

Date	Name	Domain	Focus	Task Types	Metrics	Models	Citation
2020-09-07	MMLU (Massive Multitask Language Understanding)	Multidomain	Academic knowledge and reasoning across 57 subjects	Multiple choice	Accuracy	GPT-4o, Gemini 1.5 Pro, o1, DeepSeek-R1	[1] ⇒
2023-11-20	GPQA Diamond	Science	Graduate-level scientific reasoning	Multiple choice, Multi-step QA	Accuracy	o1, DeepSeek-R1	[2] ⇒
2018-03-14	ARC-Challenge (Advanced Reasoning Challenge)	Science	Grade-school science with reasoning emphasis	Multiple choice	Accuracy	GPT-4, Claude	[3] ⇒
2025-01-24	Humanity’s Last Exam	Multidomain	Broad cross-domain academic reasoning	Multiple choice	Accuracy		[4] ⇒
2024-11-07	FrontierMath	Mathematics	Challenging advanced mathematical reasoning	Problem solving	Accuracy		[5] ⇒
2024-07-18	SciCode	Scientific Programming	Scientific code generation and problem solving	Coding	Solve rate (percent)	Claude3.5-Sonnet	[6] ⇒
2025-03-13	AIME (American Invitational Mathematics Examination)	Mathematics	Pre-college advanced problem solving	Problem solving	Accuracy		[7] ⇒
2025-02-15	MATH-500	Mathematics	Math reasoning generalization	Problem solving	Accuracy		[8] ⇒
2024-04-02	CURIE (Scientific Long-Context Understanding, Reasoning and Information Extraction)	Multidomain Science	Long-context scientific reasoning	Information extraction, Reasoning, Concept tracking, Aggregation, Algebraic manipulation, Multimodal comprehension	Accuracy		[9] ⇒
2023-01-26	FEABench (Finite Element Analysis Benchmark)	Computational Engineering	FEA simulation accuracy and performance	Simulation, Performance evaluation	Solve time, Error norm	FEniCS, deal.II	[10] ⇒
2024-07-12	SPIQA (Scientific Paper Image Question Answering)	Computer Science	Multimodal QA on scientific figures	Question answering, Multimodal QA, Chain-of-Thought evaluation	Accuracy, F1 score	Chain-of-Thought models, Multimodal QA systems	[11] ⇒

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2020-09-28	MedQA	Medical Question Answering	Medical board exam QA	Multiple choice	Accuracy	Neural reader, Retrieval-based QA systems	[12] $\Rightarrow$
2025-05-13	BaisBench (Biological AI Scientist Benchmark)	Computational Biology	Omics-driven AI research tasks	Cell type annotation, Multiple choice	Annotation accuracy, QA accuracy	LLM-based AI scientist agents	[13] $\Rightarrow$
2023-01-26	MOLGEN	Computational Chemistry	Molecular generation and optimization	Distribution learning, Goal-oriented generation	Validity percent, Novelty percent, QED, Docking score	MolGen	[14] $\Rightarrow$
2020-05-02	Open Graph Benchmark (OGB) - Biology	Graph ML	Biological graph property prediction	Node property prediction, Link property prediction, Graph property prediction	Accuracy, ROC-AUC	GCN, GraphSAGE, GAT	[15] $\Rightarrow$
2011-10-01	Materials Project	Materials Science	DFT-based property prediction	Property prediction	MAE, $R^2$	Automatminer, Crystal Graph Neural Networks	[16] $\Rightarrow$
2020-10-20	OCP (Open Catalyst Project)	Chemistry; Materials Science	Catalyst adsorption energy prediction	Energy prediction, Force prediction	MAE (energy), MAE (force)	CGCNN, SchNet, DimeNet++, GemNet-OC	[17]–[20] $\Rightarrow$
2023-06-20	JARVIS-Leaderboard	Materials Science; Benchmarking	Comparative evaluation of materials design methods	Method benchmarking, Leaderboard ranking	MAE, RMSE, Accuracy		[21] $\Rightarrow$
2022-02-22	Quantum Computing Benchmarks (QML)	Quantum Computing	Quantum algorithm performance evaluation	Circuit benchmarking, State classification	Fidelity, Success probability	IBM Q, IonQ, AQT@LBNL	[22] $\Rightarrow$
2024-10-01	CFDBench (Fluid Dynamics)	Fluid Dynamics; Scientific ML	Neural operator surrogate modeling	Surrogate modeling	L2 error, MAE	FNO, DeepONet, U-Net	[23] $\Rightarrow$
None	SatImgNet	Remote Sensing	Satellite imagery classification	Image classification	Accuracy		[24] $\Rightarrow$
2023-07-19	ClimateLearn	Climate Science; Forecasting	ML for weather and climate modeling	Forecasting	RMSE, Anomaly correlation	CNN baselines, ResNet variants	[25] $\Rightarrow$
2022-06-09	BIG-Bench (Beyond the Imitation Game Benchmark)	NLP; AI Evaluation	Diverse reasoning and generalization tasks	Few-shot evaluation, Multi-task evaluation	Accuracy, Task-specific metrics	GPT-3, Dense Transformers, Sparse Transformers	[26] $\Rightarrow$
2019-11-20	CommonSenseQA	NLP; Commonsense	Commonsense question answering	Multiple choice	Accuracy	BERT-large, RoBERTa, GPT-3	[27] $\Rightarrow$

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2019-07-24	Winogrande	NLP; Com- monsense	Winograd Schema-style pronoun reso- lution	Pronoun resolution	Accuracy, AUC	RoBERTa, BERT, GPT-2	[28] ⇒

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