

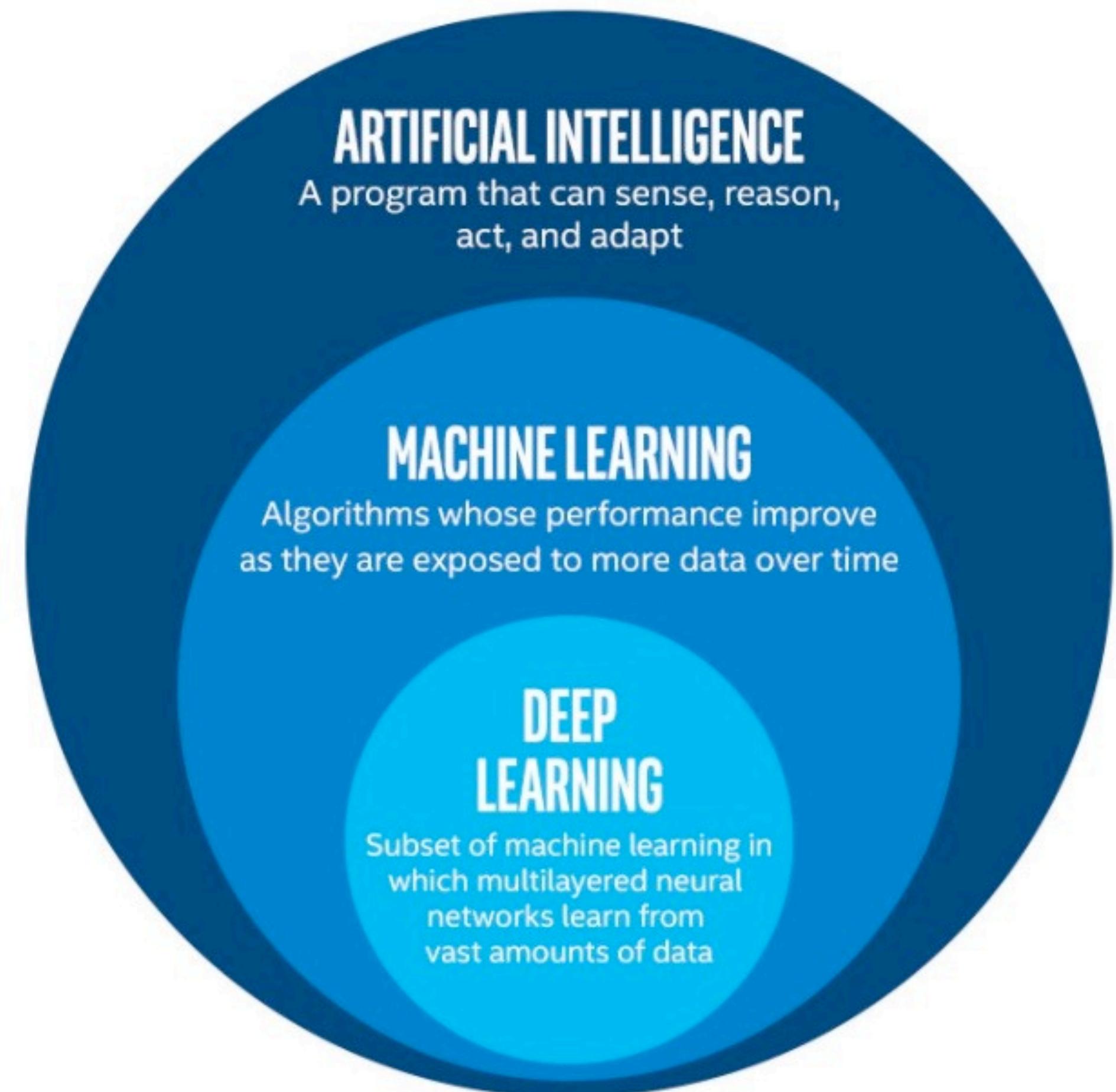
Machine learning

Autor: Cristopher McIntyre Garcia

Mini-Cours

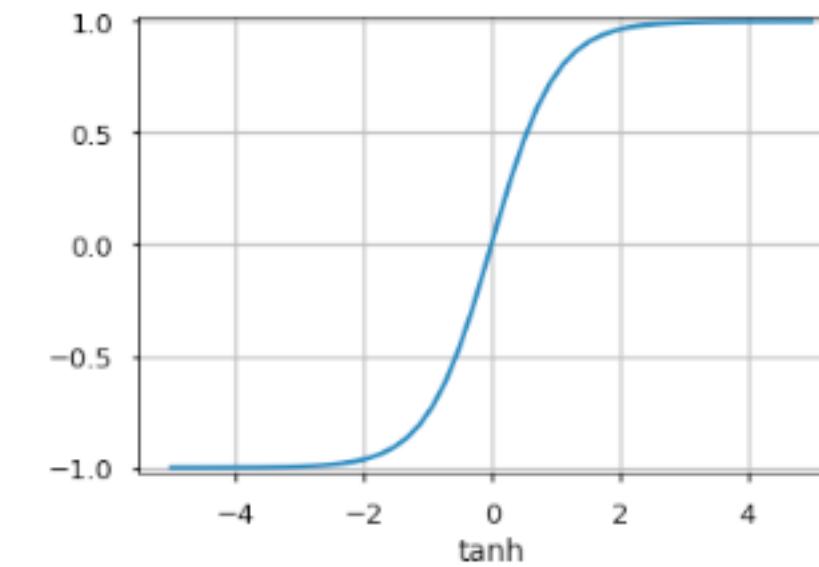
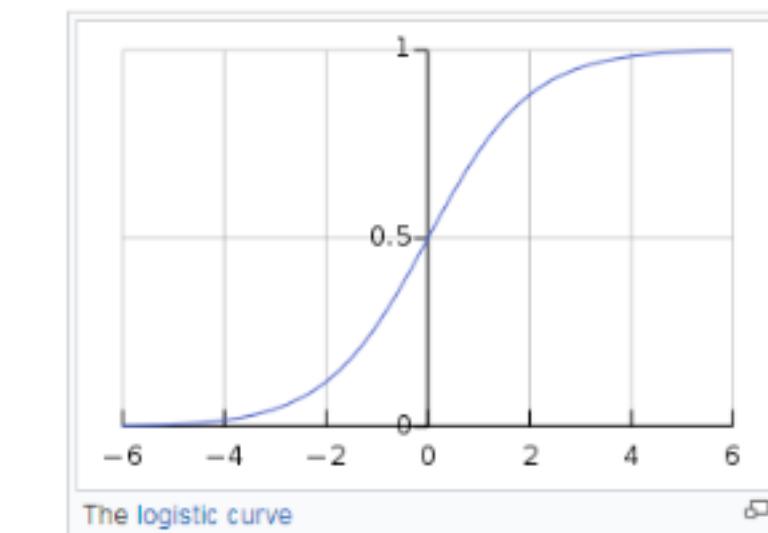
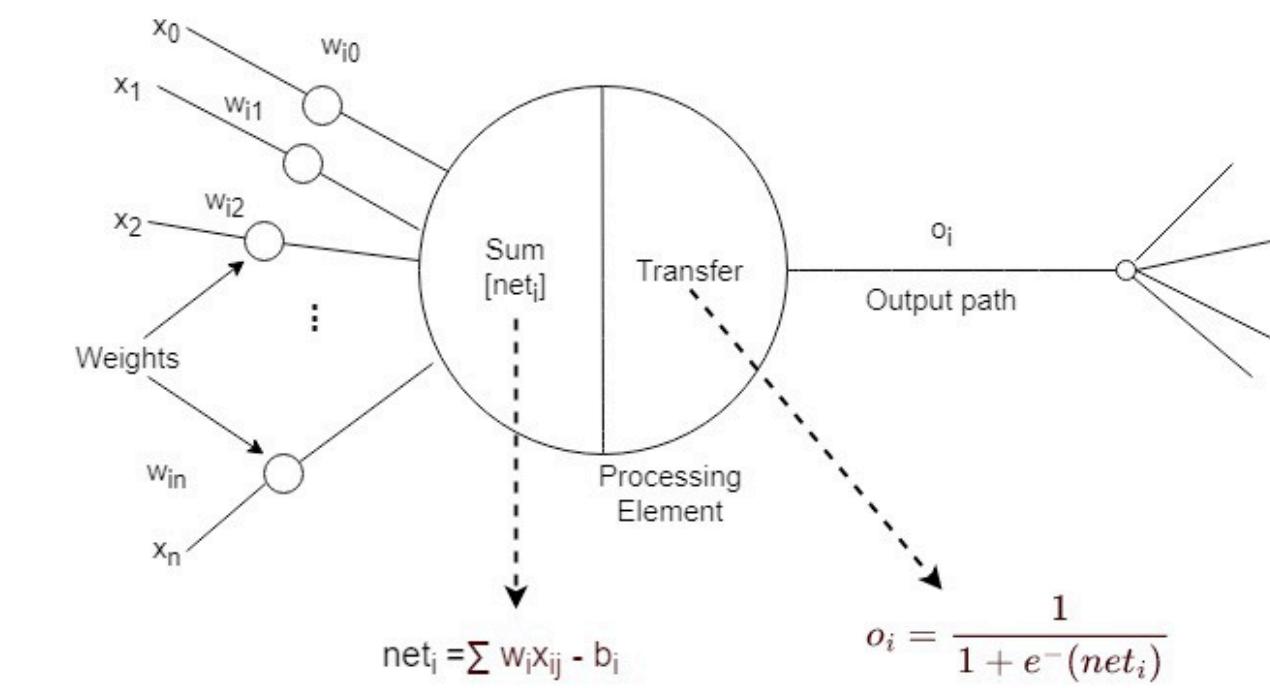
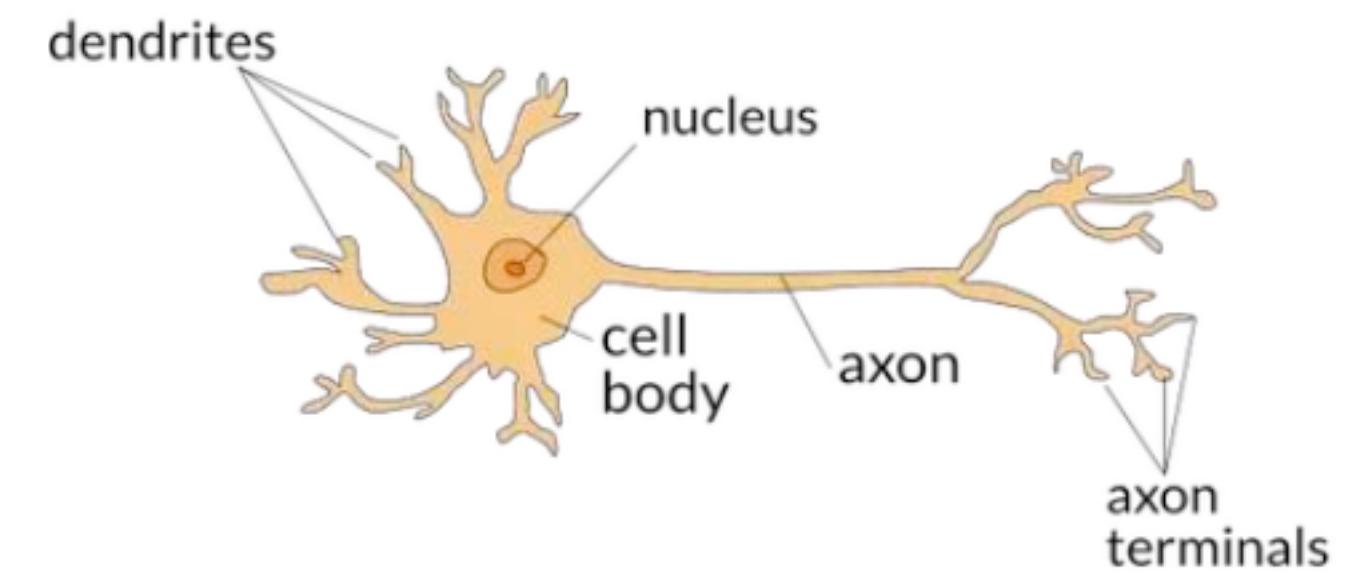
- Jour 1: Introduction à l'image digital
- Jour 2: Techniques de base de la vision artificiel et au traitement d'image
- Jour 3: Filtre et technique traditionnel avancé
- Jour 4: Introduction a l'intelligence artificiel (ML et DL)
- Jour 5: Technique modern - Conversation de nos courrent

AI vs ML vs DL



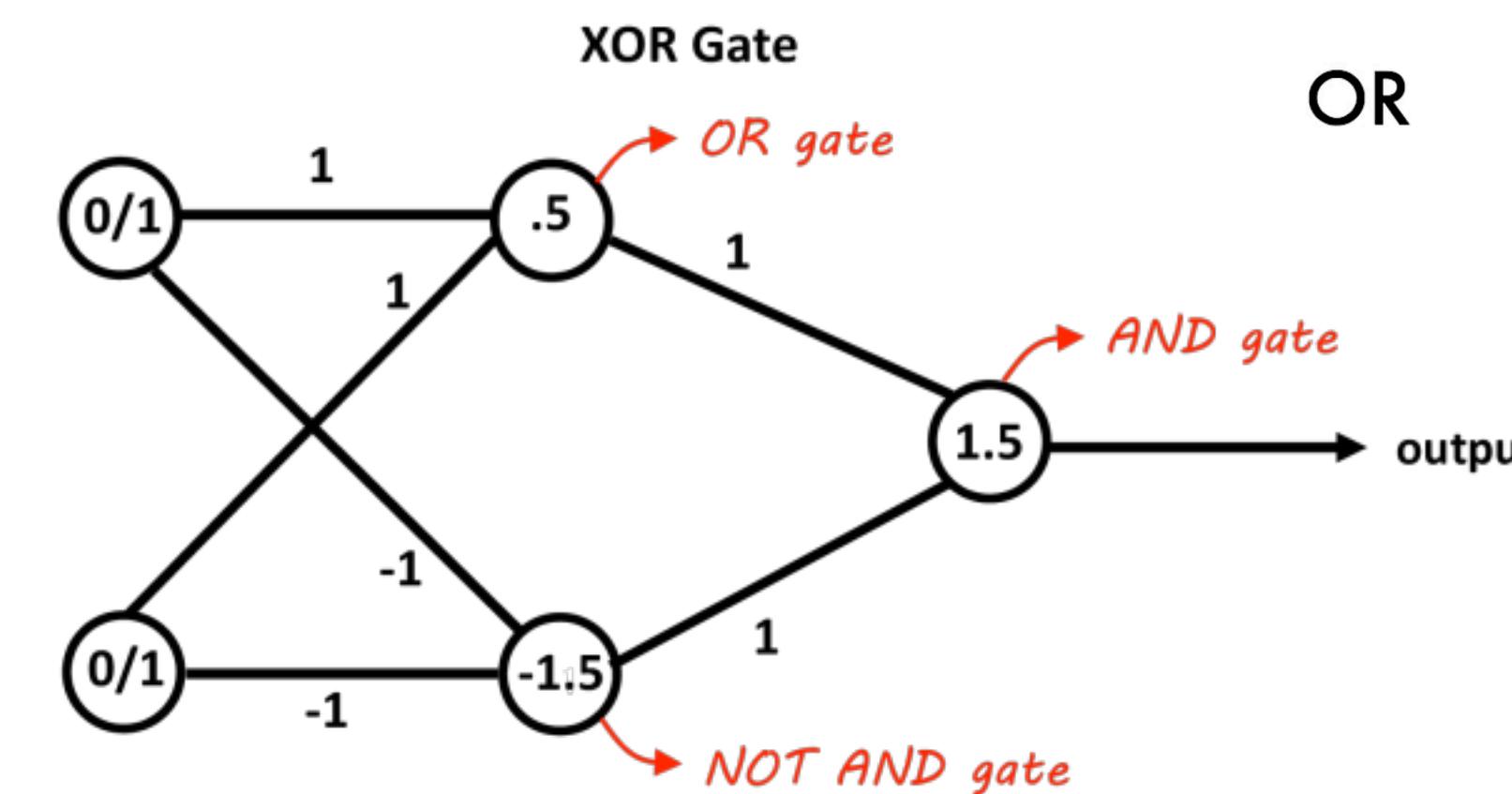
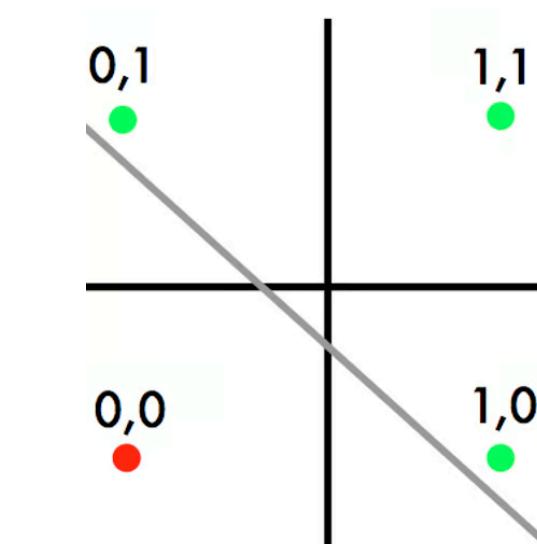
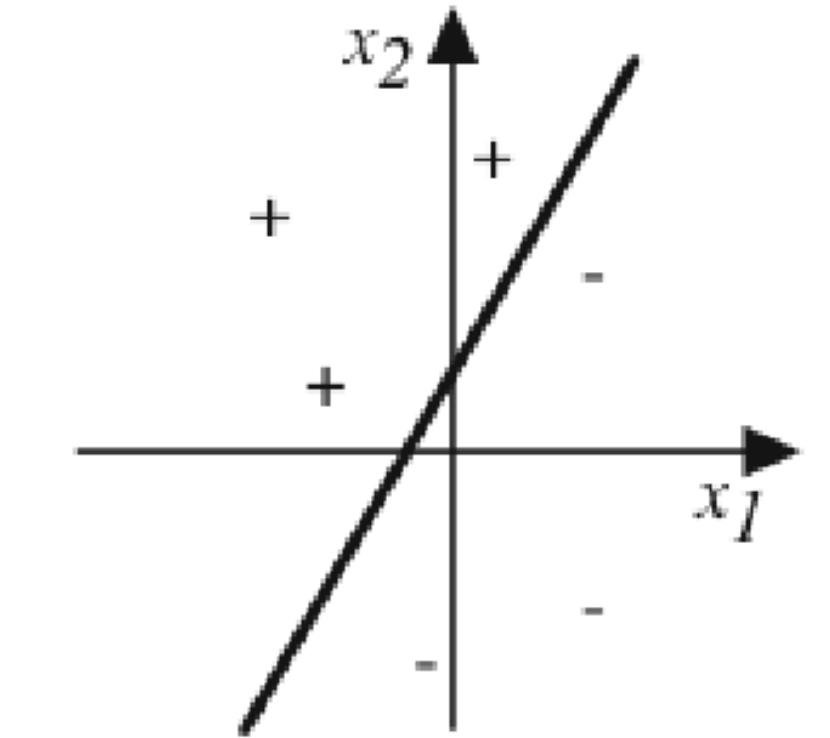
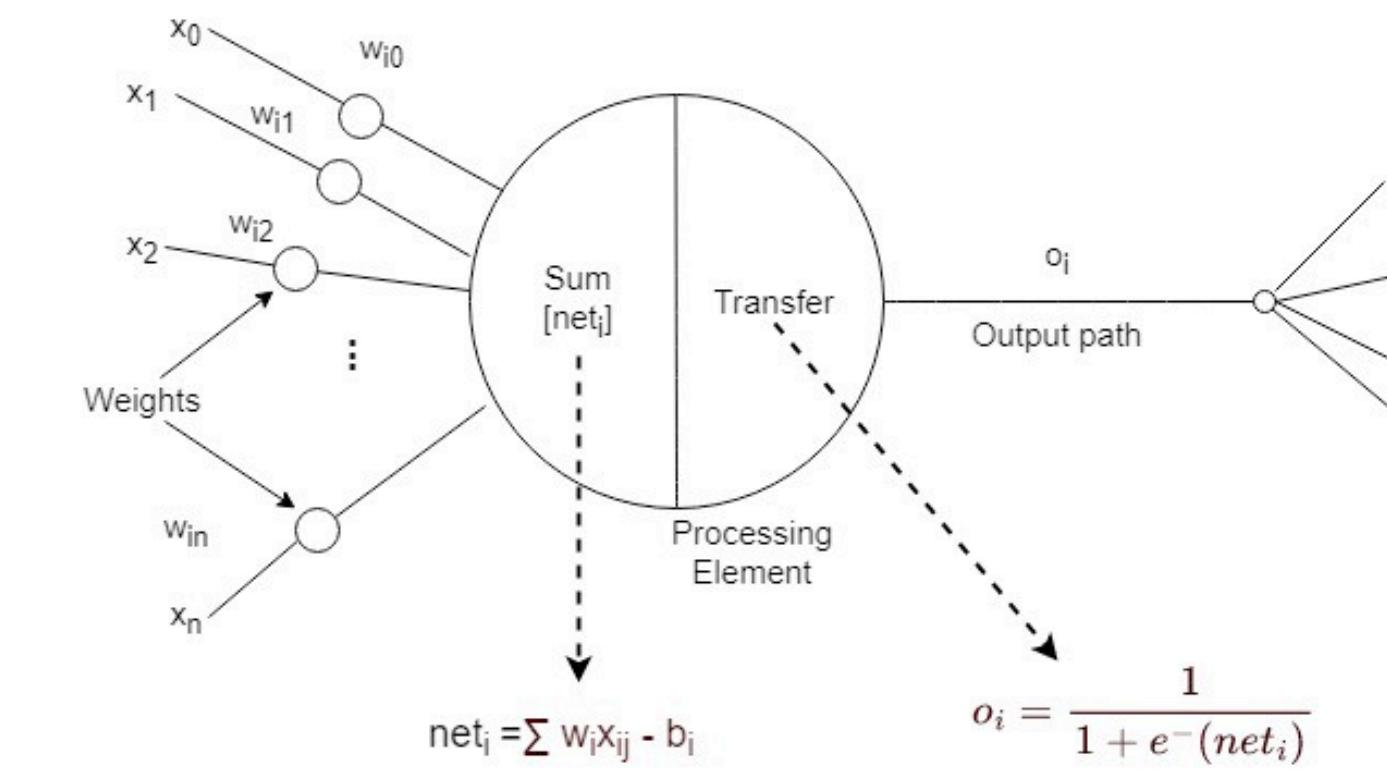
Neurone

- Entrée et sortie
- Poids calculé pendant la formation
- Sommation
- Fonction d'activation
 - Sigmoid, relu, tanh



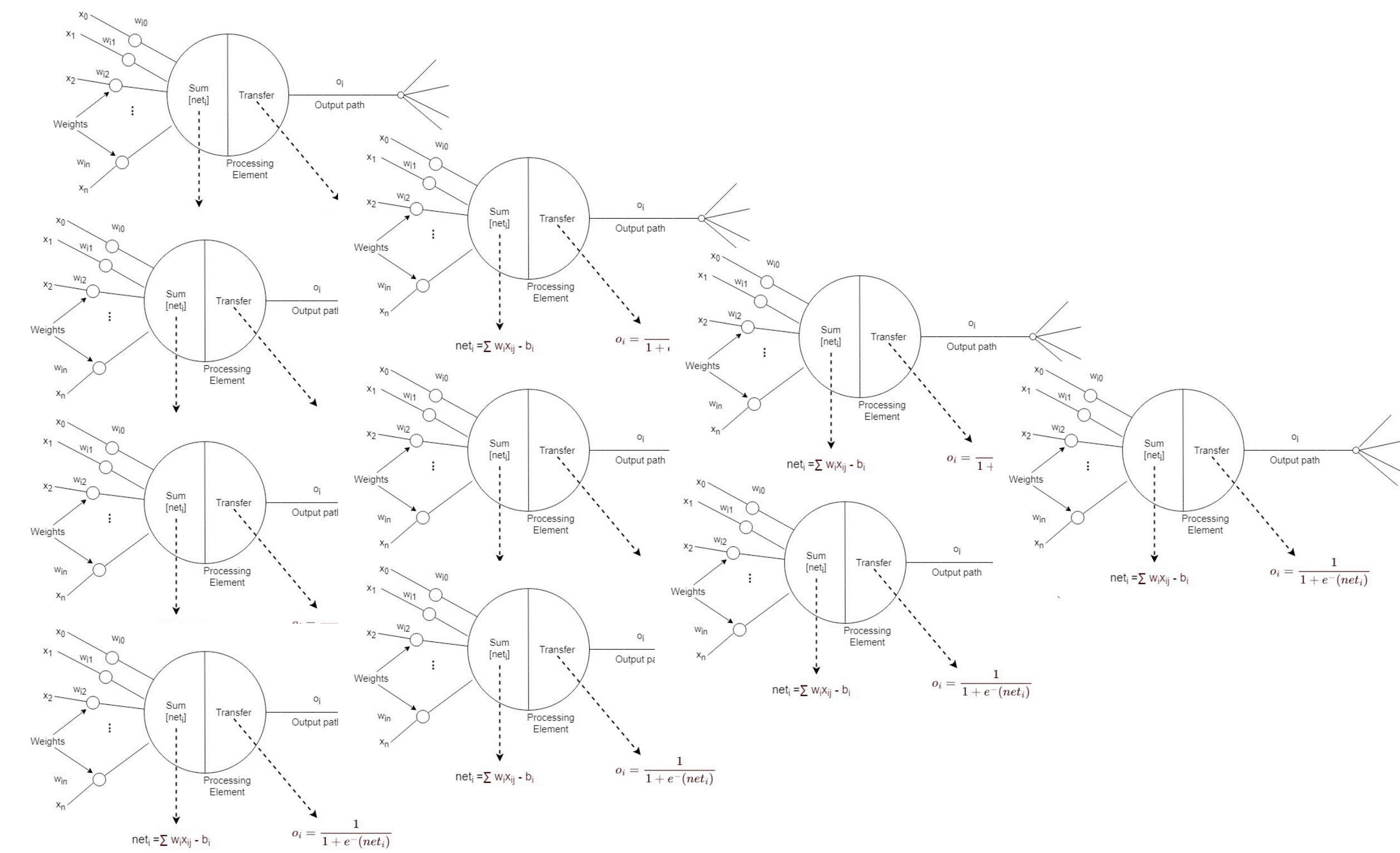
Perceptron

- Représente Limites linéaires
- Problème XOR
- Perceptron multicouche



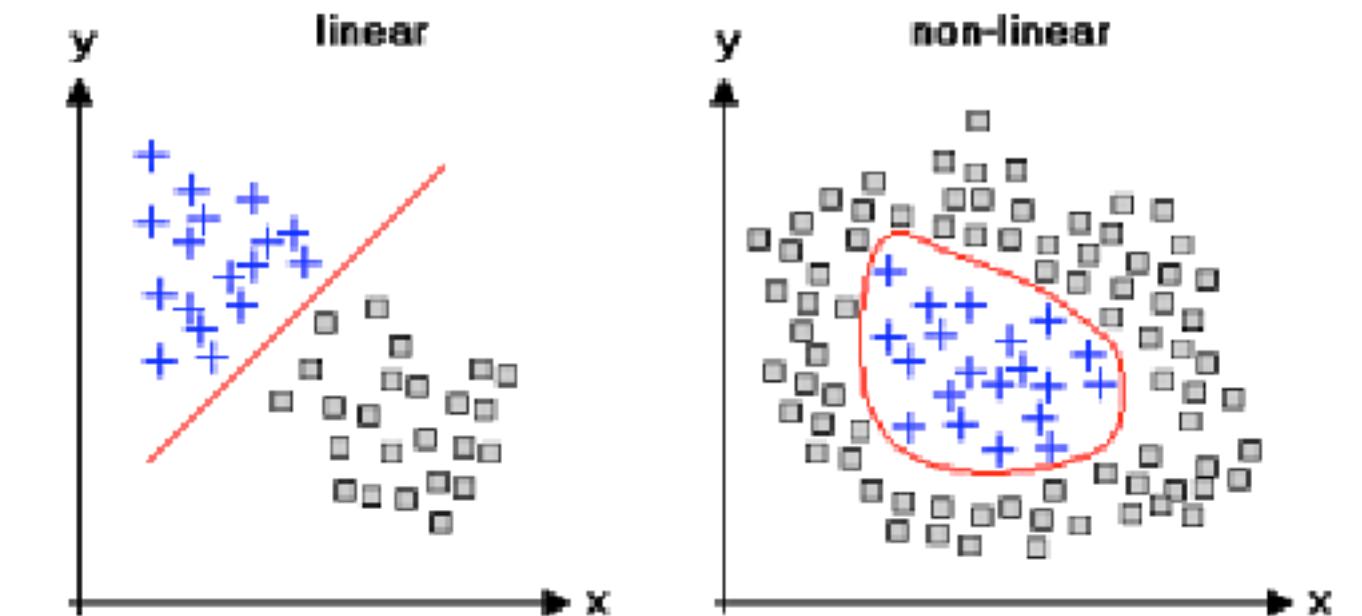
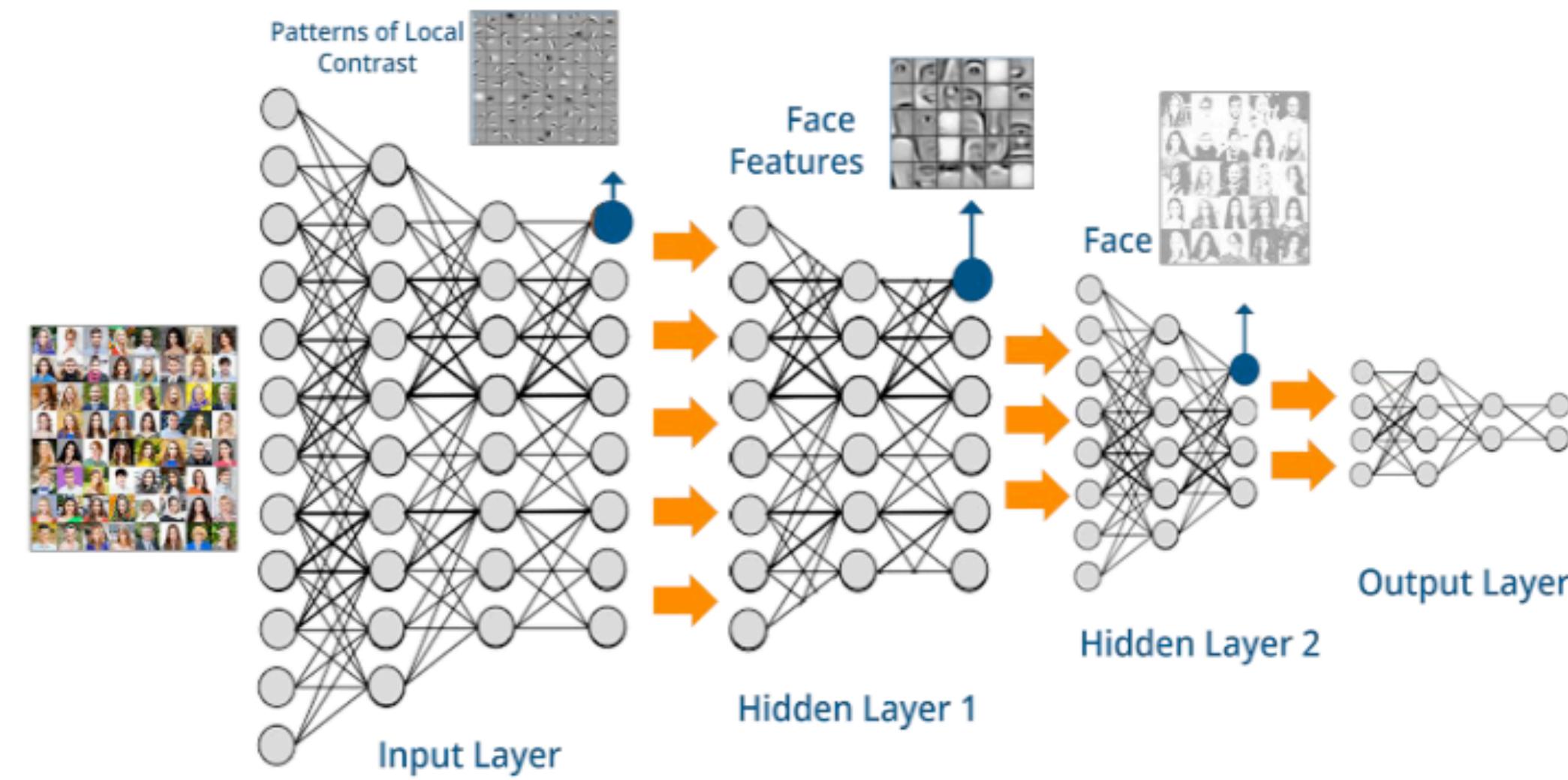
Multilayer perceptron

- Plusieurs neurone



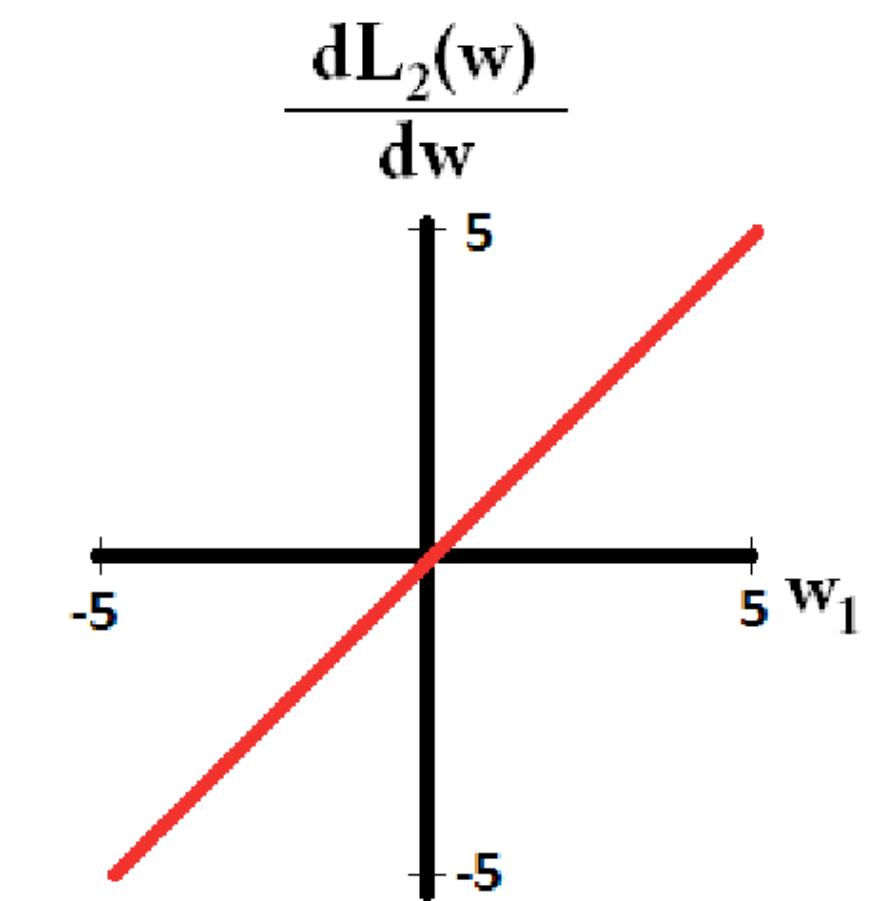
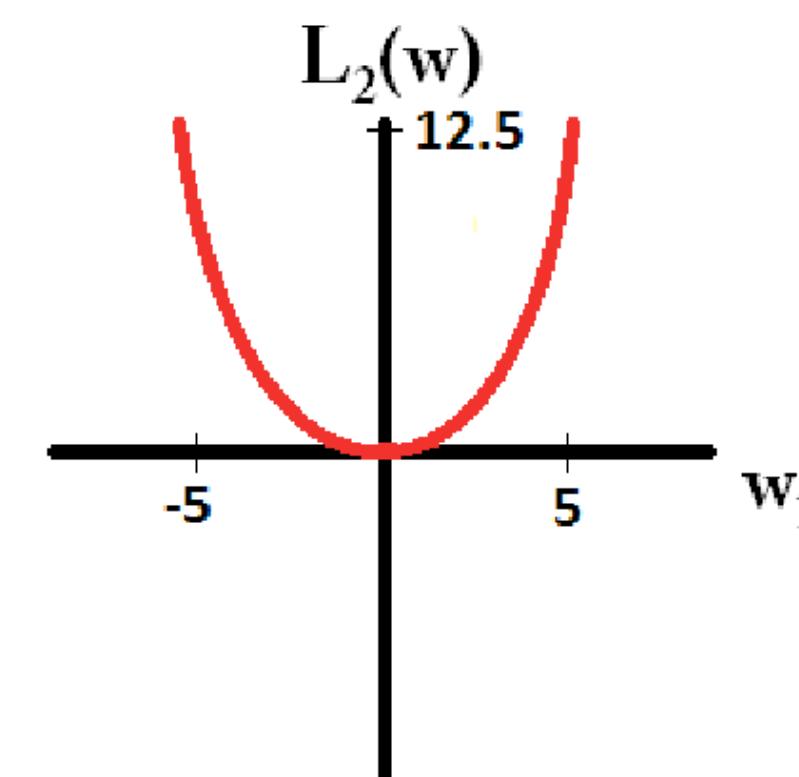
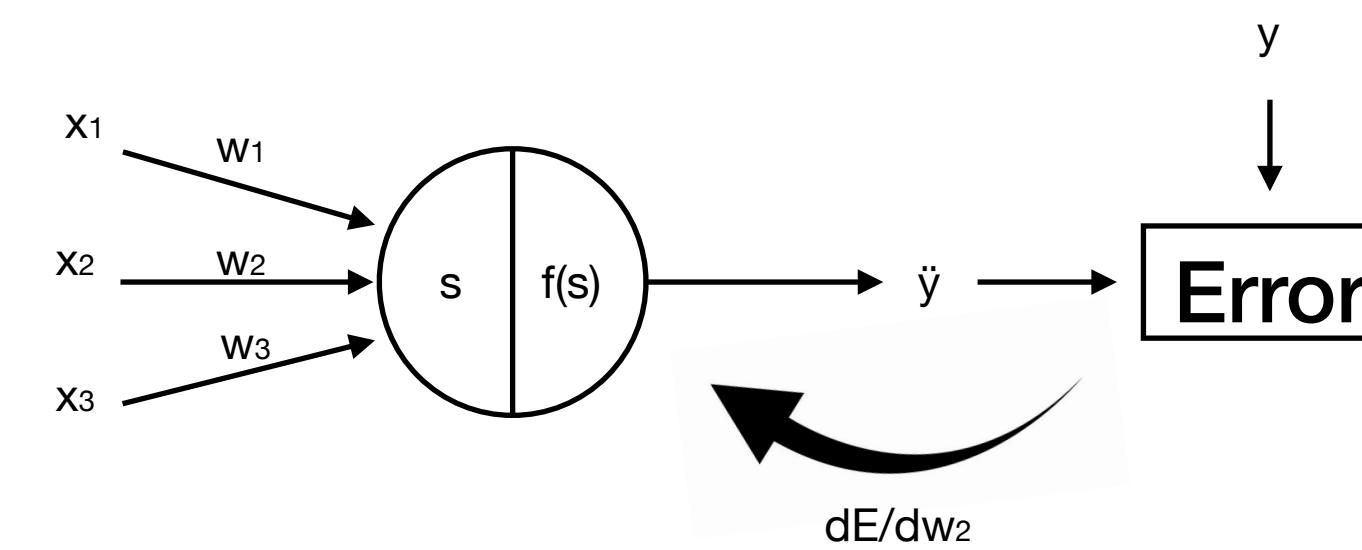
Multilayer perceptron

- Plusieurs neurone
- Caractéristiques
- Couches
- Modèle
- Formation pour plusieurs tâches



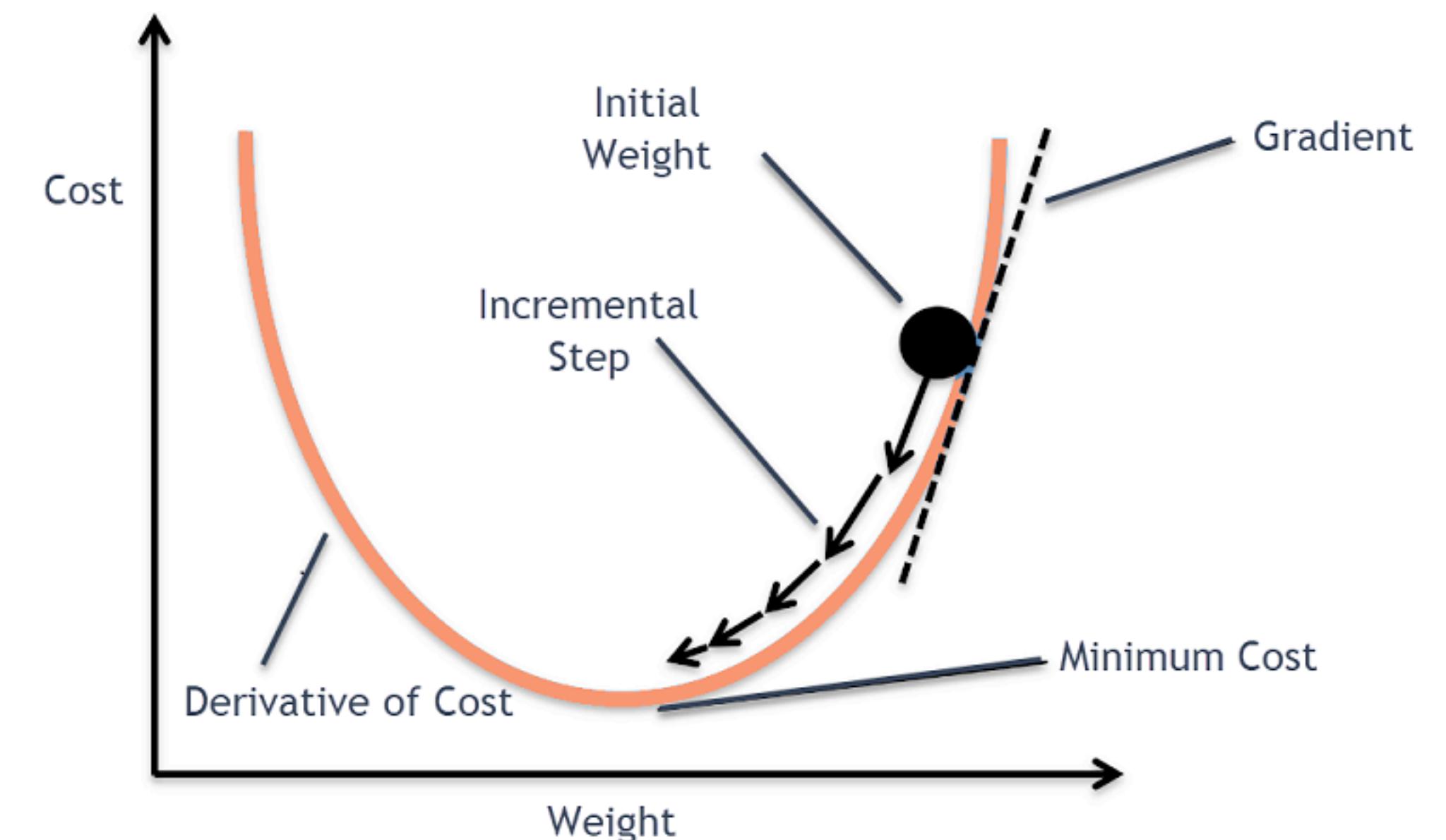
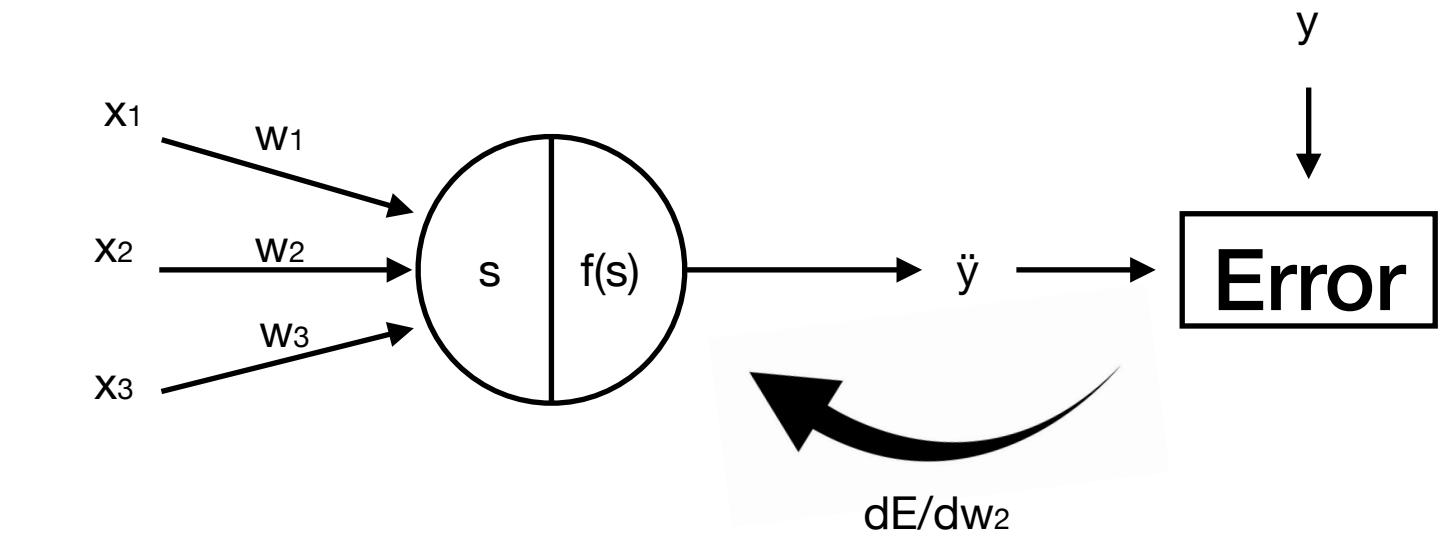
Formation

- Erreur entre vrais et prédit
 - Vrais: y
 - Prédi: \hat{y}
- Erreur L2
- Gradient (dérivé)
 - Impact du poids



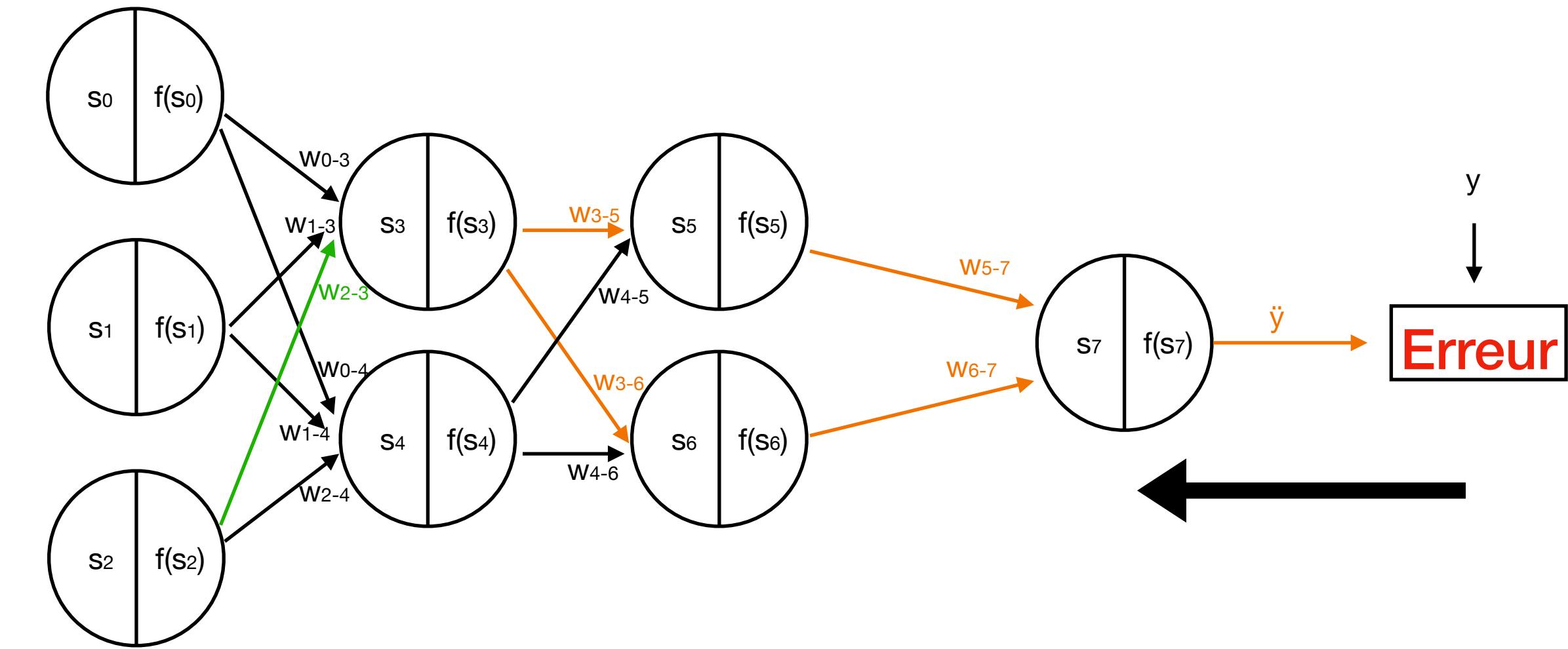
Gradient décent

- Erreur minimal
- Étapes incremental
- Gradient dirigent
- Modifie le poids à chaque étape
- Chaine



Règle de la chaîne

- Dérivé partiel
- Fonction d'erreur
- Chaîne de l'erreur au poids



$$dE/dw_{2-3} =$$

- Automatic avec Pytorch

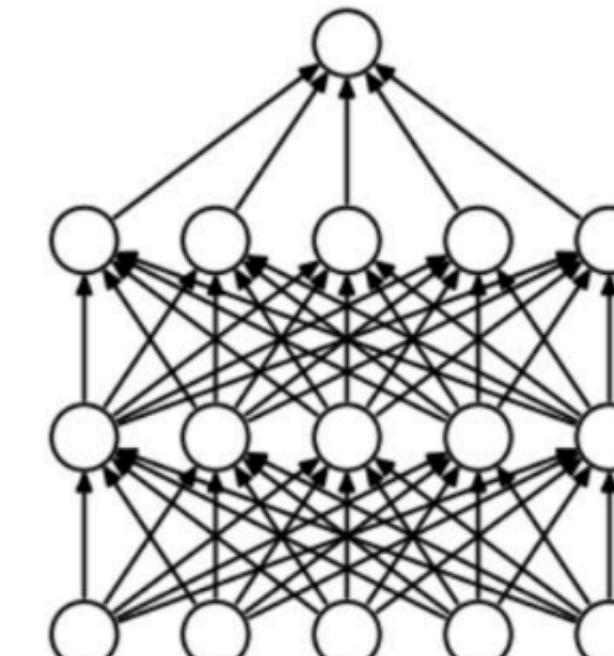
$$dE/df(s_7) \times df(s_7)/ds_7 \times$$

$$(ds_7/df(s_5) \times df(s_5)/ds_5 \times ds_5/df(s_3) + ds_7/df(s_6) \times df(s_6)/ds_6 \times ds_6/df(s_3)) \times$$

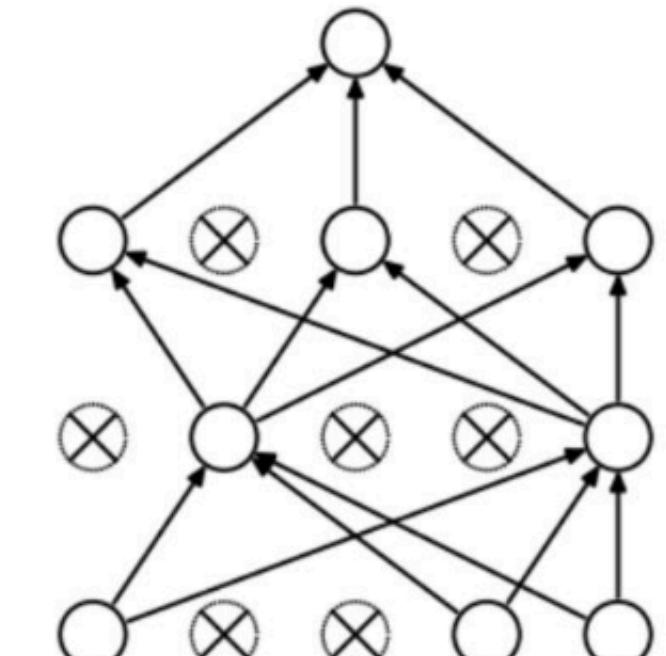
$$df(s_3)/ds_3 \times ds_3/dw_{2-3}$$

Régularisation Dropout

- Pas dépendre fortement d'un neurone particulier
- Sur-ajustement



(a) Standard Neural Net

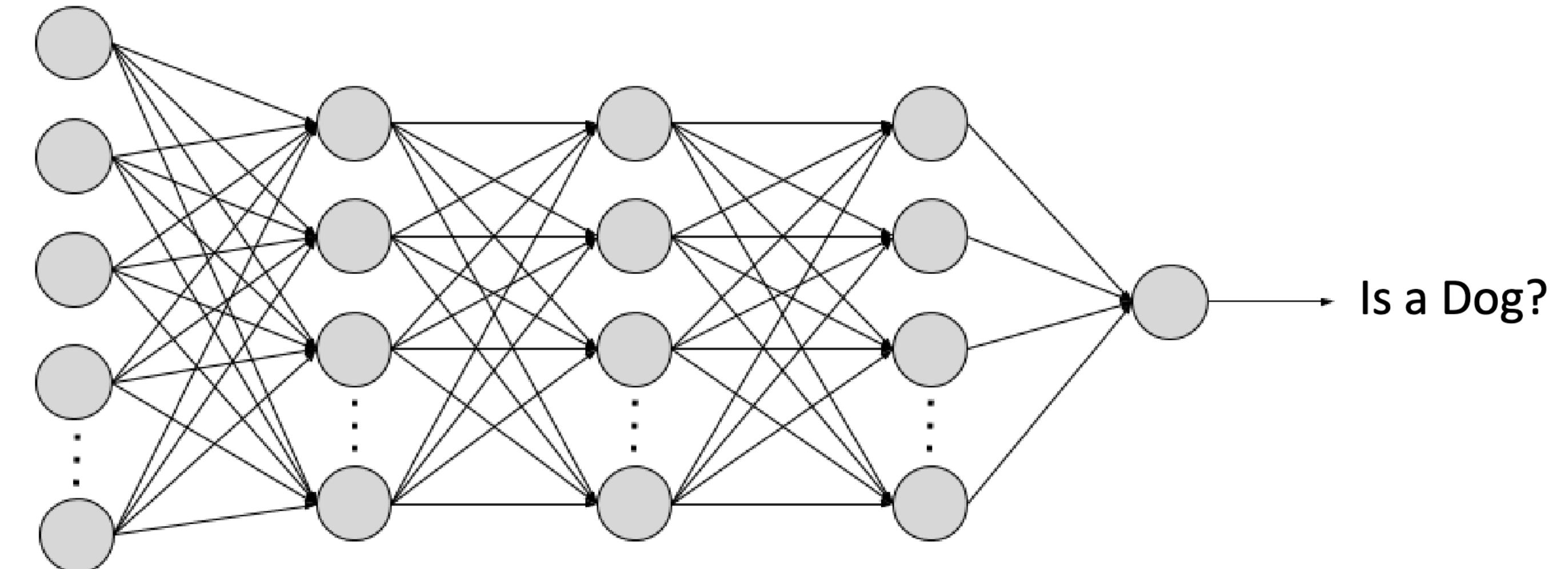
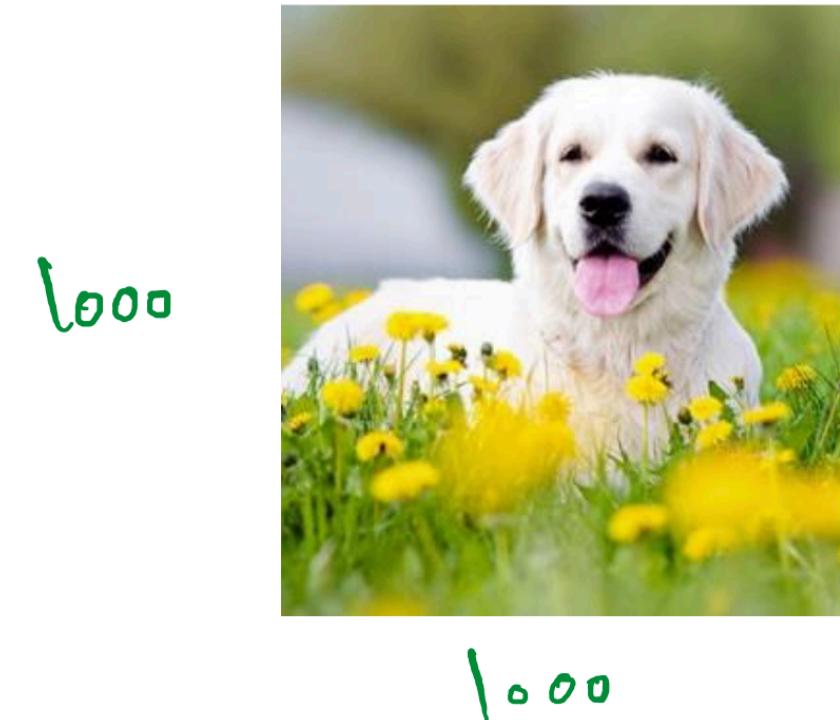


(b) After applying dropout.



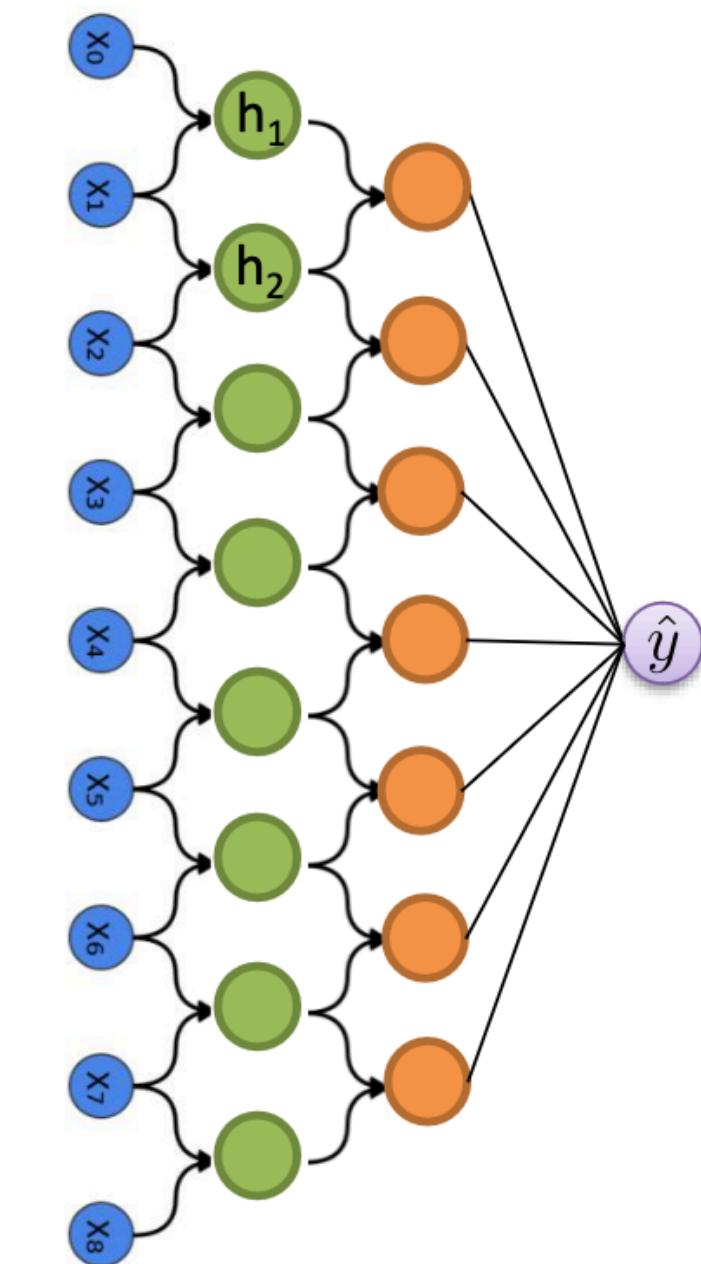
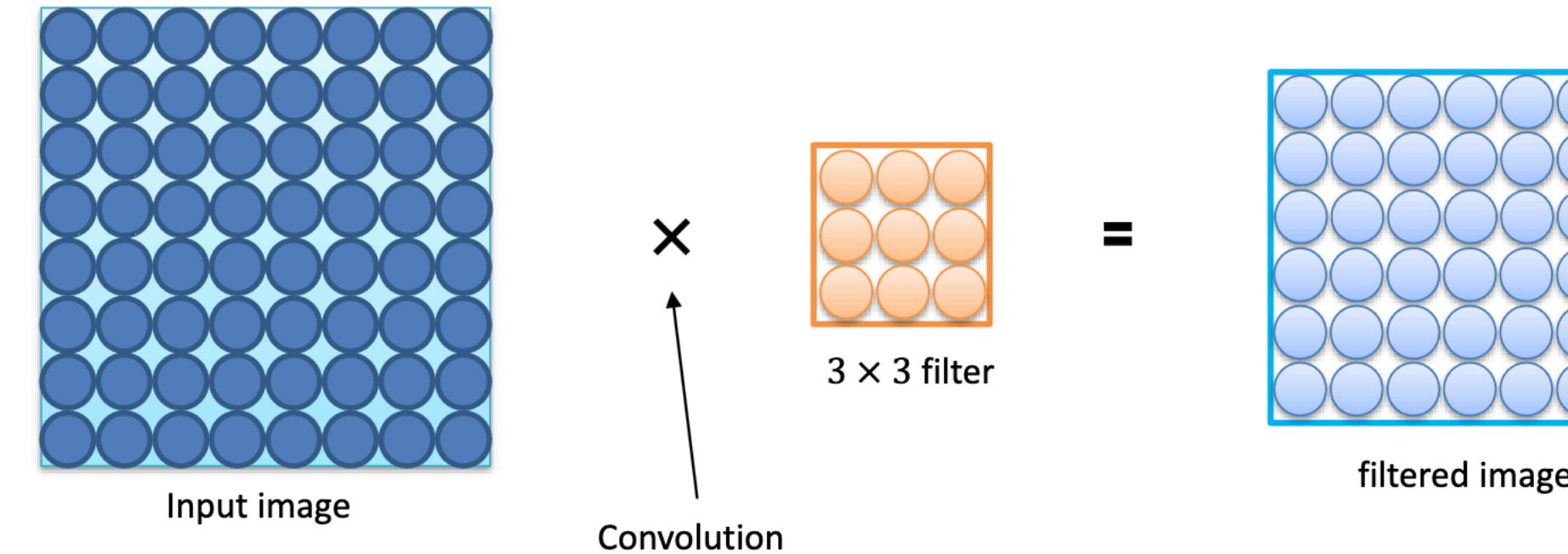
Réseau entièrement connecté sur les images

- Un neurone pour chaque pixel
- Trop de poids entre les couches
- 3 Milliard dans la première couche



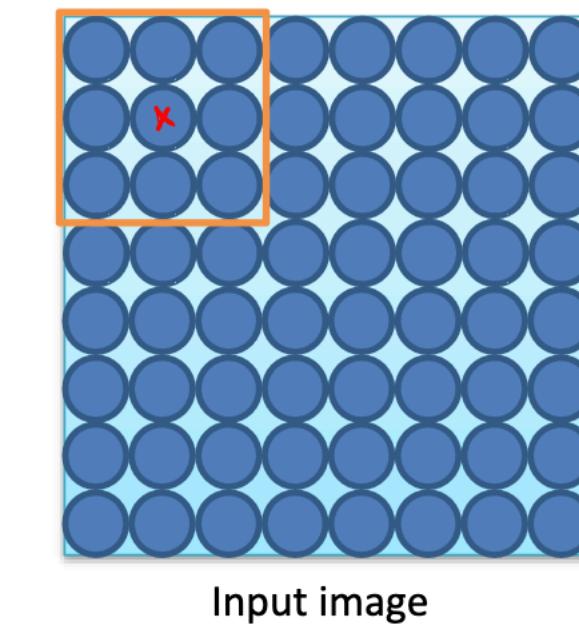
Convolutional neural network

- Connecter chaque unité cachée à un patch de neurones de la couche précédente
- Partager la matrice des paramètres
- Entrées locales

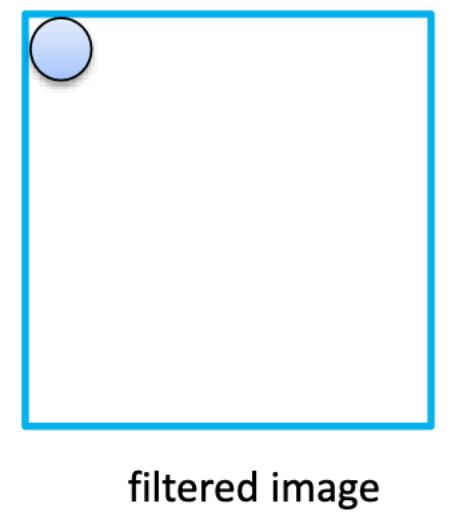
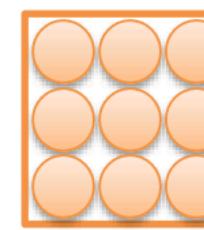


ConNet

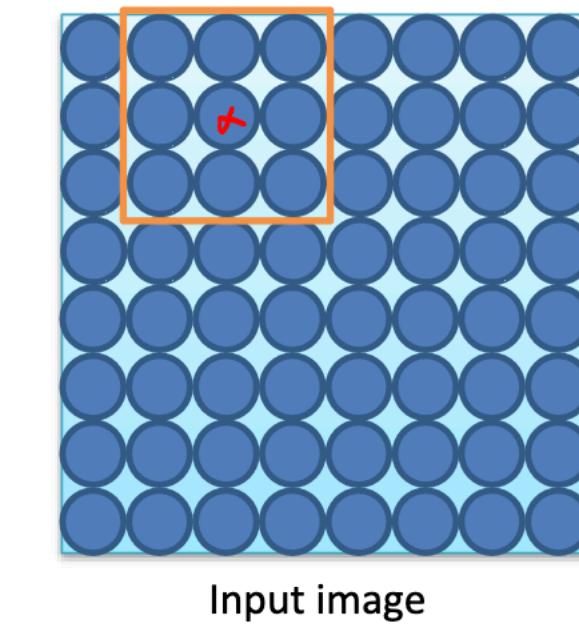
- Faites glisser le filtre
- Extraction des paramètres



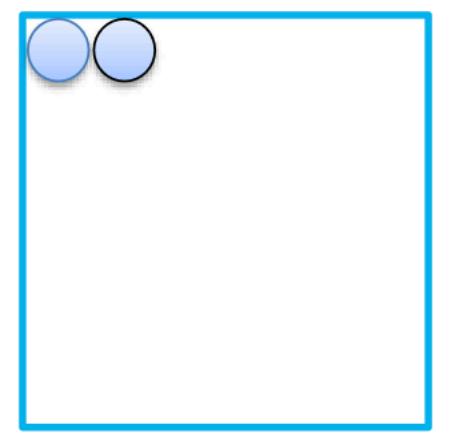
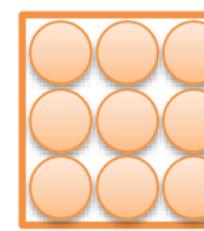
Convolution



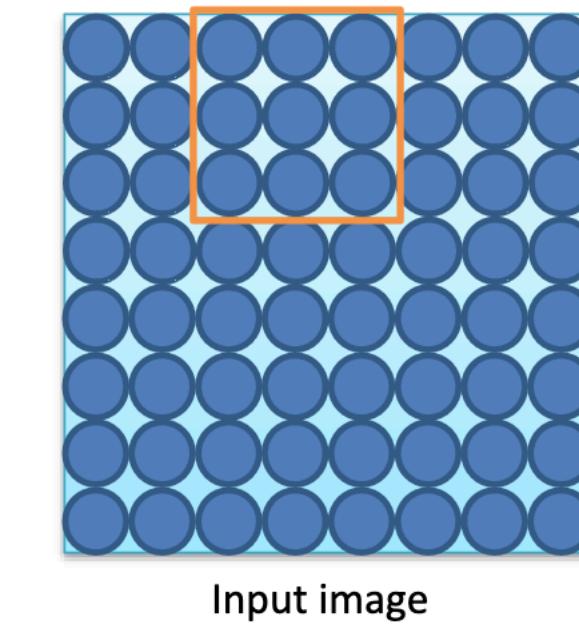
filtered image



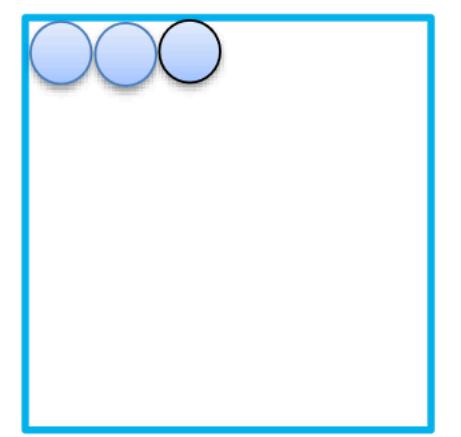
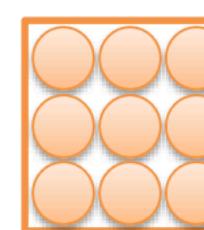
Convolution



filtered image



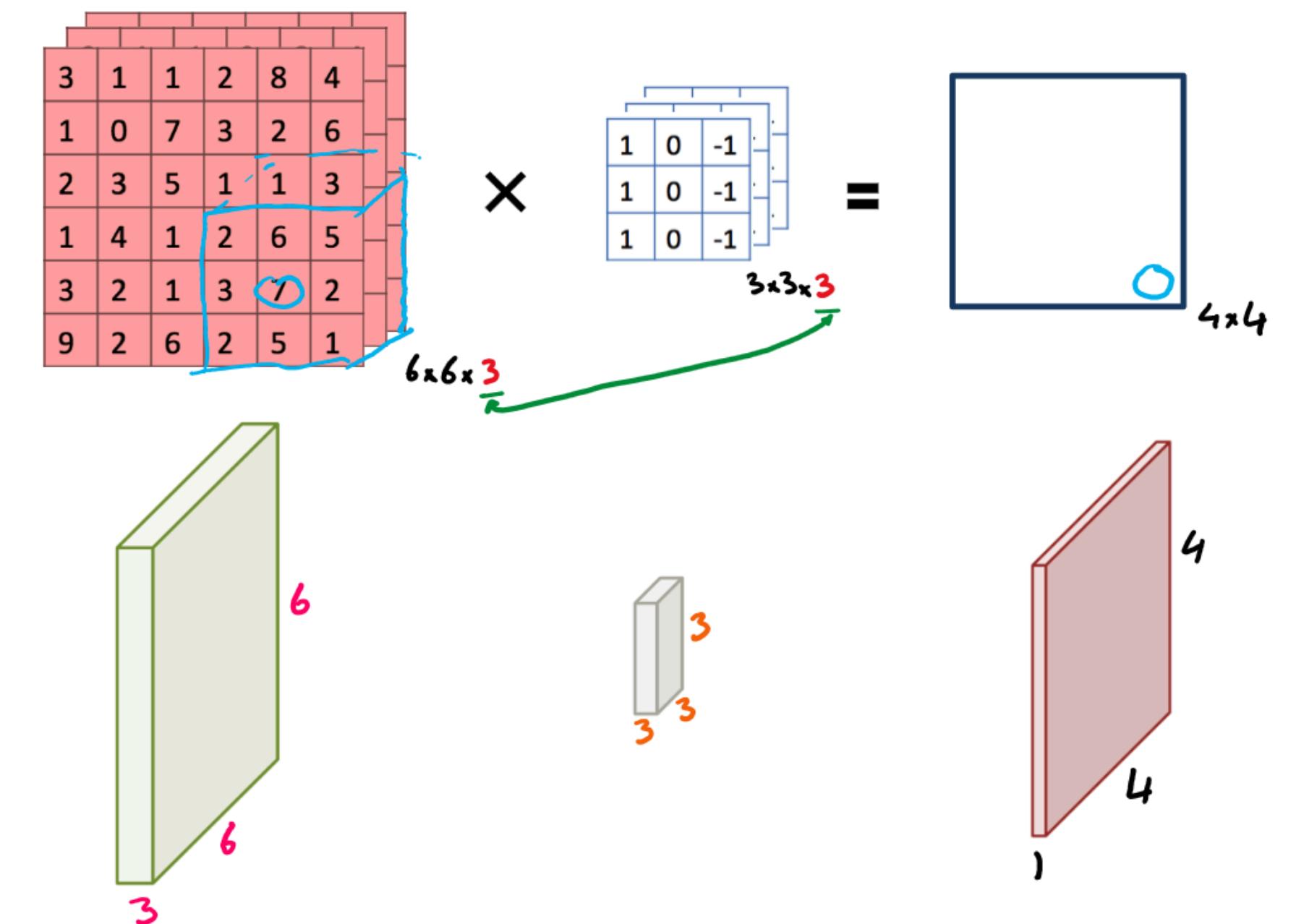
Convolution



filtered image

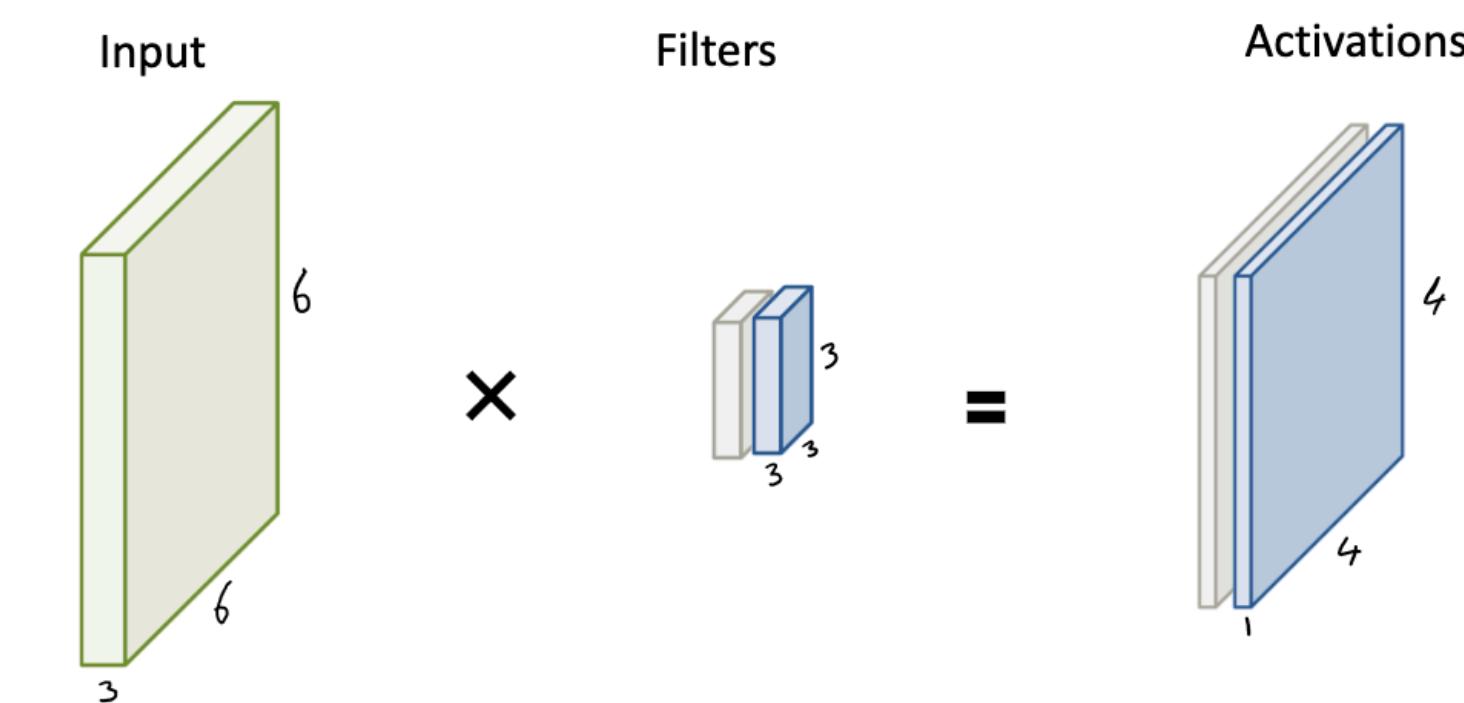
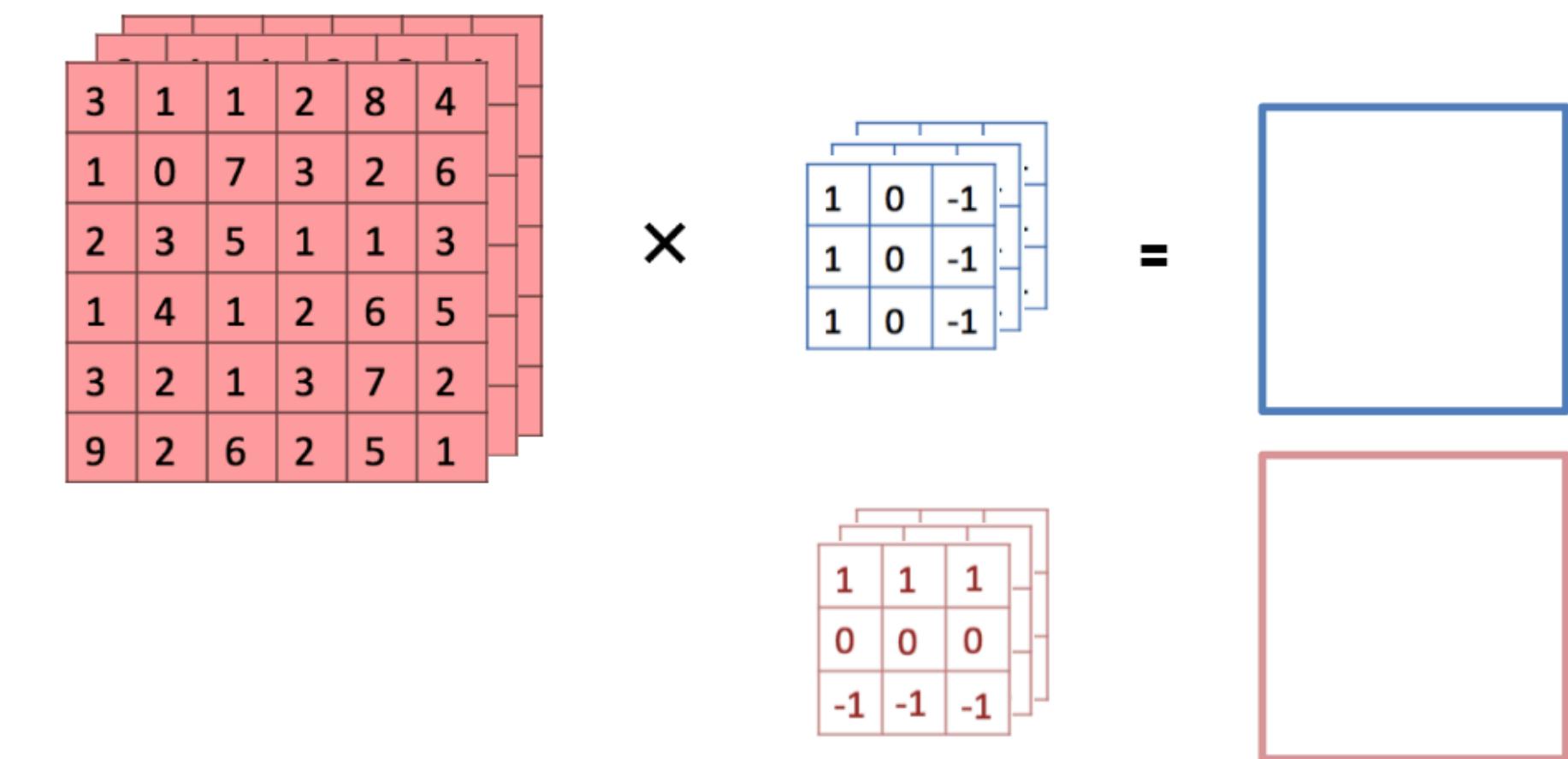
Convolution sur RGB

- Pas de glissement dans la direction du canal
- Résultat est une matrice 2D



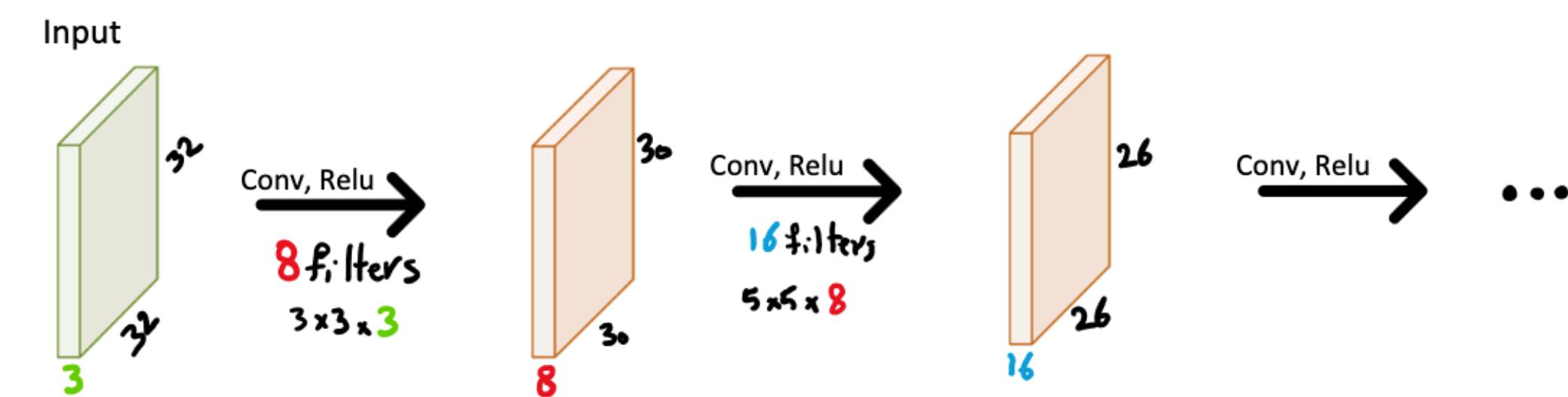
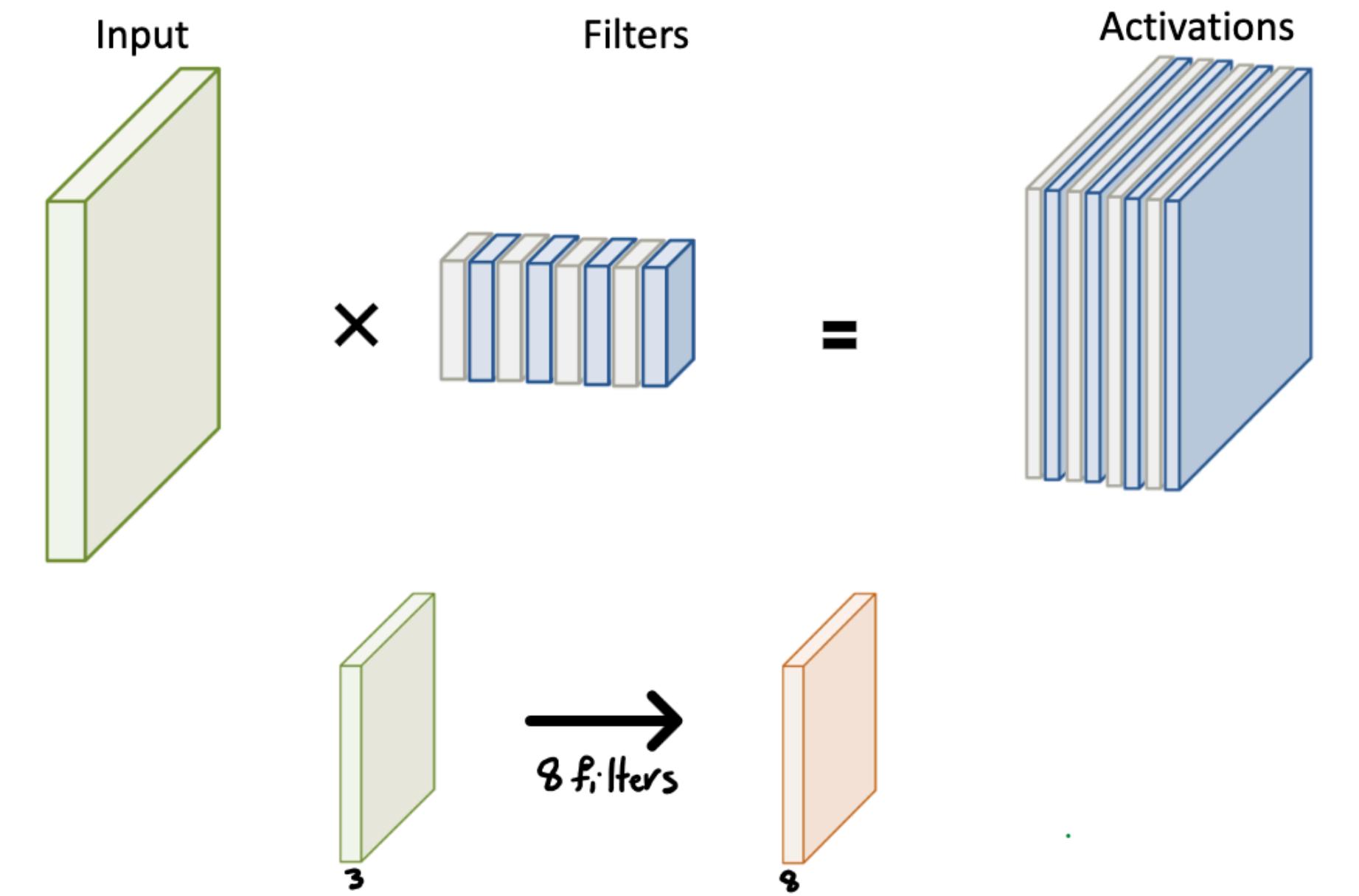
Plusieurs filtre

- Résultat augmenté
- Capture the propriété augment
- Filtre Prewitt



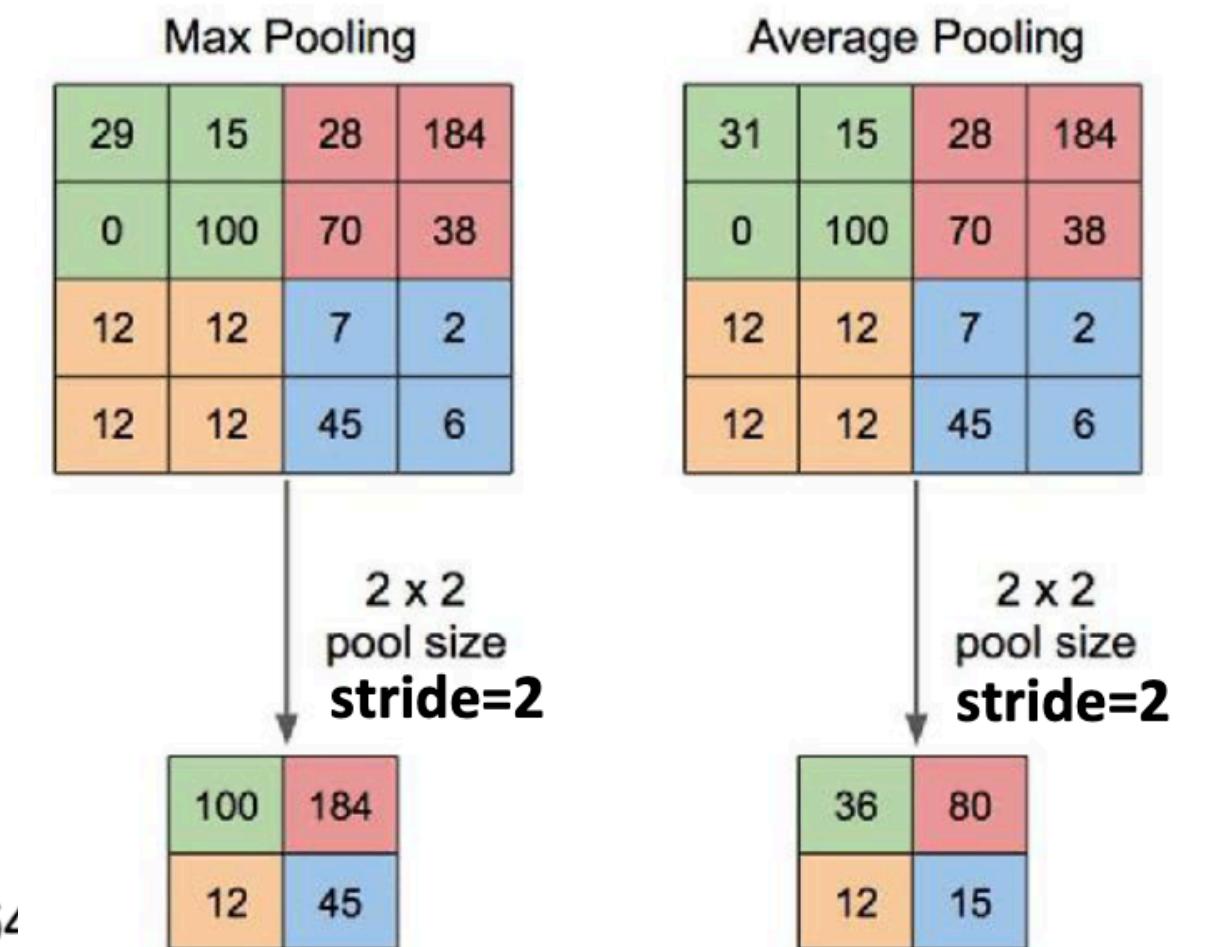
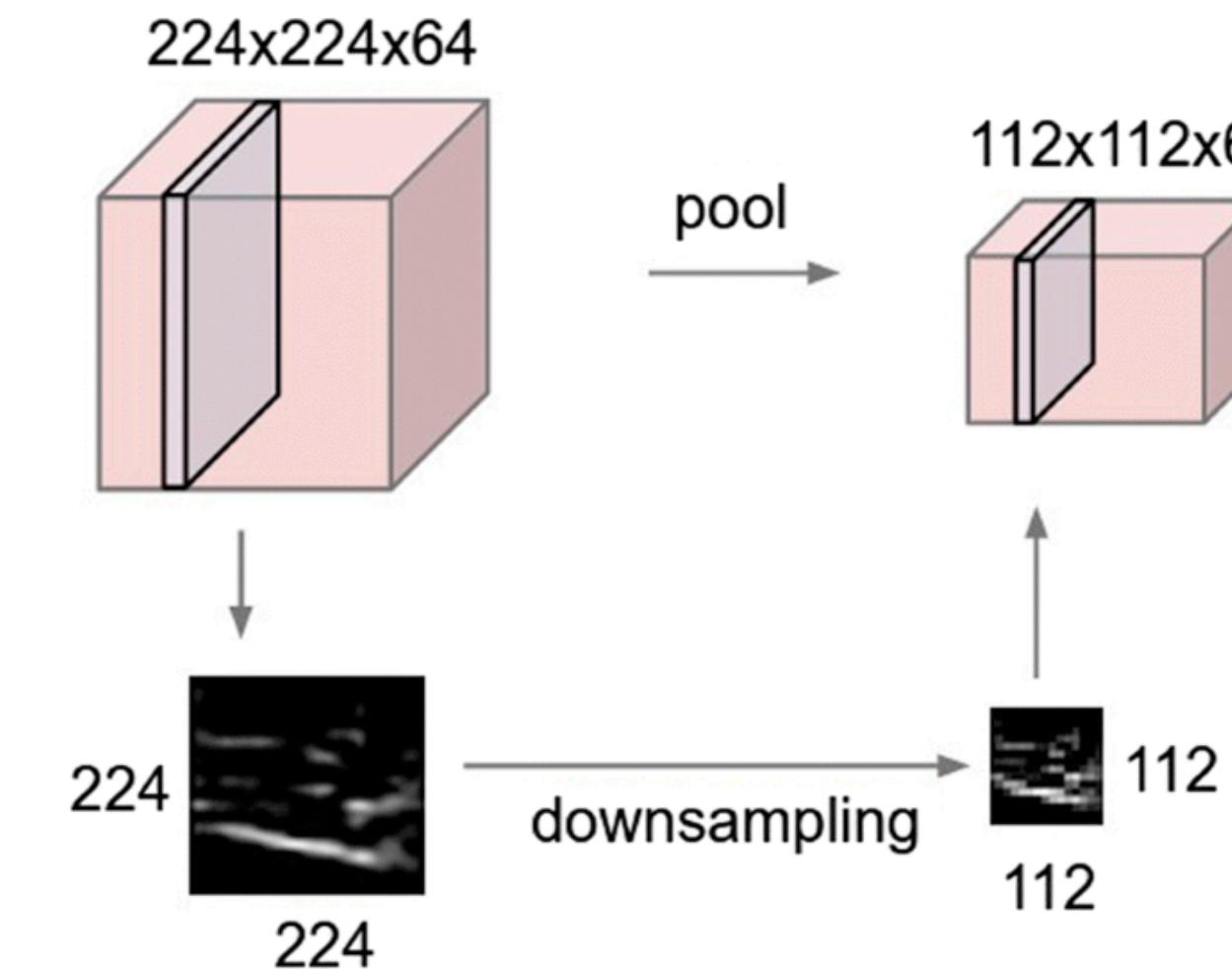
Plusieurs filtre

- Augmentation de propriété découvert

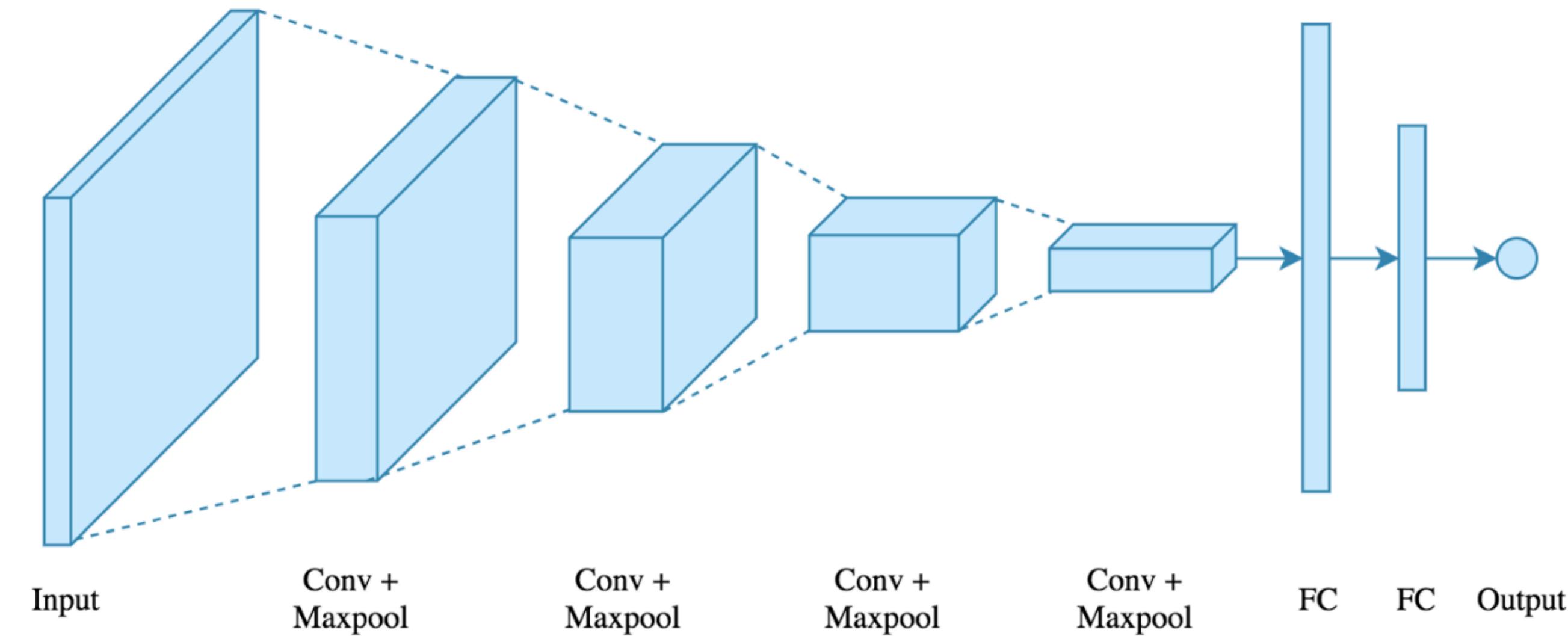


Pooling

- Une couche de pooling
- Effectue un échantillonnage descendant
- Ne modifie pas le nombre de canaux
- Pas de paramètres à apprendre

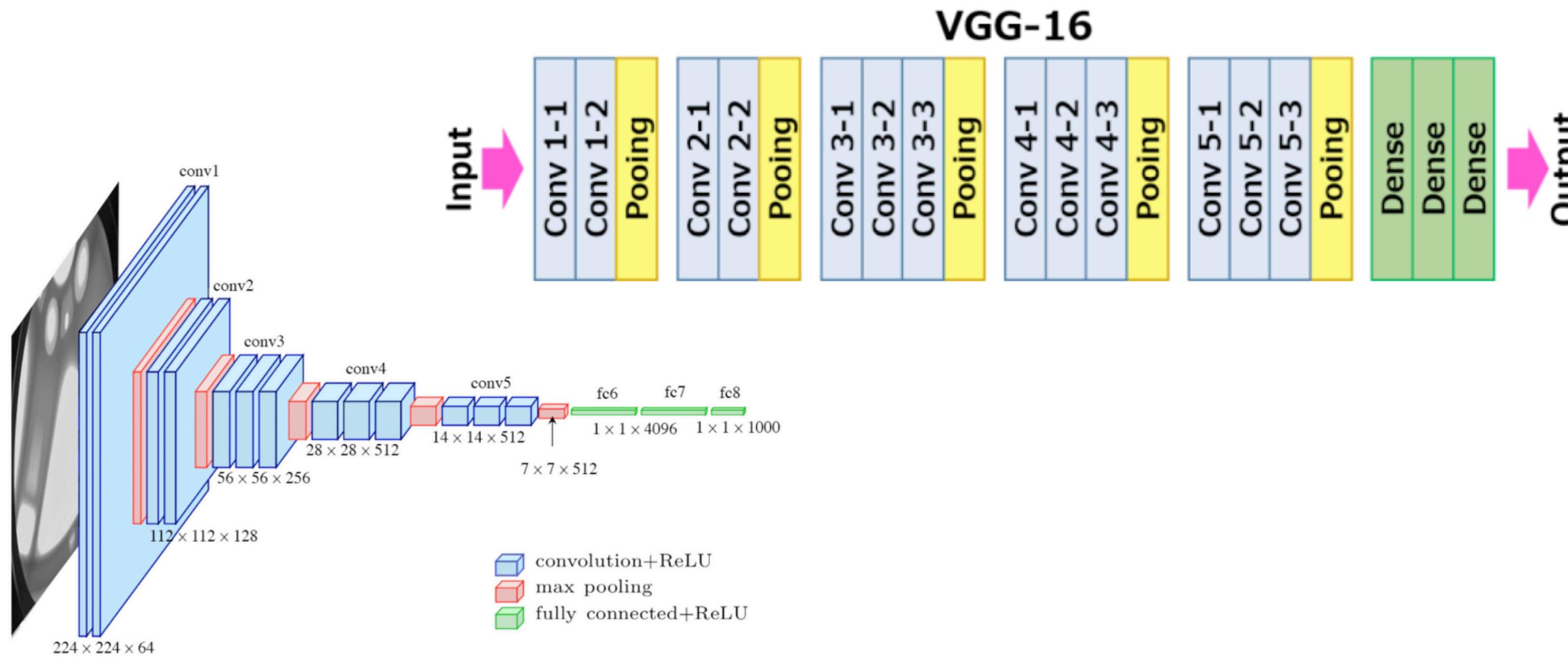
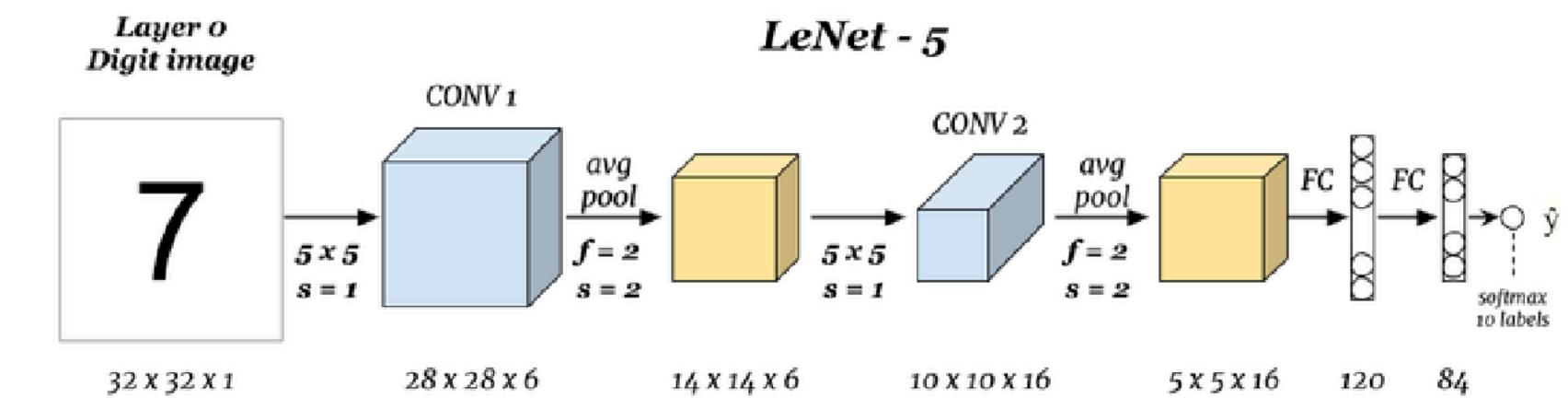
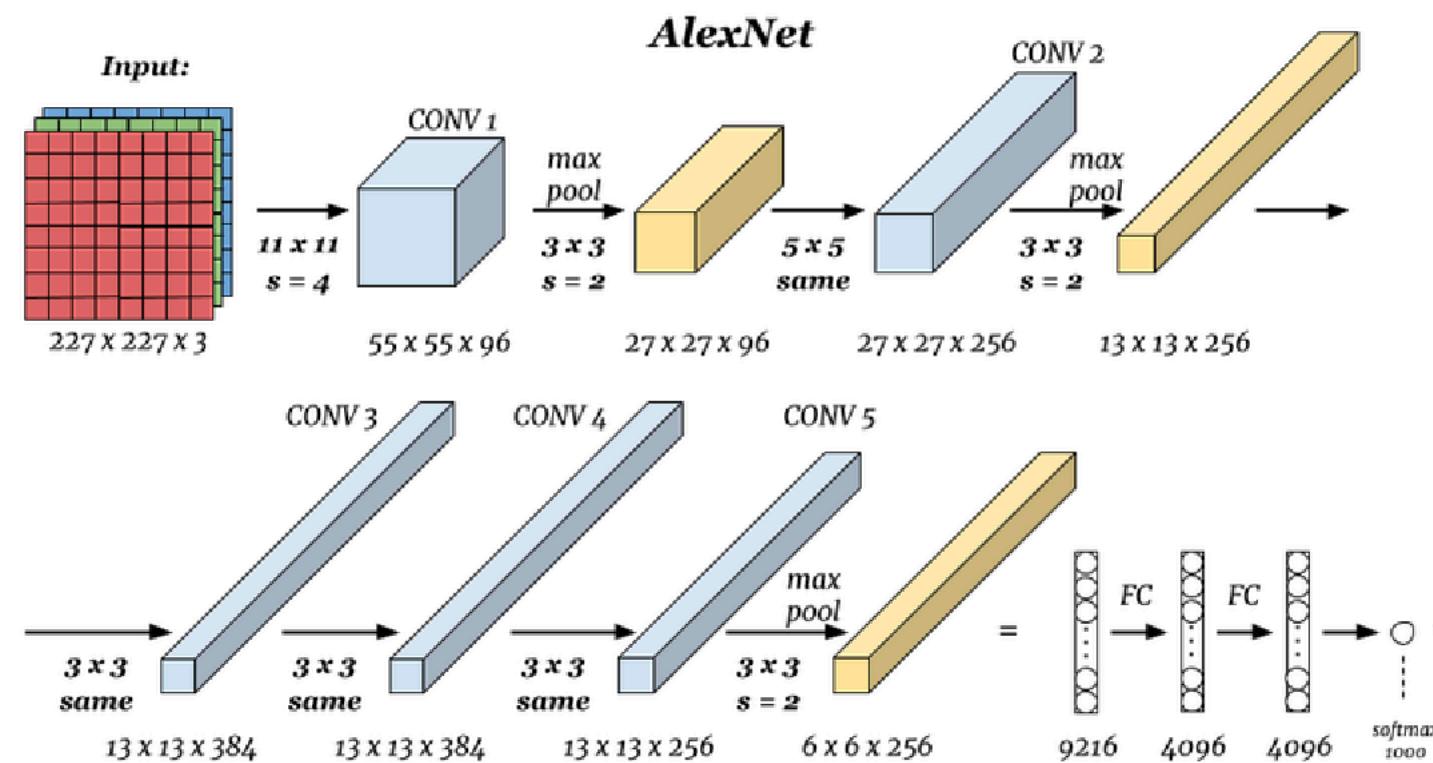


Example de ConvNet



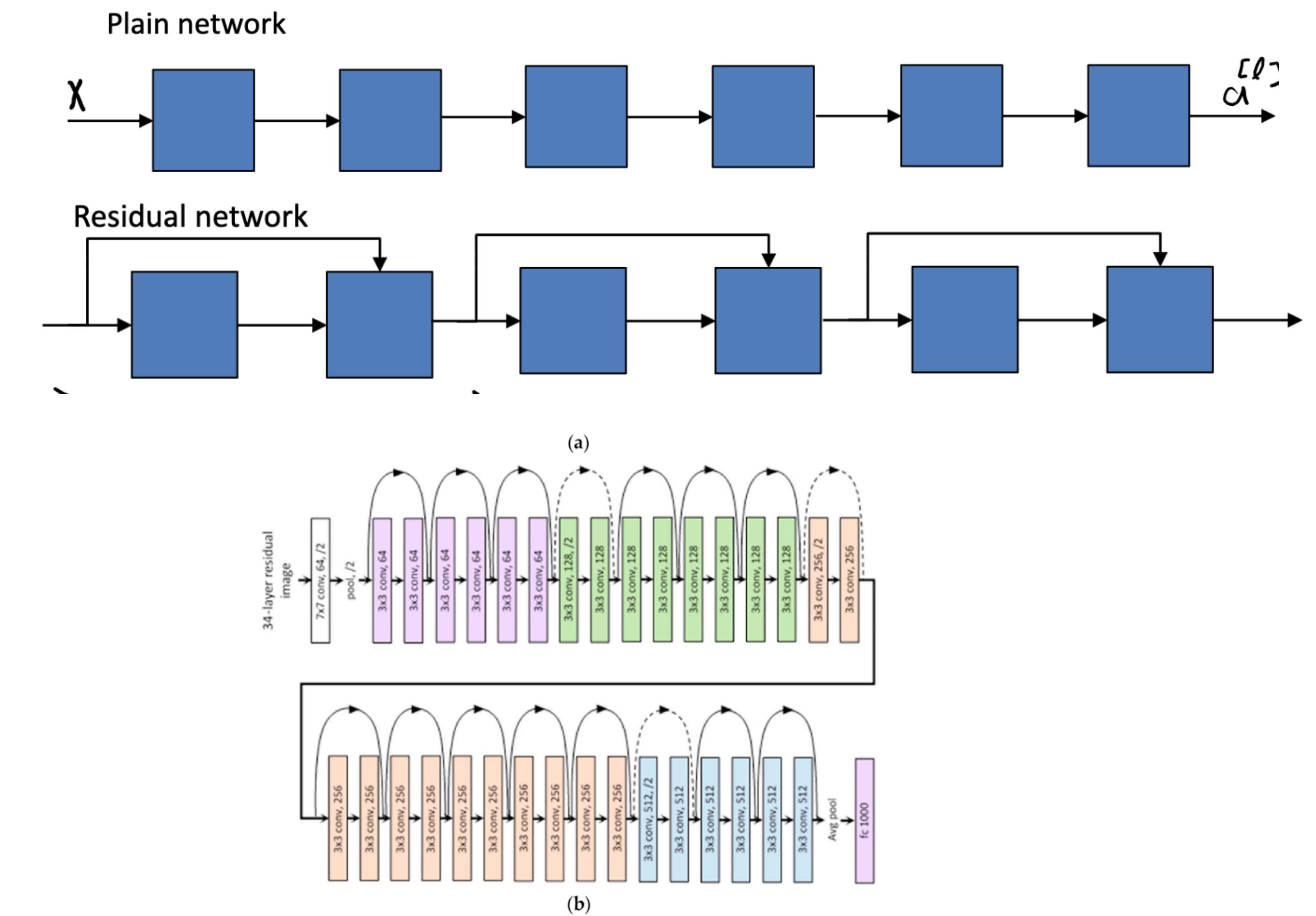
ConvNets

- LeNet
- AlexNet
- VGG

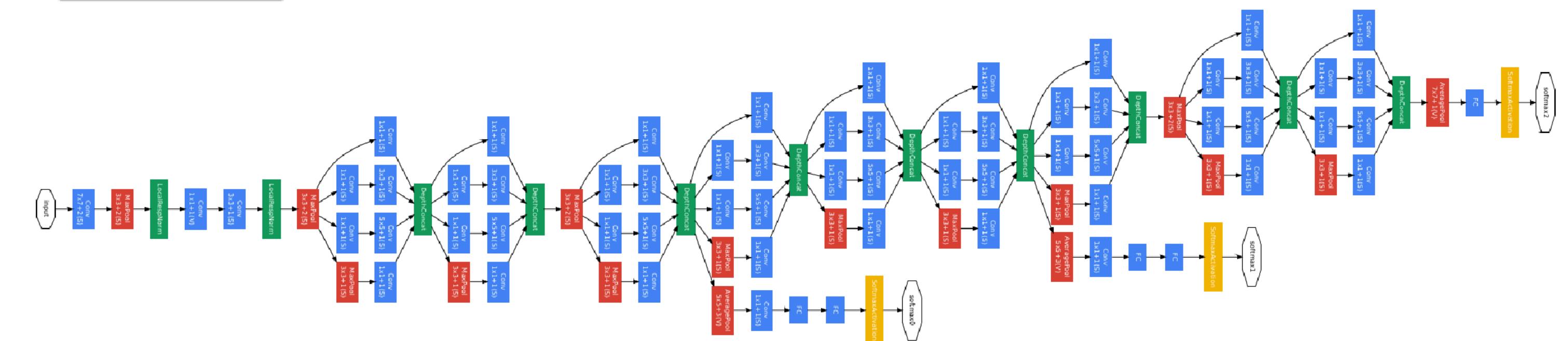
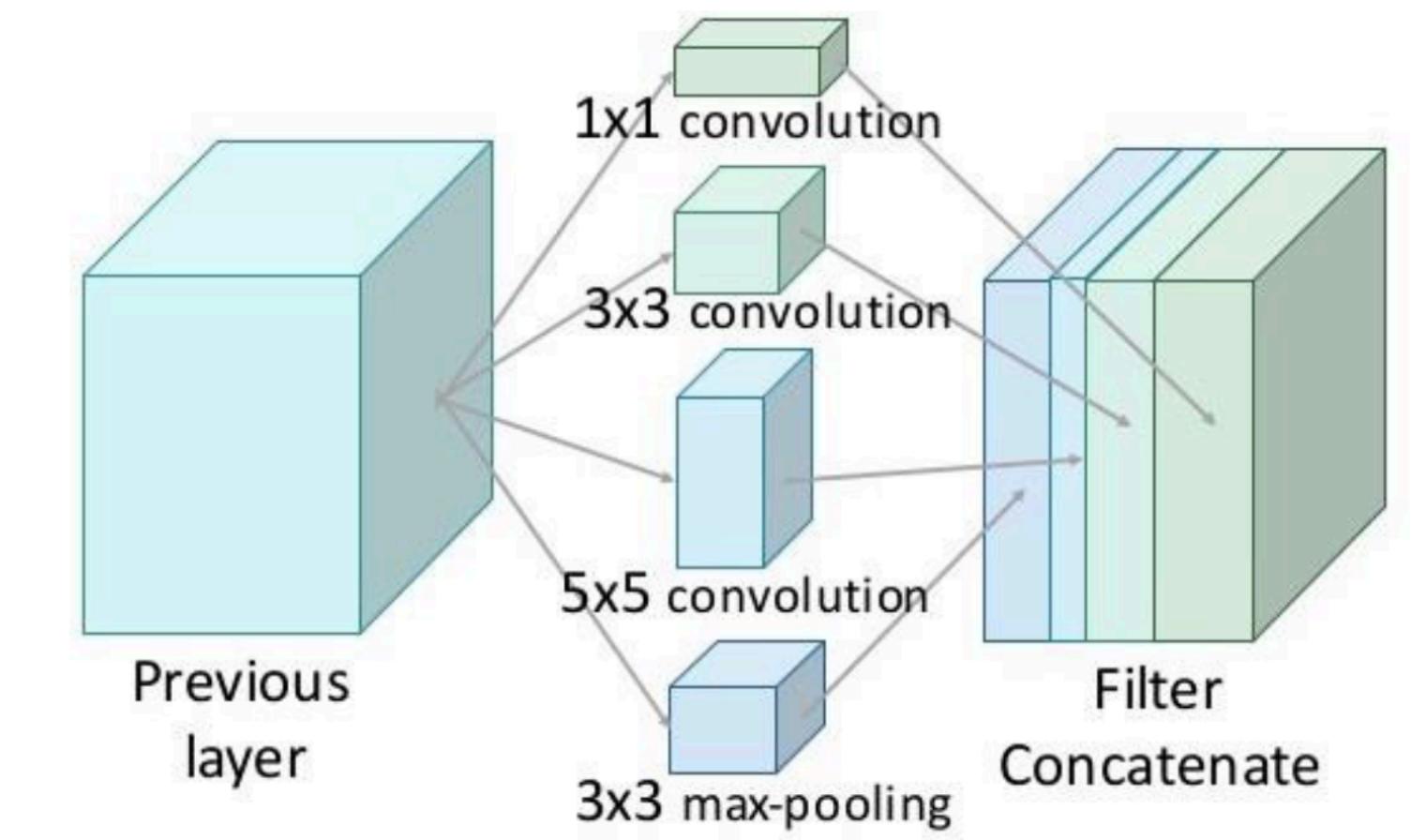
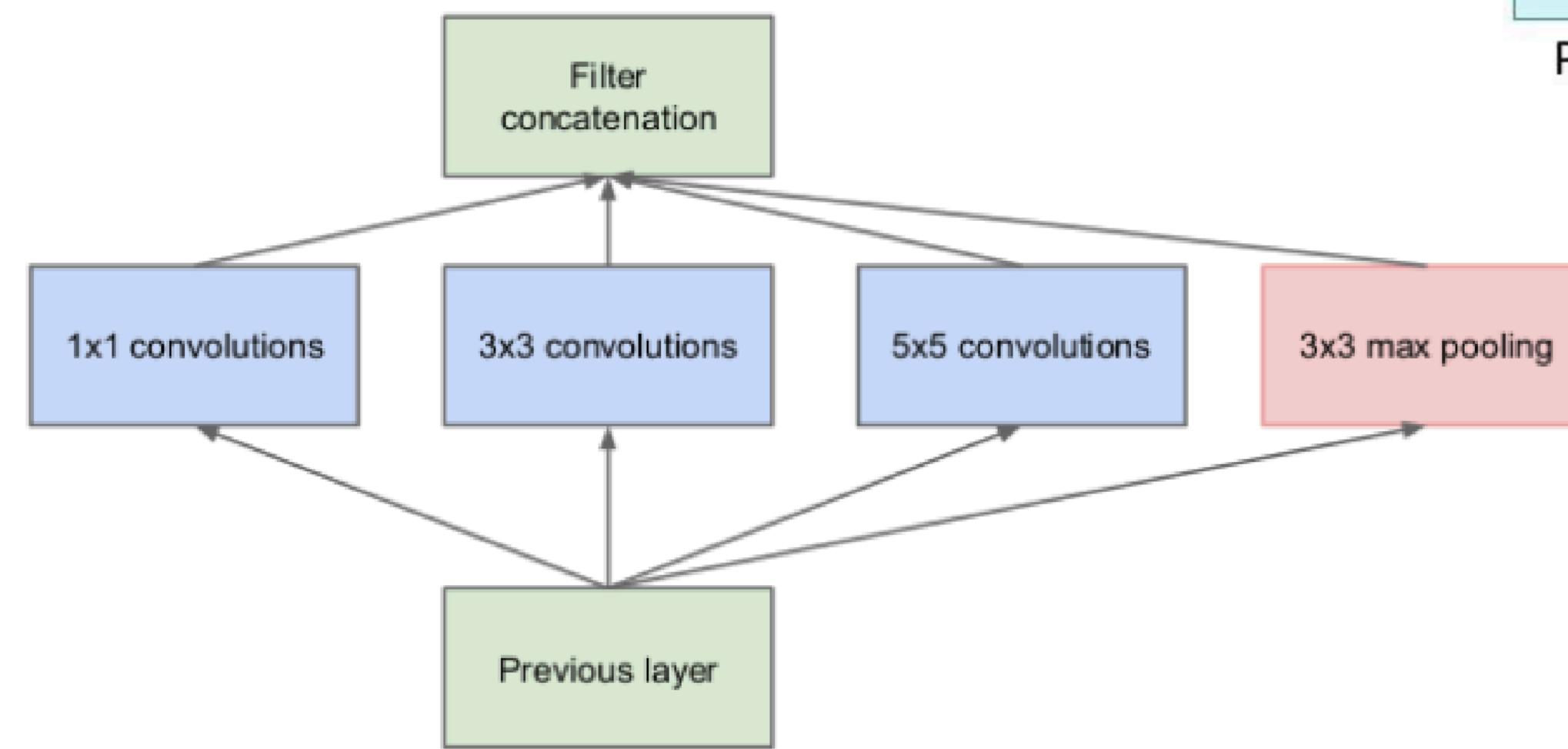


Blocs résiduels

- Réutilisation des fonctionnalités
- Stabiliser l'apprentissage
- Débit de gradient renforcé

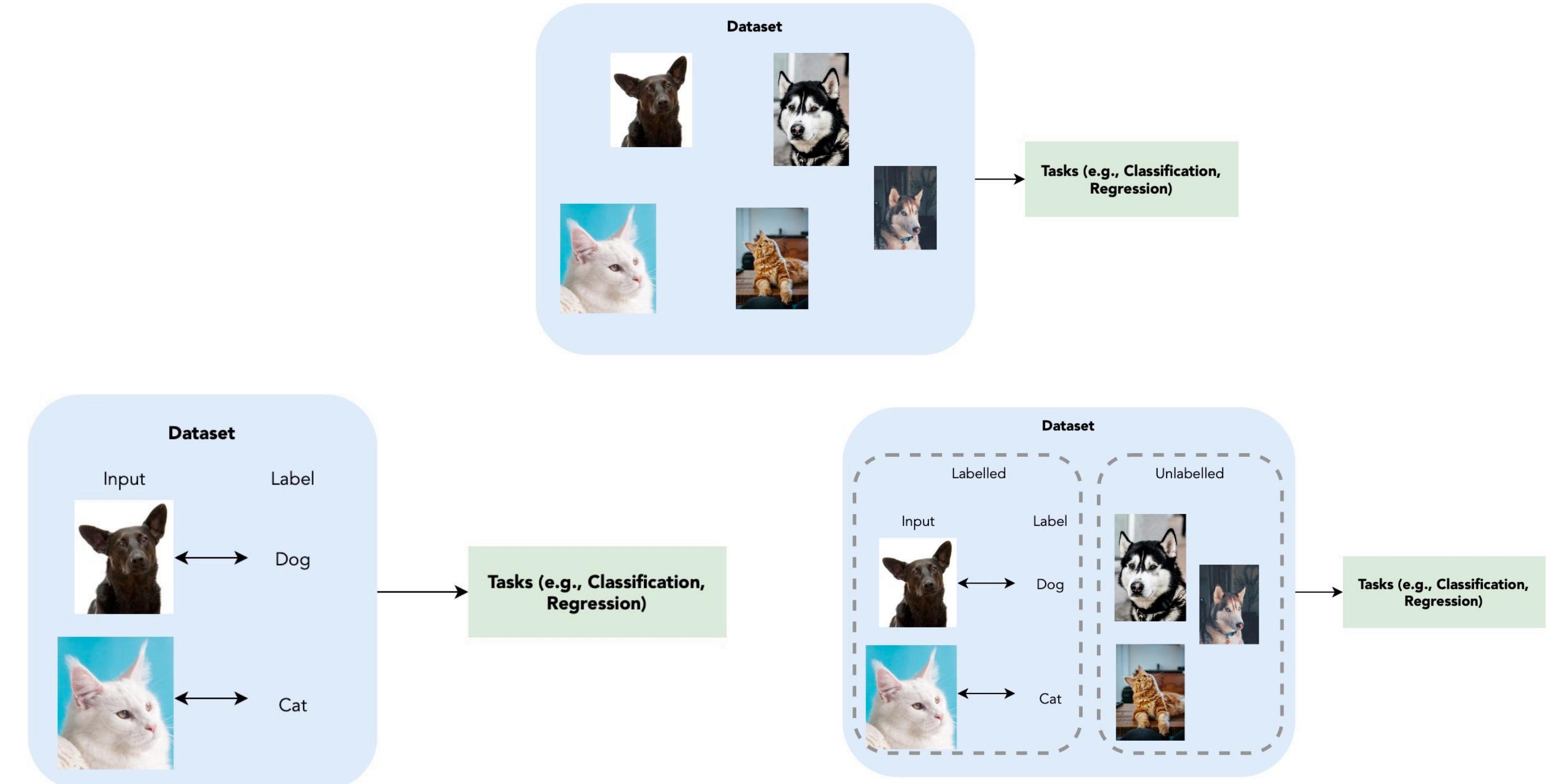


Réseau Inception



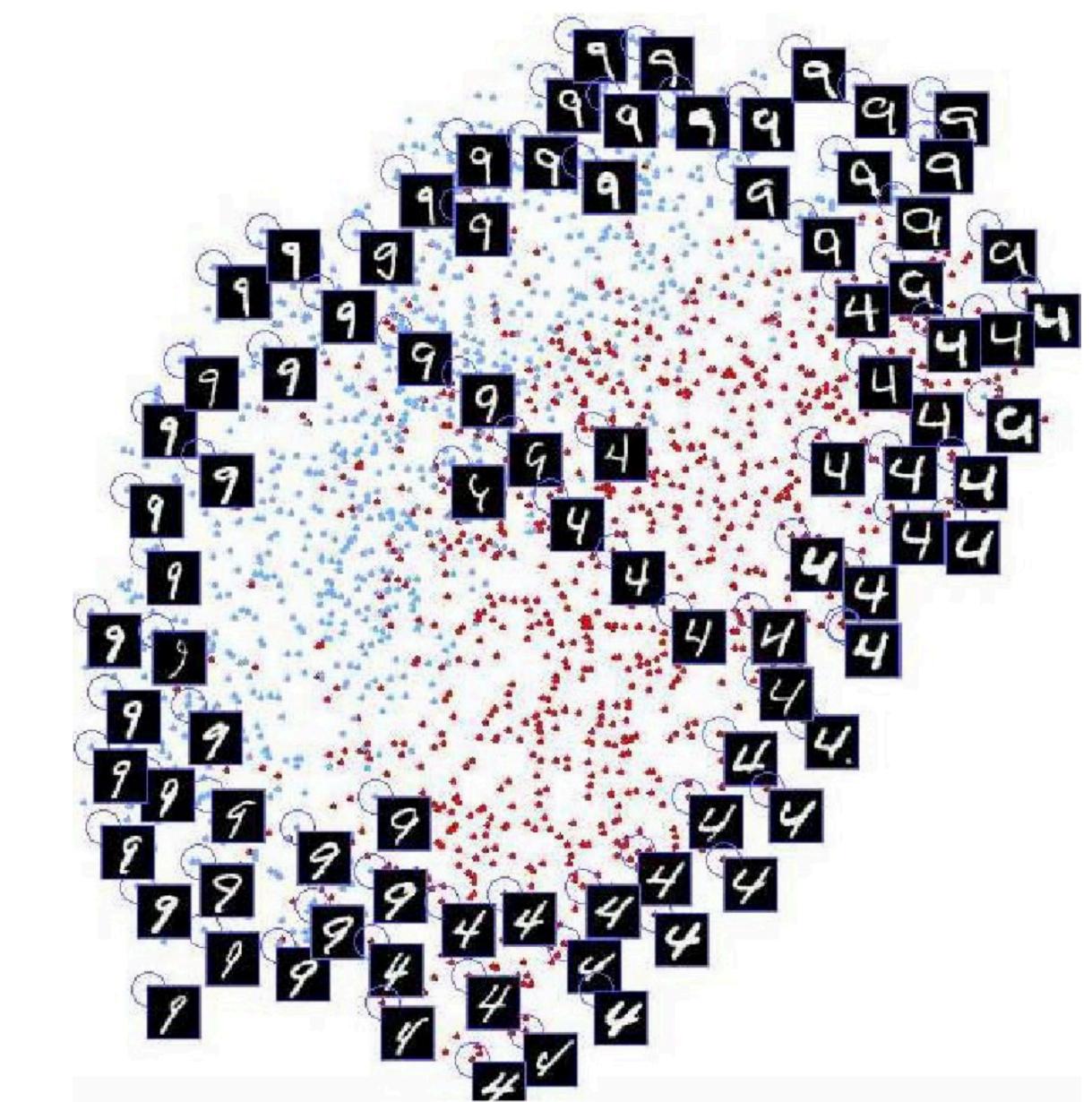
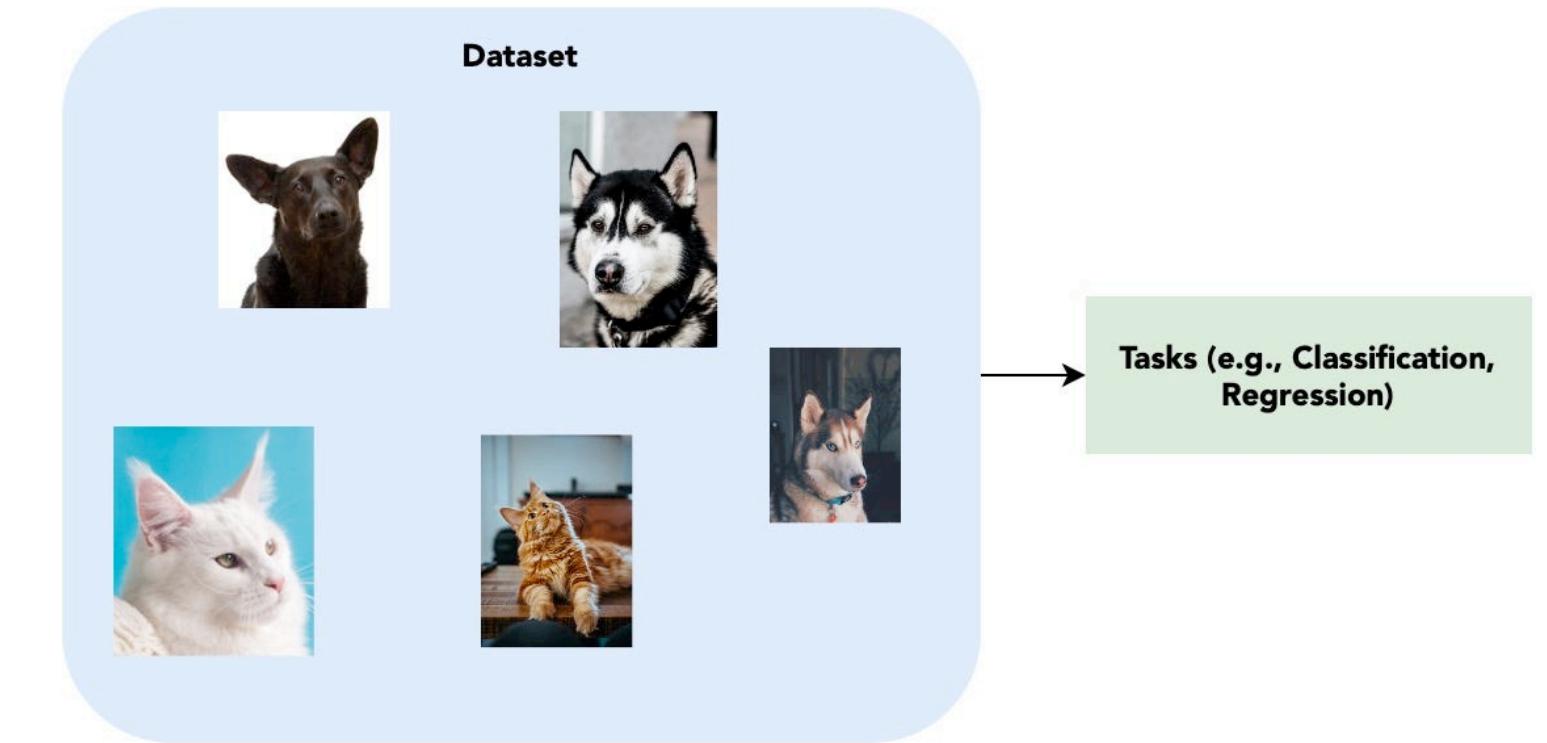
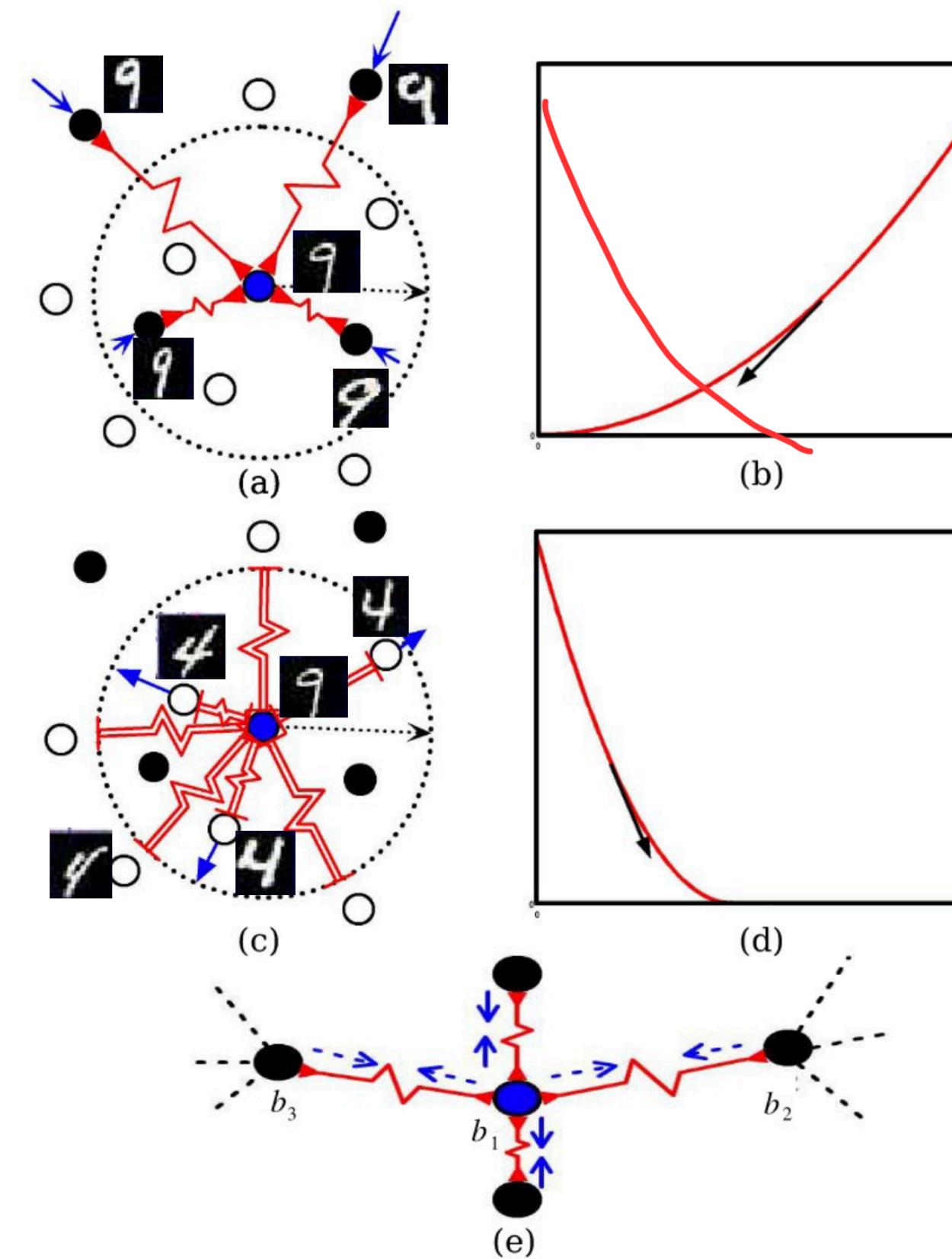
Les différents types d'apprentissage

- Supervisé
- Semi-supervisé
- Non supervisé

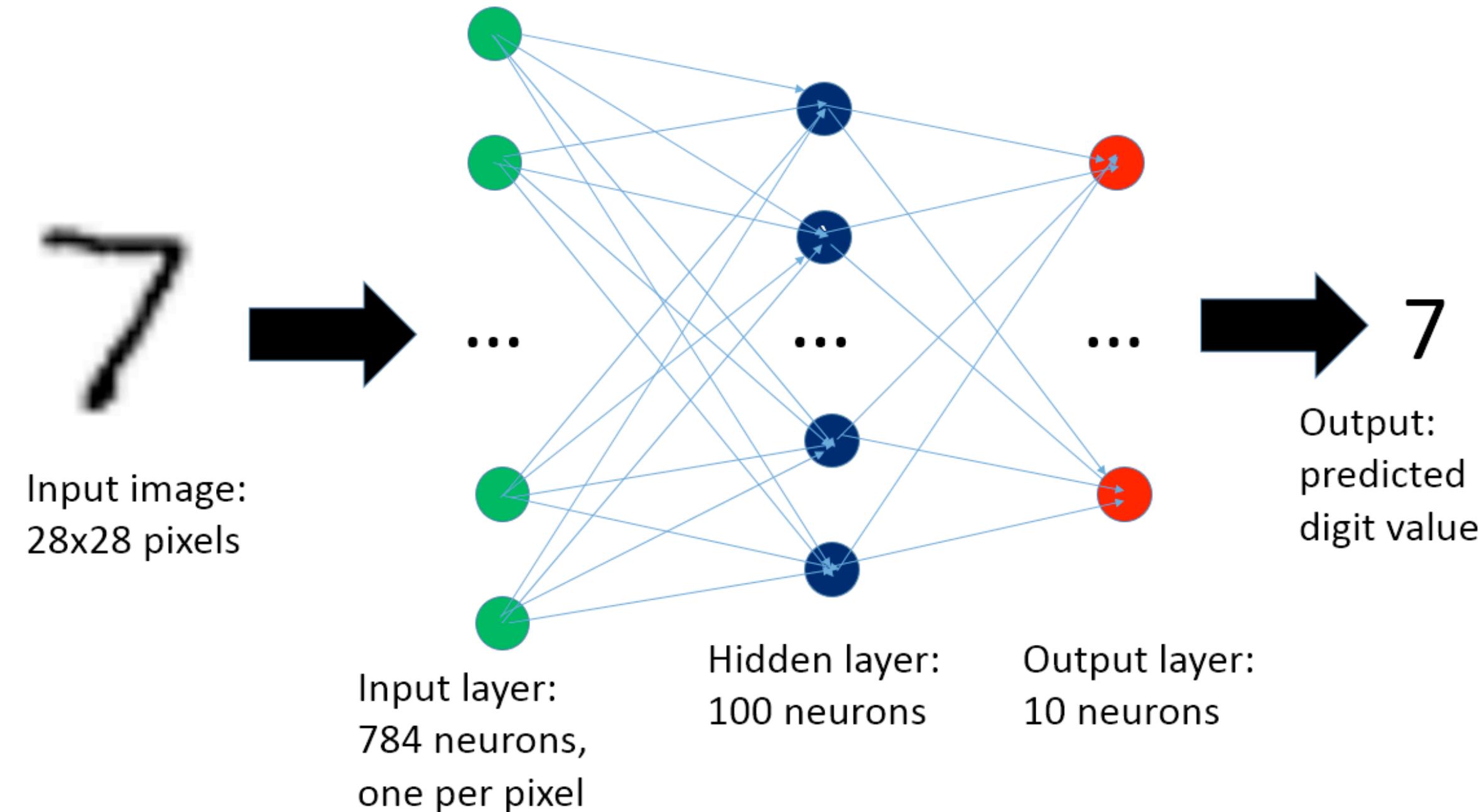


Apprentissage non supervisé

- Espace latent
- Attraction et repoussement



Applications



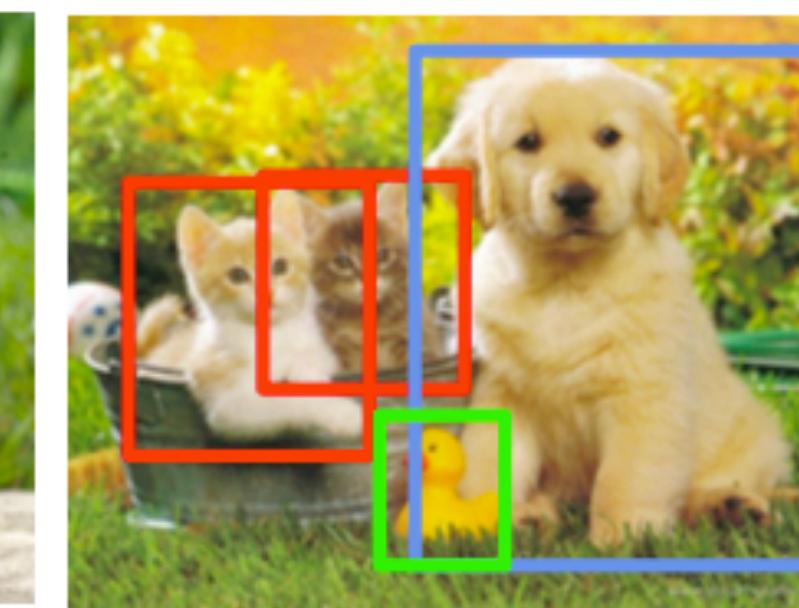
Classification



Classification + Localization



Object Detection

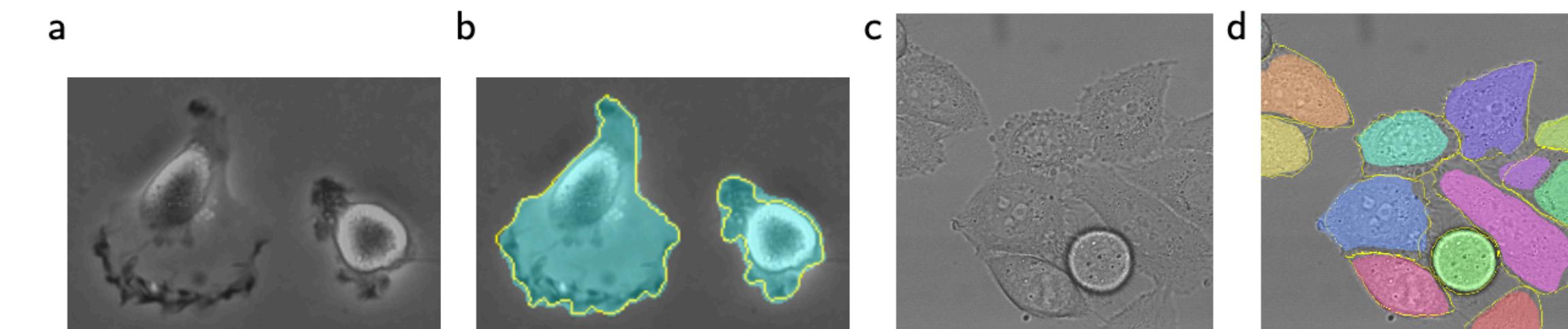
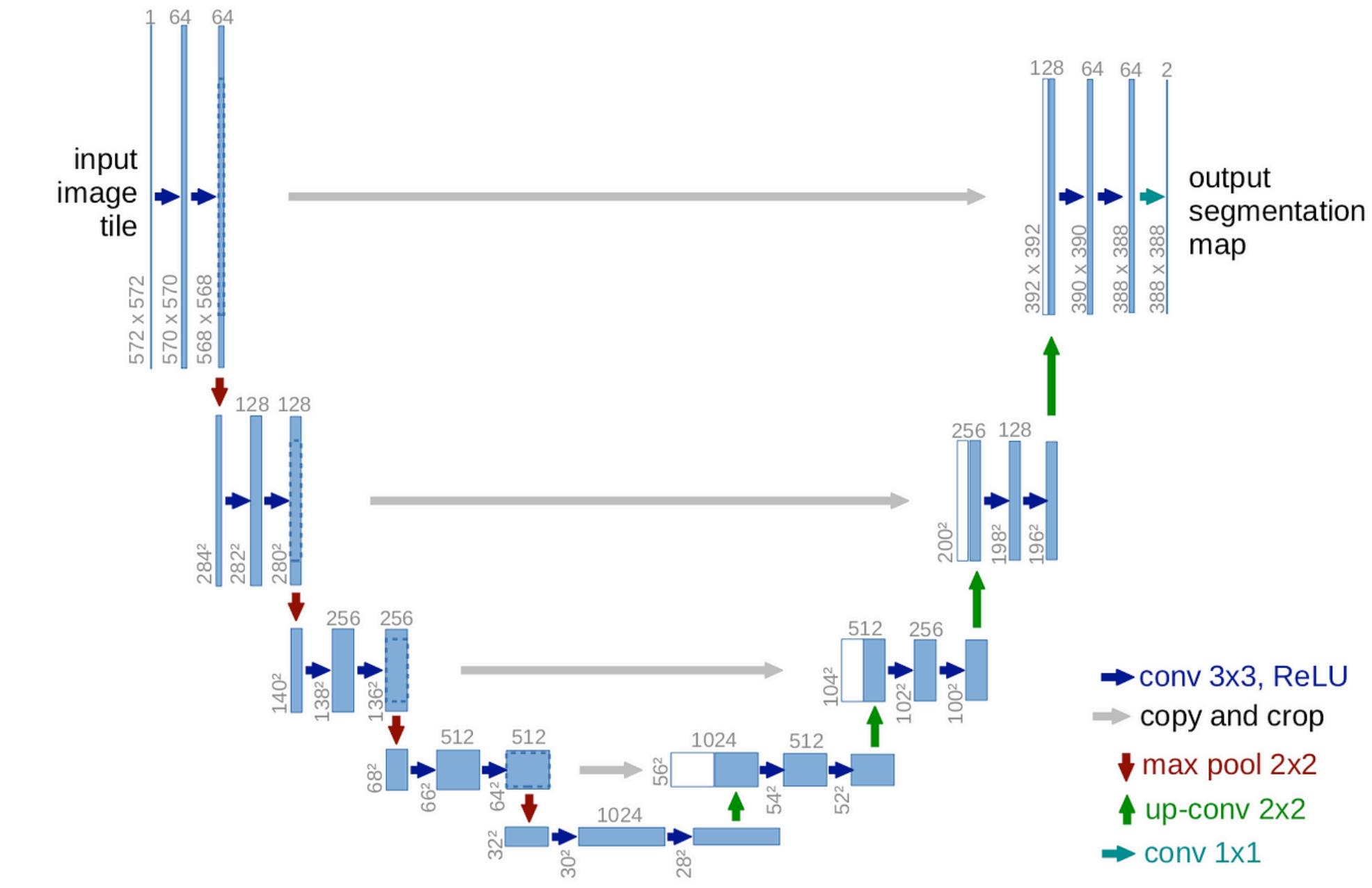


Instance Segmentation



Segmentation séquentielle

- UNet
- Image médical



Python + OpenCV



- GitHub
 - <https://github.com/cmcin019/mini-cours>
- Google Colab

Fin

- Merci