

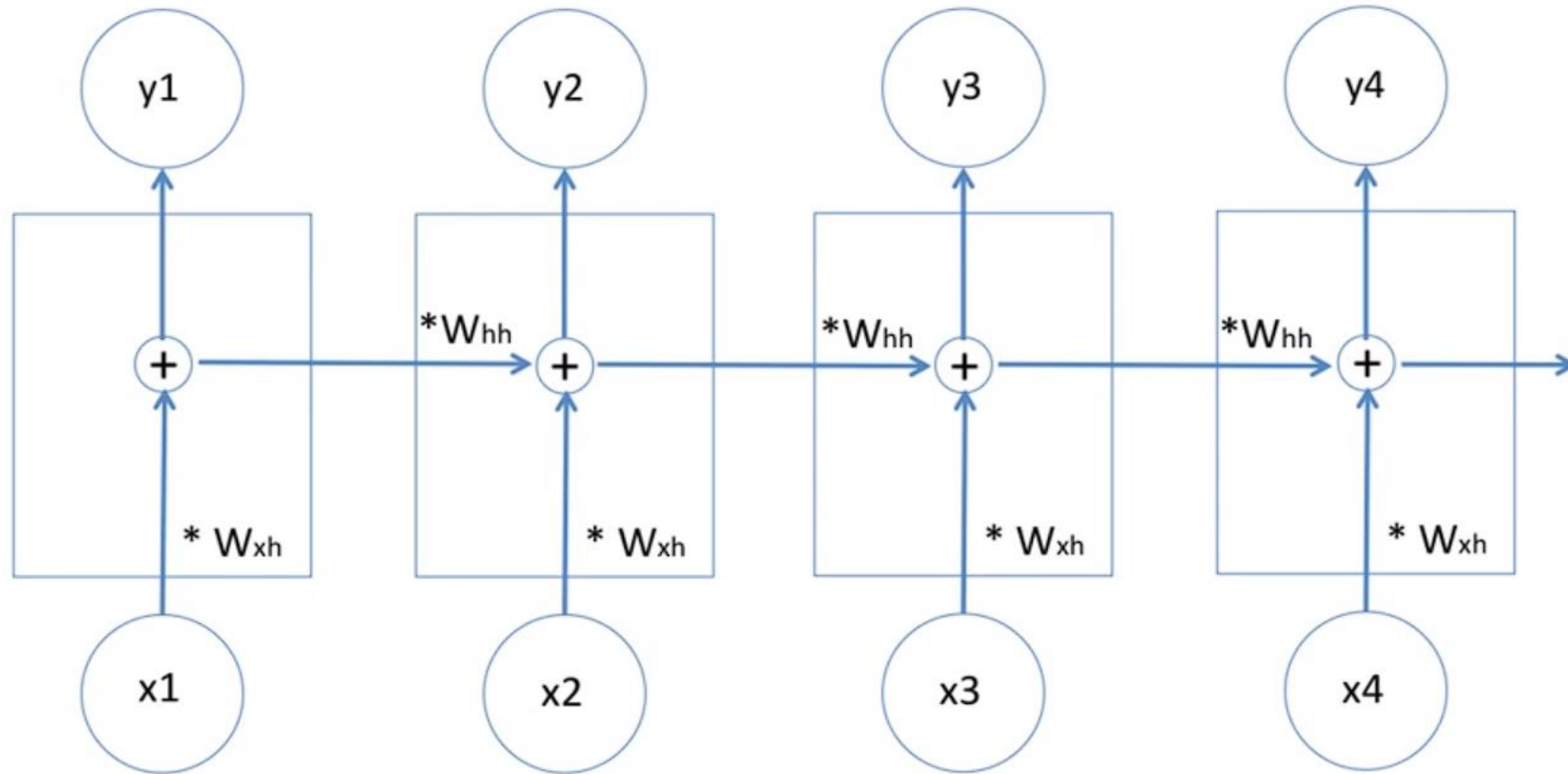
RNN

Recurrent Neural Network (RNN)

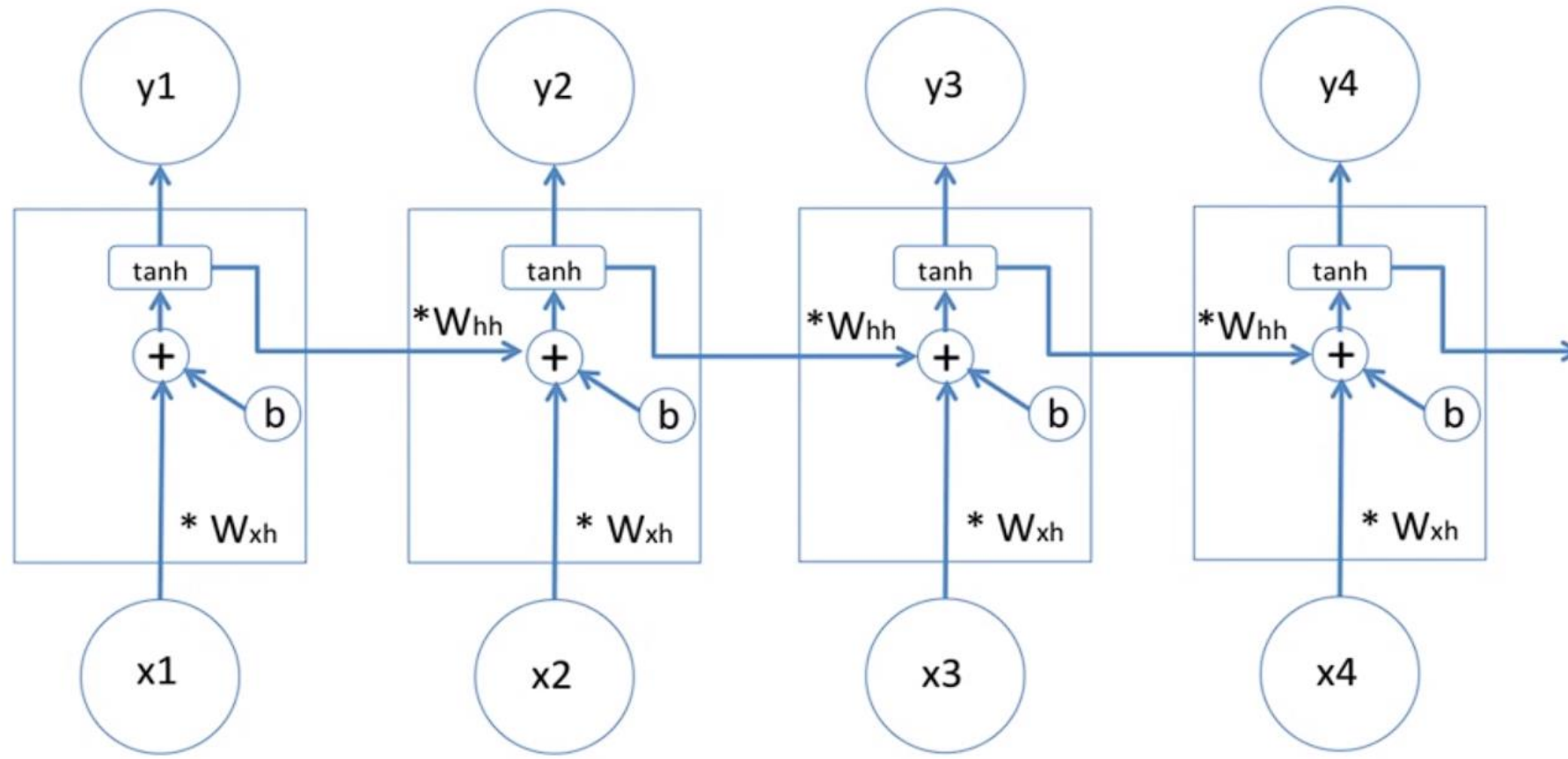
순환신경망(Recurrent Neural Network)

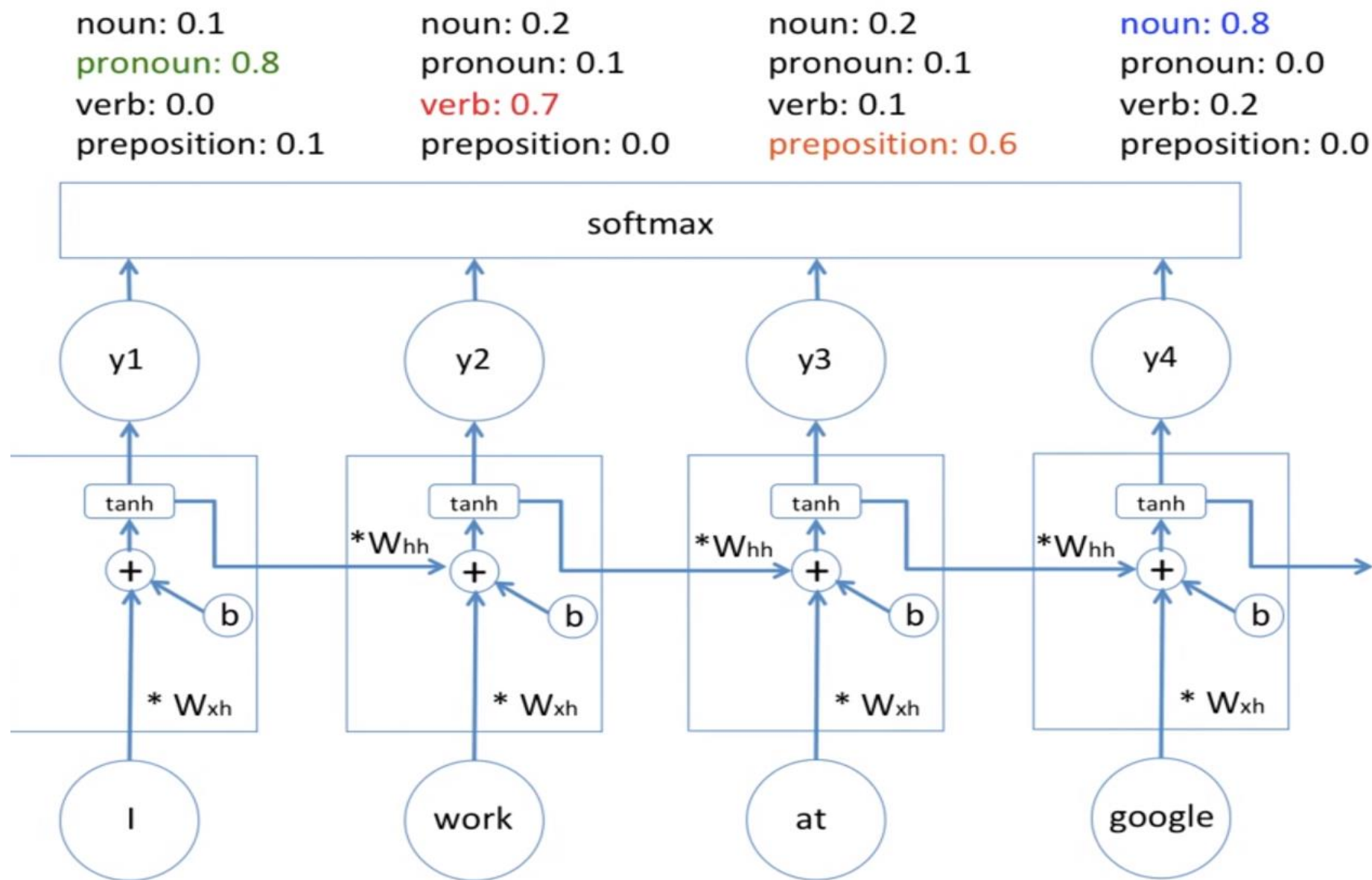
- 여러개의 데이터가 순서대로 입력되었을때 앞서 입력받은 데이터를 잠시 기억해 놓는 방법
- 모든 입력 값에 이 작업이 순서대로 이뤄지며 같은 층을 맴도는 것으로 보여 순환 신경망이라 부른다

Sequence is important for POS tagging

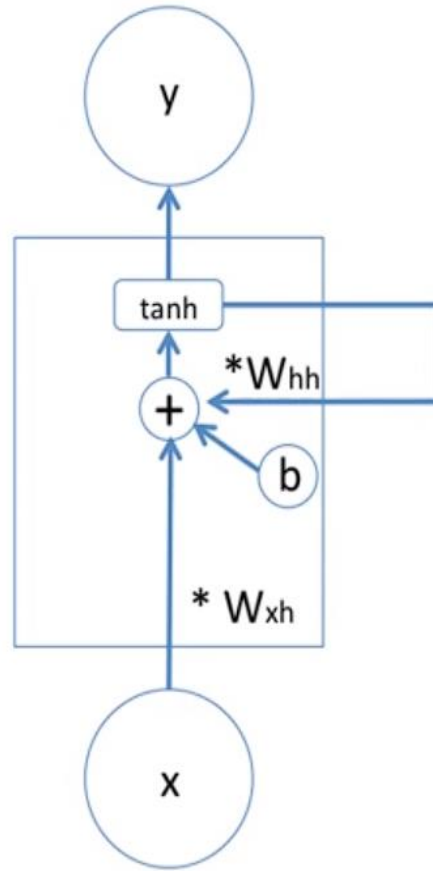


Sequence is important for POS tagging



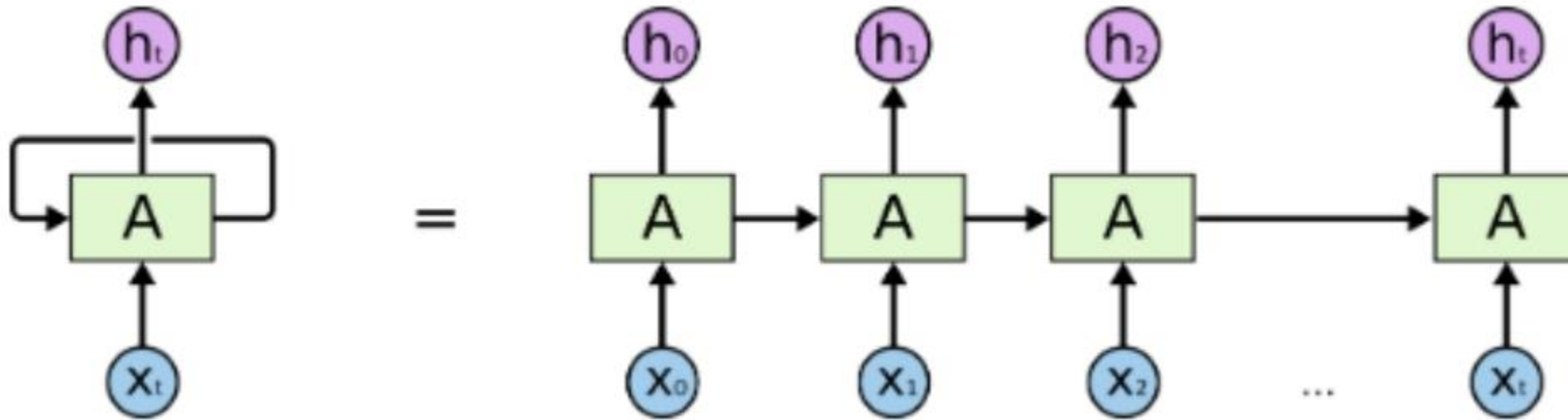


simplify of model diagram



$$h_t = \tanh(W_{xh} * x_t + W_{hh} * h_{t-1} + b)$$

Recurrent Neural Network



An unrolled recurrent neural network.

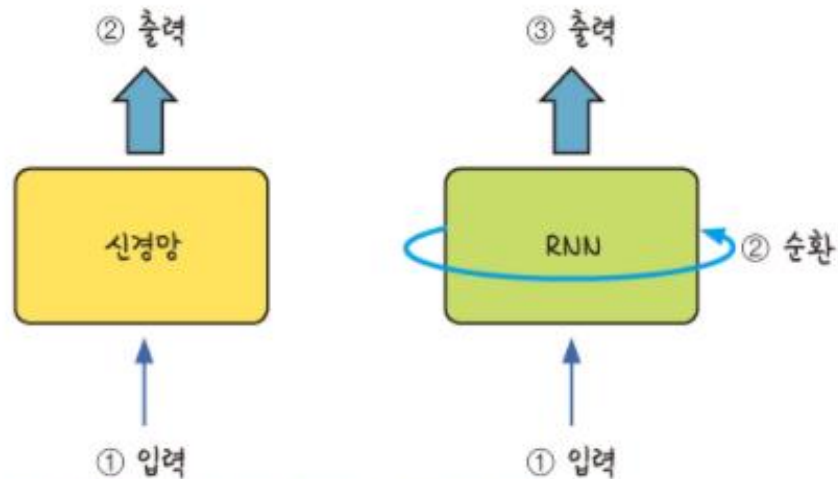


그림 17-1 일반 신경망과 순환 신경망의 차이

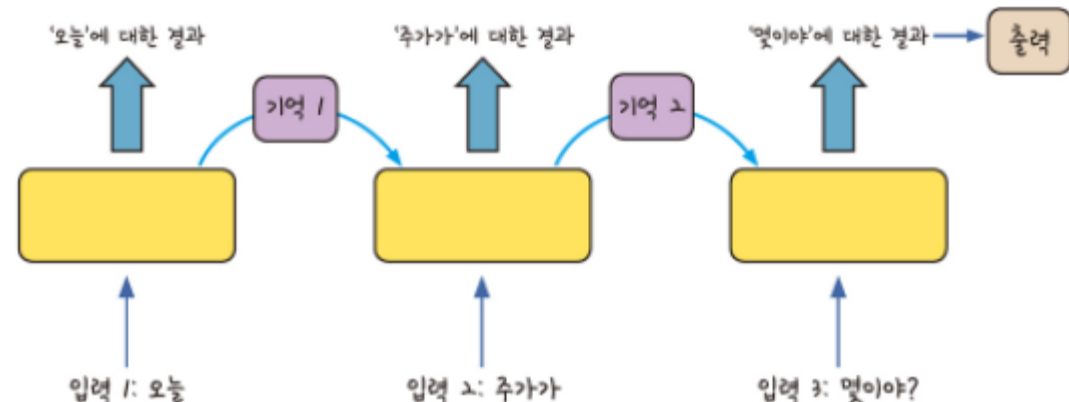


그림 17-2 "오늘 주가가 몇이야?"를 RNN이 처리하는 방식

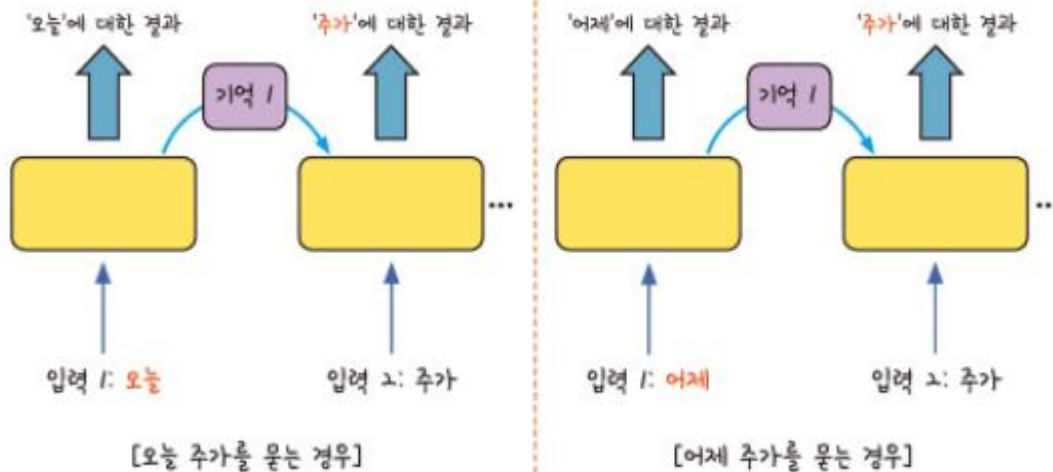
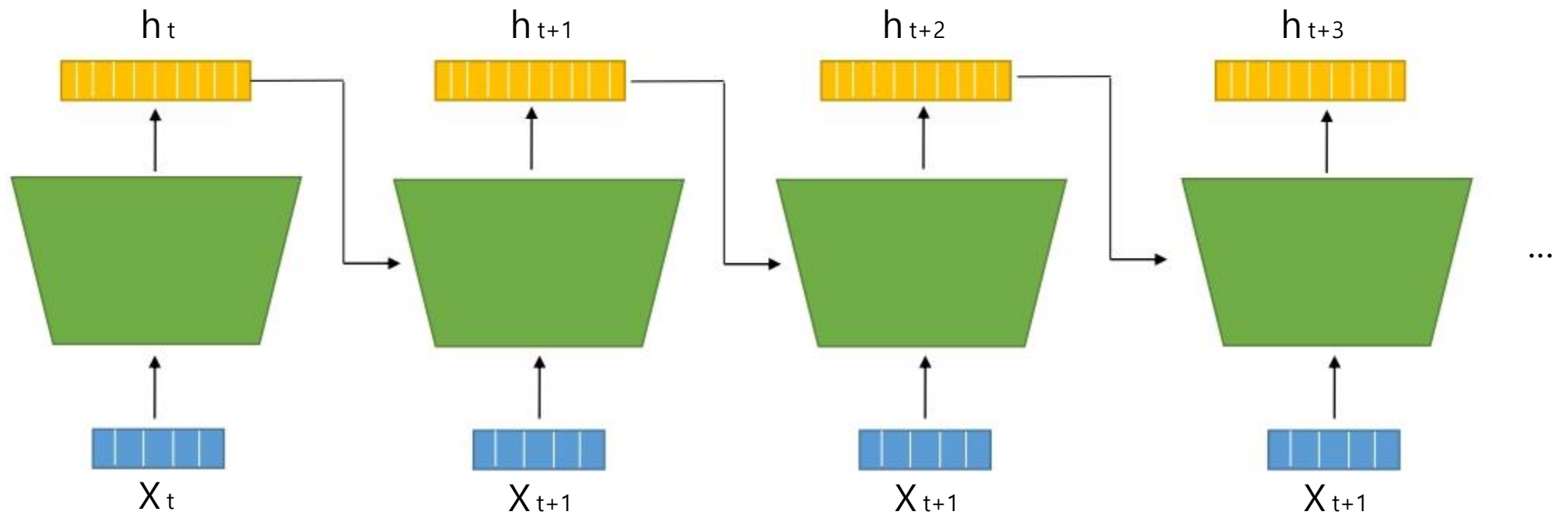


그림 17-3 RNN을 사용하는 이유

Recurrent Neural Network

Process both new inputs and model output of previous input!



Recurrent Neural Network

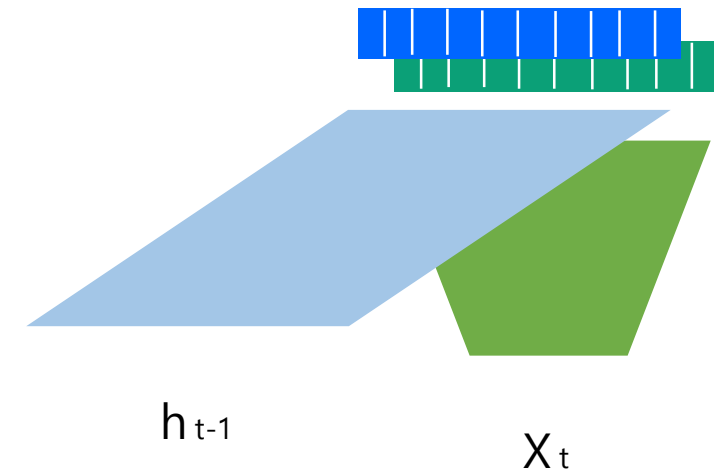
But exactly, how can we combine new input and previous output?



X_t

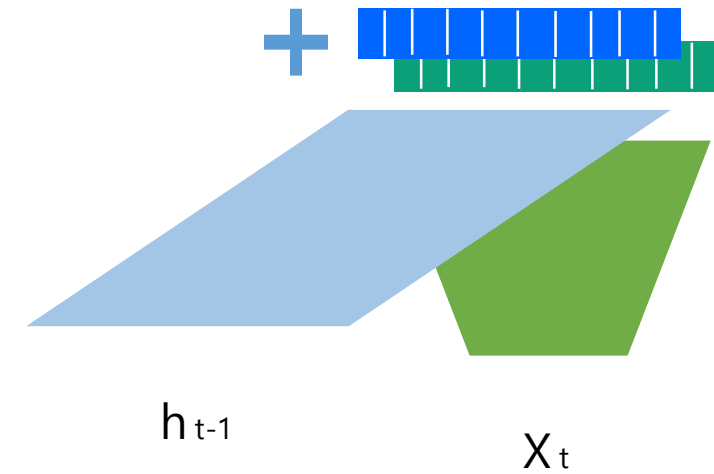
Recurrent Neural Network

But exactly, how can we combine new input and previous output?



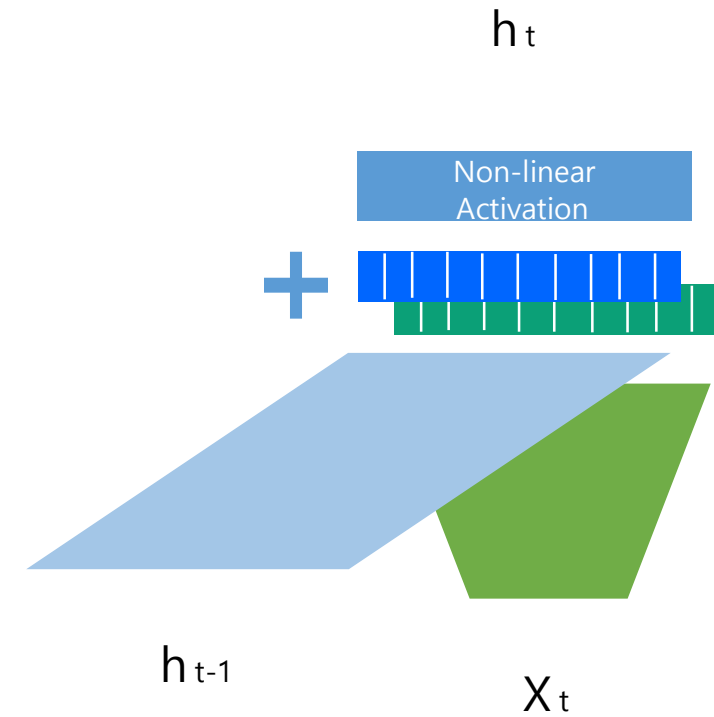
Recurrent Neural Network

But exactly, how can we combine new input and previous output?



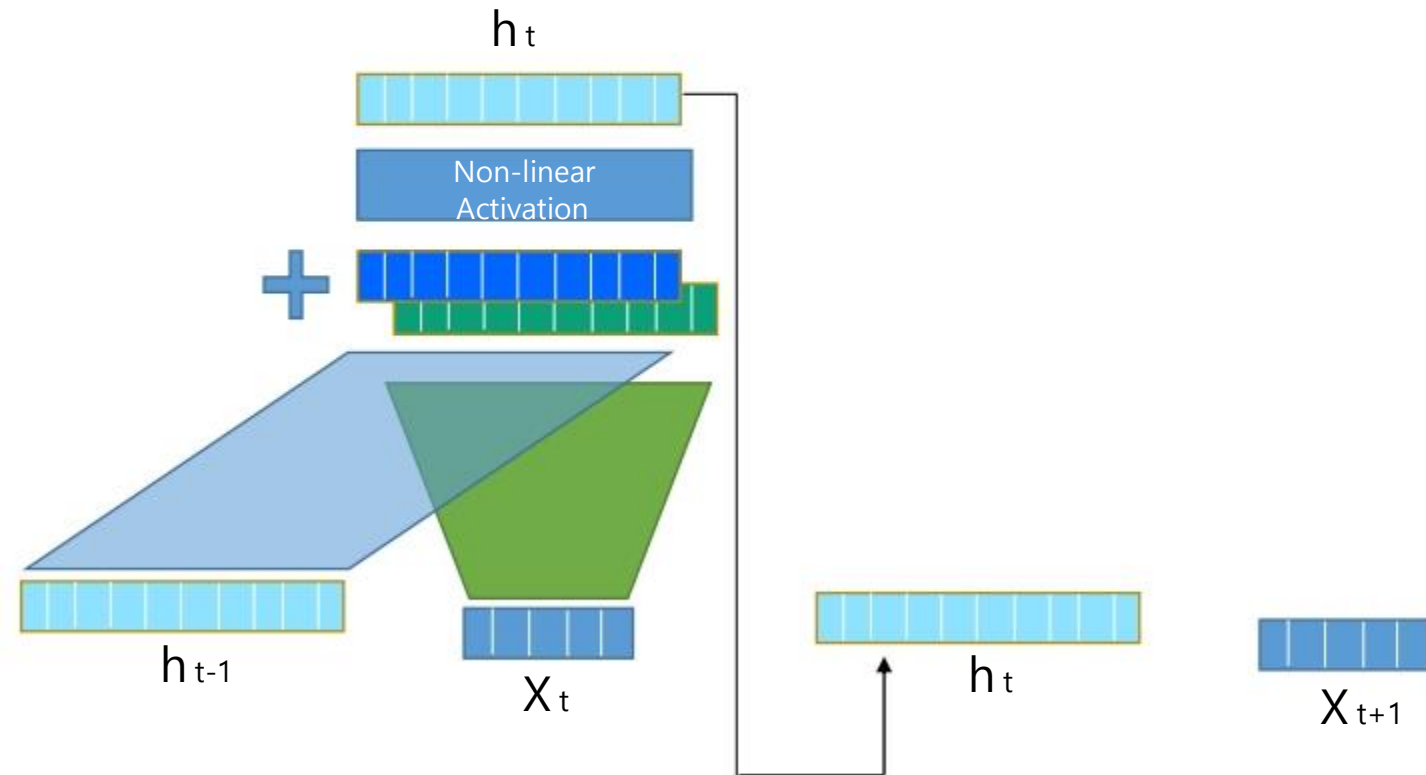
Recurrent Neural Network

But exactly, how can we combine new input and previous output?



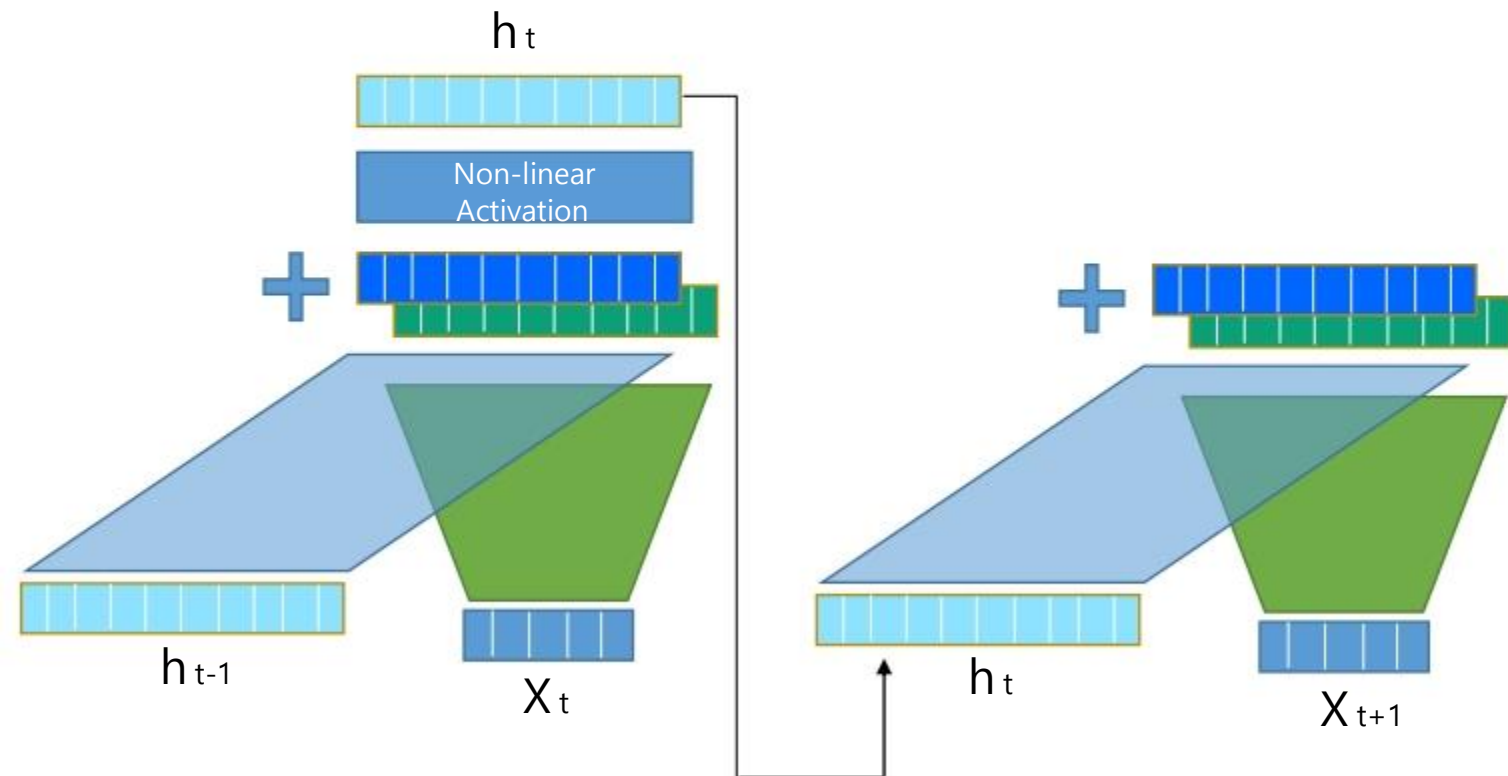
Recurrent Neural Network

But exactly, how can we combine new input and previous output?



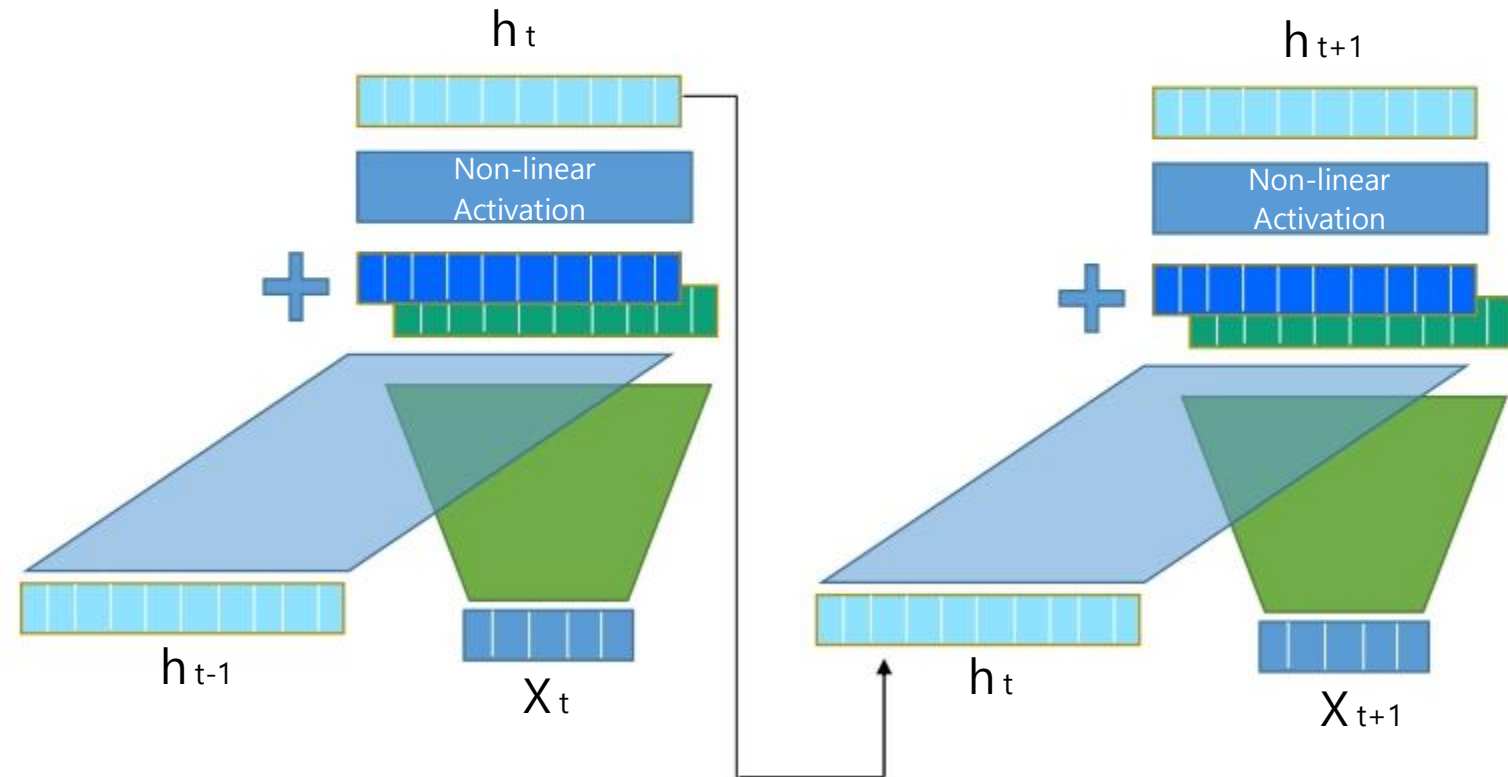
Recurrent Neural Network

But exactly, how can we combine new input and previous output?



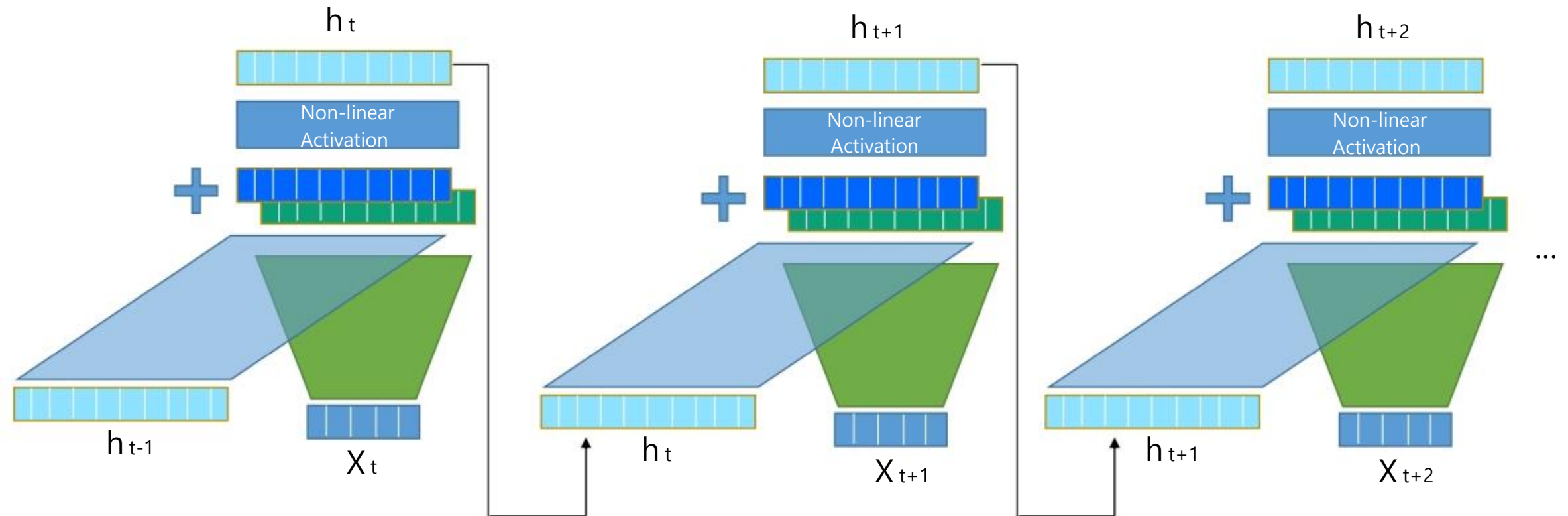
Recurrent Neural Network

But exactly, how can we combine new input and previous output?

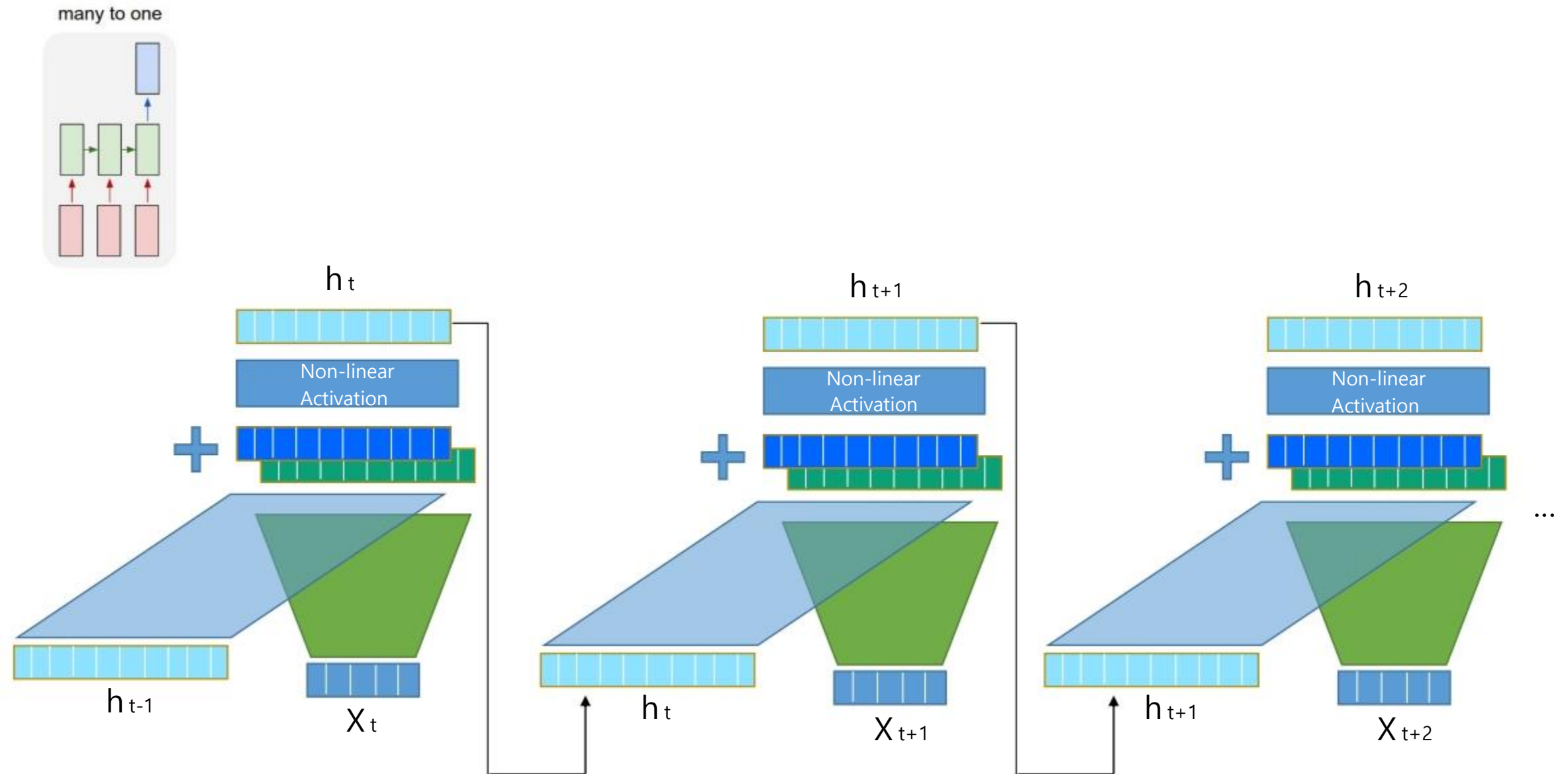


Recurrent Neural Network

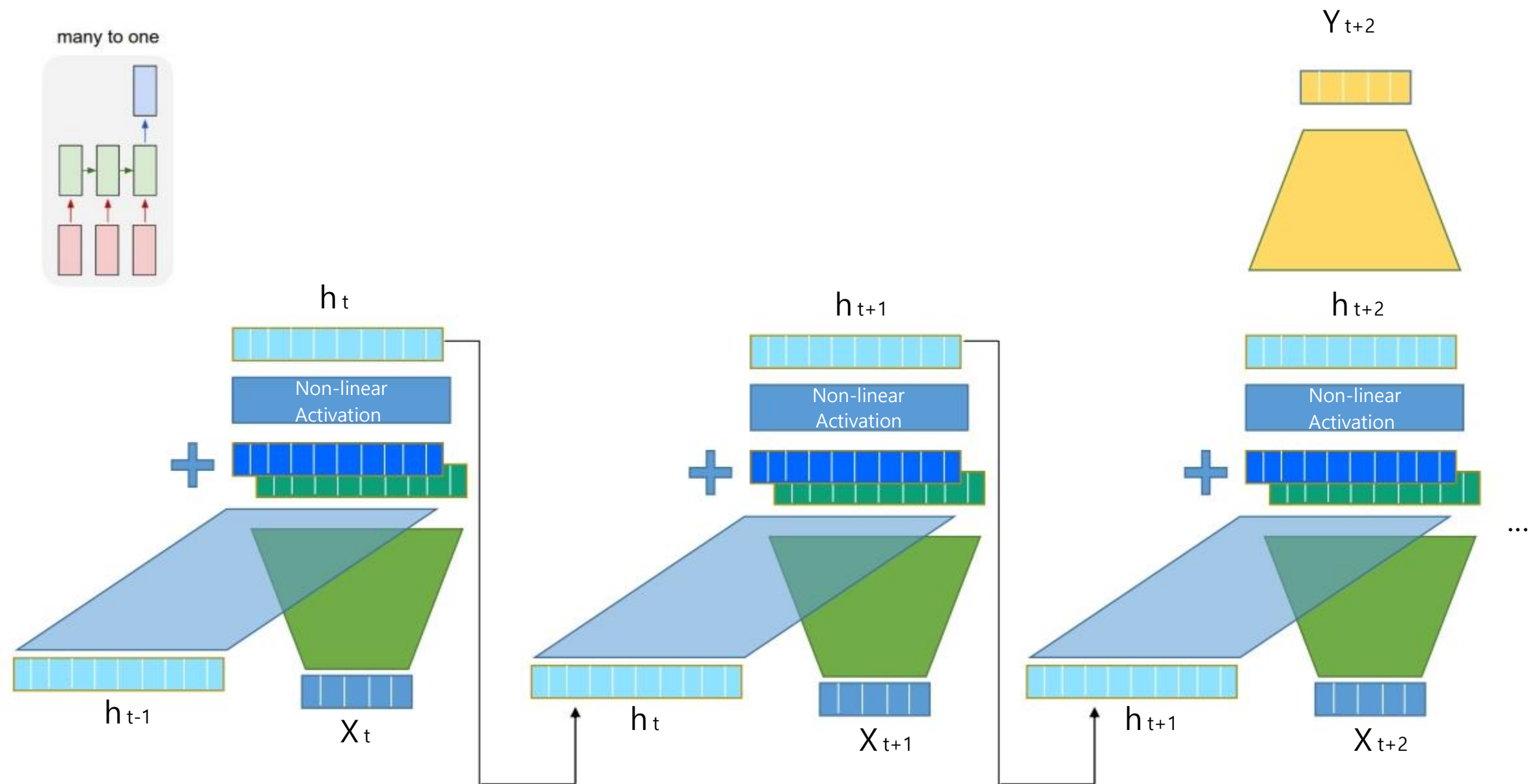
But exactly, how can we combine new input and previous output?



Recurrent Neural Network

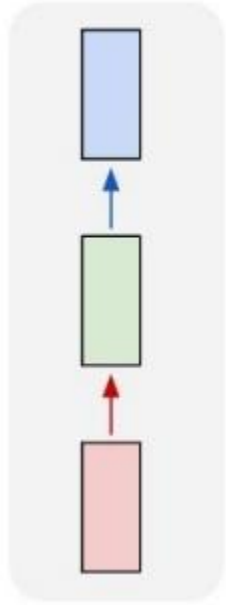


Recurrent Neural Network

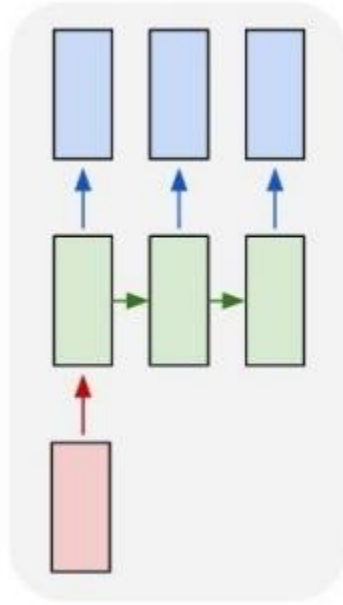


Types of Task Dealing with Sequential Data

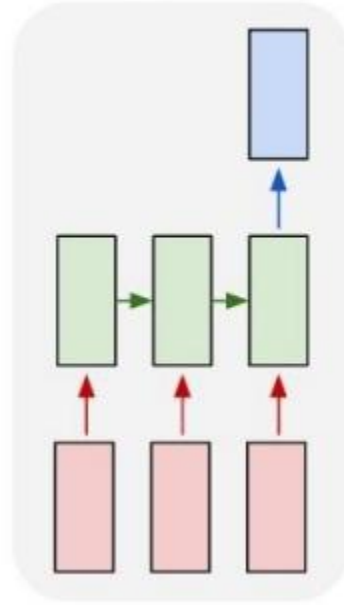
one to one



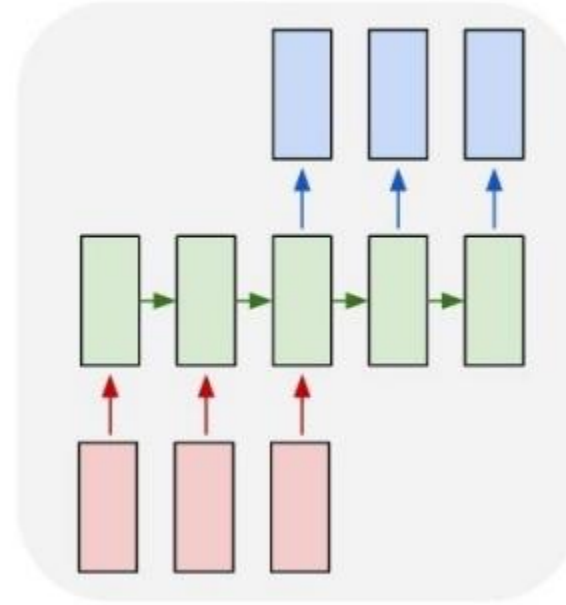
one to many



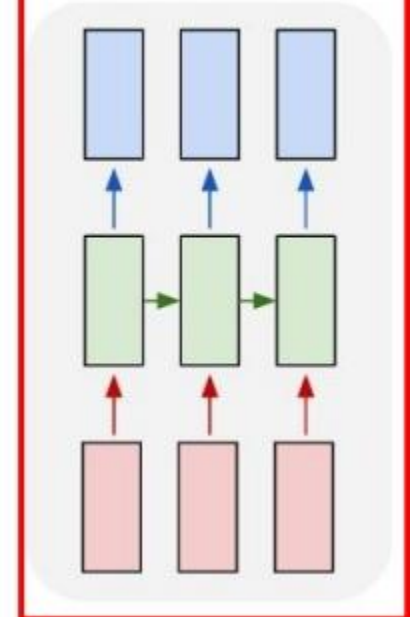
many to one



many to many

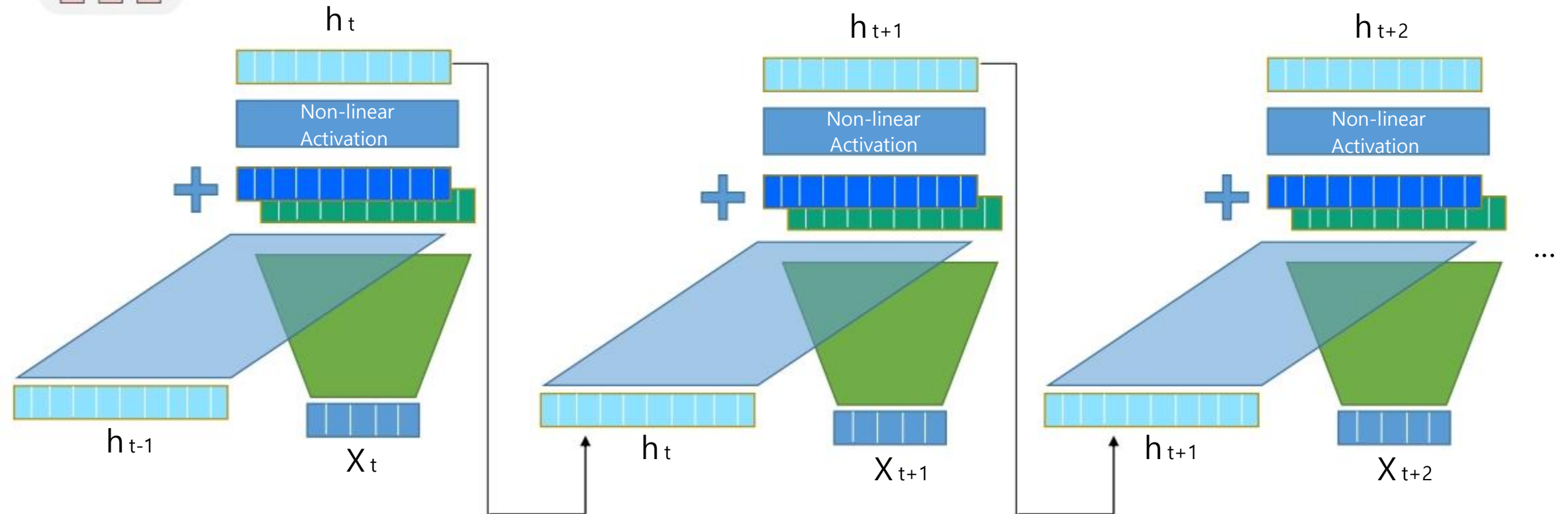
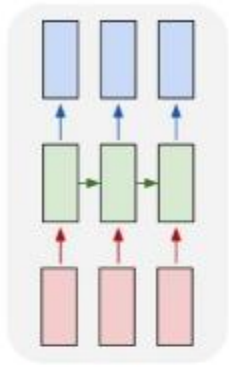


many to many

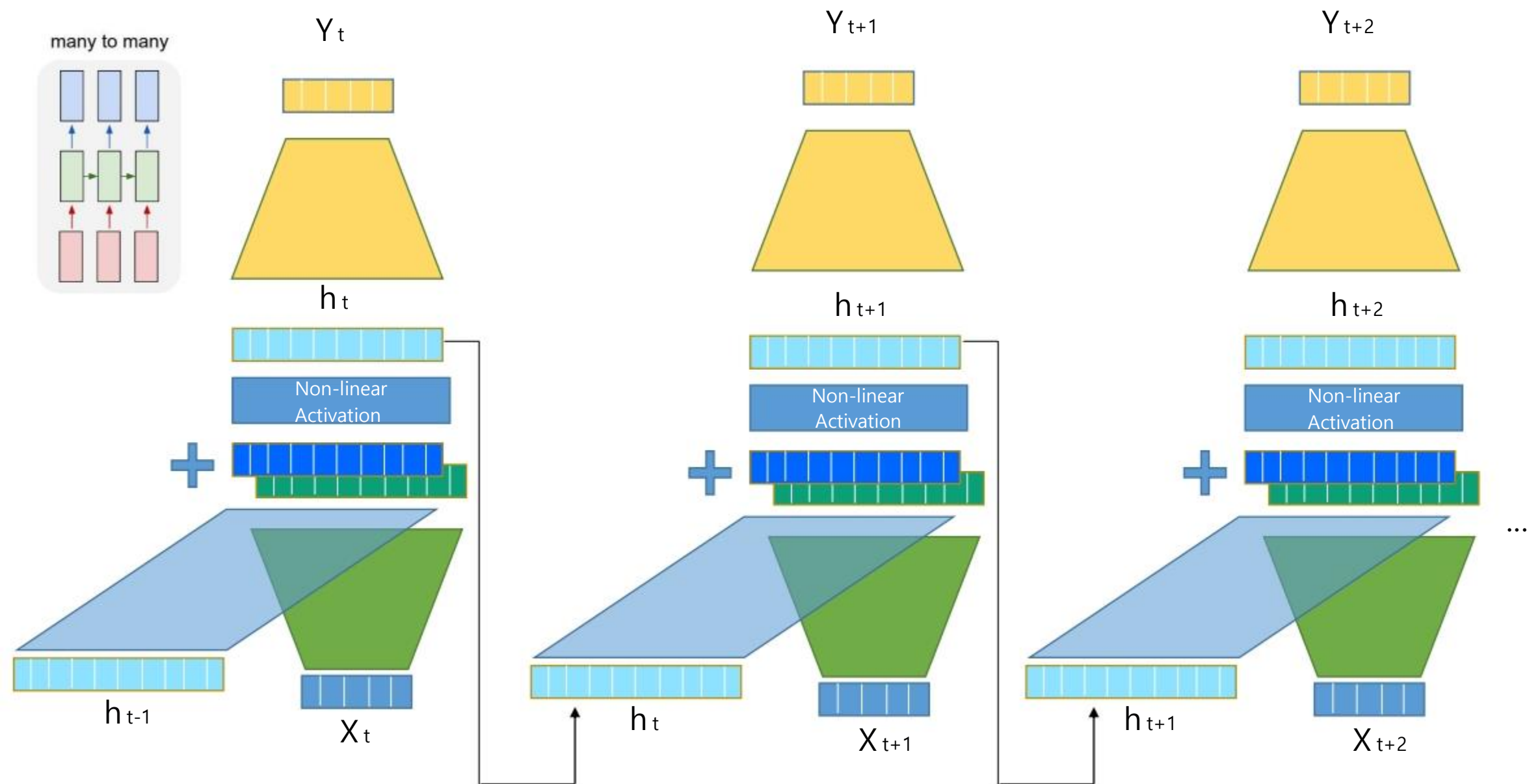


Recurrent Neural Network

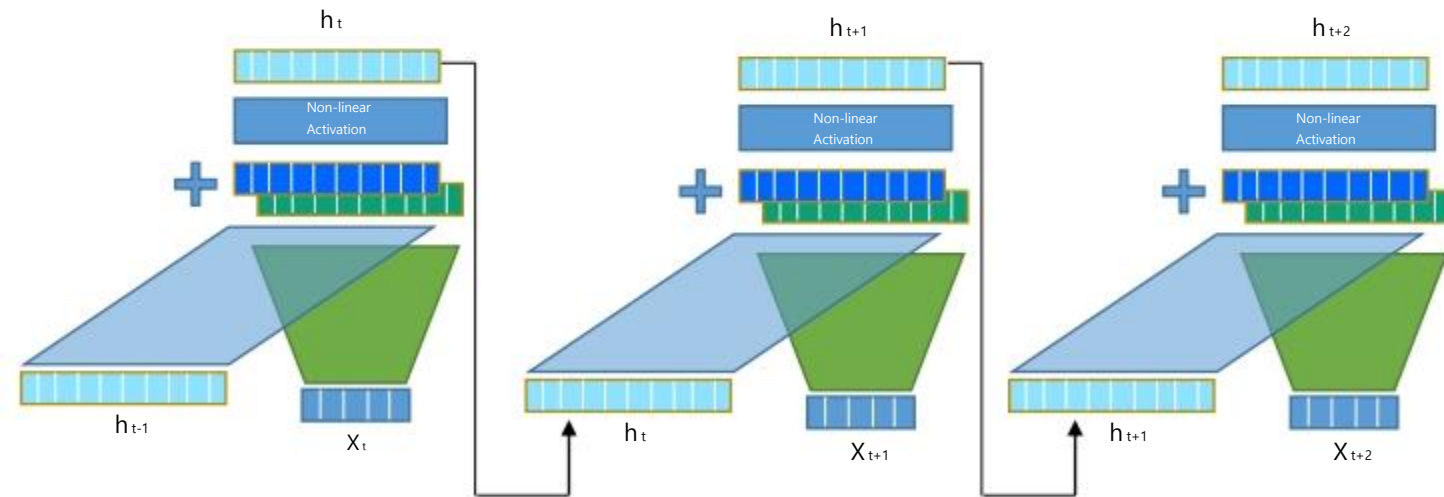
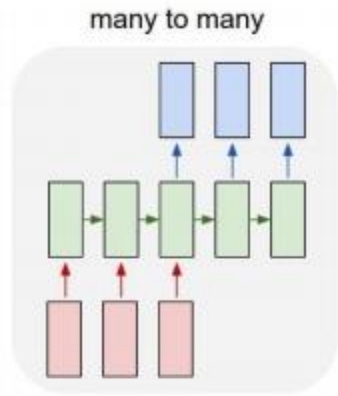
many to many



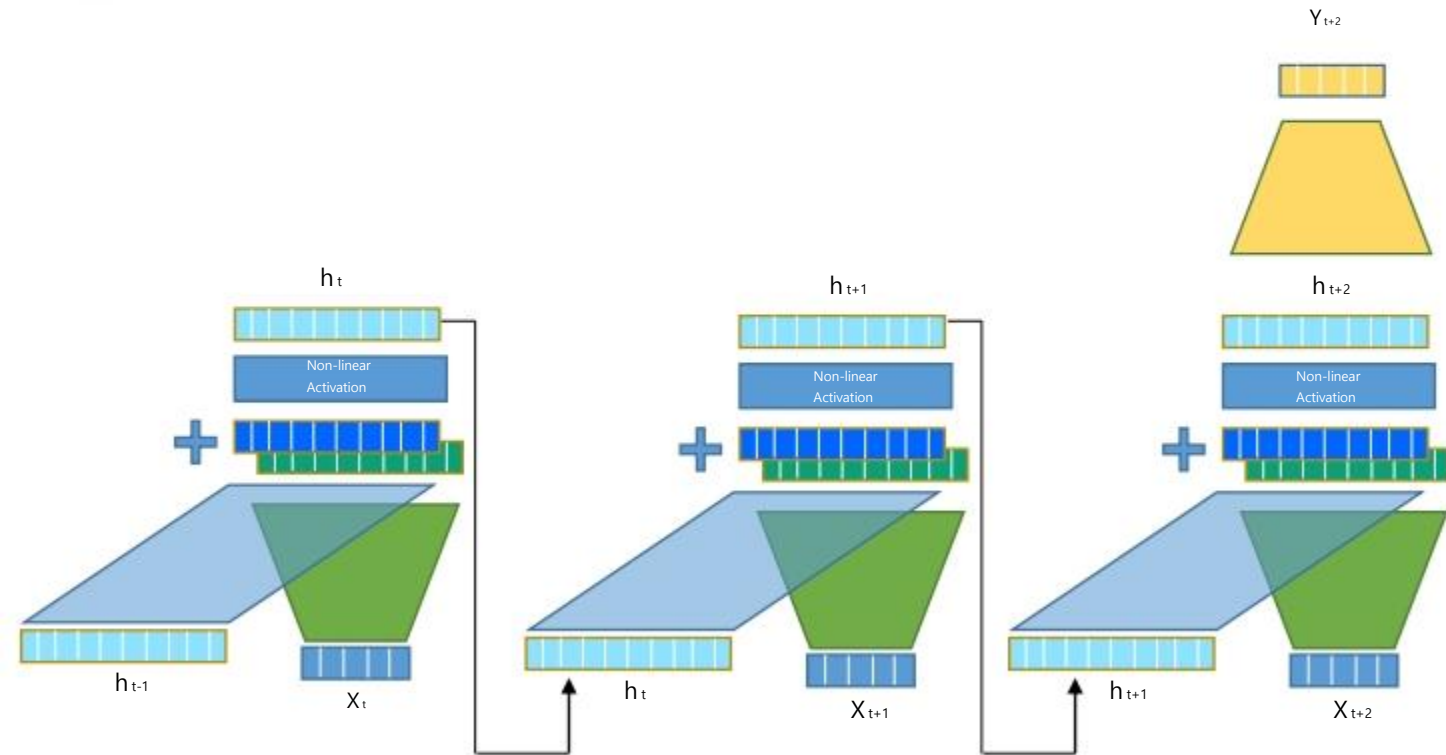
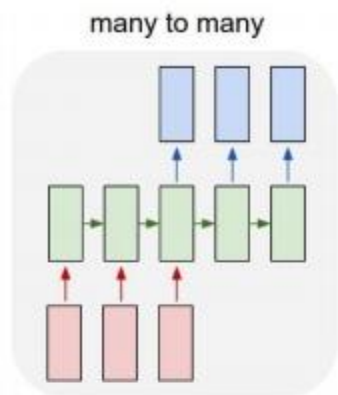
Recurrent Neural Network



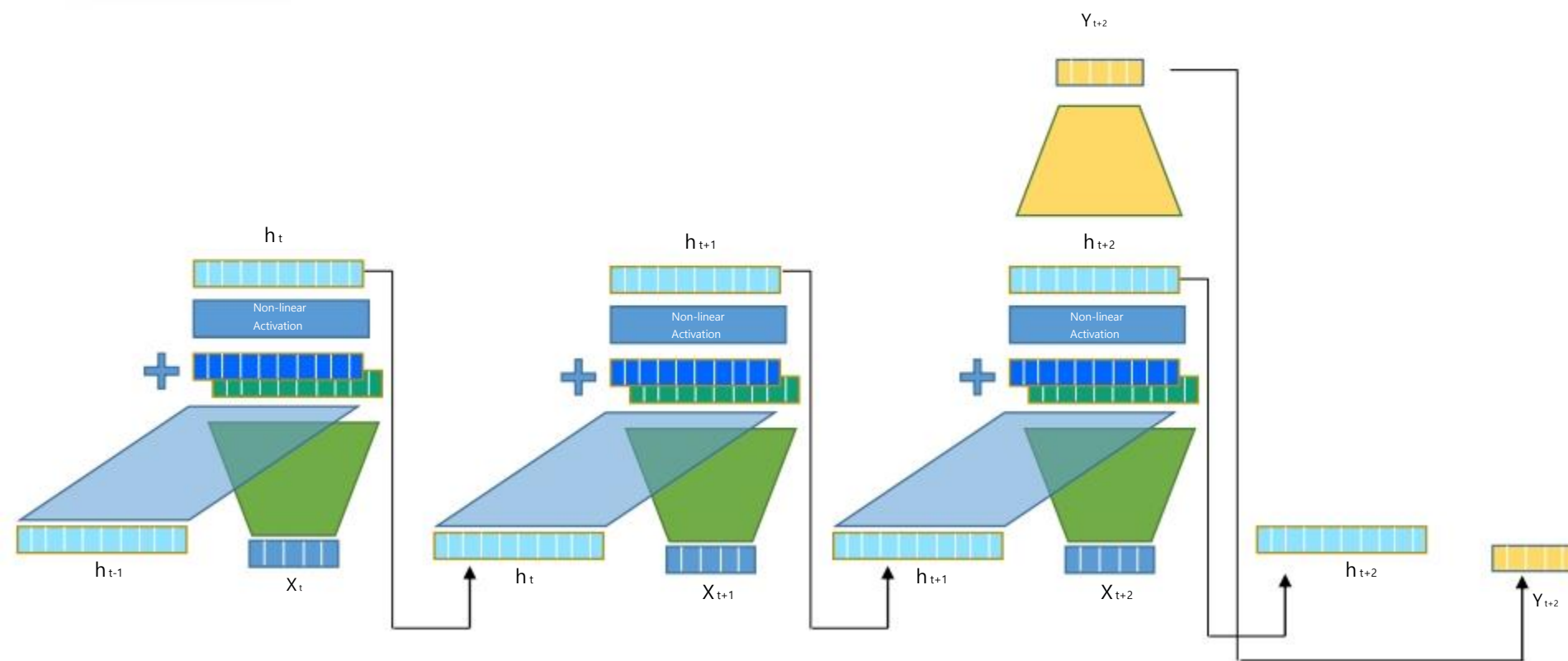
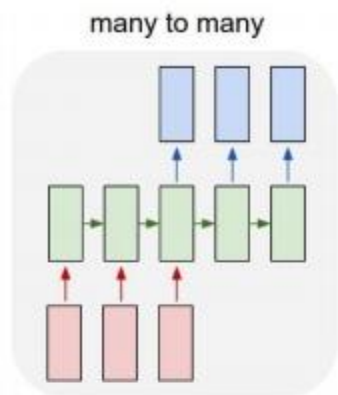
Recurrent Neural Network



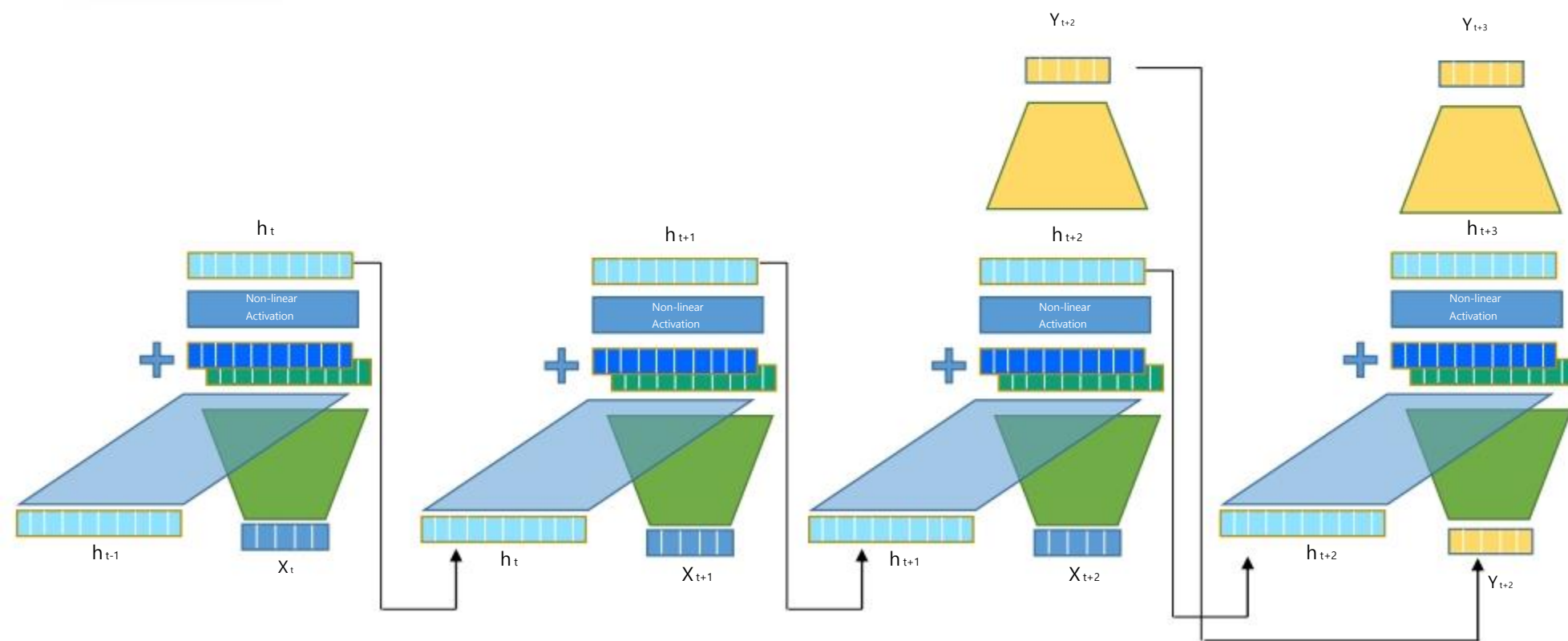
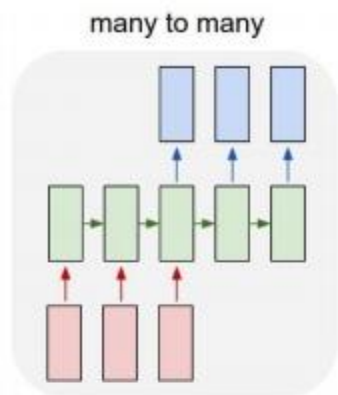
Recurrent Neural Network



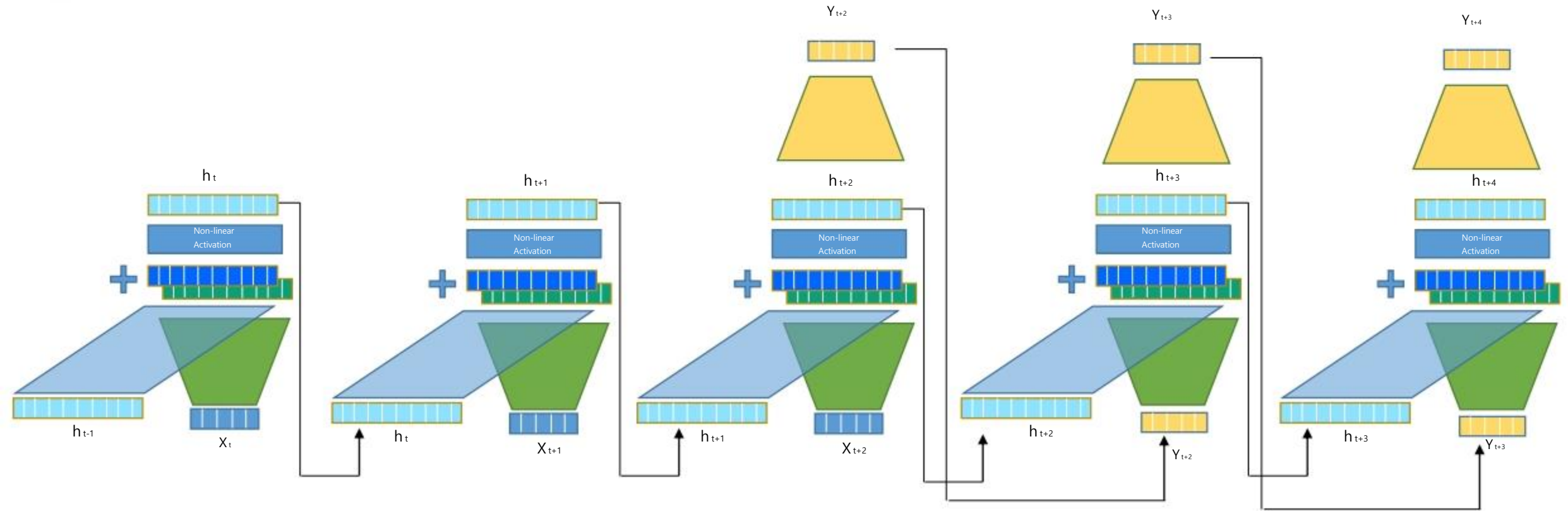
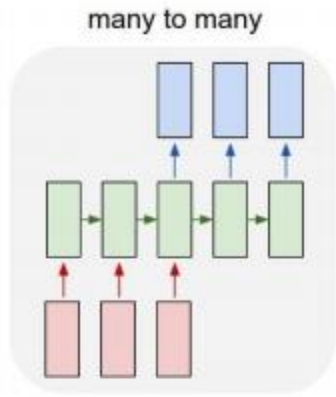
Recurrent Neural Network



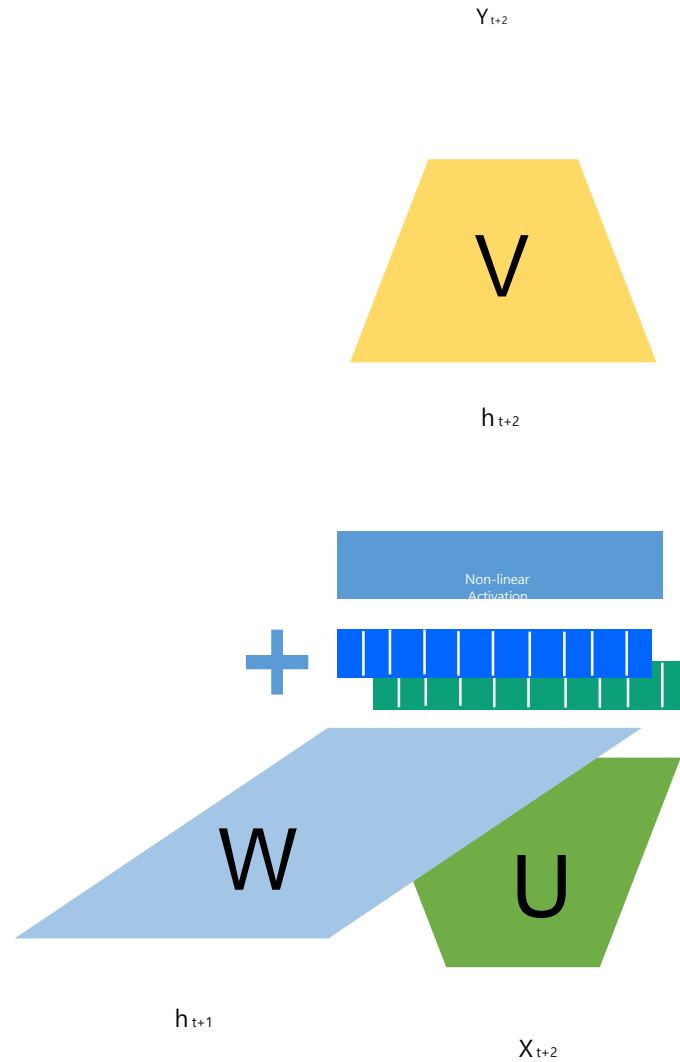
Recurrent Neural Network



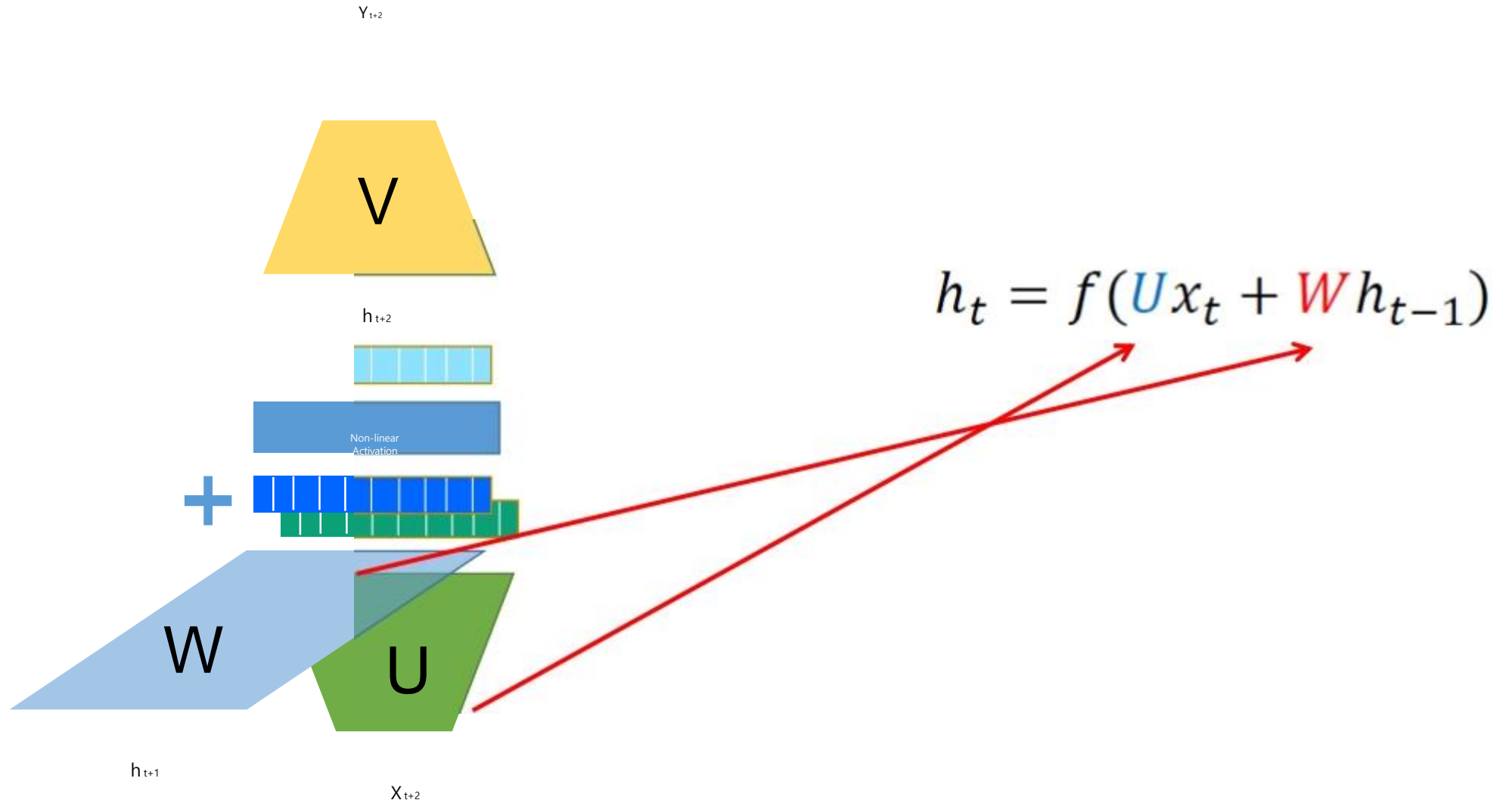
Recurrent Neural Network



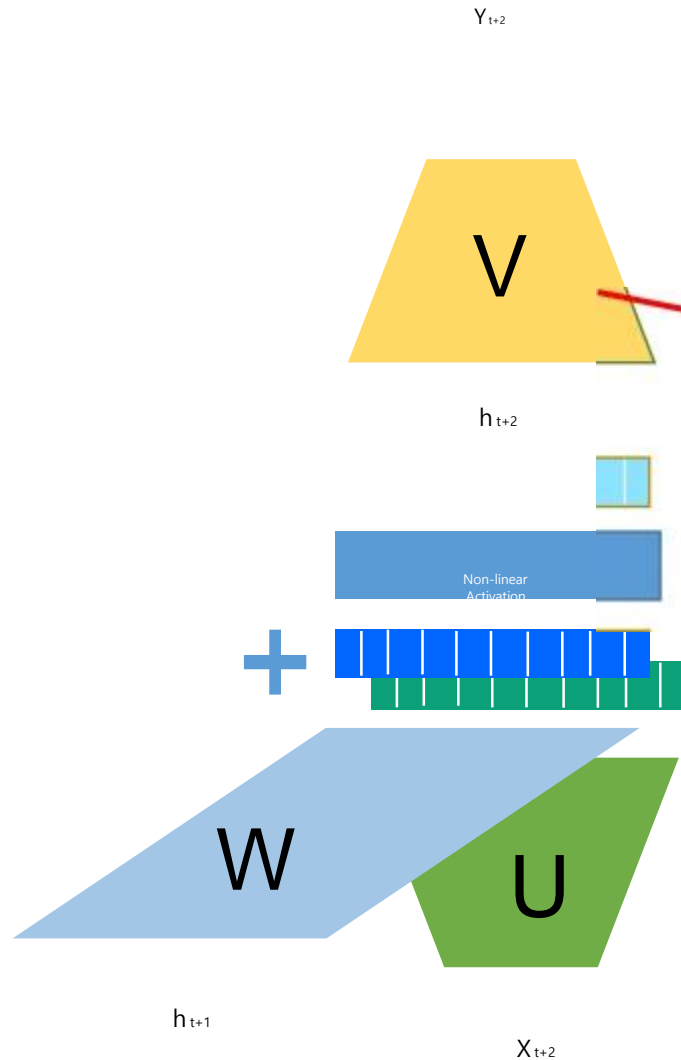
Recurrent Neural Network with Math



Recurrent Neural Network with Math



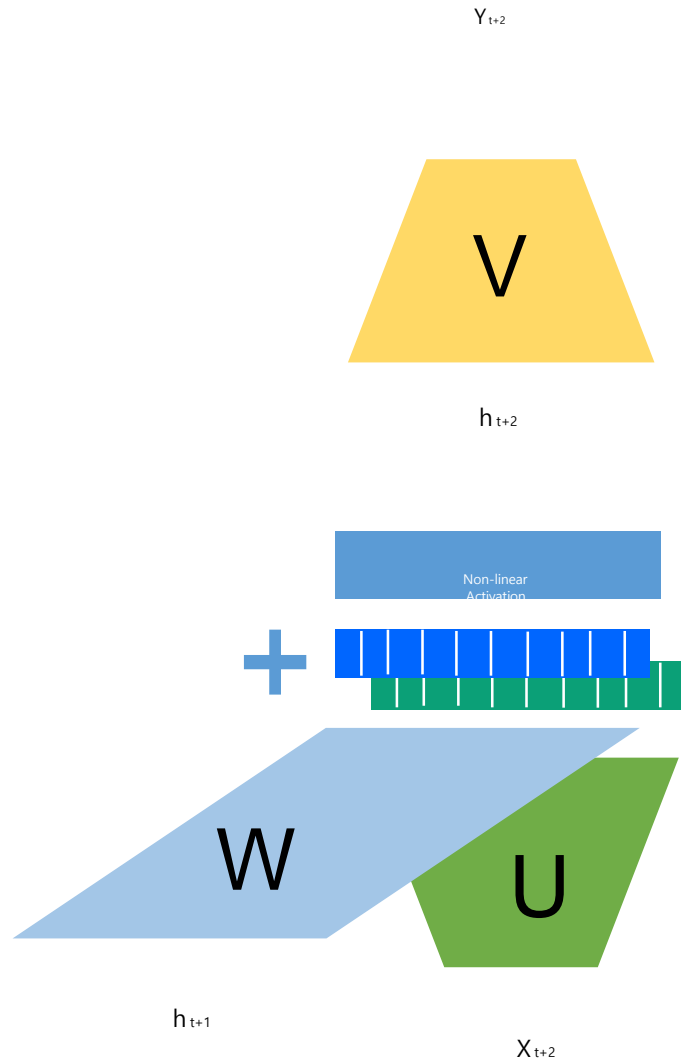
Recurrent Neural Network with Math



$$h_t = f(Ux_t + Wh_{t-1})$$

$$y_t = f(Vh_t)$$

Recurrent Neural Network with Math



$$h_t = f(Ux_t + Wh_{t-1})$$
$$y_t = f(Vh_t)$$

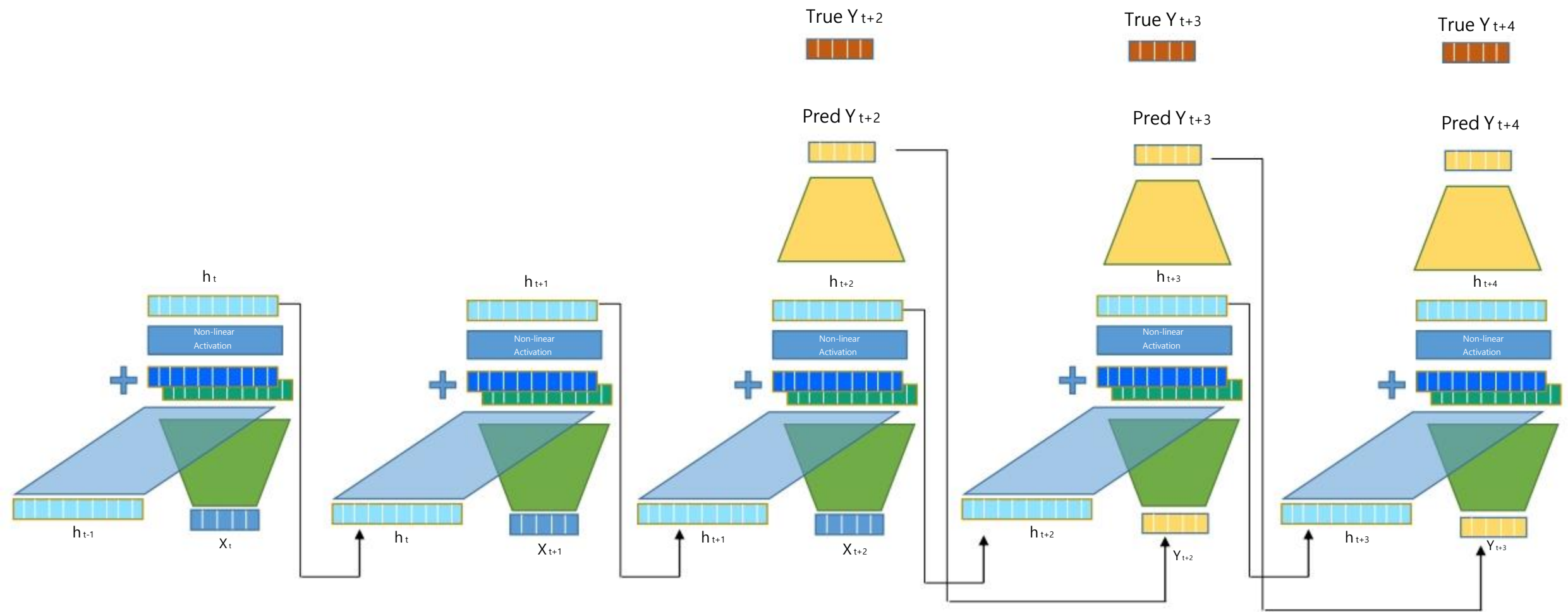
$f(x) = \tanh(x)$

$f(x) = x$

Okay, now we understand RNN model(hypothesis)

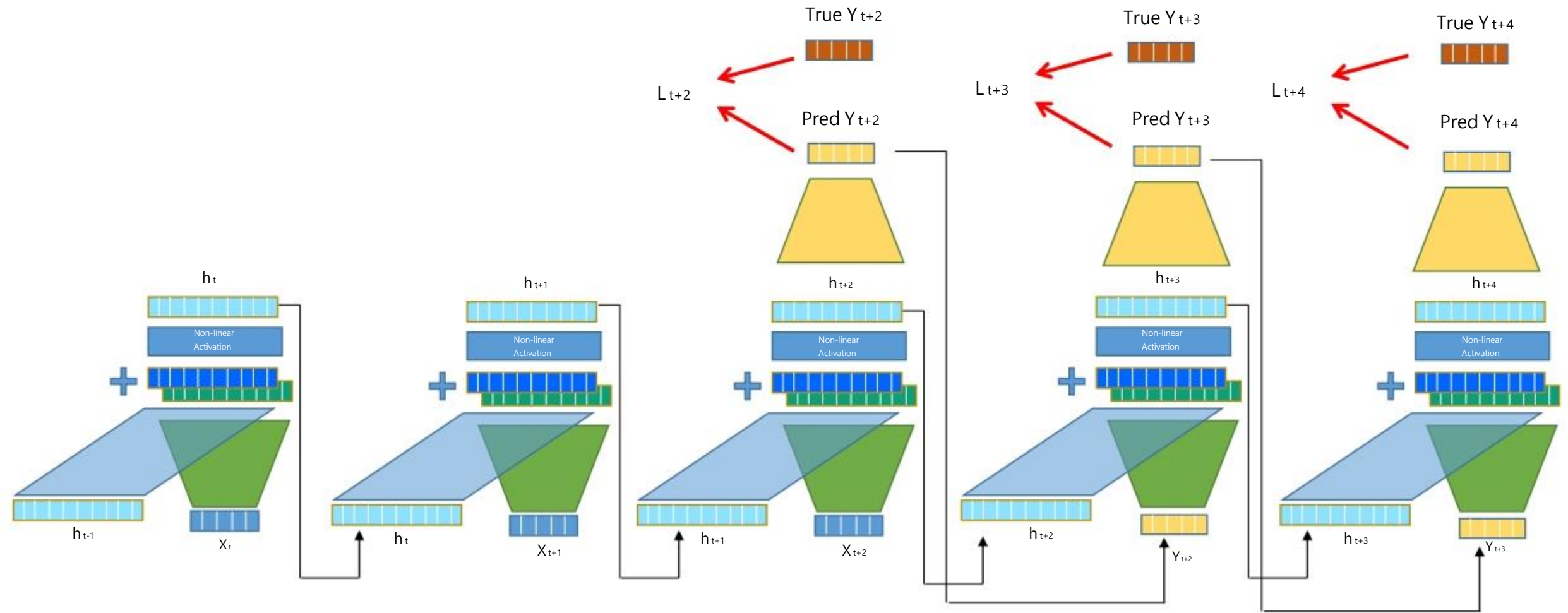
How can we evaluate it?

Calculate Loss of Recurrent Neural Network



Calculate Loss of Recurrent Neural Network

$$Loss(\theta) = \sum_t loss(y_{true,t}, y_{pred,t})$$

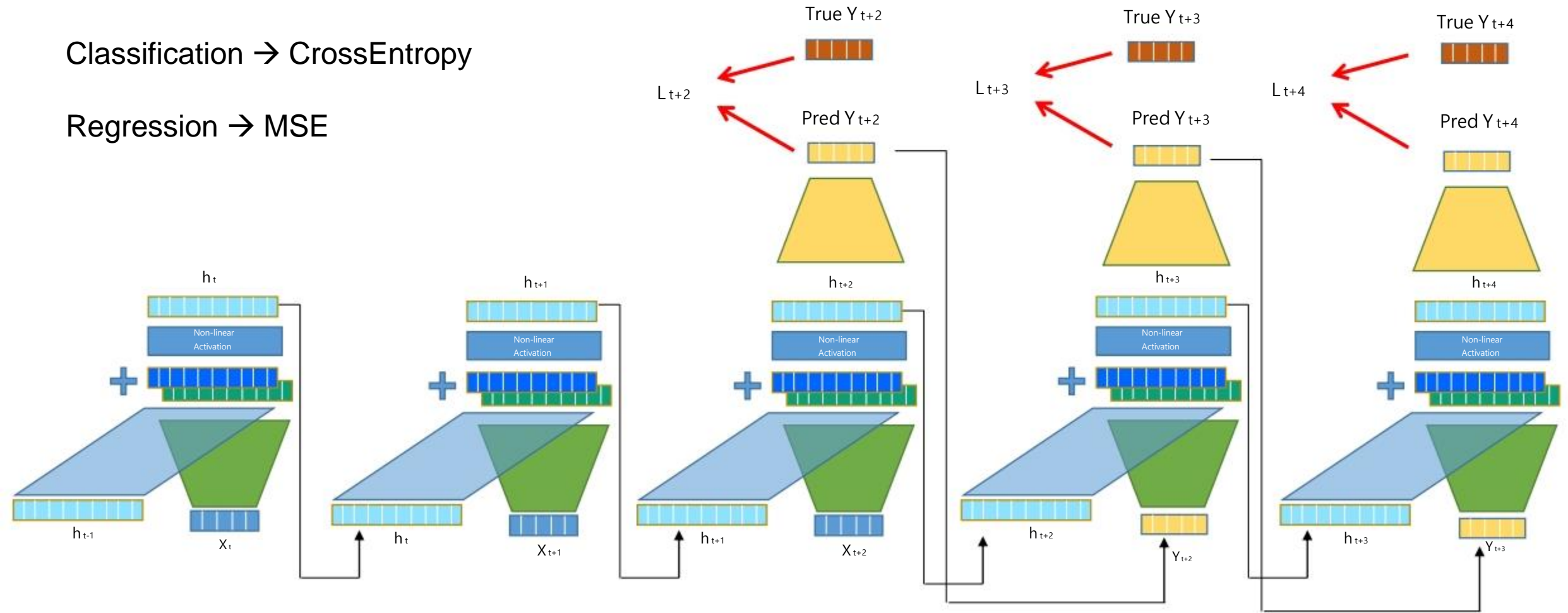


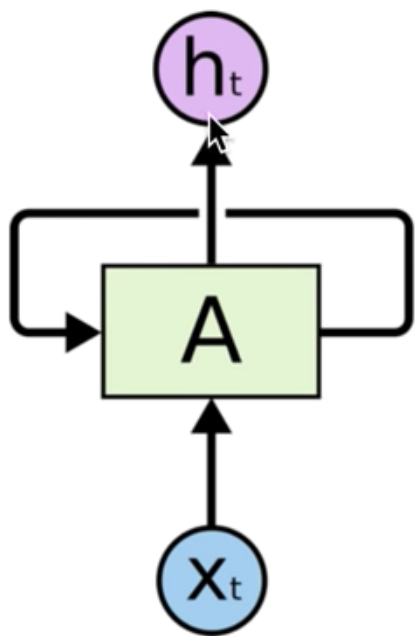
Calculate Loss of Recurrent Neural Network

$$Loss(\theta) = \sum_t loss(y_{true,t}, y_{pred,t})$$

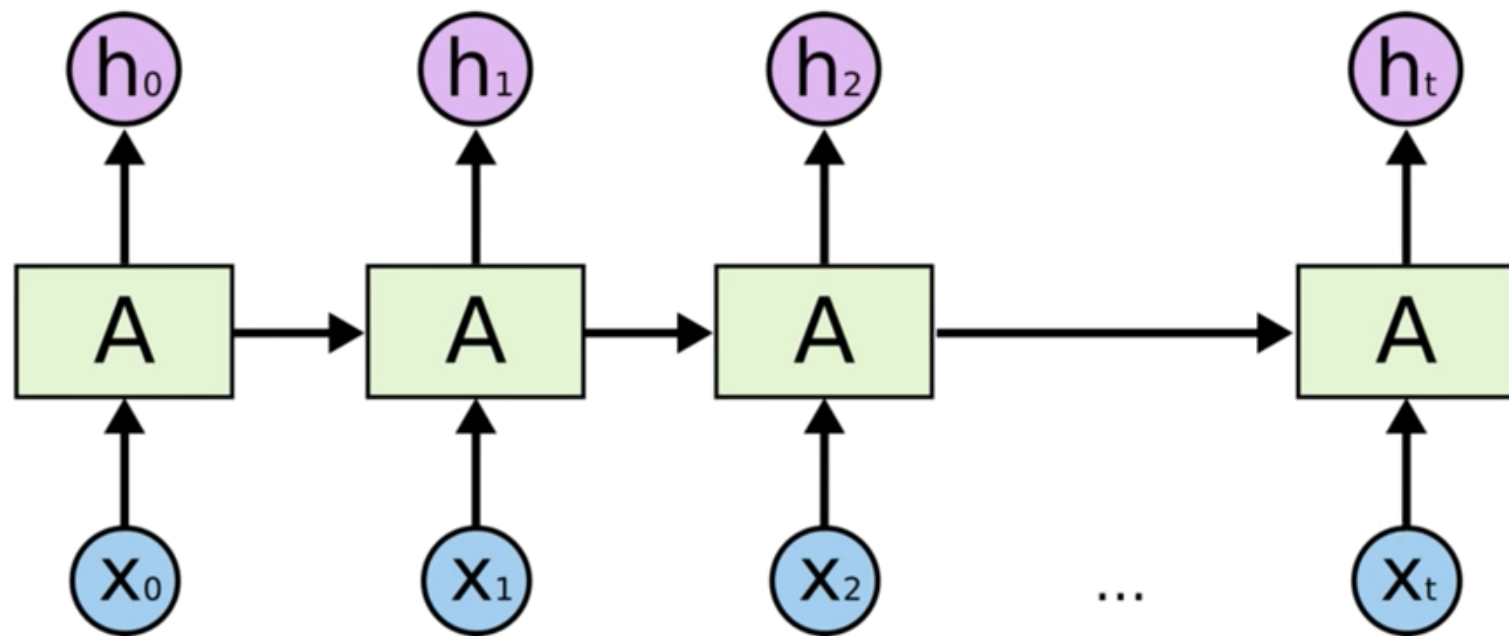
Classification → CrossEntropy

Regression → MSE

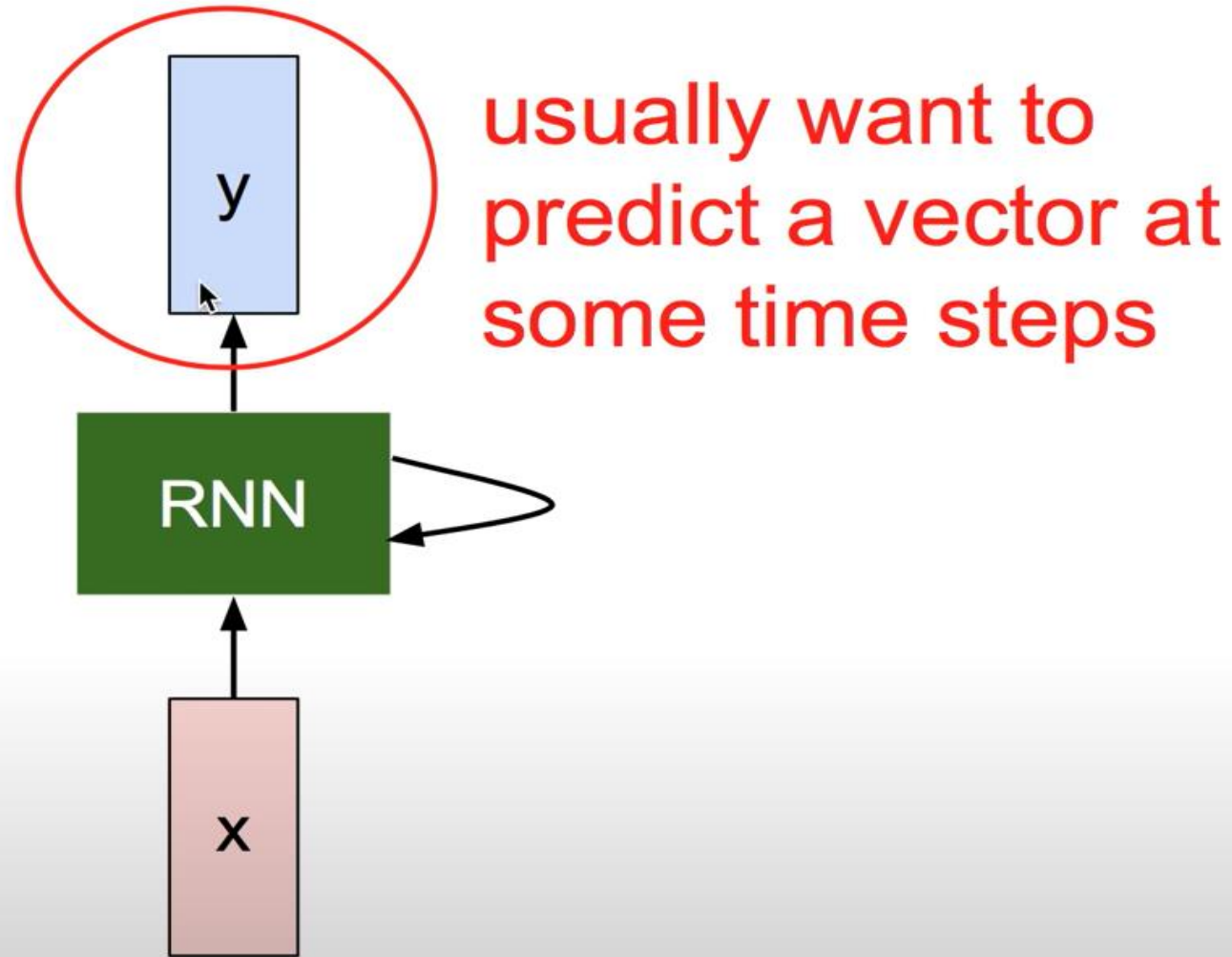




=



Recurrent Neural Network



Recurrent Neural Network

We can process a sequence of vectors \mathbf{x} by applying a recurrence formula at every time step:

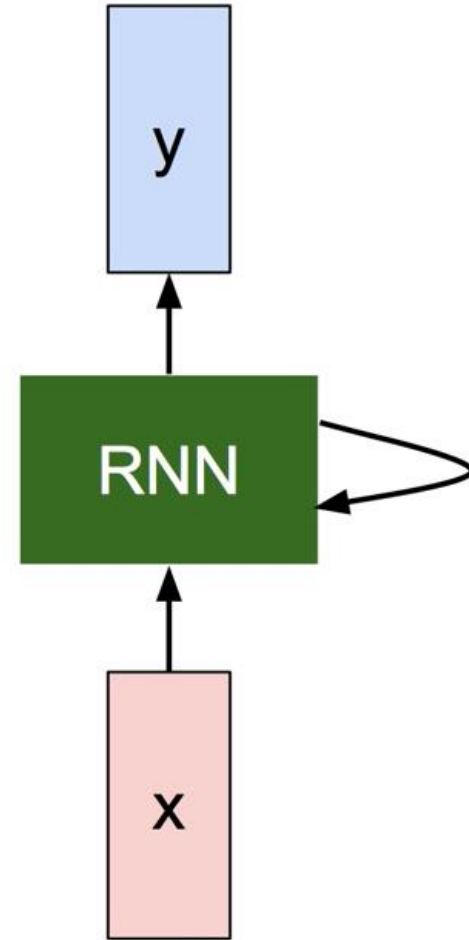
$$\boxed{h_t} = \boxed{f_W}(\boxed{h_{t-1}}, \boxed{x_t})$$

new state

some function with parameters W

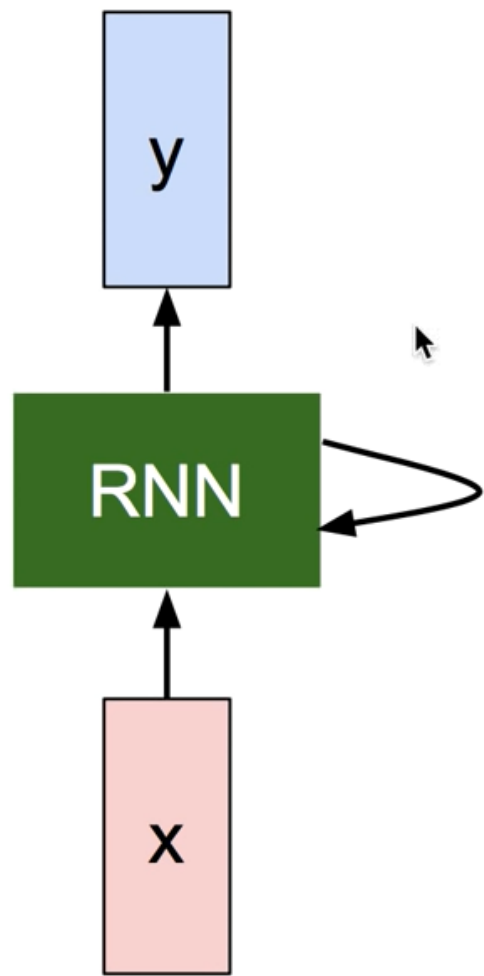
old state

input vector at some time step



(Vanilla) Recurrent Neural Network

The state consists of a single “*hidden*” vector \mathbf{h} :



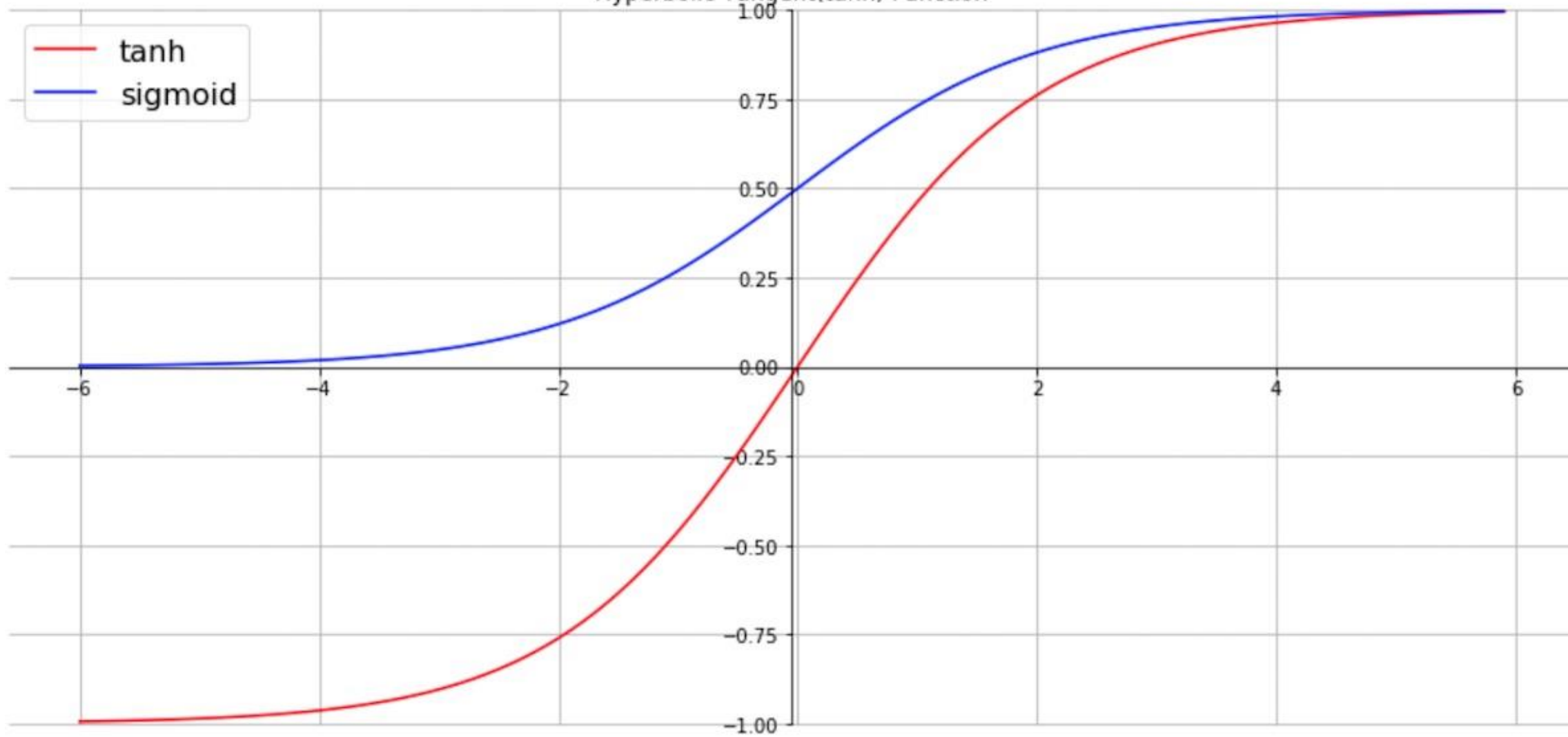
$$h_t = f_W(h_{t-1}, x_t)$$



$$h_t = \tanh(W_{hh}h_{t-1} + W_{xh}x_t)$$

$$y_t = W_{hy}h_t$$

Hyperbolic Tangent(tanh) Function

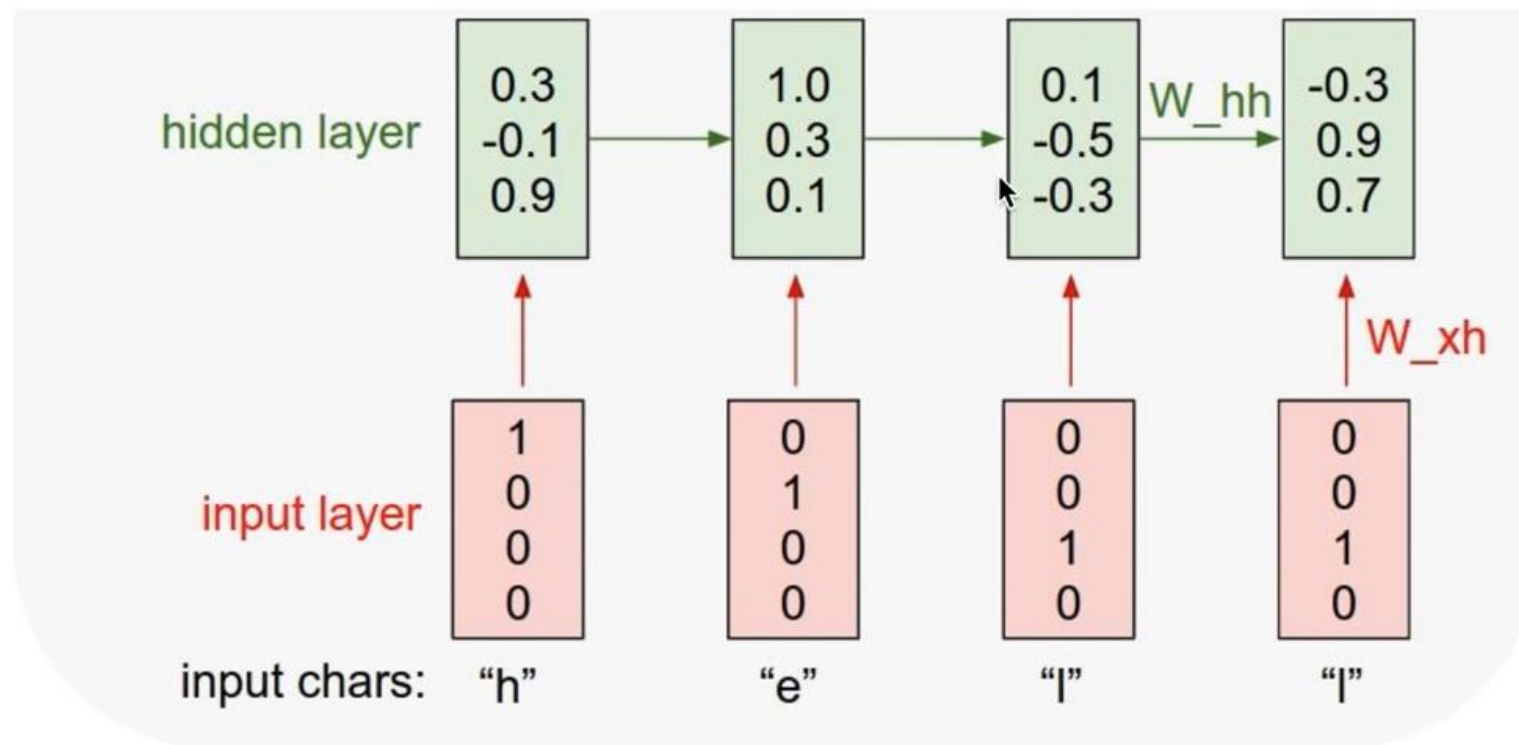


Character-level language model example

Vocabulary:
[h,e,l,o]

Example training sequence:
“hello”

$$h_t = \tanh(W_{hh}h_{t-1} + W_{xh}x_t)$$

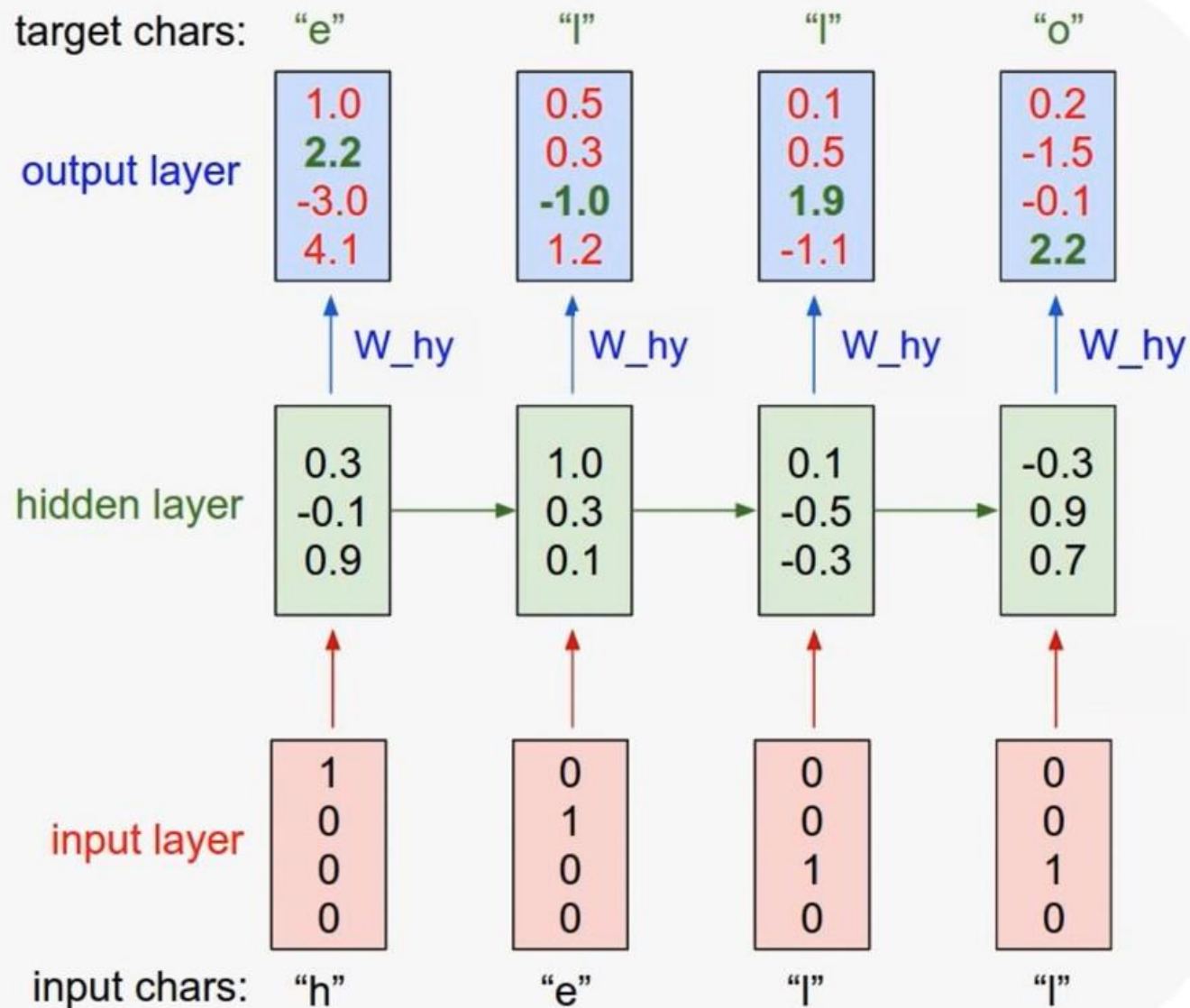


Character-level language model example

$$y_t = W_{hy}h_t$$

Vocabulary:
[h,e,l,o]

Example training sequence:
“hello”



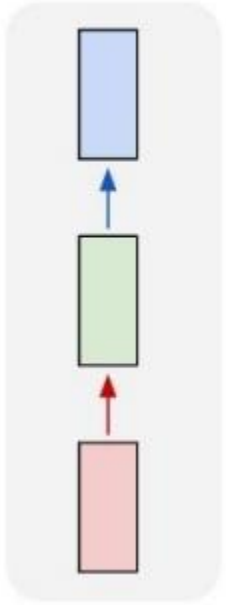
RNN applications

https://github.com/TensorFlowKR/awesome_tensorflow_implementations

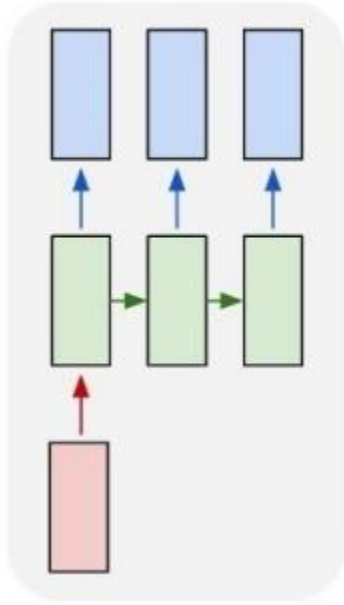
- Language Modeling
- Speech Recognition
- Machine Translation
- Conversation Modeling/Question Answering
- Image/Video Captioning
- Image/Music/Dance Generation

Types of Task Dealing with Sequential Data

one to one

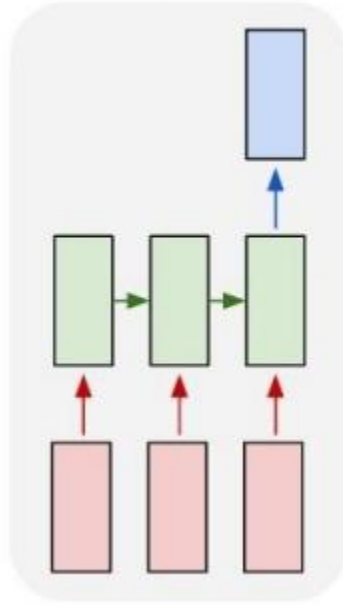


one to many



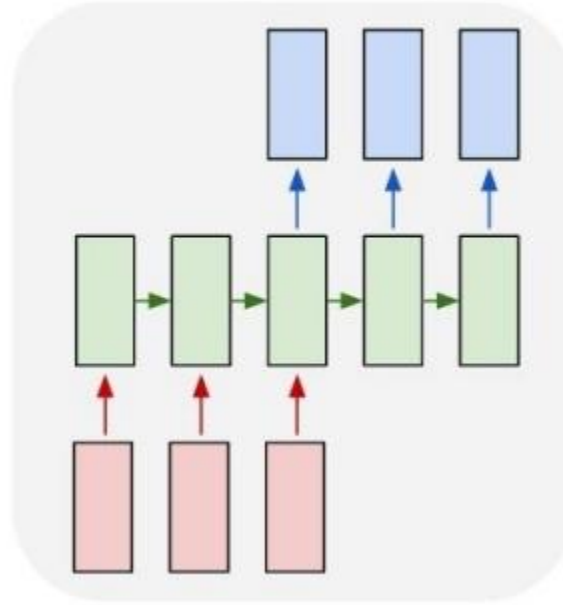
이미지로
부터 단어
추출

many to one



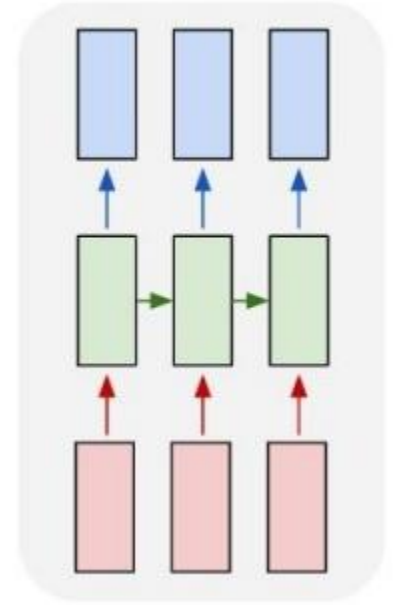
감정분석
시계열

many to many



문자열로
부터 다른
문자열

many to many

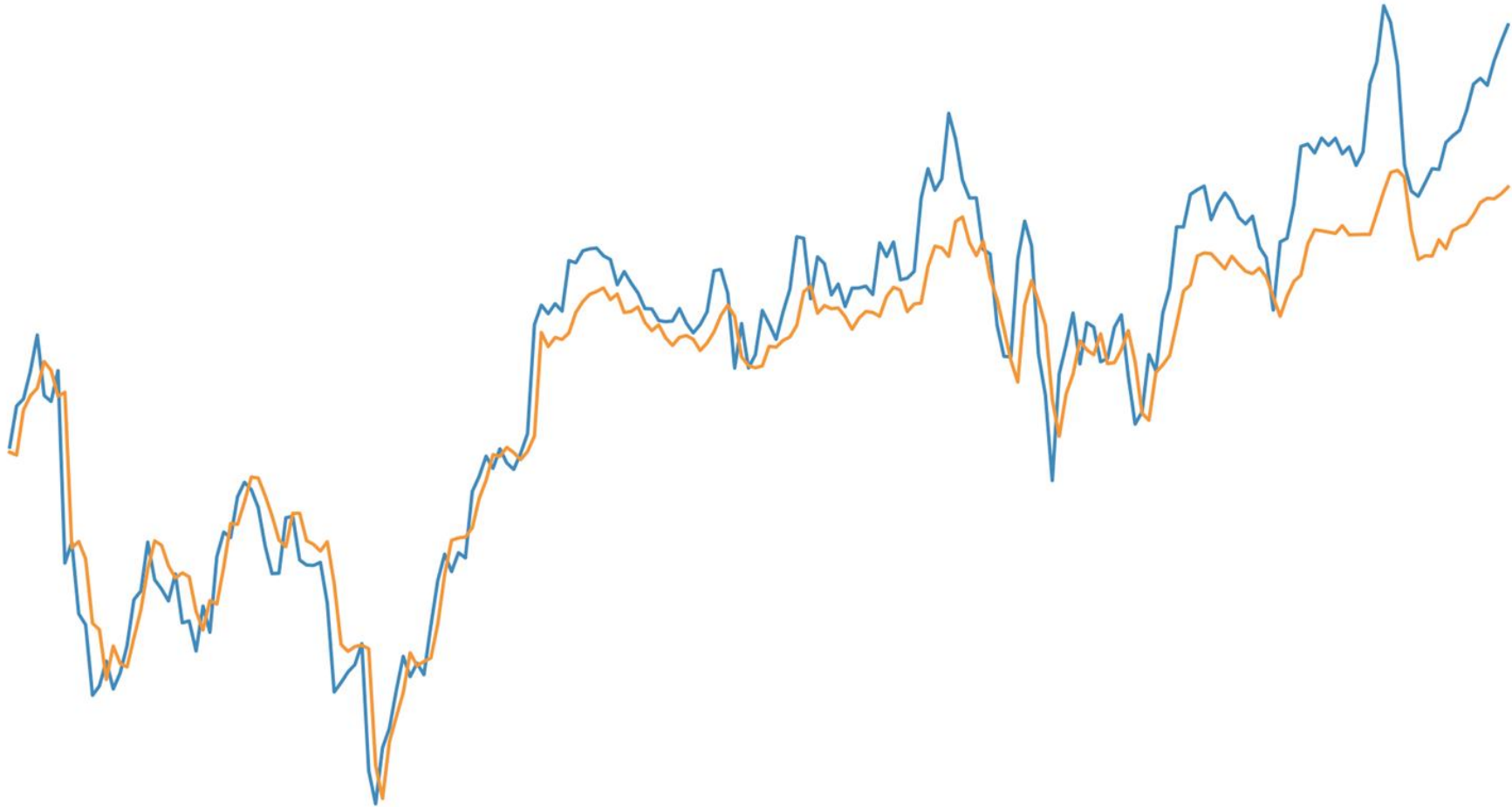


비디오프레임

LSTM, GRU

- 여러개의 데이터가 순서대로 입력되었을때 앞서 입력받은 데이터를 잠시 기억해 놓는 방법
- 기억된 데이터의 중요성을 판단하여 별도의 가중치를 줘서 다음 데이터로 넘긴다.
- 모든 입력 값에 이 작업이 순서대로 이뤄지며 같은 층을 맴도는 것으로 보여 순환 신경망이라 부른다

Time series data

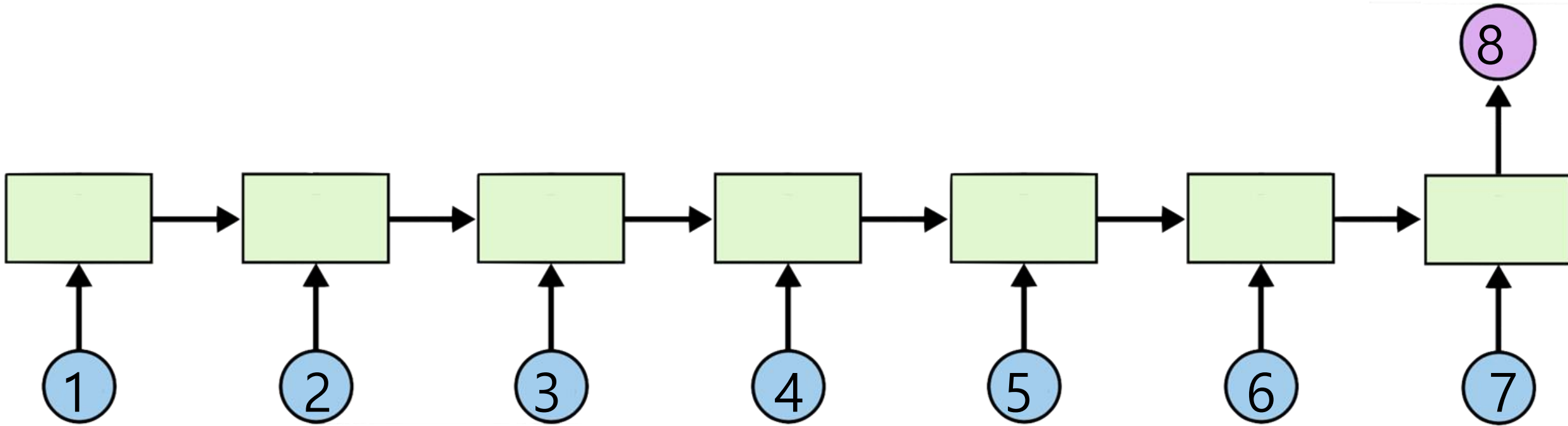


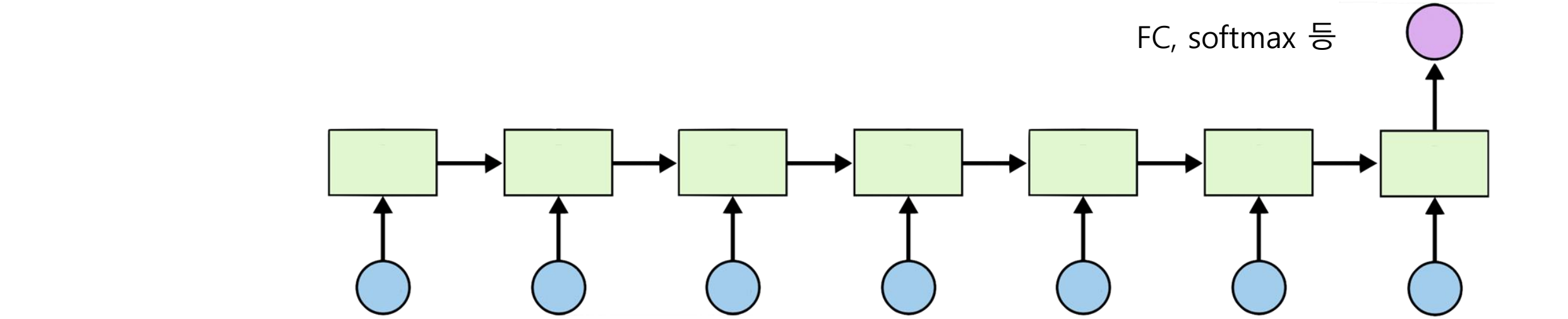
Time series data

| Open | High | Low | Volume | Close |
|------------|------------|------------|---------|------------|
| 828.659973 | 833.450012 | 828.349976 | 1247700 | 831.659973 |
| 823.02002 | 828.070007 | 821.655029 | 1597800 | 828.070007 |
| 819.929993 | 824.400024 | 818.97998 | 1281700 | 824.159973 |
| 819.359985 | 823 | 818.469971 | 1304000 | 818.97998 |
| 819 | 823 | 816 | 1053600 | 820.450012 |
| 816 | 820.958984 | 815.48999 | 1198100 | 819.23999 |
| 811.700012 | 815.25 | 809.780029 | 1129100 | 813.669983 |
| 809.51001 | 810.659973 | 804.539978 | 989700 | 809.559998 |
| 807 | 811.840027 | 803.190002 | 1155300 | 808.380005 |

'data-02-stock_daily.csv'

Many to one





| Open | High | Low | Volume | Close |
|------------|------------|------------|---------|------------|
| 828.659973 | 833.450012 | 828.349976 | 1247700 | 831.659973 |
| 823.02002 | 828.070007 | 821.655029 | 1597800 | 828.070007 |
| 819.929993 | 824.400024 | 818.97998 | 1281700 | 824.159973 |
| 819.359985 | 823 | 818.469971 | 1304000 | 818.97998 |
| 819 | 823 | 816 | 1053600 | 820.450012 |
| 816 | 820.958984 | 815.48999 | 1198100 | 819.23999 |
| 811.700012 | 815.25 | 809.780029 | 1129100 | 813.669983 |
| 809.51001 | 810.659973 | 804.539978 | 989700 | ? |
| 807 | 811.840027 | 803.190002 | 1155300 | ? |