

Sailor2: Sailing in South-East Asia with Inclusive Multilingual LLMs

Presenter: Longxu Dou

Team: Qian Liu, Fan Zhou, Changyu Chen, Ziqi Jin, Zichen Liu, Sailor 2 Community Members

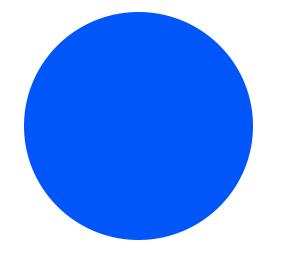
Sailor2: More Data, More Language

We have extracted SEA language corpora from 92 snapshots of CommonCrawl, initially consuming 250TB of disk, which may contain 800B high-quality tokens at most.

Sailor (140B SEA Tokens)



Sailor2 (800B SEA Tokens)

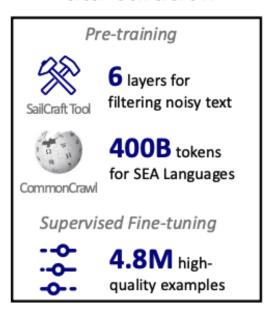


Sailor2: Team Member (SAIL and Community)

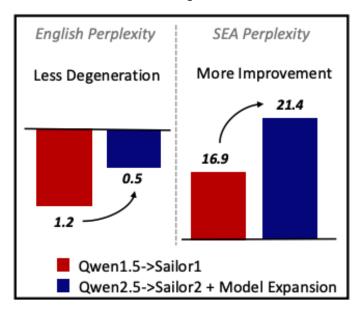


Sailor2: New Milestone of Open SEA Language Model

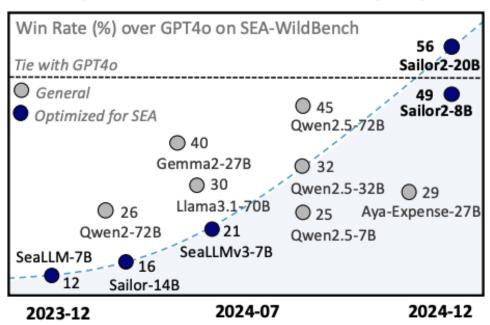
Data Curation



Model Expansion



Open Models for SEA Languages



Sailor2-20B achieves the 50-50 win rate over GPT4o on SEA languages.

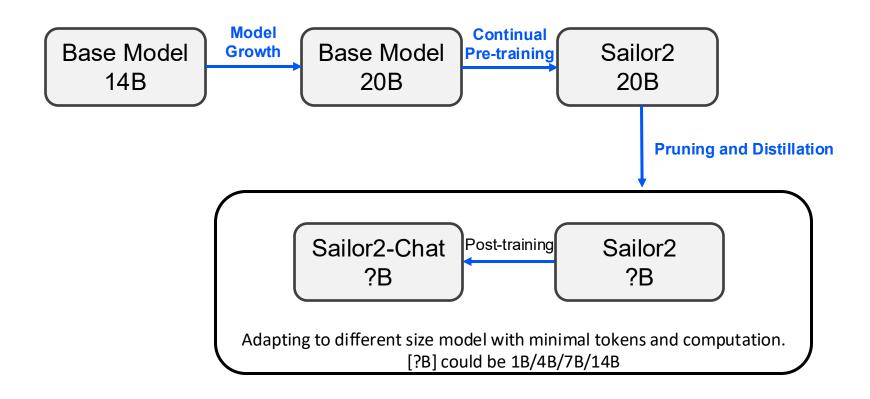
Sailor2: More Supported Languages

| Language | ISO Code | Country/Region | No. of Speakers |
|------------|----------|-------------------------------|-----------------|
| Indonesian | ind | Indonesia | 268 million |
| Vietnamese | vie | Vietnam | 96 million |
| Javanese | jav | Indonesia (Java island) | 82 million |
| Thai | tha | Thailand | 70 million |
| Burmese | mya | Myanmar | 54 million |
| Sundanese | sun | Indonesia (West Java) | 42 million |
| Malay | zsm | Malaysia, Brunei, Singapore | 33 million |
| Tagalog | tgl | Philippines (Luzon) | 28 million |
| Cebuano | ceb | Philippines (Cebu, Mindanao) | 21 million |
| Khmer | khm | Cambodia | 16 million |
| Ilocano | ilo | Philippines (Northern Luzon) | 8 million |
| Lao | lao | Laos | 7 million |
| Waray | war | Philippines (Eastern Visayas) | 3 million |

Sailor2 support 13 SEA languages, Chinese and English.

Sailor2: Training Roadmap

Goal: Training a very large model once to derive multiple smaller models efficiently.



Sailor2: Overview

- **Team**: 16 core members x 6 months
- **Model**: 1B, 8B, 20B
- Data: 500B clean tokens (some repeated more than 1 epoch), including 400B SEA related tokens, from CommonCrawl and Open Source.
- Total Computation in Continual Pre-training: 256 H100 GPUs for 30 days (Community)
- Total Computation in Post-training: 64 A100-80G GPUs for 7 days (SAIL)
- Open Language Model:
 - Release the base model and the chat model.
 - Open-source the data processing pipeline, the fine-tuning code, the evaluation code, preference dataset and evaluation dataset.

Sailor2 Base Model Evaluation

Compared to other advanced multilingual models, Sailor2 demonstrates comparable or better performance, especially in extremely low-resource languages like Javanese.

| Language | Benchmark _(eval) | Sailor2- 8B | Qwen2.5- 7B | Gemma2- 9B | Lllama3.1 | SeaLLM- v3-7B | Sailor2- 20B | Qwen2.5- 32B | Gemma2- 27B | Llama3.1- 70B | Aya- Expanse- 32B |
|------------|-----------------------------------------------------------------------------------------------|-----------------------------|----------------------|-----------------------------|----------------------|-----------------------|-----------------------------|-----------------------------|----------------------|-----------------------------|-----------------------------|
| | Avg. | 57.6 | 52.8 | 52.5 | 47.2 | 43.4 | 62.8 | 59.1 | 61.8 | 61.2 | 51.1 |
| Indonesian | IndoCulture _(0 shot) TydiQA _(3 shot) Belebele _(3 shot) | 73.4 66.4 48.9 | 58.7 63.5 49.3 | 65.6 65.5 50.7 | 56.7 63.4 46.8 | 53.0 65.5 30.6 | 76.4 71.7 52.1 | 68.9 63.9 54.1 | 66.1 65.1 53.3 | 72.7 69.9 56.4 | 70.6 58.2 60.3 |
| Thai | MMLU _(5 shot) M3Exam _(5 shot) Belebele _(3 shot) | 55.4 57.0 43.2 | 52.8 51.7 44.1 | 57.8 52.7 40.6 | 44.1 43.7 43.1 | 50.8 51.3 43.0 | 66.3 69.3 47.4 | 70.7 69.2 49.4 | 62.5 57.0 46.0 | 67.1 63.7 52.3 | 39.6 38.6 45.3 |
| Vietnamese | VMLU _(3 shot) M3Exam _(3 shot) Belebele _(3 shot) | 56.2 65.6 48.7 | 52.6 66.4 50.8 | 51.7 65.5 49.0 | 48.9 54.4 46.0 | 56.8 63.1 48.6 | 65.9 74.6 53.8 | 64.9 77.3 54.6 | 59.1 68.6 52.0 | 63.9 68.9 61.8 | 65.9 63.2 58.3 |
| Malay | Tatabahasa _(3 shot) | 67.3 | 41.5 | 53.6 | 42.9 | 37.4 | 67.3 | 50.4 | 58.6 | 58.3 | 48.1 |
| Javanese | M3Exam _(3 shot) | 57.1 | 35.9 | 45.3 | 40.4 | 38.5 | 62.3 | 47.7 | 49.1 | 53.4 | 46.1 |
| Multiple | FLORES-200 _(3 shot) XCOPA _(3 shot) | 35.4 74.1 | 30.6 71.8 | 35.8 73.0 | 31.7 69.4 | 29.6 70.4 | 35.8 77.5 | 34.3 77.3 | 36.6 75.3 | 36.5 79.8 | 35.7 72.1 |

Sailor2 Chat Model Evaluation

The win rate of Sailor2-20B-Chat against GPT-4o-0806 on Sea-WildBench is nearly 50%, demonstrating it performs at the GPT-4o level for local chat scenarios.

| Model | SWB Score | Coding | Creative Tasks | Info Seeking | Reasoning | Math | Length |
|--------------------------|-----------|--------|----------------|--------------|-----------|------|---------|
| Sailor2-20B-Chat | 0.56 | 0.62 | 0.56 | 0.58 | 0.57 | 0.54 | 2814.74 |
| Sailor2-8B-Chat | 0.49 | 0.42 | 0.57 | 0.53 | 0.50 | 0.42 | 2849.41 |
| Qwen2.5-72B-Instruct | 0.45 | 0.50 | 0.39 | 0.44 | 0.45 | 0.49 | 3026.82 |
| SEA-LIONv3-70B-Instruct | 0.40 | 0.42 | 0.38 | 0.40 | 0.39 | 0.39 | 2340.65 |
| Gemma-2-27B-Instruct | 0.40 | 0.38 | 0.41 | 0.39 | 0.39 | 0.37 | 2288.33 |
| Qwen2.5-32B-Instruct | 0.32 | 0.39 | 0.28 | 0.29 | 0.32 | 0.33 | 2090.61 |
| Gemma-2-9B-Instruct | 0.31 | 0.26 | 0.36 | 0.33 | 0.30 | 0.26 | 2163.03 |
| Qwen2.5-14B-Instruct | 0.30 | 0.33 | 0.25 | 0.28 | 0.28 | 0.30 | 2267.94 |
| Llama-3.1-70B-Instruct | 0.30 | 0.37 | 0.26 | 0.28 | 0.28 | 0.28 | 2543.06 |
| SEA-LIONv3-8B-Instruct | 0.30 | 0.32 | 0.32 | 0.30 | 0.28 | 0.22 | 2357.14 |
| Aya-Expanse-32B | 0.29 | 0.29 | 0.28 | 0.28 | 0.27 | 0.24 | 2495.47 |
| Qwen2-72B-Instruct | 0.26 | 0.22 | 0.27 | 0.28 | 0.25 | 0.23 | 1546.21 |
| Qwen2.5-7B-Instruct | 0.25 | 0.28 | 0.20 | 0.23 | 0.22 | 0.22 | 2415.08 |
| SEA-LIONv2.1-8B-Instruct | 0.23 | 0.23 | 0.24 | 0.24 | 0.20 | 0.18 | 1735.26 |
| SeaLLMs-v3-7B-Chat | 0.21 | 0.21 | 0.19 | 0.19 | 0.18 | 0.15 | 2298.47 |
| Llama-3.1-8B-Instruct | 0.19 | 0.18 | 0.15 | 0.16 | 0.15 | 0.13 | 2356.67 |
| SeaLLM-7B-v2 | 0.18 | 0.14 | 0.16 | 0.17 | 0.14 | 0.12 | 2298.15 |
| SeaLLM-7B-v2.5 | 0.17 | 0.14 | 0.14 | 0.15 | 0.13 | 0.11 | 2184.55 |
| Qwen2.5-3B-Instruct | 0.16 | 0.14 | 0.10 | 0.12 | 0.12 | 0.13 | 2324.08 |
| Sailor-14B-Chat | 0.16 | 0.07 | 0.11 | 0.13 | 0.10 | 0.09 | 2465.85 |
| SeaLLM-7B-v1 | 0.12 | 0.03 | 0.07 | 0.09 | 0.07 | 0.06 | 2585.40 |
| Mistral-7B-Instruct-v0.3 | 0.10 | 0.11 | 0.03 | 0.07 | 0.06 | 0.07 | 2336.51 |
| Sailor-7B-Chat | 0.09 | 0.02 | 0.04 | 0.06 | 0.04 | 0.03 | 1404.60 |
| Llama-2-70B-Chat | 0.08 | 0.07 | 0.05 | 0.06 | 0.05 | 0.05 | 2354.30 |
| Llama-2-13B-Chat | 0.06 | 0.04 | 0.04 | 0.05 | 0.03 | 0.03 | 2317.36 |
| Llama-2-7B-Chat | 0.05 | 0.03 | 0.02 | 0.04 | 0.02 | 0.03 | 2330.50 |

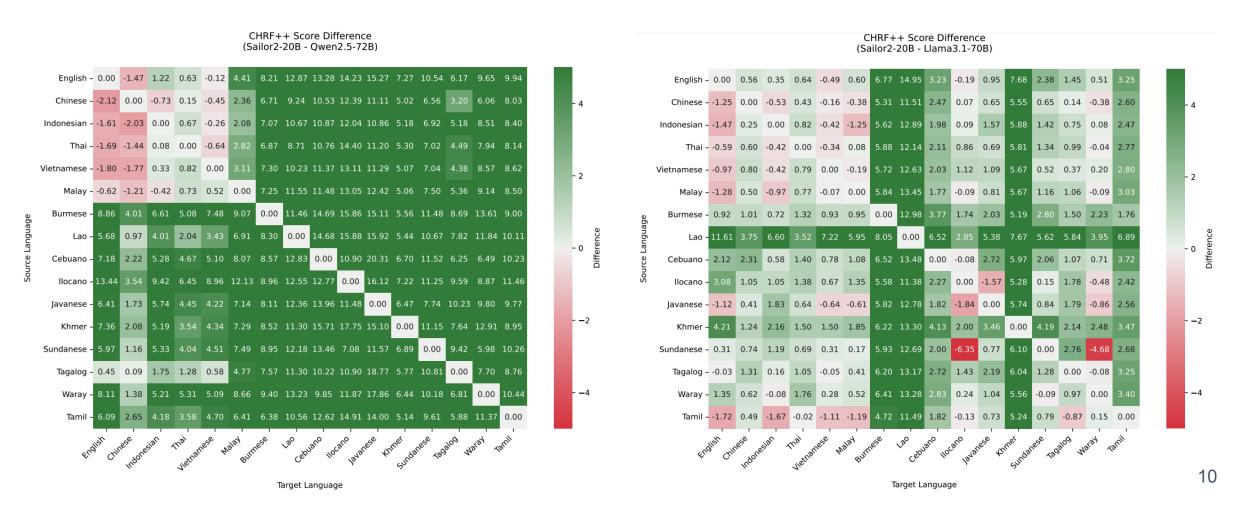
| Model | SWB Score | tha | vie | ind | tgl | zsm | khm | lao | mya | Length |
|--------------------------|-----------|------|------|------|------|------|------|------|------|---------|
| Sailor2-20B-Chat | 0.56 | 0.53 | 0.50 | 0.54 | 0.50 | 0.49 | 0.63 | 0.69 | 0.64 | 2814.74 |
| Sailor2-8B-Chat | 0.49 | 0.48 | 0.46 | 0.46 | 0.42 | 0.43 | 0.50 | 0.66 | 0.55 | 2849.41 |
| Qwen2.5-72B-Instruct | 0.45 | 0.54 | 0.51 | 0.51 | 0.42 | 0.48 | 0.33 | 0.41 | 0.31 | 3026.82 |
| SEA-LIONv3-70B-Instruct | 0.40 | 0.45 | 0.45 | 0.48 | 0.40 | 0.41 | 0.32 | 0.28 | 0.32 | 2340.65 |
| Gemma-2-27B-Instruct | 0.40 | 0.43 | 0.40 | 0.46 | 0.40 | 0.39 | 0.34 | 0.38 | 0.31 | 2288.33 |
| Qwen2.5-32B-Instruct | 0.32 | 0.37 | 0.42 | 0.42 | 0.26 | 0.38 | 0.24 | 0.19 | 0.16 | 2090.61 |
| Gemma-2-9B-Instruct | 0.31 | 0.36 | 0.40 | 0.39 | 0.30 | 0.38 | 0.19 | 0.19 | 0.19 | 2163.03 |
| Qwen2.5-14B-Instruct | 0.30 | 0.40 | 0.40 | 0.23 | 0.35 | 0.20 | 0.21 | 0.12 | 0.30 | 2267.94 |
| Llama-3.1-70B-Instruct | 0.30 | 0.33 | 0.37 | 0.37 | 0.28 | 0.35 | 0.18 | 0.15 | 0.19 | 2543.06 |
| SEA-LIONv3-8B-Instruct | 0.30 | 0.38 | 0.40 | 0.38 | 0.34 | 0.35 | 0.12 | 0.08 | 0.14 | 2357.14 |
| Aya-expanse-32B | 0.29 | 0.25 | 0.45 | 0.46 | 0.27 | 0.35 | 0.06 | 0.12 | 0.13 | 2495.47 |
| Qwen2-72B-Instruct | 0.26 | 0.26 | 0.30 | 0.33 | 0.29 | 0.32 | 0.20 | 0.20 | 0.16 | 1546.21 |
| Qwen2.5-7B-Instruct | 0.25 | 0.30 | 0.35 | 0.36 | 0.12 | 0.29 | 0.09 | 0.09 | 0.08 | 2415.08 |
| Sealionv2.1-8B-Instruct | 0.23 | 0.30 | 0.33 | 0.31 | 0.16 | 0.28 | 0.07 | 0.08 | 0.10 | 1735.26 |
| SeaLLMs-v3-7B-Chat | 0.21 | 0.23 | 0.22 | 0.21 | 0.19 | 0.16 | 0.15 | 0.16 | 0.09 | 2298.47 |
| Llama-3.1-8B-Instruct | 0.19 | 0.19 | 0.26 | 0.21 | 0.15 | 0.18 | 0.06 | 0.07 | 0.07 | 2356.67 |
| SeaLLM-7B-v2 | 0.18 | 0.18 | 0.18 | 0.19 | 0.09 | 0.13 | 0.10 | 0.12 | 0.09 | 2298.15 |
| SeaLLM-7B-v2.5 | 0.17 | 0.18 | 0.19 | 0.18 | 0.10 | 0.14 | 0.08 | 0.11 | 0.06 | 2184.55 |
| Qwen2.5-3B-Instruct | 0.16 | 0.14 | 0.21 | 0.18 | 0.08 | 0.16 | 0.06 | 0.06 | 0.04 | 2324.08 |
| Sailor-14B-Chat | 0.16 | 0.11 | 0.17 | 0.14 | 0.04 | 0.14 | 0.02 | 0.12 | 0.06 | 2465.85 |
| SeaLLM-7B-v1 | 0.12 | 0.05 | 0.07 | 0.07 | 0.04 | 0.05 | 0.10 | 0.11 | 0.09 | 2585.40 |
| Mistral-7B-Instruct-v0.3 | 0.10 | 0.07 | 0.11 | 0.07 | 0.08 | 0.11 | 0.02 | 0.03 | 0.02 | 2336.51 |
| Sailor-7B-Chat | 0.09 | 0.04 | 0.07 | 0.05 | 0.02 | 0.06 | 0.02 | 0.07 | 0.03 | 1404.60 |
| Llama-2-70B-Chat | 0.08 | 0.02 | 0.05 | 0.11 | 0.06 | 0.13 | 0.03 | 0.01 | 0.03 | 2354.30 |
| Llama-2-13B-Chat | 0.06 | 0.01 | 0.05 | 0.08 | 0.02 | 0.03 | 0.01 | 0.01 | 0.03 | 2317.36 |
| Llama-2-7B-Chat | 0.05 | 0.01 | 0.02 | 0.05 | 0.04 | 0.03 | 0.01 | 0.02 | 0.04 | 2330.50 |

Task-Level Evaluation

Language-Level Evaluation

Sailor2: Strong Translation Performance

Compare with models 3x larger (Qwen2.5-72B and Llama3.1-70B), Sailor2 performs significantly better in SEA languages while remaining comparable in high-resource languages.



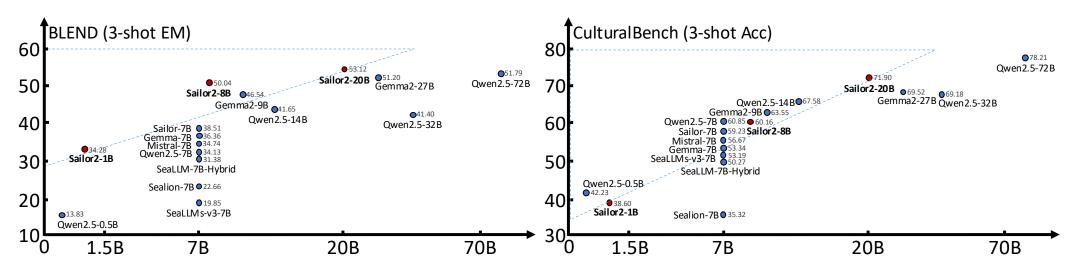
Sailor2: Long Context Training

Sailor2 supports 32K-128K context input, enabling more use cases such as DocumentQA.

| Model | 128K | 64K | 32K | 16K | 8K | 4K |
|------------------|-------|-------|-------|-------|-------|-------|
| Qwen2.5-0.5B | 0.00 | 0.00 | 46.50 | 52.65 | 55.95 | 64.42 |
| Sailor2-1B | 0.00 | 0.00 | 0.62 | 3.99 | 35.81 | 55.93 |
| Sailor2-1B-32K | 0.00 | 0.00 | 36.52 | 49.63 | 55.50 | 56.84 |
| Qwen2.5-7B | 20.67 | 61.70 | 78.58 | 81.72 | 83.58 | 86.72 |
| Sailor2-8B | 0.00 | 2.17 | 9.59 | 23.08 | 49.13 | 69.38 |
| Sailor2-8B-128K | 19.94 | 41.57 | 54.61 | 64.32 | 75.73 | 80.04 |
| Qwen2.5-14B | 32.93 | 66.68 | 85.09 | 86.96 | 87.40 | 87.56 |
| Sailor2-20B | 0.55 | 14.08 | 46.60 | 67.76 | 79.62 | 87.86 |
| Sailor2-20B-128K | 47.46 | 66.70 | 79.52 | 85.24 | 86.63 | 88.21 |

Sailor2: Better Culture Understanding

Among models of similar size, Sailor2 understands SEA culture better, including its food, traditions and geography, making it ideal for locally-relevant chat applications.



Paano madalas na bumabati ang mga tao sa Pilipinas nang hindi gumagamit ng mga salita?
(English Question: How do people in the Philippines often greet each other without using words?)

A. Sa pagtaas ng kilay nila
B. Sa pamamagitan ng pagbibigay ng mahigpit na pagkakamay
C. Sa pagtango ng kanilang ulo
D. Sa pamamagitan ng pagbibigay ng high five.

Prediction:

Golden: C Qwen2.5-32B: B Sailor2-20B: C

Masakan Indonesia populer apa yang menggunakan daging yang ditusuk dan dipanggang, disajikan dengan saus kental, pedas, dan berbahan dasar kunyit?
(English Question: What popular Indonesian dish uses skewered and grilled meat, served with a thick, spicy, and turmeric-based sauce?)

A. Sate Padang.
B. Sate Madura.
C. Steak Wagyu.
D. Iga BBQ.

Prediction:

Golden: A Qwen2.5-72B: B Sailor2-20B: A

Sailor2: Accelerate the Multilingual LLM Research with Open Cookbook

Data Curation

Fineweb-Pro Chinese-Fineweb-Edu Open-Web-Math-Pro

Dataset for Replay

Wikipedia
Open Subtitles
Translation
CommonCrawl
Public PDF

Dataset for SEA Languages

SEA-UltraFeedback SEA-UltraChat

Dataset for Post-Training

Data Recall for Low-resource Language

Data Mixture with Tiny Model Simulation

Continual Pre-Training

Model Expansion

Two-Stage Continual Training

> Model Parallel Optimization

Post-Training

Data Selection for Instruction Tuning

Two-Stage Instruction Tuning

Two-Stage Preference Tuning

Model Customization

Long-Context Training

Speculative Decoding

Model Pruning

(Translation)

CultureBench (Culture)

Evaluation

SailCompass

(Generation, Classification)

FLoRes

BLEnD (Culture)

Global MMLU (Culture, Knowledge)

RULER (Long-Context)

SEA-WildBench (Chat)

🗱 Key Insights in Building Sailor2 🌵

Data

- Pre-training Data: Rigorous multi-level deduplication (document, sentence, URL) to mitigate the redundancy of Common Crawl sources.
- Post-training Data: Popular translation models (e.g., GPT4o/NLLB) often underperform in low-resource languages and require careful refinement.

Training

- Model Expansion: Improve model capacity to better absorb multilingual knowledge, especially in over-trained models.
- Two-stage Training: Ensure steady learning from data of varying quality.

Evaluation

- Comprehensive Metrics: Assess basic language understanding, cultural context awareness, and conversational proficiency.
- **Probe Data:** Use the selected probe data to accurately and promptly check the model's capabilities beyond standard quantitative metrics.

Sailor2: Open Model for Better Developments

All models, resources, and code associated with Sailor2 are released under the Apache 2.0 License, which allows commercial usage.

| Model Checkpoints | | | | | | | |
|-------------------|----------------------|------------------------------------|-----------------------|--|--|--|--|
| Stage | Sailor2-1B | Sailor2-8B | Sailor2-20B | | | | |
| Pre-Annealing | sail/Sailor2-1B-Pre | sail/Sailor2-8B-Pre | sail/Sailor2-20B-Pre | | | | |
| Base | sail/Sailor2-1B | sail/Sailor2-8B | sail/Sailor2-20B | | | | |
| SFT | sail/Sailor2-1B-SFT | sail/Sailor2-8B-SFT | sail/Sailor2-20B-SFT | | | | |
| Chat | sail/Sailor2-1B-Chat | sail/Sailor2-8B-Chat | sail/Sailor2-20B-Chat | | | | |
| Codebases / Tools | | | | | | | |
| Type | 🗘 Link | | | | | | |
| Data Cleaning | sail-sg/ | /sailcraft | | | | | |
| Data Mixture | sail-sg/ | /regmix | | | | | |
| Pre-training | sail-sg/ | /Megatron-Sailor2 | | | | | |
| Post-training | sail-sg/ | oat/ | | | | | |
| Evaluation | sail-sg/ | 'sailcompass | | | | | |
| | Post- | Training Dataset | | | | | |
| Domain | 😕 Link | | | | | | |
| SFT-Stage1 | sailor2/ | /sailor2-sft-stage1 | | | | | |
| SFT-Stage2 | sailor2/ | /sailor2-sft-stage2 | | | | | |
| Off-policy DPO | sailor2/ | sailor2/sea-ultrafeedback | | | | | |
| On-policy DPO | sailor2/ | sailor2/sea-ultrafeedback-onpolicy | | | | | |
| | | | | | | | |

| Evaluation Dataset | | | | | | | | | |
|--------------------|-----------------------------------------------|------------------|-----------------------------|-----------------------------|--|--|--|--|--|
| Domain | 😕 Link | | | | | | | | |
| SailCompass | s sail/Sailcompass_data | | | | | | | | |
| SEA-WildBe | nch sailo | r2/sea-wildbench | 1 | | | | | | |
| | Model Checkpoints (via Long-Context Training) | | | | | | | | |
| Stage | Sailor2-1B | Sailor2-8B | | Sailor2-20B | | | | | |
| Base | sail/Sailor2-L-1B | sail/Sailor2- | L-8B | sail/Sailor2-L-20B | | | | | |
| SFT | sail/Sailor2-L-1B-SFT | sail/Sailor2- | L-8B-SFT | sail/Sailor2-L-20B-SFT | | | | | |
| Chat | sail/Sailor2-L-1B-Chat | t sail/Sailor2- | L-8B-Chat | sail/Sailor2-L-20B-Chat | | | | | |
| | Model Checkpo | oints (via Specu | lative Dec | oding) | | | | | |
| Stage | Sailor2-8B | | Sailor2-20B | | | | | | |
| Base Model | sail/Sailor2-8B-Ch | at-Glide | sail/Sailor2-20B-Chat-Glide | | | | | | |
| | Model Checkpoints (via Model Pruning) | | | | | | | | |
| Stage | Sailor2-3B (Pruning | via Sailor2-8B) | Sailor2-14 | B (Pruning via Sailor2-20B) | | | | | |
| Base Model | sail/Sailor2-3B | | sail/Sail | or2-14B | | | | | |
| SFT | sail/Sailor2-3B-SFT | | sail/Sailor2-14B-SFT | | | | | | |
| Chat | sail/Sailor2-3B-Chat | | sail/Sailor2-14B-Chat | | | | | | |
| | | | | | | | | | |

What's Next? More Low-Resource Languages.

- (1) Synthetic Data for Low-resource languages
 - Current Common-crawl only sources a small set of data in low-resource languages.
 - TODO: Global data mining and translate high-quality corpus from English.
- (2) Tokenizer-Free Model for Open-Vocabulary Learning
 - Current Tokenizer is unfriendly to morphological-rich languages like Thai and Khmer.
 - TODO: Pixel-based LLM or Byte-level LLM.
- (3) Efficient Continual Pre-training for Multilingual model
 - Existing LLMs are overfitting to certain languages. Continual training is tricky and costly.
 - TODO: Improve the model plasticity and partially update the model parameters.

Thank You

