

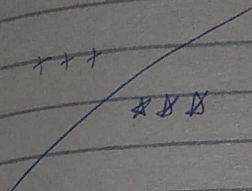
19-11-2021

week # 14 lecture # 17

Monday

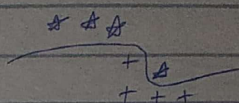
Paper Discussion

Q#1



Linear classifier

WX



Non-linear classifier.

$w_3(w_2(w_1X)) = WX$
as a linear classifier.

λ = Regularization strength

Hyperparameter

^{Regularization}
 L_2 also prefers smaller weights & also
prefers diffused weights

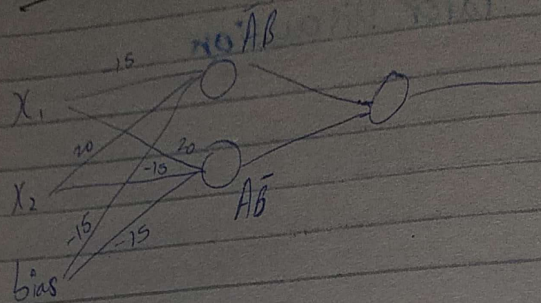
Q11) Non-linear MATHY function characteristics:

a) Don't saturate output

b) easy to compute

c) ~~Not~~ Zero centered.

Q19) XOR:-



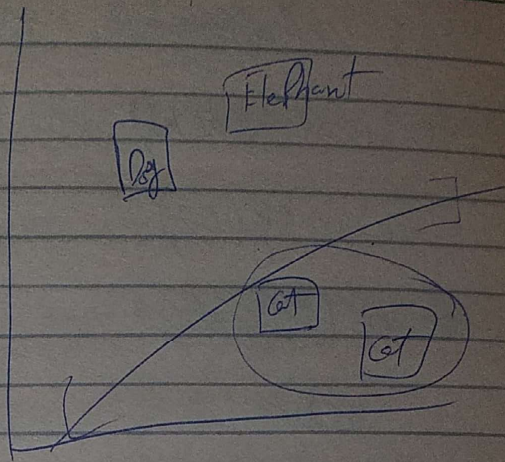
we have to ~~show~~ show what could be the weights in this NN

x_1	x_2	XOR
A	B	
0	0	0
0	1	1
1	0	1
1	1	0

$$XOR = AB + A\bar{B}$$

Q20) Interpretations of rows in w:

- ① Generic ~~image~~ image of each class
- ② Each row is a equation of line which separates this class from other classes.
- ③ weights +ve or -ve.



Q2) sigmoid + loss is

$$\sum_y -\log(p_y)$$

for multi-label classification.

Q2/3) code.

```
x = np.array([2, -1, 3])
```

```
w = np.array([[0.1, 0.2, 0.3], [-1, 1, 1], [2, 4, 0.5]])
```

```
o1 = w.dot(x)
```

Output of 3 neurons

```
def sig(x):
```

```
    return 1 / (1 + np.exp(-x))
```

```
z = h1 * sig(o1)
```

$w2 = \text{np.array}([2, 0.5, 0.4])$

02. $w1 \cdot \text{dot}(h1)$

$h2 = \text{sig}(02)$

~~if $\text{loss} < 0.01$~~

~~return~~

$t1 = y * \text{np.log}(h2)$

$t2 = (1-y) * \text{np.log}(1-h2)$

$\text{loss} = t1 + t2$

Q2) Q3) gradient code.

Assignment #3-

MNIST

No batch normalization

Optimizer:

Not gradient descent

use Adam or

dropout

from scratch.