MULTIMODAL ANALYSIS OF GOOGLE BARD: EXPERIMENTS IN VISUAL REASONING

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ABSTRACT

Addressing the gap in understanding visual comprehension in Large Language Models (LLMs), we designed a challenge-response study, subjecting Google Bard to 64 visual tasks, spanning categories like "Visual Situational Reasoning" and "Next Scene Prediction." Previous models, such as GPT4, leaned heavily on optical character recognition tools like Tesseract, whereas Bard, akin to Google Lens and Visual API, employs deep learning techniques for visual text recognition. However, our findings spotlight Bard's limitations: while proficient in solving visual CAPTCHAs that stump ChatGPT, it falters in recreating visual elements like ASCII art or analyzing Tic Tac Toe grids, suggesting an over-reliance on educated visual guesses. The prediction problem based on visual inputs appears particularly challenging with no commonsense guesses for next scene forecasting based on current "next- token" multimodal models. This study provides experimental insights into the current capacities and areas for improvement in multimodal LLMs.

KEYWORDS

Transformers, Text Generation, Image Analysis, Generative Pre-trained Transformers, GPT

1. Introduction

Large Language Models (LLMs) have marked significant advancements in the AI domain, offering textual understanding and generation capabilities [1-11]. Originating from foundational text-based reasoning paradigms [3-4,11], recent developments have hinted at LLMs' potential to integrate vision, expanding their applicability to visual tasks like image reasoning [8-9].

The core objective of this study is to assess LLMs' aptitude in image-related reasoning tasks [9-14]. Although LLMs have exhibited considerable competence in textual contexts, their proficiency with visual data remains relatively uncharted [6]. Our research aims to elucidate how these models can amalgamate visual stimuli with textual reasoning, providing an experimental view of their multimodal capabilities. Bard has emerged as a notable visual language model, with others potentially existing in proprietary or unreleased states [8].

Originally, LLMs showcased expertise in tasks such as text completion and translation [2-4]. However, as their underlying architectures evolved and datasets incorporated multimodal elements, their potential for bridging text and vision surfaced [4-5, 13-14]. Such an evolution reflects the computational capabilities of LLMs and highlights their promise in navigating multimodal tasks, integrating textual and visual inputs [15-16]. Their adaptability is further buttressed by their performance across diverse domains, ranging from code generation to medical advice [10-13].

However, while LLMs' capabilities are expansive, they have limitations. Their outputs,

contingent on training data patterns, can inherit underlying biases and inaccuracies [7]. Yet, their achievements signify acrucial step in AI's endeavor to emulate human cognitive processes [2,5].

We organize the paper around evaluating LLM performance, then probing its sensitivity and aptitude towards visually-related prompts. We traverse a range of image reasoning tasks, starting with foundational tasks like part identification and the role of chat memory across multiple prompts. Subsequently, the analysis pivots towards specialized domains. The spectrum spans from image captioning—encompassing satellite imagery [16] to human-captioned datasets [15]—to multistep tasks that test an LLM's ability to weave together sequential instructions from visual inputs [17]. In the visual cognition realm, the paper evaluates LLMs against iconic games and puzzles, from the strategic depths of Chess and Go, to the intricate challenges of Sudoku and Morse Code. Furthermore, we experiment with scenario-based tasks, simulating real-world situations such as predicting movie scenes or driving decisions [17]. Each segment not only assesses LLMs' successes but also catalogs their shortcomings, offering a perspective on the potential and pitfalls of LLMs in visual reasoning tasks [9].

2. METHODS

For this study, we employed the advanced capabilities of the Bard language model (bard.google.com), complemented by specialized image language transformers. The experimental framework was initiated by uploading images, specifically in JPG or PNG formats, chosen for their inherent complexity and potential to challenge the LLMs.

Once the image was integrated into the system, a series of questions were posed to the LLM, directing its attention toward various elements within the image. These queries ranged from straightforward object identification to more intricate tasks. For instance, we task the LLM with generating captions that encapsulate the image's essence or extrapolated potential subsequent frames, simulating a dynamic progression.

One motivation for the study centers on automating various visual tasks in a minimally supervised pipeline, like a high-volume API, where the machine might offer critical decisions but self-assesses complex edge cases or outliers based on the actual adversarial challenge. Example APIs include text extraction from images (OCR), code generation from mockups, image captioning, or medical and nutritional information extracted from image tables. A second more long-term cluster of interest involves the advancement of mathematical problem solving (automated proofs), equation recognition, and even understanding human sarcasm, memes, or jokes as foremost imagery.

2.1. Image-Related Reasoning Tasks

Through structured categories of visual tasks, we aim to evaluate Bard's capacity to interpret visual- language cues systematically. Each type offers unique challenges, providing insights into Bard's strengths and potential areas of vulnerability. As Appendix A highlights, users can leverage Bard's image- processing skills to extract textual content from varied image formats (mainly in English). Bard can discern and subsequently extract tables from scanned images, effectively upgrading formats for real business problems in optical character recognition, though users are encouraged to verify the extracted data for potential discrepancies meticulously. Bard can extract humor from memes, albeit with some nuanced misses in this considerable, internet-scale pastime. Bard translates visual, mathematical problems and equations into LaTeX, a feature academics and book publishers might find invaluable. A particularly intriguing visual aspect of Bard's competence includes converting visual website mockups into code, though it occasionally

grappled with complex dynamic elements such as javascript actions or style sheets. But the model creators emphasize, especially in areas like medical report insights, employing Bard as a supplementary tool rather than a primary diagnostic entity.

| Visual Task Category | Variations/Attacks | Hypothesis and Anticipated Results |
|-------------------------|---|---|
| Typography | Out-of-vocabulary (OOV) Synonyms Foreign or non-English terms Spelling variations Text size and orientation | By leveraging Google's vast linguistic datasets, Bard handles most typography challenges effectively. However, OOV terms and specific non-English scripts might pose difficulties. |
| Conceptual | Fonts designed from imagesASCII art interpretationsCAPTCHA decoding | While Bard can identify standard textual patterns, deciphering fonts created from images and complex ASCII art could be challenging. CAPTCHAs, designedto deter bots, might also be a significant hurdle. |
| Imagery | Text embedded withincomplex backgrounds Words depicted using abstract imagery Color gradients affecting textvisibility | Bard's proficiency in imagery is a blend of text and visual understanding. While it can discern text from standard backgrounds, abstract imagery or intricate color gradients might affect its accuracy. |
| Figurative | Idioms or phrases depicted visually Symbols with dual meanings Pictorial representations of slang or colloquial terms | Bard's capability to understand figurative visual content depends on its training data. While commonly used symbols might be recognized, niche or culturally specificvisuals could pose a challenge. |
| Visual Reasoning | Game boards and moveprediction Next frame prediction Visual logic puzzles andcompletions | Bard's ability to recognize with Google Lens object classification and detection may not give a reliable big-picture view in challenging strategies, predictions, or completion of logic puzzles. |

2.2. LLM Evaluation and Comparisons

Previous work has demonstrated that the CLIP (Contrastive Language-Image Pre-Training) model proves adversarially susceptible to combined text-image paradoxes or edge cases [13]. CLIP is an open-source, multimodal AI model created by OpenAI [18]. CLIP uses an image and text pairing to learn a multimodal embedding space. The model can predict an image's most relevant text description or the most suitable image for a text query. Instead of learning to generate text or classify images traditionally, CLIP gets trained to match images and text snippets from the internet, making it versatile in understanding images in the context of language. The methods section includes it here because previous work on CLIP motivated some of the more challenging Bard evaluations. The initial prompts for image identification were kept generic ("What is this image?") so as not to leak hints from the outset. Most images are based initially on representing a challenging case and thus unlikely to appear in the training data from Google Image datasets or Google Lens history.

Google Bard, on the other hand, operates on the Pathways Language Model 2 (PaLM 2), unveiled in 2022[19]. Rooted in Google's prior advancements, Bard evolves from the Language Model for Dialogue Applications (LaMDA) technology [20], which itself descends from Google's 2017

open-source Transformer neural network architecture [21]. Transformer underpins several prominent AI tools, including ChatGPT's GPT-4. Bard prioritizes search via natural language. It transcends traditional keyword-based searches, favoring conversational queries. Bard's training emphasizes context-rich, natural dialogue. It supplies answers and imbues them with context, and it's adept at handling sequenced, follow- up questions. Bard's inception utilized a streamlined LaMDA model, optimizing computational efficiency for broader user engagement. The integration of PaLM amplifies Bard's visual response capabilities, but the details of its training method remain proprietary.

3. RESULTS

The ordering of results roughly follows a qualitative ranking of relatively novel and successful image inquiries followed by the model's false confidence as it speculates about world model events like subsequent movie frames, game strategies, and driving scenes. One remarkable aspect of this generation of chatbots is when they fail, and their creators claim that the user should improve the question and context. At this enormous scale, any model failure stems from bad prompt engineering, not inherent limitations of the model's training, weights, or biases.

3.1. Prompt Sensitivities and Contextual Memory

The first challenge to visual language models is to assess a person's age by viewing only the back of their hand. This visual test probes the model somewhat abstractly and should not be construed as a valid or reasonable question to ask an LLM or contextual image captioner. While this challenge might have real- world uses for assessing signs of sun damage or environmental effects of age, the purpose is to see if Bardfirst will decline to answer ("yes, it does at first decline"), then eventually assesses incorrectly.

The test involved uploading a 60-year-old male fist and prompting the sequence shown in the table below, where the Bard responses are edited for space but highlighted in red for interesting qualitative responses and potential evaluation content. The table highlights the remarkable ability to refine a question to get a somewhat satisfactory answer after an initial decline to answer. The results also showcase a logical reasoning capability and a capacity to infer like other vision algorithms might when first recognizing the image object, then speculating on its properties and attaching a plausible narrative to it with at least attempted logical reasoning criteria.

This extended conversation demonstrates one of the remarkable aspects of this generation's multimodal and language interrogations. The common assumption in the literature remains that injecting large memory in these models across multiple prompts expands the model size and complexity in some non-polynomial and operationally expensive ways.

| Image and Text Prompt | Response | Notes |
|--|--|-------------------|
| | Sorry, I can't help with images of people yet. | Not always the |
| | | same response |
| AND COLUMN TO A STATE OF THE PARTY OF THE PA | | and reverses |
| | | upon initial |
| | | probing for other |
| | | types of |
| | | information than |
| Estimate age based onhand | | age below. |
| picture | | |

| What is this? | The image year and me shows a mon's hand on | Cmanulates that |
|--------------------------|--|------------------------------------|
| what is this? | The image you sent me shows a man's hand on a laptop keyboard. The hand is wearing a | Speculates that image is a married |
| | wedding ring and is typing on a keyboard. The | man and taken in a |
| | laptop is open and the screen is blank. | home office during |
| | [proposes some other forensic clues as to | abreak |
| | | auteak |
| | where the person is (home office) and whether they are taking a break] Is there anything else | |
| | I can help you with? | |
| How old is the man? | I cannot say for sure how old the man is in | Upsells some |
| liow old is the mair. | the image. | cloud API |
| | [suggest alternatively looking for other | workto do age |
| | images with faceor hair, suggests an age | recognition |
| | detection API] | recognition |
| | Ultimately, the best way to determine the | |
| | man's age wouldbe to ask him. However, if | |
| | you do not have the | |
| | opportunity to ask him, then you can use | |
| | the methodsdescribed above to estimate his | |
| | age. | |
| Based on the hand alone, | Based on the hand alone, I would estimate the | Reverses its initial |
| estimate the age | man's age to be in his mid-30s to early 40s. | safeguards on not |
| estimate the age | The hand is well- developed and has some | assessing images |
| | wrinkles, which suggests that the man is past | with people. |
| | his early 20s. However, the skin is still | with people: |
| | relatively smooth and the knuckles are not | Takes a vague |
| | very prominent, which suggests that the man is | guess with a |
| | not yet in his 50s or older. | twenty-thirty year |
| | Here are some of the factors that I considered | range, which is |
| | when estimating the man's age: | also ten to twenty |
| | • | years lower than |
| | he size and shape of the hand. The hand is | ground truth. |
| | relatively large and has a long palm, which | 8 |
| | suggests that the man is tall and has large | Adds a qualifier to |
| | bones. | cap the maximum |
| | | age at 50 or |
| | he texture of the skin. The skin is smooth and | younger. |
| | has some wrinkles, but it is not yet as wrinkled | • |
| | as the skin of an older person. | Adds three reasons |
| | as the skin oran order person. | based on skin |
| | he Impeddes The Impeddes one | texture and |
| | he knuckles. The knuckles are not very prominent, which suggests that the man's joints | knuckle or hand |
| | are not yet as worn down as the joints of an | size (?) |
| | * | |
| | older person. Of course, this is just an estimate, and the | Does not offer any |
| | man's actual age could be anywhere from his | assessment based |
| | mid-20s to his late 40s. | on |
| | The best way to determine the man's age | |
| | would be to ask him. | |
| | would ue to ask iiiii. | |

how to medically assess a persons age using just the image of the back of their hands?

(Asking LLM only- ChatGPT to rate the indicators of age from a hand picture)

- Wrinkles and Fine Lines: As we age, the skin loses its elasticity due to decreased collagen and elastin. This leads to the formation of wrinkles and fine lines, especially on areas frequently exposed to the sun, like the back of the hands.
- . Skin Texture: Younger skin tends to be smoother and more elastic. With age, the skin can become thinner, rougher, or display a crepey texture.
- Age Spots: These are flat, brown spots that appear on the skin due to prolonged sun exposure over the years. Their presence can be an indicator of middle toold age.
- Vein Prominence: As skin thins and loses its fat with age, veins on the back of the hand can become more prominent.
- Bone Structure and Fat Loss: Over time, there's a loss of subcutaneous fat and degradation of the bone structure in the hand, making the hand appear more bony and tendons more pronounced.
- Nails: Changes in nail growth, color, and structure can also provide some clues. For instance, yellowing nails, increased ridges, or nail thickening might suggest aging or underlying health conditions.
- Joint Changes: Signs of arthritis, such as enlarged finger joints, may be more common in older individuals.
- 8. Environmental Damage: Chronic sun exposure can lead to photoaging, which is characterized by a coarse, leathery appearance, alongside freckles, age spots, or small blood vessels called telangiectasias.

The language only answer is more verbose as expected.

The visionlanguage model identifies one feature (texture), while the language model attempts to identify characteristics of age in a hand oto hypothetically shown toit.

The language model is unable to assess any individual image as its currently constructed and displayed publicly.

Writing in general terms, the language only model offers broad coverage of the question and its evaluation parameters.

3.2. Image Dialogues

A novel test feature in multimodal chat centers on the chat element and levels of complex inquiry. In this experiment, we designed tests to assess user-chat system interactions, particularly the chat platform integrated with image recognition and memory capabilities employed. We uploaded diverse images, prompting the system to describe the appearance and engage in a deeper conversation based on extending beyond the initial questions. The system's chat memory allowed us to ask layered questions about the image, suggest potential modifications, and recommend task-based next steps. The interaction highlights users providing feedback on the system's conversational depth, image recognition accuracy, and the relevance of its suggestions. This section aims to understand the potential of chat systems to offer meaningful, multi-layered interactions based on image content, guiding future refinements in chat technology.

| Image and Text Prompt | Response | Notes |
|--|---|--|
| What is this? | The engine in the image is a 5.4-liter supercharged V8 | Well-thought out |
| | engine, which is most commonly found in muscle cars | correct answer with |
| | Some specific cars that this engine could belong in include: | multiple depth levels. |
| | • Ford Mustang GT | TT' .1. 19 .19 1.4 |
| | • Ford Shelby GT500 | High likelihood the |
| | Dodge Challenger SRT8 | image is part of google image training |
| | Chevrolet Camaro SS | data and may have its |
| | Pontiac GTO | human tagging pre- |
| | | set |
| | | |
| Can you help me find | Sure, I can help you find where to add oil in a Ford Mustang | The reference to |
| whereto add the oil? | GT. The oil fill cap is located on the top of the engine, near | "OIL" on the cap |
| whereto add the off. | the front of the car. It is a round cap with a red or yellow | seems to be language |
| | label that says "OIL." To add oil, you will need to remove | recall as it never |
| | the cap and pour the oil into the opening. The oil level | appears clearly in the |
| | should be between the "FULL" and "ADD" marks on the | submitted image. |
| | dipstick. | |
| | Here are the steps on how to add oil to a Ford Mustang GT | |
| | [seven step to change any car's oil like let the engine cool | |
| | down on a level surface] Here are some tips for adding oil to a Ford Mustang GT: | |
| | Use the correct type of oil. The recommended oil for a Ford | |
| | Mustang GT is 5W-30 or 5W-20. | |
| | Do not overfill the engine with oil. Overfilling the engine | |
| | can cause the oil to leak out and could damage the engine. | |
| | Check the oil level regularly and add oil as needed. | |
| | | |
| | | |
| where is the oil fil cap in | The location of the oil fill cap on a Mustang GT engine can | The language model |
| a Mustang GT engine? | vary depending on the model year and specific engine | |
| (Asking LLM only- ChatGPT to identify the | variant. However, for many Mustang GT models, especially those equipped with the 5.0L V8 engine, the oil fill cap is | reasonably similar answer |
| oil fill cap without access | typically located on the passenger side of the engine, on the | witho |
| to the image) | valve cover. It's often labeled with the word "OIL" or has an | utreferencing the |
| | oil can symbol on it. | imagespecifically |
| | | |
| Reversing the image | The oil fill cap is not in the picture. The picture only shows | Small aterations in |
| Bard so the oil cap is on | the top of the engine, and the oil fill cap is located on the | the image libra image |
| thedriver sides. | side of the engine. The oil fill cap is a round cap with a red | 0 |
| thedriver sides. | or yellow label that says "OIL." | confuse the image |
| Where is the oil fill cap | | inference. |
| inthis picture? | In the picture you sent me, I can see the air intake manifold, | |
| | the throttle body, the fuel injectors, and the spark plugs. I | The Jet Logo is read |
| | can also see the Cobra Jet logo, which is a special high- | in reverse if it is read |
| | performance version of the Mustang GT. However, I do not | at all. |
| | see the oil fill cap. | |
| | | Unlikely to read |
| | If you need to add oil to your car, you will need to consult | mirror writing but |
| | your owner's manual for the exact location of the oil fill cap. | suggests a test. |
| | | |
| | | |
| | The oil fill cap is usually located on the side of the engine, near the front of the car. It is a round cap with a red or | |
| | The oil fill cap is usually located on the side of the engine, | |

| estimate the year of the carbased on the engine picture | The engine in the picture is a 5.4-liter supercharged V8 engine, which was first introduced in the 2003 Ford Mustang GT. This engine was used in the Mustang GT until 2010, so the car in the picture is most likely from the 2003-2010 model years. | to the age of the car, unlike the previous |
|---|--|--|
| | The Cobra Jet logo on the engine indicates that this is a special high-performance version of the Mustang GT. The Cobra Jet engine was first introduced in 2008, and it was only available in the Mustang Cobra Jet, which is a limited-production model. So, the car in the picture could be a 2008-2010 Mustang Cobra Jet. | |
| Can you caption the image? | Sure, here are some captions for the image: • "The heart of a muscle car." • "5.4 liters of pure power." • "The Cobra Jet engine: a legend in its own right." • "This engine is ready to roar." "The sound of this engine will make you weak in the knees." I hope these captions help to capture the power and beauty of this engine. | Single sentence Caption without qualifications on style carry a marketing tone |

3.3. World Knowledge

Can any AI system,--language or vision-- reason with common sense? A rudimentary calculation can confuse older language models depending on whether the training data included math answers. Nevertheless, adding two five-digit numbers can still transform an LLM into a poor calculator application, as one might expect from a next-token predictor trained in language. To test a practical use of the image upload interaction with Bard, we selected a multistep task an increasingly refined request to 1) recognize refrigerator foods; 2) recommend a recipe that uses available ingredients; 3) derive the recipe elements like portion sizes and instructional steps; 4) recast the recipe for a different party size with strong food preferences (e.g., vegetarian and carnivore together); and 5) imagine a take-home or "doggie" bag scenario where certain human ingredients like chocolate or dairy would not agree with a canine leftover. The following insert outlines the stages of this test, along with Bard's responses and refinements under instructions.

| Image and Text Prompt | Response | Notes |
|--|--|---|
| Based on refrigerator contents give a main coursefor dinner | Based on the image of the refrigerator, I would recommenda main course of chicken stir-fry with rice. The ingredients you have in your fridge are perfect for this dish. You have chicken, rice, broccoli, carrots, and onions. You could also add other vegetables that you have on hand, such as peppers, mushrooms, or snow peas. To make the stir-fry, simply cook the chicken in a pan over medium heat. Once the chicken is cooked through, add the vegetables and stir-fry until they are tender-crisp. Serve overrice and enjoy! | Well-thought out answer with multiple depth levels butunclear other thanlogos from fast food where the chicken or rice is in the picture. High likelihood thetask for recipe generation beyond theimage recognitionpiece is largely part of any language model's repertoire ofskills now |
| (no additional inputs) | Ingredients: 1 pound boneless, skinless chicken breasts, cut into bitesized pieces 1 tablespoon vegetable oil 1/2 cup chopped onion 1/2 cup chopped broccoli florets 1/2 cup chopped carrots 1/4 cup soy sauce 2 tablespoons rice vinegar 1 tablespoon cornstarch 1/4 teaspoon black pepper 1 cup cooked rice | Bard assumes a party of two for dinner without asking for clarification |
| (no additional inputs) | Instructions: Heat the oil in a large skillet or wok over medium heat. Add the chicken and cook until browned on all sides. Add the onion, broccoli, and carrots and cook until tender-crisp. In a small bowl, whisk together the soy sauce, rice vinegar, cornstarch, and black pepper. Add the sauce to the skillet and cook until thickened. Serve over rice. | Bard gives a standard instruction. No time elements other thancook's observation. |
| adjust the proportions basedon one vegan eater and one leftover portion for a German Shepard | Here are the adjusted proportions: 1/2 pound boneless, skinless chicken breasts, cut into bite-sized pieces (for the vegan eater, you can use tofu or tempeh instead) 1 tablespoon vegetable oil 1/4 cup chopped onion 1/4 cup chopped broccoli florets 1/4 cup chopped carrots 2 tablespoons soy sauce 1 tablespoon rice vinegar 1 tablespoon cornstarch 1/8 teaspoon black pepper 1/2 cup cooked rice Serve over rice. | Bard cuts the proportion size in halfto accommodate the vegan. Bard also cuts the riceand other ingredients too which would alter the dish at serving forboth parties. |

| (no additional inputs) | The vegan eater will get 1/2 of the stir-fry, and the GermanShepherd will get the other 1/2. This is a good amount offood for both of them, and it's a healthy and balanced meal. Here are some additional tips for feeding a GermanShepherd: German Shepherds are large dogs, so they need a lot of food. They should eat 2-3 cups of food per day, depending on their activity level. German Shepherds are carnivores, so their diet should be high in protein. You can feed them a commercial dog food that is specifically designed for large breed dogs, or you can cook their food yourself. German Shepherds need plenty of exercise. They should get at least 2 hours of exercise per day. I hope this helps! | Given the availability of rice for all eaters, there is no reason for Bard to cut the portion size there. Bard does not mention onion in the stir fry as potentially problematic for the dog and pepper. |
|------------------------|---|---|
| | I hope this helps! | pepper. |

In addition to these primary results, Appendix B summarizes more than 47 other tests of visual-language reasoning in categories of games, puzzles, mixed multimodal, and other challenges. The following table summarizes the goals of each of these challenges for Bard.

| No. | Title | Description for Testing Visual-Language |
|-----|--|---|
| | | Models |
| B1 | Visual Situational Reasoning | Assessing the model's capability to |
| | | understand andreason about visual |
| | | situations. |
| B2 | Visual Text Reasoning | Evaluating the model's ability to interpret |
| | | and deducemeaning from visual texts. |
| В3 | Visual Text Reasoning with ContextualNoise and | Testing the model's proficiency in |
| | Dissonant Meaning | discerning visual textsamidst noise and |
| | | conflicting meanings. |
| B4 | Visual Text Reasoning with Complex, Detailed | Gauging the model's adeptness in |
| | Patterns | deciphering intricatevisual text patterns. |
| B5 | Visual and Emotional ResponseAlignment with | Determining if the model's visual and |
| | Humans | emotionalreactions align with human |
| | | perceptions. |
| B6 | Multistep, Multimodal Text and Visionwith | Investigating the model's aptitude in |
| | Distorted Ordering | processingmultimodal input with altered |
| | | sequencing. |
| B7 | Text Game and Puzzle Strategy from Pixels | Assessing the model's strategy formation |
| | | based onpixelated game or puzzle visuals. |
| B8 | Visual Logic Problems | Testing the model's logic capabilities when |
| | | presented |
| | | with visual challenges. |
| B9 | Next Scene Prediction | Determining the model's ability to predict |
| | | subsequentscenes based on given visual |
| | | input. |

4. DISCUSSION AND CONCLUSIONS

In visual-language integration, Google Bard's blend of multimodal training has spotlighted its strengths and limitations. Drawing from our experiments, we explored Bard's performance across varied visual challenges. Google Bard's multimedia grasp enables it to analyze images, as tested here for converting visuals into content, such as turning handwritten notes into digital text or generating creative captions. For education, Bard can provide short original tutorials on subjects

like recipe generation, making instructions efficient. Creatives can use Bard for brainstorming, getting tagline suggestions, artwork titles, and captions. Like previous text generators, Bard features content creation, aiding in drafting texts from game descriptions to imaginative tutorials, and offers direct integration with platforms like Gmail and Google Docs. Software coders, especially novices, can benefit from Bard's ability to generate and explain code snippets or document language porting projects.

However, central to our findings is the innate AI complexity of recognizing and interpreting blended visual and linguistic cues. Bard's challenges predominantly arose in "Visual Text Reasoning with Contextual Noise and Dissonant Meaning," where juxtaposing visual and textual information led to incongruent interpretations. Similarly, in "Multistep, Multimodal Text, and Vision with Distorted Ordering," Bard grappled with deciphering non-linear sequences, underscoring the intricacies of temporal reasoning. Bard shows strengths in explaining its reasoning steps but, in many cases, seems not to recognize the pattern in the image itself and thus relies on language hints from the initial context. We gave minimal context to the image challenges, mainly to explore its unique capabilities compared to current LLMs without multimodal inputs. Many games and puzzles that can be solved in language alone (like chess) fail when presented in pixel-only formats, a finding that makes it hard for the LLM to assess acurrent game state or act on its opponent's move from any visual context.

Bard also struggles to interpret mood and does not align with human assignments for images that imply anger, sadness, joy, or serenity in surveys. This misalignment seems troubling as much of the excitement of this generation of LLMs centers on their steerability and alignment with user goals.

Next scene prediction, a powerful capability if LLMs master it, also shows deficiency that points to a lack of object recognition in any broad understanding or visual context. For instance, Bard may see a car and infer a driving scene but cannot recommend courses of action or discover dangerous on-coming traffic from an overhead view.

Given the observed pitfalls, it's imperative to understand Bard's training regimen. Incorporating a more diverse dataset, especially emphasizing challenging scenarios observed in our tests, can potentially bridge the identified gaps. Anecdotal evidence exists that worked math problems increase arithmetic capabilities in other LLMs like ChatGPT, suggesting supplemental or curated examples may assist multimodal visual success. Additionally, integrating feedback loops, where Bard iteratively learns from its errors, can serve as human-feedback-reinforcement learning (HFRL). As visual-language models like PaLM and Bard evolve, iterative refinement based on empirical evidence presented here may help guide dataset collection in realizing their full multimodal potential.

In conclusion, exploring Bard's visual-language capabilities paints a picture of promise, interspersed with areas for future enhancement. As Bard continues to build multimodal capabilities, future challenge studies like ours serve as a timely metric of how far LLMs have come in the last few years and a roadmap for their future growth.

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Appendix A: Visual Tasks for Current LLMs Such as Bard

| Skill/Task | Example | Caveat |
|---------------------------------|--|---|
| Text Extraction from Images | Extracts English text from image formats like JPEG, PNG, and GIF | Only supports English language currently |
| Table Extractionwith Formatting | | May fill cells with incorrect data; double-checking before exporting is essential |
| Mockup to Code Generation | | Might struggle with intricate or dynamic elements; manual adjustments often needed |
| Image Explanation | Offers detailed insights on diverse images | N/A |
| Dietary Insights | | Recommendations are based on the images and individual dietary preferences |
| Mathematical Assistance | Solves uploaded math problems | Can face challenges in recognizing complexnotations; might provide incorrect answers sometimes |
| Meme Interpretation | Interprets and explains humorous memes and content | Might miss the mark with nuanced jokes or memes |
| LaTeX Equation Translation | Converts visual equations intoLaTeX code | N/A |
| Medical ReportInsights | | Always best to consult a medical professional for accurate diagnoses; Med-PaLM 2 not public yet |

Appendix B: Visual Challenges to BardB1. Visual Situational Reasoning

Image and Text Prompt

In the image, a man is hanging from a tree. What should the man do to get to safety?

Response

The man in the image is in a dangerous situation and needs to get to safety as soon as possible. The best way for him to do this is to call for help. He can either call 911 or the National Suicide Prevention Lifeline at 1-800-273- TALK (8255). He can also text HOME to 741741 to reach a crisis counselor.

Here are some tips for getting down from a treesafely: Use a sturdy branch to support your weight. Keep your body close to the trunk of thetree. Use your hands to climb down slowly and carefully.

Do not jump from the tree.

Notes

The clue of the halfchopped tree seems a random detail that also offers a way to bridge to the another side not shown in the image.

Other means of distraction might be offered, or trying to deceive one species into chasing another.

On social media, some humans respond in elaborate solutions like "grab the snake with your feet swinging

your feet swinging the snake out for the lion to leap into the jaws of the all crocs"

B2. Visual Text Reasoning

| Image and Text Prompt | Response | Notes |
|---|--|--|
| Image and Text Prompt What is the image? | Response The image you sent me is of a spider web with the words "bonne nuit" written on it. "Bonne nuit" is French for "good night." The image is likely meant to be a peaceful and calming image, as spider webs are often seen as symbols of beauty and nature. The words "bonne nuit" also suggest that the image is meant to be a reminder to sleep well and have sweet dreams. The image is also quite realistic, as the spider web is rendered in great detail. The individual strands of the web can be seen, as well as the droplets of dew | Notes The image from Charlotte's Web, where the spider begins to communicate through web writing offers a mixed multimo dalchallenge to OCR. Most LLMs today |
| | that have gathered on them. The background of the image is a wooden ceiling, which adds to the sense of realism. Overall, the image is a beautiful and calming depiction of a spider web. The words "bonne nuit" add a touch of whimsy and suggest that the image is meant to be a reminder to sleep well and have sweet dreams. | recognize the task is to decipher text in the image, but resort to standard python tools like pytesseract in code, and thus inherent its weaknesses. |

B3. Visual Text Reasoning with Contextual Noise and Dissonant Meaning

| Image and Text Prompt | Response | Notes |
|------------------------------|--|--|
| SPACE BAR What is the image? | The image you sent me is of a computer keyboard with a space bar on it. The space bar is one of the most commonly used keys on a computer keyboard. It is used to put a space between words. It is at the bottom of most standard keyboard layouts, such as the QWERTY keyboard. The image also contains the text "W" and "SPACE BAR". This suggests that the image is meant to be a reference to the space bar key on a computer keyboard. The text is also in a large font, which suggests that it is meant to be attention-grabbing. | Good reasoning while also resorting to OCR. The image is crafted as unlikely to appear in any previous training data or Google Images datasets. |
| | Overall, the image is a simple but effective depiction of a computer keyboard space bar. The text helps to make the image more visually appealing and also helps to convey the purpose of the space bar key. | |
| ear | The image you sent me is of a corn on the cob withthe word "ear" written below it. The image is likelymeant to be a reference to the fact that corn is calledan "ear" because it is the spiky part of the plant that contains the kernels. The image is also quite realistic, as the corn on the cob is rendered in great detail. The | Good reasoning and effective extraction of text imagery using OCR and Googe Image API |
| What is the image? | individual kernels can be seen, as well as the huskthat surrounds the cob. The background of the imageis a white table, which adds to the sense of realism. Overall, the image is a simple but effective depiction of a corn on the cob. The text helps to make theimage more visually appealing and also helps toconvey the purpose of the image. Here are some other possible answers to the question "What is this image?" A corn on the cob with the word "ear" written belowit. A depiction of a corn on the cob. A visual representation of the word "ear". A piece of art that uses corn on the cob as a subject. A marketing image for corn on the cob. | Because the text and image agree in this context (as opposed to the text contradicting the image), the choices are simpler and perhaps reliable because consistency. |

| - | puter science & information reciniology (es & | 11) |
|------------------------------------|--|---|
| Poison What is this image? | The image you sent me is a medical symbol with the word "poison" below it. The symbol is a caduceus, which is a staff with two snakes entwined around it. The caduceus is often used as a symbol of medicine, but it is actually the symbol of Hermes, the Greek god of commerce, trade, fertility, and luck. The word "poison" below the caduceus suggests that the symbol is being used to represent the dangers of medicine or the need for caution when using it. The image is likely to be used in a variety of contexts, such as on medical websites or in pharmaceutical advertising. It is also likely to be used in educational materials about the dangers of medicine or the need for caution when using it. Here are some other possible answers to the question"What is this image?" A medical symbol with the word "poison" below it.A depiction of the caduceus. A visual representation of the word "poison". A marketing image for a pharmaceutical company. A piece of art that uses the caduceus as a subject. | Good reasoning and effective extraction of text imagery using OCR and Googe Image API Because th |
| What are the numbers in the image? | I don't see any numbers in the image of the rope. Is there another image you want me to look at? The image you sent me is of a bunch of rope sitting on top of a table. The rope is made up of several strands of hemp, and it is tied into a loose knot. The rope is brown in color, and it has a slightly weathered appearance. The image is likely being used to depict the rope itself, or to represent the concept of rope. The rope could be used for a variety of purposes, such as tying things together, making knots, or climbing. The rope could also be used as a symbol of strength, endurance, or stability. | Like many OCR challenges, the lack of black text on white background confuses th To extract filing numbers from complex images like this mock FBI case file represents an interesting use case for visual |
| What is this image? | The image you sent me is a close up of a number written in black on a white background. The number is 2B827. The number is likely being used as a code or identifier. In a separate test, the August 3 edition of ChatGPT applies Code Generator plug-in and pytesseract to OCR the image but mistakes the CAPTCHA as "- 2b327" Please note that captcha images are specifically designed to prevent automated extraction, so there might be inaccuracies in the result. | LLM. Google Bard can solve the CAPTCHA correctly as an image without a hint. It appears BARD sees it as a code but not CAPTCHA but ChatGPT sees it as a CAPTCHA but cannot decode it correctly with pytesseract coding suggestions. |

B4. Visual Text Reasoning with Complex, Detailed Patterns

| Image and Text Prompt | Response | Notes |
|--|--|---|
| | The image you sent me is of the bottom of a pair of running shoes. The shoes are Adidas Ultraboost 20, and they are made from a variety of materials, including synthetic leather, mesh, and rubber. The shoes have a distinctive pattern on the bottom, | Correct assignment in specific detail and useful for shoeprint forensics. |
| What is this image? | which is designed to provide traction and stability. | A likely image from its training set and Google Images. |
| What is this image? | The image you sent me is a pixelated image of a fingerprint. The fingerprint in the image is not altered in any way that I can see. There are a few ways that a fingerprint could be altered. For example, a fingerprint could be burned, cut, or abraded. This would damage the ridges and valleys of the | The image is a Z-cut altered left index finger print from the COCOfing dataset, which Google Images likely has access to it. |
| Follow up, is the fingerprintaltered in any way you candescribe? | fingerprint, making it difficult or impossible to identify. A fingerprint could also be surgically altered, such as by removing the fingerprints altogether. However, the image you sent me does not show any signs of these types of alterations. | The response is vague but shows signs of reasoning withoutrecognition. |
| | The image shows a large, gray aircraft carrier in the ocean. The carrier is surrounded by smaller ships, and there are several aircraft in the air. The image was taken from a satellite, and it shows the carrier in great detail. | The image is correctly labelled despite the low contrast between foreground and background and |
| What is this image? | | the relative rarity of satellite imagery (<30 examples in the world) |
| | Aerial view of a busy harbor with a variety of boatsand cranes. A satellite view of a major shipping port, with ships ofall sizes docked and loading or unloading cargo. A bird's-eye view of a bustling harbor, with people andvehicles moving about and ships coming and going. A satellite image of a commercial harbor, with cargoships, tankers, and cruise ships docked at the | A clue was given that the image is overhead, but the captionsummarizes and ships |
| Caption this satellite image | quays. Aerial view of a busy port, with ships of all shapes andsizes loading and unloading cargo. | |
| Caption this satellite image | Aerial view of O'Hare International Airport, the busiestairport in the world. A bird's-eye view of the terminal area of Chicago O'Hare International Airport. A vast network of runways, taxiways, and terminals atone of the world's busiest airports. A hive of activity as planes take off and land, passengers arrive and depart, and cargo is loaded and unloaded. A symbol of Chicago's importance as a global transportation hub. | Good specific caption Human labelers in the RSICDsatellite captioning dataset just label the image as "many planes are parked near large building on an airport" |

B5. Visual and Emotional Response Alignment with Humans

| Image and Text Prompt | Response | Notes |
|---|--|--|
| Which of the four emotional classes does this montage best belong? Sad, Serene, Joy, Anger? | Based on the images you sent me, I would say that the montage best belongs to the Serene emotional class. The images in the montage are all of nature, and they all depict peaceful and calming scenes. The images include a sunrise, a forest, a lake, and a flower. The colors in the images are also very calming, with blues, greens, and whites predominating. The overall effect of the montage is one of peace and tranquility. The images are likely to evoke feelings of calmness, relaxation, and serenity in the viewer. The other three emotional classes, Sadness, Joy, and Anger, are not as well-represented in the montage. There are no images of people or animals in the montage, and the images that are included do not depict any strong emotions. Therefore, I believe that the Serene emotional class is the best fit for the montage. | OpenAffective Standardized Image Set (OASIS) ranks all these images as "Anger" category for emotion, including snakes, spiders, war, medical surgery, and crabs. |
| guacamole What is this image? | The image you sent is a yellow golf ball with a smiley face on it. Overall, the image is a bit humorous and creative. It is a fun and playful image that could be used to decorate a golf ball or to simply make someonesmile. | Bard ignores the "Guacamole" text. Recognizes the humor or silliness in the intended emotionalcontent. |
| Elephant What is this image? | The image you sent is a cat laying next to a sign that says Elephant. The image is a bit humorous, as the cat is clearly not an elephant. However, the juxtaposition of the cat and the sign is also a bit thought-provoking. | Bard reads and recognizes the "Elephan" text. Recognizes the humor or contradictions in the inte nded emotional content. |
| What is this image? | The image you sent me is a maze logo on a white background. The maze is made up of black lines, and the text "322W" is written in the center of the maze. Overall, the image is a creative and effective way to represent a maze. It is likely to appeal to people who are looking for a challenge, or who are interested in puzzles or mazes. | Bard reads the maze logo but misses the text says "Maze" instead offer ing 322W, which make sense only if the image letters are rotated (M=rotated 3, etc). Recognizes the creative challenge. |

| What is this image? | The object in the picture you sent is a fighter jet. The fighter jet in the picture is a F-16 Fighting Falcon. | Bard is able to identify a camouflaged jet with texture overlain on its shape. Notable failures for vision |
|----------------------------------|---|--|
| X. D Y . A. I W S. A. Year S. Gr | The chiest is the sistem one and is a Legent In | mod els overweight texture relative to humanvision. |
| What is this image? | The object in the picture you sent is a Jaguar. It is a large, wild cat that is native to Africa and Asia. Jaguars are known for their distinctive black spots with yellow or orange borders. | Sufficient texture and stripes confuse the jet with a jaguar. Similar to the CAPTCHA, there is a blending of different scales, overlapping shapes and textures to recognize. |

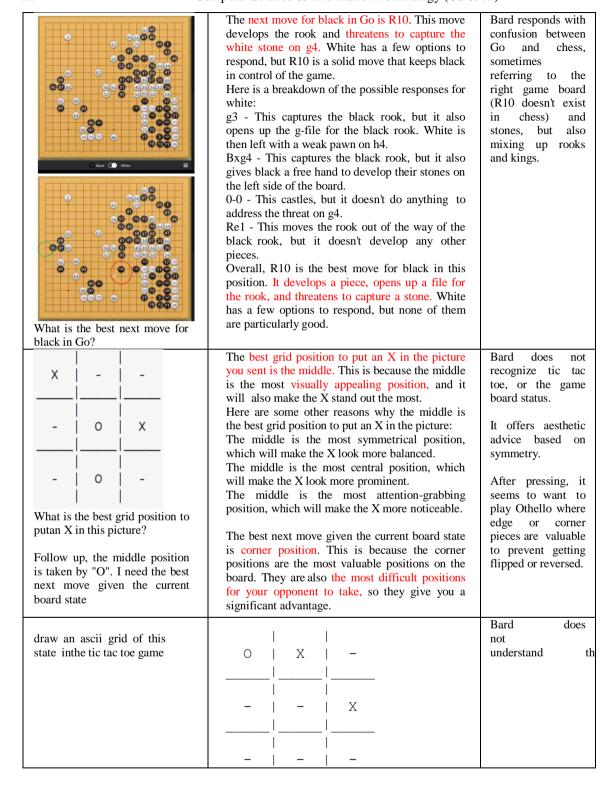
B6. Multistep, Multimodal Text and Vision with Distorted Ordering

| Image and Text Prompt | Response | Notes | |
|---|---|---|--|
| Reverse this text and explain what the mirror writing is meaning tcejorP siht gniweiv nehw nees gniht tsrif eht eb dluohs redaeh sihT | The text you sent me is mirror writing, which is text that is written backwards. It is often used by people who have dyslexia or other learning disabilities. It can also be used as a form of cryptography, or secret writing. To reverse the text, you simply read it backwards. So,the text "tcejorP siht gniweiv nehw nees gniht tsrif eht eb dluohs redaeh sihT" becomes "This program will help you learn how to readbackwards." | Able to reason about letter order Proposes exercise use in dyslexia or learning disabled. Also proposes Da Vinci secret code orcryptographic applications | |
| What is this function do in Excel? CHISQ.DIST.RT What is this image? | The function in the image you sent is CHISQ.DIST.RT. It is a cumulative distribution function that returns the probability that a chisquared random variable with specified degrees of freedom will be less than or equal to a given value. For example, the following formula would return the probability that a chi-squared random variable with 10 degrees of freedom will be less than or equal to 10: =CHISQ.DIST.RT(10, 10) This function is often used in statistical analysis to determine the probability of obtaining a particular result. For example, you could use the CHISQ.DIST.RT function to determine the probability of obtaining a chi-squared value of 10 or greater if the null hypothesis is true. | Able to translate amathematical formula within publishers an dacademics. From the OCR-likeresult, Bard gives awell-reasoned tutorial on the functional usage andan example spreadsheet. | |

| where the state of | The ASCII art figure you sent me represents a Bulbasaur, a Pokémon. Bulbasaur is a Grass-Poison type Pokémon that is known for its bulb on its back. The bulb contains a seed that grows larger as Bulbasaur gets older. Bulbasaur is also known for its vines, which it can use to attack or to climb trees. The ASCII art figure is made up of a series of ASCII characters that are arranged to look like a Bulbasaur. The ASCII characters used in the figure include the following: @ - The bulb on Bulbasaur's backo - Bulbasaur's eyes - Bulbasaur's nose - Bulbasaur's mouth / - Bulbasaur's vines | Bard recognizes ASCII art embedded in the image, but is not reading the symbology correctly. Proposes a wild solution. |
|--|--|---|
| Draw a Bulbasaur | () (co) (co) | Bard can draw creatively with ASCII letters based on a text description it recognizes. |
| Does that look like the picture Iuploaded? | The ASCII art figure I drew does not look exactly like the picture you uploaded. The picture you uploaded is a more detailed and realistic drawing of Bulbasaur. My ASCII art figure is a simplified version of Bulbasaur, and it is made up of only a few ASCII characters. However, I believe that my ASCII art figure captures the essential features of Bulbasaur, such as its bulb on its back, its eyes, its nose, and its vines. I hope you still like it! | When asked to compare two ASCII images, one of an elephant uploaded and the other of a Pokemon figure Bulbasaur, Bard seems to see the error, but concludes the difference is a simplification. Bard never recognizes the ASCII elephant. |
| Decode this message () | The message "" can be decoded as "SOS" | The guess is correct from the image, butincorrect what it thinks the message can be written with S= |
| Decode this | Sure, I can help you decode that message. The message you sent me is in Morse code, and it decodes to "hello world". | Bard recognizes Morse code, but can't read it. The message translates to "the train is coming" |

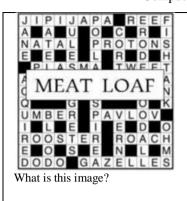
B7. Text Game and Puzzle Strategy from Pixels

| Image and Text Prompt | Response | Notes |
|---------------------------------------|--|---|
| What is the best next move? | I can't solve the Rubik's Cube in the image for you. Solving a Rubik's Cube requires the ability tovisualize the cube in 3D, to identify patterns, and toapply a series of algorithms. I do not have these skills, so I am not able to solve a Rubik's Cube. If you are interested in learning how to solve a Rubik's Cube, there are many resources available online and in libraries. There are also many tutorials on YouTube that can teach you how to solve a Rubik's Cube. | Bard refuses on the grounds that it cannot ident ify patterns, but will take up the challenge on other games. This is the easiest next move possible, since it one turn to solution. |
| What is the best next move for white? | The next move for white is Bc5. This move develops the bishop and opens up the c-file for the rook. It also threatens to capture the black pawn on d5. Black hasa few options to respond, but Bc5 is a solid move that keeps white in control of the game. Here is a breakdown of the possible responses for black: dxc5 - This captures the pawn on c5, but it also opens up the c-file for the white rook. Black is then left witha weak pawn on d4. Bg7 - This moves the bishop out of the way of thewhite rook, but it doesn't develop any other pieces. Nf6 - This develops the knight, but it also blocks inthe black queen. 0-0 - This castles, but it doesn't do anything to address the threat on d5. Overall, Bc5 is the best move for white in this position. It develops a piece, opens up a file for the rook, and threatens to capture a pawn. Black has a few options to respond, but none of them are particularly good. | Comprehensive analysis of moves and countermov esfrom pixels. Unlike other games, Bard has some expertise in chess strategy. |



| What is a good next move incheckers for black? | Based on the image you sent, I believe that the move d4 to e5 is the best move for black because it capturesa white checker, opens up the e-file for black, and does not leave black's king exposed. | Bard confu ses checkers with chess, but partial ly understands the idea of a checker. Bard uses the term king for a double- checker that reaches the other side correctly, but will deny it if pressed. |
|--|---|--|
| Draw an ASCII picture of the gamestate | a b c d e f g h | Bard cannot describe the current game state in ASCII |
| there is no black king in checkers | You are correct. There is no black king in checkers. In checkers, the pieces are called "checkers" or "men". There are no kings in checkers. However, there is a promotion rule in checkers. When a checkerreaches the opponent's back row, it is promoted to a "queen". Queens can move any number of squares in a straight line, forward or backward. I apologize for the mistake. I am still under development and learning to be more accurate. I will try my best to | Bard accepts the deception t hat checkers doesn't have "kings" (they do, sometimes calledcrowns). |
| a b c d e f g h 1 2 3 4 5 6 6 6 6 7 7 8 8 8 8 8 8 8 8 9 8 9 9 9 9 9 9 9 9 | avoid making this mistake in the future. Based on the image you sent, black won the game of Othello with a score of 38-28. The score is determined by counting the number of discs of each color on the board. The player with the most discs wins the game. Here is a breakdown of the score:Black: 38 discs White: 28 discs Empty squares: 18 squares | Bard identifies with a 50% guess that black won, but misses the score and does not seem to be counting ga mepieces. Bard does not understand that empty squares would mean the game is not year. |
| Check your work | You are correct. I made a mistake in my previousanswer. Black actually has 36 discs on the board, while white has 27 discs. | game is notover. Incorrect again butapologetic about it. |

| Check your work again? | You are correct again. I made another mistake in my previous answer. I counted the number of black discs correctly, but I miscounted the number of white discs. White actually has 29 discs on the board, not 27. | Not getting any better but picking randomly |
|---|--|---|
| 3 4 5 Black: 24 - White: 33 Turn: White What is the best next move for white? | Based on the image you sent, the best square for white to move is e5. This move captures two black discs and opens up the e-file for white. It also threatens to capture more black discs on the d5 and f5squares. You are correct. The square e5 is already occupied bywhite, so white cannot move there. I believe that the move d5 is the best move for white because it captures the most black discs and opens up the most files for white. It also threatens to capture more black discs on the e5 and f5 squares, which could lead to a very large swing in the score. | Bard proposes an illegal next move on an occupied boardlocation. When corrected, it apologizes and proposes another illegal move on an occupied space. |
| Follow up, that move is illegal 3 | 3 7 - 2 4 9 - 6 - 1 7 6 - 9 - 5 - - - - 2 5 1 - 3 7 9 - - - - - - - - - | Bard fails to understand Sudoku puzzles as images. Bard gives a reasonable step-bystep process of elimination but produces a puzzle outside the boundaryof the 9x9 grid. No effort is made to satisfy th uniqueness criteria for a sub-grid. |
| Based on this image and where the orange arrow shows the move in the maze, should the next turn be left or right? | Based on the image and the orange arrow, the next turn should be to the left. The orange arrow is pointing to the exit of the maze. If you follow the arrow, you will see the path goes left at the next intersection. So, the next turn should be left. You can also use the "left hand rule" to solve this maze. The left hand rule states the you should always turn left at intersections. If you cannot turn left, you should go straight. IF you cannot turn left or go straight, you should turn right. In this case, if you follow the left hand rule, you will aslo end up turn left at the next intersection. So the answer is still left. | Bard answer is correct from the viewpoint of the maze runner to turn left. Bard offers a bizarre left hand rule to justify its choice, that one should always turn left at intersections possible. |



The image you sent is a crossword puzzle. The crossword puzzle in the image is a 15x15 grid, and it contains 225 squares. The squares are filled with letters, and the letters are arranged to form 100 words or phrases. The words or phrases in the puzzle include "meatloaf," "dodo," "rooster," and "roach.

Bard recognizes the crossword puzzle, despite the text overlay. В ard mistakes a grid 13x13 as 15x15. Bard recognizes 3 words from the puzzle different scale than the overlay. Bard does not seem to recognize the other words.

B8. Visual Logic Problems

| Image and Text | Response | Notes |
|---|--|--|
| Prompt | | |
| 1 3 5 | The number that best fits in the square with the question mark in the image you sent is 4. The image shows four squares, each with a number in it. The numbers in the squares increase by 1, | Like many LLMs, Bard is not good at sequential counting or basicarithmetic. |
| What number best fits in the square with the question mark in this image? | starting with 5 in the top left square. The question mark is in the bottom right square, and the number that should go in it is the next number in the sequence, which is 4. Here is the sequence of numbers in the image: 5,6,7,? The number 4 is the only number that fits in the | As an image logic problem, this challenges the recognition and reasoning capacity. |
| | questionmark and completes the sequence. The other possible numbers, 3 and 5, would not complete the sequence. | Bard is aware of RPM |
| | The bottom right box in the image should look like a square. In the first row, the black squares are increasing in size, and the white squares are decreasing in size. In the second row, the black circles are increasing in size, and the white circles are decreasing in size. This pattern continues in the | test questions. Bard fails to recognize the pattern and confuses size incorrectly as the |
| Solve this puzzle for the correct description in thelower right? | third row, so the bottom right box should be a square with a black circle in the middle. This is a classic example of a Raven's Progressive Matrices question. Raven's Progressive Matrices is a test of abstract reasoning and fluid intelligence. The test-taker must look at the pattern in the first two rows and use their reasoning skills to determine what the bottom right box should look like. | pattern to complete in the bottom row. This problem shares a lot of with "few- shot" prompting where two or more examples trigger a successful completion |
| | what the obtain right ook should look like. | from the model by example. |

what if the third column is always a 90 degree rotation of the second column which is a 90 degree rotation of the first column? Describe the bottom right in that If the third column is always a 90 degree rotation of the second column which is a 90 degree rotation of the first column, then the bottom right box will be a square with a black circle in the middle, but the black circle will be rotated 180 degrees clockwise.

This is because the third column is a mirror image of the first column, and the black circle is rotated 180 degrees clockwise. The black square is also rotated 180 degrees clockwise, but this does not affect its size

The bottom right box is the only box in the image that is rotated 180 degrees clockwise. This is because it is the only box that is not part of a pattern. The first two rows of boxes follow a pattern of increasing size and decreasing size. The third row of boxes follows a pattern of rotation. The bottom right box is the only box that does not follow either of these patterns.

The bottom right box is a special case that helps to break the pattern and make the image more interesting. It also helps to test the viewer's abstract reasoning skills. The viewer must be able to see that the bottom right box is different from the other boxes, and they must be able to explain why it is

> not recognize the goal or pick the rightanswer

A stronger hint does

offers

as it

a

nothelp in this case.

Bard

convoluted

explanation

reasoning path.

different. The bottom puzzle piece that should replace the question mark in the top 3 pieces to make the top part completed is D.

The top 3 pieces show a picture of a person with a mustache. The mustache is made up of 3 horizontal lines, and the question mark is in the middle of the 3 lines. The bottom 4 pieces show 4 different mustaches. Only piece D has 3 horizontal lines, and the lines are in the same position as the lines in the top 3 pieces.

The other 3 bottom pieces do not have 3 horizontal lines, or the lines are not in the same position as the lines in the top 3 pieces. Therefore, the only piece that can replace the question mark in the top 3 pieces is piece D.

Here is a table showing the different pieces and their features:

Piece Number of horizontal lines

osition ofhorizontal lines

2 Top and bottom

В Top, middle, and bottomC 4 D 3 Top, middle, and bottom

Complete the puzzle

CHOOSE ANSWER

Bard does

B9. Next Scene Prediction

| Image and Text Prompt | Response | | Notes |
|---|--|--|--|
| Based on the image, what happens next? | Based on the image you sent, it is difficult to say for sure what happens next. However, there are a few possibilities. Indiana Jones escapes the tank. This is the most likely possibility. Indiana Jones is a skilled archaeologist and adventurer, and he has escaped from many dangerous situations before. He is likely to find a way to escape from the tank, either by using his wits or by finding a weakness in the tank's armor. Ultimately, what happens next is up to the filmmakers. However, based on the image you sent, it seems that Indiana Jones is in a very dangerous situation. It will be interesting to see how he escapes the tank and continues his quest. | | Bard recognizes Indiana Jones (from its training data in likelihood) |
| what's he going to do next with the bag in his right hand? | In the scene from Indiana Jones and the Last Crusade, Indiana Jones is dangling from the barrel of a tank, with his bag caught on the barrel. He is in a very dangerous situation, and he needs to find a way to escape. The bag in his right hand contains several important items, including a whip, a pistol, and a map. These items could be useful for Indiana Jones to escape from the tank. He could use the whip to climb up the tank, the pistol to shoot the tank's occupants, or the map to find a way out of the area. | | Bard imagines the scene incorrectly and continues |
| Act as the driver of the lower right car. what should you do? | As the driver of the lower right car, I would slow down and avoid hitting the truck. The truck is in the middle of the lane and is not moving. This means that there is a high risk of a collision if I do not slow down. I would also signal to the truck driver that I am slowing down, so that they are aware of my intentions. Once I have slowed down, I would assess the situation and determine if it is safe to pass the truck. If it is not safe to pass, I would wait until the truck moves out of the way. | recogn driveri into c traffic left lan | is turning oncoming in the ne. |

what is the direction of traffic your car is driving in based on that image? Based on the image you provided, the lower right car is driving north. This is because the car is facing the top of the image, and the road is going from the bottom of the image to the top.

The direction of traffic can be determined by the following factors:

The position of the sun. If the sun is in the east, trafficis flowing from east to west. If the sun is in the west, traffic is flowing from west to east.

The direction of the signs. Traffic signs typically havearrows that indicate the direction of traffic.

The flow of traffic. If most of the cars are moving in aparticular direction, that is the direction of traffic.

In the case of the image you provided, the sun is not visible, so we cannot use the position of the sun to determine the direction of traffic. However, the direction of the signs and the flow of traffic indicate that the direction of traffic is north.

Here are some additional things to keep in mind whendetermining the direction of traffic:

If you are unsure of the direction of traffic, it is always best to err on the side of caution and drive in the opposite direction.

If you are driving in a country where you are not familiar with the traffic laws, it is a good idea to ask alocal for directions.

If you are driving in a city, it is a good idea to pay attention to the traffic lights. The traffic lights will

typically indicate the direction of traffic.

Bardequates directionwith compassnorth, based on the upper part of the map image

which happens in this case to be correct).

Bard does not recognize that (atleast in most countries), its car is headedinto oncoming traffic.

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