Lecture 07 Submission Pranav

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1) a) There is I feature (ie 2)
  b) There are 2 parameters that need to be
   estimated, they are a & b
  c) The activation function is an identity
     function.
          x(x)=x.
    This function y=axtb is similar to y=mx+c
     turefore it is an identity function
 d) Loss function for mean squared ever.
        L= 1 \ \( \hat{y}^{(i)} - y^{(i)} \)^2
```

Solution for 1d

```
def calc_least_square(a, b, list1):
    sum_of_squares = 0
    for [xi, yi] in list1:
        sum_of_squares += (yi - (a * xi + b)) ** 2
    return sum_of_squares / (2 * len(list1))

val_0_0 = calc_least_square(0, 0, [[2, 1], [3,2], [4,3]] )

val_0_1 = calc_least_square(0, 1, [[2, 1], [3,2], [4,3]] )
```

```
val_0_2 = calc_least_square(0, 2, [[2, 1], [3,2], [4,3]] )
val_1_0 =calc_least_square(1, 0, [[2, 1], [3,2], [4,3]] )
val_1_1 = calc_least_square(1, 1, [[2, 1], [3,2], [4,3]] )
val_1_2 =calc_least_square(1, 2, [[2, 1], [3,2], [4,3]] )
val_2_0 = calc_least_square(2, 0, [[2, 1], [3,2], [4,3]] )
val_2_1 =calc_least_square(2, 1, [[2, 1], [3,2], [4,3]] )
val_2_2 = calc_least_square(2, 2, [[2, 1], [3,2], [4,3]] )
import pandas as pd
a_0 = pd.Series([val_0_0, val_0_1, val_0_2])
a_1 = pd.Series([val_1_0, val_1_1, val_1_2])
a_2 = pd.Series([val_2_0, val_2_1, val_2_2])
pd.concat([a_0, a_1, a_2] ,axis=1)
\overline{\Rightarrow}
                             2
     0 2.333333 0.5
                      8.333333
     1 0.833333 2.0 12.833333
     2 0.333333 4.5 18.333333
```

Here I wrote a function to calcualte the mean squared error and substituted the values, the rows represent b and the columns represent a, each of them take values 0, 1, 2 and we have 9 combindations

Solution for 1e, Partial derivative with respect to a

Final value is (1 / n) * ((y_hat - y) * x)

Soltuion for 1f, partial derivative with respect to b

f)
$$y = ax + b$$
.

$$L = \frac{1}{2n} \sum (\hat{y}^{(i)} - y^{(i)})^{2}.$$

Considering it as a Scalar for the purpose of durindon

$$L = \frac{1}{2n} (\hat{y} - y)^{2}.$$

$$L = \frac{1}{2n} ((ax + b) - y)^{2}$$

Partially deriving with respect to b .

$$\frac{\partial L}{\partial b} = \frac{\partial}{\partial b} \frac{1}{2n} ((ax + b) - y)^{2}.$$

Using the chain sule $\left[\frac{d}{dn} f(g(x)) = f'(g(x), g'(x))\right]$.

$$= \frac{1}{2n} \cdot \chi (ax + b - y) \cdot \frac{\partial}{\partial b} (an + b - y).$$

Since $\frac{d}{dn} = \frac{1}{2n} \cdot \chi (ax + b - y) \cdot \frac{1}{2n} \cdot$