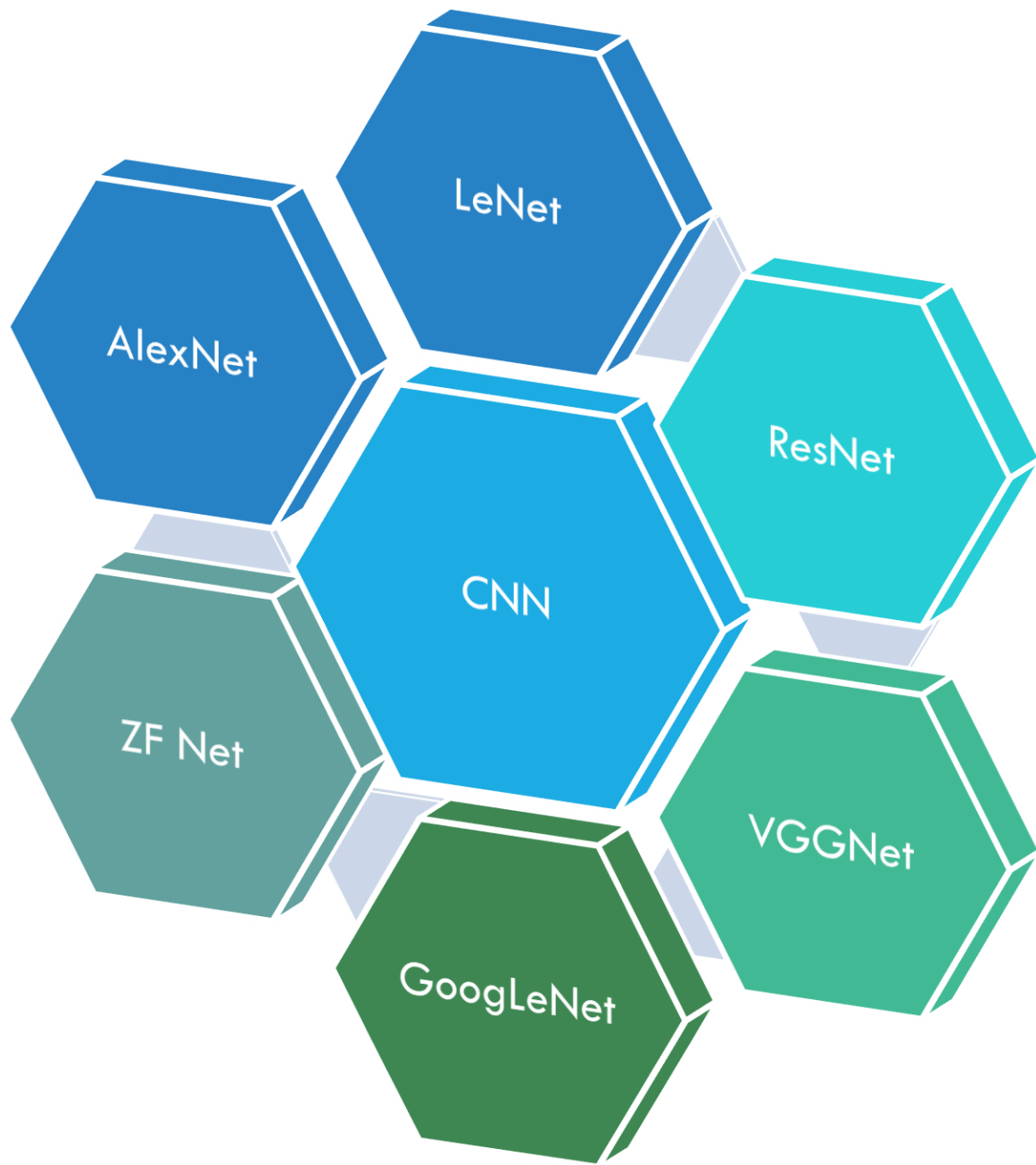


Convolutional Neural Network: Module 4 - Transfer Learning Artificial Intelligence

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Existing Networks



ImageNet Challenge

IM GENET

- 1,000 object classes (categories).
- Images:
 - 1.2 M train
 - 100k test.



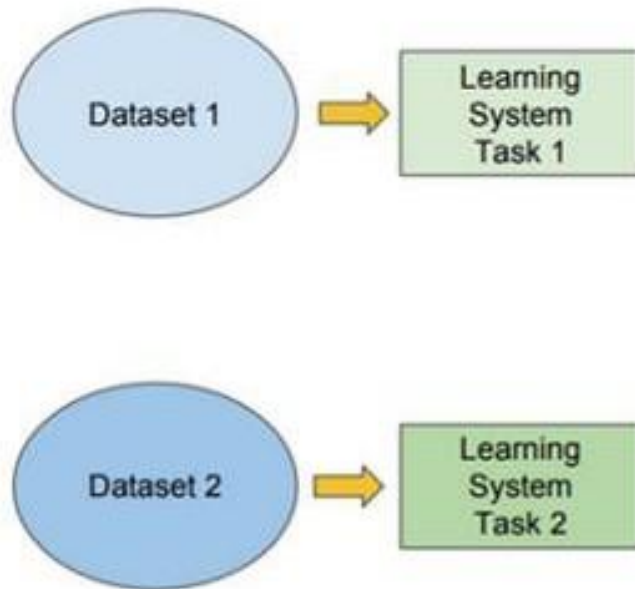
The popular ImageNet Challenge based on the ImageNet Database

Traditional ML

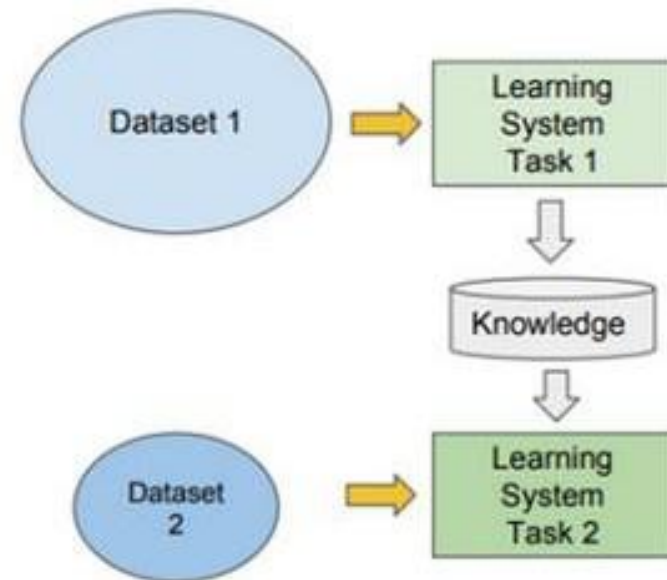
vs

Transfer Learning

- Isolated, single task learning:
 - Knowledge is not retained or accumulated. Learning is performed w.o. considering past learned knowledge in other tasks



- Learning of a new tasks relies on the previous learned tasks:
 - Learning process can be faster, more accurate and/or need less training data



Traditional Learning vs Transfer Learning

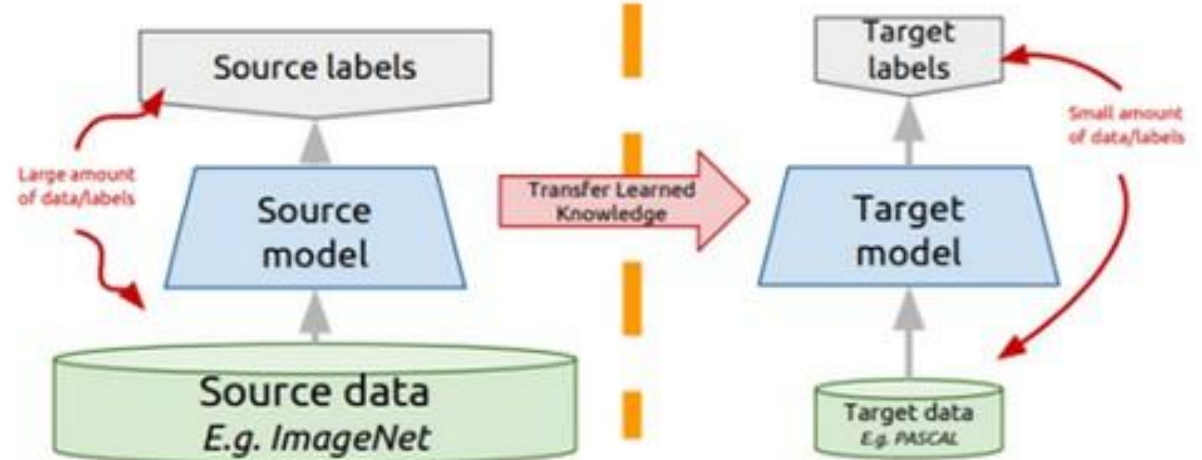
Transfer learning: idea

Instead of training a deep network from scratch for your task:

- Take a network trained on a different domain for a different **source task**
- Adapt it for your domain and your **target task**

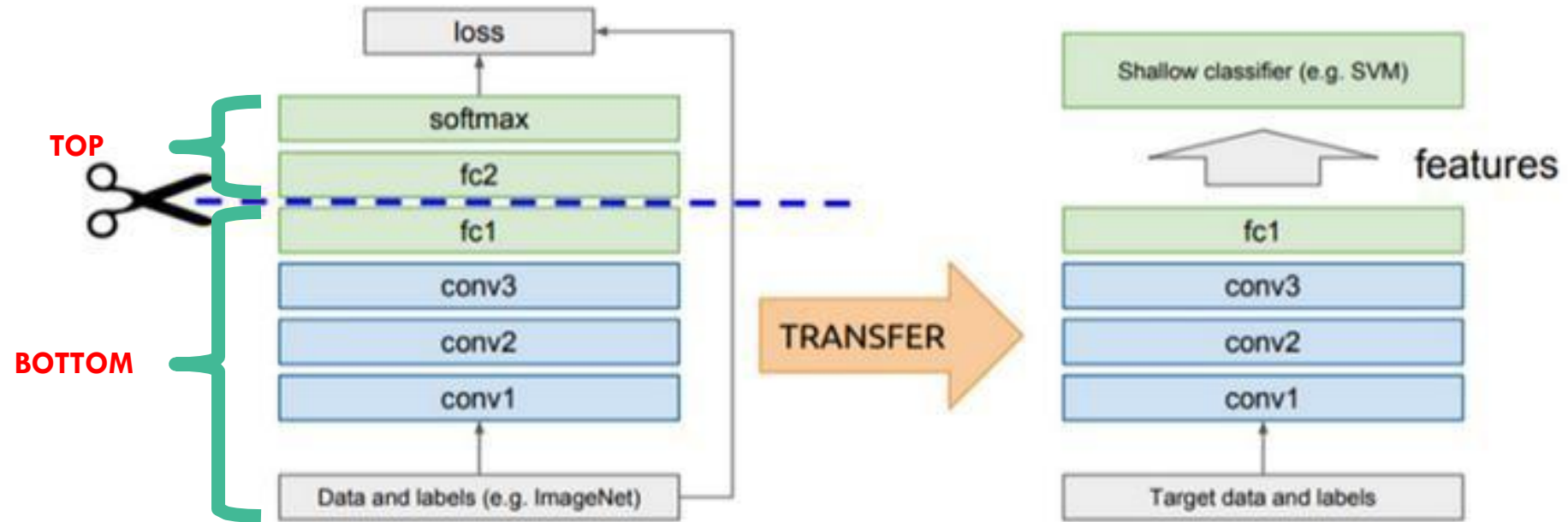
Variations:

- Same domain, different task
- Different domain, same task



Ideas for deep transfer learning

Assumes that $D_S = D_T$



Transfer Learning with Pre-trained Deep Learning Models as Feature Extractors

Fine-tuning: supervised domain adaptation

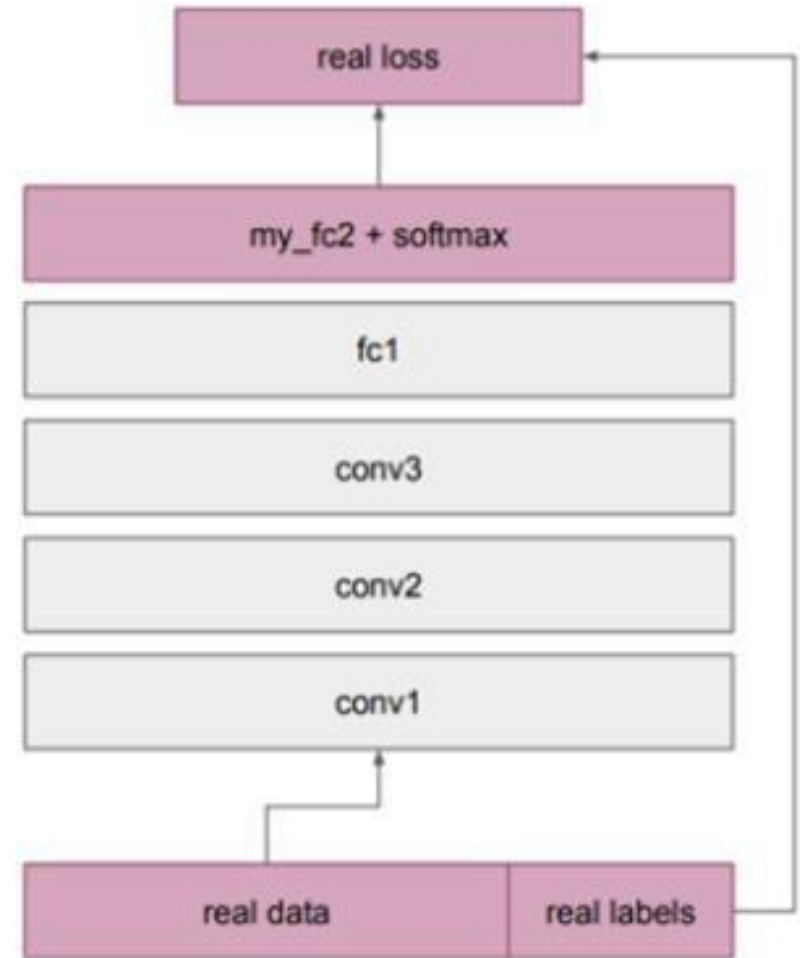
Train deep net on “nearby” task for which it is easy to get labels using standard backprop

- E.g. ImageNet classification
- Pseudo classes from augmented data
- Slow feature learning, ego-motion

Cut off top layer(s) of network and replace with supervised objective for target domain

Fine-tune network using backprop with labels for target domain until validation loss starts to increase

Aligns D_S with D_T



Freeze or fine-tune?

Bottom n layers can be frozen or fine tuned.

- **Frozen:** not updated during backprop
- **Fine-tuned:** updated during backprop

Which to do depends on target task:

- **Freeze:** target task labels are scarce, and we want to avoid overfitting
- **Fine-tune:** target task labels are more plentiful

In general, we can set learning rates to be different for each layer to find a tradeoff between freezing and fine tuning

