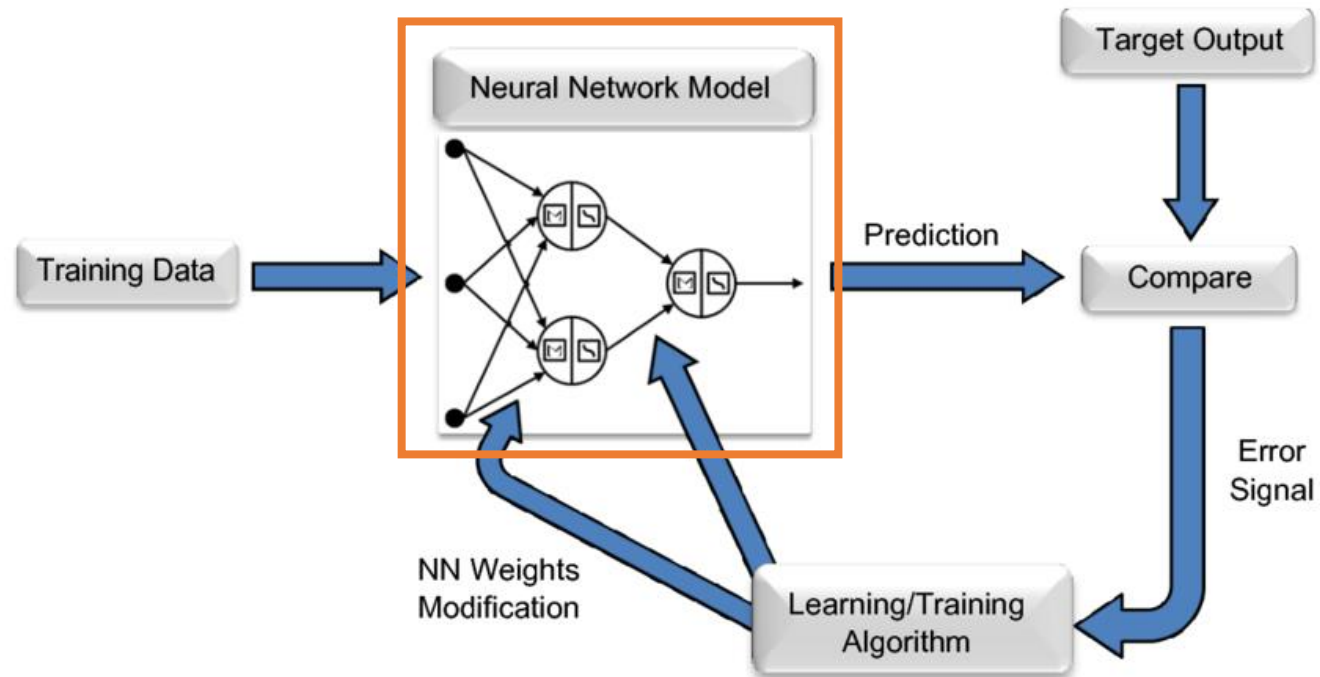


인공지능의 기초

3주차

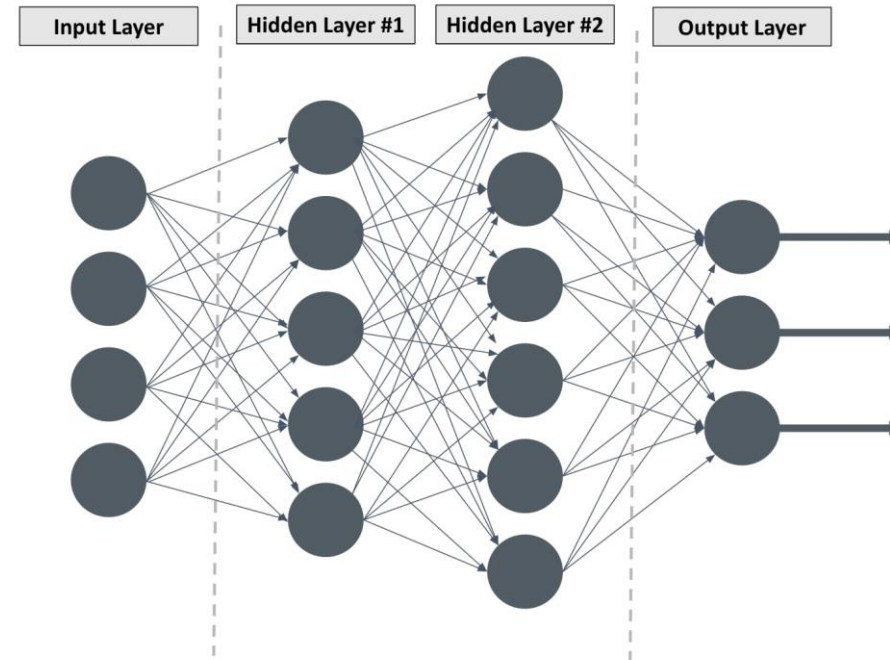
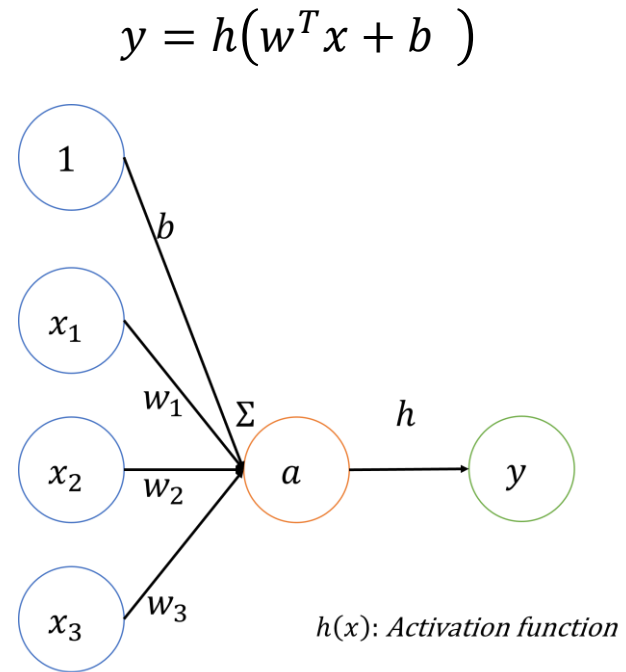
Review

- 딥러닝 학습 과정



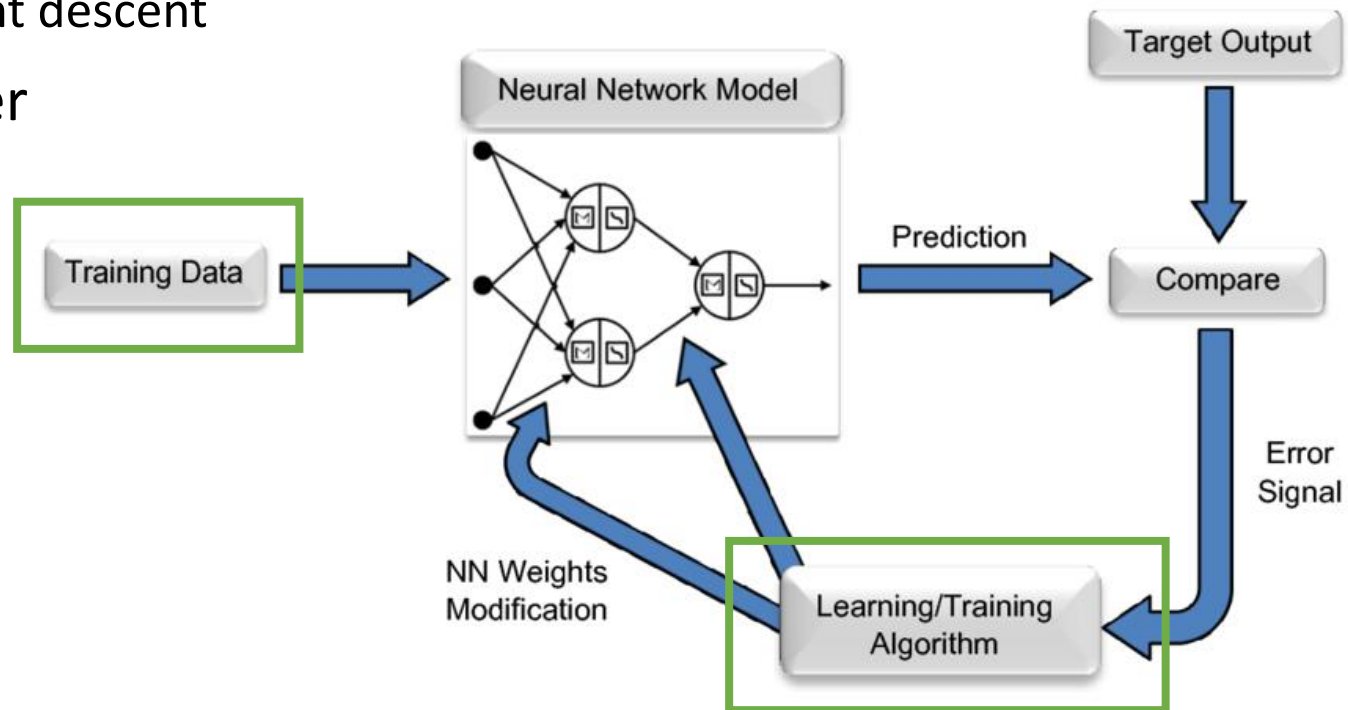
Review

- Deep Neural Network



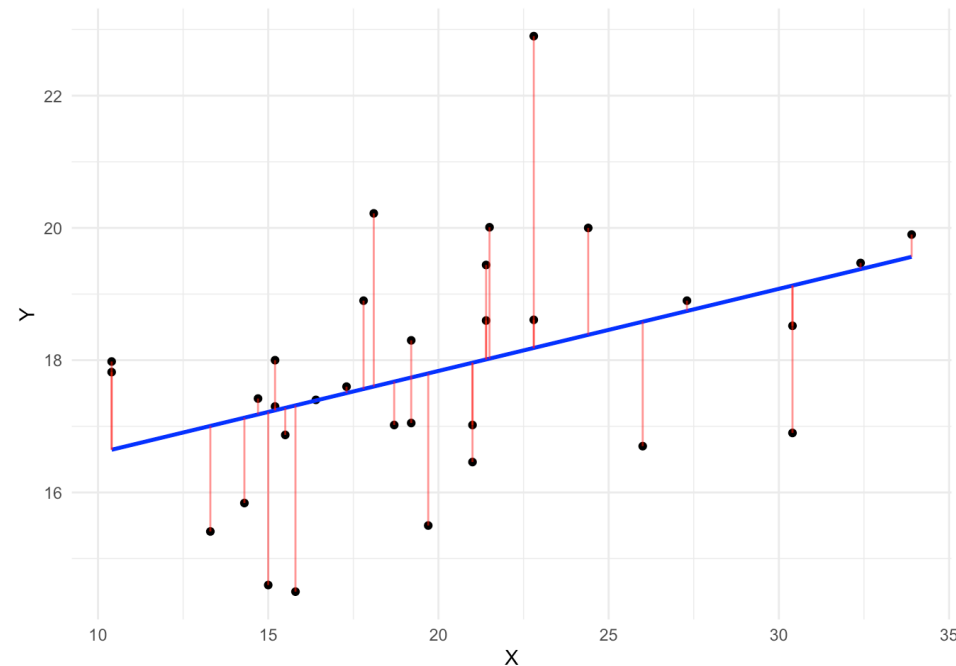
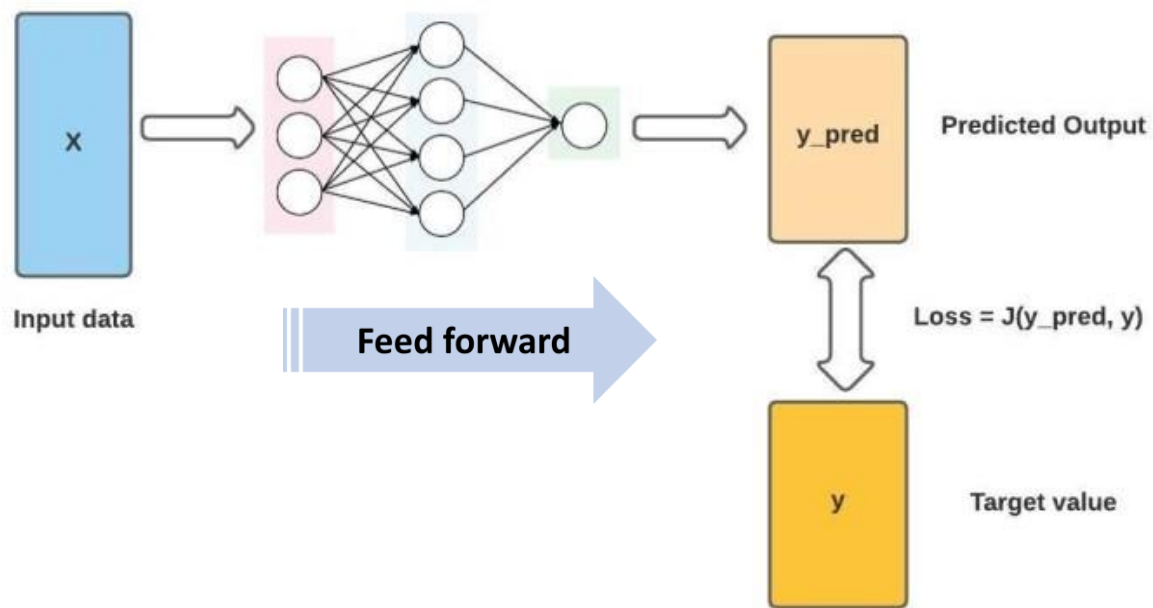
Today

- Backpropagation
 - Loss function
 - Weight update
 - Gradient descent
- Dataloader



사전지식

- Loss function
- 우리의 목표는 loss를 줄이는 것!



$$MAE = \frac{1}{N} \sum_{i=1}^n |\hat{y}_i - y_i|$$

$$MSE = \frac{1}{N} \sum_{i=1}^n (\hat{y}_i - y_i)^2$$

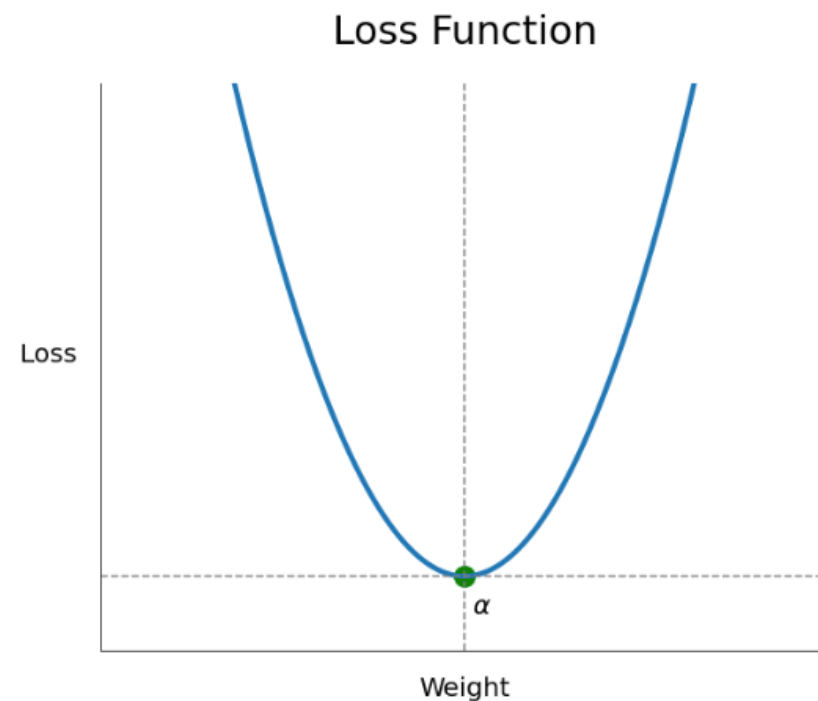
사전지식

- Weight update
- Loss를 최소화하기 위한 weight를 찾자!

$$MSE = \frac{1}{N} \sum_{i=1}^n (\hat{y}_i - y_i)^2$$



$$y = f(w, x)$$



사전지식

- Gradient descent
- 주어진 함수의 극소점을 찾는 알고리즘

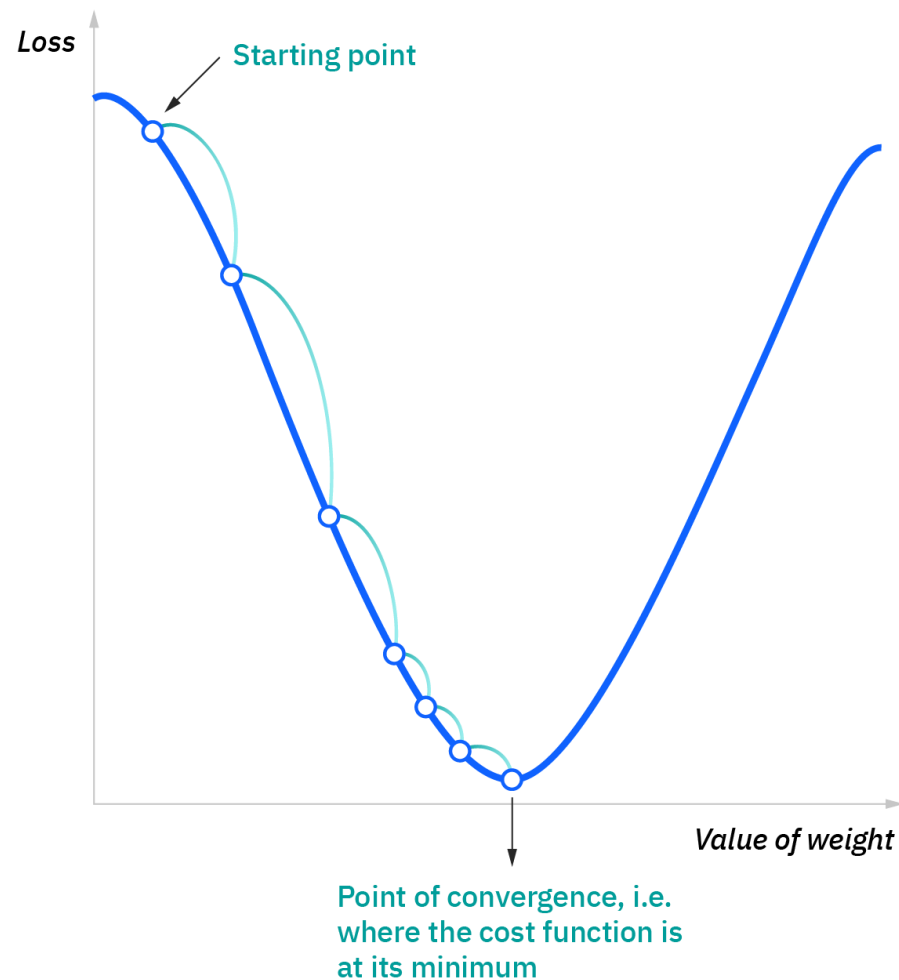
Gradient Descent

Repeat until converge {

$$w = w - \alpha \left[\frac{\partial Loss}{\partial w} \right]$$

$$b = b - \alpha \left[\frac{\partial Loss}{\partial b} \right]$$

}



Backpropagation

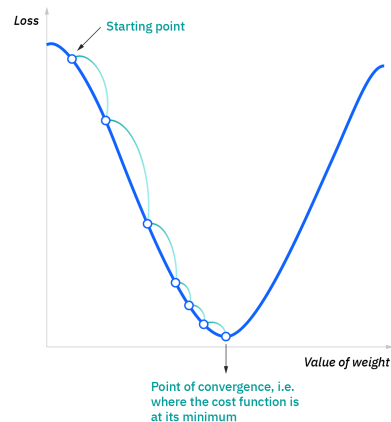
- Feedforward 과정을 통하여 입력으로 출력을 계산해낸다.
- Loss를 구한다.
- 각 weight가 Loss에 얼마나 영향을 미치는지 추론한다.
 - Chain rule을 응용한다.
- Weight를 업데이트한다. (gradient descent)

The Chain Rule

If $y = f(u)$, where $u = g(x)$

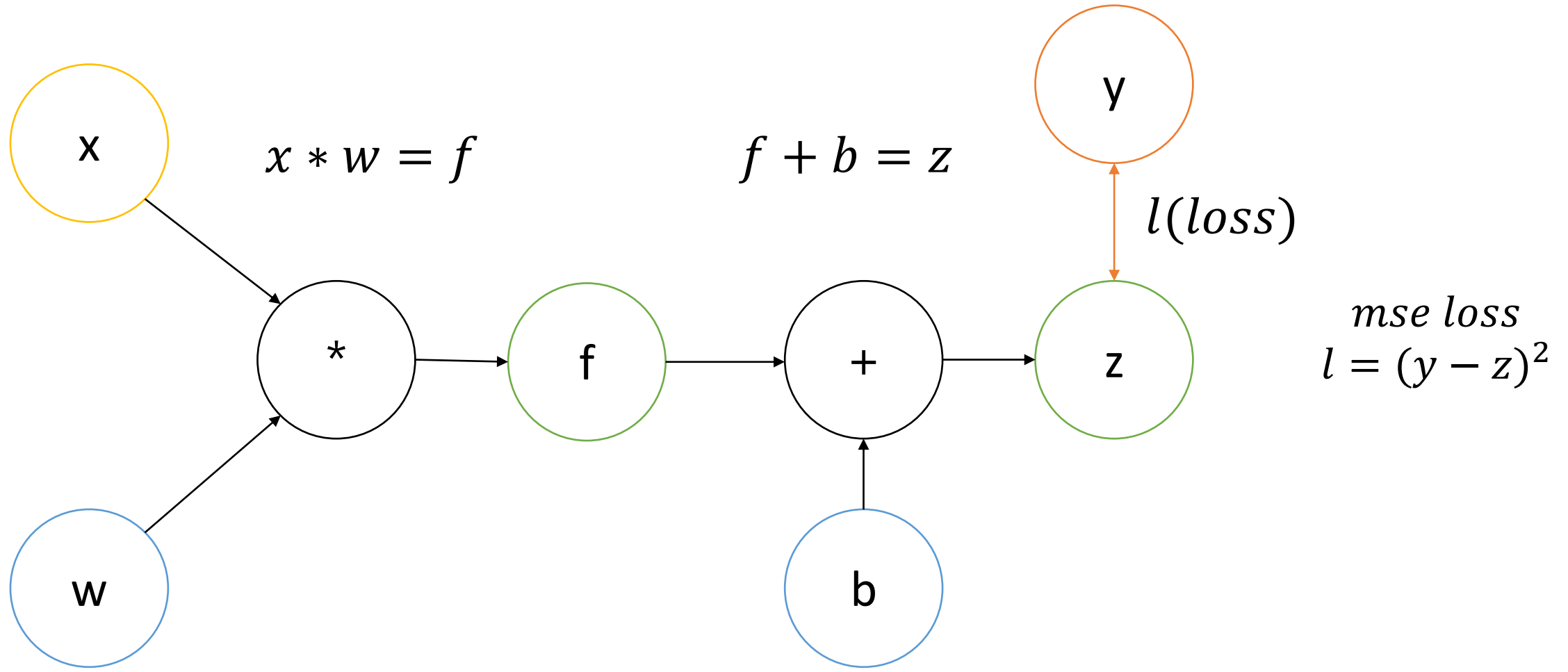
$$\frac{dy}{dx} = \frac{dy}{du} \cdot \frac{du}{dx}$$

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$$w = w - \alpha \frac{\partial loss}{\partial w}$$

Backpropagation

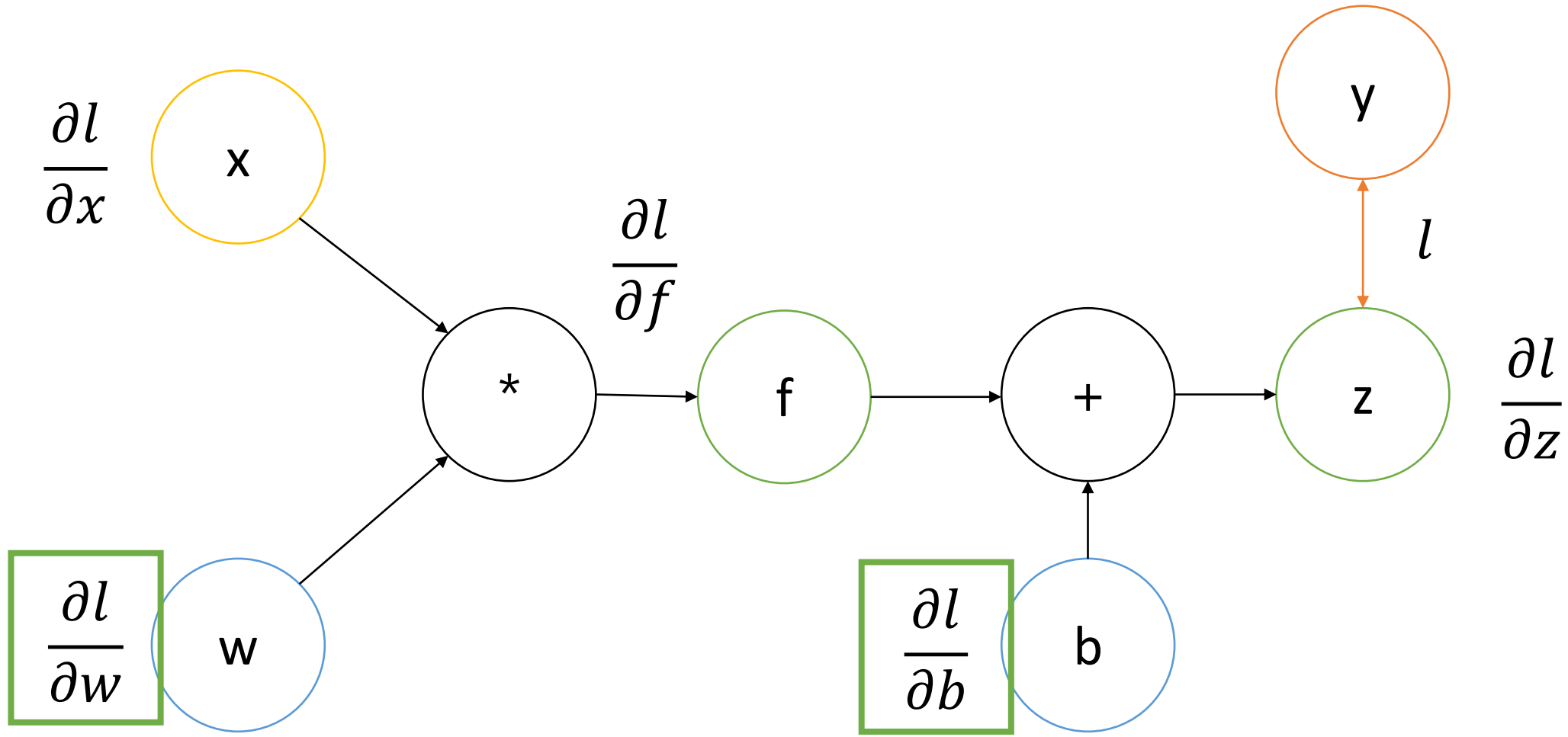


$$x * w = f$$

$$f + b = z$$

$$l = (y - z)^2$$

Backpropagation

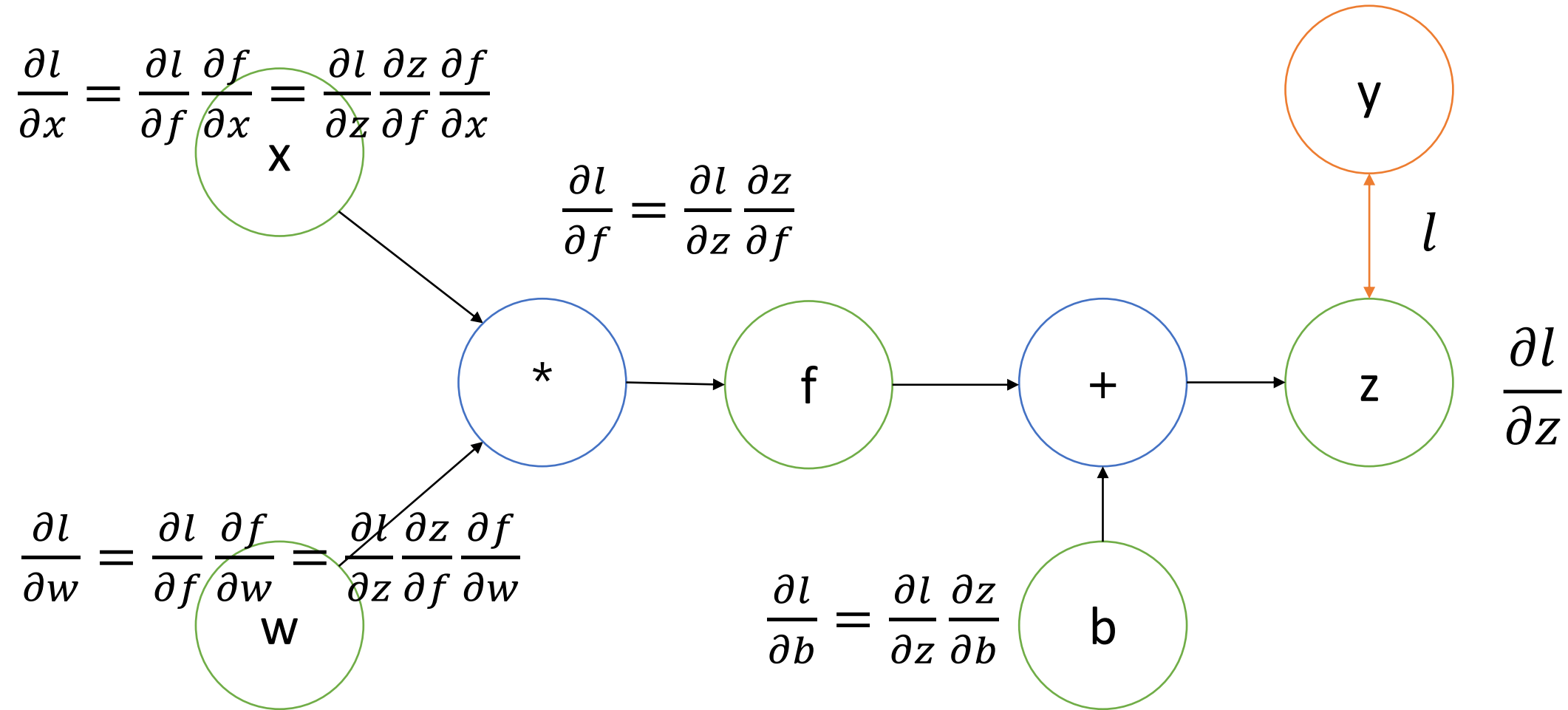


$$x * w = f$$

$$f + b = z$$

$$l = (y - z)^2$$

Backpropagation



Backpropagation

$$x * w = f$$

$$f + b = z$$

$$l = (y - z)^2$$

$$\frac{\partial f}{\partial x} = w$$

$$\frac{\partial z}{\partial f} = 1$$

$$\frac{\partial l}{\partial y} = 2y - 2z$$

$$\frac{\partial z}{\partial w} = b$$

$$\frac{\partial z}{\partial b} = 1$$

$$\frac{\partial l}{\partial z} = 2z - 2y$$

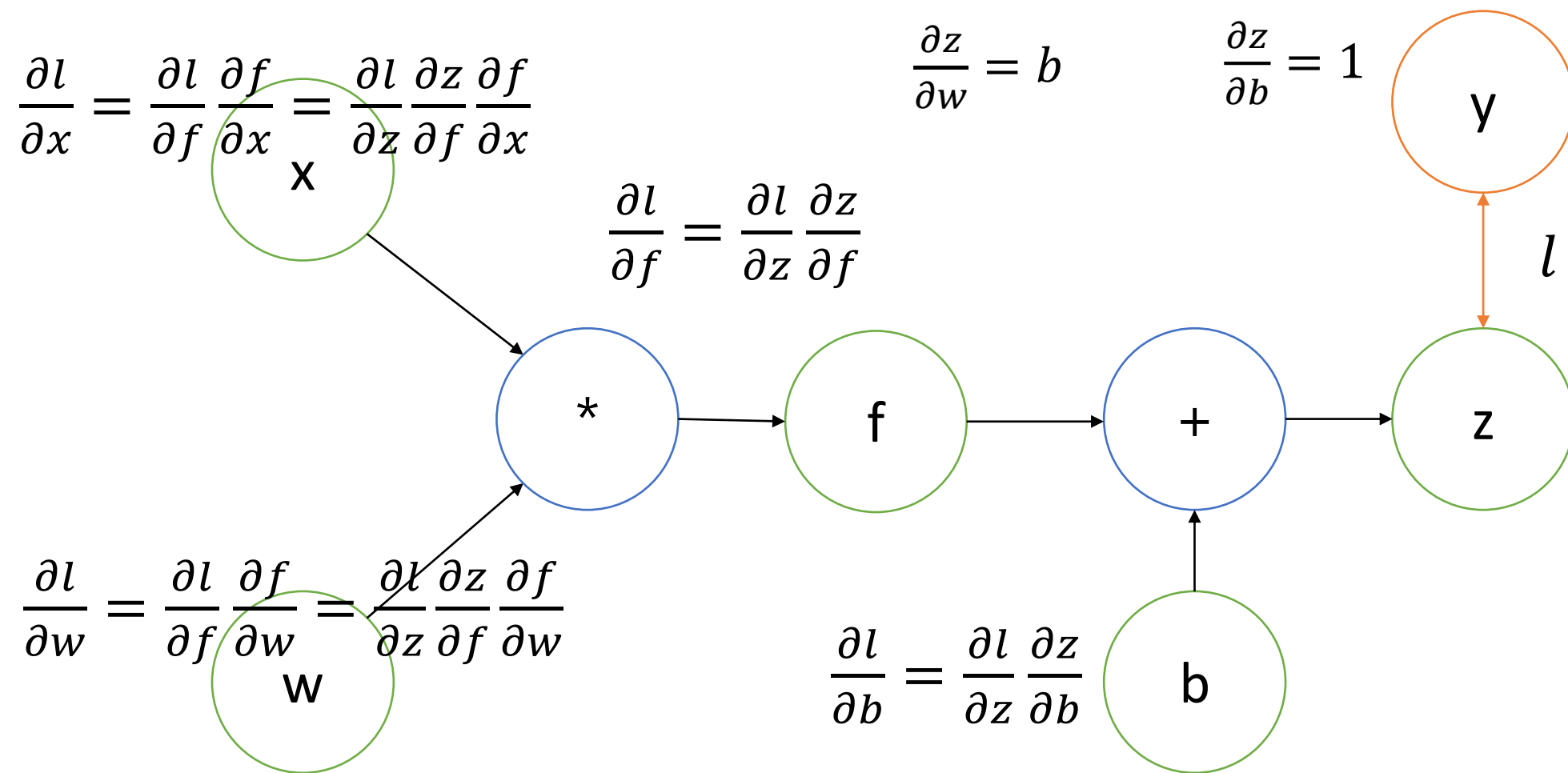
$$\frac{\partial l}{\partial x} = \frac{\partial l}{\partial f} \frac{\partial f}{\partial x} = \frac{\partial l}{\partial z} \frac{\partial z}{\partial f} \frac{\partial f}{\partial x}$$

$$\frac{\partial l}{\partial f} = \frac{\partial l}{\partial z} \frac{\partial z}{\partial f}$$

$$\frac{\partial l}{\partial w} = \frac{\partial l}{\partial f} \frac{\partial f}{\partial w} = \frac{\partial l}{\partial z} \frac{\partial z}{\partial f} \frac{\partial f}{\partial w}$$

$$\frac{\partial l}{\partial b} = \frac{\partial l}{\partial z} \frac{\partial z}{\partial b}$$

$$\frac{\partial l}{\partial z}$$



Backpropagation

$$x * w = f$$

$$f + b = z$$

$$l = (y - z)^2$$

$$\frac{\partial f}{\partial x} = w$$

$$\frac{\partial z}{\partial f} = 1$$

$$\frac{\partial l}{\partial y} = 2y - 2z$$

$$\frac{\partial z}{\partial w} = b$$

$$\frac{\partial z}{\partial b} = 1$$

$$\frac{\partial l}{\partial z} = 2z - 2y$$

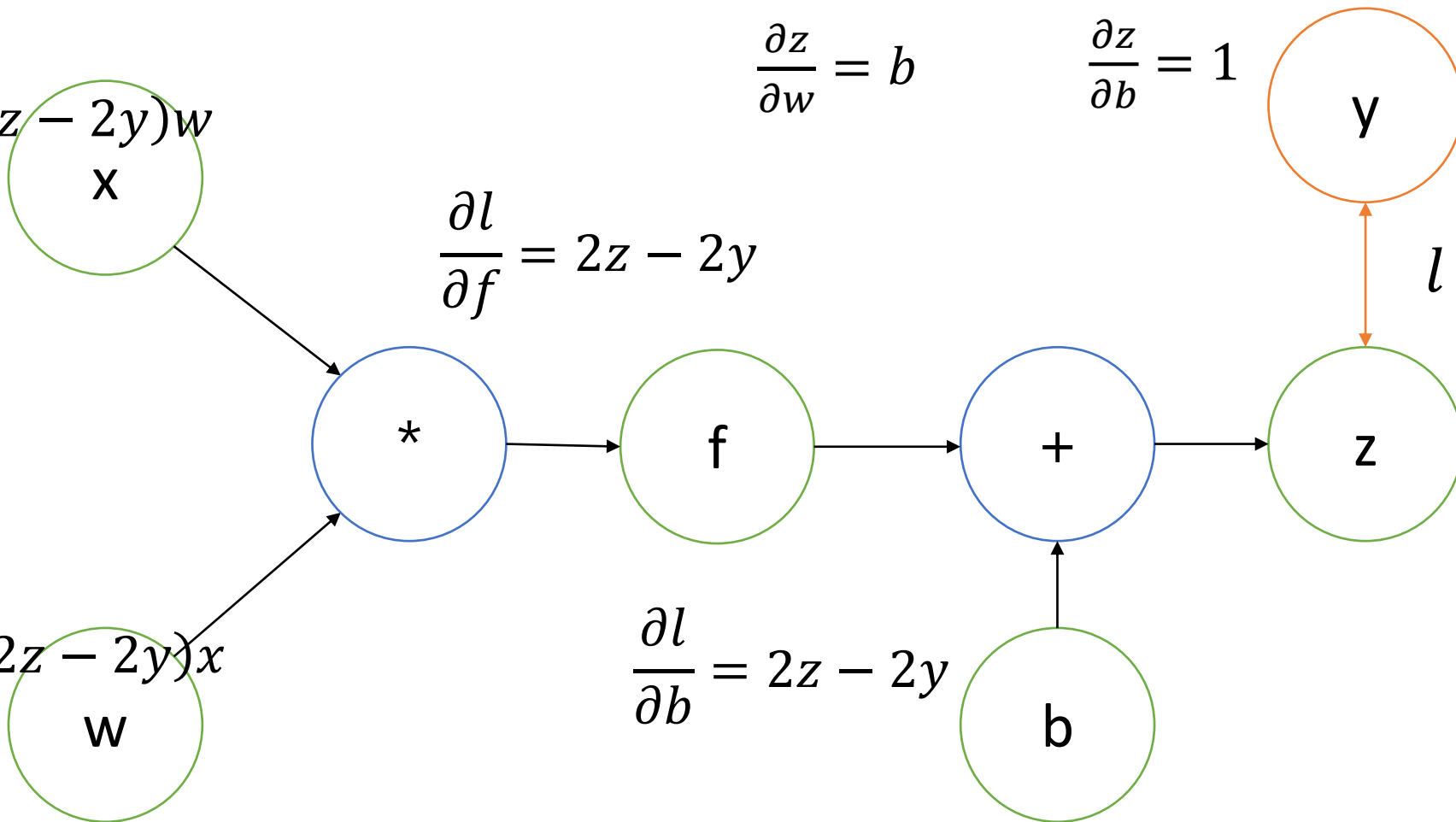
$$\frac{\partial l}{\partial x} = (2z - 2y)w$$

$$\frac{\partial l}{\partial f} = 2z - 2y$$

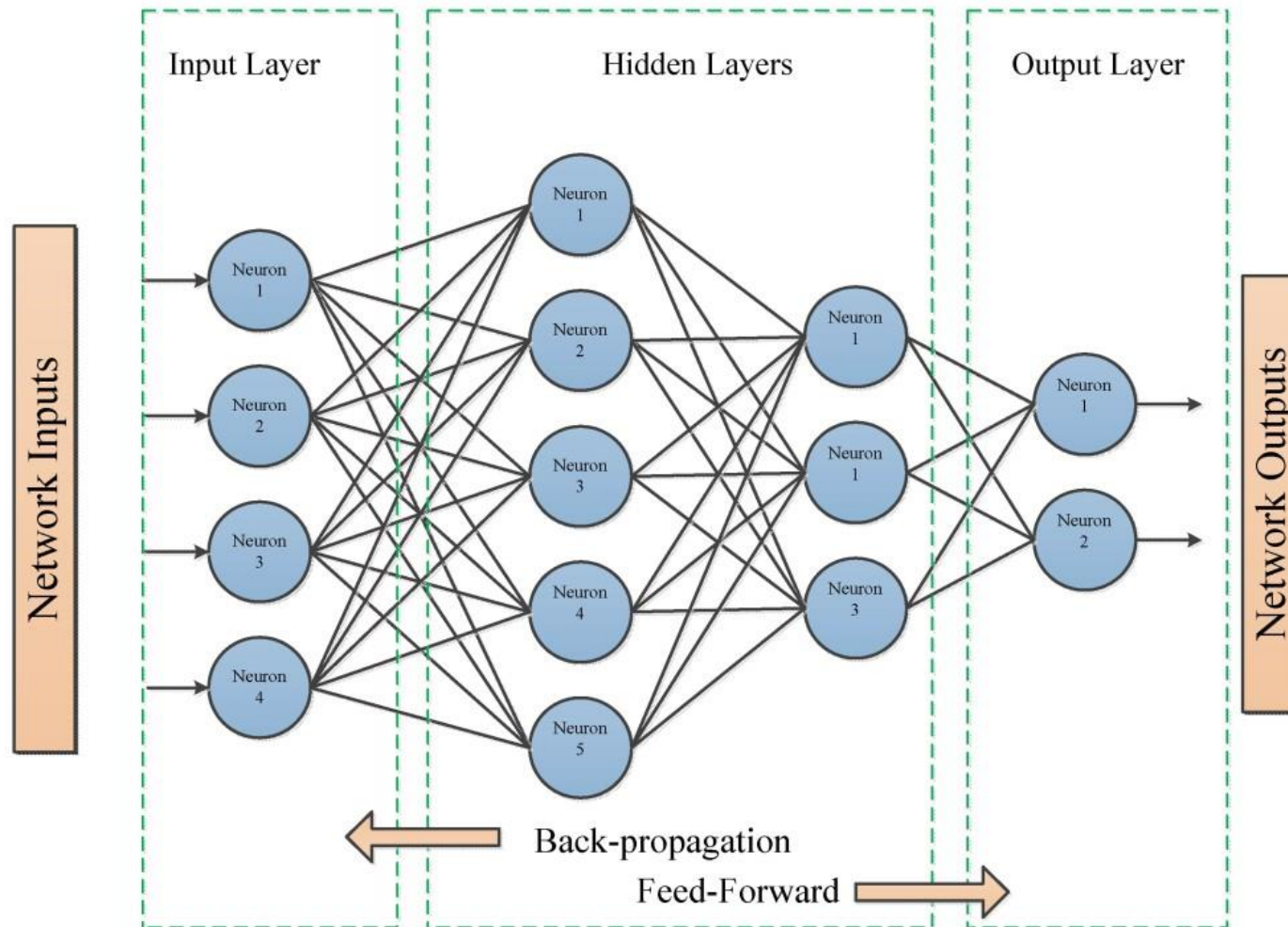
$$\frac{\partial l}{\partial z} = 2z - 2y$$

$$\frac{\partial l}{\partial w} = (2z - 2y)x$$

$$\frac{\partial l}{\partial b} = 2z - 2y$$



Backpropagation in Deep Neural Network



실습

```
import torch

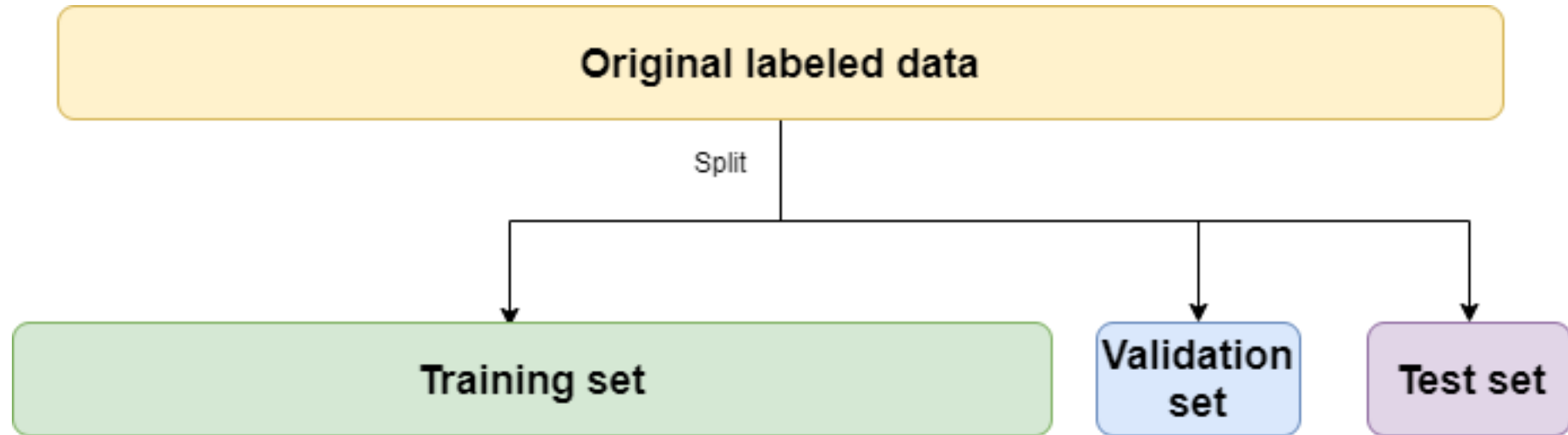
x = torch.ones(1)*2 # input tensor
y = torch.zeros(1) # expected output
w = torch.randn(1, 1, requires_grad=True)
b = torch.randn(1, requires_grad=True)
z = torch.matmul(x, w)+b
#loss = torch.nn.MSELoss(z, y, reduction='none')
loss = torch.mean((z - y)**2)

print('x: ', x)
print('w: ', w)
print('b: ', b)
print('z: ', z)
print('loss: ', loss)
```

```
loss.backward()
print('w gradient: ', w.grad)
print('b gradient: ', b.grad)
```

```
print(2*(z-y)*x)
print(2*(z-y))
```

Dataloader



실습

```
import torch
from torch.utils.data import Dataset
from torchvision import datasets
from torchvision.transforms import ToTensor
import matplotlib.pyplot as plt
```

```
training_data = datasets.FashionMNIST(
    root="data",
    train=True,
    download=True,
    transform=ToTensor()
)
```

```
test_data = datasets.FashionMNIST(
    root="data",
    train=False,
    download=True,
    transform=ToTensor()
)
```

```
labels_map = {
    0: "T-Shirt",
    1: "Trouser",
    2: "Pullover",
    3: "Dress",
    4: "Coat",
    5: "Sandal",
    6: "Shirt",
    7: "Sneaker",
    8: "Bag",
    9: "Ankle Boot",
}

figure = plt.figure(figsize=(8, 8))
cols, rows = 3, 3
for i in range(1, cols * rows + 1):
    sample_idx = torch.randint(len(training_data), size=(1,)).item()
    img, label = training_data[sample_idx]
    figure.add_subplot(rows, cols, i)
    plt.title(labels_map[label])
    plt.axis("off")
    plt.imshow(img.squeeze(), cmap="gray")
plt.show()
```

실습

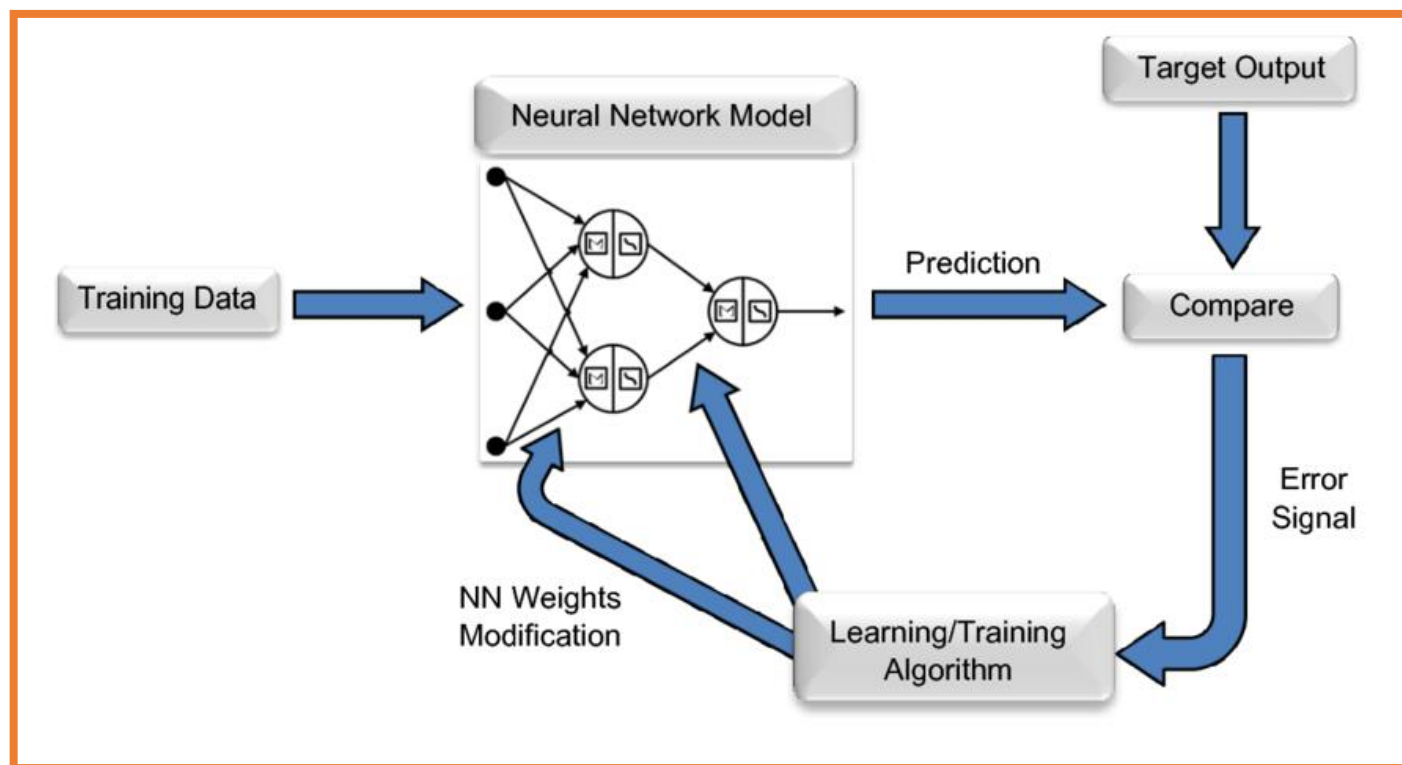
```
from torch.utils.data import DataLoader
```

```
train_dataloader = DataLoader(training_data, batch_size=64, shuffle=True)  
test_dataloader = DataLoader(test_data, batch_size=64, shuffle=True)
```

```
# Display image and label.  
train_features, train_labels = next(iter(train_dataloader))  
print(f"Feature batch shape: {train_features.size()}")  
print(f"Labels batch shape: {train_labels.size()}")  
img = train_features[0].squeeze()  
label = train_labels[0]  
plt.imshow(img, cmap="gray")  
plt.show()  
print(f"Label: {label}")
```

예고

■ Training



감사합니다

시험 잘 보세요!!

출처

- https://pytorch.org/tutorials/beginner/basics/autogradqs_tutorial.html
- https://pytorch.org/tutorials/beginner/basics/data_tutorial.html