# What Kind Of Sociologist Is Your LLM?

# Task B: LLMS Display Promising But Limited Capabilities When Explaining News Through SOCIAL Theory. **Research Process**





STEP 2.

University of Amsterdam

#### **CONSISTENCIES WITHIN MODEL**

| simulation |  |
|------------|--|



Consistency between domain of knowledge and keyword selection on Llama: the density chart shows that LLama labels keywords in the appropriate theoretical domains consistently.



Inconsistency in Article Categorization Between Models: From the start, noticeable discrepancies in how news articles are sorted into contextual categories, underscoring a mismatch between them.



**Prompting: Meta-Data** We generate relevant information about the article that the LLM uses for its chain-of-thought approach: the main protagonists of the article, the summarization of the event in 2 keywords and the epistemological discipline that best matches the mentioned event.

#### STEP 3.

**Prompting: Analysis** Based on the previous step, we then let the LLM assess 3 possibilities for the following dimensions: • Sociologist, • Concept, • Connection

In the final internal iteration the LLM selects the sociologist that best matches the event, plus the reasoning, the aspect of the text that influenced its decision, the name of the concept, the concept in keywords and the

Consistency between domain of knowledge and keyword selection on Llama: the density chart shows that LLama labels keywords in the appropriate theoretical domains consistently.



The heat map, based on association scores with different sociological concepts, shows: LLaMa shows a strong association with "Cultural Capital", "Habitus" and "Risk Society". Both models have weak associations with concepts such as "Stigmatization" and "Right to the City".

| Edward O. Wilson - | 5                                          | Z    | Bruno Latour -    | 16                                         | - 100 z |  |
|--------------------|--------------------------------------------|------|-------------------|--------------------------------------------|---------|--|
| Charles Tilly -    | 4                                          | - 20 | Manuel Castells - | 16                                         |         |  |
| George Ritzer -    | 4                                          |      | Sherry Turkle -   | 14                                         | - 50    |  |
| Talcott Parsons -  | 4                                          | - 10 | Max Weber -       | 13                                         |         |  |
| Karl Marx -        | з                                          |      | Herbert Marcuse - | 12                                         |         |  |
| Peter Blau -       | 3                                          |      | Ulrich Beck -     | 12                                         |         |  |
| Durkheim –         | з                                          |      | Frantz Fanon -    | 11                                         |         |  |
| Émile Durkheim -   | 3                                          |      | David Harvey -    | 11                                         |         |  |
| Numb               | Number of Citations<br>Number of Citations |      |                   | Number of Citations<br>Number of Citations |         |  |

Inconsistency: The models show little overlap in choosing key authors and theories to explain news phenomena, underscoring a mismatch between them.



Inconsistency in Keywords Selection Between Models: Keywords help us understand a model's choice of concepts and theories. The choice of markedly different keywords by models, mirroring differences in contextual categorization, reflects how they interpret concepts and theories.

contribution of the concept to sociology STEP 4. **Data cleaning** : Having added the LLMs'

suggestions into one additional column in the dataset, we split the contents into several distinct columns. Due to different column structures, a thorough manual cleaning step was necessary.

> STEP 5. **Final results**

Our prompt produced various dimensions which built the basis of our analysis:

Analysis 1: Comparison of main authors and concepts.

Analysis 2: Co-Occurence Network by

#### MAIN TAKE-AWAYS

# LIMITATIONS

- General assessment: Overall, using large language models (LLMs) to interpret news with theory is interesting but comes with many limitations.
- Consistency within models: Individual models are consistent in linking keyword selections with seemingly coherent concepts and social theorists.
- Inconsistency across models: The models do not consistently select the same categories and keywords for labelling news articles.
- Future research: Future studies should focus on refining prompts 1) to achieve more consistency between models keyword identification and downstream theoretical explanations, and 2) address data processing challenges, possibly through multi-step approaches 3) explore biases in the selection of theories and authors.

# • **Prompt:** CoT showed promising capabilities in identifying various dimensions in one prompt; however, our prompt was long and technical - so it didn't work on two proposed LLMs - Qwen & Gemma, limiting our analysis to two models, Mistral and LLaMa. • Technical issues & time constraints: Rigorously assessing

capability of LLMs would require manual validation of suggested sociologists and concepts, beyond our resources and time frame. • Structure of output: Despite inclusion of delimiters in the prompt to give us structured outputs, the final output varied between

articles, which led to a tedious data cleaning step.

#### categories

Analysis 3: **Co-Citation Network by** authors

Analysis 4 : Comparison of super categories

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