



Recurrent Neural Network

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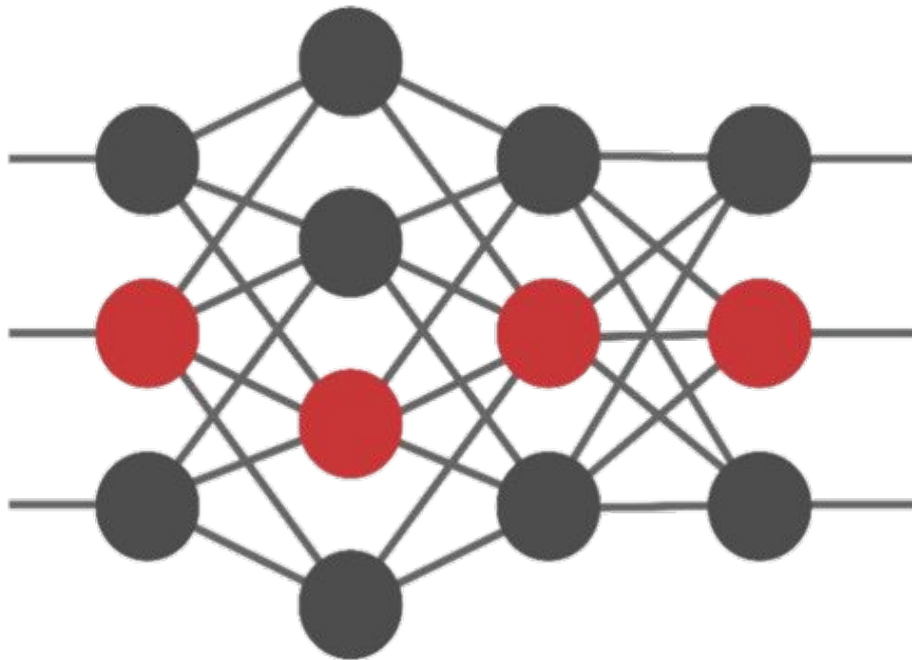
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Outline

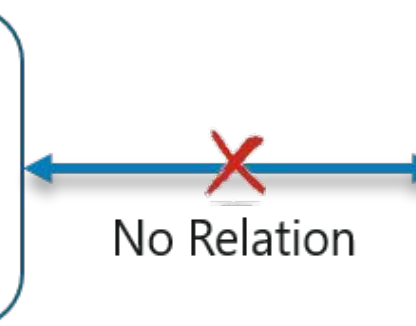
- Why Not Feed-forward Networks?
- What Are Recurrent Neural Networks?
- How To Train Recurrent Neural Networks?
- Vanishing And Exploding Gradients
- Long Short Term Memory (LSTM) Networks
- LSTM Use-Case

Why Not Feed-forward Networks?

- Consider an **image classification**, where NN trained to **classify images** of various **animals** such as a **cat** or a **dog**, the NN provides an **output** with a **corresponding label** to the image of a cat or a dog **respectively**.



Output at 't'



No Relation



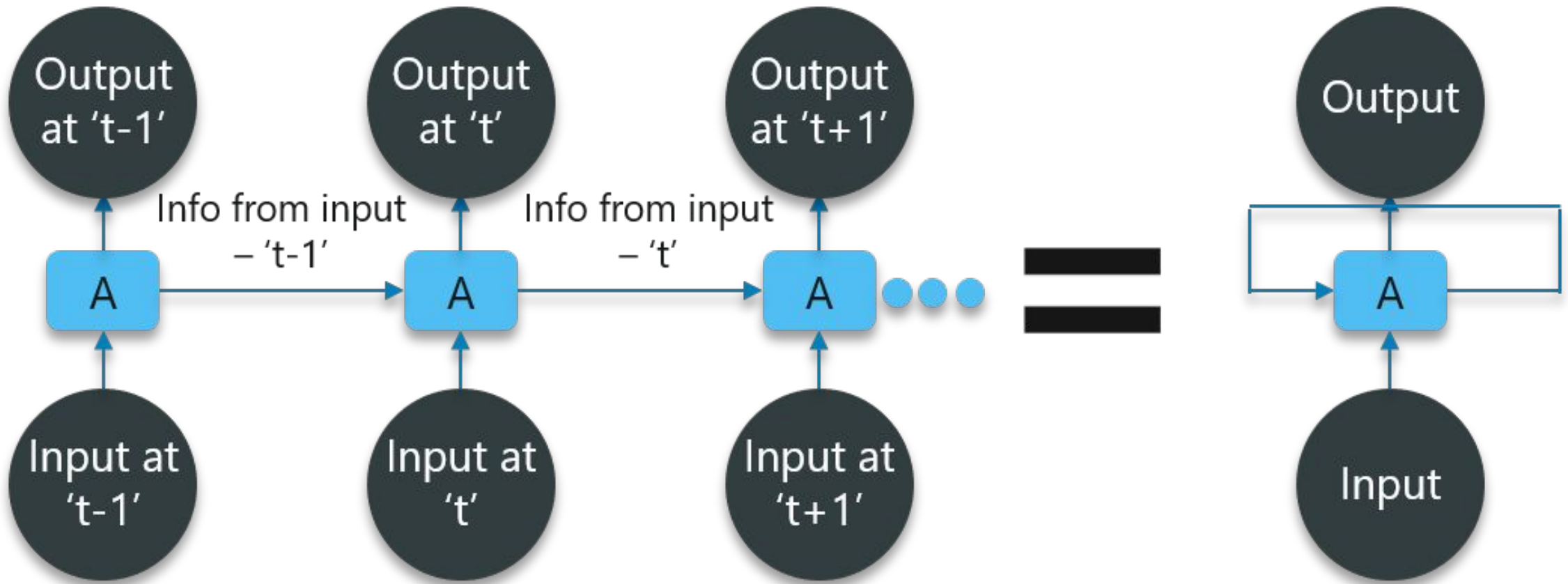
Output at 't-1'

Why Not Feed-forward Networks?...

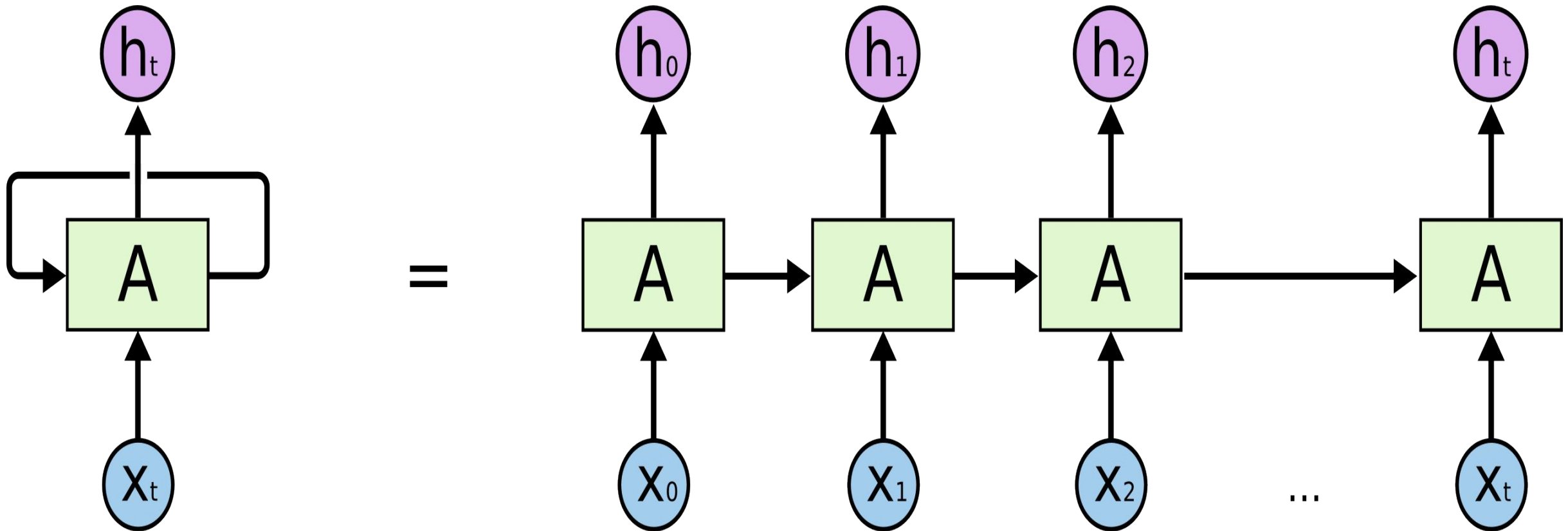
- Consider a real case of **reading a book**, where with **every page** you move forward into, you need the **understanding** of the **previous pages** to make **complete** sense of the **information**.
- With a **feed-forward network** the **new** output at time ' $t + 1$ ' has **no relation** with outputs at either time t , $t - 1$, or $t - 2$.
- So, feed-forward networks **cannot be used** when **predicting a word** in a sentence as it will have no **absolute relation** with the **previous** set of **words**. But, with **Recurrent Neural Networks**, this challenge can be **overcome**.
- A **recurrent neural network (RNN)** is a class of [artificial neural networks](#) where connections between nodes form a [directed graph](#) along a **temporal sequence**

Why Not Feed-forward Networks?...

- Consider a diagram: RNN

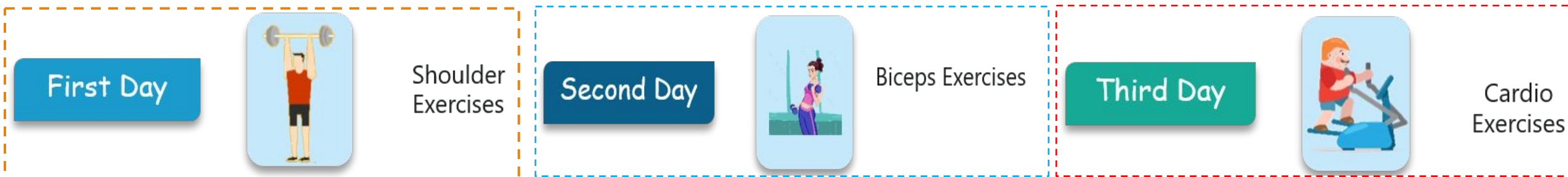


Generalized RNN



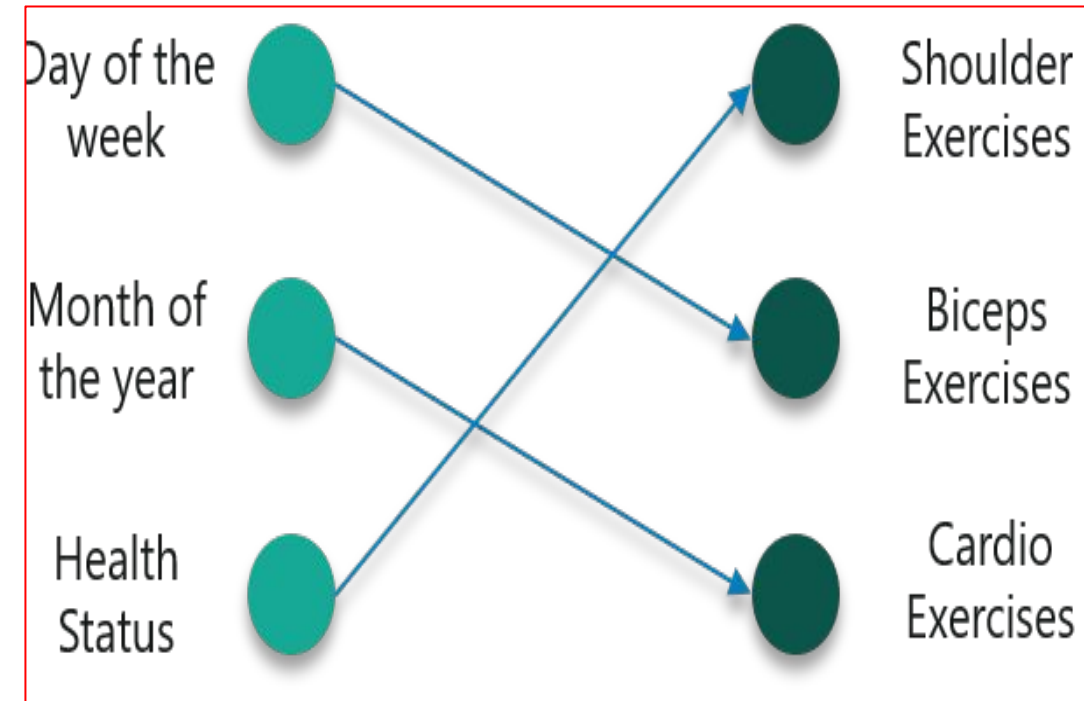
Recurrent Neural Network?

- Recurrent Networks are a **type of artificial neural network** designed to **recognize patterns** in sequences of data, such as **text**, genomes, handwriting, the spoken word, numerical **times series** data emanating from sensors, **stock markets** and **government agencies**.
- Consider an Example of **gym**, where **trainer** has given a **schedule** for workout:



Recurrent Neural Network?...

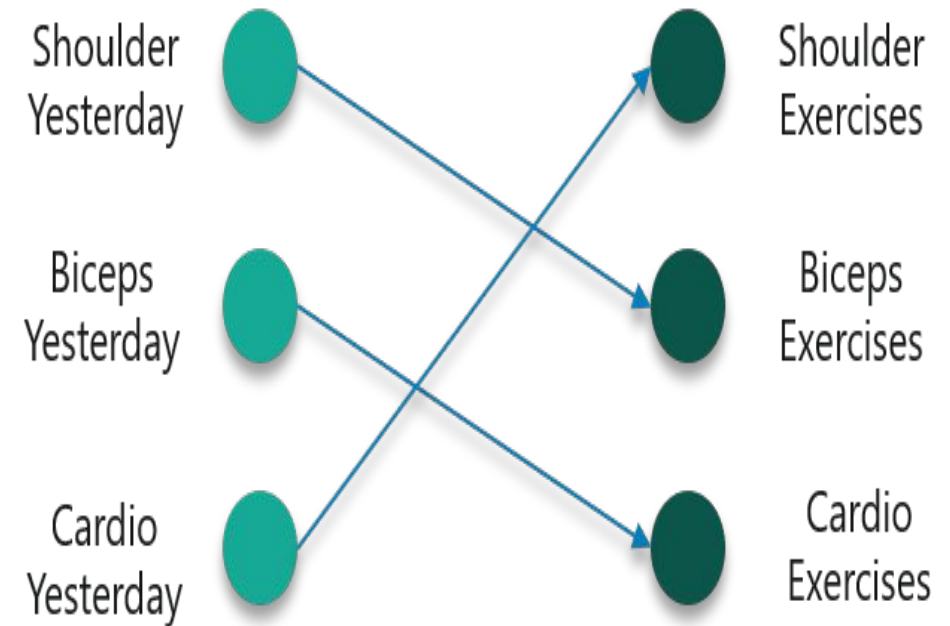
- These exercise are repeated every week in proper order.
- Let we try to predict the exercise of day using Feedforward NN.
- The inputs are **day**, **month** and the **health status**. A neural network has to be **trained** using these inputs to provide us the with the **prediction** of the **exercises**.



This is not appropriate to do

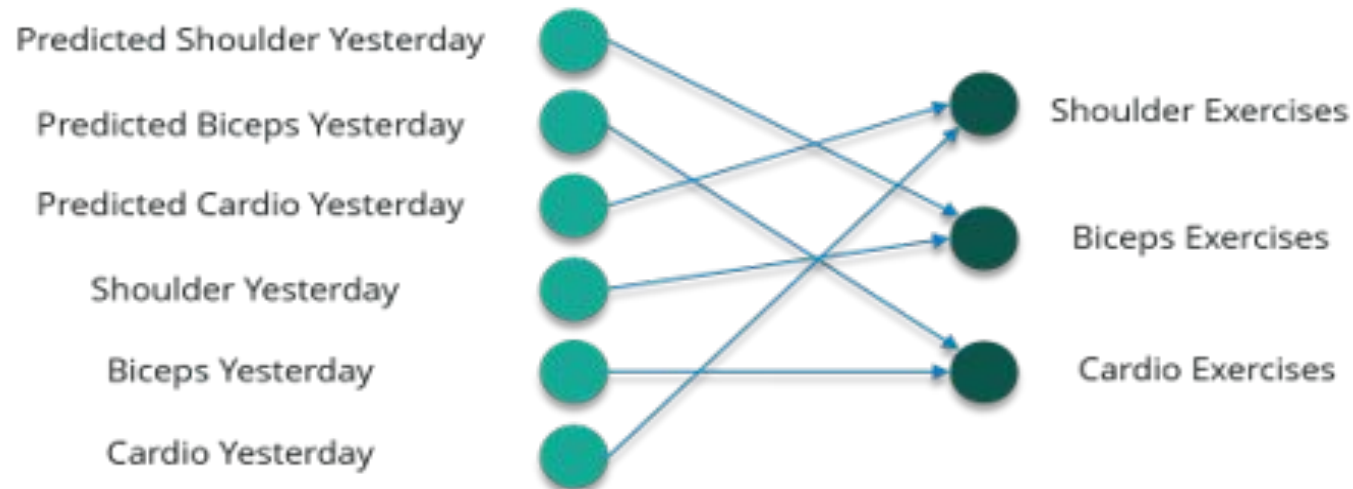
Recurrent Neural Network?...

- To **fix this**, we can make use of the concept of **Recurrent Neural Networks** can be buildup.
- Consider the **inputs** to be the **workout** done on the **previous day**.
- So if a person did a **shoulder workout** yesterday, he can do a **bicep exercise** today and this goes on for the **rest** of the **week** as well.



Recurrent Neural Network?...

- However, if you happen to **miss a day** at the gym, the data from the **previously attended timestamp** can be **considered** as shown below.
- If a **model** is trained based on the data it can **obtain** from the **previous exercises**, the output from the model will be **extremely accurate**.



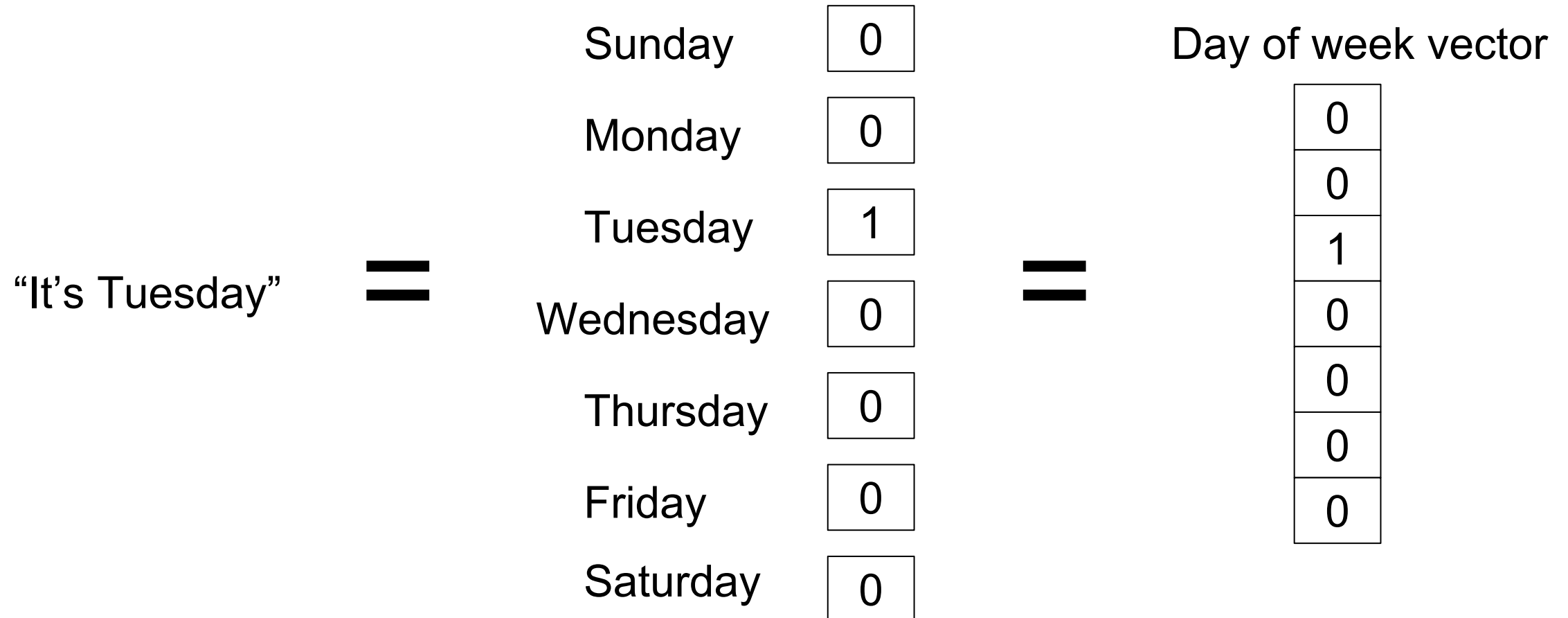
A Data as a Vector

- **Vectors are numbers which are input to the model to denote if you have done the exercise or not.**
- A vector is a list of values: *Example 1*

“High is 67 F. Low is 43 F. Wind is 13 mph. .25 inches of rain. Relative humidity is 83%.”	=	High temperature	67	=	Weather vector	
		Low temperature	43			67
		Wind speed	13			43
		Precipitation	.25			13
		Humidity	.83			.25
						.83

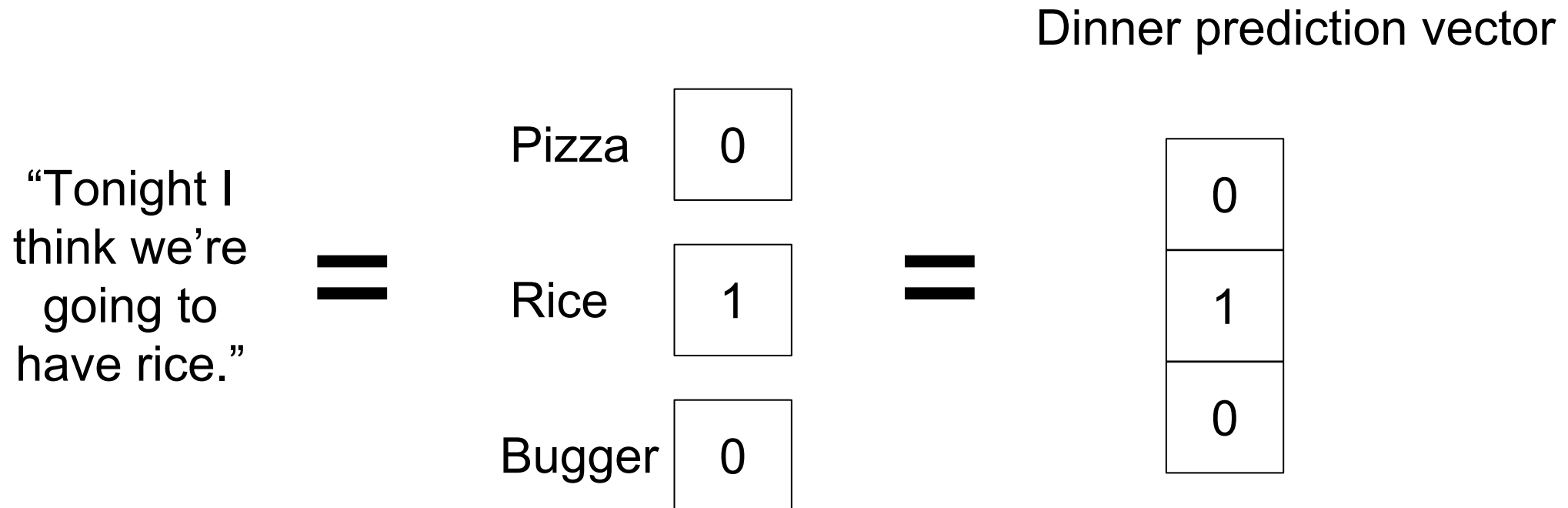
A Data as a Vector...

- A vector is a list of values: *Example 2*



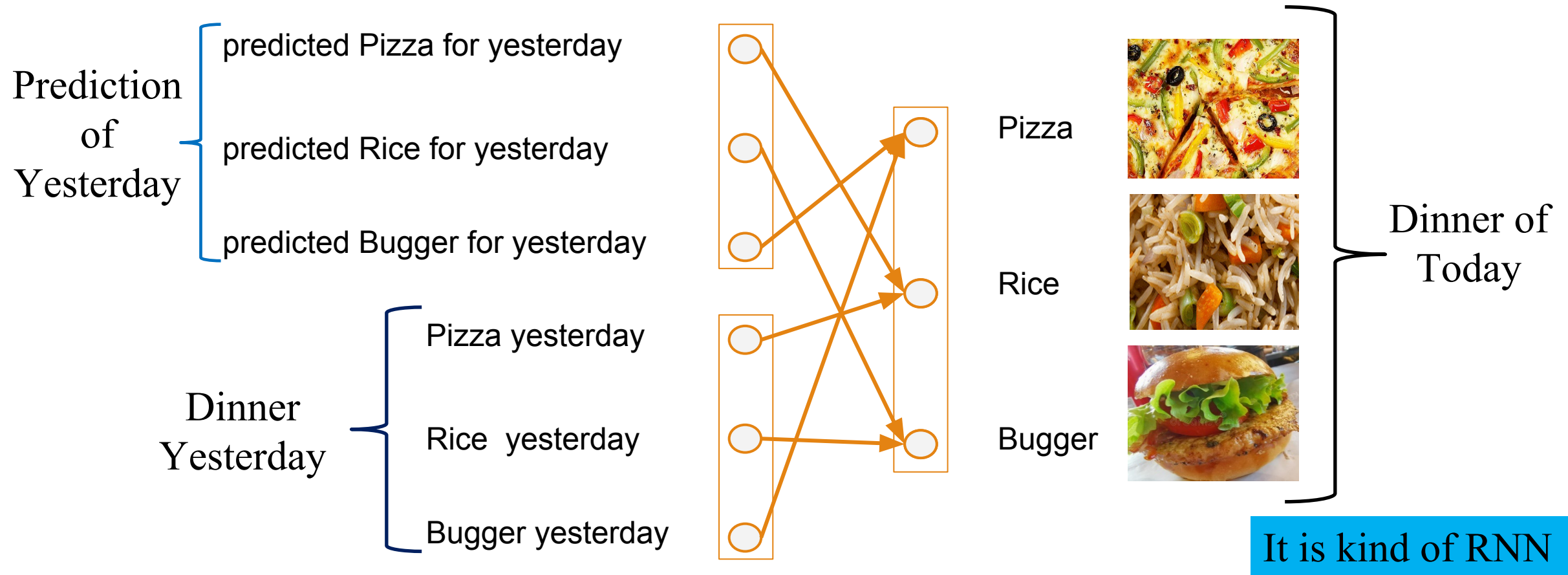
A Data as a Vector...

- A vector is a list of values: *Example 3*



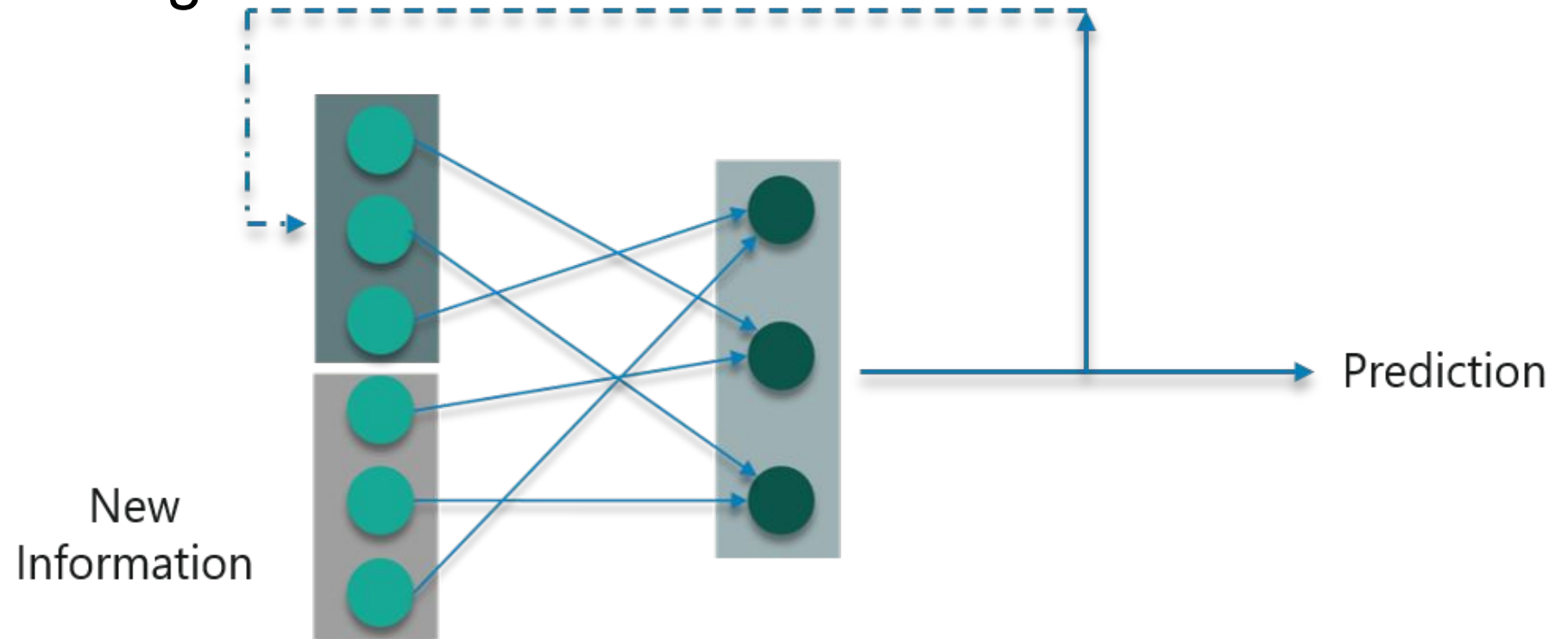
Example

- Model is more appropriate, which gives prediction dinner based on previous day dinner



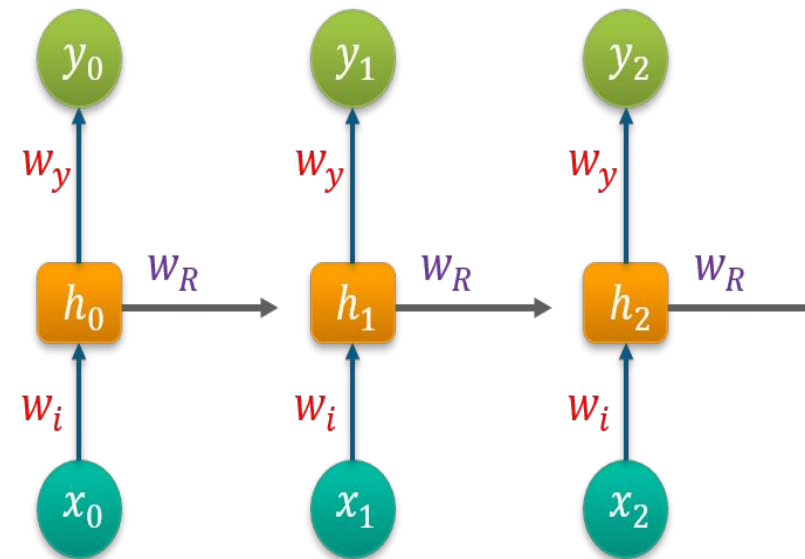
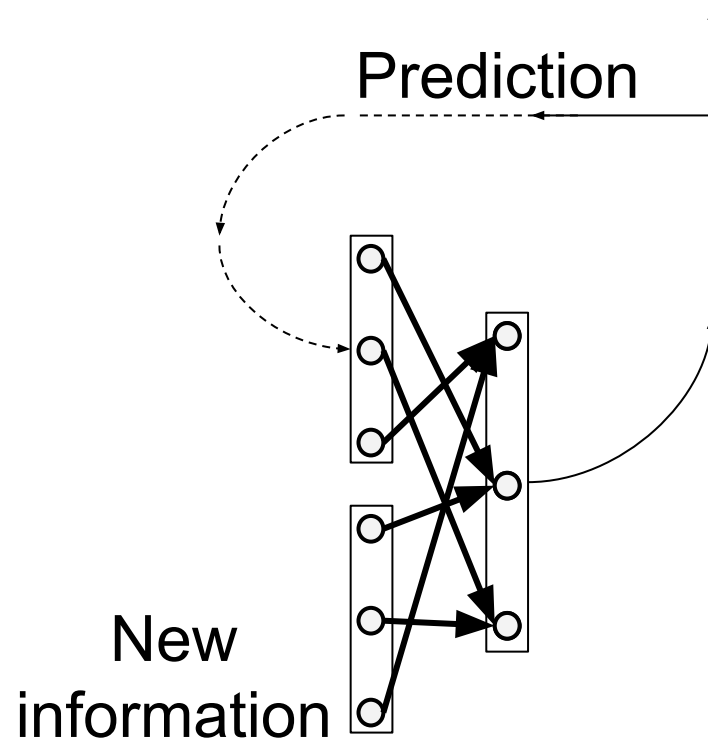
Model of these Examples

- These are all examples, a correct prediction can be made based upon previous output or state. Hence, a model can be buildup as shown in below Fig.



Model: Recurrent Neural Network

- A mathematical model can be buildup, where '**w**' to be the **weight matrix** and '**b**' being the **bias**.
- At time **t=0**, input is '**x0**' and the task is to figure out what is '**h0**'. Putting **t=0** in the **equation** and obtaining the function **h(t)** value. Next, the value of '**y0**' is found out using the **previously calculated values** when applied to the **new formula**



$$h^{(t)} = g_h (w_i x^{(t)} + w_R h^{(t-1)} + b_h)$$

$$y^{(t)} = g_y (w_y h^{(t)} + b_y)$$

Example of NN

- Consider the preparation of food items based on weather conditions.
- Consider a NN model to do this.
- Input weather is **'sunny'** then output is **'ice-cream'**
- Input weather is **'Rainy'** then output is **'French Fries'**



Example of NN...

- Consider these are represented as vector



$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

Ice Cream



$$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

French Fries



$$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Pizza



Sunny



$$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Rainy

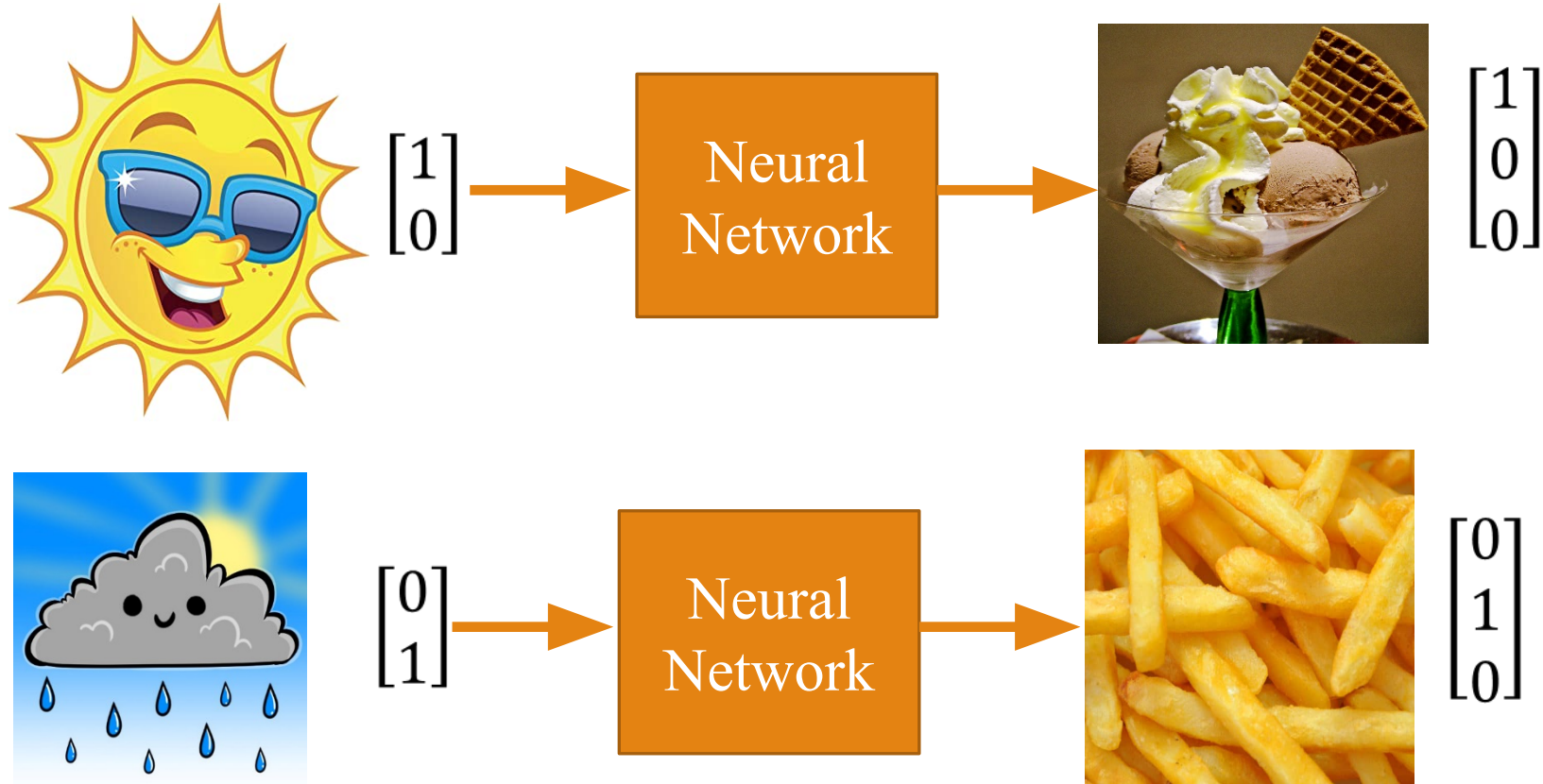


$$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

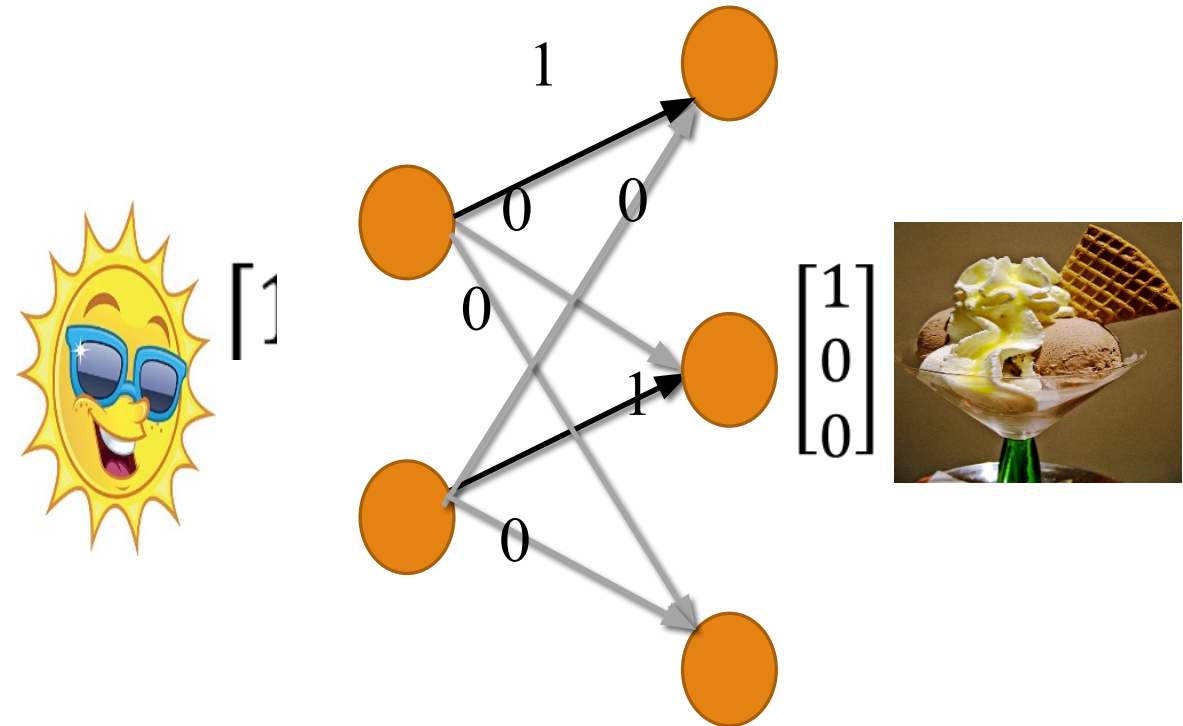
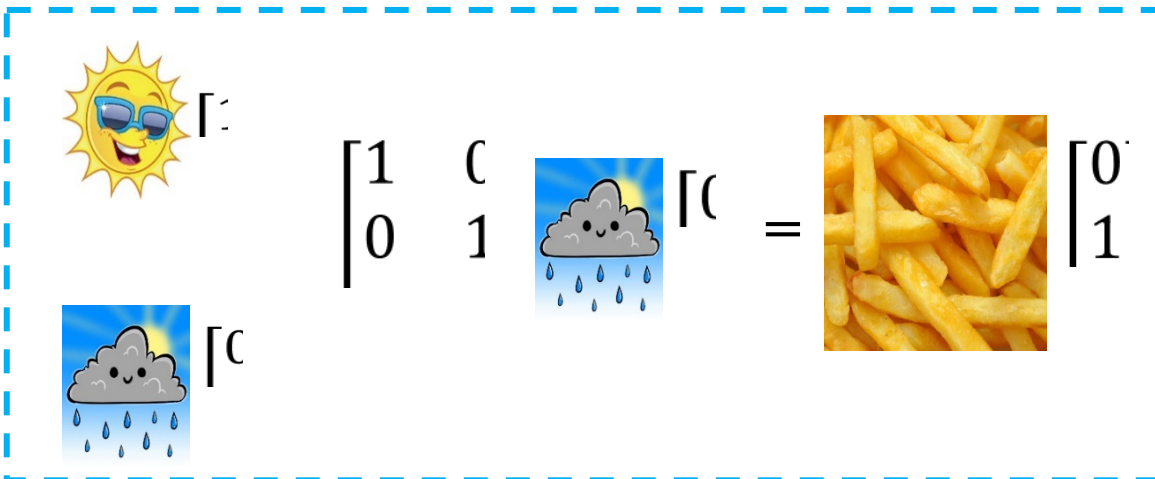
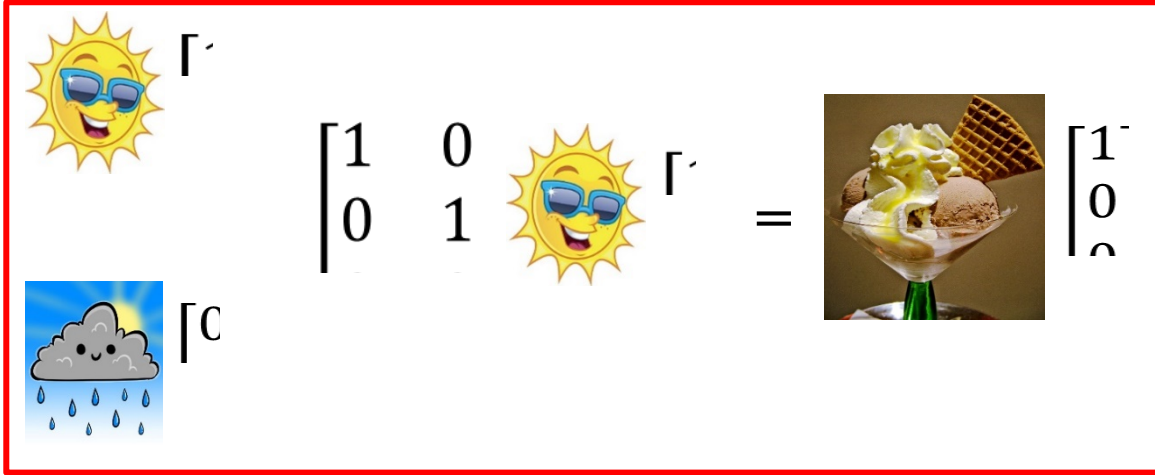


Example of NN...

- Consider these are represented as vector

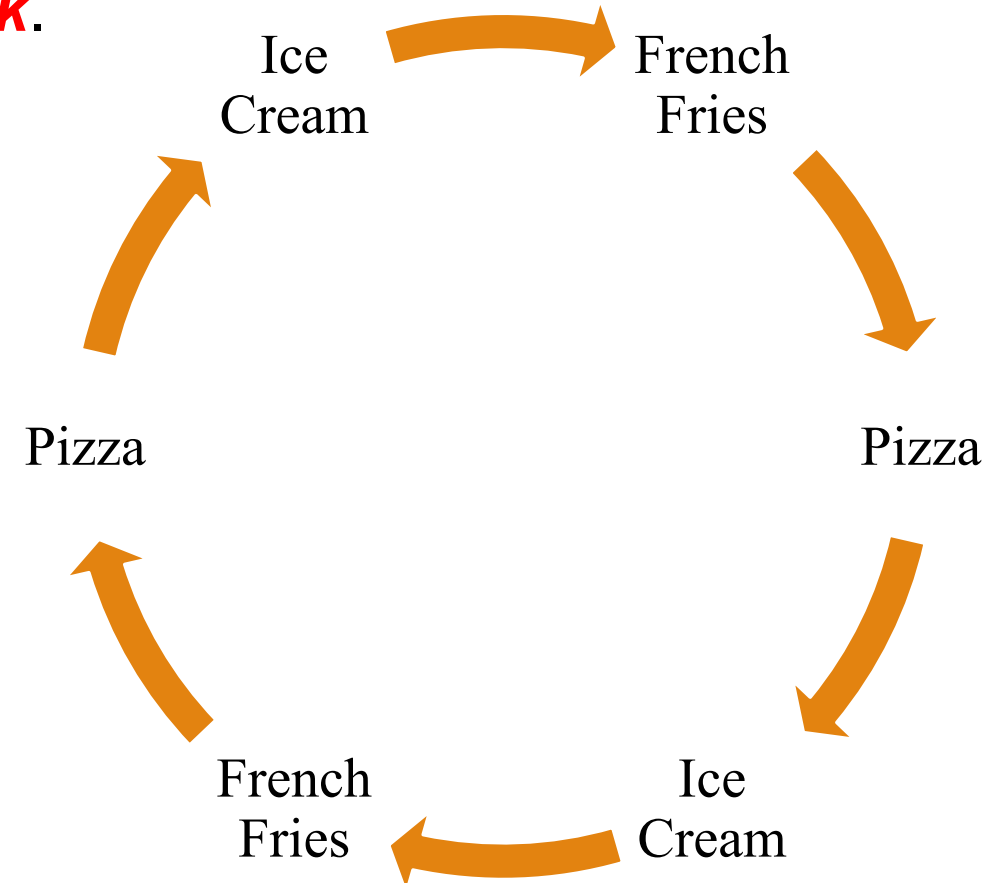


Example of NN...Representation as Vector

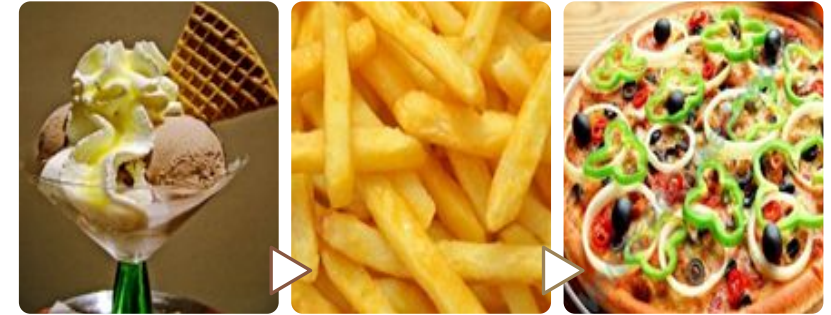
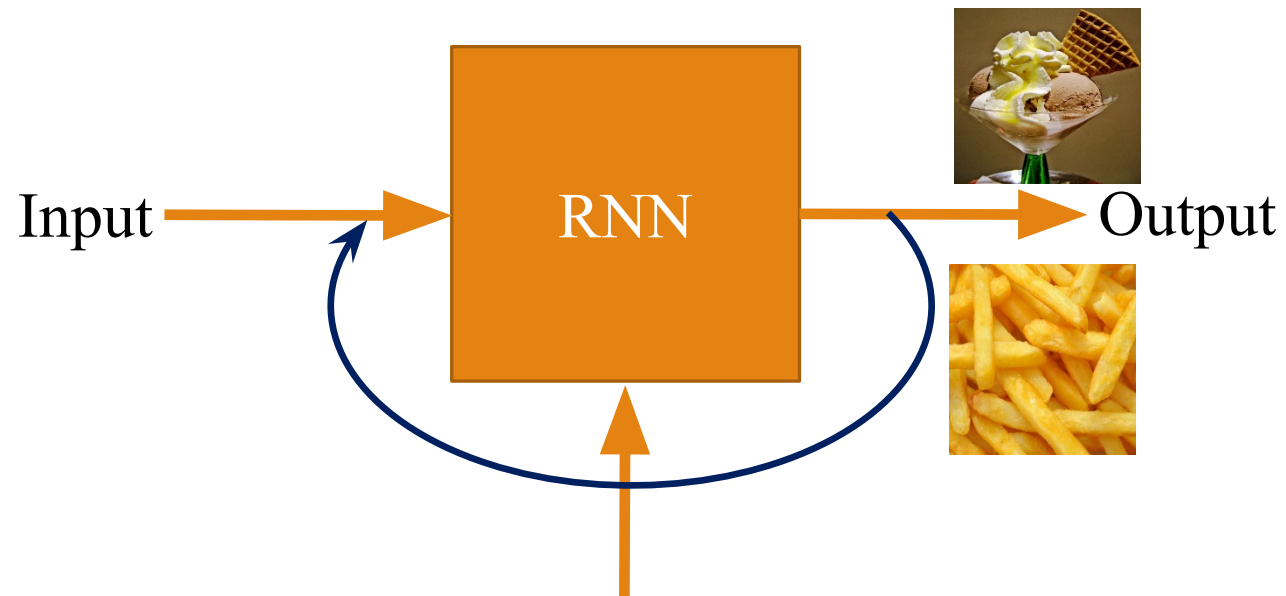


Good Example of Cooking Food

- A good cook, always have understanding what he/she has served on **previous occasions/days**. *Day wise in a week.*



Model of Cooking Food



Simple Neural Network



$$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



$$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



$$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$$



$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$



$$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

=

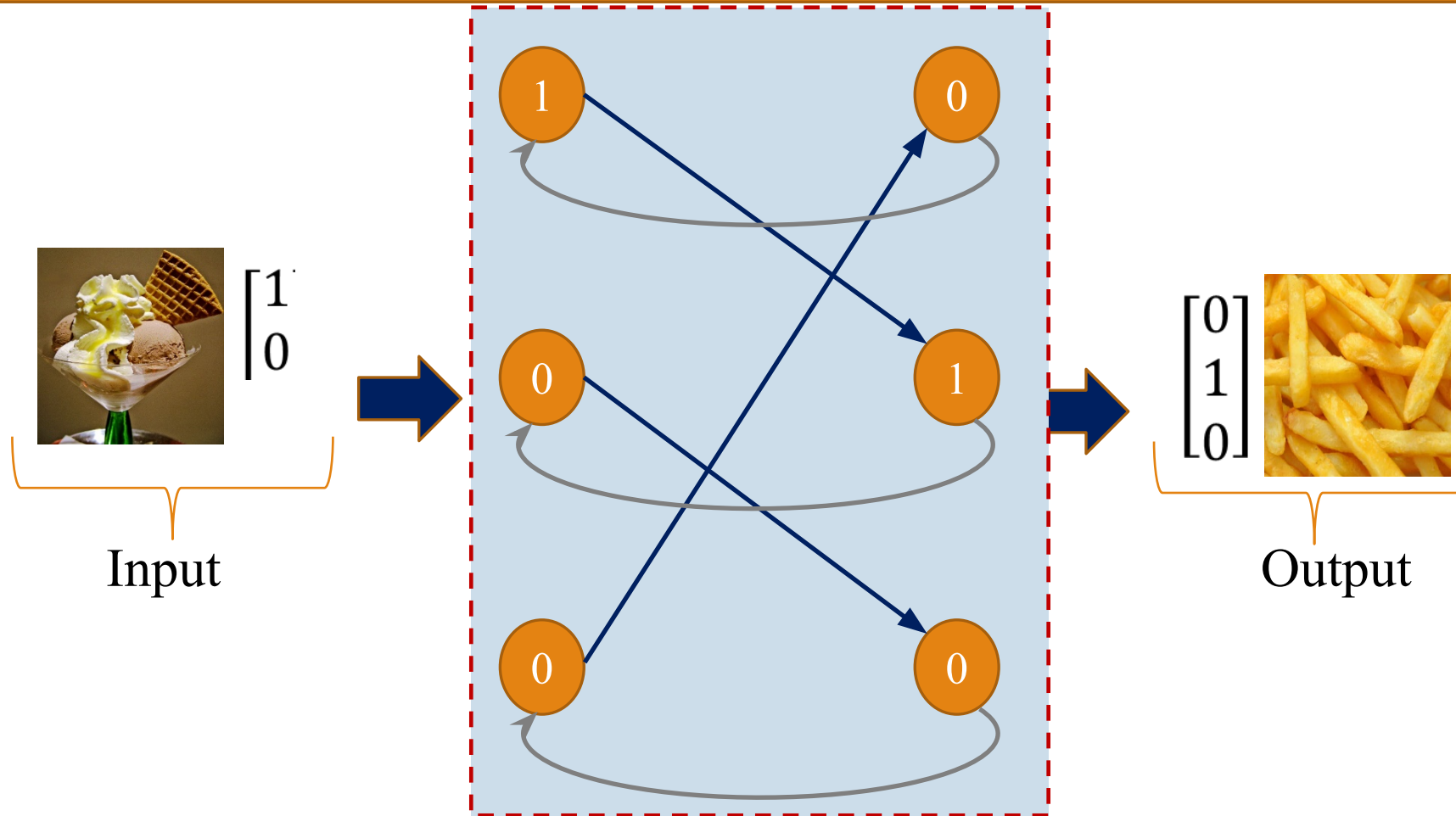


$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

=

Recurrent Neural Network



Schedule of Food with Weather

- Consider if weather is sunny then person will out and enjoy the weather and if its rainy then person will stay at home and cook the food as per schedule.

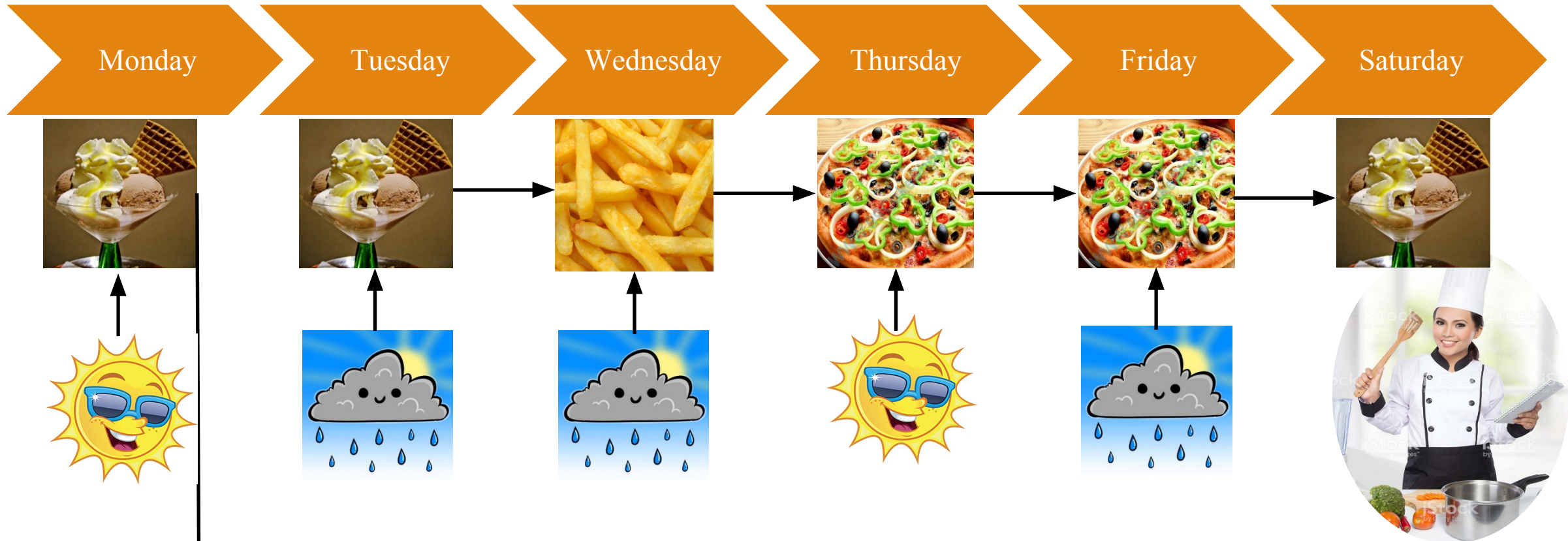


Same as yesterday

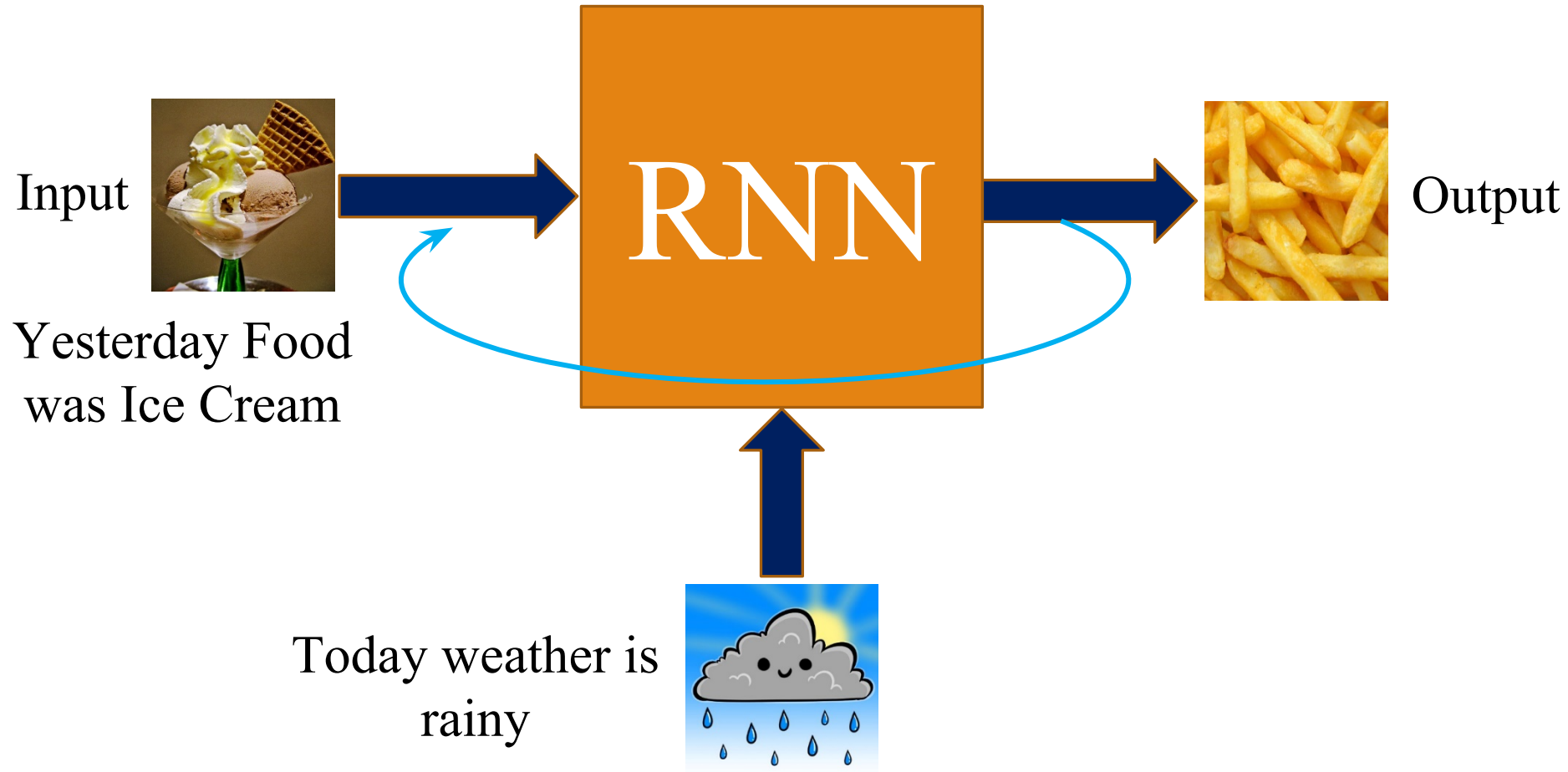


Next Dish

Cooking Schedule of Week



Model of Cooking Schedule : RNN



Vector Representation of Food Items

- Consider these are represented as vector



$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

Ice Cream



$$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

French Fries



$$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Pizza



Sunny



$$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Rainy



$$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

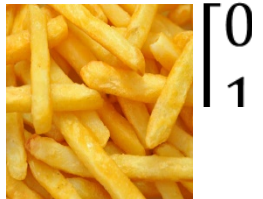


Food Matrix

- Consider these are represented as vector


 $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$

Ice Cream


 $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$

French Fries


 $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Pi

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ \hline 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

=

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$



Same

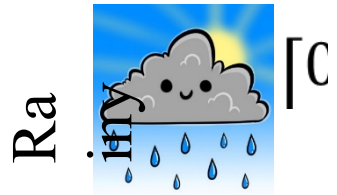
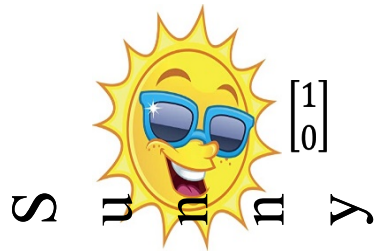
Next Day

Similarly
for other
items too

Concenation of todays and tomorrow food

Weather Matrix

- Consider these are represented as vector



$$\begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$



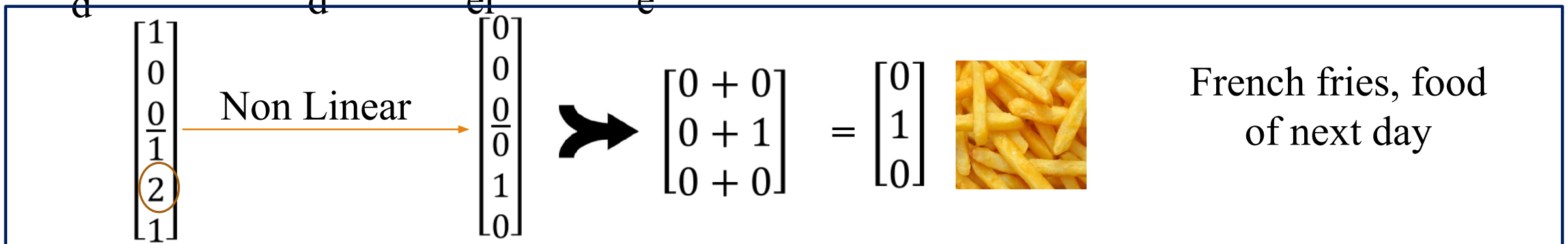
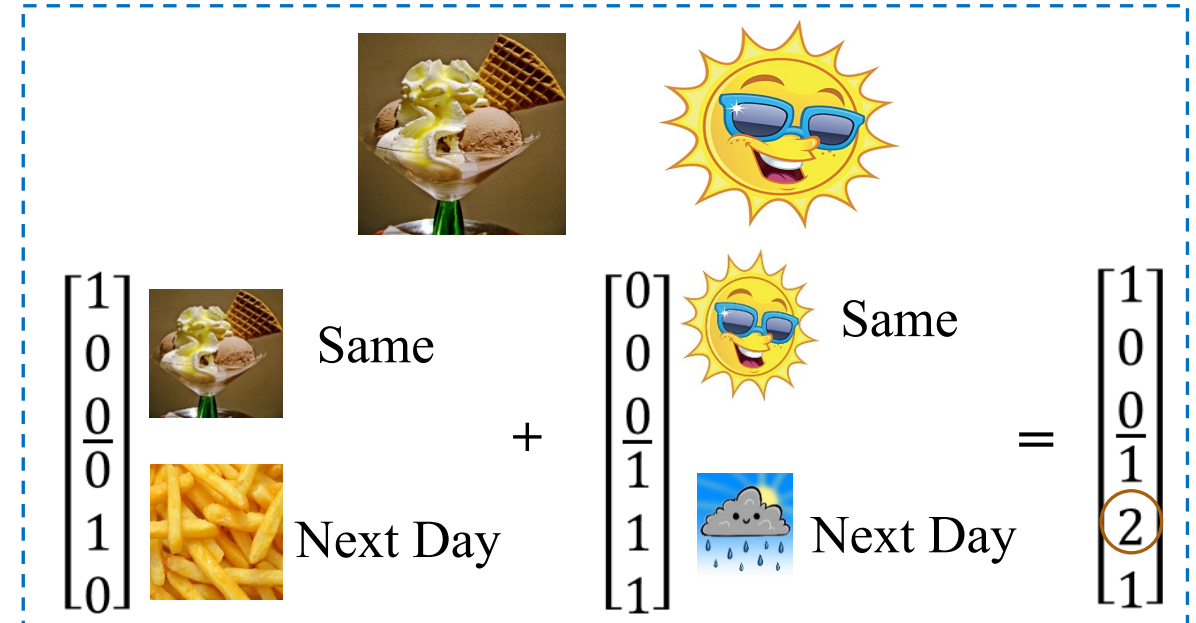
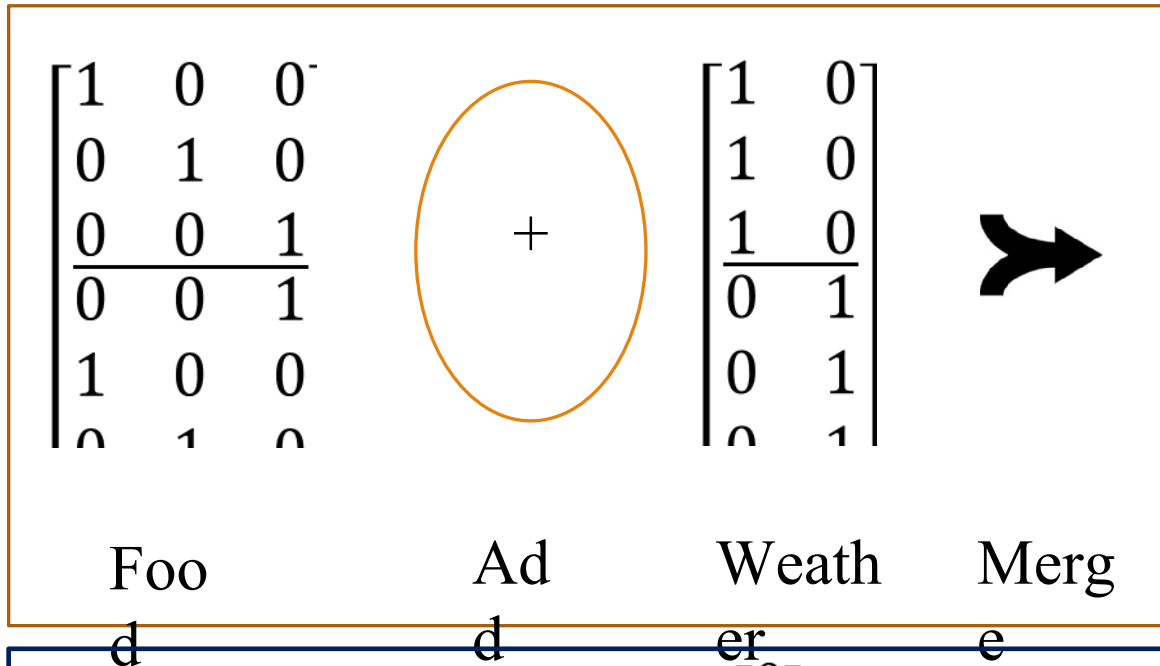
Same day



Next day

Weather Matrix

A More Complicated RNN



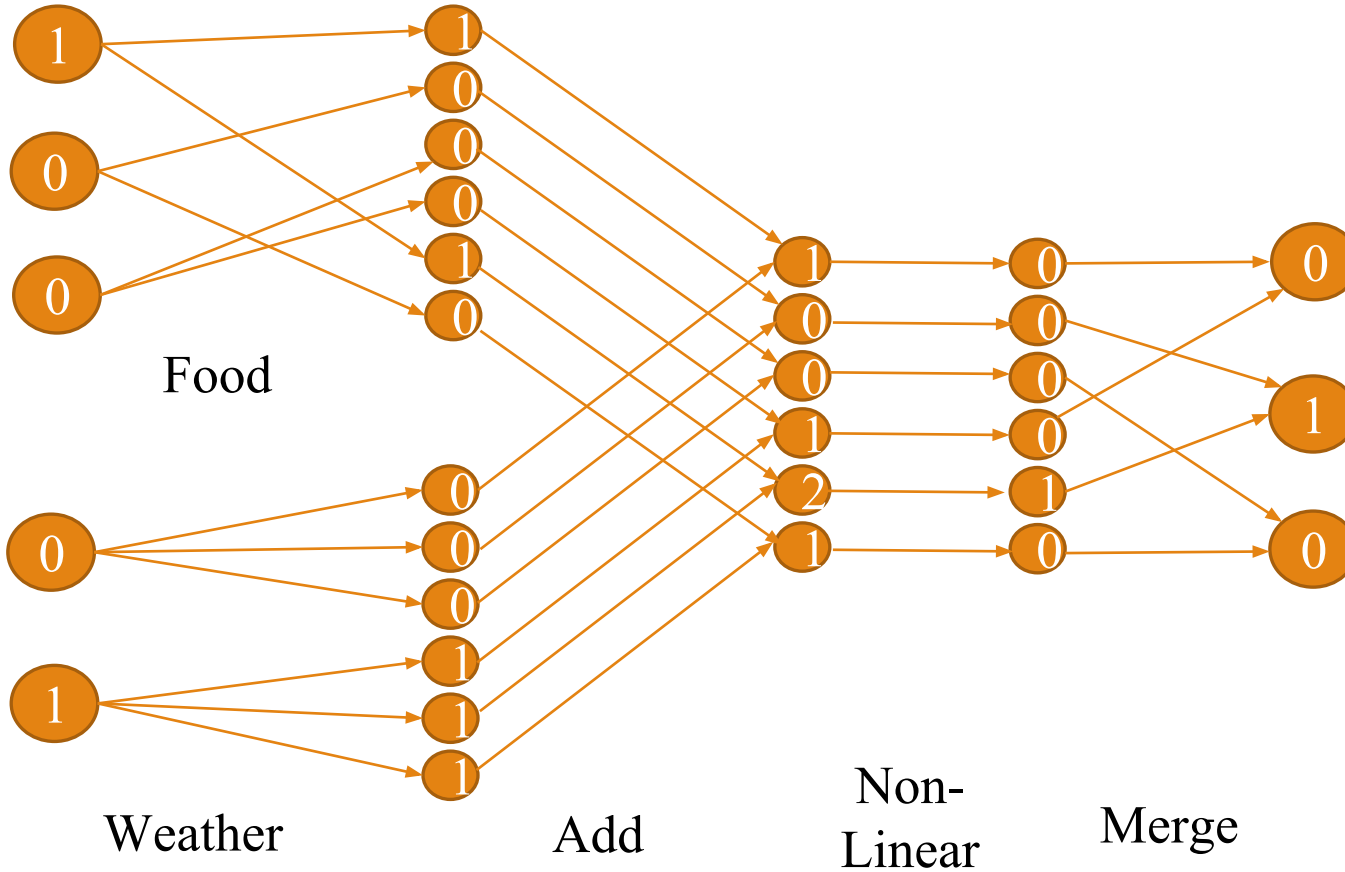
RNN Model


 $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$

Food

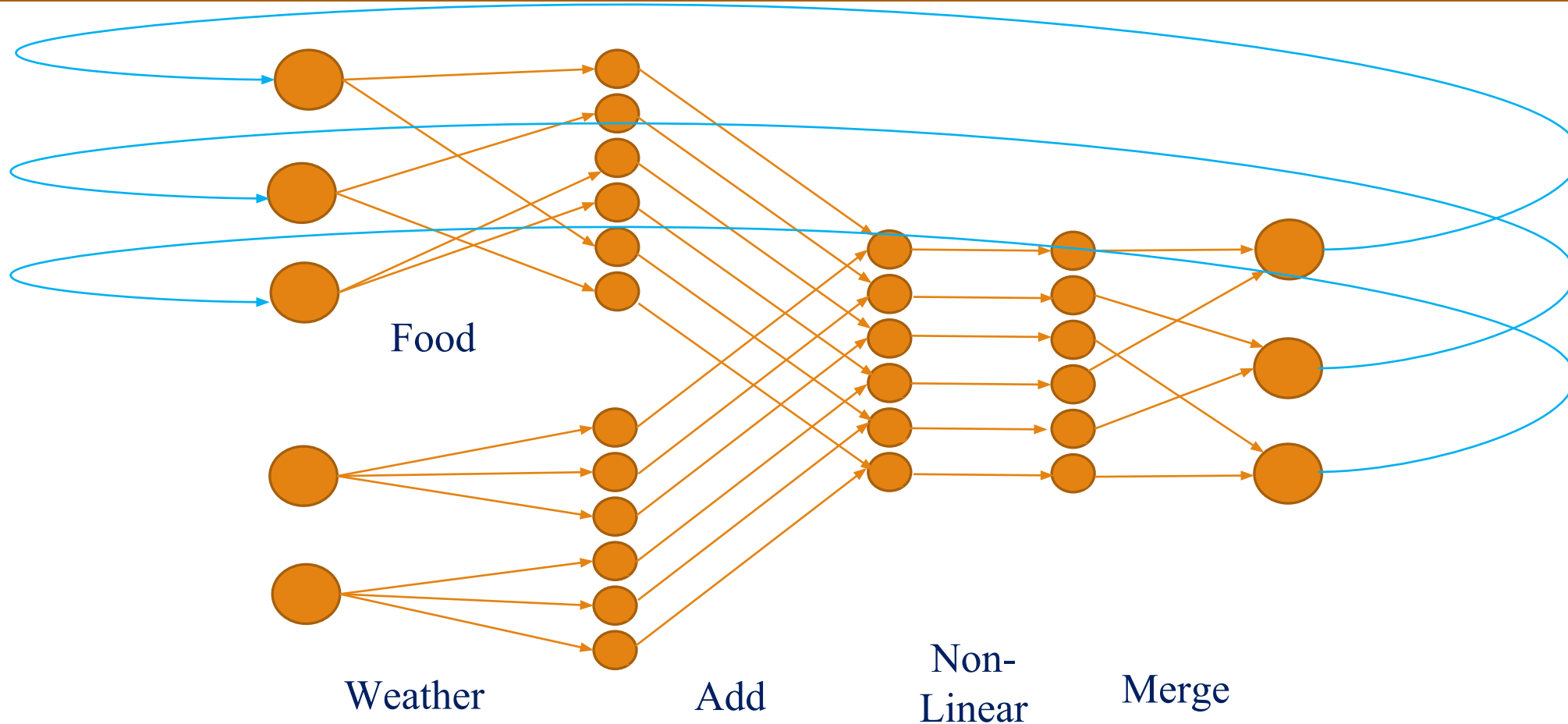

 $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$

Weather



Food
of
Next
Day

RNN Model...



Output
Vector is
fed to input
:
Recurrent

Training RNN

- To train the RNN, these weights need to be updated.

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ \hline 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ \hline 0 & 1 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 \end{bmatrix}$$

Fo

A

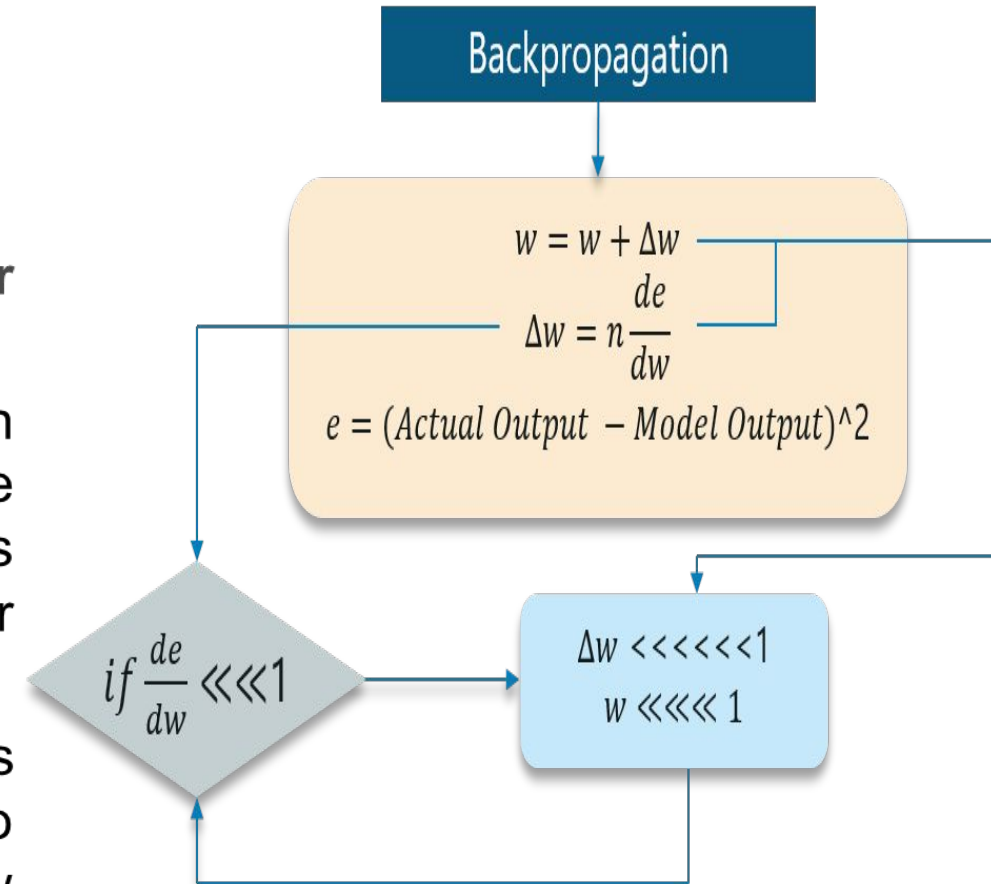
Weat

Mer

- Recurrent Neural Networks use **backpropagation algorithm** for training, but it is **applied** for every **timestamp**. It is commonly known as **Back-propagation Through Time (BTT)**.

Training RNN...

- There are **some issues** with Back-propagation such as:
 - ✓ **Vanishing Gradient**
 - ✓ **Exploding Gradient**
- **Vanishing Gradient**
 - ✓ In **back-propagation**, the **goal** is to **calculate** the **error** (e) = (Actual Output – Model Output)².
 - ✓ The **changes** in the e w.r.t the **change** in the **weight** is **calculated** and multiplied with learning rate (n), which is the **actual change** in the weight. This change in weight Δw is added to the **old set** of weight for every iterations.
 - ✓ The issue here is when change of e w.r.t. to w is multiplied with n , the value may be very less. It is also called as **vanishing gradient**. At this condition now further weights are updated.



Training RNN...

- **Exploding Gradient**

- ✓ The working of the exploding gradient is **similar** but the **weights** **here** change **drastically** instead of **negligible change**.
- ✓ To overcome with these issues, following may be done.

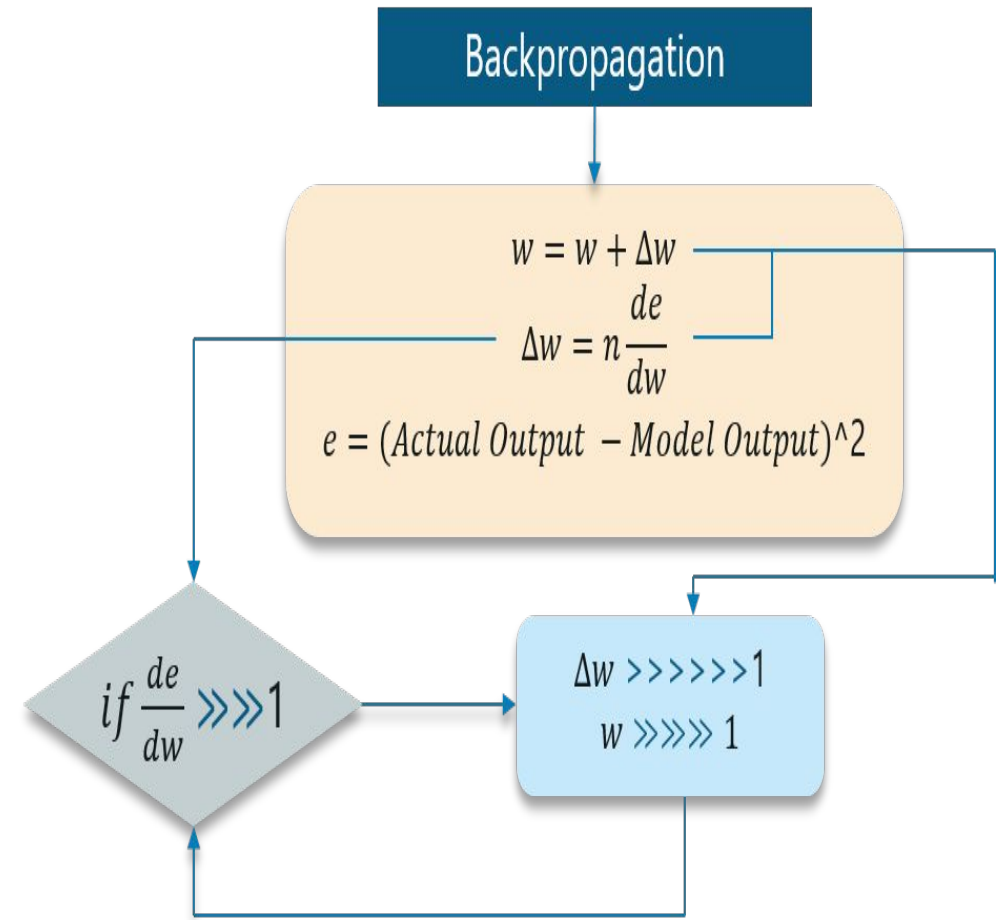
Issues and Solutions

- **Exploding Gradient**

- ✓ Clip the gradient using threshold.
- ✓ Using RMSprop optimizer, to adjust the learning rate.

- **Vanishing Gradient**

- ✓ Use ReLU activation function.
- ✓ Different network can be used such as LSTM and GRU.

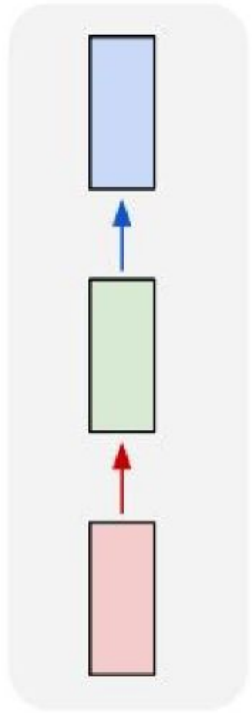


Applications

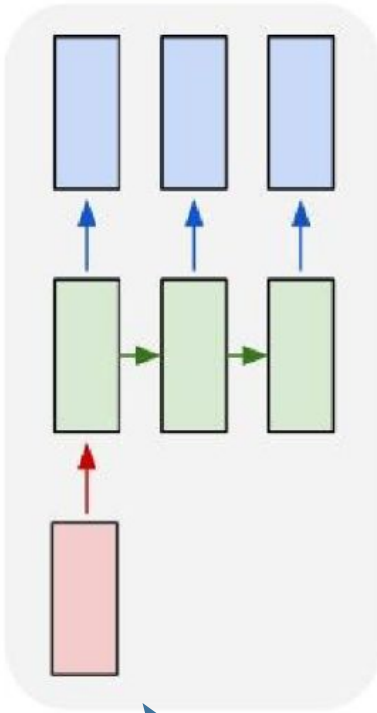
- **Recognize patterns** in sequences of data such as:
 - ✓ Stock Market prediction
 - ✓ Text Generation
 - ✓ Genomes
 - ✓ Handwriting,
 - ✓ Spoken word
 - ✓ Numerical times **series** data emanating from sensors

Applications

one to one

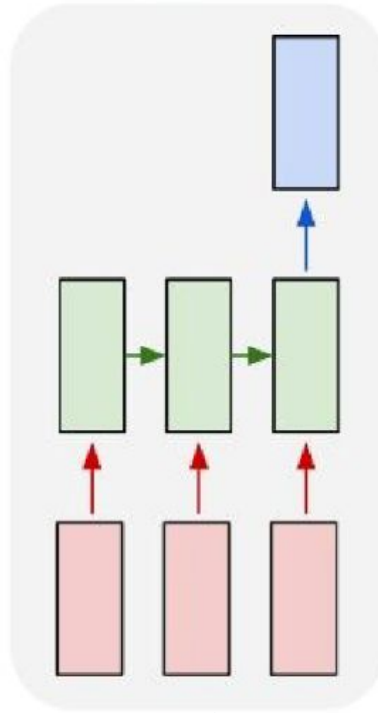


one to many



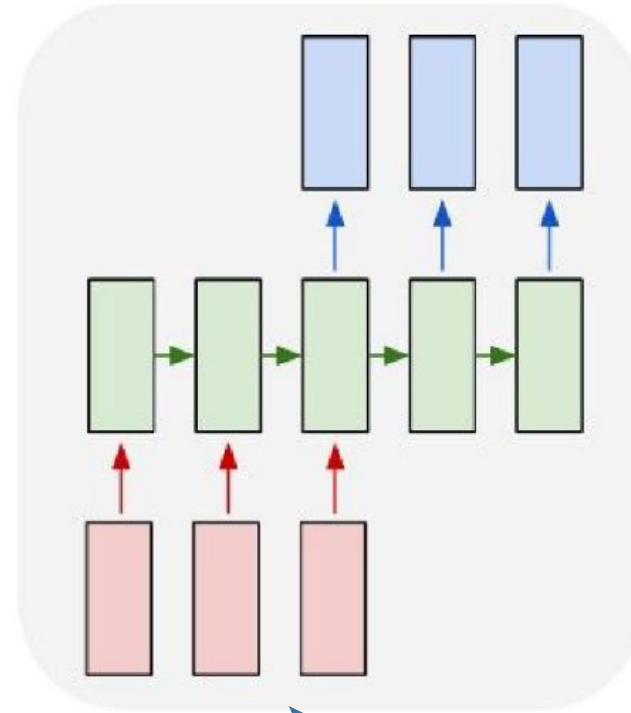
e.g. **Image Captioning**
image → sequence of words

many to one



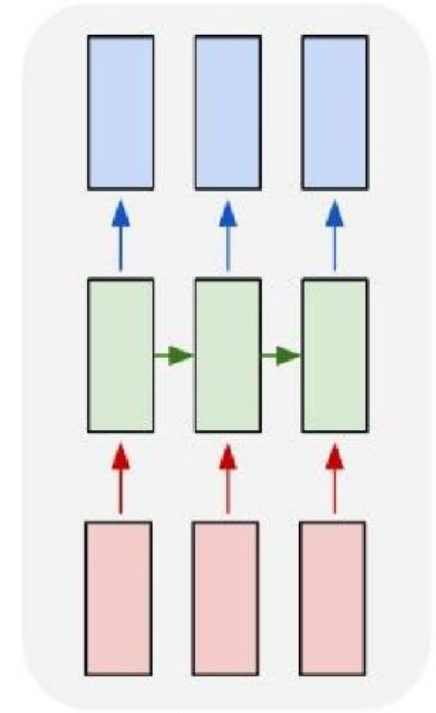
e.g. **Sentiment Classification**
sequence of words → sentiment

many to many



e.g. **Machine Translation**
seq of words → seq of words

many to many



e.g. Video classification on
frame level

References

- <https://www.youtube.com/watch?v=UNmqTiOnRfg>
- <https://www.youtube.com/watch?v=WCUNPb-5EYI>
- <https://www.youtube.com/watch?v=iX5V1WpxxkY>
- <https://www.edureka.co/blog/recurrent-neural-networks/>

Thank You

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