

③ Regression

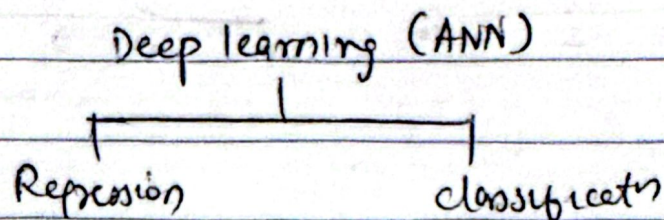
Hidden \rightarrow ReLU or variant of ReLU

o/p \rightarrow Linear Activation



separate loss function

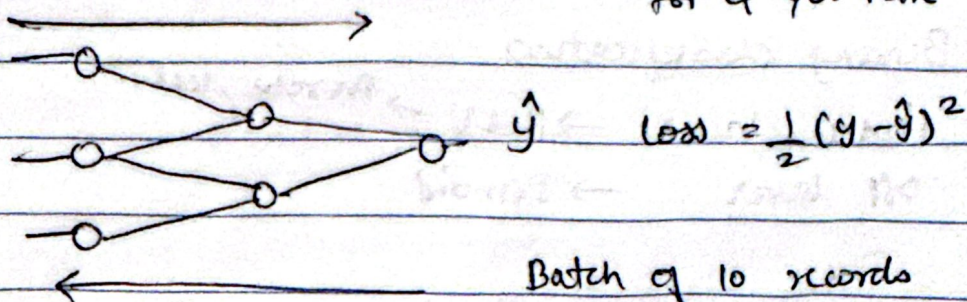
LOSS FUNCTIONS



① Regression

- MSE
- MAE
- ~~RMS~~ Huber Loss

for if you have 100 records



Batch of 10 records

then it will be cost function

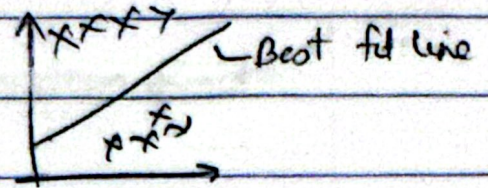
$$\text{cost} = \frac{1}{2} \sum_{i=1}^n (y - \hat{y})^2$$

① Mean Squared Error (MSE)

$$\text{loss function} = \frac{1}{2n} (y - \hat{y})^2$$

$$\text{Cost function} = \frac{1}{2} \sum_{i=1}^n (y - \hat{y})^2$$

- Not robust to outliers

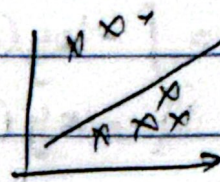


② Mean Absolute Error (MAE)

$$\text{loss function} = \frac{1}{2n} |y - \hat{y}|$$

$$\text{Cost function} = \frac{1}{2n} \sum_{i=1}^n |y - \hat{y}|$$

- Robust to outliers
- Time consuming



③ Huber loss

Combination of MSE & MAE

Hyperparameter

$$\text{loss} = \begin{cases} \frac{1}{2} (y - \hat{y})^2 & \text{if } |y - \hat{y}| \leq \delta \\ \delta |y - \hat{y}| - \frac{1}{2} \delta^2, & \text{otherwise} \end{cases}$$

outliers not present

present

② Classification

Cross Entropy — $\begin{cases} \rightarrow \text{Binary classification} & \text{cross entropy} \rightarrow \text{Binary classification} \\ \rightarrow \text{Categorical cross entropy} & \rightarrow \text{multiclass classification} \end{cases}$

① Binary cross entropy

$$\text{Loss} = -y * \log(\hat{y}) - (1-y) * \log(1-\hat{y}) \rightarrow \text{used in Log Rep}^m$$

\Downarrow

$$\text{Loss} = \begin{cases} -\log(1-\hat{y}) & \text{if } y=0 \\ -\log(\hat{y}) & \text{if } y=1 \end{cases} \quad \left. \begin{array}{l} \text{Binary} \\ \text{classification} \end{array} \right\}$$

$$\hat{y} = \frac{1}{1+e^{-x}}$$

② Categorical cross entropy

f1 f1	f2	f3	o/p	Good	Bad	Neutral
2	3	4	Good	1	0	0
5	6	7	Bad	0	1	0
8	9	10	Neutral	0	0	1

No. of categories

first we convert it into one hot encoding

$$\text{loss} = \mathcal{L}(x_i, y_i) = - \sum_{j=1}^c y_{ij} \ln(\hat{y}_{ij})$$

Here we use Softmax Activation functⁿ

learn about this

$$\sigma(z) = \frac{e^{z_i}}{\sum_{j=1}^k e^{z_j}} \quad \left. \vphantom{\frac{e^{z_i}}{\sum_{j=1}^k e^{z_j}}} \right\} \text{softmax Actn functn}$$

its applied in every output layer for multiclass classificatⁿ problem

Important

Binary classificatⁿ problem

- Use ReLU Actn functⁿ for Hidden & then sigmoid in the output layer

Multiclass classificatⁿ

- Middle layer use ReLU & the ^{output} ~~output~~ use Softmax Actn functⁿ

→ Classification

Categorical

Relu, Softmax - multiclass →

Relu, Sigmoid - Binary → Binary

→ Regression

MSE, MAE, Huber

Relu, Linear Actiⁿ function

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