# **Mastering LLMs**

A Conference For Developers & Data Scientists

### **Growing Into Something Bigger**

#### With world-class guest instructors:

- Jeremy Howard: Co-Founder Answer.Al & Fast.Al
- Sophia Yang: Head of Developer Relations, Mistral Al
- Wing Lian: Creator of Axolotl library for LLM fine-tuning
- Simon Willison: Creator of Datasette
- Mark Saroufim: PyTorch developer @ Meta
- Paige Bailey: DevRel Lead, GenAl, Google
- Shreya Shankar: LLMOps and LLM Evaluations researcher
- Zach Mueller: Lead maintainer of HuggingFace accelerate
- Bryan Bischof: Director of Al Engineering at Hex
- Jason Liu: Creator of Instructor
- Abhishek Thakur leads AutoTrain at HuggingFace.
- Johno Whitaker: R&D at AnswerAl
- Charles Frye: Al Engineer at Modal Labs
- Eugene Yan: Senior Applied Scientist @ Amazon
- Harrison Chase: CEO of LangChain
- Travis Addair: Co-Founder & CTO of Predibase
- John Berryman: Author of O'Reilly Book Prompt Engineering for LLMs
- Joe Hoover: Lead ML Engineer at Replicate
- Ben Clavié: R&D at AnswerAl



#### How To Be Successful With This Conference

- Tinker with the tools we show you -> Office Hours
- The importance of blogging. <u>Tips</u>.
- How to share your work (blogs, projects, etc)
  - Axolotl @winglian, @axolotl\_ai, @hamelhusain
  - Deepspeed/FSDP/Accelerate @TheZachMueller
  - Modal @charles\_irl

## Plan For Today

- What is Axolotl and how to use it to fine-tune model
  - Honeycomb example
  - Convo with Wing Lian
- Parallelism & HF Accelerate w/ Zach Mueller
- Fine-tuning on Modal
- Q&A

# **Modeling Choices**

Base model

LoRA vs Full Fine Tune

### Choosing a Base Model

#### Model Size



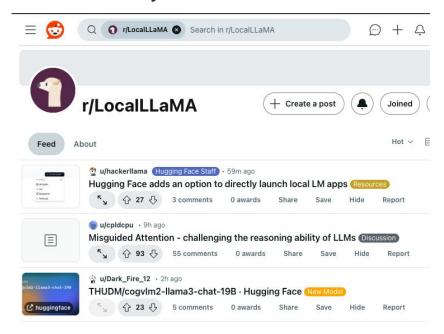
Note The base 7B model in HF transformers format

```
    meta-llama/Llama-2-13b-hf
    Text Generation - Updated Apr 17 - ≥ 319k - ♥ 542
```

Note The base 13B model in HF transformers format

Note The base 70B model in HF transformers format

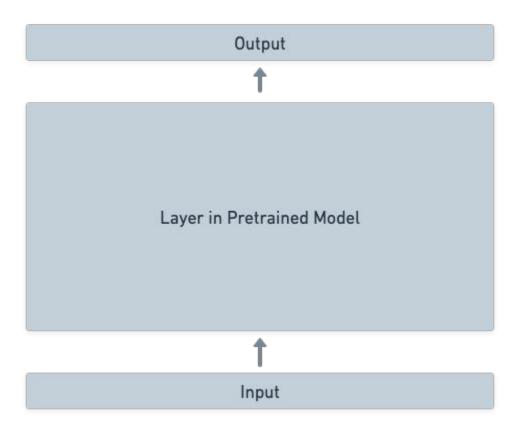
#### **Model Family**

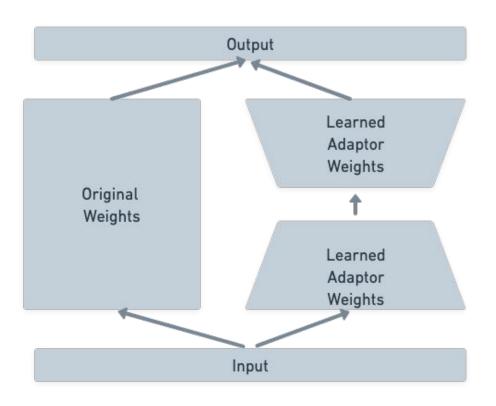


# **Modeling Choices**

Base model

LoRA vs Full Fine Tune

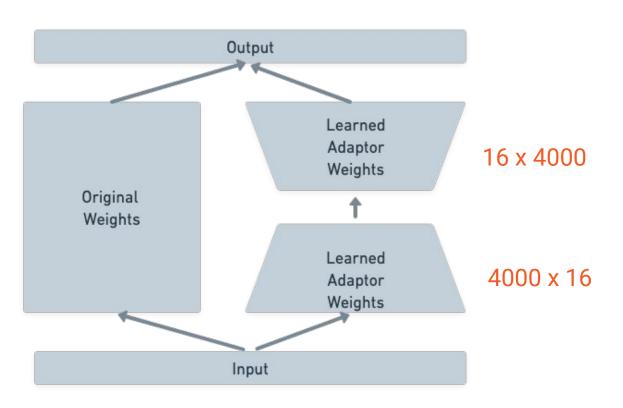


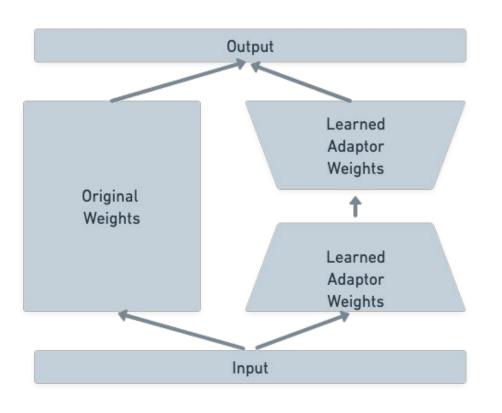


Input: 4000 dimensions

Output: 4000 dimensions

Original weights: 16M





Input: 4000 dimensions

Output: 4000 dimensions

Original weights: 16M

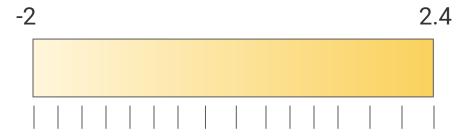
Adaptor rank: 16

LoRA weights: 2 \* 16 \* 4000

= 128,000

### QLoRA

- LoRA at lower precision
- Memory savings with possible loss in quality





WITH WITH HYPERPARAMETERS

### What Is Axolotl

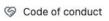
Wrapper for Hugging Face tools

Easy to use. So you can focus on your data

Best practices built-in

# **Using Axolotl**

















#### **Table of Contents**

- Introduction
- Supported Features
- Quickstart
  - Environment
    - Docker
  - Conda/Pip venv
  - o Cloud GPU Latitude.sh, JarvisLabs, RunPod
  - Bare Metal Cloud GPU
  - Windows
  - Mac
  - Google Colab
  - · Launching on public clouds via SkyPilot
  - Launching on public clouds via dstack
- Dataset
- Config
  - Train
  - Inference
  - Merge LORA to Base
  - Special Tokens

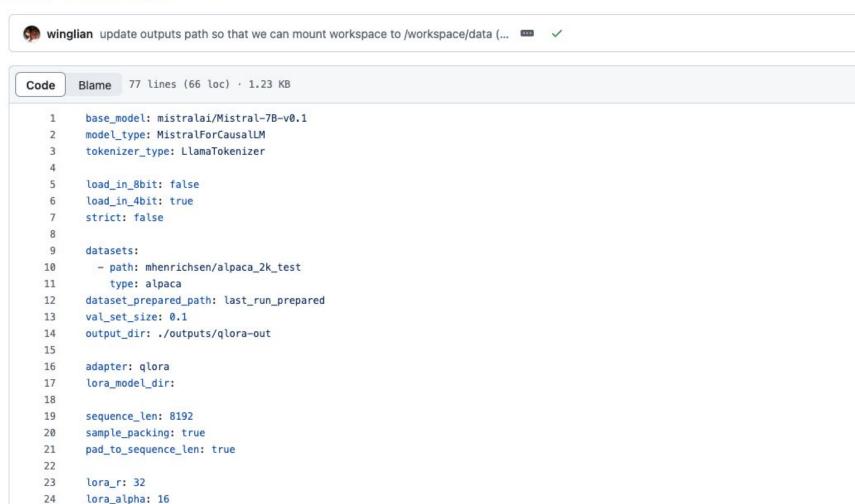


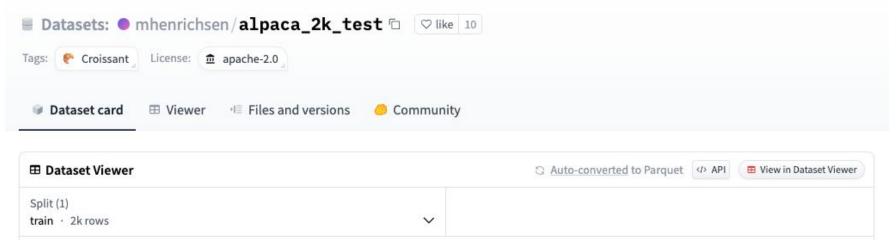
Axolotl provides a unified repository for finetuning a variety of AI models with ease

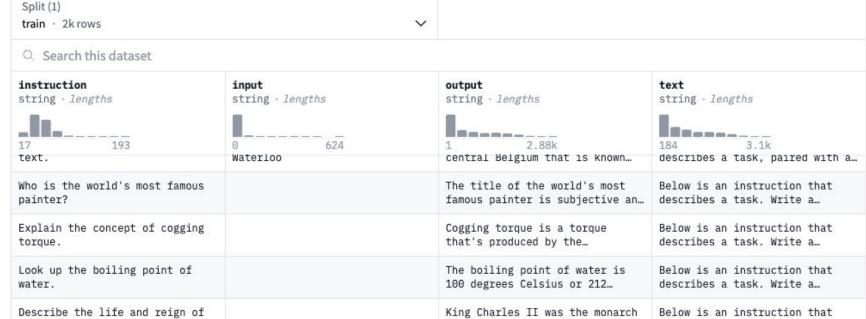
Go ahead and Axolotl questions!!

jquesnelle add save_only_model option (#1634)	✓ 702a669 · 2 days ago	(1,448 Commits
github	cloud image w/o tmux (#1628)	3 days ago
.vscode	feat: enable trl's autounwrap (#1060)	4 months ago
cicd	make sure to install causal_conv1d in docker (#1459)	2 months ago
deepspeed_configs	DBRX Model Support (#1462)	last monti
devtools	Bootstrap Hosted Axolotl Docs w/Quarto (#1429)	2 months ago
docker	fix ray install (#1630)	3 days ago
docs	docs(config.qmd): add loraplus example (#1577)	2 weeks ag
examples	update outputs path so that we can mount workspace to /	3 days ago
image	badge (#739)	7 months ago
scripts	more fixes to work with runpod + skypilot (#1629)	3 days ago

#### axolotl / examples / mistral / qlora.yml 📮







of England, Scotland, and...

describes a task. Write a...

King Charles II.

```
wandb_entity:
wandb_watch:
wandb_name:
wandb_log_model:
gradient_accumulation_steps: 4
micro_batch_size: 2
num_epochs: 1
optimizer: adamw_bnb_8bit
lr scheduler: cosine
learning_rate: 0.0002
train_on_inputs: false
group_by_length: false
bf16: auto
fp16:
tf32: false
gradient_checkpointing: true
early_stopping_patience:
resume_from_checkpoint:
local rank:
logging_steps: 1
xformers_attention:
flash_attention: true
```

wandb\_project:

Code of conduct

Apache-2.0 license

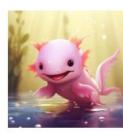






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    - Conda/Pip venv
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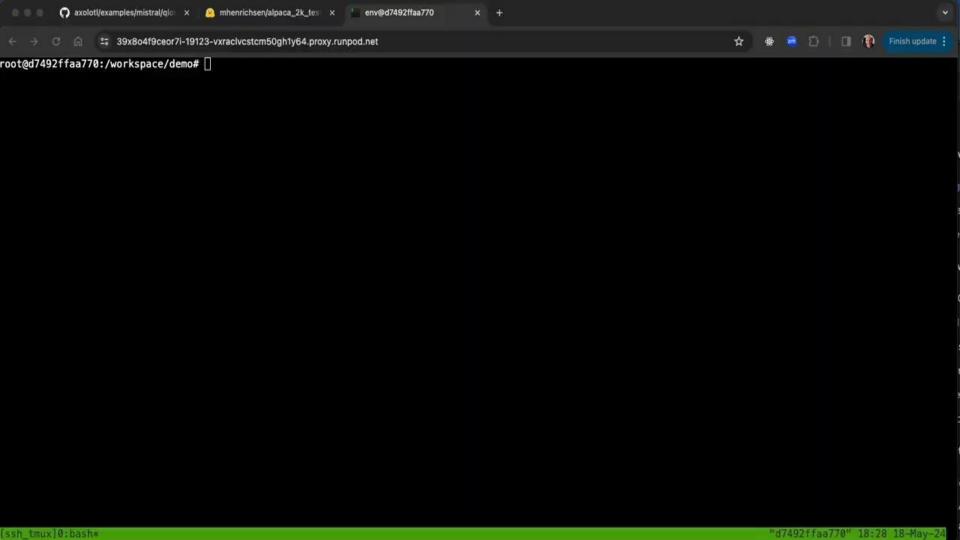
Axolotl provides a unified repository for finetuning a variety of AI models with ease

Go ahead and Axolotl questions!!

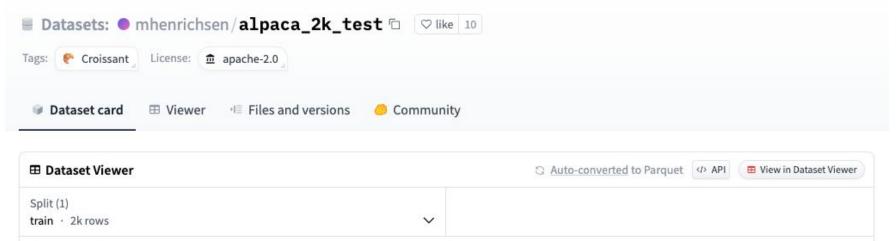
```
# preprocess datasets - optional but recommended
CUDA_VISIBLE_DEVICES="" python -m axolotl.cli.preprocess examples/openllama-3b/lora.yml
# finetune lora
accelerate launch -m axolotl.cli.train examples/openllama-3b/lora.yml
# inference
accelerate launch -m axolotl.cli.inference examples/openllama-3b/lora.yml \
    --lora_model_dir="./lora-out"
# gradio
```

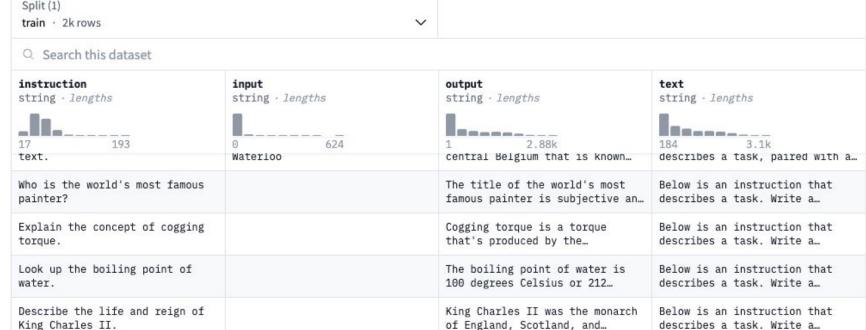
accelerate launch -m axolotl.cli.inference examples/openllama-3b/lora.yml \

--lora\_model\_dir="./lora-out" --gradio



```
<start>
Below is an instruction that describes a task, paired with an input that provi
des further context. Write a response that appropriately completes the request
### Instruction:
{instruction}
### Input:
{input}
### Response:
{output}
<end>
```





```
<start>
Below is an instruction that describes a task, paired with an input that provi
des further context. Write a response that appropriately completes the request
### Instruction:
{instruction}
### Input:
{input}
### Response:
{output}
<end>
```

```
00, 2691) es(-100, 274) the(-100, 272) request(-100, 2159) \cdot (-100, 28723) < 0 \times 0
A>(-100, 13) <0x0A>(-100, 13) ###(-100, 27332) Inst(-100, 3133) ruction(-100,
3112) :(-100, 28747) <0x0A>(-100, 13) rew(-100, 2516) rite(-100, 1967) this(-1
00, 456) sentence(-100, 12271) as(-100, 390) a(-100, 264) question(-100, 2996)
<0\times0A>(-100, 13) <0\times0A>(-100, 13) ###(-100, 27332) Input(-100, 11232) :(-100,
28747) <0\times0A>(-100, 13) My(-100, 5183) mom(-100, 1948) made(-100, 1269) me(-1)
00, 528) a(-100, 264) delicious(-100, 15992) dinner(-100, 7854) .(-100, 28723)
<0\times0A>(-100, 13) <0\times0A>(-100, 13) ###(-100, 27332) Response(-100, 12107) :(-1
00, 28747) <0x0A>(-100, 13) Did(7164, 7164) your(574, 574) mom(1948, 1948) mak
e(1038, 1038) you(368, 368) a(264, 264) delicious(15992, 15992) dinner(7854, 7
854) ?(28804, 28804) </s>(2, 2)
```

(-100, 2899) that(-100, 369) appropri(-100, 6582) ately(-100, 1999) complet(-1

# Case Study

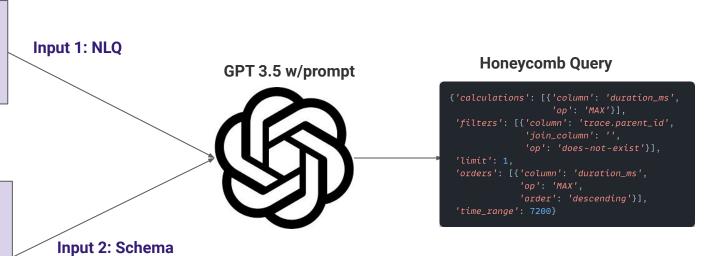
## Honeycomb - NL to Query

User: "latency distribution by status code"

#### Candidate Fields:

['trace.trace\_id', 'trace.span\_id', 'trace.parent\_id', 'duration\_ms', 'name', 'faas.instance', 'faas.id', 'filter',

'telemetry.instrumentation\_library',
'library.name', 'faas.name' ...



# Honeycomb Case Study

https://github.com/parlance-labs/ftcourse

## **Debugging Axolotl**

How-To Guides > Debugging

#### **Debugging**

How to debug Axolotl

This document provides some tips and tricks for debugging Axolotl. It also provides an example configuration for debugging with VSCode. A good debugging setup is essential to understanding how Axolotl code works behind the scenes.

#### **Table of Contents**

- General Tips
- Debugging with VSCode
  - Background
  - Configuration
  - Customizing your debugger
  - Video Tutorial
- Debugging With Docker
  - Setup
  - Attach To Container
  - Video Attaching To Docker On Remote Host

https://openaccess-ai-collective.github.io/axolotl/docs/debugging.html

## **Debugging Axolotl**

#### **General Tips**

While debugging it's helpful to simplify your test scenario as much as possible. Here are some tips for doing so:

[!Important] All of these tips are incorporated into the example configuration for debugging with VSCode below.

- 1. Make sure you are using the latest version of axolotl: This project changes often and bugs get fixed fast. Check your git branch and make sure you have pulled the latest changes from main.
- 2. Eliminate concurrency: Restrict the number of processes to 1 for both training and data preprocessing:
  - Set CUDA\_VISIBLE\_DEVICES to a single GPU, ex: export CUDA\_VISIBLE\_DEVICES=0.
  - Set dataset\_processes: 1 in your axolotl config or run the training command with --dataset\_processes=1.
- 3. Use a small dataset: Construct or use a small dataset from HF Hub. When using a small dataset, you will often have to make sure sample\_packing: False and eval\_sample\_packing: False to avoid errors. If you are in a pinch and don't have time to construct a small dataset but want to use from the HF Hub, you can shard the data (this will still tokenize the entire dataset, but will only use a fraction of the data for training. For example, to shard the dataset into 20 pieces, add the following to your axolotl config): yaml dataset: ... shards: 20
- 4. Use a small model: A good example of a small model is TinyLlama/TinyLlama-1.1B-Chat-v1.0.
- 5. Minimize iteration time: Make sure the training loop finishes as fast as possible, with these settings.

```
micro_batch_size: 1max_steps: 1val_set_size: 0
```

- Clear Caches: Axolotl caches certain steps and so does the underlying HuggingFace trainer. You may want to clear some of these caches when debugging.
  - Data preprocessing: When debugging data preprocessing, which includes prompt template formation, you may
    want to delete the directory set in dataset\_prepared\_path: in your axolotl config. If you didn't set this value, the
    default is last run prepared.
  - HF Hub: If you are debugging data preprocessing, you should clear the relevant HF cache <u>HuggingFace cache</u>, by deleting the appropriate ~/.cache/huggingface/datasets/... folder(s).
  - The recommended approach is to redirect all outputs and caches to a temporary folder and delete selected subfolders before each run. This is demonstrated in the example configuration below.

https://openaccess-ai-colle ctive.github.io/axolotl/docs /debugging.html#general-ti ps

# Questions For Wing

# Zach Mueller Accelerate / FSDP

# Training On Modal

## Why Modal

- Feels local, but its remote ("code in production")
- Massively parallel
- Python native
- Docs: <a href="https://modal.com/">https://modal.com/</a>

#### Things I've built with modal

- Transcript Summarizer
- W&B Webhook

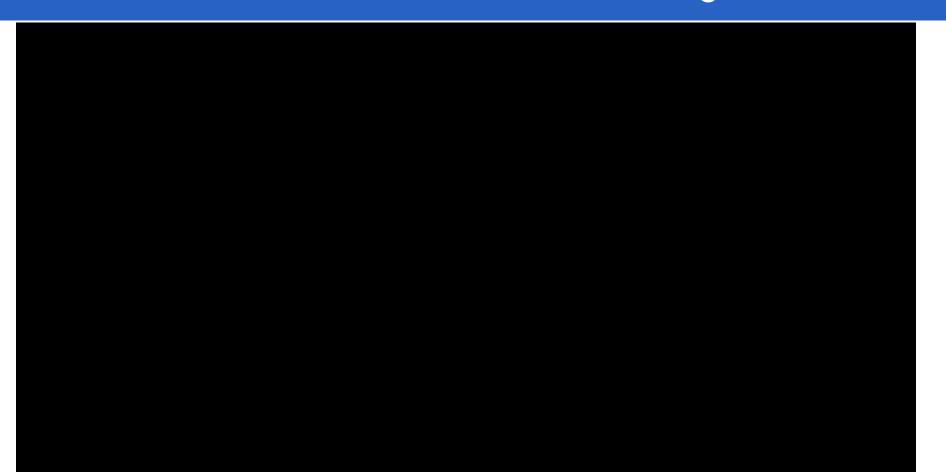
## Modal - Axolotl Fine Tuning

#### https://github.com/modal-labs/llm-finetuning

Has additional defaults / some differences

- Merges LoRA back into the base model
- Use a –data flag instead of relying on the config
- Deepspeed config comes from the axolot repo that is cloned

# Modal - Axolotl Fine Tuning



## Modal - Debug Data

https://github.com/modal-labs/llm-finetuning/blob/main/nbs/inspect\_data.ipynb

Tip: replace github.com with nbsanity.com to view notebooks

# Q & A