# Forgetability in CNNs

#### Objective:-

To go deeper into the internal understanding of CNNs via class-wise information retention.

Knowing about the extent of forgetability while training/inferring to newer information (whole domain/incremental)

Either taking it as an opportunity to attackers or as privacy preserving mechanism

## Overview about the Problem Statement

#### About:-

- Nowadays, Neural Networks have become so sophisticated that they perform specialized applications in Medical, Civil, etc. with lesser hassle and sometimes more proficient than Humans.
- But while devising these architectures, we didn't focus on the response of newer information into the already pretrained models and how the neuron-based components can be brittle/plastic towards the current annotation space, while introducing newer samples.
- This can result into failures of sensitive applied implementations, especially in the field of Face Recognition, Disease Identification, etc. where the possibilities of anomalies and outliers are highly possible.

### Initial Approach:-

- We would be focussing on the effects of different components associated with CNNs like Learnable Convolutions, Pooling Layers, Skip Connections, etc. on the results towards training/testing samples and even towards newer/incremental sources of data.
- Apart from this, it's a possibility for us to devise a mechanism to prevent the forgetability of other classes knowledge while tweeking on the characteristics of CNN focussing on target class.
- In addition to already present targeted tweeking methods (mostly for attack purposes), optionally we can go to that direction and make efforts for its usage in privacy-based defenses.

### Evaluation Scheme:-

- Apart from the traditional evaluation metrics like Accuracy, F1-Score, Support, etc.; we want to also look on the distortion of embeddings space after the target class gets affected by incremental data/ information.
- Also, we can go into possibilities of looking the statistical informations and deviations of embedding space and logits during the experimentations.
- In addition to above, if we would be able to have any existing libraries/codebase for visualizing Neural Network components, then it will be really great to look on that as well.