# HelpSteer3-Preference: Open Human-Annotated Preference Data across Diverse Tasks and Languages

Zhilin Wang, Jiaqi Zeng, Olivier Delalleau, Hoo-Chang Shin, Felipe Soares, Alexander Bukharin, Ellie Evans, Yi Dong, Oleksii Kuchaiev zhilinw@nvidia.com

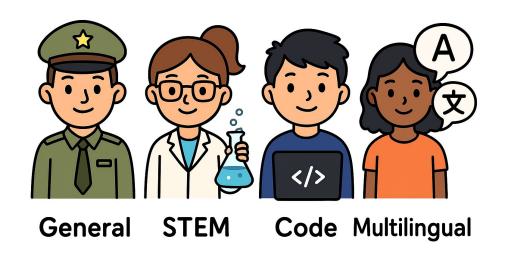
# Limitations of Existing Preference Dataset

Dataset	Released	Quality	Diverse	Multilingual	Annotator	<b>Commercial Use</b>
HH-RLHF [1]	Apr 22	×	×	×	<u>@</u>	<b>✓</b>
Open Assistant [2]	Apr 23	X	X	<b>√</b>	<b>@</b>	<b>✓</b>
UltraFeedback [3]	Oct 23			×	•••	
HelpSteer [4]	Nov 23			X	<u>©</u>	✓
Nectar [5]	Nov 23			×	•	
Skywork-Preference [6]	Oct 24	/		×	<b>9</b> +( <b>1</b> )	•
HelpSteer2-Preference [7]	Oct 24	1		×	<u>@</u>	✓
INF-ORM-Preference [8]	Dec 24	1		×	<b>2</b> +( <b>i</b> )	•
HelpSteer3-Preference (Ours)	Mar 25	1	1	1	<b>©</b>	<b>√</b>

#### Diversity of Tasks that represent LLM Usage

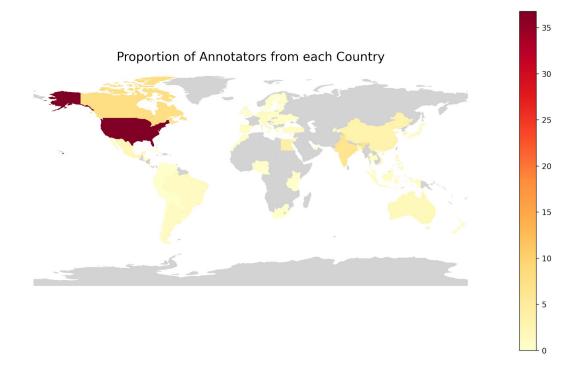
**Existing Preference Datasets are mainly working on General Chat settings** 

Other domains require expert annotators that are hard to recruit for



## Multi-Region Annotators to reflect Global Preferences

#### Annotators from more than 80 countries



## Diversity of Languages across Coding and Multilingual

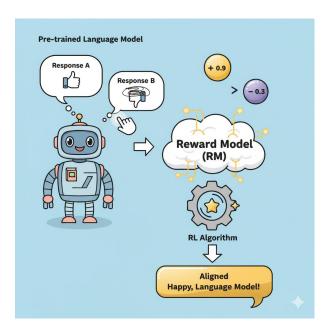
#### 14 Coding languages + 13 Natural Languages

Empowers Preference Learning beyond English-only data in current datasets

Subset	% of Samples by Language													
Code	Python	JS/HTML/CSS	C#	SQL	Java	C++	Go	C	PHP	TS	PowerShell	Rust	R	Bash
Proportion (%)	38.2	23.3	5.5	5.2	5.1	5.0	3.6	3.3	3.3	3.1	1.2	1.1	1.1	1.0
Multilingual	Chinese	Korean	French	Spanish	Japanese	German	Russian	Port.	Ital.	Viet.	Dutch	Polish	Indo	nesian
Proportion (%)	30.2	10.4	10.1	10.1	7.0	6.2	6.0	5.6	5.2	2.4	2.4	2.2	2.1	

#### **Evaluation Metrics**

RM-Bench for Reward Model in RLHF



JudgeBench for LLM-Judges



### Comparison with SOTA Reward Models

Substantially higher (>10%) RM-Bench and JudgeBench compared to existing SOTA models.

	RM-Bench							JudgeBench					
Model	Chat	Math	Code	Safety	Easy	Normal	Hard	Overall	Knowl.	Reason.	Math	Coding	Overall
Bradley-Terry Reward Models													
English RM (General + STEM + Code)	75.4	84.5	69.3	90.4	92.1	85.7	71.1	79.9	70.8	76.5	82.1	66.7	73.7
Multilingual RM	86.2	82.4	66.8	94.1	86.5	85.4	80.0	82.4	66.2	71.4	82.1	59.5	69.4
External Baselines													
Llama-3.1-Nemotron-70B-Reward	70.7	64.3	57.4	90.3	92.2	76.8	48.0	70.7	62.3	72.5	76.8	57.1	66.9
Skywork-Reward-Gemma-2-27B*	71.8	59.2	56.6	94.3	89.6	75.4	50.0	70.5	59.7	66.3	83.9	50.0	64.3
Skywork-Reward-Llama-3.1-8B*	69.5	60.6	54.5	95.7	89.0	74.7	46.6	70.1	59.1	64.3	76.8	50.0	62.3

#### Performance of RLHF models trained with Reward Models

RLHF with Llama 3.3 70B causes it to outperform GPT-4o and Sonnet-3.5 on MT Bench, Arena Hard and WildBench

	MT Bench	Arena Hard			7	VildBench		
Model	(GPT-4-Turbo)	(95% CI)	Overall	Creative	Plan.	Data Analy.	Info. Seek.	Coding
Llama-3.3-70B-Instruct (Init. Policy)	8.29	62.4 (-2.5, 2.5)	52.5	55.5	54.1	48.2	54.8	51.7
+ RLOO w/ English RM	9.24	<b>87.0</b> (-1.3, 1.3)	60.0	65.0	60.8	52.5	62.2	62.0
+ RLOO w/ Multilingual RM	8.81	69.8 (-1.9, 2.1)	55.5	58.7	56.9	50.8	58.4	54.7
+ RLOO w/ Baseline RM	9.04	80.7 (-1.7, 1.9)	58.9	63.6	60.9	53.4	61.9	57.6
External Baselines								
gpt-4o-2024-05-13	8.74	79.3 (-2.1, 2.0)	59.3	59.1	60.2	57.3	58.6	60.5
Claude-3.5-Sonnet-20240620	8.81	79.2 (-1.9, 1.7)	54.7	55.6	55.6	50.2	55.5	56.5