

Date - 14 Oct - 2024

PRE-TRAINED MODELS IN CNN

- These models are initially trained on large, general-purpose datasets like ImageNet.
- They learn to recognise various features, from simple edge to complex textures and objects.
- The architecture of pretrained models varies, but they share common traits.

1. ResNet (Residual Networks):

- it introduced by Microsoft Research, revolutionised deep learning by using residual connections to mitigate the vanishing gradient problem in deep networks.
- variants: ResNet-50, ResNet-101, ResNet-152

2. Inception (GoogleNet):

- developed by google, the inception network uses inception modules to capture multi-scale features.
- variants: inception v3, inception v4, inceptionResNet

3. VGG (Visual Geometry Group)

- developed by the visual Geometry Group at the university of oxford, VGG models are known for their simplicity and depth.
- variants: VGG-16, VGG-19

4. EfficientNet:

- Developed by Google, EfficientNet models achieve high accuracy with fewer parameters and computational resources.
- variants: EfficientNet-B0 to EfficientNet-B7.

5. DenseNet (Dense Convolutional Network):

- Developed by researchers at Cornell University, DenseNet connects each layer to every other layer in a feed forward fashion.
- Variants: DenseNet121, DenseNet-169, DenseNet-201.

6. MobileNet:

- Developed by Google, MobileNet models are designed for mobile and embedded vision applications.
- variants: MobileNetV1, MobileNetV2, MobileNetV3.

7. NASNet (Neural Architecture Search Network):

- Developed by Google using neural architecture search techniques to optimize the network structure.
- variants: NASNet-A, NASNet-B, NASNet-C.

8. Xception (Extreme Inception):

- Developed by Google, Xception is an extension of the inception architecture with depthwise separable convolutions.

9. AlexNet:

Developed by Alex Krizhevsky, AlexNet is one of the earliest deep learning models that popularized the use of CNNs in image classification.

10. Vision Transformers:

Developed by Google, Vision Transformers apply the transformer architecture, initially designed for NLP, to image classification.

TRANSFER LEARNING

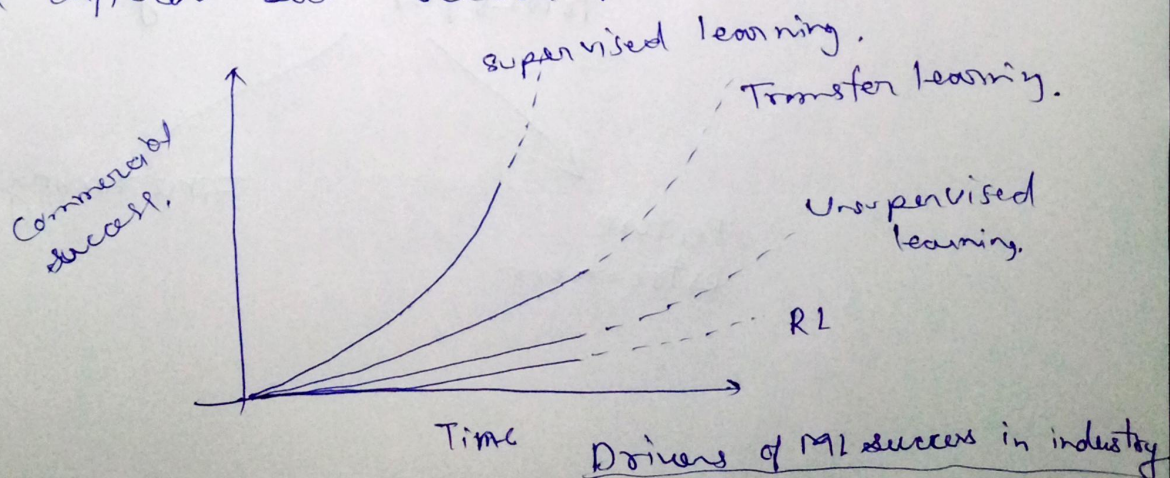
problem with training our own model:

- (i) Because Deep learning required lot of data with labeled
- (ii) it is time consuming.

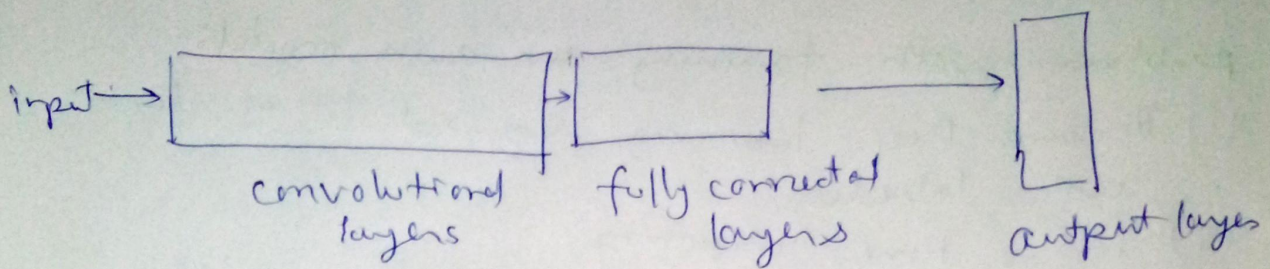
- The solution of above two statements is to use pre-trained models example VGG, ResNet, Xception.
- But there is may be chance that the we solve our problem statement using pre-trained model but our problem statement's class is not present in dataset where the pre-trained model is trained to resolve this problem we ~~use~~ use transfer learning.

Transfer learning:

Transfer learning is a research problem is ML that focuses on storing knowledge gained while solving one problem and applying it ~~to~~ to a different but related problem.



How Transfer learning works:



- remove the fully connected layer, and then build your own fully connected layer.
- freeze the convolutional layers (part) ~~has~~ became primitive features, are same in any image and weights change at the ~~last~~ convolution layer do freeze and $\frac{1}{2}$
- Training part is only done in fully connected layers
- But if, apply the fine tuning then train the last part of convolution also.

