

21) Transfer Learning

TL via Model Copying

- Transfers a trained model to another dataset/problem
- Purpose: Saves computational & data resources to go further
- When to use: ① ur problem is similar to the source model's
② ur data is insufficient ③ Source model is deep
- Process: Fine-Tuning the model: Freezes some layers (≥ 0) & Retrain remaining
more ① = more layers to freeze

Famous Pretrained CNNs

- PyTorch provides importable pretrained CNNs

① LeNet: standard convolution-pooling block

$x \rightarrow \text{Convolution} \rightarrow \text{pooling} \rightarrow \text{Convolution} \rightarrow \text{pooling} \xrightarrow{\text{Flatten}} \text{dense} \rightarrow \text{dense} \rightarrow \text{dense}$
(28×28) ($6 @ 28 \times 28$) ($6 @ 14 \times 14$) ($16 @ 10 \times 10$) ($16 @ 5 \times 5$) ($120, 84$) ($84, 10$) ($10, 1$)

② AlexNet: recognizes images in 1000 categories w/ human-level performance

- Model trained on 2 separate GPUs & Convolutions stacked together w/ ReLU
- more suitable for transfer learning

③ VGGNet: similar to AlexNet

④ ResNet: similar to AlexNet but some layers have residual/skipped connections^{to other layers}

TL via Autoencoder

- Steps: ① Pretrain AE model ② Transfer encoder part of AE
③ Attach it to prediction layers (ex. FFN) ④ Fine-Tune this model
- Pros & Cons: ① Learn features tailored to ur data
② Can reuse data from Step 1 & 4 without overfitting
③ Do not necessarily learn features relevant for prediction