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TaskGalaxy: Scaling Multi-modal Instruction Fine-tuning

with Tens of Thousands Vision Task Types

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Introduction

Multi-modal Instruction Fine-tuning

- Task:** Enable LLMs models to understand visual inputs and perform instruction-following outputs--dialogue capabilities based on visual data

Format:

Predicted = MLLM(Input; θ), where Input = {<images>, <text>}

- Core Challenge:** Poor task diversity limits the generalization ability of MLLMs, leading to biased outputs.

Motivation

Limitations of Existing Multi-modal Fine-tuning Datasets

Insufficient Task-specific Data:

The poor diversity of VQA tasks leads to MLLMs being exposed to out-of-distribution(OOD) situations. 🤖

Expensive, Time-consuming, Labor-intensive Annotations:

Involvement of a large number of specialized personnel, lack of automation. 🤖

Rich Knowledge and Strong Ability of MLLMs

- GPT-4o:** Powerful text understanding, visual-text understanding, reasoning abilities
- CLIP:** Robust text-to-image alignment capabilities
- GLM-4v, InternVL-Chat, InternVL2:** Strong visual understanding and question answering ability, skilled at solving multimodal tasks

Contribution

- A novel multi-modal instruction fine-tuning dataset, **TaskGalaxy**, which contains **tens of thousands** of vision task types and approximately 413k samples, addressing the limitation of task diversity in existing datasets
- An almost **fully automated** pipeline for creating a comprehensive fine-tuning dataset of diverse task types was designed, which can be **flexibly expanded** by incorporating high-quality images, task types, and question-answer samples
- Incorporating TaskGalaxy into the fine-tuning of LLaVA-v1.5 and InternVL-Chat-v1.0 resulted in **improvements across all 16 benchmarks** compared to fine-tuning with the original data, proving expanding the diversity of visual task types and high-quality question-answer pairs associated with these tasks significantly enhances the generalization capabilities of multimodal models

Pipeline — TaskGalaxy

Task Types

- 100+ task type seeds → 19,227 task types

Cases

Evaluation

Benefits of high task type coverage in TaskGalaxy for the SFT stage

More Task Types, the Better !

TaskGalaxy: Leading the Way in Instruction-Tuning Performance

Model	Method	Benchmarks									
		MME	MMB	MMB ^{CN}	POPE	LLaVA ^W	MMVet	TQA	SQA	MathVista	
LLaVA-v1.5-7B	Baseline	1476	63.29	56.45	86.30	47.70	24.70	57.59	68.77	28.20	
	ShareGPT-4V	1501	65.97	59.10	86.29	49.20	29.00	57.56	70.60	28.20	
	LLaVA-OneVision	1251	59.79	52.84	83.90	51.20	29.60	52.99	73.19	28.20	
	ALLaVA-4V	1474	60.13	55.39	84.21	38.00	27.00	53.77	70.05	29.20	
	Cambrian-1	1494	61.08	54.46	85.46	52.00	25.70	53.17	71.03	29.10	
	TaskGalaxy	1520	66.62	59.43	86.40	52.30	28.60	58.08	71.06	29.30	
	ChartQA	AI2D	Q-Bench	Q-Bench ^{CN}	HalluBench	SEED	MMMU	Average (w/o MME)	43.62		
	Baseline	14.40	25.29	24.89	31.26	47.95	38.62	19.70	43.62		
	ShareGPT-4V	17.84	27.08	26.22	32.51	48.79	59.26	15.60	44.48		
	LLaVA-OneVision	18.72	27.95	24.48	33.51	47.74	30.06	17.00	41.68		
InternVL-Chat-v1.0-7B	Baseline	1481	60.22	53.01	84.17	43.70	26.80	52.61	67.71	33.00	
	TaskGalaxy	1512	65.03	57.91	86.23	52.30	30.10	56.15	68.88	30.10	
	ChartQA	AI2D	Q-Bench	Q-Bench ^{CN}	HalluBench	SEED	MMMU	Average (w/o MME)	47.17		
	Baseline	14.12	35.92	42.89	43.73	51.94	59.06	26.90	47.17		
	ShareGPT-4V	14.52	35.59	46.69	36.38	52.36	47.24	30.30	42.48		
	LLaVA-OneVision	13.76	22.75	40.08	42.89	53.39	40.87	24.60	40.20		
	ALLaVA-4V	12.99	28.28	42.87	44.16	51.41	48.36	27.30	42.94		
	Cambrian-1	16.00	36.69	48.00	41.33	54.63	56.24	30.60	46.98		
	TaskGalaxy	15.16	37.69	48.21	46.32	53.00	60.44	32.80	49.63		
	ChartQA	AI2D	Q-Bench	Q-Bench ^{CN}	HalluBench	SEED	MMMU	Average (w/o MME)	47.17		

The Benefits of Chain-of-Thought

Model	Method	Benchmarks									
		MME	MMB	LLaVA ^W	MathVista	ChartQA	Q-Bench	MMMU	Average (w/o MME)		
LLaVA-v1.5-7B	Baseline	1506	64.69	53.0	26.7	14.72	26.08	16.6	44.46		
	Baseline-max.5	1506	65.80	53.4	27.3	20.20	36.48	17.4	46.61		
	Baseline-max.5 (CoT)	1523	66.72	64.7	27.9	20.96	43.27	19.3	47.92		

TaskGalaxy remains strong on advanced architecture.

Model	Method	Benchmarks									
		MME	MMB	MMB ^{CN}	POPE	LLaVA ^W	MMVet	TQA	SQA	MathVista	
InternVL-Chat-V2.0-8B	Baseline	1336	68.32	66.46	86.30	63.20	46.17	66.24	90.58	50.10	
	TaskGalaxy	1565	73.88	70.79	86.90	62.85	48.86	70.49	92.71	52.31	
	ChartQA	AI2D	Q-Bench	Q-Bench ^{CN}	HalluBench	SEED	MMMU	Average (w/o MME)	47.17		
	Baseline	76.64	75.88	57.79	56.98	57.51	62.72	40.50	65.86		