

(LLM notes)
(Also see 106)

RAG (Retrieval-Augmented Generation)

- is a method in LLMs to access external knowledge ^(Real time) rather than relying on only what they were trained on
- it does it by combining search (retrieval) with language gen
- why: after training cut off date LLMs don't know anything and can hallucinate facts RAG solves this by
 - retrieving relevant info from external knowledge base
 - generating answer using that retrieved info (w/ LLM)

via vector similarity from DB's

A RAG chain is those structured sequence of steps a RAG

pipeline uses Ex: input → Embedder → Retriever → Filter → LLM → Output

Fine tuning: involves using a pre trained model on specific dataset to adapt it to a particular task making it better for that task but limited to training knowledge, hence RAG

(See page 104) ⇒ Scaling Laws for LLM

- describe how performance improves as model size (params), data set and compute (FLOPs) change

- More compute → Bigger model + more data → Better performance (up to a point)

- loss (perplexity) decreases as log-linearly as model size ↑, dataset size ↑ and compute ↑ until diminishing returns or overfitting

- key points / laws • more params = Better capacity to learn complex patterns

• more data = prevents overfitting, improves generalization • more training steps = more updates

• more compute = enables longer training, more updates = more training (up to a point)

Balance (Optimal ratio between)

• model size • data size • training compute

- if model too big for dataset: overfit
- too much data for small model: underfit

Limits

- eventually you hit diminishing returns
- need better architecture (transformers, MoE)
- alignment & fine tuning (RLHF, RAG)

- This Limit of LLM is getting better rapidly then slowing down → is the "Bending curve" and Sol is mentioned above

