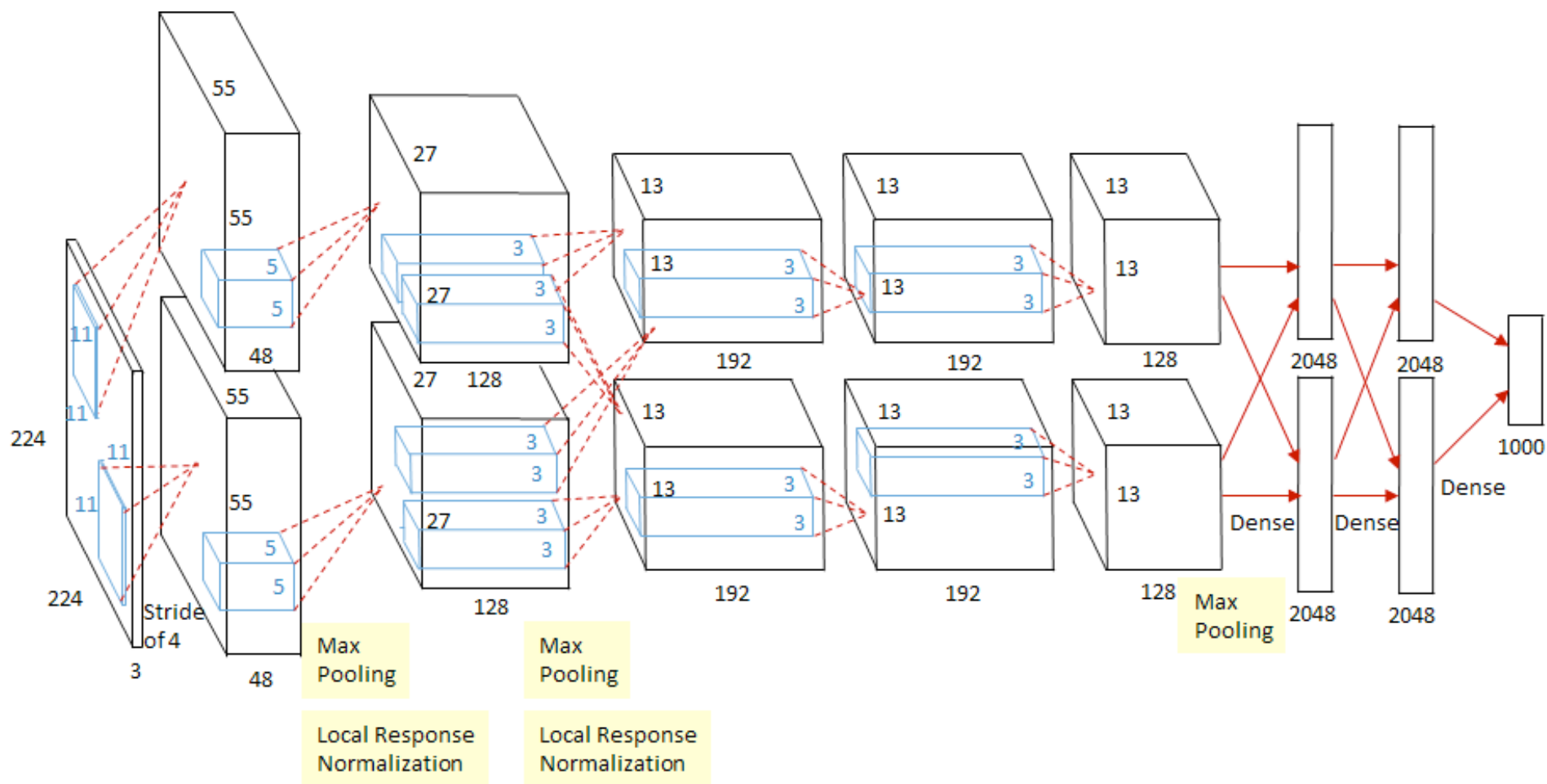


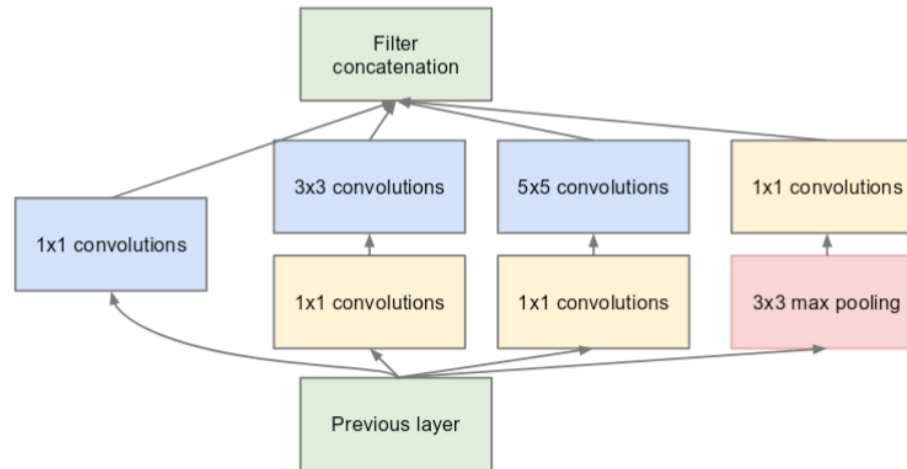
Whirlwind tour of modern deep learning

Vision Model

Alexnet (2012) - Conv. + GPU



Inception 2014-2016

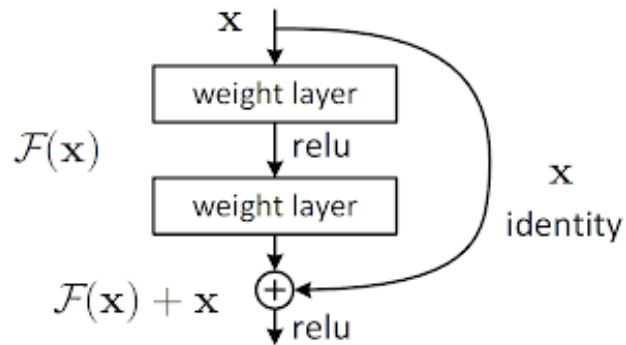


(b) Inception module with dimension reductions

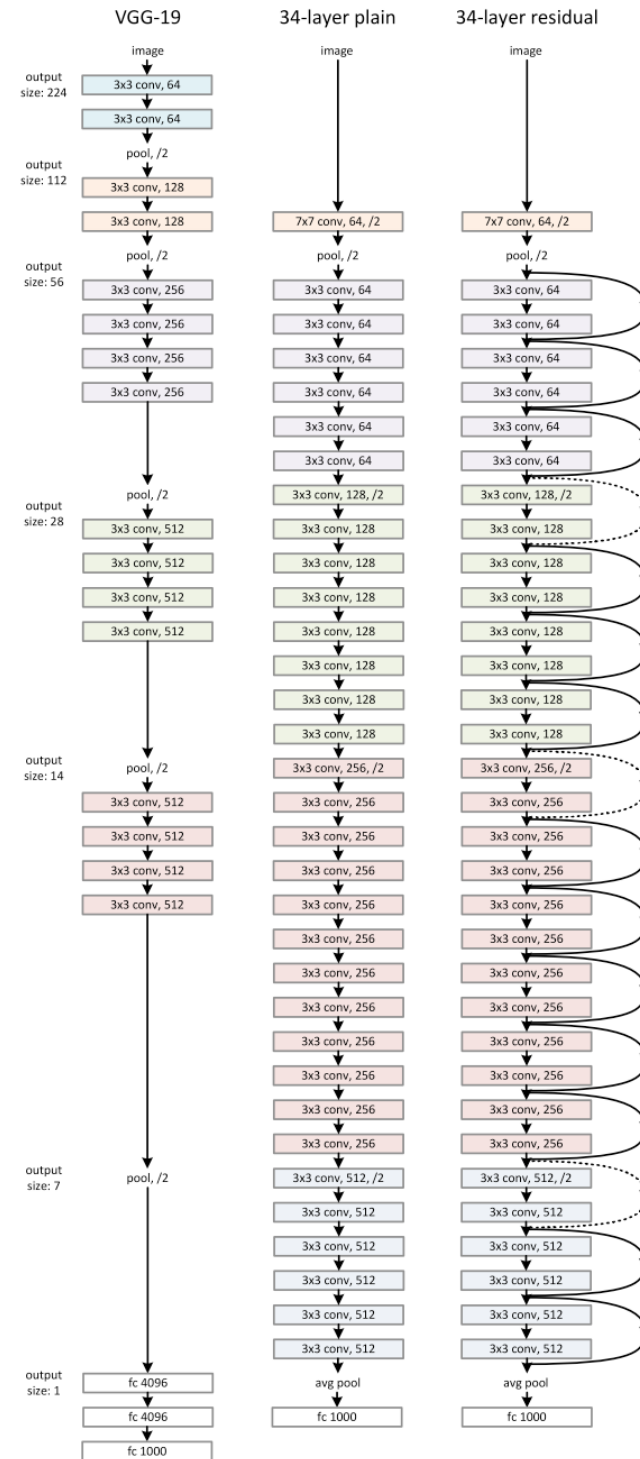
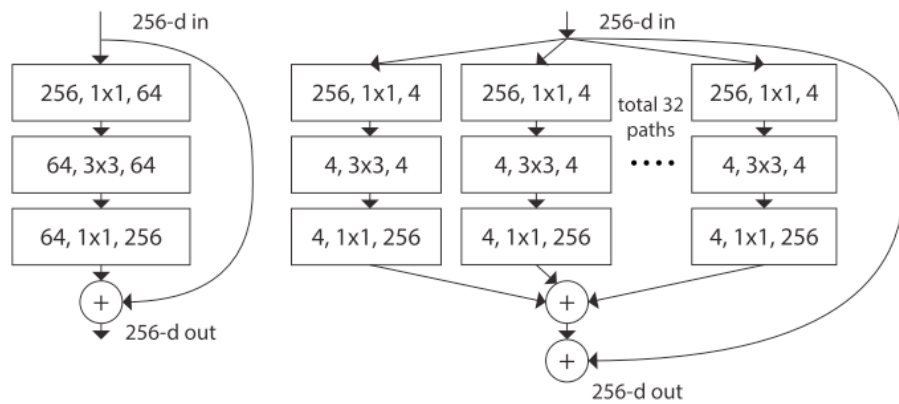
GoogLeNet



Resnet 2016



ResXnet 2016



<https://keras.io/api/applications/#usage-examples-for-image-classification-models>

Neural style transfer

https://www.tensorflow.org/tutorials/generative/style_transfer

<https://medium.com/tensorflow/neural-style-transfer-creating-art-with-deep-learning-using-tf-keras-and-eager-execution-7d541ac31398>

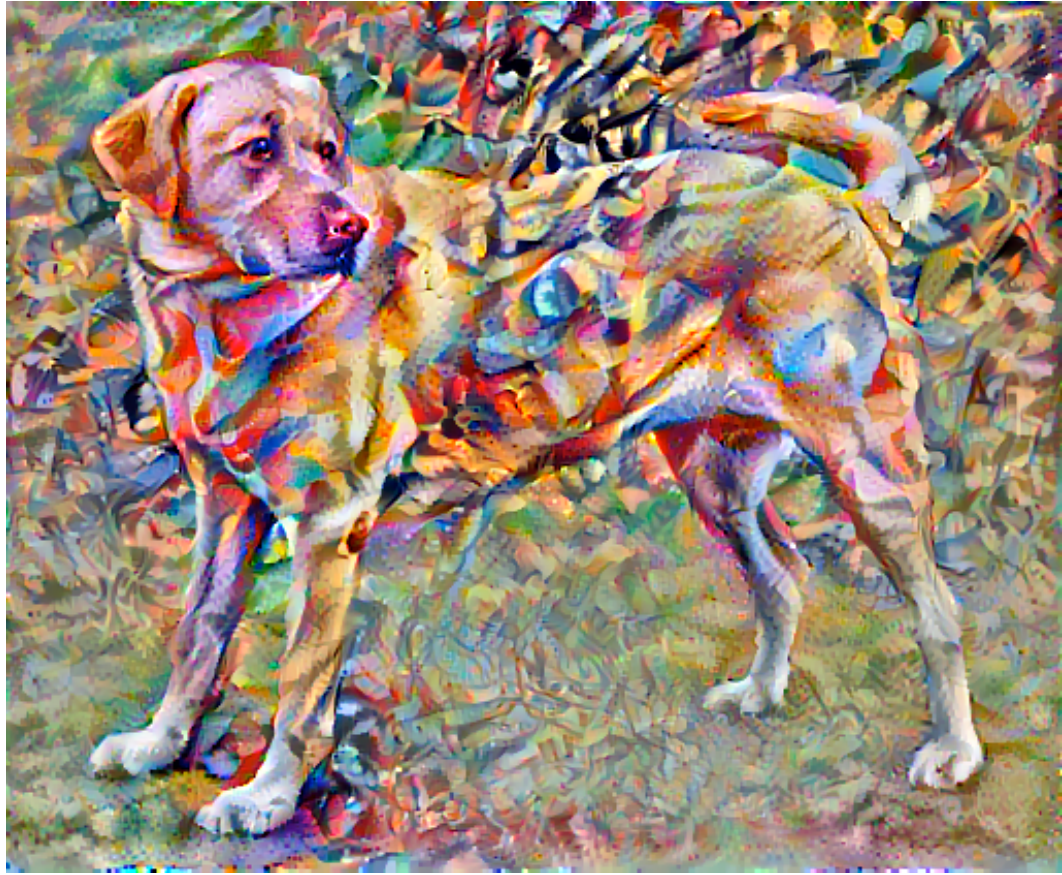
Content

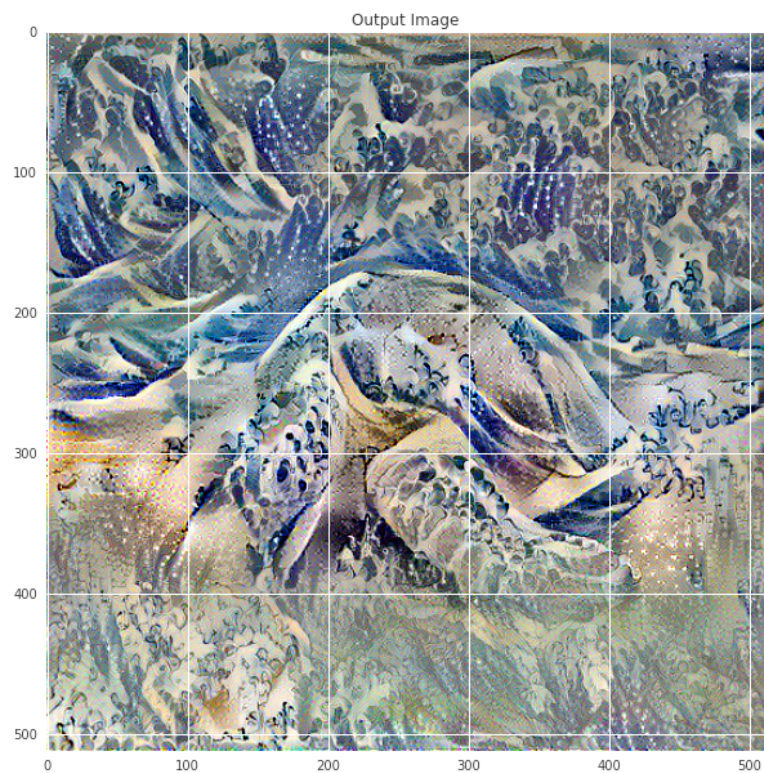
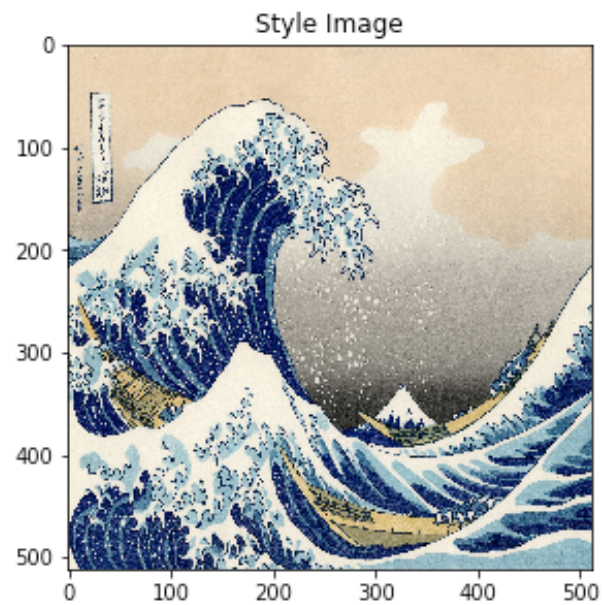
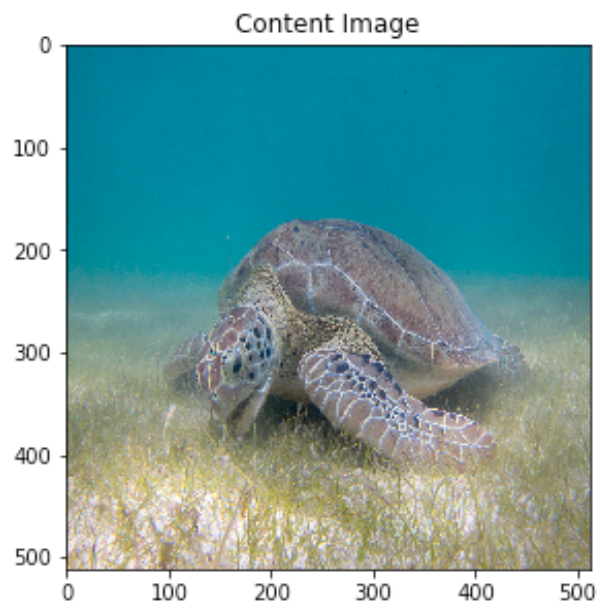


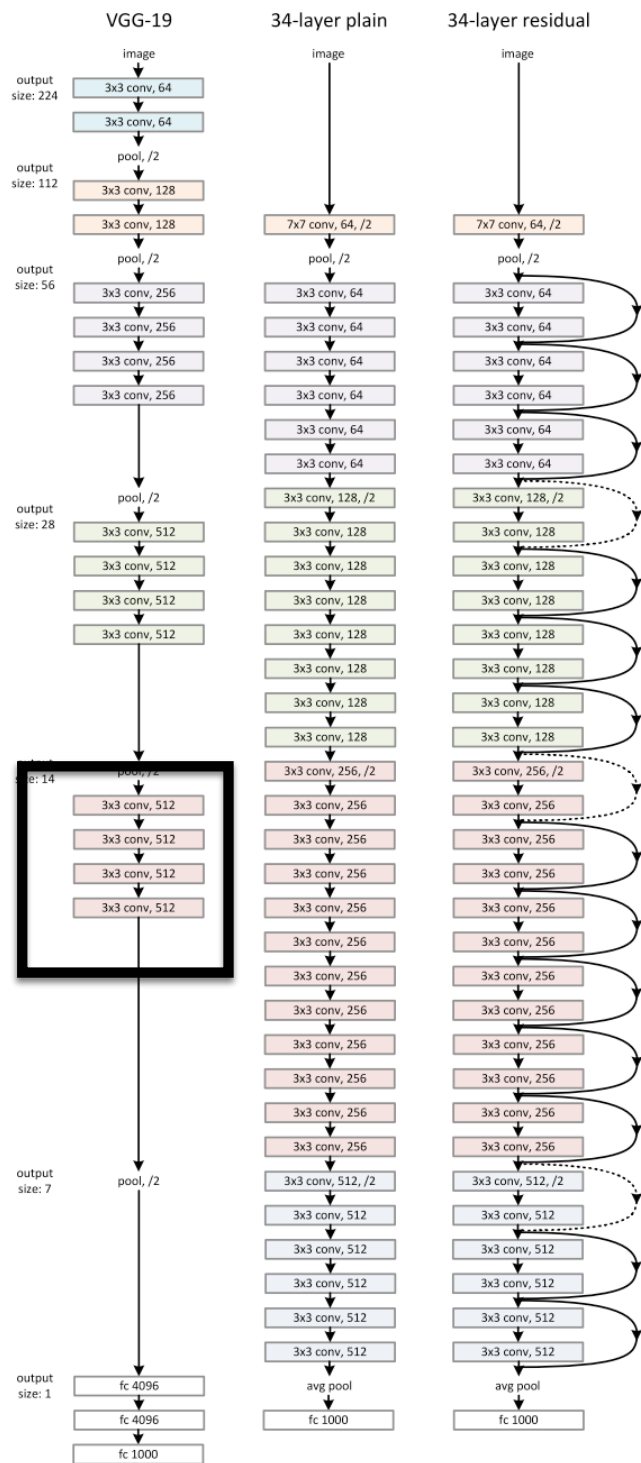
+

Style







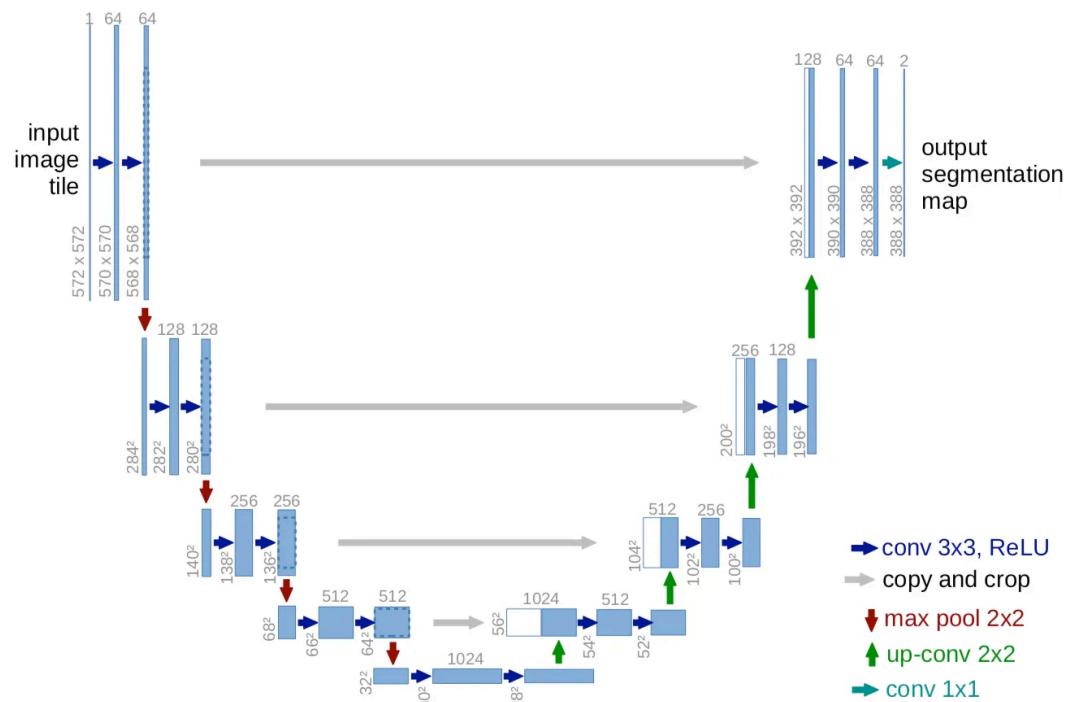


Content: means
Style: Correlations between filter

$$L_{content}^l(p, x) = \sum_{i,j} (F_{ij}^l(x) - P_{ij}^l(p))^2$$

$$E_l = \frac{1}{4N_l^2 M_l^2} \sum_{i,j} (G_{ij}^l - A_{ij}^l)^2$$

$$L_{style}(a, x) = \sum_{l \in L} w_l E_l$$



<https://towardsdatascience.com/unet-line-by-line-explanation-9b191c76baf5>

<https://towardsdatascience.com/understanding-semantic-segmentation-with-unet-6be4f42d4b47>

Natural Language Processing

Word2Vec: Map words to vector space

<https://www.tensorflow.org/tutorials/text/word2vec>

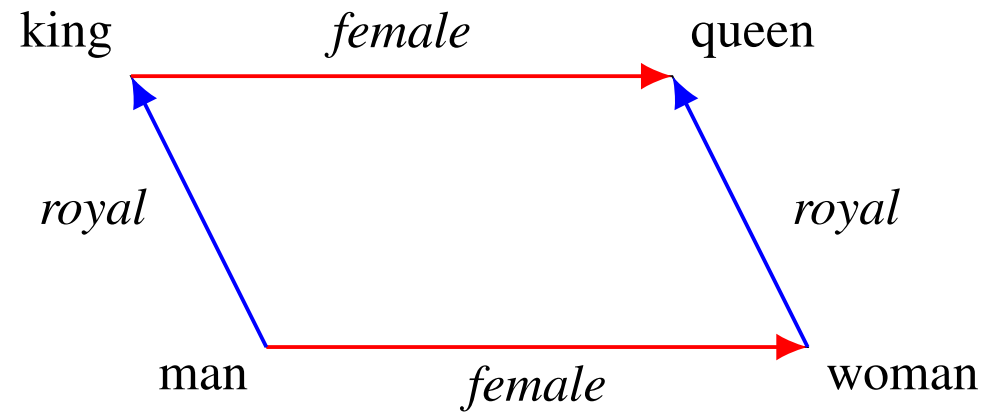
Window Size	Text	Skip-grams
2	[The <u>wide</u> road shimmered] in the hot sun.	wide, the wide, road wide, shimmered
	The [wide road <u>shimmered</u> in the] hot sun.	shimmered, wide shimmered, road shimmered, in shimmered, the
	The wide road shimmered in [the hot <u>sun</u>].	sun, the sun, hot
3	[The <u>wide</u> road shimmered in] the hot sun.	wide, the wide, road wide, shimmered wide, in
	[The wide road <u>shimmered</u> in the hot] sun.	shimmered, the shimmered, wide shimmered, road shimmered, in shimmered, the shimmered, hot
	The wide road shimmered [in the hot <u>sun</u>].	sun, in sun, the sun, hot

$$\frac{1}{T} \sum_{t=1}^T \sum_{-c \leq j \leq c, j \neq 0} \log p(w_{t+j} | w_t)$$

$$p(w_O | w_I) = \frac{\exp \left(v'_{w_O} \top v_{w_I} \right)}{\sum_{w=1}^W \exp \left(v'_w \top v_{w_I} \right)}$$

Talk about negative sampling

Word analogies

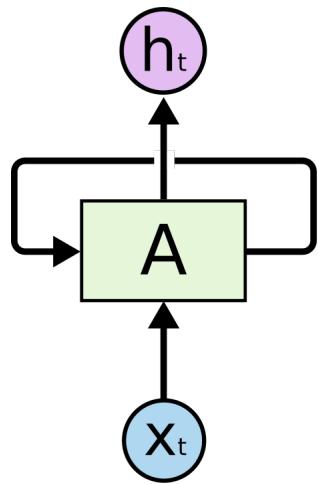
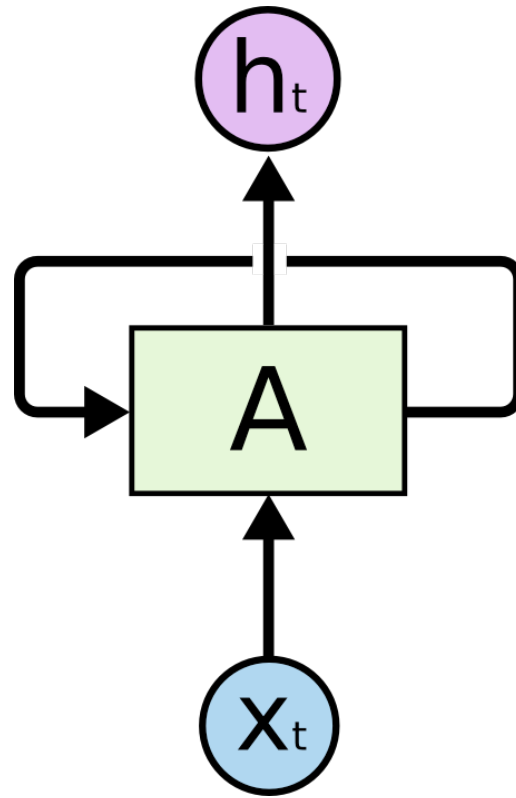


<https://p.migdal.pl/2017/01/06/king-man-woman-queen-why.html>

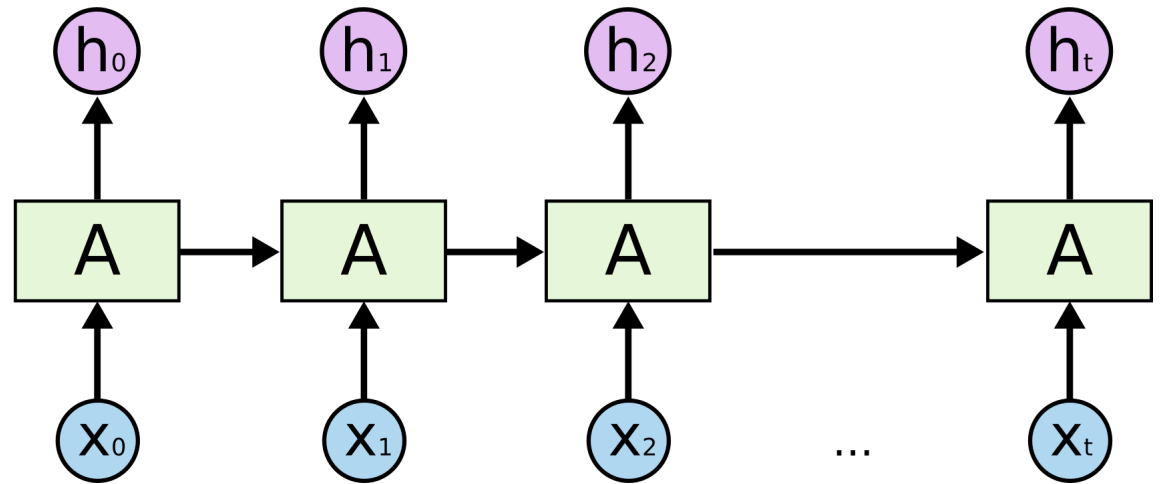
[https://pytorch.org/tutorials/beginner/nlp/
sequence_models_tutorial.html](https://pytorch.org/tutorials/beginner/nlp/sequence_models_tutorial.html)

Recurrent Neural Networks

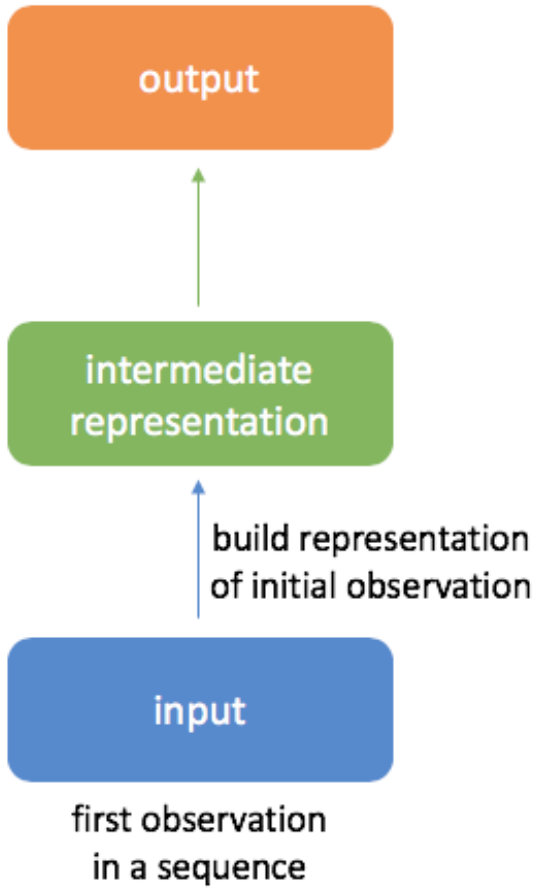
[https://www.jeremyjordan.me/introduction-to-recurrent-neural-
networks/](https://www.jeremyjordan.me/introduction-to-recurrent-neural-networks/)



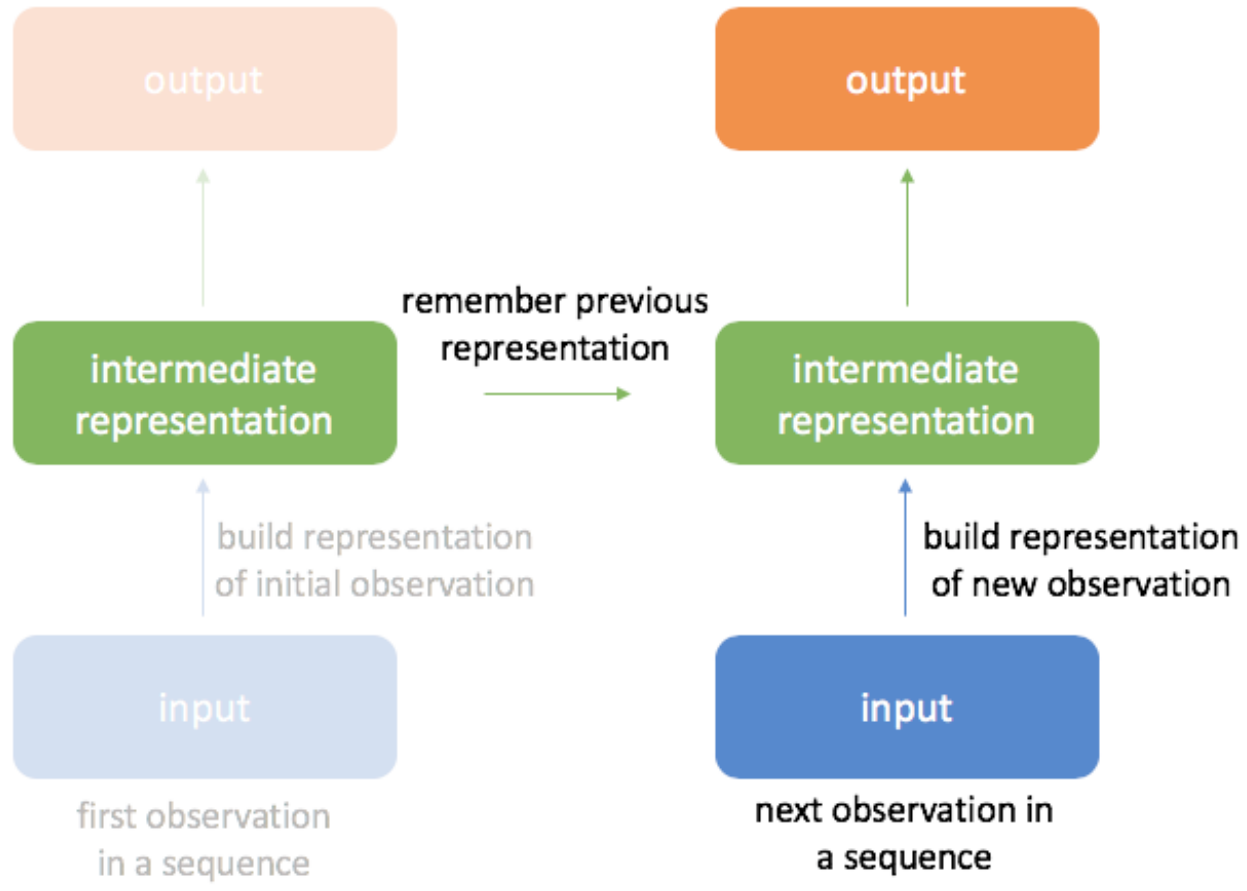
=

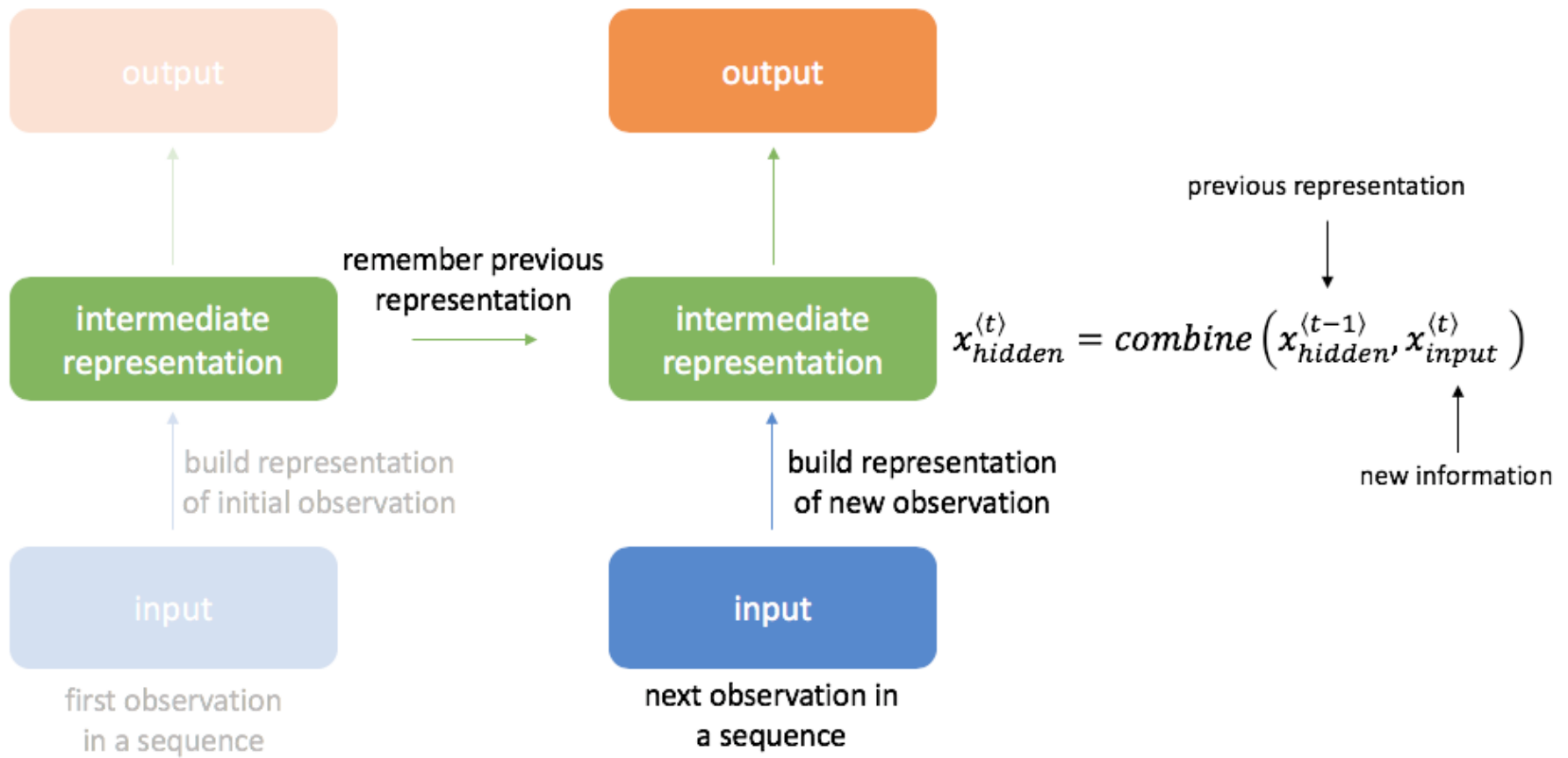


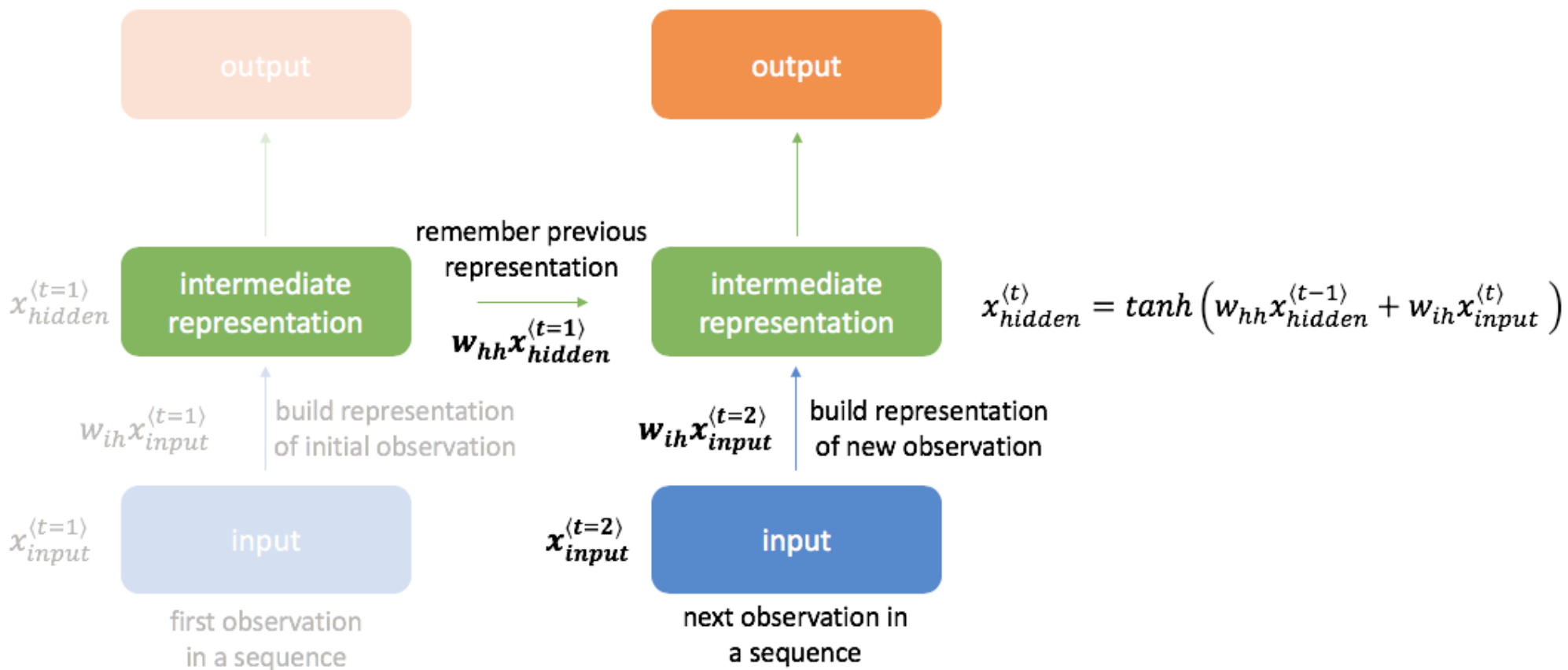
Non-sequential



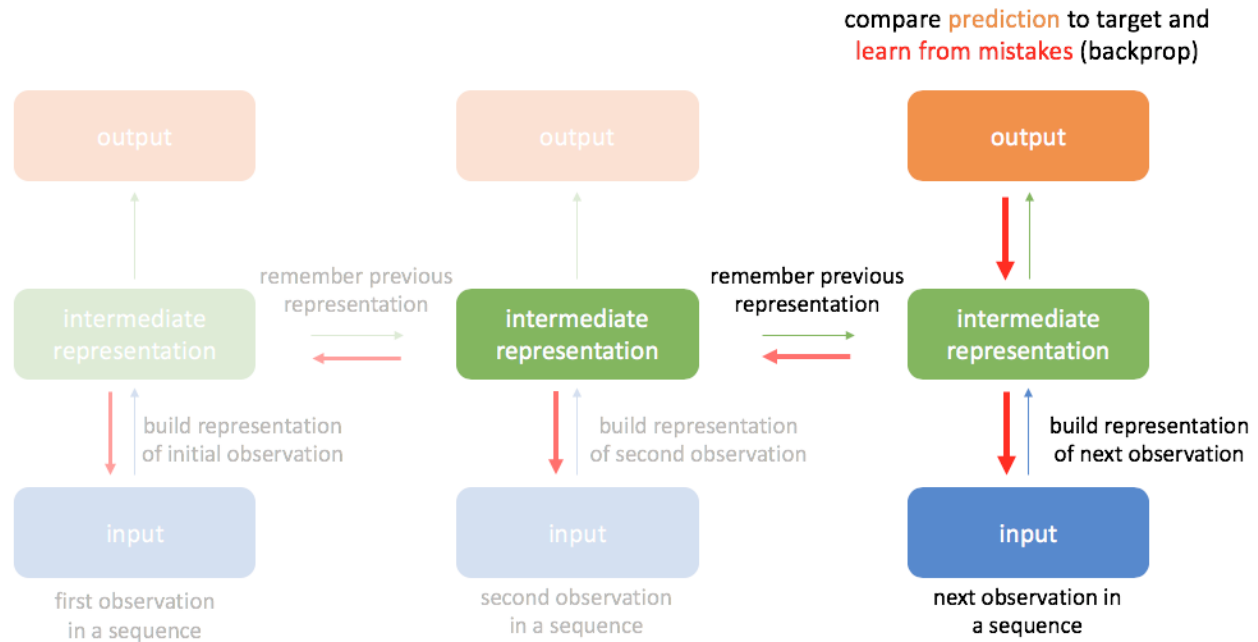
Sequential





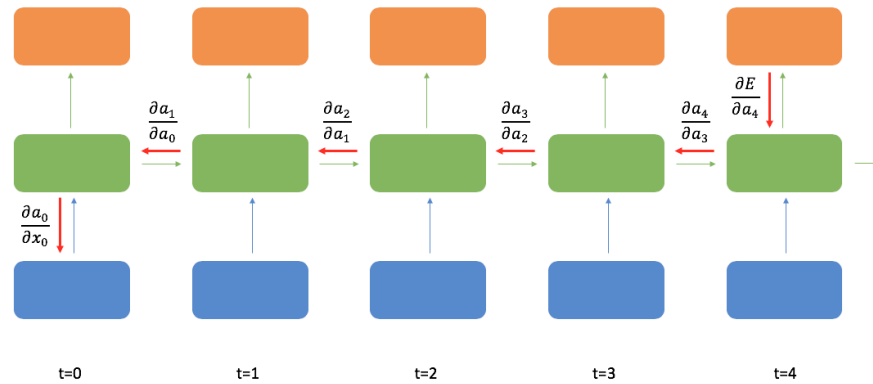


Backprop

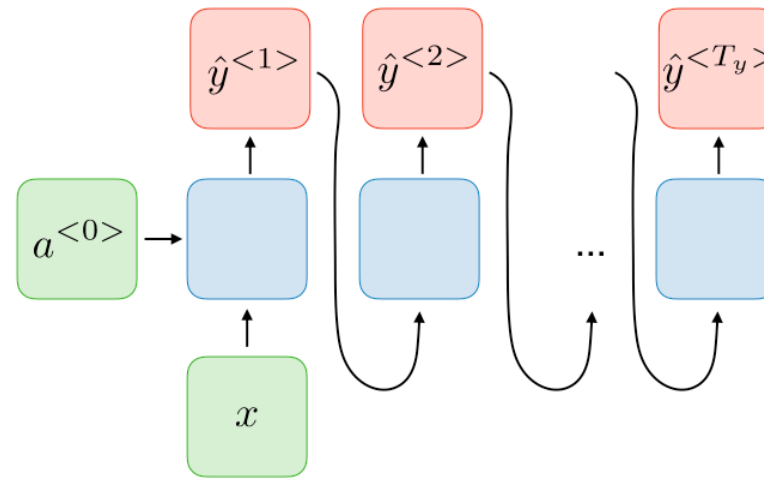


Notation:

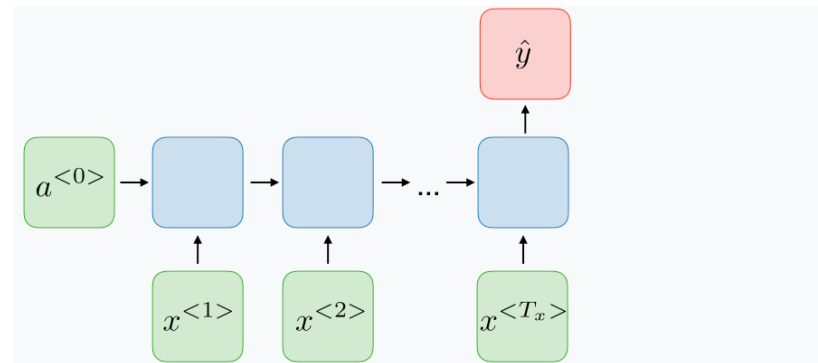
$a_t = x_{hidden}^{(t=t)}$ (the output of a recurrent layer)
 $x_t = x_{input}^{(t=t)}$ (the input to a recurrent layer)



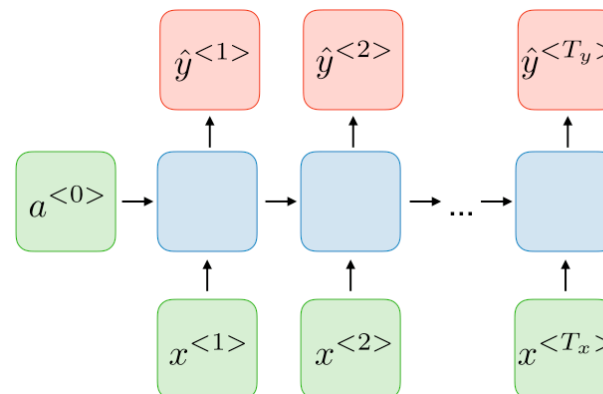
One-to-many



Many-to-one



Many-to-Many

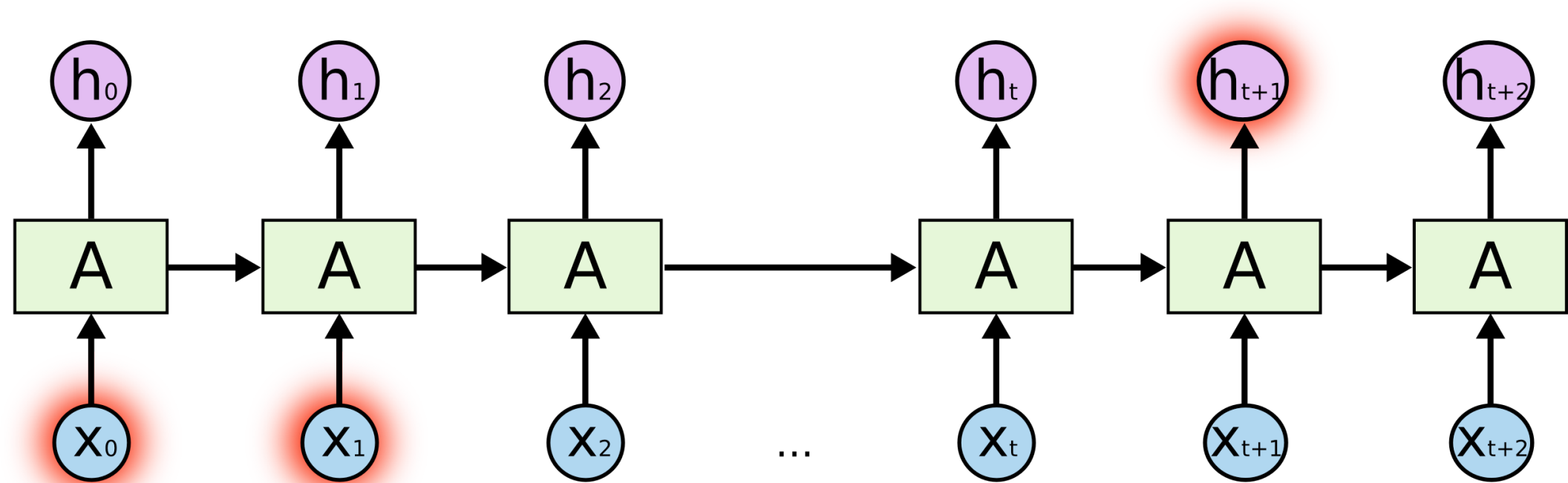


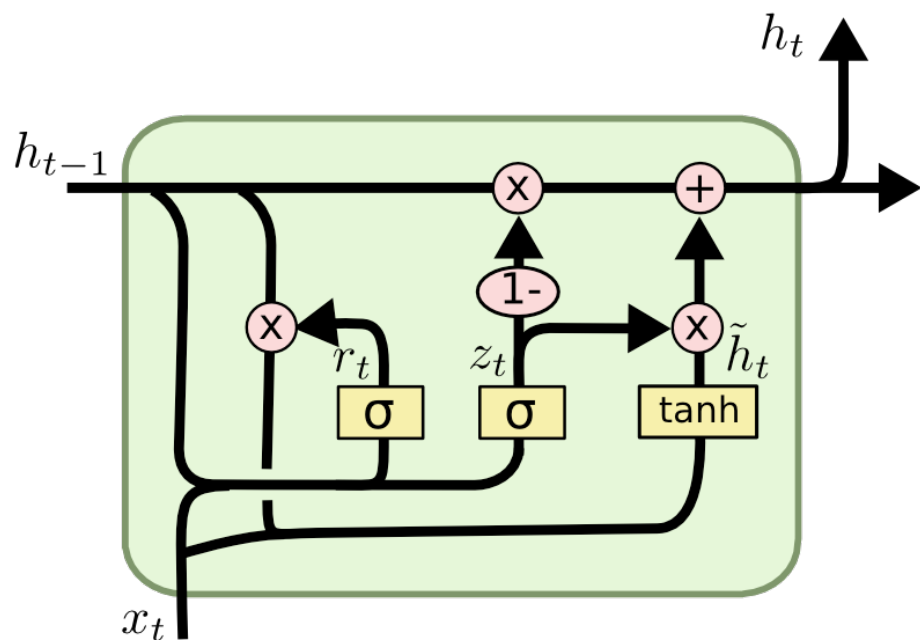
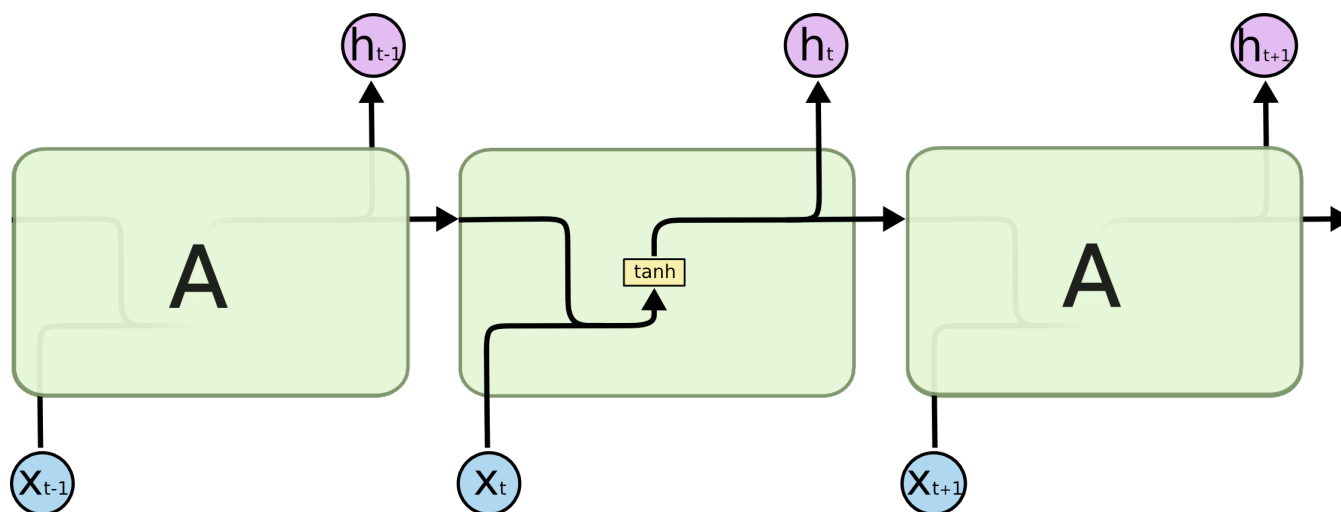
LSTM- Networks

[https://pytorch.org/tutorials/beginner/nlp/
sequence_models_tutorial.html](https://pytorch.org/tutorials/beginner/nlp/sequence_models_tutorial.html)

<https://colah.github.io/posts/2015-08-Understanding-LSTMs/>

How do you remember things for long times?
Many layers....





$$z_t = \sigma(W_z \cdot [h_{t-1}, x_t])$$

$$r_t = \sigma(W_r \cdot [h_{t-1}, x_t])$$

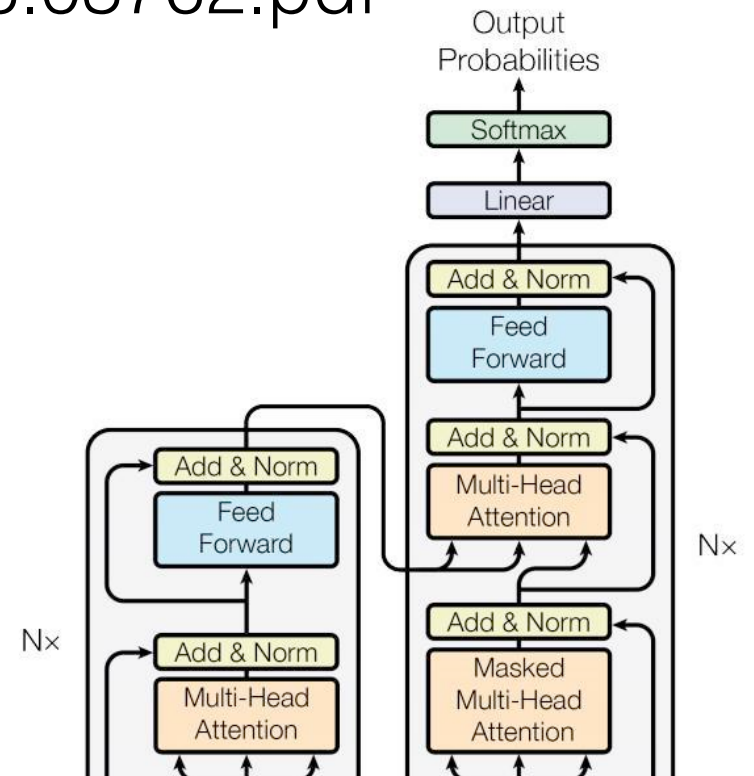
$$\tilde{h}_t = \tanh(W \cdot [r_t * h_{t-1}, x_t])$$

$$h_t = (1 - z_t) * h_{t-1} + z_t * \tilde{h}_t$$

Transformers

https://pytorch.org/tutorials/beginner/transformer_tutorial.html

<https://arxiv.org/pdf/1706.03762.pdf>



Transformers: Language Translation