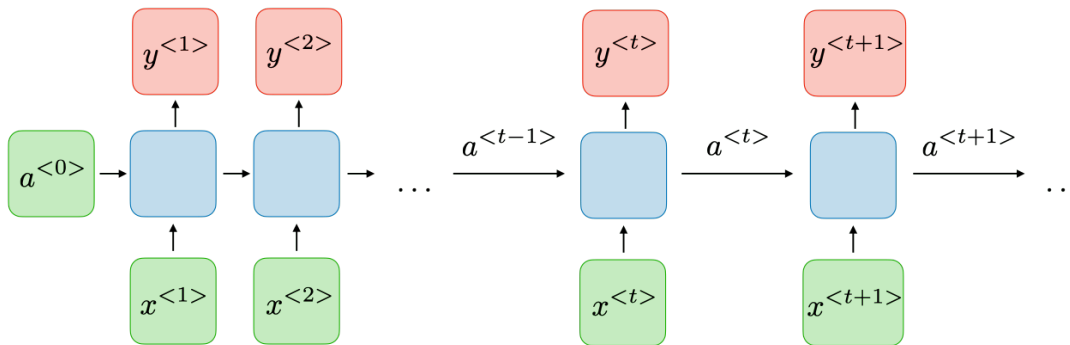


# Recurrent Neural Network(RNN)

A Recurrent Neural Network is a Neural network that is specialised for processing sequence of data.

$x(t) = x(1), x(2), x(2), \dots, x(n)$  where,  $t \in [1, n]$



$$a^{<t>} = g1(W_{aa}a^{<t-1>} + W_{ax}x^{<t>} + b_a)$$

$$y^{<t>} = g2(W_{ya}a^{<t>} + b_y)$$

Where:

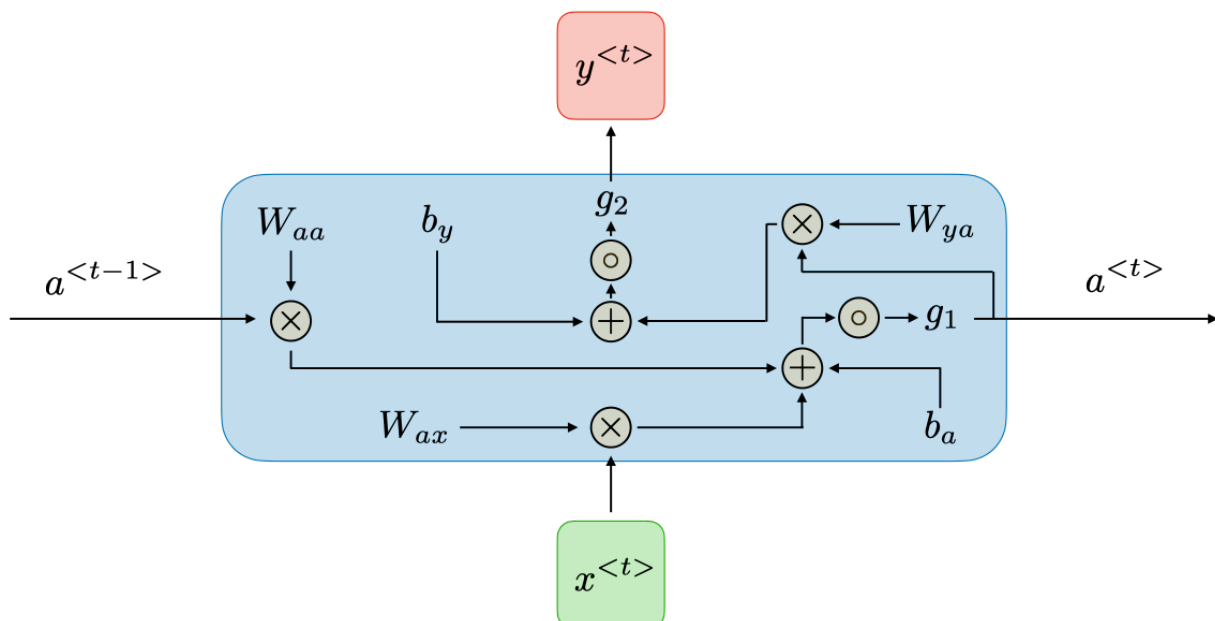
$t$  : timestamp

$a^{<t>}$  : activation

$y^{<t>}$  : output

$g1, g2$  : activation function

$W_{aa}, W_{ax}, W_{ya}, b_a, b_y$  : are the coefficients



Advantages	Drawbacks
Possibility of processing input of any length	Computation being slow
Model size not increase with size of input	Difficult of accessing information from a long time ago
Computation takes into account historical information	Cannot consider any future input for the current state
Weights are shared across time	

## Applications of RNNs

Type of RNN	Illustration	Example
<b>One-to-One</b> $T_x(input) = T_y(output) = 1$		Traditional Neural network
<b>One-to-Many</b> $T_x = 1, T_y > 1$		Music generation
<b>Many-to-One</b> $T_x > 1, T_y = 1$		Sentiment Classification
<b>Many-to-Many</b> $T_x = T_y$		Name entity recognition
<b>Many-to-Many</b> $T_x \neq T_y$		Machine Translation

## Loss Function

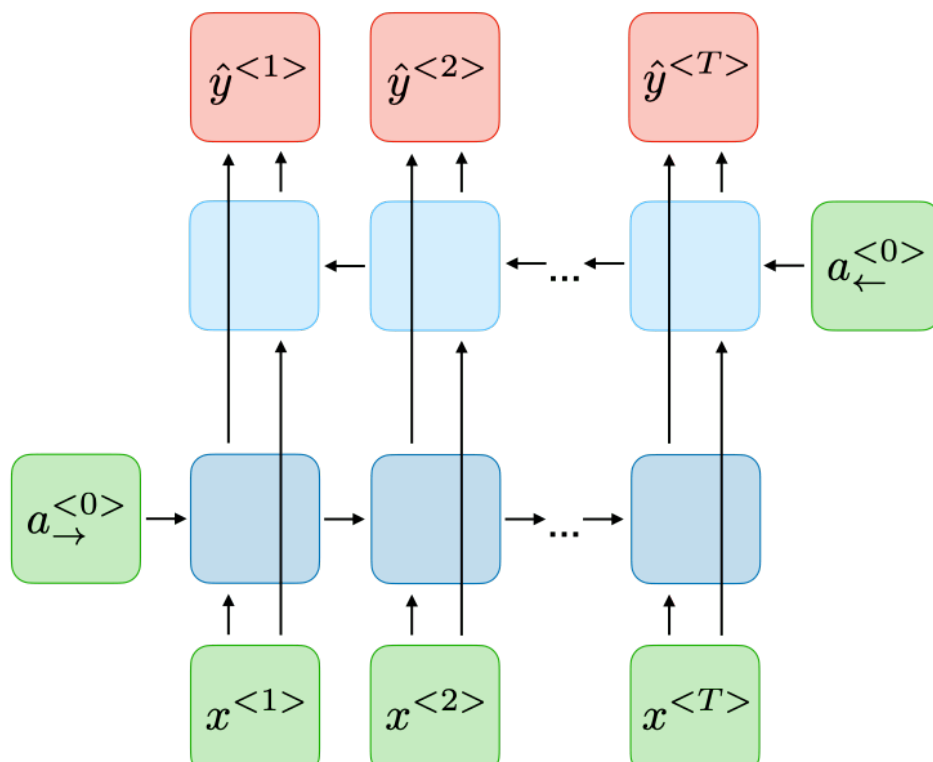
$$E^{(t)} = \Sigma(actual^{<t>} - predicted^{<t>})^2$$

## Back propagation

$$E'^{(t)} = \sum \frac{dE^{(t)}}{dw} \big|_t$$

## Variants of RNN

### 1. Bidirectional



## 2. Deep RNN

