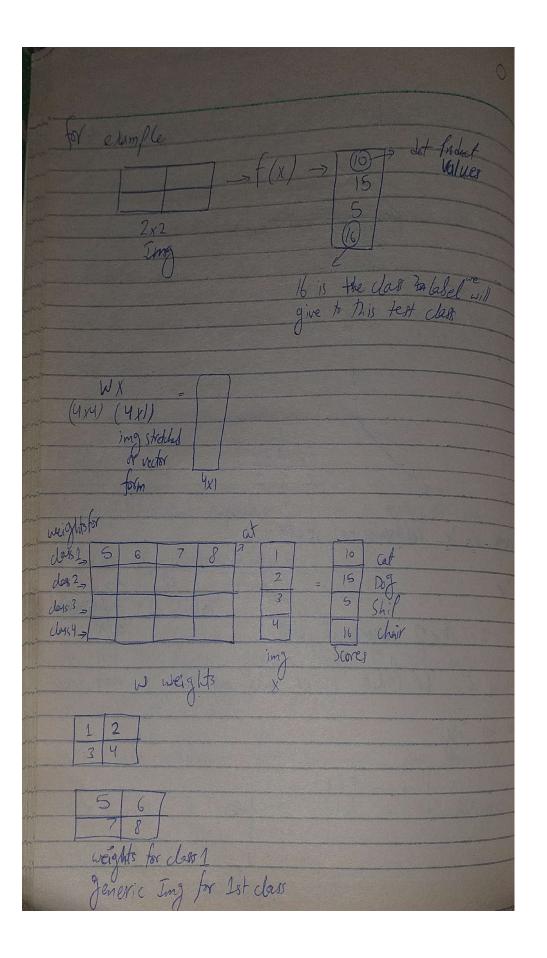
20-9-2021 Tuesday
First stretch the image 3 Jimensiand ing to 1 Dimensional array
3 0 255 255 0 0 0 0 0 0 0 0 0
6 Stretched to 1D array
0 0 255 0 255 255 0 0 0 0 0 0
Now, element wise subtraction of images from test set &
Training set images
e.g 10 0 5 6 5
0 10 0 10 0
= [10] -10 5 -4 5
Abs = [10 10 5 4 5]
10+10+5+4+5=34
\$100 C 100 C
L, 1P1-P21
L2 (P, -P2)2 Penalizes those Pixels that have

large Listance In L, distance & of 2 Pixels is 10, 34 Selavately while in Lz distance, same is 100, 16 difference to in first is 6, while in 12 is 84. Hyperparameters: Value of Kis add bcz at even we get ties. K can be 1, 3, 5, 7, 9 ... we fut these values for K & check accuracy on each value & Pick best accuracy for that dataset Accuracy = No. of correctly Predicted

No. of total Images/Total Predicted Images Actual labels Presides Lasels Accuracy = 5 = 55% K=1 in Training set, Accuracy = 100%

So, we don't use training set to check accuracy. we divide train set into 2 sets a) Train set a used for Predictations of the b) Malication Set - used to check Accuracy. Also K can't be greater than no. of imags in the train set Test set - unseen data Problems: 1 Computationally expensive (Slow in testing for large data se 2 curse of Dimensionality. Linear Classifier: we try to learn generic ing of all the dasses we want to Predict. Then take difference from her unseen ing from these generic ing & whoter is closest will be that image class.

This we use dot Product as our distance metric Ind3 5000 18,000 Image can be refresented as vector like 27 Limensiand 2x2x3 = 12 Limensional Point vector Let Product mans thise ing class is more closer to test imag 4 no. of classes



mate inf argmax(wx)
index of more value in column vector is the class Y=mx+C if C=0, line Passes through 0)C1 X2 X3 X4 mill + M2 X2+ M3 X3+ M4 X4 bias is added to learn more sleific Accurate images