

```
In [1]: import numpy as np
```

```
In [2]: def ReLU(z):
    return np.maximum(z, 0)

def d_ReLU(z):
    return 1.0 * (z > 0)

def sigmoid(z):
    return 1.0 / (1.0 + np.exp(-z))

def d_sigmoid(z):
    return sigmoid(z) * (1 - sigmoid(z))

# Using log10
def binary_cross_entropy(y_hat, y_pred):
    return -1 * (y_hat * np.log10(y_pred) + (1 - y_hat) * np.log10(1 - y_pred))

def d_binary_cross_entropy(y_hat, y_pred):
    return -1 * (y_hat / y_pred - (1 - y_hat) / (1 - y_pred))

def max_pool(input, kernel, stride):
    z = np.zeros(((input.shape[0] - kernel) // stride + 1, (input.shape[1] - kernel) // stride + 1))
    idx = []

    for x in range(z.shape[0]):
        for y in range(z.shape[1]):
            i = x * kernel
            j = y * kernel
            block = input[i : i + kernel, j : j + kernel]
            z[x, y] = np.max(block)
            index = np.add(np.unravel_index(block.argmax(), block.shape), (i, j))
            idx.append(index)

    return z, idx

# Not using stride or padding for simplicity
def convolution(input, kernel, bias):
    output = np.zeros((input.shape[0] - kernel.shape[0] + 1, input.shape[1] - kernel.shape[1] + 1))
    for i in range(output.shape[0]):
        for j in range(output.shape[1]):
            output[i, j] = np.sum(input[i : i + kernel.shape[0], j : j + kernel.shape[1]])
    return output

# Simple visualization function
def printMatrix(A, text):
    print(text)
    print(A)
    print("----")
```

Forward pass

```
In [3]: input = np.array([[1, 2, 0, 2, 1],
                          [0, 1, 1, 2, 1],
                          [1, 2, 0, 1, 0],
                          [1, 2, 1, 2, 1],
                          [0, 0, 3, 2, 3]])

kernel = np.array([[1, -1],
                   [-1, 1]])

bias_conv = 1.0
```

```
printMatrix(input, "Input image")
printMatrix(kernel, "Kernel")
```

```
Input image
[[1 2 0 2 1]
 [0 1 1 2 1]
 [1 2 0 1 0]
 [1 2 1 2 1]
 [0 0 3 2 3]]
```

```
---
Kernel
[[ 1 -1]
 [-1  1]]
---
```

```
In [4]: z_conv = convolution(input, kernel, bias_conv)

a_relu = ReLU(z_conv)

a_max_pool, idx_max = max_pool(a_relu, kernel.shape[0], stride=2)

a_flatten = a_max_pool.flatten()

printMatrix(z_conv, "Conv")
printMatrix(a_relu, "ReLU")
printMatrix(a_max_pool, "Max Pool")
printMatrix(a_flatten.reshape(1,4), "Flatten")
```

```
Conv
[[ 1.  3.  0.  1.]
 [ 1. -1.  1.  1.]
 [ 1.  2.  1.  1.]
 [ 0.  5. -1.  3.]]
```

```
---
ReLU
[[1. 3. 0. 1.]
 [1. 0. 1. 1.]
 [1. 2. 1. 1.]
 [0. 5. 0. 3.]]
```

```
---
Max Pool
[[3. 1.]
 [5. 3.]]
```

```
---
Flatten
[[3. 1. 5. 3.]]
---
```

Fully Connected layer

```
In [5]: weight_fcnn = np.array([1, 1, -1, 1])
bias_fcnn = 0

z_fcnn = weight_fcnn.dot(a_flatten) + bias_fcnn
a_fcnn = sigmoid(z_fcnn)
loss = binary_cross_entropy(1, a_fcnn)
print(f"Activation FCNN: {a_fcnn}")
print(f"Binary cross-entropy loss: {loss}")
```

```
Activation FCNN: 0.8807970779778823
Binary cross-entropy loss: 0.05512413479491803
```

```
In [6]: printMatrix(bias_conv, "Bias Conv")
```

```
printMatrix(kernel, "Kernel")
printMatrix(weight_fcnn, "Weight FCNN")
printMatrix(bias_fcnn, "Bias FCNN")
```

```
Bias Conv
1.0
---
Kernel
[[ 1 -1]
 [-1  1]]
---
Weight FCNN
[ 1  1 -1  1]
---
Bias FCNN
0
---
```

Backward Pass

```
In [7]: learning_rate = 0.1
```

```
In [8]: delta = d_binary_cross_entropy(1, a_fcnn) * d_sigmoid(z_fcnn)
print(f"dL / d(bias_fcnn) = {delta}")
bias_fcnn = bias_fcnn - learning_rate * delta

delta = delta * a_flatten
print(f"dL / d(weight_fcnn) = {delta}")
weight_fcnn = weight_fcnn - learning_rate * delta

delta = delta.reshape(a_max_pool.shape)
printMatrix(delta, "Reshaped derivative")

dL / d(bias_fcnn) = -0.11920292202211769
dL / d(weight_fcnn) = [-0.35760877 -0.11920292 -0.59601461 -0.35760877]
Reshaped derivative
[[-0.35760877 -0.11920292]
 [-0.59601461 -0.35760877]]
---
```

```
In [9]: calculate_delta = np.zeros(a_relu.shape)

for idx, grad in zip(idx_max, delta.flatten()):
    calculate_delta[idx[0], idx[1]] = grad

delta = calculate_delta
printMatrix(a_relu, "a_relu")
printMatrix(delta, "dL/d(a_relu)")

delta = d_ReLU(z_conv) * delta
printMatrix(delta, "dL/d(z_conv)")

grad_b1 = delta.sum()
print(f"dL / d(grad_b1) = {grad_b1}")
bias_conv = bias_conv - learning_rate * grad_b1
```

```

a_relu
[[1. 3. 0. 1.]
 [1. 0. 1. 1.]
 [1. 2. 1. 1.]
 [0. 5. 0. 3.]]
---
dL/d(a_relu)
[[ 0.          -0.35760877  0.          -0.11920292]
 [ 0.           0.           0.           0.          ]
 [ 0.           0.           0.           0.          ]
 [ 0.          -0.59601461  0.          -0.35760877]]
---
dL/d(z_conv)
[[ 0.          -0.35760877  0.          -0.11920292]
 [ 0.           0.           0.           0.          ]
 [ 0.           0.           0.           0.          ]
 [ 0.          -0.59601461  0.          -0.35760877]]
---
dL / d(grad_b1) = -1.4304350642654122

```

```

In [10]: grad_kernel = np.zeros(kernel.shape)

grad_kernel[0,0] = np.sum(input[0:4, 0:4] * delta)
grad_kernel[0,1] = np.sum(input[0:4, 1:5] * delta)
grad_kernel[1,0] = np.sum(input[1:5, 0:4] * delta)
grad_kernel[1,1] = np.sum(input[1:5, 1:5] * delta)

kernel = kernel - learning_rate * grad_kernel
printMatrix(grad_kernel, "dL/d(grad_kernel)")

dL/d(grad_kernel)
[[-2.86087013 -1.0728263 ]
 [-1.31123214 -3.33768182]]
---

```

Updated values

```

In [11]: printMatrix(weight_fcnn, "Updated Weight FCNN")
printMatrix(bias_fcnn, "Updated Bias FCNN")
printMatrix(kernel, "Updated Kernel")
printMatrix(bias_conv, "Updated Bias Conv")

Updated Weight FCNN
[ 1.03576088  1.01192029 -0.94039854  1.03576088]
---
Updated Bias FCNN
0.011920292202211769
---
Updated Kernel
[[ 1.28608701 -0.89271737]
 [-0.86887679  1.33376818]]
---
Updated Bias Conv
1.1430435064265412
---

```