Could you briefly share some of the key problems your team has solved using Al technologies?

1. Correctness: How accurately the model performs its intended function and produces the expected outputs.

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2. Robustness: The model's ability to produce correct results when exposed to new, unseen, or noisy data and resist adversarial input perturbations.

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3. Efficiency: A trained Al model's ability to produce outputs with minimal time, computational power, and energy.

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4. Fairness: It ensures that the model's decisions do not contain discrimination or bias against certain demographic groups or protected attributes.

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5. Explainability: The ability to explain the model's decision-making processes.

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6. Privacy: It ensures that personal and sensitive data related to the Al model (e.g., training data) is protected from unauthorized access.

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7. Scalability: It measures the model's ability to maintain acceptable performance when its workload or the amount of data it handles increases.

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8. Deployability: It measures the ease with which an Al model can be integrated into existing environments, workflows, or systems while maintaining its performance and functionality.

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Which quality attribute do you consider most important to ensure before deployment? Does this priority remain the same across all your models, or does it differ based on factors such as model type or application domain?

Is the AI software development workflow that your team adopted significantly different from the classical software development workflow, for example, Scrum or Waterfall? Can you briefly describe your workflow?

Have you utilized self-collected data, purchased data from other companies, or open-source datasets to create training/testing datasets? What factors influenced your decision-making process in selecting these sources?

What components of the Al model development process are fully automated, and what components require human intervention?

Could you provide examples of how you would determine which attribute of an Al software (such as correctness, fairness) to prioritize based on the client's requirements? How would you align the client's needs with these aspects?

Do you have experience that you had limited data to work with? How did you and your team ensure that your Al models were still reliable?

Usually, the data distribution of the target environment of AI software will drift over time. For example, let's consider an AI model for predicting house prices in a certain area. Over time, factors like the local economy, population growth, or even unforeseen events like natural disasters could change the housing market dynamics. The model that was initially trained on past data may no longer provide accurate predictions because the data distribution it was trained on has drifted from the current reality.

In the context of your recent projects, do you believe that they are or could be affected by data drifting over time? If so, how would you approach this issue to ensure that your Al models remain high quality as the underlying data changes? How frequently data update is needed?

What methods will your team use to validate and clean data in order to ensure data quality? Will this process influence the quality of the Al project?

If you could envision a tool to assist in addressing data quality issues in your Al projects, what key features or capabilities would you desire? How would this ideal tool facilitate your work and improve the overall quality of your data?

In the projects that you have developed, when do you prefer to develop models from scratch and when do you prefer to fine-tune a pretrained model to assure project quality? What affects your decision-making process?

Under the context that you decided to utilise a pre-trained model:

- Which source do you want to search for a suitable PTM? (e.g. Hugging Face, GitHub, PyTorch Hub etc).
- How to find a suitable model architecture?
- How to find a suitable model size & precision? The bigger the better?

Are Al-specific logging libraries, such as mlflow and wandb, being used in your project to track the quality of Al software? If so, how do you use them?

What strategies would you usually employ to select the features that will be involved in the model input in order to obtain a high quality? What will influence your decision-making?

If the project deadline is tight, how would you prioritize the evaluation criteria and why?

What strategies are typically employed to ensure that quality outcomes obtained in the development environment can accurately reflect the expected quality outcomes in the deployment environment?

What are the main differences between checking the quality of Al software before deployment and after deployment?

Do you use any method to generate/expand datasets? How do you ensure the quality of the generated data?

What process would your team usually use to organise the data labelling process in order to obtain good data quality while using a relatively low cost?

Have your team managed to obtain feedback from the end users? To what extent can end users' feedback help to improve the Al project's quality?

In the field of Al products, gaining customer trust is paramount to the successful deployment and adoption of the solutions. The users need to trust not only the accuracy but also the reliability, fairness, and transparency of the models. Could you share your strategies or specific actions taken to build and maintain this trust in your Al products?