

## [Day 4] Unit 4: "Domain Specific LLMs"



Solving domain

- SyBIM: offers solution (ai assist.) at every layer

Specific Problem

- Cyber security pressures: constant emergence of attacks, quant work (operational tol.), talent shortage.

\* manual tasks can be automated

- Challenges: limited data to train on.

Creating Sec LMs

- targeted training approach, needs multi-lang cap, cyber sec. training; supervised fine tuning

- Evaluation: compare output to experts answers

\* human evals play crit. role.

- in-context learning: app built, train model how to use platform.

- RAGI: allow model to get current info

- Flexible & planning framework: planning & exec of complex tasks

- Goal: central platform to transform sys. security & sanit.

- LLMs can understand & apply med. concepts

- ways to be used.

- patients getting tailored responses.

- understanding & categorizing urgency.

- personalized patient experience.

- provide real-time feedback.

- Up-to-date knowledge.

- Emphasis on safe practices - abs. sur safe & effect.

Med PalM 2

- first ai to pass med. license exams.

- quantity vs. quality, can't be validated

- Multi-modal operations

ensemble

- trained: using med & AI data

- of course, chain of thought,

\* ensemble learning: learns from itself, reasoning & answer

[CodeLangs]

- fine tuning can be used from classic NLP to stylized gen.

- Available fine tuning model: Tuned Model, create.

- Download dataset → Preprocess data (pn-processing) + apply pnprocessing to training & data sets → Sample data → evaluate

Fine tuning a  
Custom Model

## Use a custom modul

- calling `train` specify model tuning hyperparam.
  - epoch-count: how many times to loop through data
  - batch-size: how many rows to process in a single step
  - learning-rate: defines the scaling factor for updating model weights @ each step.
- When search grounding = und, model returns entire metadate that includes links to search suggestions, supp. docs & info on how supp. doc was used
- Grounding supports in metadate provide a way to correlate grounding chunks and w/ the generated output text.
- Search w/ tools

## [linstrum]

- Efficient strategies for specializing LLMs: but method is task n task basis, look @ similar problem
  - \* experiment!
- Fine-tuning issues: worst case = too far, randomly do specific task.
  - \*imp n have thorough eval.
  - ↳ does is task in distribution of model?
    - paper: "inverse curse"
- Challenges in finetuning: conflicting methods to approach security, different data sets for tasks which are sensitive.
  - \* citations provide confidence
- MLLM: emphasis on guidance.
- Tradeoff of fine-tuning:
  - \* fine-tun, \* chance to lose memory on other tasks.
- In-context learning very effective.
- Retrieval w/ caching systems in place.
- Reducing token count (min value token savings)
- C,D,C,B,

Pop Quiz