Deep Learning for Social Sciences

Assignment 4: Fine-tuning Large Language Models for Political Text Classification

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26 June 2025

To be handed in by July 16th 23:59

1 Overview

In this assignment, you will explore fine-tuning techniques for large language models using real parliamentary speech data. Unlike traditional machine learning approaches that rely on hand-crafted features, modern language models can learn rich representations directly from text through fine-tuning on domain-specific data.

The assignment uses the ParlaMint parliamentary corpus, containing speeches from various European parliaments. You will predict political party affiliation and ideological orientation from speech text, comparing fine-tuned model performance against pretrained ones to understand when domain adaptation provides additional predictive power.

Complete all tasks for receiving points. Maximum points: 30. Bonus points available but will not exceed 30 total.

1.1 Dataset Information

- Source: ParlaMint parliamentary corpus (CLARIN repository)
- Country: Several countries available, I suggest trying with Austria, but you can choose the country you prefer
- Files: Speech transcripts (.txt) and metadata (.tsv) files
- Data type: Parliamentary speeches with speaker metadata
- Target variables: Political party affiliation, ideological orientation
- Features: Speech text, party affiliation, speaker role, date, session info
- Time period: Focus on the last 5 years of available data for temporal consistency

Available at: https://www.clarin.si/repository/xmlui/handle/11356/1912

2 Tasks

2.1 Data Import, Preprocessing and Exploratory Analysis (5 points)

- Load and parse both speech transcripts and metadata files
- Create a unified dataset matching speeches to political parties and political orientation

- Focus on the last 5 years of available data to ensure temporal consistency
- Implement data cleaning and preprocessing:
 - Remove speeches shorter than 50 words
 - Filter out procedural and administrative speeches
 - Handle missing or incomplete party information
 - Remove duplicate speeches
- Perform comprehensive exploratory data analysis:
 - Analyze the distribution of speech lengths and political parties
 - Compute basic text statistics (vocabulary size, average speech length, etc.)
 - Visualize party distribution and temporal patterns
 - Create word clouds or frequency distributions for different parties

2.2 Model Training and Comparison (10 points)

Train and evaluate fine-tuned language models for two classification tasks:

2.2.1 Task A: Political Party Prediction

- Target: Multi-class classification of political party affiliation
- Hypothesis: Different parties use distinct vocabulary and rhetorical patterns
- Base model: Use a small pre-trained generative language model (e.g. Qwen, Gemma, Llama ...)
- Features: Raw speech text (no manual feature engineering)

2.2.2 Task B: Ideological Orientation Prediction

- Target: Multi-class classification (left/center/right etc)
- Hypothesis: Ideological positions manifest in language use and topic focus
- Base model: Same architecture as Task A
- Features: Raw speech text

For each task:

- Build a suitable dataset for instruction fine-tuning
- Use the pretrained model for the task as a baseline
- Implement fine-tuning using LoRA (Low-Rank Adaptation)
- Use appropriate train/validation/test splits (e.g. 70%/15%/15%)
- Report comprehensive performance metrics
- Include confusion matrices and classification reports

2.3 Report Writing (15 points)

Write a comprehensive report (max 3 pages) that includes:

2.3.1 Structure:

- **Introduction**: Problem description, political science context of the country analyzed, and technical approach
- Data Analysis: Dataset description, preprocessing steps, exploratory findings
- Methods: Model architectures, fine-tuning procedures
- Results: Performance comparison tables, confusion matrices, error analysis
- Discussion and Conclusions:
 - Discuss potential applications and limitations
 - Key findings and methodological insights

3 Bonus Tasks

You can use up to an additional half page for each bonus task.

3.1 Bonus Task 1: Cross-temporal Evaluation (1 bonus points)

- Train models on historical data and test on recent speeches
- Analyze how political language evolution affects model performance
- Discuss implications for temporal generalization in political NLP

3.2 Bonus Task 2: Attention Analysis and Interpretability (2 bonus points)

- Implement attention visualization techniques (e.g. SHAP Values)
- Identify which words/phrases are most important for party classification
- Analyze whether the model learns politically meaningful patterns
- Compare learned patterns with political science knowledge

Note: This assignment reflects real-world challenges in computational political science. Focus on understanding when and why domain-specific fine-tuning improves performance over general baselines, rather than achieving perfect accuracy.