

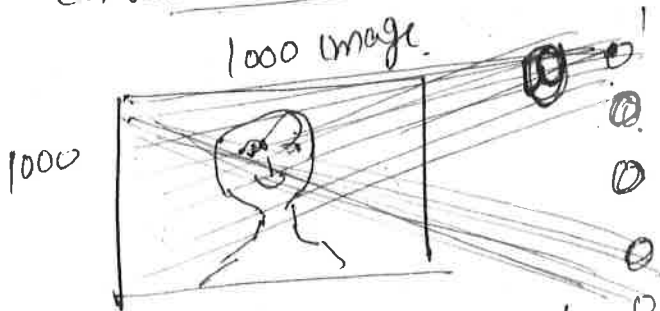
Deep Learning (Machine Learning)

①

Convolutional Neural Networks

April 29, 2017

COL 774



Too many parameters.

If we have 1000 hidden units

1 billion parameters
Too many!

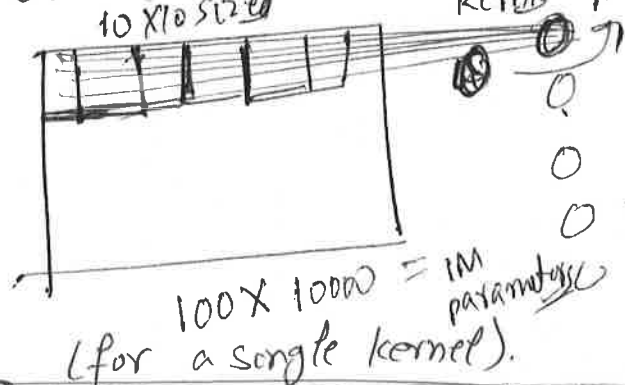
②

Idea:

Image has a spatial arrangement.

→ Exploit compositionality
→ Face is made of multiple parts.

Use localized kernels.



③

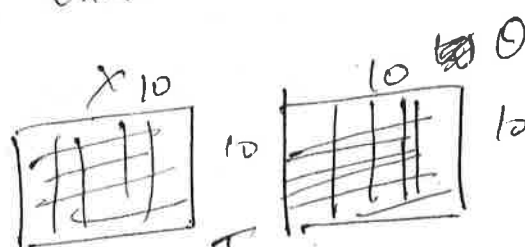
Further - the parameters are tied:

$10 \times 10 = 100$ parameters + 1 for bias

There can be multiple

such kernels (depth d) \rightarrow $(10 \times 10 \times d) + 1$ parameters

Each kernel learns a different kind of pattern. Given a kernel w.



filters for depth

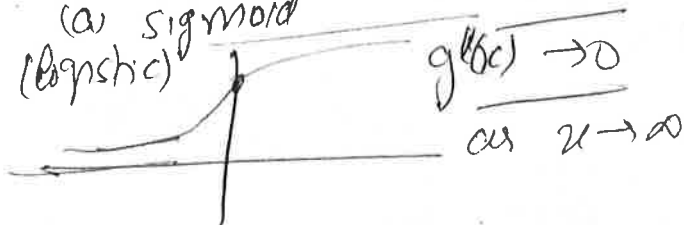
④

$$P(x) = g(OTx)$$

g function.

$$g(x) = \frac{1}{1 + e^{-x}}$$

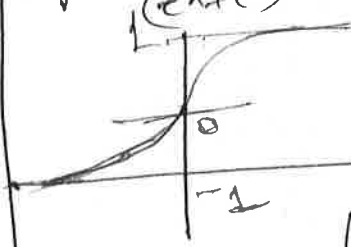
(a) sigmoid (logistic)



(b) tanh: $g(x)$

$$g'(x) \rightarrow 0 \text{ as } x \rightarrow \infty$$

$$g'(x) = \frac{e^x - e^{-x}}{(e^x + e^{-x})^2}$$

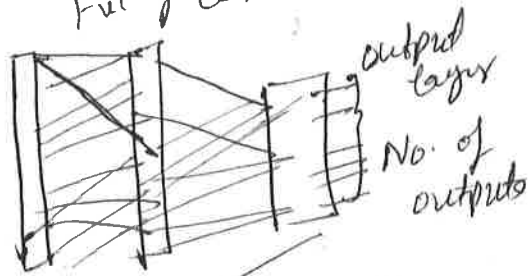


(c)

$$g(x) = \ln(1 + e^x)$$

$$g'(x) = \frac{e^x}{1 + e^x}$$

Fully Connected Layer



implements softmax

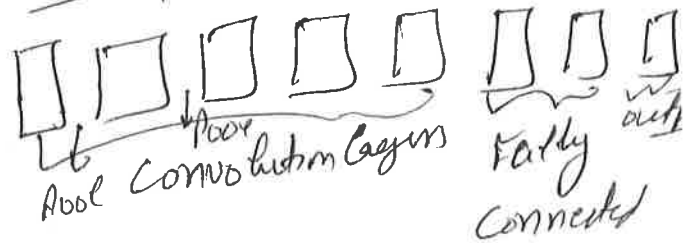
$$O(l) = \frac{e^{x_l}}{\sum_{k=1}^L e^{x_k}}$$

Diagram illustrating the softmax function. It shows a box containing the formula $O(l) = \frac{e^{x_l}}{\sum_{k=1}^L e^{x_k}}$. The denominator is written as "sum of e^{x_k} for $k=1$ to L ".

generalization of logistic

L : Number of labels

Alexnet: [2012]



Dropout: (hidden layer)

Randomly omit each to competition with $p=0.5$

Avoids overfitting

Same