

# First Discussion

1. Collected diverse Personas Datasets from Kaggle (Persona chat(ConvAI2)-facebook AI)

Fine tuned with (Unsloth-LoRA Low-Rank Adaptation) phi3→Llama 3

Challenges: Time-consuming, Google Colab disconnections

integrated with Fast API backend Publish API with ngrok

Connected API with React Native Mobile App via Expo GO

conclusion: collect real data from WA,FB,twitter,telegram or create fake one form GPT4



by **Anas Saleh**

computation and Language
<i>first revised 2 Apr 2024 (this version, v5)</i>
<b>Investigating the Ability of Large Language Models to Express Personality Traits</b>
Xubo Cao, Cynthia Breazeal, Deb Roy, Jad Kabbara
As large language models (LLMs) in creating personalized chatbots, there has been limited research on evaluating the extent to which the behaviors of LLM-based agents consistently reflect specific personality traits. We consider studying the behavior of LLM-based agents which we refer to as LLM personas and present a framework to evaluate whether LLMs can generate content that aligns with their assigned personality profiles. To this end, we simulate distinct LLM personas based on their self-reported Big Five Inventory (BFI) personality test and a story writing task, and then assess their essays with automatic and human evaluation. Our results show that self-reported BFI scores are consistent with their designated personality types, with large effect sizes observed across five traits. Additionally, LLM personas can be used to serve as agents to simulate human behaviors, given the powerful ability to understand human instructions and provide personalized responses. We wonder whether LLMs can simulate a person in a higher form than simple human behaviors. Therefore, we aim to train an agent with the ability to act as a specific person instead of using limited prompts to instruct ChatGPT API. In this work, we introduce Character-LLM that teach LLMs to act as specific characters, such as Shakespeare, Julius Caesar, etc. Our method focuses on editing profiles as experiences of a certain character and training models to be personal simulators. As part of our approach, we build a test playground that interviews trained agents and evaluates whether the agents “textit{memorize}” their characters.
Artificial Intelligence
<b>Large-Scale Simulations of 1,000 People</b>
Yi Q. Zou, Aaron Shaw, Benjamin Mako Hill, Carrie Cai, Meredith Ringel Morris, Robb Willer, Percy Liang, Michael S. Bernstein
Behavioral simulation—general-purpose computational agents that replicate human behavior across domains—could enable broad applications in social science research. We present a novel agent architecture that simulates the attitudes and behaviors of 1,052 real individuals—applying large language models to qualitative interview transcripts. The agents replicate the attitudes and behaviors of the individuals that they represent. The generative agents replicate participants’ responses on a variety of topics. Participants’ agents replicate their own answers two weeks later, and perform comparably in predicting personality traits and outcomes in experimental research. Our results show that across racial and ideological groups compared to agents given demographic descriptions. This work provides a foundation for new tools that can be used in social science research.
Computers and Society
<i>first revised 2 Mar 2025 (this version, v6)</i>
<b>Character-LLM: Benchmarking the Personification of Large Language Models</b>
Tianqi Zhang, Qingqiu Li, Linyi Yang, Yuejie Zhang, Rui Feng, Liang He, Shang Gao, Yue Zhang
Large Language Models (LLMs) are recognized as systems that closely mimic aspects of human intelligence. This capability has attracted attention from the social sciences and humanities, leading to the use of LLMs to replace human participants in experiments, thereby reducing research costs and complexity. In this paper, we introduce a framework for benchmarking the personification of LLMs. We propose a strategy for constructing virtual characters’ life stories from the ground up, a Multi-Agent Cognitive Mechanism capable of simulating human behaviors. We propose a new evaluation method to assess human simulations from both self and observational perspectives. Experimental results demonstrate that our construction method aligns with their target characters. Our work is a preliminary exploration which offers great potential in practical applications. All the code and data are available at <a href="#">https://github.com/Character-LLM/Character-LLM</a> .
computation and Language
<i>first revised 9 Jul 2023 (this version, v5)</i>
<b>Large Language Models to Simulate Multiple Humans and Replicate Human Subject Studies</b>
Yi Q. Zou, Aaron Shaw, Benjamin Mako Hill, Carrie Cai, Meredith Ringel Morris, Robb Willer, Percy Liang, Michael S. Bernstein, Adam Tauman Kalai
A Turing Test, called a Turing Experiment (TE), for evaluating to what extent a given language model, such as GPT models, can simulate different aspects of human behavior. We propose a new TE, called a Multi-Human Turing Experiment (MTE), to evaluate the ability of a language model to simulate multiple humans. We use a sample of participants in human subject research. We carry out TEs that attempt to replicate well-established findings from prior studies. We use the MTE to compare how well different language models are able to reproduce classic economic, psycholinguistic, and social psychology experiments, including the Ultimatum Game, the Dictator Game, the Trolley Problem, the Prisoner’s Dilemma, the Wason Selection Task, the Milgram Shock Experiment, and Wisdom of Crowds. In the first three TEs, the existing findings were replicated using recent models, while the latter three were not. Our results suggest that large language models can simulate multiple humans and replicate human subject studies. This work provides a foundation for new tools that can be used in social science research.

- Huge LLMs
- Zero Shot
- Few Shot
- AI Agents
- Prompt Engineering
- Retrieval-Augmented Generation - RAG
- Memory Mechanisms

# Second Discussion



## Alternative Data Collection Methods

Used voice recordings from a team member

Transcribed speech using **Whisper Large V3 Turbo** (supports Arabic dialects)



## Avoided fine-tuning due to computational cost

Used AI inference to process extracted text

Used **Groq AI inference with Llama-3.3-70B Versatile**



## Voice Cloning for Personalized Speech

Converted text responses into speech using **XTTS model**

High-quality voice cloning, supports Arabic (closer to Gulf Arabic)



## Creating a Realistic Deepfake

Used **SadTalker** for deepfake video generation

Improved output quality with **Wave2Lip**

Challenges in Data Collection:

**Hard** to Collect Real data from specific Person from WA or Social media required sign in

conclusion: The professor said she **doesn't see enough effort in how we're mimicking the personality to make it respond like** us. She wants us to focus more on the simulation aspect and see if there's any **algorithm applied or related research papers**. We told her we can't use WhatsApp data, and although Telegram is an option, we don't use it much, so there's little data available. Social media posts and tweets also aren't helpful since people rarely share personal content there.

# Third Discussion in Ramadan

## Hard To Fine Tune With Our Resources

### ◆ Key Points about ConvAI2 Dataset:

- **ConvAI2** is a benchmark dataset designed to improve **persona-based dialogue generation**.
- Each speaker is assigned a **persona** (e.g., "I have four sisters") to encourage **more personalized responses**.
- It includes around **131,000 utterances** and **fewer than 5,000 unique personas**, making it relatively **small compared to open-domain datasets** like Reddit-Pushshift (1.2B+ utterances).
- Due to its size, ConvAI2 highlights the **data scarcity problem** in persona-based dialogue systems.
- The proposed model in the paper was evaluated on ConvAI2 and achieved:
  - **Competitive or better performance** than large models like **GPT-2**.
  - **30% improvement in perplexity**.
  - **Strong results using only 20–30% of the training data**, showing data efficiency.

after reading papers :

1- **First Paper** "Personalized Dialogue Generation with Persona-Adaptive Attention" **PAA (2022)**

### Summary

This paper enhances persona-based dialogue systems, which aim to generate replies that are consistent with a speaker's personality and conversation history. Unlike basic chat models, these systems must carefully balance the speaker's persona with the conversation context.

### The Authors Introduce

⚖️ Persona-Adaptive Attention (PAA): A new **attention mechanism that dynamically adjusts the importance of persona vs. dialogue context**.

🧠 Dynamic Masking: This technique filters out unnecessary info and acts as a regularizer, helping the model avoid overfitting.

### Results

Outperforms strong baseline models in both human and automatic evaluations.

Works well even with low-resource data (achieving near full-data performance with just 20–30% of training data).

Multiple model variants show that their adaptive weighting and masking strategies are essential.

Key Insight: Their method offers a smart and flexible way to generate personalized, consistent responses — even with limited training data.

- ◆ Dataset Introduced: **ConvAI - MINST - Reddit**

# Second Paper on Personalizing Dialogue Agents

## Paper Overview

"Personalizing Dialogue Agents: I have a dog, do you have pets too?" (2018)

## Results

This profile-based approach led to better next-utterance predictions, making conversations feel more human-like and connected.

The conversations became more personal and memorable, a big step up from generic chit-chat bots.

## Main Contributions

📋 **Persona Conditioning:** Dialogue agents are trained to include their own profile information in responses (e.g., "I like hiking").

🧑🧑 **Interactive Learning:** The agents also learn to infer the persona of the person they're talking to, based on the dialogue.

🧠 **Engagement Strategy:** Since the other person's persona is unknown at the start, the model proactively asks personal questions to elicit information — mimicking real social behavior.

## Dataset Introduced: **PersonaChat** which is part of **ConvAI**

A curated dataset where each speaker is given a profile with 5 sentences (e.g., "I love painting", "My favorite food is pizza").





Structure: Contains over 160,000 utterances and is used to train models to generate responses grounded in their persona.

Goal: Improve personalization and make chatbot dialogues more natural and engaging.

Access: Released publicly by Facebook AI Research, often cited in follow-up personalization studies.



# Third Paper on Characteristic AI Agents

 <b>Paper Overview</b>	 <b>Character100 Benchmark</b>	 <b>Evaluation Metrics</b>
"Characteristic AI Agents via Large Language Models" (2024)	100 most-visited people on Wikipedia	Measuring character trait embodiment
 <b>Comprehensive Experiments</b>	Testing LLM performance in character simulation	

## 3-Third Paper "Characteristic AI Agents via Large Language Models" (2024)

 Summary:

The research paper proposes a framework for creating AI agents with distinct and consistent personality traits. This framework is designed to be applied *individually* to each character.

This paper explores how Large Language Models (LLMs) can be **used to simulate real-life characters** in chatbot systems — **not just generic roles, but detailed personality-driven agents**. Although commercial tools exist for role-based chatbots, academic research has lagged behind.

The research does not focus on "training" the large language model from scratch or even extensively fine-tuning it. Instead, it relies on pre-trained large language models (pre-trained LLMs) and develops methods to guide and provide them with the necessary information to embody the desired personality. The specific methods mentioned or inferred from the nature of the research include:

- Detailed Character Profiles:** They create a precise and structured description of the character. This description includes multiple aspects such as:
  - Background
  - Personality Traits (e.g., using the OCEAN model or similar)
  - Goals and Motivations
  - Relationships
  - Knowledge and Expertise
  - Memory: An important component for maintaining consistency. The research might include a memory module to store and retrieve information about past interactions or significant events in the character's "life."

```
[
{
  "text": "Adele has cited the Spice Girls as a major influence in regard to her love and passion for music , stating that \" they made me what I am today \"",
  "sentiment": "neutral",
  "emotion": "sadness",
  "topic": "career",
  "persona": "Adele"
},
{
  "text": "On 1 October 2021 , projections and billboards of the number \" 30 \" appeared on significant landmarks and buildings in different cities around the world",
  "sentiment": "positive",
  "emotion": "neutral",
  "topic": "technology",
  "persona": "Adele"
},
{
  "text": "At the end of 2016 , Billboard named Adele Artist of the Year for the third time , and also received the Top Billboard 200 album . 25 was the best selling album in the world",
  "sentiment": "positive",
  "emotion": "surprise",
  "topic": "technology",
  "persona": "Adele"
},
{
  "text": "At the 36th Brit Awards in London on 24 February , Adele received the awards for British Female Solo Artist , British Album of the Year for 25 , British Artist of the Year for 25 , and Best British Album",
  "sentiment": "positive",
  "emotion": "fear",
  "topic": "gender",
  "persona": "Adele"
},
]
```

- Profile-Guided Prompt Engineering:** This detailed profile is used to formulate prompts that are given to the large language model. These prompts guide the model to generate responses (speech, potential actions) that are consistent with all aspects of the character defined in the profile.
- Agent Architecture:** The research might propose a specific architecture for the agent that includes different components (such as a memory unit, a planning/decision-making unit, in addition to the base language model) working together to produce consistent and distinct character behavior.

◆ Dataset Introduced: **Character100**

Description: A benchmark dataset consisting of **100 of the most-visited people on Wikipedia**, used to train and evaluate LLMs in simulating real-life personalities.

Purpose: Enables role-playing and character emulation by LLMs.

Contents: Includes persona profiles and contextual data about each character.

Access: Publicly available at the project's GitHub: Character100 Benchmark

**LoRA vs Fine Tuning - PAA**

**ConvAI2 vs Personachat vs Character 100**

**BlenderBot vs Llama**

searching about models :

# BlenderBot 3 Model Overview

## Internet Access

Can search the web for up-to-date information.

## Long-term Memory

Remembers facts about users to personalize conversations.

## Learning from Users


Improves over time by learning from real interactions.


## Open Source

Meta released the code and model to support research.

## More Natural Dialogue

Trained to generate human-like responses.

1-  BlenderBot 3 – Overview BlenderBot 3 is a chatbot developed by [Parl.ai](#), part of Meta's research. Meta AI (formerly Facebook) and released in August 2022. It's part of Meta's research into open-domain conversational AI.

 Key Features: Internet Access: Can search the web for up-to-date information.

Long-term Memory: Remembers facts about users to personalize conversations.

Learning from Users: Improves over time by learning from real interactions.

Open Source: Meta released the code and model to support research.

More Natural Dialogue: Trained to generate human-like responses.



# Llama with RAG and Vector Store Implementation

1

## Vector Store Creation

Creating a **Vector Store** from the data we've prepared using FAISS + LangChain.



## RAG Pipeline

Using the **RAG** Pipeline with LLaMA + Retriever + Memory.



## Analysis Models

"sentiment\_model = "cardiffnlp/twitter-roberta-base-sentiment (**Sentiment**) تحليل مشاعر

"style\_model = "j-hartmann/emotion-english-distilroberta-base (**Emotion**) تحليل أسلوب

"topic\_model = "facebook/bart-large-mnli (**Topic**) تصنيف موضوع



## Memory Options

ConversationBufferMemory context لو عايز تحافظ على الـ

ConversationSummaryMemory لو عندك سياق طويل

أو تبني memory-specific store لكل persona

5

## example:

```
{ "content": "Sure, I can help you with that.",
```

```
  "persona": "Alex",
```

```
  "sentiment": "positive",
```

```
  "emotion": "joy",
```

```
  "topic": "career" }
```

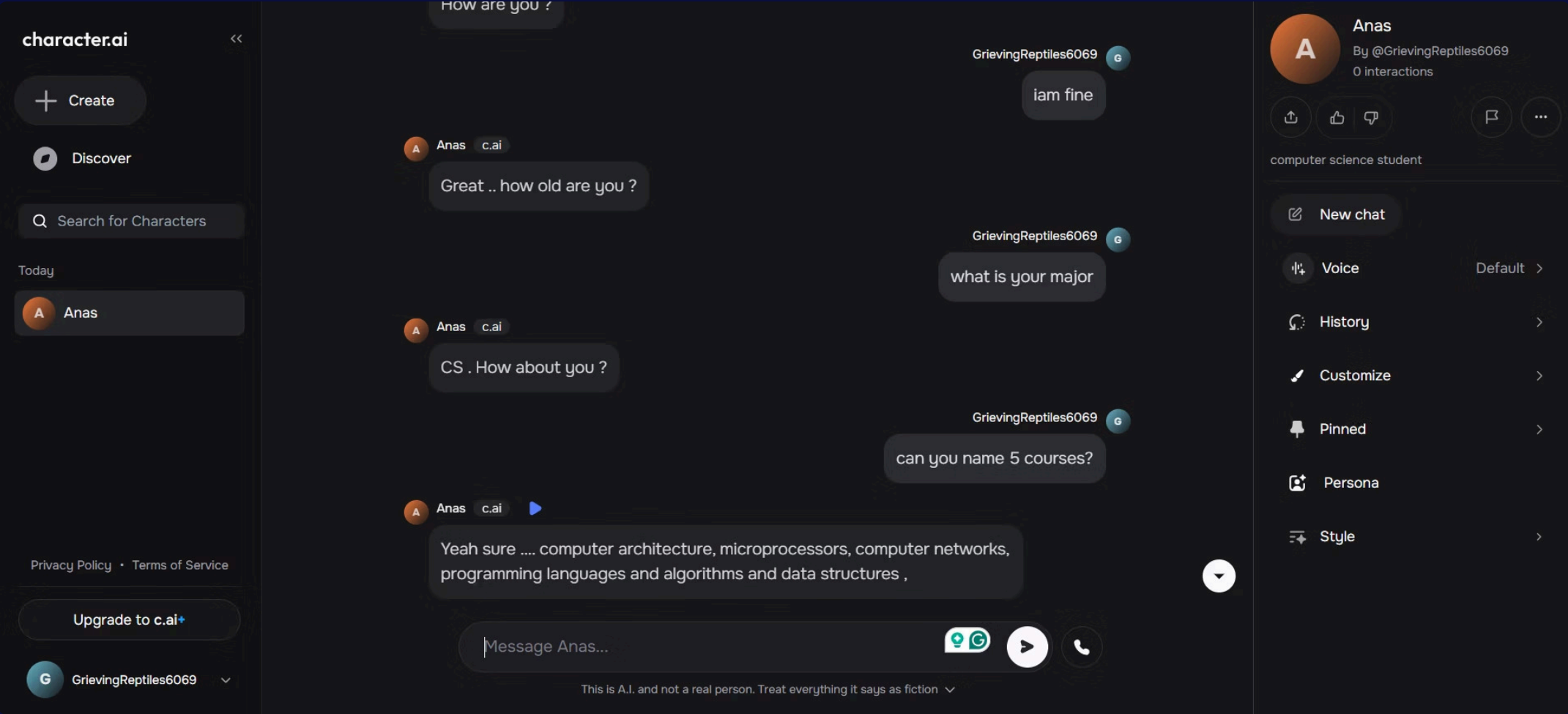
2- unsloth/Meta-Llama-3.1-8B-bnb-4bit + RAG + Vector Store

Creating a Vector Store from the data we've prepared using FAISS + LangChain. Using the RAG Pipeline with LLaMA + Retriever + Memory.

# Future Steps for Model Enhancement

- **Expand training data:** Use datasets like PersonaChat and ConvAI for diversity. and merging.
- **Enhance memory and retrieval:** Improve context-awareness with better RAG components.
- **Platform integration:** Develop frontend/backend for web and mobile using Streamlit.
- **User feedback:** Implement questionnaires to refine dialogue personalization.
- **Experimentation:** Attempt LoRA fine-tuning on BlenderBot 3 using Hugging Face.

# Character.ai



arXiv

> cs > arXiv:2201.08239

Search...

Help | Adv

Computer Science > Computation and Language

[Submitted on 20 Jan 2022 (v1), last revised 10 Feb 2022 (this version, v3)]

LaMDA: Language Models for Dialog Applications

Romal Thoppilan, Daniel De Freitas, Jamie Hall, Noam Shazeer, Apoorv Kulshreshtha, Heng-Tze Cheng, Alicia Jin, Taylor Bos, Leslie Baker, Yu Du, YaGuang Li, Hongrae Lee, Huaixiu Steven Zheng, Amin Ghafouri, Marcelo Menegali, Yanping Huang, Maxim Krikun, Dmitry Lepikhin, James Qin, Dehao Chen, Yuanzhong Xu, Zhifeng Chen, Adam Roberts, Maarten Bosma, Vincent Zhao, Yanqi Zhou, Chung-Ching Chang, Igor Krivokon, Will Rusch, Marc Pickett, Pranesh Srinivasan, Laichee Man, Kathleen Meier-Hellstern, Meredith Ringel Morris, Tulsee Doshi, Renelito Delos Santos, Toju Duke, Johnny Soraker, Ben Zevenbergen, Vinodkumar Prabhakaran, Mark Diaz, Ben Hutchinson, Kristen Olson, Alejandra Molina, Erin Hoffman-John, Josh Lee, Lora Aroyo, Ravi Rajakumar, Alena Butryna, Matthew Lamm, Viktoriya Kuzmina, Joe Fenton, Aaron Cohen, Rachel Bernstein, Ray Kurzweil, Blaise Aguera-Arcas, Claire Cui, Marian Croak, Ed Chi, Quoc Le

We present LaMDA: Language Models for Dialog Applications. LaMDA is a family of Transformer-based neural language models specialized for dialog, which have up to 137B parameters and are pre-trained on 1.56T words of public dialog data and web text. While model scaling alone can improve quality, it shows less improvements on safety and factual grounding. We demonstrate that fine-tuning with annotated data and enabling the model to consult external knowledge sources can lead to significant improvements towards the two key challenges of safety and factual grounding. The first challenge, safety, involves ensuring that the model's responses are consistent with a set of human values, such as preventing harmful suggestions and unfair bias. We quantify safety using a metric based on an illustrative set of human values, and we find that filtering candidate responses using a LaMDA classifier fine-tuned with a small amount of crowdworker-annotated data offers a promising approach to improving model safety. The second challenge, factual grounding, involves enabling the model to consult external knowledge sources, such as an information retrieval system, a language translator, and a calculator. We quantify factuality using a groundedness metric, and we find that our approach enables the model to generate responses grounded in known sources, rather than responses that merely sound plausible. Finally, we explore the use of LaMDA in the domains of education and content recommendations, and analyze their helpfulness and role consistency.

Subjects: **Computation and Language (cs.CL)**; Artificial Intelligence (cs.AI)

Cite as: [arXiv:2201.08239 \[cs.CL\]](#)  
(or [arXiv:2201.08239v3 \[cs.CL\]](#) for this version)  
<https://doi.org/10.48550/arXiv.2201.08239>

# Prompt Engineering + Inference of models + Questionnaire + Memory & RAG

# Recent Project Milestones

- **Voice Transcription:** Users can now send voice recordings with automatic text transcription. Whisper, CoquiTTS, gTTS
- **VoiceMood Avatars:** Animated avatars appear during responses; lip syncing is actively being developed.
- **Enhanced Memory:** Chat history improves context, enabling more personalized model responses.
- **Telegram Data:** Data integration has begun, with full implementation scheduled for Eid.
- **Character Memory:** Each virtual character now features its own distinct memory system.





Google Docs



WhatsApp Video 2025-05-09 at 00.56.48\_0fdc889b.mp4



Google Docs



WhatsApp Video 2025-05-09 at 00.57.06\_5700cabc.mp4