

Offre n°2024-07866

PhD Position F/M Code and Proof Generation with Large Language Models

Le descriptif de l'offre ci-dessous est en Anglais

Type de contrat :CDD

Niveau de diplôme exigé :Bac + 5 ou équivalent

Fonction :Doctorant

Mission confiée

Generative AI is gaining momentum and has raised significant interest in tackling more and more problems from linguistics, maths, commonsense reasoning, biology, physics, etc.

Transformers introduced in [1] have quickly become the state-of-the-art neural network architecture for sequence processing with applications ranging from natural language processing and computer vision to code generation [2]. Transformers performances scale with the number of parameters and the number of training data [3], and with modern GPU/TPU chips it is now possible to train very large Transformers models with billions of parameters.

Large Language Models (LLMs), like GPT-4, are extremely large Transformers models trained for natural processing tasks on huge datasets containing billions of words. After the initial training, LLMs can be specialized for a specific task using various techniques:

- Fine-tuning consists in adjusting the parameters of an LLM by re-training the model, or part of the model, on a specialized dataset starting from pre-trained parameters. In addition, direct preference optimization [4] can fine-tune LMs to align with human preferences, achieving precise control of the behavior of LLMs.
- Prompt augmentation techniques leverage the capabilities of general-purpose LLMs to learn and adapt by adding context directly in the user input thanks to a prompt. Retrieval Augmented Generation (RAG) [5] is an advanced form of prompt augmentation where, given a prompt, relevant data are retrieved from an external database, and added to the original prompt.

Beyond natural language, general-purpose LLMs quickly demonstrated emergent programming abilities due to the presence of code in the training dataset. There has been an explosion of specialized LLMs either entirely trained or fine-tuned on code: AlphaCode [6], StarCoder [7], Codex [2], CodeT5 [8], Code LLaMa [9], etc. Researchers are only beginning to explore the capabilities of LLMs for software development and many challenges need to be addressed.

In this thesis, we will explore new research in neural code generation and applications to formal verification. To improve the reliability of LLM based code assistants, we will explore possible interactions between the LLM and external tools like a Python interpreter, a test framework, or a proof assistant. While LLM based interactive tools are only nascent, they have the potential to improve software development at every level.

LLMs have shown promise in proving formal theorems using interactive theorem provers (ITP) such as Isabelle, Lean or Coq. While full proof automation remains challenging, one of our goals in this thesis is to build a tool to enable the triple interaction human-ITP-LLM for Coq. We will explore various fine-tuning and prompt augmentation techniques in this context and then focus more precisely on the verification of generated code. We want to use an LLM to formalize a specification in Coq, and generate both the corresponding code, and a proof of correctness using existing formalized semantics. The proof assistant then tries the proof to accept or reject a program, and the human can validate the formal specification, or refine it if necessary.

****References:****

- [1] Attention Is All You Need, Vaswani et al., 2017 <https://arxiv.org/abs/1706.03762>
- [2] Evaluating Large Language Models Trained on Code, Chen et al., 2021 <https://arxiv.org/abs/2107.03374>
- [3] Training Compute-Optimal Large Language Models, Hoffmann et al., 2022 <https://arxiv.org/abs/2203.15556>
- [4] Direct Preference Optimization: Your Language Model is Secretly a Reward Model, Rafailov et al., 2023 <https://arxiv.org/abs/2305.18290>
- [5] Retrieval-Augmented Generation for Knowledge-Intensive NLP Tasks, Lewis et al., 2020 <https://arxiv.org/abs/2005.11401>

- [6] Competition-Level Code Generation with AlphaCode, Li et al., 2022 <https://arxiv.org/abs/2203.07814>
- [7] StarCoder: may the source be with you!, Li et al., 2023 <https://arxiv.org/abs/2305.06161>
- [8] CodeT5: Identifier-aware Unified Encoder-Decoder Models for Code Understanding and Generation, Wang et al., 2021, <https://arxiv.org/abs/2109.00859>
- [9] Code Llama: Open Foundation Models for Code, Rozière et al., 2023 <https://arxiv.org/abs/2308.12950>

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Informations générales

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- **Ville :** Paris
- **Centre Inria :** [Centre Inria de Paris](#)
- **Date de prise de fonction souhaitée :** 2024-09-01
- **Durée de contrat :** 3 ans
- **Date limite pour postuler :** 2024-07-21

Contacts

- **Équipe Inria :** [ARGO](#)
- **Directeur de thèse :**
Lelarge Marc / Marc.Lelarge@inria.fr

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