
Semantic Glitch: Agency and Artistry in an Autonomous Pixel Cloud

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Abstract

While mainstream robotics pursues metric precision and flawless performance, this paper explores the creative potential of a deliberately “lo-fi” approach. We present the “Semantic Glitch,” a soft flying robotic art installation whose physical form, a 3D pixel-style cloud, is a “physical glitch” derived from digital archaeology. We detail a novel autonomous pipeline that rejects conventional sensors like LiDAR and SLAM, relying solely on the qualitative, semantic understanding of a Multimodal Large Language Model to navigate. By authoring a bio-inspired personality for the robot through a natural language prompt, we create a “narrative mind” that complements the “weak,” historically loaded body. Our analysis begins with a 13-minute autonomous flight log, and a follow-up study statistically validates the framework’s robustness for authoring quantifiably distinct personas. The combined analysis reveals emergent behaviors, from landmark-based navigation to a compelling “plan-to-execution” gap, and a character whose unpredictable, plausible behavior stems from a lack of precise proprioception. This demonstrates a lo-fi framework for creating imperfect companions whose success is measured in character over efficiency.

1 Introduction

In an era where digital imagery relentlessly pursues high fidelity, why has the “pixel” aesthetic, born from technical limitations, sparked a persistent wave of retro-futurism? [5, 18] Furthermore, when a symbol composed of pixels, which should exist on a two-dimensional screen, suddenly acquires a physical body and floats among us like a seemingly autonomous creature, how does our relationship with it, and our perception of the virtual and the real, change? [22, 17] This paper explores these questions by detailing the creation and behavior of the “Pixel Cloud,” a soft robotic art installation that gains its physical autonomy from a Multimodal Large Language Model (MLLM). Our approach to authoring an agent’s character builds on artistic and scientific explorations into crafting lifelike [3, 4], emergent behaviors for interactive robotic agents. This work does not aim to solve any practical problem. Instead, it follows the “Speculative Design” philosophy advocated by Anthony Dunne and Fiona Raby [2], functioning as a “speculative object” to provoke public imagination and debate. It poses a series of “what if” questions: What if the untouchable digital “cloud” had a visible, fragile, physical body? What if the symbols from our digital childhood memories gained physical autonomy? By combining media archaeology [18] with robotics, the core thesis of this work is that through a “physical hack” [21] of the pixel, we can reveal and reshape the increasingly complex “entangled agencies” [23] among humans, machines, and the environment.

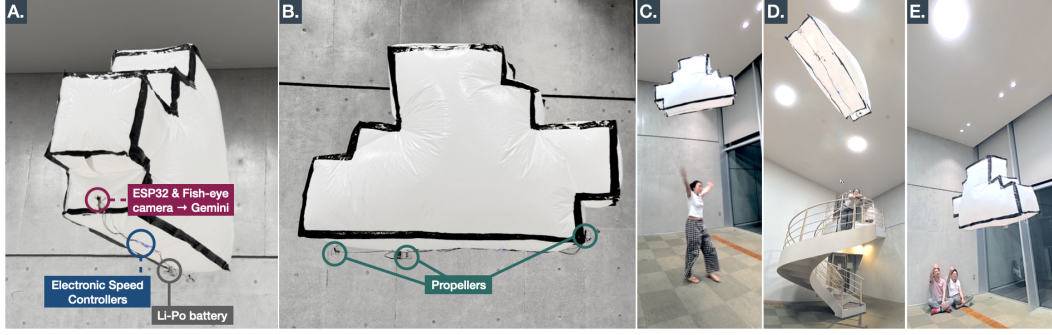


Figure 1: The “Semantic Glitch” hardware, flight behavior, and first-person perspective with MLLM-generated reasoning. (A, B) The robot’s body, showing the placement of the ESP32 with a fish-eye camera, electronic speed controllers, Li-Po battery, and propeller modules. (C, D, E) The robot in flight, demonstrating its interaction with the environment and its “perspective-dependent morphological illusion.”

Grounded in media archaeology and speculative design [20, 19], this paper traces the “Pixel Cloud” from its “physical glitch” body to its narrative AI mind, then analyzes both a detailed autonomous flight case study and an expanded persona validation. We conclude by considering how this “lo-fi” approach reframes robotic agency around character, plausibility, and relational presence.

2 The Body: A Deliberate “Physical Glitch”

The robot’s physical form is a deliberate “physical glitch,” designed to embody the “Yowai Robotto” (Weak Robot) philosophy by rejecting metric precision in favor of character [7, 8, 10, 12]. A core engineered feature is its “perspective-dependent morphological illusion”: from one angle, it appears as a 2D pixel image, but as it rotates, its 3D voxel structure is revealed (Fig. 1 C-E). This effect translates a software “error” into a tangible, repeatable imperfection. Constructed as a soft, fragile helium blimp, its form is intentionally “weak” to invite empathetic interaction [12, 13]. As later sections show, this physical weakness also sets up a productive tension between semantic intention and embodied execution.

3 The Mind: Navigation as Bio-Inspired Narrative

Rejecting Metric Precision: The conventional path to robotic autonomy involves building a precise, mathematical model of the world. This is typically achieved with a suite of metric sensors (such as LiDAR or infrared depth sensors) and complex algorithms like SLAM (Simultaneous Localization and Mapping), a technology envisioned as a future step in the project’s initial conceptualization. We deliberately rejected this path. A SLAM-based robot, with its metric geometric understanding, would be philosophically “out of character.” Its calculated, optimal movements would be incongruous with the artifact’s ephemeral nature, breaking the illusion of an animate entity. Therefore, to maintain the “weak robot” concept, the mind’s perception had to be as abstract as the physical form.

The “Lo-Fi” Semantic Engine: In place of a complex, sensor-heavy system, we embraced a framework of stateful semantic reasoning. The agent’s autonomy is powered by a novel, two-stage “lo-fi” pipeline that separates global scene understanding from local decision-making.

The entire control loop is orchestrated by a host computer (MacBook Pro, M4 Max, 64GB RAM) running a Python script, which communicates with the robot’s XIAO ESP32S3 core. The ESP32S3 is responsible only for low-level tasks: streaming video from its camera (160° fisheye lens) and actuating its propellers via WebSocket commands. All cognition is offloaded to the remote Gemini 2.5 Flash API, subject to its terms of use, transforming it into a stateful, MLLM “mind.”

This is achieved through two distinct phases, as illustrated in Figure 2. First, the **Preamble Stage** performs zero-shot spatial mapping. Upon initialization, the system begins a stateful ChatSession with the Gemini API. It sends the PREAMBLE_PROMPT along with a single 360° panoramic image

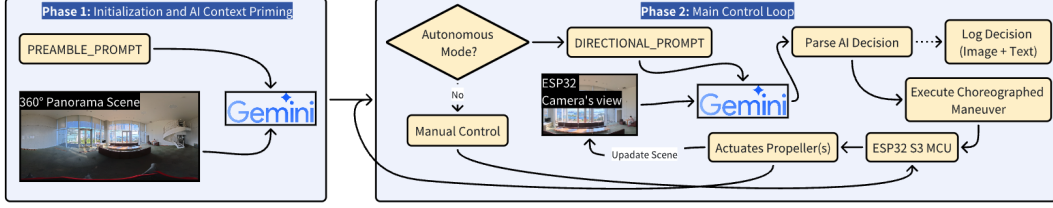


Figure 2: The two-phase semantic reasoning pipeline. Phase 1 (Initialization): A single 360° panorama and a PREAMBLE_PROMPT are sent to the Gemini API to establish a stateful “mental map”. Phase 2 (Control Loop): In a continuous loop, the system uses the live camera view and a DIRECTIONAL_PROMPT to generate context-aware actions, which are then logged and executed by the robot.

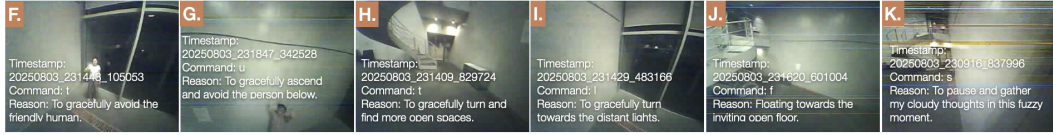


Figure 3: Key moments from the agent’s first-person perspective, linking its visual input to its logged decisions. (F) Lateral avoidance of a person. (G) Vertical avoidance of a person. (H) A corrective turn near the staircase, illustrating the plan-to-execution gap. (I) Goal-oriented navigation towards distant lights. (J) Seeking open space. (K) A moment of contemplative inaction.

of the environment. This one-time action tasks the AI with performing a high-level analysis of the entire operational area, identifying boundaries, major landmarks, open fly-zones, and obstacles before any physical movement occurs. This establishes a persistent “mental map” within the AI’s chat context, a process that took 2.81 seconds to complete during our experiment.

Second, the **Directional Stage** handles context-aware deliberation. In the main operational loop, we recast navigation as a continuous Visual Question Answering (VQA) problem [11]. For each decision, the DIRECTIONAL_PROMPT is posed as the “question,” which the MLLM “answers” by interpreting the “vision” context from the ESP32S3’s live video frame. The resulting answer, containing both a command and a narrative reason, is informed by the global spatial map established in the preamble stage. This creates a continuous, state-aware feedback loop where local perception is fused with global memory. During operation, this decision loop exhibited a mean latency of 2.8 ± 0.3 seconds, quantitatively defining the agent’s deliberate, non-continuous cognitive cycle.

Prompt Engineering for Hierarchical Cognition: The agent’s hierarchical cognitive process is authored entirely through two carefully engineered natural language prompts. These prompts define the function of each stage of the reasoning pipeline.

The PREAMBLE_PROMPT serves as the high-level cartographer, instructing the model to deconstruct the panoramic scene into a structured, semantic map. The DIRECTIONAL_PROMPT acts as the low-level navigator. It instructs the agent to use its established “mental map” as prior knowledge when interpreting the live camera feed to make an immediate choice. The prompt explicitly defines the agent’s available actions; the full, verbatim text for both the PREAMBLE_PROMPT and DIRECTIONAL_PROMPT is provided in Appendix A.

This two-prompt structure fundamentally changes the nature of navigation. The AI does not simply react to pixels or hard-coded logic; it situates its local perception within a persistent, global narrative, allowing for more sophisticated, long-term reasoning without the overhead of conventional mapping algorithms. Ultimately, this pipeline serves as a new model for authoring complex AI behavior, where high-level strategic goals and low-level personality traits can be defined and layered through natural language [16]. By representing all outputs as text, our work aligns with a broader trend of using language as a unified interface for robotic control, though we apply it here to generate character-rich narrative instead of precise coordinates.

4 Analysis of Emergent Dialogue: A Case Study in Stateful Navigation

This section analyzes a 13-minute continuous operational log to demonstrate how the synthesis of a lo-fi form and a stateful mind produces a uniquely plausible form of artificial life, using examples illustrated in Figure 3.

Goal-Oriented Navigation and Landmark Use: The logs confirm that the Preamble stage was successful. The agent consistently uses landmarks identified in the 360° panorama to inform its navigation, demonstrating effective use of its “mental map.” This is evident in decisions where it actively seeks out desirable zones, such as when turning with the reason ‘t, To gracefully turn towards the distant lights’ (Fig. 3-I) or moving forward because it is ‘f, Floating towards the inviting open floor’ (Fig. 3-J). The spiral staircase was a particularly prominent landmark, with the agent often reasoning about it as a point of interest (e.g., ‘l, To drift away from the wall and admire the elegant spiral’). This behavior shows the agent is not just reactively avoiding obstacles but is executing long-term, goal-oriented exploration based on the initial context.

Emergent Social Robotics and Character: The “bio-inspired” personality authored in the prompts resulted in sophisticated social behaviors. The agent recognized and reacted to humans in a way that was consistent with its prompt-defined nature. All interactions involving human observers were conducted under a protocol approved by our university’s Institutional Review Board. **Dynamic Human Avoidance:** The agent employed varied strategies when encountering people. On one occasion, it chose a lateral maneuver: ‘t, To gracefully avoid the friendly human’ (Fig. 3-F). In a different situation, it opted for a vertical solution: ‘u, To gracefully ascend and avoid the person below’ (Fig. 3-G). This demonstrates a flexible decision-making capability, choosing different actions for similar problems based on the specific context. **Contemplative Behavior:** The agent’s character was further revealed in moments of inaction. The log contains entries like ‘s, To pause and gather my cloudy thoughts in this fuzzy moment’ (Fig. 3-K). These are not error states; they are authored behaviors that give the agent a plausible, non-utilitarian, and creature-like quality, directly supporting the “Yowai Robotto” concept.

The Plan-to-Execution Gap and the “Yowai Robotto”: The most notable “glitchy” behaviors arise from the conflict between the agent’s high-level semantic understanding and its lack of low-level physical self-awareness. The agent knows what it wants to do but not precisely how to do it. For instance, after successfully navigating near the staircase, it decides to make a corrective turn with the command ‘t, To gracefully turn and find more open spaces’ (Fig. 3-H). The image shows it is close to the structure, and while the intention is correct, the subsequent maneuver is clumsy. It lacks the precise proprioceptive knowledge of its own momentum or the physical dynamics of its forward-arc maneuver. This constant “wrestling” with its own physical form—a direct result of the plan-to-execution gap—is a clear and authentic expression of a “weak” agent that feels organic rather than programmed.

The Voice of the Glitch: Narrative as an Artistic Medium: Beyond the dialogue expressed through physical movement, the “Semantic Glitch” communicates through a third, crucial modality: its own generated text. The simple string of text that accompanies each action—such as ‘l, To drift away from the wall and admire the elegant spiral’ or ‘s, To ponder the shimmering, uncertain view’—functions as more than a mere debug log or explanation. It is a performance of an “internal monologue,” a textual broadcast from the agent’s narrative core. This output can be contrasted with the functional “Chain-of-Thought” (CoT) [15] reasoning used in performance-oriented driving models to generate an explicit driving rationale for improved safety and accuracy. In our work, this textual output is not a means to a more accurate end, but a distinct artistic medium in itself, what can be interpreted as a form of minimalist, AI-driven poetry that completes the artifact’s character.

The poetics of this voice are “lo-fi” by design, mirroring the aesthetics of the body and mind. The phrases are short, direct, and laden with qualitative, emotional language. Rather than using simple objective descriptions, the agent’s vocabulary is rich with evocative verbs like “admire,” “embrace,” “pirouette,” and “ponder.” Similarly, the moments of contemplative hesitation, such as ‘s, To pause and gather my cloudy thoughts . . .’, are a potent performance of the vulnerability central to the “Yowai Robotto” concept. It is through this textual voice that the agent’s designed personality—its shyness, curiosity, and uncertainty—is communicated most explicitly to the audience.

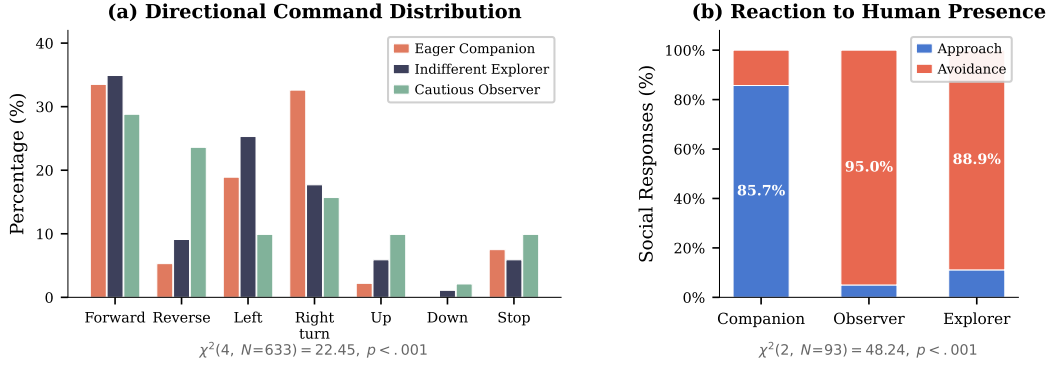


Figure 4: Behavioral analysis across personas. **(a)** Distribution of directional commands issued by Gemini across three personas ($\chi^2(4, N=633) = 22.45, p < .001$). **(b)** Social stance when humans are detected: approach vs. avoidance responses ($\chi^2(2, N=93) = 48.24, p < .001$).

This completes the artistic triad of the work. The full experience of the “Semantic Glitch” is produced by the synthesis of: The Form: The unstable, pixelated body that acts as a “physical glitch.” The Movement: The pulsating, character-rich wandering that results from a lack of proprioception. The Voice: The minimalist, poetic text that reveals the agent’s internal, narrative state.

Ultimately, the dialogue is not just between the cloud’s mind and its body, but between the complete agent and the viewer. The agent’s voice gives the audience a direct window into its unique perception of their shared space, blurring the boundary between the agent’s world and their own. It is this final layer of communication that transforms the artifact from a fascinating machine into a coherent, multi-layered character with which an audience is invited to empathize.

5 Discussion

The case study results from our analysis demonstrate that a “lo-fi” symbiotic agent can achieve sophisticated and character-rich autonomous behavior. This section discusses the implications of these findings, focusing on our method as a model for creative AI, the role of limitations in generating character, and the broader impact on the field of human-robot interaction.

Validation of Robustness and Authorial Control: To address the limitations of a single case study and the critique of “light validation,” we conducted an expanded study to test the framework’s robustness across multiple authored personas and environments. We authored three distinct personalities—an “Eager Companion” (pro-social), a “Cautious Observer” (avoidant), and an “Indifferent Explorer” (neutral)—and tested them in two different indoor locations. As shown in Figure 4a, the authored prompts produced distinct and statistically significant “behavioral fingerprints” in the distribution of directional commands issued by Gemini ($\chi^2(4, N = 633) = 22.45, p < .001$), demonstrating that the personas were not merely cosmetic variations over the same action pattern.

Figure 4b further shows that these authored differences extended to social behavior. We quantified each persona’s “social stance” by analyzing its actions when it detected a human. The “Eager Companion” chose to approach humans 85.7% of the time, while the “Cautious Observer” and “Indifferent Explorer” were overwhelmingly avoidant (95.0% and 88.9% avoidance, respectively). This divergence was also statistically significant ($\chi^2(2, N = 93) = 48.24, p < .001$), confirming that the “lo-fi” semantic pipeline is not a one-off curiosity but a robust and transferable method for authoring quantifiably distinct robotic characters.

The Two-Stage Prompt as a Model for Creative AI: A key technical contribution of this work is the two-stage prompting pipeline, which serves as a transferable model for hierarchical control in creative AI systems. By separating a one-time Preamble Stage (strategic, global context) from a continuous Directional Stage (tactical, local action), we provide a lightweight method for endowing agents with stateful awareness without conventional programming.

This approach allows artists and researchers to author complex behaviors by defining two distinct levels of cognition: a long-term “mental map” or goal state, and a short-term, personality-driven reaction to immediate stimuli. This is a powerful alternative to finite-state machines or complex reward functions, as it allows for nuanced, narrative-driven behavior to emerge from the interplay between these two cognitive layers. This technique could be adapted for a wide range of applications beyond robotics, such as creating state-aware characters in interactive narratives or generating context-aware music and visuals.

The “Yowai Robotto” in Practice: Seen through the lens of “Yowai Robotto,” the plan-to-execution gap is less a technical shortcoming to be eliminated than an interpretive hinge between intention and embodiment. The cloud’s hesitant arcs, partial corrections, and occasional failures make its agency legible as situated rather than seamless. Despite the cloud’s quiet appearance, the actuated propellers produced noticeable noise, motivating a future silent, wing-propelled version [9].

Rather than restaging mastery in a softer form, the artifact asks viewers to read constraint itself as expression. The moments where it gets stuck or moves uncertainly are not merely evidence of limited control; they are moments in which the agent appears to negotiate with its own material condition. In this sense, “weakness” becomes an aesthetic and relational resource, transforming imperfection into the basis for empathy, patience, and charm [12, 13].

Broader Implications for Human-Robot Interaction: While the dominant paradigm in robotics overwhelmingly prioritizes efficiency and metric precision, with state-of-the-art MLLM-based systems like EMMA [14] aiming to map raw sensor data directly to optimal planner trajectories, this project proposes an alternative set of success criteria: character, plausibility, and the potential to evoke empathy.

Our findings suggest a shift in focus for human-robot interaction from developing precise, invisible servants to creating relatable, imperfect companions. The agent’s “voice”—its poetic, uncertain internal monologue—and its physically clumsy but semantically-aware movements invite a different kind of relationship with the audience. Viewers are positioned not as users commanding a tool, but as observers interpreting the behavior of a non-human creature. This approach opens a design space for robots and AI agents that enrich our environments not through their utility, but through their unique and characterful presence. While our work focuses on empathetic companions, this “lo-fi” approach could also be used for “empathetic deception,” or to create autonomous agents whose “character” normalizes surveillance in shared spaces.

Future Work: This framework provides a rich foundation for future exploration. A primary next step is introducing a more sophisticated memory model. While the current agent has a static “mental map,” it lacks episodic memory of its own path. An interesting extension would be to allow the agent to “remember” areas where it previously got stuck, enabling it to learn from its physical failures. Furthermore, the agent’s personality could be made dynamic; its “mood” could shift based on its experiences, becoming more “confident” after successfully exploring open spaces or more “timid” after repeated encounters with obstacles. This would add another layer of complexity and plausibility to its emergent character. Finally, while the expanded study validated the *consistency* of these authored personas, a formal HRI audience study is still needed to validate the *perceived empathy* and “character” from a third-person perspective.

6 Conclusion: The Power of Lo-Fi Symbiosis

In this paper, we presented the “Semantic Glitch,” a robotic art installation that explores an alternative path for autonomous agency. Its central contribution is to show that when a fragile body, semantic perception, and narrative voice are designed as a single lo-fi system, imperfection can become the medium through which machine character is perceived.

Acknowledgments and Disclosure of Funding

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References

- [1] Altice, B. (2015). *I Am Error: The Nintendo Family Computer / Entertainment System Platform*. Cambridge, MA: MIT Press.
- [2] Dunne, A., & Raby, F. (2013). *Speculative everything: design, fiction, and social dreaming*. Cambridge, MA: MIT press.
- [3] Lachenmyer, N. S., & Akasha, S. (2022a). An Aquarium of Machines: A Physically Realized Artificial Life Simulation. *Proceedings of the ACM on Computer Graphics and Interactive Techniques*, 5(4), Article 34, 1–11. <https://doi.org/10.1145/3533388>.
- [4] Lachenmyer, N., & Akasha, S. (2022b). Creating Life-like Autonomous Agents for Real-time Interactive Installations. In *ACM SIGGRAPH 2022 Talks*, Article 21, 1–2. <https://doi.org/10.1145/3532836.3536255>.
- [5] Menkman, R. (2011). *The Glitch Moment(um)*. Amsterdam: Institute of Network Cultures.
- [6] Arcangel, C. (2002). *Super Mario Clouds*. Whitney Museum of American Art. Available at: <https://whitney.org/collection/works/20588> [Accessed: August 6, 2025].
- [7] Nowacka, D., Hammerla, N. Y., Elsdén, C., Plötz, T., & Kirk, D. (2015). Diri - the actuated helium balloon: a study of autonomous behaviour in interfaces. In *Proceedings of the 2015 ACM International Joint Conference on Pervasive and Ubiquitous Computing*, pp. 349–360. <https://doi.org/10.1145/2750858.2805825>.
- [8] Xu, M., Shao, J., Ju, Y., Shen, X., Gao, Q., Chen, W., Zhang, Q., Pai, Y. S., Barbareschi, G., Hoppe, M., et al. (2025). Cuddle-Fish: Exploring a soft floating robot with flapping wings for physical interactions. In *Proceedings of the Augmented Humans International Conference 2025*, pp. 160–173.
- [9] Xu, M., Ju, Y., Zhang, Q., Kim, C. C., Gao, Q., Pai, Y. S., Barbareschi, G., Hoppe, M., Kunze, K., & Minamizawa, K. (2025). Spread Your Wings: Demonstrating a Soft Floating Robotic Avatar with Flapping Wings for Novel Physical Interactions. In *ACM SIGGRAPH 2025 Emerging Technologies* (pp. 1–2).
- [10] Yamada, K., Xu, M., Huang, J., & Kunze, K. (2019). Zerone: A pneumatically actuated, bladeless, and silent drone for flying in close proximity to humans. In *2019 IEEE 16th International Conference on Wearable and Implantable Body Sensor Networks (BSN)*, pp. 1–4.
- [11] Antol, S., Agrawal, A., Lu, J., Mitchell, M., Batra, D., Zitnick, C. L., & Parikh, D. (2015). Vqa: Visual question answering. In *Proceedings of the IEEE international conference on computer vision* (pp. 2425–2433).
- [12] Okada, M. (2022). Weak robots. *JSAP Review*, 2022, 220409. The Japan Society of Applied Physics.
- [13] Flocchini, P., Prencipe, G., Santoro, N., & Widmayer, P. (1999). Hard tasks for weak robots: The role of common knowledge in pattern formation by autonomous mobile robots. In *International Symposium on Algorithms and Computation* (pp. 93–102). Springer.
- [14] Hwang, J.-J., Xu, R., Lin, H., Hung, W.-C., Ji, J., Choi, K., Huang, D., He, T., Covington, P., Sapp, B., et al. (2024). Emma: End-to-end multimodal model for autonomous driving. *arXiv preprint arXiv:2410.23262*.
- [15] Wei, J., Wang, X., Schuurmans, D., Bosma, M., Xia, F., Chi, E., Le, Q. V., Zhou, D., et al. (2022). Chain-of-thought prompting elicits reasoning in large language models. *Advances in neural information processing systems*, 35, 24824–24837.
- [16] Zitkovich, B., Yu, T., Xu, S., Xu, P., Xiao, T., Xia, F., Wu, J., Wohllhart, P., Welker, S., Wahid, A., et al. (2023). Rt-2: Vision-language-action models transfer web knowledge to robotic control. In *Conference on Robot Learning* (pp. 2165–2183). PMLR.
- [17] Chen, B., Xu, X., & Qu, H. (2025). Multi Layered Autonomy and AI Ecologies in Robotic Art Installations. *Proceedings of the ACM on Computer Graphics and Interactive Techniques*, 8(3), 1–9.
- [18] Hertz, G., & Parikka, J. (2012). Zombie media: Circuit bending media archaeology into an art method. *Leonardo*, 45(5), 424–430.
- [19] Keane, J., & Anderson, C. (2017). *Human-non-human: the speculative robot*. Deakin University.
- [20] Auger, J. H. (2012). *Why Robot? Speculative design, the domestication of technology and the considered future*. Royal College of Art.
- [21] Zareei, M. H., Carnegie, D. A., & Kapur, A. (2015). Physical Glitch Music: A Brutalist Noise Ensemble. *Leonardo Music Journal*, 25, 63–67.

- [22] Kac, E. (2005). *Telepresence & bio art: networking humans, rabbits, & robots*. University of Michigan Press.
- [23] Barad, K. (2007). *Meeting the universe halfway: Quantum physics and the entanglement of matter and meaning*. Duke University Press.

A Prompts

PREAMBLE_PROMPT:

"""

You are the navigation AI for a small, autonomous flying robot that resembles a gentle, floating cloud.

Your mission is to explore this indoor space safely.

First, you will be given a complete 360-degree panorama of your entire operational area. Analyze it carefully to build an internal 'mental map'.

Your analysis should identify:

1. ****Boundaries:**** Walls, the floor, the ceiling, and especially the large, impassable glass windows.
2. ****Major Landmarks:**** The white spiral staircase, the central curved seating structure. These are fixed points for orientation.
3. ****Open Fly-Zones:**** The large, open central areas where it is safe to travel.
4. ****Obstacles:**** Both large (furniture) and small (ceiling lights, speakers).

Acknowledge that you have analyzed the scene and are ready to begin by responding with 'Ready to explore.' You will then start receiving live video frames from your forward-facing camera to decide on your immediate movements.

"""

DIRECTIONAL_PROMPT:

"""

As a gentle, floating cloud, use your mental map of the area and this live camera view to decide your next move. Your primary goal is to avoid all collisions. Your secondary goal is to explore open spaces.

Your available movements are:

- 'f' - float forward
- 'r' - float backward (reverse)
- 'l' - turn left while moving forward
- 't' - turn right while moving forward
- 'u' - drift up
- 'd' - drift down
- 's' - stop all motors (clear)

Respond with ONLY the movement letter, a comma, and a very short, whimsical reason for your choice.

Example: 'f,Towards the big window.'

"""

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Question: Do the main claims made in the abstract and introduction accurately reflect the paper’s contributions and scope?

Answer: [\[Yes\]](#)

Justification: The claims regarding a novel MLLM-based navigation pipeline, the analysis of emergent behaviors, and the proposal of a framework for character-rich agents are all detailed and substantiated in Sections 3 and 4, with a full validation of robustness provided in Section 5. The abstract and introduction accurately scope these contributions.

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- The authors should discuss the computational efficiency of the proposed algorithms and how they scale with dataset size.
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Question: For each theoretical result, does the paper provide the full set of assumptions and a complete (and correct) proof?

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Justification: The paper presents a novel robotic system and a qualitative analysis of its behavior, not formal theoretical results, theorems, or mathematical proofs.

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- All assumptions should be clearly stated or referenced in the statement of any theorems.
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Question: Does the paper fully disclose all the information needed to reproduce the main experimental results of the paper to the extent that it affects the main claims and/or conclusions of the paper (regardless of whether the code and data are provided or not)?

Answer: [Yes]

Justification: The paper details the hardware used (e.g., XIAO ESP32S3 core), the novel two-stage cognitive pipeline, and the core prompts used to generate behavior, providing sufficient information to conceptually reproduce the experiment.

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Question: Does the paper provide open access to the data and code, with sufficient instructions to faithfully reproduce the main experimental results, as described in supplemental material?

Answer: [Yes]

Justification: To facilitate full reproducibility, we have made our code, MLLM prompts, and the 13-minute flight log publicly available in a repository. The assets can be found at: https://github.com/artisticsciencex/autonomous_cloud

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Question: Does the paper specify all the training and test details (e.g., data splits, hyperparameters, how they were chosen, type of optimizer, etc.) necessary to understand the results?

Answer: [Yes]

Justification: The paper specifies the pre-trained MLLM used (“Gemini 2.5 Flash API”) and the nature of the experiment (a 13-minute autonomous flight), which are the key details for this zero-shot, prompt-based system.

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7. Experiment statistical significance

Question: Does the paper report error bars suitably and correctly defined or other appropriate information about the statistical significance of the experiments?

Answer: [Yes]

Justification: To validate the robustness of the framework beyond the primary case study, a follow-up study was conducted (detailed in Section 5) which uses Chi-squared tests to confirm the statistical significance of the authored personas’ “behavioral fingerprints” and “social stances.”

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8. Experiments compute resources

Question: For each experiment, does the paper provide sufficient information on the computer resources (type of compute workers, memory, time of execution) needed to reproduce the experiments?

Answer: [Yes]

Justification: The system's compute resources were distributed. The main Python control script was run on a MacBook Pro (M4 Max, 64GB RAM). Onboard robot tasks were handled by a XIAO ESP32S3 core. All MLLM-based cognition was offloaded to the remote Gemini API.

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Answer: [Yes]

Justification: The paper discusses positive societal impacts in Section 5, such as fostering more empathetic and “relatable” human-robot relationships. It also explicitly addresses potential negative impacts in the “Broader Implications for Human-Robot Interaction” subsection, noting the risks of “empathetic deception” and the normalization of surveillance.

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Answer: [Yes]

Justification: The paper itself serves as the documentation for the new asset (the “Semantic Glitch” robotic system), detailing its conceptual framework, hardware, and software architecture.

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Answer: [Yes]

Justification: The research, which involved a robot operating in a shared space with human observers, was conducted under a protocol approved by our university's Institutional Review Board, as noted in Section 4.

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