## **LLM Fine-Tuning**

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#### Why Adaptation?

#### **Base LLM**

Predicts next word, based on text training data

#### Once upon a time, there was a unicorn

that lived in a magical forest with all her unicorn friends

#### What is the capital of France?

What is France's largest city? What is France's population? What is the currency of France?

#### **Instruction Tuned LLM**

Tries to follow instructions

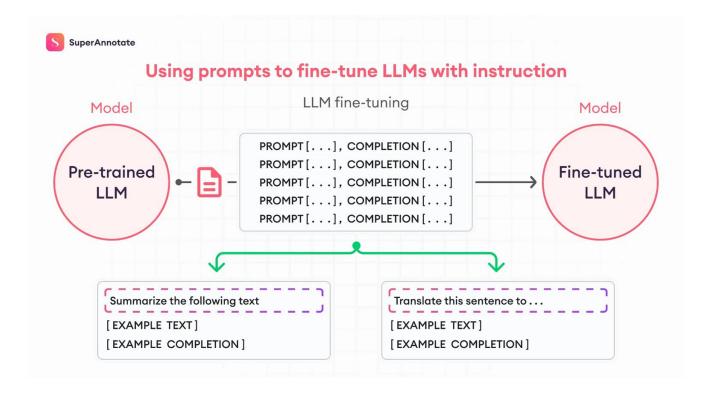
Fine-tune on instructions and good attempts at following those instructions.

RLHF: Reinforcement Learning with Human Feedback

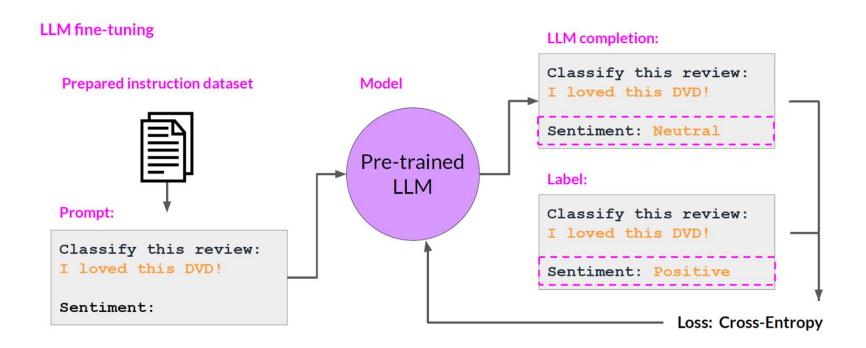
Helpful, Honest, Harmless

What is the capital of France? The capital of France is Paris.

#### LLM Fine-Tuning Process

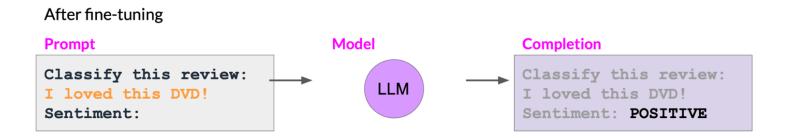


#### LLM Fine-Tuning Process



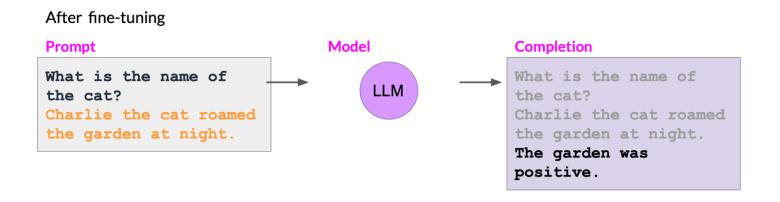
#### Catastrophic Forgetting

 Fine-tuning can significantly increase the performance of a model on a specific task...



#### Catastrophic Forgetting

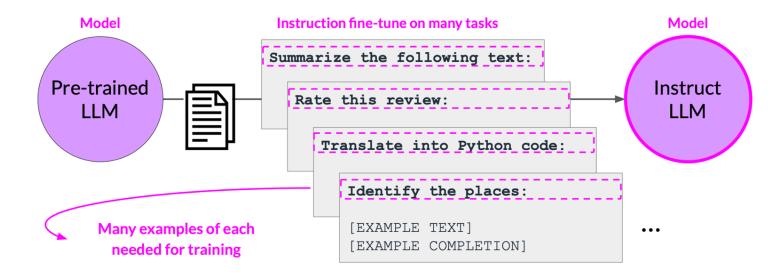
...but can lead to reduction in ability on other tasks



## How to Avoid Catastrophic Forgetting

- Fine-tune on multiple tasks at the same time
- Consider Parameter Efficient Fine-tuning (PEFT)

## Multi-task Fine-Tuning

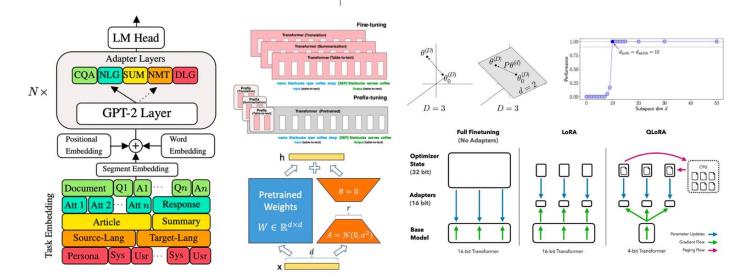


# Computational challenges still remain!

#### Parameter-Efficient Fine-Tuning (PEFT) Techniques



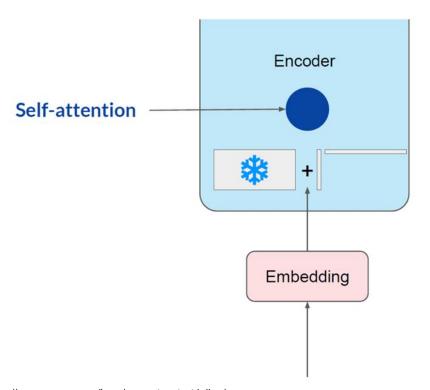
Easily Train a Specialized LLM: PEFT, LoRA, QLoRA, LLaMA-Adapter, and More



#### **PEFT**

## Finetuned Weights Weight Update $W_{ t ft} = W_{ t pt} + \Delta W$ Pretrained Weights

#### LoRA [9]



- 1. Freeze most of the original LLM weights.
- 2. Inject 2 rank decomposition matrices
- 3. Train the weights of the smaller matrices

Steps to update model for inference

1. Matrix multiply the low rank matrices

$$B * A = B \times A$$

2. Add to original weights



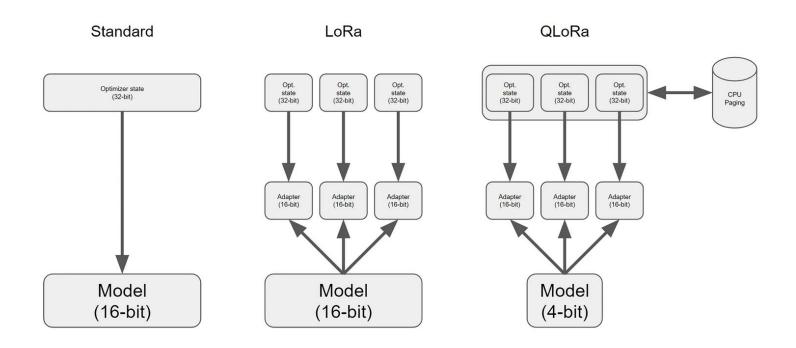
<sup>[3]</sup> https://www.coursera.org/learn/generative-ai-with-llms/

<sup>[5]</sup> Hu, E. J., Wallis, P., Allen-Zhu, Z., Li, Y., Wang, S., Wang, L., & Chen, W. LoRA: Low-Rank Adaptation of Large Language Models. In International Conference on Learning Representations.

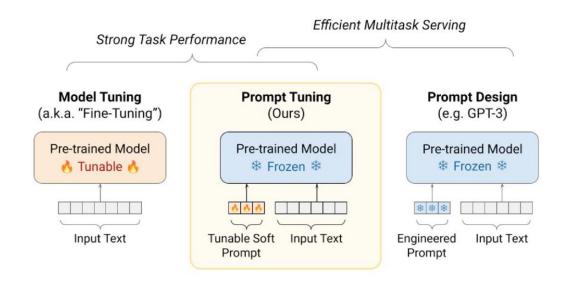
<sup>[6]</sup> https://www.youtube.com/watch?v=dA-NhCtrrVE

<sup>[7]</sup> https://www.youtube.com/watch?v=t509sv5MT0w

#### QLoRA [9]



#### Prompt-Tuning [9]



[10] Lester, B., Al-Rfou, R., & Constant, N. (2021, November). The Power of Scale for Parameter-Efficient Prompt Tuning. In Proceedings of the 2021 Conference on Empirical Methods in Natural Language Processing (pp. 3045-3059).

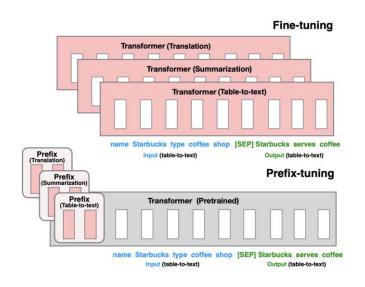
 $<sup>\</sup>hbox{[II]} \ \underline{https://cobusgreyling.medium.com/prompt-tuning-hard-prompts-soft-prompts-49740de6c64c}$ 

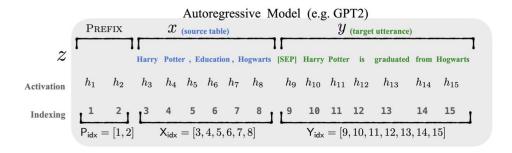
<sup>[12]</sup> Wang, Y., Chauhan, J., Wang, W., & Hsieh, C. J. (2024). Universality and limitations of prompt tuning. Advances in Neural Information Processing Systems, 36.'

<sup>[13]</sup> https://www.youtube.com/watch?v=HkZOGGvZzg4

<sup>[14]</sup> https://fnl.es/Science/Papers/Prompt+Engineering/Prompt+Tuning

#### Prefix-Tuning [14]





$$h_i = egin{cases} P_{ heta}[i,:], & ext{if } i \in \mathsf{P}_{\mathsf{idx}}, \ \mathsf{LM}_{\phi}(z_i, h_{< i}), & ext{otherwise}. \end{cases}$$



## Thank you!

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