

Deep Learning

MSDS 631

Multi-Task, Data Augmentation,
Transfer Learning

Michael Ruddy

Questions?

- From last lecture?

Overview

- Small Datasets
- Transfer Learning
- Data Augmentation
- Multi-Task Learning

Small Datasets

- Not enough data
 - Expensive to obtain
 - Too big to store

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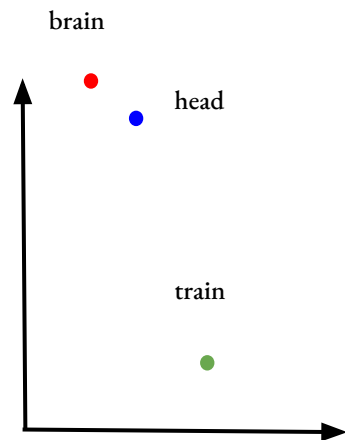
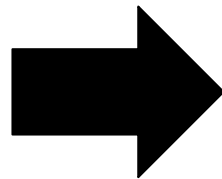
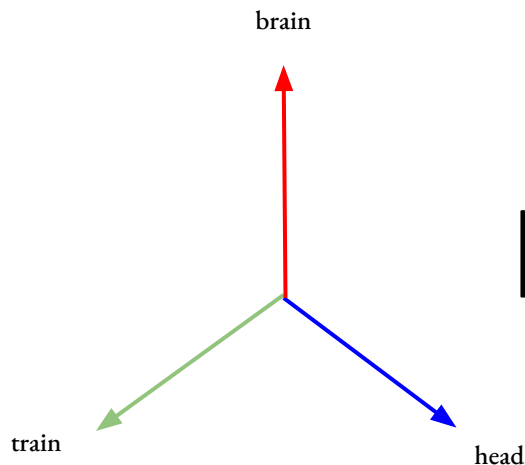
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- Multi-Task Learning

Transfer Learning

- Transfer features from one task/dataset to another
- Word Embeddings
 - Learn from context (lots of data)
 - Use for all sort of NLP tasks (maybe less data/labels)

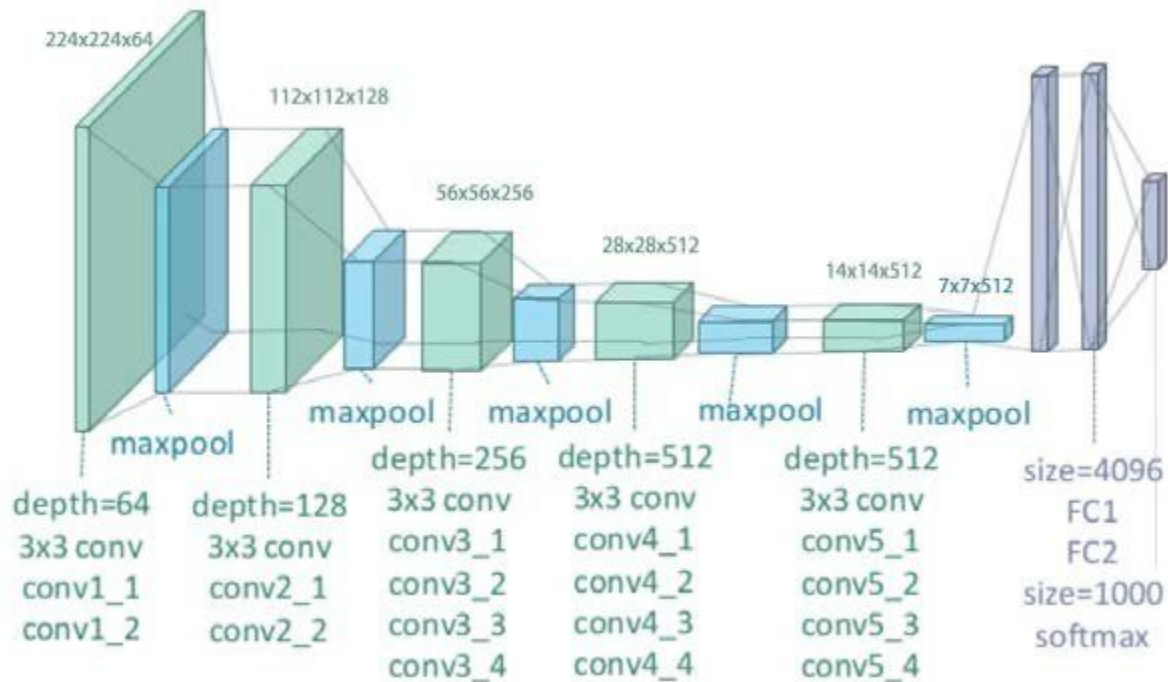
$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$
brain	head	train



Transfer Learning

- Transfer features from one task/dataset to another
- Word Embeddings
- Train a CNN on ImageNet: multi-class problem with 1000 classes
 - Lots of images!

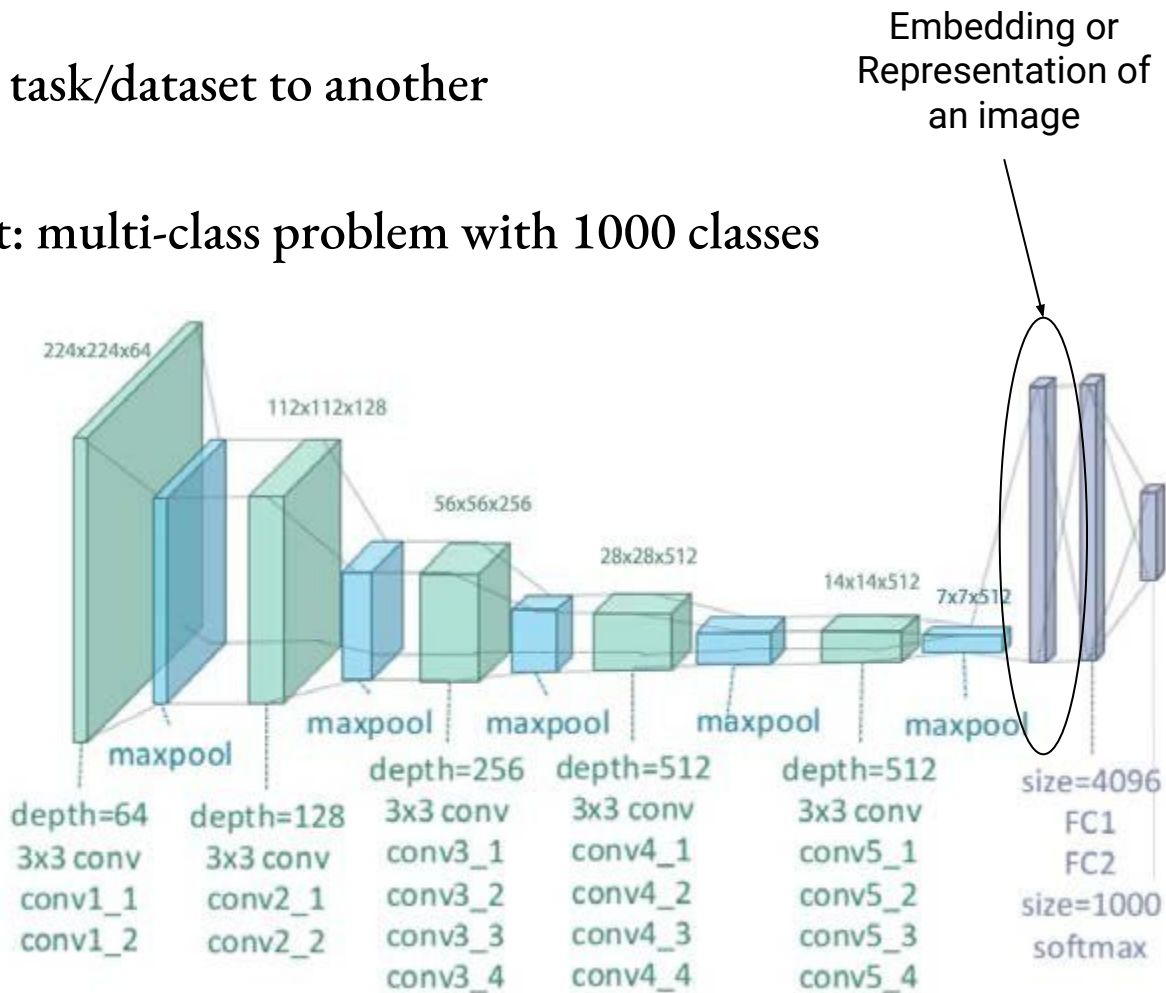
VGG-19



Transfer Learning

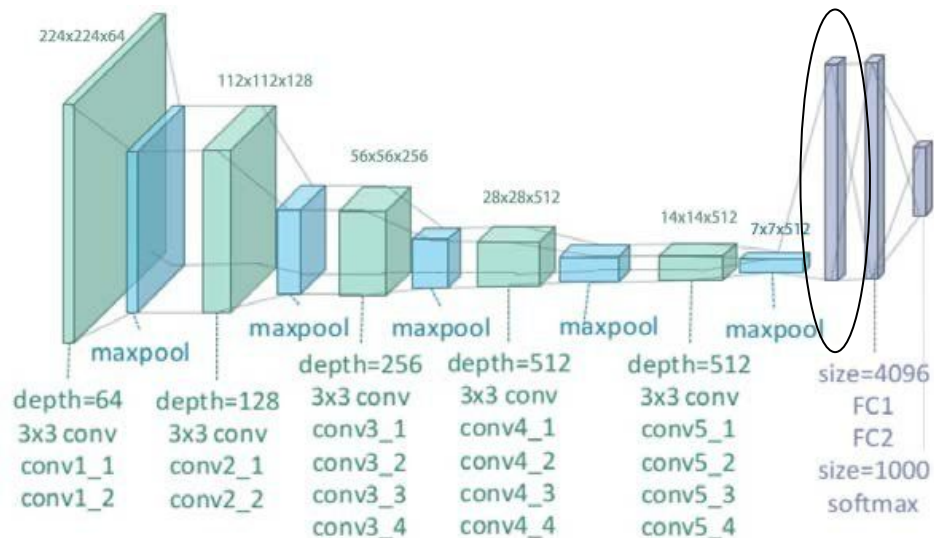
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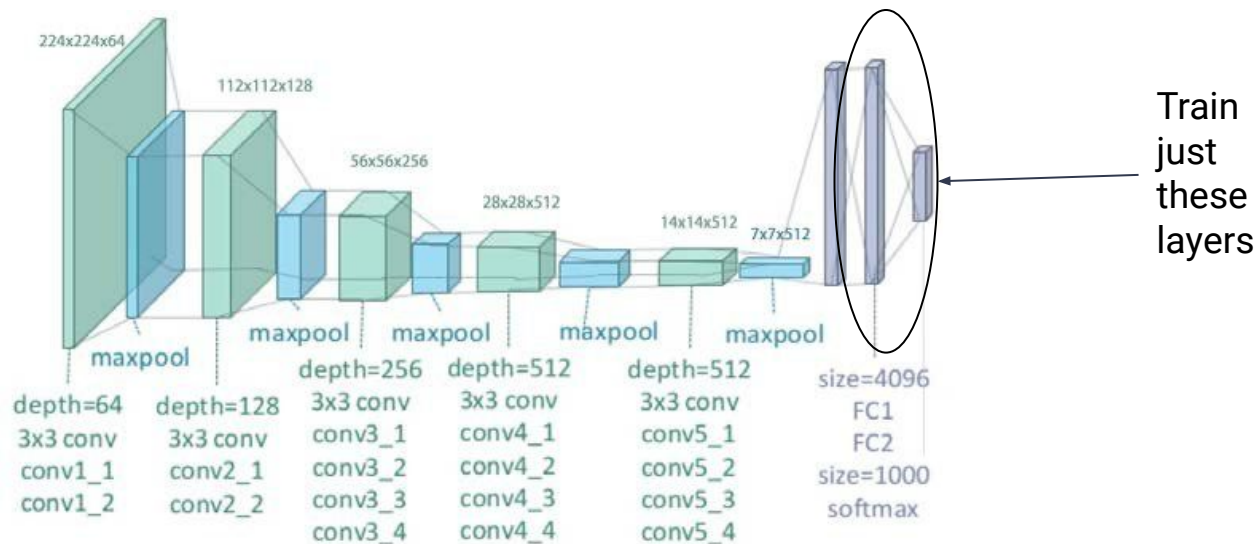
Transfer Learning

- Transfer features from one task/dataset to another
- Word Embeddings
- Train a CNN on ImageNet: multi-class problem with 1000 classes
- Two approaches
 - Freeze your embedding/representation



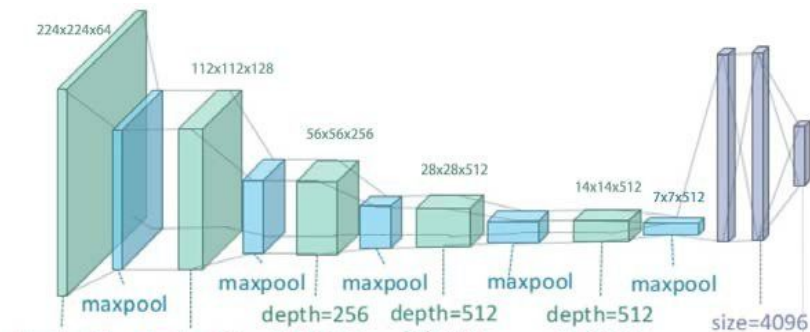
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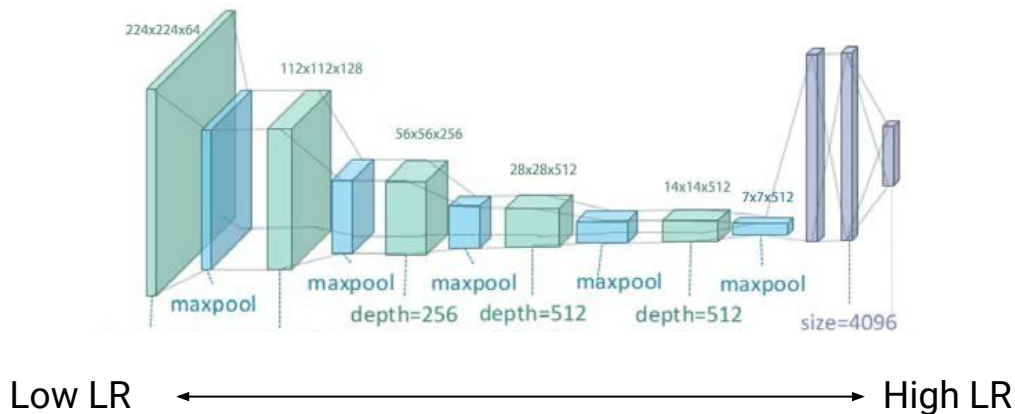
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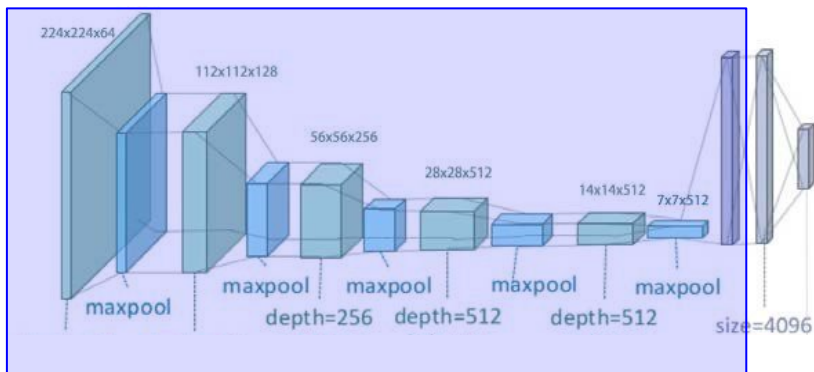
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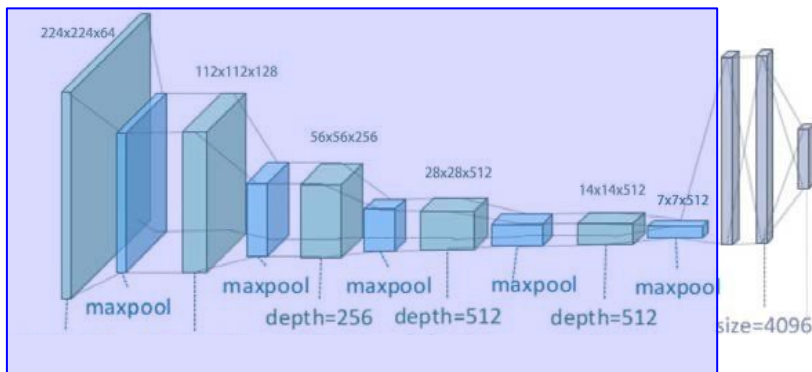
Epoch 0



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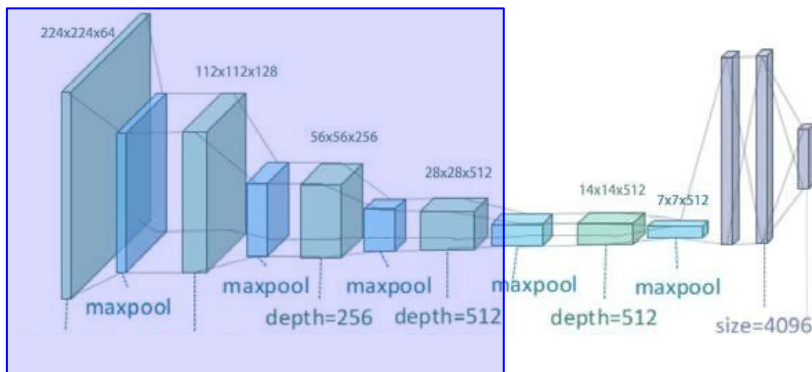
Epoch 1



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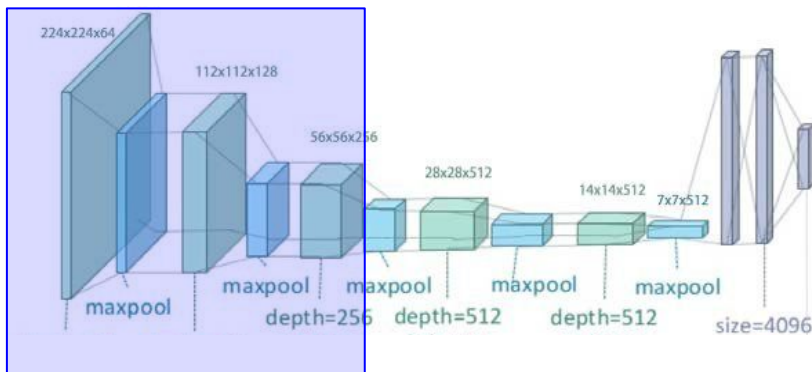
Epoch 2



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Epoch 3



Data Augmentation

- Idea: when you have a small dataset, just create more!
- Create synthetic data
- Transform your original data

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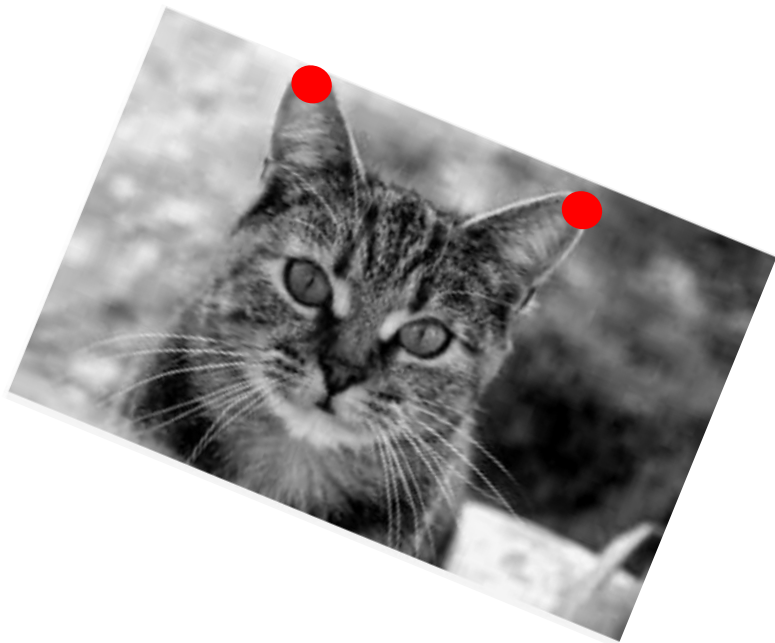
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 - Idea: model learns a bit of Euclidean geometry this way

Data Augmentation

- Idea: when you have a small dataset, just create more!
- Create synthetic data
- Transform your original data
 - Encode invariance/equivariance for a model
 - Idea: model learns a bit of Euclidean geometry this way
 - Make your model less sensitive to noise



This is still a cat...

Multi-Task Learning

- Two tasks where computed features can be shared
 - For example: Classification + Localization

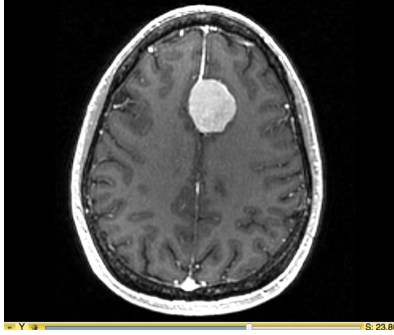
Multi-Task Learning

- Two tasks where computed features can be shared
 - For example: Classification + Localization
- Usually helpful when there is less data than desired somewhere
 - More labels on the same inputs potentially lets you learn “more” about the image
 - Maybe one task has a lot of data and another similar task does not

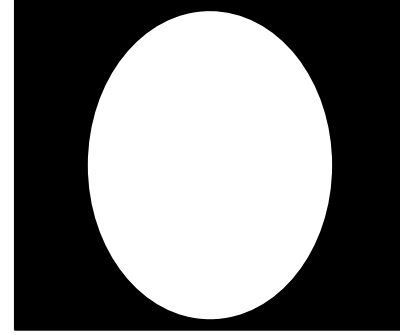
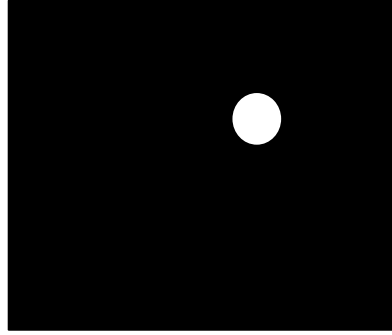
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- Careful!
 - Don't want a model that is just mediocre at both
 - One task can take over

Multi-Task Learning

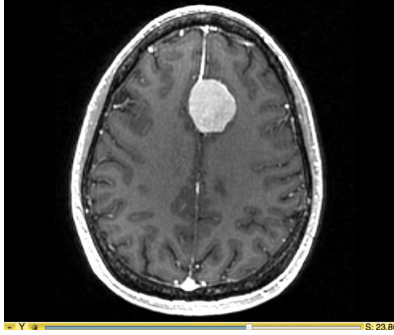


Input

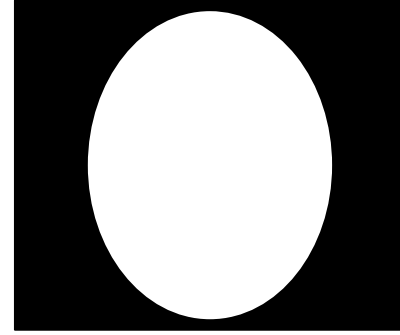
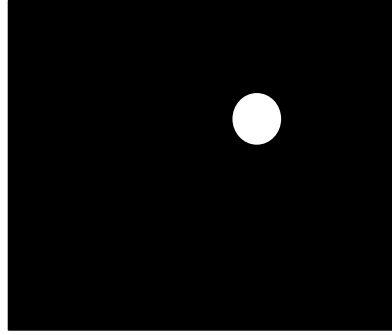


labels

Multi-Task Learning



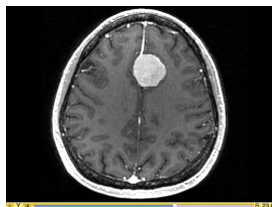
Input



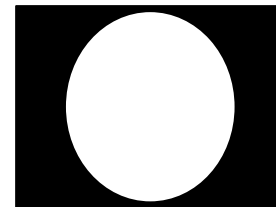
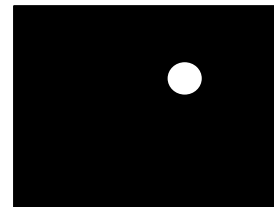
labels

Benign or Malignant

Multi-Task Learning

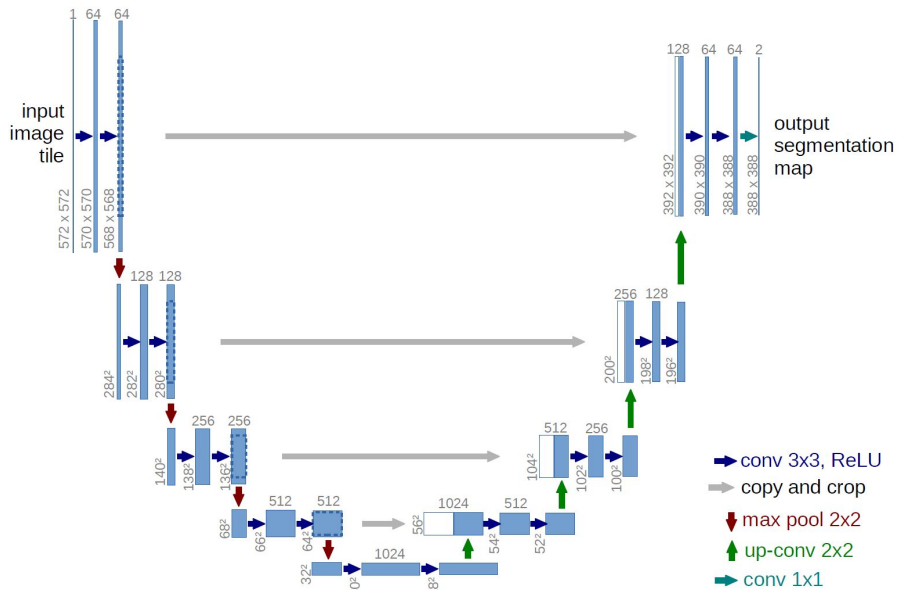


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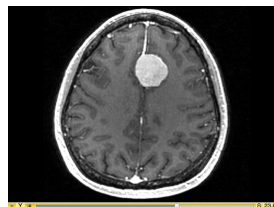


labels

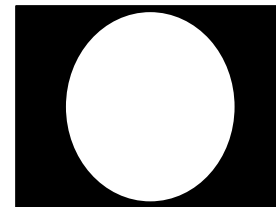
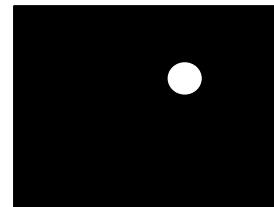
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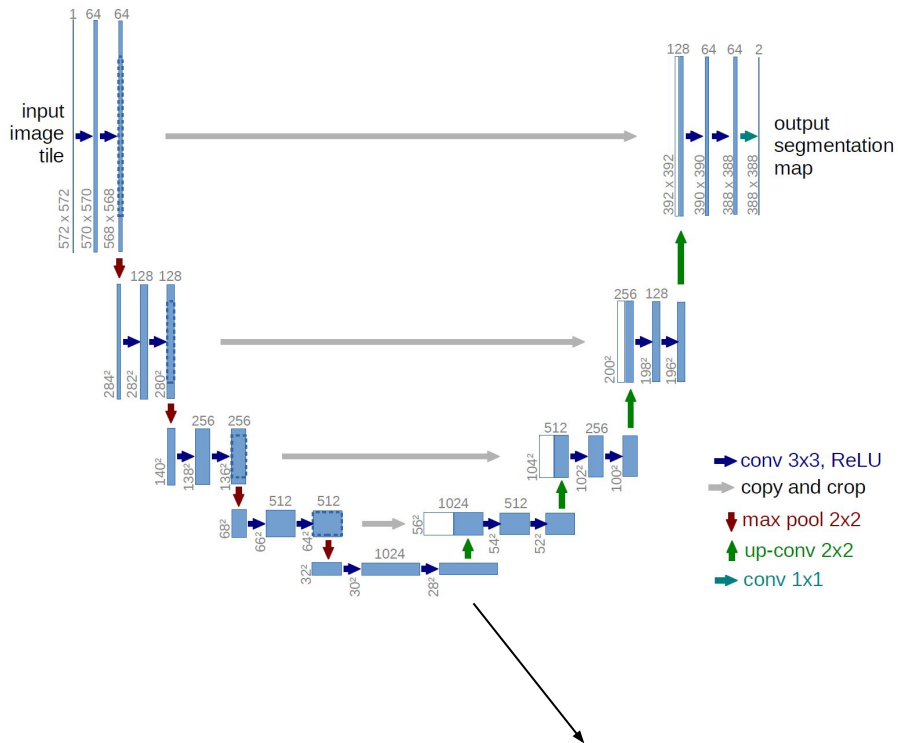


Input



labels

Benign or
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Unroll to Linear Layer then Classification

Summary

- Transfer Learning
 - Transfer features learned from one dataset/task to another
- Data Augmentation
 - Augment your dataset
 - Synthetic Data
 - Transformed Data
 - Encode invariance/equivariance to nuisance transformations
- Multi-Task Learning
 - Leverage other tasks to improve the target task
 - Like simultaneous transfer learning!