

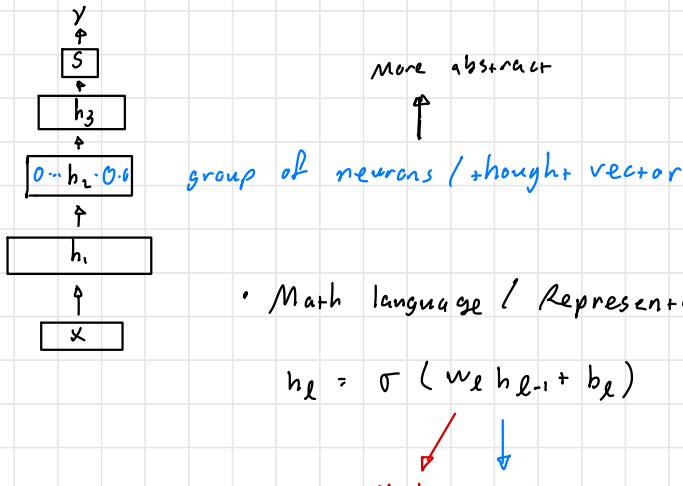
# Lecture 17



- Deep Learning

- Slogan 1 : approximator
- Slogan 2 : learned computation / algorithm
- Slogan 3 : team of vectors

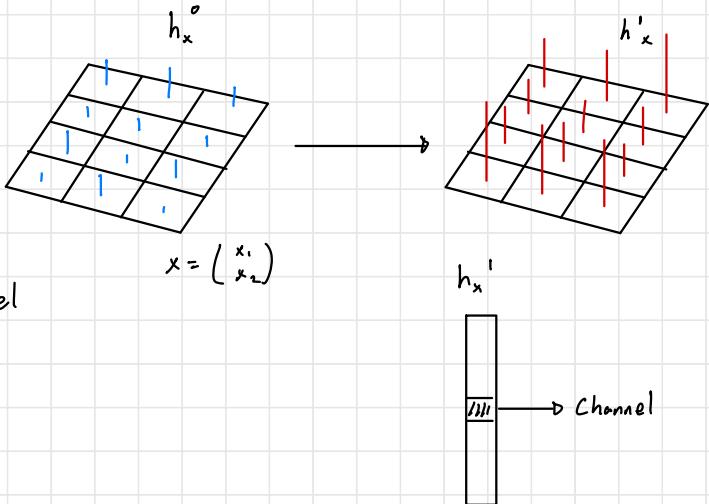
- MLP :



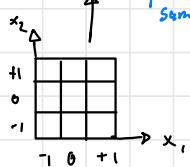
• Embedding / Distributed Representation:

- Convolutional Neural Network (CNN / ConvNet)

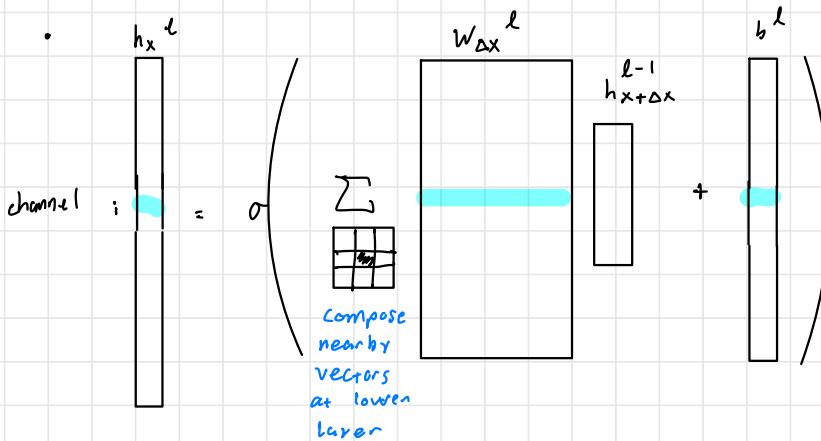
$x$ : image

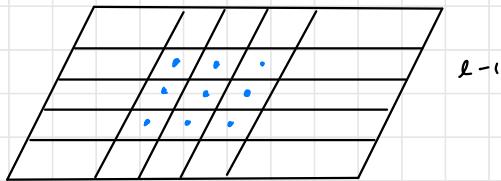
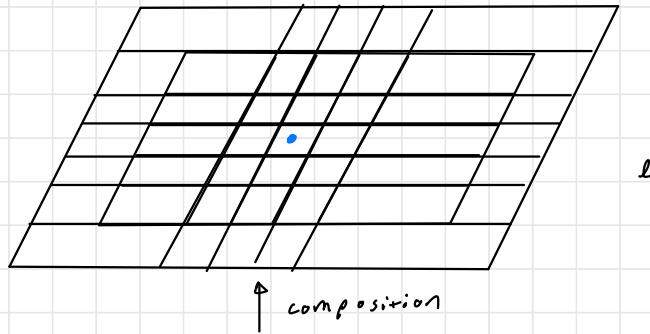


$$h_x^l = \sigma \left( \sum_{\Delta x \in \Delta} W_{\Delta x}^l h_{x+\Delta x}^{l-1} + b^l \right)$$

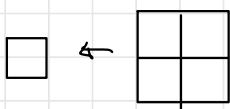


- $W_{\Delta x}^l$  : kernel  $3 \times 3$





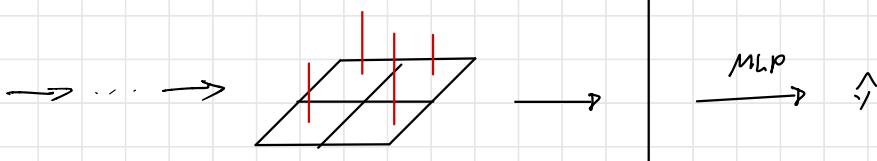
- Subsampling :  $h_x^{l, \text{sub}} = h_{2x}^l$



- Can also do Averaging / Max Pooling

$h^l$  thought for whole image

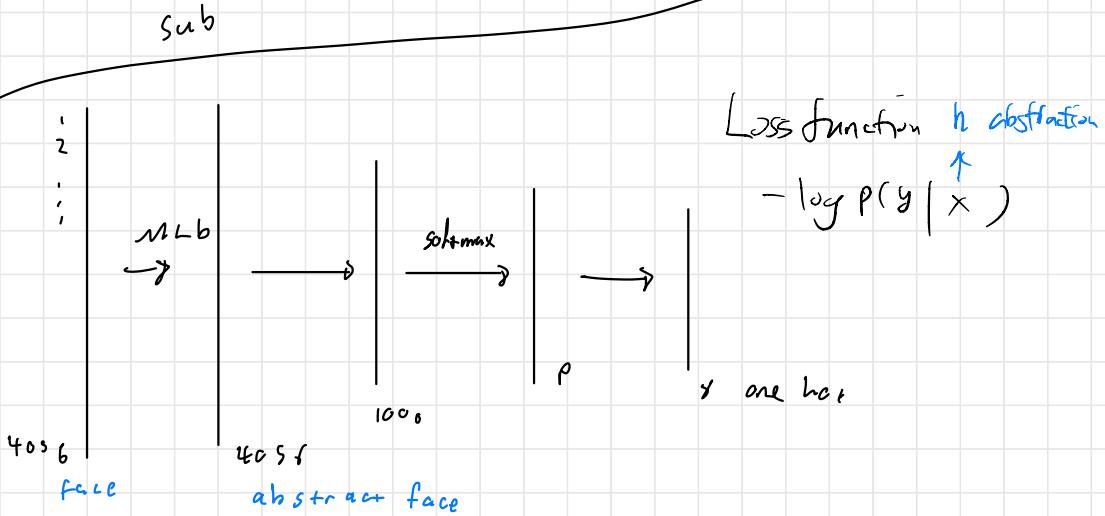
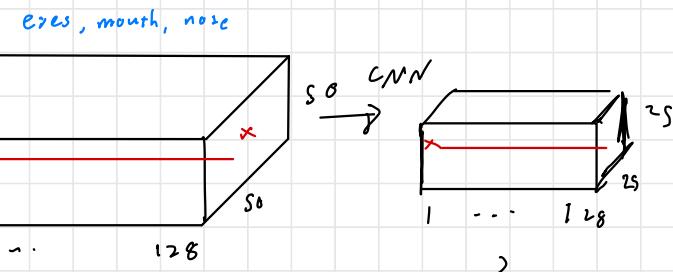
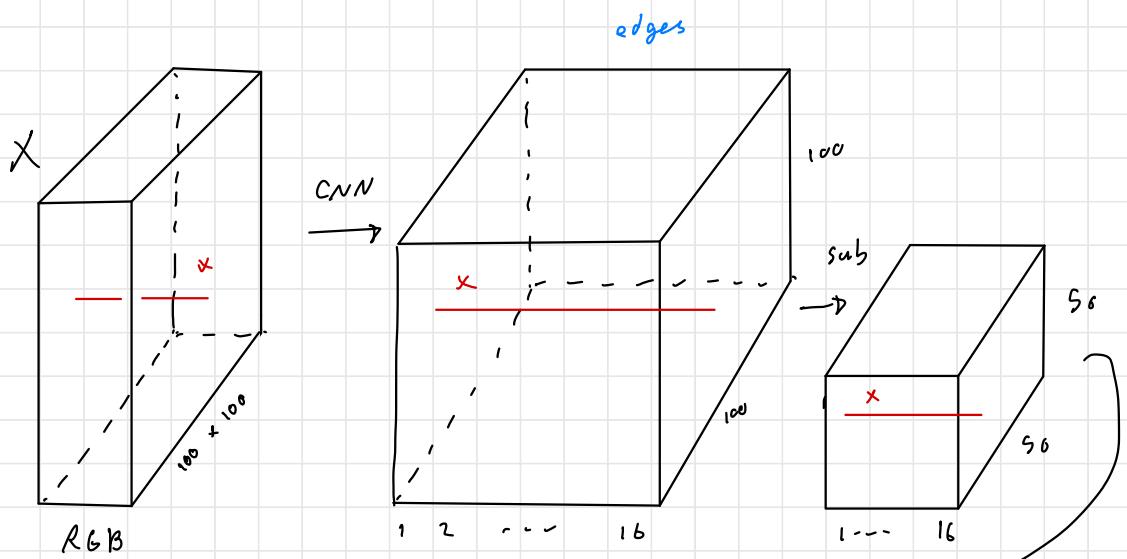
- After Subsampling :



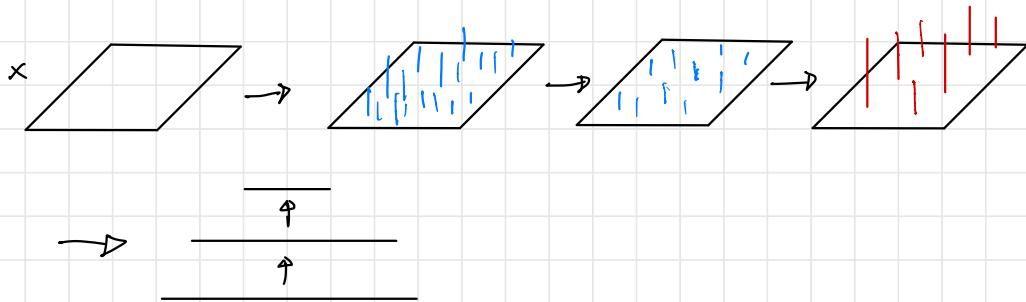
conv layers

fully connected layer,

$$h^l = \sigma \left( \sum_x w_x h_x^{l-1} + b^l \right)$$

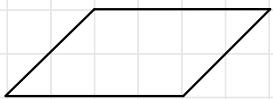
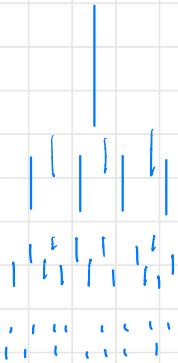


• Plot in terms of vectors:

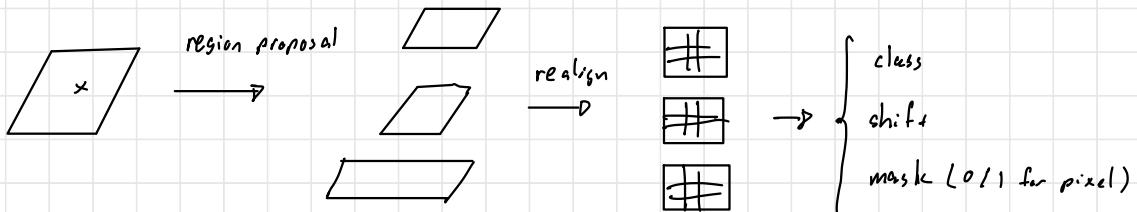


↑ compose & abstract

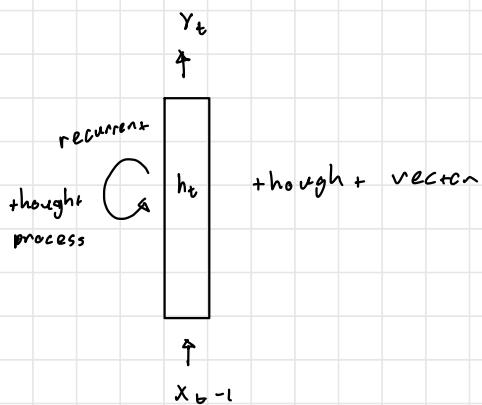
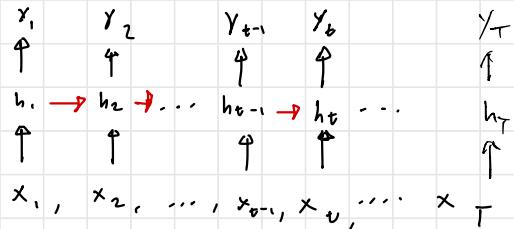
A team of vectors  
formation



• detection:



- Image : Spatial
- What about temporal?
- Recurrent Neural Network (RNN)



- Math Language :

$$h_t = \tanh \left( w_h h_{t-1} + w_x x_t + b \right)$$

Concatenation

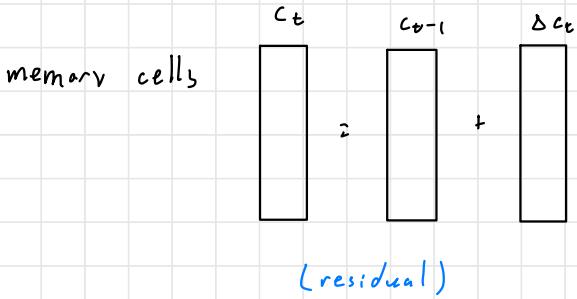
Diagram illustrating the computation of  $h_t$ :

- A vertical rectangle  $h_t$  is shown.
- An upward arrow labeled  $x_{t-1}$  points to a vertical rectangle  $h_{t-1}$ .
- A blue arrow labeled  $\downarrow$  points to the term  $w_h h_{t-1}$  in the equation.
- The equation shows  $h_t$  as the tanh of the sum of  $w_h h_{t-1}$ ,  $w_x x_t$ , and  $b$ .
- A small square  $x_t$  is shown below  $w_x$ .

$$\frac{e^x - e^{-x}}{e^x + e^{-x}}$$



- long term memory : in  $\Theta = (w, b, \text{ # layers})$
- short term memory : in  $h_t$
- long short-term memory : (LSTM)



$$\Delta c_t = \tanh \left( W \begin{pmatrix} h_{t-1} \\ x_t \end{pmatrix} + b \right)$$

$$h_t = \tanh \left( \begin{pmatrix} \square \\ c_t \end{pmatrix} \right)$$

• Gates:

forget      input      output

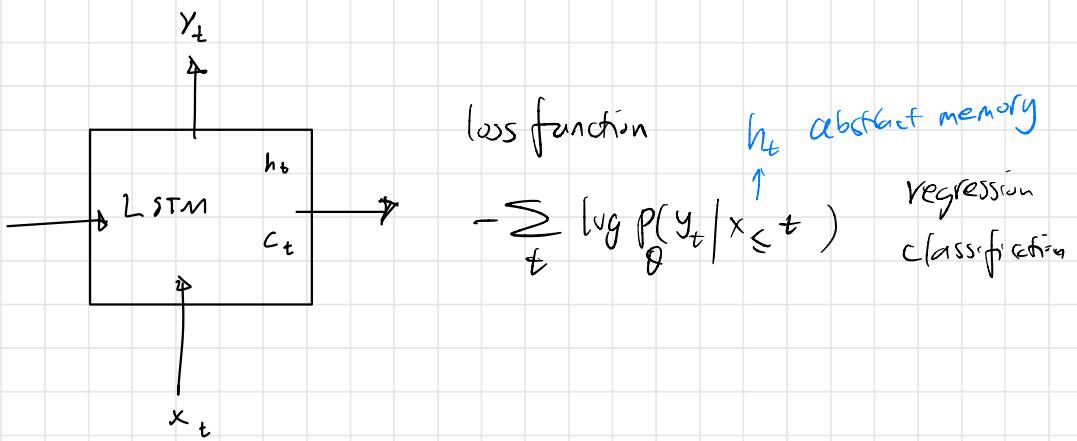
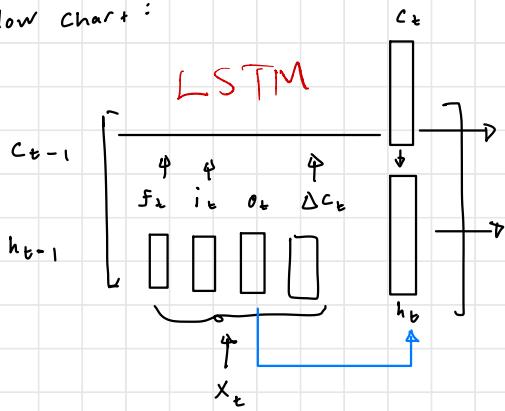
$F_t$	$i_t$	$O_t$	$= \text{sigmoid} \left( w h_{t-1} + b \right)$
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Each of them  $O_t$

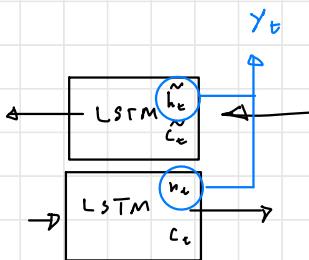
$$C_t = F_t \odot C_{t-1} + i_t \odot \Delta C_t$$

$$h_t = \tanh O_t \odot C_t$$

• Flow chart :



• Bidirectional LSTM :



Sequential