Generative AI project lifecycle

Embarking on the journey of developing and deploying a Language Model (LM) powered application requires a structured approach. In this article, we'll explore the generative AI project lifecycle, providing insights into the crucial stages from conception to deployment. By understanding this framework, you'll gain valuable intuition for making informed decisions, overcoming challenges, and building robust infrastructure for your application.

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Scope	Select	Adapt and align model		Application integration	
Define the use case	Choose an existing model or pretrain your own	Prompt engineering	Evaluate	and deploy model for inference	Augment model and build LLM- powered applications
		Fine-tuning			
		Align with human feedback			

Defining Scope:

The cornerstone of any project lies in defining its scope accurately. With LMs, versatility varies based on model size and architecture. It's essential to delineate the specific function your LM will serve in your application. Whether it's multifaceted tasks or specialized functions like named entity recognition, precision in scope definition can save time and computational resources.

• Model Selection:

Once the scope is defined, the next decision revolves around choosing between training a model from scratch or leveraging existing base models. Typically, starting with pre-trained models is preferred, though custom training may be necessary in certain scenarios.

Understanding the considerations behind this choice is pivotal for efficient project management.

Assessing Performance and Training:

With the chosen model, the focus shifts to assessing its performance and potential additional training. Prompt engineering and in-context learning are initial strategies to optimize model performance. Fine-tuning through supervised learning becomes relevant if the model falls short of desired performance. This iterative process sets the stage for refining the model to meet application requirements.

• Ensuring Ethical Behaviour:

As LMs evolve, ensuring their alignment with human preferences becomes paramount. Techniques like reinforcement learning with human feedback aid in shaping models to behave ethically and responsibly. Evaluation metrics play a crucial role in assessing model performance and alignment with desired outcomes.

• Deployment and Optimization:

Deploying the optimized model into the application infrastructure marks a significant milestone. Optimization for deployment ensures efficient resource utilization and enhances user experience. Additionally, considering supplementary infrastructure requirements is essential to address inherent limitations of LMs, such as information invention and limited reasoning abilities.

• Overcoming Limitations:

Addressing fundamental limitations of LMs requires innovative solutions. Techniques for mitigating these challenges, such as overcoming information invention tendencies and enhancing reasoning capabilities, are vital for advancing LM applications.

Mastering the generative AI project lifecycle equips developers with the
necessary framework to navigate the complexities of LM-powered
applications. By adhering to a structured approach from defining scope to
deployment and beyond, developers can harness the full potential of LMs
while addressing ethical considerations and overcoming inherent

limitations. As you embark on your project journey, remember to iterate through each stage, refining your approach to meet evolving requirements and ensure success.