



UNIVERSIDAD DE LAS AMÉRICAS PUEBLA

ENGINEERING SCHOOL

DEPARTMENT OF INFORMATICS, ELECTRONICS AND MECHATRONICS

Desarrollo de Aplicaciones Móviles (LIS4012)

Project Progress

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Tasks Accomplished by Each Team Member by March 6th

Team member 1: Cecilia Soriano Tochimani

Cecilia has been focused on developing the app's user interface using SwiftUI in Xcode. She has designed and implemented various views, ensuring a seamless and intuitive user experience. Her work includes creating layouts, animations, and navigation flows to make the app visually appealing and user-friendly. Additionally, she has been refining the app's responsiveness and accessibility to ensure a high-quality user experience across different iOS devices.

Tasks done:

- The splash frame has been completed.
- The home view is in progress and has advanced to 30% completion.

Team member 2: Melisa Sampieri Espinoza

Melisa has been working on integrating the BERT pretrained model into the app for emotion recognition and tone analysis. She is responsible for adapting the model to fit the specific needs of the app, leveraging her experience with BERT. Additionally, she has been testing different hyperparameter configurations to optimize performance. She is also exploring ways to connect the model's output to the app's views for real-time suggestions while ensuring efficiency and accuracy.

Tasks done:

- Trained multiple versions of the BERT model to evaluate which performs best for our application.

Team member 3: Luis Felipe Jarquin Romero

Luis has been focused on integrating the OpenAI API to enhance the app's functionality, particularly for text rewriting and tone refinement. He is working on establishing a secure connection between the app and the API to ensure smooth data processing. Additionally, he is optimizing the API's performance to provide quick and accurate responses to user inputs. To achieve this, he has collaborated with another team member to research the best implementation strategies and ensure compliance with security standards. He has also been involved in approving and reviewing the progress made by other team members to ensure alignment with project goals.

Tasks done:

- Conducted research with a teammate on how to effectively implement the OpenAI API.

- Approved and reviewed the progress of different implementation aspects to ensure consistency and efficiency and responsible to make the documentation

Tasks Accomplished by Each Team Member by March 13th

Team member 1: Cecilia Soriano Tochimani

Cecilia has been focused on developing the app's user interface using SwiftUI in Xcode. She has designed and implemented various views, ensuring a seamless and intuitive user experience. Her work includes creating layouts, animations, and navigation flows to make the app visually appealing and user-friendly. Additionally, she developed another view for the app, keeping it clean and easy on the eyes.

Tasks done:

- Fixed errors in the home view.
- Implemented a feature where the identified sentiment is highlighted with color, fulfilling one of the app's requirements.

Team member 2: Melisa Sampieri Espinoza

Melisa worked on converting the BERT model into a CoreML so that it can be implemented in Swift UI.

Tasks done:

- Find versions of BERT to include it in the application
- https://huggingface.co/docs/transformers/model_doc/distilbert#tfdistilbertformaskedlm
- <https://apple.github.io/coremltools/docs-guides/source/convert-tensorflow-2-bert-transformer-models.html>
- <https://huggingface.co/dlaiymani/bert-base-ner-coreml>
- <https://github.com/huggingface/swift-coreml-transformers/tree/master>

Team member 3: Luis Felipe Jarquin Romero

Luis researched methods for cybersecurity. A log in is obviously necessary and there may be messages that the user would like to further protect. Also, it is important to encrypt users' passwords. After considering Argon2, PBKDF2, and bcrypt, we decided to go with bcrypt because of how easy it is to implement while maintaining strong security [1]. It is much faster than Argon2, as the hashing is done in less than one second [2].

Tasks done:

- Deciding on bcrypt for Swift UI called SwiftBcrypt.
- Updating documentation on GitHub.

Justification of the project

In developing VibeCheck, we focused on building a mobile-first experience that delivers real-time emotion recognition and tone improvement suggestions. Our priority was ensuring the app could perform fast, accurate, and privacy-conscious natural language processing (NLP) across multiple languages and platforms. To support this, we explored, evaluated, and integrated a range of transformer-based models, culminating in the selection and fine-tuning of compact yet capable pre-trained models.

We experimented with several BERT-based architectures and trained different model versions tailored specifically for emotion recognition. This iterative approach allowed us to evaluate trade-offs in model performance, size, and inference speed, ensuring the best fit for both mobile deployment and user experience. We successfully converted our selected model to CoreML, enabling seamless integration within our SwiftUI-based interface and allowing on-device processing—critical for maintaining user privacy and reducing server load.

Beyond just adopting existing technologies, we invested in optimizing real-world usability: we integrated sentiment visualization into the UI, added secure authentication with bcrypt for password management, and implemented OpenAI's GPT-4o mini model to support text rewriting and suggestion features under budget constraints.

Our work reflects a hands-on approach: evaluating model architectures, implementing training pipelines, handling model conversion and optimization, and embedding NLP functionality directly into the app. These efforts ensured that VibeCheck remains fast, responsive, and secure, while still providing high-quality emotion and tone analysis across a diverse user base.

Pretrained models

DistilBERT, a distilled version of BERT: smaller, faster, cheaper and lighter. DistilBERT is a small, fast, cheap and light Transformer model trained by distilling BERT base. It has 40% less parameters than google-bert/bert-base-uncased, runs 60% faster while preserving over 95% of BERT's performances as measured on the GLUE language understanding benchmark. VibeCheck,

requires emotion recognition in text and automated suggestions for tone improvement while being multilingual and efficient. DistilBERT is a great choice for use case due to the following reasons:

- Lightweight & Efficient for Mobile Applications
- Emotion Recognition & Sentiment Analysis
- Multilingual Capabilities
- Pre-trained Model with Fine-Tuning Flexibility.
- Compatibility with iOS (CoreML & TensorFlow Lite)
- Low Latency & Fast Inference
- Integration with API-based Workflows

DistilBERT is a smaller, faster, and cheaper variant of BERT, making it ideal for mobile applications where computational resources are limited. It retains 97% of BERT's accuracy while being 60% faster and 40% smaller, which reduces latency in real-time applications. Since the app requires real-time text analysis and inline suggestions, using a pre-trained DistilBERT model can save time and computational resources while allowing further fine-tuning for your specific needs. DistilBERT models can be converted to ONNX, TensorFlow Lite, or CoreML, making them compatible with iOS devices. DistilBERT can be deployed on-device (using CoreML/TensorFlow Lite) rather than a cloud-based API, ensuring privacy and compliance with data regulations.

- General Language Understanding Evaluation (GLUE) Benchmark: DistilBERT retains 97% of BERT's performance across multiple tasks, indicating its robust language understanding capabilities.
- IMDb Sentiment Analysis: DistilBERT achieves an accuracy of 92.82%, closely matching BERT's 93.46%. Wikipedia+5ar5iv+5Medium+5
- Stanford Question Answering Dataset (SQuAD) 1.1: DistilBERT attains an F1 score of 85.8 and an Exact Match (EM) score of 77.7, compared to BERT's 88.5 F1 and 81.2 EM scores. Medium+1ar5iv+1
- Stanford Sentiment Treebank (SST-2): A fine-tuned version of DistilBERT on SST-2 achieves an accuracy of 91.3%, whereas BERT reaches 92.7%.

BERT is designed to pre-train deep bidirectional representations from unlabeled text by jointly conditioning on both left and right context in all layers. As a result, the pre-trained BERT model can be fine-tuned with just one additional output layer to create state-of-the-art models for a wide range of tasks, such as question answering and language inference, without substantial task-specific architecture modifications.

VibeCheck mobile application focuses on emotion recognition, sentiment analysis, and tone improvement for social media posts. Since the app needs high-accuracy NLP processing, BERT (Bidirectional Encoder Representations from Transformers) is an excellent choice.

- BERT's Strength in Context Understanding.

- High Accuracy for Emotion & Sentiment Analysis.
- Pre-Trained Models for Faster Implementation.
- Multilingual Support.
- Ability to Generate Text Suggestions.
- Compatibility with iOS and Backend APIs.
- Secure & Private Processing.

BERT is a bidirectional transformer model, meaning it understands the full context of a word based on surrounding words. This is crucial for tone and emotion detection, where meaning can shift based on context. Multilingual BERT is trained on 104 languages, making it ideal for detecting tone across different linguistic and cultural contexts. BERT can be fine-tuned for text generation and paraphrasing, allowing it to rewrite user input with a more positive or neutral tone.

While BERT is heavier than DistilBERT, it can still be optimized for mobile using TensorFlow Lite or ONNX. BERT can be deployed in the cloud (reducing mobile device load) while maintaining secure API communication.

- BERT achieved 92.31% accuracy in classifying movie reviews as positive or negative.
- When analyzing tweets about airline services, a BERT-based model attained 87.1% accuracy.
- BERT-based models reached 93% accuracy in identifying sentiment in pandemic-related tweets.

These results demonstrate BERT's state-of-the-art performance in natural language processing, making it an excellent choice for VibeCheck's emotion recognition and text analysis features.

Implemented Views

Using SwiftUI and Xcode, the planned screens from the prototype were successfully implemented. Additionally, some features that were not originally considered were added, such as the ability to click on a recommendation and view the saved text.

For future updates, there are plans to change the color of the pop-up that appears when clicking on the underlined text and add the functionality to edit the text.



Figura 1. Home view

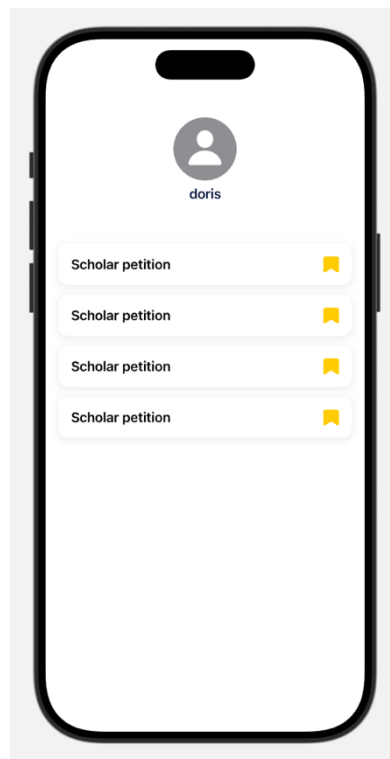


Figura 2. Profile View

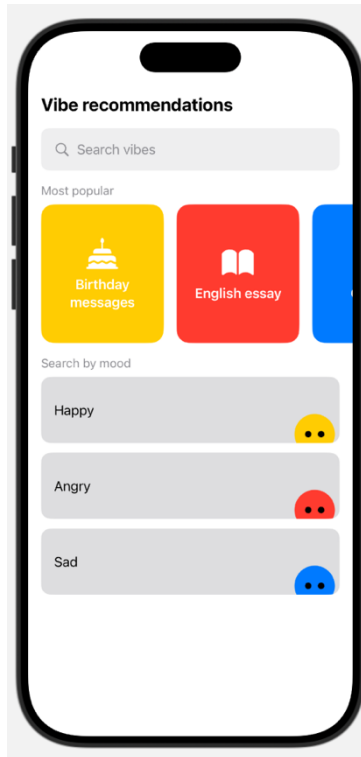


Figura 3. Recommendations View

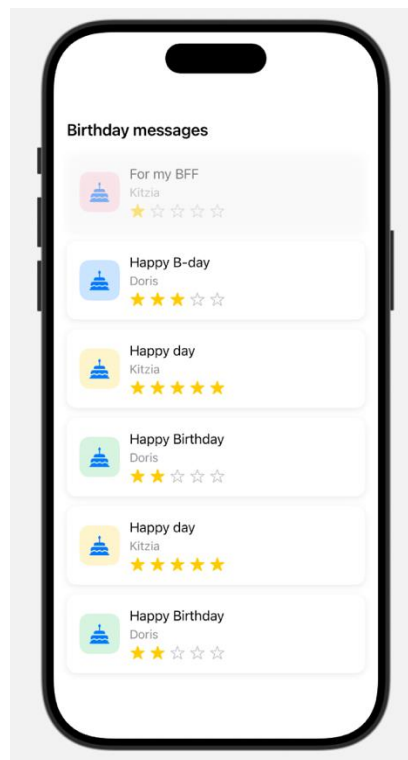


Figura 4. Recommendations Part 2 View



Figura 5. Save text View

API Choice

We chose the GPT-4o mini model for the OpenAI API because of how affordable it is considering our limited budget.

GPT-4.5	GPT-4o	GPT-4o mini
Largest GPT model designed for creative tasks and agentic planning, currently available in a research preview 128k context length	High-intelligence model for complex tasks 128k context length	Affordable small model for fast, everyday tasks 128k context length
Price	Price	Price
Input: \$75.00 / 1M tokens	Input: \$2.50 / 1M tokens	Input: \$0.150 / 1M tokens
Cached input: \$37.50 / 1M tokens	Cached input: \$1.25 / 1M tokens	Cached input: \$0.075 / 1M tokens
Output: \$150.00 / 1M tokens	Output: \$10.00 / 1M tokens	Output: \$0.600 / 1M tokens

Summary

User Interface Development

- Designed and implemented various views in SwiftUI.
- Fixed errors in the home view.
- Added a feature that highlights identified sentiments with colors.

Machine Learning Integration

- Trained multiple BERT model versions for emotion recognition.
- Converted the BERT model to CoreML for SwiftUI integration.
- Researched different BERT implementations and available models.

Backend & Security

- Researched and selected bcrypt for password encryption in Swift.
- Investigated OpenAI API implementation for text processing.
- Updated project documentation on GitHub.

Future Development

The next phase of development for VibeCheck will focus on building out the backend infrastructure to support scalable and secure natural language processing. This includes integrating our emotion recognition models into a robust backend system and developing custom APIs to handle real-time text analysis, tone suggestions, and user interactions. The backend will be responsible for receiving user input, running inference using our fine-tuned models, and returning structured responses that the mobile frontend can display efficiently. Additionally, we plan to implement API endpoints for account management, sentiment tracking history, and multilingual support. Depending on performance and privacy considerations, we will deploy models either on-device via CoreML or server-side using optimized frameworks like TensorFlow or ONNX. This backend development will ensure the application is modular, maintainable, and ready for future scaling, including cloud deployment, user analytics, and integration of additional NLP features.

References:

1. T. P. Batubara, S. Siburian, and R. T. A. Pasaribu, "Analysis Performance BCrypt Algorithm to Improve Password Security from Brute Force," *J. Phys.: Conf. Ser.*, vol. 1811, no. 1, p. 012129, 2021. Available: <https://doi.org/10.1088/1742-6596/1811/1/012129>.
2. Stytc, "Argon2 vs. bcrypt vs. scrypt: Which password hashing algorithm should you use?" *Stytc Blog*, May 2022. Available: <https://stytc.com/blog/argon2-vs-bcrypt-vs-scrypt>.

GitHub link: <https://github.com/LuisAtMcGill/vibecheck.git>