

# Sequence Modeling

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# Sequential Data Modeling

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- ▶ **Sequential Data**

- ▶ Most of data are sequential
- ▶ Speech, Text, Image, ...

- ▶ **Deep Learnings for Sequential Data**

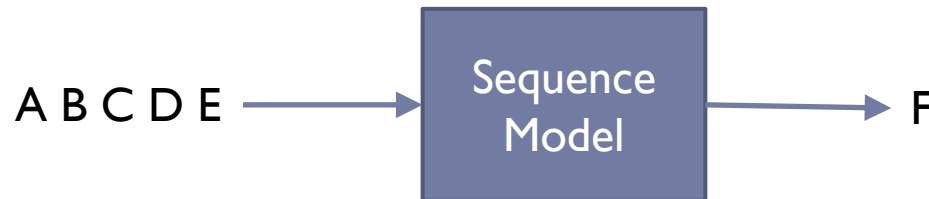
- ▶ **Convolutional Neural Networks (CNN)**
  - ▶ Try to find local features from a sequence
- ▶ **Recurrent Neural Networks: LSTM, GRU**
  - ▶ Try to capture the feature of the past

# Sequential Data Modeling

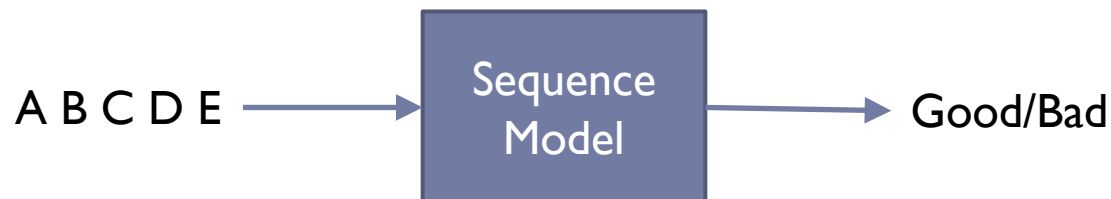
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## ▶ Three Types of Problems

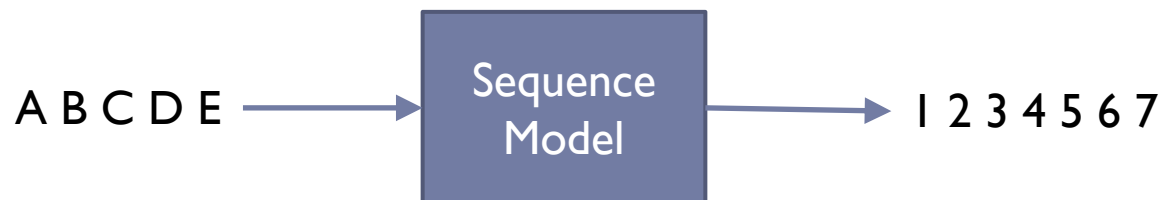
### ▶ Next Step Prediction



### ▶ Classification



### ▶ Sequence Generation



# Sequential Data Modeling

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## ▶ Sequence Generation

### ▶ Machine Translation

This is a very good wine → C'est un très bon vin

### ▶ Speech Recognition



→ This is a very good wine

### ▶ Image Caption Generation



→ A bird is flying

# Types of Processes

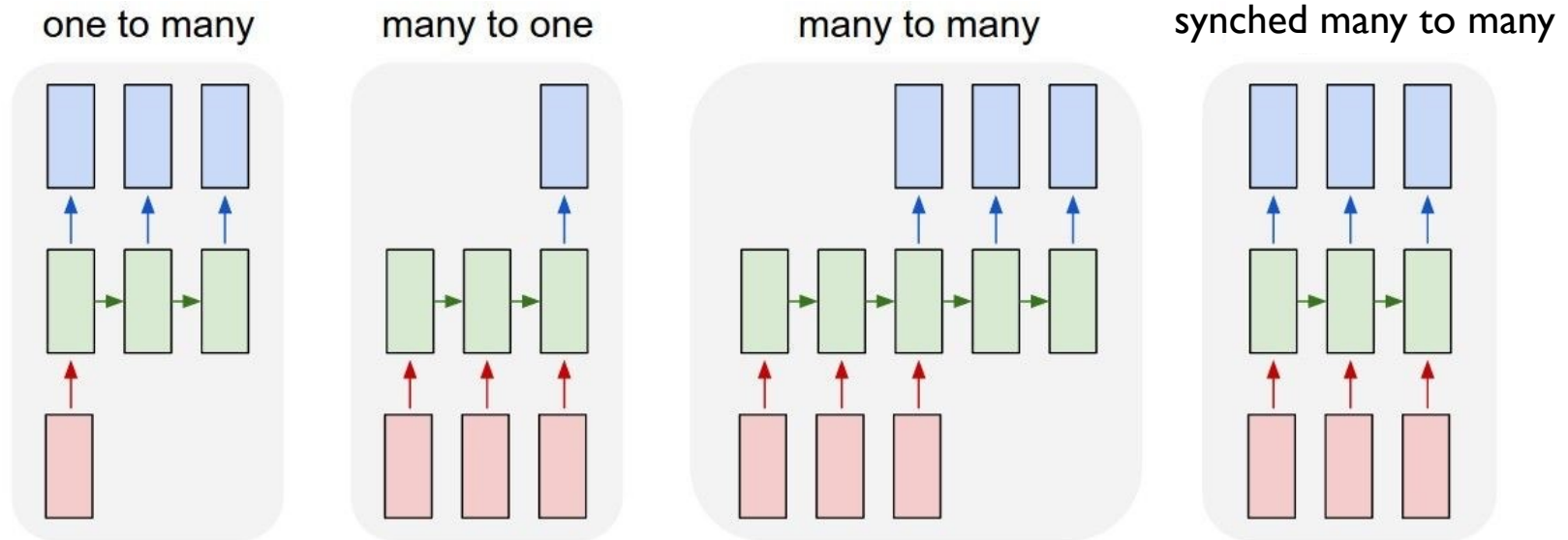
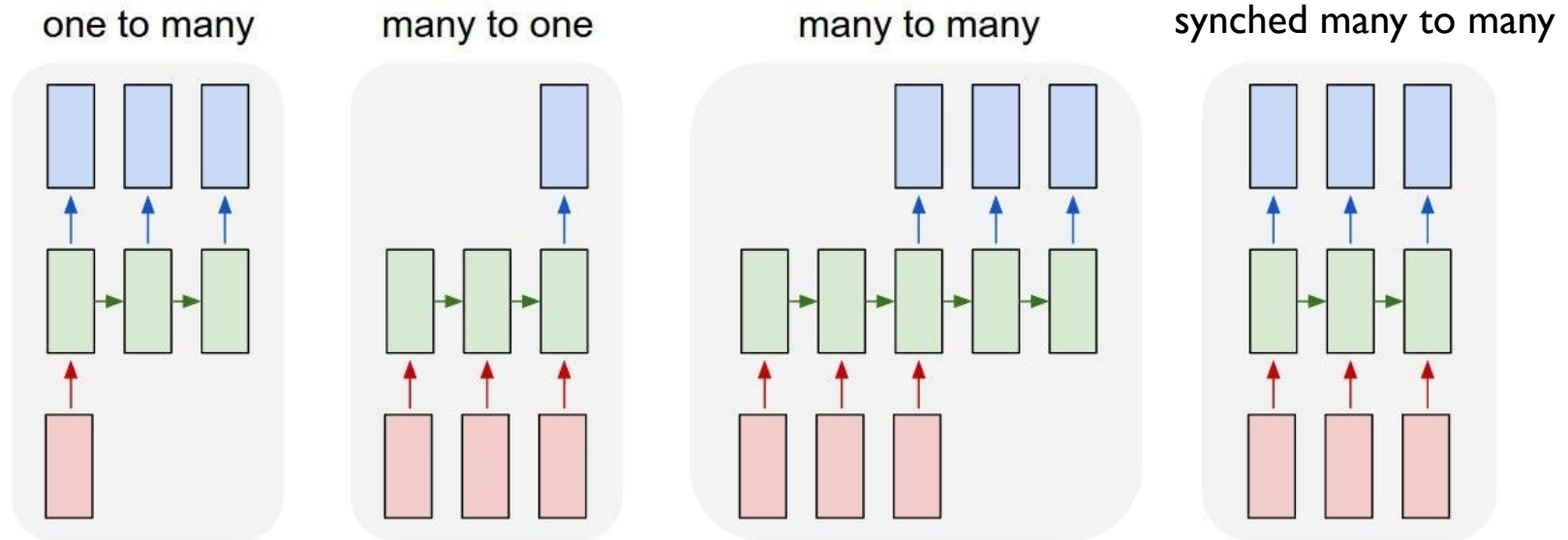


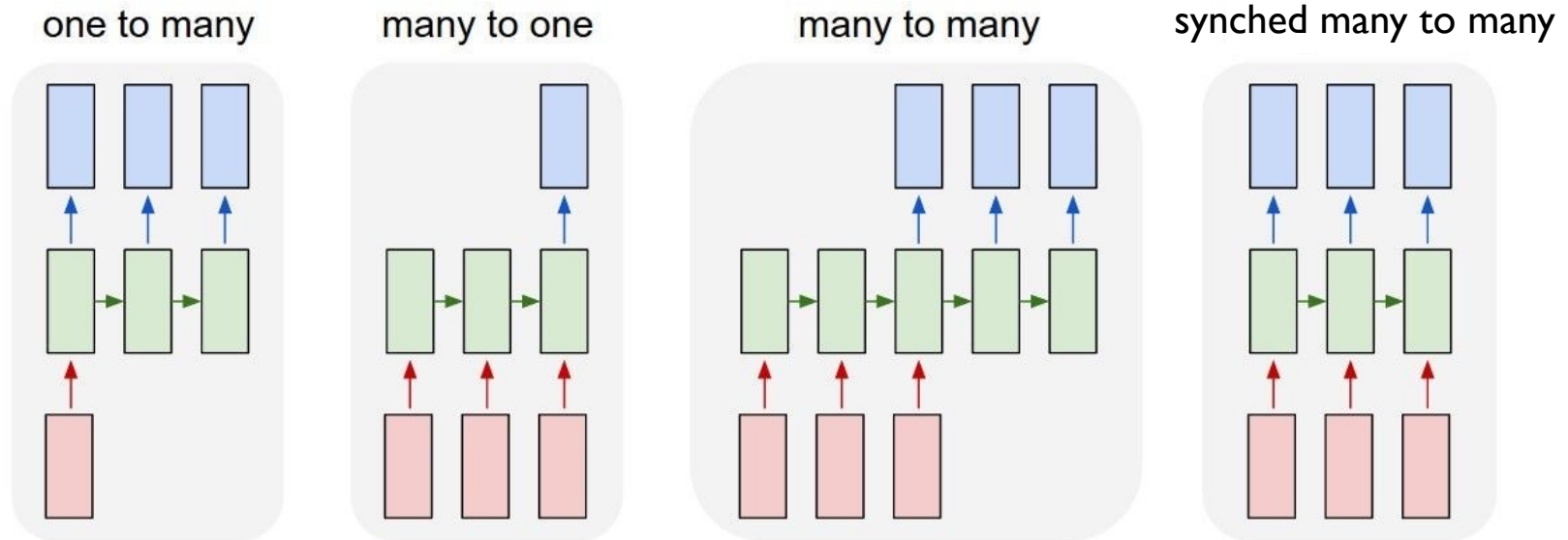
Image Captioning  
Image → sequence of words

# Types of Processes



→ Sentiment Classification  
sentence → sentiment

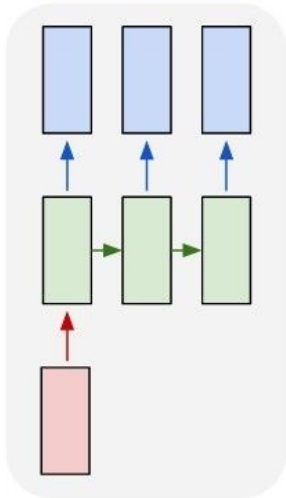
# Types of Processes



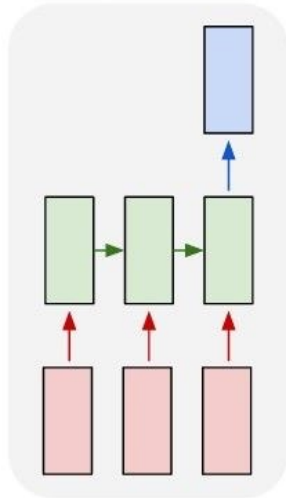
Machine Translation  
sentence → sentence

# Types of Processes

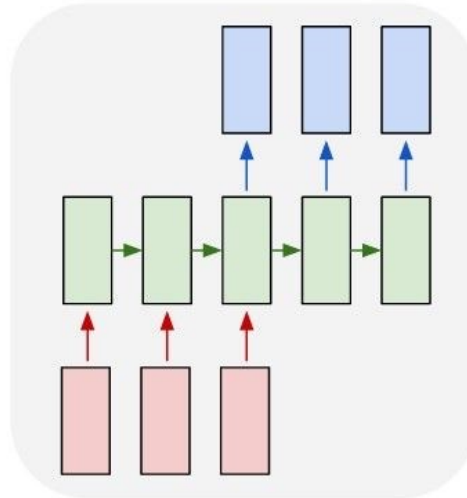
one to many



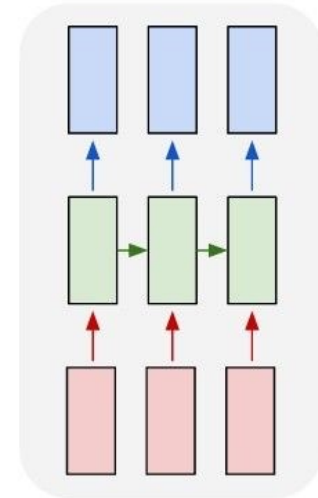
many to one



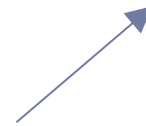
many to many



synched many to many



Stock Price Prediction,  
Prediction of next word

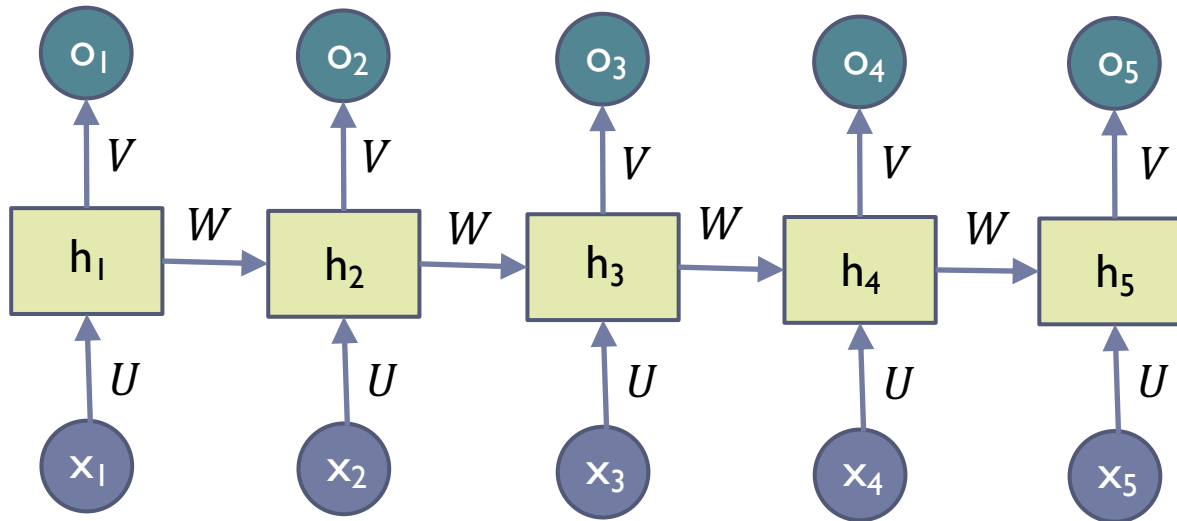




# Synced Many to Many

## ▶ Training

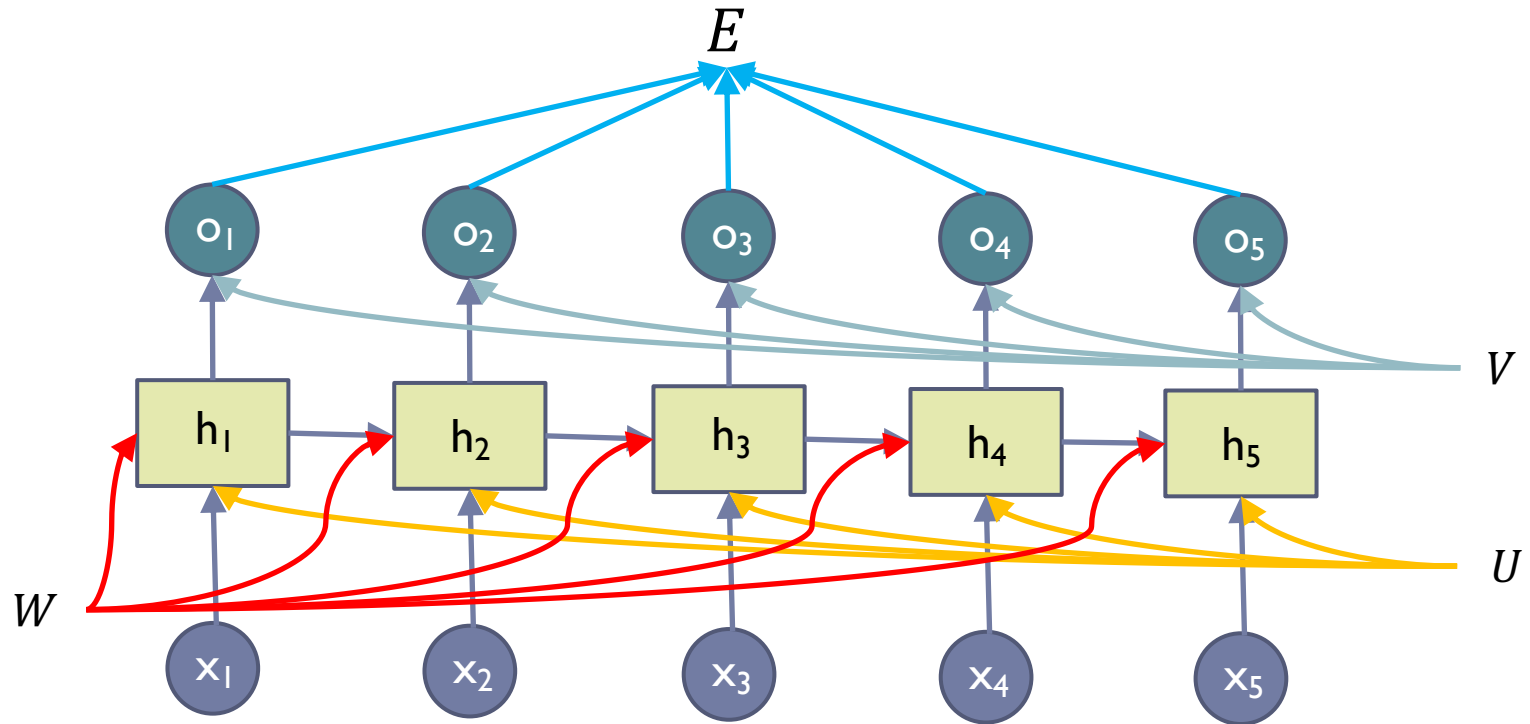
$$x_1 x_2 x_3 \cdots x_n \rightarrow y_1 y_2 y_3 \cdots y_n$$



$$E = \sum_{i=1}^n (y_i - o_i)^2$$

# Synced Many to Many

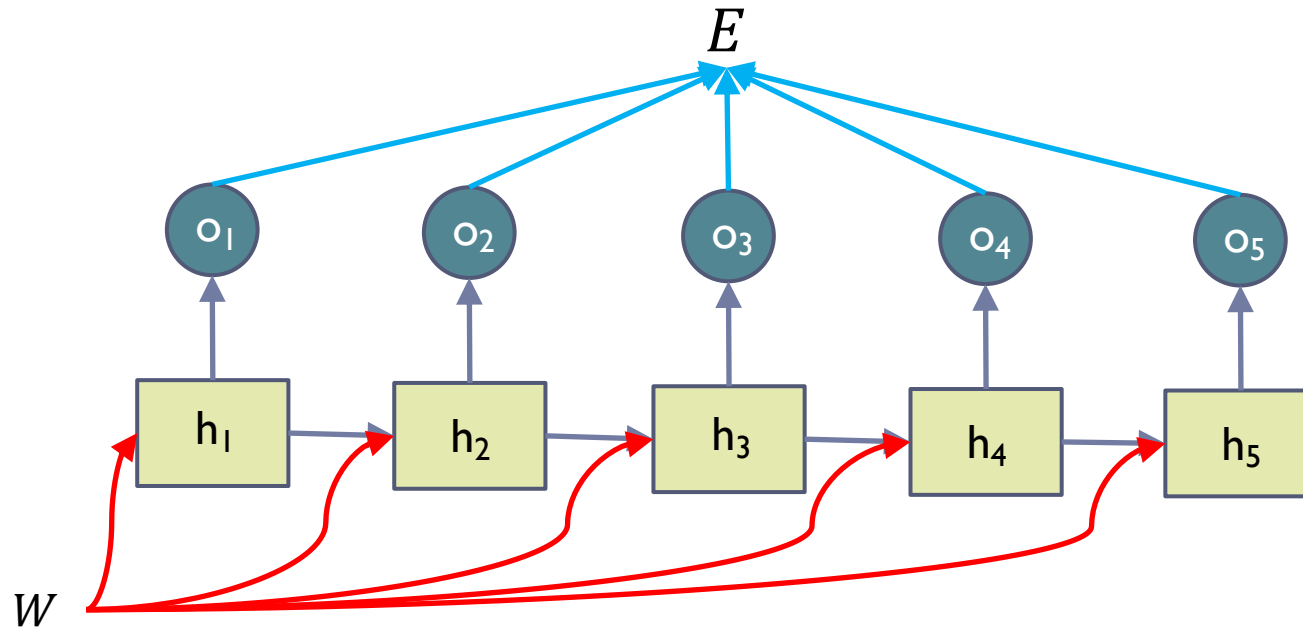
## ► Training



$$\frac{\partial E}{\partial w} = ?$$

# Synced Many to Many

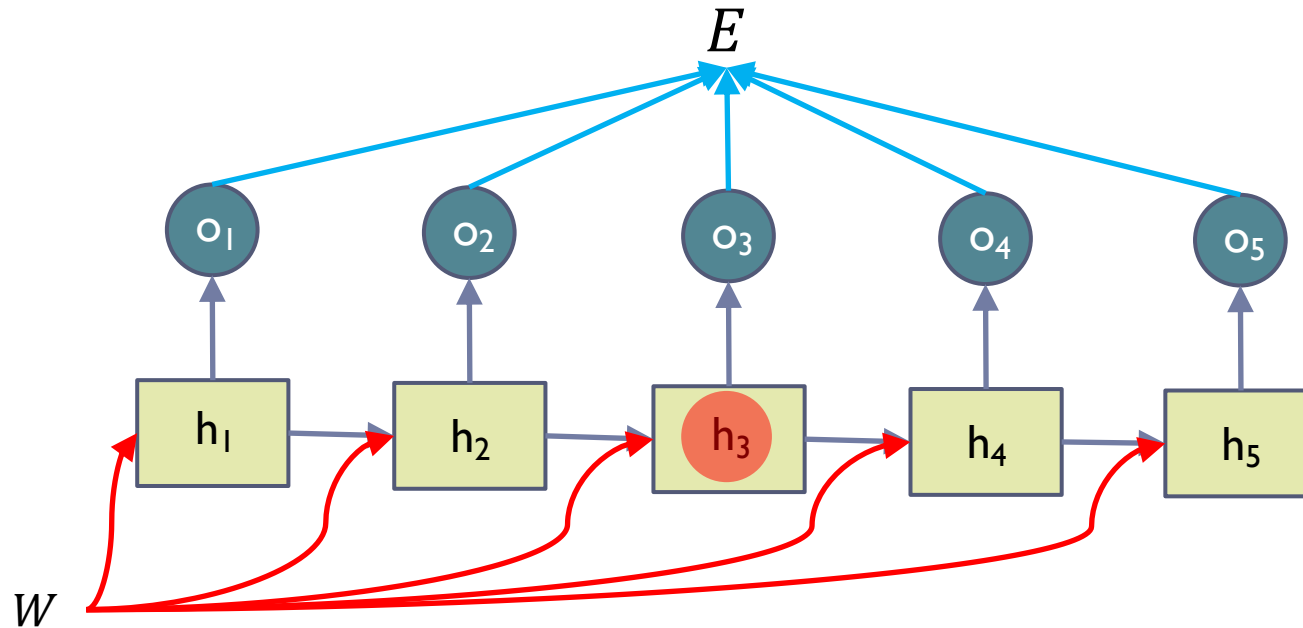
## ► Training



$$\frac{\partial E}{\partial w} = \sum_{i=1}^n \frac{\partial E}{\partial h_i} \frac{\partial h_i}{\partial w}$$

# Synced Many to Many

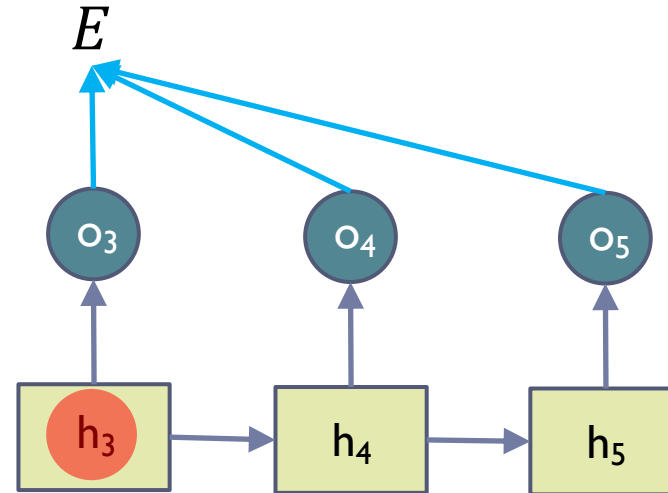
## ▶ Training



$$\frac{\partial E}{\partial w} = \sum_{i=1}^n \frac{\partial E}{\partial h_i} \frac{\partial h_i}{\partial w}$$

# Synced Many to Many

## ▶ Training



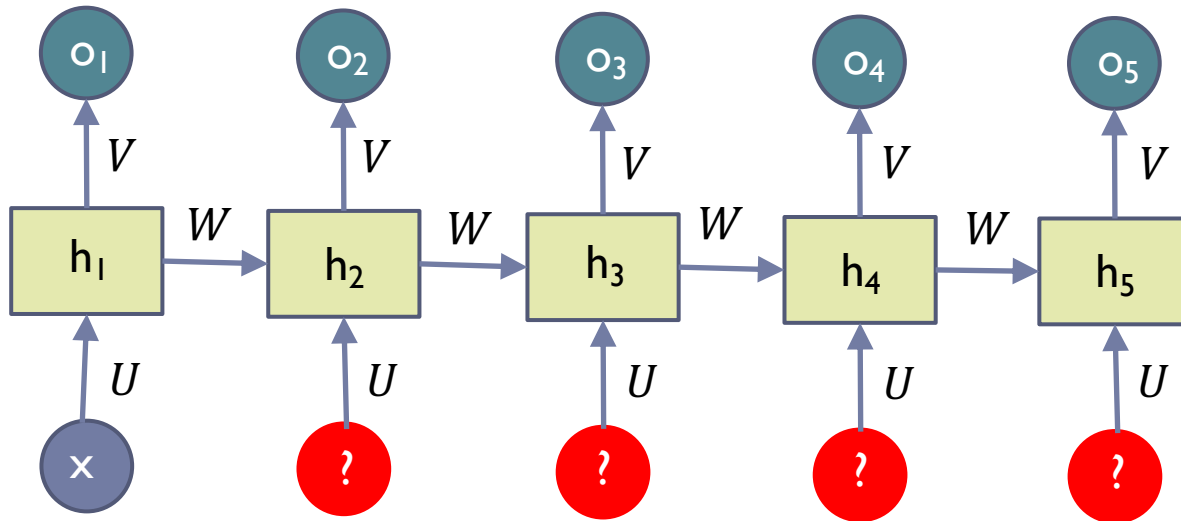
$$\frac{\partial E}{\partial h_i} = \frac{\partial E}{\partial o_i} \frac{\partial o_i}{\partial h_i} + \frac{\partial E}{\partial h_{i+1}} \frac{\partial h_{i+1}}{\partial h_i}$$

$$\frac{\partial E}{\partial w} = \sum_{i=1}^n \frac{\partial E}{\partial h_i} \frac{\partial h_i}{\partial w}$$

# One to Many

## ▶ Training

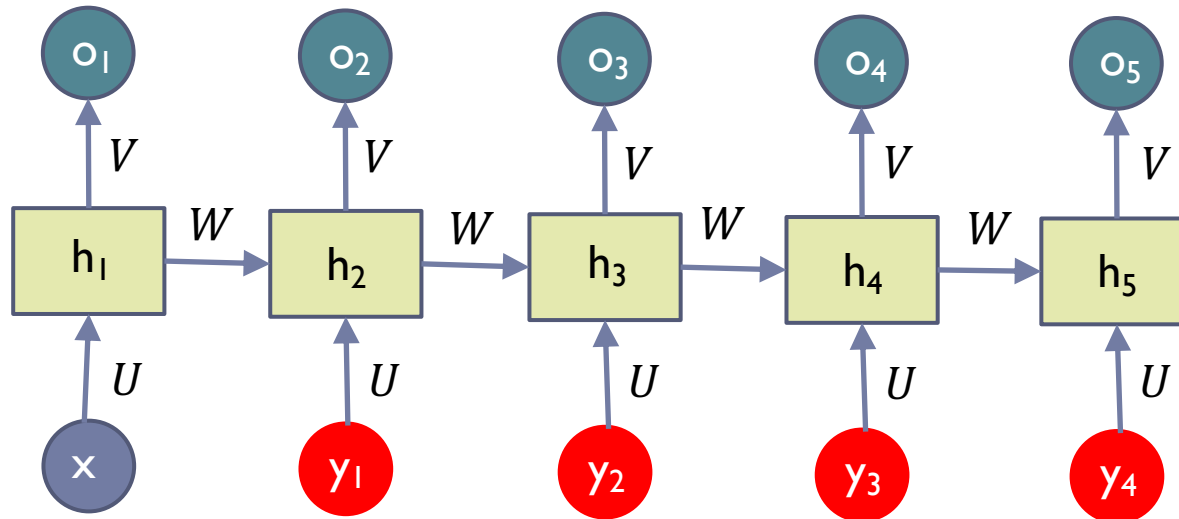
$$x \rightarrow y_1 y_2 y_3 \cdots y_n$$



# One to Many

## ▶ Training

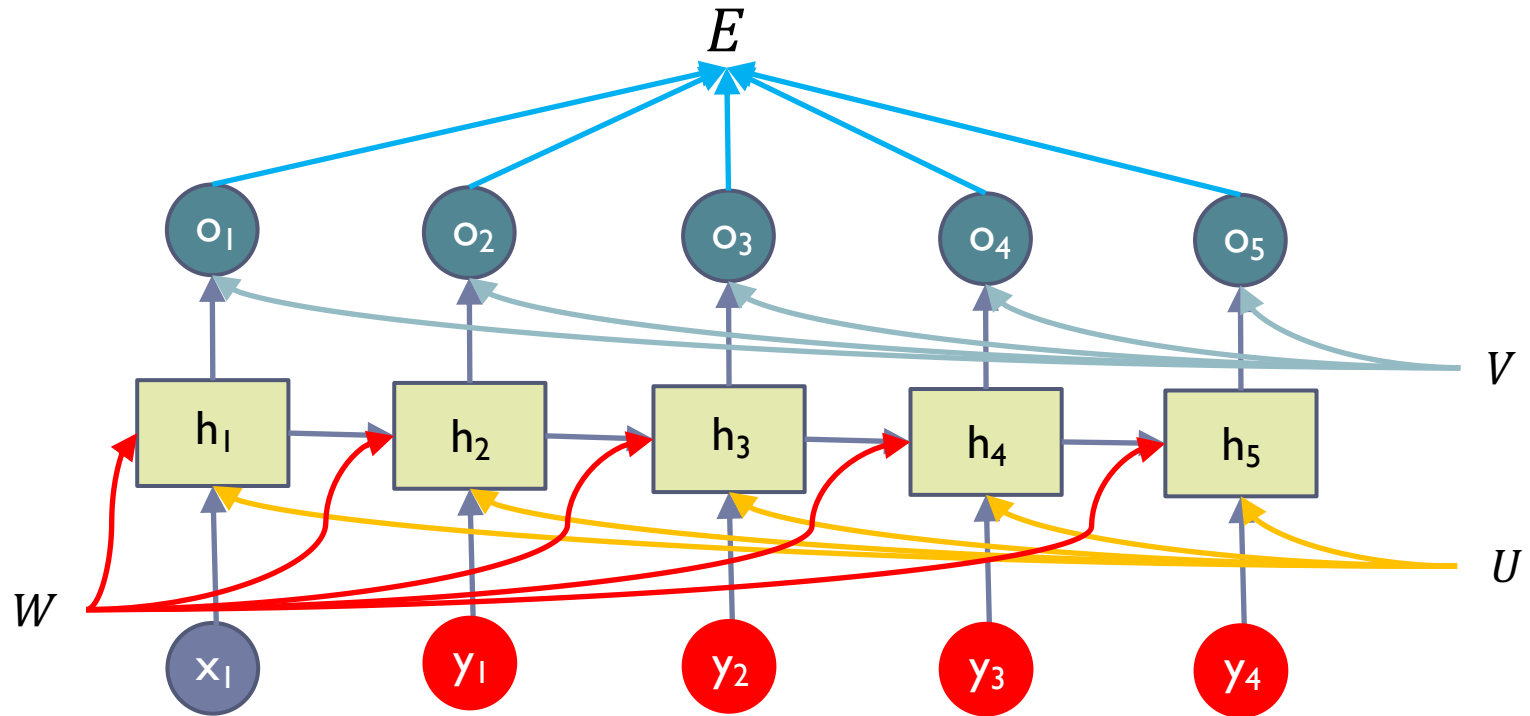
$$x \rightarrow y_1 y_2 y_3 \cdots y_n$$



$$E = \sum_{i=1}^n (y_i - o_i)^2$$

# One to Many

## ► Training



$$\frac{\partial E}{\partial w} = \sum_{i=1}^n \frac{\partial E}{\partial h_i} \frac{\partial h_i}{\partial w}$$

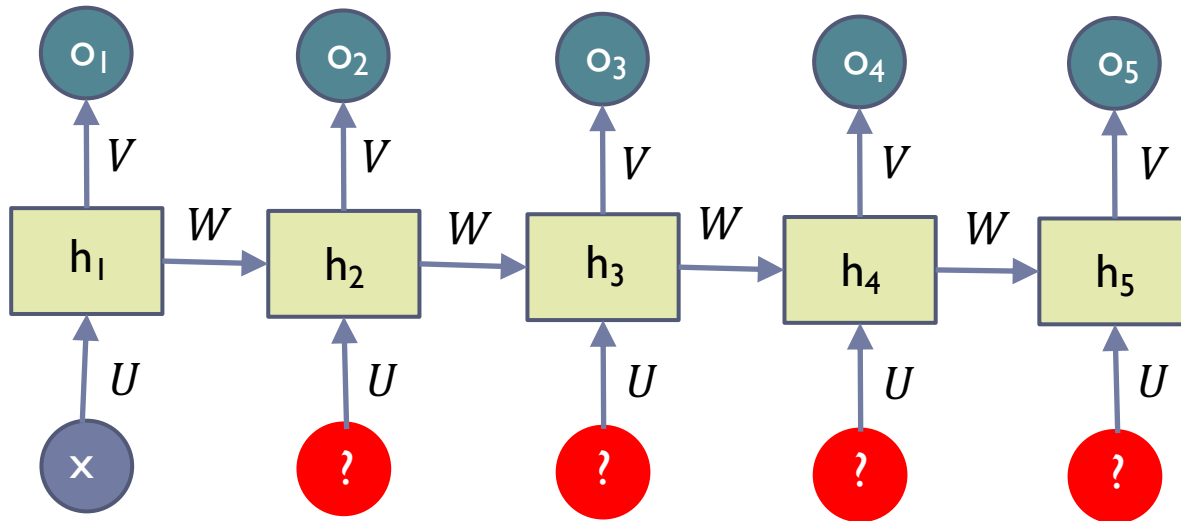
$$\frac{\partial E}{\partial h_i} = \frac{\partial E}{\partial o_i} \frac{\partial o_i}{\partial h_i} + \frac{\partial E}{\partial h_{i+1}} \frac{\partial h_{i+1}}{\partial h_i}$$



# One to Many

## ▶ Testing

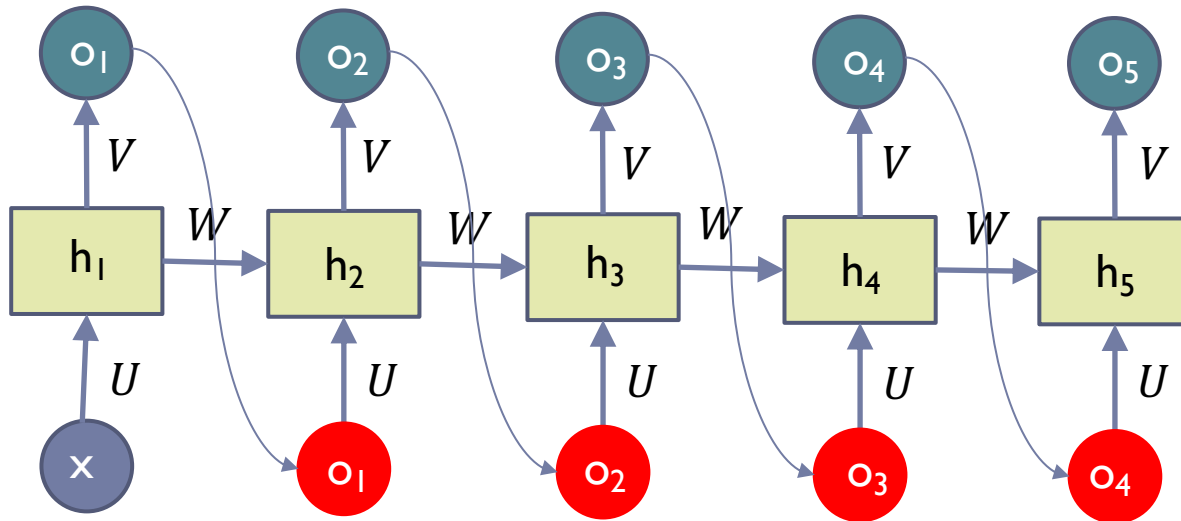
$x \rightarrow ??????$



# One to Many

## ▶ Testing

$x \rightarrow ???????$

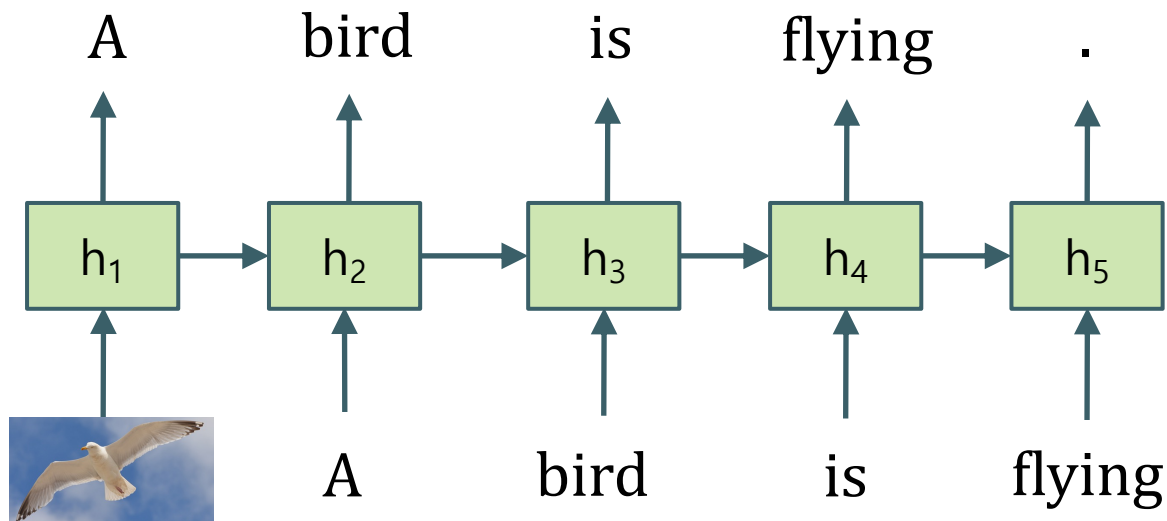


# One to Many

## ▶ How to Handle Images and Texts

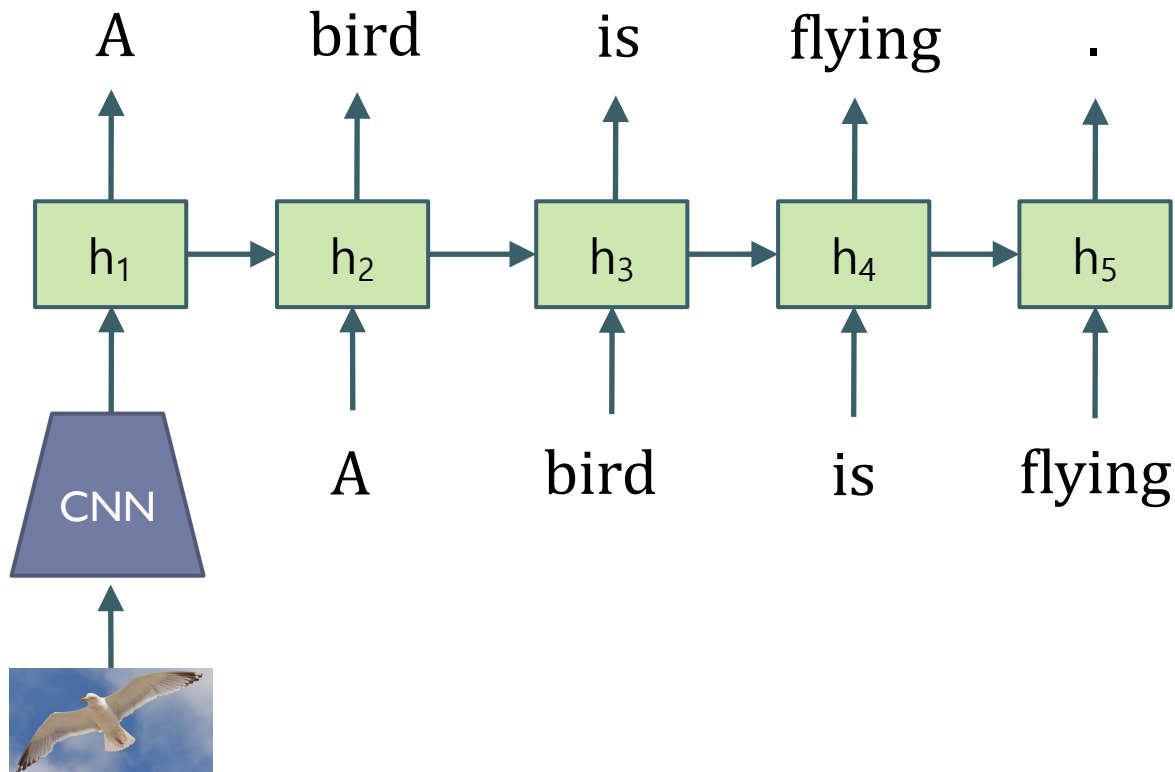


→ A bird is flying.



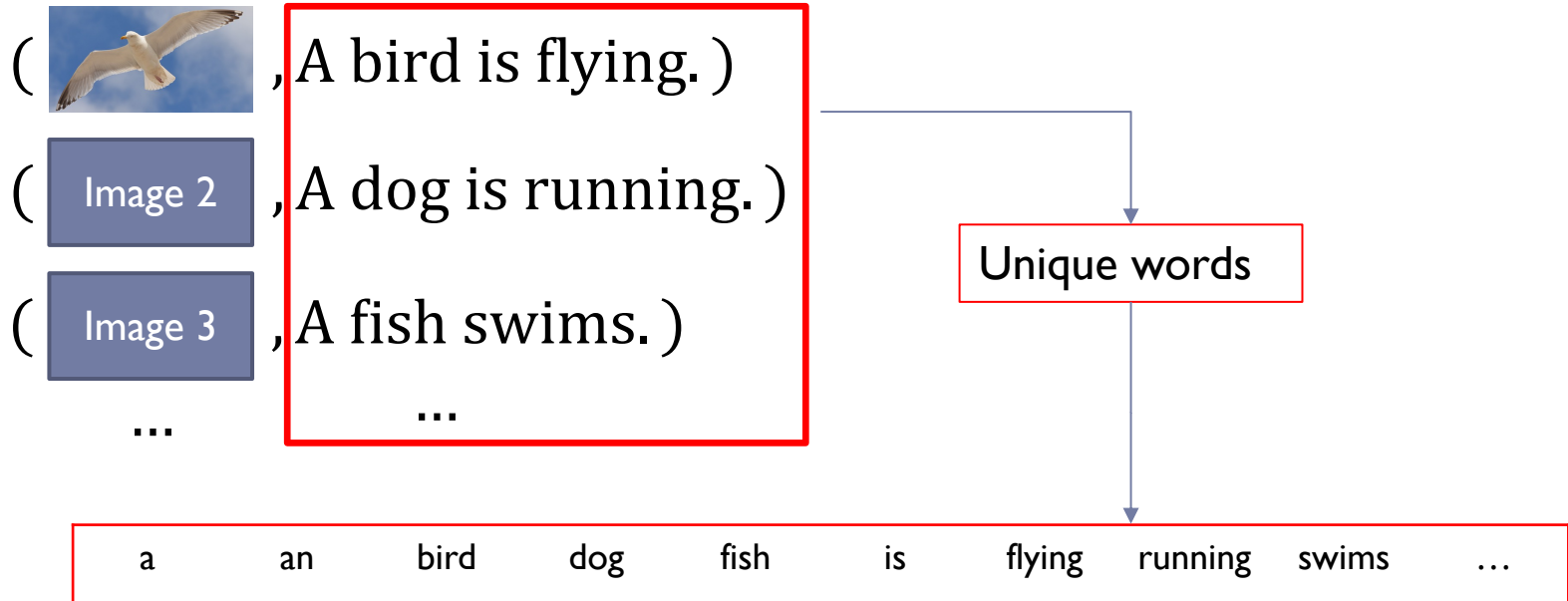
# One to Many

## ► How to Handle Images and Texts



# One to Many

## ▶ How to Handle Images and Texts



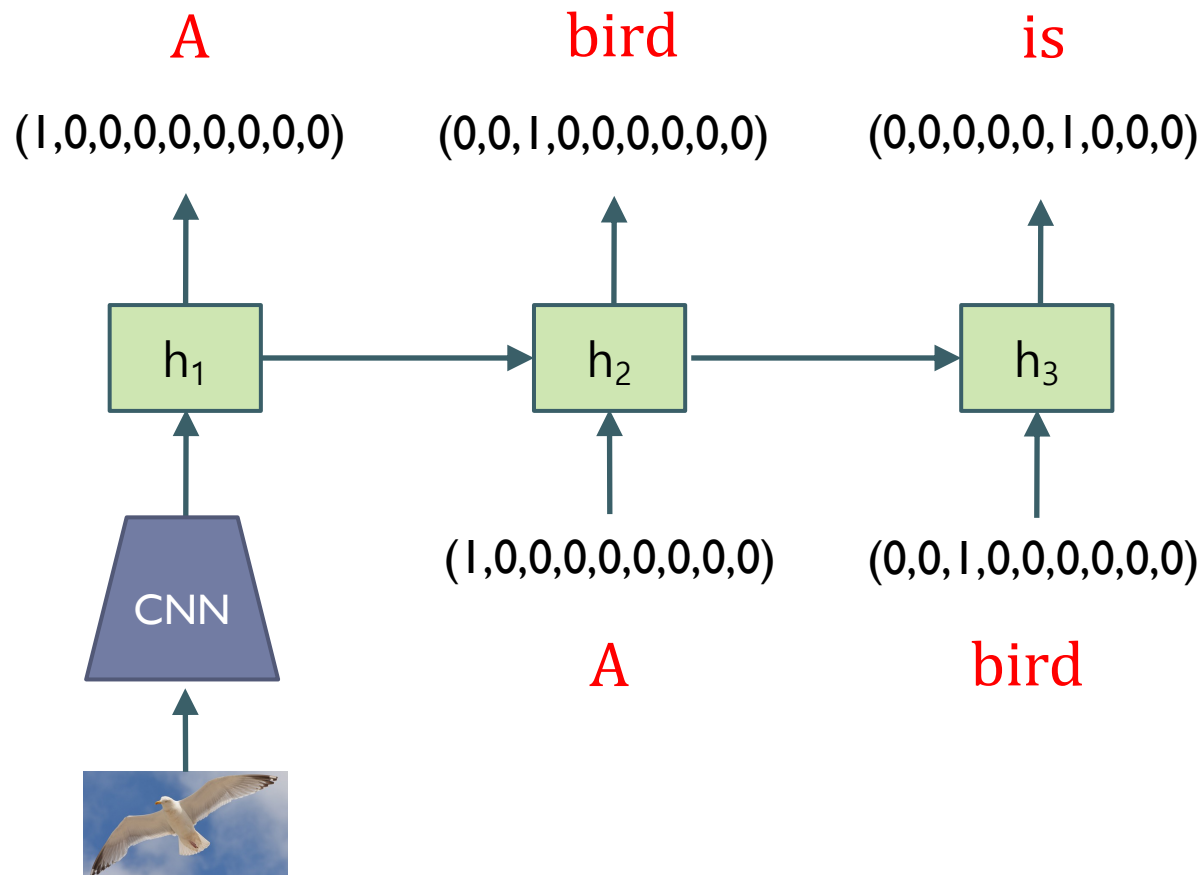
A   ( 1 , 0 , 0 , 0 , 0 , 0 , 0 , 0 , 0 , 0 , ... )

bird   ( 0 , 0 , 1 , 0 , 0 , 0 , 0 , 0 , 0 , 0 , ... )

is   ( 0 , 0 , 0 , 0 , 0 , 1 , 0 , 0 , 0 , 0 , ... )

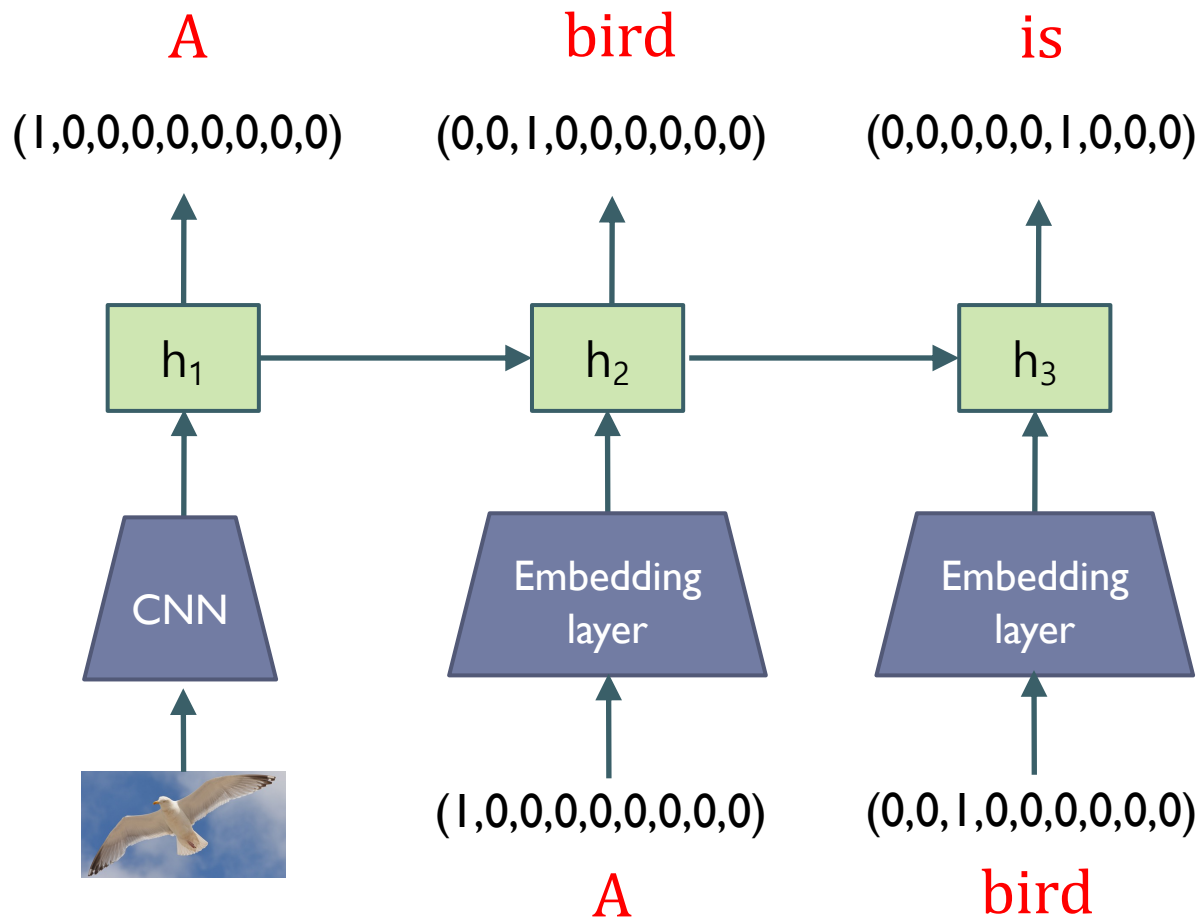
# One to Many

## ► How to Handle Images and Texts



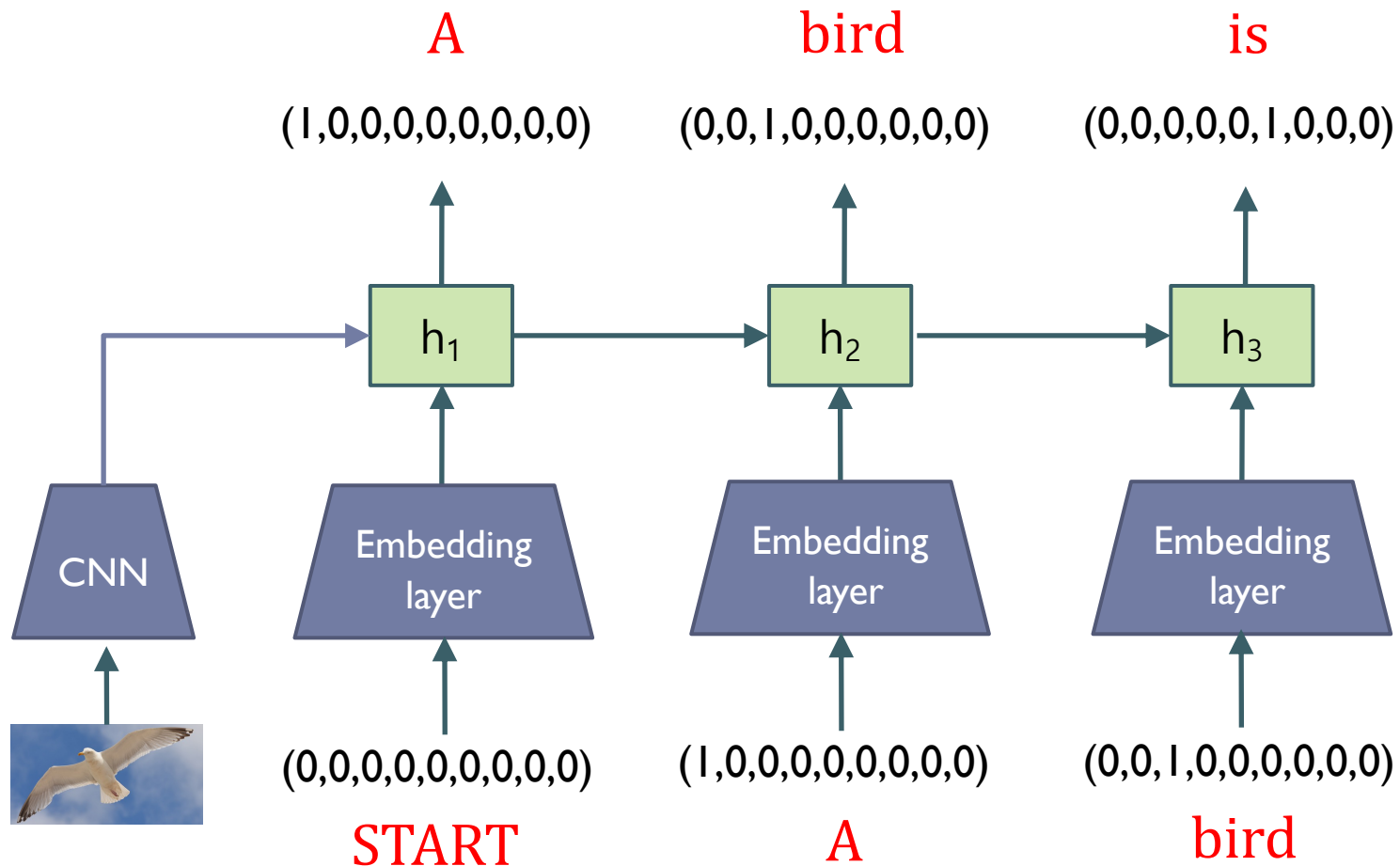
# One to Many

## ► How to Handle Images and Texts



# One to Many

## ► How to Handle Images and Texts

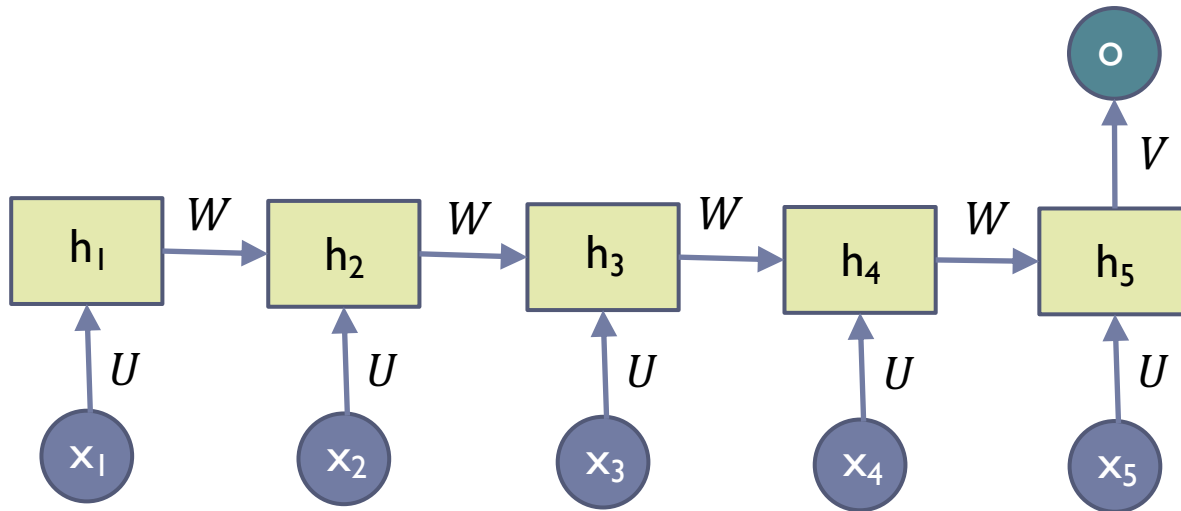




# Many to One

## ▶ Training

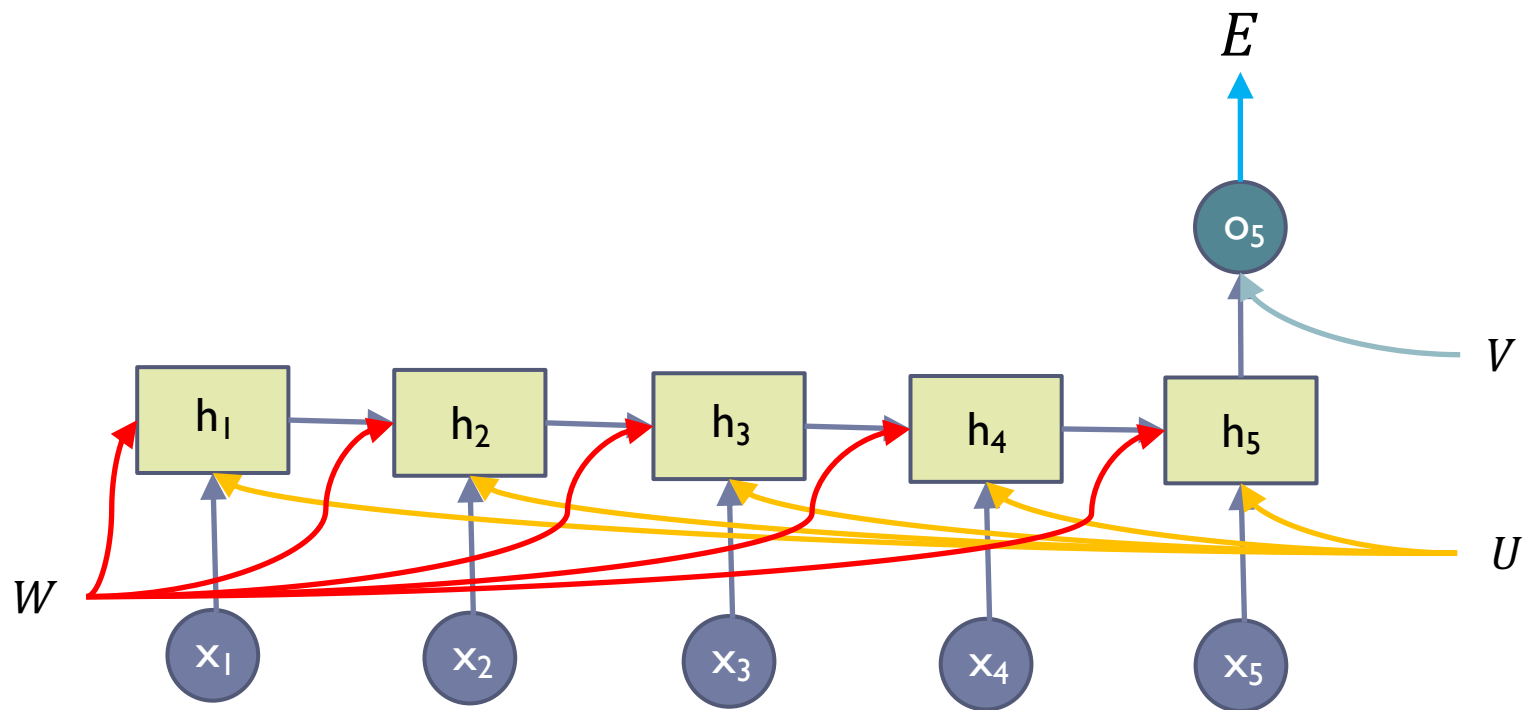
$$x_1 x_2 x_3 \cdots x_n \rightarrow y$$



$$E = (y - o)^2$$

# Many to One

## ▶ Training



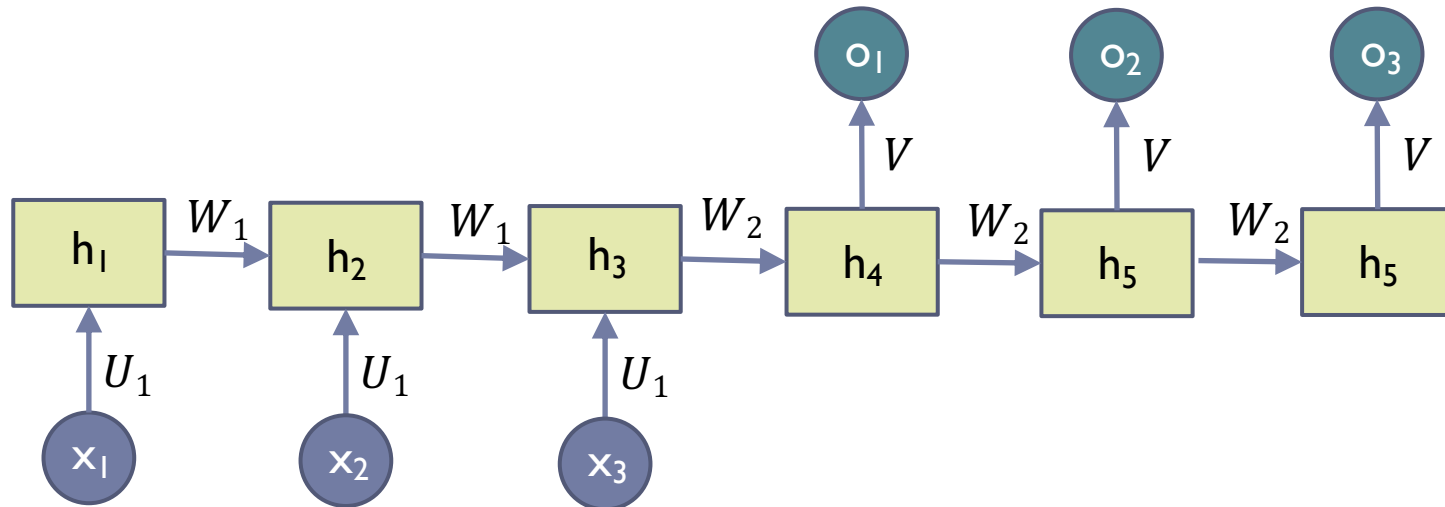
$$\frac{\partial E}{\partial w} = \sum_{i=1}^n \frac{\partial E}{\partial h_i} \frac{\partial h_i}{\partial w}$$

$$\frac{\partial E}{\partial h_i} = \frac{\partial E}{\partial h_{i+1}} \frac{\partial h_{i+1}}{\partial h_i}$$

# Many to Many

## ▶ Training

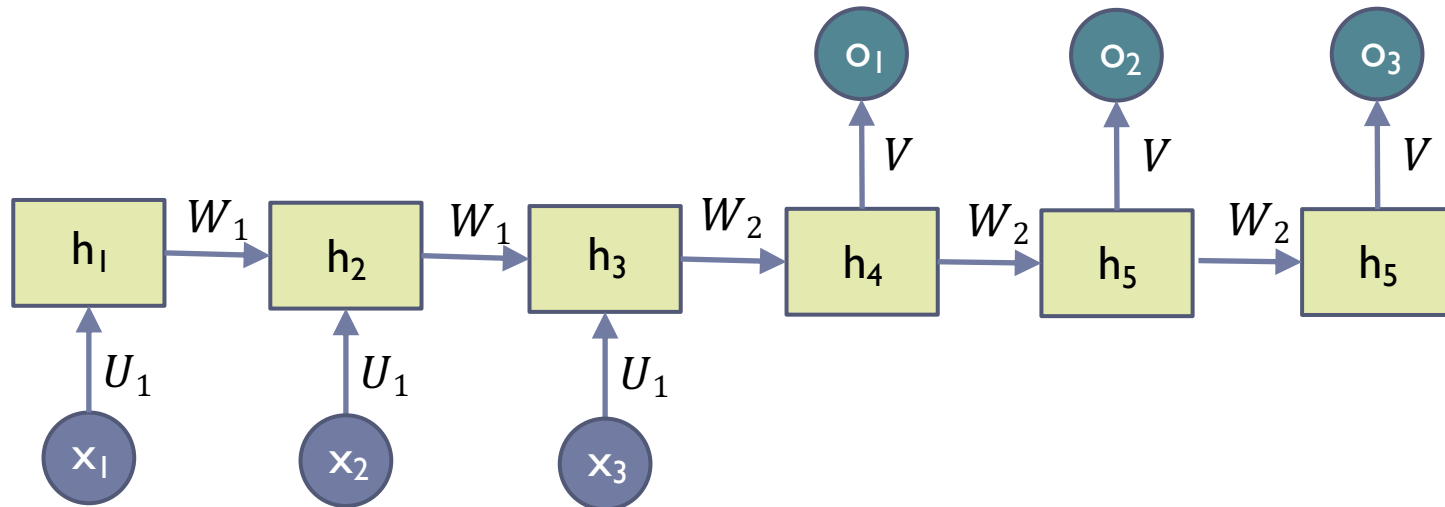
$$x_1 x_2 x_3 \cdots x_n \rightarrow y_1 y_2 y_3 \cdots y_n$$



# Many to Many

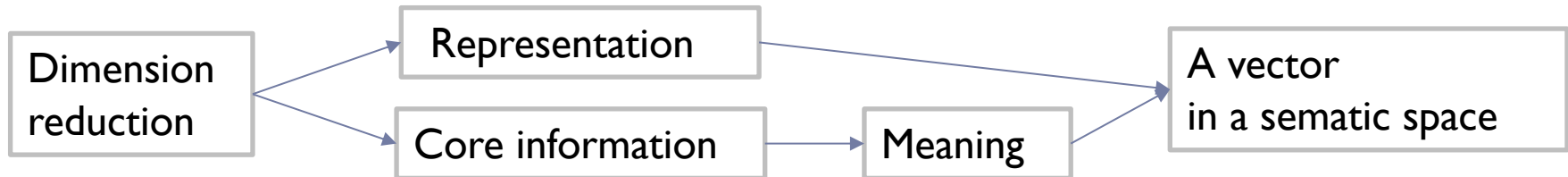
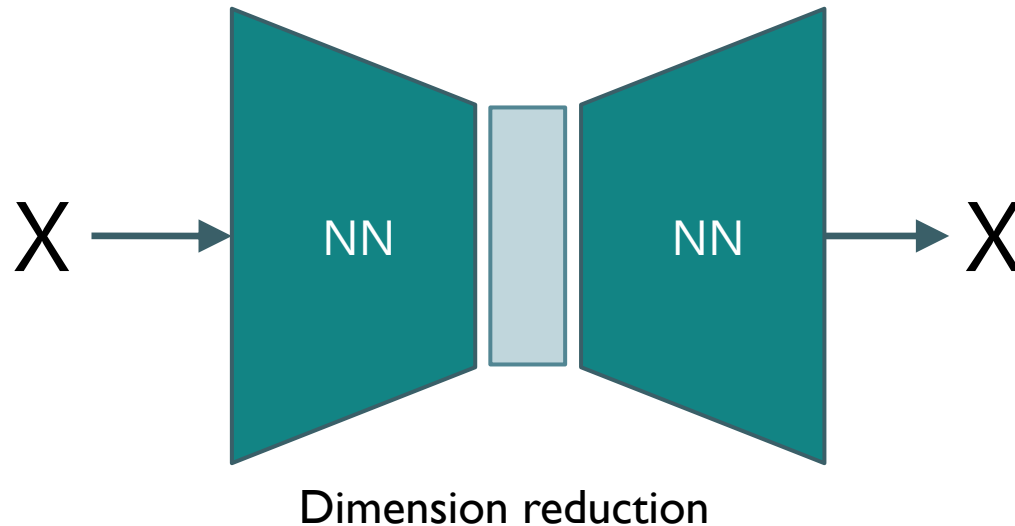
## ▶ Training

$$x_1 x_2 x_3 \cdots x_n \rightarrow y_1 y_2 y_3 \cdots y_n$$



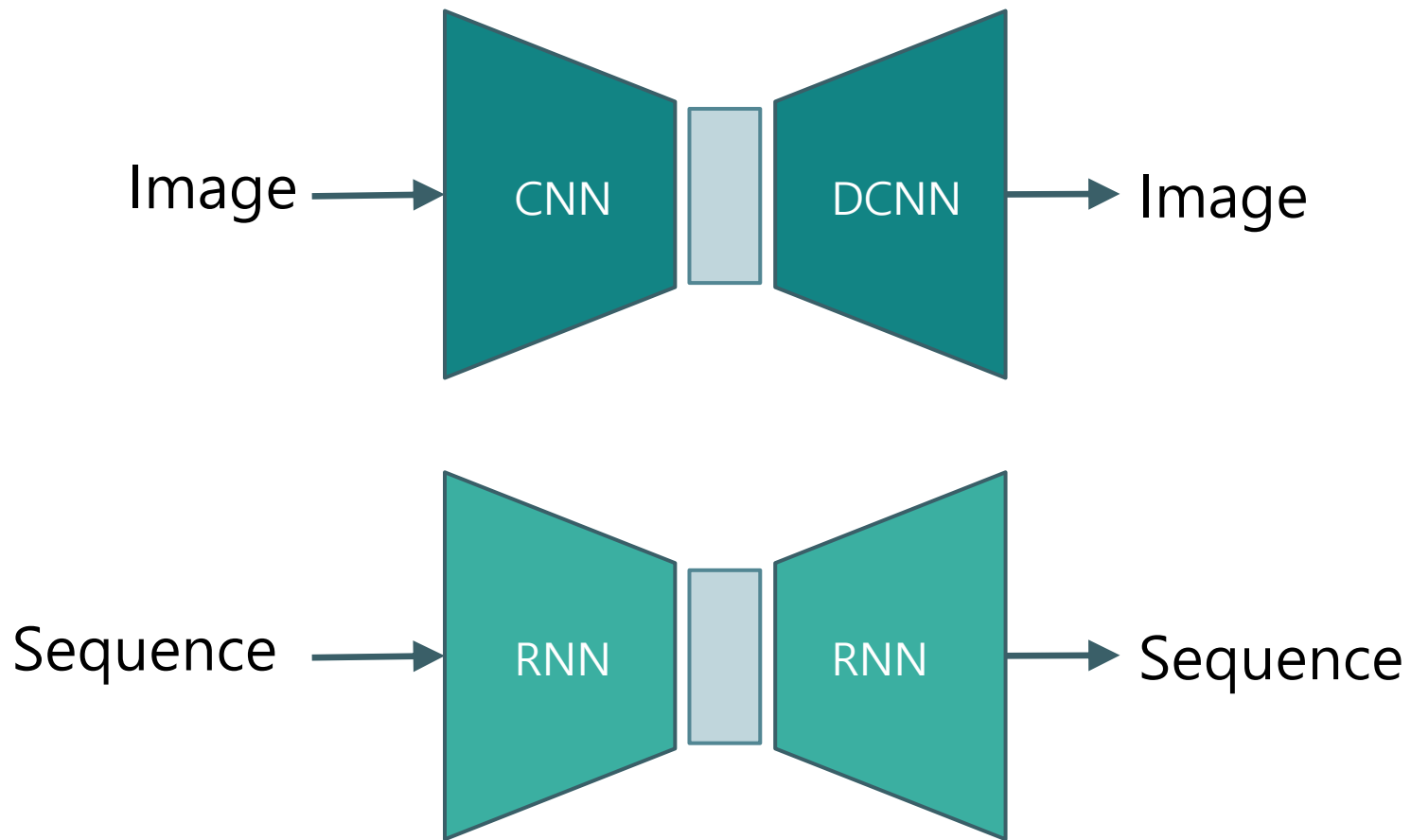
# Many to Many

## ▶ AutoEncoder



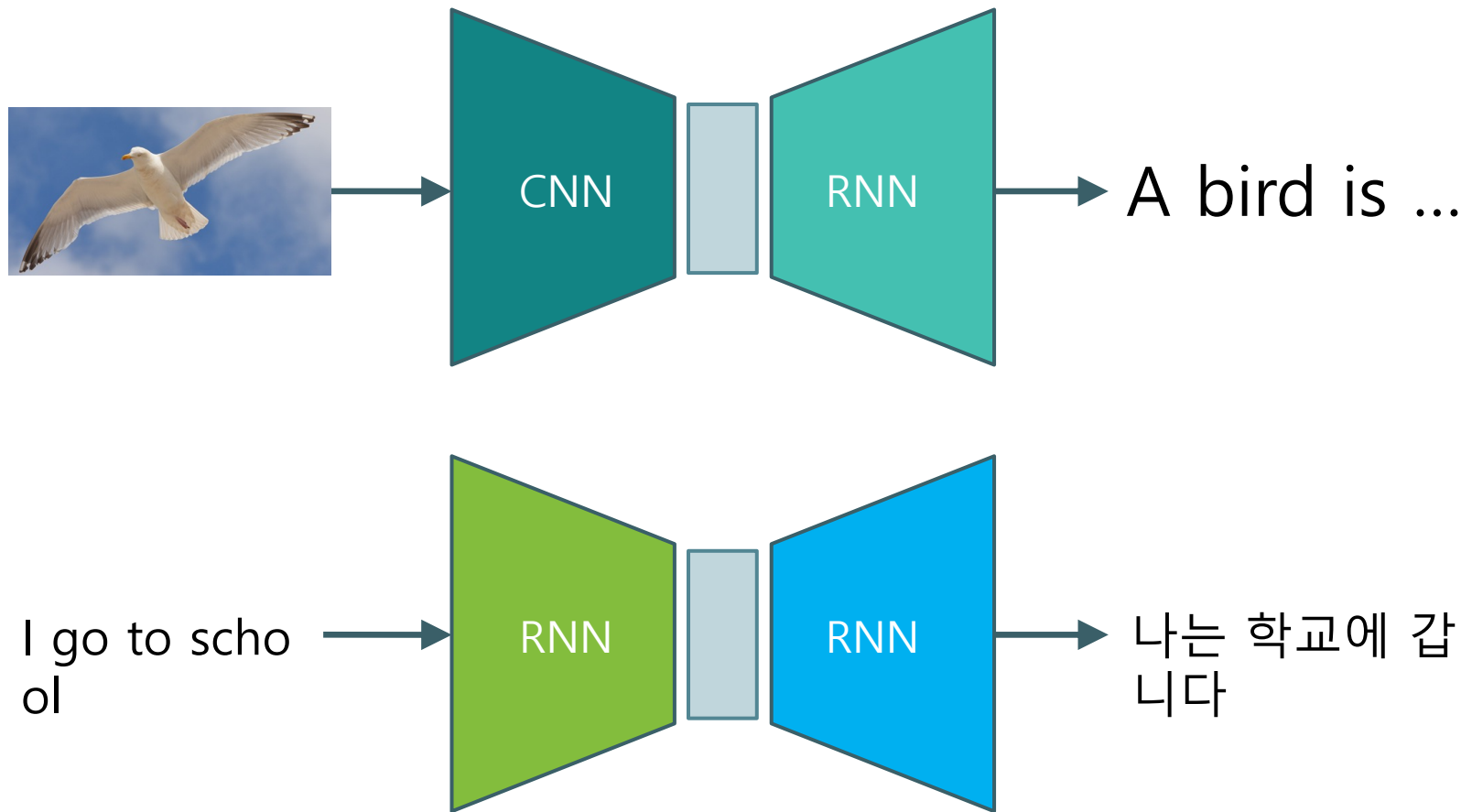
# Many to Many

## ▶ AutoEncoder



# Many to Many

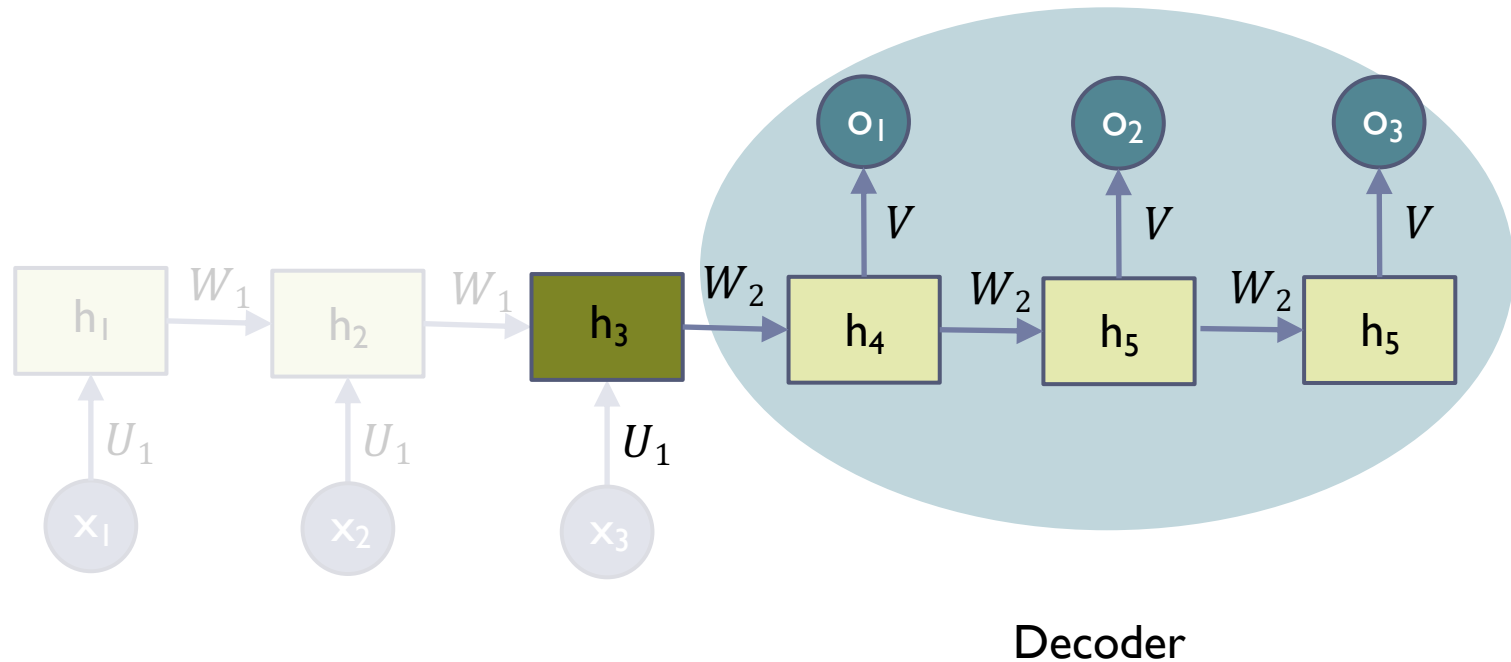
## ▶ AutoEncoder



# Many to Many

## ▶ Training

$$x_1 x_2 x_3 \cdots x_n \rightarrow y_1 y_2 y_3 \cdots y_n$$



Encoder

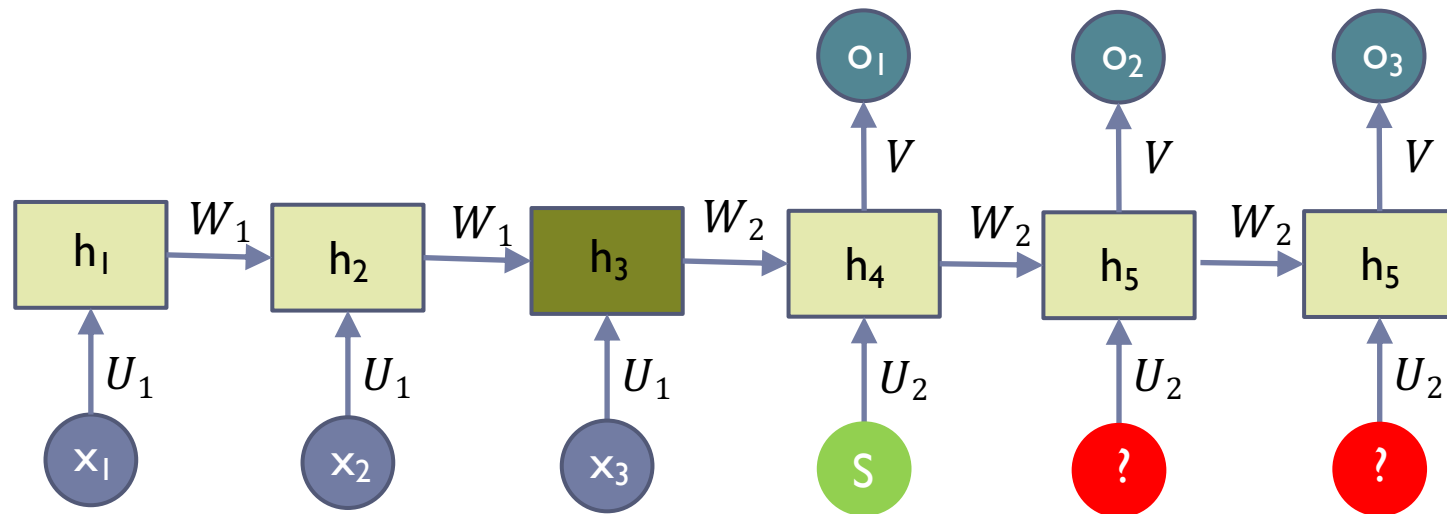
Decoder



# Many to Many

## ▶ Training

$$x_1 x_2 x_3 \cdots x_n \rightarrow y_1 y_2 y_3 \cdots y_n$$

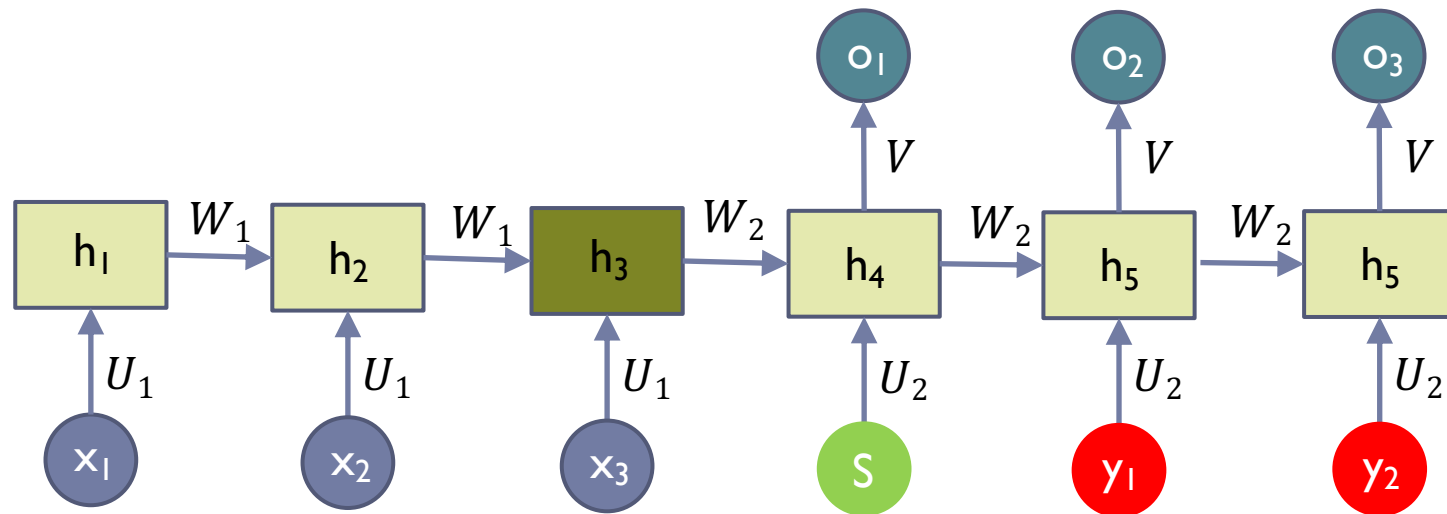


$$E = \sum_{i=1}^n (y_i - o_i)^2$$

# Many to Many

## ► Training

$$x_1 x_2 x_3 \cdots x_n \rightarrow y_1 y_2 y_3 \cdots y_n$$



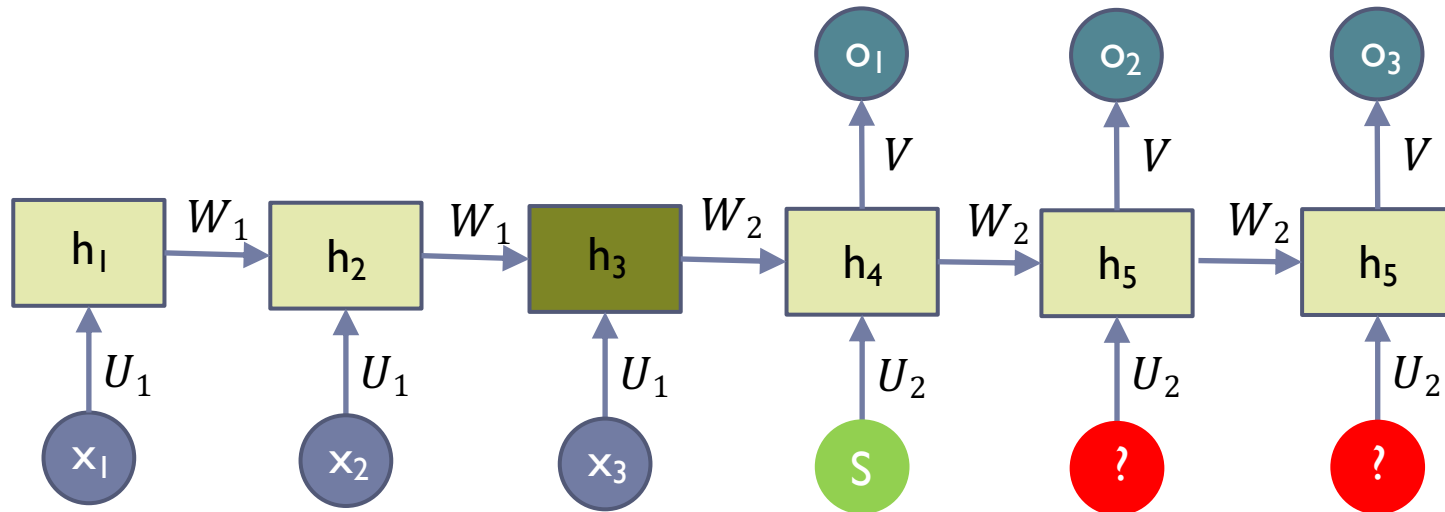
$$\frac{\partial E}{\partial w} = ??$$

Combination of [Many to One] and [One to Many]

# Many to Many

## ▶ Testing

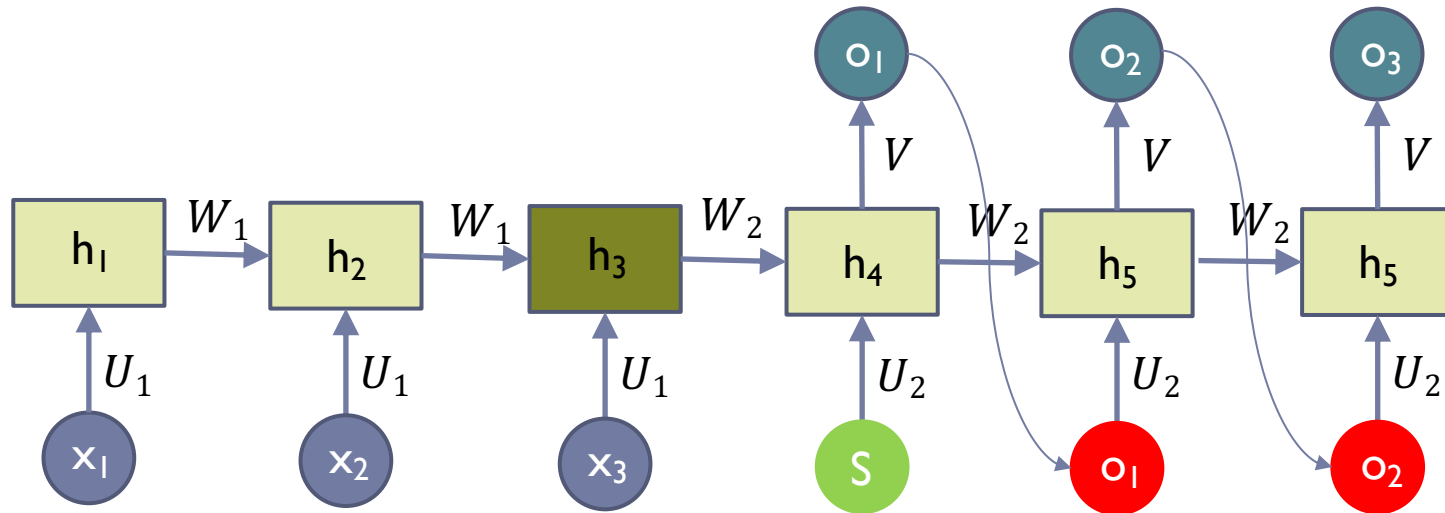
$$x_1 x_2 x_3 \cdots x_n \rightarrow ? ? ? ? ?$$



# Many to Many

## ▶ Testing

$$x_1 x_2 x_3 \cdots x_n \rightarrow ? ? ? ? ?$$

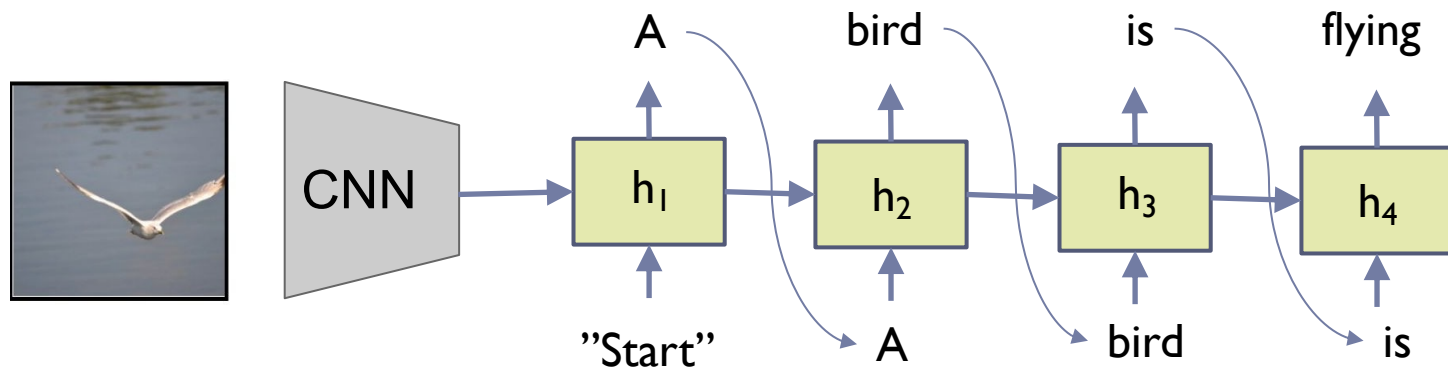


# Example

## ▶ One to Many

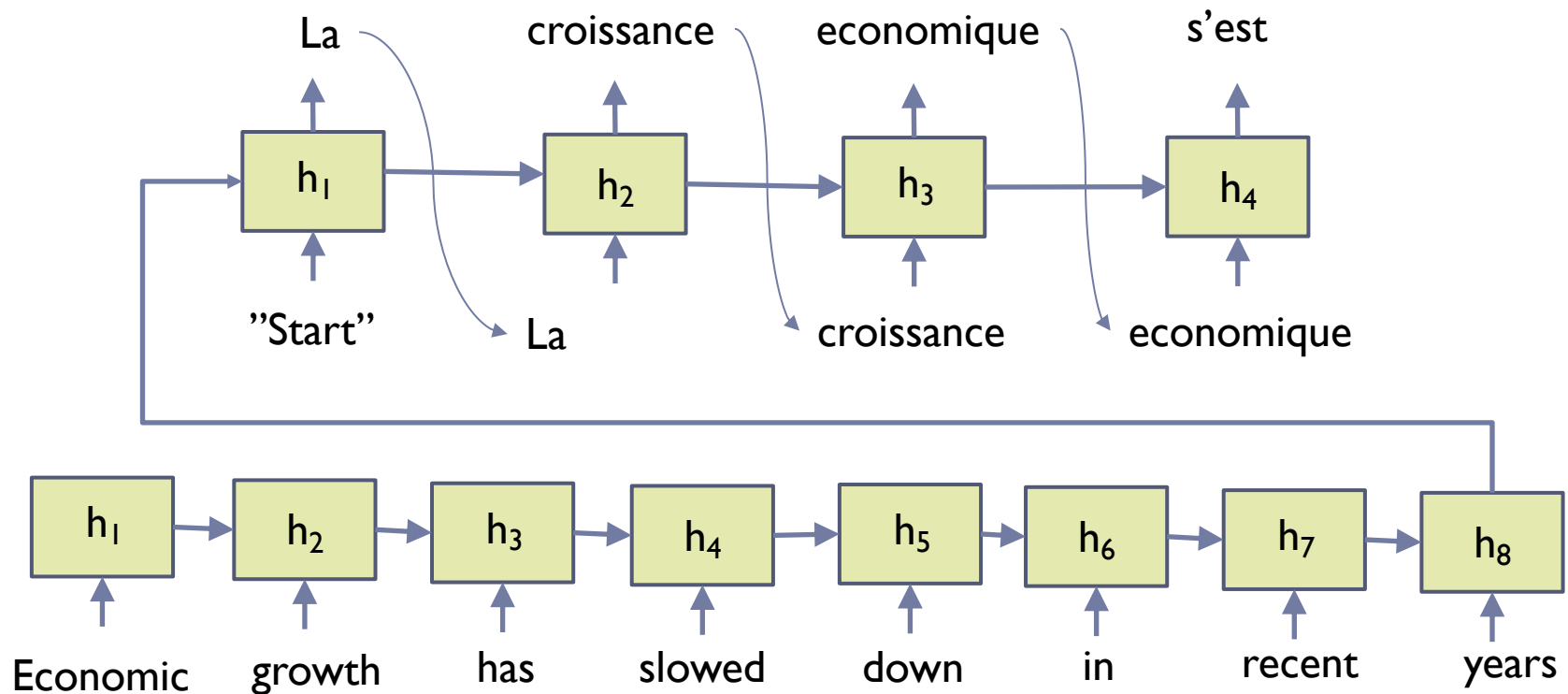
### ▶ Caption Generation

- ▶ Image is represented by a CNN
- ▶ Word Embedding at the input layer
- ▶ Softmax at the output layer



# Example

- ▶ **Many to Many**
  - ▶ Word Embedding



# Question and Answer