

# CMSC 510 – L20

## Regularization Methods for Machine Learning



### Part 20b: Skip connections

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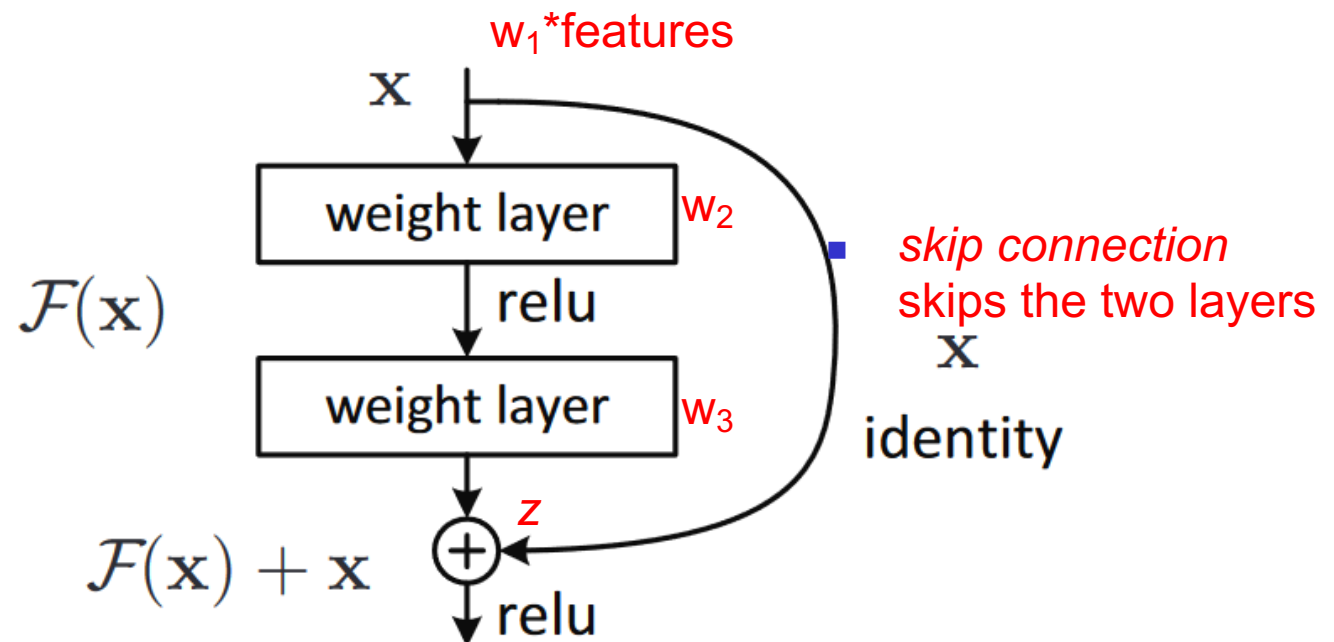
# Multi-layer networks

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- A neural network with ReLU activation is essentially:
  - $Y = \text{ReLU}(W_3 \text{ReLU}(W_2 \text{ReLU}(W_1 X)))$
- We see terms like  $z = w_{3ij} * w_{2kl} * w_{1mn} * x$
- The derivative of  $z$  over  $w_{2kl}$  is  $w_{3ij} * w_{1mn} * x$
- If  $w_{3ij}$  is small  
the gradient “signal” telling  $w_{2kl}$  in which way to change  
does not reach  $w_{2kl}$

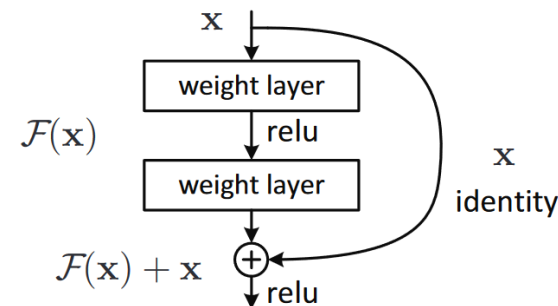
# Skip connections

- $z = w_{3ij} * w_{2kl} * w_{1mn} * \text{features}$ 
  - If  $w_{3ij}$  or  $w_{2kl}$  is small the gradient “signal” telling  $w_{1mn}$  in which way to change does not reach  $w_{1mn}$
- Solution: link  $w_{1mn}$  to  $z$  directly – via a *skip connection*



# Skip connections

- Solution: link  $w_{1mn}$  to  $z$  directly – via a skip connection
- Without skip connection:
  - We see terms like  $z = w_{3ij} * w_{2kl} * w_{1mn} * \text{input}$
  - The derivative of  $z$  over  $w_{1mn}$  is  $w_{3ij} * w_{2kl} * \text{input}$
- With skip connection:
  - Connect  $w_{1mn} * x$  directly to  $z$ , in addition to going through  $w_3$  and  $w_2$
  - We have  $z = w_{3ij} * w_{2kl} * w_{1mn} * \text{input} + w_{1mn} * \text{input}$   
$$z = (w_{3ij} * w_{2kl} + 1) * w_{1mn} * \text{input}$$
  - The derivative of  $z$  over  $w_{1mn}$  is  $w_{3ij} * w_{2kl} * \text{input} + \text{input}$

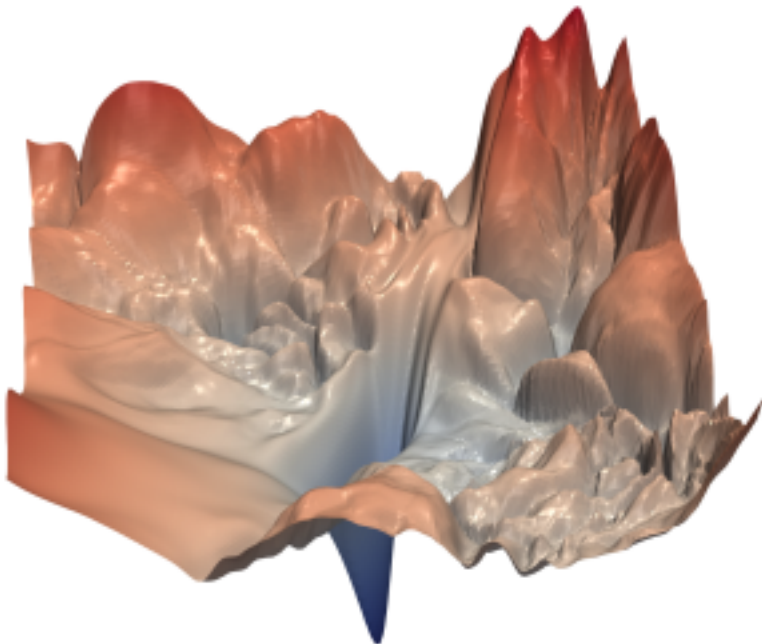


# ResNet

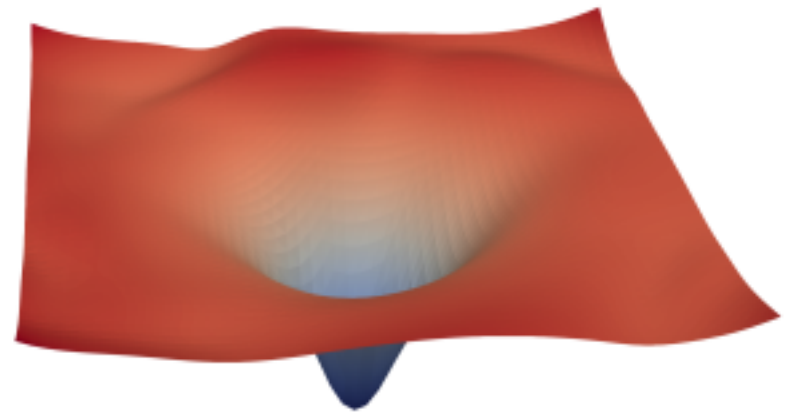
- Adding skip connections makes the loss landscape smoother

$$z = (w_{3ij} * w_{2kl} + 1) * w_{1mn} * x$$

less reliance just on multiplication



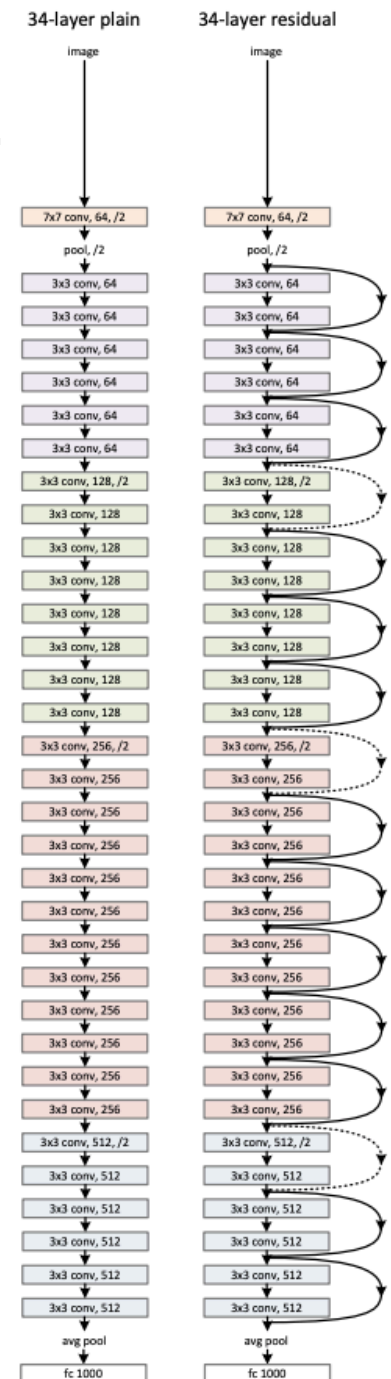
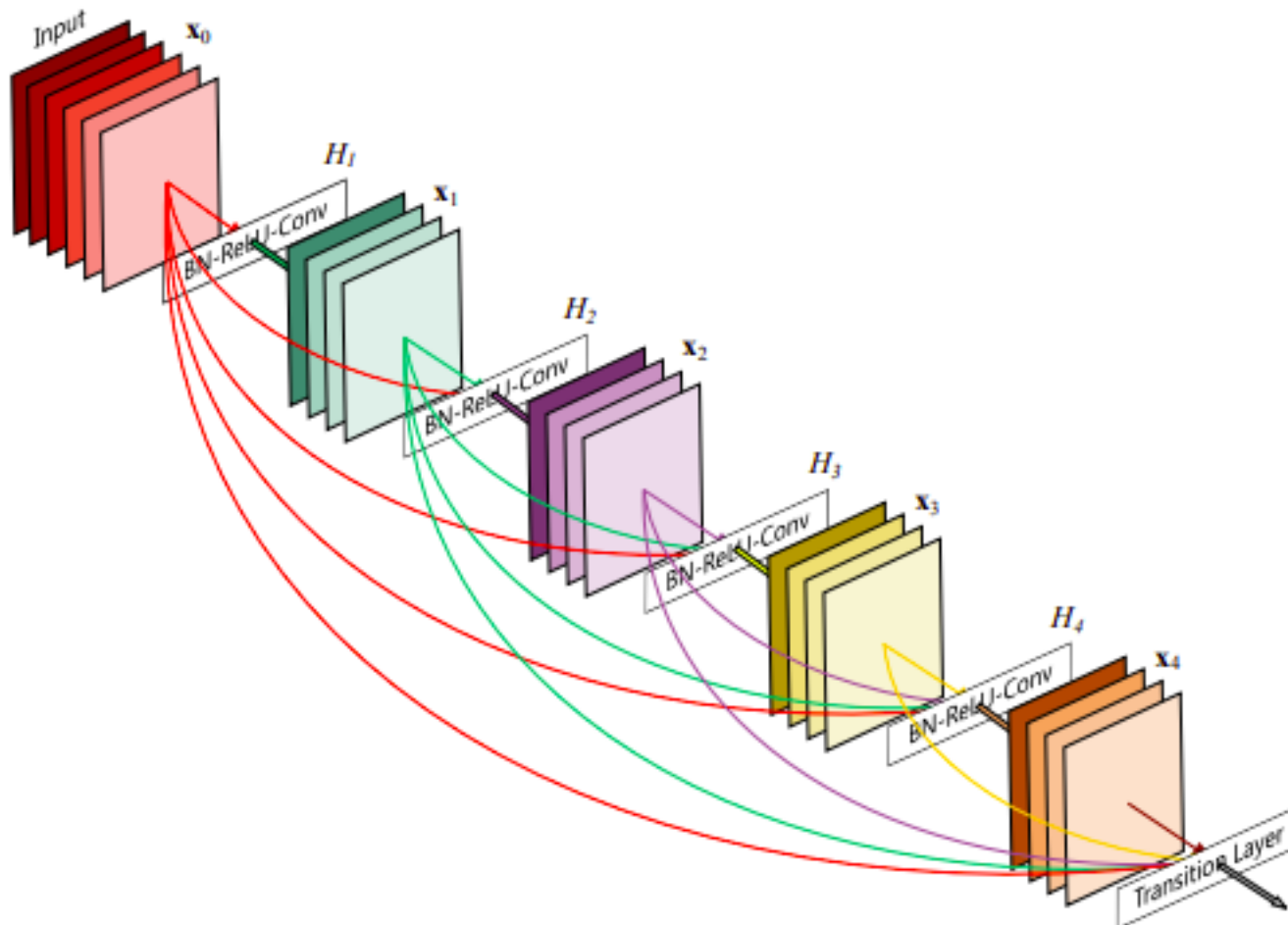
(a) without skip connections



(b) with skip connections

# Skip connections

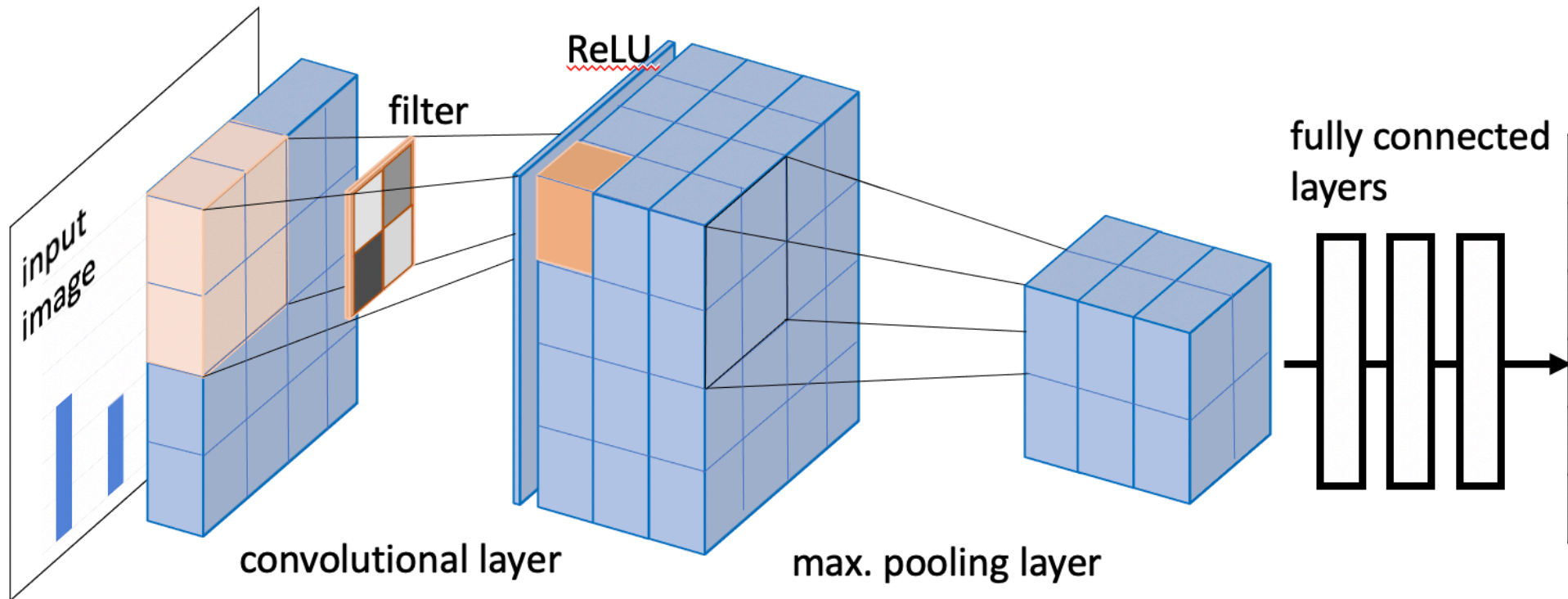
- Modern network architectures often involve many skip connections



# U-Net architecture

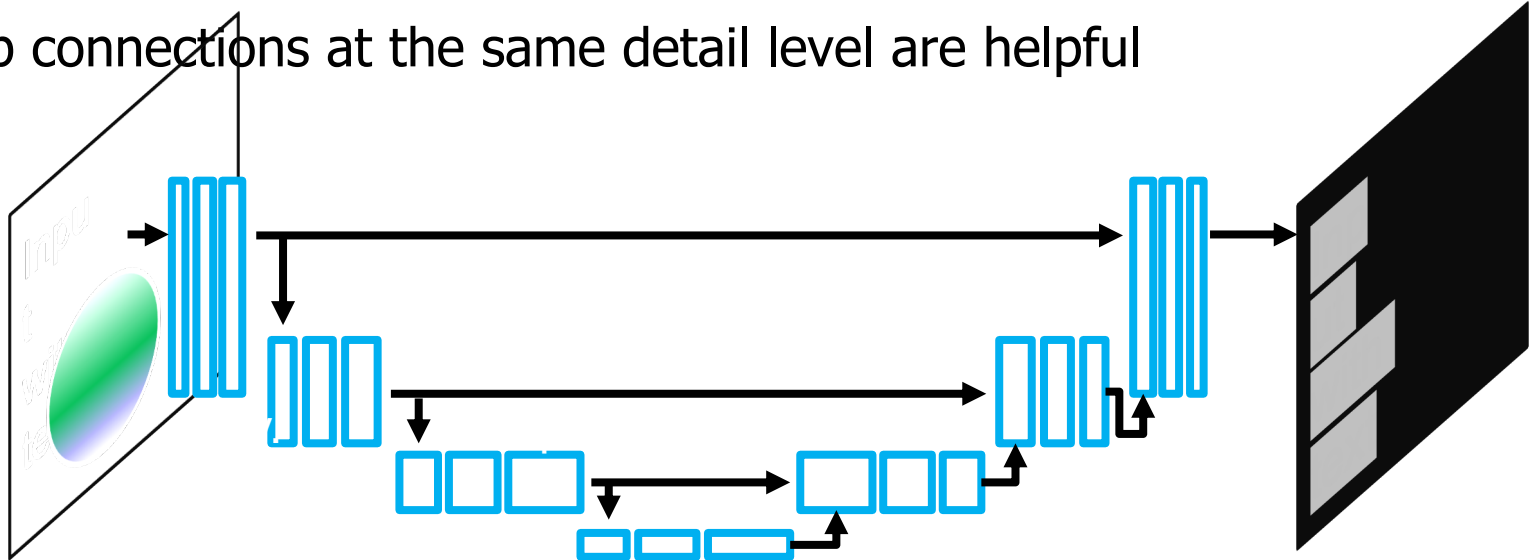
ConvNet:

downsampling via maxpooling



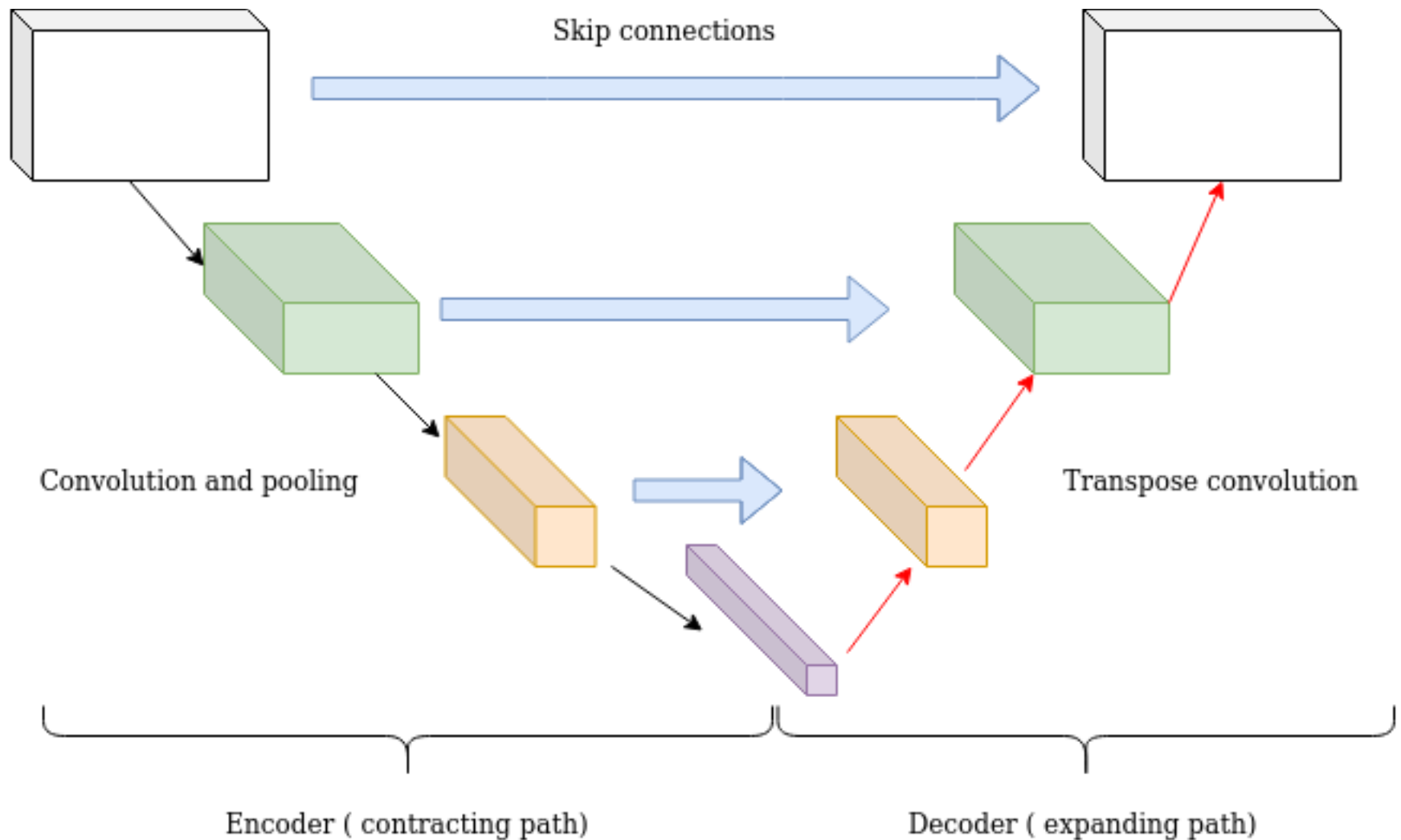
# U-Net architecture

- Skip connections are used in image-to-image networks
- E.g. U-Net architecture
  - Convnet downsampling (via max pooling)
    - From detail to coarse-grained
  - Upsampling
    - Generate detail from coarse-grained
  - Skip connections at the same detail level are helpful





# U-Net architecture



# Skip connections - summary

- Skip connection can help with optimization
- They can also help with preserving details (information) from earlier layers, e.g. in U-Net architecture
- Networks with skip connections are often called residual networks (ResNets)
  - They only need to learn the residual (the "delta")  $f(x)-x$  not the whole  $f(x)$  from scratch

