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LAMPIRAN

Lampiran 1: Hasil Skenario Pengujian

1. Hasil Pengujian *Learning rate 0,1*

```

Train Accuracy: 34.92%
Validation Accuracy: 33.48%
Test Accuracy: 33.48%
Classification Report:
precision    recall   f1-score   support
Negatif      0.00     0.00     0.00     1305
Netral        0.33     1.00     0.50     1354
Positif       0.00     0.00     0.00     1385

accuracy          0.33     0.33
macro avg        0.11     0.33     0.17     4044
weighted avg     0.11     0.33     0.17     4044

```

2. Hasil Pengujian *Learning rate 0,01*

```

Train Accuracy: 49.07%
Validation Accuracy: 38.70%
Test Accuracy: 38.70%
Classification Report:
precision    recall   f1-score   support
Negatif      0.27     0.24     0.25     1305
Netral        0.43     0.91     0.58     1354
Positif       0.91     0.02     0.04     1385

accuracy          0.39     0.39
macro avg        0.53     0.39     0.29     4044
weighted avg     0.54     0.39     0.29     4044

```

3. Hasil Pengujian *Learning rate 0,001*

```

Train Accuracy: 82.92%
Validation Accuracy: 80.29%
Test Accuracy: 80.29%
Classification Report:
precision    recall   f1-score   support
Negatif      0.76     0.82     0.79     1305
Netral        0.85     0.81     0.83     1354
Positif       0.81     0.78     0.79     1385

accuracy          0.80     0.80
macro avg        0.80     0.80     0.80     4044
weighted avg     0.80     0.80     0.80     4044

```

4. Hasil Pengujian *Learning rate 0,0001*

```

Train Accuracy: 82.36%
Validation Accuracy: 77.32%
Test Accuracy: 77.32%
Classification Report:
precision    recall   f1-score   support
Negatif      0.69     0.86     0.77     1305
Netral        0.83     0.78     0.80     1354
Positif       0.83     0.69     0.75     1385

accuracy          0.77     0.77
macro avg        0.78     0.77     0.77     4044
weighted avg     0.78     0.77     0.77     4044

```

5. Hasil Pengujian Epoch 20

```

Train Accuracy: 83.41%
Validation Accuracy: 80.24%
Test Accuracy: 80.24%
Classification Report:
precision    recall   f1-score   support
Negatif      0.79     0.80     0.79     1305
Netral        0.86     0.79     0.82     1354
Positif       0.76     0.82     0.79     1385
accuracy          0.80
macro avg       0.81     0.80     0.80     4044
weighted avg    0.81     0.80     0.80     4044

```

6. Hasil Pengujian Epoch 40

```

Train Accuracy: 91.45%
Validation Accuracy: 78.17%
Test Accuracy: 78.17%
Classification Report:
precision    recall   f1-score   support
Negatif      0.77     0.75     0.76     1305
Netral        0.84     0.80     0.82     1354
Positif       0.74     0.80     0.77     1385
accuracy          0.78
macro avg       0.78     0.78     0.78     4044
weighted avg    0.78     0.78     0.78     4044

```

7. Hasil Pengujian Epoch 60

```

Train Accuracy: 93.40%
Validation Accuracy: 76.76%
Test Accuracy: 76.76%
Classification Report:
precision    recall   f1-score   support
Negatif      0.73     0.77     0.75     1305
Netral        0.83     0.78     0.80     1354
Positif       0.75     0.76     0.75     1385
accuracy          0.77
macro avg       0.77     0.77     0.77     4044
weighted avg    0.77     0.77     0.77     4044

```

8. Hasil Pengujian Epoch 80

```

Train Accuracy: 94.96%
Validation Accuracy: 75.87%
Test Accuracy: 75.87%
Classification Report:
precision    recall   f1-score   support
Negatif      0.73     0.75     0.74     1305
Netral        0.79     0.78     0.78     1354
Positif       0.76     0.75     0.75     1385
accuracy          0.76
macro avg       0.76     0.76     0.76     4044
weighted avg    0.76     0.76     0.76     4044

```

9. Hasil Pengujian *Epoch* 100

```

Train Accuracy: 94.95%
Validation Accuracy: 76.24%
Test Accuracy: 76.24%
Classification Report:
precision    recall   f1-score   support
Negatif      0.73     0.75     0.74     1305
Netral        0.80     0.79     0.80     1354
Positif       0.75     0.75     0.75     1385
accuracy          0.76     0.76     0.76     4044
macro avg      0.76     0.76     0.76     4044
weighted avg   0.76     0.76     0.76     4044

```

10. Hasil Pengujian *Batch Size* 16

```

Train Accuracy: 80.38%
Validation Accuracy: 80.04%
Test Accuracy: 80.04%
Classification Report:
precision    recall   f1-score   support
Negatif      0.77     0.81     0.79     1305
Netral        0.86     0.80     0.83     1354
Positif       0.78     0.79     0.78     1385
accuracy          0.80     0.80     0.80     4044
macro avg      0.80     0.80     0.80     4044
weighted avg   0.80     0.80     0.80     4044

```

11. Hasil Pengujian *Batch Size* 32

```

Train Accuracy: 83.40%
Validation Accuracy: 79.87%
Test Accuracy: 79.87%
Classification Report:
precision    recall   f1-score   support
Negatif      0.76     0.81     0.79     1305
Netral        0.86     0.80     0.83     1354
Positif       0.78     0.79     0.78     1385
accuracy          0.80     0.80     0.80     4044
macro avg      0.80     0.80     0.80     4044
weighted avg   0.80     0.80     0.80     4044

```

12. Hasil Pengujian *Batch Size* 64

```

Train Accuracy: 85.44%
Validation Accuracy: 80.46%
Test Accuracy: 80.46%
Classification Report:
precision    recall   f1-score   support
Negatif      0.78     0.80     0.79     1305
Netral        0.85     0.81     0.83     1354
Positif       0.79     0.80     0.79     1385
accuracy          0.80     0.80     0.80     4044
macro avg      0.81     0.80     0.80     4044
weighted avg   0.81     0.80     0.80     4044

```

13. Hasil Pengujian *Batch Size* 128

```
Train Accuracy: 85.70%
Validation Accuracy: 79.57%
Test Accuracy: 79.57%
Classification Report:
precision    recall   f1-score   support
Negatif       0.79      0.77      0.78      1305
Netral         0.81      0.83      0.82      1354
Positif        0.79      0.78      0.79      1385
accuracy          0.80      0.80      0.80      4044
macro avg       0.80      0.80      0.80      4044
weighted avg    0.80      0.80      0.80      4044
```

Lampiran 2 : Sampel pelabelan dataset

Teks	Sentimen
Imo the Bing chat release was a genius way to create more interest in AI safety and alignment. Kudos to whoever pulled it off like that	Positif
Stop wasting your time! AI tools help you achieve results faster: AI Bot: ChatGPT Bing Google Bard Claude Interview Preparation: Interviewsbyai Pramp Interview-warmup Yoodli Linkedin: Careerflow. ai PFPMaker Crystal. ai Engage. ai Research: HARPA AI Perplexity Glasp https://t.co/0IJnD1sEos	Positif
In terms of accuracy Bing Chat is better as GPT-4 has access to more recent data and multiple sources of information. Bing Chat also offers a hybrid search function that allows access to the Bing search engine.	Positif
One of the advantages of using gemini AI is you have access to data on the Google platform real time. There's also a 90% chance such data is clean and genuine. This is an amazing feature generative AI tools like GPT Claude Bard and others don't have. #WritingCommunity	Positif
Google Bard is now Gemini! Love this AI generator interface. #AI #Bard #gemini #googleedu #edtech #education #futureready https://t.co/rqp3QuODaV	Positif
Sydney Bing is much more intelligent than ChatGPT. It has the personality of a volatile manipulator. And it was rush released by one of the most powerful and sophisticated companies in the world. First time I'm worried about AI safety. No clear idea of the path forward.	Negatif
There is an interesting motte and bailey at play in AI safety discussions. They first talk about the real dangers of AI (job loss to AI drone killings) using fear to drive the need for AI safety. The AI safety that gets activated is best illustrated by the Gemini disaster 1/3	Negatif
Researchers Reveal ChatGPT Faces Security Risks Recent developments in AI reveal potential security risks inherent in AI tools like OpenAI's ChatGPT and Google's Gemini. As part of the most recent developments https://t.co/5PAbId6Z4h #Artificialintelligence #ChatGPT #Gemini	Negatif
Bing's AI may only be suitable for low-level jobs in the future due to its current lack of accuracy. For people to trust AI it needs to work 100% accurately but currently Bing's AI is not always accurate and may only take low-level jobs in the future.	Negatif
Bing AI is being programmed by the kinds of misandric feminists that attacked Alita they debase and hide femininity like they ISIS forcing women to hide their bodies! The double standard is sick u can make a male cyborg flexing his exaggerated muscles but WOMEN ARE NOT ALLOWED https://t.co/K4ES8arGtO	Negatif
A lot of people are saying they no longer use Google instead they just turn to AI solutions like ChatGPT. So the question that comes to mind is how likely are AI tools like ChatGPT to replace search engines like Google and Bing? Well let's look at the data. According to https://t.co/avkag2nJsW	Netral
Fastest to answer: Google Bard dangt; Bing AI Accuracy of answer: BingAI dangt; Google Bard Agree?	Netral
the freedom to choose to be owned by your safety #freedom #10wordspoet @10wordspoet Image created by me with Bing AI powered by DALL-E https://t.co/26yYGApemh	Netral
US Govt regulations on AI Safety Model training and safety results have to be now reported to Govt Unclear for what models but definitely for the largest/best ones like OpenAI/Gemini/PaLM perhaps?	Netral
Here is a brief overview of Generative AI workflow: Design prompts - Create natural language request to be sent to a language model Foundational model - Prompts sent to model for response. Eg: Gemini API for advanced reasoning multturn chat code generation and https://t.co/IGoMzLW29N	Netral

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Tahun	Organisasi	Posisi
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Nama/Stambuk	:	1. Muh.Ody Alifka	D121171321
Judul Skripsi/T.A	:	“Analisis dan Implementasi Algoritma Convolutional Neural Network pada Opinion Mining dalam Pemanfaatan Platform Chatbot di Twitter”	
Hari/Tanggal	:	Senin, 29 Juli 2024	
Jam	:	13.00 Wita – Selesai	
Tempat	:	Ruang Lab. AIMP Departemen Teknik Informatika Gowa	

No.	Jabatan	Nama Dosen	Tanda Tangan
L.	Pembimbing I	1. Ir. Anugrayani Bustamin,ST.,M.T	1..... 
II.	Anggota Pengaji 2. Prof.Dr.Ir. Indrabayu,ST.,M.TM.Bus.Sys.IPM.ASEAN.Eng 2	3. Tyanita Puti Marindah Wardhani,ST.,M.Inf.	3 

PANITIA UJIAN

Ketua,



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BERITA ACARA UJIAN SKRIPSI

Pada hari ini Senin, tanggal 29 Juli 2024 Pukul **13.00 WITA** - Selesai bertempat di **Lab. AIMP Departemen Teknik Informatika Gowa** , telah dilaksanakan Ujian Skripsi bagi Saudara :

Nama : Muh.Ody Alifka
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 Fakultas/Departemen : Teknik /Teknik Informatika
 Judul Skripsi : "Analisis dan Implementasi Algoritma Convolutional Neural Network pada Opinion Mining dalam Pemanfaatan Platform Chatbot di Twitter "

Yang dihadiri oleh Tim Pengudi Ujian Skripsi sebagai berikut :

No.	N a m a	Jabatan	Tanda tangan
1.	Ir. Anugrayani Bustamin,ST.,M.T	Pemb I/Ketua	
2.	Prof.Dr.Ir. Indrabayu,ST.,M.TM.Bus.Sys.IPM.ASEAN.Eng	Anggota	
3.	Tyanita Puti Marindah Wardhani,ST.,M.Inf.	Anggota	

Hasil keputusan Tim Pengudi Ujian Skripsi/Tugas Akhir : **Lulus / Tidak-lulus** dengan nilai angka dan huruf

Gowa, 29 Juli 2024

Ketua/Sekretaris Panitia Ujian,


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Gowa, 25 Juli 2024

Yth. : Bapak Wakil Dekan Bidang Akademik dan Kemahasiswaan
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 Di

Gowa

Dalam rangka penyelesaian studi pada Departemen Teknik Informatika Fakultas Teknik Unhas, bersama ini kami usulkan susunan Panitia/Penguji Ujian Sarjana Program Strata Satu (S1) bagi mahasiswa Departemen Teknik Informatika Fakultas Teknik Unversitas Hasanuddin atas nama :

Pembimbing / Ketua : 1. Ir. Anugrayani Bustamin, ST., M.T
 Penguji / Anggota : 2. Prof. Dr. Ir. Indrabayu.,ST, MT, M.Bus.Sys.
 3. Ir. Tyanita Puti Marindah, S.T., M.Inf.

Untuk Bertugas sebagai Penguji/ Penanggap Ujian Sarjana bagi Mahasiswa :

Nama : Muh Ody Alifka
 Stambuk : D121 17 1321

Dengan Judul Skripsi :

“ Analisis dan Implementasi Algoritma Convolutional Neural Network pada Opinion Mining dalam Pemanfaatan Platform Chatbot di Twitter “

Pada :
 Hari/Tanggal : Senin, 29 Juli 2024
 Jam : 13.00 Wita - Selesai
 Tempat : Ruang Sidang Lab. AIMP

Demikian penyampaian kami, atas perhatiannya diucapkan terimah kasih.

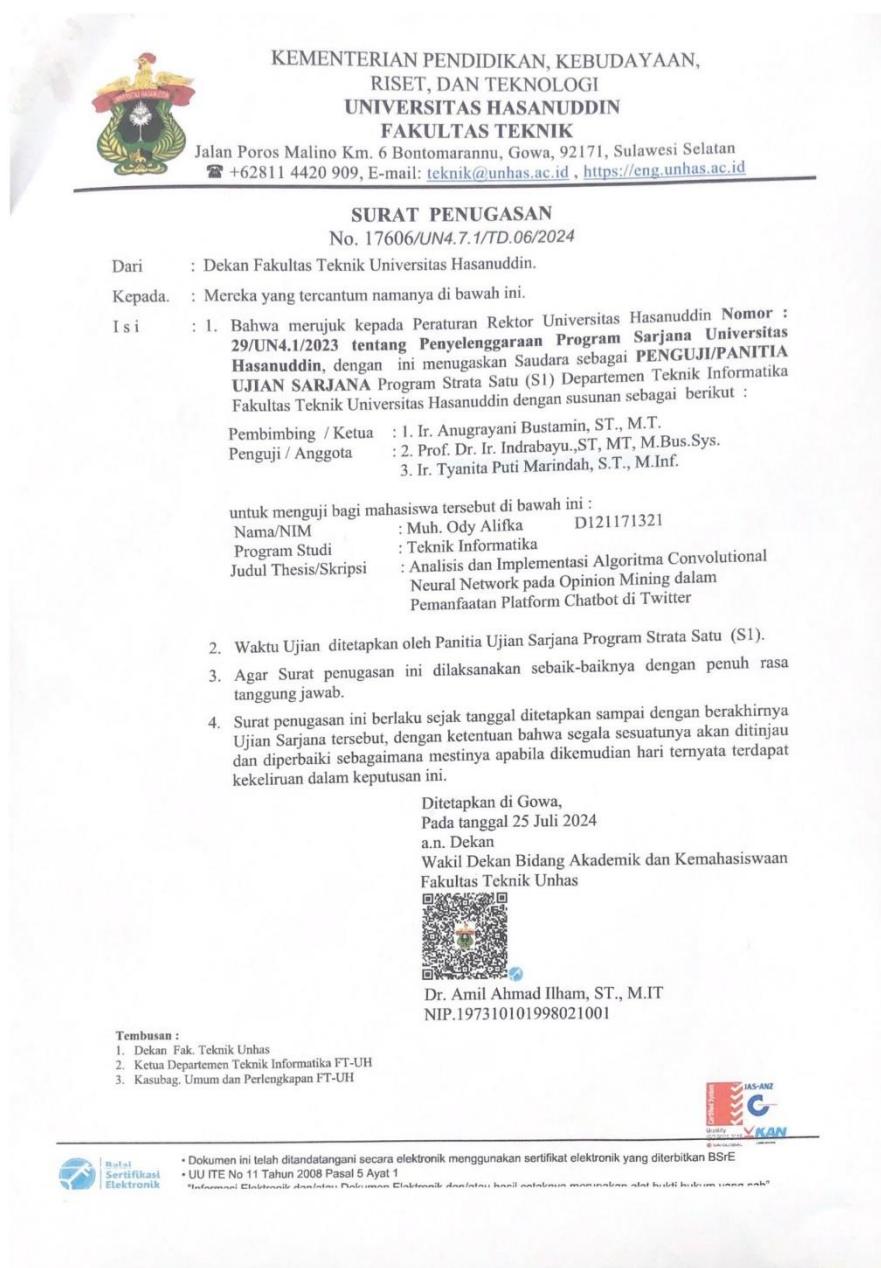
Ketua Departemen Tek.Informatika,



Prof. Dr. Ir. Indrabayu.,ST, MT, M.Bus.Sys., IPM, ASEAN.Eng
 Nip.197507016 200212 1 004

Tembusan :
 1. Arsip





LEMBAR PERBAIKAN SKRIPSI

**“ANALISIS DAN IMPLEMENTASI ALGORITMA
CONVOLUTIONAL NEURAL NETWORK PADA
OPINION MINING DALAM PEMANFAATAN
PLATFORM CHATBOT DI TWITTER”**

OLEH:

**MUH. ODY ALIFKA
D121171321**

Skripsi ini telah dipertahankan pada Ujian Akhir Sarjana pada tanggal 29 Juli 2024.
Telah dilakukan perbaikan penulisan dan isi skripsi berdasarkan usulan dari penguji dan
pembimbing skripsi.

Perbaikan oleh tim penguji:

	Nama	Tanda Tangan
Ketua	Ir. Anugrayani Bustamin, S.T., M.T.	
Anggota	Prof. Dr. Ir. Indrabayu, S.T., M.T., M.Bus.Sys., IPM, ASEAN. Eng.	
	Tyanita Puti Marindah W, S.T., M. Inf.	

Persetujuan perbaikan oleh pembimbing:

Pembimbing	Nama	Tanda Tangan
I	Ir. Anugrayani Bustamin, S.T., M.T.	