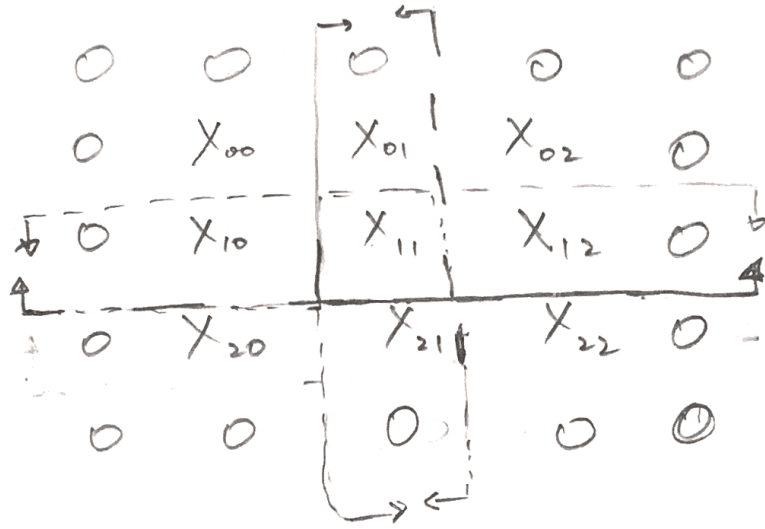


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1.1

zero-pad size = 1

stride = 2

y shape = (4, 1)

X shape = (9, 1)

→ A shape = (4, 9)

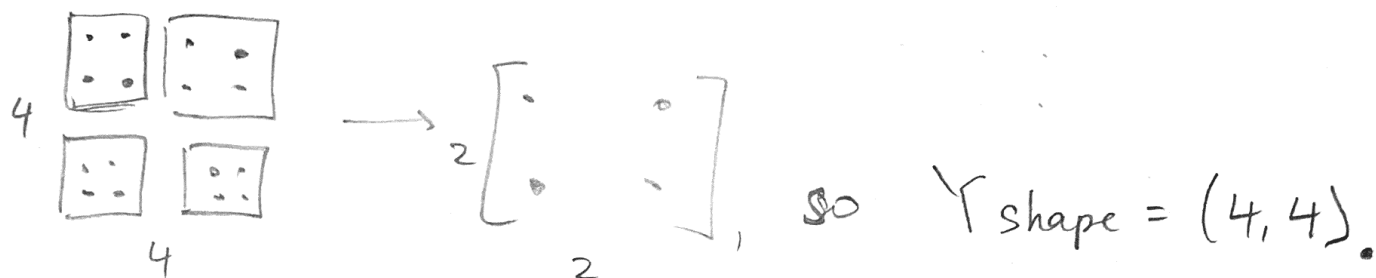
$$A = \begin{bmatrix} X_{00} & X_{01} & X_{02} & X_{10} & X_{11} & X_{12} & X_{20} & X_{21} & X_{22} \\ W_{11} & W_{12} & 0 & W_{21} & W_{22} & 0 & 0 & 0 & 0 \\ 0 & W_{10} & W_{11} & 0 & W_{20} & W_{21} & 0 & 0 & 0 \\ 0 & 0 & 0 & W_{01} & W_{02} & 0 & W_{11} & W_{12} & 0 \\ 0 & 0 & 0 & 0 & W_{00} & W_{01} & 0 & W_{10} & W_{11} \end{bmatrix}$$

$$\text{So, } y = A \cdot X$$

$$\begin{bmatrix} y_{00} \\ y_{01} \\ y_{10} \\ y_{11} \end{bmatrix} = \begin{bmatrix} W_{11} & W_{12} & 0 & W_{21} & W_{22} & 0 & 0 & 0 & 0 \\ 0 & W_{10} & W_{11} & 0 & W_{20} & W_{21} & 0 & 0 & 0 \\ 0 & 0 & 0 & W_{01} & W_{02} & 0 & W_{11} & W_{12} & 0 \\ 0 & 0 & 0 & 0 & W_{00} & W_{01} & 0 & W_{10} & W_{11} \end{bmatrix} \begin{bmatrix} X_{00} \\ X_{01} \\ X_{02} \\ X_{10} \\ X_{11} \\ X_{12} \\ X_{20} \\ X_{21} \\ X_{22} \end{bmatrix}$$

1-2 |  $W_{\text{shape}} = (2, 2)$      $X_{\text{shape}} = (2, 2)$ .    stride 2, no pad

So a forward convolution takes  $(r, c)$ , does stride 2  
w/  $2 \times 2$  kernel, no pad, to get  $2 \times 2$  output.



Must use same kernel and stride to convert

input dim  $(2+2p', 2+2p')$  to  $(4, 4)$

$$4 = ((2+2p') - 2) / 2 + 1 \quad \text{from} \quad ((W+2P) - F) / S + 1$$

$(3 = p')$  So zero pad  $\times$  w/ 3.

$y =$  Convolve  $2 \times 2$ , w/ 2 stride on the following:



$4 \times 4 \rightarrow 16 \times 1$  row major

$$y = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & X_{00}W_{11} & X_{01}W_{10} & 0 \\ 0 & X_{10}W_{01} & X_{11}W_{00} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Now  $A$  has shape  $(16 \times 4)$

Need to find  $A$  s.t.  $y_{\text{row}} = A \begin{bmatrix} X_{00} \\ X_{01} \\ X_{10} \\ X_{11} \end{bmatrix} \rightarrow$

$y =$

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ w_{11} x_{00} \\ w_{10} x_{01} \\ 0 \\ 0 \\ w_{01} x_{10} \\ w_{00} x_{11} \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$\downarrow$   
 $y$

$=$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & w_{11} & 0 & 0 & 0 \\ 0 & 0 & w_{10} & 0 & 0 \\ 0 & 0 & 0 & w_{01} & 0 \\ 0 & 0 & 0 & 0 & w_{00} \end{bmatrix} \begin{bmatrix} x_{00} \\ x_{01} \\ x_{10} \\ x_{11} \\ 0 \end{bmatrix}$$

$5 \times 4$

$=$

$\downarrow$   
 $A$

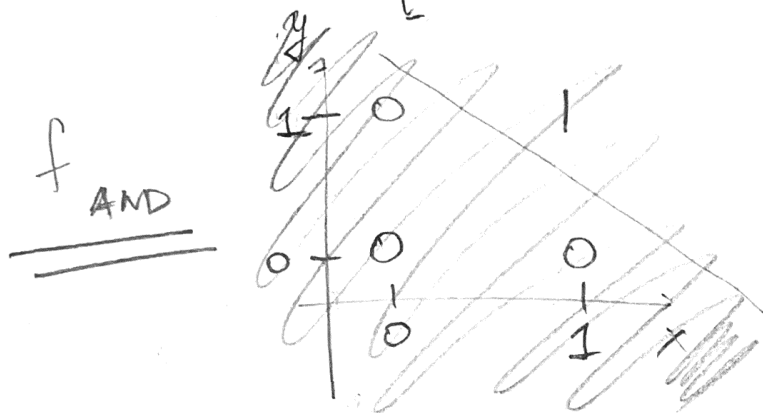
$$\begin{bmatrix} x_{00} \\ x_{01} \\ x_{10} \\ x_{11} \end{bmatrix}$$

$\downarrow$   
 $x$

2.1

$$w \in \mathbb{R}^2, b \in \mathbb{R}, x \in \{0,1\}^2$$

$$f(x) = \begin{cases} 1 & w^T x + b \geq 0 \rightarrow w^T x \geq -b \\ 0 & w^T x + b < 0 \rightarrow w^T x < -b \end{cases}$$



$$w_1(1) + w_2(1) \geq -b \quad (a)$$

$$w_1(0) + w_2(0) < -b \quad (b)$$

$$w_1(0) + w_2(1) < -b \quad (c)$$

$$w_1(1) + w_2(0) < -b \quad (d)$$

from (b),  $0 < -b$ , so  $b < 0$ . Let's fix  $b = -1$

then (a) becomes  $w_1 + w_2 \geq 1$

(c) "  $w_2 < 1$

(d) "  $w_1 < 1$

Letting  $w_1 = w_2 = 0.6$

satisfies these 3 inequalities.

$$w_{\text{AND}} = \begin{pmatrix} 0.6 \\ 0.6 \end{pmatrix} \quad b_{\text{AND}} = -1.0$$

f OR

$w_1(0) + w_2(0) < -b \rightarrow 0 < -b, b < 0$ , Again fix  $b = -1$

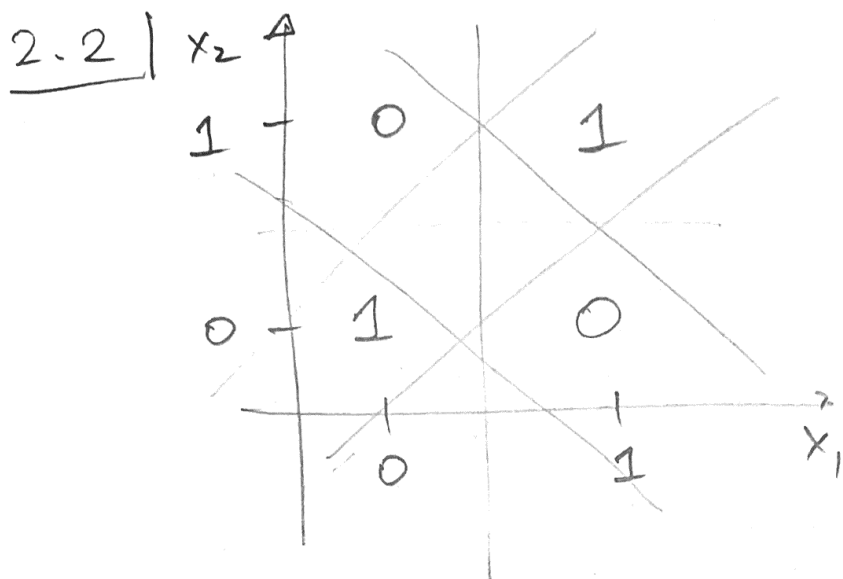
$$w_1(0) + w_2(1) \geq 1 \rightarrow w_2 \geq 1 \quad (a)$$

$$w_1(1) + w_2(0) \geq 1 \rightarrow w_1 \geq 1 \quad (b)$$

$$w_1(1) + w_2(1) \geq 1 \rightarrow w_1 + w_2 \geq 1 \quad (c)$$

Let  $w_1 = w_2 = 1$ , satisfies a, b, c.

$$w_{\text{OR}} = \begin{pmatrix} 1.0 \\ 1.0 \end{pmatrix} \quad b_{\text{OR}} = -1.0$$



$$w_1(0) + w_2(0) \geq -b \rightarrow 0 \geq -b \rightarrow b \geq 0 \quad (a)$$

$$w_1(1) + w_2(1) \geq -b \rightarrow w_1 + w_2 \geq -b \quad *$$

$$w_1(1) + w_2(0) < -b \rightarrow w_1 < -b \quad **$$

$$w_1(0) + w_2(1) < -b \rightarrow w_2 < -b \quad ***$$

from \*\*, \*\*\*,  $w_1 + w_2 < -2b$

$$\Rightarrow w_1 + w_2 + 2b < 0. \quad (b)$$

from \*,  $w_1 + w_2 + b \geq 0$

Let  $X = w_1 + w_2 + b$ , then  $X \geq 0$ . (c)

from (b),  $(w_1 + w_2 + b) + b < 0, \Rightarrow X + b < 0$

from (a), (b), we know  $X \geq 0, b \geq 0$ , so  $X + b \geq 0$ .

this contradicts (b) ( $X + b < 0$ )

So " $\Leftrightarrow$ " cannot be represented using linear model of given form.

$$\underline{3.1} \quad \sigma(\cdot) = f_1(\cdot) = \left| \vec{W}^{(1)} \vec{x} + \vec{b} \right| = \left| \underset{d \times d}{2I} \cdot \underset{d \times 1}{\vec{x}} + \underset{d \times 1}{\vec{b}} \right|_{(d \times 1)}$$

Look at each element in  $|2I\vec{x} + \vec{b}|$

call it  $| (2(1)x_i + b_i) | \quad \forall i \in \{1 \dots d\}$ .

$b_i = -1$ , (this is given).

We only care about  $O = (0, 1)^{d \times 1}$    
  $\swarrow$  open range

So  $|2x_i - 1| < 1$ .

$$-1 < 2x_i - 1 < 1$$

$$\frac{0}{2} < \frac{2x_i}{2} < \frac{2}{2}$$

$$0 < x_i < 1$$

So, for each element,  $(0, 1)$  is the only input region that can be mapped to output region  $(0, 1)$  with given  $W^{(1)}, b^{(1)}$ ,  
each element of  $\sigma(\cdot)$  has 1 input regions.

So, total of  $1^d$  input regions = 1 input region

$\sigma(\cdot)$  is a bijection, since  $\sigma^{-1}(\cdot)$  exists.

$$\sigma^{-1}(\cdot) = \frac{1}{2} I (\vec{x} - \vec{b})$$

3.2]  $f \circ g(\cdot)$  identifies  $n_g \cdot n_f$  regions onto  $(0,1)^d$

3.3] from the explanation at the top of section 3-Depth, each layer  $h^{(r)}$  has  $d$  elements, each of which identifies 2 region inputs, so each layer has  $2^d$  regions that are identified.

Since the entire net has  $L$  layers, and from the result of 3.2, composition of functions identifies a number of regions equal to the product of the number of regions identified by each composed function,

Then the  $L$  layer net is a composition of  $L$  functions and each layer has  $\underbrace{2 \cdot 2 \cdots 2}_d$  or  $2^d$  identified regions

$$\underline{\underline{L \text{ layer net}}} = \underbrace{2^d \cdot 2^d \cdots 2^d}_L = \boxed{2^{Ld}}$$