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# Archaeology of Intelligent Machines: Comparing Romanian Language Usage in Romania with Romanian Usage in the Diaspora

#### 1st Semester of 2024-2025

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#### **Abstract**

This study focuses on observing the linguistic contact between Romanian and the majority language in various diaspora regions. By employing advanced natural language processing techniques such as tokenization, vectorization, and statistical analysis, the project identified nuanced patterns reflecting how Romanian interacts and adapts to the linguistic environments of host countries. The findings highlight sociocultural dynamics and offer insights into the evolving linguistic identity of Romanian speakers abroad.

### 1 Introduction

The relationship between Romanian as written in Romania and in the diaspora represents a fascinating linguistic and cultural study. Sharing a common language base, these variations evolve under distinct cultural, geographical, and social influences. This project aims to capture and analyze these differences systematically.

This study aims to:

- Observe the linguistic contact between Romanian and the majority language in each region.
- Identify unique linguistic features and adaptations across contexts.
- Quantify stylistic and semantic differences using advanced NLP techniques.
- Analyze socio-cultural influences that shape language use in the diaspora.

We developed a comprehensive NLP pipeline for this analysis, incorporating text normalization, tokenization, diacritic restoration, and advanced statistical evaluations. The analysis is enriched by comparisons with previous studies and detailed corpus evaluations.

We chose this subject because we found the context of Romanian elections intriguing and wanted

to explore whether significant differences exist in how these events are covered in diaspora articles compared to those in Romania.

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### **Contributions**

- Alex: Developed the NLP pipeline, including data preprocessing, tokenization, and statistical analysis.
- Medeea: Focused on data collection and visualization. Curated the corpus from diverse regional sources and generated graphs to highlight linguistic patterns.
- Sara: Conducted the interpretive analysis and comparative study. Analyzed the sociocultural implications of linguistic patterns and drafted the findings and conclusions sections.

#### 2 Approach

#### **Approach**

To complete this project, we employed a series of statistical and natural language processing (NLP) techniques to analyze regional variations in written Romanian. Below, I detail the approach and methods used.

## **Data Collection and Corpus Details**

The code and dataset are hosted in a Git repository. https://github.com/mariamedeea/ Romanian-in-Different-Regions

#### **Software Tools Used**

- Programming Language: Python
- **Libraries:** pandas, NumPy, NLTK, scikit-learn, spaCy, and matplotlib
- Environment: Google Colab for computational processing and visualization

### **Training and Processing Time**

The processing time varied by task:

- Text preprocessing: 1.5 hour per region
- Feature extraction (bigram, trigram computation, TF-IDF): 30 minutes
- Generating visualizations: 15 minutes
- total NLP Pipeline execution: 15 minutes

## Machine Learning and Optimization Techniques

While this project was predominantly statistical, we employed TF-IDF (term frequency-inverse document frequency) for lexical analysis for identifying characteristic words for regions (Figure 2). Preprocessing included stop-word removal, stemming, and lemmatization to improve text standardization.

## **Evaluation Report**

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The evaluation focused on comparing linguistic patterns across regions. Key findings included:

- **Distribution of POS Tags:** Regions demonstrated distinct usage patterns, e.g., one region favored verbs while another favored nouns.
- Anagram Analysis: Highlighted unique morphological traits by region.
- Bigram/Trigram Analysis: Identified commonly co-occurring phrases and syntactic structures.
- Characteristic Words for Regions: Derived using TF-IDF, showing lexical uniqueness.
- **Loanwords:** Assessed for frequency and type, indicating cultural influences.
- Average Text Length: Regions with richer descriptions or narratives had longer texts.

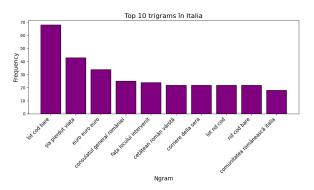
#### Visualisation

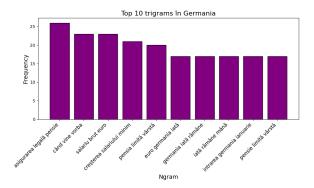
Below is a visualisation illustrating the findings:

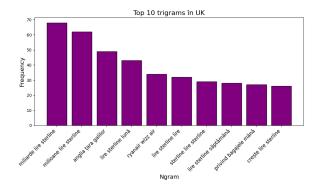
## 3 Findings and Insights

## **Regional Linguistic Patterns**

- Germany:
  - Strong influence of geography, culture, and politics (e.g., *Renania*, *Bundestag*).







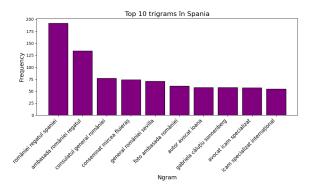


Figure 1: Trigrams

09	<ul> <li>Administrative topics are prominent, such as <i>Kindergeld</i> (child benefits).</li> <li>Romania:</li> </ul>	4 Limitations	149
10		While the findings are robust, several limitations	150
11		were noted:	151
12	- Dominance of media and news-related	• Dependence on word-level features, limiting	152
13	terms (e.g., DCNews, Blinken).	contextual depth.	153
14	<ul> <li>Focus on global and local events.</li> </ul>	<ul> <li>Corpus diversity was constrained by the avail-</li> </ul>	154
15	• Italy:	ability of textual data from specific regions.	155
16	- Influence of local culture and administra-	<ul> <li>Advanced models such as transformers were</li> </ul>	156
17	tion (e.g., Modena, Carabinieri).	not utilized, which could provide richer in-	157
18	<ul> <li>Social and work-related topics are evi-</li> </ul>	sights.	158
19	dent.	Future work will address these limitations by	159
20	• Spain:	expanding corpus diversity and incorporating state-	160
21	- Geographic and cultural connection (e.g.,	of-the-art NLP techniques.	161
22	Canare, Picasso).	5 Conclusions and Future Work	162
23	- Translation and integration topics high-	This project successfully examined the linguistic	163
24	lighted.	adaptations of Romanian in diaspora contexts, high-	164
25	• UK:	lighting the socio-cultural dynamics influencing	165
		language use. Key takeaways include:	166
26 27	<ul> <li>Economic and administrative focus (e.g., <i>GBP</i>, <i>HMRC</i>).</li> </ul>	Romanian exhibits significant adaptability, in-	167
28	<ul><li>References to health services (<i>NHS</i>) and</li></ul>	fluenced by the majority language in host	168
29	geography.	countries.	169
		Regional variations offer a window into cul-	170
30	General Observations	tural integration and identity.	171
31	• Distinct cultural and administrative terms re-		
32	flect local adaptation.	• NLP pipelines are effective in quantifying and	172
33	• Media and online influence are significant	visualizing linguistic patterns.	173
34	across all regions.	Future directions include:	174
35	<ul> <li>Romanian diaspora integrates into local issues</li> </ul>	<ul> <li>Incorporating social media data for more dy-</li> </ul>	175
36	while maintaining ties to Romania.	namic analyses.	176
37	Regional Linguistic Insights	• Using transformers and contextual embed-	177
38	• Romania: Diverse vocabulary; focus on local	dings for deeper insights.	178
39	and global news.	<ul> <li>Expanding the study to spoken language anal-</li> </ul>	179
40	• Germany & UK: Practical terms related to	ysis.	180
41	economy and administration dominate.	References	181
42	• Italy & Spain: Balance of cultural and social	• Susan Sanders, Tom Dotz, Tom Hoobyar	
43	integration with Romanian identity.	(2013). "NLP: The Essential Guide"	182 183
44	Overall Patterns		

• Linguistic adaptation is visible through local

• Romanian identity remains strong across all

influences in diaspora regions.

analyzed texts.

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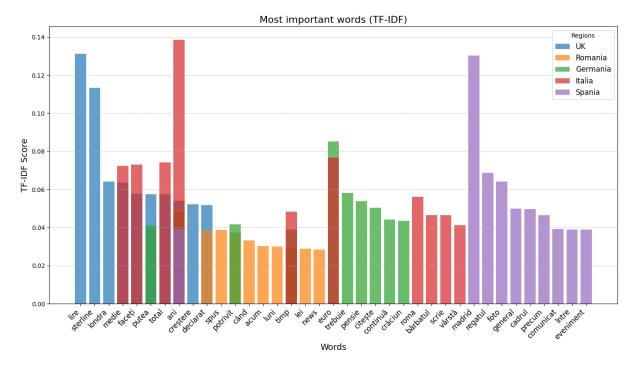


Figure 2: TF-IDF

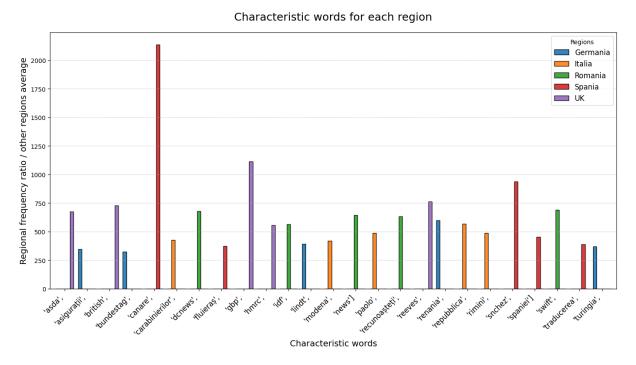


Figure 3: Characteristic Words