



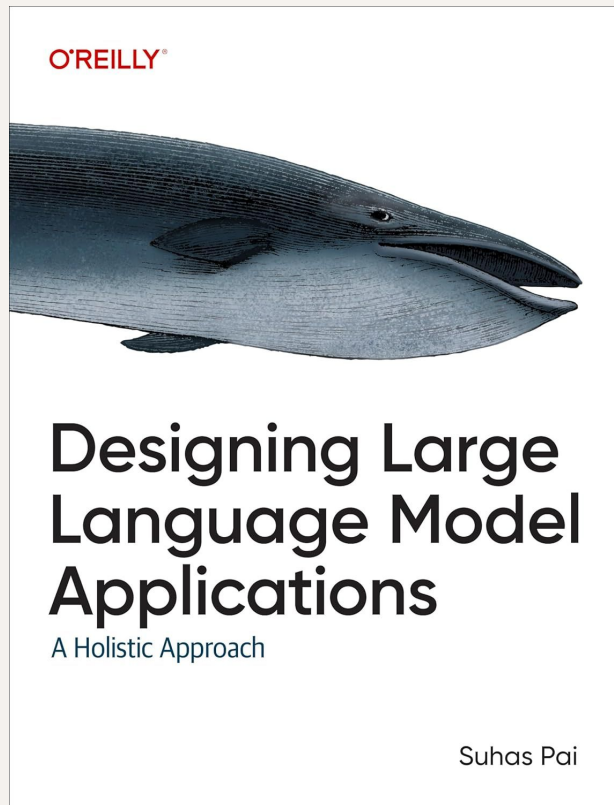
# Continual Learning in LLMs

- Suhas Pai, CTO, Hudson Labs



# About Me

- CTO & ML Research @ Hudson Labs (<https://hudson-labs.com>)
- Book 'Designing LLM Applications' published by O'Reilly Media
- Led/contributed to various open-source LLMs (BLOOM, Aurora-M etc.)
- Chair, TMLS (Toronto Machine Learning Society), MLOPS World conference
- Leading an independent research group 'llm-playbooks'



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**Why does catastrophic forgetting happen?**

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# Learning Rate strategies for pre-training LLMs

- Cosine Annealing with warmup
- Convergence to stable equilibrium

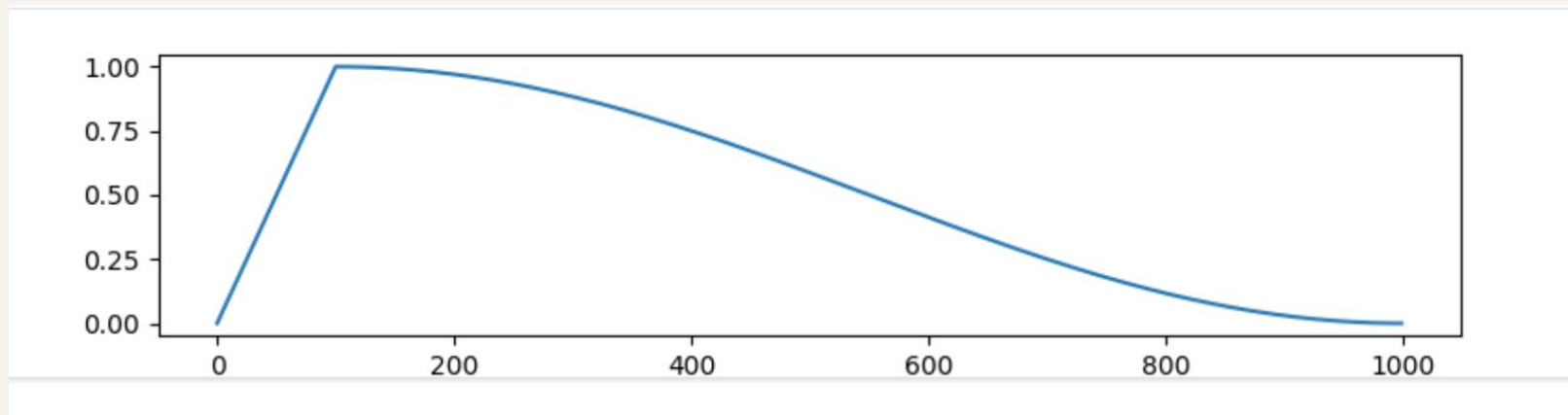


Image Credit: Hugging Face

# Cosine Schedule - Hugging Face

## Parameters

- **optimizer** (`~torch.optim.Optimizer`) — The optimizer for which to schedule the learning rate.
- **num\_warmup\_steps** (`int`) — The number of steps for the warmup phase.
- **num\_training\_steps** (`int`) — The total number of training steps.
- **num\_cycles** (`float`, *optional*, defaults to 0.5) — The number of waves in the cosine schedule (the defaults is to just decrease from the max value to 0 following a half-cosine).
- **last\_epoch** (`int`, *optional*, defaults to -1) — The index of the last epoch when resuming training.

# Problems with cosine annealing

- Rewarming from minimum values causes instability and forgetting
- Need to know number of training steps in advance
- WSD (Warmup-Stable-Decay) is gaining more prominence these days

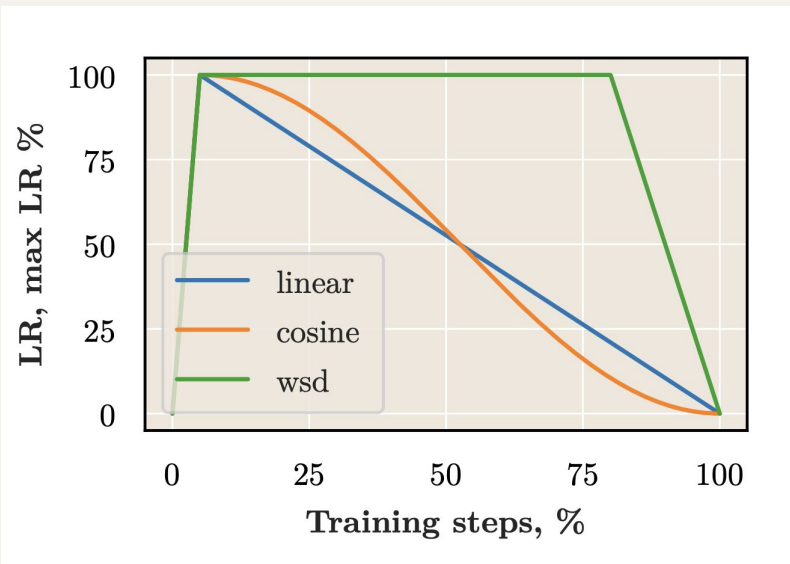
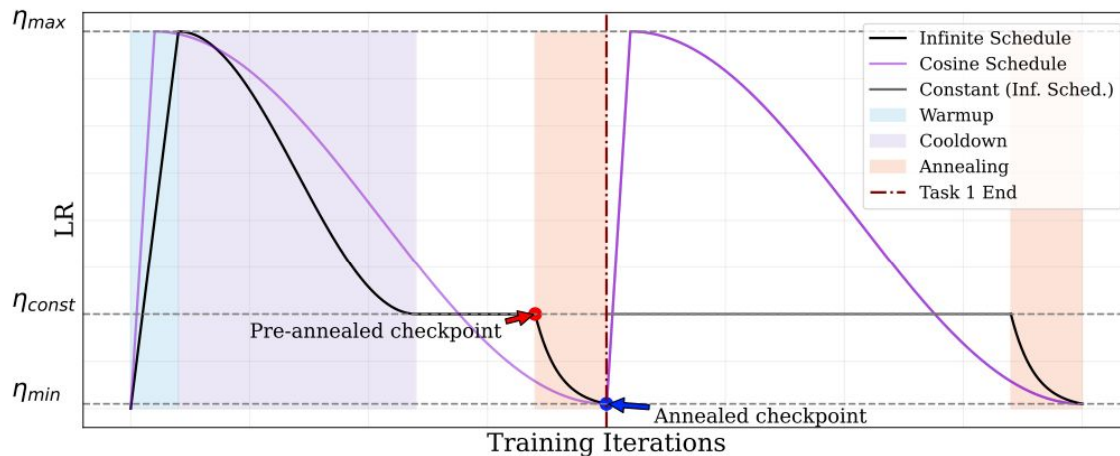


Image Credit: Dremov et al. (Aug 2025)

# Infinite Learning Schedule

- Initial warmup
- Decay
- Plateau
- Rapid annealing



Singh et al. (Sep 2025)



# Demo



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Should you continually learn over the base model or the instruction tuned model?

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# Catastrophic forgetting in instruct models

- Continued pre-training on instruction tuned models degrades instruction following capability
- Instruction tuning should be performed after continued pre-training
- Instruction residuals can be used to prevent repetitively performing instruction tuning after every round of continued pre-training

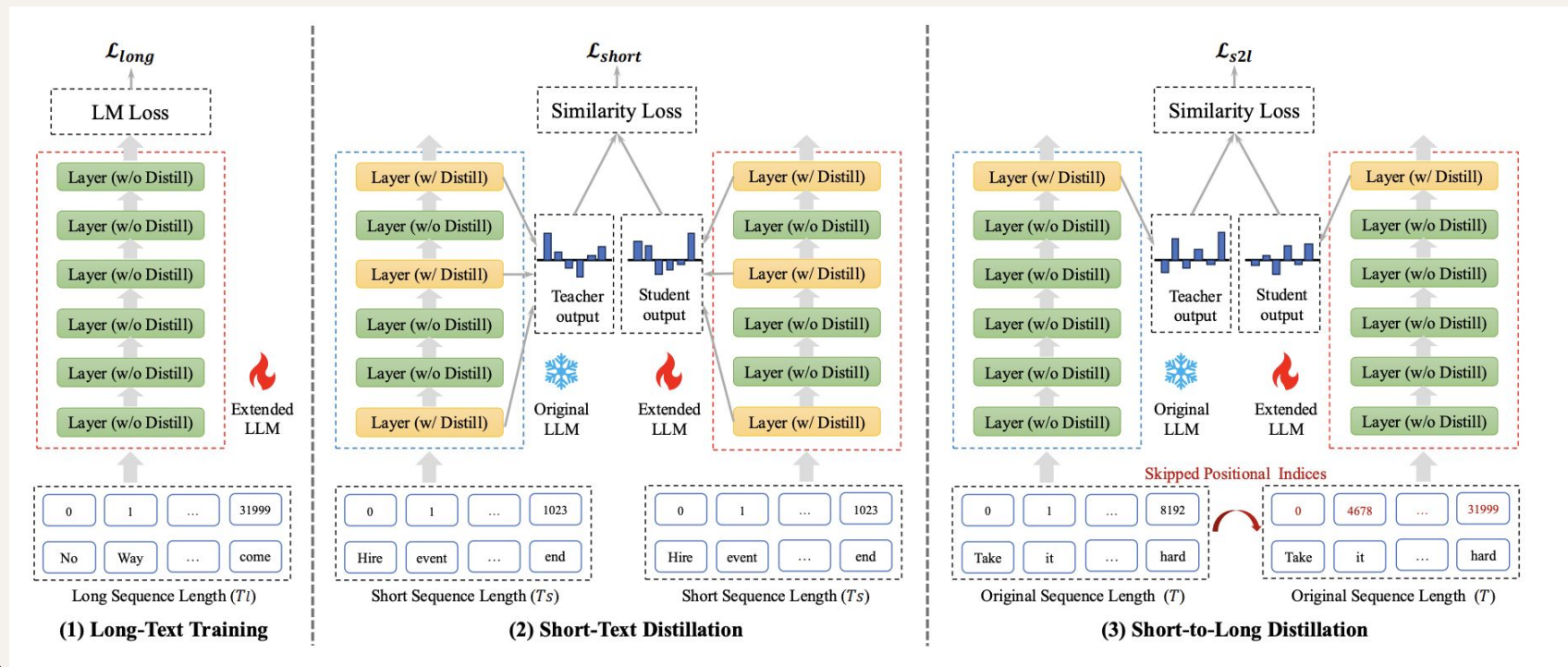
$$\Theta_r^{v1} = \theta_i^{d1v1} - \theta_b^{d1}.$$

Jindal et al. (Oct 2024)

# Catastrophic forgetting in long-context models

- Training on long-context data reduces performance on short-text data
- Rudimentary methods involve replaying short-context data during long-context training

# Long Context Pre-training with Restoration Distillation



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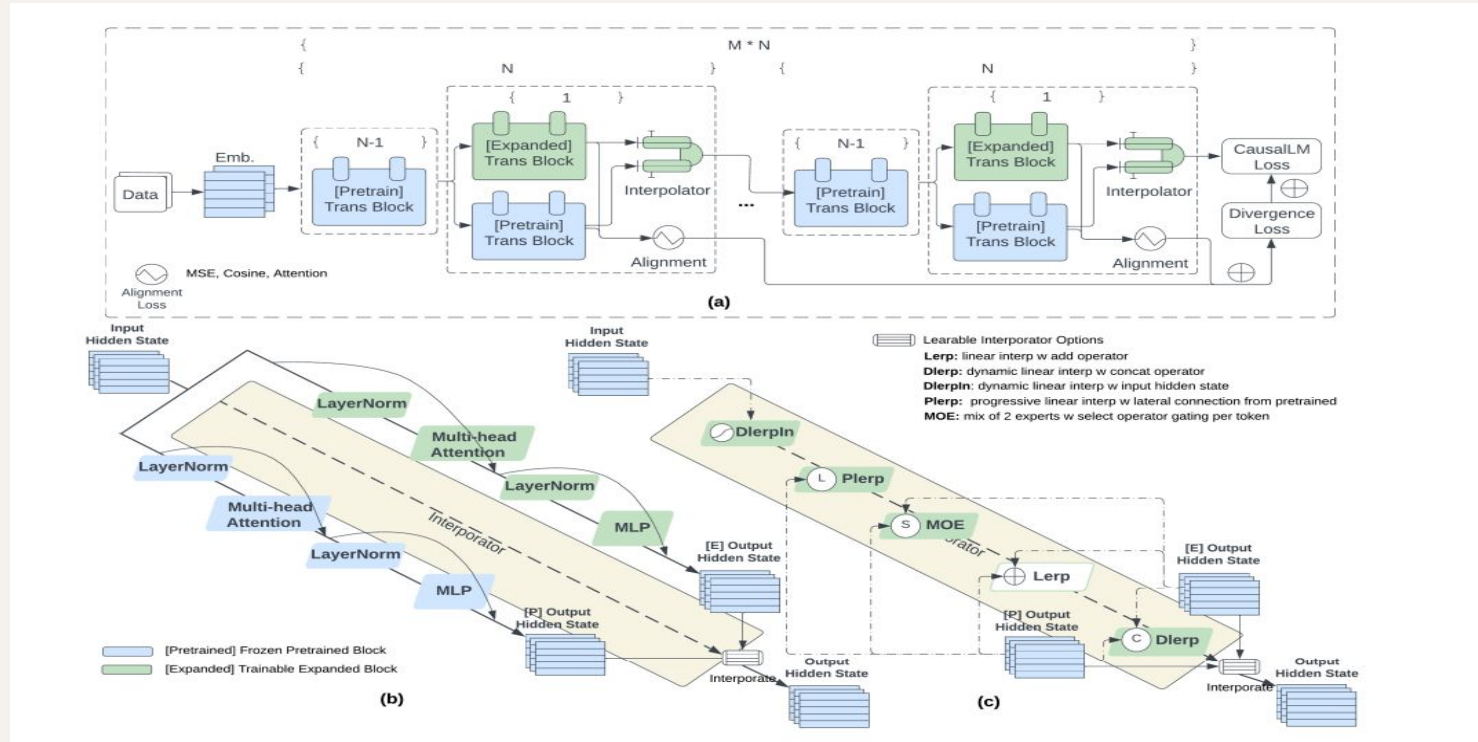
# Updating Parameters

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# Interpolating base and extended models

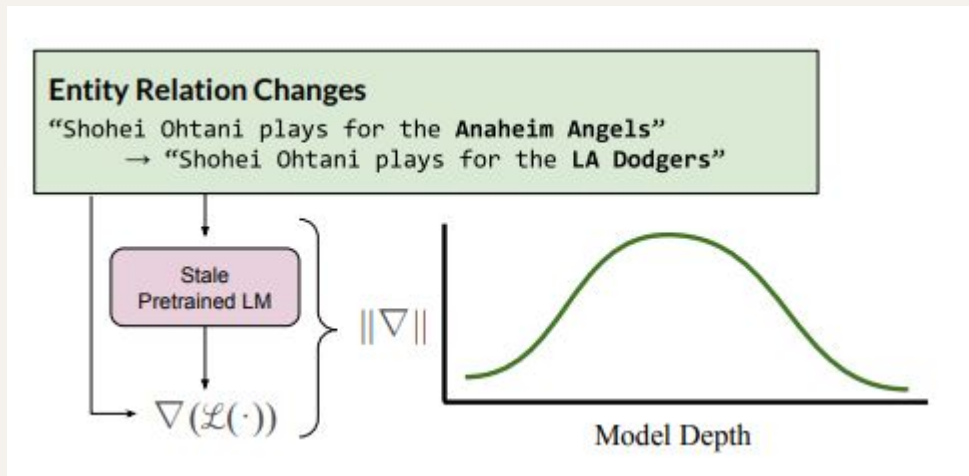
- Linear Interpolation
- Dynamic Linear Interpolation
- Dynamic Linear Interpolation on input hidden-states
- Progressive Linear Interpolation
- Mixture of Experts Gating

# Interpolating intermediate states



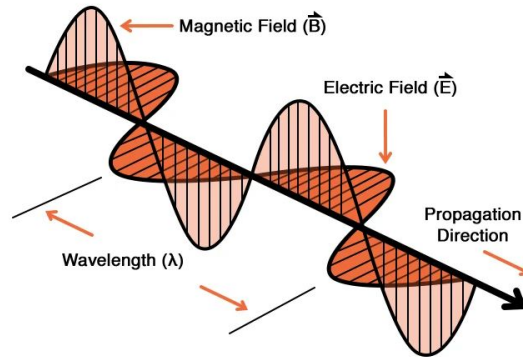
# Model Editing

- Large gradient norms observed in certain layers when predicting entity tokens





# Trivia Time: Connect the following



# Contact

- [piesauce.substack.com](https://piesauce.substack.com)
- <https://x.com/piesauce>
- <https://www.linkedin.com/in/piesauce/>