



Image Classification ✨

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Workshop Plan

01

**Datasets &
Evaluation**

02

**CNN
Architectures**

03

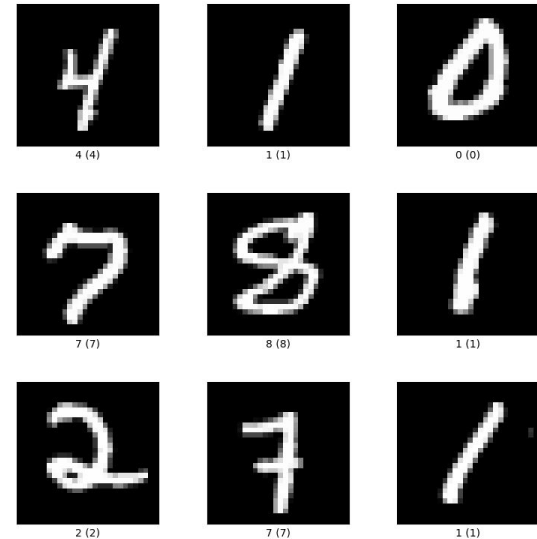
**Transfer
Learning**

01

Datasets & Evaluation

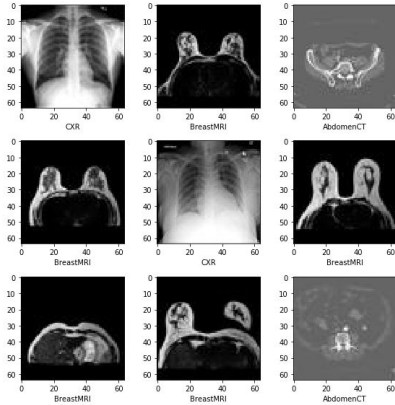
MNIST & Variants

- The "**Hello World**" of image classification.
- Consists of **28x28** grayscale images of handwritten digits (**0-9**)
- Contains **60,000** training images and **10,000** testing images.



MNIST & Variants

Medical MNIST



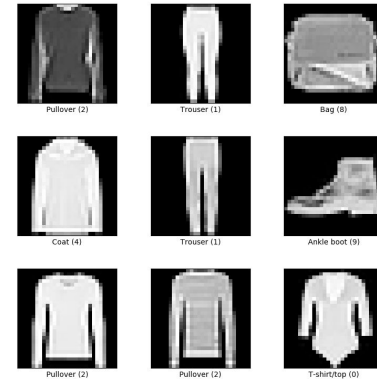
[Learn more](#)

Sign Language MNIST



[Learn more](#)

Fashion MNIST



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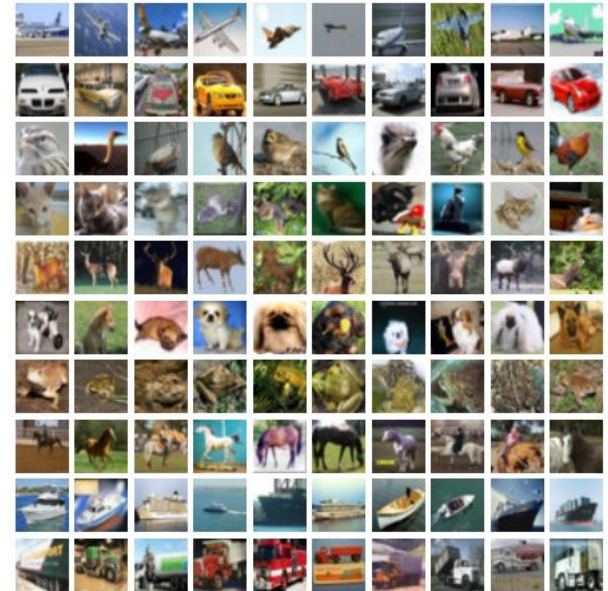
CIFAR

CIFAR-10

- Consists of **32x32 colored** images in **10 classes**, with **6000** images per class.
- The classes include **common objects** such as airplanes, cars, birds, cats, and more.

CIFAR-100

- Similar to CIFAR-10 but contains **100 classes**, with **600** images per class.



ImageNet

- One of the most widely used and **influential datasets** in computer vision.
- Consists of over **14 million** images.
- Has **+20,000** classes.
- Each image is annotated with **bounding boxes** and **hierarchical labels**.



Evaluation

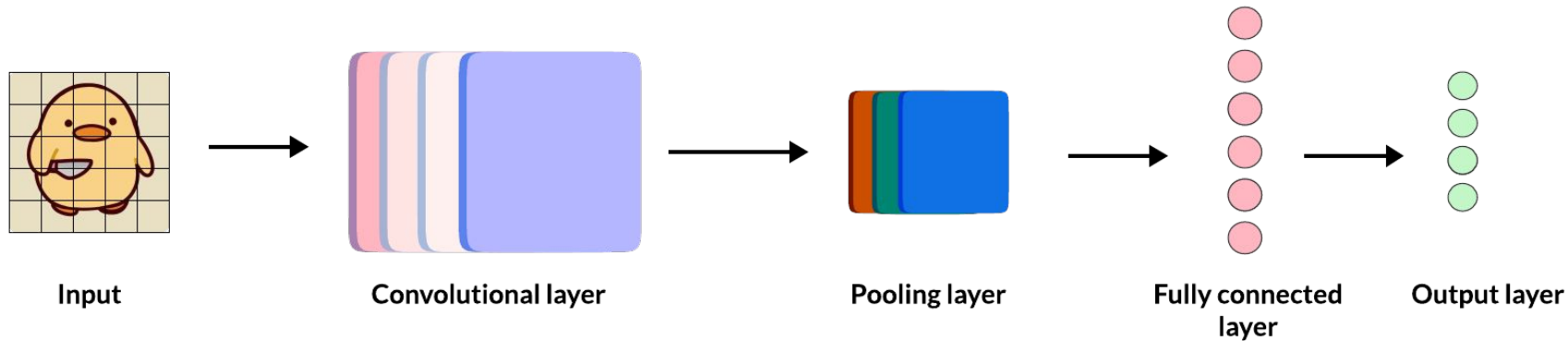
- As any other **classification** task, **image classification** models can be evaluated using the classic metrics :
 - Accuracy
 - Precision
 - Recall
 - F1 Score
 - Cross-entropy Loss
 -
- However, you might hear about some confusing terms. I got your back !
 - **Top-1 Accuracy** : The percentage of images for which the correct class is predicted as the **top prediction**, a.k.a **ol' plain accuracy** :)
 - **Top-k Accuracy** : The percentage of images for which the correct class is predicted within the **top k predictions**.

02

CNN **Architectures**

CNN Architectures

In the previous workshop, you learned what a simple **convolutional neural network architecture** is !

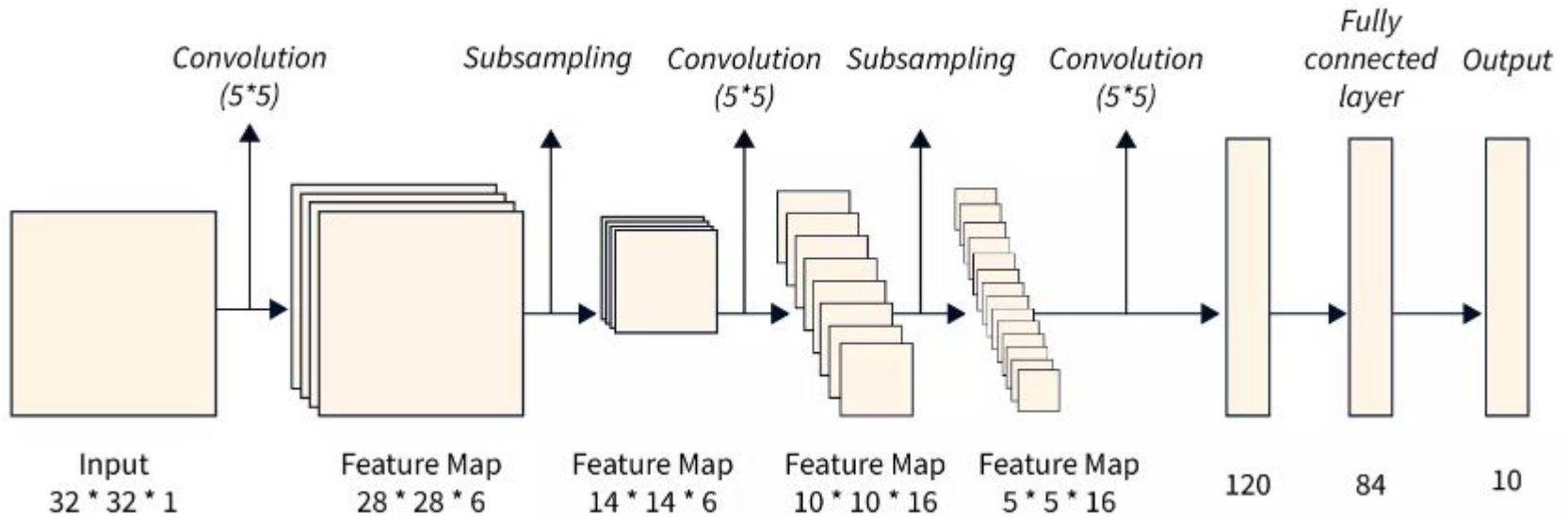


CNN Architectures

- The **number**, **arrangement**, **parameters** and **connections** between these components allow us to define various **CNN architectures** that work well according to the **dataset**, **task** and **hardware**.
- Let's explore some famous **CNN architectures** designed to tackle different complexities of **image classification** tasks



LeNet

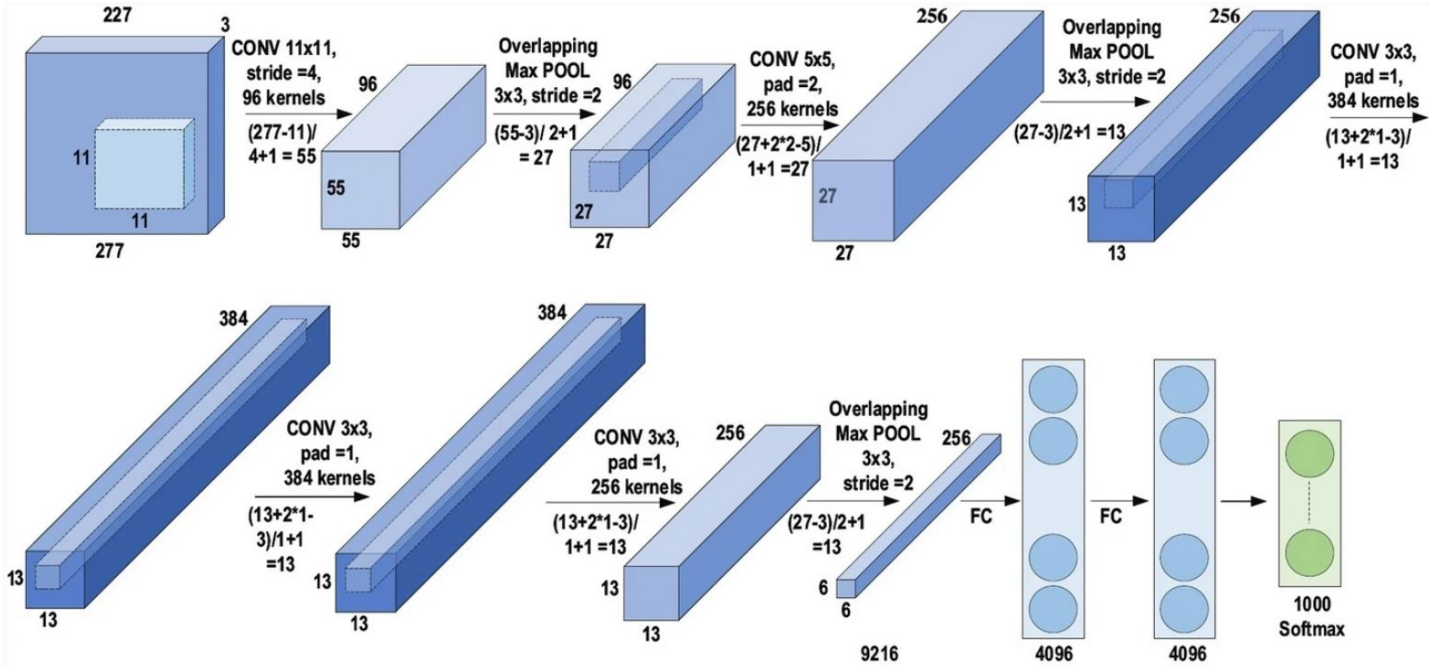


LeNet

- Consists of **convolutional layers** followed by **pooling layers** and **fully connected layers**. Yeah, the grandma of CNNs :-)
- Initially used for handwritten digit recognition tasks.
- Pioneered the concept of **trainable CNNs**.



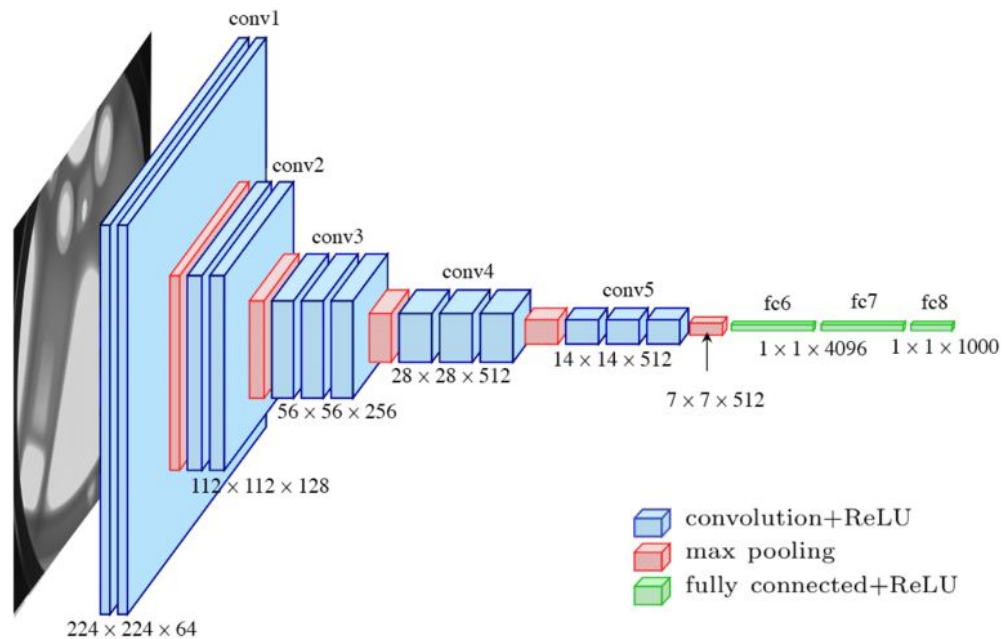
AlexNet



AlexNet

- Significantly **deeper** than previous networks at the time.
- Uses **ReLU** activation functions, **dropout** for regularization, and **data augmentation** techniques. She got it all 😎
- **Won the ImageNet Large Scale Visual Recognition Challenge (ILSVRC) in 2012, sparking the deep learning revolution ✨**

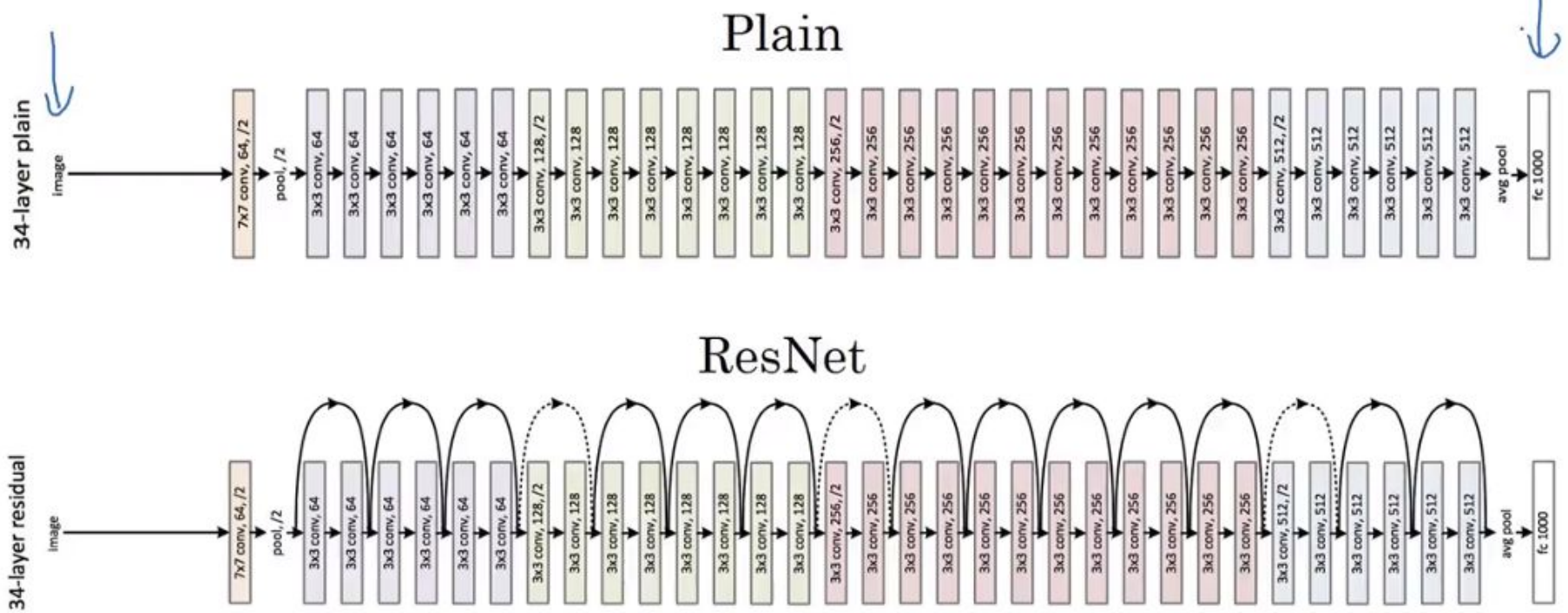
VGGNet



VGGNet

- Known for its simplicity and uniform architecture, consisting of multiple 3x3 convolutional layers.
- Available in different variants with varying depths (e.g., VGG16, VGG19).
- Achieved competitive performance on ImageNet classification.

ResNet



ResNet

- Addressed the **vanishing gradient problem** by utilizing **skip connections** (or residual connections) to enable training of very deep networks.
- Available in various **depths**, such as ResNet-18, ResNet-50, ResNet-101, and ResNet-152.
- Achieved **state-of-the-art** results on various image classification benchmarks.

There's **more** to explore !

ZFNet

GoogLeNet

EfficientNet

SENet

Xception

ResNeXt

DenseNet

As you can see, there **isn't** a limit to how many options we can explore. Crafting a **good architecture** is thus a complicated task, that requires a lot of **time** and **expertise**, and a non-ending loop of **trials** and **errors** !

So instead of starting **from scratch** with every new task, we can use the existing **AWESOME models** that AI scientists have came up with and **tweak** them **according to our needs**



03

Transfer Learning

Transfer Learning

We call the existing awesome models : **pretrained models**.

- A **pretrained model** is a saved network that was **previously trained** on a **large dataset**, typically on a large-scale image classification task.
- **Intuition** : if a model is trained on a **large** and **general enough** dataset, this model will effectively serve as a **generic model** of the visual world.

So we can then take advantage of them without having to start from scratch by training a large model on a large dataset.



Transfer Learning

We can use transfer learning in two ways :

1. **Using a pretrained model for feature extraction:** this means borrowing a pretrained model's knowledge to extract relevant features from new data, enhancing the performance of tasks without modifying the original model's parameters.
2. **Fine-tuning a pretrained model:** this involves adjusting the parameters of a pretrained model to better suit a new task or dataset. By fine-tuning, we allow the model to learn specific patterns of the new data while retaining the valuable knowledge it gained during pretraining.



Talk is cheap. Show me the code.

Linus Torvalds

“ quote fancy

Install linux, windows is for children

Thank you for attending

any questions ?



Thank you for attending

any questions ?

Don't be shy, ask

