

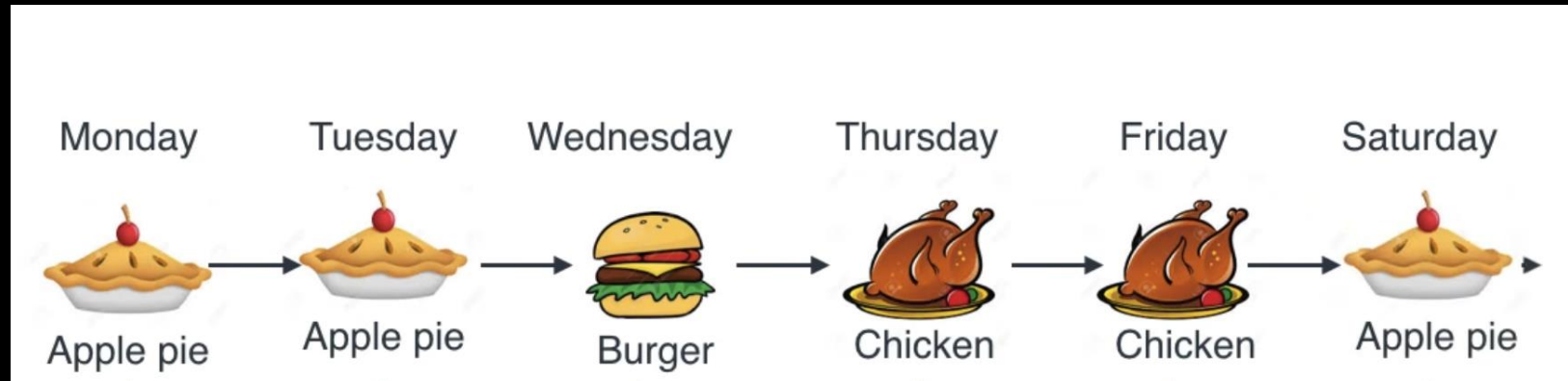
Redes Recurrentes

Las redes neuronales recurrentes (RNN) son un tipo de red neuronal diseñada para procesar datos secuenciales.

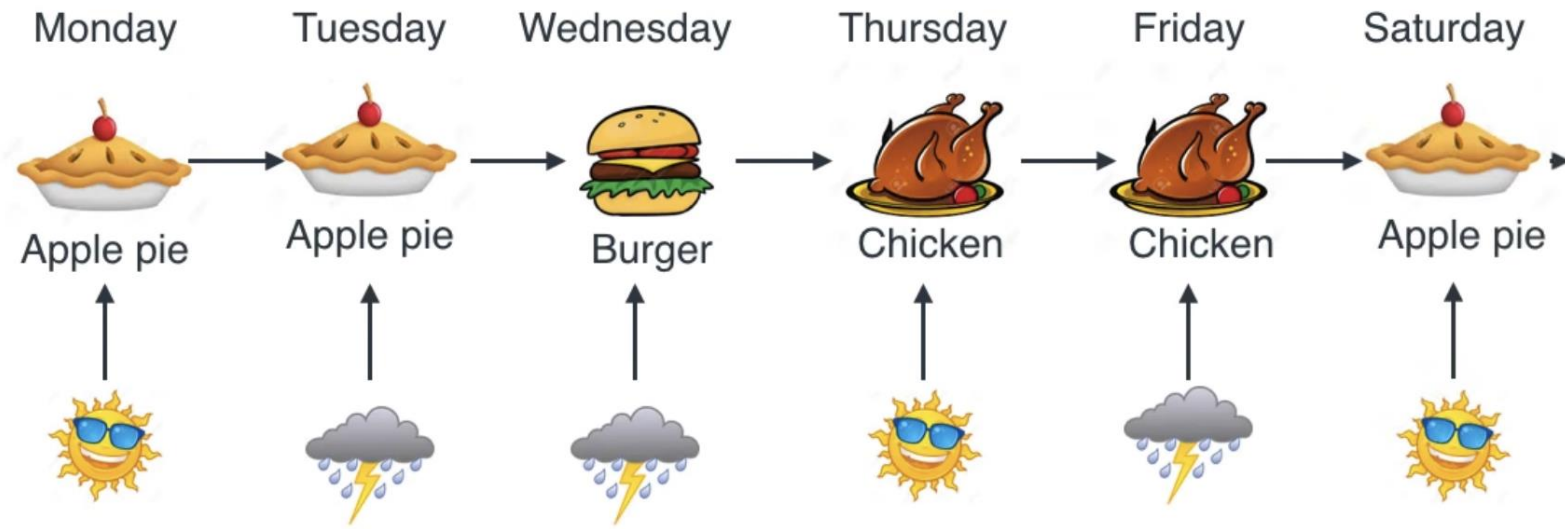
Ideal para problemas en los que la información de estados previos influye en la salida actual.

Permiten modelar dependencias temporales en series de tiempo, lenguaje y otras secuencias.

Como se decide el menu del dia?



Como se decide el menu del dia?



Por suerte a
alguien se le
ocurrio registrar
esto



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



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



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



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



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

Supongamos una preimera version del
esquema de las cartas, muy sencilla

Regla: soleado, contento, torta
nublado, hamburguesa

Matematica de la regla

Neural Network

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

$$\begin{bmatrix} 0 \\ 1 \end{bmatrix} \text{ ☁️⚡️ }$$

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

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

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

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

Matematica de la regla

Neural Network

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

$$\begin{bmatrix} 0 \\ 1 \end{bmatrix} \text{ ☁️⚡️ }$$

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \text{ ☁️⚡️ } =$$

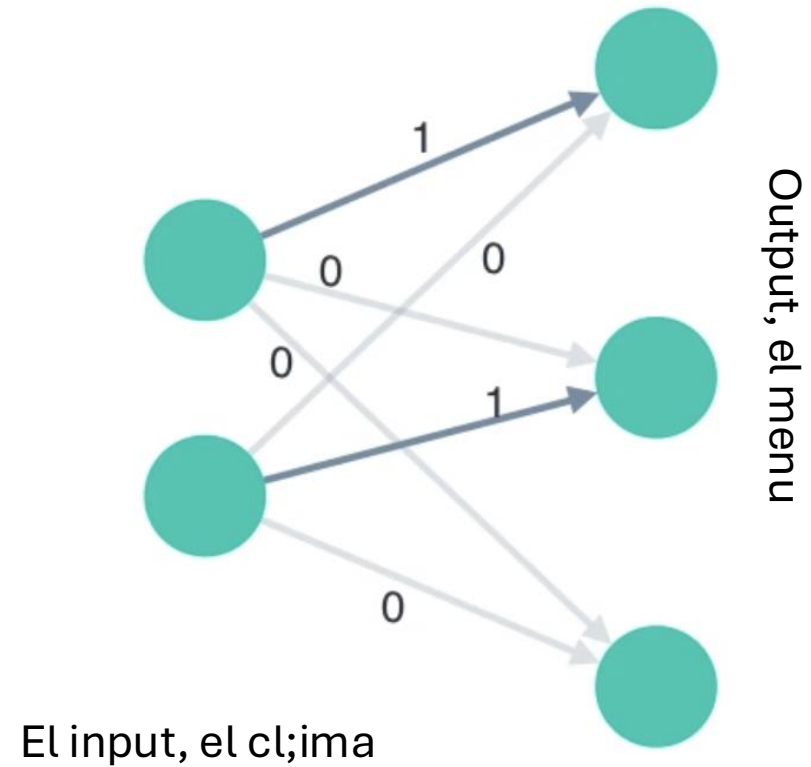
$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \text{ 🥧 }$$

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

La matematica, escrita como una matriz

Neural Network

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

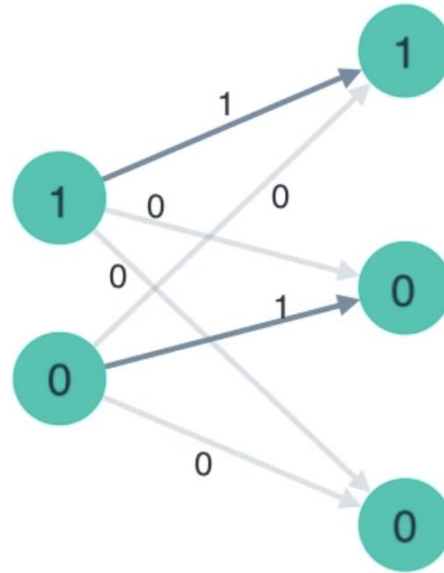


Neural Network

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



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



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



El cocinero TOC



Apple pie



Burger



Chicken



Simple Neural Network

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

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

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

$$\begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \text{ 🥧 } = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \text{ 🍔 }$$

Simple Neural Network

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



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



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



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

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



=

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



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



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

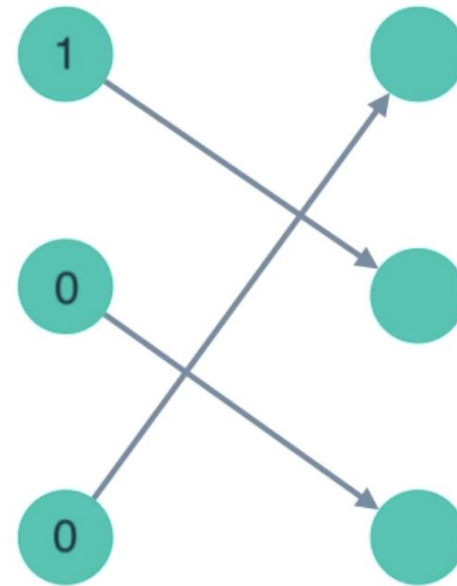


Simple (Recurrent) Neural Network

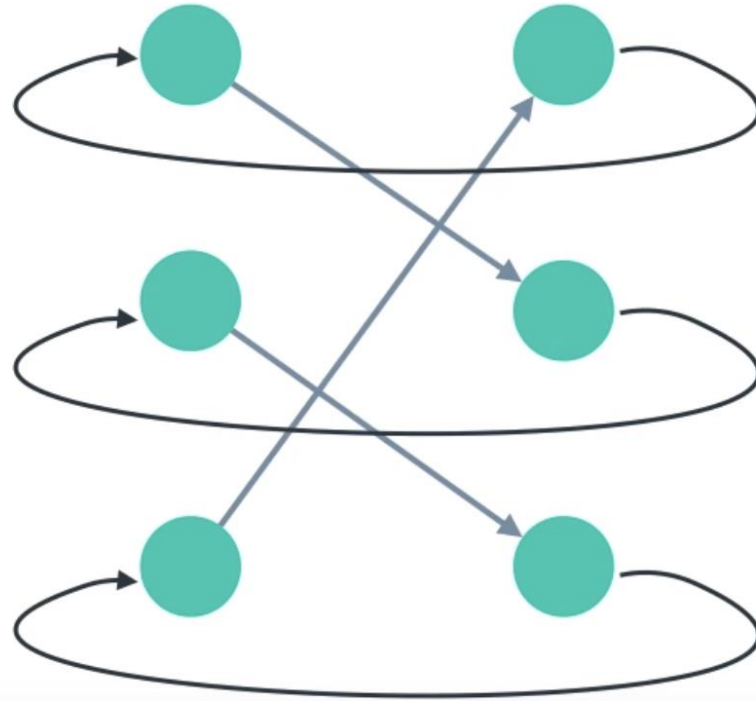
$$\begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$



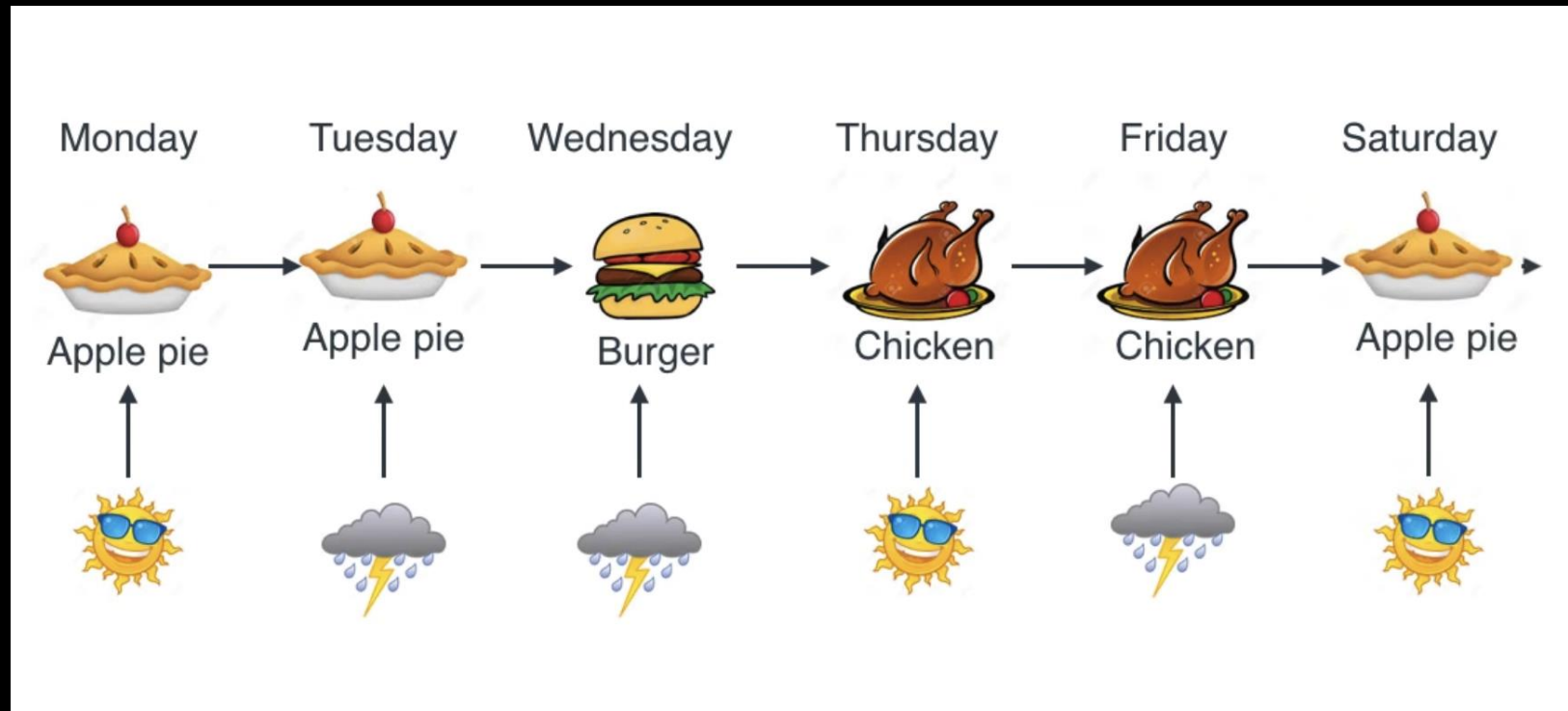
$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$



Simple (Recurrent) Neural Network

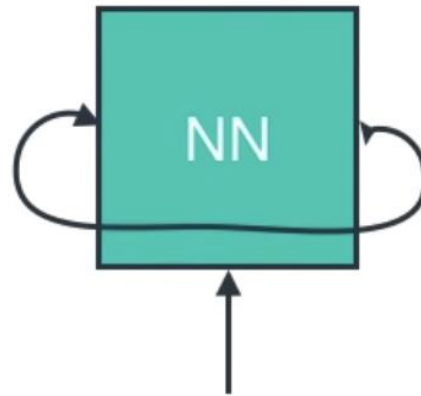


El cocinero que se consiguió una vida



Hay un orden, pero esta afectado por el clima

Recurrent Neural Network



Vectors



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



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



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



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



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

Vamos a necesitar unas prescripciones matematicas mas ricas...

More Complicated RNN

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

Food

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

Weather

Food

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \begin{img alt="Pie" data-bbox="235 290 312 380"/>$$

$$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \begin{img alt="Hamburger" data-bbox="245 465 303 555"/>$$

$$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \begin{img alt="Roast chicken" data-bbox="238 635 309 735"/>$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ \hline 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \begin{img alt="Pie" data-bbox="562 450 638 540"/> = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \hline 0 \\ 1 \\ 0 \end{bmatrix}$$

Food

Food

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


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


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


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

Food



=

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



Same



Next day

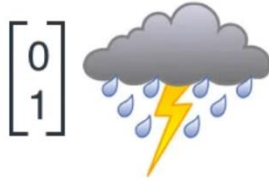
Weather

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

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

Weather

Weather



$$\begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ \hline 0 & 1 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} \text{Sun} \end{bmatrix} =$$

Weather

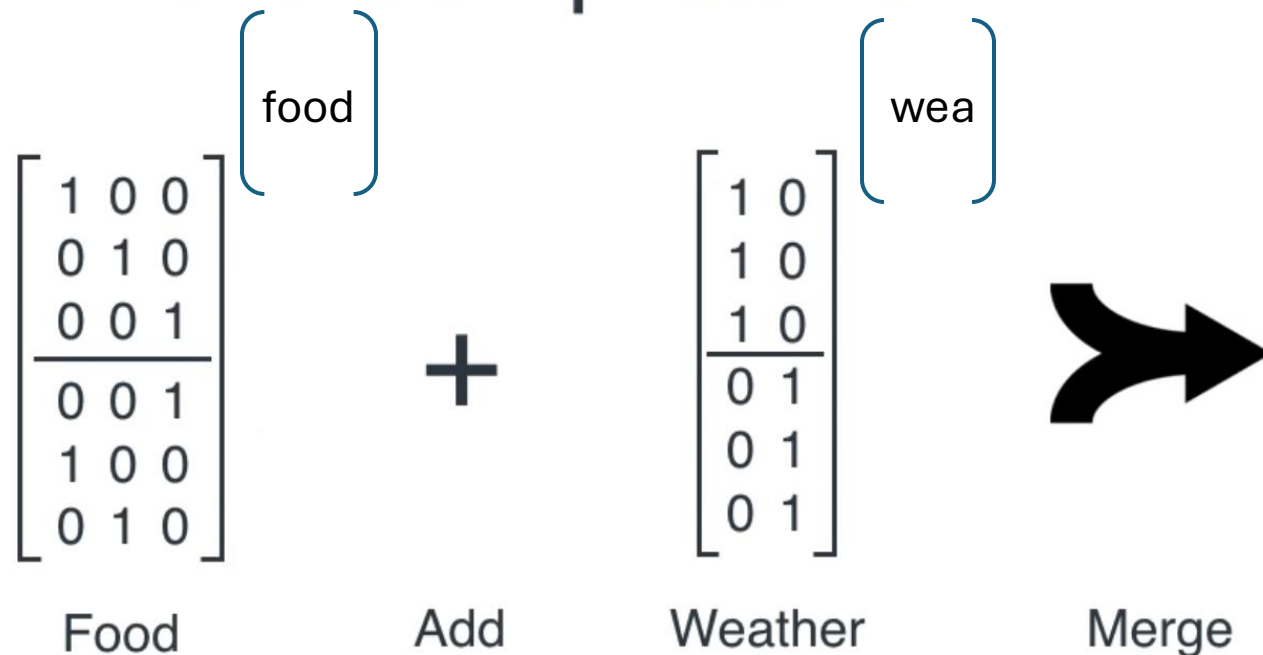
$$\begin{bmatrix} 1 \\ 1 \\ 1 \\ \hline 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} \text{Sun} \\ \text{Sun} \\ \text{Sun} \\ \text{Cloud} \\ \text{Cloud} \\ \text{Cloud} \end{bmatrix} \begin{matrix} \text{Same} \\ \text{Same} \\ \text{Same} \\ \text{Next day} \\ \text{Next day} \\ \text{Next day} \end{matrix}$$

Weather

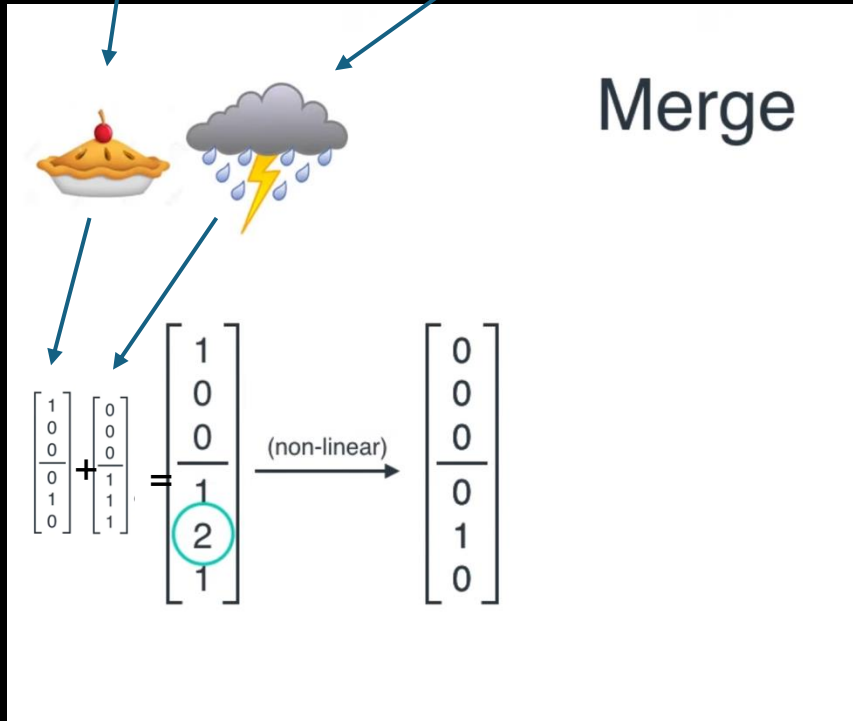
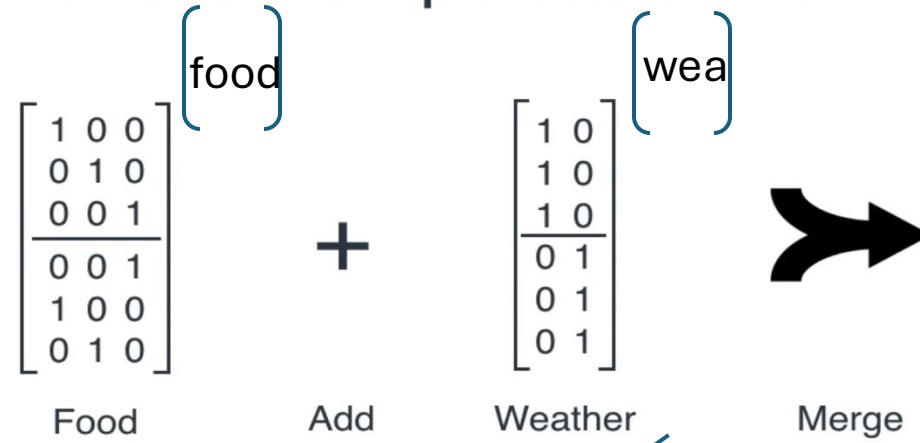
$$\begin{array}{l} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ ☀️ } \\ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \text{ ☁️⚡️ } \end{array} \quad \begin{array}{c} \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ \hline 0 & 1 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \text{ ☁️⚡️ } \\ \text{Weather} \end{array} = \begin{array}{l} \begin{bmatrix} 0 \\ 0 \\ 0 \\ \hline 1 \\ 1 \\ 1 \end{bmatrix} \text{ ☀️ } \text{ Same} \\ \text{☁️⚡️ } \text{ Next day} \end{array}$$

Vamos a necesitar unas prescripciones matematicas mas ricas...

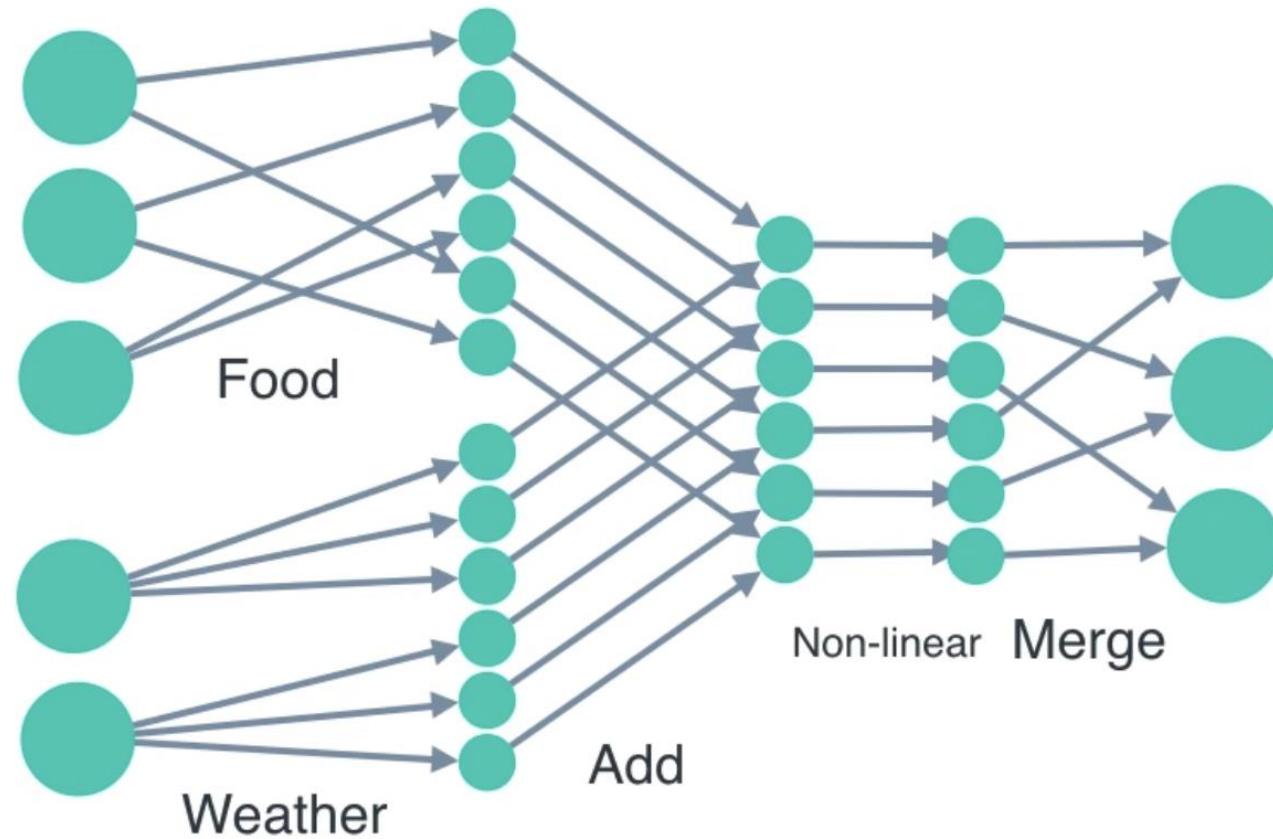
More Complicated RNN



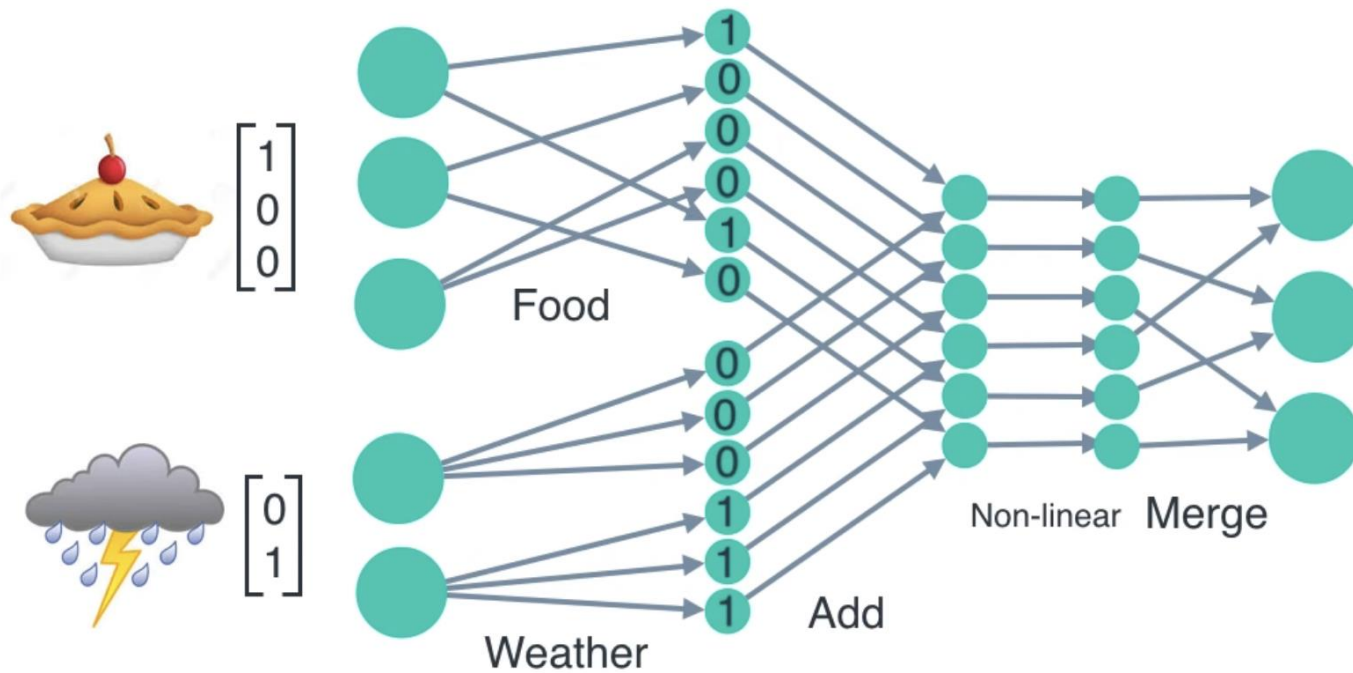
More Complicated RNN



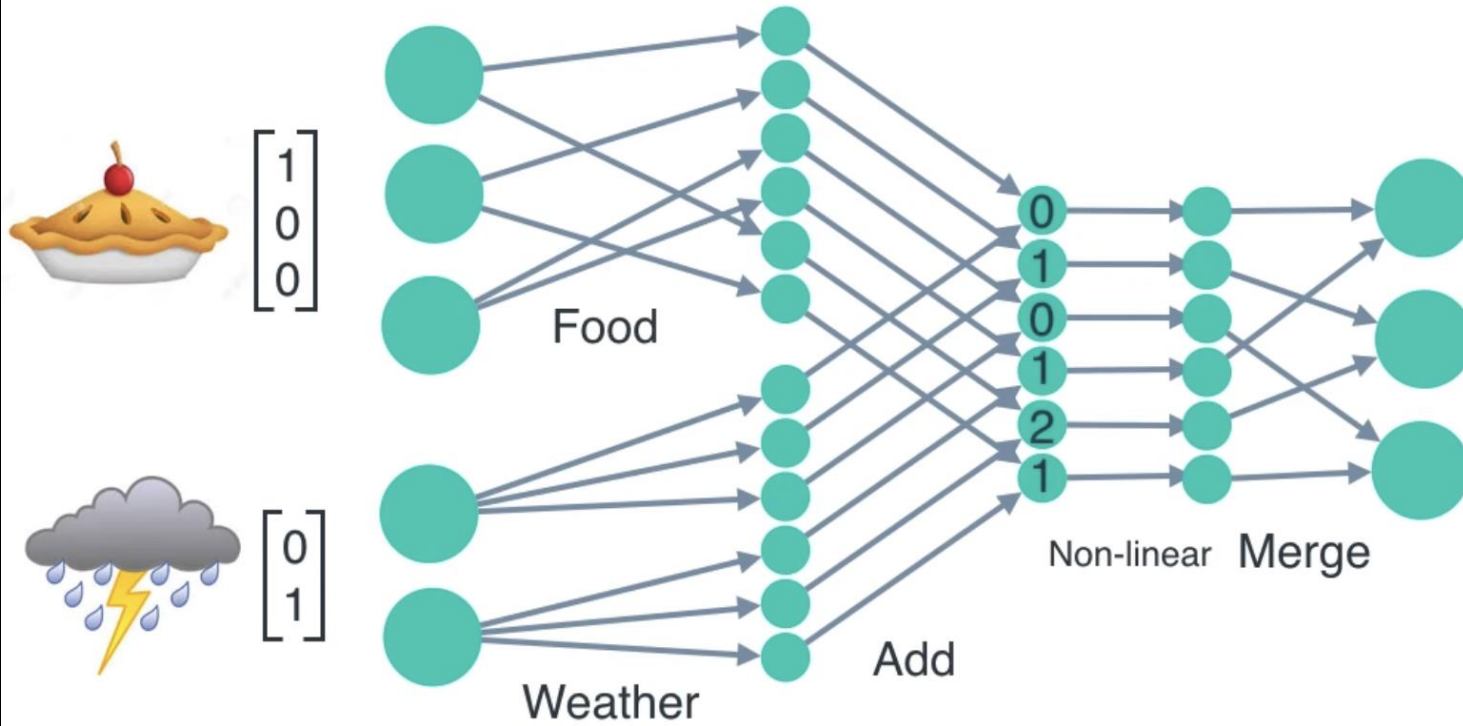
Recurrent Neural Network



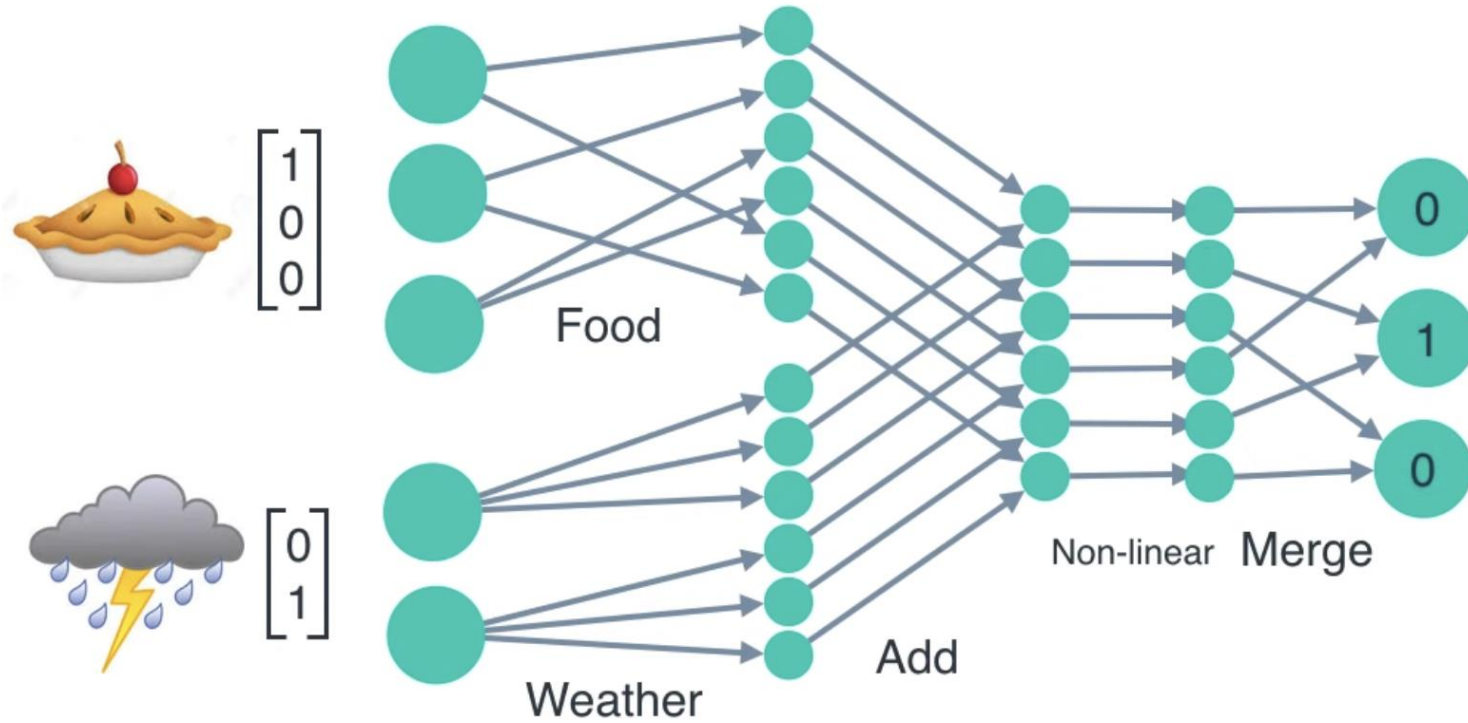
Recurrent Neural Network

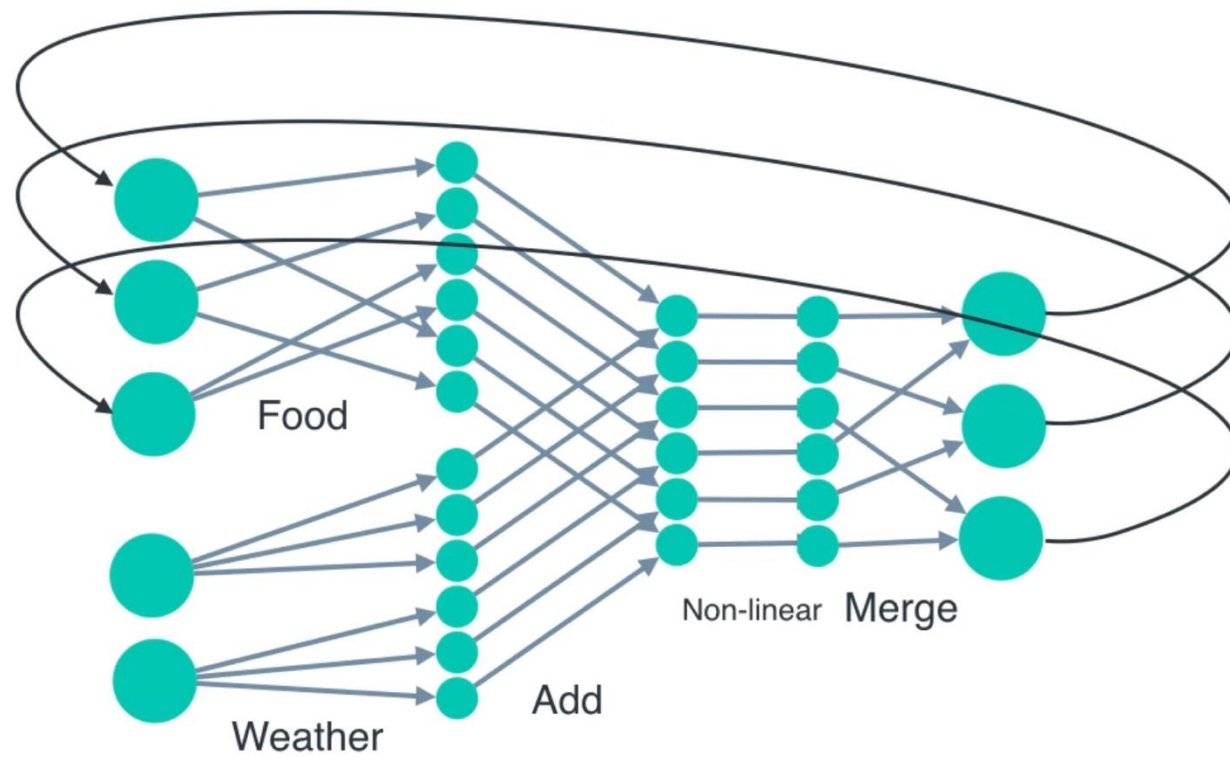


Recurrent Neural Network

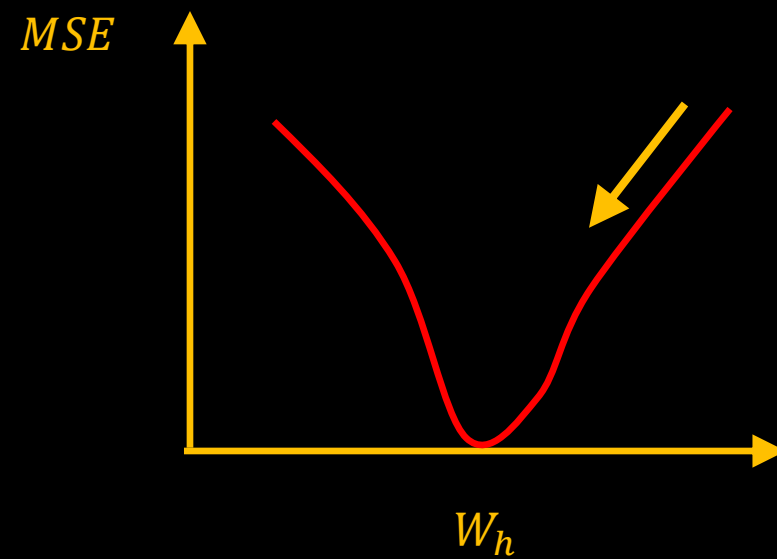


Recurrent Neural Network

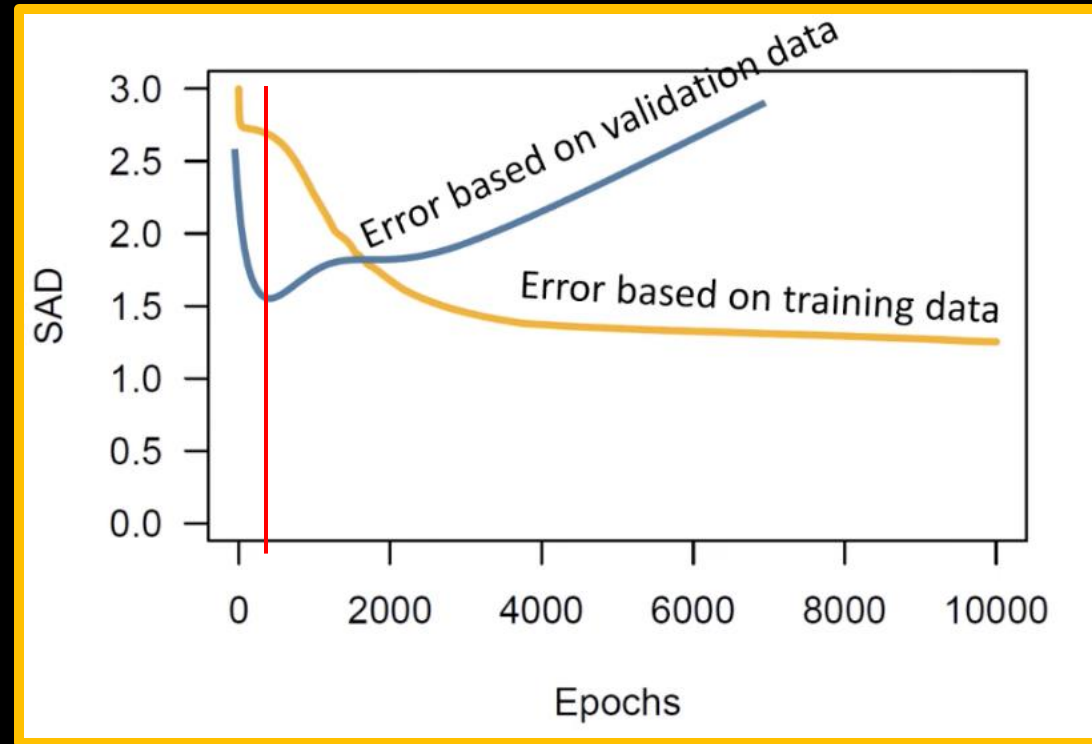




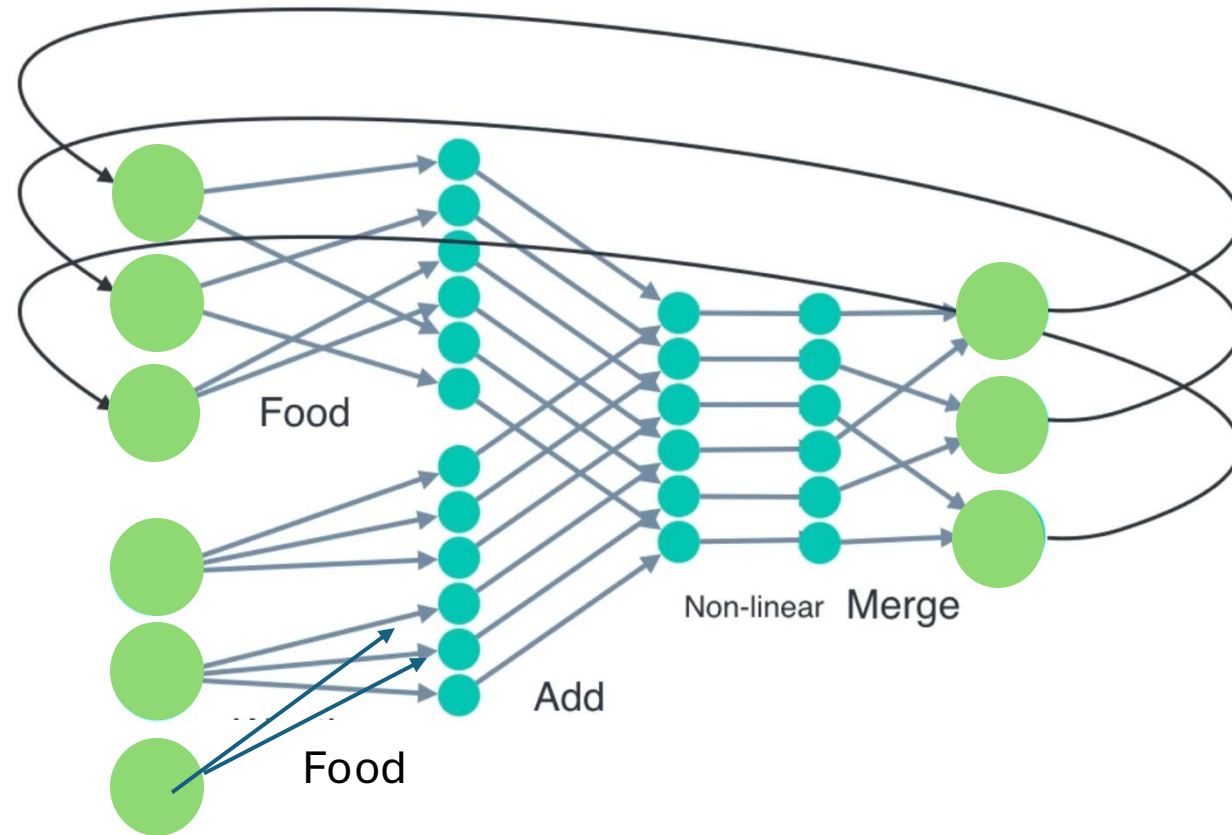
$$MSE = \sum_1^9 (\hat{y}_i - y_i)^2$$

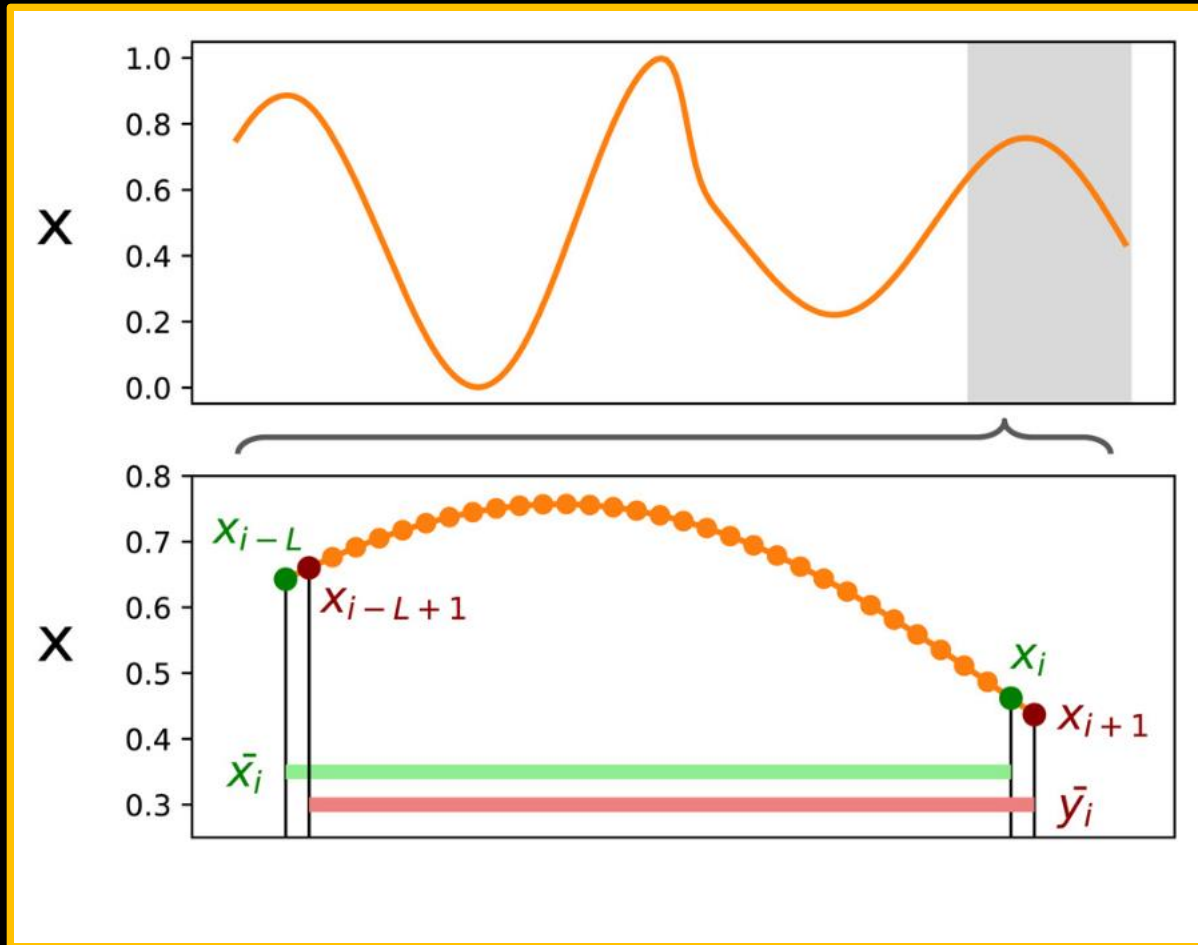


No tiene sentido y mas alla de la linea roja



Una señal temporal (vectorial) y punto...



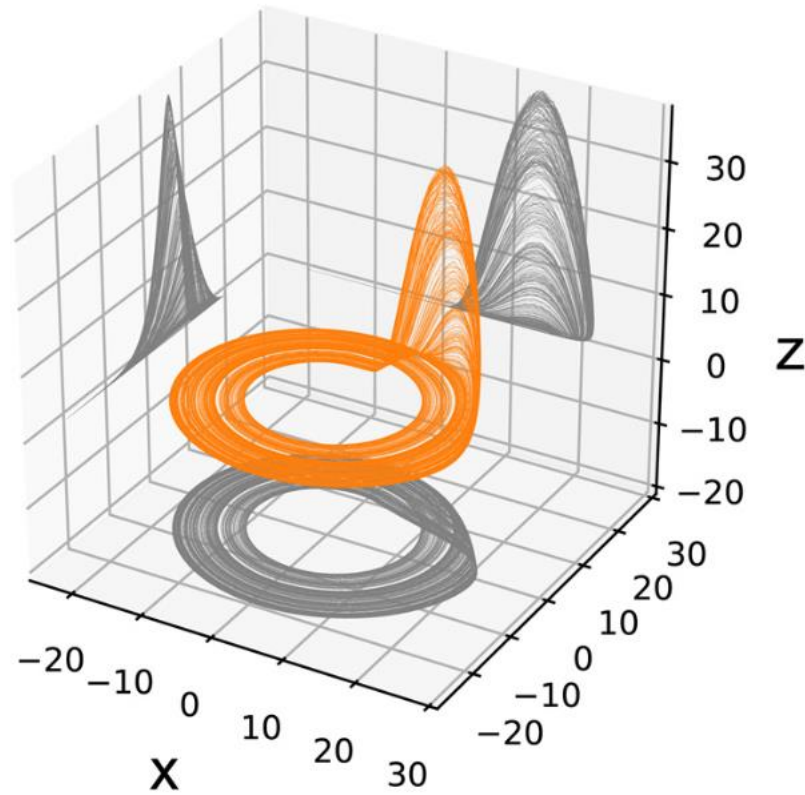


Mas bien unas cuarenta coordenadas, no 3, en este ejemplo...

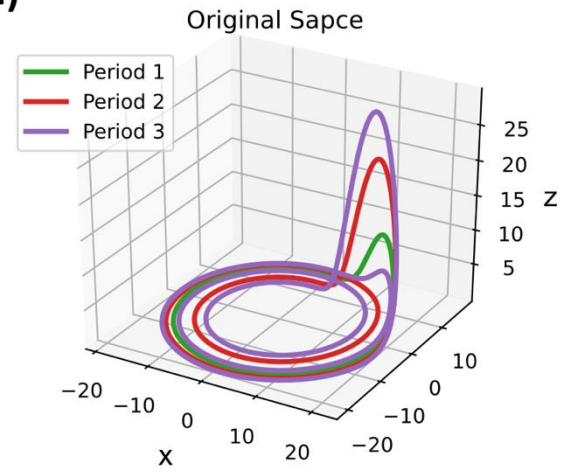
$$\frac{dx}{dt} = -y - z$$

$$\frac{dy}{dt} = x + ay$$

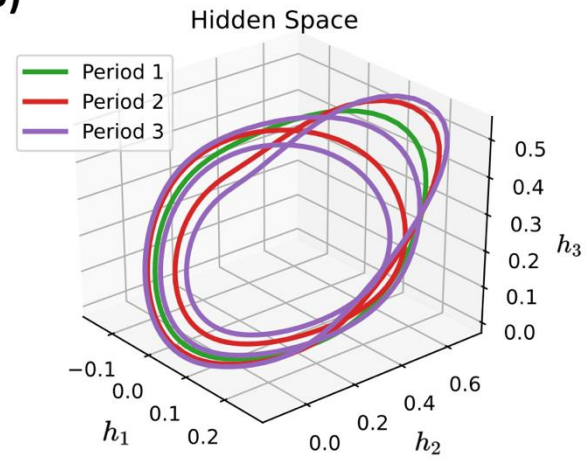
$$\frac{dz}{dt} = b + z(x - c)$$



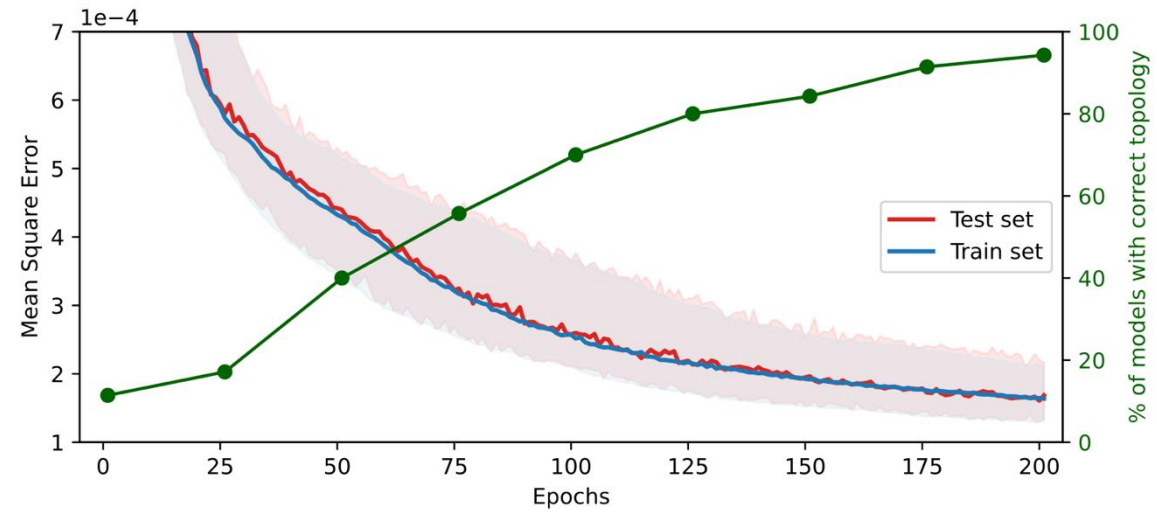
a)



b)



c)



1) RNN Simple

 Keras

```
model = Sequential()  
model.add(SimpleRNN(5, activation='tanh', input_shape=(look_back, n_features)))  
model.add(Dense(1, activation='linear'))  
model.compile(optimizer='adam', loss='mse')
```

Importante: Los datos de entrada deben tener la forma:

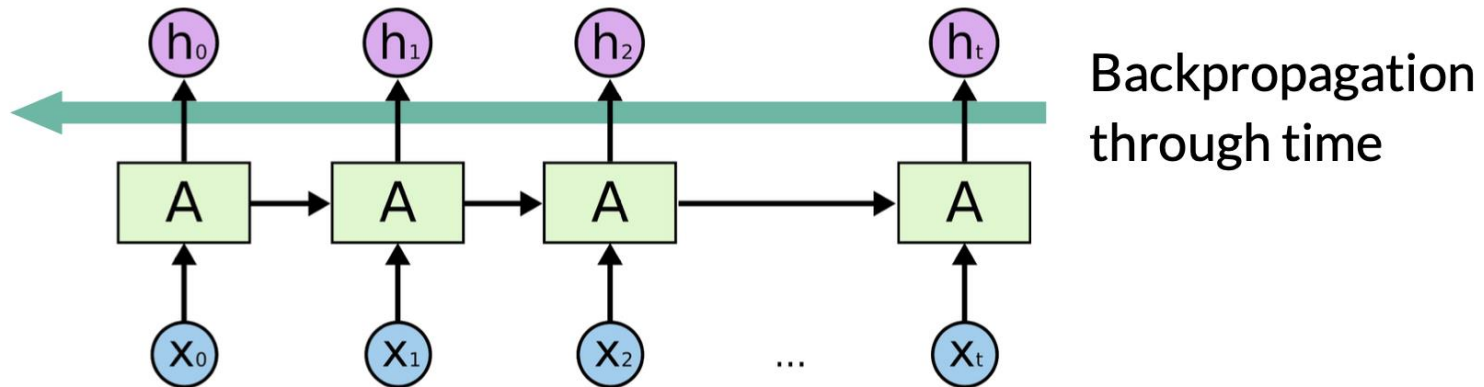
`X.shape = [num_samples, look_back, n_features]`

En la señal anterior, num_samples es del orden de 40000, los look back unos 40, y las features 1

1) RNN Simple

**Vanishing gradient
problem**

Problema! Les cuesta
aprender dependencias de
largo plazo.



Attention Is All You Need

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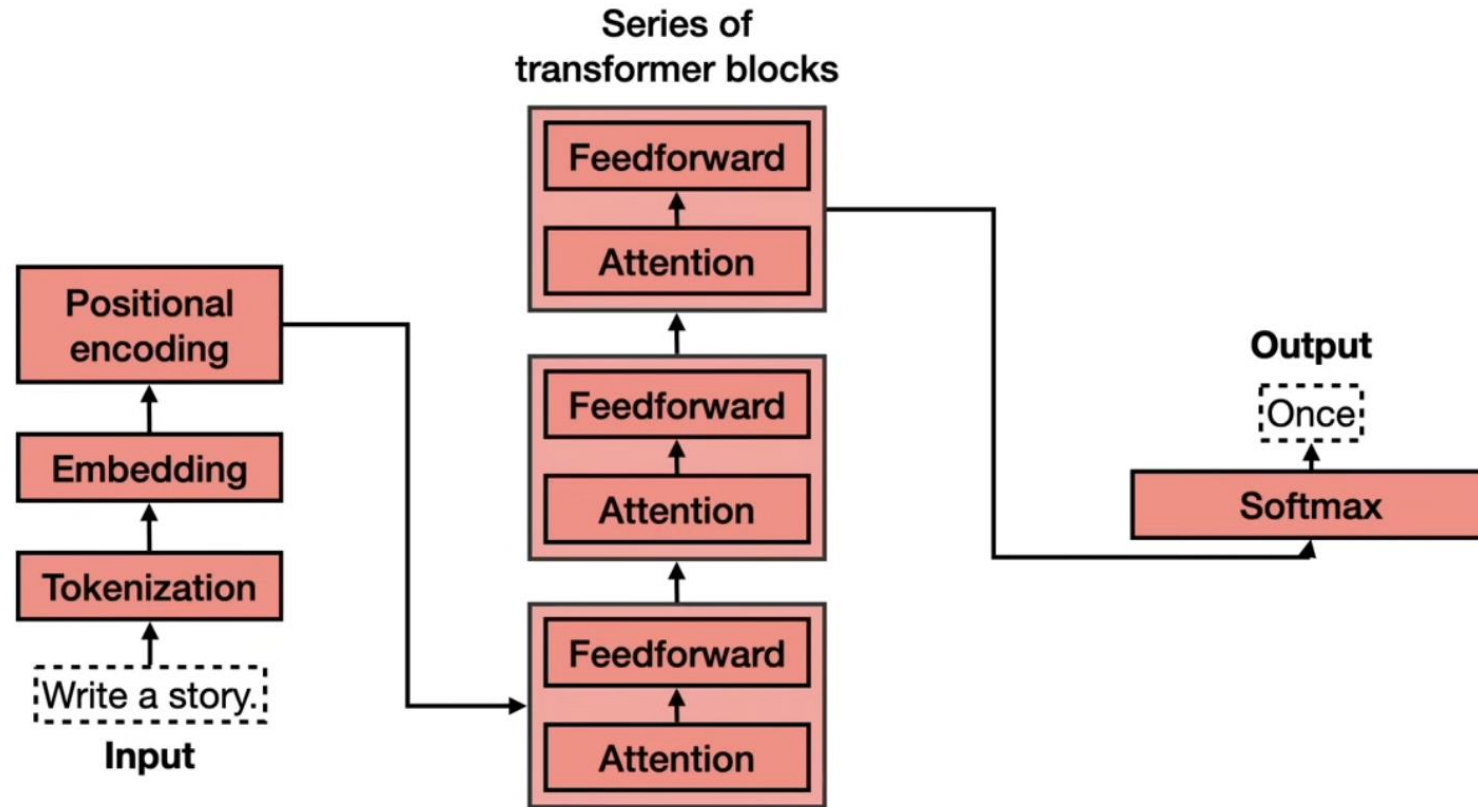
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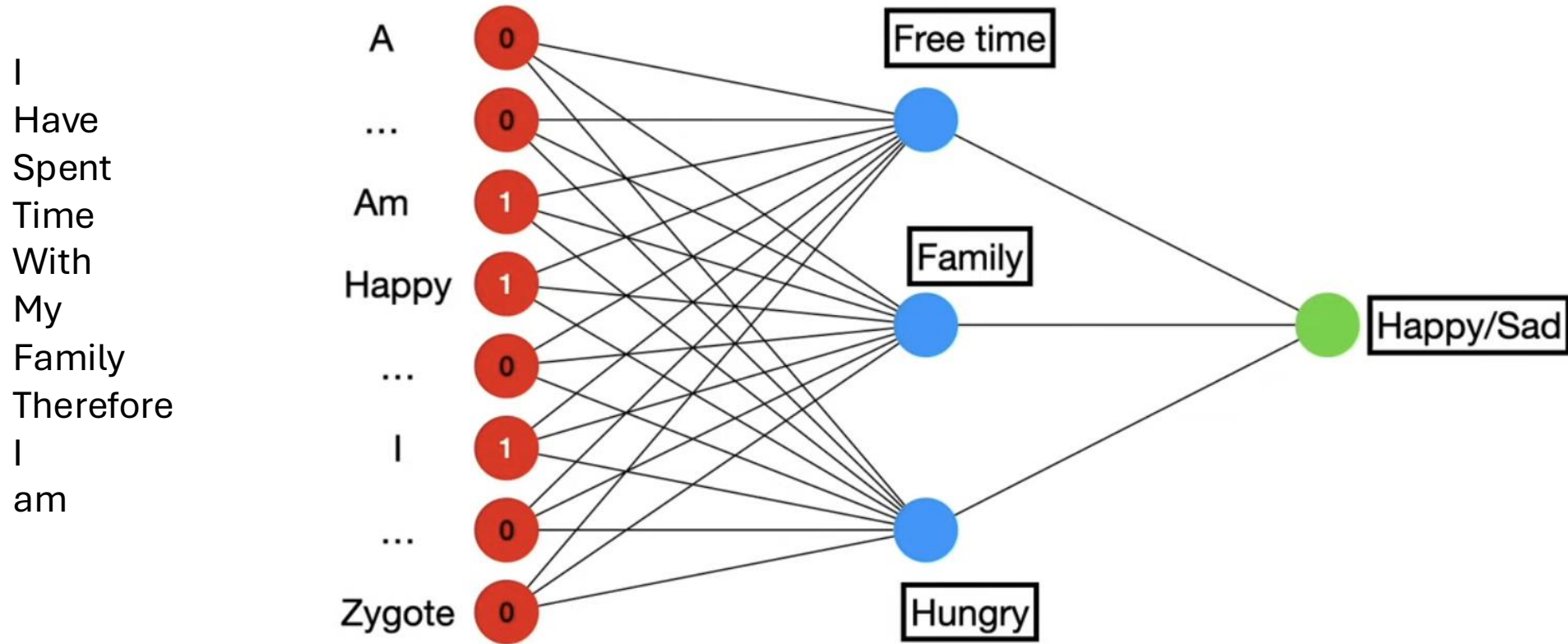
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Transformers... prediccion, palabra por palabra.



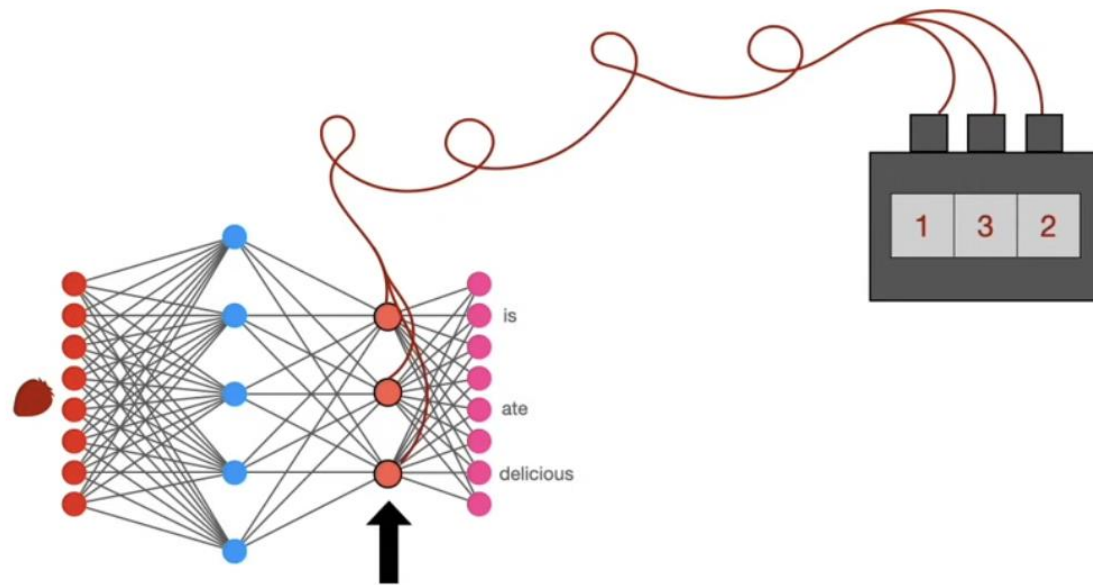
La construccion de un embedding “semantico”

Sea una red “all to all”,
preparada para predecir la proxima palabra



Un modo de hacer embeddings

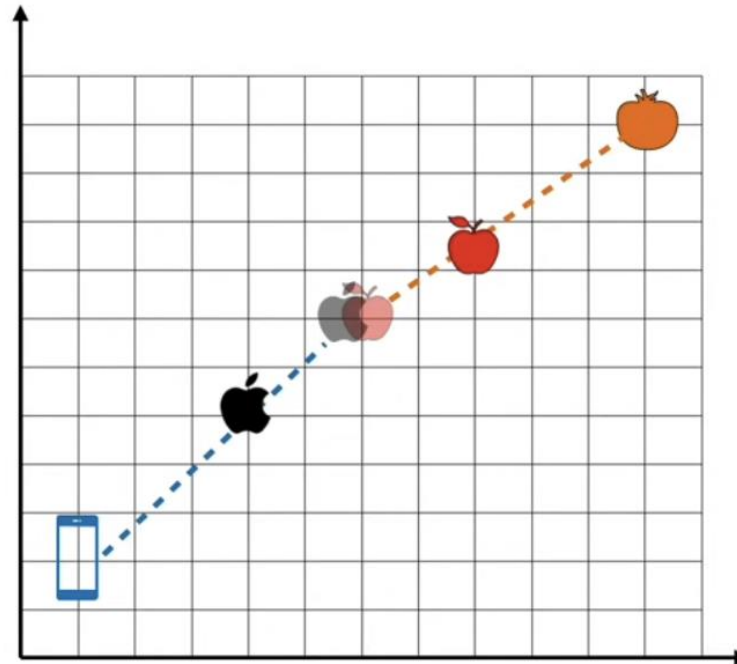
Representacion via los valores numericos de la capa anteultima



Word	Numbers		
Strawberry	1	3	2
Apple	1.1	2.9	2.2

Attention

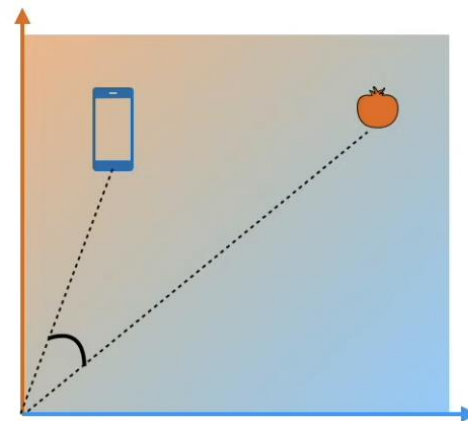
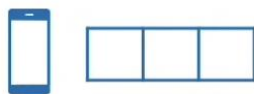
**Un modo de que las palabras sean afectadas por el contexto:
Alterar el embedding de las palabras segun las demas del contexto**



please buy an **apple** and an **orange**

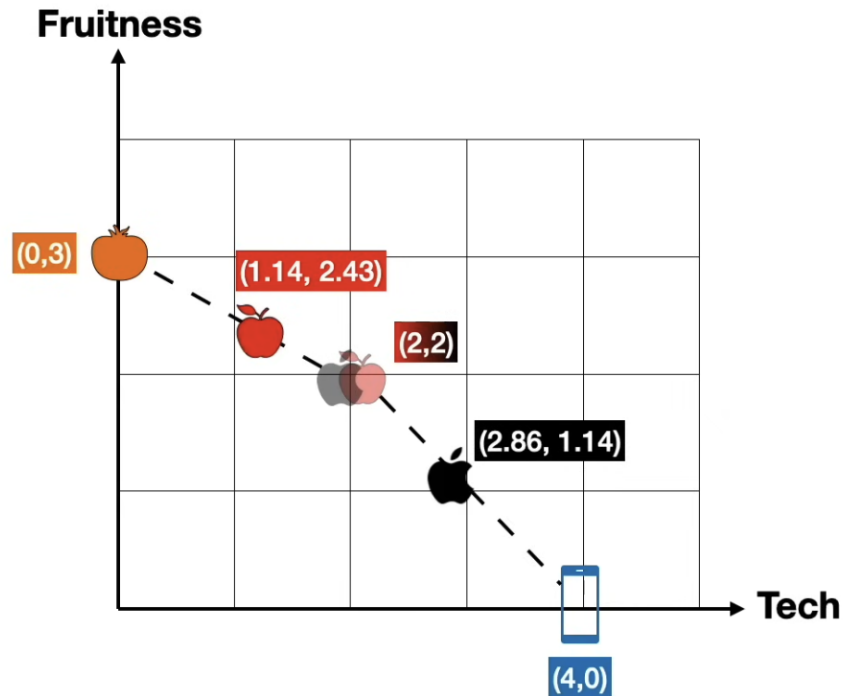
apple unveiled the new **phone**

Similarity



$$\text{Similarity}(\text{tomato}, \text{smartphone}) = \begin{bmatrix} \square & \square & \square \end{bmatrix} \begin{bmatrix} \square \\ \square \\ \square \end{bmatrix}$$

Attention refiere a moverse en el espacio de embedding
en la direccion que sugiera la similitud



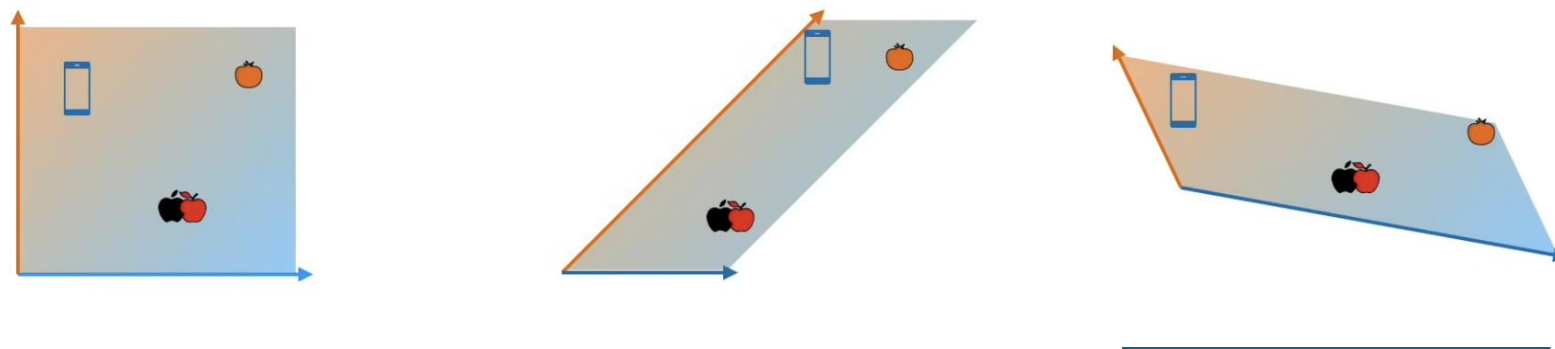
an **apple** and an orange

$$\text{Apple} \rightarrow 0.43 \text{ Orange} + 0.57 \text{ Apple}$$

an **apple** phone

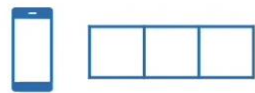
$$\text{Apple} \rightarrow 0.43 \text{ Phone} + 0.57 \text{ Apple}$$

Ahora, para mover según el contexto,
es mejor realizar una transformación lineal

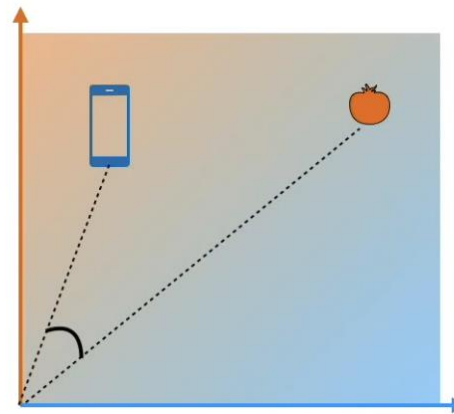


(donde el corrimiento funcione mejor)

Similarity

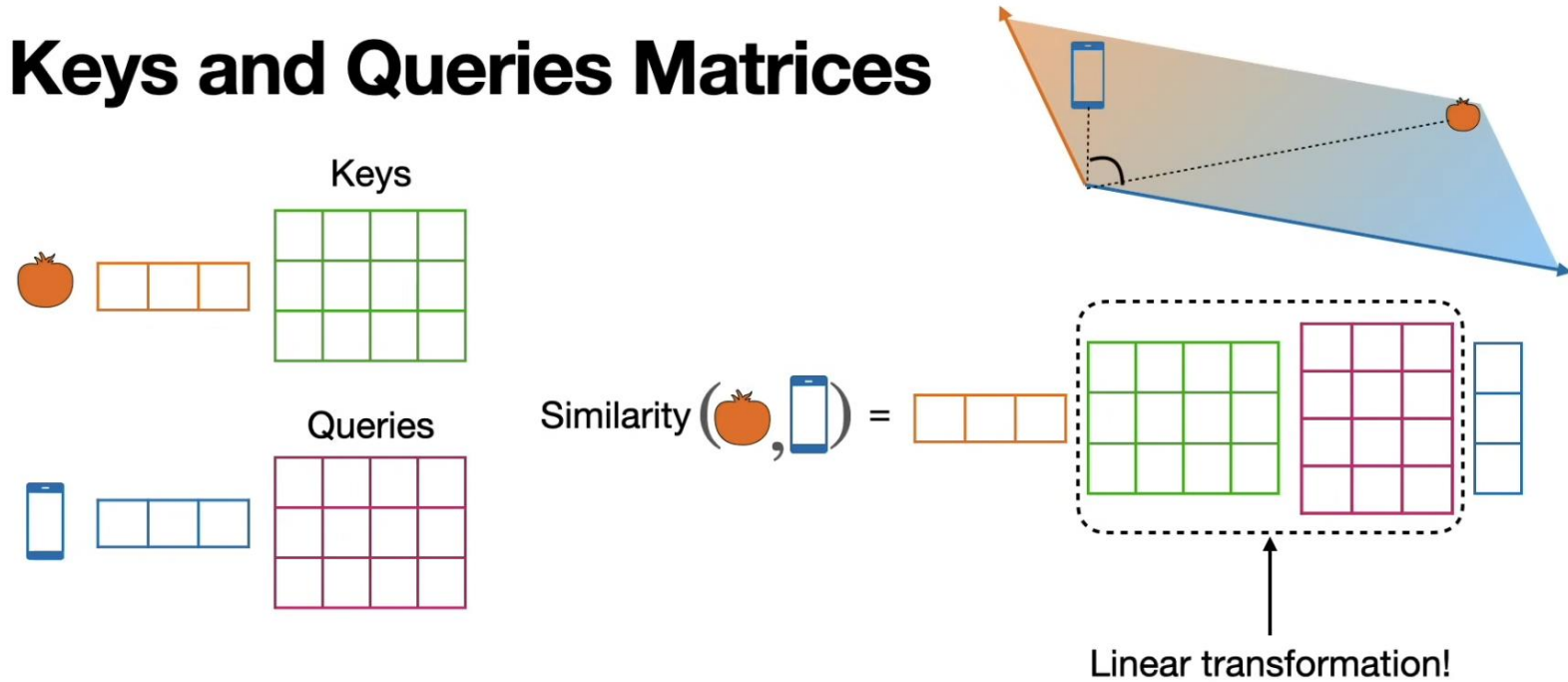


$$\text{Similarity}(\text{tomato}, \text{smartphone}) = \begin{bmatrix} \square & \square & \square \end{bmatrix} \begin{bmatrix} \square \\ \square \\ \square \end{bmatrix}$$

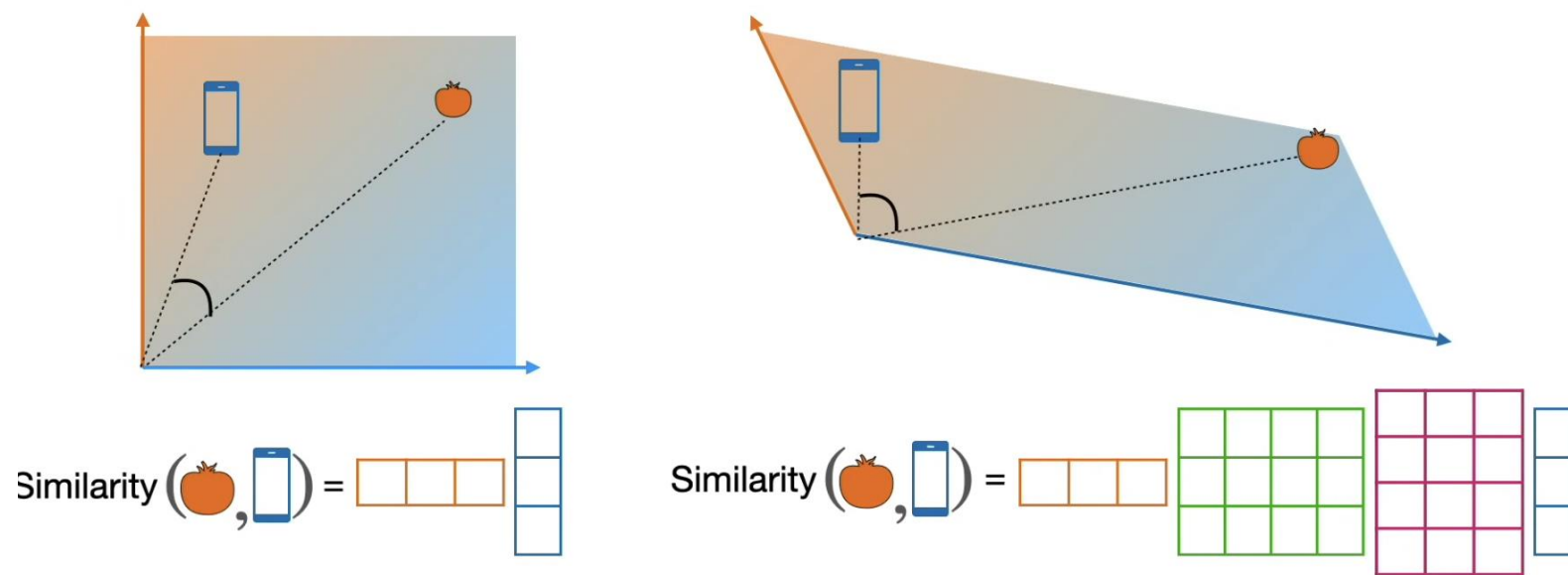


Un nuevo embedding

Keys and Queries Matrices

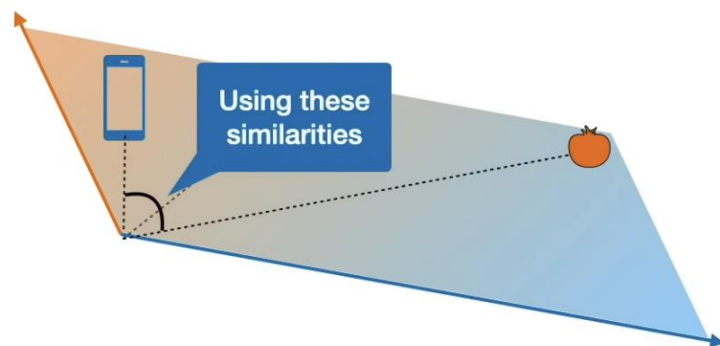


Similarity on a transformed embedding



Y la idea es que attention funcione mejor en este nuevo embedding

Values matrix

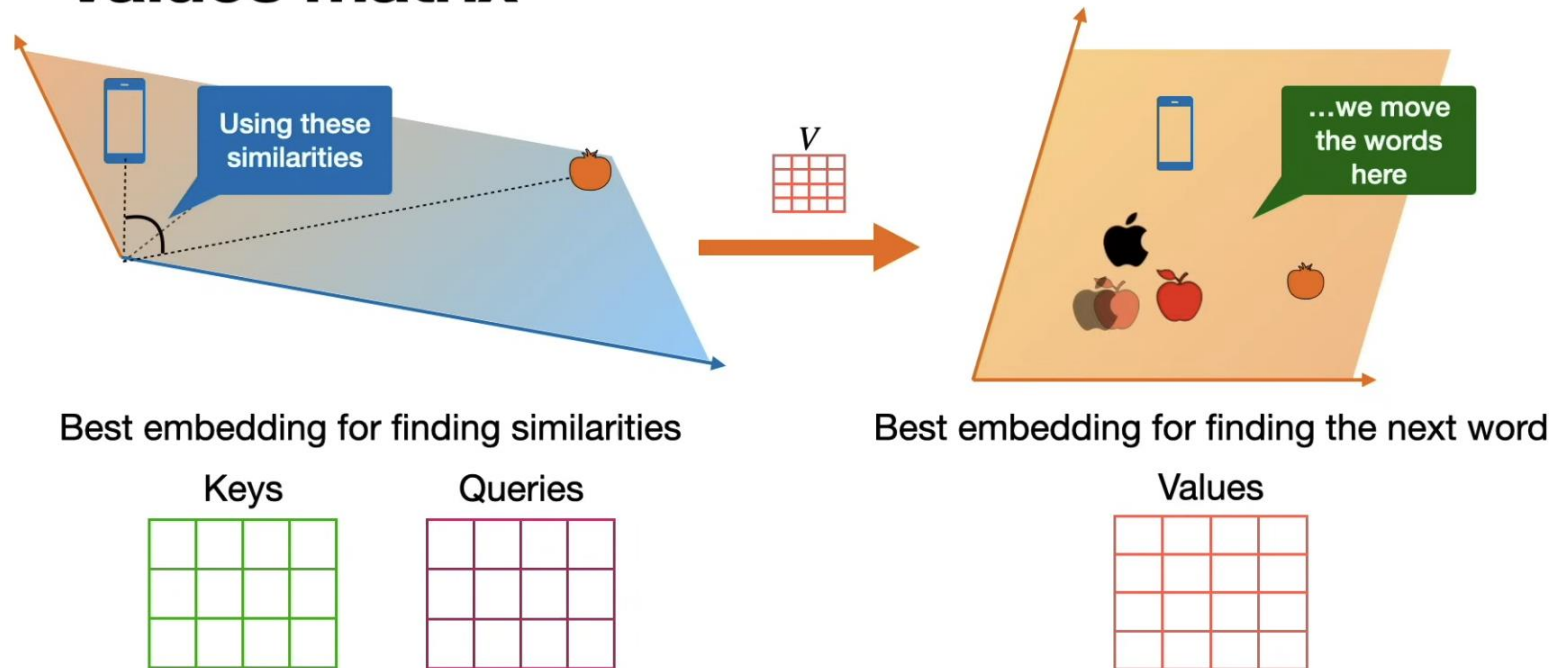


Best embedding for finding similarities



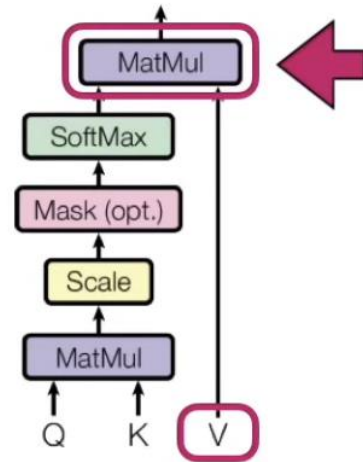
Best embedding for finding the next word

Values matrix

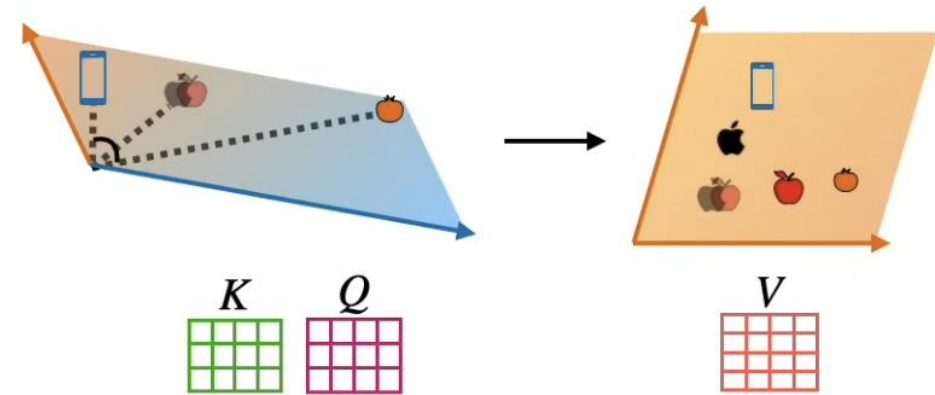


Self-attention

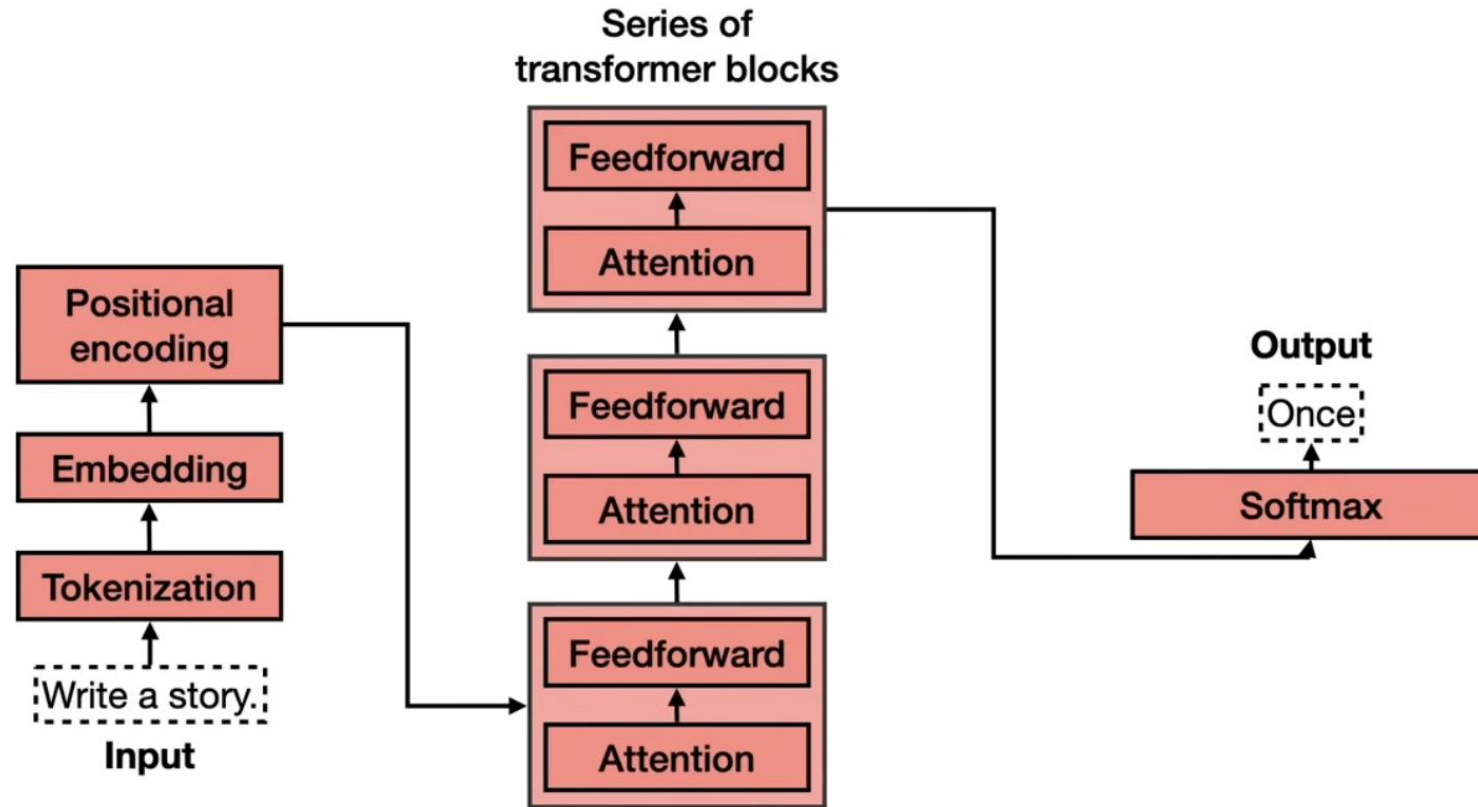
Scaled Dot-Product Attention

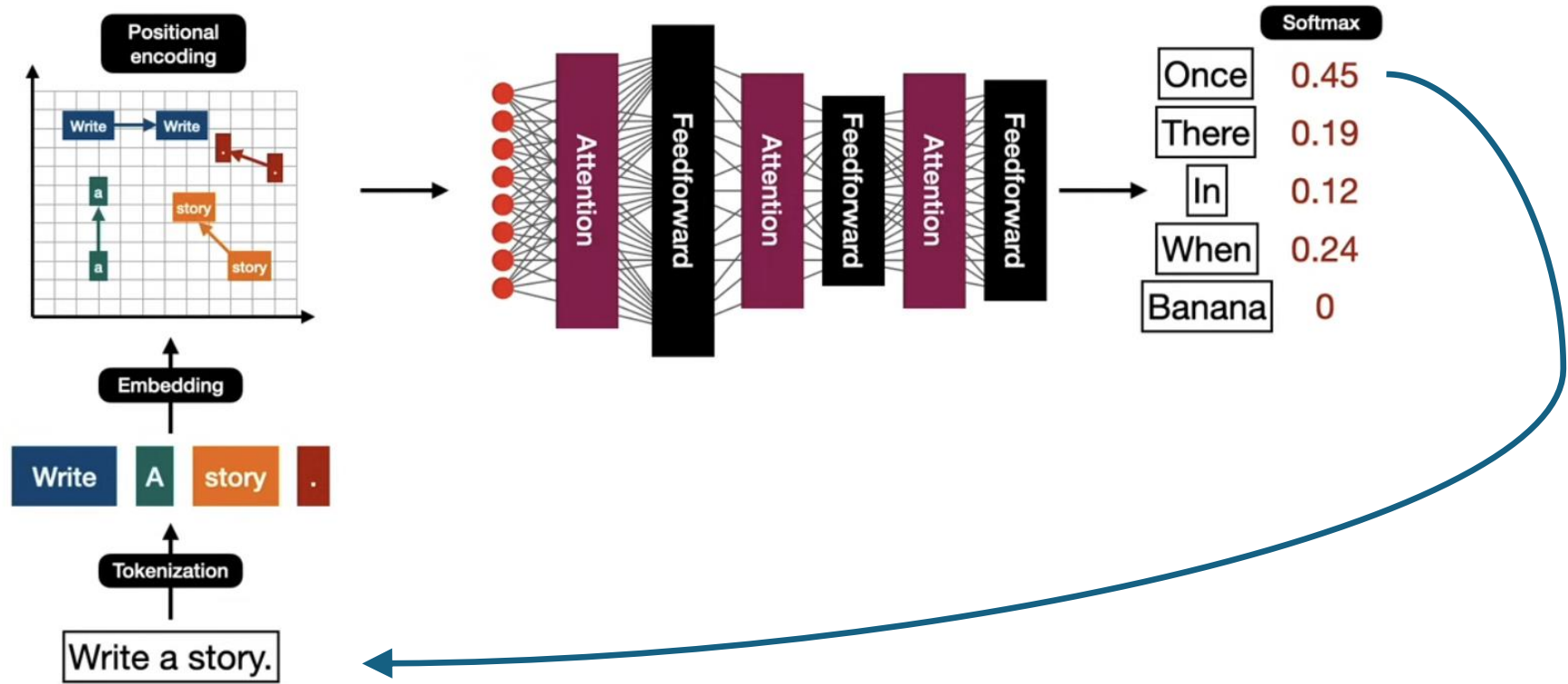


$$\text{Attention}(Q, K, V) = \text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right) V$$



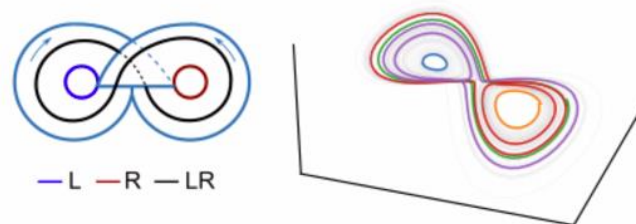
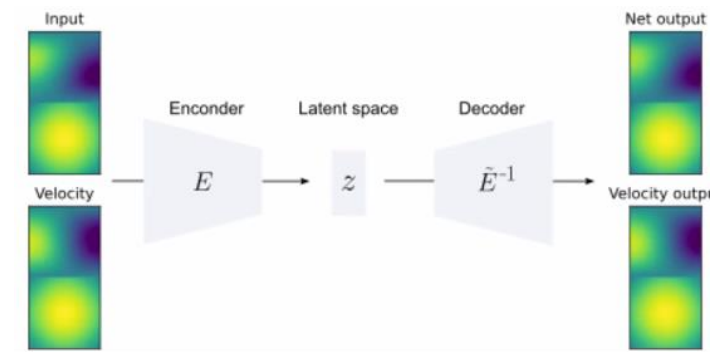
Transformers... prediccion, palabra por palabra.



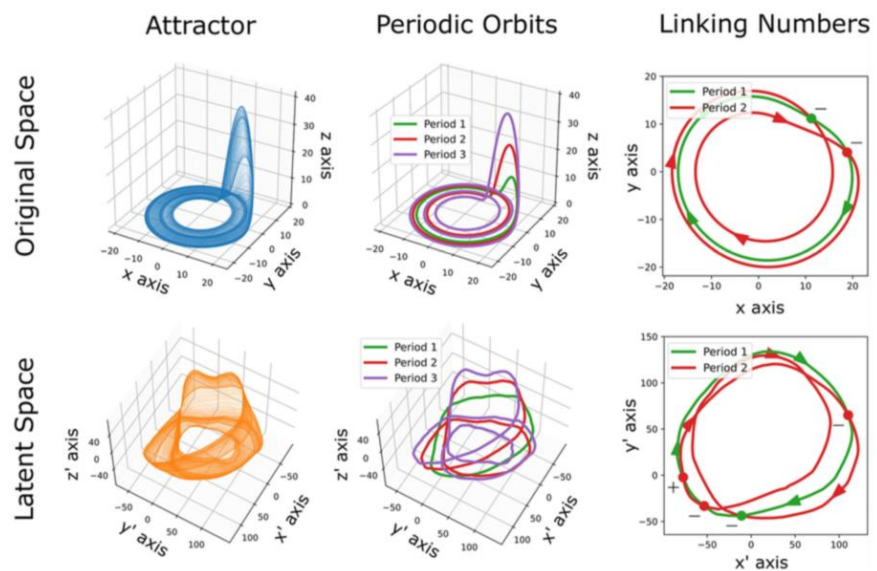


The picture that emerges is that the neural networks used in different implementations of IA projects, do recreate the original flow of the dynamics.

AE trained with movie



AE trained with temporal data



RNN trained with temporal data

