

Key Linguistic Markers for Differentiating Schizophrenia, Schizoaffective, and Bipolar Disorders

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Topic and research question of the current work

- Analysis of spontaneous speech samples from patients with **schizophrenia** (SZ), **schizoaffective** (SAD) and **bipolar disorders** (BD), and healthy controls (HC)
- Which **linguistic features** are the most helpful in distinguishing between patient groups and controls?
- How effectively can we **automatically differentiate** between these groups based on a set of linguistic features?

Structure of the presentation

1. Brief description of the disorders in question
2. Corpus compilation and processing steps
 - Methods and analysis
3. Results: significance tests and machine learning experiments
4. Discussion of the results

Schizophrenia-bipolar spectrum disorders

- **Schizophrenia** (henceforth SZ): characterized by symptoms of delusions, hallucinations, disorganized speech or behavior, and impaired cognitive ability
- **Schizoaffective disorder** (henceforth SAD): mixed psychotic (hallucinations or delusions) and affective symptoms (mood episodes) → intermediate position between BD and SZ in the schizophrenia-bipolar spectrum
- **Bipolar disorder** (BD): characterized by episodes of mania, hypomania, and alternating or intertwining episodes of depression, possibly psychosis
- Common feature of the three disorders: cognitive deficits, impaired executive function → impaired verbal function

The corpus

- Hungarian database recorded by the Prevention of Mental Illnesses Interdisciplinary Research Group (University of Szeged) led by István Szendi
- **Spontaneous** speech recordings
- Task: describe the previous day
- Manually transcribed
- Part of the HuMenDisCo corpus (Szabó et al. 2023)

	SZ	SAD	BD	Control	All
Participants/Texts	27	14	15	21	77
Age; M(SD)	38.80(10.17)	41.43(9.73)	49.08(8.67)	36.42(10.49)	40.63(10.71)

Table 1: Basic demographic data of the corpus

Corpus processing steps

- Automatic linguistic analysis with magyarlanc (Zsibrita et al., 2013)
- The texts were split into sentences, tokenized, and the tokens were lemmatized and assigned a proper part-of-speech and morphological tag (lemmatization is especially important in the case of morphologically rich languages such as Hungarian)
- 17 basic **statistical features** (e.g., number of sentences, number and frequency of distinct lemmas compared to the number of words)
- 10 **speech-based features** (e.g., number of pauses and hesitations)
- 87 **morphosyntactic features** (e.g., parts-of-speech tags, number and frequency of superlative adjectives)
- Statistical analysis: pairwise t-tests for each feature and transcript
- Automatic classification: a random forest classifier of the WEKA package (Hall et al., 2009)

Results of the significance tests

- The ratio of **unknown words** (with an “unknown” POS tag) significantly higher in all patient groups than in the HC group
- SZ: fewest amount of sentences and tokens, BD: highest amount
- Unfilled **pauses** are significantly different in all patient groups compared to HC: SZ having a higher and BD and SAD having a lower rate
- SZ and SAD: somewhat more **function words** than HC and BD
- BD: highest rate of conjunctions → more complex sentences
- SZ: significantly lower rate of proper nouns than HC and SAD

Results of the machine learning experiments

1. Comparing all four groups
 - Overall accuracy was **50.65%**, outperforming the baseline (35.06%)
 - Most useful feature set: **speech-based** (59.74%)
 - Excluding one feature set: best result without statistical features (54.55%)

Feature type	Only	Without
Statistical	25.974	55.8442
Speech-based	59.7403	44.1558
Morphological	49.3506	54.5455
Syntactic	29.8701	55.8442
All	50.6494	50.6494

Table 2: Results of ablation analysis

Results of the machine learning experiments

2. Comparing two groups: HC vs. patient groups

- Overall accuracy was **75.32%**
- Most valuable feature set: **speech-based** (88.31%)
- Excluding one feature set: best result without syntactic (77.92%)

Feature type	Only	Without
Statistical	61,039	76,6234
Speech-based	88,3117	68,8312
Morphological	70,1299	75,3247
Syntactic	64,9351	77,9221
All	75,32474	75,3247

Table 3: Results of ablation analysis (HC vs. patient groups)

Discussion I: Comparison between HCs and patients

- Overall accuracy of 75.32%
- **Speech-based** feature set: most valuable (accuracy of 88.31%); better than all the features together
- When omitting **syntactic** features → the model performed notably better than in any other cases (77,92%) → syntactic information may not be as crucial or relevant for the specific task; it underscores the importance of feature selection
- Importance of the “**unknown**” **POS tag** (significantly lower rate in HC) → neologisms or disfluencies

Discussion II: Comparison of all four groups

- Overall accuracy of 50.65%
- Effectiveness of **speech-based** features again (accuracy of 59.74%)
 - Similarities between SZ and HC (higher rate of pauses and hesitations) and between BD and SAD (lower rate of pauses and hesitations)
- When omitting **statistical** features, the result outperformed the overall accuracy (54.54%)

Conclusions and limitations

- Analysis of spontaneous speech samples from patients with **schizophrenia** (SZ), **schizoaffective** (SAD) and **bipolar disorders** (BD), as well as healthy **controls** (HC) in Hungarian
- Rich linguistic feature set
- **Automatic classification** between said groups based on linguistic differences
- **Speech-based** features proved to be the most effective in both the classification tasks
- We can get meaningful results even without deep linguistic analysis
- Some of the results of **the ablation analysis** underscored the importance of feature selection
- The algorithm is more successful at distinguishing patients from HC
- Limitation: smaller sample size

References

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