Proposal

Defenses Against Adversarial Attacks

With the development and application of machine learning techniques, more and more applications adapt machine learning algorithms, such as self-driving car, face recognition, and stock prediction. However, the machine learning algorithms has potential and inevitable risk of false negative or false positive, which might cause large loss or risk in commercial applications (autonomous cars). Adversarial machine learning is a technique employed in the field of machine learning which attempts to fool models through malicious input.[[1]](https://en.wikipedia.org/wiki/Adversarial_machine_learning#cite_note-1)This technique can be applied for a variety of reasons, the most common being to attack or cause a malfunction in standard machine learning models.

Basically, I would like to cover several aspects of defense of adversarial attacks. For example, defense against Poisoning attacks and defense against adversarial attacks in deep neural networks. There are several ways to defense the adversarial attacks, such as distillation, Fortified networks and PeerNets which can exploit peer wisdom against adversarial attacks. In the proposal, I would like to discuss these ideas in more details

References:

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