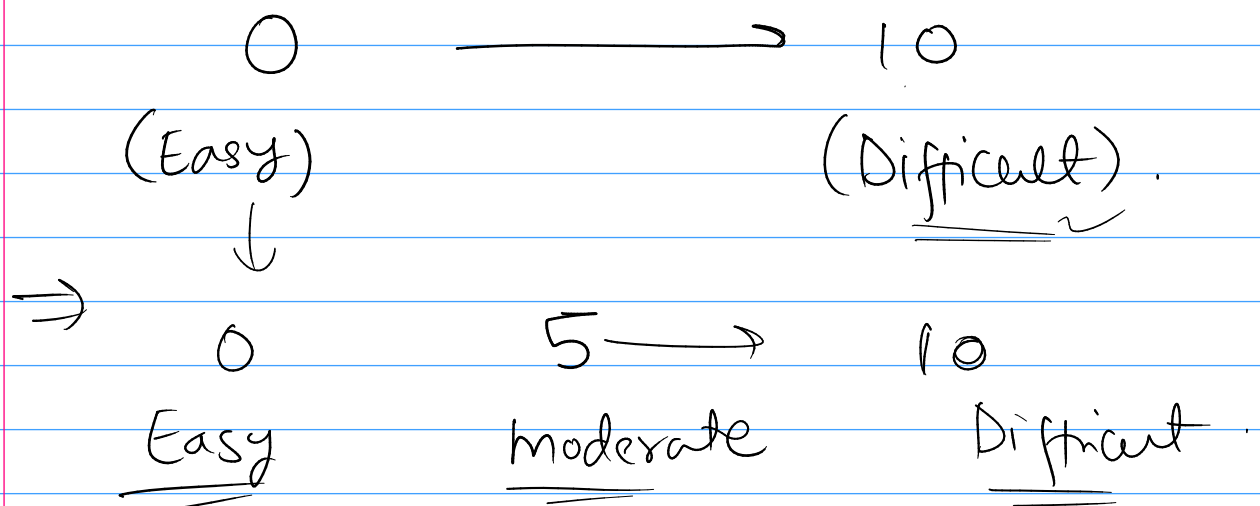


⇒ Rate the QP :-



1. How a loss function is different from metric function? Write 3 metric functions with there working.

Loss	Metric.
↓	↓
Used to Update weights.	just to evaluate <u>model</u> .

mae , mse , accuracy , binary
categorical - A , row -

2. What is maxpooling operation in Convnets? Explain.

↓

Down Sample your image $60 \times 60 \Rightarrow 30 \times 30$
(2,2)

↳ Reduces requirement of more
time for training :-

3. Calculate categorical cross-entropy loss for $y_{\text{true}} = [[0, 1, 0], [1, 0, 0], [0, 0, 1]]$, $y_{\text{pred}} = [[0.1, 0.8, 0.1], [0.1, 0.1, 0.8], [0.1, 0.1, 0.8]]$. Hint: $\ln(0.8) = -0.2231$, $\ln(0.1) = -2.3025$.

$$\text{Loss} = - \sum y_{\text{true}} \ln y_{\text{pred}}$$

$$= - (1 \times \ln(0.8) + 1 \times \ln(0.1) + 1 \times \ln(0.8))$$

$$= - (-0.2231 - 2.3025 - 0.2231)$$

$$= 2.7487 \quad \checkmark$$

0	0	1	1	0	0	0
0	0	1	1	0	0	0
0	0	1	1	0	0	0
0	0	1	1	0	0	0
0	0	1	1	0	0	0
0	0	1	1	0	0	0
0	0	1	1	0	0	0

Image Data

1	0	-1
1	0	-1
1	0	-1

Kernel

4. Do one convolution operation with the given kernel on the above image data. Given: **Stride = 2**

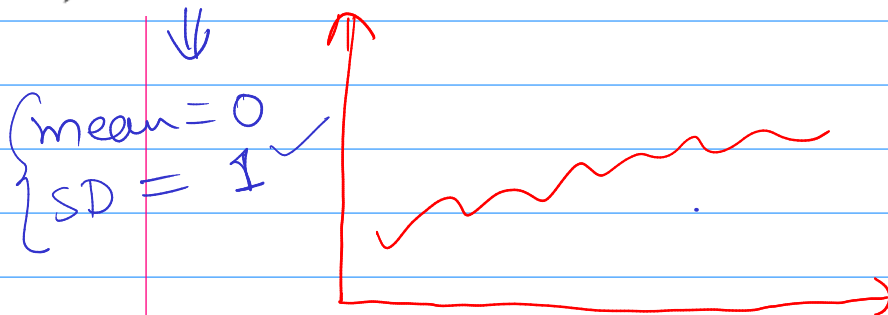
-3	3	0
-3	3	0
-3	3	0



training
⇒

test Data

5. Define these terms: a) Temporal Leak b) Redundancy in data c) Optimisation d) Generalisation e) Normalisation of data.



t in seconds.

t = 0, 1, 2, 3, ...

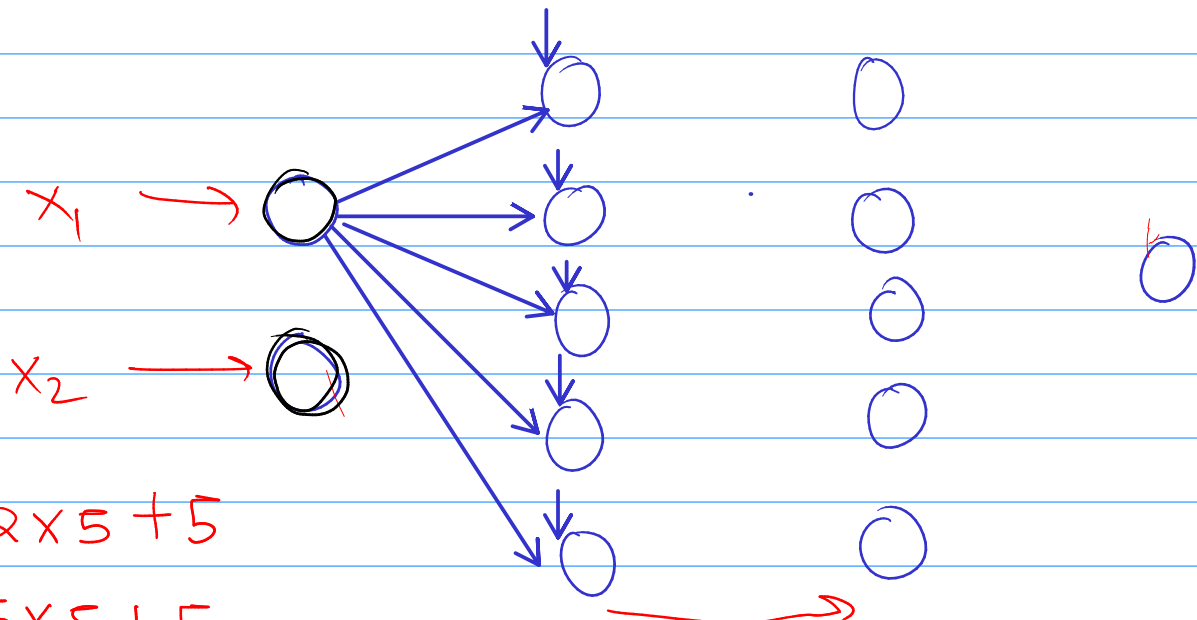
1, 3, 0, 2

1. Make a custom loss function which returns $(y_true - y_pred)^4$ as loss. Also write the compile statement for the model. Hint : $tf.pow()$, $tf.subtract()$

```
def custom_loss(y_true, y_pred):
    2.5/3 sub = tf.subtract(y_true, y_pred)
    return tf.pow(sub, 4)
```

```
model.compile(optimizer='rmsprop',
              ② loss=custom_loss,
              metrics=['mae'])
```

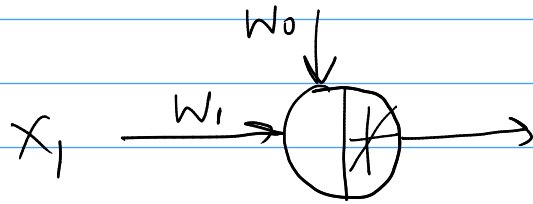
2. A dense neural network is formed in 2 - 5 - 5 - 1 fashion. Calculate the total number of trainable parameters (weights).



$$\begin{aligned}
 &2 \times 5 + 5 \\
 &+ 5 \times 5 + 5 \\
 &+ 1 \times 5 + 1
 \end{aligned}$$

51 ✓

3. A loss function depends upon two weights w_0 and w_1 as $\text{loss} = w_0 + w_1$. Find out the direction in which this loss can be minimised. *Hint: Gradient Descent*



$$y = w_1 x_1 + w_0$$

$$0 = (w_1 x_1 + w_0)$$

$$x_1 = 1$$

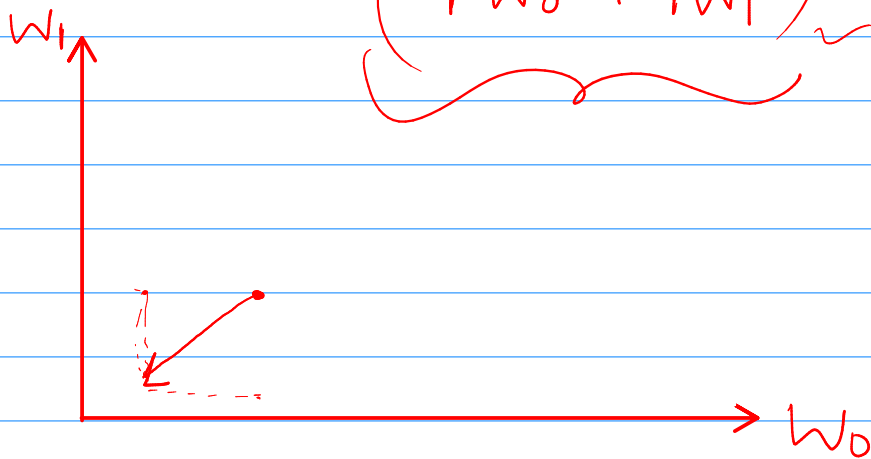
$$\text{loss} = -(w_0 + w_1)$$

direction

$$\Rightarrow -\left(\frac{\partial}{\partial w_0} \hat{w}_0 + \frac{\partial}{\partial w_1} \hat{w}_1 + \dots\right) \text{loss}$$

\Rightarrow

$$-(1 \hat{w}_0 + 1 \hat{w}_1)$$



$$w' = \begin{bmatrix} 0 \\ 0 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

4. WiFi network strength was measured by changing distance from router (D), router antenna length (L) and manufacturer (M) (Nokia = 0, TP-Link = 1, D-Link = 2) and a table was prepared. Write a code/pseudo-code (Neural Network) for predicting Wifi network strength at some set of values (D', L', M').

D	L	M	WiFi
1	10cm	0	-2dB
2	-	2	-

Regression or Classification

Binary

Multi-class.

