

ML/DL for Everyone Season2

with PYTORCH

Linear Regression

Code: <https://github.com/deeplearningzerotoall/PyTorch>

Slides:

Video:

Lecturer: 이승재 (Princeton University) <seungjaeryanlee@gmail.com>



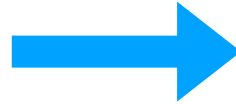
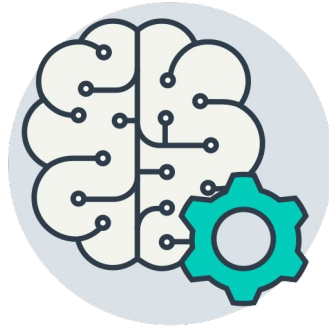
다루고자 하는 주제

- Data definition
- Hypothesis
- Compute loss
- Gradient descent

Data definition

What would be the grade if I study 4 hours?

4
hours



?
points

Prediction

Hours (x)	Points (y)
1	2
2	4
3	6
4	?

Training dataset

Test dataset

Data definition

```
x_train = torch.FloatTensor([[1], [2], [3]])  
y_train = torch.FloatTensor([[2], [4], [6]])
```

$$X_{\text{train}} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \quad Y_{\text{train}} = \begin{pmatrix} 2 \\ 4 \\ 6 \end{pmatrix}$$

Hours (x)	Points (y)
1	2
2	4
3	6

- 데이터는 `torch.tensor!`
- 입력 따로, 출력 따로!
 - 입력: `x_train`
 - 출력: `y_train`
 - 입출력은 `x, y` 로 구분

Hypothesis

$$y = Wx + b$$

Weight

Bias

Hypothesis

```
x_train = torch.FloatTensor([[1], [2], [3]])  
y_train = torch.FloatTensor([[2], [4], [6]])
```

```
W = torch.zeros(1, requires_grad=True)  
b = torch.zeros(1, requires_grad=True)  
hypothesis = x_train * W + b
```

- Weight 와 Bias 0으로 초기화
 - 항상 출력 0을 예측
- requires_grad=True
 - 학습할 것이라고 명시

$$y = Wx + b$$

Compute loss

Mean Squared Error (MSE)

$$\text{cost}(W, b) = \frac{1}{m} \sum_{i=1}^m \left(\underbrace{H(x^{(i)})}_{\text{Prediction}} - \underbrace{y^{(i)}}_{\text{Target}} \right)^2$$

Mean

Compute loss

```
x_train = torch.FloatTensor([[1], [2], [3]])  
y_train = torch.FloatTensor([[2], [4], [6]])
```

```
W = torch.zeros(1, requires_grad=True)  
b = torch.zeros(1, requires_grad=True)  
hypothesis = x_train * W + b
```

```
cost = torch.mean((hypothesis - y_train) ** 2)
```

- `torch.mean` 으로 평균 계산!
- 한 줄인데 읽기 편한 코드 :)

$$\frac{1}{m} \sum_{i=1}^m \left(H(x^{(i)}) - y^{(i)} \right)^2$$

Gradient descent

```
x_train = torch.FloatTensor([[1], [2], [3]])  
y_train = torch.FloatTensor([[2], [4], [6]])
```

```
W = torch.zeros(1, requires_grad=True)  
b = torch.zeros(1, requires_grad=True)  
hypothesis = x_train * W + b
```

```
cost = torch.mean((hypothesis - y_train) ** 2)
```

```
optimizer = optim.SGD([W, b], lr=0.01)  
  
optimizer.zero_grad()  
cost.backward()  
optimizer.step()
```

- `torch.optim` 라이브러리 사용
 - `[W, b]` 는 학습할 `tensor`들
 - `lr=0.01` 은 learning rate
- 항상 붙어다니는 3줄
 - `zero_grad()` 로 gradient 초기화
 - `backward()` 로 gradient 계산
 - `step()` 으로 개선

Full training code

```
x_train = torch.FloatTensor([[1], [2], [3]])
y_train = torch.FloatTensor([[2], [4], [6]])

W = torch.zeros(1, requires_grad=True)
b = torch.zeros(1, requires_grad=True)

optimizer = optim.SGD([W, b], lr=0.01)

nb_epochs = 1000
for epoch in range(1, nb_epochs + 1):
    hypothesis = x_train * W + b
    cost = torch.mean((hypothesis - y_train) ** 2)

    optimizer.zero_grad()
    cost.backward()
    optimizer.step()
```

한번만

1. 데이터 정의
2. Hypothesis 초기화
3. Optimizer 정의

반복!

1. Hypothesis 예측
2. Cost 계산
3. Optimizer 로 학습

What's Next?

- How does gradient descent minimize cost?