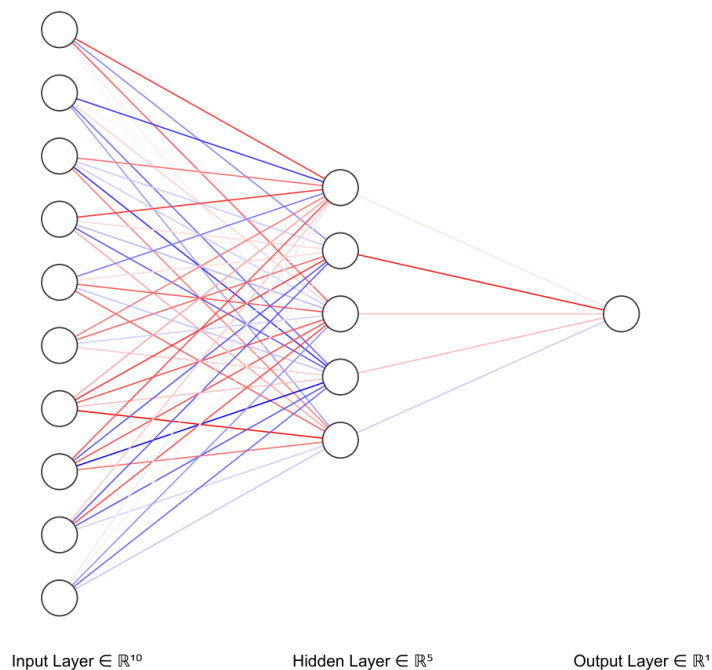


Neural Network from scratch



Dot product of Vector and Matrix

$$c_k = \sum_i a_i b_{ik}$$

```
In [1]: def vec_dot_mat(a, b):  
        c = []  
        for k in range(len(b[0])):  
            s = 0  
            for i in range(len(a)):  
                s += a[i] * b[i][k]  
            c.append(s)  
        return c
```

Random matrix

```
In [2]: from random import uniform  
  
def rand_matrix(i, j):  
    return [[uniform(-1, 1) for _ in range(j)] for _ in range(i)]
```

Let us take,

- Input matrix X of size (10, 10)
- True weight1 matrix W_{t1} of size (10, 5)
- True bias1 b_{t1}
- True weight2 matrix W_{t2} of size (5, 1)
- True bias1 b_{t2}

```
In [3]: x = rand_matrix(10, 10)  
wt1 = rand_matrix(10, 5)  
bt1 = uniform(-1, 1)
```

```
wt2 = rand_matrix(5, 1)
bt2 = uniform(-1, 1)
```

We have to find out below weights and biases

```
In [4]: w1 = rand_matrix(10, 5)
        b1 = uniform(-1, 1)
        w2 = rand_matrix(5, 1)
        b2 = uniform(-1, 1)
```

$$\text{forward}_l = \sum_k \left[\left[\sum_j x_j \cdot w1_{jk} + b1 \right] \cdot w2_{kl} \right] + b2$$

```
In [5]: def forward(x, w1, w2, b1, b2):
        z1 = [i+b1 for i in vec_dot_mat(x, w1)]
        z2 = [i+b2 for i in vec_dot_mat(z1, w2)]
        return z2
```

```
In [6]: y = [forward(i, wt1, wt2, bt1, bt2) for i in x]
```

Modification of weight

```
In [7]: def mod_w(w, index, h):
        w[index[0]][index[1]] += h
        return w
```

$$\text{loss} = \frac{1}{N} \sum \left\{ y_l - \sum_k \left[\left[\sum_j x_j \cdot w1_{jk} + b1 \right] \cdot w2_{kl} \right] + b2 \right\}^2$$

```
In [8]: def loss(x, y, w1, w2, b1, b2):
        l = 0
        yp = forward(x, w1, w2, b1, b2)
        for i in range(len(y)):
            l += (y[i] - yp[i]) ** 2
        l /= len(y)
        return l
```

$$\text{overall_loss} = \frac{1}{N^2} \sum \sum \left\{ y_{il} - \sum_k \left[\left[\sum_j x_{ij} \cdot w1_{jk} + b1 \right] \cdot w2_{kl} \right] + b2 \right\}^2$$

```
In [9]: def overall_loss(x, y, w1, w2, b1, b2):
        yp = [forward(i, w1, w2, b1, b2) for i in x]
        l = 0
        for i in range(len(y)):
            for j in range(len(y[0])):
                l += (y[i][j] - yp[i][j]) ** 2
        l /= len(y) * len(y[0])
        return l
```

$$\text{grad}_w = \frac{\text{loss}(\text{mod}_w(w, (i, j), h)) - \text{loss}(\text{mod}_w(w, (i, j), -h))}{2h}$$

$$\text{grad}_b = \frac{\text{loss}(b + h) - \text{loss}(b - h)}{2h}$$

$$w1 \leftarrow w1 - lr \cdot \Delta w1$$

$$w2 \leftarrow w2 - lr \cdot \Delta w2$$

$$b1 \leftarrow b1 - lr \cdot \Delta b1$$

$$b2 \leftarrow b2 - lr \cdot \Delta b2$$

```
In [10]: def grad(x, y, w1, w2, b1, b2, h, lr):
    dw1 = []
    for i in range(len(w1)):
        temp = []
        for j in range(len(w1[0])):
            temp.append((loss(x, y, mod_w(w1, (i, j), h), w2, b1, b2) \
                - loss(x, y, mod_w(w1, (i, j), -h), w2, b1, b2)) / (2 * h))
        dw1.append(temp)

    dw2 = []
    for i in range(len(w2)):
        temp = []
        for j in range(len(w2[0])):
            temp.append((loss(x, y, w1, mod_w(w2, (i, j), h), b1, b2) \
                - loss(x, y, w1,
                    mod_w(w2, (i, j), -h), b1, b2)) / (2 * h))
        dw2.append(temp)

    db1 = (loss(x, y, w1, w2, b1+h, b2) \
        - loss(x, y, w1, w2, b1-h, b2)) / (2 * h)
    db2 = (loss(x, y, w1, w2, b1, b2+h) \
        - loss(x, y, w1, w2, b1, b2-h)) / (2 * h)

    for i in range(len(w1)):
        for j in range(len(w1[0])):
            w1[i][j] -= lr * dw1[i][j]

    for i in range(len(w2)):
        for j in range(len(w2[0])):
            w2[i][j] -= lr * dw2[i][j]

    b1 -= lr * db1
    b2 -= lr * db2

    return w1, w2, b1, b2
```

```
In [11]: epochs = 1000
lr = 0.01
h = 0.001
opt_w_b = (w1, w2, b1, b2)
lowest_loss = overall_loss(x, y, w1, w2, b1, b2)
print('Initial loss: {}'.format(lowest_loss))
for _ in range(epochs):
    for x_row, y_row in zip(x,y):
        w1, w2, b1, b2 = grad(x_row, y_row, w1, w2, b1, b2, h, lr)
        l = overall_loss(x, y, w1, w2, b1, b2)
        if l < lowest_loss:
            lowest_loss = l
            opt_w_b = (w1, w2, b1, b2)
print('Lowest loss: {}'.format(lowest_loss))
print('Final loss: {}'.format(overall_loss(x, y, w1, w2, b1, b2)))
```

Initial loss: 13.28357350279802

Lowest loss: 7.2511179192521535e-06

Final loss: 7.2511179192521535e-06

Rust code

```
use rand::Rng;

fn main() {
    let mut rng = rand::thread_rng();

    fn vec_dot_mat(a:&Vec<f64>, b:&Vec<Vec<f64>>) -> Vec<f64> {
        let mut c = vec![];
        for k in 0..b[0].len() {
            let mut s = 0.0;
            for j in 0..a.len() {
                s += a[j] * b[j][k];
            }
            c.push(s);
        }
        c
    }

    fn rand_matrix(i:usize, j:usize) -> Vec<Vec<f64>> {
        let mut rng = rand::thread_rng();
        let mut m = vec![vec![0.0;j];i];
        for a in 0..i {
            for b in 0..j {
                m[a][b] = rng.gen_range(-1.0..=1.0);
            }
        }
        m
    }

    let x = rand_matrix(10, 10);
    let wt1 = rand_matrix(10, 5);
    let bt1 = rng.gen_range(-1.0..=1.0);
    let wt2 = rand_matrix(5, 1);
    let bt2 = rng.gen_range(-1.0..=1.0);

    let mut w1 = rand_matrix(10, 5);
    let mut b1 = rng.gen_range(-1.0..=1.0);
    let mut w2 = rand_matrix(5, 1);
    let mut b2 = rng.gen_range(-1.0..=1.0);

    fn forward(x:&Vec<f64>, w1:&Vec<Vec<f64>>, w2:&Vec<Vec<f64>>, b1:&f64,
b2:&f64) -> Vec<f64> {
        let z1:Vec<f64> = vec_dot_mat(x,
w1).iter().map(|i|i+b1).collect();
        let z2 = vec_dot_mat(&z1, w2).iter().map(|i|i+b2).collect();
        z2
    }

    let y:Vec<Vec<f64>> = x.iter().map(|i| forward(i, &wt1, &wt2, &bt1,
&bt2)).collect();

    fn mod_w(w:&Vec<Vec<f64>>, index:(usize, usize), h:f64) ->
Vec<Vec<f64>>{
        let mut w1 = w.clone();
        w1[index.0][index.1] += h;
        w1
    }
}
```

```

    fn loss(x:&Vec<f64>, y:&Vec<f64>, w1:&Vec<Vec<f64>>,
w2:&Vec<Vec<f64>>, b1:&f64, b2:&f64) -> f64 {
    let mut l = 0.0;
    let yp = forward(x, w1, w2, b1, b2);
    for i in 0..y.len() {
        l += (y[i] - yp[i]).powi(2);
    }
    l /= y.len() as f64;
    l
}

    fn overall_loss(x:&Vec<Vec<f64>>, y:&Vec<Vec<f64>>, w1:&Vec<Vec<f64>>,
w2:&Vec<Vec<f64>>, b1:&f64, b2:&f64) -> f64 {
    let mut yp = vec![];
    for i in x {
        let temp = forward(i, w1, w2, b1, b2);
        yp.push(temp);
    }
    let mut l = 0.0;
    for i in 0..y.len() {
        for j in 0..y[0].len() {
            l += (y[i][j] - yp[i][j]).powi(2);
        }
    }
    l /= (y.len() * y[0].len()) as f64;
    l
}

    fn grad(x:&Vec<f64>, y:&Vec<f64>, w1:&Vec<Vec<f64>>,
w2:&Vec<Vec<f64>>, b1:&f64, b2:&f64, h:f64) -> (Vec<Vec<f64>>,
Vec<Vec<f64>>, f64, f64) {
    let mut dw1 = vec![];
    for i in 0..w1.len() {
        let mut temp = vec![];
        for j in 0..w1[0].len() {
            temp.push((loss(x, y, &mod_w(w1, (i, j), h), w2, b1, b2) -
loss(x, y, &mod_w(w1, (i, j), -h), w2, b1, b2)) / (2.0 * h));
        }
        dw1.push(temp);
    }

    let mut dw2 = vec![];
    for i in 0..w2.len() {
        let mut temp = vec![];
        for j in 0..w2[0].len() {
            temp.push((loss(x, y, w1, &mod_w(w2, (i, j), h), b1, b2) -
loss(x, y, w1, &mod_w(w2, (i, j), -h), b1, b2)) / (2.0 * h));
        }
        dw2.push(temp);
    }

    let db1 = (loss(x, y, w1, w2, &(b1+h), b2) - loss(x, y, w1, w2, &
(b1-h), b2)) / (2.0 * h);
    let db2 = (loss(x, y, w1, w2, b1, &(b2+h)) - loss(x, y, w1, w2,
b1, &(b2-h))) / (2.0 * h);

    (dw1, dw2, db1, db2)
}

    let epochs = 1000;
    let lr = 0.01;

```

```

let h = 0.001;
let mut opt_w_b = (w1.clone(), w2.clone(), b1.clone(), b2.clone());
let mut lowest_loss = overall_loss(&x, &y, &w1, &w2, &b1, &b2);
println!("Initial loss: {}", lowest_loss);
for _ in 0..epochs {
    for index in 0..x.len() {
        let (dw1, dw2, db1, db2) = grad(&x[index], &y[index], &w1,
&w2, &b1, &b2, h);
        for i in 0..w1.len() {
            for j in 0..w1[0].len() {
                w1[i][j] -= lr * dw1[i][j];
            }
        }

        for i in 0..w2.len() {
            for j in 0..w2[0].len() {
                w2[i][j] -= lr * dw2[i][j];
            }
        }

        b1 -= lr * db1;
        b2 -= lr * db2;
    }
    let l = overall_loss(&x, &y, &w1, &w2, &b1, &b2);
    if l < lowest_loss {
        lowest_loss = l;
        opt_w_b = (w1.clone(), w2.clone(), b1.clone(), b2.clone());
    }
}
println!("Lowest loss: {}", lowest_loss);
println!("Final loss: {}", overall_loss(&x, &y, &w1, &w2, &b1, &b2));
// println!("{:?}", opt_w_b);
}

```

output:

```

Initial loss: 14.449090232922213
Lowest loss: 0.000005498838681749966
Final loss: 0.000005498838681749966

```