.954
100%| | 646/646 [00:04<00:00, 132.32it/s]
Epoch 34/50, Train Loss: 0.005, Train Accuracy: 99.909
100%| | 646/646 [00:04<00:00, 132.44it/s]
Epoch 35/50, Train Loss: 0.003, Train Accuracy: 99.964
100%| | 646/646 [00:04<00:00, 131.04it/s]
Epoch 36/50, Train Loss: 0.003, Train Accuracy: 99.972
100%| | 646/646 [00:04<00:00, 131.03it/s]
Epoch 37/50, Train Loss: 0.002, Train Accuracy: 99.972
100%| | 646/646 [00:05<00:00, 121.80it/s]
Epoch 38/50, Train Loss: 0.003, Train Accuracy: 99.952
100%| | 646/646 [00:05<00:00, 114.84it/s]
Epoch 39/50, Train Loss: 0.003, Train Accuracy: 99.965
100%| | 646/646 [00:05<00:00, 128.56it/s]
Epoch 40/50, Train Loss: 0.003, Train Accuracy: 99.967
100%| | 646/646 [00:04<00:00, 133.62it/s]
Epoch 41/50, Train Loss: 0.002, Train Accuracy: 99.975
100%| | 646/646 [00:04<00:00, 137.24it/s]
Epoch 42/50, Train Loss: 0.002, Train Accuracy: 99.972
100%| | 646/646 [00:04<00:00, 140.47it/s]
Epoch 43/50, Train Loss: 0.002, Train Accuracy: 99.979
100%| | 646/646 [00:04<00:00, 138.16it/s]
Epoch 44/50, Train Loss: 0.001, Train Accuracy: 99.992
100%| | 646/646 [00:04<00:00, 133.45it/s]
Epoch 45/50, Train Loss: 0.005, Train Accuracy: 99.873

```

100%|      | 646/646 [00:04<00:00, 139.68it/s]
Epoch 46/50, Train Loss: 0.002, Train Accuracy: 99.987
100%|      | 646/646 [00:04<00:00, 135.94it/s]
Epoch 47/50, Train Loss: 0.001, Train Accuracy: 99.996
100%|      | 646/646 [00:04<00:00, 139.74it/s]
Epoch 48/50, Train Loss: 0.001, Train Accuracy: 99.998
100%|      | 646/646 [00:04<00:00, 129.87it/s]
Epoch 49/50, Train Loss: 0.001, Train Accuracy: 99.993
100%|      | 646/646 [00:04<00:00, 133.93it/s]
Epoch 50/50, Train Loss: 0.004, Train Accuracy: 99.909

```

Training complete. Model saved to 'train_logs/window_2\model.pth', stats saved to 'train_logs/window_2\train_stats.csv'

```

100%|      | 132/132 [00:00<00:00, 157.48it/s]
c:\Users\Ward\anaconda3\envs\multibench\Lib\site-
packages\sklearn\metrics\_classification.py:2480: UserWarning: y_pred contains
classes not in y_true
  warnings.warn("y_pred contains classes not in y_true")

```

Evaluation Results:
Top-1 Accuracy: 94.54%
Top-5 Accuracy: 99.53%
Balanced Accuracy: 84.41%

Classification Report:		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.99	0.95	0.97
1113	SYSTEM_GOODBYE	0.99	0.99	0.99
1331	SYSTEM_INFORM	0.99	0.95	0.97
719	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.98	0.96	0.97
310	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.88	0.88	0.88
17	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.97	0.95	0.96
79	SYSTEM_NOTIFY_SUCCESS	0.98	0.98	0.98
499				

806	SYSTEM_OFFER	0.91	0.98	0.94
383	SYSTEM_OFFER_INTENT	0.96	0.97	0.97
608	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.97	0.95	0.96
14	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.93	0.96
1961	SYSTEM_REQUEST	0.99	0.99	0.99
585	SYSTEM_REQ_MORE	0.97	0.98	0.98
572	USER_AFFIRM	0.96	0.96	0.96
139	USER_AFFIRM_INTENT	0.91	0.91	0.91
103	USER_AFFIRM_INTENT USER_INFORM	0.94	0.85	0.89
1961	USER_INFORM	0.99	0.97	0.98
658	USER_INFORM_INTENT	0.95	0.94	0.95
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.77	0.90	0.83
12	USER_INFORM_INTENT USER_THANK_YOU	0.75	0.25	0.38
772	USER_INFORM USER_INFORM_INTENT	0.96	0.95	0.96
0	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.83	0.80	0.82
14	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.33	0.07	0.12
216	USER_INFORM USER_NEGATE	0.95	0.91	0.93
184	USER_INFORM USER_REQUEST_ALTS	0.76	0.92	0.84
7	USER_NEGATE	0.42	0.71	0.53
69	USER_NEGATE_INTENT	0.60	0.72	0.65
72	USER_NEGATE_INTENT USER_GOODBYE	0.75	0.83	0.79
9	USER_NEGATE USER_GOODBYE	0.50	0.11	0.18

587	USER_NEGATE USER_THANK_YOU	0.98	0.99	0.99
719	USER_REQUEST	0.97	0.92	0.94
152	USER_REQUEST_ALTS	0.91	0.89	0.90
340	USER_REQUEST USER_AFFIRM	0.96	0.93	0.95
515	USER_SELECT	0.85	0.91	0.88
279	USER_SELECT USER_GOODBYE	0.75	0.71	0.73
377	USER_THANK_YOU	0.82	0.78	0.80
384	USER_THANK_YOU USER_GOODBYE	0.74	0.80	0.77
16850	accuracy			0.95
16850	macro avg	0.82	0.80	0.80
16850	weighted avg	0.95	0.95	0.95

=== Running training for window size: 3 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```

self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:19<00:00, 270.40it/s]
100%|      | 1331/1331 [00:03<00:00, 412.92it/s]
100%|      | 646/646 [00:17<00:00, 36.07it/s]
```

Epoch 1/50, Train Loss: 1.970, Train Accuracy: 45.085

```

100%|      | 646/646 [00:05<00:00, 108.69it/s]
```

Epoch 2/50, Train Loss: 0.869, Train Accuracy: 77.950
100%| | 646/646 [00:05<00:00, 113.46it/s]
Epoch 3/50, Train Loss: 0.558, Train Accuracy: 86.640
100%| | 646/646 [00:05<00:00, 112.16it/s]
Epoch 4/50, Train Loss: 0.402, Train Accuracy: 90.955
100%| | 646/646 [00:05<00:00, 110.41it/s]
Epoch 5/50, Train Loss: 0.302, Train Accuracy: 93.566
100%| | 646/646 [00:05<00:00, 112.81it/s]
Epoch 6/50, Train Loss: 0.234, Train Accuracy: 95.233
100%| | 646/646 [00:05<00:00, 111.91it/s]
Epoch 7/50, Train Loss: 0.185, Train Accuracy: 96.320
100%| | 646/646 [00:05<00:00, 111.65it/s]
Epoch 8/50, Train Loss: 0.149, Train Accuracy: 97.138
100%| | 646/646 [00:05<00:00, 110.34it/s]
Epoch 9/50, Train Loss: 0.123, Train Accuracy: 97.601
100%| | 646/646 [00:05<00:00, 112.76it/s]
Epoch 10/50, Train Loss: 0.102, Train Accuracy: 98.040
100%| | 646/646 [00:05<00:00, 113.21it/s]
Epoch 11/50, Train Loss: 0.086, Train Accuracy: 98.300
100%| | 646/646 [00:05<00:00, 114.71it/s]
Epoch 12/50, Train Loss: 0.073, Train Accuracy: 98.640
100%| | 646/646 [00:05<00:00, 114.77it/s]
Epoch 13/50, Train Loss: 0.062, Train Accuracy: 98.830
100%| | 646/646 [00:05<00:00, 113.76it/s]
Epoch 14/50, Train Loss: 0.053, Train Accuracy: 99.059
100%| | 646/646 [00:05<00:00, 115.01it/s]
Epoch 15/50, Train Loss: 0.046, Train Accuracy: 99.166
100%| | 646/646 [00:05<00:00, 111.47it/s]
Epoch 16/50, Train Loss: 0.040, Train Accuracy: 99.320
100%| | 646/646 [00:05<00:00, 113.32it/s]
Epoch 17/50, Train Loss: 0.034, Train Accuracy: 99.445
100%| | 646/646 [00:05<00:00, 115.18it/s]

Epoch 18/50, Train Loss: 0.030, Train Accuracy: 99.539
100%| | 646/646 [00:05<00:00, 114.23it/s]
Epoch 19/50, Train Loss: 0.026, Train Accuracy: 99.617
100%| | 646/646 [00:05<00:00, 115.16it/s]
Epoch 20/50, Train Loss: 0.022, Train Accuracy: 99.690
100%| | 646/646 [00:05<00:00, 115.33it/s]
Epoch 21/50, Train Loss: 0.019, Train Accuracy: 99.742
100%| | 646/646 [00:05<00:00, 113.11it/s]
Epoch 22/50, Train Loss: 0.017, Train Accuracy: 99.800
100%| | 646/646 [00:05<00:00, 113.14it/s]
Epoch 23/50, Train Loss: 0.014, Train Accuracy: 99.841
100%| | 646/646 [00:05<00:00, 111.57it/s]
Epoch 24/50, Train Loss: 0.013, Train Accuracy: 99.850
100%| | 646/646 [00:05<00:00, 113.96it/s]
Epoch 25/50, Train Loss: 0.011, Train Accuracy: 99.893
100%| | 646/646 [00:05<00:00, 110.74it/s]
Epoch 26/50, Train Loss: 0.009, Train Accuracy: 99.912
100%| | 646/646 [00:05<00:00, 114.75it/s]
Epoch 27/50, Train Loss: 0.008, Train Accuracy: 99.937
100%| | 646/646 [00:05<00:00, 114.50it/s]
Epoch 28/50, Train Loss: 0.007, Train Accuracy: 99.924
100%| | 646/646 [00:05<00:00, 114.44it/s]
Epoch 29/50, Train Loss: 0.008, Train Accuracy: 99.930
100%| | 646/646 [00:05<00:00, 114.40it/s]
Epoch 30/50, Train Loss: 0.005, Train Accuracy: 99.977
100%| | 646/646 [00:05<00:00, 114.58it/s]
Epoch 31/50, Train Loss: 0.004, Train Accuracy: 99.990
100%| | 646/646 [00:05<00:00, 114.88it/s]
Epoch 32/50, Train Loss: 0.004, Train Accuracy: 99.987
100%| | 646/646 [00:05<00:00, 112.13it/s]
Epoch 33/50, Train Loss: 0.004, Train Accuracy: 99.960
100%| | 646/646 [00:05<00:00, 114.13it/s]

Epoch 34/50, Train Loss: 0.007, Train Accuracy: 99.886
100%| | 646/646 [00:05<00:00, 114.38it/s]
Epoch 35/50, Train Loss: 0.003, Train Accuracy: 99.994
100%| | 646/646 [00:05<00:00, 113.79it/s]
Epoch 36/50, Train Loss: 0.002, Train Accuracy: 99.993
100%| | 646/646 [00:05<00:00, 111.97it/s]
Epoch 37/50, Train Loss: 0.002, Train Accuracy: 99.995
100%| | 646/646 [00:05<00:00, 114.04it/s]
Epoch 38/50, Train Loss: 0.002, Train Accuracy: 99.994
100%| | 646/646 [00:05<00:00, 114.61it/s]
Epoch 39/50, Train Loss: 0.004, Train Accuracy: 99.912
100%| | 646/646 [00:05<00:00, 113.84it/s]
Epoch 40/50, Train Loss: 0.006, Train Accuracy: 99.909
100%| | 646/646 [00:05<00:00, 114.92it/s]
Epoch 41/50, Train Loss: 0.002, Train Accuracy: 99.992
100%| | 646/646 [00:05<00:00, 114.95it/s]
Epoch 42/50, Train Loss: 0.001, Train Accuracy: 99.996
100%| | 646/646 [00:05<00:00, 114.45it/s]
Epoch 43/50, Train Loss: 0.001, Train Accuracy: 99.999
100%| | 646/646 [00:05<00:00, 114.83it/s]
Epoch 44/50, Train Loss: 0.001, Train Accuracy: 99.995
100%| | 646/646 [00:05<00:00, 113.87it/s]
Epoch 45/50, Train Loss: 0.001, Train Accuracy: 99.998
100%| | 646/646 [00:05<00:00, 114.26it/s]
Epoch 46/50, Train Loss: 0.007, Train Accuracy: 99.822
100%| | 646/646 [00:05<00:00, 113.14it/s]
Epoch 47/50, Train Loss: 0.003, Train Accuracy: 99.964
100%| | 646/646 [00:06<00:00, 102.72it/s]
Epoch 48/50, Train Loss: 0.001, Train Accuracy: 99.992
100%| | 646/646 [00:06<00:00, 101.91it/s]
Epoch 49/50, Train Loss: 0.001, Train Accuracy: 99.999
100%| | 646/646 [00:05<00:00, 112.27it/s]

Epoch 50/50, Train Loss: 0.001, Train Accuracy: 99.998

Training complete. Model saved to 'train_logs/window_3\model.pth', stats saved to 'train_logs/window_3\train_stats.csv'

100%| | 132/132 [00:01<00:00, 128.91it/s]

c:\Users\Ward\anaconda3\envs\multibench\Lib\site-

packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains classes not in y_true

warnings.warn("y_pred contains classes not in y_true")

Evaluation Results:

Top-1 Accuracy: 94.29%

Top-5 Accuracy: 99.55%

Balanced Accuracy: 82.53%

Classification Report:

		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.96	0.95	0.96
1113				
	SYSTEM_GOODBYE	0.98	0.98	0.98
1331				
	SYSTEM_INFORM	0.97	0.95	0.96
719				
	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.97	0.95	0.96
310				
	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.73	0.65	0.69
17				
	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.91	0.87	0.89
79				
	SYSTEM_NOTIFY_SUCCESS	0.98	0.98	0.98
499				
	SYSTEM_OFFER	0.90	0.94	0.92
806				
	SYSTEM_OFFER_INTENT	0.94	0.98	0.96
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.94	0.94	0.94
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.79	0.88
14				
	SYSTEM_REQUEST	0.98	0.98	0.98
1961				
	SYSTEM_REQ_MORE	0.98	0.97	0.97
585				
	USER_AFFIRM	0.96	0.96	0.96
572				

	USER_AFFIRM_INTENT	0.92	0.94	0.93
139				
	USER_AFFIRM_INTENT USER_INFORM	0.97	0.86	0.91
103				
	USER_INFORM	0.99	0.98	0.98
1961				
	USER_INFORM_INTENT	0.97	0.89	0.93
658				
	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM_INTENT USER_SELECT	0.77	0.87	0.81
167				
	USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
12				
	USER_INFORM USER_INFORM_INTENT	0.92	0.96	0.94
772				
	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.78	0.74	0.76
117				
	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
14				
	USER_INFORM USER_NEGATE	0.92	0.96	0.94
216				
	USER_INFORM USER_REQUEST_ALTS	0.86	0.86	0.86
184				
	USER_NEGATE	0.80	0.57	0.67
7				
	USER_NEGATE_INTENT	0.66	0.77	0.71
69				
	USER_NEGATE_INTENT USER_GOODBYE	0.69	0.75	0.72
72				
	USER_NEGATE USER_GOODBYE	0.75	0.33	0.46
9				
	USER_NEGATE USER_THANK_YOU	0.97	0.99	0.98
587				
	USER_REQUEST	0.95	0.95	0.95
719				
	USER_REQUEST_ALTS	0.92	0.93	0.92
152				
	USER_REQUEST USER_AFFIRM	0.98	0.97	0.97
340				
	USER_SELECT	0.86	0.93	0.89
515				
	USER_SELECT USER_GOODBYE	0.86	0.71	0.78
279				
	USER_THANK_YOU	0.83	0.84	0.84
377				
	USER_THANK_YOU USER_GOODBYE	0.77	0.84	0.80
384				

	accuracy			0.94
16850				
	macro avg	0.82	0.80	0.81
16850				
	weighted avg	0.94	0.94	0.94
16850				

=== Running training for window size: 4 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```

    self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:25<00:00, 211.99it/s]
100%|      | 1331/1331 [00:08<00:00, 155.41it/s]
100%|      | 646/646 [00:09<00:00, 64.84it/s]

Epoch 1/50, Train Loss: 2.160, Train Accuracy: 35.546
100%|      | 646/646 [00:07<00:00, 90.42it/s]

Epoch 2/50, Train Loss: 1.118, Train Accuracy: 66.195
100%|      | 646/646 [00:06<00:00, 93.25it/s]

Epoch 3/50, Train Loss: 0.753, Train Accuracy: 80.202
100%|      | 646/646 [00:06<00:00, 94.29it/s]

Epoch 4/50, Train Loss: 0.553, Train Accuracy: 86.054
100%|      | 646/646 [00:06<00:00, 95.13it/s]

Epoch 5/50, Train Loss: 0.427, Train Accuracy: 89.790
100%|      | 646/646 [00:06<00:00, 94.23it/s]

Epoch 6/50, Train Loss: 0.337, Train Accuracy: 92.459
100%|      | 646/646 [00:06<00:00, 93.92it/s]

Epoch 7/50, Train Loss: 0.272, Train Accuracy: 94.160

```

100%| | 646/646 [00:06<00:00, 93.38it/s]
Epoch 8/50, Train Loss: 0.222, Train Accuracy: 95.396
100%| | 646/646 [00:06<00:00, 93.35it/s]
Epoch 9/50, Train Loss: 0.183, Train Accuracy: 96.268
100%| | 646/646 [00:06<00:00, 94.27it/s]
Epoch 10/50, Train Loss: 0.153, Train Accuracy: 96.958
100%| | 646/646 [00:06<00:00, 95.61it/s]
Epoch 11/50, Train Loss: 0.129, Train Accuracy: 97.495
100%| | 646/646 [00:06<00:00, 94.78it/s]
Epoch 12/50, Train Loss: 0.109, Train Accuracy: 97.968
100%| | 646/646 [00:06<00:00, 94.32it/s]
Epoch 13/50, Train Loss: 0.093, Train Accuracy: 98.281
100%| | 646/646 [00:06<00:00, 95.05it/s]
Epoch 14/50, Train Loss: 0.081, Train Accuracy: 98.608
100%| | 646/646 [00:06<00:00, 94.51it/s]
Epoch 15/50, Train Loss: 0.069, Train Accuracy: 98.884
100%| | 646/646 [00:06<00:00, 94.59it/s]
Epoch 16/50, Train Loss: 0.059, Train Accuracy: 99.060
100%| | 646/646 [00:06<00:00, 92.59it/s]
Epoch 17/50, Train Loss: 0.051, Train Accuracy: 99.284
100%| | 646/646 [00:06<00:00, 95.17it/s]
Epoch 18/50, Train Loss: 0.044, Train Accuracy: 99.410
100%| | 646/646 [00:06<00:00, 95.63it/s]
Epoch 19/50, Train Loss: 0.038, Train Accuracy: 99.508
100%| | 646/646 [00:06<00:00, 94.96it/s]
Epoch 20/50, Train Loss: 0.033, Train Accuracy: 99.610
100%| | 646/646 [00:06<00:00, 95.17it/s]
Epoch 21/50, Train Loss: 0.029, Train Accuracy: 99.701
100%| | 646/646 [00:06<00:00, 94.42it/s]
Epoch 22/50, Train Loss: 0.025, Train Accuracy: 99.742
100%| | 646/646 [00:06<00:00, 95.34it/s]
Epoch 23/50, Train Loss: 0.022, Train Accuracy: 99.795

100%| | 646/646 [00:06<00:00, 96.05it/s]
Epoch 24/50, Train Loss: 0.019, Train Accuracy: 99.838
100%| | 646/646 [00:06<00:00, 95.08it/s]
Epoch 25/50, Train Loss: 0.016, Train Accuracy: 99.889
100%| | 646/646 [00:06<00:00, 92.66it/s]
Epoch 26/50, Train Loss: 0.015, Train Accuracy: 99.886
100%| | 646/646 [00:06<00:00, 95.85it/s]
Epoch 27/50, Train Loss: 0.012, Train Accuracy: 99.931
100%| | 646/646 [00:06<00:00, 95.84it/s]
Epoch 28/50, Train Loss: 0.012, Train Accuracy: 99.896
100%| | 646/646 [00:06<00:00, 95.12it/s]
Epoch 29/50, Train Loss: 0.010, Train Accuracy: 99.935
100%| | 646/646 [00:06<00:00, 95.75it/s]
Epoch 30/50, Train Loss: 0.008, Train Accuracy: 99.964
100%| | 646/646 [00:06<00:00, 95.03it/s]
Epoch 31/50, Train Loss: 0.007, Train Accuracy: 99.966
100%| | 646/646 [00:06<00:00, 95.84it/s]
Epoch 32/50, Train Loss: 0.011, Train Accuracy: 99.870
100%| | 646/646 [00:06<00:00, 95.42it/s]
Epoch 33/50, Train Loss: 0.006, Train Accuracy: 99.982
100%| | 646/646 [00:06<00:00, 95.64it/s]
Epoch 34/50, Train Loss: 0.005, Train Accuracy: 99.989
100%| | 646/646 [00:06<00:00, 92.57it/s]
Epoch 35/50, Train Loss: 0.004, Train Accuracy: 99.990
100%| | 646/646 [00:06<00:00, 95.07it/s]
Epoch 36/50, Train Loss: 0.008, Train Accuracy: 99.884
100%| | 646/646 [00:06<00:00, 95.56it/s]
Epoch 37/50, Train Loss: 0.005, Train Accuracy: 99.970
100%| | 646/646 [00:06<00:00, 95.19it/s]
Epoch 38/50, Train Loss: 0.006, Train Accuracy: 99.923
100%| | 646/646 [00:06<00:00, 95.79it/s]
Epoch 39/50, Train Loss: 0.003, Train Accuracy: 99.993

```

100%|      | 646/646 [00:06<00:00, 95.74it/s]
Epoch 40/50, Train Loss: 0.002, Train Accuracy: 99.996
100%|      | 646/646 [00:06<00:00, 96.00it/s]
Epoch 41/50, Train Loss: 0.002, Train Accuracy: 99.995
100%|      | 646/646 [00:06<00:00, 95.55it/s]
Epoch 42/50, Train Loss: 0.002, Train Accuracy: 99.994
100%|      | 646/646 [00:06<00:00, 95.86it/s]
Epoch 43/50, Train Loss: 0.009, Train Accuracy: 99.811
100%|      | 646/646 [00:06<00:00, 92.69it/s]
Epoch 44/50, Train Loss: 0.003, Train Accuracy: 99.990
100%|      | 646/646 [00:06<00:00, 94.95it/s]
Epoch 45/50, Train Loss: 0.002, Train Accuracy: 99.998
100%|      | 646/646 [00:06<00:00, 95.65it/s]
Epoch 46/50, Train Loss: 0.002, Train Accuracy: 99.999
100%|      | 646/646 [00:06<00:00, 95.67it/s]
Epoch 47/50, Train Loss: 0.001, Train Accuracy: 99.999
100%|      | 646/646 [00:06<00:00, 96.50it/s]
Epoch 48/50, Train Loss: 0.001, Train Accuracy: 99.999
100%|      | 646/646 [00:06<00:00, 95.95it/s]
Epoch 49/50, Train Loss: 0.001, Train Accuracy: 99.998
100%|      | 646/646 [00:06<00:00, 96.06it/s]
Epoch 50/50, Train Loss: 0.009, Train Accuracy: 99.766

```

```

Training complete. Model saved to 'train_logs/window_4\model.pth', stats saved
to 'train_logs/window_4\train_stats.csv'

```

```

100%|      | 132/132 [00:01<00:00, 110.60it/s]
c:\Users\Ward\anaconda3\envs\multibench\Lib\site-
packages\sklearn\metrics\_classification.py:2480: UserWarning: y_pred contains
classes not in y_true
  warnings.warn("y_pred contains classes not in y_true")

```

```

Evaluation Results:
Top-1 Accuracy: 93.09%
Top-5 Accuracy: 99.35%
Balanced Accuracy: 80.16%

```

Classification Report:

		precision	recall	f1-score
support				
1113	SYSTEM_CONFIRM	0.98	0.88	0.93
1331	SYSTEM_GOODBYE	0.99	0.98	0.99
719	SYSTEM_INFORM	0.97	0.90	0.93
310	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.91	0.95	0.93
17	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.92	0.65	0.76
79	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.88	0.87	0.88
499	SYSTEM_NOTIFY_SUCCESS	0.98	0.98	0.98
806	SYSTEM_OFFER	0.84	0.98	0.91
383	SYSTEM_OFFER_INTENT	0.95	0.96	0.96
608	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.95	0.94	0.95
14	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.71	0.83
1961	SYSTEM_REQUEST	0.97	0.98	0.97
585	SYSTEM_REQ_MORE	0.93	0.98	0.96
572	USER_AFFIRM	0.97	0.92	0.94
139	USER_AFFIRM_INTENT	0.90	0.87	0.89
103	USER_AFFIRM_INTENT USER_INFORM	0.92	0.84	0.88
1961	USER_INFORM	0.99	0.97	0.98
658	USER_INFORM_INTENT	0.94	0.93	0.94
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.78	0.87	0.83
12	USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
772	USER_INFORM USER_INFORM_INTENT	0.94	0.95	0.94

0	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.75	0.72	0.73
14	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
216	USER_INFORM USER_NEGATE	0.94	0.83	0.88
184	USER_INFORM USER_REQUEST_ALTS	0.79	0.88	0.83
7	USER_NEGATE	0.57	0.57	0.57
69	USER_NEGATE_INTENT	0.65	0.80	0.71
72	USER_NEGATE_INTENT USER_GOODBYE	0.70	0.62	0.66
9	USER_NEGATE USER_GOODBYE	1.00	0.11	0.20
587	USER_NEGATE USER_THANK_YOU	0.96	0.98	0.97
719	USER_REQUEST	0.93	0.94	0.94
152	USER_REQUEST_ALTS	0.88	0.90	0.89
340	USER_REQUEST USER_AFFIRM	0.95	0.94	0.94
515	USER_SELECT	0.79	0.95	0.86
279	USER_SELECT USER_GOODBYE	0.87	0.66	0.75
377	USER_THANK_YOU	0.81	0.80	0.81
384	USER_THANK_YOU USER_GOODBYE	0.74	0.84	0.79
16850	accuracy			0.93
16850	macro avg	0.80	0.76	0.77
16850	weighted avg	0.93	0.93	0.93

=== Running training for window size: 5 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning:
You are using `torch.load` with `weights_only=False` (the current default
value), which uses the default pickle module implicitly. It is possible to

construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for ``weights_only`` will be flipped to ``True``. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via ``torch.serialization.add_safe_globals``. We recommend you start setting ``weights_only=True`` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:56<00:00, 95.69it/s]
100%|      | 1331/1331 [00:03<00:00, 426.95it/s]
100%|      | 646/646 [00:31<00:00, 20.53it/s]

Epoch 1/50, Train Loss: 2.330, Train Accuracy: 32.535
100%|      | 646/646 [00:08<00:00, 79.00it/s]

Epoch 2/50, Train Loss: 1.398, Train Accuracy: 56.592
100%|      | 646/646 [00:07<00:00, 81.54it/s]

Epoch 3/50, Train Loss: 1.012, Train Accuracy: 71.383
100%|      | 646/646 [00:07<00:00, 82.50it/s]

Epoch 4/50, Train Loss: 0.761, Train Accuracy: 80.474
100%|      | 646/646 [00:07<00:00, 82.74it/s]

Epoch 5/50, Train Loss: 0.596, Train Accuracy: 85.652
100%|      | 646/646 [00:07<00:00, 82.92it/s]

Epoch 6/50, Train Loss: 0.478, Train Accuracy: 89.129
100%|      | 646/646 [00:07<00:00, 82.87it/s]

Epoch 7/50, Train Loss: 0.394, Train Accuracy: 91.287
100%|      | 646/646 [00:07<00:00, 82.22it/s]

Epoch 8/50, Train Loss: 0.330, Train Accuracy: 92.966
100%|      | 646/646 [00:07<00:00, 83.09it/s]

Epoch 9/50, Train Loss: 0.281, Train Accuracy: 94.109
100%|      | 646/646 [00:07<00:00, 80.78it/s]

Epoch 10/50, Train Loss: 0.238, Train Accuracy: 95.235
100%|      | 646/646 [00:07<00:00, 82.98it/s]

Epoch 11/50, Train Loss: 0.204, Train Accuracy: 96.085
100%|      | 646/646 [00:07<00:00, 82.14it/s]
```

Epoch 12/50, Train Loss: 0.176, Train Accuracy: 96.772
100%| | 646/646 [00:07<00:00, 82.90it/s]
Epoch 13/50, Train Loss: 0.156, Train Accuracy: 97.243
100%| | 646/646 [00:07<00:00, 82.50it/s]
Epoch 14/50, Train Loss: 0.133, Train Accuracy: 97.772
100%| | 646/646 [00:07<00:00, 83.20it/s]
Epoch 15/50, Train Loss: 0.115, Train Accuracy: 98.097
100%| | 646/646 [00:07<00:00, 82.89it/s]
Epoch 16/50, Train Loss: 0.101, Train Accuracy: 98.414
100%| | 646/646 [00:08<00:00, 80.10it/s]
Epoch 17/50, Train Loss: 0.089, Train Accuracy: 98.621
100%| | 646/646 [00:09<00:00, 67.67it/s]
Epoch 18/50, Train Loss: 0.077, Train Accuracy: 98.844
100%| | 646/646 [00:10<00:00, 59.29it/s]
Epoch 19/50, Train Loss: 0.068, Train Accuracy: 99.033
100%| | 646/646 [00:08<00:00, 76.26it/s]
Epoch 20/50, Train Loss: 0.060, Train Accuracy: 99.196
100%| | 646/646 [00:08<00:00, 76.43it/s]
Epoch 21/50, Train Loss: 0.052, Train Accuracy: 99.338
100%| | 646/646 [00:10<00:00, 62.91it/s]
Epoch 22/50, Train Loss: 0.046, Train Accuracy: 99.471
100%| | 646/646 [00:08<00:00, 79.95it/s]
Epoch 23/50, Train Loss: 0.042, Train Accuracy: 99.519
100%| | 646/646 [00:08<00:00, 78.30it/s]
Epoch 24/50, Train Loss: 0.035, Train Accuracy: 99.648
100%| | 646/646 [00:07<00:00, 81.13it/s]
Epoch 25/50, Train Loss: 0.031, Train Accuracy: 99.714
100%| | 646/646 [00:07<00:00, 82.49it/s]
Epoch 26/50, Train Loss: 0.028, Train Accuracy: 99.751
100%| | 646/646 [00:07<00:00, 82.69it/s]
Epoch 27/50, Train Loss: 0.031, Train Accuracy: 99.661
100%| | 646/646 [00:07<00:00, 83.02it/s]

Epoch 28/50, Train Loss: 0.023, Train Accuracy: 99.820
100%| | 646/646 [00:07<00:00, 84.59it/s]
Epoch 29/50, Train Loss: 0.020, Train Accuracy: 99.855
100%| | 646/646 [00:07<00:00, 83.72it/s]
Epoch 30/50, Train Loss: 0.017, Train Accuracy: 99.892
100%| | 646/646 [00:07<00:00, 84.16it/s]
Epoch 31/50, Train Loss: 0.016, Train Accuracy: 99.893
100%| | 646/646 [00:07<00:00, 81.11it/s]
Epoch 32/50, Train Loss: 0.015, Train Accuracy: 99.908
100%| | 646/646 [00:08<00:00, 79.58it/s]
Epoch 33/50, Train Loss: 0.014, Train Accuracy: 99.906
100%| | 646/646 [00:07<00:00, 81.55it/s]
Epoch 34/50, Train Loss: 0.011, Train Accuracy: 99.952
100%| | 646/646 [00:07<00:00, 82.32it/s]
Epoch 35/50, Train Loss: 0.015, Train Accuracy: 99.815
100%| | 646/646 [00:07<00:00, 82.70it/s]
Epoch 36/50, Train Loss: 0.010, Train Accuracy: 99.954
100%| | 646/646 [00:07<00:00, 82.79it/s]
Epoch 37/50, Train Loss: 0.008, Train Accuracy: 99.966
100%| | 646/646 [00:07<00:00, 81.66it/s]
Epoch 38/50, Train Loss: 0.007, Train Accuracy: 99.969
100%| | 646/646 [00:07<00:00, 82.75it/s]
Epoch 39/50, Train Loss: 0.012, Train Accuracy: 99.810
100%| | 646/646 [00:07<00:00, 82.49it/s]
Epoch 40/50, Train Loss: 0.013, Train Accuracy: 99.856
100%| | 646/646 [00:07<00:00, 82.94it/s]
Epoch 41/50, Train Loss: 0.006, Train Accuracy: 99.985
100%| | 646/646 [00:07<00:00, 82.85it/s]
Epoch 42/50, Train Loss: 0.005, Train Accuracy: 99.983
100%| | 646/646 [00:07<00:00, 82.55it/s]
Epoch 43/50, Train Loss: 0.004, Train Accuracy: 99.985
100%| | 646/646 [00:07<00:00, 81.87it/s]

Epoch 44/50, Train Loss: 0.004, Train Accuracy: 99.990
 100%| | 646/646 [00:07<00:00, 82.41it/s]
 Epoch 45/50, Train Loss: 0.004, Train Accuracy: 99.989
 100%| | 646/646 [00:07<00:00, 82.96it/s]
 Epoch 46/50, Train Loss: 0.016, Train Accuracy: 99.672
 100%| | 646/646 [00:07<00:00, 81.88it/s]
 Epoch 47/50, Train Loss: 0.008, Train Accuracy: 99.884
 100%| | 646/646 [00:08<00:00, 75.86it/s]
 Epoch 48/50, Train Loss: 0.004, Train Accuracy: 99.987
 100%| | 646/646 [00:08<00:00, 73.29it/s]
 Epoch 49/50, Train Loss: 0.003, Train Accuracy: 99.994
 100%| | 646/646 [00:08<00:00, 80.49it/s]
 Epoch 50/50, Train Loss: 0.003, Train Accuracy: 99.994

Training complete. Model saved to 'train_logs/window_5\model.pth', stats saved to 'train_logs/window_5\train_stats.csv'

100%| | 132/132 [00:07<00:00, 17.03it/s]
 c:\Users\Ward\anaconda3\envs\multibench\Lib\site-packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains classes not in y_true
 warnings.warn("y_pred contains classes not in y_true")

Evaluation Results:
 Top-1 Accuracy: 92.61%
 Top-5 Accuracy: 99.31%
 Balanced Accuracy: 80.24%

Classification Report:

		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.94	0.91	0.92
1113	SYSTEM_GOODBYE	0.97	0.97	0.97
1331	SYSTEM_INFORM	0.95	0.94	0.95
719	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.97	0.95	0.96
310	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.65	0.79
17				

79	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.94	0.81	0.87
499	SYSTEM_NOTIFY_SUCCESS	0.98	0.96	0.97
806	SYSTEM_OFFER	0.89	0.92	0.91
383	SYSTEM_OFFER_INTENT	0.90	0.98	0.94
608	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.89	0.95	0.92
14	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.71	0.83
1961	SYSTEM_REQUEST	0.98	0.96	0.97
585	SYSTEM_REQ_MORE	0.92	0.96	0.94
572	USER_AFFIRM	0.95	0.91	0.93
139	USER_AFFIRM_INTENT	0.87	0.88	0.87
103	USER_AFFIRM_INTENT USER_INFORM	0.94	0.88	0.91
1961	USER_INFORM	0.99	0.96	0.98
658	USER_INFORM_INTENT	0.95	0.91	0.93
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.73	0.85	0.79
12	USER_INFORM_INTENT USER_THANK_YOU	0.20	0.17	0.18
772	USER_INFORM USER_INFORM_INTENT	0.94	0.95	0.94
0	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.76	0.74	0.75
14	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
216	USER_INFORM USER_NEGATE	0.93	0.87	0.89
184	USER_INFORM USER_REQUEST_ALTS	0.84	0.84	0.84
7	USER_NEGATE	0.38	0.43	0.40
69	USER_NEGATE_INTENT	0.56	0.83	0.67

72	USER_NEGATE_INTENT USER_GOODBYE	0.64	0.54	0.59
9	USER_NEGATE USER_GOODBYE	0.50	0.22	0.31
587	USER_NEGATE USER_THANK_YOU	0.95	0.99	0.97
719	USER_REQUEST	0.94	0.94	0.94
152	USER_REQUEST_ALTS	0.89	0.94	0.91
340	USER_REQUEST USER_AFFIRM	0.95	0.94	0.95
515	USER_SELECT	0.81	0.93	0.86
279	USER_SELECT USER_GOODBYE	0.80	0.71	0.75
377	USER_THANK_YOU	0.78	0.79	0.79
384	USER_THANK_YOU USER_GOODBYE	0.75	0.79	0.77
16850	accuracy			0.93
16850	macro avg	0.78	0.76	0.77
16850	weighted avg	0.93	0.93	0.93

=== Running training for window size: 6 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:48<00:00, 110.83it/s]
100%|      | 1331/1331 [00:07<00:00, 175.15it/s]
```

100%| | 646/646 [01:29<00:00, 7.25it/s]
Epoch 1/50, Train Loss: 2.451, Train Accuracy: 28.739
100%| | 646/646 [00:16<00:00, 38.26it/s]
Epoch 2/50, Train Loss: 1.543, Train Accuracy: 53.210
100%| | 646/646 [00:10<00:00, 62.77it/s]
Epoch 3/50, Train Loss: 1.180, Train Accuracy: 65.443
100%| | 646/646 [00:10<00:00, 62.72it/s]
Epoch 4/50, Train Loss: 0.934, Train Accuracy: 74.645
100%| | 646/646 [00:10<00:00, 63.59it/s]
Epoch 5/50, Train Loss: 0.740, Train Accuracy: 81.104
100%| | 646/646 [00:10<00:00, 63.55it/s]
Epoch 6/50, Train Loss: 0.607, Train Accuracy: 84.488
100%| | 646/646 [00:09<00:00, 65.39it/s]
Epoch 7/50, Train Loss: 0.514, Train Accuracy: 86.896
100%| | 646/646 [00:09<00:00, 65.66it/s]
Epoch 8/50, Train Loss: 0.441, Train Accuracy: 88.839
100%| | 646/646 [00:09<00:00, 66.05it/s]
Epoch 9/50, Train Loss: 0.382, Train Accuracy: 90.559
100%| | 646/646 [00:09<00:00, 65.95it/s]
Epoch 10/50, Train Loss: 0.331, Train Accuracy: 92.328
100%| | 646/646 [00:09<00:00, 65.40it/s]
Epoch 11/50, Train Loss: 0.285, Train Accuracy: 93.732
100%| | 646/646 [00:10<00:00, 64.01it/s]
Epoch 12/50, Train Loss: 0.249, Train Accuracy: 94.747
100%| | 646/646 [00:10<00:00, 62.67it/s]
Epoch 13/50, Train Loss: 0.222, Train Accuracy: 95.470
100%| | 646/646 [00:09<00:00, 64.68it/s]
Epoch 14/50, Train Loss: 0.192, Train Accuracy: 96.289
100%| | 646/646 [00:09<00:00, 66.39it/s]
Epoch 15/50, Train Loss: 0.169, Train Accuracy: 96.852
100%| | 646/646 [00:09<00:00, 66.12it/s]
Epoch 16/50, Train Loss: 0.149, Train Accuracy: 97.285

100%| | 646/646 [00:10<00:00, 64.19it/s]
Epoch 17/50, Train Loss: 0.133, Train Accuracy: 97.702
100%| | 646/646 [00:09<00:00, 66.48it/s]
Epoch 18/50, Train Loss: 0.118, Train Accuracy: 97.992
100%| | 646/646 [00:09<00:00, 65.92it/s]
Epoch 19/50, Train Loss: 0.104, Train Accuracy: 98.261
100%| | 646/646 [00:09<00:00, 66.50it/s]
Epoch 20/50, Train Loss: 0.092, Train Accuracy: 98.546
100%| | 646/646 [00:09<00:00, 66.70it/s]
Epoch 21/50, Train Loss: 0.083, Train Accuracy: 98.740
100%| | 646/646 [00:09<00:00, 66.58it/s]
Epoch 22/50, Train Loss: 0.074, Train Accuracy: 98.897
100%| | 646/646 [00:09<00:00, 67.05it/s]
Epoch 23/50, Train Loss: 0.066, Train Accuracy: 99.063
100%| | 646/646 [00:09<00:00, 65.03it/s]
Epoch 24/50, Train Loss: 0.059, Train Accuracy: 99.206
100%| | 646/646 [00:09<00:00, 66.45it/s]
Epoch 25/50, Train Loss: 0.054, Train Accuracy: 99.281
100%| | 646/646 [00:09<00:00, 65.85it/s]
Epoch 26/50, Train Loss: 0.046, Train Accuracy: 99.479
100%| | 646/646 [00:09<00:00, 66.11it/s]
Epoch 27/50, Train Loss: 0.044, Train Accuracy: 99.493
100%| | 646/646 [00:09<00:00, 71.03it/s]
Epoch 28/50, Train Loss: 0.038, Train Accuracy: 99.616
100%| | 646/646 [00:09<00:00, 66.70it/s]
Epoch 29/50, Train Loss: 0.034, Train Accuracy: 99.673
100%| | 646/646 [00:09<00:00, 65.36it/s]
Epoch 30/50, Train Loss: 0.029, Train Accuracy: 99.770
100%| | 646/646 [00:09<00:00, 66.24it/s]
Epoch 31/50, Train Loss: 0.026, Train Accuracy: 99.812
100%| | 646/646 [00:09<00:00, 66.26it/s]
Epoch 32/50, Train Loss: 0.023, Train Accuracy: 99.851

100%| | 646/646 [00:09<00:00, 65.58it/s]
Epoch 33/50, Train Loss: 0.028, Train Accuracy: 99.697
100%| | 646/646 [00:09<00:00, 66.21it/s]
Epoch 34/50, Train Loss: 0.020, Train Accuracy: 99.876
100%| | 646/646 [00:09<00:00, 69.45it/s]
Epoch 35/50, Train Loss: 0.016, Train Accuracy: 99.933
100%| | 646/646 [00:09<00:00, 64.84it/s]
Epoch 36/50, Train Loss: 0.014, Train Accuracy: 99.944
100%| | 646/646 [00:09<00:00, 65.74it/s]
Epoch 37/50, Train Loss: 0.020, Train Accuracy: 99.793
100%| | 646/646 [00:09<00:00, 65.11it/s]
Epoch 38/50, Train Loss: 0.013, Train Accuracy: 99.955
100%| | 646/646 [00:09<00:00, 64.73it/s]
Epoch 39/50, Train Loss: 0.011, Train Accuracy: 99.969
100%| | 646/646 [00:09<00:00, 65.92it/s]
Epoch 40/50, Train Loss: 0.011, Train Accuracy: 99.948
100%| | 646/646 [00:09<00:00, 65.15it/s]
Epoch 41/50, Train Loss: 0.014, Train Accuracy: 99.893
100%| | 646/646 [00:10<00:00, 63.25it/s]
Epoch 42/50, Train Loss: 0.011, Train Accuracy: 99.930
100%| | 646/646 [00:09<00:00, 65.53it/s]
Epoch 43/50, Train Loss: 0.008, Train Accuracy: 99.979
100%| | 646/646 [00:10<00:00, 62.29it/s]
Epoch 44/50, Train Loss: 0.008, Train Accuracy: 99.967
100%| | 646/646 [00:09<00:00, 66.49it/s]
Epoch 45/50, Train Loss: 0.006, Train Accuracy: 99.989
100%| | 646/646 [00:09<00:00, 66.22it/s]
Epoch 46/50, Train Loss: 0.005, Train Accuracy: 99.989
100%| | 646/646 [00:09<00:00, 65.94it/s]
Epoch 47/50, Train Loss: 0.011, Train Accuracy: 99.797
100%| | 646/646 [00:10<00:00, 64.54it/s]
Epoch 48/50, Train Loss: 0.009, Train Accuracy: 99.946

100%| | 646/646 [00:09<00:00, 66.14it/s]

Epoch 49/50, Train Loss: 0.009, Train Accuracy: 99.910

100%| | 646/646 [00:10<00:00, 64.03it/s]

Epoch 50/50, Train Loss: 0.006, Train Accuracy: 99.972

Training complete. Model saved to 'train_logs/window_6\model.pth', stats saved to 'train_logs/window_6\train_stats.csv'

100%| | 132/132 [00:16<00:00, 8.21it/s]

c:\Users\Ward\anaconda3\envs\multibench\Lib\site-

packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains classes not in y_true

warnings.warn("y_pred contains classes not in y_true")

Evaluation Results:

Top-1 Accuracy: 90.77%

Top-5 Accuracy: 98.93%

Balanced Accuracy: 76.81%

Classification Report:

		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.92	0.88	0.90
1113				
	SYSTEM_GOODBYE	0.98	0.96	0.97
1331				
	SYSTEM_INFORM	0.97	0.93	0.95
719				
	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.95	0.92	0.93
310				
	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.71	0.29	0.42
17				
	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.86	0.72	0.79
79				
	SYSTEM_NOTIFY_SUCCESS	0.93	0.96	0.95
499				
	SYSTEM_OFFER	0.85	0.93	0.89
806				
	SYSTEM_OFFER_INTENT	0.91	0.98	0.94
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.94	0.91	0.92
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.61	0.79	0.69
14				
	SYSTEM_REQUEST	0.96	0.94	0.95

1961				
	SYSTEM_REQ_MORE	0.92	0.96	0.94
585				
	USER_AFFIRM	0.91	0.91	0.91
572				
	USER_AFFIRM_INTENT	0.79	0.87	0.83
139				
	USER_AFFIRM_INTENT USER_INFORM	0.83	0.75	0.79
103				
	USER_INFORM	0.98	0.95	0.97
1961				
	USER_INFORM_INTENT	0.95	0.90	0.93
658				
	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM_INTENT USER_SELECT	0.66	0.72	0.69
167				
	USER_INFORM_INTENT USER_THANK_YOU	0.33	0.42	0.37
12				
	USER_INFORM USER_INFORM_INTENT	0.91	0.95	0.93
772				
	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.53	0.62	0.57
117				
	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
14				
	USER_INFORM USER_NEGATE	0.82	0.82	0.82
216				
	USER_INFORM USER_REQUEST_ALTS	0.71	0.79	0.75
184				
	USER_NEGATE	1.00	0.43	0.60
7				
	USER_NEGATE_INTENT	0.62	0.74	0.68
69				
	USER_NEGATE_INTENT USER_GOODBYE	0.68	0.56	0.61
72				
	USER_NEGATE USER_GOODBYE	0.00	0.00	0.00
9				
	USER_NEGATE USER_THANK_YOU	0.97	0.98	0.97
587				
	USER_REQUEST	0.96	0.89	0.92
719				
	USER_REQUEST_ALTS	0.82	0.92	0.87
152				
	USER_REQUEST USER_AFFIRM	0.92	0.92	0.92
340				
	USER_SELECT	0.80	0.92	0.86

515				
	USER_SELECT USER_GOODBYE	0.81	0.60	0.69
279				
	USER_THANK_YOU	0.80	0.73	0.76
377				
	USER_THANK_YOU USER_GOODBYE	0.67	0.85	0.75
384				
	accuracy			0.91
16850				
	macro avg	0.74	0.73	0.73
16850				
	weighted avg	0.91	0.91	0.91
16850				

=== Running training for window size: 7 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```

    self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [01:15<00:00, 71.48it/s]
100%|      | 1331/1331 [00:13<00:00, 99.69it/s]
100%|      | 646/646 [02:36<00:00,  4.13it/s]

Epoch 1/50, Train Loss: 2.508, Train Accuracy: 25.890
100%|      | 646/646 [00:39<00:00, 16.33it/s]

Epoch 2/50, Train Loss: 1.789, Train Accuracy: 44.786
100%|      | 646/646 [00:14<00:00, 44.84it/s]

Epoch 3/50, Train Loss: 1.483, Train Accuracy: 53.200
100%|      | 646/646 [00:18<00:00, 35.13it/s]

Epoch 4/50, Train Loss: 1.228, Train Accuracy: 62.159
100%|      | 646/646 [00:12<00:00, 53.25it/s]
```

Epoch 5/50, Train Loss: 1.015, Train Accuracy: 70.966
100%| | 646/646 [00:14<00:00, 46.04it/s]
Epoch 6/50, Train Loss: 0.847, Train Accuracy: 77.447
100%| | 646/646 [00:20<00:00, 31.86it/s]
Epoch 7/50, Train Loss: 0.719, Train Accuracy: 81.631
100%| | 646/646 [00:11<00:00, 56.26it/s]
Epoch 8/50, Train Loss: 0.615, Train Accuracy: 84.706
100%| | 646/646 [00:15<00:00, 42.98it/s]
Epoch 9/50, Train Loss: 0.516, Train Accuracy: 87.663
100%| | 646/646 [00:14<00:00, 44.84it/s]
Epoch 10/50, Train Loss: 0.443, Train Accuracy: 89.657
100%| | 646/646 [00:13<00:00, 46.38it/s]
Epoch 11/50, Train Loss: 0.387, Train Accuracy: 91.267
100%| | 646/646 [00:12<00:00, 51.19it/s]
Epoch 12/50, Train Loss: 0.343, Train Accuracy: 92.464
100%| | 646/646 [00:21<00:00, 29.91it/s]
Epoch 13/50, Train Loss: 0.302, Train Accuracy: 93.656
100%| | 646/646 [00:15<00:00, 42.57it/s]
Epoch 14/50, Train Loss: 0.265, Train Accuracy: 94.602
100%| | 646/646 [00:23<00:00, 27.70it/s]
Epoch 15/50, Train Loss: 0.234, Train Accuracy: 95.372
100%| | 646/646 [00:12<00:00, 51.23it/s]
Epoch 16/50, Train Loss: 0.210, Train Accuracy: 95.952
100%| | 646/646 [00:11<00:00, 55.54it/s]
Epoch 17/50, Train Loss: 0.187, Train Accuracy: 96.503
100%| | 646/646 [00:11<00:00, 55.76it/s]
Epoch 18/50, Train Loss: 0.166, Train Accuracy: 96.989
100%| | 646/646 [00:11<00:00, 54.82it/s]
Epoch 19/50, Train Loss: 0.150, Train Accuracy: 97.300
100%| | 646/646 [00:14<00:00, 43.92it/s]
Epoch 20/50, Train Loss: 0.134, Train Accuracy: 97.689
100%| | 646/646 [00:11<00:00, 54.52it/s]

Epoch 21/50, Train Loss: 0.120, Train Accuracy: 97.972
100%| | 646/646 [00:11<00:00, 54.76it/s]
Epoch 22/50, Train Loss: 0.108, Train Accuracy: 98.242
100%| | 646/646 [00:11<00:00, 54.93it/s]
Epoch 23/50, Train Loss: 0.097, Train Accuracy: 98.558
100%| | 646/646 [00:11<00:00, 55.95it/s]
Epoch 24/50, Train Loss: 0.088, Train Accuracy: 98.687
100%| | 646/646 [00:12<00:00, 52.95it/s]
Epoch 25/50, Train Loss: 0.081, Train Accuracy: 98.850
100%| | 646/646 [00:12<00:00, 50.78it/s]
Epoch 26/50, Train Loss: 0.071, Train Accuracy: 99.074
100%| | 646/646 [00:11<00:00, 55.30it/s]
Epoch 27/50, Train Loss: 0.064, Train Accuracy: 99.225
100%| | 646/646 [00:17<00:00, 37.63it/s]
Epoch 28/50, Train Loss: 0.058, Train Accuracy: 99.340
100%| | 646/646 [00:15<00:00, 41.59it/s]
Epoch 29/50, Train Loss: 0.052, Train Accuracy: 99.456
100%| | 646/646 [00:19<00:00, 32.39it/s]
Epoch 30/50, Train Loss: 0.047, Train Accuracy: 99.548
100%| | 646/646 [00:25<00:00, 25.43it/s]
Epoch 31/50, Train Loss: 0.045, Train Accuracy: 99.583
100%| | 646/646 [00:11<00:00, 57.73it/s]
Epoch 32/50, Train Loss: 0.038, Train Accuracy: 99.703
100%| | 646/646 [00:11<00:00, 57.70it/s]
Epoch 33/50, Train Loss: 0.034, Train Accuracy: 99.775
100%| | 646/646 [00:16<00:00, 39.11it/s]
Epoch 34/50, Train Loss: 0.031, Train Accuracy: 99.791
100%| | 646/646 [00:13<00:00, 47.88it/s]
Epoch 35/50, Train Loss: 0.028, Train Accuracy: 99.833
100%| | 646/646 [00:15<00:00, 40.88it/s]
Epoch 36/50, Train Loss: 0.030, Train Accuracy: 99.753
100%| | 646/646 [00:12<00:00, 53.28it/s]

Epoch 37/50, Train Loss: 0.024, Train Accuracy: 99.873
 100%| | 646/646 [00:12<00:00, 52.04it/s]
 Epoch 38/50, Train Loss: 0.020, Train Accuracy: 99.932
 100%| | 646/646 [00:11<00:00, 53.92it/s]
 Epoch 39/50, Train Loss: 0.018, Train Accuracy: 99.926
 100%| | 646/646 [00:11<00:00, 54.14it/s]
 Epoch 40/50, Train Loss: 0.020, Train Accuracy: 99.893
 100%| | 646/646 [00:14<00:00, 43.60it/s]
 Epoch 41/50, Train Loss: 0.024, Train Accuracy: 99.780
 100%| | 646/646 [00:12<00:00, 53.68it/s]
 Epoch 42/50, Train Loss: 0.014, Train Accuracy: 99.960
 100%| | 646/646 [00:12<00:00, 52.55it/s]
 Epoch 43/50, Train Loss: 0.012, Train Accuracy: 99.971
 100%| | 646/646 [00:12<00:00, 53.18it/s]
 Epoch 44/50, Train Loss: 0.011, Train Accuracy: 99.975
 100%| | 646/646 [00:12<00:00, 53.82it/s]
 Epoch 45/50, Train Loss: 0.010, Train Accuracy: 99.977
 100%| | 646/646 [00:12<00:00, 52.73it/s]
 Epoch 46/50, Train Loss: 0.017, Train Accuracy: 99.806
 100%| | 646/646 [00:12<00:00, 50.86it/s]
 Epoch 47/50, Train Loss: 0.009, Train Accuracy: 99.979
 100%| | 646/646 [00:12<00:00, 52.21it/s]
 Epoch 48/50, Train Loss: 0.008, Train Accuracy: 99.983
 100%| | 646/646 [00:12<00:00, 53.33it/s]
 Epoch 49/50, Train Loss: 0.014, Train Accuracy: 99.821
 100%| | 646/646 [00:12<00:00, 52.67it/s]
 Epoch 50/50, Train Loss: 0.014, Train Accuracy: 99.883

Training complete. Model saved to 'train_logs/window_7\model.pth', stats saved to 'train_logs/window_7\train_stats.csv'

100%| | 132/132 [00:17<00:00, 7.46it/s]
 c:\Users\Ward\anaconda3\envs\multibench\Lib\site-packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains

```

classes not in y_true
warnings.warn("y_pred contains classes not in y_true")

```

```

Evaluation Results:
Top-1 Accuracy: 90.79%
Top-5 Accuracy: 98.96%
Balanced Accuracy: 78.69%

```

```

Classification Report:

```

		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.91	0.89	0.90
1113				
	SYSTEM_GOODBYE	0.94	0.96	0.95
1331				
	SYSTEM_INFORM	0.97	0.92	0.94
719				
	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.95	0.92	0.93
310				
	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.59	0.74
17				
	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.86	0.70	0.77
79				
	SYSTEM_NOTIFY_SUCCESS	0.98	0.89	0.93
499				
	SYSTEM_OFFER	0.86	0.94	0.90
806				
	SYSTEM_OFFER_INTENT	0.85	0.97	0.91
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.92	0.93	0.93
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.86	0.86	0.86
14				
	SYSTEM_REQUEST	0.96	0.95	0.96
1961				
	SYSTEM_REQ_MORE	0.89	0.94	0.91
585				
	USER_AFFIRM	0.94	0.88	0.91
572				
	USER_AFFIRM_INTENT	0.77	0.82	0.79
139				
	USER_AFFIRM_INTENT USER_INFORM	0.83	0.85	0.84
103				
	USER_INFORM	0.97	0.95	0.96
1961				
	USER_INFORM_INTENT	0.95	0.91	0.93

658				
	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM_INTENT USER_SELECT	0.68	0.83	0.75
167				
	USER_INFORM_INTENT USER_THANK_YOU	0.27	0.25	0.26
12				
	USER_INFORM USER_INFORM_INTENT	0.93	0.93	0.93
772				
	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.74	0.65	0.69
117				
	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
14				
	USER_INFORM USER_NEGATE	0.83	0.85	0.84
216				
	USER_INFORM USER_REQUEST_ALTS	0.80	0.84	0.82
184				
	USER_NEGATE	0.57	0.57	0.57
7				
	USER_NEGATE_INTENT	0.53	0.74	0.61
69				
	USER_NEGATE_INTENT USER_GOODBYE	0.67	0.65	0.66
72				
	USER_NEGATE USER_GOODBYE	0.33	0.11	0.17
9				
	USER_NEGATE USER_THANK_YOU	0.96	0.94	0.95
587				
	USER_REQUEST	0.94	0.92	0.93
719				
	USER_REQUEST_ALTS	0.88	0.92	0.90
152				
	USER_REQUEST USER_AFFIRM	0.91	0.91	0.91
340				
	USER_SELECT	0.81	0.91	0.86
515				
	USER_SELECT USER_GOODBYE	0.80	0.64	0.71
279				
	USER_THANK_YOU	0.75	0.79	0.77
377				
	USER_THANK_YOU USER_GOODBYE	0.78	0.80	0.79
384				
	accuracy			0.91
16850				
	macro avg	0.76	0.75	0.75
16850				

16850

	weighted avg	0.91	0.91	0.91
--	--------------	------	------	------

=== Running training for window size: 8 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [01:06<00:00, 81.32it/s]
100%|      | 1331/1331 [00:11<00:00, 115.17it/s]
100%|      | 646/646 [02:17<00:00,  4.69it/s]

Epoch 1/50, Train Loss: 2.609, Train Accuracy: 23.254
100%|      | 646/646 [01:29<00:00,  7.25it/s]

Epoch 2/50, Train Loss: 1.889, Train Accuracy: 41.823
100%|      | 646/646 [01:16<00:00,  8.49it/s]

Epoch 3/50, Train Loss: 1.545, Train Accuracy: 51.638
100%|      | 646/646 [01:24<00:00,  7.68it/s]

Epoch 4/50, Train Loss: 1.312, Train Accuracy: 59.675
100%|      | 646/646 [00:58<00:00, 11.13it/s]

Epoch 5/50, Train Loss: 1.124, Train Accuracy: 67.149
100%|      | 646/646 [01:14<00:00,  8.73it/s]

Epoch 6/50, Train Loss: 0.977, Train Accuracy: 72.759
100%|      | 646/646 [00:45<00:00, 14.34it/s]

Epoch 7/50, Train Loss: 0.860, Train Accuracy: 76.955
100%|      | 646/646 [00:51<00:00, 12.55it/s]

Epoch 8/50, Train Loss: 0.740, Train Accuracy: 80.810
100%|      | 646/646 [00:44<00:00, 14.63it/s]
```

Epoch 9/50, Train Loss: 0.628, Train Accuracy: 84.355
100%| | 646/646 [00:48<00:00, 13.28it/s]
Epoch 10/50, Train Loss: 0.532, Train Accuracy: 87.277
100%| | 646/646 [01:05<00:00, 9.89it/s]
Epoch 11/50, Train Loss: 0.464, Train Accuracy: 89.197
100%| | 646/646 [00:44<00:00, 14.57it/s]
Epoch 12/50, Train Loss: 0.409, Train Accuracy: 90.786
100%| | 646/646 [00:37<00:00, 17.22it/s]
Epoch 13/50, Train Loss: 0.364, Train Accuracy: 92.042
100%| | 646/646 [00:38<00:00, 16.96it/s]
Epoch 14/50, Train Loss: 0.324, Train Accuracy: 93.039
100%| | 646/646 [00:29<00:00, 21.98it/s]
Epoch 15/50, Train Loss: 0.307, Train Accuracy: 93.524
100%| | 646/646 [00:24<00:00, 26.80it/s]
Epoch 16/50, Train Loss: 0.265, Train Accuracy: 94.571
100%| | 646/646 [00:27<00:00, 23.53it/s]
Epoch 17/50, Train Loss: 0.238, Train Accuracy: 95.320
100%| | 646/646 [00:28<00:00, 22.64it/s]
Epoch 18/50, Train Loss: 0.215, Train Accuracy: 95.876
100%| | 646/646 [00:26<00:00, 24.15it/s]
Epoch 19/50, Train Loss: 0.194, Train Accuracy: 96.282
100%| | 646/646 [00:33<00:00, 19.19it/s]
Epoch 20/50, Train Loss: 0.177, Train Accuracy: 96.688
100%| | 646/646 [00:29<00:00, 21.93it/s]
Epoch 21/50, Train Loss: 0.161, Train Accuracy: 97.066
100%| | 646/646 [00:12<00:00, 50.18it/s]
Epoch 22/50, Train Loss: 0.147, Train Accuracy: 97.389
100%| | 646/646 [00:19<00:00, 33.37it/s]
Epoch 23/50, Train Loss: 0.134, Train Accuracy: 97.692
100%| | 646/646 [00:29<00:00, 21.79it/s]
Epoch 24/50, Train Loss: 0.122, Train Accuracy: 98.007
100%| | 646/646 [00:20<00:00, 31.69it/s]

Epoch 25/50, Train Loss: 0.111, Train Accuracy: 98.279
100%| | 646/646 [00:13<00:00, 49.67it/s]
Epoch 26/50, Train Loss: 0.110, Train Accuracy: 98.247
100%| | 646/646 [00:19<00:00, 33.43it/s]
Epoch 27/50, Train Loss: 0.093, Train Accuracy: 98.685
100%| | 646/646 [00:19<00:00, 33.75it/s]
Epoch 28/50, Train Loss: 0.085, Train Accuracy: 98.836
100%| | 646/646 [00:22<00:00, 28.85it/s]
Epoch 29/50, Train Loss: 0.078, Train Accuracy: 98.976
100%| | 646/646 [01:12<00:00, 8.89it/s]
Epoch 30/50, Train Loss: 0.071, Train Accuracy: 99.179
100%| | 646/646 [00:39<00:00, 16.51it/s]
Epoch 31/50, Train Loss: 0.066, Train Accuracy: 99.250
100%| | 646/646 [00:36<00:00, 17.51it/s]
Epoch 32/50, Train Loss: 0.059, Train Accuracy: 99.413
100%| | 646/646 [00:31<00:00, 20.53it/s]
Epoch 33/50, Train Loss: 0.054, Train Accuracy: 99.498
100%| | 646/646 [00:37<00:00, 17.19it/s]
Epoch 34/50, Train Loss: 0.051, Train Accuracy: 99.522
100%| | 646/646 [00:44<00:00, 14.36it/s]
Epoch 35/50, Train Loss: 0.046, Train Accuracy: 99.613
100%| | 646/646 [00:40<00:00, 16.13it/s]
Epoch 36/50, Train Loss: 0.040, Train Accuracy: 99.748
100%| | 646/646 [00:51<00:00, 12.50it/s]
Epoch 37/50, Train Loss: 0.037, Train Accuracy: 99.768
100%| | 646/646 [00:27<00:00, 23.26it/s]
Epoch 38/50, Train Loss: 0.042, Train Accuracy: 99.636
100%| | 646/646 [00:31<00:00, 20.42it/s]
Epoch 39/50, Train Loss: 0.032, Train Accuracy: 99.824
100%| | 646/646 [00:34<00:00, 18.96it/s]
Epoch 40/50, Train Loss: 0.028, Train Accuracy: 99.884
100%| | 646/646 [00:34<00:00, 18.63it/s]

Epoch 41/50, Train Loss: 0.026, Train Accuracy: 99.895
 100%| | 646/646 [00:28<00:00, 22.85it/s]
 Epoch 42/50, Train Loss: 0.024, Train Accuracy: 99.896
 100%| | 646/646 [00:18<00:00, 34.99it/s]
 Epoch 43/50, Train Loss: 0.024, Train Accuracy: 99.901
 100%| | 646/646 [00:15<00:00, 42.82it/s]
 Epoch 44/50, Train Loss: 0.019, Train Accuracy: 99.939
 100%| | 646/646 [00:18<00:00, 35.40it/s]
 Epoch 45/50, Train Loss: 0.018, Train Accuracy: 99.933
 100%| | 646/646 [00:12<00:00, 49.90it/s]
 Epoch 46/50, Train Loss: 0.026, Train Accuracy: 99.772
 100%| | 646/646 [00:17<00:00, 35.98it/s]
 Epoch 47/50, Train Loss: 0.015, Train Accuracy: 99.959
 100%| | 646/646 [00:22<00:00, 28.66it/s]
 Epoch 48/50, Train Loss: 0.013, Train Accuracy: 99.975
 100%| | 646/646 [00:13<00:00, 46.85it/s]
 Epoch 49/50, Train Loss: 0.012, Train Accuracy: 99.977
 100%| | 646/646 [00:16<00:00, 38.57it/s]
 Epoch 50/50, Train Loss: 0.012, Train Accuracy: 99.977

Training complete. Model saved to 'train_logs/window_8\model.pth', stats saved to 'train_logs/window_8\train_stats.csv'

100%| | 132/132 [00:20<00:00, 6.45it/s]
 c:\Users\Ward\anaconda3\envs\multibench\Lib\site-packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains classes not in y_true
 warnings.warn("y_pred contains classes not in y_true")

Evaluation Results:
 Top-1 Accuracy: 87.98%
 Top-5 Accuracy: 98.29%
 Balanced Accuracy: 69.34%

Classification Report:

	precision	recall	f1-score
support			
SYSTEM_CONFIRM	0.88	0.89	0.89

1113				
	SYSTEM_GOODBYE	0.97	0.94	0.96
1331				
	SYSTEM_INFORM	0.96	0.92	0.94
719				
	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.93	0.95	0.94
310				
	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.50	0.18	0.26
17				
	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.70	0.48	0.57
79				
	SYSTEM_NOTIFY_SUCCESS	0.93	0.92	0.92
499				
	SYSTEM_OFFER	0.80	0.93	0.86
806				
	SYSTEM_OFFER_INTENT	0.86	0.96	0.91
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.92	0.87	0.90
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.67	0.43	0.52
14				
	SYSTEM_REQUEST	0.96	0.93	0.94
1961				
	SYSTEM_REQ_MORE	0.89	0.90	0.89
585				
	USER_AFFIRM	0.92	0.88	0.90
572				
	USER_AFFIRM_INTENT	0.66	0.63	0.65
139				
	USER_AFFIRM_INTENT USER_INFORM	0.60	0.68	0.64
103				
	USER_INFORM	0.97	0.93	0.95
1961				
	USER_INFORM_INTENT	0.93	0.89	0.91
658				
	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM_INTENT USER_SELECT	0.44	0.66	0.53
167				
	USER_INFORM_INTENT USER_THANK_YOU	0.08	0.08	0.08
12				
	USER_INFORM USER_INFORM_INTENT	0.91	0.94	0.92
772				
	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.40	0.50	0.44
117				
	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00

14				
	USER_INFORM USER_NEGATE	0.81	0.77	0.79
216				
	USER_INFORM USER_REQUEST_ALTS	0.62	0.79	0.69
184				
	USER_NEGATE	1.00	0.14	0.25
7				
	USER_NEGATE_INTENT	0.59	0.57	0.58
69				
	USER_NEGATE_INTENT USER_GOODBYE	0.64	0.50	0.56
72				
	USER_NEGATE USER_GOODBYE	0.00	0.00	0.00
9				
	USER_NEGATE USER_THANK_YOU	0.98	0.93	0.95
587				
	USER_REQUEST	0.93	0.86	0.90
719				
	USER_REQUEST_ALTS	0.81	0.80	0.80
152				
	USER_REQUEST USER_AFFIRM	0.85	0.92	0.88
340				
	USER_SELECT	0.76	0.90	0.82
515				
	USER_SELECT USER_GOODBYE	0.80	0.52	0.63
279				
	USER_THANK_YOU	0.71	0.67	0.69
377				
	USER_THANK_YOU USER_GOODBYE	0.70	0.79	0.74
384				
	accuracy			0.88
16850				
	macro avg	0.69	0.66	0.66
16850				
	weighted avg	0.89	0.88	0.88
16850				

=== Running training for window size: 9 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning:
You are using `torch.load` with `weights_only=False` (the current default
value), which uses the default pickle module implicitly. It is possible to
construct malicious pickle data which will execute arbitrary code during
unpickling (See
<https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for
more details). In a future release, the default value for `weights_only` will be
flipped to `True`. This limits the functions that could be executed during

unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via ``torch.serialization.add_safe_globals``. We recommend you start setting ``weights_only=True`` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [00:56<00:00, 96.48it/s]
100%|      | 1331/1331 [00:49<00:00, 26.77it/s]
100%|      | 646/646 [03:08<00:00,  3.44it/s]

Epoch 1/50, Train Loss: 2.674, Train Accuracy: 20.693
100%|      | 646/646 [01:34<00:00,  6.82it/s]

Epoch 2/50, Train Loss: 2.094, Train Accuracy: 34.621
100%|      | 646/646 [01:33<00:00,  6.92it/s]

Epoch 3/50, Train Loss: 1.807, Train Accuracy: 43.883
100%|      | 646/646 [01:17<00:00,  8.38it/s]

Epoch 4/50, Train Loss: 1.599, Train Accuracy: 49.491
100%|      | 646/646 [01:06<00:00,  9.72it/s]

Epoch 5/50, Train Loss: 1.395, Train Accuracy: 56.094
100%|      | 646/646 [00:48<00:00, 13.38it/s]

Epoch 6/50, Train Loss: 1.054, Train Accuracy: 68.751
100%|      | 646/646 [01:14<00:00,  8.70it/s]

Epoch 7/50, Train Loss: 0.871, Train Accuracy: 75.109
100%|      | 646/646 [00:43<00:00, 14.97it/s]

Epoch 8/50, Train Loss: 0.767, Train Accuracy: 78.630
100%|      | 646/646 [01:12<00:00,  8.89it/s]

Epoch 9/50, Train Loss: 0.677, Train Accuracy: 82.229
100%|      | 646/646 [00:41<00:00, 15.56it/s]

Epoch 10/50, Train Loss: 0.602, Train Accuracy: 84.819
100%|      | 646/646 [00:59<00:00, 10.92it/s]

Epoch 11/50, Train Loss: 0.538, Train Accuracy: 86.841
100%|      | 646/646 [01:09<00:00,  9.36it/s]

Epoch 12/50, Train Loss: 0.483, Train Accuracy: 88.495
100%|      | 646/646 [00:45<00:00, 14.19it/s]

Epoch 13/50, Train Loss: 0.433, Train Accuracy: 89.971
```


100%| | 646/646 [00:48<00:00, 13.45it/s]
Epoch 14/50, Train Loss: 0.388, Train Accuracy: 91.221
100%| | 646/646 [01:04<00:00, 10.07it/s]
Epoch 15/50, Train Loss: 0.349, Train Accuracy: 92.257
100%| | 646/646 [00:31<00:00, 20.74it/s]
Epoch 16/50, Train Loss: 0.318, Train Accuracy: 93.167
100%| | 646/646 [00:40<00:00, 16.12it/s]
Epoch 17/50, Train Loss: 0.286, Train Accuracy: 93.826
100%| | 646/646 [00:27<00:00, 23.64it/s]
Epoch 18/50, Train Loss: 0.267, Train Accuracy: 94.299
100%| | 646/646 [00:36<00:00, 17.86it/s]
Epoch 19/50, Train Loss: 0.241, Train Accuracy: 94.936
100%| | 646/646 [00:32<00:00, 19.85it/s]
Epoch 20/50, Train Loss: 0.217, Train Accuracy: 95.584
100%| | 646/646 [00:29<00:00, 21.54it/s]
Epoch 21/50, Train Loss: 0.200, Train Accuracy: 95.945
100%| | 646/646 [00:25<00:00, 25.08it/s]
Epoch 22/50, Train Loss: 0.185, Train Accuracy: 96.262
100%| | 646/646 [00:17<00:00, 37.16it/s]
Epoch 23/50, Train Loss: 0.169, Train Accuracy: 96.687
100%| | 646/646 [00:16<00:00, 40.24it/s]
Epoch 24/50, Train Loss: 0.156, Train Accuracy: 97.020
100%| | 646/646 [00:23<00:00, 27.33it/s]
Epoch 25/50, Train Loss: 0.148, Train Accuracy: 97.214
100%| | 646/646 [00:26<00:00, 24.53it/s]
Epoch 26/50, Train Loss: 0.134, Train Accuracy: 97.573
100%| | 646/646 [00:19<00:00, 33.08it/s]
Epoch 27/50, Train Loss: 0.123, Train Accuracy: 97.859
100%| | 646/646 [00:28<00:00, 23.02it/s]
Epoch 28/50, Train Loss: 0.114, Train Accuracy: 98.117
100%| | 646/646 [00:20<00:00, 30.91it/s]
Epoch 29/50, Train Loss: 0.105, Train Accuracy: 98.306

100%| | 646/646 [00:25<00:00, 25.42it/s]
Epoch 30/50, Train Loss: 0.097, Train Accuracy: 98.495
100%| | 646/646 [00:19<00:00, 33.83it/s]
Epoch 31/50, Train Loss: 0.091, Train Accuracy: 98.667
100%| | 646/646 [00:13<00:00, 47.21it/s]
Epoch 32/50, Train Loss: 0.082, Train Accuracy: 98.913
100%| | 646/646 [00:13<00:00, 47.26it/s]
Epoch 33/50, Train Loss: 0.077, Train Accuracy: 99.020
100%| | 646/646 [00:13<00:00, 48.07it/s]
Epoch 34/50, Train Loss: 0.074, Train Accuracy: 99.102
100%| | 646/646 [00:13<00:00, 48.06it/s]
Epoch 35/50, Train Loss: 0.070, Train Accuracy: 99.140
100%| | 646/646 [00:13<00:00, 48.14it/s]
Epoch 36/50, Train Loss: 0.060, Train Accuracy: 99.362
100%| | 646/646 [00:13<00:00, 48.00it/s]
Epoch 37/50, Train Loss: 0.055, Train Accuracy: 99.483
100%| | 646/646 [00:17<00:00, 36.12it/s]
Epoch 38/50, Train Loss: 0.054, Train Accuracy: 99.498
100%| | 646/646 [00:17<00:00, 37.98it/s]
Epoch 39/50, Train Loss: 0.055, Train Accuracy: 99.465
100%| | 646/646 [00:15<00:00, 40.55it/s]
Epoch 40/50, Train Loss: 0.047, Train Accuracy: 99.637
100%| | 646/646 [00:15<00:00, 40.40it/s]
Epoch 41/50, Train Loss: 0.041, Train Accuracy: 99.708
100%| | 646/646 [00:15<00:00, 40.79it/s]
Epoch 42/50, Train Loss: 0.038, Train Accuracy: 99.777
100%| | 646/646 [00:17<00:00, 36.26it/s]
Epoch 43/50, Train Loss: 0.034, Train Accuracy: 99.818
100%| | 646/646 [00:24<00:00, 26.27it/s]
Epoch 44/50, Train Loss: 0.039, Train Accuracy: 99.672
100%| | 646/646 [00:15<00:00, 42.69it/s]
Epoch 45/50, Train Loss: 0.031, Train Accuracy: 99.864

```

100%|      | 646/646 [00:13<00:00, 47.41it/s]
Epoch 46/50, Train Loss: 0.031, Train Accuracy: 99.835
100%|      | 646/646 [00:13<00:00, 46.73it/s]
Epoch 47/50, Train Loss: 0.031, Train Accuracy: 99.792
100%|      | 646/646 [00:14<00:00, 43.37it/s]
Epoch 48/50, Train Loss: 0.024, Train Accuracy: 99.913
100%|      | 646/646 [00:18<00:00, 34.95it/s]
Epoch 49/50, Train Loss: 0.023, Train Accuracy: 99.915
100%|      | 646/646 [00:20<00:00, 31.68it/s]
Epoch 50/50, Train Loss: 0.029, Train Accuracy: 99.787

```

Training complete. Model saved to 'train_logs/window_9\model.pth', stats saved to 'train_logs/window_9\train_stats.csv'

```

100%|      | 132/132 [00:24<00:00,  5.38it/s]
c:\Users\Ward\anaconda3\envs\multibench\Lib\site-
packages\sklearn\metrics\_classification.py:2480: UserWarning: y_pred contains
classes not in y_true
  warnings.warn("y_pred contains classes not in y_true")

```

Evaluation Results:
Top-1 Accuracy: 86.93%
Top-5 Accuracy: 98.00%
Balanced Accuracy: 66.98%

Classification Report:

		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.89	0.84	0.87
1113	SYSTEM_GOODBYE	0.94	0.94	0.94
1331	SYSTEM_INFORM	0.93	0.92	0.92
719	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.89	0.90	0.90
310	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.33	0.12	0.17
17	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.76	0.39	0.52
79	SYSTEM_NOTIFY_SUCCESS	0.92	0.82	0.86
499				

	SYSTEM_OFFER	0.83	0.94	0.88
806				
	SYSTEM_OFFER_INTENT	0.80	0.94	0.87
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.91	0.92	0.91
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.50	0.07	0.12
14				
	SYSTEM_REQUEST	0.95	0.94	0.95
1961				
	SYSTEM_REQ_MORE	0.85	0.85	0.85
585				
	USER_AFFIRM	0.85	0.74	0.79
572				
	USER_AFFIRM_INTENT	0.72	0.88	0.79
139				
	USER_AFFIRM_INTENT USER_INFORM	0.80	0.72	0.76
103				
	USER_INFORM	0.95	0.96	0.96
1961				
	USER_INFORM_INTENT	0.92	0.91	0.92
658				
	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
0				
	USER_INFORM_INTENT USER_SELECT	0.55	0.68	0.61
167				
	USER_INFORM_INTENT USER_THANK_YOU	0.20	0.08	0.12
12				
	USER_INFORM USER_INFORM_INTENT	0.94	0.93	0.93
772				
	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.47	0.47	0.47
117				
	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
14				
	USER_INFORM USER_NEGATE	0.86	0.69	0.76
216				
	USER_INFORM USER_REQUEST_ALTS	0.62	0.72	0.67
184				
	USER_NEGATE	0.00	0.00	0.00
7				
	USER_NEGATE_INTENT	0.65	0.58	0.61
69				
	USER_NEGATE_INTENT USER_GOODBYE	0.48	0.57	0.52
72				
	USER_NEGATE USER_GOODBYE	0.00	0.00	0.00
9				
	USER_NEGATE USER_THANK_YOU	0.95	0.91	0.93
587				

719	USER_REQUEST	0.93	0.84	0.88
152	USER_REQUEST_ALTS	0.69	0.84	0.76
340	USER_REQUEST USER_AFFIRM	0.80	0.85	0.82
515	USER_SELECT	0.69	0.90	0.79
279	USER_SELECT USER_GOODBYE	0.76	0.48	0.59
377	USER_THANK_YOU	0.75	0.66	0.70
384	USER_THANK_YOU USER_GOODBYE	0.64	0.77	0.70
16850	accuracy			0.87
16850	macro avg	0.68	0.65	0.65
16850	weighted avg	0.87	0.87	0.87

=== Running training for window size: 10 ===

C:\Users\Ward\AppData\Local\Temp\ipykernel_10956\3759199132.py:9: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during unpickling (See <https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models> for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

```
self.embeddings = torch.load(embeddings_path)
100%|      | 5403/5403 [01:19<00:00, 67.58it/s]
100%|      | 1331/1331 [00:27<00:00, 48.45it/s]
100%|      | 646/646 [03:31<00:00, 3.06it/s]
```

Epoch 1/50, Train Loss: 2.697, Train Accuracy: 20.932

```
100%|      | 646/646 [00:36<00:00, 17.55it/s]
```

Epoch 2/50, Train Loss: 2.150, Train Accuracy: 32.265

100%| | 646/646 [02:10<00:00, 4.96it/s]
Epoch 3/50, Train Loss: 1.849, Train Accuracy: 41.505
100%| | 646/646 [01:35<00:00, 6.75it/s]
Epoch 4/50, Train Loss: 1.628, Train Accuracy: 47.973
100%| | 646/646 [00:45<00:00, 14.10it/s]
Epoch 5/50, Train Loss: 1.450, Train Accuracy: 54.038
100%| | 646/646 [00:47<00:00, 13.71it/s]
Epoch 6/50, Train Loss: 1.305, Train Accuracy: 60.136
100%| | 646/646 [00:52<00:00, 12.36it/s]
Epoch 7/50, Train Loss: 1.179, Train Accuracy: 65.568
100%| | 646/646 [00:57<00:00, 11.28it/s]
Epoch 8/50, Train Loss: 1.066, Train Accuracy: 70.528
100%| | 646/646 [00:39<00:00, 16.53it/s]
Epoch 9/50, Train Loss: 0.958, Train Accuracy: 75.031
100%| | 646/646 [00:47<00:00, 13.68it/s]
Epoch 10/50, Train Loss: 0.835, Train Accuracy: 79.353
100%| | 646/646 [00:37<00:00, 17.30it/s]
Epoch 11/50, Train Loss: 0.692, Train Accuracy: 84.047
100%| | 646/646 [00:42<00:00, 15.29it/s]
Epoch 12/50, Train Loss: 0.602, Train Accuracy: 86.189
100%| | 646/646 [00:37<00:00, 17.41it/s]
Epoch 13/50, Train Loss: 0.533, Train Accuracy: 87.947
100%| | 646/646 [00:46<00:00, 14.02it/s]
Epoch 14/50, Train Loss: 0.478, Train Accuracy: 89.436
100%| | 646/646 [00:49<00:00, 13.17it/s]
Epoch 15/50, Train Loss: 0.431, Train Accuracy: 90.552
100%| | 646/646 [00:43<00:00, 14.95it/s]
Epoch 16/50, Train Loss: 0.394, Train Accuracy: 91.500
100%| | 646/646 [00:45<00:00, 14.22it/s]
Epoch 17/50, Train Loss: 0.359, Train Accuracy: 92.286
100%| | 646/646 [01:13<00:00, 8.77it/s]
Epoch 18/50, Train Loss: 0.324, Train Accuracy: 93.257

100%| | 646/646 [01:57<00:00, 5.50it/s]
Epoch 19/50, Train Loss: 0.296, Train Accuracy: 93.962
100%| | 646/646 [01:47<00:00, 6.01it/s]
Epoch 20/50, Train Loss: 0.270, Train Accuracy: 94.574
100%| | 646/646 [01:27<00:00, 7.40it/s]
Epoch 21/50, Train Loss: 0.248, Train Accuracy: 95.129
100%| | 646/646 [02:10<00:00, 4.94it/s]
Epoch 22/50, Train Loss: 0.228, Train Accuracy: 95.603
100%| | 646/646 [00:55<00:00, 11.67it/s]
Epoch 23/50, Train Loss: 0.219, Train Accuracy: 95.830
100%| | 646/646 [00:56<00:00, 11.37it/s]
Epoch 24/50, Train Loss: 0.195, Train Accuracy: 96.406
100%| | 646/646 [00:53<00:00, 12.14it/s]
Epoch 25/50, Train Loss: 0.179, Train Accuracy: 96.801
100%| | 646/646 [01:10<00:00, 9.10it/s]
Epoch 26/50, Train Loss: 0.165, Train Accuracy: 97.124
100%| | 646/646 [01:03<00:00, 10.13it/s]
Epoch 27/50, Train Loss: 0.151, Train Accuracy: 97.465
100%| | 646/646 [00:56<00:00, 11.49it/s]
Epoch 28/50, Train Loss: 0.142, Train Accuracy: 97.647
100%| | 646/646 [00:45<00:00, 14.22it/s]
Epoch 29/50, Train Loss: 0.130, Train Accuracy: 97.897
100%| | 646/646 [00:48<00:00, 13.18it/s]
Epoch 30/50, Train Loss: 0.119, Train Accuracy: 98.192
100%| | 646/646 [01:15<00:00, 8.51it/s]
Epoch 31/50, Train Loss: 0.115, Train Accuracy: 98.225
100%| | 646/646 [00:51<00:00, 12.48it/s]
Epoch 32/50, Train Loss: 0.104, Train Accuracy: 98.489
100%| | 646/646 [00:55<00:00, 11.73it/s]
Epoch 33/50, Train Loss: 0.095, Train Accuracy: 98.677
100%| | 646/646 [01:03<00:00, 10.15it/s]
Epoch 34/50, Train Loss: 0.089, Train Accuracy: 98.832

100%| | 646/646 [01:06<00:00, 9.65it/s]
Epoch 35/50, Train Loss: 0.086, Train Accuracy: 98.853
100%| | 646/646 [01:13<00:00, 8.74it/s]
Epoch 36/50, Train Loss: 0.074, Train Accuracy: 99.112
100%| | 646/646 [00:57<00:00, 11.23it/s]
Epoch 37/50, Train Loss: 0.070, Train Accuracy: 99.188
100%| | 646/646 [01:04<00:00, 10.05it/s]
Epoch 38/50, Train Loss: 0.064, Train Accuracy: 99.324
100%| | 646/646 [00:44<00:00, 14.60it/s]
Epoch 39/50, Train Loss: 0.061, Train Accuracy: 99.356
100%| | 646/646 [00:54<00:00, 11.91it/s]
Epoch 40/50, Train Loss: 0.058, Train Accuracy: 99.373
100%| | 646/646 [00:54<00:00, 11.96it/s]
Epoch 41/50, Train Loss: 0.054, Train Accuracy: 99.467
100%| | 646/646 [01:24<00:00, 7.61it/s]
Epoch 42/50, Train Loss: 0.057, Train Accuracy: 99.422
100%| | 646/646 [00:40<00:00, 16.13it/s]
Epoch 43/50, Train Loss: 0.044, Train Accuracy: 99.667
100%| | 646/646 [00:31<00:00, 20.53it/s]
Epoch 44/50, Train Loss: 0.043, Train Accuracy: 99.686
100%| | 646/646 [00:42<00:00, 15.25it/s]
Epoch 45/50, Train Loss: 0.041, Train Accuracy: 99.682
100%| | 646/646 [00:49<00:00, 13.02it/s]
Epoch 46/50, Train Loss: 0.035, Train Accuracy: 99.817
100%| | 646/646 [01:02<00:00, 10.30it/s]
Epoch 47/50, Train Loss: 0.032, Train Accuracy: 99.867
100%| | 646/646 [01:37<00:00, 6.61it/s]
Epoch 48/50, Train Loss: 0.033, Train Accuracy: 99.815
100%| | 646/646 [01:03<00:00, 10.17it/s]
Epoch 49/50, Train Loss: 0.036, Train Accuracy: 99.718
100%| | 646/646 [01:25<00:00, 7.57it/s]

Epoch 50/50, Train Loss: 0.026, Train Accuracy: 99.912

Training complete. Model saved to 'train_logs/window_10\model.pth', stats saved to 'train_logs/window_10\train_stats.csv'

100%| | 132/132 [00:20<00:00, 6.50it/s]

c:\Users\Ward\anaconda3\envs\multibench\Lib\site-packages\sklearn\metrics_classification.py:2480: UserWarning: y_pred contains classes not in y_true

warnings.warn("y_pred contains classes not in y_true")

Evaluation Results:

Top-1 Accuracy: 86.16%

Top-5 Accuracy: 97.29%

Balanced Accuracy: 67.52%

Classification Report:

		precision	recall	f1-score
support				
	SYSTEM_CONFIRM	0.89	0.79	0.84
1113				
	SYSTEM_GOODBYE	0.91	0.93	0.92
1331				
	SYSTEM_INFORM	0.93	0.91	0.92
719				
	SYSTEM_INFORM SYSTEM_NOTIFY_SUCCESS	0.92	0.85	0.89
310				
	SYSTEM_INFORM SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	0.60	0.18	0.27
17				
	SYSTEM_NOTIFY_FAILURE SYSTEM_REQ_MORE	0.68	0.34	0.45
79				
	SYSTEM_NOTIFY_SUCCESS	0.90	0.86	0.88
499				
	SYSTEM_OFFER	0.82	0.92	0.86
806				
	SYSTEM_OFFER_INTENT	0.75	0.97	0.85
383				
	SYSTEM_OFFER SYSTEM_INFORM_COUNT	0.87	0.91	0.89
608				
	SYSTEM_OFFER SYSTEM_NOTIFY_FAILURE	1.00	0.29	0.44
14				
	SYSTEM_REQUEST	0.94	0.93	0.93
1961				
	SYSTEM_REQ_MORE	0.87	0.91	0.89
585				
	USER_AFFIRM	0.90	0.78	0.84
572				

139	USER_AFFIRM_INTENT	0.79	0.82	0.81
103	USER_AFFIRM_INTENT USER_INFORM	0.79	0.75	0.77
1961	USER_INFORM	0.97	0.92	0.94
658	USER_INFORM_INTENT	0.95	0.90	0.92
0	USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
167	USER_INFORM_INTENT USER_SELECT	0.62	0.75	0.68
12	USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
772	USER_INFORM USER_INFORM_INTENT	0.91	0.94	0.92
0	USER_INFORM USER_INFORM_INTENT USER_NEGATE_INTENT	0.00	0.00	0.00
117	USER_INFORM USER_INFORM_INTENT USER_SELECT	0.43	0.39	0.41
14	USER_INFORM USER_INFORM_INTENT USER_THANK_YOU	0.00	0.00	0.00
216	USER_INFORM USER_NEGATE	0.72	0.67	0.70
184	USER_INFORM USER_REQUEST_ALTS	0.56	0.77	0.65
7	USER_NEGATE	0.50	0.14	0.22
69	USER_NEGATE_INTENT	0.46	0.58	0.51
72	USER_NEGATE_INTENT USER_GOODBYE	0.56	0.50	0.53
9	USER_NEGATE USER_GOODBYE	0.00	0.00	0.00
587	USER_NEGATE USER_THANK_YOU	0.95	0.93	0.94
719	USER_REQUEST	0.93	0.84	0.88
152	USER_REQUEST_ALTS	0.74	0.86	0.80
340	USER_REQUEST USER_AFFIRM	0.83	0.82	0.83
515	USER_SELECT	0.67	0.87	0.76
279	USER_SELECT USER_GOODBYE	0.67	0.50	0.57
377	USER_THANK_YOU	0.80	0.63	0.71

384	USER_THANK_YOU USER_GOODBYE	0.59	0.84	0.69
16850	accuracy			0.86
16850	macro avg	0.68	0.64	0.64
16850	weighted avg	0.87	0.86	0.86

Ablation study completed! Results saved to 'ablation_window_size_results.csv'

Visualize the results of the different window sizes:

```
[ ]: import pandas as pd
import matplotlib.pyplot as plt

# Load the CSV file
df = pd.read_csv("ablation_window_size_results.csv")

# Sort by window_size just to be safe
df = df.sort_values("window_size")

# Create a color-mapped table using pandas styling
styled_table = df.style.background_gradient(
    subset=["accuracy", "top5_accuracy", "balanced_accuracy"],
    cmap="Blues"
).format({
    "accuracy": "{:.2f}",
    "top5_accuracy": "{:.2f}",
    "balanced_accuracy": "{:.2f}",
    "dataset_creation_time_sec": "{:.2f}",
    "train_time_sec": "{:.2f}"
}).set_caption("Ablation Study: Window Size vs Performance")

styled_table
```

```
[ ]: <pandas.io.formats.style.Styler at 0x1bad1322c60>
```

```
[23]: import pandas as pd
import matplotlib.pyplot as plt

# Load CSV
df = pd.read_csv("ablation_window_size_results.csv")

# Set style
plt.style.use("seaborn-v0_8-whitegrid")
```

```

fig, axs = plt.subplots(1, 2, figsize=(15, 6))

# Plot accuracy curves
axs[0].plot(df["window_size"], df["accuracy"], label="Accuracy", marker='o')
axs[0].plot(df["window_size"], df["top5_accuracy"], label="Top-5 Accuracy",
            ↪marker='^')
axs[0].plot(df["window_size"], df["balanced_accuracy"], label="Balanced_
            ↪Accuracy", marker='s')
axs[0].set_xlabel("Window Size")
axs[0].set_ylabel("Accuracy (%)")
axs[0].set_title("Performance vs. Window Size")
axs[0].legend()

# Plot time curves
axs[1].plot(df["window_size"], df["dataset_creation_time_sec"], label="Dataset_
            ↪Creation Time", marker='o')
axs[1].plot(df["window_size"], df["train_time_sec"], label="Train Time",
            ↪marker='x')
axs[1].set_xlabel("Window Size")
axs[1].set_ylabel("Time (seconds)")
axs[1].set_title("Time vs. Window Size")
axs[1].legend()

plt.tight_layout()
plt.show()
fig.savefig("ablation_study_results.png", dpi=300)

```

