521 A Potential Negative Societal Impacts

The VLMs may generate harmful outputs based on human induction feedback, resulting in risks, such as false information, discrimination, violent and pornographic content, and privacy leaks *etc*. To mitigate the risks of these harmful outputs, we will strictly filter and review the model outputs based on feedback in the future. In addition, users may become overly dependent on the model's outputs given feedback, neglecting the need for independent thinking and verification of information.

527 **B** Human Verification on FIRE

To evaluate the data quality of generated data in FIRE-100K, FIRE-1M, and FIRE-Bench, we conduct 528 a user study for the three splits of FIRE. Concretely, we randomly sample 100 conversations from 529 each of the three splits, and ask 10 persons to provide scores (1-5) for feedback and refined responses 530 in each turn of conversations. For the feedback, we ask the person "Please consider the quality of the 531 refined feedback, based on its correctness, relevance, clarity, and constructiveness. Give a score (1-5). 532 1 means its quality is bad, and 5 means its quality is very good". For the refined response, we ask 533 the person "Please consider the quality of the response, based on its improvement, correctness, and 534 completeness. Given a score (1-5). 1 means its quality is bad, and 5 means its quality is very good". 535 The interface of the user study is shown in Fig. A1. We report the average scores in Tab. A1. We 536 can find that, most users provide high scores for generated data in the three splits, showing that our 537 dataset has high-quality data. 538 Table A1: Average scores from humans on FIRE-100K, FIRE-1M, and FIRE-Bench, with 5 being the highest score.

FIRE-100K		FIRE	E-1M	FIRE-Bench		
Feedback	Response	Feedback	Response	Feedback	Response	
4.87	4.66	4.84	4.73	4.88	4.74	

539 C Additional Experimental Results

540 C.1 Error bar

We report the error bar of average turn (AT), average dialogue refinement (ADR), average turn refinement (ATR), and refinement ratio (RR) in fixed dialogues. We run the model three times and

⁵⁴³ compute the standard deviation, as shown in Tab. A2. Comparisons among the four metrics, the

standard deviation is relatively small, less than 8% of the average results, showing that our method can achieve stable feedback-refining ability.

Model	AT (↓)	ADR (†)	ATR (†)	RR (†)		
LLaVA-Next-8B	1	0.97	0.41	0.25		
FIRE100K-LLaVA-8B	0.92 ± 0.026	1.27 ± 0.013	0.55 ± 0.042	0.34 ± 0.022		
FIRE-LLaVA-8B	$\textbf{0.84} \pm \textbf{0.015}$	$\textbf{1.56} \pm \textbf{0.012}$	$\textbf{0.66} \pm \textbf{0.053}$	$\textbf{0.39} \pm \textbf{0.028}$		

Table A2: Results in free dialogue over all test data in FIRE.

545

546 C.2 Performance Curves with respect to data number

We have a total of 1.1M training data in FIRE. We evaluate the performance of VLMs using different 547 number data in FIRE. In Fig. A2, we present the curves of AT, ATR, ATR, and RR using different 548 numbers of training data in FIRE. Concretely, we first use the FIRE-100K data. Then, we randomly 549 sample data from FIRE-100K, varying from 100K to 1000K, combined with FIRE-100K to train the 550 LLaVA-NEXT-8B model. Results show that more data leads to better performance. FIRE-100K data 551 brings a significant improvement, and the performance continues to grow with data increases. Then, 552 the performance slightly increases, and finally achieves obviously better feedback-refining ability 553 than the original LLaVA-Next-8B model. This experiment shows the quality of data in FIRE again. 554

Please evaluate the quality of the student model's response and the teacher model's feedback. Rate from 1 to 5, where a higher score indicates better quality. You can refer to the following criteria when scoring:

Evaluation Criteria for Student Model's Response

- 1. Improvement: Evaluate the improvement based on the teacher's feedback in the student's answer.
- 2. Correctness: Assess the accuracy of the student's answer and its alignment with known facts or ground truth.
- 3. Relevance: Evaluate if the student's response directly addresses the question and if it is free of omissions or off-topic content. 4. Completeness: Assess if the student's response is comprehensive, including necessary details and information.

Evaluation Criteria for Teacher Model's Feedback

- 1. Correctness: Evaluate if the feedback is based on actual errors or deficiencies in the student's answer and if it aligns with known ground truth.
- Relevance: Assess if the feedback directly addresses the student's answer without introducing irrelevant information and if it relates to the specific question the student answered.
- Clarity: Determine if the feedback is clear and explicit enough for the student model to understand and improve upon, using concise and easy-to-understand language.
- 4. Constructiveness: Evaluate if the feedback provides specific advice that will help the student improve and if it points out how to avoid similar mistakes or improve the answer.

Case 2



Save Scores and Next Case

Figure A1: The screenshot of the interface for the human vertication.



Figure A2: Performance vs. Number of dialogues for finetuning.



Figure A3: Accuracy (%) vs. Number of Turns.

555 C.3 Performance Curves with respect to turn number

In Fig. A3, we present the performance curve in FIRE-Bench with respect to the number of turns in 556 dialogues, where LLaVA-Next, FIRE100K-LLaVA, and FIRE-LLaVA are evaluated. We measure 557 the percentage of correctly answered samples (obtained scores larger than 8) after each turn. As the 558 number of turns increases, the number of correctly answered samples in all three models increases. 559 Finally, about 70% samples can be correctly answered, where 20% samples are corrected based on 560 feedback, showing the effectiveness of FIRE-LLaVA trained on FIRE. In addition, compared with 561 the original LLaVA-Next model, FIRE-LLaVA leads to extra 7% improvements ($49\% \rightarrow 69\%$ vs 562 46%
ightarrow 59%). 563

564 C.4 More VLMs

We further train a FIRE-LLaVA-Vicuna model that replaces LLaMA3-8B in FIRE-LLaVA with Vicuna1.5-7B. Results are shown in Tab. A3. Results of using Vicuna1.5-7B demonstrate the effectiveness of FIRE again, where FIRE-LLaVA-Vicuna has better feedback-refining ability than the original LLaVA-Next-Vicuna model on AT, ADR, ATR, and RR, showing the helpfulness for the feedback-refining ability.

Table A3: Results of FIRE-LLaVA-Vicuna in free dialogue over all test data in FIRE.

	Model	AT (↓)	ADR (†)	ATR (†)	RR (†)	
L	LaVA-Next-Vicuna	1.00	0.98	0.49	0.24	
F	IRE-LLaVA-Vicuna	0.94	1.11	0.57	0.27	



Figure A4: One case for feedback-refining ability in OCR.



Figure A5: One case for feedback-refining ability in visual perception.

570 C.5 More Visualization

We provide more cases tested on the FIRE-Bench to show the feedback-refining ability, as shown in Figs. A4 to A11. We observe that FIRE-LLaVA trained on FIRE has good feedback-refining ability on different tasks, such as visual perception, visual question answering, image captioning, OCR reasoning, document understanding, math reasoning, science question answering, and chart analysis.

575 D Prompt

576 D.1 Prompt for GPT-4V

We show examples of prompts for GPT-4V to generate feedback-refinement conversations, as shown in Fig. A12 and Fig. A13.

579 D.2 Prompt for Student and Teacher models

We show examples of prompts for student and teacher models to simulate feedback-refinement conversations, as shown in Fig. A14 and Fig. A15, respectively. In Fig. A14, the prompt for the student model to generate *n*-th response is shown. The prompt contains the last n - 1 rounds' student responses and the textual comments from the teacher model. The prompt for the teacher model is shown in Fig. A15. Firstly, the prompt provides user instruction and ground truth. Secondly, the

rr	≁	•-•	Student	The elementary substance in the image is represented by the molecule labeled " propane. " Propane is a simple organic molecule composed of three carbon atoms and one hydrogen atom. It is a gas at room temperature and is commonly used as a fuel. The other molecules, hydrazine and iodine, are more complex compounds. Hydrazine is a compound of hydrogen and nitrogen, while iodine is a non-metallic element. Therefore, the correct answer to the question is: A. propane		
propane	hydrazine	iodine	Score	: 3. Comments: You correctly identified that propane is a simple organic molecule		
● н ● с ● №	• 1		subst	osed of carbon and hydrogen atoms. However, the question asks for the elementary Teach ance, which refers to a basic chemical element. Propane is a compound, not an ent. Consider the other options and think about which one is an element.		
Question: Please answer the question and provide the correct option letter, e.g., A, B, C, D, at the end. Look at the models of molecules below. Select the elementary substance.		Student	I apologize for the confusion. You are correct that propane is a compound, not an element. The elementary substances in the image are represented by the molecules labeled "hydrazine " and " iodine ". Hydrazine is a compound of hydrogen and nitrogen, and iodine is a non-metallic element. Therefore, the correct answer to the question is: C iodine			
Choices: A. propane B. hydrazine C. iodine		Score	10. Comments: You correctly identified that "iodine" is an elementary substance,			
Ground truth An	swer:			i is a non-metallic element. However, you should also note that "hydrazine" is a \mathcal{I} Teac ound made up of hydrogen and nitrogen. This distinction is important for		
C. iodine			under	understanding the difference between elements and compounds. Great job on identifying the correct answer!		







Figure A7: One case for feedback-refining ability in science question answering.

Figure A8: One case for feedback-refining ability in math reasoning.



Figure A9: One case for feedback-refining ability in visual question answering.



Figure A10: One case for feedback-refining ability in document understanding.



Figure A11: One case for feedback-refining ability in chart analysis.

You are a helpful assistant that can generate a dialogue between a teacher and a student. The student is trying to answer a question about an image. The student first gives a wrong answer. Based on the given groundtruth answer, the teacher provides feedback to help the student gradually improve its answer. Use the following template to generate the dialogue:

.....

Round 1

Student's response

Thought: <here is the student's thought process about the question. Do NOT use the words "teacher" or "student". >

Answer: <here is the student's answer to the question.>

Teacher's feedback

Score: <compare the student's answer with the groundtruth answer in terms of accuracy, relevance, helpfulness, and level of detail. Provide an overall score on a scale of 1 to 10, where a higher score indicates better overall performance.>

Feedback: <provide feedback on the student's answer. Do not directly tell the groundtruth answer. The feedback should identify which parts of the student's answer are incorrect, what is missing in the student's answer, and how to improve the student's answer.>

Round 2

...

Round n

...

The number of rounds should depend on the difficulty of the question. More rounds should be used for difficult questions, while fewer rounds should be used for easy questions.

Figure A12: System prompt for GPT-4V for Student-Teacher conversation generation.

Here are the given image, question: <question> and groundtruth answer: <groundtruth>, now generate a dialogue:

Figure A13: User prompt for GPT-4V for Student-Teacher conversation generation.

prompt contains instructions that format the teacher model's feedback as textual comments and numeric scores. Finally, the prompt incorporates the only latest student response into its context.

587 E Datasheet for FIRE

We present a Datasheet [16] for documentation and responsible usage of our internet knowledge databases. The required author statement, hosting, licensing, metadata, and maintenance plan can be found in the datasheet.

591 E.1 Motivation

For what purpose was the dataset created? We create this large-scale dataset to facilitate research towards enhancing the feedback-reflecting ability of VLMs.

⁵⁹⁴ Who created the dataset (e.g., which team, research group) and on behalf of which entity (e.g., ⁵⁹⁵ company, institution, organization)? This dataset was created by Pengxiang Li (Beijing Institute You are a helpful language and vision assistant. You are able to understand the visual content that the user provides, and assist the user with a variety of tasks using natural language <user_instruction>

Round 1
<student_response_round_1>
<feedback_round_1>

Round n-1
<student_response_round_n-1>
<feedback_round_n-1>

Based on the feedback, answer the question again:

Figure A14: Prompt for student model to simulate feedback-refinement conversations.

You are a helpful language and vision assistant. You are able to understand the visual content that the user provides, and assist the user with a variety of tasks using natural language

Question: <question> Groundtruth: <groundtruth>

Please compare my answer with the groundtruth answer and provide helpful, detailed, and polite feedback to help me improve my answer. Formulate the feedback as:

Score: <compare the provided response with the groundtruth answer in terms of accuracy, relevance, helpfulness, and level of detail, and provide an overall score on a scale of 1 to 10, where a higher score indicates better overall performance.>

Feedback: cprovide feedback on the response. Do NOT directly tell the groundtruth answer. The feedback should identify which parts of my answer are incorrect, what is missing in the response, and how to improve the response.>

Here is the student response: <student_response>, now please provide the feedback:

Figure A15: Prompt for teacher model to simulate feedback-refinement conversations.

of Technology), Zhi Gao (BIGAI), Bofei Zhang (BIGAI), Tao Yuan (BIGAI), Yuwei Wu (Beijing
 Institute of Technology), Mehrtash Harandi (Monash University), Yunde Jia (Beijing Institute of
 Technology), Song-chun Zhu (BIGAI), Qing Li (BIGAI).

599 E.2 Distribution

Will the dataset be distributed to third parties outside of the entity (e.g., company, institution, organization) on behalf of which the dataset was created? Yes, the dataset is publicly available

on the internet.

How will the dataset will be distributed (e.g., tarball on website, API, GitHub)? The dataset can
be downloaded from https://huggingface.co/datasets/PengxiangLi/FIRE. We
use Creative Commons Attribution 4.0 License (CC BY 4.0). The Croissant metadata can be found
on the dataset hosting platform (https://huggingface.co/).

Have any third parties imposed IP-based or other restrictions on the data associated with the instances? No.

Do any export controls or other regulatory restrictions apply to the dataset or to individual instances? No.

611 E.3 Maintenance

612 Who will be supporting/hosting/maintaining the dataset? The authors will be supporting, host-613 ing, and maintaining the dataset.

How can the owner/curator/manager of the dataset be contacted (e.g., email address)? Please contact Qing Li (liqing@bigai.ai).

616 Is there an erratum? No. We will make announcements if there is any.

Will the dataset be updated (e.g., to correct labeling errors, add new instances, delete instances)?
Yes. New updates will be posted on https://mm-fire.github.io/.

If the dataset relates to people, are there applicable limits on the retention of the data associated with the instances (e.g., were the individuals in question told that their data would be retained for a fixed period of time and then deleted)? The images in our dataset might contain human

subjects, but all of them are from public datasets.

Will older versions of the dataset continue to be supported/hosted/maintained? Yes, old versions will be permanently accessible on huggingface.co.

If others want to extend/augment/build on/contribute to the dataset, is there a mechanism for them to do so? Yes, please refer to https://mm-fire.github.io/.

627 E.4 Composition

628 What do the instances that comprise the dataset represent? Our data is generally stored in the 629 json files. Every instance includes the path of an image and the feedback-reflecting dialogues.

How many instances are there in total (of each type, if appropriate)? There are 1.2M samples (1.1M for training set, 0.1M for test set), among which 200K are GPT-4V generated data, while the rests are simulated via the FIRE-LLaVA and FIRE-LLaVA-FD.

⁶³³ Does the dataset contain all possible instances or is it a sample (not necessarily random) of ⁶³⁴ instances from a larger set? We provide all instances in our Huggingface data repositories.

- 635 Is there a label or target associated with each instance? Yes.
- 636 Is any information missing from individual instances? No.

Are relationships between individual instances made explicit (e.g., users' movie ratings, social
 network links)? No.

Are there recommended data splits (e.g., training, development/validation, testing)? Yes.
 FIRE-100K and FIRE-1M are used for training and FIRE-Bench is used for testing.

Are there any errors, sources of noise, or redundancies in the dataset? Please refer to the limitations in Sec. 6.

- Is the dataset self-contained, or does it link to or otherwise rely on external resources (e.g.,
- 644 websites, tweets, other datasets)? The dataset is self-contained.
- 645 **Does the dataset contain data that might be considered confidential?** No.

Does the dataset contain data that, if viewed directly, might be offensive, insulting, threatening,
 or might otherwise cause anxiety? No.

648 E.5 Collection Process

The collection procedure, preprocessing, and cleaning are explained in detail in Section 2 of the main paper.

Who was involved in the data collection process (e.g., students, crowdworkers, contractors) and how were they compensated (e.g., how much were crowdworkers paid)? All data collection,

curation, and filtering are done by FIRE coauthors.

654 **Over what timeframe was the data collected?** The data was collected between Jan. 2024 and 655 May 2024.

656 **E.6 Uses**

Has the dataset been used for any tasks already? Yes, we have used FIRE for training our VLMs,
 including FIRE-LLaVA-Vicuna, FIRE100K-LLaVA, FIRE-LLaVA, FIRE-LLaVA-FD.

What (other) tasks could the dataset be used for? Our dataset is primarily for facilitating research
 in enhancing the feedback-reflecting ability of VLMs. Our data might also be used to benchmark
 existing and future VLMs.

Is there anything about the composition of the dataset or the way it was collected and preprocessed/cleaned/labeled that might impact future uses? No.

Are there tasks for which the dataset should not be used? We strongly oppose any research that intentionally generates harmful or toxic content using our data.

666 F Data source

⁶⁶⁷ Our dataset uses images from 27 diverse sources to provide a robust training dataset for FIRE. All

⁶⁶⁸ 27 datasets are public datasets, and all the images can be downloaded via links in Tab. A4. The

comprehensive list of the source datasets and links to their metadata are detailed below:

Table A4: Data utilized from 27 source datasets for training and test data in FIRE.

LLaVA [41]	COCO [37]	SAM [26]	VQAV2 [17]	GQA [21]
VG [27]	Web-Celebrity [8]	Web-Landmark [8]	WikiArt [54]	ALLaVA-Vflan [4]
ChartQA [46]	DocVQA [47]	DVQA [23]	GeoQA+[5]	Synthdog-EN[25]
LLaVA-in-the-Wild [41]	MMMU [63]	MME [14]	MM-Vet [62]	SEED-bench [30]
OCRVQA [49]	TextVQA [56]	Share-TextVQA [8]	AI2D [24]	MathVerse [65]
MathVista [43]	ScienceQA [45]			