Project overview:

The main purpose of this project is to develop an AI powered threat detection with LLM explanaition and real-time monitoring. The solution should be scalable and deployable in different environments.

Development solution:

The datasets used (with their correspondant 3 models) are from

3 AI models, each one is trained on a different dataset (IoT Malicious Detection Dataset - https://www.kaggle.com/datasets/agungpambudi/network-malware-detection-connection-analysis

Malware Detection in Network Traffic Data -

https://www.kaggle.com/datasets/agungpambudi/network-malware-detection-connection-analysis

And Network Traffic Data-Malicious Activity Detection

https://www.kaggle.com/datasets/advaitnmenon/network-traffic-data-malicious-activity-detection

The choice of these 3 different datasets is to create 3 models, each one trained on specific field of data, although they have a lot of common columns, the goal is to combine them together into our pipeline to have 3 models each one trained on a different domain of packets for a better prediction/packets classification together.

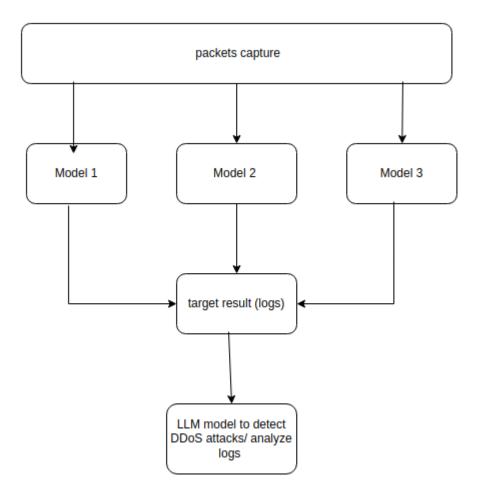
Prerequisites used to start developing:

2 kaggle notebooks already publically available, we didn't invent the wheel from scratch to see what will work and what won't.

https://www.kaggle.com/code/rem4000/xgboost-iot-malicious-detection-99-99-accuracy

https://www.kaggle.com/code/istiakahammedeee/explainable-ai-techniques-for-intrusion-detection

Setting up the project architecture:



Since the 3 models will be combined (works in parrallel), they should have a unified input, which are common columns to extract from each of the 3 datasets.

The models testing / accuracy is based on its training on these common columns only.

The models will be saved and imported to the pipeline once trained and validated.

Each model will give a prediction to the given packet, the logs will be inserted and analyzed by an LLM model to detect any sort of redendency or a DDoS attack or threat.

Setting up environments and testing

After training and saving the models from the notebooks, we test the models on:

Development enivornment: running script locally and generate a flow of packets to capture.

Staging environment: running the script from a virtual machine and capture packets sent from kali linux VM.

Production env: simulating real deployment on azure by checking the results after deployment

Steps:

After developing the 3 notebooks and saving the models,

Staging is to test locally between 2 VMs, simulating threat attacks with Kali linux

Example: ICMP flood., detector = detector1.py

Model execution

Attack simulation from the Kali VM

