



PERFORMANCE EVALUATION OF INDOBERT AND LDA FOR SENTIMENT ANALYSIS AND TOPIC MODELING ON HALODOC USER REVIEWS

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STATISTICS AND DATA SCIENCE STUDY PROGRAM SCHOOL OF DATA SCIENCE, MATHEMATICS, AND INFORMATICS IPB UNIVERSITY BOGOR 2025







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Bogor, June 2025

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ABSTRAK

DHIYA KHALISHAH TSANY SUWARSO. Evaluasi Kinerja IndoBERT dan LDA untuk Analisis Sentimen dan Pemodelan Topik pada Ulasan Pengguna Aplikasi Halodoc. Dibimbing oleh KHAIRIL ANWAR NOTODIPUTRO dan SACHNAZ DESTA OKTARINA.

Pertumbuhan teknologi mendorong adopsi layanan kesehatan digital, seperti Halodoc yang memungkinkan pengguna melakukan konsultasi dokter dan pengiriman obat secara daring. Untuk memahami persepsi pengguna terhadap layanan ini, penelitian ini melakukan analisis sentimen menggunakan IndoBERT dan pemodelan topik menggunakan Latent Dirichlet Allocation (LDA) terhadap alasan pengguna Halodoc di Google Play Store. Penelitian ini bertujuan mengevaluasi kinerja IndoBERT dalam mengklasifikasikan sentimen ulasan (positif, netral, negatif) serta mengidentifikasi topik-topik utama dalam ulasan positif dan negatif menggunakan LDA. Hasil penelitian menunjukkan bahwa IndoBERT dengan *class weight* memberikan performa terbaik dalam kondisi kelas tidak seimbang, dengan macro F1-score sebesar 0,8230 dan balanced accuracy sebesar 0,8274. Model ini mengklasifikasikan sentimen positif dan negatif dengan baik, tetapi performa pada sentimen netral rendah karena data terbatas, sifatnya ambigu, dan kosakata yang tumpang tindih. Selanjutnya, pemodelan topik dilakukan terhadap ulasan positif dan negatif. Model LDA dengan tokenisasi bigram menghasilkan nilai koherensi tertinggi dan topik yang lebih mudah diinterpretasikan karena mempertahankan pasangan kata yang bermakna. Pemilihan parameter K, α , dan β turut memengaruhi kualitas topik yang dihasilkan. Lima topik terbentuk dari ulasan positif dan tujuh topik dari ulasan negatif, dengan nilai koherensi masing-masing sebesar 0,6797 dan 0,6634. Temuan ini memberikan wawasan dalam memahami kebutuhan pengguna dan meningkatkan kualitas layanan Halodoc.

Kata kunci: analisis sentimen, halodoc, IndoBERT, *Latent Dirichlet Allocation* (LDA), pemodelan topik.

ABSTRACT

DHIYA KHALISHAH TSANY SUWARSO. Performance Evaluation of IndoBERT and LDA for Sentiment Analysis and Topic Modeling on Halodoc User Reviews. Supervised by KHAIRIL ANWAR NOTODIPUTRO and SACHNAZ DESTA OKTARINA.

Technological growth has driven the adoption of digital health services such as Halodoc, which allows users to consult doctors and receive medications online. To understand user perceptions of this service, this project conducts sentiment analysis using IndoBERT and topic modeling using Latent Dirichlet Allocation (LDA) on Halodoc user reviews from Google Play Store. This project aims to evaluate IndoBERT performance in classifying review sentiments (positive, neutral, negative) and to identify key topics in positive and negative reviews using LDA. Results show that IndoBERT with class weight achieved the best performance under class imbalance conditions, with macro F1-score of 0.8230 and balanced accuracy of 0.8274. The model effectively classified positive and negative sentiments, but struggled with neutral sentiment due to limited data, its ambiguity, and overlapping vocabulary. Subsequently, topic modeling was applied to the positive and negative reviews. LDA with bigram tokenization produced the highest coherence scores and more interpretable topics, by preserving meaningful word pairings. Parameter tuning of K, α , and β influenced topic quality. Five topics emerged from positive reviews and seven from negative reviews, with coherence scores of 0.6797 and 0.6634, respectively. These findings offer valuable insights into user needs and can support service improvement for Halodoc.

Keywords: halodoc, IndoBERT, Latent Dirichlet Allocation (LDA), sentiment analysis, topic modeling.





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DHIYA KHALISHAH TSANY SUWARSO

Skripsi to complete the requirement for graduation of Bachelor Degree in Statistics and Data Science Study Program

STATISTICS AND DATA SCIENCE STUDY PROGRAM
SCHOOL OF DATA SCIENCE, MATHEMATICS, AND INFORMATICS
IPB UNIVERSITY
BOGOR
2025



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Analysis and Topic Modeling on Halodoc User Reviews

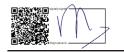
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Date of Thesis Defense: June 3rd 2025

Date of Graduation:





PREFACE

Praise and gratitude are sincerely expressed to Allah subhanaahu wa ta'ala for all His blessings, which have enabled the completion of this scientific work. The theme chosen for the project conducted from January 2025 to June 2025 is sentiment analysis and topic modeling, titled "Performance Evaluation of IndoBERT and LDA for Sentiment Analysis and Topic Modeling on Halodoc User Reviews".

The completion of this scientific work is due to various support and assistance from different parties. Therefore, the author would like to thank all parties who have helped in the completion of this project, including:

- 1. The author's parents (Gatot Suwarso and Popy Puspitasari), siblings, and extend family for their boundless support and prayers.
- 2. The author's advisors, Prof. Dr. Ir. Khairil Anwar Notodiputro, M.S. and Sachnaz Desta Oktarina, S.Stat, M.Agr.Sc., Ph.D. who have assisted, guided, directed, and supported during the preparation of this project.
- 3. The external examiner, Aulia Rizki Firdawanti, S.Stat., M.Si., who provided critical evaluation and thoughtful recommendations, which have substantially contributed to enhancing the clarity of this project.
- 4. The moderators, Akbar Rizki, S.Stat., M.Si., and Gerry Alfa Dito, S.Si., M.Si., who have offered constructive feedback and insightful suggestions during the colloquium and seminar, contributing to the improvement of this project.
- 5. All the lecturers and staff of the Department of Statistics who have imparted valuable knowledge and supported all the author's needs during the coursework and the preparation of this scientific work.
- 6. All friends, especially Angel, Diva, Kheni, Reyzha, Fuadiyah, Kamilah, Adiba, Arfiah, Oktavia, Azra and all peers from Statistics Batch 58, for their invaluable support and solidarity throughout this academic journey.
- 7. All parties who have assisted the author in preparing this scientific work, who cannot be mentioned one by one.

The author acknowledges that this scientific work is still not perfect. Therefore, the author apologizes if there are still any mistakes or weaknesses. May this scientific work benefit those who need it and contribute to the advancement of knowledge.

Bogor, June 2025

Dhiya Khalishah Tsany Suwarso







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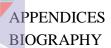
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